Modelling of gas-liquid precipitation systems

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Abstract

The performance of gas-liquid precipitation in industrial reactors, such as bubble columns, is determined by the interplay between multiphase fluid dynamics, gas-liquid reaction engineering and crystallisation mechanisms such as nucleation, growth and agglomeration. The aim of this work is to develop efficient computational methods for dealing with these issues.

The coupling of turbulent mixing and chemical phenomena lies at the heart of multiphase reaction engineering, but direct CFD approaches are usually confronted with excessive computational demands. A hybrid approach is proposed here, where the quantification of mixing is accomplished through averaging the flow and concentration profiles resulting from a CFD flow field calculation and a computational ("virtual") tracer experiment. These results form the basis for a mapping of the CFD grid into a generalised compartmental model, where the chemistry calculations can be carried out efficiently. A phenomenological model of mass transfer and chemical reaction is employed to simulate the interfacial phenomena in gas-liquid reactors, but the main concept lends itself to a variety of chemical reaction engineering problems.

A first-order finite element scheme is proposed for solving the one-dimensional dynamic population balance equation with nucleation, growth, aggregation and breakage. The numerical scheme features improved stability by taking into account the convective nature of the growth term, and enhanced conservation of moments at non-uniform grids by identifying and mitigating the sources of conservation error in the evaluation of the aggregation source term. Exhaustive benchmarking is carried out against analytical solutions, and in all cases tested the method proves capable of producing accurate results at a reasonable CPU time.

Finally, the theoretical and computational methods developed in this work are applied to simulate the precipitation of CaCO₃ in a bubble column reactor. Dynamic measurements of the crystal size distribution (CSD) are carried out, and the evolution of the process is adequately reproduced by the model.
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Notation

Index notation is used throughout. Repeated indices imply summation while single indices denote iteration, i.e.,

$$\frac{\partial y_i}{\partial x_i} = \frac{\partial y_1}{\partial x_1} + \frac{\partial y_2}{\partial x_2} + \ldots + \frac{\partial y_n}{\partial x_n}$$

$$\frac{\partial}{\partial x_i} = \left[ \frac{\partial}{\partial x_1}, \frac{\partial}{\partial x_2}, \ldots, \frac{\partial}{\partial x_n} \right]$$

Although effort was taken to ensure a consistent notation, due to the fact that different chapters deal with different problems, occasionally the same symbol may be used to denote different variables in different chapters (e.g. index $i$ denotes chemical species in chapters 2, 4 but element boundaries in chapter 3). The notation is consistent within each chapter, however, and symbols are explained when first encountered.

The substantial derivative is defined as:

$$\frac{D}{Dt} = u_i \frac{\partial}{\partial x_i} + \frac{\partial}{\partial t}$$

Some of the acronyms frequently used are:

**CFD** Computational Fluid Dynamics

**CSD** Crystal Size Distribution

**PBE** Population Balance Equation
Introduction

"If it is not permitted to us to penetrate to a complete knowledge concerning the motion of fluids, it is not to mechanics, or to the insufficiency of the known principles of motion, that we must attribute the cause. It is analysis itself which abandons us here, since all the theory of the motion of fluids has just been reduced to the solution of analytical formulae."

Leonhard Euler, 1755

During the last decades, engineering and process design have been revolutionized by the introduction of computational methods. Computational fluid dynamics (CFD) have emerged as a powerful, fully predictive tool for the design of processes and equipment where fluid mechanics play a key role. The application of CFD to chemical reaction engineering has lagged, however, and is confronted with unique challenges. The coupling of CFD with chemical phenomena gives rise to issues not usually encountered in purely hydrodynamic applications. These issues are both physical and computational, and are unlikely to be resolved in the near future. Chemical phenomena occur at scales much smaller than those characteristic of hydrodynamic phenomena, and the computational resources required to resolve these scales are unlikely to be available soon, unless unexpected breakthroughs in computer hardware are due.

This leads to the inevitable conclusion that hybrid methods are required, if we are to proceed from the realm of empiricism and make use of the currently available computational resources, without pursuing unfeasible targets. To develop them we should relax our requirements for fundamental models and settle for approaches that, though fully predictive, are partly based on phenomenological concepts. Actually, an element of phenomenology is always inevitable; even purely hydrodynamic problems require turbulence modelling. The first point of focus of this work is therefore the
proposition of a hybrid methodology for chemical reactor modelling, that makes use of both CFD and reaction engineering concepts.

Statistical concepts are inherent in a variety of reaction engineering problems: crystals, bubbles, droplets, polymers and cells are some examples of entities encountered in multiphase reactors that cannot be adequately described by a single value but require a distribution. The most rigorous method to account for such processes is the population balance, an integro-differential equation that describes the evolution of a particle size distribution. This equation can only be solved numerically, and methods often encounter problems regarding stability and conservation of the moments. The second point of focus is the development of a finite element scheme that makes certain advances towards resolving these problems.

Precipitation is a crystallisation process driven by supersaturation induced by chemical reactions - instead of cooling, evaporation of solvent etc. Gas-liquid precipitation is therefore the formation of a solid crystalline product via chemical reactions involving a gas and a liquid phase. The analysis of this process is a truly interdisciplinary problem, as it must account for a unique combination of physical and chemical phenomena: multiphase fluid dynamics, that determine several key reaction engineering variables such as the gas hold-up, chemical reactions may occur at both the gas-liquid interface and bulk, and various particulate mechanisms such as nucleation, growth and agglomeration play a role in determining the crystal size distribution (CSD) of the product. Therefore it encompasses all the fundamental problems which the aforementioned computational methods aim to address, and provides an excellent case study of applying them. The precipitation of CaCO$_3$ via the reaction of gaseous CO$_2$ and dissolved Ca(OH)$_2$ is considered here.

This thesis is structured as follows. Chapter 1 reviews the state of the art in four distinct fields that are relevant to the material of this work:

**Gas-liquid reaction engineering:** The analysis of mass transfer and chemical reaction, and the development of multiphase reactor models.

**Multiphase fluid dynamics:** The contribution of CFD to the analysis of multiphase flow; current status and problems.

**Population balance:** Emphasis will be on numerical solution techniques and the
problems that they face.

**Crystallisation:** Mechanistic models describing the crystallisation mechanisms (nucleation, growth and agglomeration), that must be integrated in the population balance.

In chapter 2, the hybrid methodology for multiphase reactor modelling is presented. A phenomenological model of mass transfer and chemical reaction for use with bubble column reactors is also employed, though the main concepts are applicable to a variety of reaction engineering problems. A $k-\varepsilon$-based code is employed for the CFD simulations, but the method is not restricted to it and will benefit from any future developments in turbulence modelling. The application of the method is demonstrated with a case study (CO$_2$ absorption in a bubble column).

The finite element method for solution of the population balance equation is developed in chapter 3. The accuracy of a numerical method can be tested by comparing against analytical solutions that have been found for special cases. Such benchmarking is carried out against solutions for size-dependent growth and nucleation, size-dependent aggregation with a variety of kernels, binary and multiple breakage with uniform and parabolic daughter particle distributions and combinations of the above, respectively.

Finally, these computational methods are applied to the gas-liquid precipitation problem (chapter 4). Experiments on CaCO$_3$ precipitation via reaction of CO$_2$ and Ca(OH)$_2$ in a 21lt bubble column are conducted and simulated with the aid of the model. Concluding remarks follow (chapter 5), while the computational implementation of the methods in this work is presented in the appendices in the form of Fortran 90 programs.
Chapter 1

Literature review

1.1 Gas-liquid reaction engineering

1.1.1 Fundamentals of gas-liquid mass transfer with chemical reaction

The interaction of mass transfer with chemical reactions at an interface lies at the heart of multiphase reaction engineering. The following discussion will outline the fundamental equations and the levels of approximation on which the current models are based. The fundamental mass transport equation for a chemical component can be written as 1:

\[
\frac{\partial C_i}{\partial t} + \frac{\partial}{\partial j} (u_j C_i) = \frac{\partial J_{ij}}{\partial j} + \sum_k r_k(C_i)
\]  

(1.1)

where \( J_{ij} \) is the diffusion of species \( i \) in the direction \( j \), usually described by Fick's law. For a rigorous derivation of the transport equations see e.g. Landau and Lifshitz (1959) or Bird, Stewart and Lightfoot (1960). When eq. 1.1 is applied to the gas-liquid interface, the hydrodynamic conditions are usually not known, so it must be coupled with the momentum (Navier-Stokes) equations:

\[
\frac{\partial (\rho u_j)}{\partial t} + \frac{\partial}{\partial j} (\rho u_j u_i) = \frac{\partial \tau_{jl}}{\partial j} - \frac{\partial p}{\partial j} + F_b
\]  

(1.2)

where \( \tau_{jl} \) is the stress tensor, while \( F_b \) stands for body forces such as gravity, if present. The microscopic scale of interfacial phenomena means that a discretisation

\footnote{In this section index \( i \) stands for chemical species, \( j \) or \( l \) for spatial dimension \( (x,y,z) \) and \( k \) for chemical reaction.}
in the sub-micron range is required to resolve them. Moreover, the location and shape of the interface is usually not known and must be determined by the solution. Therefore, few attempts have been made to resolve mass transfer along with the interfacial velocity field.

The early work of Lochiel and Calderbank (1964) considered mass transfer from bubbles as diffusion with a moving boundary and derived solutions for it using the approximations of boundary layer theory. Such models cannot be applied in many real situations, because they oversimplify both the velocity field and interface shape. Recent advances in CFD and interface tracking techniques have enabled the numerical simulation of free surface flow, and recently Shyy, Udaykumar, Rao and Smith (1997) have provided a review of numerical methods for simulation of interfaces. Liao and McLaughlin (2000) have solved the problem of mass transfer from bubbles numerically. Although such approaches are valuable for providing insight into the interfacial phenomena, they have no practical use for large scale reactor simulation, because that would imply resolving the whole domain with a microscopic grid. For the time being they are restricted to the simulation of single bubbles or small groups.

In order to formulate a numerically tractable interfacial mass transfer model with potential for coupling with complex reaction systems, simplification of the hydrodynamics is necessary. The most idealised hydrodynamic conditions emerge when the interface is stagnant, i.e. $u_j = 0$. In that case eq. 1.1 is reduced to the following:

$$\frac{\partial C_i}{\partial t} = \frac{\partial}{\partial j} J_{ij} + \sum_k r_j(C_i)$$  \hspace{1cm} (1.3)

If the interface is flat and very long in both lateral dimensions, the above equation would be one-dimensional in terms of space. By applying Fick’s law for the diffusive flux, we arrive at the following:

$$\frac{\partial C_i}{\partial t} = D \frac{\partial^2 C_i}{\partial x^2} + \sum_k r_k(C_i)$$  \hspace{1cm} (1.4)

This is a set of two-dimensional parabolic PDEs. Danckwerts (1970) identified this model with the case of absorption into a semi-infinite quiescent liquid and described experiments to which it can be directly applied. In multiphase reaction engineering, however, interest lies mainly in dispersed turbulent flows, which offer...
rapid mass transfer rates. Since resolution of interfacial hydrodynamics in such flows is very difficult, the proposed models attempt to draw analogies with idealised cases where hydrodynamics play no role (i.e. $u_j = 0$) such as in eq. 1.4. It follows that they are to a large degree phenomenological; yet they are often used due to their tractability.

1.1.1.1 Approximate models of absorption in turbulent dispersed flows

The film model is based on the concept that the gas must be dissolved in the surface by molecular conduction before it can be transferred to the interior by convection (Danckwerts, 1970). Interphase mass transfer is assumed to takes place within a thin layer next to the surface under steady state conditions. Eq. 1.4 is then reduced to the following system of ODEs:

$$D \frac{d^2 C_i}{dx^2} + \sum_k r_k(C_i) = 0$$  \hspace{1cm} (1.5)

If mass transfer resistance is in both gas and liquid phases, a two-film model is needed; however, this study is concerned with liquid-phase reactions where gas phase resistance is usually negligible (Deckwer, 1992). In the following, we shall refer to the liquid side of the interface. Initial conditions for the volatile components are estimated by an equilibrium relationship, such as Henry’s law, while at the end of the film the profiles must approach the bulk concentrations. The only parameter, the film thickness, must be estimated experimentally. The concept of the film model is not very realistic, as it implies an unnatural discontinuity of concentration near the surface; moreover, its solution for the case of physical absorption, assuming that the bulk is in steady state operation, shows the liquid side mass transfer coefficient $k_l$ to be proportional to the diffusion coefficient $D_i$, while experimental evidence rather suggests proportionality to $\sqrt{D}$ (Astarita, 1967). Nevertheless, it has been used in a number of studies due to its computational advantage of involving ordinary differential equations rather than partial, as do the other models.

A more realistic approach was proposed by Higbie (1935) and is known as the penetration model. Its physical reasoning is that elements or ‘parcels’ of liquid are assumed to be brought from the bulk to the surface by the flow pattern, and to remain there for a small amount of time, during which they behave like rigid
Literature review

bodies. The contact time is so small that the depth of gas penetration into the liquid element is negligible compared to the depth of the element itself. This way, it draws an analogy between gas absorption into agitated liquids and absorption into a quiescent semi-infinite liquid, and eq. 1.4 can be applied directly.

One initial and two boundary conditions are required. The initial conditions express that in the beginning of the contact time the liquid element has the bulk concentrations everywhere. On the gas side of the interface, Dirichlet boundary conditions are imposed for the volatile component, specifying its concentration by an equilibrium condition such as Henry’s law, while zero-flux conditions can be applied to non-volatile species. The assumption of small penetration depth and the approximation of the liquid element as a semi-infinite quiescent liquid are implemented by imposing the bulk side boundary conditions at an infinite distance from the interface:

\[ t > 0, x \to \infty \Rightarrow C_i = C_{i,\text{bulk}} \]  

(1.6)

The penetration model accounts for an important feature of the absorption mechanism: instead of being stationary, the interfacial layers are being constantly replenished. This replenishment can be either due to turbulence, as in the case of an agitated tank, or due to laminar flow over a packing or around a bubble, in which case the fluid is mixed after having crossed the bubble or solid object. The penetration model replaces the parameter of the film thickness with the contact time. In many cases it can only be empirically determined, but sometimes, such as in the case of rising bubbles, it can be estimated on the basis of physical arguments. On the other hand, the fact that partial differential equations are involved means that it is much more difficult to obtain solutions. For physical absorption (without reactions) the following analytical solution has been derived (Danckwerts, 1970):

\[ C_i = (C_i^* - C_{i,\text{bulk}}) \cdot \text{erf} \left( \frac{x}{2\sqrt{D\tau}} \right) + C_{i,\text{bulk}} \]  

(1.7)

Toor and Marchello (1958) proposed the following modification to the penetration model: the interface is assumed to have a fixed depth, such that the concentration gradients are zero when approaching the bulk. This has the same effect as the original Higbie model on the following condition:

\[ \tau \leq \frac{s^2}{D} \]  

(1.8)
This implies that it is more appropriate for short contact times. In effect it imposes the bulk-side boundary conditions at a finite distance $\delta$ instead of infinity, which is a major advantage for obtaining numerical solutions.

A shortcoming of the penetration model is that it assumes all elements to remain on the surface for exactly the same amount of time. To account for this, the Danckwerts model of surface renewal (Danckwerts, 1970) introduces the concept of age distribution: the contact time is not the same for all elements but follows a distribution instead. Eq. 1.4 still applies, but the absorption rate is calculated by integration over all element ages. Assuming that the probability of an element to be renewed is independent of its age, the average absorption rate would be:

$$J = s \int_{0}^{\infty} J e^{-\theta} d\theta$$

Eq. (1.9)

In essence, though, the surface renewal theory replaces the parameter of the contact time with that of renewal rate $s$. More complicated models include the eddy diffusivity theory, due to King (1966), which treats the problem of absorption in turbulent flows by splitting the diffusion coefficient into molecular and eddy diffusivity. This, however, increases the number of parameters to two. Another two-parameter approach is the surface-rejuvenation model. Two-parameter models have been seldom applied in numerical studies, because their higher degree of mechanistic insight is countered by the increased uncertainty in the parameter estimation.

### 1.1.1.2 Effect of chemical reaction on mass transfer

Solution of eqs. (1.4) or (1.5) becomes very complicated in the presence of chemical reactions (chemisorption), as reaction kinetics are highly non-linear. A comprehensive survey of gas-liquid reactions may be found in Doraiswamy and Sharma (1983). In some cases it is possible to decouple these phenomena by comparing their relative rates through the Hatta number, defined as:

$$Ha = \frac{\text{rate of reaction}}{\text{rate of absorption}}$$

(1.10)

The exact expression of the Hatta number depends upon the particular reaction system. According to this comparison, the following cases or absorption regimes can be identified:
• If the chemical reactions are much slower than mass transfer (kinetic control) then the situation can be considered as physical absorption, with chemical reactions occurring subsequently in the bulk. In this case, analytical solutions for mass transfer in both the film and penetration models can be derived.

• When the reactions are instantaneous (diffusion control) the reactants are assumed to diffuse up to a reaction plane situated at some distance inside the interface, where reactions happen instantaneously. This is another simplified case where mass transfer can be decoupled from chemical reactions and analytical solution is possible.

• If the reactions are fast but not instantaneous, they will occur simultaneously with mass transfer along the interface. This is clearly the most complex case, as reaction kinetics must be included in the mass transfer equations. Astarita (1967) and Danckwerts (1970) have obtained analytical solutions for some simple reactions (e.g. zero order, first order irreversible), but in the general case of a complex arbitrary reaction system numerical treatment is necessary. Versteeg, Kuipers, Van Beckum and Van Swaaij presented such a treatment for the single reversible (1989a) and the parallel reversible (1989b) reaction problems. Their solutions was based on both the film and penetration models, and a spatial transformation was introduced to impose the infinity boundary condition of the later. A similar treatment by Glasscock and Rochelle (1989) included the eddy diffusivity model. Van Swaaij and Versteeg (1992) presented a review of the topic and emphasised the necessity for numerical approach.

Another dimensionless modulus used to describe the interaction of absorption with chemical reactions is the enhancement factor:

\[ E_f = \frac{\text{rate of chemisorption}}{\text{rate of physical absorption}} \]  

In the fast reaction regime chemisorption is faster than physical absorption, because the consumption of the volatile component as it is being absorbed increases the driving force for mass transfer; absorption is thus "enhanced". Analytical expressions for the enhancement factor cannot be derived but for the simplest cases, where chemisorption can be resolved.
The rigorous analysis of gas-liquid phenomena presented in this chapter refers to the microscopic interfacial region. We shall now see how these models can be implemented in the modelling of industrial multiphase reactors.

1.1.2 Gas-liquid reactor models

The flow patterns that arise in industrial-scale gas-liquid reactors are so complex that detailed modelling of interfacial phenomena is not possible. Due to this complexity design methods remained for a long time experimental. The simplest way to include mass transfer in a reactor model is in the form of overall coefficients:

\[ J = k_{a}(C^* - C) \]  

(1.12)

Experimental approaches for determining overall mass transfer coefficients have been reviewed by, among others, Shah, Kelbar and Godbole (1982) and Deckwer (1992). Despite dimensionless analysis, empirical correlations for the mass transfer coefficient are rooted to the equipment used for extracting them and their reliability during scale-up is limited. Moreover, this approach can only be applied to steady state problems or simple reaction systems where reaction enhancement can be evaluated analytically. The complexity increases dramatically when mass transfer is accompanied by a fast chemical reaction. In this case mass transfer and chemical reactions are closely coupled, and the reaction enhancement effect can only be studied in detail by application of one of the gas-liquid absorption theories described above.

With respect to the hydrodynamic aspects of the problem, empirical stagewise or axial dispersion models have been frequently used. Deckwer, Hallensleben and Popovic (1980) employed the cell model with backflow to simulate physical absorption of CO\textsubscript{2} in a bubble column. Wachi and Morikawa (1986) applied the same model to the chlorination of ethylene, a problem of absorption with instantaneous chemical reaction. In the later, an analytical solution of the penetration model for the specific reaction system was employed to estimate the reaction enhancement, but the mass transfer coefficient was given by an empirical correlation. Schluter, Steiff and Weinspach (1992) adopted a similar approach for modelling mass transfer and reaction with both the cell model with backflow and the axial dispersion.
Wachi and Jones (1991) and Hostomský and Jones (1995) employed the film and penetration model respectively to model mass transfer, reaction and precipitation in a stirred tank reactor with a flat interface. A similar implementation of the penetration model was recently proposed to evaluate reaction enhancement in stirred vessels by Van Elk, Borman, Kuipers and Versteeg (2000), and the same authors made a stability and bifurcation analysis of the behaviour of gas-liquid reactors based on this approach (1999).

CFD models of gas-liquid reactors have appeared during the last decade, but are usually restricted to pure hydrodynamic simulation. Joshi (2001) has recently reviewed the current state of the art in the CFD modelling of bubble columns. Severe computational obstacles are met, however, when the coupling of gas-liquid mass transfer and hydrodynamics is considered. Torvik and Svendsen (1990) developed a steady-state two-dimensional CFD model of a gas-liquid reactor, while Fleischer, Becker and Eigenberger (1995, 1996) proposed an one-dimensional dynamic model for a thin and tall bubble column. In these studies, however, interfacial mass transfer and chemical reaction were decoupled. Recently Zahradnik et al. (2001) have applied the "network of zones" approach to gas-liquid reactor modelling. This approach, originating from the analysis of mixing in stirred tanks (Mann and Hackett, 1988), offers a computationally attractive alternative to the more expensive CFD models.

1.2 CFD modelling of gas-liquid flow

1.2.1 Conventional models and Computational Fluid Dynamics

Initially, hydrodynamic studies of bubble columns focused on experimental measurements and simple analytical models. The main design parameters were identified to be the gas superficial velocity, size, height-to-diameter ratio and the type and operation of sparger. These parameters were cast into dimensionless terms and correlated experimentally to the quantities that determine the performance of the reactor, such as slip velocity, liquid circulation rate, phase distribution and interfacial area. Such correlations along with analytical models were reviewed by Shah et al. (1981), Dekker (1992), Kastanek et al. (1993) and Joshi, Ranade et al. (1990). Due to the complexity of the phenomena underlying multi-phase flow however, experimental
correlations lack wide predictive potential and remain tied to the equipment used for their determination.

One of the first theoretical methods for the analysis of multi-phase flow was the drift flux model, proposed by Wallis (1969). Several studies during the 80's such as Koide et al. (1984), Jones (1985) and Chisti, Halard and Moo-Young (1987) attempted to predict the liquid circulation rate based on energy balances, while turbulence and other features that result in energy losses were implemented in the form of empirical coefficients. These models contained simple algebraic equations and were useful for interpreting experimental data, though quantitative agreement was difficult to achieve.

The developments in digital computers and numerical methods during the 70's made possible the numerical solution of the fundamental differential equations of fluid dynamics in single-phase flows, yet this approach could not be easily extended to multi-phase systems. In contrary to single-phase systems, a gap in our fundamental knowledge of the physics underlying multi-phase flow, especially of interfacial phenomena, meant that it would remain an active area of research. Ishii (1975) made one of the first attempts to establish the appropriate form of the fundamental equations governing multi-phase flows, and discussed various kinds of statistical averaging (time, space, ensemble) through which they can be derived. Stewart and Wendroff (1984) presented a review of the theoretical advances to-date, and concluded that fundamental knowledge was still lacking. Insofar as the computational aspects are concerned Spalding (1980) proposed the Inter-Phase Slip Algorithm (IPSA), an efficient way for numerical solution of the Eulerian-Eulerian formulation of the multi-phase flow equations, that became the industry standard. In the following years, some multi-phase flow problems were approached with computational methods.

It was mostly during the 90's, however, that computational modelling of bubble columns in particular was investigated. Schwarz and Turner (1988) and Svendsen (1992) carried out steady-state calculations from the Eulerian-Eulerian viewpoint. Sokolichin and Eigenberger (1994 I,II) introduced dynamic modelling and compared it to steady state; subsequent researchers, such as Boisson and Malin (1996), Sanyal et al. (1999) also explored the transient approach. Delnoij et al. (1997) discussed the application of the Volume of Fluid (VOF), a method for interface tracking, to
bubble rise studies. A recent review of the state of the art was presented by Jakobsen et al. (1997). Most recently, Pfleger, Gomes et al. (1999) studied the applicability of turbulence models to gas-liquid flow, while Thakre and Joshi (1999) discussed the drag force formulation.

1.2.2 Overview of computational models

In contrary to single-phase fluid flow, where it is generally accepted that the Navier-Stokes equations constitute an accurate model of the phenomena involved (including turbulence, in terms of instant velocity), the physics underlying multi-phase flows are not yet completely understood, despite considerable experimental and theoretical effort. A major difficulty lies in the fact that the structure of multi-phase flows can take completely different forms. In the case of vertical gas-liquid flow, for instance, the basic regimes have been identified as bubbly flow, slug flow, churn-turbulent flow, annular flow and mist flow, the main parameter that determines the regime being the gas superficial velocity (or the slip velocity, in case the liquid too is moving). In general it is not possible to fully predict the flow structure from a fundamental model; instead, the flow regime to be simulated must be born in mind in order to formulate an appropriate multi-phase flow model. Some models are very regime-specific, based on the particularities of each structure, while others are more general; however, a-priori knowledge of the regime seems indispensable at present.

Multi-phase models for gas-liquid flows can be generally classified into three types:

1. Phenomenological models that comprise algebraic equations based on simplifying assumptions and mechanistic insight.

2. Semi-fundamental models, derived from the Navier-Stokes through statistical averaging procedures that, however, give rise to terms whose closure must be carried out empirically.

3. Fundamental models, based on the Navier-Stokes equations, that require tracking of all interfaces.

Models of the first type were the only option at an age when differential models were too expensive to compute; yet they can sometimes be combined with a CFD
code or assist in estimating parameters. They will not be dealt with in this discussion. The third class of models is applicable in separated flows, or for modelling microscopic phenomena in dispersed flows; it will be briefly discussed in the next section. At present, large-scale CFD modelling of dispersed flows can only be carried out with models of the second type.

1.2.3 Fundamental models with interface tracking

When a single set of Navier-Stokes equations is used to describe a multiphase flow field, the variables exhibit a discontinuous change at the interfaces, that must in general be implemented with appropriate jump conditions. The main complexity in this kind of analysis arises from the fact that the location of the discontinuities is not known a priori, and a method for interface tracking (i.e. simulating the detailed shape of the interface) is required. Such methods have been reviewed by Shyy, Udaykumar, Rao and Smith (1997). Whatever the method, however, an essential prerequisite for this level of approximation is that the grid be fine enough to resolve the structure of the interface. This limits the applicability of this class of models to the following situations:

- Separated flows, e.g. annular flows, channel flows etc.
- Microscopic studies of dispersed flows, where single or few bubbles are being simulated.

In the case of multiphase reactors the model can only be used for microscopic simulation of bubbles or droplets, involving extremely fine grids. In the latter case, numerical diffusion may spoil the shape of the interface. This problem may be faced with techniques such as the one proposed by Hirt and Nichols (1981), which locates the interface and then moves fluid to either side of it. The incorporation of surface tension is a problem, as the exact orientation of the interface is not known with precision. Delnoij, Kuipers and van Swaaij (1997) applied this method (commonly referred to as Volume of Fluid) to simulate the formation and trajectory of a single bubble through a quiescent liquid. The reported computation time is rather high, and certainly the method is unsuitable for modelling and scaling up a bubble column as it operates at the microscopic scale.
1.2.4 Eulerian-Lagrangian model

In this model the liquid phase is treated as a continuum and the bubbles as discrete particles, in a way similar to the tracking of solid particles in combustion models. The continuous phase is viewed from an Eulerian perspective, i.e. with respect to a fixed coordinate system, and the complete Navier-Stokes equations are applied using a variable density, averaged on a volume fraction basis. For the particulate (bubble) phase a Lagrangian perspective is adopted, solving a momentum equation for each individual bubble. The momentum equation for the bubbles is written, using the substantial derivative, as:

\[
\frac{D\rho_g u_{g,j}}{Dt} = \sum F
\]  

This is the simplest form of Newton’s law, and \( \sum F \) is the sum of forces acting on each bubble. These may include: gravity force, drag force, pressure force, virtual mass, Basset history term and transversal lift (Jakobsen et al., 1997).

With this type of model, not only phase distribution but also individual bubble trajectories can be calculated. Furthermore, a distribution of bubble sizes can easily be introduced, since each bubble is accommodated by a separate equation, and even bubble-bubble interactions can be incorporated. Thus, this type of model is the most detailed among multi-fluid models for simulating bubbly flows. However, it has an apparent disadvantage: when a large number of bubbles is to be simulated computational time can be excessive, as a separate transport equation has to be solved for each one of them. Thus, it has been used mainly for fundamental studies of hydrodynamic phenomena, or to assess the validity of closure laws (such as equations for drag coefficients) to be used in an Eulerian-Eulerian two fluid model (Delnoij et al., 1997). Several studies have employed the Eulerian-Lagrangian model to simulate bubble columns, notably Lapin and Lubbert (1994) and Sokolichin et al. (1997). The correct form of the fluid-particle and fluid-fluid forces is still disputed, though. It has also been suggested (Delnoij et al., 1997) that this type of model, combined with a volume tracking model, could be used to simulate the churn flow in the heterogeneous bubble regime.
1.2.5 Eulerian-Eulerian model

In this model, the two phases are treated as interpenetrating continua. The interpretation of this concept is that the values of the variables at each point are what we would measure on average if we were to carry out a large number of experiments with the same conditions. Consequently, a complete set of Navier-Stokes equations is written for each phase. The transport equation of a scalar $\varphi$ takes the form:

$$\frac{\partial (u_j \rho \epsilon \varphi)}{\partial t} + \frac{\partial}{\partial x_j} (\epsilon \rho u_j \varphi) = \frac{\partial}{\partial j} (\epsilon J_{\varphi,j}) + \rho \cdot S_{\varphi} + K \cdot \Delta \varphi + \dot{m} \cdot \varphi$$

where $j$ denotes the spatial dimensions. This equation is derived via ensemble averaging procedures; one may consult Ishii (1975), Stewart and Wendroff (1981), Drew (1983) and Drew and Passman (1998) for details of the derivation. The first four terms represent accumulation, convection, diffusion and production or consumption (the source term $- S_{\varphi}$), and are identical to their single-phase counterparts apart for being multiplied with the local volumetric fraction ($\epsilon$) of the phase in question. The last two terms in the right-hand side of the equation represent interfacial transport (usually proportional to the difference in property $\varphi$ between the two phases, hence generally depicted as $k \Delta \varphi$) and transport due to thermodynamic change of phase.

It is important to distinguish between the two: in a gas-liquid flow, for instance, the former could be used to describe gas-liquid mass transfer due to absorption, while the latter would denote condensation or evaporation.

The two aforementioned terms establish the link between the two sets of equations describing each phase. The model introduces, though, new parameters in order to describe interfacial transfer. Interfacial momentum transfer is always present in the form of drag force, while mass transfer does not always take place. For every property that is transferred through the interface a closure law has to be included in the model. The determination of these closure laws is, as it will be shown, one of the most important issues when applying Eulerian-Eulerian models. Another issue that arises in this approach is the size of the control volume compared to the bubbles. If the bubbles are sufficiently smaller than the control volume, then the concept of interpenetrating continua and of each control volume containing a certain fraction of each phase is valid; if coalescence phenomena are significant, however, the larger bubbles formed may well grow bigger than the control volume. This issue also plays...
a role in the parametrisation of the stress tensor (Jakobsen et al., 1997).

The Eulerian-Eulerian type of model is less computationally intensive than the Eulerian-Lagrangian, and therefore the only one that can resolve the large-scale flow structures in an industrial bubble columns (Delnoij et al., 1997).

1.2.5.1 Interfacial forces

The main shortcoming of the Eulerian-Eulerian two-fluid model is the lack of universally accepted forms of the closure laws, i.e. expressions for the interfacial forces. Many features of the flow field depend on them, in particular the gas phase distribution. Considerable debate exists over which forces should be taken into account, and how they should be modelled. At present, the model has to rely on experimental evidence.

**Drag force.** The drag force is the dominant interfacial force appearing in the momentum equations. For this reason, most simulation studies in the literature neglect the other forces (virtual mass, lift etc.). In modelling it, the first issue to be resolved is whether bubble-bubble interactions should be taken into account. These interactions are not only through collisions but also due to interference of their flow fields. If the gas fraction is low, as is the case in the current study, then these interactions can be considered negligible (Jakobsen et al., 1997) and the equation for the drag force on isolated bubbles may be used. For a single bubble this equation takes the form:

\[
F_{D,j} = C_D \frac{\pi}{4} d^2 \frac{1}{2} \rho L (u_{G,j} - u_{L,j}) |u_{G,j} - u_{L,j}|
\] (1.15)

Thus, the drag force per unit volume of the bubble swarm will be given by:

\[
F_{D,j} = C_D \frac{3}{4d} \varepsilon G \rho L (u_{G,j} - u_{L,j}) |u_{G,j} - u_{L,j}|
\] (1.16)

The main problem in this approach is the estimation of the drag coefficient \(C_D\). Evidence suggests that it is a complex function of the Reynolds \((N_{Re})\), Eötvos \((Eo)\) and Morton \((M)\) numbers (Jakobsen et al., 1997). Furthermore, it is likely to be affected by the presence of other bubbles (Thakre and Joshi, 1999). Only in the case of creeping flow, when just the form drag is accountable, an analytical solution is possible, resulting in the well-known Stokes law. For other cases, a variety of
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correlations for calculating $C_D$ have been proposed, and some of them are presented on table 1.

An alternative approach is to assume that the bubbles have reached terminal velocity. In this case, the force balance requires that vertical forces (drag, weight and buoyancy) should sum up to zero.

$$F_{D,j} = W - B \Rightarrow C_D \frac{\pi}{4} d^2 \frac{1}{2} \rho_L (u_{G,j} - u_{L,j}) |u_{G,j} - u_{L,j}| = V (\rho_G - \rho_L) g \quad (1.17)$$

To allow for the fact that the slip velocity is bound to have a range rather than a unique value, the following correction can be applied, provided that an estimate of the average slip velocity is known:

$$F_{D,j} = \varepsilon_G (\rho_{G,j} - \rho_{LG,j}) g \frac{u_{G,j} - u_{L,j}}{u_{G,j} - u_{L,j}} \quad (1.18)$$

This approach has been adopted by Schwarz and Turner (1988), Joshi et al. (1998) and Sokolichin and Eigenberger (1994, 1999). Using a common slip velocity of 0.2 m/s, $\rho_L = 1000$ kg/l and $\rho_L >> \rho_G$ the equation reduces to a constant:

$$F_{D,j} = 4.9 \cdot 10^4 \varepsilon_G (u_{G,j} - u_{L,j}) \quad (1.19)$$

This gives a constant terminal velocity for all bubbles, which compares quite well with experimental data. Yet this approach is unrealistic, as the slip velocity is not constant and depends on the particular gas-liquid system (Thakre and Joshi, 1999). In some cases, though, it has been reported to give better solution than the more complex methods (Jakobsen et al., 1997).

Another notable contribution to the estimation of drag force is the drift flux model of Wallis (1969). This can be used to yield the terminal rise velocity and subsequently the drag force, making use of experimental measurements of the gas void fraction. Thakre and Joshi (1999) presented such a methodology, where the drift flux model was expanded to account for the radial profile of liquid velocity.

**Virtual or Added Mass force.** When a bubble is accelerated in the liquid, the surrounding liquid is accelerated as well. This results in an additional force, proportional to the relative accelerations of the phases:

$$F_{a,j} = -\varepsilon_g C_a \rho_l \frac{D}{D_t} (u_{g,j} - u_{l,j}) \quad (1.20)$$
Figure 1.1: Values of the drag coefficient computed by various correlations.

<table>
<thead>
<tr>
<th>$C_D$</th>
<th>Range</th>
<th>Named after:</th>
</tr>
</thead>
<tbody>
<tr>
<td>$24/N_{Re}$</td>
<td>$0 &lt; N_{Re} &lt; 0.2$</td>
<td>Stokes</td>
</tr>
<tr>
<td>$24/N_{Re}(1 + 0.15N_{Re}^{0.687})$</td>
<td>$0 &lt; N_{Re} &lt; 500 - 1000$</td>
<td>Schiller and Nauman</td>
</tr>
<tr>
<td>$24/N_{Re}(1 + 0.1N_{Re}^{0.75})$</td>
<td>$&gt;&gt;$</td>
<td>Ishii and Zuber</td>
</tr>
<tr>
<td>$24/N_{Re} + 5.48N_{Re}^{0.573}$ + 0.36</td>
<td>$&gt;&gt;$</td>
<td>Ihme et al.</td>
</tr>
<tr>
<td>0.44</td>
<td>$500 - 1000 &lt; N_{Re} &lt; 1 - 2 \times 10^5$</td>
<td>Newton</td>
</tr>
<tr>
<td>$2/3E_0^{1/2}$</td>
<td>$&gt;&gt;$, distorted bubbles</td>
<td>Ishii and Zuber</td>
</tr>
<tr>
<td>$4 \frac{dg}{3} \frac{\Delta \rho}{\rho} u_T^2$</td>
<td>$&gt;&gt;$, ellipsoidal bubbles</td>
<td>Grace</td>
</tr>
<tr>
<td>$\frac{2}{\rho g} M^{-0.149} (J - 0.857)$</td>
<td>spherical cap bubbles</td>
<td>Grace</td>
</tr>
<tr>
<td>$8/3$</td>
<td>$500 \leq N_{Re} &lt; 5000$, ellips.</td>
<td>Johansen and Boysan</td>
</tr>
<tr>
<td>$0.622/(1/E_0 + 0.235)$</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Table 1.1: Some correlations for calculation of the drag coefficient.
The coefficient $C_a$ has to be determined experimentally. According to Jakobsen et al. (1997), a value of 0.5 can be theoretically justified for rigid particles, but it is likely to decrease with increasing gas fractions. Sokolichin and Eigenberger (1994) report that 0.25 is a good value for bubbles while 0.5 is more suitable for solid particles.

**Lift or Magnus force.** This rotational force is exerted on rigid particles moving in a nonuniform flow field. It is given as the curl of the continuous phase velocity:

$$F_{i,j} = \varepsilon_g C_i (u_{g,j} - u_{i,j}) \times \left( \frac{\partial}{\partial x_j} \times u_{i,j} \right)$$

(1.21)

According to Svendsen (1992), this could account for the effect of bubbles accumulating in the centre in uniformly aerated bubble columns. The role of this force, though, was questioned by Sokolichin and Eigenberger (1994).

**Other forces.** Another force, not usually mentioned in most literature simulation studies, is the wall lubrication force. It exists in a region less than 5 bubble diameters from the wall, meaning that it can be seen only in sufficient fine grids. Considering, though, the fact that the control volume has to be substantially larger than the bubbles for the Eulerian-Eulerian model to be reliable, this force will probably be negligible. The Saffman force is another lateral force; Jakobsen et al. (1997) claimed that it is included in the formula for the Magnus force. Thakre and Joshi (1999) also followed this concept.

### 1.2.5.2 Bubble size distribution

Bubble size distribution has been studied by, among others, Kocamustafaogullari and Ishii (1983, 1995) who introduced the concept of "interfacial area concentration", and Bauer and Eigenberger (1999) who used a modified approach of "bubble number density". The later consists of a number balance for the total number of bubbles, accounting for its alteration by coalescence or redispersion phenomena. Once the number of bubbles has been determined, a mean diameter is applied to calculate the interfacial area. For a detailed estimation of the surface area, though, the most complete approach would be the population balance, which will be considered in detail later.

A population balance model needs to be supplied with functions describing bub-
ble coalescence and break-up kinetics, as well as with the initial distribution of the bubbles that have just emerged from the sparger. Though semi-empirical models of the former have appeared (e.g. Prince and Blanch, 1990; Luo and Svendsen, 1996) they are by no means definitive. Some researchers such as Tan and Harris (1986) and Geary and Rice (1991) have attempted to predict the bubble size as a function of the sparger geometry, but they consider only single holes and quiescent liquids. In a multi-hole sparger, the turbulent conditions arising from the bubble swarm would not allow for a single bubble size. In very high gas flowrates (heterogeneous flow) the bubble size may be dominated by inter-particle effects, but in homogeneous flow the initial distribution that arises from the sparger may prevail. Given the considerable uncertainty in our knowledge of both kinetics and initial conditions, the extra effort required for the solution of a population balance (which introduces a new independent variable associated with the bubble size, thus increasing the dimensionality of the problem) is probably not justified at present.

1.2.5.3 Turbulence

Turbulence is one of the major unresolved issues in fluid mechanics, and the state of the art in multi-phase flows is even more vague than in single-phase. Although it is generally accepted that the Navier-Stokes equations give an accurate account of fluid flow, even when turbulence is present, the small scale and high frequency of turbulent eddies mean that a numerical solution aiming to predict them would require extremely small spacial and temporal steps, which are out of reach of the current computational resources; moreover, turbulent flows are very sensitive to boundary conditions (a characteristic of chaotic phenomena), which can be neither determined with accuracy nor controlled. Therefore phenomenological models of turbulence have been developed, which attempt to predict probabilistic aspects of the flow. The following classes of models can be identified:

Direct Numerical Simulation (DNS): The direct solution of the Navier-Stokes equations, at very small scales and high frequencies, in order to resolve all turbulence scales. At present limited to low Reynolds numbers and simplified geometries, and unlikely to be of practical use in the near future.
Large Eddy Simulation (LES): A spatial filtering technique that resolves the larger eddies but uses models for the effect of the smaller (subgrid) scale turbulence.

Time-averaging techniques: Time-averaging results in equations that describe the mean properties of the flow, which is usually all that the engineer needs, but gives rise to a closure problem, as further terms are generated (products of fluctuations). Several closure approaches have been proposed, the most common ones being those based on the eddy-viscosity concept. The $k - \varepsilon$ model (Launder and Spalding, 1972) is currently the standard in industrial CFD - it employs two further differential equations to determine the effect of the unclosed terms on the mean properties. Other options are second-moment closures (Reynolds stress models) which employ equations for the unclosed terms themselves.

Probability density methods (PDF): These methods attempt to determine the entire distribution of the turbulent properties, usually at single points in space and time. They are especially advantageous for reactive flows.

These approaches have been applied mainly to single-phase flows. According to Jakobsen et al. (1997), large eddy simulation should lend itself better to extrapolation into multi-phase flows, as its concepts are closer to the mechanisms prevailing in multi-phase flow. Such an attempt has not been done yet, because it is too expensive to compute. Therefore, most studies on gas-liquid flows either assume laminar flow, or use one of the single-phase turbulence models for the continuous (liquid) phase, usually the $k-\varepsilon$, either intact or with slight modifications. Sokolichin and Eigenberger (1994) defended the laminar approach, on the basis that the $k-\varepsilon$ model cannot conceptually be extended to gas-liquid flow, and that it would dampen the transient character of the flow structure. In their results, the laminar simulation was shown to predict some vortices that were missed from a standard $k-\varepsilon$ implementation. Pfleger, Gomes et al. (1999) conducted a numerical/experimental study focusing on this point: they concluded that, though the $k-\varepsilon$ does dampen the chaotic behaviour exhibited by the laminar model, converging to a periodic oscillation instead, it represented better the experimental data on radial velocity. Two ways in
which the k-ε model can be modified to account for the particularities of gas-liquid flow are to include turbulence caused by bubble wakes (bubble-induced turbulence) and turbulent dispersion of the gas phase.

In general, large bubbles are expected to enhance turbulence, while small bubbles are thought to suppress it (Boisson and Malin, 1996). Bubble-induced turbulence was modelled by Sato (1981). It introduces a new term to the effective viscosity of the liquid phase, in addition to the contributions of the laminar and eddy viscosity terms:

\[ \mu_{eff} = \mu_{lam} + \mu_{t} + \mu_{B} \]  

(1.22)

Though it is generally accepted that bubble-induced turbulence plays a certain role, this approach to modelling it is still debatable. Pfleger, Gomes et al. (1999) did not include it in their simulations; Boisson and Malin (1996) and Svendsen et al. (1992) did.

Turbulent dispersion of the gas phase can have an effect on the radial velocity and fraction profiles. It is usually modelled via a transport equation for the gas fraction, including diffusion through eddy diffusivity. Again, this approach is by no means established. Pfleger, Gome et al. (1999) tested it in conjunction with the k-ε versus a plain k-ε model and found the compliance with experimental data to deteriorate.

1.3 The population balance

1.3.1 Origin and formulation of the PBE

During the last century, the population balance equation (PBE) has been established as a rigorous way of describing a variety of particulate systems such as cell populations, droplet and bubble swarms, granular solids, crystals, polymers, coagulating aerosols and even astrophysical systems. In all of these situations, the aim is to predict the evolution of the distribution of one or more properties that characterise the individual particulate entities, and the dynamic PBE is in essence a number balance equation describing this evolution (fig. 1.2).

The PBE originates in the work of Smoluchowski (1917), whose discrete equation for pure aggregation stemmed out of his study of Brownian motion. The equation
was later cast in a continuous form and extended to include the process of breakage. Thus, in the 1960's Hulbert and Katz (1964) and Randolph and Larson (1988) formulated the PBE for nucleation and growth phenomena. Subsequent milestones such as Drake’s comprehensive survey (1972), the analysis of Ramkrishna and Borwanker (1973) and Ramkrishna’s review (1985) did much to rationalise the PBE and establish it as a mainstream method for the analysis of particulate systems. In its most general form, the continuous PBE is a dynamic transport equation that describes the temporal evolution of population density as a result of four particulate mechanisms: nucleation, growth, aggregation and breakage, as well as transport due to the flow field:

\[
\frac{\partial n(x_i, t)}{\partial t} = -\frac{\partial (u_i \cdot n(x_i, t))}{\partial x_i} + D \frac{\partial^2 n(x_i, t)}{\partial x_i^2} + B_{\text{nuc}} + B_{\text{agg}} - D_{\text{agg}} + B_{\text{br}} - D_{\text{br}} \quad (1.23)
\]

Written thus, the PBE refers to a state space comprised of time, physical space, and an additional number of dimensions that represent the characteristic properties of the particles whose distribution we wish to calculate, often referred to as internal coordinates\(^2\). The gradient term in eq. 1.23, apart from input/output due to convective flow (differentiation in physical space), represents growth (differentiation in the particle size domain) or any other mechanism that operates on the population density in a convective way, as well as input/output due to convective flow. Similarly the diffusion term may represent both diffusion in physical space and eff-

\(^2\)Both spatial and internal coordinates in eq. 1.23 are denoted by \(x_i\).

Figure 1.2: Evolution of the population density.
effective diffusion of the population density, though the latter has usually no physical interpretation.

Unfortunately, the PBE yields analytical solutions for only a few special cases. Numerical solution of the PBE remains a considerable challenge, however, largely due to the integral form of the aggregation and breakage terms, and most approaches seem to have been tailored to specific applications (Rawlings et al., 1993). The aim of this work is to develop a numerical scheme that is capable of dealing effectively with an arbitrary particulate process, rather than being derived to match the needs of individual problems, while maintaining simplicity and speed. Particular emphasis will be put on stability and proper conservation of the moments.

1.3.2 Numerical solution of the PBE

Solutions to the discrete Smoluchowski equation are out of the scope of this paper; those interested may consult Drake’s review (1972). We will only mention that its exact solution can be found by direct numerical simulation of every individual particle - a computationally expensive approach, feasible only for relatively small populations. Monte Carlo simulations are another class of methods that will not be considered here; they are generally more efficient for multivariate problems, where solutions to the continuous version become too involved (Ramkrishna, 1985).

Analytical solutions of the continuous population balance have been derived only for a few idealised cases (see e.g. Scott, 1968, Drake, 1972, Ramabhadran et al., 1976, or Hostomsky, 1986); they are very useful, though, for benchmarking numerical solutions. One of the oldest and most widespread methods of solving it is the method of moments (Hulburt and Katz, 1964; Randolph and Larson, 1988), which is based on the transformation:

\[
L^k(f(n,v)) = \int_0^\infty v^k f(n(v),v)dv
\]  

(1.24)

By applying this transformation, it is sought to generate terms of the form:

\[
m_{<k>} = \int_0^\infty v^k n(v)dv
\]  

(1.25)

which is identified as the k-th moment of the distribution. The most important moments are the zeroth, which expresses the total number of particles, and the first,
which is proportional to the total mass of the system (when particle volume is used as the independent variable). Thus the PBE is transformed from a partial integro-differential equation in terms of population density into a differential equation in terms of the $k$-th moment. Although this method does provide the exact solution for the moments, it does not retrieve the entire distribution. In some special cases the distribution may be retrieved from the moments if its form is known ad hoc, but generally this process is unstable (Randolph and Larson, 1988). Moreover, not all formulations of the PBE can be reduced to moment equations, since some of the terms resulting from the transformation may not exhibit the form of eq. (1.24) - in fact, this is usually the case with formulations that include agglomeration and breakage. Thus the use of moments is limited to special cases, usually involving nucleation and growth only.

- Methods that retrieve the distribution by approximating the solution with a series of trial functions, whose coefficients are to be determined so that their sum will satisfy the PBE. When these functions are global, the scheme is a variation of the method of weighted residuals; in essence, the method of moments also belongs to this class (Ramkrishna, 1985). When the functions are local we have a variation of the finite element method.

- Methods that discretise the spectrum of the independent variable into a number of intervals and subsequently use the mean-value theorem to transform the continuous PBE into a series of equations in terms of either number of particles - termed the M-I approach - or average population density in each class - the M-II approach. These methods are often referred to as "Discretised Population Balances" or DPBs.

Weighted residual methods with global functions were among the first to be tried, and one may consult Ramkrishna (1985) for a review of early attempts. It was soon realised, however, that global functions cannot capture the features of an arbitrary-shaped distribution, especially if it exhibits sharp changes and discontinuities. If a-priori knowledge of the shape of the resulting distribution is available, the trial functions can be tailored to accommodate it; in that case, the method converges and may even be computationally attractive. If generality and flexibility are
required, however, the trial functions must clearly be local. Finite element methods approximate the solution with piecewise low order polynomials that are only locally non-zero, and are thus flexible and capable of capturing highly irregular solutions. The first notable attempt to solve the PBE with finite elements was carried out by Gelbard and Seinfeld (1978), who considered both orthogonal collocation and spline collocation for the solution of a dynamic PBE with nucleation, growth and aggregation.

Meanwhile, an alternative approach emerged: the discretisation of the domain of the independent variable. As pointed out by Gelbard and Seinfeld (1978) this approach, if straightforward, leads to severe errors unless a uniform grid is employed - unrealistic for anything but systems exhibiting a very small size range. For processes such as crystallisation or coagulation of aerosols, where large particles can be created very soon via aggregation, a non-uniform, preferably exponential grid must be employed. Bleck (1970) pointed out that the computational effort can be severely reduced if the grid assumes a geometric progression of \((2^j)\) - this way, the number of combinations of particles that must be considered to evaluate the aggregation terms is substantially reduced.

The next challenge for advocates of DPBs was to find a way of conserving the moments in such a grid. Batterham et al. (1981) modified this approach so that it conserves exactly the first moment with respect to volume (i.e. the total mass) by dividing particles that were created at non-nodal points into fragments whose volume would lie at the nearby nodal points. This artificial way of conserving mass, however, introduced errors in all the other moments, as well as in the predicted population density. The most established method in this series, proposed by Hounslow et al. (1988), is capable of conserving exactly both total mass and number of particles, as well as accommodating simultaneous growth and aggregation. All these methods are, however, constrained with respect to the choice of grid, as they are based on exploiting the properties of the geometric progression. Recent works by Litster et al. (1995) and Kumar and Ramkrishna (1996a,b,c) have succeeded in extending the concept of DPBs to finer grids; the latter also offers the advantage of conserving any two chosen moments. Also notable are the "method of classes" due to Marchal et al. (1988), which treats aggregation as a chemical reaction, and the work of Hill
After Gelbard and Seinfeld's pioneering work (1978), finite element and other function approximation methods seemed to attract fewer attention than DPBs; even the above authors made their own contribution to the DPBs (Gelbard et al., 1980). The main reasons for this trend are assumed to be the excessive, for that time, computational demands raised by these methods as well as the complexity of their implementation, in particular when applied to aggregation and breakage problems. Steemson and White (1988) used spline collocation to solve steady state problems of nucleation, growth and size dispersion; Eyre et al. (1988) also employed spline collocation, complemented with an adaptive grid. The recent advances in computational power, however, have brought finite element methods back to the forefront. Nicmanis and Hounslow (1998) solved the steady-state PBE with nucleation, growth, aggregation and breakage using collocation and Galerkin finite elements with cubic Lagrangian trial functions. The approach of Liu and Cameron (2001) to the dynamic PBE is also a function approximation method that uses wavelets to represent the solution, and focuses on accurate prediction of discontinuities. Recent reviews have been contributed by Rawlings et al. (1993), Costoglou and Karabelas (1994), Vanni (2000) and Lee (2001).

1.4 Crystallisation and precipitation

1.4.1 Overview

Crystallisation is one of the oldest, and yet most widely employed, unit operations. It is employed for the production of a large variety of crystalline products such as inorganic salts, chemicals and pharmaceuticals, as well as for separation and purification purposes. Comprehensive reviews of research in this field have been contributed by Nyvlt, Söhnle, Matuchova and Broul (1985), Randolph and Larson (1988), Söhnle and Garside (1992), Rawlings, Miller and Witkowski (1993), Mullin (2001), Mersmann et al. (2001), and Jones (2002).

According to Mullin (2001), the following crystallisation techniques can be distinguished:

- Solution crystallisation, the formation of a crystalline phase in a supersatu-
• Precipitation, fast crystallisation often driven by chemical reactions.

• Melt crystallisation, the separation of a multicomponent liquid mixture by cooling close to freezing point.

• Recrystallisation, the dissolution of an already crystallised material into fresh solvent and subsequent crystallisation by cooling, to remove soluble impurities.

• Sublimation, the production of a crystalline phase by direct condensation of a supersaturated vapour.

The driving force of crystallisation is supersaturation, which determines both its thermodynamics and kinetics. A solution can exist in one of three states: stable (no supersaturation) where crystallisation is impossible, metastable (low supersaturation) where crystallisation is unlikely to occur spontaneously but can easily be induced by a seed, and labile, where spontaneous crystallisation is probable but not inevitable (Mullin, 2001). Supersaturation also affects the rates at which crystallisation mechanisms proceed. A number of ways to express supersaturation have been proposed, such as the following: (Nyvlt et al., 1985)

\[
\Delta C = C - C^* \\
S = \frac{C}{C^*} \\
\sigma = \frac{\Delta C}{C^*} = S - 1
\]

Here \(C^*\) is the saturation concentration. A solution can become supersaturated in several ways, such as cooling to reduce the solubility, solvent evaporation to increase solute concentration, and addition of a co-solvent (salting out). Of particular interest in this work is reaction crystallisation, or precipitation, where supersaturation is generated by a chemical reaction between two soluble reactants that produces a sparingly soluble material. Precipitation is distinguished from other crystallisation techniques, in that its analysis requires a combination of crystallisation and reaction engineering principles, and a fair amount of literature is dedicated to it (e.g. David and Klein, in Mersmann, 2001; Söhnel and Garside, 1992). Precipitation is charac-
terised by high degree of supersaturation and fast kinetics that usually lead to the production of smaller crystals.

Apart from the usual unit operation objectives such as yield and efficiency, the analysis of crystallisation processes is also concerned with the prediction of the physical properties of the crystal product, such as its morphology and CSD. These properties are determined by the relative strength and interaction of several fundamental crystallisation mechanisms, such as nucleation, crystal growth and agglomeration. The ability to predict and control the physical properties of the product is a major advantage to the designer, since it grants the potential for product engineering. Thus the effort to establish kinetic models of crystallisation mechanisms has been at the core of crystallisation research since its very onset, yet definitive expressions still eludes the efforts of workers in the field. What follows is a brief review of the theories available to date, with emphasis on those that enable the formulation of kinetic models.

1.4.2 Nucleation

Though from a thermodynamic point of view the crystalline state is stable in a supersaturated solution, it is the nucleation kinetics that determine whether and at what rate crystals will be formed. Nucleation is a step of key importance to the outcome of the crystallisation process, as it can have a profound effect on both crystal morphology and CSD. However, it is also one of the least understood due to the difficulties encountered in isolating it experimentally from other phenomena, as well as the variety of fundamentally different mechanisms that are collectively called "nucleation" (Rawlings et al., 1993).

It is usually classified into: (Mullin, 2001)

- Primary: Formation of crystal nuclei directly from the dissolved species, driven by supersaturation. Further divided in:
  - Homogeneous: Spontaneous formation of nuclei.
  - Heterogeneous: Induced by foreign particles.
- Secondary: Generation of nuclei from existing crystals, e.g. by attrition.
Of these only homogeneous nucleation has been approached by way of rigorous physical and mathematical reasoning, in the form of the so-called classical nucleation theory. This are based on thermodynamic concepts, such as the free energy required for nuclei formation. The free energy is maximum for a certain size of nuclei, called critical size, and subsequently decreases with size. The nucleation rate is estimated from the rate at which arrangements of molecules ("embryos") exceed the critical size, and is expressed by an Arrhenius-type equation. Very high supersaturation is required to overcome the energy maximum and cause spontaneous nucleation. For this reason, homogeneous nucleation is prevalent only in precipitation, where very high supersaturation levels are common.

The rate laws resulting from classical nucleation theory have seldom been applied for scale-up of industrial processes because, apart from the assumptions inherent in them, homogeneous nucleation is not a common phenomenon in industrial processes. Carefully conducted experiments in very clean equipment are required to ensure that nucleation occurs solely via the homogeneous mechanism. On the other hand the energy barrier to be overcome is much lower in heterogeneous and secondary nucleation, and actual crystallisers contain lots of crystals and impurities. For this reason, they will not be discussed here; for a review of them one may consult, among others, Nielsen (1964) or Mullin (2001).

Foreign surfaces and particles act as a catalyst in the nucleation process. The supersaturation levels required for heterogeneous nucleation are much lower than those for the homogeneous mechanism, but still too high for it to dominate in processes other than precipitation. Furthermore, primary nucleation is very difficult to predict and scale-up, because the very high levels of supersaturation it requires, when generated, cause a burst of crystals that is very difficult to control (Rawlings et al., 1993). Since heterogeneous nucleation is not easy to formalise, empirical power-law functions of supersaturation are usually employed to express actual nucleation rates:

\[ B_0 = k_n S^n \]  

(1.29)

In most industrial crystallisation processes apart from precipitation, secondary nucleation is the prevailing source of nuclei (Rawlings et al., 1993). In contrast with primary nucleation, secondary nucleation is much easier to control. Comprehensive reviews of the underlying mechanisms have been contributed by Botsaris (1968),
Garside and Davey (1980), Larson (1984) and Mersmann, Sangl, Kind and Pohlish (1988). Secondary nucleation is also modelled with empirical power law rules, that associate its rate with the two main driving forces, supersaturation and the total crystal mass:

\[ B_0 = k_n S^{n_1} M_c^{n_2} \]  

(1.30)

Nucleation is affected by several environmental factors and hydrodynamic factors. Agitation generally enhances nucleation, and may cause solutions were otherwise stable to nucleate. However, very intense agitation has been found to reduce nucleation in some cases, possibly by disrupting the molecular clusters (Mullin, 2001). As a result, scale-up of nucleation rates is mostly empirical, based on experiments conducted in conditions as close as possible to the actual ones.

1.4.3 Crystal growth

Crystal growth is a broad field on its own, and comprehensive surveys may be found in Garside (1984) and Mullin (2001). From a kinetic point of view, crystal growth may be viewed as a two-step process: diffusion of molecules or ions to the crystal surface, and incorporation into the lattice, though the reality is likely to be more complicated (Mullin, 2001). Most of the chemical engineering literature treats the diffusion step with an empirical mass-transfer coefficient, while alternative approaches include treating the diffusion through a stagnant film, or applying some form of boundary layer theory (Rawlings et al., 1993).

The incorporation of new molecules into the lattice has been phenomenologically described by several theories, most of which can be found in Mullin (2001). Early theories were based on thermodynamic arguments, but predicted lower growth rates than those experimentally observed. The BCF theory due to Burton, Carberra and Frank (1951) overcame this problem by assuming that growth occurs at a screw dislocation. Birth and spread models (O’Hara and Reid, 1973; van der Eerden, Bennema and Cherepanova, 1978) postulates that growth develops from surface nucleation that can occur at edges, corners or on faces of a crystal.

For engineering purposes, however, the semiempirical power law has become the standard representation of the growth rate (Rawlings et al., 1993), with an
Arrhenius-type expression to denote temperature dependence:

\[ G_t = k_g e^{\frac{-E_g}{RT}} S^g \]  

(1.31)

where \( G_t \) is growth with respect to a linear dimension (such as diameter, or radius) and \( S \) is a measure of supersaturation. The fact that size does not enter the expression is commonly referred to as McCabe's \( \Delta L \) law; this is not always the case, however. Deviations from this are interpreted either as size-dependent growth or growth rate dispersion. Size dependent growth may be described by empirical functions such the following (Randolph and Larson, 1988):

\[ G = G_0(1 + \gamma t)^\alpha \]  

(1.32)

Growth rate dispersion is interpreted as fluctuations in growth rate due to reasons such as variations in crystal structure, hydrodynamic environment, local supersaturation etc. (Randolph and White, 1977) and is usually modelled with the aid of a diffusion-like term.

1.4.4 Agglomeration

The importance of agglomeration in crystallisation processes is becoming more and more evident, a fact reflected in the recent literature. Agglomeration can have a profound effect in the CSD, and may determine the physical properties of the end product, as well as its potential for separation; depending on the application, either its promotion or prevention may be desirable. Its mathematical analysis is faced with severe hindrances, however.

Terms such as "aggregation", "flocculation", "coagulation" and "agglomeration" have been applied interchangeably to describe the tendency of particles to cluster together (Mullin, 2001), a phenomenon that appears in many different fields such as the study of aerosols and colloidal systems. In crystallisation, however, the formation of such clusters involves two clearly distinct phenomena: the clustering of crystals due to collisions and interparticle attractive forces, and the cementation of the cluster into a single particle by way of solid bridges, due to crystal growth in its interior. Therefore a clear, formal distinction between the physical states of the clusters during the various stages of the process is required. The most clear formalism is
probably due to ASTM (1985), also used by Hartel et al. (1986), and it will be adopted in this study. According to it, three kinds of clusters must be distinguished:

- Flocculates: groups of particles held together by interparticle forces.
- Aggregates: Tighter-bound clusters, existing during the transition from flocculates to agglomerates.
- Agglomerates: Fully cemented clusters, unified by crystalline material grown in the interior.

Some crystal groups resulting from malgrowth phenomena (e.g. twins or dendrites) may have the same appearance as agglomerates under microscopic examination, and are called primary agglomerates (Mersmann, 2001). This is a fundamentally different phenomenon, however, and in this section we shall refer only to agglomeration that results from inter-particle clustering.

The first stage of agglomeration, i.e. the formation of flocculates due to collisions and interparticle attraction, is akin to similar phenomena occurring in colloids and aerosols. Thus the starting point for the formulation of a mathematical description of agglomeration are the theories developed in those fields, most importantly the pioneering work of Smoluchowski (1917; see Chandrasekhar, 1945, for an account of Smoluchowski's work in English). The second step, however, i.e. the transformation into an agglomerate due to inter-cluster growth, is unique to crystallisation processes and has a dominating influence. As a result, theories that are directly transposed from aerosol collection to crystallisation often fail to describe crystal agglomeration. On the other hand, the formulation of models that properly account for both steps faces severe technical obstacles due to the different nature of the phenomena involved in each stage, as well as the lack of appropriate experimental data. To gain insight into the different stages of an agglomeration process through experiments, let alone isolate them, is a formidable task.

Many factors influence an agglomeration process (see Mersmann, 2001, for a detailed list), but the most important ones can be grouped under the following:

- Hydrodynamics, which determine the motion of particles in a suspension, and therefore the rate and efficiency of collisions.
Literature review

- Interparticle forces, which govern the initial flocculation stage.
- Solute concentration (supersaturation), which induce growth of inter-cluster solid material, thus transforming the aggregate into an agglomerate.

A brief review of the knowledge available so far in each one of these fields follows.

1.4.4.1 Hydrodynamics

Any clustering process is second order, induced by collisions of particles:

\[ \frac{dN}{dt} = \beta N^2 \]  

(1.33)

Thus the first thing to consider is the frequency of collisions, which must be derived from hydrodynamic arguments. The first theoretical study of aggregation problems is probably due to Smoluchowski (1916). He considered the problem of perikinetic aggregation, i.e. the clustering of particles due to thermal (Brownian) motion, and approached it as a diffusion process. Assuming ideal spheres, he arrived at the discrete population balance equation:

\[ \frac{dn(v)}{dt} = \frac{1}{2} \sum_{v'=0}^{v'} k_v(v', v - v')n(v')n(v - v')dv' - n(v) \sum_{v'=0}^{\infty} k_v(v', v)n(v')dv' \]  

(1.34)

The kernel he derived for perikinetic aggregation (see Chandrasekhar, 1945, or Drake, 1972, for the derivation) is:

\[ k_v(v', v - v') = \frac{2kT}{3\mu} \left( \frac{(r(v') + r(v - v'))^2}{r(v')r(v - v')} \right) \]  

(1.35)

By using the approximation:

\[ \frac{(r(v') + r(v - v'))^2}{r(v')r(v - v')} \approx 4 \]

Smoluchowski was able to reduce eq. 1.35 to the size-independent kernel, and obtain an analytical solution for the infinite set of eqs. 1.34. This approximation is conceptually valid for aggregation of a system initially monodisperse, for a short amount of time. However, the size-independent has been used extensively in the crystallisation literature, due to its simplicity and the mutual-cancellation of other effects.
Perikinetic motion is prevalent only for very small particles (< 1μm). When collisions are induced by hydrodynamic forces, aggregation is termed orthokinetic. Smoluchowski (1917) was also the first to derive a hydrodynamic kernel for orthokinetic aggregation, considering the case of shear-induced collisions:

\[ k_\alpha (v', v - v') = \frac{4}{3} \gamma (v'^{1/3} + (v - v')^{1/3})^3 \] (1.36)

Orthokinetic is likely to be the dominant aggregation mechanism in crystallisation and precipitation processes. Two mechanistic kernels have been proposed to describe orthokinetic aggregation in a turbulent flow field. The first one (diffusion mechanism) assumes that the aggregation rate is determined by number of encounters of particles entrained by turbulent eddies (Drake, 1972):

\[ k_\alpha (v', v - v') = 0.31 \left( \frac{\xi}{\nu} \right)^{1/2} (v'^{1/3} + (v - v')^{1/3})^3 \] (1.37)

The diffusion mechanism is dominant for fine particles (< 10μm). For larger particles the inertial mechanism, has been proposed, according to which collision occur due to particles of different sizes moving at different velocities (Drake, 1972):

\[ k_\alpha (v', v - v') = 0.2 \frac{\rho c^{3/4}}{\rho_0 L^{5/4}} \left[ v'^{1/3} + (v - v')^{1/3} \right] \left[ (v'^{1/3} + (v - v')^{1/3})_2 + (v - v')^{1/3} \right] E_i \] (1.38)

where \( \rho \) is the density of the particles, \( \rho_0 \) the density of the gas and \( E_i \) an efficiency factor. These kernels are, however, more applicable to the coagulation of aerosols.

Aggregation can also be induced during gravitational settling, as particles of different sizes move at different velocities; the size dependency is similar to eq. 1.38. Several empirical kernels have also been proposed, such as the Thompson kernel; they may be found in Drake (1972) or Beckman and Farmer (1984).

The influence of hydrodynamics is likely to extend to other aspects of the agglomeration process than the collision frequency, however. It has been observed that agglomerates do not form during the induction time when crystals are very small (Mersmann, 2001), and do not exceed a maximum size. The former effect is accounted by most size-dependent kernels, but the latter seems to point to a disruption mechanism of the large aggregates, which is most likely hydrodynamic. Larger aggregates are subjected to higher stresses and are less likely to sustain them. So far only two hydrodynamic models of disruption have appeared, due to David et al.
(1991) and Mumtaz et al. (1997). The latter assumes a linear shear field, while the former employs turbulent quantities, but both rely on simplifying assumptions, as the real phenomenon is bound to be very complex.

1.4.4.2 Interparticle forces

The starting point of an agglomeration process is the formation of a flocculate, i.e. the clustering of two crystals during a collision due to interparticle forces. These arise from various interactions between both quantum-scale and macroscopic particles, but the latter are likely to dominate. One of the most established theories of interparticle forces in colloidal systems is the DLVO theory (Derjaguin and Landau, 1941; Verwey and Overbeek, 1948). The DLVO theory concentrates on two kinds of forces that dominate: Van der Waals and electrostatic. Several non-DLVO forces may also be accountable, such as solvation, structural and hydration forces (Mersmann, 2001).

Interparticle force theories are usually applied to study the aggregation of colloidal systems (e.g. Melis, Verduyn, Storti, Morbidelli and Baldyga, 1999), but have seldom been incorporated into crystal agglomeration rate models. The reason is that their role is confined to the initial stage of agglomeration, i.e. the formation of flocculates. The solid bridges that develop shortly afterwards give rise to lattice forces, which are much stronger, and it is probably those that dominate the overall agglomeration process (Bramley, Hounslow, Newman, Paterson and Pogessi, 1997).

1.4.4.3 Solute concentration

Completion of the second step in the agglomeration process, that is the cementation of the aggregate, requires crystal growth in certain spots in the interior of the cluster, called solid bridges or cementation sites. Growth requires the presence of supersaturation, and may be controlled by diffusion or surface integration. The experiments of Bramley et al. (1997) on precipitation of calcium oxalate showed that the agglomeration rate is better correlated with the limiting ion concentration, rather than the supersaturation. This implies that diffusion into the cementation site is the more likely mechanism, at least for similar systems. Mumtaz et al. (1997) proposed a 2-D diffusion-reaction model of crystal growth in the cementation site.
Collier et al. (1999) and Hounslow et al. (2001) have presented extensive experimental data on calcite and calcium oxalate that show aggregation and growth rates to be roughly proportional, all other things being equal. This supports the view that supersaturation must be reflected in the formulation of the aggregation kernel.

1.4.4.4 Agglomeration kinetics

Research in aggregation processes dates from Smoluchowski’s pioneering work (1917), and has produced valuable insights in fields as diverse as aerosols, colloids etc. The application of this knowledge in the agglomeration of crystals, however, is still at a primitive stage. Among the first studies were those of Hartel et al. (1988) and Hartel and Randolph (1988) who studied the agglomeration of calcium oxalate, a system of medical importance (formation of urinary stones). They employed an MSMPR crystalliser and a Couette-Taylor reactor in series, in an attempt to isolate the agglomeration from the nucleation and growth phenomena.

As mentioned before, agglomeration of crystals is a two-step process. The first step involves the collision of particles (crystals or aggregates) that results in the formation of a flocculate. This process is akin to the coagulation (collection) of aerosol and colloidal systems, and theories originating in these fields can be inferred. This step is essentially a second-order process:

\[
\frac{dN}{dt} = \beta N^2
\]  

(1.39)

where \( N \) is the total number of particles. The continuous population balance describing an aggregation process is:

\[
\frac{dn(v)}{dt} = \frac{1}{2} \int_{v'}=v \, k_a(v', v - v')n(v)n(v - v')dv' - n(v) \int_{v'=0}^{v'=\infty} k_a(v', v)n(v')dv' 
\]

(1.40)

The simplest way to express the kernel, \( \beta \), is as a size-dependent function that expresses the frequency of collisions, derived hydrodynamically or empirically:

\[
\beta_a = k_a \beta(v, v')
\]  

(1.41)

Apart from frequency, the formation of the initial clusters (flocculates) also depends on the efficiency of collisions. Efficiency is generally defined as the ratio of the
number of collisions that lead to the formation of a flocculate to the total number of collisions, and depends on interparticle forces and hydrodynamic factors. Thus the kernel is often described as:

$$\beta_a = k_a b_{eff} \beta(u, u')$$  \hspace{1cm} (1.42)

The second step is the growth of crystalline material between the clusters at spots called cementing sites. This step differentiates crystal agglomeration from other clustering processes, and few attempts at its mechanistic description have appeared so far. Most notable are the contributions of David et al. (1991) and Mumtaz et al. (1997).

David et al. (1991) associated this step with a "sticking probability" of the crystals (as opposed to the "encounter probability" term used by them to describe the first step), and proposed that it is related to the time required to build a solid bridge. Assuming growth of a cylindrical bridge between two spherical crystals, they arrived at an expression that shows the agglomeration kernel to be inversely proportional to the growth rate, as well as introducing a further size dependency:

$$t_{bridge} = \frac{d_s}{f_s G}$$  \hspace{1cm} (1.43)

where $d_s$ is the diameter of the smaller particle and $f_s$ is the shape factor. They also assumed that the growth of a bridge may be interrupted if the two participating particles are dissociated by the turbulence, and attempted an estimation of the dissociation probability on the basis of turbulence parameters. The latter indicated that the time scale during which turbulent dissociation may occur is much smaller than the time scale required to build a bridge, and the authors concluded that agglomeration requires rapid growth of the bridge.

The model of David et al. (1991) does not consider the growth of the bridge in detail. Mumtaz et al. (1997) modelled the deposition of ions into a cementation site by means of a 2-D diffusion-reaction model. The solid bridge was approximated by two cones, building up from either sides of the site and meeting at a singular point. The interruption of the cementation process was related to the tensile stress applied on the bridge due to a linear shear field for a certain collision orientation, and the yield stress of the crystalline material.
The cementation of an aggregate is a complex phenomenon, however, and both approaches rely on simplifying assumptions. Though a couple of spheres may be a relevant model when considering the aggregation of two individual crystals, observed agglomerates usually comprise many individual crystals, which means that most aggregation events occur between crystals and aggregates or between aggregates themselves. An aggregate can be envisaged as a network of particles, linked with bridges (cementation cites) that were formed at different times. Each cementation site has therefore different strength and probability of rupturing, depending on its age of formation. Though the tensile forces applied on the bridge linking an isolated couple of spheres may be prone to the description of Mumtaz et al. (1997), the stresses to which the interior bridges are being subjected are probably very different. In a large aggregate the stress distribution on its surface is bound to be uneven, and this would subject the interior bridges to high shear forces; it may indeed be the cause that prevents the formation of very large agglomerates. The approach of David et al. (1991) also assumes that the aggregate does not change in volume until the solid bridge is completed. Furthermore, the hydrodynamic stresses to which is an aggregate is subjected during its life time are likely to be more complex than the linear shear field of Mumtaz et al., or the turbulent dissociation principle of David et al..

In the works of Mumtaz et al. (1997) and of Hounslow et al. (2001), the concept of aggregation efficiency is extended to include the cementation step. In that context an effective collision is one that survives all obstacles until the formation of a completed agglomerate. Based on an exhaustive set of experimental data on stirred tanks, encompassing two systems (calcium carbonate and calcium oxalate) and a variety of different conditions, Hounslow et al. (2001) proposed a correlation for the overall efficiency of the agglomeration process. His correlation shows a linear dependency on growth and a strong hydrodynamic effect. Particle size does not enter directly, but only the mean size in the whole vessel has an influence. As a result, the predicted CSDs are similar in shape to those arising from the size independent model.

Inherent in these approaches (that employ the concept of efficiency for the whole process) is the notion that agglomeration is a single event that either succeeds or not
in creating a stable, cemented agglomerate. This notion, imported from the aerosol literature, is however inconsistent with one particular aspect of the agglomeration process. A cluster resulting from an initially successful collision may exist, for a considerable amount of time, in a state where its bridges are partly cemented to a different extent each. During that time it may participate in collisions, and thus affect the population balance by adding up to the population density at its size class, even though the bridge initially responsible for it may eventually rupture and split it into fragments - different from the initial ones. To take this into account one needs to consider agglomeration as a the overall result of two separate events: aggregation and disruption. Such a process would be described by the following population balance equation:

\[
\frac{dn(v)}{dt} = \frac{1}{2} \int_{v' = 0}^{v' = v} k_a(v', v - v') n(v') n(v - v') dv' - \int_{v' = 0}^{v' = \infty} k_a(v', v) n(v') dv' - n(v) \int_{v' = 0}^{v' = \infty} k_a(v', v) n(v') dv' + \int_0^\infty \rho(v, v') S(v') n(v', t) dv' - S(v)n(v)
\]

(1.44)

This approach introduces two further issues: determining the disruption kernel and the size distribution of the fragments resulting from a disruption event. Hartel and Randolph (1988), who adopted this approach, assumed that disruption is size-dependent and that each event results in the generation of two equal-sized particles. There seems to be no theoretical justification for the latter.

The aggregation-disruption approach was also explored by Wachi and Jones (1991; 1995). They assumed that disruption is carried out in an attritionary manner, i.e. every breakage event involves the removal of a single crystal. To attain this, they introduced the crystal number \(S\) (i.e. the number of crystals in an aggregate) in the population balance:

\[
\frac{dn(v, S)}{dt} = \frac{1}{2} \int_{S=1}^\infty \int_{v'=0}^{v'=v} k_a(v', S) n(v - v', S) dS dv' - \int_{S=1}^\infty \int_{v'=0}^{v'=\infty} k_a(v', S) dS dv' - n(v) \int_{S=1}^\infty \int_{v'=0}^{v'=\infty} k_a(v', S) dS dv' + \int_{S=1}^\infty \rho(v, v', S) S(v', S) n(v', S) dS' - S(v)n(v)
\]

(1.45)
The numerical analysis of this dual-coordinate population balance was studied by Wojcic and Jones (1998). Another advantage of this approach is that it allows for more accurate calculation of secondary particle characteristics (e.g. specific interfacial area), as the number of crystals within an aggregate is also being predicted. Zauner and Jones (2001) returned to the original aggregation-disruption population balance (eq. 1.44, but employed a parabolic daughter particle distribution (Hill and Ng, 1995; see also chap.3), also on the grounds of the attritionary mechanism.

However, there are several conceptual as well as practical problems involved in the aggregation-disruption approach. An aggregate contains bridges (cementation cites) of various ages and therefore strengths. It may be that the age of each bridge determines the probability of the aggregate being ruptured at that point, regardless of whether it links the aggregate to a single particle or not. For such an approach to be realised, tracking of individual particles in a Lagrangian manner would be necessary. The Wachi and Jones approach yields the number of individual crystals in an aggregate, but not their age since being entrapped in it. Moreover, it would be exceedingly difficult, if not impossible, to measure rates for aggregation and disruption in isolation, and hence obtain an estimate for the constants in eq. 1.44. Breakage rates obtained by subjecting pre-prepared agglomerates to stirring, such as those reported by Wojcic and Jones (1997), refer to the attrition of fully cemented agglomerates, rather than the weak aggregates that govern the crucial intermediate stage in the agglomeration process.

1.4.5 Mathematical modelling of crystallisation processes

The need to predict and control the physical properties of the product in industrial crystallisers, in particular the CSD, and to associate it with the design and operation parameters, meant that a rigorous mathematical framework for the description of crystallisation processes was required. Early researchers in the field, most notably Randolph (1964) and Randolph and Larson (1988), identified the population balance framework as ideally suited to this purpose.

As mentioned in section 1.3, the population balance is an integro-partial differential equation that describes the evolution of population density with respect to one or more independent variables that characterise a particulate system. In the
In section 1.3 it was stated that in the general case, the PBE cannot be solved analytically. To simplify the problem and obtain a tractable model, the following assumptions are possible:

- Steady state operation
- Size-independent growth
- No inter-particle phenomena (agglomeration, breakage)
- Spatial uniformity of concentrations and CSD
- Particles follow the streamlines of the flow

Several levels of approximation arise by adopting certain of these assumptions. Some of them result in analytical solutions that facilitate the analysis of experimental results but have limited applicability; others can be solved via complicated and time-consuming numerical methods; and others are simply intractable given the present computational resources. These levels will be outlined in the following discussion.

**1.4.5.1 MSMPR theory**

The first attempts at modelling crystallisation processes by means of population balance focused on the simplest and most common type crystalliser, the mixed-
suspension mixed-product removal (MSMPR). The MSMPR is the equivalent of a CSTR reactor, in that both the concentrations of chemical species and the product CSD are assumed to be spatially uniform. In its simplest form, MSMPR theory makes use of all of the assumptions stated above, resulting in a population balance of the form:

\[ G_t \frac{dn(L)}{dL} + \frac{n(L)}{\tau} = 0 \] (1.46)

In eq. 1.46 it is also assumed that the input to the reactor is clear liquor, i.e. free of particles. Nucleation in this case may enter as an initial condition, the population of the "embryo" crystals:

\[ L = L_0 \Rightarrow n(L) = n_{L_0} \] (1.47)

The analytical solution of eq. 1.46 is the well-known MSMPR equation (Randolph and Larson, 1988):

\[ n = n_{L_0} e^{-\frac{L}{L_0}} \] (1.48)

Thus the product of an MSMPR crystalliser exhibits an exponential CSD. Due to its simplicity and the availability of analytical solution, the MSMPR setup is often employed in experimental studies, in particular for the extraction of crystallisation kinetics. More complex crystallisation systems may exhibit size-dependent growth and inter-particle phenomena, in which case analytical solution is usually not possible. The absence of the time coordinate, however, still makes analysis of results and kinetics extraction much easier. Collier and Hounslow (1997) have proposed a method for extracting growth and agglomeration kinetics from MSMPR experiments; Hostomsky (1987) has derived a semi-analytical (based on Taylor expansion) solution for nucleation and size-independent agglomeration in an MSMPR.

### 1.4.5.2 Batch crystallisers

In the case of batch crystallisers time enters as an independent coordinate, and solution of a partial differential equation is necessary. Another complication is the fact that the solute species' concentrations, and hence supersaturation, are not constant either. At this point it must be recalled that crystallisation kinetics are functions of supersaturation, and vary with time too. This implies that the population balance must be coupled with the species' mass balances. The resulting equations in terms
of particle size are:

\[
\frac{\partial n(L,t)}{\partial t} = \frac{\partial (G_i(C_i)n(L,t))}{\partial L} \tag{1.49}
\]
\[
\frac{dC_i}{dt} = \sum_j r_j C_i \tag{1.50}
\]

where \( r_j \) denotes consumption due to either chemical reactions (in precipitation) or crystal growth. Eq. 1.49 must be complemented by two initial conditions for the population density and one for each species’ mass balance:

\[
t = 0 \Rightarrow n(L,t) = n_{t_0} \tag{1.51}
\]
\[
L = L_0 \Rightarrow n(L,t) = n_{L_0} \tag{1.52}
\]
\[
t = 0 \Rightarrow C_i = C_{i,t_0} \tag{1.53}
\]

Note that the terms \( r_j \) may refer to both chemical reactions (in precipitation) and consumption due to crystal mass deposition. Though the two-dimensional hyperbolic equation eq. 1.49 can be solved by the method of characteristics, the species’ mass balances may involve non-linear reaction rates and in general analytical solution is not possible.

The next option is the method of moments (Hulburt and Katz, 1964). The \( k \)-th moment of the population density with respect to particle size is defined as:

\[
m_k = \int_0^\infty L^k n(L,t) dL \tag{1.54}
\]

The zeroth, second and third moments with respect to size have a physical meaning: they are proportional to the total number, surface area and volume (mass) of crystals, respectively. The moment transformation, already defined by eq. 1.24, is written below in terms of particle size:

\[
L^k(f(n(L,t),L)) = \int_0^\infty L^k f(n(L,t),L,t) dL \tag{1.55}
\]

By applying this transformation to the population balance, it is hoped that every term in the equation can be arranged into a function of the moments of the population density. Unfortunately, this is only possible when no inter-particle phenomena are present and the growth function assumes specific forms. In the case
of size-independent growth, the following infinite set of ODEs results:

$$\frac{dm_k}{dt} = kGm_{k-1} + L_0 \cdot B_0$$

(1.56)

To obtain a closed set of equations the first four moments are required, (starting from the zeroth) as the third moment denotes the production of crystal mass, thus establishing the coupling with the species' mass balance. Eqs. 1.56, together with eqs. 1.52, thus form a closed set of ODEs that can be numerically integrated to yield the temporal evolution of the moments in a batch crystalliser. The mean crystal size can then be extracted from the moments, according to the equation (Randolph and Larson, 1988):

$$\bar{L}_{k_1,k_2} = (\frac{m_{k_1}}{m_{k_2}})^{1-k}$$

(1.57)

This general expression includes all possible definitions of mean size, such as the volume-average and area-average (Sauter-mean). In conclusion, it must be pointed out the method of moments cannot yield the CSD, but only averaged properties of it. Numerical methods are required to retrieve the full distribution.

### 1.4.5.3 Inter-particle phenomena

The description of inter-particle particulate phenomena, such as agglomeration and breakage, through a size-based population balance is very complicated. Hounslow, Ryall and Marshall (1988) presented a size-based population balance:

$$\frac{\partial n(L,t)}{\partial t} = -\frac{\partial(G_1(C_1)n(L,t))}{\partial L} + \frac{L^2}{2} \int_0^L \beta[(L^3 - \lambda^3)^{\frac{1}{3}}, \lambda][n[(L^3 - \lambda^3)^{\frac{1}{3}}, t]n(\lambda, t)]d\lambda - n(L, t)\int_0^\infty \beta(L, \lambda)n(\lambda, t) d\lambda$$

(1.58)

The above authors have also presented a numerical method for its solution. The method of moments is not applicable, as not all of the terms arising from the transformation can be arranged into moments.

Inter-particle phenomena conserve volume, hence the volume-based expressions for agglomeration shown in section 1.4.4.1 are much simpler. Analytical solutions for these models are more common in the aerosol literature (e.g. Scott, 1968; Ramabhadrnan et al., 1976; Drake, 1972) and are not very useful in crystallisation, because they usually assume that agglomeration is the only mechanism and require
an exponential initial distribution. An exception is the semi-analytical solution of Hostomsky (1987) for nucleation and size-independent agglomeration, which was developed to describe an MSMPR crystallization experiment. In chapter 3, a numerical method for solving volume-based population balance will be developed.

1.4.5.4 Spatially distributed crystallisers

Having considered time-dependency, and interparticle phenomena, the last assumption to drop is that of spatial uniformity. This is one of the most recent and least developed fields in crystallisation research. The main ways in which spatially distributed properties may affect the outcome of crystallisation are:

- Size classification. This implies that particles do not follow the streamlines of the flow, and as a result the CSD is not the same at all points.

- Non-uniform supersaturation field. This may be due to variations in concentration or temperature. As a result crystallisation kinetics, which are functions of supersaturation, exhibit spatial variations that have a profound effect on CSD.

- Hydrodynamic effects in interparticle phenomena. Agglomeration, which usually occurs via the orthokinetic mechanism, and breakage are heavily dependent on local hydrodynamics. A non-uniform flow field can severely affect the course of these processes.

Size classification has been considered by early researchers (e.g. Randolph and Larson, 1988), but only in an empirical manner. The most common example is classified product removal, i.e. the removal of particles above or below a certain size. This may either arise naturally or be enforced (e.g. by the installation of a screening device in the outflow), and can be easily taken into account by introducing a classification factor in the model, which must be determined experimentally. Variations of CSD within the reactor have seldom been considered, however, and particles are usually assumed to follow the streamlines of the flow.

Dropping the last assumption would imply considering the particles as a separate phase, and coupling the population balance with the Navier-Stokes equations. If
size classification is also to be taken into account rigorously, a separate momentum balance must be solved for each particle size, resulting in a virtually intractable computational problem. In systems such as precipitation where the particles are very small (< 10μm), though, the interaction between particles and flow can be neglected. Since in this work we are dealing with such a system, no further consideration of this issue will be attempted.

Spatial variations in supersaturation can be very important in precipitation. In case of very fast reactions, "hot spots" of high supersaturation values may be generated around points of reactant input. Growth and nucleation kinetics are very sensitive to supersaturation, with high values of it usually favouring nucleation. This results in more crystals being produced, whose growth causes a rapid depletion of supersaturation, culminating in a CSD shifted towards the lower end of the size spectrum.

Pohorecki and Baldyga (1983, 88) were among the first to proposed models for taking into account the spatial distribution of concentration. Tavare (1986) presented a comprehensive review of mixing models with relevance to precipitation. Generally, such models fall in either of two categories:

- Integration of population balance with the Navier-Stokes equations.
- Solution of the population balance within a compartmental model, custom-derived for a specific equipment, based on hydrodynamic considerations and possibly CFD.

Models of the first kind are usually based on the moments transformation, in order to minimise the number of extra differential equations that need to be solved alongside the Navier-Stokes equations; such are the models of Wei and Garside (1997) and van Leeuwen, Bruinsma and Van Rosmalen (1996). Recently Paschedag, Tornatora, Ausner and Signorino (2002) et al. have coupled the CFD with a discretised population balance comprising 40 classes. In these works the coupling between CFD and population balance is usually one-way, i.e. the Navier-Stokes equations are solved for each new time step, and the resulting concentrations are employed for propagating the PBE. Excessive computation time is the main problem faced by this approach.
Models of the second kind sometimes employ CFD for calculating average hydrodynamic properties of compartments (Zauner and Jones, 2000). The compartmental model is custom-derived for a specific kind of equipment, usually a stirred tank. In this work, this approach will benefit from a generalised compartmental model, that can be tuned to an arbitrary geometry.
Chapter 2

Hybrid CFD - reaction engineering approach

2.1 Introduction

2.1.1 Present challenges in multiphase reactor modelling

The performance of gas-liquid reactors relies on the combined outcome of multiphase fluid dynamics, interfacial mass transfer and chemical reactions. The detailed, simultaneous analysis of the above phenomena is out of reach given the currently available computing power, partly because they take place at different scales. Furthermore most of these phenomena are non-linear, and empirical models cannot be extrapolated too far from the set of design parameters and operating conditions under which they were extracted. As a result, multiphase reaction engineering is still confronted with major challenges from a modelling point of view.

The quantitative description of mixing is a major issue of reaction engineering. Early research attempted to quantify mixing without knowledge of the flow field, and a major step in this direction was taken when Danckwerts (1953,58) introduced the concept of Residence Time Distribution (RTD). The RTD arising from tracer experiments can be used to identify situations that lie between the two extremes of plug flow and complete mixing, and a variety of compartmental models, such as the tank-in-series and the cell model with backflow (or its continuous equivalent, the axial dispersion model) have been proposed to accommodate them (Mecklenburg
and Hartland, 1975; Nauman and Buffham, 1983). Also notable is the "network-of-zones" approach of Mann and Hackett (1988), a description of the mixing hydrodynamics that offers a computationally inexpensive alternative to CFD.

More fundamental information about mixing can be obtained through the Navier-Stokes equations which, though, require an excessive amount of computation to resolve all turbulence scales by direct numerical simulation (DNS). Shortcuts with variable levels of detail have been developed, ranging from the commonly used in engineering turbulent viscosity-based approaches such as the $k-\varepsilon$ (Launder and Spalding, 1972) to the much more accurate Large Eddy Simulation (LES, see e.g. Pope, 2000), which, though sparingly used today, may become the industry standard in the future. The coupling of fluid dynamics with complex and often non-linear chemical reaction systems, however, results in a steep increase of computational time, that often prohibits the use if this approach for reactor scale-up purposes.

In the last decades, the reaction engineering community has started to explore the prospect of employing CFD in the modelling of multiphase reactors. Torvik and Svendsen (1990) made one of the first attempts to model mixing and reaction in a bubble column with CFD, based on the $k-\varepsilon$ model, while mass transfer and chemical reaction were still dealt with by assuming constant reaction enhancement. Ranade (1995), Delnoij, Kuipers and van Swaaij (1997), Van den Akker (1997), Kuipers and van Swaaij (1997), and Dudukovic, Larachi and Mills (1999) among others, have reviewed the current state of the art regarding the application of CFD to chemical reaction engineering. The reviews of Jakobsen, Sannaes, Grevskott and Svendsen (1997) and Joshi (2001) focus on CFD modelling of gas-liquid flows, with relevance to bubble column reactors.

The direct application of CFD to chemical processes faces several obstacles, however. Even in single phase reactors, chemical reactions are described by highly non-linear terms that often cause numerical instabilities. More complex phenomena, such as interfacial mass transfer with chemical reaction, are encountered in multiphase or catalytic reactors, while in cases such as precipitation and biochemical engineering additional equations, such as the population balance, must be solved alongside the Navier-Stokes equations. The computational resources required are often prohibitive, while the complexity of the problems that arise from the coupling
of the fluid dynamics with the chemical phenomena means that the systems must be treated case-by-case.

2.1.2 The need for hybrid approach

Recently, "hybrid" approaches have emerged as an alternative. In those CFD is employed only for hydrodynamic simulation, while the chemical phenomena are resolved in a custom-built compartmental model. Although this decoupling cannot be applied to cases where the coupling between hydrodynamics and chemistry is very strong, such as in combustion, many chemical reactors with a liquid bulk phase fall into this category. Bauer and Eigenberger (1999) used a "zone model" to study a bubble column reactor; Bezzo, Macchietto and Pantelides (2000) developed an interface of communication between the gPROMS modelling software and a commercial CFD code; Zauner and Jones (2000) used a segregated feed model in conjunction with CFD to study precipitation in a stirred tank. Though these approaches represent essentially a compromise, their lifetime is likely to be quite long, as the full CFD solution would require a radical breakthrough in computer technology.

The approach proposed here belongs to this class. In its context, the detailed hydrodynamic information on meso- and macro- mixing that CFD provides is mapped into a more tractable compartmental model, suitable for efficient calculation of chemical reactions and other nonlinear phenomena (e.g. precipitation), while still incorporating the essential features of the geometry and operating conditions of the reactor. The main concept of the approach, the mapping of the CFD grid into the generalised compartmental model, is introduced in the next section. This is followed by a procedure for linking it to $k - \epsilon$ - based CFD, and an application to bubble column reactors by superimposing a phenomenological model of mass transfer with chemical reaction. The basic methodology is not limited to these, however, and may be used in conjunction with more advanced turbulence models, or other phenomenological reaction engineering models.
2.2 The generalised compartmental model

2.2.1 Basic concepts

An outline of the approach can be provided as follows: first, the CFD simulations are carried out on a grid that incorporates the detailed geometry of the reactor. Subsequently, a compartmental model is derived via mapping the CFD grid into a coarser network of compartments. The compartmental model, being much more computationally tractable than the CFD, is finally used for the reaction engineering calculations, by superposing on it the chemical reactions, population balance and any other complex models that may be desired. The implicit assumption is that the coupling between the hydrodynamics and the chemical reactions is one-way, i.e. the reactions have no significant feedback on the hydrodynamics. In gas-liquid reactors this is usually the case, as long as the reactions are not highly exothermal and the reaction product does not affect the mixture rheology (e.g. polymerisation, dense solid phase precipitation).

The mapping concept is best illustrated through the example of fig. 2.1, where five subdomains of the CFD grid are mapped into a compartmental model that retains only the fluxes between the original subdomains. The compartmentalisation must be based on physical reasoning, and will be discussed in more detail later; at the moment it will be pointed out that in fig. 2.1 compartment 5, entirely enclosed within 4, may be devised to contain a vortex that results in dead space. Another point that will become clear later is that the dimensionality of the CFD grid is irrelevant, i.e. a 3-D grid may also be mapped in the same way.

It is important to stress the difference in the physical interpretation of a compartment between the RTD models and the present approach. Whereas in RTD models a compartment arises from the fitting of an RTD curve but does not necessarily correspond to a physical part of the reactor, in the present approach the compartments are generated from the CFD grid, and are thus tightly bound to the geometry of the particular piece of equipment.

Conventional compartmental models rely on RTD experiments for the determination of their parameters. Rather than that, in this approach the parameters of the compartmental model are determined from CFD. Therefore the approach is fully
Figure 2.1: Mapping of a CFD grid into a compartmental network.
predictive and does not require experimental input at any of its stages. Experimental validation is required only with respect to the CFD model, and this is an issue that will not concern us here. The methodology that will be proposed for connecting the compartmental model to the CFD does not restrict the approach to a particular kind of CFD model; therefore, if more accurate models become available in the future, they may be used instead to obtain more accurate results.

The link between the CFD and the compartmental model is provided in the form of the virtual tracer experiment. This is a simulation of the dispersion of a passive tracer carried out with both the CFD and the compartmental model, in a manner that indicates to which extent the latter captures the mixing effects of the actual reactor, as simulated by the former. Once the compartmental model is able to carry out the virtual tracer experiment and yield approximately the same results as the CFD, the designer can be assured that it describes adequately the mixing behaviour of the reactor and proceed to introduce into it the chemical reactions and any other complex phenomena whose direct coupling with CFD would be impracticable.

To automate the mapping process we introduce the generalised compartmental model, a network of fully-mixed compartments with many degrees of freedom, that may be reduced into a specific compartmental model when its parameters are specified (via the CFD). The parameters are the number of compartments, the way they are connected and the values of the flowrates. The exact shape of the compartments and their boundaries is of no essence, since they are spatially uniform; instead, all that matters is the way they are connected.

**2.2.2 Topology of the compartmental model**

Let $K$ be the number of compartments in the network. Each compartment will be identified by the compartment index, $k$:

$$ k \in \{1, 2, ..., K\} $$  \hspace{1cm} (2.1)

Each compartment is allowed to have boundaries with a maximum of $L$ neighbours, the indices of which are established with the mapping:

$$ I : (k, l) \rightarrow I(k, l) $$  \hspace{1cm} (2.2)
With each compartment \( k \) we associate a maximum of \( L \) connectivity flowrates, denoting flow towards its \( l-th \) neighbour. Consequently, a pair of opposite direction flowrates is associated with each boundary. This two-way connectivity provides a means of introducing local turbulent exchange.

The inverse map of \( I(k, l) \) yields the indices of the \( L \) (at most) compartments that have an outflow towards \( k \):

\[
I' : (k, l) \rightarrow I'(k, l) \tag{2.3}
\]

The one-dimensional array of compartment indices, together with the maps \( I(k, l) \) and \( I'(k, l) \) establish the topology of the network, i.e. the way the cells are connected. Only \( I(k, l) \) is required to specify the network; \( I'(k, l) \) may be easily determined by the program.

The dimensionality of the CFD grid space is not retained in the compartmental model, which contains information only about how the compartments are linked. For example, it makes no difference whether the neighbours of a compartment are in the \( x \), \( y \) or \( z \) direction, since all of them would be mapped in the same way. An important consequence of this is that CFD grids of any dimensionality may be mapped. In the example of fig. 2.1 the mapping of a 2-D grid was illustrated, but a 3-D one would be mapped in exactly the same way. The one-dimensional compartment index \( k \) is arbitrary, while all topological information is contained in \( I(k, l) \) and \( I'(k, l) \).

To describe multi-phase reactors, separate compartments are employed for each phase. In this work the structure of the network will be the same for all phases, to allow the implementation of mass exchange between the phases via the penetration model. Thus the building block of a two-phase compartmental network comprises the two compartments of each phase, in contact with up to \( L \) other compartments (fig. 2.2). The following notation is introduced:

\( C_{ikp} \) Concentrations of chemical species \( i \), spatially averaged over compartment \( k \) at phase \( p \).

\( V_{kp} \) Volume of compartment \( k \) at phase \( p \).

\( Q_{kp}^{in} \) Inflow into the system (i.e. the whole reactor) at compartment \( k \), at phase \( p \).
Figure 2.2: The building block of the multi-phase generalised compartmental model (two-phase).

$Q_{kp}^{\text{out}}$ Outflow from the system at compartment $k$, at phase $p$.

$Q_{kp}^{(k,l)}$ Flow from compartment $k$, at phase $p$, towards its $l$-th neighbour, the compartment $I(k,l)$.

$Q_{I'(k,l),p}^{k}$ Flow into compartment $k$, at phase $p$, from its $l$-th neighbour, the compartment $I'(k,l)$.

The total number of parameters thus are:

- Number of cells (1)
- Volumes of cells ($K$)
- Connectivity flowrates, maximum $[K \times L]$

The total number of parameters that may be arbitrarily specified (i.e. degrees of freedom) is however constrained by the mass balances, as it will be shown in the next section.
2.2.3 Constraints

All of the parameters mentioned above may not be arbitrary specified. To begin with, the number of cells is inferred from the specification of the cell volumes. Moreover, the cell volumes in a phase must sum up to the total volume of the system in that phase:

\[ V_p = \sum_{k=1}^{K} V_{kp} \]  

(2.4)

For every cell in a phase the mass balance must be satisfied, resulting in \( K \) further constraints:

\[ \sum_{l=1}^{L} Q_{kp}^{l(k,l)} + Q_{kp}^{\text{out}} = \sum_{l=1}^{L} Q_{kp}^{k(l,k,l)p} + Q_{kp}^{\text{in}} \]  

(2.5)

2.2.4 Mass balances

A compartment \( k \) has flowrates to and from \( L \) neighbouring cells, and there may also be input or output from the whole system at a location in the vicinity of \( k \) (fig. 2.2). The dynamic mass balance for the concentration of a chemical species \( i \) can be written as:

\[
\frac{dC_{ikp}(t)}{dt} = \frac{1}{V_{kp}} \left[ -Q_{kp}^{\text{out}} \cdot C_{ikp}(t) - \sum_{l=1}^{L} Q_{kp}^{l(k,l)} \cdot C_{ikp}(t) + Q_{kp}^{\text{in}} \cdot C_{ikp}(t) + \sum_{l=1}^{L} Q_{kp}^{l(k,l)p} \cdot C_{i(l,k,l)p}(t) + f(C_{i}(t)) \right]
\]  

(2.6)

were \( f \) is the reaction term (\( j \) is the reaction index):

\[ f(C_{i}(t)) = \sum_{j=1}^{J} r_{j}(C_{ikp}(t)) \]

This balance may be written more concisely as:

\[
\frac{dC_{ikp}(t)}{dt} = W_{\text{out}} \cdot C_{ikp}(t) + W_{\text{in}} + f(C_{i})
\]  

(2.7)

where:

\[ W_{\text{out}} = -\frac{1}{V_{kp}} \left( \sum_{l=1}^{L} Q_{kp}^{l(k,l)} + Q_{kp}^{\text{out}} \right) \]

\[ W_{\text{in}} = +\frac{1}{V_{kp}} \sum_{l=1}^{L} Q_{kp}^{l(k,l)p} \cdot C_{i(l,k,l)p}(t) + \frac{1}{V_{kp}} \cdot C_{ikp}(t) \cdot Q_{kp}^{\text{in}} \]
Note that, in the case of the gas phase, the reaction term may include the consumption due to chemical absorption. As this is a system of ordinary differential equations, initial conditions must be specified for its solution:

\[ t = 0 \rightarrow C_{ikp} = C_{ikp}^0 \]  \quad (2.8)

### 2.2.5 Link with CFD

We shall now describe how the generalised compartmental model gets linked to the geometry and operating conditions of a specific reactor via a CFD simulation. As it has been mentioned, the topology of the compartmental model is specified through a mapping of the CFD grid cells onto the compartments. Apart from the topology, we need to determine the flowrates between the compartments.

In this work we shall distinguish between mixing due to the mean (time-averaged) flow field, and turbulent local mixing i.e. any localised mixing effects, caused by turbulence at scales higher than the Kolmogorov. To describe the former we shall use the mean flow information that is yielded by a CFD code supplied with a turbulence model based on time averaging, such as the $k - \epsilon$.

To account for the latter we shall employ the virtual tracer experiment, i.e. compute the dispersion of a passive scalar using the above CFD code. The flow and scalar concentration fields that result from CFD will subsequently be used to derive a compartmental model, on which the chemical reactions can be superimposed provided that they have no significant feedback on the hydrodynamics.

#### 2.2.5.1 Overall tuning algorithm

The overall process of constructing and fine-tuning the generalised compartmental network can be summarised by the following algorithm:

**Step 1** Carry out the flow field computation and virtual tracer experiment with CFD.

**Step 2** Determine the network structure and estimate connectivity flowrates based on mean flow.

**Step 3** Introduce two-way flowrates to account for local turbulent exchange.
Step 4 Evaluate the convergence between the compartmental model and the CFD tracer experiment. If poor, return to step 3. If not possible at all, return to step 2 and modify the structure.

Step 5 Having established a network that adequately describes the dispersion, superimpose the reaction engineering models to account for complex reaction systems, gas-liquid mass transfer, micromixing etc.

2.2.5.2 Determination of network structure and flowrates

At this point, the mapping of the CFD grid into the compartmental model can be made on the basis of a qualitative inspection of the flow and concentration fields yielded by CFD, based on the designer's intuition. Future work could consider automating the process through a quantitative algorithm. A few heuristic rules such as the following could be used to guide the decision:

- The flow should not change direction along a boundary.
- Streamlines should cross similar distances within a certain compartment.
- Regions of uniform concentration may be accommodated by a single compartment.

The implication of the first rule is that the all main flow features giving rise to macromixing, such as circulation loops, must be accounted for by the initial structure (apart from small vortices, which may be fully contained within a single compartment). The second rule suggests that all fluid elements entering a compartment will have similar residence times at that compartment. The third rule suggests that turbulent exchange does not have to be accounted for by the inter-compartmental flowrates alone, but may be inherent in the network itself.

The flowrate at any compartmental boundary can be calculated from the CFD results by summing up the flows through the CFD cell faces that comprise that boundary. From the equation of continuity, the total flow along any finite surface (such as a compartment) is:

$$ J_j = \oint \rho u_j dA_j $$

(2.9)
Thus the flowrates between compartments can be calculated from the CFD results by summing up the flows through the CFD cell faces that comprise that boundary. Such information is readily available in most CFD codes, as the term $p u_j d A_j$ must be calculated to form the convective part of the coefficient matrix. The network thus generated features one-way connectivity; the two-way is reserved for the introduction of turbulent mixing. This will be carried out by adjusting the two-way flowrates until the virtual tracer experiment yields the same results with both the CFD and compartmental model. In the present work, this adjustment was carried out by trial and error; future work may consider more systematic methods of attaining it.

2.2.5.3 The virtual tracer experiment

As we saw in the preceding paragraph, the main flow and circulation patterns in the reactor are accounted for in the compartmental model by averaging the mean flow information yielded by the CFD, to yield the connectivity flowrates in the former. The network thus generated, however, would account only for the mixing due to the large-scale features (such as the circulation between riser and downcomer in a loop reactor), while leaving out the local mixing effects. A tubular reactor, for instance, would be reduced to a tank-in-series rather than an axial dispersion equivalent model. To incorporate local mixing into the compartmental model, we shall employ the virtual tracer experiment.

The main idea behind the concept of the virtual tracer experiment is that both the CFD and compartmental model must be able to simulate the dispersion of a passive scalar (tracer) and yield similar results. This will signify that the compartmental model captures the turbulent mixing behaviour of the actual reactor (up to the level of accuracy exhibited by the CFD model) and may be safely employed for the reaction engineering analysis. Several modifications of the two-way flowrates between compartments may be carried out so that the compartmental model converges with the CFD with respect to the simulation of the virtual tracer experiment; if that is not at all possible, it signifies a fundamental inadequacy of the compartmental structure, and that the network should be constructed anew.

The virtual tracer experiment must be designed in such a way that it is indicative
of the turbulent mixing during the actual reactor operation. In particular, the tracer must be introduced in a way that emulates the feeding policy of the reactor. In the case of a continuous flow reactor, where the reactant is fed through an inlet, this can be straightforwardly achieved via a stepwise introduction of the tracer at the inlet point. For a semi-batch bubble column, however, this goal is not so straightforward to attain, as the reactant is fed along the bulk at a variable rate, depending on the local presence of bubbles. The best way to emulate this feeding policy seems to be the stepwise introduction of the tracer into every cell compartment of the CFD compartmental model, as a step change at a value proportional to the local gas-liquid interfacial area. This can be computed on the basis of the local volume fraction and mean bubble diameter \( d_b \):

\[
\dot{S} \sim \frac{e_g \cdot V_{cell}}{(\pi d_b^3/6) \cdot V_{cell}} \cdot \pi d_b^2 \frac{6e_g}{d_b}
\]

(2.10)

An advantage of this approach is that the structure of the compartmental model will be maintained as simple as possible, by accounting only for the most important mixing events that determine reactor performance during actual operation. To illustrate this point, consider the following example. If a tracer is introduced at a single point in a bubble column, steep gradients are likely to be generated around that point. In order to capture this effect with the compartmental model, several small compartments will be required to accommodate that region. These gradients will not arise during the actual operation, however, because the introduction of the reactant is not that localised; in fact, it takes place throughout the whole of the riser section, resulting in a much more smooth concentration distribution. By introducing the tracer in a way that emulates the actual feeding policy, we ensure that the compartmentalisation process will result in the smallest number of compartments necessary to describe adequately the mixing effects that are of importance during actual reactor operation, and nothing more than that.

In the experimental RTD approach a passive tracer is introduced in an impulse, step or periodic manner, and the RTD is extracted from concentration measurements at the output. The shape of the RTD curve is indicative of a certain behaviour (e.g. plug flow, dead space), and a compartmental model is subsequently devised to emulate this behaviour. It is important to note, however, that there is no one-to-one correspondence between the detailed reactor design and RTD; in principle,
an infinity of reactors could yield the same RTD. The virtual tracer experiment, on the other hand, yields the detailed spatial distribution of the tracer. Since the CFD contains the detailed geometry and operating conditions of the reactor, it follows that the result of the virtual tracer experiment and hence of the compartmental model derived on its basis uniquely corresponds to the specific reactor under consideration.

Finally let us compare the present approach to the earlier hybrid models. These relied on predictions of turbulent quantities via CFD, such as turbulent kinetic energy and dissipation rate, which were used to estimate the flowrates between compartments through phenomenological equations. As Montante, Lee, Brucato and Yianneskis (2001) have shown, these quantities are severely underestimated by $k-\varepsilon$ based simulations. On the other hand, this methodology is based on mean flow information and the virtual tracer experiment, and thus avoids the shortcomings of the $k-\varepsilon$ CFD simulations.

2.2.5.4 Extraction of mixing information from $k-\varepsilon$ based CFD

In principle, the results from any CFD simulation may be averaged to parametrise the generalised compartmental model. Multiphase flow models were reviewed in chapter 1.2.2. Among the main kinds of models that can be employed (volume tracking, Eulerian-Lagrangian, Eulerian-Eulerian) it was concluded that the Eulerian-Eulerian approach is best suited for resolving the main flow characteristics in large scale equipment within a reasonable computational time. With respect to turbulence modelling, for engineering-scale gas-liquid flow computations so far only the $k-\varepsilon$ model has been applied and tested, though it is unlikely to be the most suitable (Jakobsen et al., 1997). For this reason, it has been employed for the case study presented here to illustrate the approach. The fact that the methodology is not linked to a specific CFD model, however means that it will benefit from any future developments in the field.

The $k-\varepsilon$ model (Launder and Spalding, 1972) belongs to a family of models that are derived through Reynolds averaging, a temporal filter that decomposes the hydrodynamic variables into a mean and a fluctuation:

$$\phi = \bar{\phi} + \phi'$$

(2.11)
Reynolds averaging generates several terms of the kind $\overline{\phi_1 \phi_2}$, i.e. mean values of products of fluctuations of different variables, resulting in a closure problem. The approach of the $k-\varepsilon$ model is to model the effect of these terms on the mean flow field by two extra transport equations, one for the turbulent kinetic energy ($k$) and one for the energy dissipation rate ($\varepsilon$). These equations and their derivation may be found in Launder and Spalding (1972). The mean flow field that results encompasses the gross features of the flow such as recirculation zones, large-scale vortices giving rise to dead space, bypassing routes etc, that give rise to macromixing. Further, localised mixing effects are revealed when one computes the field of a scalar quantity.

In order to establish a correspondence between the hydrodynamic information yielded by the $k-\varepsilon$ and the mixing concepts used in reaction engineering, we shall distinguish between mixing due to the mean (time-averaged) flow field, and local turbulent exchange i.e. any localised mixing effects, lateral or axial, caused by turbulence at scales higher than the Kolmogorov. The former is fully described by the mean flow field, while the latter can be revealed through CFD by the computational equivalent of a tracer experiment. No mention is made to micromixing, which is not at all captured by the $k-\varepsilon$ model. Its incorporation would require further phenomenological models, such as those of Baldyga, Bourne and Hearn (1997).

2.2.5.5 Superposition of reaction engineering models

Once a compartmental network that properly accounts for mixing has been established, various mechanistic reaction engineering models can be superimposed on it. Although the present work focuses on bubble column reactors, the main concept lends itself to other potential reaction engineering applications, such as:

- Single-phase reactors with complex nonlinear reaction systems, whose direct coupling with CFD would pose serious computational problems.
- Stirred tank reactors with fast reactions, where micromixing models can be superimposed in the compartments located at the inlets.
- Precipitation, where additional population balance equations describing crystal growth, agglomeration or breakage must be solved.
• Multiphase non-catalytic reactors (e.g. bubble columns), where interfacial mass transfer and chemical reaction models can be accommodated within the compartments. These models must be resolved at the interfacial scale, and therefore cannot be directly incorporated in a CFD code.

• Catalytic and biochemical reactors, where adsorption models or cell population balances may be needed.

Some of the reaction engineering models required for the above applications require extra hydrodynamic information: e.g. a gas-liquid mass transfer model requires knowledge of gas hold-up, or a particle agglomeration model (in precipitation) may be a function of the energy dissipation rate. This extra information may be obtained from the CFD simulations in the same way that the mixing information was extracted, i.e. by averaging the distributed information yielded by CFD within the compartments.

The present work focuses on bubble column reactors, and a way of introducing interfacial mass transfer and chemical reactions in the network according to the penetration theory will be demonstrated. This way, the absorption enhancement due to a fast, but not instantaneous, chemical reaction system can be properly taken into account. It must be noted, however, that the main concept lends itself to other potential reaction engineering applications, such as micromixing in stirred tanks and precipitation.

2.3 Application to bubble column reactors

2.3.1 Mixing in bubble column reactors

The traditional chemical reaction engineering approach to mixing is to envisage it as the sum of two processes: macromixing, i.e. the mixing of segregated fluid elements due to the gross features of the flow field, and micromixing, i.e. molecular-level transfer of material between the elements, thus bringing the reactants into contact (Danckwerts, 1958). An intermediate mixing scale (mesomixing) has also been suggested (Balodyga, Bourne and Hearn, 1997) to account for the mixing of the inlet plume with its surroundings. This approach has been developed mainly
for stirred tanks, where incomplete mixing can create very steep gradients in the case of fast reactions. This is not the case with gas-liquid reactors, however, for the following reasons:

- A very fast reaction would be exhausted at the interface.
- In case of a reaction that proceeds in the bulk the gradients are not as steep as in stirred tanks, due to the fact that the feed is not localised around an inlet point but distributed along the whole riser section.

The main causes of mixing in bubble columns are liquid circulation due to forced or natural convection, and lateral turbulent exchange of regions with high gas holdup (and hence reactant feed) with regions with low gas holdup. Thus no phenomenological model of micromixing is required; instead, a model of interfacial mass transfer and chemical reaction must be employed to link the compartments of the liquid phase with those of the gas phase.

2.3.2 Model of interfacial mass transfer and chemical reaction

Apart from the coupling of bulk-phase chemical reactions with hydrodynamics, the simulation of multiphase reactors requires the solution of additional models that account for interphase transport. In the case of gas-liquid reactors, conventional ways of accounting for mass transfer were reviewed in section 1.1.2. It was concluded that research so far has concentrated on problems where chemical reactions and mass transfer in the interface can be decoupled or simplified. This is possible in the following situations:

- Reactions too slow or instantaneous, that do not occur in the same time frame as mass transfer.
- Steady-state problems, where reaction enhancement is constant.
- Simple reaction systems, for which analytical solutions to the diffusion-reaction equation can be found.

Unless the time scales of a gas-liquid reaction and mass transfer differ by orders of magnitude (instantaneous or slow reaction regime), however, these two phenomena
are intricately coupled. The consumption of absorbent at the interface increases the
driving force for mass transfer, a phenomenon known as reaction enhancement, and
the amount of absorption can only be resolved by detailed solution of the diffusion
- reaction equations that result from a phenomenological model such as the film,
penetration and surface renewal. These models were reviewed in sec. 1.1.1.1.

2.3.2.1 Implementation of the penetration model

Among the most established absorption models, the penetration theory will be em­
ployed here because it is conceptually suited to the mechanism of bulk-interface
interaction in dispersed multiphase flows: a bubble moving with a slip velocity rel­
ative to the bulk is in contact with a surrounding element of bulk fluid for a short
amount of time, and subsequently overcomes it. The only parameter, the contact
time, can be theoretically estimated from the bubble mean diameter and terminal
velocity:

\[ \tau = \frac{d_b}{u_b} \]  

(2.12)

These concepts are implemented here according to the following scheme: the
liquid element surrounding the bubble and the bulk are considered as two separate
dynamic reactors that operate independent of each other and interact at discrete
time intervals. In the beginning of the contact time, the interface is being detached
from the bulk. Subsequently, the species mass balances equations are solved in
each interfacial and bulk reactor. When overcome by the bubble, the liquid element
returns to the bulk and gets mixed with it. Not all the liquid elements within a
cell will contact the gas simultaneously, but since the contact time is very small
compared to the process time the situation is equivalent to having a very large
interfacial element emerging from the bulk, returning to be mixed and then re-
emerging in the surface.

The concept of the penetration theory (Higbie, 1935) has previously been em­
ployed for modelling flat interface (Hostomsky and Jones, 1995) and stirred tank
gas-liquid reactors (Van Elk, Borman, Kuipers and Versteeg, 2000). It offers a
major advantage from a computational point of view: the bulk and the interfacial
phenomena are decoupled, and each one can solved independent of each other, inter­
acting via the mixing process. The new concentrations that arise from the mixing
are the initial conditions for the next contact time integration.

The main assumptions inherent in this implementation are:

- Isothermal operation. The inclusion of an energy balance is straightforward, provided that the temperature change is not strong to to the point of inducing a fundamental change in the hydrodynamics of the system.

- Dilute suspension, if the reaction product is a solid.

- Homogeneous bubbly flow, i.e. no bubble coalescence. This assumption allows the use of a mean bubble diameter and terminal velocity for the estimation of the contact time in the penetration model.
2.3.3 Interface equations

While the equations describing the bulk are being integrated along a contact time, mass transfer and chemical reactions take place at the interfacial elements, each one isolated from the others and the bulk network according to the penetration model concept. The latter assumes unsteady-state diffusion without convection on one spatial dimension (note that the notation of the compartmental model, such as indices referring to compartments and phases, does not apply to this section - the interfacial element is treated in isolation):

\[
\frac{\partial C_i(x,t)}{\partial t} = D \frac{\partial^2 C_i(x,t)}{\partial x^2} + f_i(C_i(x,t)) \tag{2.13}
\]

where \( f_i(C_i(x,t)) \) stands for the non-linear reaction terms in the balance of component \( i \):

\[
f_i(C_i(x,t)) = \sum_{j=1}^{J} r_j(C_i(x,t))
\]

At the gas side of the interface \( x = 0 \) the concentration of the absorbed component can be explicitly specified by an equilibrium relation such as Henry’s law, while zero flux is the boundary condition for the non-volatile components:

Volatile species:

\[
x = 0, t > 0 \Rightarrow C_i(x,t) = \frac{yP}{H} \tag{2.14}
\]

Non-volatile species:

\[
x = 0, t > 0 \Rightarrow \frac{dC_i(x,t)}{dx} = 0 \tag{2.15}
\]

To impose boundary conditions at the bulk side we adopt the concept of Toor and Marchello (1958), according to which they can be specified at a fixed distance \( x = \delta \) from the interface, in contrast to the cumbersome \( x \rightarrow \infty \) proposed by the original Higbie model. This is valid on the condition that the concentration gradients are zero approaching the bulk at the given distance, certified by the following criterion:

\[
\tau \leq \frac{\delta^2}{D} \tag{2.16}
\]

For 1-10 mm bubbles with a terminal slip velocity of 20-30 cm/s this condition is easily satisfied with an interfacial length of 10-100\( \mu \)m. The boundary conditions
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at the bulk side are the bulk concentrations at the beginning of the contact time, and therefore constitute the initial conditions as well:

\[ x = \delta, t > 0 \Rightarrow C_i = C_{i, \text{bulk}} \]  
\[ t = 0, x > 0 \Rightarrow C_i = C_{i, \text{bulk}} \]

(2.17)

(2.18)

In the end of a contact time, the interfacial elements and the bulk cells are getting mixed. The concentrations that result from the mixing calculation constitute the initial conditions for integration of the bulk in the compartmental model, as well as the interface over the next contact time. The total amount of each component present in the interface is calculated from the concentration distribution at the end of the contact time as:

\[ [C_{i, \text{tot}}]_{\text{int}} = \int_0^\delta C_i(x, \tau) \, dx \]  

(2.19)

The initial conditions for the next contact time integration are then calculated as:

\[ C_{i, \text{klp}}^0 = \frac{[C_{i, \text{tot}}]_{\text{int}} + C_{iklp} \cdot V_{klp}}{V_{klp}^\text{int} + V_{klp}} \]  

(2.20)

2.4 Computational implementation

2.4.1 Generalised compartmental model

In our implementation, the set of ODEs (eqs. 2.7) that describe the bulk of each compartment are integrated with the LSODE set of subroutines, employing an Adams method. LSODE is available in the public domain via www.netlib.org, and is discussed by Hindmarsh (1983).

Successful tuning of the generalised compartmental model to the CFD requires, as it will be shown in the next chapter, consideration of many different structures for successive refinement. Hence the success and usability of the method relies on an efficient computational implementation that facilitates the generation of structures as much as possible. In our code, all network characteristics are defined through the input files, in a comprehensive way that allows quick and easy modification. Each structure is checked for satisfaction of the mass balance (eq. 2.5) before proceeding.

The code has been programmed in FORTRAN 90, which allows elementary object oriented programming through the use of derived data types (structures). A
structure is constructed to describe the network compartments, while all properties associated with them (volumes, flowrates, concentrations etc.) are defined as elements of the structure, corresponding to the matrix notation employed here. Only a few seconds are required for simulating tracer experiments, as these involve just convection of material without reaction terms. Once the network has been tuned to the CFD results, the reactions can be superimposed. The runs with reaction are still very fast; about 30 compartments were employed for the case study presented here compared with thousands of CFD cells.

2.4.2 Solution of interfacial PDEs

Since the interfacial elements are isolated during the contact time, their equations describing them are not coupled, and therefore there is no need for an iterative solution. Each element can be treated individually, by integrating the system of two-dimensional non-linear parabolic PDEs that describe the interfacial phenomena. The discretised equations are here generated from integral forms of the differential equations, in a manner similar to the finite volume method method (Patankar, 1982). Functions of the nodal values may be used to approximate the variation of each term over the element, but eventually the solution constitutes of the nodal values only. The derivation of the discretised equation begins by integrating eq. (2.13) over an one-dimensional control volume:

\[
\int_{x}^{x+dx} \frac{\partial C_i(x,t)}{\partial t} dx = \int_{x}^{x+dx} D \frac{\partial^2 C_i(x,t)}{\partial x^2} dx + \int_{x}^{x+dx} f(C_i(x,t)) dx
\]  

(2.21)

Diffusion terms are approximated with central differences. The grid and notation employed for the spatial and temporal discretisation is shown in fig. 2.4. Regarding the time integration, the following generalised implicit formula will be employed here:

\[
\int_{t-1}^{t} C_{ip} dt = (aC_{ipt} + (1 - a)C_{ipt-1}) \delta t
\]  

(2.22)

This reduces to the fully implicit scheme for \( \alpha = 1 \), the explicit scheme for \( \alpha =
Cells employed for the calculation of cell $p$ at time $t$ in the implicit temporal integration scheme

$t-1$

$w,p,e$: cell centres
$W,E$: cell faces

Figure 2.4: Notation used for the derivation of the discretised equations at the interface.

Let $\alpha = 0.5$. After some manipulation we obtain:

$$C_{i,p,t}\delta x - C_{i,p,t-1}\delta x =$$

$$D \frac{C_{i,x,t}}{\delta x} \delta x - \frac{2D}{\delta x} \delta x (\alpha C_{i,p,t} + (1 - \alpha) C_{i,p,t-1}) + D \frac{C_{i,y,t}}{\delta x} \delta x + \int_{t-1}^{t} (f_{i,p}\delta x) dt \quad (2.23)$$

To proceed, the non-linear term $f$ must be linearised. Newton's method is employed, with numerical calculation of the partial derivatives by perturbing $f$ around $C_i$ at the previous time point:

$$\left. \frac{\partial f_{i,p}}{\partial C_i} \right|_{t-1} = \frac{f(C_i) - f(C_i + \delta C_i)}{\delta C_i} \bigg|_{t-1} \quad (2.24)$$

See Appendix A for more details on the linearisation scheme and the calculation of the coefficients.

2.4.3 CFD modelling of vertical dispersed gas-liquid flows

The CFD model was implemented using the CFX code (version 4.3, AEA Technology, Harwell, UK), which solves the multiphase Eulerian-Eulerian model with the IPSA (Inter-Phase Slip Algorithm - Spalding, 1980). Since the Eulerian-Eulerian model will be adopted in this study, it will now be considered in more detail. This
section will discuss several currently unresolved or debatable issues associated with its application, with relevance to the objectives of the present study.

2.4.3.1 Grid and time stepping

Several computational studies have compared transient with steady-state and two-dimensional with three-dimensional simulations of bubble columns (e.g. Sokolichin and Eigenberger, 1994; Jakobsen et al., 1997). It was found that only three-dimensional transient simulations are able of revealing the chaotic nature of such flows. In this study, though, CFD simulation is employed as a tool for estimating essentially average values of hydrodynamic quantities, and forms part of a tool intended for design and scale-up of equipment within a reasonable amount of computational time. Therefore the approach adopted here is a two-dimensional dynamic model that can be resolved in an ordinary workstation within a couple of days. The steady-state approach was also considered, but was found to be unstable and exhibit severe problems with mass conservation, at least in our solver (CFX 4.3). This is to be expected since several of the phenomena involved are highly dynamic, in particular the introduction of gas that causes an expansion of the column during the first seconds, and the accumulation of bubbles in the downcomer. A simulation of about 10 real-time seconds was found to be adequate for the main flow features to stabilise. The time stepping policy was to initiate with very small time steps (0.001 - 0.005 sec), and gradually switch to 0.01 sec once the main features of the flow had been established. Convergence was evaluated on the basis of the residuals, and especially of the continuity equation residual, as compared to the total mass in the reactor. It was ensured that the residual per time step would be less than 0.01% of the total mass.

A bubble column is a cylindrical axisymmetric geometry but it is well known that the motion of rising bubbles is swirling, and this can only be captured with a 3-D model (Sokolichin and Eigenberger, 1995). As in this study CFD is used to predict averaged quantities of gas hold-up and liquid flow, phase distribution along the cross-section is of little interest and 2-D modelling is considered to be sufficient. This also rules out the need for including non-drag forces that give rise to the swirling motion. Regarding the grid, Eulerian-Eulerian gas-liquid flow simulations have an inherent
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limitation with respect to grid refinement: if the cells are similar to or smaller than the bubbles, the model loses its physical basis (Jakobsen et al., 1997; Sokolichin and Eigenberger, 1994). Therefore grids with cells having dimensions in the order of 1 - 0.5 cm were investigated. Further refinement did not have significant effect, while the physical validity of the model would be challenged. Thus a grid of 0.5 cm cells was considered sufficiently fine, given that some flow details such as swirling motion would be lost anyway due to the two-dimensional reduction.

2.4.3.2 Boundary conditions

Standard no-slip conditions are imposed on the column walls and draft tube. The introduction and removal of the gas phase is a problem, however. An exact specification of gas inlet velocity would imply that its flow is fully developed at the sparger holes, which is clearly unrealistic. Moreover, the sparger holes are extremely small (0.5 - 1mm) and the grid is not fine enough to resolve them. Therefore, it is considered preferable to insert the gas as a source in the continuity equation at the cells next to the boundary. No momentum source was applied, as jetting is unlikely to occur at low superficial velocities and homogeneous flow.

To remove the gas, CFX 4.3 suggests the use of a "degassing" boundary condition, which means to model the free surface as a solid wall and remove via a sink any gas that reaches that point with an upward axial velocity. We consider this approach unrealistic, however, because it is an artificial way to model the free surface and, most important, because it does not reflect the initial expansion of the water column. As gas is being introduced, water level rises, and this can cause problems in the continuity equation if the free surface is modelled as a wall. This may be mitigated by introducing some dispersed gas as an initial condition, but that would require an a priori knowledge of the gas holdup. The approach adopted in this study is to leave some space filled with gas above the water column in the initial conditions for the volume fraction, and let the free surface expand as the gas is being introduced. Of course, the exact shape of the free surface cannot be precisely simulated with a two-fluid model, but it is of no relevance here. The boundary condition is imposed on the gas end of the column, and sets the pressure there equal to atmospheric. Thus the inserted gas is left to exit naturally, by increasing the pressure above the
column. The validity of the approach can be easily checked by a mass balance between the inlet gas flowrate and the amount that exits from the pressure boundary - in the simulations shown here these were approximately the same.

2.4.3.3 Additional phenomenological models

Another important issue is the turbulence modelling: the applicability of the $k - \varepsilon$ model to multi-phase flows has been the subject of several studies (Schwarz and Turner, 1998; Pfleger et al., 1999; Sokolichin and Eigenberger, 1999). They seem to conclude that a combination of a $k - \varepsilon$ model to describe the liquid phase with a laminar gas phase yields results closer to experimental evidence than other tested combinations (laminar-laminar, $k - \varepsilon - k - \varepsilon$). Following this, a standard $k - \varepsilon$ model is applied here to the liquid phase only. It must be noted that turbulence modelling in multi-phase models is far from definitive, as the $k - \varepsilon$ model was initially developed for single-phase flows. For this reason, many authors supplement it with additional models for the turbulent dispersion of the gas phase, which however introduce more unknown parameters. In this work, however, the simplest approach was chosen.

Two a priori assumptions must be made to use the Eulerian-Eulerian multiphase model. Regarding the bubble size, only a mean value is being used (in accordance to the reaction engineering model), and it can be estimated empirically. The second assumption concerns the drag force, which is responsible for the coupling between the equations either phase. The drag force for a bubble swarm is given by eq. 1.16. Correlations for determining the drag force coefficient were reviewed in chapter 1. Among them, the Schwarz and Turner approach (eq. 1.19, used with comparative success in several studies of vertical bubbly flows such as Sokolichin and Eigenberger (1994), is adopted for the purposes of this study.

2.5 Case study: Parallel-consecutive reactions in a draft-tube bubble column

In the example that follows, the usage of this methodology will be demonstrated via a gas-liquid reaction system (absorption of CO$_2$ into alkali solution) taking place in a draft tube bubble column. It must be stressed, however, that the particular
CFD implementation described below does not constitute a part of the methodology; in principle any CFD model can be used in conjunction with the generalised compartmental model.

2.5.1 Validation of CFD model

Before integrating with the reaction engineering model, the hydrodynamic predictions of the CFD simulation will be compared with the experimental data of Wachi, Jones and Elson (1991). In that work gas hold-up in both riser and downcomer sections and mean liquid circulation were measured in an internal draft tube bubble column (fig. 2.5), using the piezometric and hot pulse injection methods respectively. The column was 2.5 m high, 0.22 m inside diameter and 85 l in capacity, and gas was injected via a perforated plate in the bottom of the inner annulus (riser). Different liquids, gas flowrates and internal draft tubes were tested in that work, but the data presented here for comparison are for the air-water system and a 0.12 m diameter draft tube. Fig. 2.6 shows a comparison between experimental and simulated gas hold-up at five different flowrates.

The shortcomings of the Eulerian-Eulerian approach with respect to predicting flow details and radial phase distribution) have been pointed out by many authors, such as Jakobsen (2000). In this framework, however, CFD is employed for calculating averaged properties, and in this context the model seems to be adequate. At
Figure 2.6: Gas hold-up in the riser and downcomer, comparison of CFD model with the experiments of Wachi et al. (1991).

In the riser, agreement between CFD and experiment is excellent and may be considered quantitative at the low gas flowrates tested, which explore the whole range of homogeneous bubbly flow. A slight deviation at the highest flowrates may indicate the beginning of the transition to the heterogeneous regime. Gas entrapment in the downcomer is, however, underpredicted, a fact possibly related to the turbulence modelling which, for multiphase flow, is still at a primitive stage. The model would be more accurate if applied to external loop reactors, where gas is totally separated at the top. In this case, however, holdup in the riser is more important, as the entrapped bubbles in the downcomer are soon depleted from the reactant component and play no important role in the reaction engineering.

The compartmentalisation will be carried out for the steady state. For this reason, the concentration difference between two points in the column (one in the riser and one in the downcomer) was monitored until a pseudo-steady state was reached. The virtual tracer experiment was then carried out in a manner resembling...
the feeding policy of bubble columns, eq. 2.10.

2.5.2 Network structure and mean flowrates

Fig. 2.7 shows the flow field, tracer concentration and volume fraction profiles. Note that the column head is occupied by gas, as it can be seen in the volume fraction contour plot, and therefore the tracer concentration in that region is of no importance. Fig. 2.8 shows radial concentration profiles in the riser at different axial points. By inspecting these results, we draw the following conclusions regarding the structure of the compartmental model:

- The liquid phase flow field is dominated by a circulation loop, characteristic of airlift reactors, due to the differential gas holdup between the riser and downcomer. This loop is the main macromixing mechanism and must be accounted for by the structure of the network.

- The downcomer resembles a plug flow, with no observable gradient in the lateral direction. Therefore there is no need for lateral compartments in that region.

- The volume fraction plot (fig. 2.8) shows the gas to be concentrated in the middle of the inner annulus and reduce towards the wall. Consequently the reactant is fed mostly throughout the central region of the riser, and disperses towards the walls mainly through turbulent mixing. As a result, a lateral concentration gradient is generated in the riser.

The change in the direction of the flow, from upward in the riser to downward in the downcomer, suggests the use of two vertical layers of cells for the riser and downcomer, respectively. The fact that the gas holdup is concentrated in the middle of the inner annulus suggests that the riser be further splitted into 2 vertical layers, one for the gas-rich and one for the gas-poor region. Since the reactant is introduced to the bulk from wherever gas is present, this way we emulate the feeding policy of the reactor. Based on the above, a network comprising 10 horizontal and 3 vertical layers of cells is constructed. The cells are initially linked with flowrates reflecting only the convective mean flow, averaged along the cell boundaries. This network is depicted in fig. 2.9.
Figure 2.7: CFD results (left to right): Volume fraction, mean velocity profile (m/sec) and dimensionless scalar concentration (resulting from the virtual tracer experiment), axisymmetric simulation, column axis on the right.
Figure 2.8: Radial gas holdup profiles in the column at four axial locations.
2.5.3 Turbulent exchange

Figs. 2.10, 2.11 show the axial concentration profiles in the two riser layers and in the downcomer. It is evident that this network yields a rather poor description of the riser concentrations, due to the lack of direct material exchange between the two vertical layers of riser compartments. The network is thus refined to the one shown in fig. 2.12, featuring lateral turbulent exchange. Agreement between the network and CFD is significantly improved, as shown by figs. 2.10, 2.11. The discrepancy is now in the order of $1 - 4\%$, and is deemed acceptable for the purpose of this study. We are now ready to proceed with the superposition of the chemical phenomena.

2.5.4 Superposition of chemical reactions

Assume the following system of mass transfer accompanied by a system of parallel – consecutive reactions:

$$\text{CO}_2^{(aq)} + \text{OH}^- \rightleftharpoons \text{HCO}_3^- \quad [i]$$
$$\text{HCO}_3^- + \text{OH}^- \rightleftharpoons \text{CO}_3^{\text{aq}} \quad [ii]$$

Kinetics and equilibrium parameters are taken from Astarita (1967) and are shown in table 2.1. Fig. 2.15 shows the molar fractions of $\text{CO}_2^{(aq)}$, $\text{HCO}_3^-$, $\text{CO}_3^{\text{aq}}$ at equilibrium conditions, obtained by for different values of pH in the range 7-13. At pH=12-13 the equilibrium is totally shifted towards $\text{CO}_3^{\text{aq}}$; below pH=11, $\text{HCO}_3^-$ exists in significant amounts, and around pH=10 its coexists with $\text{CO}_3^{\text{aq}}$ at equal concentrations, while at pH=9 the molar fraction of $\text{CO}_3^{\text{aq}}$ is less than 0.1. As pH approaches the neutral range, $\text{CO}_2^{(aq)}$ dominates.

<table>
<thead>
<tr>
<th>Parameter, units</th>
<th>Value</th>
<th>Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>Equilibrium constant of sub-reaction i, $K_i$, m$^3$ mol$^{-1}$</td>
<td>$6.1 \cdot 10^4$</td>
<td>Astarita, 1967</td>
</tr>
<tr>
<td>Equilibrium constant of sub-reaction ii, $K_{ii}$, m$^3$ mol$^{-1}$</td>
<td>5.88</td>
<td>&gt;&gt;</td>
</tr>
<tr>
<td>Kinetic constant of sub-reaction i, $k_i$, mol$^{-1}$ m$^3$ s$^{-1}$</td>
<td>5.0</td>
<td>&gt;&gt;</td>
</tr>
<tr>
<td>Kinetic constant of sub-reaction ii, $k_{ii}$, mol$^{-1}$ m$^3$ s$^{-1}$</td>
<td>$1 \cdot 10^4$</td>
<td>&gt;&gt;</td>
</tr>
<tr>
<td>Henry’s constant, $H$, atm m$^3$ mol$^{-1}$</td>
<td>0.0346</td>
<td>&gt;&gt;</td>
</tr>
</tbody>
</table>

Table 2.1: Parameter values and their sources.

Fig. 2.16 shows a sample of the concentration profiles in the interface, as resulting by the application of the penetration model. Figs. 2.17, 2.18 show the time course of
Figure 2.9: Compartmental network with two vertical layers of riser cells, to emulate the feeding policy of a bubble column.
Figure 2.10: Concentration profiles in the riser, network of fig. 2.9.

Figure 2.11: Concentration profiles in the downcomer, network of fig. 2.9.
Figure 2.12: The network of 2.9 is improved to account for lateral turbulent exchange in the riser.
Figure 2.13: Concentration profiles in the riser, network of fig. 2.12.

Figure 2.14: Concentration profiles in the downcomer, network of fig. 2.12.
Figure 2.15: Equilibrium concentrations at different values of pH for the system $CO_2 - HCO_3^- - CO_3^{2-}$

the concentration and pH profiles, averaged over the whole column. In the beginning of the process almost all of the absorbed gas is converted to $CO_3^{2-}$. The increase in $CO_3^{2-}$ concentration soon reaches a peak however, and is subsequently being reduced, while $HCO_3^-$ starts to increase. This event is reflected in the pH curve by a change in slope, and can be explained by the shift of equilibrium of reaction (ii) in favour of $HCO_3^-$ occurring around pH 11. At lower pH values the equilibrium of reaction (i) shifts towards $CO_2$, and another change in the pH slope is observed. After this point, chemical reactions become negligible and the effect is that of physical absorption. Later absorption stops and pH becomes constant, as there is no driving force due to the high bulk concentration of dissolved $CO_2$. The shape of the pH curve is in agreement with the experiments of Fleischer et al. (1996).

In fig. 2.19 we see a comparison of the total amount of $CO_2$ absorbed in the end of the chemical process to the amount that would have been absorbed during physical absorption. It is much higher in the chemical process, partly but not only due to the enhancement of mass transfer by the reaction at the interface. In the chemical process, the bulk conversion of $CO_2$ to $CO_3^{2-}$ and $HCO_3^-$ keeps the concentration of dissolved $CO_2$ in the bulk close to zero, thus maintaining the mass transfer driving force at high levels, until the time of the pH drop. On the other hand, the physical
Figure 2.16: A sample of simulated concentration profiles in the interface absorption process is sooner halted due to the accumulation of CO$_2$ in the bulk. This example illustrates the need for rigorous dynamic modelling in the analysis of batch absorption processes.
Figure 2.17: Evolution of pH.

Figure 2.18: Evolution of chemical species’ concentrations.
Figure 2.19: Physical and chemical absorption.
Chapter 3

Finite element scheme for the time-dependent population balance

3.1 The need for a new approach

The population balance equation (PBE) lies at the core of particulate process modelling, yet its numerical solution continues to pose severe challenges. In chapter 1 the existing numerical methods for solving the population balance were reviewed and classified into two categories: methods based on a function approximation of the solution (e.g. weighted residual, finite element) and methods that divide the domain of the independent variable into classes, commonly referred to as DPB (Discretised Population Balances). The main conclusions from the review can be summarised in the following:

Discretised population balances:

- are custom-derived methods tailored to solve individual problems efficiently but lack generality,
- are praised mainly for their ease of use and computational efficiency, and
- are engineered to predict some moments (usually one or two) exactly, but other moments may exhibit severe errors, as may do their reproduction of the distribution.
Finite element scheme for the time-dependent population balance

Function approximation methods:

- are a general and very flexible class of numerical methods and as such, when applied to the PBE, they can easily cope with any possible formulation, e.g. size-dependent growth, different agglomeration kernels etc.,
- retrieve the entire distribution with detail,
- are more difficult to implement and more computationally demanding, but
- their prediction of any moment is subject to errors.

In light of these conclusions, in this work we set out to derive a finite element scheme that will exhibit the following features:

Generality. The dynamic PBE with nucleation, growth, dispersion, aggregation and breakage will be considered, the only simplifications being spatial uniformity and one-dimensional domain of the particulate variable. The individual numerical needs of each one of these processes will be identified, and suggestions will be made for configuring the finite element scheme to optimally accommodate them.

Stability. To achieve this we shall employ linear trial functions, and collocate them in a way that reflects the convective nature of the growth term. This way, even problems with fronts such as the dynamic nucleation case can be dealt with accurately.

Conservation of moments. A good numerical method must be conservative, if inaccurate, even in a coarse grid. The schemes used in solid and fluid dynamics usually possess this property, but in the PBE conservation is not so straightforward due to presence of the integral terms. The possible sources of error will be identified, and ways of mitigating them will be proposed.

Ease of implementation. Owing to the use of linear trial functions, the resulting scheme is very simple and easy to incorporate in a computer code.

Balance of accuracy and speed. As it will be demonstrated by the test cases, very accurate results can be achieved within reasonable computational time,
even in desktop computers. The lower accuracy of the linear trial functions is compensated by the use of a finer grid.

### 3.2 Formulation of the PBE

In chapter 1, a general form of the population balance was written as:

\[
\frac{\partial n(x_i, t)}{\partial t} = - \frac{\partial (u_i \cdot n(x_i, t))}{\partial x_i} + D \frac{\partial^2 n(x_i, t)}{\partial x_i^2} + \\
+ B_{\text{nuc}} + B_{\text{agg}} - D_{\text{agg}} + B_{\text{br}} - D_{\text{br}} \quad (3.1)
\]

The independent variables \( x_i \) in the PBE are of two types:

1. The dimensions of physical space, \( \{x_i, i = 1, 2, 3\} \).

2. The properties of the population whose distribution we wish to calculate.

These two types are sometimes called external and internal coordinates (Randolph and Larson, 1988). The first issue to be discussed before proceeding with the formulation is the choice of internal variable. In crystallisation, it is a measure of particle size. Previous work (Gelbard and Seinfeld, 1978; Hounslow et al., 1988; Randolph and Larson, 1988;) has often favoured the use of particle diameter, due to the fact that it is the variable most commonly measured by particle sizing equipment. Moreover, certain particulate processes such as crystallisation often exhibit size independent growth, which facilitates the computation when a linear measure of size is used as independent variable. In this work, however, we have chosen to formulate the PBE in terms of particle volume in order to emphasise conservation through correct discretisation of aggregation and breakage terms. In these two processes volume is being conserved, and therefore it is the natural choice of an independent variable in the PBE. Furthermore, if particle diameter were chosen, the expression for the aggregation would include shape factors that are system-dependent. In what follows \( x_i \) will refer to the spatial coordinates \( (i = 1, 2, 3) \), while \( v \) to particle volume.

A spatially distributed PBE implies coupling with fluid dynamics. Such a coupling raises extreme computational demands and has only been accomplished through oversimplification of the PBE (see chapter 1). For this reason, the PBE will
be formulated for a spatially uniform system, by integrating eq. 1.23 over a uniform
finite domain:
\[
\frac{1}{V} \int \int \int \left[ \frac{\partial n(v, t)}{\partial t} + \sum_{i} \frac{\partial}{\partial x_i} \left[ u_i \cdot n(x_i, t) \right] - D \frac{\partial^2}{\partial x_i^2} \left[ n(x_i, t) \right] - B_{\text{nu}} - B_{\text{agg}} + D_{\text{agg}} - B_{\text{br}} + D_{\text{br}} \right] dV = 0
\]  
(3.2)

The velocity with respect to the internal dimension (volume) is in fact the growth rate. Thus the convection term can be split into differentiation along physical space and differentiation along the volume coordinate. The physical interpretation of the former is the change in the distribution due to inflow and outflow of particles with different distributions, while the latter denotes change due to growth.

\[
\frac{\partial}{\partial x_i} \left[ u_i \cdot n(v, t) \right] = \frac{\partial}{\partial x_i} \left[ u_i \cdot n(v, t) \right] + \frac{\partial}{\partial \tau} \left[ G(v, t) \cdot n(v, t) \right]
\]  
(3.3)

The integration over a spatially uniform domain reduces convection in physical space into an input - output term:

\[
\frac{1}{V} \int \int \int \frac{\partial}{\partial x_i} \left[ u_i \cdot n(v, t) \right] dV = \frac{n_{\text{in}}(v, t) - n(v, t)}{\tau}
\]  
(3.4)

where \( n_{\text{in}}(v, t) \) is a known function denoting the input distribution.

Similarly, the diffusion term can be divided into diffusion in physical space and diffusion with respect to particle volume. The former can be neglected, since we are considering a spatially uniform domain. The latter is occasionally employed to account for growth rate dispersion, a phenomenon sometimes observed in crystallisation processes, where particles of the same size exhibit fluctuations in their growth rate. This phenomenon, occurring mainly in crystallisation, is attributed to the following reasons (Randolph and White, 1974): a) spatial effects (e.g. non-uniform supersaturation field, size classification due to flow field ) and b) variations in particle structure. Diffusional transport implies the transport of a property due to the presence of gradients, but rather than this, it is clear that the fluctuations mentioned above occur due to another variable not being included in the model. Indeed, spatial effects can be taken into account by coupling PBE with fluid dynamics, while variations in particle structure can be described by extra internal coordinates. Diffusion will not be included in our formulation, but a note will be made on how it could be accounted for if required.
The remaining terms are not functions of the spatial coordinates, and therefore remain intact after the spatial integration. Nucleation is the appearance of particles at certain size \((v_0)\), and is incorporated via a Dirac delta function. This means, however, that a sharp change in the profile will occur around the nuclei size, and the numerical method must be very stable to capture it.

\[
B_{\text{nuc}} = B_0 \delta(v - v_0) \tag{3.5}
\]

Aggregation is described by the continuous Smoluchowski equation (see chapter 1). The birth term denotes generation of particles of size \(v\) due to aggregation of any combination of smaller particles:

\[
B_{\text{agg}} = \frac{1}{2} \int_0^v \beta_a(v', v - v') n(v', t) n(v - v', t) dv' dv' \tag{3.6}
\]

The meaning of the aggregation kernel has been explained in chapter 1. The death term represents the disappearance of particles of size \(v\) due to their aggregation with any other particle:

\[
n(v, t) \int_0^\infty \beta_a(v, v') n(v', t) dv' \tag{3.7}
\]

The breakage birth term denotes the rate at which particles of size \(v\) are being generated due to breakage of any particle \(v' > v\). Thus integration from \(v\) to \(\infty\) is required:

\[
B_{\text{br}} = \int_v^\infty \beta_b N_p(v') b(v/v') n(v') dv' \tag{3.8}
\]

The term in the integral has to account for three things:

1. The rate at which breakage events occur. This is accommodated by the breakage kernel: \(\beta(v)\).
2. The number of daughter particles generated in each breakage event (degree of fragmentation), \(N_p(v)\).
3. The size distribution of these particles, represented by the function \(b(v/v')\).

The number of daughter particles and their size distribution are accommodated by the functions \(k(v')\) and \(b(v/v')\) respectively, and are subject to the following
Finite element scheme for the time-dependent population balance

Constraints (Ramkrishna, 1985):

\[ b(v/v') = 0 \forall v > v' \quad (3.9) \]

\[ \int_0^{v'} b(v/v') dv = 1 \quad (3.10) \]

\[ k(v') \int_0^{v'} vb(v/v') dv = 1 \quad (3.11) \]

The breakage death term is the rate at which particles of size \( v \) break into fragments of smaller size, and is represented simply by a polynomial term:

\[ D_{br} = -\beta_b(v)n(v) \quad (3.12) \]

Thus we arrive at the form of the PBE we shall deal with:

\[ \frac{\partial n(v, t)}{\partial t} = \frac{n_n(v, t) - n(v, t)}{\tau} - \frac{\partial}{\partial v} [G(v) \cdot n(v, t)] + B_0 \delta(v - v_0) + \]

\[ + \frac{1}{2} \int_0^v \beta_a(v - v', v') n(v - v', t) n(v', t) dv' - n(v, t) \int_0^\infty \beta_a(v, v') n(v', t) dv' + \]

\[ + \int_v^{\infty} \beta_b N_p(v') b(v/v') n(v') dv' - \beta_b(v) n(v) \quad (3.13) \]

It remains to define the boundary conditions. Eq. 3.13 is a first-order hyperbolic differential equation, and as such requires two initial conditions. The initial condition for the time domain is straightforward:

\[ n(v, 0) = n_0(v) \quad (3.14) \]

With respect to the volume coordinate, the most physically realistic choice is to state that there are no particles of zero volume:

\[ n(0, t) = 0 \quad (3.15) \]

The volume domain must then start from zero (rather than nuclei volume). Randolph and Larson (1988) proposed instead the condition \( n(v_0, t) = B_0/G \) for nucleation - growth problems, which implies that the number of nuclei is determined by a balance between nucleation and growth. This condition is not valid for aggregation problems, because in that case nuclei may also be reduced via the aggregation mechanism, and moreover it implies that nuclei are of zero size, which is not physically realistic (Rojkowski, 1990).
3.3 Preliminary considerations

3.3.1 Truncation of the domain

Since the domain of the volume variable is infinite, to carry out computations it must be truncated to a finite value. This implies that the tail of the distribution is being cut off, and the \( i \)-th moment is underpredicted by the amount:

\[
\int_{v_{tr}}^{\infty} v^n n(v, t) dv
\]

Clearly the underprediction is more significant the higher the order of the moment. Gelbard and Seinfeld (1978) provide analytical expressions of this error for cases that have an analytical solution, while Nicmanis and Hounslow (1998) provide a systematic procedure for selecting the upper limit in steady-state PBE, based on the second moment. To select the upper limit for a dynamic simulation, however, one needs insight into the effects of the truncation:

- In aggregation problems, particles forming aggregates bigger than the truncation point fall out of the computational domain, thus causing the moments to reduce. At the same time aggregation of particles from the domain with particles from the excluded tail is not being considered, which has a positive effect on the moments. The prevailing effect is a negative one, however, and in a pure aggregation problem the truncation error is easily spotted by a reduction in the first moment.

- In combined aggregation - breakage problems the effect of truncation is more difficult to estimate, as aggregates from the neglected tail can also break into particles within the computational domain.

- In growth problems the values of the population density that lie within the computational domain are not affected by the excluded tail. However, the whole body of distribution is moving towards bigger sizes, and once it reaches the truncation point the solution is clearly of no value.

Based on these remarks, we can deduce some heuristics for truncating the domain. For a dynamic simulation, the best practice is to use the moments to monitor
the solution and determine whether the truncation has any effect. Aggregation-breakage problems may be monitored as above on the basis of the first moment. In growth-dominated problems, the values of the population density that lie within the computational domain are not affected by the excluded tail. Since the whole body of distribution is moving towards bigger sizes, however, once it reaches the truncation point the solution is clearly of no value. Usually the size range to be covered can be estimated beforehand from the growth kinetics.

3.3.2 Grid generation

The choice of grid is very important, as it has profound effect on both accuracy and conservation of the scheme. It will be shown later that uniform meshes are more conservative; however, they cannot provide adequate accuracy, except for volume-independent growth problems (a rather unrealistic case) where the distribution moves with a constant speed. In aggregation and volume-dependent growth problems very big particles can be rapidly created, meaning that a long range of volume must be covered, while nucleation creates a large number of particles in the smaller size ranges, thus giving rise to steep gradients that require a locally refined grid. Non-uniform meshes can suffer from errors in the conservation of moments, however, and it is the intention of this work to provide a more conservative algorithm. Most of the test cases to be presented here have been solved on an exponential grid generated with the following series:

$$v_i = v_{\min} + v_{\min} \cdot \frac{1 - \alpha^{i-1}}{1 - \alpha}$$  \hspace{1cm} (3.17)

The volume $v_{\min}$ should correspond to the smallest particles present, i.e. the nuclei. By varying the coefficients $v_0$ an $\alpha$, one can generate grids with an increasing degree of coarseness. A preliminary criterion to facilitate conservation in non-uniform meshes is:

$$v_i \leq 2v_{i-1}$$  \hspace{1cm} (3.18)

This is because the biggest particle that can be formed by particles of sizes up to $v_{i-1}$ is $2v_{i-1}$, hence if the next mesh point is bigger than $2v_{i-1}$ that the distribution will fall to zero at a place between the two points. To satisfy this with grids generated by eq. 3.17, the first two intervals are kept equal.
For best performance the grid must be tailored to the needs of the problem, while an adaptive grid is probably the only way to cope efficiently with different cases and long-time simulations. Remeshing algorithms, however, are out of the scope of this work. Previous researchers (Gelbard and Seinfeld, 1978; Hounslow et al., 1988) have often employed scaled variables to facilitate the prediction of population density at the lower end of the size spectrum. However, as Nicmanis and Hounslow (1998) have shown, this renders the correct prediction of large particles and of the moments more difficult, as errors in the scaled domain are amplified when transforming the solution back to the original coordinates. We subscribe to this view, and therefore the method presented here is developed over an unscaled domain.

3.4 The main steps of the method

The dynamic PBE is a hyperbolic integro-differential equation with respect to time and the particulate coordinate – in this case the particle volume. The finite element method is used for the discretisation of the particulate coordinate, thus transforming the PBE into a system of ordinary differential equations with respect to discretised values of the population density. Temporal integration is subsequently carried out to compute the values in the new time point. The approach can be summarised in the following steps:

Step 1 Approximation of the solution with trial functions.

Step 2 Evaluation of integral terms. The aggregation birth, death and breakage birth terms must be evaluated via numerical integration.

Step 3 Formulation of weighted residual equations. The trial functions are substituted into the rest of the terms, and any derivatives are analytically elaborated. The resulting function, called the residual, is multiplied by appropriate weighting functions, depending on which version of the finite element method is desired - common choices are collocation, Galerkin and subdomain. The residual is integrated over each element, analytically or numerically.

Step 4 Integration of the resulting ODEs along time, which can be carried out with any initial value ODE integration method.
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Step 1: Approximation of the solution with trial functions

Step 2: Evaluation of integral terms

Step 3: Evaluation of the residual

Step 4: Propagation of the approximated solution along time

Figure 3.1: The main steps of the method.

It is useful to compare here this procedure with the implementation of the finite element method in more familiar applications, such as solid and fluid mechanics. The common practice in those cases is to derive a weak form of the original equation by using the divergence theorem, in order to lower the minimum required order of the approximating functions and to provide a means of introducing Neumann boundary conditions. This step is not required here, because the PBE is only first order with respect to both time and volume and the boundary conditions are all of Dirichlet type. Moreover, the discretised equations are often derived in terms of local variables defined within each element by a coordinate transformation. Although this practice does simplify the derivations, it cannot be applied to the PBE because of the aggregation source term which, as we shall see, involves values of the independent variable that are defined within different and often isolated elements. Finally, the application of the finite element method to an integro-differential equation requires an extra step, the evaluation of the integrals with the trial functions. It will be shown that this step is a potential source of errors in the conservation of moments if carried out in a straightforward manner.

The main steps of the method will now be elaborated.
3.4.1 Approximation of the solution

In finite element methods the solution is approximated with trial functions, which are low order Lagrangian or Hermitian polynomials that are non-zero only over a small part of the domain. These functions are written in such a way that their undetermined coefficients are the values of the solution at the nodal points. Previous work on the PBE employed mostly cubic splines (Gelbard and Seinfeld, 1978; Eyre and Wright, 1988; Steemson and White, 1988;) or high-order Lagrangian polynomials (Nicmanis and Hounslow, 1998). The scheme presented here, however, will be based on linear trial functions for two main reasons. Higher order trial functions require more evaluations and interpolations for the same number of nodal points, while first order functions can accommodate a much finer grid in the same amount of computational time - note that higher order does not necessarily imply higher accuracy, especially when the solution is irregular. Secondly, the use of linear functions enhances stability for growth problems. Furthermore, these functions simplify considerably the scheme and facilitate the analysis of conservation errors.

Two linear trial functions are required:

\[ f_{1,2}(v) = \left( \frac{v - v_{i-1}}{v_i - v_{i-1}}, \frac{v_i - v}{v_i - v_{i-1}} \right) \]

(3.19)

The solution over an element is then approximated as:

\[ n(v)_{[i-1,i]} = n_{i-1} \cdot \frac{v_i - v}{v_i - v_{i-1}} + n_i \cdot \frac{v - v_{i-1}}{v_i - v_{i-1}} \]

(3.20)

With linear trial functions each nodal point is also an element boundary. If the domain is divided into \( N \) elements, these correspond to \( N + 1 \) nodal points. In what follows, we shall use index \( i \) to denote a nodal point or an element bounded by \( (i - 1, i] \) (point 0 corresponds to the lower boundary).

3.4.2 Evaluation of integral terms

The three integral terms of the PBE (aggregation birth, aggregation death, breakage birth) are the main potential sources of conervation error in the numerical solution. Note that conservation is not a synonym of accuracy, which can be improved mainly by refining the grid. The aim of a conservative numerical method is to conserve the moments of the distribution (the most important being the first, which denotes
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conservation of mass) even in a coarse grid. To attain conservation, the sources generated by the discretisation must be balanced by the sinks over all elements. In the following, two potential sources of conservation error will be identified:

- Error due to grid non-uniformity.
- Error due to non-conservative quadrature.

These errors will be elaborated below, and ways of mitigating them will be proposed.

3.4.2.1 Aggregation source term

The aggregation source term is:

\[
B_{agg}(v, t) = \frac{1}{2} \int_{0}^{v} \beta_a(v', v - v') n(v', t) n(v - v', t) dv' \tag{3.21}
\]

The aggregation source term links each element with all the elements preceding it. To evaluate it at an element boundary \( v_m \), the integral may be split into a sum of contributions from all preceding elements:

\[
B_{agg}(v_m, t) = \sum_{i=1}^{m} \left[ \frac{1}{2} \int_{v_{i-1}}^{v_i} \beta_a(v', v - v') n(v', t) n(v - v', t) dv' \right] \tag{3.22}
\]

To calculate the contribution from each preceding element \([i-1,i]\) we need the products of the nodal values at its boundaries, \([n_{i-1}, n_i]\) with the values at the complementary points, i.e. the points \([v_m - v_{i-1}, v_m - v_i]\). Suppose that these lie in a distant element \( k \). Substituting the trial functions into eq. 3.22 we obtain:

\[
B_{agg}(v_m, t) = \sum_{i=1}^{m} \left\{ \frac{1}{2} \int_{v_{i-1}}^{v_i} \beta_a(v', v_m - v') \left[ n(v_{i-1}, t) \frac{v_i - v'}{v_i - v_{i-1}} + n(v_i, t) \frac{v' - v_{i-1}}{v_i - v_{i-1}} \right] \right\} \tag{3.23}
\]

Further elaboration of eq. 3.23 is not possible, however, because it involves values of the independent variable that lie in two distinct, possibly isolated elements. This is the reason why the common practice of transformation into local element
coordinates cannot be applied to the PBE. Another important point is that in a non-uniform grid the complementary points are not, generally, element boundaries, and may not even lie within the same element. In that case, their nodal values must be interpolated from the boundaries of the elements where they belong. This is a potential source of conservation error, best understood via a geometric representation of the aggregation source term. Fig. 3.2 shows the distribution in the element $i$ and the two neighbouring elements where the complementary points lie. By determining the nodal values by interpolation, the resulting integration corresponds to aggregation of the particles represented by the area of the following tetrahedra:

\[
\left[ (v_i, 0), (v_{i-1}, 0), (v_{i-1}, n(v_{i-1})), (v(i), n(v_i)) \right] + \\
\left[ (v - v_{i-1}, 0), (v - v_i, 0), (v - v_i, n(v - v_i)), (v - v_{i-1}, n(v - v_{i-1})) \right]
\]

Figure 3.2: Geometrical representation of the conservation error arising in non-uniform grids.

In the case of fig. 3.2, however, the second tetrahedron includes a triangle of non-existing particles, which is the source of the conservation error. The correct aggregating particles are defined by the areas $(I_1, I'_1)$ and $(I_2, I'_2)$. To obtain a conservative algorithm, the interval defined by the mirror points will be splitted
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into a number of sub-intervals, separated by the inter-element boundaries that are located within that interval. If \( K - 1 \) boundaries are contained, the integral over each element in eq. 3.23 is splitted into a sum of \( K \) contributions from the resulting sub-intervals:

\[
B_{agg}(v_m, t) = \frac{1}{2} \sum_{i=1}^{m} \sum_{k=1}^{K} \int_{v_{i,k}}^{v_{i,k-1}} \beta(v', v_m - v') n(v_m - v', t) n(v', t) dv'
\]  

(3.23)

The integrand of eq. 3.23 is non-linear, meaning that numerical integration must be applied. This can be another source of conservation error, since any errors in it will result in an imbalance between the aggregation birth and death terms. To eliminate this error, the integration must be exact. Since the integrand comprises the product of the two trial functions, both first order polynomials, and the aggregation kernel, if the latter is linear exact integration can be accomplished by adopting the following rule:

\[
\langle \text{order of quadrature} \rangle = \langle \text{order of kernel} \rangle + 2
\]  

(3.25)

Figure 3.3: The aggregation source term is of higher order than the approximated population density (which is linear). To obtain exact quadrature, a number of points must be interpolated within each interval.

Fig. 3.3 shows the curve representing the aggregation source term, superimposed over the approximated solution at an element. Since the end-points of the elements are readily available, a Newton-Cotes closed formula is preferable. The second-order
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formula, the familiar Simpson’s rule, is also exact for third-order polynomials and needs only one interpolation. The use of linear trial functions proves advantageous here, as higher order functions would require more interpolations. In the case of the constant kernel, Simpson’s rule yields:

\[ B_{agg}(t) = \frac{1}{2} k_a \sum_{i=1}^{N} \sum_{k=1}^{K} (v_{i,k} - v_{i,k-1}) \cdot \left[ \frac{1}{6} n(v_{i,k-1}, t)n(v_{k-1}, t) + \frac{4}{6} n(v_{i,k-1,1/2}, t)n(v_{k-1,1/2}, t) + \frac{1}{6} n(v_{i,k}, t)n(v_{k}, t) \right] \]

(3.26)

Finally, a note will be made about non-linear kernels. In that case, exact integration can be attained via Gaussian quadrature, by using the kernel as weight function. Press et al. (1992) describe a procedure for computing the coefficients of Gaussian quadrature with arbitrary weight functions. This process is rather complex, however, and for practical purposes a Newton-Cotes formula is usually acceptable, as the test cases will demonstrate.

3.4.2.2 Aggregation sink term

The aggregation sink term is easier to treat, since its integrand involves just the product of the population density and the aggregation kernel:

\[ D_{agg}(t) = n(v, t) \int_{0}^{\infty} \beta(v - v', v)n(v', t)dv' \]

(3.27)

As above, the evaluation should follow the rule:

\[ \langle \text{order of quadrature} \rangle = \langle \text{order of kernel} \rangle + 1 \]

(3.28)

3.4.2.3 Breakage source term

The breakage source term can be written as:

\[ \int_{v}^{\infty} \beta_s b'(v/v')n(v')dv' \]

(3.29)

where \( b'(v/v') \) is a function that incorporates both degree of fragmentation and size distribution of fragments. To maintain conservation, integration should follow the same rules proposed above for aggregation: the quadrature should ideally be
exact for the combination of trial functions, breakage kernel and daughter particles
distribution:

\[
\langle \text{order of quadrature} \rangle = \langle \text{order of kernel} \rangle + \langle \text{order of distribution function} \rangle + 1
\]  
(3.30)

### 3.4.3 Formulation of weighted residuals

So far, the integral terms of the PBE have been evaluated by way of numerical
integration. The weighted residual expressions over each element are formulated by
multiplying the PBE with an appropriate weight function and integrating:

\[
\int_{v_{i-1}}^{v_i} \left[ w(v) \cdot \frac{\partial n(v,t)}{\partial t} \right] dv = \int_{v_{i-1}}^{v_i} \left[ w(v) \cdot \frac{n_{in}(v,t) - n(v,t)}{\tau} \right] dv -
\]

\[
\int_{v_{i-1}}^{v_i} \left\{ w(v) \cdot \frac{\partial}{\partial v} [G(v) \cdot n(v,t)] \right\} dv - \int_{v_{i-1}}^{v_i} \{ w(v) \cdot [B_0 \delta(v - v_0)] \} dv -
\]

\[
+ \int_{v_{i-1}}^{v_i} [w(v) \cdot B_{agg}(v,t)] dv - \int_{v_{i-1}}^{v_i} [w(v) \cdot (D_{agg}(v,t))] dv -
\]

\[
+ \int_{v_{i-1}}^{v_i} [w(v) \cdot B_{br}(v,t)] dv - \int_{v_{i-1}}^{v_i} [w(v) \cdot (\beta_{br}(v) \cdot n(v,t))] dv
\]  
(3.31)

Common choices of weight functions are:

- **Collocation:** \( w(v) = \delta(v - v_k) \) where \( k \) is a nodal point  
  \hspace{1cm} (3.32)

- **Subdomain:** \( w(v) = 1 \)  
  \hspace{1cm} (3.33)

- **Galerkin:** \( w(v)c_i(v) = \frac{v - v_{i-1}}{v_i - v_{i-1}} \cdot \frac{v_i - v}{v_i - v_{i-1}} \)  
  \hspace{1cm} (3.34)

Collocation is an evaluation of the residual at the nodal points. The other
schemes require integration of the residual over a whole interval, and are thus more
expensive to compute. Moreover integral schemes cannot be combined with the
algorithm proposed above for conservative evaluation of the aggregation source term
because they require the integrand to be continuous. For these reasons, we choose
to employ the collocation scheme.
3.4.3.1 Growth term

Substituting the trial functions in the growth term, we obtain:

\[
\int_{v_{i-1}}^{v_i} \left\{ w(v) \cdot \frac{\partial}{\partial v} \left[ G(v) \cdot n(v, t) \right] \right\} dv = \\
\int_{v_{i-1}}^{v_i} \left\{ \delta(v - v_k) \frac{\partial}{\partial v} \left[ G(v) \cdot \left( n_{i-1}(t) \frac{v_i - v}{v_i - v_{i-1}} + n_i(t) \frac{v - v_{i-1}}{v_i - v_{i-1}} \right) \right] \right\} dv \quad (3.35)
\]

In orthogonal collocation on finite elements (Finlayson, 1975), the trial functions are of higher than first order and the evaluation is performed on points interior to each element. As a result, information flows to the collocation point from both ends of the element. This strategy is not efficient for discretising the growth term, because growth is a convective mechanism. This implies that the population at a point is affected only by upstream points - particles may grow only from smaller to bigger sizes. The use of higher order schemes where information flows in both directions can result in instabilities and oscillations. Here we may draw an analogy with CFD, where upwind differencing of the convection term is employed when the flow is convection dominated. Since the PBE has to dampen the effect of growth, in order to attain an equivalent effect the following discretisation is suggested (fig. 3.4):

\[
\int_{v_{i-1}}^{v_i} \left\{ \delta(v - v_k) \frac{\partial}{\partial v} \left[ G(v) \cdot \left( n_{i-1}(t) \frac{v_i - v}{v_i - v_{i-1}} + n_i(t) \frac{v - v_{i-1}}{v_i - v_{i-1}} \right) \right] \right\} dv = \\
\frac{n_i(t)}{v_i - v_{i-1}} - \frac{n_{i-1}(t)}{v_i - v_{i-1}} + n_i(t) \cdot \frac{\partial}{\partial v} (G(v_i)) \quad (3.36)
\]

An advantage of the finite element method is that it can cope with any size-dependent growth mechanism, given the volumetric growth function \( G(v) \). Note that for the common case of linear size-independent growth, the volumetric growth is size-dependent:

\[
G_i = \frac{dL}{dt} = \frac{\partial \left( \frac{v/v_k}{v} \right)^{1/3}}{\partial v} \cdot \frac{dv}{dt} = \frac{1}{3k_i^{1/3}} v^{-2/3} G(v) \quad (3.37)
\]
3.4.3.2 Nucleation term

Nucleation is implemented as a point source at the first nodal point. With finite elements, however, it is impossible to have true point sources, as trial functions are valid over a whole interval. Care must be taken, therefore, in order to make sure that the nucleation rate $B_0(t)$ reflects the number of particles produced in the first two intervals that are affected and has units of $(time^{-1}volume^{-1})$:

$$\int_{v_{i-1}}^{v_i} [\delta(v - v_0) \cdot B_0(t)] dv = \frac{B_0}{\frac{1}{2}(v_1 - v_0)} + \frac{B_0}{\frac{1}{2}(v_2 - v_1)}$$  \hspace{1cm} (3.38)

3.4.4 Temporal Integration

Adopting the collocation scheme, and in the absence of diffusion, the steps described so far result in the transformation of the original integro-differential dynamic PBE into a set of ordinary differential equations of the form:

$$\frac{\partial n_i(t)}{\partial t} = \frac{n_{i+1}(t) - n_i(t)}{\tau} - \frac{n_i(t)}{v_i - v_{i-1}} + \frac{n_{i-1}(t)}{v_i - v_{i-1}} - n_i(t) \cdot \frac{\partial}{\partial v} (G(v_i)) + \delta(v - v_0) \cdot \left[ \frac{B_0}{\frac{1}{2}(v_1 - v_0)} + \frac{B_0}{\frac{1}{2}(v_2 - v_1)} \right] +$$

$$+ B_{agg}(v_i, t) - D_{agg}(v_i, t) + B_{br}(v_i, t) - \beta_{br}(v_i) \cdot n_i(t)$$  \hspace{1cm} (3.39)
Having performed the pointwise evaluations of the PBE terms at the nodal points, we have succeeded in transforming the original partial integro-differential equation into a set of ordinary differential equations, where the dependent variables are the nodal values. The integration of these ODEs along time is bound to introduce further errors in terms of both accuracy and conservation, that can only be mitigated by employing an accurate integration algorithm. The results presented here were obtained using the LSODE integration package (Hindmarsh, 1983), using the Adams class of methods with adaptive time-stepping. The source code of LSODE is available in public domain (e.g. via netlib.org). In most cases, the error in the first moment was maintained below 1%. The FORTRAN 90 programming language has been used for the computational implementation.

3.5 Results and discussion

The method will now be tested against analytical solutions of the PBE for simplified problems. The CPU times reported correspond to an average desktop PC (Pentium III 933 MHz).

3.5.1 Aggregation

The continuous time-dependent PBE for aggregation only is:

\[
\frac{\partial n(v,t)}{\partial t} = \frac{1}{2} \int_0^v b_a(v,v') n(v-v',t) n(v',t) dv' - n(v,t) \int_0^\infty b_a(v,v') n(v',t) dv' - \int_0^\infty b_a(v,v') n(v',t) dv'
\]  

(3.40)

Most analytical solutions have been derived for the exponential initial distribution:

\[
n(v,0) = \frac{N_0}{v_0} e^{-\frac{v}{v_0}}
\]  

(3.41)

According to eq. 3.41, at \( v = 0 \) the distribution assumes a finite value \( n(0,0) = \frac{N_0}{v_0} \). This is incompatible with the initial condition used in our formulation \( n(0,0) = 0 \), and is further unrealistic, as there are no particles of zero size. In the test cases, the initial distribution in the simulation starts from zero and assumes the values of the exponential distribution at the size of the nuclei. This results in a slight
disagreement with the analytical solution at sizes close to zero, more evident in the aggregation-growth case (IV-2).

The grid to be employed in the pure aggregation cases is non-uniform, following the exponential series of eq. 3.17. The parameters, summarised in table 3.1, were chosen so that aggregation may proceed to a large extent. This can be quantified by the index of aggregation, defined by Hounslow (1990) on the basis of the zeroth moment:

\[ I_{agg}(t) = 1 - \frac{m_0(t)}{m_{0,t}} \]  

(3.42)

3.5.1.1 Case I-1: Size-independent kernel

This kernel can be used as an approximation of Smoluchowski’s equations for Brownian coagulation (perikinetic aggregation - see chapter 1). It is defined as:

\[ \beta_a(v - v', v') = k_a \]  

(3.43)

For the exponential initial distribution, Scott (1968) derived the following solution:

\[ n(v) = k_a \frac{4N_0}{(T + 2)^2} \cdot \exp\left( -\frac{v}{v_0} \cdot \frac{2}{T + 2} \right) \]  

(3.44)

where \( N_0 \) and \( v_0 \) are constants that denote the initial number of particles and average volume, respectively, and \( T \) is a dimensionless time defined as:

\[ T = k_a N_0 t \]  

(3.45)

A comparison of the numerical and analytical solutions for the population density, plotted on logarithmic axes in fig. 3.5, shows excellent agreement. The conservation of the first moment is also maintained at a good level, as shown in table 3.1, despite the use of a fairly coarse grid.

3.5.1.2 Case I-2: Sum (Golovin) kernel

This kernel is defined by:

\[ \beta_a(v - v', v') = k_a(v + v') \]  

(3.46)

A solution for this kernel, again with the exponential initial distribution (eq. 3.41), has also been provided by Scott:

\[ n(v) = \frac{N_0(1 - T)}{vT^{1/2}} \cdot \exp(-(1 + T)v') \cdot I_1(2v'T^{1/2}) \]  

(3.47)
where $J_1$ is the modified Bessel function of the first kind of order one. The dimensionless time here is defined as:

$$T = b_0 N_0 v_0 t$$  \hfill (3.48)

Results for the population density are plotted in fig. 3.6, while the moments and aggregation index are shown in table 3.1. As with the case of constant kernel, agreement is very good.

### 3.5.1.3 Case 1-3: Fluid shear and gravitational kernels

These kernels, describing orthokinetic aggregation and gravitational settling respectively, were discussed in chapter 1.

The shear kernel is defined as:

$$\beta_a(v - v', v') = k_a(v^{1/3} + v'^{1/3})^3$$  \hfill (3.49)

As we saw in chapter 1, this kernel corresponds to Smoluchowski’s theory of orthokinetic aggregation. Another kernel derived with physical arguments is the
gravitational coalescence (Drake, 1972):

$$\beta_a(v - v', v') = k_a(v^{1/3} + v'^{1/3})^2 \cdot v^{2/3} + v'^{2/3}$$  \hspace{1cm} (3.50)

Although there are no analytical solutions for these kernels that we are aware of, they will be considered here due to their physical significance. Furthermore, it is interesting to see how the method behaves in the presence of a non-linear kernel. The solution may still be partly evaluated on the basis of the conservation of the first moment, which must remain constant in any aggregation process.

As discussed earlier, the aggregation source term with a non-linear kernel requires a custom-derived Gaussian quadrature to be integrated exactly. Before taking the effort to derive such a rule, however, it is worth to try first the Newton-Cotes quadrature used in the linear kernels. Table 3.1 shows the conservation of the first moment for both of these kernels, after employing Simpson’s rule for evaluating the aggregation source term. Evidently the error is negligible, and this implementation should be sufficient for most cases. The steep increase in CPU time is due to the need for computing fractional exponents. In fig. 3.7, the distributions resulting from these kernels are compared with those of the sum and constant kernels. Aggregation
seems to be more intense with the Smoluchowski and gravitational kernels, a fact that can be deduced by the form of these kernels.

![Graph showing aggregation, constant, sum, fluid shear and gravitational kernels, simulation results.]

Figure 3.7: Aggregation, constant, sum, fluid shear and gravitational kernels, simulation results.

### 3.5.2 Nucleation-growth

Nucleation-growth problems have been studied extensively by, among others, Randolph and Larson (1988). The dynamic PBE for nucleation-growth with flow is:

\[
\frac{\partial n(v, t)}{\partial t} = \frac{n_{in}(v, t) - n(v, t)}{\tau} - \frac{\partial}{\partial v} (G(v) \cdot n(v, t)) + B\delta(v - v_0)
\]  

(3.51)

As this is a hyperbolic equation, the main challenge for a numerical method is to capture the steep moving fronts and shocks that may emerge. Note that the analytical solutions for nucleation-growth have been derived by incorporating nucleation as a boundary condition \( n(v_0, t) = B_0/G \), while in this method nucleation enters as a point source. Parameters of the simulations are shown in table 3.2.
Case I-1

<table>
<thead>
<tr>
<th>Kernel</th>
<th>$k_a$</th>
<th>$k_a^*$</th>
<th>$k_a(v + v')$</th>
<th>$k_a^*(v + v')$</th>
<th>shear (eq. 23)</th>
<th>gravity</th>
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<td>1</td>
<td>1</td>
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<td>$I_{agg}$, %</td>
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<td>86</td>
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<td>1.5</td>
<td>2</td>
<td>20</td>
</tr>
</tbody>
</table>

Table 3.1: Results for the aggregation cases.

3.5.2.1 Case II-1: Size-independent linear growth and nucleation with flow

Most analytical solutions refer to size-independent growth, for which case eq. 3.51 is simplified to:

$$\frac{\partial n(v, t)}{\partial t} = \frac{n_{in}(v, t) - n(v, t)}{\tau} - G\frac{\partial n(v, t)}{\partial v} + B\delta(v - v_0)$$  \hspace{1cm} (3.52)

Note that in reality, growth can be independent only with respect to a linear size variable, and the case of volume-independent growth is considered here only for validation purposes. At steady-state, the solution to eq. 3.52 is the well-known MSMPR equation (eq. 1.48), written here in terms of volume:

$$n(v, \infty) = \frac{B}{G} \cdot e^{-v/G\tau}$$  \hspace{1cm} (3.53)

Fig. 3.8 shows the time-dependent numerical solution and the steady-state analytical solution. After several residence times, the two curves merge. Comparison is very good, apart from the first two points, whose deviation has been explained above. Table 3.2 shows the numerical parameters employed in this case.

3.5.2.2 Case II-2: Size-dependent growth and nucleation with flow

A commonly used size-dependent growth function is (Randolph and Larson, 1988):
Figure 3.8: Steady-state size-independent nucleation-growth, comparison with the analytical solution of Randolph and Larson (1988).

\[ G = G_0 (1 + \gamma v)^\beta \]  

(3.54)

This function involves the growth rate of nuclei, \( G_0 \), and two empirical parameters. The above authors also provide a steady-state solution for this class of functions:

\[ n(v, \infty) = K \cdot \frac{B}{G} \cdot (1 + \gamma v)^{-\beta} \exp \left[ -\frac{(1 + \gamma v)^{-\beta}}{G_0 \tau \gamma (1 - b)} \right] \]  

(3.55)

where

\[ K = \exp \left[ \frac{1}{G_0 \tau \gamma (1 - b)} \right] \]  

(3.56)

Fig. 3.9 shows the numerical solution at increasing values of the residence time. After 10 residence times, the simulation approximates the analytical solution at steady state.

### 3.5.2.3 Case II-3: Batch Size-independent nucleation-growth

The analytical time-dependent solution to eq. 3.51 for size-independent growth, no flow and no particles present at \( t=0 \), has been derived by Hounslow et al. (1988):

\[ n(v, t) = \frac{B}{G} \cdot H(t - \frac{v}{G}) \]  

(3.57)
Finite element scheme for the time-dependent population balance

where $H(\cdot)$ is the step function. This is a very ill-conditioned problem, as the solution is simply a discontinuity that moves with time. Such a discontinuity is very difficult to capture, and may become a source of numerical instability. The method of Hounslow et al. (1988), being of zero order, proved stable enough, but its prediction of the front was smeared by numerical dispersion. This effect can only be mitigated by the use of a finer grid, but in that method the grid is constrained to follow a geometrical progression. The finite element method, on the other hand, may employ a uniform grid, which is the most suitable for this case.

Fig. 3.10 compares numerical and analytical results. The stability of the method is owning to the first-order trial functions and the upwind discretisation of growth. Dispersion around the discontinuity is inevitable, since this is a first order method, but it can be controlled by refining the grid. A smaller time step was also chosen to mitigate the dispersion, but CPU time is still in the order of seconds.

3.5.3 Breakage

The equation for breakage without flow is:
Finite element scheme for the time-dependent population balance

Figure 3.10: Batch size-independent nucleation-growth, comparison with the analytical solution of Hounslow et al. (1988)

\[
\frac{\partial n(v, t)}{\partial t} = \int_{v'}^{\infty} b(v') b(v/v') n(v', t) dv' - (3b(v) n(v, t))
\]  

(3.58)

The functions \(N_0(v)\) and \(b(v/v')\) denoting the degree of fragmentation and size distribution of daughter particles are often merged into a single function, \(b'(v/v')\). The possibilities for this function are numerous, but in this work we shall consider three representative cases: uniform binary breakage, parabolic binary breakage and multiple fragmentation. Parameters are shown in table 3.3.

<table>
<thead>
<tr>
<th>Case</th>
<th>I-1</th>
<th>I-2</th>
</tr>
</thead>
<tbody>
<tr>
<td>Growth kernel</td>
<td>(G_0)</td>
<td>(G_0(1+\gamma v)) (G_0)</td>
</tr>
<tr>
<td>(B_0)</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>(G_0)</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>(\gamma)</td>
<td>-</td>
<td>1</td>
</tr>
<tr>
<td>(\beta)</td>
<td>-</td>
<td>1</td>
</tr>
<tr>
<td>(t (\tau))</td>
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<tr>
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<td>0.5</td>
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<td>200</td>
</tr>
<tr>
<td>Size range</td>
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<td>0.1-10</td>
</tr>
<tr>
<td>CPU (sec)</td>
<td>1</td>
<td>1.5</td>
</tr>
</tbody>
</table>

Table 3.2: Nucleation-growth cases, parameters and results.
3.5.3.1 Case III-1: Uniform binary breakage

Uniform breakage implies that fragments of every size have the same probability to form. This situation may arise e.g. in polymer chains which, being one-dimensional, are as likely to break at any point. If the breakage is also binary, it can be represented by the following overall function:

\[ b'(v/v') = \frac{2}{v'} \]  

(3.59)

It can be easily confirmed that eq. 3.59 satisfies the constraints given by eqs. 3.9 - 3.11. Ziff and McGrady (1985) derived an analytical time-dependent solution for this case, assuming a first-order size-dependent kernel \( \beta(v) = k_b \cdot v \):

\[ n(v, t) = \exp(-tv^2 - v) \cdot [1 + 2t(1 + v)] \]  

(3.60)

Fig. 3.5.3.1 shows the initial distribution and the distributions resulting from breakage at different times. A comparison between the simulation and the analytical solution shows that the approximation is very good.

Figure 3.11: Batch uniform binary breakage, comparison of simulation and analytical solution of Ziff and McGrady (1985).
### 3.5.3.2 Case III-2: Parabolic binary breakage

Rigid particles are more likely to break into unequal pieces. This can be accounted for via the use of a parabolic "U-shape" function to describe the daughter particle size distribution. An overall function describing parabolic binary breakage has been suggested by Hill and Ng (1995):

\[
b(v, v') = \frac{24}{v'^3} \left(1 - \frac{hv'}{2}\right) \left(v^2 - vv' + \frac{v'^2}{4}\right) + h
\] (3.61)

For \(0 < h < 2/v\), this function represents a concave parabolic profile with decreasing steepness. At \(h = 2/v\) the function is equal to the uniform binary breakage, while for \(2/v < h < 3/v\) the function represents a convex parabola (an unlikely case). Outside this bounds the function assumes unrealistic negative values.

To the authors' knowledge there is no general analytical solution except for \(h=2\) in which case the simulation approximates very well the solution for uniform binary breakage (eq. 3.59); moreover, the first moment is properly conserved in all three cases. As expected, parabolic breakage favours smaller and bigger particle sizes in the expense of the intermediate ones, resulting in a slightly bimodal distribution (fig. 3.5.3.2).

### 3.5.3.3 Case III-3: Multiple breakage (fragmentation)

Ziff (1991) derived analytical solutions for a class of fragmentation problems where the number of fragments and their size distribution are described by:

\[
b'(v/v') = \frac{a \gamma}{v'} \left(\frac{v}{v'}\right)^{\gamma-2} + \frac{(1-a)\delta}{v'} \left(\frac{v}{v'}\right)^{\delta-2}
\] (3.62)

where \(a, \gamma, \delta\) are parameters. The average number of fragments produced can be determined by the following:

\[
\overline{N}(v') = \int_0^{v'} b(v/v')dv
\] (3.63)

Ziff derived complete time-dependent solutions for a subset of this class, defined by:

\[
b'(v/v') = \frac{\delta \lambda}{\delta - \lambda} \left(\frac{v^{\lambda-2}}{v'} - \frac{v^{\delta-2}}{v'}\right)
\] (3.64)

\[
\beta_b(v) = k_b \cdot v^\lambda
\] (3.65)
Finite element scheme for the time-dependent population balance

Figure 3.12: Batch parabolic breakage for \( h=0,1,2 \) and comparison with analytical solution (Ziff and McGrady, 1985) for \( h=2 \).

For a monodisperse initial condition, his solution is:

\[
\begin{align*}
    n(v, t) &= I^{-1} \exp(-tI^\lambda)\delta(v - I) + \lambda \delta tv\delta^2 \int_0^I v^{\lambda-\delta-1} \exp(-tv^\lambda)dv' \\
    n(v, 0) &= I^{-1}\delta(v - I)
\end{align*}
\]  

(3.66)

(3.67)

Fig. 3.5.3.3 shows the numerical and analytical solution the case of ternary fragmentation \((\lambda = 2, \delta = 3)\). A uniform grid was employed for this simulation. To implement the monodisperse initial condition, an initial distribution resembling a spike had to be employed, resulting in a rather ill-conditioned problem; nevertheless solution was easily obtained.

3.5.4 Combined problems

Combined problems (involving combinations of the mechanisms) are the most difficult to tackle, and very few analytical solutions have been found to compare with. At least two such cases are known to the authors.
3.5.4.1 Case IV-1: Steady-state nucleation - aggregation

Hostomsky (1987) derived a semi-analytical solution for steady-state nucleation and size-independent aggregation with flow and no particles in the feed. Under these conditions, the PBE may be written as:

\[
\frac{1}{2} b_a \int_0^v n(v' \to v) n(v') dv' - n(v) b_a \int_0^\infty n(v') dv' + B_0 \delta(v - v_0) - \frac{n(v)}{\tau} = 0
\]  

(3.68)

The solution is:

\[
n(v) = \left( \frac{1 + 2\alpha}{4\pi} \right)^{1/2} \cdot \left[ \frac{\alpha / (\alpha + 1/2)}{b_a \tau v^{3/2}} \right]^{v/v_0}
\]  

(3.69)

where

\[
\alpha = b_a B_0 \tau
\]

Hostomsky’s solution is semi-analytical because its derivation involves a Taylor series expansion, and therefore subject to error. It holds for \( v > 10v_0 \) and is more accurate for higher volumes. Hostomsky provided also an equation for the zeroth moment:

\[
M_0 = \frac{(2b_a B_0 \tau^2 + 1)^{1/2} - 1}{b_a \tau}
\]  

(3.70)
Table 3.3: Breakage cases, parameters and results.

Fig. 3.14 compares this solution with the numerical simulation. A discrepancy in the first points is due to the semi-analytical solution, which does is inaccurate at that range. At $v > 10v_0$ agreement is quite good, and the zeroth moment is also well predicted (table 3.4).

3.5.4.2 Case IV-II: Batch aggregation-growth

This case was studied by Ramabhadran et al. (1976). For a size-independent aggregation kernel, a first-order linear size-dependent growth term ($G = k_g v$) and the exponential initial distribution they derived the following analytical solution:

$$n(v,t) = \frac{N_0}{v_0}(1 - T^2) \exp \left( -\frac{2\Lambda T}{1 - T} \right) \exp \left[ -\frac{v}{v_0}(1 - T) \exp \left( -\frac{2\Lambda T}{1 - T} \right) \right]$$

where:

$$\Lambda = \frac{\sigma_1}{\beta_0 N_0}$$
Figure 3.14: Steady-state aggregation (constant kernel) and nucleation, comparison with semi-analytical solution of Hostomsky (1987).

\[ T = 1 - \frac{2}{2 + \tau/\Lambda} \]

\[ \tau = k_B t \]

Fig. 3.15 shows the analytical and numerical results. A discrepancy in the beginning is, as explained above, due to the deviation of the initial condition used in the simulation from eq. 3.17.

An interesting feature in this case is the speed with which the distribution moves towards the larger sizes, due to a combination of very rapid growth and aggregation. Fig. 3.16 shows the mass distribution moving rapidly towards bigger sizes. A larger grid is required in this case, in order to cover sufficiently the whole range over which the distribution is moving.

3.5.5 Discussion

Our test cases indicated that the CPU time demanded from an average desktop PC for nucleation-growth problems is at about the same order of magnitude as the real time being simulated, while for problems involving aggregation and/or breakage it increases roughly by an order of magnitude due to the integral terms (e.g. a few
Finite element scheme for the time-dependent population balance

Table 3.4: Combined cases parameters and results.

<table>
<thead>
<tr>
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<th>IV-1</th>
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<tr>
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<tr>
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<td>-</td>
</tr>
<tr>
<td>CPU (sec)</td>
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<td>7</td>
</tr>
</tbody>
</table>

minutes of CPU are required for a few seconds of real time). Direct comparison with CPU times reported by other authors is not straightforward because they refer to different computers, but an evaluation can be attempted based on the mechanics of each method. The use of linear elements, in conjunction with our strategy for evaluating the integral terms, makes the number of computations and interpolations substantially fewer than those required for higher-order elements (as shown, for most aggregation kernels just one interpolation is enough). Thus the method is expected to be much faster than higher-order finite element collocation methods, such as cubic spline or orthogonal collocation, while integral formulations such as Galerkin are even more involved. Higher order methods may be more accurate, but the CPU time saved due to the first-order elements can be spent on refining the grid, a practice that is guaranteed to improve accuracy. DPBs, especially those based on a geometric series (such as the Hounslow method) should be even faster, because their way of evaluating the integral terms requires fewer computations. However, the important conclusion is that the CPU demands raised by this new method are affordable even with desktop PCs, while a workstation would be fast enough for any real case.
Figure 3.15: Batch aggregation (constant kernel) and growth (size-dependent), comparison of simulation and analytical solution of Ramabhadran et al., (1976).

Figure 3.16: Evolution of first moment (mass) in the aggregation-growth problem - comparison of simulation and analytical solution of Ramabhadran et al. (1976).
Chapter 4

Gas-liquid precipitation of CaCO$_3$ in a bubble column reactor

4.1 Introduction

In this chapter, the theoretical and computational methods that were developed so far will be applied to the analysis of gas-liquid precipitation. Being a complex process whose outcome is determined by the interplay of fluid dynamics, chemical reactions and particulate processes, gas-liquid precipitation is a case study that fully demonstrates the need for a hybrid modelling approach. Moreover, accurate population balance modelling is essential in order to predict the crystal size distribution (CSD) of the precipitate, thus granting potential for product engineering.

4.1.1 Industrial importance

Gas-liquid precipitation is the formation of a solid crystalline product via chemical reactions involving a gas and a liquid phase. A variety of products, mostly inorganic crystals and speciality chemicals, are produced in this way. Wachi and Jones (1992) listed several such processes:

- Ammonium phosphate ((NH$_4$)$_2$HPO$_4$), from gaseous NH$_3$ and phosphoric acid (H$_3$PO$_4$)
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

- Ammonium sulphate ((NH₄)₂SO₄), from NH₃(ɡ) and sulphuric acid (H₂SO₄)
- Barium carbonate (BaCO₃), by carbonation of either barium hydroxide (Ba(OH)₂) or barium sulphate (BaSO₄)
- Calcium carbonate, by carbonation of lime (Ca(OH)₂) solution
- Calcium fluoride (CaF₂) by calcium nitrate (CaNO₃) and hydrogen fluoride (HF)
- Gypsum (CaSO₄), from sulphur oxide (SO₂) and lime (Ca(OH)₂)
- Geothite (FeOOH) by air oxidation of alkaline suspension of ferrous hydroxide (Fe(OH)₂)
- Potassium chloride (KCl), by ammoniation of sea bitterns containing chlorides of potassium, sodium and calcium
- Sodium bicarbonate (NaHCO₃) by carbonation of salt and ammonia
- Strontium carbonate (SrCO₃) by carbonation of strontium hydroxyde (Sr(OH)₂)
- Terephthalic acid (COOH(C₆H₄)COOH) by air oxidation of p-xylene in acetic acid with cobalt and manganese salts of metal bromide at 200 °C and 400 psi. The precipitated crystals are used for the manufacture of Dacron™ and Terylene™.

Despite its industrial importance (Jones, 2002), gas-liquid precipitation has only recently become the objective of rigorous study. A comprehensive review of the state of the art has been provided by Wachi and Jones (1995). The objective of the analysis and mathematical modelling is to provide means of controlling the precipitate properties, such as the crystal size distribution (CSD), by adjusting the process design and operation parameters.

4.1.2 Present challenges

Gas-liquid reactive precipitation was first studied in laboratory-scale reactors (fig. 4.1, (b) with a flat gas-liquid interface. Experimental and modelling studies on such equipment have been conducted by Yagi and co-workers (1986), Wachi and Jones...
Gas-liquid precipitation of CaCO$_3$ in a bubble column reactor

(1991) and Hostomsky and Jones (1995). Other laboratory-scale devices that have been employed, mainly for investigating the kinetics of gas-liquid precipitation, are an agitated tank with the gas being dispersed in the form of bubbles (Yagi, Iwazawa, Sonobe, Matsubara and Hikita, 1984; Yagi, 1990; Kotaki and Tsuge, 1990) and the Couette-Taylor reactor (Jung, Kang, Kim and Choi, 2000). The latter was employed for studying of morphology of precipitated CaCO$_3$ by careful control of solution composition and shear.

Industrial reactors for gas-liquid precipitation are shown in fig. 4.1 (a,c,d). A proper analysis of such equipment requires consideration of hydrodynamics, which determine several key reaction engineering parameters such as the mass transfer rate. Furthermore, prediction of the CSD can be accomplished by solution of the population balance equation. The coupling of the above requires excessive computational resources, and it has been realised (Wachi and Jones, 1995) that a stagewise model should be employed rather than a differential one.

The complexity is further increased by the fact that such processes are usually operated in batch or semi-batch mode, which implies that dynamic modelling is required. As the concentrations of the reacting species vary with time, supersaturation levels and crystallisation rates also vary. Consequently the population balance is strongly coupled with the reacting species’ mass balances, thus ruling out analytical solutions even in the simplest cases (e.g. nucleation-growth, no agglomeration).

4.1.3 Modelling of gas-liquid precipitation with the method of moments

Earlier work (Rigopoulos and Jones, 2001) employed a stage-wise dynamic model for the simulation of a tall bubble column carrying out CaCO$_3$. Gas-liquid mass transfer and chemical reaction was accounted for via an implementation of the penetration theory similar to the one in this work. In fact, that model can be viewed as a specific case of the generalised compartmental model presented in chapter 2. Fig. 4.2 shows the representation of the bubble column in that work.

The population balance was formulated with crystal diameter as the independent coordinate (eq. 1.49), and coupled with the mass balances of the chemical species within each fully-mixed compartment. Mean crystal size was predicted with the
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

Figure 4.1: Reactors for gas-liquid precipitation - (a) bubbling stirred tank, (b) flat interface stirred tank, (c) draft tube bubble column, (d) spray column (reprinted from Jones, 2002)
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

Figure 4.2: Stage-wise model for a draft-tube bubble-column reactor (Rigopoulos and Jones, 2001)

method of moments. This method, discussed in sec. 1.4.5.2, leads to a closed set of equations when nucleation and growth only are considered; it is not suitable for incorporating inter-particle events such as agglomeration and breakage.

The final model consisted of the following set of ODEs, describing the evolution of the moments and chemical species’ concentrations within each compartment:

\[
\frac{d n_i^j(t)}{dt} = \frac{m_{i-1,j} - m_i(t)}{\tau} + G(C_i(t)) \cdot j m_{i-1}(t) \quad (4.1)
\]

\[
\frac{d C_i^j(t)}{dt} = \frac{C_{i,i-1} - C_i^j(t)}{\tau} + \sum_{k=1}^{K} r_k(C_{1,i}(t), \ldots C_{n,i}(t)) \quad (4.2)
\]

The zeroth, first, second and third moments were resolved. These quantities, as explained in sec. 1.4.5.2, denote the total number, length, surface area and volume of the crystals and comprise the minimum set of moments that leads to a closed system.

4.1.4 Deficiencies of the moments solution

Comparison with experimental results from an earlier work by Jones, Wachi and Delannoy (1992) is shown in fig. 4.3. The main trends were adequately predicted, but several questions such as the rapid increase in mean size after the pH drop, possibly attributed to agglomeration, were left unanswered. This was due to the fact that both the experimental results and model predictions reported only mean values of the CSD.

Agglomeration seems to be pronounced during gas-liquid precipitation, as the
experiments of Kotaki et al. (1990) have shown. It has a profound effect in the CSD, and may determine the physical properties of the end product as well as its potential for separation; depending on the application, either its promotion or prevention may be desirable. Its mathematical analysis is complex, however: it requires detailed solution of the population balance, and moreover no definitive kinetic model is available.

In order to answer these questions new experiments of CaCO₃ precipitation are reported here, during which the entire CSD is retrieved via laser light scattering. In addition, the conservative finite element discretisation method of the dynamic volume-based population balance that was developed in chapter 3 will be employed to simulate the evolution of the CSD. Thus deeper insight can be drawn into the gas-liquid precipitation process and the underlying mechanisms; furthermore this process, encompassing a variety of phenomena and characteristic scales, provides an ideal vehicle for testing the theoretical methods developed in the preceding chapters.
4.2 Experimental part

The objective of this section is to describe the equipment and analytical methods employed in the experimental work, and to draw several clues regarding the precipitation mechanism from a preliminary inspection of the experimental results. These clues will prove to be of vital importance for configuring the kinetic model of CaCO₃ precipitation. It must be stressed that the present work does not constitute an exhaustive study of CaCO₃ precipitation kinetics. Such a study would require isolation of individual mechanisms, such as growth or agglomeration, which is not possible in a bubble column, an equipment whose simplicity of design gives rise to extraordinary complexity in the physical and chemical phenomena. A large body of literature has been devoted to that problem, employing special equipment such as the Couette-Taylor (Jung, Kang, Kim and Choi, 2000) and the Poiseuille flow crystalliser (Mumtaz and Hounslow, 2000). Rather than this, our purpose is to explore whether the complex hydrodynamic, reaction engineering and crystallisation phenomena taking place in an industrial type of reactor, such as the bubble column, can be investigated with the aid of a mathematical model. Precipitation kinetics tend to be system-dependent and unpredictable, and virtually no systematic study has been done on bubble columns; literature values will be used wherever possible, and a semi-empirical procedure for adjusting the remaining parameters will be outlined.

4.2.1 Experimental setup

4.2.1.1 Gas-liquid precipitation rig

The experimental work was conducted using a bubble column reactor, equipped with an internal concentric draft tube to enhance circulation and mixing. The outer cylinder was 0.77 m high, with internal diameter 0.215 m; the corresponding dimensions of the draft tube were 0.45 m and 0.135 m. Gas (a mixture of CO₂ and N₂) was injected through a sparger ring situated at the bottom of the inner annulus. In each experiment, the column was filled with approximately 21 lt of Ca(OH)₂ solution (Analar grade Ca(OH)₂ - deionised water). Before commencing the experiment, pure N₂ gas was being injected for some time to enhance the dissolution of the Ca(OH)₂ by intense circulation and mixing.
Since the aim was to observe the dynamic behaviour of the process, during the course of each experiment regular samples were taken and analysed immediately for pH, concentration of Ca\(^{++}\) and crystal size distribution. The CSD analysis was carried out with a laser scattering instrument (Malvern Mastersizer S), equipped with lenses valid for measuring particles in the range 0.05 - 900 μm. The measurements were essentially off-line, but since the CaCO\(_3\) suspension could be examined in the Mastersizer without any need for further processing and each measurement took only a few seconds to complete, no significant changes in the CSD could occur during the sampling procedure. To gain more insight into the crystal structure and morphology, selected samples were dried and further analysed with Scanning Electron Microscopy (JEOL JSM-820). These analytical methods will be further discussed below.

### 4.2.1.2 Measurement of crystal size via laser light scattering

The determination of particle size by laser scattering is based on the interaction between an incident light source and the solid particles. The amounts of reflection, refraction, and absorption of the incident wave depend on the refractive index of the particle, its geometry, and the wavelength of the incident light (Rawlings et al., 1993). By the use of Fourier and Reverse Fourier optics, this scattering is
Gas-liquid precipitation of CaCO$_3$ in a bubble column reactor

Imaged to an array of detectors at the focal plane of the optics. There is a direct relationship between the distribution of the scattered light energy on these detectors and the particle size distribution which gives rise to it. This method has become the preferred standard in many industries for characterization and quality control. The applicable range according to ISO 13320 is 0.13000μm (Malvern Instruments, 2001).

Instruments consist of:

- A laser as a source of coherent intense light of fixed wavelength. He-Ne gas lasers (=0.63m) are the most common as they offer the best stability (especially with respect to temperature) and better signal to noise than the higher wavelength laser diodes. It is to be expected when smaller laser diodes can reach 600nm and below and become more reliable that these will begin to replace the bulkier gas lasers.

- A suitable detector. Usually this is a slice of photosensitive silicon with a number of discrete detectors. It can be shown that there is an optimum number of detectors (16-32) increased numbers do not mean increased resolution. For the photon correlation spectroscopy technique (PCS) used in the range 1nm to 1m approximately, the intensity of light scattered is so low that a photomultiplier tube, together with a signal correlator is needed to make sense of the information.

- Some means of passing the sample through the laser beam. In practice it is possible to measure aerosol sprays directly by spraying them through the beam. This makes a traditionally difficult measurement extremely simple. A dry powder can be blown sprayed into the environment. Particles in suspension can be measured by recirculating the sample in front of the laser beam.

Older instruments and some existing instruments rely only on the Fraunhofer approximation which assumes that the particle is much larger than the wavelength of light employed (ISO 13320 defines this as being greater than 40λ i.e. 25m when a 625nm He-Ne laser is used), that all sizes of particle scatter with equal efficiencies and that the particle is opaque and transmits no light. These assumptions are never correct for many materials and for small material they can give rise to errors.
approaching 30% especially when the relative refractive index of the material and medium is close to unity. When the particle size approaches the wavelength of light the scattering becomes a complex function with maxima and minima present. The latest instruments, such as the Mastersizer S used in this work, use the full Mie theory which completely solves the equations for interaction of light with matter. This allows completely accurate results over a large size range (0.02 – 2000μm typically - 0.05 – 900μm in the case of the Mastersizer S). The Mie theory assumes the volume of the particle as opposed to Fraunhofer which is a projected area prediction. It is valid for all particulate materials in all transparent media liquids, gases or even transparent solids (Malvern Instruments, 2001).

4.2.1.3 Examination of crystal morphology via Scanning Electron Microscopy

Electron microscopy is a rather straightforward technique to determine the size and shape of microscopic particles. It can also reveal information on their composition and internal structure, e.g. by detecting the characteristic x-rays that are produced by the interaction of the electrons with matter, or by analysing how the electrons are diffracted. Three types of electron microscope are commonly encountered: transmission electron microscopy (TEM), scanning electron microscopy (SEM) and scanning transmission electron microscopy (STEM) which combines the two previous modes of operation.

In this work, SEM technology was employed for exploring crystal morphology. SEM is carried out by scanning the surface via a narrow electron beam and detecting the yield of either secondary or backscattered electrons as a function of the position of the primary beam. Contrast is caused by the orientation: parts of the surface facing the detector appear brighter than parts of the surface with their surface normal point away from the detector. The secondary electrons have mostly low energies (5–50eV) and originate from the surface region of the sample. Backscattered electrons come from deeper and carry information on the composition of the sample, because heavy elements are more efficient scatterers and appear brighter in the image. Dedicated SEM instruments have resolutions of about 5 nm. The main difference between SEM and TEM is that SEM sees contrast due to the topology and composition
of a surface, whereas the electron beam in TEM projects all information on the mass it encounters in a two-dimensional image, which, however, is of sub-nanometer resolution (Niemantsverdriet, 1993).

4.2.2 Results and discussion

4.2.2.1 Time course of precipitation

Typical plots of the temporal evolution of pH, [Ca\(^{++}\)], mean crystal size and CSD are shown in Figs. 4.5, 4.7, 4.6. The CSD is plotted as:

\[
\begin{align*}
\text{abscissa} & : \frac{\text{volume \% in interval}}{\text{interval length (diameter unit)}} \quad \left[\frac{1}{\mu m}\right] \\
\text{ordinate} & : \text{particle diameter} \quad [\mu m]
\end{align*}
\]

The main features may be summarised as:

- Most of the precipitation occurs during a period of high pH (11-12), with [Ca\(^{++}\)] being consumed at a slightly decaying rate. This period is followed by a rapid fall in the pH, while [Ca\(^{++}\)] has reached a minimum. Subsequently, [Ca\(^{++}\)] increases again, indicating re-dissolution of some of the crystal mass.

- After an induction period, the mean crystal size increases at a slow rate. As soon as the pH drops a rapid increase occurs, eventually stabilising around 5 \(- 6\mu m\).

- The first CSD measurement (immediately after the induction period) is dominated by a single bell-shaped peak around 1\(\mu m\). The peak becomes gradually sharper, with a tail on its right. At \(t=240\) sec a new peak appears at the size range 0.1\(\mu m \) \(- 1\mu m\). At the end of the precipitation, signified by the pH drop, the bimodal distribution remains.

4.2.2.2 Product morphology

Fig. 4.8 shows a large-scale view of the precipitate at an early stage of the process (soon after the end of the induction time), evidently comprising mostly single crystals. A closer view (figs. 4.9, 4.10) reveals that rombohedric calcites form the majority of the product crystals. Occasional agglomerates are also present, and in
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

Figure 4.5: Evolution of mean size during a typical experiment

fig. 4.11 we get a close view at one of them. It is interesting that agglomerates appear before the pH drop, indicating that high pH does not completely prevent agglomeration, even if it has an inhibitory effect. The initial agglomerates have sharp, clearly defined edges, and the single crystals that comprise them can be clearly identified.

At a later stage, after the pH drop, the product is dominated by agglomerates (figs. 4.12, 4.13), confirming the increase in mean size. Meanwhile the edges of the agglomerates become smoother and smoother (fig. 4.14), until we reach a stage where only agglomerates exist, exhibiting an almost amorphous shape (fig. 4.15). At the same time the concentration of the precipitate is evidently lower.

4.2.3 Interpretation of results

The equilibrium of the CO₂ - OH⁻ system must be born in mind when attempting to interpret these results. In chapter 2, we plotted (fig. 2.15) the equilibrium concentrations of [CO₃²⁻], [HCO₃⁻] and [CO₂,aq], the prevailing species at the alkaline pH region, calculated with the data of Astarita (1967). CO₃²⁻, which contributes to the precipitation process, is present at a significant proportion only at pH > 10. At intermediate pH, HCO₃⁻ prevails; later, only [CO₂,aq] is present.

The increase in mean size after the pH drop is in accordance with previous
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

Figure 4.6: Evolution of CSD during a typical experiment
Gas-liquid precipitation of CaCO$_3$ in a bubble column reactor

Figure 4.7: Evolution of pH and Ca$^{++}$ during a typical experiment

Figure 4.8: Large-scale view of the precipitate, shortly after the induction time
Figure 4.9: A closer view of the single product crystals (a)

Figure 4.10: A closer view of the single product crystals (b)
Gas-liquid precipitation of CaCO$_3$ in a bubble column reactor

Figure 4.11: Close view of an agglomerate formed at an early stage of the process.

Figure 4.12: View of the precipitate at a later stage of the process (a).
Figure 4.13: View of the precipitate at a later stage of the process (b)

Figure 4.14: Close view of an agglomerate, at a later stage of the process
experiments (Jones, Wachi and Delannoy, 1992; Wachi and Jones, 1995) and was previously explained in the following way: During the high pH stage particles experience slow growth, while agglomeration is being inhibited by the electrical charge on the particles' surface. Once the pH has fallen, the mean size increases rapidly due to agglomeration (Wachi and Jones, 1991). This interpretation, however, leaves an important question unanswered: how does agglomeration occur after the depletion of supersaturation, which is necessary to create solid bridges and unify the loose aggregates. Furthermore, the experiments of Kotaki and Tsuge (1990) on the same system clearly evidenced agglomeration at high pH; our own experiments also detected agglomerates during early stages (fig. 4.11).

The evolution of crystal shape provides vital clues for the understanding of the process. The loss of the sharp edges in the agglomerates, coinciding with the reappearance of free Ca^{++} after the pH drop suggests that the crystals are redissolving. Edges exhibit more rapid dissolution, thus leading to the almost amorphous material of fig. 4.15. The redissolution can be explained by the shift of the equilibrium from CO_3^{2-} towards HCO_3^- . At high pH, when CO_3^{2-} is favoured, a balance is maintained between the migration of ions from the crystal lattice to the surrounding solution

Figure 4.15: An agglomerate during a very late stage of the process.
and deposition of ions in the lattice. After the pH drop, however, the CO$_3^{2-}$ that migrates is converted to HCO$_3^-$ and, as a result, the balance is disrupted. The overall dissolution process can be described by the following reaction:

\[
CaCO_3 + H_2O \leftrightarrow Ca^{++} + HCO_3^- + OH^-
\]

The dissolution of a crystal is proportional to its surface area. Smaller crystals with a higher surface area to volume ratio are going to disappear faster, leaving only the larger agglomerates in the system. This explains the fact that agglomerates dominate the precipitate after pH drop, despite the fact that supersaturation has been depleted. The increase in the mean crystal size after the pH drop is therefore only an apparent growth.

Bimodality in the CSD is usually explained by agglomeration or breakage. Though agglomerates have indeed been detected in the precipitate through SEM, the fact that the second peak appears at the small size range indicates that agglomeration is not the cause. Disruption of the weak aggregates is another candidate, but it must be ruled out by the fact that the new peak appears at sizes smaller than the size of the single crystals that comprise the agglomerates ($\approx 1 \mu m$).

The crystals comprising the new peak must, therefore, be produced by a new wave of nucleation-growth. This could be the effect of secondary nucleation, which is proportional to the existing crystal mass (Botsaris, 1976; Garside and Davey, 1980). Size-independent growth imposes a third-order polynomial increase in crystal mass with respect to time. It is reasonable, then, that at some point a sharp increase in crystal mass will rapidly generate a large amount of secondary nuclei, thus inducing a new wave of crystal growth.

To confirm this speculation, an experiment was conducted during which the introduction of CO$_2$ was halted just before the new peak was expected to appear. N$_2$ continued to be introduced, maintaining the total gas flowrate constant, to ensure that the turbulent conditions would remain unchanged. After 30 min the distribution remained unchanged (i.e. single peak), indicating that disruption or breakage alone of the existing crystals does not account for the new peak. As soon as CO$_2$ was introduced again the new peak appeared, thus proving that it consists of freshly grown particles and does not occur without the presence of supersaturation.
4.3 Process simulation

4.3.1 Kinetics of CaCO₃ precipitation

4.3.1.1 Reaction system

The reaction of CO₂ into solutions of alkalis has been studied extensively, as it is both a commonly encountered process (gas cleaning, carbonate salts production, geophysical precipitation) as well as a classical example of the interaction between mass transfer and chemical reactions. For that reason it was used in chapter 2 to demonstrate the methodology for gas-liquid reactor modelling. Its kinetics will be examined in more detail in this chapter, in order to form a detailed kinetic model of CaCO₃ precipitation. Extensive reviews of this reaction system have been contributed by Danckwerts and Sharma (1966), Astarita (1967), Danckwerts (1970) and Juvecar and Sharma (1972).

The first step to consider in a CO₂ absorption system is the gas-liquid equilibrium, which determines the boundary condition for the volatile species:

\[ \text{CO}_2(g) \rightarrow \text{CO}_2(aq) \quad [i] \]

It is now generally accepted that the main resistance to mass transfer lies in the liquid phase (Juvekar and Sharma, 1972). Thus, step 1 can be approached by an equilibrium equation such as Henry's law:

\[ Y_{CO_2} \cdot P = \mathcal{R} \cdot [CO_2(aq)] \quad (4.3) \]

In a system including several ionic species, however, it is advisable to take into account the influence of ionic strength (Danckwerts, 1970).

\[ \log \frac{H}{H_w} = - \sum_i J_i h_i \quad (4.4) \]

\[ h_i = h_+ + h_- + h_g \quad (4.5) \]

The values for the contributions of cations (h⁺), anions (h⁻), gas (h₉) and compounds (hᵢ) are taken from Pohorecki and Moniuk (1988).

The kinetics of CO₂ absorption into alkali solutions are determined by the conversion of CO₂(aq) into HCO₃⁻, a reaction that proceeds at a fast, but finite rate.
This is followed by an instantaneous ionic reaction:

\[
CO_{2(aq)} + OH^- \rightleftharpoons HCO_3^-
\]  \[\text{[ii]}\]

\[
HCO_3^- + OH^- \rightleftharpoons CO_3^{2-}
\]  \[\text{[iii]}\]

The precipitation of CaCO_3 may be written concisely as:

\[
Ca^{++} + CO_3^{2-} \rightleftharpoons CaCO_3(s)
\]  \[\text{[iv]}\]

The differential equations describing the evolution of concentrations can be written as:

\[
\frac{d[CO_{2(aq)}]}{dt} = r_1 - r_2
\]  \[4.6\]

\[
\frac{d[OH^-]}{dt} = r_2 - r_3
\]  \[4.7\]

\[
\frac{d[HCO_3^-]}{dt} = r_2 - r_4
\]  \[4.8\]

\[
\frac{d[Ca^{++}]}{dt} = r_4
\]  \[4.9\]

The rate \( r_1 \) is the rate of mass transfer. It can be calculated from the penetration theory according to the methodology presented in Chapter 2: after the end of each contact time, the new concentration profiles in the interface are integrated along the penetration depth.

\( r_2, r_3 \) are the kinetic rates of steps [ii] and [iii] respectively, given by the following equations:

\[
r_2 = k_2[CO_{2(aq)}][OH^-] - \frac{k_2}{K_2}[HCO_3^-]
\]  \[4.10\]

\[
r_3 = k_3[HCO_3^-][OH^-] - \frac{k_3}{K_3}[CO_3^{2-}]
\]  \[4.11\]

The values of the kinetic and equilibrium constants are due to Astarita (1967), are shown in table 2.1. The kinetic constant of the instantaneous reaction (iii) is approximated by a very high value. The ionic equilibrium of the CO_2 - OH^- system may involve several ionic species, depending on the pH. In this work we are interested in the alkaline range, where CO_3^{2-}, which contributes to the precipitation process, is present at a significant proportion. Fig. 2.15 in chapter 2 shows the molar fractions of CO_{2(aq)}, HCO_3^-, CO_3^{2-} at equilibrium conditions, as a function of pH, obtained by for different values of pH in the range 7-13.
Gas-liquid precipitation of CaCO$_3$ in a bubble column reactor

Expressing the precipitation step by means of the chemical equation (iv) obscures the underlying crystallisation mechanisms that determine its rate. The rate of CaCO$_3$(s) production is essentially that of volumetric crystal growth, but this is size-dependent even when the linear growth is size-independent (McCabe's $\delta L$ law). To obtain the rate of change for the whole crystal mass, we need to integrate the volumetric growth function over the whole range of individual crystal volumes:

$$r_A = \frac{d[CaCO_3(s)]}{dt} = \int_{v_0}^{v_\infty} G(v)n(v) dv \cdot \frac{p_{CaCO_3}}{MW_{CaCO_3}} \quad (4.12)$$

Note how the CSD enters this expression, in the form of the population density $n(v)$. Thus the rate of crystal mass production is inherently coupled with the population balance, and to estimate it we need a complete kinetic model of precipitation that accounts for all crystallisation mechanisms, namely nucleation, crystal growth and interparticle phenomena such as agglomeration and breakage. In this work, the population balance will be discretised with the aid of the conservative finite element scheme developed in chapter 3. That method transforms the integro-partial differential PBE into a series of ordinary differential equations, in terms of the nodal values of population densities (eq. 3.39). Thus the final system of mass and population balance ODEs to be solved takes the form:

$$\frac{dn(v_i,t)}{dt} = f_1(n(v_1), n(v_2)...n(v_n), C_1, C_2,...C_n) \quad (4.13)$$

$$\frac{dC_i(t)}{dt} = f_2(n(v_1), n(v_2)...n(v_n), C_1, C_2,...C_n) \quad (4.14)$$

The next task is to select the kinetic laws to describe the rates of the crystallisation mechanisms.

### 4.3.1.2 Supersaturation

As we saw in chapter 1, most of the crystallisation mechanisms are heavily dependent on superaturation. This is particularly true of precipitation, where the high degree of supersaturation gives rise to rapid nucleation and growth rates. The role of supersaturation dependency is pronounced in batch experiments, where the species' concentrations vary with time. Consider a general ionic reaction of the type:

$$\alpha A^{\beta^+} + \beta B^{a^-} \rightleftharpoons A_\alpha B_{\beta(s)}$$
Among the various measures of supersaturation reviewed in chap. 1, the most appropriate one for ionic precipitation processes is the relative supersaturation, $\lambda_s$:

$$\lambda_s = \frac{C_A C_B}{K_{sp}}$$

(4.15)

where $K_{sp}$ is the solubility product:

$$K_{sp} = [A^{\beta+}]^\alpha [B^{\alpha-}]^\beta$$

(4.16)

It is generally agreed that the driving force of supersaturation is best related to $\alpha + \beta \sqrt{\lambda_s} - 1$ (Bhatia, 1989). Furthermore, in an ionic solution, it is preferable to express supersaturation in terms of activities (Chakraborty and Bhatia, 1997; Collier and Hounslow, 1999). Thus the measure of supersaturation to be used in this work is:

$$\sqrt{\lambda_s} - 1 = \sqrt{\frac{\alpha_{Ca^{\beta+}} \alpha_{CO^{\alpha-}}}{K_{sp}}} - 1$$

(4.17)

Activities are calculated according to the Debye-Huckel equation, with a modification proposed by Davies (1962):

$$\alpha_i = \gamma_i \cdot C_i$$

(4.18)

$$\log(\gamma_i) = -A_{DH} z_i^2 \left( \frac{\sqrt{I}}{1 + \sqrt{I}} - 0.3I \right)$$

(4.19)

where $I$ is the ionic strength, given in terms of the concentration and charge on each ionic species:

$$I = \frac{1}{2} \sum_i C_i z_i^2$$

(4.20)

4.3.1.3 Crystallisation kinetics

The so-called inverse problem in crystallisation, i.e. the determination of crystallisation kinetics from CSD data, is very difficult to solve due to the fact that several mechanisms (nucleation, primary and secondary, growth, agglomeration, breakage) may be simultaneously active, and all of them are non-linear functions of supersaturation. The situation becomes more difficult in a batch experiment, where different mechanisms may prevail during different time periods. Most methods for kinetics extraction are based on steady-state experiments and study of a single mechanism in isolation.
Gas-liquid precipitation of \( \text{CaCO}_3 \) in a bubble column reactor

Growth can be isolated by means of carefully conducted experiments, e.g. by using seeds and maintaining supersaturation within the metastable zone to suppress nucleation. Crystal growth of calcite has been extensively studied, and relative consensus seems to exist. The growth of an ionic crystal, such as \( \text{CaCO}_3 \), may be limited by either diffusion of ions to the crystal surface, or their subsequent reaction and integration into the lattice. If the precipitated crystals are small (< 20\( \mu \)m), sparingly soluble and in a well mixed suspension, the diffusion is bound to be very fast and the surface integration is usually the rate limiting step (Collier and Hounslow, 1999). Several researchers (Hostomsky and Jones, 1991; Collier and Hounslow, 1999; Tai and Hsu, 2001) have confirmed that in precipitation of calcite surface integration is indeed the rate limiting step. The following expression is usually employed to describe growth rate kinetics:

\[
G = k_g (\lambda_s - 1)^2
\]

(4.21)

The experiments of the above authors seem to agree on the order of magnitude and, to a certain degree, on the values of the parameter \( k_g \). The value used in this work is that of Collier and Hounslow (1999), shown in table 4.1.

Nucleation, on the other hand, is much more difficult to study due to the variety of mechanisms (homogeneous, heterogeneous, secondary) that may be responsible for it. A review of these mechanisms has been presented in chapter 1. Its kinetics are usually studied by examining the number of crystals produced during the induction time. The more rigorous classical theories of homogeneous nucleation are seldom applicable to industrial equipment, and empirical correlations are usually preferred. Primary nucleation depends mainly on supersaturation, and a power law of the following form is commonly encountered:

\[
B_0 = k_{n1} (\lambda_s - 1)^{kn2}
\]

(4.22)

Secondary nucleation is induced by the existing crystals (Botsaris, 1976; Garside and Davey, 1980), and power laws describing it must involve the crystal mass as well:

\[
B_0 = k_{ns}(\lambda_s - 1)^{pn1}M_c^{pn2}
\]

(4.23)

Although several studies on calcium carbonate have appeared (Pakter, 1968; Swinney, Stevens and Peters, 1982; Koutsoukos and Kontoyannis, 1983), their results
cannot be extrapolated to equipment of different type and scale, where nucleation may occur via a different mechanism. Moreover, relatively few data have been reported on gas-liquid systems. Most important is the fact that during a batch process, different nucleation mechanisms may prevail during different periods. In our case, the experiments have indicated that in the beginning high supersaturation levels induce primary nucleation, but later, secondary nucleation gives rise to a new wave of crystal growth. The two mechanisms are fundamentally different, and may operate independent of each other. To account for these facts, our overall nucleation model consists of the sum of the two models, primary and secondary:

\[ B_0 = k_{n1}(\lambda_s - 1)^{kn2} + k_{n3}(\lambda_s - 1)^{kn4}M_c^{kn5} \] (4.24)

The determination of agglomeration kinetics is confronted with the problem of distinguishing it from growth. Usually apparent growth is reported, i.e. the combined effect of growth and agglomeration. Bramley et al. (1996) showed that, in the case of size independent growth, agglomeration can be decoupled from growth, but their method applies to MSMPR experiments.

In chapter 1 agglomeration models were reviewed. It was concluded that no mechanistic model has so far succeeded in accounting for both aspects of the agglomeration process, i.e. the formation of flocculates due to collisions and interparticle attraction and the subsequent growth of crystalline material between the clusters at spots called cementing sites. The main obstacles were indentified to be the following:

- Particles of various states (floculates, aggregates of various ages, firm agglomerates) are likely to be present in a crystal agglomerating system. CSD measurements, however, give no information on these states.

- The interparticle phenomena that modify the CSD are also various. Unsuccessful collisions, disruption of aggregates and breakage of firm agglomerates are all possible, depending on the conditions, and it is very difficult to distinguish between these events.

- Current mechanistic models tend to consider an aggregate as a two-particle system, bridged at a cementing site; in reality, however, it may be a multi-particle cluster with bridges of various ages and strengths.
The size-independent kernel seems to be in a good agreement with most agglomeration experiments in stirred tanks, and is used extensively. This agreement, however, is likely to be due to mutual cancellation of competing mechanisms, and we do not know whether it would be maintained in an arbitrary environment.

As a result, agglomeration kinetics are heavily system-dependent and difficult to scale-up, let alone apply to different kinds of equipment. The dependency of the agglomeration kernel on hydrodynamics has been investigated mainly in stirred tanks (Hounslow et al., 2000), whose flow field is very different from that of bubble columns. For this reason, no explicit dependency on hydrodynamics will be employed here; instead, the agglomeration constant will be adjusted to fit experiments with different gas flowrates.

Among the most common theoretical agglomeration kernels, we found the fluid shear kernel (orthokinetic aggregation) to provide the best fit of our experiments. This is in good agreement with the theory, but in contrast with most of the experimental data on stirred tanks, which showed the size-independent kernel to be more suitable (Hostomsky and Jones, 1991; Hounslow et al., 2000). It must be born in mind, however, that when agglomeration is the only inter-particle event in the population balance, the results refer to the net effect of an agglomeration-disruption equilibrium. Stirred tanks exhibit higher local shear forces, and therefore disruption of the weak, not fully cemented aggregates (especially the bigger ones) is likely to occur more often, yielding an apparent size-independent agglomeration. On the other hand, the more gentle agitation in the bubble column may be the reason why the effect of the size-dependent kernel is brought out.

MSMPR experiments (Hounslow et al., 2000) have shown agglomeration to be roughly proportional to growth. The underlying cause of both phenomena is supersaturation, and during a batch experiment supersaturation is varying. It seems appropriate, then, to correlate agglomeration to supersaturation directly. Given the second order dependency of growth on supersaturation, we may infer that a similar dependency to the agglomeration model. Our formulation thus is:

$$\beta_a(v', v - v') = k_a(\lambda - 1)^2[v'^{\frac{1}{2}} + (v - v')^{\frac{1}{2}}]^3$$ (4.25)
Apart from the growth kinetics, the remaining parameters must be adjusted to fit the experiments. This is especially difficult in a batch experiment, where different mechanisms may prevail at different time periods, and each mechanism is a non-linear function of supersaturation. In our study the determination of the parameters was guided by: a) the mean size of the crystals immediately after the induction period, which is determined by the balance between primary nucleation and growth, b) the evolution of of $Ca^{++}$ concentration, and c) the appearance of the second peak at the low size range, which indicates the effect of secondary nucleation.

<table>
<thead>
<tr>
<th>Parameter, units</th>
<th>Value</th>
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<tr>
<td>Solubility product, $K_{sp}$, mol$^2$ m$^{-6}$</td>
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<tr>
<td>Growth const., $k_g$, m/s</td>
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<tr>
<td>$k_n_1$ (eq. 4.24), s$^{-1}$</td>
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<td>$k_n_2$ (eq. 4.24)</td>
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<tr>
<td>$k_n_3$ (eq. 4.24), kg$^{-1}$</td>
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<td>$k_n_4$ (eq. 4.24)</td>
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<tr>
<td>$k_n_5$ (eq. 4.24)</td>
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<tr>
<td>Agglomeration constant, $k_a$, m$^3$/s, at $0.6 \cdot 10^{-4}$ m$^3$/s gas flowrate</td>
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<tr>
<td>Agglomeration constant, $k_a$, m$^3$/s, at $1.1 \cdot 10^{-4}$ m$^3$/s gas flowrate</td>
<td>1.8</td>
</tr>
</tbody>
</table>

Table 4.1: Parameter values used in the simulations.

### 4.3.2 Modelling methodology

Modelling precipitation in a gas-liquid reactor requires sufficient knowledge of its hydrodynamics. A direct coupling of chemical reactions and the population balance with a CFD code would raise extreme computational demands, however; so far, it has been attempted only with the method of moments (e.g. Wei and Garside, 1997). Moreover, distributed information about the fluid dynamic variables is excessive for a reaction engineering analysis of the process. Usually, knowledge of several key parameters such as gas hold-up and liquid circulation averaged over regions or in the whole reactor, is sufficient.

In this work, the problem will be approached with the aid of theoretical methods developed in the preceding chapters. In chapter 2, we presented a methodology for integrating hydrodynamics with complex chemical phenomena that cannot be easily
integrated with CFD. Precipitation obviously fits this description, as the popula-
tion balance equation that describes it requires, after discretisation, a considerable
number of differential equations to be solved alongside the hydrodynamics. Thus
the advantages of the methodology are well demonstrated by this problem. Fur-
thermore, in chapter 3 we developed a numerical method for solving the population
balance equation, with particular relevance to problems including agglomeration
and/or breakage. This method will be employed here for the solution of the popu-
lation balance in \( \text{CaCO}_3 \) precipitation.

Wachi and Jones (1991a) and Hostomsky and Jones (1994), who studied pre-
cipitation in a laboratory reactor with a fiat gas-liquid interface, accounted in their
models for precipitation occurring in the interface as well as in the bulk. However,
the laboratory reactor used in their experiments was poorly mixed (1.3 lt, 100-400
rpm), resulting in very low values of the gas-liquid mass transfer coefficient, and
subsequently very high values of the contact time (2-16 sec). Under these conditions
precipitation in the interface is accountable, but in a bubble column, as well as in
most industrial equipment, the contact time is lower by several orders of magnitude
(0.02 sec for 5 mm bubbles moving at a slip velocity of 0.25 m/sec). Therefore the
effective interfacial volume is rather negligible compared to the bulk, and it seems
reasonable to neglect any particulate phenomena at the interface, thus achieving a
valuable reduction in CPU time.

4.3.2.1 CFD modelling and virtual tracer experiment

In gas-liquid reactors the interaction between the flow field and the chemical reac-
tions is usually one-way, i.e. the gas-liquid flow establishes the above key parameters
that determine the rate of reactions, but the reactions have no significant feedback
effect on the hydrodynamics. In the case of \( \text{CaCO}_3 \) precipitation this is a reasonable
assumption, provided that the suspension is dilute and the reactant gas is mixed
with an inert component, so that the bubble size does not change significantly due to
reactant depletion (as e.g. in the study of Fleicher et al., 1995). In our experiments
the solids content was less than 1% and the \( \text{CO}_2 / \text{N}_2 \) ratio in the input gas ranged
from 0.01 to 0.2, thus ensuring that the above assumption is valid.

The draft-tube bubble column used in the experiments resembles that of chapter
2, and its simulation with CFX was carried out in a similar manner. The virtual tracer experiment was also conducted along the lines of section 2.2.5.3, i.e. the passive scalar was introduced proportional to the local volume fraction (eq. 2.10). The results are shown in fig. 4.16.

The tracer concentration contours show a more or less uniform concentration throughout the column (note that the upper part of the column is occupied by gas, as the volume fraction contour shows, and the tracer concentration in that region is meaningless). This must be attributed to the short height of the column: due to intense circulation around the draft tube, the reactor functions as a CSTR. Minor concentration gradients in the top cannot be expected to play a major role in the outcome. This was confirmed by the experiments, where sampling in the riser and downcomer did not show any concentration difference beyond the margins of experimental error. Furthermore, the uncertainty in precipitation kinetics means that any fine detail in the hydrodynamic part of the model would be wasted. For this reason a single compartment will be employed to represent the column bulk.

### 4.3.2.2 Superposition of precipitation

In the preceding section it was decided that nucleation, both primary and secondary, size-independent growth and size-dependent agglomeration must be included in the model. The corresponding population balance equation, written in terms of particle volume as independent coordinate (in accordance with the method of chapter 3) is:

\[
\frac{\partial n(v, t)}{\partial t} = \frac{n_{in}(v, t) - n(v, t)}{\tau} - \frac{\partial}{\partial v} \left[ G(v) \cdot n(v, t) \right] + B_0 \delta(v - v_0) + \frac{1}{2} \int_0^v \beta_a(v - v', v') n(v - v', t)n(v', t) dv' - n(v, t) \int_0^\infty \beta_a(v, v') n(v', t) dv' 
\]

(4.26)

This equation refers to a finite, fully-mixed domain, and can therefore be applied to the compartments of the generalised compartmental model. By applying the numerical method of chapter 3, it can be transformed into a series of ordinary differential equations (eqs. 3.39). Thus the final system of ODEs to be integrated has the form:

\[
\frac{dC_i^B}{dt} = \sum_{k=1}^K r_k(C_i^B, n_j) 
\]

(4.27)

\[
\frac{dn_j}{dt} = f_j(C_i^B, n_j) 
\]

(4.28)
Figure 4.16: Left: shaded contours of volume fraction, right: passive scalar concentration (axisymmetric simulation, column axis on the right).
Here $f_j(C_j^B, n_j)$ stands for the right-hand side of the ODEs into which the original population balance is transformed via the finite element discretisation. Initial conditions are calculated from the mixing of bulk and interface at the end of the previous contact time. Note that the mass balances and the ODEs resulting from the population balance are coupled via the source terms. The coupling results because:

- precipitation kinetics are strongly dependent on supersaturation.

- The rate of consumption of the precipitating species is the rate of volumetric crystal growth, and must be calculated from the population balance.

### 4.3.3 Simulation results

Fig. 4.19 shows a comparison of the simulated and experimental results for the evolution of pH and calcium during the precipitation experiments. Overall, the time course of the process is well simulated. The measured pH curve seems to stabilize at a higher pH (7-8) than the predicted one, which falls down to pH 5-6. This is probably due to the kinetic model of the CO$_2$ - OH$^-$ system, which included only those sub-reactions that are important in the high pH regime. At near-neutral pH, other ionic species apart from the ones considered ($CO_5^-$, $HCO_5^-$) may be present; in addition, the pH meter was calibrated for the alkaline region. Any further discrepancies may be attributed to either the kinetics or the gas hold-up predictions.

The Ca$^{++}$ plot is indicative of the precipitation phenomena. During the induction time, i.e. the time during which the nuclei are formed and grow to observable size, consumption of substrate is minimal and Ca$^{++}$ is constant. Subsequently, a rapid drop in Ca$^{++}$ concentration indicates high growth rate. This is due to very high supersaturation levels, generated by the accumulation of CO$_5^-$ during the induction period. Rapid growth implies rapid consumption, and supersaturation soon falls to low levels, resulting in lower growth rate. This rate is almost constant till the pH drop, indicating a balance between consumption and generation of supersaturation. The redissolution of the crystals, resulting in the reappearance of free Ca$^{++}$, is not included in the model. The above phenomena are reflected in the supersaturation plot, shown in fig. 4.20: rapid increase during the induction time is followed by a rapid drop. Then supersaturation is constant, until dropping to zero.
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

due to depletion of Ca⁺⁺.

Figures 4.21, 4.22, 4.23, 4.24, 4.25, 4.26 show comparisons of the experimental CSD measurements and the simulation results. In most cases, the main trends are reproduced, though agreement is not quantitative. The first CSD measurement usually exhibits the highest deviation from the modelling results, partly due to experimental error. This measurement was usually taken immediately after the induction time, when the solution was still too dilute, and the obscuration of the Mastersizer S was below the limit required for a reliable measurement; large foreign particles would easily shift the CSD towards larger sizes. This has happened in fig. 4.23, where the first measurement showed a much higher proportion of big particles than subsequent ones, which is clearly unrealistic. The cumulative undersize distributions (figs. 4.22, 4.24, 4.26) reveal that the deviation occurs mainly at the larger sizes, sometimes greater even than 20μm, but such crystals were seldom observed by the SEM, which supports the possibility of the CSD being biased by foreign particles. The intermediate and final measurements, though, are in much better agreement with the predictions. Most important, the combination of secondary nucleation and agglomeration succeeds in reproducing the bimodality in the CSD.

The lack of quantitative agreement can be attributed to the uncertainty in the precipitation kinetics, since the reaction engineering model was validated by the concentration predictions. This was to be expected, as the precipitation mechanisms have not been adequately studied in the gas-liquid environment. The modelling of agglomeration seems to pose the most difficult obstacles, due to the lack of a definitive mechanistic model of the cementation step. This problem is especially pronounced in this study, due to two reasons: a) in a batch process the agglomeration kinetics are time-varying, depending on supersaturation; and b) the hydrodynamic environment that determines the aggregation-disruption equilibrium (or aggregation efficiency) in gas-liquid flows is fundamentally different from that in stirred tanks. The agglomeration model used in this study takes supersaturation into account, and thus achieves reasonable prediction of the extent and time scale of the phenomena during the semi-batch process. The sustaining or disruption of the weak aggregates is not mechanistically incorporated, however, and this may be the reason why the exact shape of the CSD is not accurately reproduced.
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

Figure 4.17: Experimental and simulated evolution of pH and [Ca⁺⁺]: [Ca(OH)₂]
= 3 mol/m³, 0.00001 : 0.0001 m³/s CO₂:N₂

Figure 4.18: Experimental and simulated evolution of pH and [Ca⁺⁺]: [Ca(OH)₂]
= 4 mol/m³, 0.00001 : 0.0001 m³/s CO₂:N₂
Figure 4.19: Experimental and simulated evolution of pH and [Ca\textsuperscript{++}]: c)\([\text{Ca(OH)}\textsubscript{2}] = 4 \text{ mol/m}^3, 0.00001 : 0.00005 \text{ m}^3/s \text{ CO}_2: \text{N}_2\)

Figure 4.20: Simulation of the evolution of supersaturation in a typical experiment.
Figure 4.21: Experimental and simulated CSD, logarithmic plot. \([\text{Ca(OH)}_2] = 3 \text{ mol/m}^3, 0.00001 : 0.0001 \text{ m}^3/s \text{ CO}_2: \text{N}_2\).
Figure 4.22: Experimental and simulated CSD, cumulative undersize plot.

\[ [\text{Ca(OH)}_2] = 3 \text{ mol/m}^3, 0.00001 : 0.0001 \text{ m}^3/s \text{ CO}_2: \text{N}_2. \]
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

Figure 4.23: Experimental and simulated CSD, logarithmic plot. [Ca(OH)₂] = 4 mol/m³, 0.00001 : 0.0001 m³/s CO₂:N₂
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

Figure 4.24: Experimental and simulated CSD, cumulative undersize plot.

\[ [\text{Ca(OH)}_2] = 4 \text{ mol/m}^3, 0.00001 : 0.0001 \text{ m}^3/s \text{ CO}_2 : \text{N}_2 \]
Gas-liquid precipitation of CaCO₃ in a bubble column reactor

Figure 4.25: Experimental and simulated CSD, logarithmic plot. \([\text{Ca(OH)}_2] = 4 \text{ mol/m}^3, 0.00001 : 0.00005 \text{ m}^3/\text{s CO}_2:\text{N}_2\).
Gas-liquid precipitation of CaCO$_3$ in a bubble column reactor

Figure 4.26: Experimental and simulated CSD, cumulative undersize plot.

$[\text{Ca(OH)}_2] = 4 \ mol/m^3, 0.00001 : 0.00005 \ m^3/s \ \text{CO}_2:\text{N}_2$. 
Chapter 5

Conclusions

This work has addressed fundamental issues in two fields of chemical engineering: the merging of fluid dynamics and chemical phenomena in multiphase reactors, and the population balance modelling of particulate processes. With respect to the first, a hybrid methodology that integrates CFD and reaction engineering principles was proposed; with respect to the second a numerical method based on finite elements was developed, that resolves several issues regarding conservation and stability.

The advances made in these fields were subsequently applied to a problem whose analysis is a case study of both: gas-liquid precipitation, i.e. the formation of a crystalline solid product via the reaction of a gas and a liquid. The conjunction of CFD, reaction engineering and population balance analysis has yielded a fully predictive and yet tractable model, that could be used for process design and scale-up purposes if supplied with accurate data on precipitation kinetics. Moreover, the detailed predictions of the evolution of crystal size distribution and concentrations, have yielded insight into the nature of the underlying phenomena. The main findings in each one of these fields will now be summarised.

5.1 Hybrid modelling of multiphase reactors

The main concept of this methodology is that mixing, both due to mean flow and turbulent exchange at the meso- and macro- scale, can be accounted for via averaging the flow and concentration fields resulting from a CFD flow simulation and a computational tracer experiment, and using these results to parametrise a generalised
Conclusions

compartmental model. The approach introduces no new phenomenological concepts, and its accuracy depends on the accuracy of the CFD (whose main phenomenological aspects are the turbulence and multiphase flow models) and the fine-tuning between the CFD and compartmental model tracer simulations. The former will benefit from any future advances; the latter depends on the time and effort the process designer can afford to invest. Application to bubble column reactors was accomplished here by superimposing a phenomenological model of interfacial mass transfer and chemical reaction on the generalised compartmental model. The main concept, however, is more broad and may be applied to several other fields such as stirred tank reactors (in conjunction with phenomenological micromixing models), biochemical and precipitation reactors (in conjunction with population balance models).

5.2 Solution of the population balance with finite elements

In this work, a finite element scheme was proposed for solving dynamic population balance problems. The PBE and boundary conditions were formulated to accommodate an arbitrary problem involving any combination of the four main particulate mechanisms: nucleation, growth, aggregation and breakage. The numerical scheme consists of collocation on linear elements with an "upwind" propagation of first-order derivatives (i.e. growth) that result in enhanced stability. Integral terms are evaluated with simple Newton-Cotes quadrature, and a three-point formula (Simpson's rule) requiring just one interpolation was found to be sufficient for most kernels. Three sources of error in the conservation of moments were identified: quadrature error, effect of non-uniform grid, and error due to temporal integration, respectively. A rule for determining the order of quadrature was proposed to eliminate the first source of error, while an accurate way of evaluating the aggregation source term, based on a geometric interpretation of it, was developed to mitigate the second. Although the error cannot be completely eliminated, the results demonstrated that if the above rules are followed, it will be negligible.

The dynamic method was capable of reaching steady-state solutions with ease and speed. Nicmanis and Hounslow (1998), who studied the steady-state PBE
with finite elements, reported numerical problems such as matrix ill-conditioning in
growth problems which could only be resolved by resorting to Galerkin methods, as
well as difficulties in inverting the dense matrices resulting from the discretisation.
On the other hand the dynamic method, guided by the solution at the previous time
step, proved very easy to converge even with collocation and explicit integration, in
which case no matrix inversion is required. This is a commonly encountered situation
in numerical analysis: a steady-state problem is easier to solve by considering the
dynamic case.

When comparing any finite element scheme with DPBs, it must be born in
mind that the mechanisms of aggregation and breakage are essentially discrete, and
therefore DPBs are a more natural choice for them. Moreover, being zero order,
DPBs can accommodate more easily certain situations that are ill-conditioned for
higher order methods (e.g. monodisperse initial conditions). The value of the finite
element method is better realised when considering combined problems that involve
both discrete and continuous mechanisms, in which case the generality of the method
allows it to tackle a variety of different problems without being tailored to the needs
of individual ones. The scheme proposed here features this generality while retaining
some of the advantages of DPBs (stability and ability to deal with ill-conditioned
problems) due to its low-order.

5.3 Gas-liquid precipitation of CaCO₃

In this work population balance, reaction engineering and hydrodynamic principles
were integrated into a modelling approach for the gas-liquid precipitation process.
By accounting for all of the above, the model aims to provide tools for analysis and
scale-up of industrial type equipment used for this class of processes. The approach
was subsequently used to simulate the precipitation of CaCO₃ by CO₂ absorption
into lime, and experiments on a laboratory-scale rig were carried out for comparison.
The main findings can be summarised in the following:

- The increase in mean size after the pH drop is due to the disappearence of the
  smaller crystals by dissolution.

- Agglomeration starts to take place at relatively high pH and proceeds to a large
Conclusions

extent, possibly because the aggregates are less frequently disrupted than in stirred tanks. For the same reason, the shear (orthokinetic) kernel succeeded in reproducing the net result of the aggregation-disruption processes better than the size-independent.

- Bimodality in the CSD was shown, by means of both experiment and model, to be the result of a new wave of nucleation-growth induced by the existing crystals (secondary nucleation).

The conjunction of the penetration theory and the CFD predictions of the gas hold-up seem to yield adequate description of the reactor performance, as the prediction of the concentrations' evolution shows. Therefore the the discrepancies in the CSD predictions can be attributed mainly to the uncertainty in precipitation kinetics. Systematic studies of the kinetics, as well as improved mechanistic models of agglomeration (especially in relation to hydrodynamics) are required to obtain quantitative agreement, but the main trends in the CSD can be explained and reproduced with the aid of the model.

5.4 Suggestions for future work

The work undertaken here could be followed up in several directions. To begin with, the hybrid CFD-reaction engineering methodology presented in chapter 2 is general and its main concept lends itself to other potential reaction engineering applications, such as:

- Single-phase reactors with complex nonlinear reaction systems, whose direct coupling with CFD would pose serious computational problems.

- Stirred tank reactors with fast reactions, where micromixing models can be included in the compartments around the inlets.

- Catalytic and biochemical reactors, where adsorption models or cell population balances may be employed.

In our case studies, the configuration of the generalised compartmental model on the basis of the CFD results was carried out through an iterative procedure led
Conclusions

by heuristics. Future work could explore the automation of the procedure, which would contribute further time-saving to the design stage.

The finite-element numerical method for solution of the time-dependent population balance is also applicable to a variety of physical and chemical systems, apart from crystallisation and precipitation. This is owed to the volume-based formulation, which excludes system-specific elements such as shape factors from the model, and to its stability and conservation properties that enable it to tackle both nucleation-growth and agglomeration-breakage problems. Some other systems that can be described by population balance equations are mentioned are: polymers, aerosols, emulsions, bubble swarms and cell populations. Future work could explore the applicability of the method to other systems.

Finally, our case study of CaCO₃ precipitation has shown that the conjunction of fluid dynamics, reaction engineering and crystallisation models within the framework of a hybrid approach is feasible and can lead to fully-predictive scale-up tools. Such tools must be supplied with reliable constitutive equations and kinetic data, however, and such data are currently lacking, regarding crystallisation mechanisms and agglomeration in particular. Future work in this field could focus on the development of reliable kinetic models, both mechanistic and empirical, for the crystallisation mechanisms so that fully predictive quantitative scale-up can be made possible.
Appendix A

Solution of the interface PDEs

A straightforward point-iterative linearisation is bound to encounter stability problems due to the non-linear reaction terms, while full computation of the Jacobian would require considerable resources. This complexity will be avoided by the use of an iterative scheme where the equations for each chemical component and moment are being solved sequentially, while on the same time utilising gradient information. The source term is linearised as:

\[
\int_{t-1}^{t} (f_{i,p} \delta x) dt = \\
= \int_{t-1}^{t} \left[ f_{i,p} |_{t-1} + \sum_{j \neq i} \frac{\partial f_{i,p}}{\partial C_j} \left[ C_{j,t} - C_{j,t-1} \right] + \frac{\partial f_{i,p}}{\partial C_i} \left[ C_{i,t} - C_{i,t-1} \right] \right] dt \delta x
\]

An advantage of this scheme is that the term:

\[
\frac{\partial f_{i,p}}{\partial C_{i,p}} \left|_{t-1} \right. \int_{t-1}^{t} C_{i,p} \delta x dt
\]

The linearised form is:

\[
C_{i,p,t} \delta x - C_{i,p,t-1} \delta x = D \frac{\delta^2}{\delta x^2} C_{i,e,t} - 2D \delta t a \frac{\delta}{\delta x} C_{i,p,t} - 2D \delta t (1-a) \frac{\delta}{\delta x} C_{i,p,t-1} + \\
+ D \frac{\delta^2}{\delta x^2} C_{i,w,t} + f_{i,p} |_{t-1} \delta x dt + \sum_{j \neq i} \frac{\partial f_{i,p}}{\partial C_j} |_{t-1} \left( C_{j,p,t} - C_{j,p,t-1} \right) \delta x dt + \\
+ \frac{\partial f_{i,p}}{\partial C_{i,p}} \left|_{t-1} \right. \int_{t-1}^{t} C_{i,p,t} \delta x dt - \frac{\partial f_{i,p}}{\partial C_{i,p}} \left|_{t-1} \right. C_{i,p,t-1} \delta x dt
\]

(A.2)
At this point all terms in the discretised equation can be assembled to form three coefficients \((\alpha_w, \alpha_p, \alpha_e)\) to the variables at the three neighbouring points \((C_{j,p,t}, C_{j,e,t}, C_{j,w,t})\) at the new time point, and a constant term \((b)\). Thus a tridiagonal matrix is formed, whose solution yields the values of the variable at the new time point:

\[
\begin{bmatrix}
   a_{1p} & a_{le} \\
   \vdots & \vdots & \ddots \\
   a_{kw} & a_{kp} & a_{ke} & \ddots \\
   \vdots & \ddots & \ddots & \ddots \\
   a_{nw} & a_{np}
\end{bmatrix}
\begin{bmatrix}
   C_{i1} \\
   \vdots \\
   C_{ik} \\
   \vdots \\
   C_{in}
\end{bmatrix}
= \begin{bmatrix}
   b_1 \\
   \vdots \\
   b_k \\
   \vdots \\
   b_n
\end{bmatrix}
\]  

(A.3)

The coefficients are yielded by the expressions:

\[
\begin{align*}
\alpha_p &= \delta x + \frac{2D\delta t a}{\delta x} - \frac{\partial f_{i,p}}{\partial C_{i,p}} \bigg|_{t-1} a \delta x \delta t & (A.4) \\
\alpha_w &= -D \frac{\delta t}{\delta x} & (A.5) \\
\alpha_e &= -D \frac{\delta t}{\delta x} & (A.6)
\end{align*}
\]

\[
b = +C_{i,p,t-1} \delta x - \frac{2D\delta t(1-a)}{\delta x} C_{i,p,t-1} + f_{i,p} \bigg|_{t-1} \delta x \delta t + \\
+ \sum_{j \neq i} \frac{\partial f_{j,p}}{\partial C_{j,p}} \bigg|_{t-1} \left( C_{j,p,t} - C_{j,p,t-1} \right) \delta x \delta t - a \frac{\partial f_{i,e}}{\partial C_{i,e}} \bigg|_{t-1} C_{i,p,t-1} \delta x \delta t
\]  

(A.7)
Appendix B

Source files and user subroutines for the CFX 4.3 code

B.1 CFX source file

The following is a sample CFX source file used for modelling a bubble column. For information on the format of CFX source files, one should consult the CFX manual (AEA Technology, 1999).

/************************************************************/
/* BUBBLE COLUMN */
/* BY STELIGS RIGGPOULOS */
/* 24/06/2002 */
/*************************************************************/

/******************** General settings ********************/

>>CFXF3D
>>SET LIMITS
   TOTAL REAL WORK SPACE 60000000
   TOTAL INTEGER WORK SPACE 10000000
>>OPTIONS
   TWO DIMENSIONS
   RECTANGULAR GRID
   CYLINDRICAL COORDINATES
   AXIS INCLUDED
   AXISSYMMETRIC MODIFICATION
   TURBULENT FLOW
   BOUYANT FLOW
   NUMBER OF PHASES 2
   TRANSIENT FLOW
   USER SCALAR EQUATIONS 1
>>USER FORTRAN
    USRTRN

---

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Source files and user subroutines for the CFX 4.3 code

USRINT
USRSRC

>>PHASE NAMES
PHASE1 'WATER'
PHASE2 'GAS'

/* Define the grid *********************************************/

>>MODEL TOPOLOGY

#CALC
Bottom_I = 5;
Main_I = 195;
Top_I = 40;
Riser_J = 10;
Dcomer_J = 8;
GasInlet_I = 1;
GasInlet1_J = 1;
GasInlet2_J = 4;
RiserTop_I = Bottom_I+Main_I;
Total_I = Bottom_I+Main_I+Top_I;
Total_J = Riser_J+Dcomer_J;
Total_K = 1;
BEFORE_DT_I = Bottom_I-1;
AFTER_DT_J = Riser_J+1;
BEFORE_DT_J = Riser_J-1;
AFTER_DT_I = RiserTop_I+1;
GAS_BOUNDARY = 6;
AFTER_GAS_BOUNDARY = 6+1;
#ENDCALC
#ENDCALC

>>CREATE BLOCK
BLOCK NAME 'BUBBLE.COLUMN'
BLOCK DIMENSIONS #Total_I #Total_J #Total_K

>>CREATE PATCH
PATCH NAME 'TOP'
PATCH TYPE 'PRESSURE BOUNDARY'
BLOCK NAME 'BUBBLE.COLUMN'
HIGH I

>>CREATE PATCH
PATCH NAME 'DRAFT-TUBE'
PATCH TYPE 'THIN SURFACE'
BLOCK NAME 'BUBBLE.COLUMN'
PATCH LOCATION #Bottom_I #RiserTop_I #Riser_J #Riser_J 1 1
HIGH J

>>CREATE PATCH
PATCH NAME 'CENTRE-LINE'
PATCH TYPE 'SYMMETRY PLANE'
BLOCK NAME 'BUBBLE.COLUMN'
LOW J

>>CREATE PATCH
PATCH NAME 'SPARGER'
PATCH TYPE 'USER3D'
BLOCK NAME 'BUBBLE.COLUMN'
PATCH LOCATION #GasInlet_I #GasInlet_I #GasInlet1_J #GasInlet2_J 1 1

>>CREATE PATCH
PATCH NAME 'DOWNCOMER'
PATCH TYPE 'USER3D'
BLOCK NAME 'BUBBLE_COLUMN'
PATCH LOCATION 1 #RiserTop_I #Riser_J #Total_J 1 1
>>CREATE PATCH
PATCH NAME 'RISER'
PATCH TYPE 'USER3D'
BLOCK NAME 'BUBBLE_COLUMN'
PATCH LOCATION 1 #RiserTop_I 1 #Riser_J 1 1

>>CREATE PATCH
PATCH NAME 'TOTAL.COLUMN'
PATCH TYPE 'USER3D'
BLOCK NAME 'BUBBLE_COLUMN'
PATCH LOCATION 1 #Total_I 1 #Total_J 1 1

/* Model data and properties *************************************/

>>MODEL DATA
>>SET INITIAL GUESS
>>INPUT FROM FILE
READ DUMP FILE
>>TITLE
PROBLEM TITLE 'DRAFT-TUBE BUBBLE COLUMN'
>>AMBIENT VARIABLES
PHASE NAME 'WATER'
VOLUME FRACTION 0.0
>>AMBIENT VARIABLES
PHASE NAME 'GAS'
VOLUME FRACTION 1.0
END

/* »DIFFERENCING SCHEME
VOLUME FRACTION 'UPWIND' */
>>PHYSICAL PROPERTIES
#CALC
DENL = 1000.0;
VISL = 1.0E-3;
DENG = 1.2;
VISG = 1.8E-5;
#ENDCALC

>>BUOYANCY PARAMETERS
GRAVITY VECTOR -9.81 0.0 0.0
BUOYANCY REFERENCE DENSITY #DENL
>>FLUID PARAMETERS
PHASE NAME 'WATER'
DENSITY #DENL
VISCOISITY #VISL
>>FLUID PARAMETERS
PHASE NAME 'GAS'
DENSITY #DENG
VISCOISITY #VISG

/* »DIFFERENCING SCHEME
VOLUME FRACTION 'UPWIND' */

>>TRANSIENT PARAMETERS
>>FIXED TIME STEPPING
TIME STEPS 5000.01

iscrim data and user subroutines for the CFX 4.3 code 185

>>SCALAR PARAMETERS
>>DIFFUSIVITIES
/* PHASE NAME 'WATER'*/
USER SCALAR1 1.0E-4

iscrim data and user subroutines for the CFX 4.3 code 185

/* Turbulence and multiphase model *************************************/

>>TURBULENCE PARAMETERS
>>TURBULENCE MODEL
Source files and user subroutines for the CFX 4.3 code

PHASE NAME 'WATER'
TURBULENCE MODEL 'K-EPSILON'
/*PARTICLE INDUCED TURBULENCE 'SATO'*/

>>TURBULENCE MODEL
PHASE NAME 'GAS'
TURBULENCE MODEL 'LAMINAR'
/*PARTICLE INDUCED TURBULENCE 'SATO'*/

>>TURBULENT PRANDTL NUMBER
PHASE NAME 'WATER'
USER SCALAR 1.0

>>MULTIPHASE PARAMETERS
>>PHASE DESCRIPTION
PHASE NAME 'WATER'
LIQUID
CONTINUOUS
>>PHASE DESCRIPTION
PHASE NAME 'GAS'
GAS
DISPERSE
MEAN DIAMETER 5.0E-3
MINIMUM VOLUME FRACTION 1.0E-8

>>MULTIPHASE MODELS
>>MOMENTUM
INTER PHASE TRANSFER
/**TURBULENCE
HOMOGENEOUS*/

>>INTER PHASE TRANSFER MODELS
>>MOMENTUM
FIRST PHASE NAME 'WATER'
SECOND PHASE NAME 'GAS'
>>PARTICLE DRAG MODEL
FLOW REGIME 'NEWTON'
NEWTON COEFFICIENT 0.66
>>NON DRAG FORCES
TURBULENT DISPERSION FORCE

/*************************************************************************/

/* Input of gas ************************************************************/

>>SOURCES
#CALC
FMASS = 0.0002;
#ENDCALC
PATCH NAME 'SPARGER'
PHASE NAME 'GAS'
TOTAL FOR PHASE
PRESSURE #FMASS 0.0
VOLUME FRACTION #FMASS 0.0

**************************************************************************/

/*************************************************************************/

/* Solver options **********************************************************/

>>SOLVER DATA
>>PROGRAM CONTROL
#CALC
IMON = 40;
JMON = 5;
KMON = 1;
#ENDCALC
Source files and user subroutines for the CFX 4.3 code

OUTPUT MONITOR BLOCK 'BUBBLE_COLUMN'
OUTPUT MONITOR POINT #IMON #JMON #KMON
MINIMUM NUMBER OF ITERATIONS 3
MAXIMUM NUMBER OF ITERATIONS 40
MASS SOURCE TOLERANCE 1.0E-7

/ * >>false Timesteps
U VELOCITY 0.01
V VELOCITY 0.01 */

>>Under relaxation factors
ALL EQUATIONS 0.1

/*-------------------------------------------------------------*/

/* Actual geometry and dimensions *****************************/

>>create grid

#CALC
Bottom_length = 0.05;
Main_length = 1.95;
Top_length = 0.4;
Riser_width = 0.06;
Dcomer_width = 0.05;

Idx = (Bottom_length+Main_length+Top_length)/Total_I;
Jdx = (Riser_width+Dcomer_width)/Total_J;
Kdx = 2*3.14159;

#ENDCALC

>>Simple grid

BLOCK NAME 'BUBBLE_COLUMN'
DX #Total_I*#Idx
DY #Total_J*#Jdx
DZ #Total_K*#Kdx

/*-------------------------------------------------------------*/

/* Boundary conditions ******************************************/

>>Model boundary conditions

>>Pressure boundaries

PATCH NAME 'TOP'
PRESSURE 0.0

/*-------------------------------------------------------------*/

/* Output files ***********************************************

>>Output options

>>Print options

>>What
ALL VARIABLES

>>Animation data
FILE NAME 'ANIM'
Z 3.14
TIME INTERVAL 0.01

PHASE NAME 'WATER'
USER SCALAR1

/*-------------------------------------------------------------*/

>>STOP
B.2 Fortran user subroutines

CFX allows the use of Fortran user subroutines to customize certain elements of the solution. The following ones were used here:

**USRTRN**: Prints user-specified information.

**USRINT**: Specifies initial conditions.

**USR SRC**: Used to introduce the passive scalar for the virtual tracer experiment.

```fortran
SUBROUTINE USRTRN(U, V, W, P, VFRAC, DEN, VIS, TE, ED, RS, T, H, RF, SCAL,
                    XP, YP, ZP, VOL, AREA, VPSR, ARPSR, WPACT, CONV, TPT,
                    IBLK, IPVERT, IPWGN, IPFACT, IPMODP, IPMODB, IPFACB,
                    WORK, IWORK, CWORK)
```

---

**User subroutine to allow users to modify or monitor the solution at the end of each time step**

**IMPORTANT**

**Users may only add or alter parts of the subroutine within the designated user areas**

---

This subroutine is called by the following subroutines:

**CUSR TRNMOD**

---

**SUBROUTINE ARGUMENTS**

- **U** - U component of velocity
- **V** - V component of velocity
- **W** - W component of velocity
- **P** - Pressure
- **VFRAC** - Volume fraction
Source files and user subroutines for the CFX 4.3 code

C  DEN - DENSITY OF FLUID
C  VIS - VISCOSITY OF FLUID
C  TE - TURBULENT KINETIC ENERGY
C  ED - EPSILON
C  RS - REYNOLD STRESSES
C  T  - TEMPERATURE
C  H  - ENTHALPY
C  RF - REYNOLD FLUXES
C  SCAL - SCALARS (THE FIRST 'NCONC' OF THESE ARE MASS FRACTIONS)
C  XP - X COORDINATES OF CELL CENTRES
C  YP - Y COORDINATES OF CELL CENTRES
C  ZP - Z COORDINATES OF CELL CENTRES
C  VOL - VOLUME OF CELLS
C  AREA - AREA OF CELLS
C  VPOR - POROUS VOLUME
C  ARPOR - POROUS AREA
C  WFACT - WEIGHT FACTORS
C  CONV - CONVECTION COEFFICIENTS
C  IPT - 1D POINTER ARRAY
C  IBLK - BLOCK SIZE INFORMATION
C  IPVERT - POINTER FROM CELL CENTERS TO 8 NEIGHBOURING VERTICES
C  IPNODN - POINTER FROM CELL CENTERS TO 6 NEIGHBOURING CELLS
C  IPFACN - POINTER FROM CELL CENTERS TO 6 NEIGHBOURING FACES
C  IPNODF - POINTER FROM CELL FACES TO 2 NEIGHBOURING CELL CENTERS
C  IPNODB - POINTER FROM BOUNDARY CENTERS TO CELL CENTERS
C  IPFACB - POINTER FROM BOUNDARY CENTERS TO BOUNDARY FACES
C  WORK - REAL WORKSPACE ARRAY
C  IWORK - INTEGER WORKSPACE ARRAY
C  CWORK - CHARACTER WORKSPACE ARRAY
C
C SUBROUTINE ARGUMENTS PRECEDED WITH A 'X' ARE ARGUMENTS THAT MUST
C BE SET BY THE USER IN THIS ROUTINE.
C NOTE THAT OTHER DATA MAY BE OBTAINED FROM CFX-4 USING THE
C ROUTINE GETADD, FOR FURTHER DETAILS SEE THE VERSION 4
C USER MANUAL.
C
C***********************************************************************
C
LOGICAL LDEN, LVIS, LTURB, LTEMP, LBUOY, LSCAL, LCOMP
+ , LRECT, LCYN, LAXIS, LPOROS, LTRANS
C
CHARACTER(*) CWORK
C
C++++++++++++++++ USER AREA 1 +++++++++++++++++++++++++++++++++++++++++
C AREA FOR USERS EXPLICITLY DECLARED VARIABLES
C CHARACTER*20 FNAME
C
C++++++++++++++++ END OF USER AREA 1 ++++++++++++++++++++++++++++++++++
C
COMMON
+ /ALL/ NBLOCK, NCCELL, NBDRY, NNODE, NFACE, NVERT, NDIM
+ /ALLWRK/ NWS, NWIS, NWSF, IWFRE, IWIFRE, IWCFRE
+ /ADDMIS/ NPHERE, NSCAL, NVAR, NPROP
+ , NVAR, NPROP, NDXNN, NDGEO, NDOREF, NILIST, NRLIST, NTOPOL
+ /CHKUSR/ IVERS, IUCALL, IUSED
+ /CONC/ NCONC
+ /DEVICE/ NREAD, NWRITE, NRDISK, NWDISK
+ /DDOM/ ILEN, ILEN
+ /LOGIC/ LDEN, LVIS, LTURB, LTEMP, LBUOY, LSCAL, LCOMP
+ , LRECT, LCYN, LAXIS, LPOROS, LTRANS
+ /MLTGRD/ MLEVEL, NLEVEL, ILEVEL
+ /SGLDBL/ IFLGPR, ICHKPR
+ /SPARM/ SMALL, SORMAX, NITER, INDPRI, MAXIT, NODREF, NODMON
+ /TIMUSR/ DTUSR
Source files and user subroutines for the CFX 4.3 code

+ /TRANSI/ NSTEP, KSTEP, MF, INCORE
+ /TRANS/ TIME, UT, DTINVF, TPROC

C
C++++++++++++++++ USER AREA 2 ++++++++++++++++++++++++++++++++++++++++++++++++++++++
C-- AREA FOR USERS TO DECLARE THEIR OWN COMMON BLOCKS
C THESE SHOULD START WITH THE CHARACTERS 'UC' TO ENSURE
C NO CONFLICT WITH NON-USER COMMON BLOCKS
C
C++++++++++++++++ END OF USER AREA 2 ++++++++++++++++++++++++++++++++++++++++++
C
C DIMENSION + U(NNODE,NPHASE), V(NNODE,NPHASE), W(NNODE,NPHASE), P(NNODE,NPHASE)
+ VFRAC(NNODE,NPHASE), DEN(NNODE,NPHASE), VIS(NNODE,NPHASE)
+ TE(NNODE,NPHASE), ED(NNODE,NPHASE), RS(NNODE,NPHASE,6)
+ T(NNODE,NPHASE), H(NNODE,NPHASE), AF(NNODE,NPHASE,4)
+ S(NNODE,NPHASE,NSCAL)
DIMENSION + XP(NNODE), YP(NNODE), ZP(NNODE)
+ VOL(NCELL), AREA(NFACE,3), VPOR(NCELL), ARPOR(NFACE,3)
+ NFACET(NFACE), CONV(NFACE, NPHASE)
+ IPT(*), IBLK(5,NBLOCK)
+ IPVERT(NCELL,8), IPN0DN(NCELL,6), IPFACN(NCELL,6), IPN0DF(NFACE,4)
+ IPNODB(NBLOCK,4), IPFACB(NBLOCK)
+ IWORK(*), WORK(*), CWORK(*)
DIMENSION SGNWL(6)
C
C++++++++++++++++ USER AREA 3 ++++++++++++++++++++++++++++++++++++++++++++++++++++++
C-- AREA FOR USERS TO DIMENSION THEIR ARRAYS

C-- AREA FOR USERS TO DEFINE DATA STATEMENTS

C-- END OPERATION
C
C++++++++++++++++ END OF USER AREA 3 ++++++++++++++++++++++++++++++++++++++++++
C
DATA SGNWL / 1.0, 1.0, 1.0, -1.0, -1.0, -1.0 /
C
C-- STATEMENT FUNCTION FOR ADDRESSING
IF (I,J,K)=IPT((K-1)*ILEN*JLEN+(J-1)*ILEN+I)
C
C-- VERSION NUMBER OF USER ROUTINE AND PRECISION FLAG
C
IVERS=3
ICHKPR = 1
C
C++++++++++++++++ USER AREA 4 ++++++++++++++++++++++++++++++++++++++++++++++++++++++
C-- TO USE THIS USER ROUTINE FIRST SET IUSED=1
C
C IUSED=1
C
C++++++++++++++++ END OF USER AREA 4 ++++++++++++++++++++++++++++++++++++++++++
C
IF (IUSED.EQ.0) RETURN
C
C-- FRONTEND CHECKING OF USER ROUTINE
IF (IUCALL.EQ.0) RETURN
C
C++++++++++++++++ USER AREA 5 ++++++++++++++++++++++++++++++++++++++++++++++++++++++
C
C END OPERATION
C
C-----------------------------------------------------------------
C CALCULATIONS IN THE BEGINNING
C
IF(KSTEP.EQ.1) THEN
    CALL IPREC('BUBBLE_COLUMN','BLOCK','CENTRES',IPT,ILEN,JLEN,KLEN,+
        CWORK,IWORK)
    FULLCEL = 0
    DO 101 I=1,ILEN
        INODE = IP(I,4,1)
        IF (VFRAC(INODE,2).LT.0.5) THEN
            FULLCEL = FULLCEL + 1
        ENDIF
    101 CONTINUE
    NODE2=IP(2,1,1)
    NODE3=IP(3,1,1)
    LIQLEVELI = FULLCEL * (XP(NODE3)-XP(NODE2))
    WRITE(81,*) LIQLEVELI
ENDIF
C
---------------------------------------------------------------
C CALCULATIONS IN THE END
---------------------------------------------------------------
C
IF(KSTEP.EQ.NSTEP) THEN
C
CALL IPALL('DOWNCOMER PATCH','CENTRES',IPT,NPT,+
        CWORK,IWORK)
C
SUMVOL = 0.0
SUMVG = 0.0
SUMLIQUIDVELD = 0.0
DO 201 I=1,NPT
    INODE=IPT(I)
    SUMVOL = SUMVOL+ VOL(INODE)
    SUMVG = SUMVG + VFRAC(INODE,2)*VOL(INODE)
    SUMLIQUIDVELD = SUMLIQUIDVELD + U(INODE,1)
201 CONTINUE
RGDOWN = SUMVG/SUMVOL*100.0
AVELIQVELD = SUMLIQUIDVELD/NPT
C
/*CALCULATE GAS HOLDUP AND AVERAGE VELOCITIES IN RISER*/
C
CALL IPALL('RISER','*','PATCH','CENTRES',IPT,NPT,+
        CWORK,IWORK)
C
SUMVOL = 0.0
SUMVG = 0.0
SUMGASVEL = 0.0
SUMLIQUIDVEL = 0.0
DO 301 I=1,NPT
    INODE=IPT(I)
    SUMVOL = SUMVOL+VOL(INODE)
    SUMVG = SUMVG + VFRAC(INODE,2)*VOL(INODE)
    SUMGASVEL = SUMGASVEL + U(INODE,2)
    SUMLIQUIDVEL = SUMLIQUIDVEL + U(INODE,1)
301 CONTINUE
RGRISE = SUMVG/SUMVOL*100.0
AVEGASVEL = SUMGASVEL/NPT
C
AVELIQUIDVEL = SUMLIQUIDVEL/NPT
SLIPVEL = AVEGASVEL - AVELIQUIDVEL

C******************************************************************************
C
C******************************************************************************
C
C-----CALCULATE GAS HOLDUP IN COLUMN
C
CALL IPALL('TOTAL_COLUMN','*','PATCH','CENTRES',IPT,NPT,
+ CWORK,IWORK)
C
SUMVOL = 0.0
SUMVG = 0.0
SUMGASVEL = 0.0
SUMLIQUIDVEL = 0.0
DO 401 I=1,NPT
   INODE=IPT(I)
   SUMVOL = SUMVOL+VOLCINODE
   SUMVG = SUMVG +VFRAC(INODE,2)*VOL(INODE)
401 CONTINUE
RGTOTAL = SUMVG/SUMVOL*100.0

C******************************************************************************
C
C******************************************************************************
C
C-----CALCULATE AVERAGE TURBULENT ENERGY DISSIPATION
C
CALL IPALL('TOTAL.COLUMN PATCH','CENTRES',IPT,NPT,
+ CWORK,IWORK)
C
SUMED =0.0
DO 501 I=1,NPT
   INODE=IPT(I)
   SUMED = SUMED+ED(INODE,1)
501 CONTINUE
AVEED = SUMED/NPT

C******************************************************************************
C
C******************************************************************************
C
C-----CALCULATE POSITION OF FREE SURFACE
C
CALL IPREC('BUBBLE_COLUMN','BLOCK','CENTRES',IPT,ILEN,JLEN,KLEN,
+ CWORK,IWORK)
FULLCEL = 0
DO 700 I=1,ILEN
   INODE = IP(I,4,1)
   IF (VFRACCINODE,2).LT.0.5) THEN
      FULLCEL = FULLCEL + 1
   ENDIF
700 CONTINUE
NODE1=IP(1,1,1)
NODE2=IP(2,1,1)
RLIQLEVEL = FULLCEL * (XP(NODE2)-XP(NODE1))
DO 730 I=1,ILEN
   DO 720 J=1,JLEN
      DO 710 K=1,KLEN
         INODE = IP(I,J,K)
710 CONTINUE
720 CONTINUE
730 CONTINUE
TOTALIQ = TOTALIQ + VOL(INODE)*VFRAC(INODE,1)
IF (VFRAC(INODE,2).LT.0.5) THEN
   REALIQ = REALIQ + VOL(INODE)*VFRAC(INODE,1)
ENDIF
710 CONTINUE
720 CONTINUE
730 CONTINUE

WRITE(NWRITE,*) P(IP(1,10,1),1)
C
C******************************************************************
C
C PREPARE OUTPUT FILE
C
OPEN (81, FILE='../RESULTS')
WRITE(81,*) '****** GAS HOLDUP ******'
WRITE(81,*) ' ' 
WRITE(81,*) ' IN DOWNCOMER (%) = ', RGDOWN
WRITE(81,*) ' IN RISER (%) = ', RGRISE
WRITE(81,*) ' ' 
WRITE(81,*) ' TOTAL (%) = ', RGTOTAL
WRITE(81,*) ' ' 
WRITE(81,*) ' TURBULENT ENERGY DISSIPATION = ', AVEED
WRITE(81,*) ' ' 
WRITE(81,*) ' ****** AVERAGE VELOCITIES ******'
WRITE(81,*) ' ' 
WRITE(81,*) ' GAS PHASE VELOCITY ', AVEGASVEL
WRITE(81,*) ' ' 
WRITE(81,*) ' AVERAGE LIQUID VELOCITY IN RISER ', AVELIQUIDVEL
WRITE(81,*) ' ' 
WRITE(81,*) ' AVERAGE LIQUID VELOCITY IN DOWNCOMER ', AVELIQVELD
WRITE(81,*) ' ' 
WRITE(81,*) ' SLIP VELOCITY ', SLIPVEL
WRITE(81,*) ' ' 
WRITE(81,*) ' LIQUID LEVEL ', RLIQLEVEL
WRITE(81,*) ' ' 
WRITE(81,*) ' TOTAL LIQUID ', TOTALLIQ
WRITE(81,*) ' ' 
WRITE(81,*) ' LIQUID UNDER FREE SURFACE ', REALIQ

C
CENDIF
C***************************************************************************
C
C RETURN
C
END

C***************************************************************************
C
C SUBROUTINE USRINIT(U,V,W,P,VFRAC,DEN,VIS,TE,ED,RS,T,H,RF,SCAL
+   ,CONV,XC,YC,ZC,XP,YP,ZP
+   ,VOL,AREA,VPOR,ARPOR,WFACT,DISMAL,IP
+   ,IBLK,IPVERT,IPNODN/IPFACN,IPNODB/IPFACB
+   ,WORK,WORK,CWORK)
C
C***************************************************************************
C
C UTILITY SUBROUTINE FOR USER-SUPPLIED INITIAL FIELD.
C
C***************************************************************************
C
C THIS SUBROUTINE IS CALLED BY THE FOLLOWING SUBROUTINE
C
C USR INIT
Source files and user subroutines for the CFX 4.3 code

C
C******************************************************************************
C
C SUBROUTINE ARGUMENTS
C
C U - U COMPONENT OF VELOCITY
C V - V COMPONENT OF VELOCITY
C W - W COMPONENT OF VELOCITY
C P - PRESSURE
C VFRAC - VOLUME FRACTION
C DEN - DENSITY OF FLUID
C VIS - VISCOSITY OF FLUID
C TE - TURBULENT KINETIC ENERGY
C ED - EPSILON
C RS - REYNOLD STRESSES
C T - TEMPERATURE
C H - ENTHALPY
C RF - REYNOLD FLUXES
C SCAL - SCALARS (THE FIRST 'NCONC' OF THESE ARE MASS FRACTIONS)
C CONV - CONVECTION COEFFICIENTS
C XC - X COORDINATES OF CELL CORNERS
C YC - Y COORDINATES OF CELL CORNERS
C ZC - Z COORDINATES OF CELL CORNERS
C XP - X COORDINATES OF CELL CENTRES
C YP - Y COORDINATES OF CELL CENTRES
C ZP - Z COORDINATES OF CELL CENTRES
C VOL - VOLUME OF CELLS
C AREA - AREA OF CELLS
C VPOR - POROUS VOLUME
C ARPOR - POROUS AREA
C WFACT - WEIGHT FACTORS
C DISWAL - DISTANCE OF CELL CENTRE FROM WALL
C
C IPT - 1D POINTER ARRAY
C IBLK - BLOCK SIZE INFORMATION
C IPVERT - POINTER FROM CELL CENTERS TO 8 NEIGHBOURING VERTICES
C IPMNDN - POINTER FROM CELL CENTERS TO 6 NEIGHBOURING CELLS
C IPFACN - POINTER FROM CELL CENTERS TO 6 NEIGHBOURING FACES
C IPMNDF - POINTER FROM CELL FACES TO 2 NEIGHBOURING CELL CENTERS
C IPNFAC - POINTER FROM BOUNDARY CENTERS TO CELL CENTERS
C IPFACB - POINTER FROM BOUNDARY CENTERS TO BOUNDARY FACES
C
C WORK - REAL WORKSPACE ARRAY
C IWORK - INTEGER WORKSPACE ARRAY
C CWORK - CHARACTER WORKSPACE ARRAY
C
C SUBROUTINE ARGUMENTS PRECEDED WITH A ' * ' ARE ARGUMENTS THAT MUST
C BE SET BY THE USER IN THIS ROUTINE.
C
C LOGICAL VARIABLE LRDISK IN COMMON BLOCK ILOGC INDICATES WHETHER
C THE RUN IS A RESTART AND CAN BE USED SO THAT INITIAL INFORMATION
C IS ONLY SET WHEN STARTING A RUN FROM SCRATCH.
C
C NOTE THAT OTHER DATA MAY BE OBTAINED FROM CFX-F3D USING THE
C ROUTINE GETADD, FOR FURTHER DETAILS SEE THE VERSION 4
C USER MANUAL.
C
C********************************************************************************
LOGICAL LDEN, LVIS, LTURB, LTEMP, LBUOY, LSCAL, LCOMP
+ , LRECT, LCYN, LAXIS, LPOROS, LTRANS
LOGICAL LRDISK, LWDISK

C CHARACTER*(*) CWORK

C++++++++++++++++ USER AREA 1 +++++++++++++++++++++++++++++++++++++++++
C AREA FOR USERS EXPLICITLY DECLARED VARIABLES
C
C++++++++++++++++ END OF USER AREA 1 ++++++++++++++++++++++++++++++++++
C
COMMON
+ /ALL/ NBLOCK, NCELL, NBDRY, NNODE, NFACE, NVERT, NDIM
+ /ALLWK/ NRWS, NIWS, NCWS, IWFRE, IWPRE, IWPRE
+ /ADDIMS/ NPHASE, NSCAL, NVAR, NPROP
+ , NDVAR, NPROP, NDXNN, NDEGEOM, NCODE, NLIST, NLIST, NTOPOL
+ /CHUSR/ IVERS, IUCALL, IUSED
+ /DEVICE/ NREAD, NWRITE, NWDISK, NDISK
+ /IDOM/ ILEM, JLEN
+ /ILOGC/ LRDISK, LWDISK
+ /LOGIC/ LDEN, LVIS, LTURB, LTEMP, LBUOY, LSCAL, LCOMP
+ , LRECT, LCYN, LAXIS, LPOROS, LTRANS
+ /MLTGRD/ MLEVEL, NLEVEL, ILEVEL
+ /SGLDBL/ IFLGPR, ICHKPR
+ /TRANSI/ NSTEP, KSTEP, MF, INCORE
+ /TRANSR/ TIME, DT, DTINVF, TPARM

C++++++++++++++++ USER AREA 2 +++++++++++++++++++++++++++++++++++++++++
C AREA FOR USERS TO DECLARE THEIR OWN COMMON BLOCKS
C THESE SHOULD START WITH THE CHARACTERS 'UC TO ENSURE
C NO CONFLICT WITH NON-USER COMMON BLOCKS
C
C++++++++++++++++ END OF USER AREA 2 ++++++++++++++++++++++++++++++++++
C
DIMENSION
+ U(NNODE, NPHASE), V(NNODE, NPHASE), W(NNODE, NPHASE)
+ P(NNODE, NPHASE), VFRAC(NNODE, NPHASE)
+ , TE(NNODE, NPHASE), ED(NNODE, NPHASE), RS(NNODE, NPHASE, 6)
+ , T(NNODE, NPHASE), H(NNODE, NPHASE), RF(NNODE, NPHASE, 4)
+ , SCAL(NNODE, NPHASE, NSCAL)
+ , DEN(NNODE, NPHASE), VIS(NNODE, NPHASE), CONV(NFACE, NPHASE)
DIMENSION
+ IC(NVERT), TC(NVERT), XP(NNODE), ZP(NNODE), ZP(NNODE)
+ , VOL(NCELL), AREA(NFACE, 3), VFOR(NCELL), ARFOR(NFACE, 3)
+ , NFACT(NFACE), DISWAL(NCELL)
DIMENSION
+ IPT(*), IBLK(5, NBLOCK)
+ , IPVERT(NCELL, 6), IPMOD(NCELL, 6), IPFACN(NCELL, 6), IPMODF(NFACE, 4)
+ , IPMODL(NBDRT, 4), IFACE(NBDRY)
DIMENSION
+ IWORK(NIWS), WORK(NRWS), CWORK(NCWS)

C++++++++++++++++ USER AREA 3 +++++++++++++++++++++++++++++++++++++++++
C AREA FOR USERS TO DIMENSION THEIR ARRAYS
C
C++++++++++++++++ AREA FOR USERS TO DEFINE DATA STATEMENTS
C++++++++++++++++ END OF USER AREA 3 ++++++++++++++++++++++++++++

C STATEMENT FUNCTION FOR ADDRESSING
IP(I,J,K)=IPT((K-1)*ILEN*JLEN+(J-1)*ILEN+I)

C VERSION NUMBER OF USER ROUTINE AND PRECISION FLAG

IVERS=3
ICHKPR = 1

C++++++++++++++++ END OF USER AREA 4 ++++++++++++++++++++++++++++++++++

C TO USE THIS USER ROUTINE FIRST SET IUSED=1

IUSED=1

C++++++++++++++++ END OF USER AREA 5 +++++++++++++++++++++++++++++++++

IF (IUSED.EQ.0) RETURN

C FRONTEND CHECKING OF USER ROUTINE
IF (IUCALL.EQ.0) RETURN

C++++++++++++++++ END OF USER AREA 5 +++++++++++++++++++++++++++++++++

RETURN
END

RETURN
END
SUBROUTINE USR SRC (IEQN, ICALL, CHAME, CALIAS, AM, SP, SU, CONV + , U, V, W, P, VFAC, DEN, VIS, TE, ED, RS, T, H, RF, SCAL + , XP, YP, ZP, VOL, AREA, VPO, ARPO, WFACT, IPT + , IBLK, IPVERT, IPNOD, IPFACN, IPNODF, IPNODB, IPFACB + , WORK, IWORK, CWORK)

C+++++++++++++++++++++++++++++++++++

C UTILITY SUBROUTINE FOR USER-SUPPLIED SOURCES

C >>> IMPORTANT <<<
C >>> USER MAY ONLY ADD OR ALTER PARTS OF THE SUBROUTINE WITHIN <<<
C >>> DESIGNATED USER AREAS <<<

C+++++++++++++++++++++++++++++++++++

C THIS SUBROUTINE IS CALLED BY THE FOLLOWING SUBROUTINES
C CUSR SCDF SCDS SCED SCENRG SCHF SCMOM SCPCE SCSCAL
C SCTE SCVF

C+++++++++++++++++++++++++++++++++++

C CREATED
C 08/03/90 ADB
C MODIFIED
C 04/03/91 ADB ALTERED ARGUMENT LIST.
C 28/08/91 IRH NEW STRUCTURE
C 28/09/91 IRH CHANGE EXAMPLE + ADD COMMON BLOCKS
C 10/02/92 PHA UPDATE CALLED BY COMMENT, ADD RF ARGUMENT,
C CHANGE LAST DIMENSION OF RS TO 6 AND IVERS TO 2
C 03/06/92 PHA ADD PRECISION FLAG AND CHANGE IVERS TO 3
C 23/11/93 CSH EXPLICITLY DIMENSION IPVERT ETC.
Source files and user subroutines for the CFX 4.3 code

C 07/12/93 NSW INCLUDE CONV IN ARGUMENT LIST AND CHANGE IVERS
C TO 4
C 03/02/94 PHA CHANGE FLOW3D TO CFDS-FLOW3D
C 03/03/94 FHW CORRECTION OF SPELLING MISTAKE
C 08/03/94 NSW CORRECT SPELLING
C 09/08/94 NSW CORRECT SPELLING
C MOVE 'IF(IUSED.EQ.0) RETURN' OUT OF USER AREA.
C INCLUDE COMMENT ON MASS SOURCES.
C 19/12/94 NSW CHANGE FOR CFX-F3D
C 02/07/97 NSW UPDATE FOR CFX-4
C
C*******************************************************************************
C SUBROUTINE ARGUMENTS
C
C IEQN  - EQUATION NUMBER
C ICALL - SUBROUTINE CALL
C CNAME - EQUATION NAME
C CALLAS - ALIAS OF EQUATION NAME
C AM    - OFF DIAGONAL MATRIX COEFFICIENTS
C SU    - SU IN LINEARISATION OF SOURCE TERM
C SP    - SP IN LINEARISATION OF SOURCE TERM
C CONV  - CONVECTION COEFFICIENTS
C U     - U COMPONENT OF VELOCITY
C V     - V COMPONENT OF VELOCITY
C W     - W COMPONENT OF VELOCITY
C P     - PRESSURE
C VFRAC - VOLUME FRACTION
C DEN   - DENSITY OF FLUID
C VIS   - VISCOSITY OF FLUID
C TE    - TURBULENT KINETIC ENERGY
C ED    - EPSILON
C RS    - REYNOLD STRESSES
C T     - TEMPERATURE
C H     - ENTHALPY
C RF    - REYNOLD FLUXES
C SCAL  - SCALARS (THE FIRST 'NCONC' OF THESE ARE MASS FRACTIONS)
C XP    - X COORDINATES OF CELL CENTRES
C YP    - Y COORDINATES OF CELL CENTRES
C ZP    - Z COORDINATES OF CELL CENTRES
C VOL   - VOLUME OF CELLS
C AREA  - AREA OF CELLS
C VPOR  - POROUS VOLUME
C ARPOR - POROUS AREA
C WFACT - WEIGHT FACTORS
C IPT   - 1D POINTER ARRAY
C IBLK  - BLOCK SIZE INFORMATION
C IPVERT - POINTER FROM CELL CENTERS TO 8 NEIGHBOURING VERTICES
C IPFACN - POINTER FROM CELL CENTERS TO 6 NEIGHBOURING CELLS
C IPFACM - POINTER FROM CELL CENTERS TO 6 NEIGHBOURING FACES
C IPFACF - POINTER FROM CELL FACES TO 2 NEIGHBOURING CELL CENTERS
C IPFACB - POINTER FROM BOUNDARY CENTERS TO CELL CENTERS
C IPFAFC - POINTER FROM BOUNDARY CENTERS TO BOUNDARY FACES
C WORK - REAL WORKSPACE ARRAY
C IWORK - INTEGER WORKSPACE ARRAY
C CWORK - CHARACTER WORKSPACE ARRAY
C
C SUBROUTINE ARGUMENTS PRECEDED WITH A '*' ARE ARGUMENTS THAT MUST
C BE SET BY THE USER IN THIS ROUTINE.
C
C NOTE THAT WHEN USING MASS SOURCES, THE FLOWS THROUGH MASS FLOW
C BOUNDARIES ARE UNCHANGED. THE USER SHOULD THEREFORE INCLUDE AT
C LEAST ONE PRESSURE BOUNDARY FOR SUCH A CALCULATION.
C
C NOTE THAT OTHER DATA MAY BE OBTAINED FROM CFX-4 USING THE
C ROUTINE GETADD, FOR FURTHER DETAILS SEE THE VERSION 4
Source files and user subroutines for the CFX 4.3 code

C USER MANUAL.
C
C*******************************************************************************
C
LOGICAL LDEN, LVIS, LTURB, LTEMP, LBUGY, LSCAL, LCOMP
+ , LRECT, LCYN, LAXIS, LPGRGS, LTRANS
C
CHARACTER*(*) CWGRK
CHARACTER CNAME*6, CALIAS*24
C
C+++++++++++++++++ USER AREA 1 +++++++++++++++++++++++++++++++++++++++++
C AREA FOR USERS EXPLICITLY DECLARED VARIABLES
C
C+++++++++++++++++ END GF USER AREA 1 ++++++++++++++++++++++++++++++++++
C
COMMON
+ /ALL/ NBLOCK, NCELL, NBODY, NNODE, NFACE, NVERT, NDIM
+ /ALLWRK/ NWRS, NWKS, NCWS, IWRFE, IWFE, IWCFRE
+ /ADDIMG/ NPHASE, NSCAL, NVAR, NFGDP
+ , NDVAR, NPDNP, NDNN, NDDEG, NDcoef, NILIST, NRLIST, NTOPOL
+ /CHKUSR/ IVERS, IUCALL, IUSED
+ /DEVICE/ NREAD, NWRITE, NRDISK, NWDISK
+ /IDUM/ ILEN, JLEN
+ /LOGIC/ LDEN, LVIS, LTURB, LTEMP, LBUGY, LSCAL, LCOMP
+ , LRECT, LCYN, LAXIS, LPGRGS, LTRANS
+ /MGTGRD/ MLEVEL, NLEVEL, ILEVEL
+ /SGLDBL/ IFLGPR, ICHKPR
+ /SPARM/ SMALL, SGRMAX, NITER, INDPI, MAXIT, NGDREF, NDMIN
+ /TRANSI/ NSTEP, KSTEP, MF, INCORE
+ /TRANSR/ TIME, DT, DTINVF, TPARM
C
C+++++++++++++++++ USER AREA 2 +++++++++++++++++++++++++++++++++++++++++
C AREA FOR USERS TO DECLARE THEIR OWN COMMON BLOCKS
C THESE SHOULD START WITH THE CHARACTERS ‘UC’ TO ENSURE
C NO CONFLICT WITH NON-USER COMMON BLOCKS
C
C+++++++++++++++++ END OF USER AREA 2 ++++++++++++++++++++++++++++++++++
C
DIMENSION AM(NCELL, 6, NPHASE), SP(NCELL, NPHASE), SU(NCELL, NPHASE)
+ , CONV(NFACE, NPHASE)
C
DIMENSION
U(NNODE, NPHASE), V(NNODE, NPHASE), W(NNODE, NPHASE), P(NNODE, NPHASE)
+ , VFRAC(NNODE, NPHASE), DEN(NNODE, NPHASE), VIS(NNODE, NPHASE)
+ , T(NNODE, NPHASE), H(NNODE, NPHASE), AF(NNODE, NPHASE, 4)
+ , SCAL(NNODE, NPHASE, NSCAL)
C
DIMENSION
XP(NNODE), YP(NNODE), ZP(NNODE)
+ , VOL(NCELL), AREA(NFACE, 3), VPOR(NCELL), ARPOR(NFACE, 3)
+ , NFAC(NFACE)
+ , IPT(5), IBLK(5, NBLOCK)
+ , IPFVERT(NCELL, 4), IPMDON(NCELL, 6), IPFACN(NCELL, 6), IPMDP(NFACE, 4)
+ , IPMDN(NBODY, 4), IPFACNB(NODY)
+ , TWORK(*), WORK(*), CWORK(*)
C
C+++++++++++++++++ USER AREA 3 ++++++++++++++++++++++++++++++++++++++++++
C AREA FOR USERS TO DIMENSION THEIR ARRAYS
C
C+++++++++++++++++ END OF USER AREA 3 ++++++++++++++++++++++++++++++++++
C
C STATEMENT FUNCTION FOR ADDRESSING
IP(T, J, K) = IPT((K-1)*ILEN*JLEN+(J-1)*ILEN+I)
C
C VERSION NUMBER OF USER ROUTINE AND PRECISION FLAG
IVERS=4
ICHKPR = 1

C+++++++++++++++++ USER AREA 4 +++++++++++++++++++++++++++++++++++++++++
C---- TO USE THIS USER ROUTINE FIRST SET IUSED=1
C
IUSED=1
C
C+++++++++++++++++ END OF USER AREA 4 ++++++++++++++++++++++++++++++++++++++
C
IF (IUSED.EQ.0) RETURN
C
C---- FRONTEND CHECKING OF USER ROUTINE
IF (IUCALL.EQ.0) RETURN
C
C---- ADD TO SOURCE TERMS
IF (ICALL.EQ.1) THEN
C
ENDIF
C
C+++++++++++++++++ END OF USER AREA 5 ++++++++++++++++++++++++++++++++++++++
ENDIF
C
C---- OVERWRITE SOURCE TERMS
IF (ICALL.EQ.2) THEN
C
C+++++++++++++++++ USER AREA 6 ++++++++++++++++++++++++++++++++++++++++++
C
C
ENDIF
RETURN
END
Appendix C

Fortran 90 source code

Several Fortran programs have been developed to implement the computational methods presented in this work. The programs are divided into three groups:

**PBE:** A set of subroutines for numerical solution of the population balance equation according to the finite element scheme presented in chap. 3.

**Network:** The computational implementation of the generalised compartmental model (chap. 2).

**Interface:** A PDE solver for the equations describing the interface (chap. 2).

Apart from the above, several auxiliary sets of subroutines are used to define problem-specific models:

**Reaction:** Defines the reaction kinetics.

**Utilities:** Provides functions for variables such as supersaturation, ionic strength etc.

**Bulk:** Provides a link with the ODE solver, in our case the LSODE (Hindmarsh, 1983).

**Main:** Initiates the computation.

C.1 PBE

```fortran
!****************************************************************************
!
! FILE PBE_module.f90
!
! Calculates the nucleation, growth, aggregation and breakage source terms
! for temporal integration of the population densities at the nodal points of the grid
! discretised with the first-order conservative finite element collocation method
!
! This file contains:
! - Module PBE_module: all general declarations
! - Subroutine PBE_source: source terms due to finite element discretisation
! - Subroutine mesh: generates mesh and performs all preliminary operations
! - Subroutine agg_kernel_lib: library of aggregation kernels
! - Subroutine br_kernel_lib: library of breakage kernels
```

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! - Subroutine PBE_initdata: read values of particulate parameters
! - Subroutine PBE_initarrays: read values of particulate parameters and initialise distributions
!
! By Stelios Rigopoulos
! 11/01/2002
! 18/01/2002: first version of modular program
! 19/04/2002 new version
!
!*******************************************************************************

module PBE_module

*******************************************************************************

declaration of common variables related to grid and kernels
!
! by Stelios Rigopoulos
! 17/12/2001
! 19/04/2002 new version
!
!*******************************************************************************

implicit none
real(8) :: agg_kernel, br_kernel_const, br_kernel_exp, br_dist_parabol
real(8) :: br_dist_gamma, br_dist_delta, br_dist_alpha
real(8) :: Nuc_ss_exp, G_ss_exp, Nuc_ss_coeff, G_ss_coeff
real(8) :: inc_v_nuc, v_0, threshold
real(8) :: NO, v0
real(8) :: G_coeff1, G_coeff2, G_coeff3, nuc, agg_kernel_const
integer :: agg_kernel_type, breakage_kernel_type, breakage_dist_type, growth_kernel_type
integer :: speedupPBE, meshflag, initflag, inputflag
integer :: m
real(8), allocatable, dimension(:) :: v, dv, v_exp13, v_exp23
integer, allocatable, dimension(:) :: vhalf
integer, allocatable, dimension(:, :, :) :: comp
real(8), allocatable, dimension(:, :, :) :: node, nodec
real(8), allocatable, dimension(:, :, :) :: agg_kernel_birth
real(8), allocatable, dimension(:, :, :) :: agg_kernel_death
real(8), allocatable, dimension(:, :, :) :: br_kernel_birth

end module PBE_module

*******************************************************************************

*******************************************************************************

subroutine PBESource(index, ni, Agg_birth, Agg_death, B_Birth, BJDeath, Growth, Nuc_Birth)

*******************************************************************************

! Calculates the source term for temporal integration of the population balance
! Implementation for collocation on finite elements with linear base functions
! Data, kernels etc to be supplied by module "kinetics"
! Discrete pointwise evaluation of aggregation birth term
!
! This implementation is for:
! Any volume-dependent growth function if provided
! Any up to third order linear aggregation kernel if provided
! Any up to fifth order linear breakage kernel
! Higher or nonlinear order kernels: may be subject to error
!
! How kernels are set:
!  - agg_kemel_order: order of the aggregation kernel, if linear can handle up to 3rd
!  - br_kemel_order: total order of the breakage birth term = kernel + distribution + 1
!
* Aggregation kernels
  - Volume-independent kernel: set agg_kemel_type to 1
  - Sum kernel: set agg_kemel_type to 2
  - Product kernel: set agg_kemel_type to 3
  - Smoluchowski kernel: set agg_kemel_type to 4
  - Gravitational kernel: set agg_kemel_type to 5
*
* Growth kernels
  - Volume-independent kernel: set growth_kemel_type to 1
  - Volume-deoendent due to $G(v) = v^n$: set growth_kemel_type to 2
  - Volume-dependent due to Randolph and Larson function: set growth_kemel_type to 3
*
* Breakage kernels
  - Power law kernel, uniform binary breakage: set breakage_kemel_type to 1
  - Power law kernel, parabolic binary breakage: set breakage_kemel_type to 2
  - Power law kernel, multiple breakage according to simplified Ziff function: 3
  - Power law kernel, multiple breakage according to generalised Ziff function: 4
  - Exponential kernel: set breakage_kemel_type to 5
*
*
* * * * * * * * *

By Stelios Rigopoulos

Development history:
30/10/2001: initial version
16/11/2001: improved conservation algorithm
02/12/2001: added breakage
03/12/2001: added product kernel
05/12/2001: new structure
11/12/2001: corrections
18/01/2002: new version for modular program
*
* Main variables
*
! m number of points
! index point to be collocated
! ni population density
! ni_in inlet population density
! v independent variable - volume
! dv lengths of intervals
*
	***************************************************************

use PBE_module
!use reaction_module !for n
implicit none

integer, intent(in) :: index
real(8), intent(out) :: Agg_Birth,Agg_Death,B_birth,B_Death,Growth,Nuc_Birth
real(8), dimension(0:m), intent(in) :: ni
!real(8), dimension(n), intent(in) :: C

integer :: i,k
integer :: si
real(8) :: term1,G_term1,G_term2
real(8) :: ni,n3,nk1,nk3,v2,n2,vk2,nk2,dv13
!real(8) :: G_coeffl,G_coeff2,G_coeff3,nuc
integer tflag
common tflag

!---------------------------------------------------------------------------------------------------------------------------

!Growth term
!Dependency on environmental variables
!Call Growth_coeff(C)
!---------------------------------------------------------------------------------------------------------------------------
if (growth_kernel_type==1) then
  !Volume-independent growth
  G_term1 = G_coeff1
  G_term2 = 0.
else if (growth_kernel_type==2) then
  !Volume-dependent growth, simple power law
  G_term1 = G_coeff1*(v(index)**G_coeff2)
  G_term2 = G_coeff1*G_coeff2*(v(index)**(G_coeff2-1.))
else if (growth_kernel_type==3) then
  !Volume-dependent growth, Randolph and Larson function
  G_term1 = G_coeff1*(1 + G_coeff2*v(index))**G_coeff3
  G_term2 = G_coeff3*G_coeff1*G_coeff2*((1.+G_coeff2*v(index))**(G_coeff3-1.))
end if
Growth = - G_term1*ni(index)/dv(index) - G_term2*ni(index)
if (index>1) then
  Growth = Growth + G_term1*ni(index-1)/dv(index) + dv(index-1)
end if
!---------------------------------------------------------------------------------------------------------------------------

!Nucleation term
!Dependency on environmental variables
!Call Nucleation_rate(C)
!---------------------------------------------------------------------------------------------------------------------------
if (index==1) then
  Nuc_Birth = Nuc*(2./(dv(1)+dv(2)))
else
  Nuc_Birth = 0.
end if
!---------------------------------------------------------------------------------------------------------------------------

!Aggregation term
!Dependency on environmental variables
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!Call Aggregation_rate(C,agg_kernel)
!
!---------------------------------------------------------
!
!Aggregation death term
Agg_Death = 0.
if ((agg_kernel_const>0.0).and.(tflag==1)) then
  do i=1,m
    if (agg_kernel_type==1) then
      Agg_Death = Agg_Death + dv(i) * 5.D-1 * ( ni(i-l) + ni(i))
    else
      dv13 = v(i)-v(i-1)
v2 = 0.5*(v(i-1)+v(i))
n2 = ni(i-l)*v(i)/dv13 - ni(i-l)*v2/dv13 + ni(i)*v2/dv13 - ni(i)*v(i-l)/dv13
      Agg_Death = Agg_Death + dv(i) * (  (1./6.)*ni(i-l)*agg_kernel_death(index,i,1) 
        & + (4./6.)*n2*agg_kernel_death(index,i,2) + 
        & (1./6.)*ni(i)*agg_kernel_death(index,i,3) )
    end if
  end do
Agg_Death = - Agg_Death*agg_kernel*ni(index)
end if
!
!---------------------------------------------------------
!
!Aggregation birth term
Agg_Birth = 0.
if ((index>0).and.(agg_kernel_const>0.0).and.(tflag==1)) then
  !Integrate over intervals
  do i=1,vhalf(index)
    if (i==vhalf(index)) then
      if ( abs( v(index)*0.5-v(vhalf(index)-1) )<threshold ) then
        !no need to integrate element vhalf because v(index)/2 falls on a nodal point
        exit
      else
        si = 1
        n1 = ni(i-l)
n3 = ni(i-l)*v(i)/dv(i) - ni(i-l)*node(index,i,si,3)/dv(i) + 
        & ni(i)*node(index,i,si,3)/dv(i) - ni(i)*v(i-l)/dv(i)
nk3 = ni(i-l)*v(i)/dv(i) - ni(i-l)*nodec(index,i,si,3)/dv(i) + 
        & ni(i)*nodec(index,i,si,3)/dv(i) - ni(i)*v(i-l)/dv(i)
nk1 = n3
        n2 = n1*node(index,i,si,3)/dv(i) - n1*node(index,i,si,2)/dv(i) + 
        & n3*node(index,i,si,2)/dv(i) - n3*node(index,i,si,1)/dv(i)
nk2 = nk1*nodec(index,i,si,3)/dv(i) - nk1*nodec(index,i,si,2)/dv(i) + 
        & nk2*nodec(index,i,si,2)/dv(i)
Fortran 90 source code

```
& nk3*nodec(index,i,si,2)/dvk(index,i,si) - nk3*nodec(index,i,si,1)/dvk(index,i,si)
Agg_birth = Agg_birth + dvi(index,i,si)*
& ( (1./6.)*ni*mk3*agg_kernel_birth(index,i,si,1) +
& (4./6.)*n2*mk2*agg_kernel_birth(index,i,si,2) +
& (1./6.)*n3*mk1*agg_kernel_birth(index,i,si,3) )
end if
else

!Integrate over sub-intervals
si = 0

do k=comp(index,i,2),comp(index,i,1)

!Define the sub-interval limits
si = si + 1

!Case 1: if sub-interval is a whole interval
if (((comp(index,i,2)<k).and.(k<comp(index,i,1))) then
  nk1 = ni(k-1)
  nk3 = ni(k)
  n3 = ni(1-I)*v(i)/dv(i) - ni(I)*node(index,i,si,3)/dv(i) +
  & ni(i)*node(index,i,si,1)/dv(i) - ni(i)*v(i)/dv(i)
  n1 = ni(I)*v(i)/dv(i) - ni(I)*node(index,i,si,3)/dv(i) +
  & ni(i)*node(index,i,si,1)/dv(i) - ni(i)*v(i)/dv(i)
!Case 2: if both points lie in the same interval
else if (comp(index,i,1)==comp(index,i,2)) then
  n1 = ni(I-1)
  n3 = ni(i)
  nk1 = ni(k-1)*v(k)/dv(k) - ni(k-1)*node(index,i,si,1)/dv(k) +
  & ni(k)*node(index,i,si,1)/dv(k) - ni(k)*v(k)/dv(k)
  nk3 = ni(k-1)*v(k)/dv(k) - ni(k-1)*node(index,i,si,3)/dv(k) +
  & ni(k)*node(index,i,si,3)/dv(k) - ni(k)*v(k)/dv(k)
!Case 3: if sub-interval is the upper part of the primary interval
else if (k==comp(index,i,2)) then
  n1 = ni(I-1)*v(i)/dv(i) - ni(I-1)*node(index,i,si,1)/dv(i) +
  & ni(i)*node(index,i,si,1)/dv(i) - ni(i)*v(i)/dv(i)
  n3 = ni(i)
  nk1 = ni(k-1)*v(k)/dv(k) - ni(k-1)*node(index,i,si,1)/dv(k) +
  & ni(k)*node(index,i,si,1)/dv(k) - ni(k)*v(k)/dv(k)
  nk3 = ni(k)
!Case 4: if sub-interval is the lower part of the primary interval
else if (k==comp(index,i,1)) then
  n1 = ni(I-1)
  n3 = ni(I-1)*v(i)/dv(i) - ni(I-1)*node(index,i,si,3)/dv(i) +
  & ni(i)*node(index,i,si,3)/dv(i) - ni(i)*v(i)/dv(i)
  nk1 = ni(k-1)
  nk3 = ni(k-1)*v(k)/dv(k) - ni(k-1)*node(index,i,si,3)/dv(k) +
  & ni(k)*node(index,i,si,3)/dv(k) - ni(k)*v(k)/dv(k)
else
  write(*,'*') 'severe exception, quitting'
```

write(41,*), 'exception occurred at the sub-interval integration'
stop
end if

n2 = n1*node(index,i,si,3)/dvi(index,i,si) - n1*node(index,i,si,2)/dvi(index,i,si) + &
   n3*node(index,i,si,2)/dvi(index,i,si) - n3*node(index,i,si,1)/dvi(index,i,si)

nk2 = nk1*nodec(index,i,si,3)/dvk(index,i,si) - nk1*nodec(index,i,si,2)/dvk(index,i,si) + &
   nk3*nodec(index,i,si,2)/dvk(index,i,si) - nk3*nodec(index,i,si,1)/dvk(index,i,si)

Agg_birth = Agg_birth + dvi(index,i,si)*
            ( (1./6.)*ni(index,i,si,1) + &
             (4./6.)*n2*br_kernel_birth(index,i,si,2) + &
             (1./6.)*n3*br_kernel_birth(index,i,si,3) )
end do
end if
end do

! Total birth term
Agg_Birth = Agg_Birth*agg_kernel
end if

!-----------------------------------------------------------------------

! Breakage

! Dependency on environmental variables

! Call Breakage_rate(C,agg_kernel)

!-----------------------------------------------------------------------

! Breakage birth term

B_birth = 0.

if (abs(br_kernel_const)>0.) .and. (index<m) .and. (tflag==D) then
  do i=(index+i),m
    dv13 = v(i)-v(i-1)
    v2 = 0.5*(v(i-1)+v(i))
    n2 = ni(i-1)*v(i)/dv13 - ni(i-1)*v2/dv13 + ni(i)*v2/dv13 - ni(i)*v(i-1)/dv13
    B_birth = B_birth + dv13* ( (1./6.)*ni(i-1)*br_kernel_birth(index,i,1) + &
      (4./6.)*n2*br_kernel_birth(index,i,2) + &
      (1./6.)*ni(i)*br_kernel_birth(index,i,3) )
  end do
end if

B_Birth = B_Birth*br_kernel_const

!-----------------------------------------------------------------------

! Breakage Death term

B_Death = 0.

if (abs(br_kernel_const)>0.) .and. (tflag==D) then
  if (breakage_kernel_type/=5) then
B_Death = - br_kernel_const*ni(index)*((v(index)))**(br_kemel_exp))
else if (breakage_kernel_type==6) then
  B_Death = - ni(index)*br_kernel_const*exp(br_kernel_exp*v(index))
end if
end if

! For scaled variables
! Agg_Birth = Agg_Birth * 5.23E-25
! Agg_Death = Agg_Death * 5.23E-25
! Growth = Growth * ((5.23E-25)**(-1./3.))

end subroutine PBESource

*******************************************************************************

subroutine mesh()
*******************************************************************************

! Subroutine mesh
! Generates/reads the mesh
! Also determines complimentaries
!
! By Stelios Rigopoulos
! 31/10/2001 - initial version
! 17/11/2001 - added location of complimentaries
!
! m number of points
! ni population density
! v independent variable - volume
! dv lengths of intervals
! v_nuc size of nuclei - to be supplied by module kinetics
! comp(index,i,1) element with point which, when added to v(i-1), will form v(index)
! comp(index,i,2) element with point which, when added to v(i), will form v(index)
! inc factor by which the exponential grid is generated
!
*******************************************************************************

use PBEjmodule
implicit none

integer :: i,index,k,k1,k2,l,si
real(8) :: v_comp,temp,v_half
integer :: maxdim

write(*,*) 'Creating mesh...'
allocate (v(0:m))
allocate (dv(m))
allocate (vhalf(m))
allocate (comp(m,m,2))
! Create mesh
if (meshflag==1) then

! Option 1: Exponential mesh
   v(0) = v_0
   v(1) = v_nuc /10000.
   v(2) = v(1)+v_nuc !D-1 !2. !v_nuc
   do i=3,m
      v(i) = v_nuc + v_nuc*(inc**real(i-1))/(1-0)
   end do
else if (meshflag==2) then

! Option 2: Uniform mesh
   v(0) = v_0
   v(1) = v_nuc
   do i=2,m
      v(i) = v(i-1) + inc
   end do
else if (meshflag==3) then

! Option 3: read from disk
   do i=0,m
      Read(63,*) v(i)
   end do
end if

! Element length
   do i=1,m
      dv(i) = v(i)-v(i-1)
   end do

! The following operations need to be carried out only when aggregation is present
if (abs(aggl_kemel_const)>threshold) then

! Initialise necessary arrays
write(*,*) 'Calculating sub-intervals...'
maxdim = 2
allocate (node(m,m,maxdim,3))
allocate (nodec(m,m,maxdim,3))
allocate (dvk(m,m,maxdim))
allocate (dvi(m,m,maxdim))
allocate(agg_kemel_birth(m,m,maxdim,3))
allocate(agg_kemel_death(m,m,3))
do index = l,m
  do i=l,m
    comp(index,i,1) = 0.
    comp(index,i,2) = 0.
    do si=l,maxdim
      node(index,i,si,1) = 0.
      node(index,i,si,2) = 0.
      node(index,i,si,3) = 0.
      nodec(index,i,si,1) = 0.
      nodec(index,i,si,2) = 0.
      nodec(index,i,si,3) = 0.
      dvi(index,i,si) = 0.
      dvk(index,i,si) = 0.
      agg_kernel_birth(index,i,si,1) = 0.
      agg_kernel_birth(index,i,si,2) = 0.
      agg_kernel_birth(index,i,si,3) = 0.
    end do
    agg_kernel_death(index,i,1) = 0.
    agg_kernel_death(index,i,2) = 0.
    agg_kernel_death(index,i,3) = 0.
  end do
end do

! Locate the complementary points
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end if
end do

doi=1,index

! Determine complimentaries
k = 1
v_comp = v(index)-v(i-l)
do
if (abs(v_comp-v(k))<threshold) then
    comp(index,i,1) = k
    exit
else if (v_comp<v(k)) then
    comp(index,i,1) = k
    exit
else
    k = k+1
end if
end do
k = 1
v_comp = v(index)-v(i)
do
if (abs(v_comp-v(k))<threshold) then
    comp(index,i,2) = k + 1
    exit
else if (v_comp<v(k)) then
    comp(index,i,2) = k
    exit
else
    k = k+1
end if
end do
end do

!----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------

! Integrate over intervals
do index=l,m

do i=1,vhalf(index)
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if (i==vhalf(index)) then
  if ( abs( v(index)*0.5-v(vhalf(index)-1) )<threshold ) then
    ! no need to integrate element vhalf because v(index)/2 falls on a nodal point
    exit
  else
    si=1
    node(index,i,si,1)= v(i-1)
    node(index,i,si,3)=v(index)*0.5
    nodec(index,i,si,1) = node(index,i,si,3) = v(index)*0.5 + ( v(index)*0.5-v(i-1) )
    dvk(index,i,si) = nodec(index,i,si,3) - nodec(index,i,si,1)
    dvi(index,i,si) = node(index,i,si,3) - node(index,i,si,1)
    nodec(index,i,si,2) = 0.5*(nodec(index,i,si,1)+nodec(index,i,si,3))
    node(index,i,si,2) = 0.5*(node(index,i,si,1)+node(index,i,si,3))
  end if
else
  ! Integrate over sub-intervals
  si = 0
  do k=comp(index,i,2),comp(index,i,1)
    ! Define the sub-interval limits
    si = si + 1
    ! Case 1: if sub-interval is a whole interval
    if ((comp(index,i,2)<k).and.(k<comp(index,i,1))) then
      nodec(index,i,si,1)=v(k-1)
      nodec(index,i,si,3)=v(k)
      node(index,i,si,1) = v(i) - ( v(k) - (v(index)-v(i)) )
      node(index,i,si,3) = v(i) - ( v(index)-v(i) )
    ! Case 2: if both points lie in the same interval
    else if (comp(index,i,1)==comp(index,i,2)) then
      node(index,i,si,1)= v(i-1)
      node(index,i,si,3)= v(i)
      nodec(index,i,si,1) = v(index)-v(i)
      nodec(index,i,si,3) = v(index) - v(i-1)
    ! Case 3: if sub-interval is the upper part of the primary interval
    else if (k==comp(index,i,2)) then
      node(index,i,si,1)= v(i) - (v(k) - (v(index)-v(i)))
      node(index,i,si,3)= v(index)-v(i)
      nodec(index,i,si,1) = v(index) - v(k)
    ! Case 4: if sub-interval is the lower part of the primary interval
    else if (k==comp(index,i,1)) then
      node(index,i,si,1)= v(i-1)
```fortran
node(index,i,si,3) = v(i-1) + ((v(index)-v(i-1))-v(k-1))
nodec(index,i,si,1) = v(k-1)
nodec(index,i,si,3) = v(index) - v(i-1)
else
    write(*,*) 'severe exception, quitting'
    write(*,*) 'exception occurred at the calculation of sub-interval nodes'
    stop
end if

dvk(index,i,si) = nodec(index,i,si,3) - nodec(index,i,si,1)
dvi(index,i,si) = node(index,i,si,3) - node(index,i,si,1)
nodec(index,i,si,2) = 0.5*(nodec(index,i,si,1)+nodec(index,i,si,3))
node(index,i,si,2) = 0.5*(node(index,i,si,1)+node(index,i,si,3))
end do
end if
end do
end do
!--------------------------------------------------------------

! Calculate kernels

! Value of kernel for birth term
write(*,*) 'computing aggregation kernels...'
do index=1,m
    do i=1,vhalf(index)
        if (i==vhalf(index)) then
            if ( abs( v(index)*0.5-v(vhalf(index)-1) )<threshold ) then
                ! no need to integrate element vhalf because v(index)/2 falls on a nodal point
                exit
            else
                si=1
                call agg_kernel_lib(index,i,si,1)
            end if
        else
            si = 0
            do k = comp(index,i,2),comp(index,i,1)
                si = si + 1
                call agg_kernel_lib(index,i,si,1)
            end do
        end if
    end do
end do
```
end do  
  ! Value of kernel for death term  
  do index=1,m
    do i=1,m
      si=0
      call agg_kernel_lib(index,i,si,0)
    end do
  end do 
end do  
  ! Optional - min aggregation size  
  do index=1,m
    if (v(index)<((i./6.)*3.14*(1.5D-6)**3.)) then 
      ! call agg_kernel_min(index,1,maxdim,0)
    end if
  end do 
end do 
do index=1,m
  do i=1,m
    if (((v(i)+v(index))<((i./6.)*3.14*(1.5D-6)**3.))) then 
      ! call agg_kernel_min(index,0,maxdim,i)
    end if
  end do
end do  
!------------------------------------------------------------------------------------------------------------------------
write(*,*) 'preliminary aggregation operations complete'
if (abs(br_kemel_const)>threshold) then
do index=1,m-1
  do i=(index+1),m
    call br_kemel_lib(index,i)
  end do
end do
write(*,*) 'preliminary breakage operations complete'
end if

end subroutine mesh

!**********************************************************************

Subroutine agg_kernel_lib(index,i,si,flag)

!**********************************************************************

use PBEjmodule
implicit none
integer, intent(in) :: i,index,si,flag
integer :: k
real(8) :: term,eff,Expfac,expexp
real(8) :: lamda,gamma

Expfac = 5.010
Expexp = 0.65
lamda = 3.05
gamma = 1.4

if (flag==1) then
  !Value of kernel for birth term
  if (agg_kernel_type==1) then
    !Volume-independent kernel
    agg_kernel_birth(index,i,si,1) = 1.
    agg_kernel_birth(index,i,si,2) = 1.
    agg_kernel_birth(index,i,si,3) = 1.
  else if (agg_kernel_type==2) then

! Sum kernel
agg_kernel_birth(index,i,si,1) = node(index,i,si,1) + nodec(index,i,si,3)
agg_kernel_birth(index,i,si,2) = node(index,i,si,2) + nodec(index,i,si,2)
agg_kernel_birth(index,i,si,3) = node(index,i,si,3) + nodec(index,i,si,1)

else if (agg_kernel_type==3) then
  ! Product kernel
  agg_kernel_birth(index,i,si,1) = node(index,i,si,1) * nodec(index,i,si,3)
  agg_kernel_birth(index,i,si,2) = node(index,i,si,2) * nodec(index,i,si,2)
  agg_kernel_birth(index,i,si,3) = node(index,i,si,3) * nodec(index,i,si,1)

else if (agg_kernel_type==4) then
  ! Fluid shear kernel
  term = ( node(index,i,si,1)**1.3 + nodec(index,i,si,3)**1.3 )
  agg_kernel_birth(index,i,si,1) = term*term*term
  term = ( node(index,i,si,2)**1.3 + nodec(index,i,si,2)**1.3 )
  agg_kernel_birth(index,i,si,2) = term*term*term
  term = ( node(index,i,si,3)**1.3 + nodec(index,i,si,1)**1.3 )
  agg_kernel_birth(index,i,si,3) = term*term*term

else if (agg_kernel_type==5) then
  ! Differential force kernel
  agg_kernel_birth(index,i,si,1) = k
  & ( (node(index,i,si,1)**1.3 + nodec(index,i,si,3)**1.3)**2 * &
  & abs ( (node(index,i,si,1)**2.3 - nodec(index,i,si,3)**2.3) ) )
  agg_kernel_birth(index,i,si,2) = k
  & ( (node(index,i,si,2)**1.3 + nodec(index,i,si,2)**1.3)**2 * &
  & abs ( (node(index,i,si,2)**2.3 - nodec(index,i,si,2)**2.3) ) )
  agg_kernel_birth(index,i,si,3) = k
  & ( (node(index,i,si,3)**1.3 - nodec(index,i,si,1)**1.3)**2 * &
  & abs ( (node(index,i,si,3)**2.3 - nodec(index,i,si,3)**2.3) ) )

else if (agg_kernel_type==6) then
  ! Thompson kernel
  agg_kernel_birth(index,i,si,1) = k
  & (node(index,i,si,1)-nodec(index,i,si,3))**2 / node(index,i,si,1)+nodec(index,i,si,3)
  agg_kernel_birth(index,i,si,2) = k
  & (nodec(index,i,si,1)-node(index,i,si,2))**2 / nodec(index,i,si,2)+node(index,i,si,2)
  agg_kernel_birth(index,i,si,3) = k
  & (node(index,i,si,3)-nodec(index,i,si,3))**2 / nodec(index,i,si,3)+nodec(index,i,si,3)

else if (agg_kernel_type==7) then
  ! Fluid shear kernel, volume-dependent efficiency
  term = ( node(index,i,si,1)**1.3 + nodec(index,i,si,3)**1.3 )
  agg_kernel_birth(index,i,si,1) = term*term*term / (node(index,i,si,1)+nodec(index,i,si,3))
  term = ( node(index,i,si,2)**1.3 + nodec(index,i,si,2)**1.3 )
  agg_kernel_birth(index,i,si,2) = term*term*term / (node(index,i,si,2)+nodec(index,i,si,2))
  term = ( node(index,i,si,3)**1.3 + nodec(index,i,si,1)**1.3 )
agg_kernel_birth(index,i,si,3) = term*term*term/(node(index,i,si,3)+nodec(index,i,si,1))

else if (agg_kernel_type==8) then

!Constant kernel, volume-dependent efficiency
agg_kernel_birth(index,i,si,1) = l./v(index)
agg_kernel_birth(index,i,si,2) = l./v(index)
agg_kernel_birth(index,i,si,3) = l./v(index)

else if (agg_kernel_type==9) then

!Fluid shear kernel, volume-dependent efficiency function

term = ( node(index,i,si,1)**(1./3.)+nodec(index,i,si,3)**(1./3.) )
! eff = 1. - ( (node(index,i,si,1)+nodec(index,i,si,3))/Vcrit**Expcri2 )
! eff = (1./v(index))**1.
eff = exp(-Expfac*v(index)**expexp)
if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,1) = term*term*term*eff

term = ( node(index,i,si,2)**(1./3.)+nodec(index,i,si,2)**(1./3.) )
! eff = 1. - ( (node(index,i,si,2)+nodec(index,i,si,2))/Vcrit**Expcri2 )
! eff = (1./v(index))**1.
eff = exp(-Expfac*v(index)**expexp)
if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,2) = term*term*term*eff

term = ( node(index,i,si,3)**(1./3.)+nodec(index,i,si,1)**(1./3.) )
! eff = 1. - ( (node(index,i,si,3)+nodec(index,i,si,1))/Vcrit**Expcri2 )
! eff = (1./v(index))**1.
eff = exp(-Expfac*v(index)**expexp)
if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,3) = term*term*term*eff

else if (agg_kernel_type==10) then

!Empirical kernel

term = ( node(index,i,si,1)**(1./3.)+nodec(index,i,si,3)**(1./3.) )

if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,1) = term*eff

! End if

agg_kernel_birth(index,i,si,2) = term*eff

agg_kernel_birth(index,i,si,3) = term*eff

if (eff<0.) then
  eff = 0.
end if

else if (agg_kernel_type==5) then

agg_kernel_birth(index,i,si,3) = term*term*term/(node(index,i,si,3)+nodec(index,i,si,1))

else if (agg_kernel_type==9) then

!Constant kernel, volume-dependent efficiency
agg_kernel_birth(index,i,si,1) = l./v(index)
agg_kernel_birth(index,i,si,2) = l./v(index)
agg_kernel_birth(index,i,si,3) = l./v(index)

else if (agg_kernel_type==10) then

!Empirical kernel

term = ( node(index,i,si,1)**(1./3.)+nodec(index,i,si,3)**(1./3.) )

if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,1) = term*eff

! End if

agg_kernel_birth(index,i,si,2) = term*eff

agg_kernel_birth(index,i,si,3) = term*eff

if (eff<0.) then
  eff = 0.
end if

else if (agg_kernel_type==10) then

!Empirical kernel

term = ( node(index,i,si,1)**(1./3.)+nodec(index,i,si,3)**(1./3.) )

if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,1) = term*eff

! End if

agg_kernel_birth(index,i,si,2) = term*eff

agg_kernel_birth(index,i,si,3) = term*eff

if (eff<0.) then
  eff = 0.
end if

else if (agg_kernel_type==10) then

!Empirical kernel

term = ( node(index,i,si,1)**(1./3.)+nodec(index,i,si,3)**(1./3.) )

if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,1) = term*eff

! End if

agg_kernel_birth(index,i,si,2) = term*eff

agg_kernel_birth(index,i,si,3) = term*eff

if (eff<0.) then
  eff = 0.
end if

else if (agg_kernel_type==10) then

!Empirical kernel

term = ( node(index,i,si,1)**(1./3.)+nodec(index,i,si,3)**(1./3.) )

if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,1) = term*eff

! End if

agg_kernel_birth(index,i,si,2) = term*eff

agg_kernel_birth(index,i,si,3) = term*eff

if (eff<0.) then
  eff = 0.
end if

else if (agg_kernel_type==10) then

!Empirical kernel

term = ( node(index,i,si,1)**(1./3.)+nodec(index,i,si,3)**(1./3.) )

if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,1) = term*eff

! End if

agg_kernel_birth(index,i,si,2) = term*eff

agg_kernel_birth(index,i,si,3) = term*eff

if (eff<0.) then
  eff = 0.
end if

else if (agg_kernel_type==10) then

!Empirical kernel

term = ( node(index,i,si,1)**(1./3.)+nodec(index,i,si,3)**(1./3.) )

if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,1) = term*eff

! End if

agg_kernel_birth(index,i,si,2) = term*eff

agg_kernel_birth(index,i,si,3) = term*eff

if (eff<0.) then
  eff = 0.
end if

else if (agg_kernel_type==10) then

!Empirical kernel

term = ( node(index,i,si,1)**(1./3.)+nodec(index,i,si,3)**(1./3.) )

if (eff<0.) then
  eff = 0.
end if

agg_kernel_birth(index,i,si,1) = term*eff

! End if

agg_kernel_birth(index,i,si,2) = term*eff

agg_kernel_birth(index,i,si,3) = term*eff

if (eff<0.) then
  eff = 0.
end if
Fortran 90 source code

! eff = 0.
end if
agg_kernel_birth(index,i,si,3) = term*eff
else
    write(*,*) 'invalid data, fatal'
    stop
end if

else if (flag==0) then

! Value of kernel for death term
if (agg_kernel_type==1) then

! Constant kernel
    agg_kernel_death(index,i,1) = 1.
    agg_kernel_death(index,i,2) = 1.
    agg_kernel_death(index,i,3) = 1.
else if (agg_kernel_type==2) then

! Sum kernel
    agg_kernel_death(index,i,1) = v(i-1)+v(index)
    agg_kernel_death(index,i,2) = 0.5*(v(i-1)+v(i)) + v(index)
    agg_kernel_death(index,i,3) = v(i)+v(index)
else if (agg_kernel_type==3) then

! Product kernel
    agg_kernel_death(index,i,1) = v(i-1)*v(index)
    agg_kernel_death(index,i,2) = 0.5*(v(i-1)*v(i)) * v(index)
    agg_kernel_death(index,i,3) = v(i)*v(index)
else if (agg_kernel_type==4) then

! Fluid shear kernel
    term = v(i-1)**(l./3.)+v(index)**(l./3.)
    agg_kernel_death(index,i,1) = term*term*term
    term = (0.5*(v(i-1)+v(i)))**(l./3.)+v(index)**(l./3.)
    agg_kernel_death(index,i,2) = term*term*term
    term = v(i)**(l./3.)+v(index)**(l./3.)
    agg_kernel_death(index,i,3) = term*term*term
else if (agg_kernel_type==5) then

! Differential force kernel
    agg_kernel_death(index,i,1) = &
    & ((v(i-1)**(1./3.))*v(index)**(1./3.))**2. * &
    & abs ((v(i-1)**(2./3.)-v(index)**(2./3.)))
    agg_kernel_death(index,i,2) = &
    & ((0.5*(v(i-1)+v(i)))**(1./3.))*v(index)**(1./3.))**2. * &
    & abs (((0.5*(v(i-1)+v(i)))**(2./3.)-v(index)**(2./3.)))
    agg_kernel_death(index,i,3) = &
    & (v(i)**(1./3.))*v(index)**(1./3.))**2. * &
& abs ((v(i)**(2./3.)-v(index)**(2./3.)))

else if (agg_kernel_type==6) then
    ! Thompson kernel
    agg_kernel_death(index,i,1) = &
    & ((v(i-1)-v(index))**2.)/(v(i-1)+v(index))
    agg_kernel_death(index,i,2) = &
    & ((0.5*(v(i-1)+v(i))-v(index))**2.)/(0.5*(v(i-1)+v(i))+v(index))
    agg_kernel_death(index,i,3) = &
    & ((v(i)-v(index))**2.)/(v(i)+v(index))

else if (agg_kernel_type==7) then
    ! Fluid shear kernel, volume-dependent efficiency
    term = v(i-1)**(l./3.)+v(index)**(l./3.)
    agg_kernel_death(index,i,1) = term*term*term/(v(index)+v(i-1))
    term = (0.5*(v(i-1)+v(i)))**(l./3.)+v(index)**(l./3.)
    agg_kernel_death(index,i,2) = term*term*term/(v(index)+(0.5*(v(i-1)+v(i))))
    term = v(i)**(l./3.)+v(index)**(l./3.)
    agg_kernel_death(index,i,3) = term*term*term/(v(index)+v(i))

else if (agg_kernel_type==8) then
    ! Constant kernel, volume-dependent efficiency
    agg_kernel_death(index,i,1) = 1./(v(index)+v(i-1))
    agg_kernel_death(index,i,2) = 1./(v(index)+(0.5*(v(i-1)+v(i))))
    agg_kernel_death(index,i,3) = 1./(v(index)+v(i))

else if (agg_kernel_type==9) then
    ! Fluid shear kernel, volume-dependent efficiency function
    term = v(i-1)**(l./3.)+v(index)**(l./3.)
    ! eff = 1. - (v(i-1)+v(index))/Vcrit**Expncrit
    ! eff = (1./(v(i-1)+v(index)))**1.
    eff = exp(-Expfac*(v(i-1)+v(index))*expexp)
    if (eff<0.) then
        eff = 0.
    end if
    agg_kernel_death(index,i,1) = term*term*term*eff
    term = (0.5*(v(i-1)+v(i)))**(l./3.)+v(index)**(l./3.)
    ! eff = 1. - (0.5*(v(i-1)+v(i))+v(index))/Vcrit**Expncrit
    ! eff = (1./(0.5*(v(i-1)+v(i))+v(index)))**1.
    eff = exp(-Expfac*(0.5*(v(i-1)+v(i))+v(index))*expexp)
    if (eff<0.) then
        eff = 0.
    end if
    agg_kernel_death(index,i,2) = term*term*term*eff
    term = v(i)**(l./3.)+v(index)**(l./3.)
    ! eff = 1. - (v(i)+v(index))/Vcrit**Expncrit
    ! eff = (1./(v(i)+v(index)))**1.
    eff = exp(-Expfac*(v(i)+v(index))*expexp)
    if (eff<0.) then
        eff = 0.
    end if
    agg_kernel_death(index,i,3) = term*term*term*eff
else if (agg_kernel_type==10) then

    ! Empirical kernel
    term = ( v(i-1)**(l./3.)*v(index)**(l./3.) )
    eff = exp(-lamda*v(i-1)**(l./3.)) * 1./( v(i-1)+v(index) )**(gamma )
    if (eff<0.) then
        eff = 0.
    end if
    agg_kernel_death(index,i,l) = term*eff

    term = ( 0.5*(v(i-1)+v(i)))**(l./3.)*v(index)**(l./3.)
    eff = exp(-lamda*(0.5*(v(i-1)+v(i)))**(l./3.)) * & 1./( ((0.5*(v(i-1)+v(i)))+v(index) )**(gamma )
    if (eff<0.) then
        eff = 0.
    end if
    agg_kernel_death(index,i,2) = term*eff

    term = ( v(i)**(l./3.)*v(index)**(l./3.) )
    eff = exp(-lamda*v(i)**(l./3.)) * 1./( v(i)+v(index) )**(gamma )
    if (eff<0.) then
        eff = 0.
    end if
    agg_kernel_death(index,i,3) = term*eff

else
    write(*,♦) 'invalid data, fatal'
    stop
end if

end subroutine agg_kernel_lib

*****************************************************************************
*****************************************************************************
Subroutine agg_kernel_max(index,flag,maxdim,i1)
*****************************************************************************
*****************************************************************************
! Upper limit of aggregation kernel
! By Stelios Rigopoulos
! 28/05/2001 - initial version
!*****************************************************************************
use PBEjmodule
implicit none
integer, intent(in) :: index,flag,maxdim
integer, intent(in) :: il
integer :: i,si
*****************************************************************************
if (flag==1) then
  do i=1,m
    do si=1,maxdim
      agg_kernel_birth(index,i,si,1) = 0.
      agg_kernel_birth(index,i,si,2) = 0.
      agg_kernel_birth(index,i,si,3) = 0.
    end do
  end do
else
  agg_kernel_death(index,i1,1) = 0.
  agg_kernel_death(index,i1,2) = 0.
  agg_kernel_death(index,i1,3) = 0.
end if
end subroutine agg_kernel_max
agg_kernel_death(index,i1,1) = 0.
agg_kernel_death(index,i1,2) = 0.
agg_kernel_death(index,i1,3) = 0.

end if
end subroutine agg_kernel_min

!******************************************************************************

Subroutine br_kernel_lib(index,i)

!******************************************************************************
!
! Library of breakage kernels
!
! By Stelios Rigopoulos
!
! 18/04/2001 - initial version
!
!******************************************************************************

use PBEjmodule
implicit none
integer, intent(in) :: i,index
real(8) :: terml,v2

v2 = 0.5*(v(i-l)+v(i))

if (breakage_dist_type==1) then
  !Binary breakage, uniform fragmentation
  br_kernel_birth(index,i,1) = 2.*(l./v(i-l))
  br_kernel_birth(index,i,2) = 2.*(l./v2)
  br_kernel_birth(index,i,3) = 2.*(l./v(i))
else if (breakage_dist_type==2) then
  !Binary breakage, parabolic distribution optimised for attrition
  br_kernel_birth(index,i,1) &=
    ( 24 * ( v(index)**2. - v(index)*v(i-l) + (v(i-l)**2.)/4. ) / (v(i-l)**3.) )
  br_kernel_birth(index,i,2) &=
    ( 24 * ( v(index)**2. - v(index)*v2 + (v2**2.)/4. ) / (v2**3.) )
  br_kernel_birth(index,i,3) &=
    ( 24 * ( v(index)**2. - v(index)*v(i) + (v(i)**2.)/4. ) / (v(i)**3.) )
else if (breakage_dist_type==3) then
  !Multiple breakage according to simplified Ziff function
  terml=br_dist_delta*br_kernel_exp/(br_dist_delta-br_kernel_exp)
  br_kernel_birth(index,i,1) &=
    ( v(index)**2. - v(index)*v(i) + (v(i)**2.)/4. ) / (v(i)**3.)
else if (breakage_dist_type==4) then
  !Binary breakage, parabolic distribution optimised for attrition
  br_kernel_birth(index,i,1) &=
    ( 24 * ( v(index)**2. - v(index)*v(i-l) + (v(i-l)**2.)/4. ) / (v(i-l)**3.) )
  br_kernel_birth(index,i,2) &=
    ( 24 * ( v(index)**2. - v(index)*v2 + (v2**2.)/4. ) / (v2**3.) )
  br_kernel_birth(index,i,3) &=
    ( 24 * ( v(index)**2. - v(index)*v(i) + (v(i)**2.)/4. ) / (v(i)**3.) )
else if (breakage_dist_type==3) then
  !Multiple breakage according to simplified Ziff function
  terml=br_dist_delta*br_kernel_exp/(br_dist_delta-br_kernel_exp)
  br_kernel_birth(index,i,1) &=
    ( v(index)**2. - v(index)*v(i) + (v(i)**2.)/4. ) / (v(i)**3.)
br_kernel_birth(index,i,2) = &
  (1/v2*(terml) * (v(index)/v2)**(br_kernel_exp-2.)-(v(index)/v2) &
  **(br_dist_delta-2.))

br_kernel_birth(index,i,3) = &
  (1/v(i)*(terml) * (v(index)/v(i))**br_kernel_exp-2.-(v(index)/v(i)) &
  **(br_dist_delta-2.))

else if (breakage_dist_type==4) then
  !multiple breakage according to generalised Ziff function

  br_kernel_birth(index,i,1) = &
  ((br_dist_alpha*br_dist_gamma/v(i-l))*(v(index)/v(i-l))**(br_dist_gamma-2.) &
  + ((l-br_dist_alpha)*br_dist_delta/v(i-l))*(v(index)/v(i-l))**(br_dist_delta-2.))

  br_kernel_birth(index,i,2) = &
  ((br_dist_alpha*br_dist_gamma/v2)*(v(index)/v2)**(br_dist_gamma-2.) &
  + ((l-br_dist_alpha)*br_dist_delta/v2)*(v(index)/v2)**(br_dist_delta-2.))

  br_kernel_birth(index,i,3) = &
  ((br_dist_alpha*br_dist_gamma/v(i))*(v(index)/v(i))**(br_dist_gamma-2.) &
  + ((l-br_dist_alpha)*br_dist_delta/v(i))*(v(index)/v(i))**(br_dist_delta-2.))

else if (breakage_dist_type==5) then
  !binary breakage, generalised parabolic distribution

  br_kernel_birth(index,i,1) = &
  (24*(1-br_dist_parabol/2.)* (v(index)**2.-v(index)*v(i-l)+(v(i-l)**2.))/4. &
  + v(i-l)**3. + br_dist_parabol/v(i-l))

  br_kernel_birth(index,i,2) = &
  (24*(1-br_dist_parabol/2.)* (v(index)**2.-v(index)*v2+(v2**2.))/4. &
  + v2**3. + br_dist_parabol/v2)

  br_kernel_birth(index,i,3) = &
  (24*(1-br_dist_parabol/2.)* (v(index)**2.-v(index)*v(i)+(v(i)**2.))/4. &
  + v(i)**3. + br_dist_parabol/v(i))

else if (breakage_dist_type==6) then
  !erosion

  if ((i==index-l).or.(i==D)) then
    br_kernel_birth(index,i,1) = 1.
    br_kernel_birth(index,i,2) = 1.
    br_kernel_birth(index,i,3) = 1.
  else
    br_kernel_birth(index,i,1) = 0.
    br_kernel_birth(index,i,2) = 0.
    br_kernel_birth(index,i,3) = 0.
  end if
else
  write(*,*) 'invalid data, fatal'
  stop
end if

else if (breakage_kernel_type==1) then
Fortran 90 source code

!Power law kernel

br_kernel_birth(index,i,1) = br_kernel_birth(index,i,1) * (v(i-1)**(br_kernel_exp))
br_kernel_birth(index,i,2) = br_kernel_birth(index,i,2) * (v(i)**(br_kernel_exp))
br_kernel_birth(index,i,3) = br_kernel_birth(index,i,3) * (v(i)**(br_kernel_exp))

else if (breakage_kernel_type==2) then

!Exponential kernel

br_kernel_birth(index,i,1) = br_kernel_birth(index,i,1) * exp(br_kernel_exp*v(i-1))
br_kernel_birth(index,i,2) = br_kernel_birth(index,i,2) * exp(br_kernel_exp*v2)
br_kernel_birth(index,i,3) = br_kernel_birth(index,i,3) * exp(br_kernel_exp*v(i))

else

write(*,*) 'invalid data, fatal'
stop
end if

end subroutine br_kernel_lib

**************************************************************

**************************************************************

subroutine PBE_initdata()

! Initialises kinetic data
! By Stelios Rigopoulos
! 09/02/2002
!
use PBE_module
implicit none
integer i,k,l

!Read all kinetic etc. data

do i=1,8
    read(31,*)
end do

read(31,5002) Agg_kemel_type
read(31,5001) aggl_kernel_const
read(31,5002) breakage_kemel_type
read(31,5002) breakage_dist_type
read(31,5001) br_kernel_const
read(31,5001) br_kernel_exp
read(31,5002) br_dist_parabol
read(31,5001) br_dist_gamma
read(31,5001) br_dist_delta
read(31,5001) br_dist_alpha
read(31,5002) growth_kernel_type
!read(31,5001) G_coeff1
!read(31,5001) G_coeff2
!read(31,5001) G_coeff3
!read(31,5001) Nuc

doi=1,3
    read(31,*)
end do

read(31,5001) Nuc_ss_coeff
read(31,5001) Nuc_ss_exp
read(31,5001) G_ss_coeff
read(31,5001) G_ss_exp

doi=1,3
    read(31,*)
end do

read(31,5002) meshflag
read(31,5002) m
read(31,5001) inc
read(31,5001) v_nuc
read(31,5001) v_0

doi=1,3
    read(31,*)
end do

read(31,5001) threshold
!read(31,5002) speedupPBE

!----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------

5001 format(29X,F20.10)
5002 format(29X,I10)
5003 format(I6,F20.10)

end subroutine PBE_initdata

!**********************************************************************************************************************************************************************************************************

!**********************************************************************************************************************************************************************************************************

subroutine PBE_initarrays(ni,nin)

!**********************************************************************************************************************************************************************************************************

!**********************************************************************************************************************************************************************************************************

!**********************************************************************************************************************************************************************************************************

!**********************************************************************************************************************************************************************************************************

!**********************************************************************************************************************************************************************************************************

use PBEjmodule

implicit none

real(8), dimension(0:m), intent(out) :: ni,nin

integer i,k,l

!----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------
! Initialise population density arrays
do i=1,m
    ni(i) = 0.0
    nin(i) = 0.0
end do
ni(0) = 0.0
nin(0) = 0.0

! Initial distribution
do i=1,6
    read(34,*)
end do
read(34,5002) initflag
ni(0) = 0. ! Initial condition
ni(m) = 0.
if (initflag==0) then
    ! Zero initial distribution
    do i=1,m-1
        ni(i) = 0. ! Underflow check
    end do
else if (initflag==1) then
    ! Exponential initial distribution
    do i=1,5
        read(34,*)
    end do
    read(34,5001) NO
    read(34,5001) VO
    do i=1,m-1
        if ((i>1).and.(ni(i-1)<threshold)) then
            ni(i) = 0. ! Underflow check
        else
            ni(i) = NO/VO * exp(-(v(i))/VO)
        end if
    end do
else if (initflag==2) then
    ! Read from disk
do i=1,12
   read(34,*)
end do
read (34,5002) k
do i=1,k
   read(34,5003) l,ni(l)
end do
end if

!-----------------------------------------------------------------------------------------------------------------------
!

! Input distribution

! do i=1,6
!   read(35,*)
!end do
read(35,5002) inputflag
nin(0) = 0. ! Initial condition
nin(m) = 0.
if (inputflag==0) then
  ! Zero initial distribution
  do i=1,m-1
    nin(i) = 0. ! Underflow check
  end do
else if (inputflag==1) then
  ! Exponential initial distribution
  do i=1,5
    read(35,*)
  end do
read(35,5001) NO
read(35,5001) VO
  do i=1,m-1
    if ((i>1).and.(nin(i-1)<threshold)) then
      nin(i) = 0. ! Underflow check
    else
      nin(i) = NO/VO * exp(-(v(i))/VO)
    end if
  end do
else if (inputflag==2) then
   !Read from disk
   do i=1,12
      read(35,*)
   end do
   read (35,5002) k
   do i=1,k
      read(35,5003) l,nin(l)
   end do
end if

!----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------
!
! Write initial and input distribution
!
! do i=1,m
!    write (111,*) v(i),ni(i)
! end do
! do i=1,m
!    write (113,*) v(i),nin(i)
! end do
!
!------------------------------------------------------------------------------------------------------------------------
5001 format(29X,F20.10)
5002 format(29X,I10)
5003 format(I6,F20.10)
end subroutine FBE_initarrays

C.2  Network

*****************************************************************************
!* FILE Network_module.f90
!* ! By Stelios Rigopoulos
!* ! Contains:
!* - module network_module
!* - subroutine network_initdata
!* - subroutine gas_initarrays
!* - subroutine network
!* ! 05/02/2002: initial version
!* 18/02/2002: new version - all in one subroutine
!* 24/02/2002: first complete version
!* 26/02/2002: corrected for output in gas flow
!* 21/06/2002: new version for reaction network
! ATTENTION
!
! The following must be manually inserted:
! Dimensions in classes (network module)
! Dimension in Cin in functions F and bulk (reaction module)
!
*******************************************************************************

module network_module
*******************************************************************************

!*******************************************************************************
!
* By Stelios Rigopoulos
* 18/02/2002: initial version
*
*******************************************************************************

implicit none

*******************************************************************************

real(8) :: total_time
real(8) :: Pressure
real(8) :: Temperature
real(8) :: R_constant
real(8) :: d_bubble
real(8) :: contact_time
real(8) :: interface_depth
real(8) :: rellerr_conc,rellerr_pdensity
integer :: bulk_lin_itmax
integer :: interval_writeup

edm module network_module

*******************************************************************************

*******************************************************************************

subroutine network_initdata
*******************************************************************************

* By Stelios Rigopoulos
* 18/02/2002: initial version
*
*******************************************************************************

use network_module

implicit none

integer k

*******************************************************************************

doo k=1,6

    read(81,*)
end do
Fortran 90 source code

read(81,1003) total_time
read(81,1003) Pressure
read(81,1003) Temperature
read(81,1003) R_constant
read(81,1003) d_bubble
read(81,1003) contact_time
read(81,1003) interface_depth
read(81,1003) relerr_conc
read(81,1003) relerr_density
read(81,1001) bulk_lin_itmax
read(81,1001) interval_writeup
read(81,*)

1003 format(29X,F20.10)
1001 format(29X, I10)

end subroutine network_initdata

!*******************************************************************************************************

subroutine gas_initarrays(y_mol,n)
!*******************************************************************************************************

! Initialises arrays
! By Stelios Rigopoulos
! 09/02/2002
!
!*******************************************************************************************************

implicit none

integer, intent(in) :: n
real(8), dimension(n,2), intent(inout) :: y_mol

integer i

! Initial concentrations

rewind(83)
doi=l,8
  read(83,*)
end do
do i=1,n
  read(83,5001) y_mol(i,1),y_mol(i,2)
end do

5001 format(29X,F10.8,13X,F5.3)

end subroutine gas_initarrays

!*******************************************************************************************************

!*******************************************************************************************************
subroutine network()

!**********************************************************************
!
! By Stelios Rigopoulos
! 05/02/2002: initial version
! 18/02/2002: new version - all in one subroutine
!
! Generation of the reaction engineering network
! Controls the program flow
! Temporal loop according to the penetration theory
!
!**********************************************************************

use PBE_module
use reaction_module
use network_module

implicit none

!---------------------------------------------------------

!General declarations

!integer, allocatable, dimension(:,,:) :: comp
!real(8), allocatable, dimension() :: v,dv
!integer, allocatable, dimension() :: vhalf
real(8), allocatable, dimension() :: pdensity_in
real(8), allocatable, dimension() :: concentration_in
real(8), allocatable, dimension() :: mol
!real(8), allocatable, dimension() :: Dif
real(8) :: total_M0,total_M1
real(8) :: totalmol
integer :: read_flow_flag
integer :: cells_X
integer :: cells_Y
integer :: i,j,k,l,k2
!integer :: m,n
integer :: index_ct,N_of_ct,bulk_lin_iter
integer :: iter_coup,maxiter_coup
integer :: converge
integer :: count
real(8) :: time
real(8) :: integration_time
real(8) :: weight_in,weight_out
real(8) :: dM0
real(8) :: void
real(8) :: ni_temp,C_temp,temp,temp_2
real(8) :: vol_growthrate
real(8) :: tot_moles,tot_conc,tot_volume

!---------------------------------------------------------

!Specification of the generic class that describes the building block
!of the reaction engineering network

!Subclasses
type volume_class
  real(8) total
  real(8) bulk
  real(8) int
  real(8) gas
end type volume_class

type left_class
  integer X
  integer Y
  real flow
  real flow_2
end type left_class

type right_class
  integer X
  integer Y
  real flow
  real flow_2
end type right_class

type up_class
  integer X
  integer Y
  real flow
  real flow_2
end type up_class

type down_class
  integer X
  integer Y
  real flow
  real flow_2
end type down_class

!Main class

type cell_class

!Location of cell
integer :: X
integer :: Y

!Directions of mass exchange
type (left_class) :: left
type (right_class) :: right
type (up_class) :: up
type (down_class) :: down

!Input and cell concentrations in both phases
real(8), dimension(6) :: Cin,C
real(8), dimension(6) :: C_ave,C_temp
real(8), dimension(6) :: Cin_2,C_2

!Input and cell population densities
real(8), dimension(0:100) :: ni,nin
real(8), dimension(0:100) :: ni_ave,ni_temp
! Gas molar fractions
real(8), dimension(6,2) :: y_mol

! Amount of absorption of each gaseous component
real(8), dimension(6) :: absorption

! Total moles in the interface
real(8), dimension(6) :: mol_interface

! Moments
real(8), dimension(0:1) :: Moment_bulk

! Volume, total and of each phase
type (volume_class) :: volume

! Flow in and out of the system
real(8) :: flow_in
real(8) :: flow_out
real(8) :: flow_2_in
real(8) :: flow_2_out

! Interfacial area
real(8) :: A_interface

! Mean diameter of particles
real(8) :: Dmean

! Temporary storage of population at nuclei size
real(8) :: nil_prev

end type cell_Class

! Define the compartments as an array of objects of the cell_class
type(cell_class), allocatable, dimension(:,,:) :: cell

!-------------------------------------------------------------------------------------------------------------------------------------------------------
! Main program
!
!-------------------------------------------------------------------------------------------------------------------------------------------------------
!
! Initialisation

m=size(cell(1,1)%ni)-1
n=size(cell(1,1)%C)

! Allocate global arrays
allocate (v(0:m))
allocate (dv(m))
allocate (whalf(m))
allocate (compt(m,m,2))
allocate(pdensity_in(0:m))
allocate(concentration_in(n))
allocate(mol(n))
call file_open(1)

! Initialise population balance parameters
call PBE_initdata()

! Create mesh - uniform for all cells
!call mesh()
!Read environmental etc. parameters
call network_initdata()
!Initiate parameters related to interface
call interface_initdata()
!Initialise reaction engineering parameters
call reaction_initdata()
!Initialise bulk integration
call integrator_initdata()

!---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------

!Create network
!Read network parameters
read(81,1001) cells_X
read(81,1001) cells_Y
allocate(cell(cells_Y,cells_X))
!Generate topology
do i=1,cells_Y !rows iteration
  do j=1,cells_X !columns iteration
    !Location of cell
    cell(i,j)%X = j
    cell(i,j)%Y = i
    !Location and connectivity of left cell
    if (j==1) then
      cell(i,j)%left%X = 0
      cell(i,j)%left%Y = 0
    else
      cell(i,j)%left%X = j-1
      cell(i,j)%left%Y = i
    end if
    !Location and connectivity of right cell
    if (j==cells_X) then
      cell(i,j)%right%X = 0
      cell(i,j)%right%Y = 0
    else
      cell(i,j)%right%X = j+1
      cell(i,j)%right%Y = i
    end if
  end do
end do
! Location and connectivity of up cell
if (i==l) then
  cell(i,j)%up%X = 0
  cell(i,j)%up%Y = 0
else
  cell(i,j)%up%X = j
  cell(i,j)%up%Y = i-1
end if

! Location and connectivity of down cell
if (i==cells_Y) then
  cell(i,j)%down%X = 0
  cell(i,j)%down%Y = 0
else
  cell(i,j)%down%X = j
  cell(i,j)%down%Y = i+1
end if
end do
end do

1003 format(29X,F20.10)
1001 format(29X,110)

! Read/generate flows
read(81,*)
read(81,1113) read_flow_flag
if (read_flow_flag==1) then
  do i=1,8
    read(81,*)
  end do
  i=0
  j=0
  do k=1,(cells_Y*cells_X)
    read(81,1110,advance = 'no') i,j
    read(81,1111) &
      & cell(i,j)%left%flow, cell(i,j)%right%flow, &
      & cell(i,j)%up%flow, cell(i,j)%down%flow, &
      & cell(i,j)%left%flow_2, cell(i,j)%right%flow_2, &
      & cell(i,j)%up%flow_2, cell(i,j)%down%flow_2, &
      & cell(i,j)%flow_in, cell(i,j)%flow_out, &
      & cell(i,j)%volume%bulk, cell(i,j)%volume%gas
  end do
end if

1113 format(29X,I10)  
1110 format (I1,14.14,1X)  

! Mass balance check

do i=1,cells_Y  
  do j=1,cells_X  
    temp = - (cell(i,j)%left%flow - cell(i,j)%right%flow) &  
      & - (cell(i,j)%up%flow - cell(i,j)%down%flow) &  
      & - cell(i,j)%flow_out

    temp_2 = - (cell(i,j)%left%flow_2 - cell(i,j)%right%flow_2) &  
      & - (cell(i,j)%up%flow_2 - cell(i,j)%down%flow_2) &  
      & - cell(i,j)%flow_2_out

    temp = temp + cell(i,j)%flow_in  
    temp_2 = temp_2 + cell(i,j)%flow_2_in

    if ((cell(i,j)%left%Y>0) .and. (cell(i,j)%left%X>0)) then
      temp = temp + cell(cell(i,j)%left%Y,cell(i,j)%left%X)%flow
      temp_2 = temp_2 + cell(cell(i,j)%left%Y,cell(i,j)%left%X)%flow_2
    end if

    if ((cell(i,j)%right%Y>0) .and. (cell(i,j)%right%X>0)) then
      temp = temp + cell(cell(i,j)%right%Y,cell(i,j)%right%X)%flow
      temp_2 = temp_2 + cell(cell(i,j)%right%Y,cell(i,j)%right%X)%flow_2
    end if

    if (cell(i,j)%up%Y>0) .and. (cell(i,j)%up%X>0)) then
      temp = temp + cell(cell(i,j)%up%Y,cell(i,j)%up%X)%flow
      temp_2 = temp_2 + cell(cell(i,j)%up%Y,cell(i,j)%up%X)%flow_2
    end if

    if ((cell(i,j)%down%Y>0) .and. (cell(i,j)%down%X>0)) then
      temp = temp + cell(cell(i,j)%down%Y,cell(i,j)%down%X)%flow
      temp_2 = temp_2 + cell(cell(i,j)%down%Y,cell(i,j)%down%X)%flow_2
    end if

    if (abs(temp)>1.0D-8) then
      write(*,*) 'incorrect flows in the bulk phase, mass balance not satisfied'
      write(*,*) 'cell: ',i,j,' residual: ',abs(temp)
      write(*,*) 'press any key to continue'
      pause
    end if

    if (abs(temp_2)>1.0D-8) then
      write(*,*) 'incorrect flows in the dispersed phase, mass balance not satisfied'
      write(*,*) 'cell: ',i,j,' residual: ',abs(temp)
      write(*,*) 'press any key to continue'
      pause
  end do
end do
end if
end do
end do

! Initialise population densities, concentrations and other variables
do i=1,cells_Y
  do j=1,cells_X
    call reaction_initarrays(cell(i,j)%XC,cell(i,j)%XCin)
    call gas_initarrays(cell(i,j)%Xy_mol,n)
    do k=1,n
      cell(i,j)%XC_2(k) = cell(i,j)%Xy_mol(k,1)*Pressure/(R_constant*Temperature)
      cell(i,j)%XCin_2(k) = cell(i,j)%XC_2(k)
    end do
    do k=0,m
      cell(i,j)%ni(k) = 0.
      cell(i,j)%nin(k) = 0.
      cell(i,j)%ni_temp(k) = 0.
      cell(i,j)%ni_ave(k) = cell(i,j)%ni(k)
    end do
    do k=1,n
      cell(i,j)%c_temp(k) = 0.
      cell(i,j)%c_ave(k) = cell(i,j)%c(k)
    end do
    do k=1,n
      cell(i,j)%mol_interface(k) = 0.
    end do
    do k=1,n
      cell(i,j)%absorption(k) = 0.
    end do
    do k=1,1
      cell(i,j)%Moment_bulk = 0.
    end do
    cell(i,j)%ni1_prev = 0.
    cell(i,j)%Dmean = 0.
    cell(i,j)%XA_interface = 6. * cell(i,j)%volume%gas/d_bubble
    cell(i,j)%volume%int = cell(i,j)%XA_interface * interface_depth
    ! Correct volume of bulk by subtracting the interface
    cell(i,j)%volume%total = cell(i,j)%volume%bulk + cell(i,j)%volume%gas
Fortran 90 source code

```fortran

    cell(i,j)%volume%bulk = cell(i,j)%volume%bulk - cell(i,j)%volume%int

end do
end do

vol_growthrate = 0.

! Impulse
! cell(2,1)%C(1) = 1.

!----------------------------------------------------------------------

! Time loop

N_of_ct = int(total_time/contact_time)
time = 0.
count = 0

! Main loop (over contact times)

! Initial write-up
writeC*,*) ' Beginning of calculation'
write(*,*) '  '

do index_ct=i,N_of_ct

! Time step write-up
write(*,*) '  '
writeC*,*) 'time is ', time
write(*,*) '  '

! Step feed

! do i=1,cells_Y
!    cell(i,i)%C(1) = cell(i,i)%C(1) + 0.01*contact_time
!
! end do

    cell(1,1)%C(1) = cell(1,1)%C(1) + 0.01*contact_time * 1.2
    cell(2,1)%C(1) = cell(2,1)%C(1) + 0.01*contact_time * 1.1
    cell(3,1)%C(1) = cell(3,1)%C(1) + 0.01*contact_time * 1.1
    cell(4,1)%C(1) = cell(4,1)%C(1) + 0.01*contact_time * 1.
    cell(5,1)%C(1) = cell(5,1)%C(1) + 0.01*contact_time * 1.
    cell(6,1)%C(1) = cell(6,1)%C(1) + 0.01*contact_time * 1.
    cell(7,1)%C(1) = cell(7,1)%C(1) + 0.01*contact_time * 0.9
    cell(8,1)%C(1) = cell(8,1)%C(1) + 0.01*contact_time * 0.9
    cell(9,1)%C(1) = cell(9,1)%C(1) + 0.01*contact_time * 0.8
    cell(10,1)%C(1) = cell(10,1)%C(1) + 0.01*contact_time * 0.7
    cell(10,2)%C(1) = cell(10,2)%C(1) + 0.01*contact_time * 0.2

!----------------------------------------------------------------------

! Step 1: Detachment - calculation of interface

! Interface write-up
writeC*,*) 'Calculating interfaces...' 
write(*,*) '  '

do i=1,cells_Y
    do j=1,cells_X
        ...
    end do
end do
```

if (cell(i,j)%volume%int > 0.) then
    Call Interface_element(cell(i,j)%C,cell(i,j)%A_interface,&
                        & cell(i,j)%y_mol,cell(i,j)%mol_interface)
end if
end do
end do

! Interface write-up -------
write(*,*) 'Interfaces calculated'
write(*,*) ', '

!---------------------------------------------------------------

! Step 2: Calculation of bulk by iteration over cells

!Bulk write-up --------
write(*,*) 'Calculating bulk...'
!---------------------------------------------------------------

bulk_lin_iter = 0

do ! Linearisation

!Temporary storage of bulk concentrations and population densities
! to use until convergence of C_ave is achieved

do i=1,cells_Y
    do j=1,cells_X
        do l=0,m
            cell(i,j)%ni_temp(l) = cell(i,j)%ni(l)
        end do
        do k=1,n
            cell(i,j)%C_temp(k) = cell(i,j)%C(k)
        end do
    end do
    temp = 0.
do i=1,cells_Y
    do j=1,cells_X

! Calculate inputs and outputs for cell (X,Y)
!---------------------------------------------

    weight_in = (1./cell(i,j)%volume%bulk))

! Input population density --------
! Includes the flows from the respective compartments
do k=0,m
   pdensity_in(k) = 0.
   !Contribution from left
   if ((cell(i,j)%left%Y>0).and.(cell(i,j)%left%X>0)) then
      pdensity_in(k) = pdensity_in(k) + cell(cell(i,j)%left%Y,cell(i,j)%left%X)%ni_ave(k) * &
                     cell(cell(i,j)%left%Y,cell(i,j)%left%X)%right%flow
   end if
   !Contribution from right
   if ((cell(i,j)%right%Y>0).and.(cell(i,j)%right%X>0)) then
      pdensity_in(k) = pdensity_in(k) + cell(cell(i,j)%right%Y,cell(i,j)%right%X)%ni_ave(k) * &
                     cell(cell(i,j)%right%Y,cell(i,j)%right%X)%left%flow
   end if
   !Contribution from up
   if ((cell(i,j)%up%Y>0).and.(cell(i,j)%up%X>0)) then
      pdensity_in(k) = pdensity_in(k) + cell(cell(i,j)%up%Y,cell(i,j)%up%X)%ni_ave(k) * &
                     cell(cell(i,j)%up%Y,cell(i,j)%up%X)%down%flow
   end if
   !Contribution from down
   if ((cell(i,j)%down%Y>0).and.(cell(i,j)%down%X>0)) then
      pdensity_in(k) = pdensity_in(k) + cell(cell(i,j)%down%Y,cell(i,j)%down%X)%ni_ave(k) * &
                     cell(cell(i,j)%down%Y,cell(i,j)%down%X)%up%flow
   end if
   !Contribution from input to the system
   pdensity_in(k) = pdensity_in(k) + cell(i,j)%nin(k) * cell(i,j)%flow_in
end do

do k=0,n
   concentration_in(k) = 0.
   !Contribution from left
   if ((cell(i,j)%left%Y>0).and.(cell(i,j)%left%X>0)) then
      concentration_in(k) = concentration_in(k) + cell(cell(i,j)%left%Y,cell(i,j)%left%X)%C_ave(k) * &
                           cell(cell(i,j)%left%Y,cell(i,j)%left%X)%right%flow
   end if
   !Contribution from right
   if ((cell(i,j)%right%Y>0).and.(cell(i,j)%right%X>0)) then
      concentration_in(k) = concentration_in(k) + cell(cell(i,j)%right%Y,cell(i,j)%right%X)%C_ave(k) * &
                           cell(cell(i,j)%right%Y,cell(i,j)%right%X)%left%flow
   end if
   !Contribution from up
   if ((cell(i,j)%up%Y>0).and.(cell(i,j)%up%X>0)) then
      concentration_in(k) = concentration_in(k) + cell(cell(i,j)%up%Y,cell(i,j)%up%X)%C_ave(k) * &
                           cell(cell(i,j)%up%Y,cell(i,j)%up%X)%down%flow
   end if
   !Contribution from down
   if ((cell(i,j)%down%Y>0).and.(cell(i,j)%down%X>0)) then
      concentration_in(k) = concentration_in(k) + cell(cell(i,j)%down%Y,cell(i,j)%down%X)%C_ave(k) * &
                           cell(cell(i,j)%down%Y,cell(i,j)%down%X)%up%flow
   end if
   !Input concentration -----------
end do
Fortran 90 source code

& cell(cell(i,j)%right%Y,cell(i,j)%right%X)%left%flow
end if

'Contribution from up

if ((cell(i,j)%up%Y>0).and.(cell(i,j)%up%X>0)) then
    concentration_in(k) = concentration_in(k) + cell(cell(i,j)&
    &%up%Y,cell(i,j)%up%X)%C_ave(k) + &
    & cell(cell(i,j)%up%Y,cell(i,j)%up%X)%down%flow
end if

'Contribution from down

if ((cell(i,j)%down%Y>0).and.(cell(i,j)%down%X>0)) then
    concentration_in(k) = concentration_in(k) + cell(cell(i,j)&
    &%down%Y,cell(i,j)%down%X)%C_ave(k) + &
    & cell(cell(i,j)%down%Y,cell(i,j)%down%X)%up%flow
end if

'Contribution from input to the system

concentration_in(k) = concentration_in(k) + cell(i,j)%Cin(k) * cell(i,j)%flow_in
end do

!The minus of output is in subroutine RHSide

weight_out = (l./(cell(i,j)%volume%bulk))*&
    & cell(i,j)%left%flow + &
    & cell(i,j)%right%flow + &
    & cell(i,j)%up%flow + &
    & cell(i,j)%down%flow

!Calculate new bulk concentrations

integration_time = contact_time
iter_coup = 0

Call Bulk(cell(i,j)%ni_temp,pdensity_in,cell(i,j)%C_temp,concentration_in,&
    & weight_in,weight_out,integration_time,&
    & cell(i,j)%moment_bulk,vol_growthrate)

if (iter_coup>maxiter_coup) then
    maxiter_coup = iter_coup
end if

!Calculate new bulk concentrations

end do
end do

!Update the average concentrations and densities and check for convergence

converge = 1
do i=1,cells_Y
  do j=1,cells_X
    do k=1,n
      temp = cell(i,j)%X_ave(k)
      cell(i,j)%X_ave(k) = 5.D-1 * (cell(i,j)%X(k)+cell(i,j)%X_temp(k))
      if (((cell(i,j)%X_ave(k)>0.) .and. & & (abs(temp-cell(i,j)%X_ave(k))/cell(i,j)%X_ave(k))>relerr_conc)) then
        converge = 0
      end if
    end do
    do k=1,m
      temp = cell(i,j)%ni_ave(k)
      cell(i,j)%ni_ave(k) = 5.D-1 * (cell(i,j)%ni(k)+cell(i,j)%ni_temp(k))
      if (((cell(i,j)%ni_ave(k)>0.) .and. & & (abs(temp-cell(i,j)%ni_ave(k))/cell(i,j)%ni_ave(k))>relerr_density)) then
        converge = 0
      end if
    end do
  end do
end do
! If linearisation did not converge, continue the linearisation loop
if (converge == 0) then
  bulk_lin_iter = bulk_lin_iter + 1
  if (bulk_lin_iter>bulk_lin_itmax) then
    write(*,*) 'no convergence of network linearisation after iterations ', bulk_lin_iter
    stop
  else
    go to 500
  end if
end if
! If converged, accept new values and exit loop
do i=1,cells_Y
  do j=1,cells_X
    do k=1,n
      cell(i,j)%X(k) = cell(i,j)%X_temp(k)
      cell(i,j)%X_ave(k) = cell(i,j)%X_ave(k)
    end do
  end do
do k=l,m
    cell(i,j)xni(k) = cell(i,j)xni_temp(k)
    cell(i,j)xni_ave(k) = cell(i,j)xni_temp(k)
end do
end do

end do
temp = 0.
exit !from linearisation loop
500 continue !the linearisation - if not converged
end do

!Bulk write-up ------------
write(*,*) 'Bulk calculated'
write(*,*) 'Iterations needed for cells coupling: ', bulk_lin_iter
write(*,*) 'Iterations (max) needed for substrate-PSD coupling: ', maxiter_coup
!
!----------------------------------------------------------
!
End of bulk calculation
!

!Step 3: Mixing of interface and bulk
do i=1,cells_Y
    do j=1,cells_X
        if (cell(i,j)volume%int > 0.) then
            do k=1,n
                cell(i,j)C(k) = ( cell(i,j)xmol_interface(k) + cell(i,j)C(k)*cell(i,j)volume%bulk ) &
                / (cell(i,j)volume%bulk + cell(i,j)volume%int)
            end do
        end if
    end do
end do

do i=1,cells_Y
    do j=1,cells_X
        total_MO = total_MO + cell(i,j)moment_bulk(0)
        total_Ml = total_Ml + cell(i,j)moment_bulk(1)*cell(i,j)volume%bulk
    end do
end do

!Results write-up --------
write(*,*) 'Total particles: ', total_MO
write(*,*) 'Total volume of particulate: ', total_Ml
write(*,*) 'Mean concentrations: ', (Mean_C(k),k=1,n)
!
!----------------------------------------------------------
Update flows to assist convergence of next bulk linearisation

do i=1,cells_Y
  do j=1,cells_X
    do k=1,n
      cell(i,j)%c_ave(k) = cell(i,j)%c(k)
    end do
    do k=1,m
      cell(i,j)%mi_ave(k) = cell(i,j)%mi(k)
    end do
  end do
end do

! Step 4: Gas

do i=1,cells_Y
  do j=1,cells_X
    if (cell(i,j)%volume%gas > 0.) then
      ! Molar mass balance
      totalmol = 0.
      do k2=1,n
        mol(k2) = cell(i,j)%c_2(k2) * cell(i,j)%volume%gas
      end do
      ! Flow in from neighboring cells
      if (((cell(i,j)%left%Y>0).and.(cell(i,j)%left%X>0)) then
        mol(k2) = mol(k2) + cell(cell(i,j)%left%Y,cell(i,j)%left%X)%c_2(k2) * &
        & cell(cell(i,j)%left%Y,cell(i,j)%left%X)%right%flow_2 * contact_time
      end if
      if (((cell(i,j)%right%Y>0).and.(cell(i,j)%right%X>0)) then
        mol(k2) = mol(k2) + cell(cell(i,j)%right%Y,cell(i,j)%right%X)%c_2(k2) * &
        & cell(cell(i,j)%right%Y,cell(i,j)%right%X)%left%flow_2 * contact_time
      end if
      if (((cell(i,j)%down%Y>0).and.(cell(i,j)%down%X>0)) then
        mol(k2) = mol(k2) + cell(cell(i,j)%down%Y,cell(i,j)%down%X)%c_2(k2) * &
        & cell(cell(i,j)%down%Y,cell(i,j)%down%X)%up%flow_2 * contact_time
      end if
      if (((cell(i,j)%up%Y>0).and.(cell(i,j)%up%X>0)) then
        mol(k2) = mol(k2) + cell(cell(i,j)%up%Y,cell(i,j)%up%X)%c_2(k2) * &
        & cell(cell(i,j)%up%Y,cell(i,j)%up%X)%down%flow_2 * contact_time
end if

! Flow into the system
mol(k2) = mol(k2) + cell(i,j)%Cin_2(k2) * cell(i,j)%flow_2_in * contact_time

! Absorption
mol(k2) = mol(k2) - cell(i,j)%mol_interface(k2)

! Flow out of the system
mol(k2) = mol(k2) - cell(i,j)%C_2(k2) * cell(i,j)%flow_2_out * contact_time

! Flow out to neighbouring cells
mol(k2) = mol(k2) - cell(i,j)%C_2(k2) * &
& (cell(i,j)%left%flow_2 + cell(i,j)%right%flow_2 + &
& cell(i,j)%up%flow_2 + cell(i,j)%down%flow_2) &
& * contact_time

if (mol(k2)<0.) then
    mol(k2) = 0.
end if

totalmol = totalmol + mol(k2)
end do

! Calculate new volume
void = cell(i,j)%volume%gas
cell(i,j)%volume%gas = totalmol*R_constant*temperature/pressure

! Calculate new interfacial area
cell(i,j)%A_interface = 6.* cell(i,j)%volume%gas/d_bubble

! Calculate new Concentration
do k2=l,n
    cell(i,j)%C_2(k2) = mol(k2)/cell(i,j)%volume%gas
end do

! Gas flow correction

end if
end do
end do

! Final write-up
write(*,*) ' Total particles: ',  total_MO
write(*,*) ' Total volume of particulate: ',  total_Ml
write(*,*) ' Mean concentrations: ',  (Mean_C(k),k=1,n)
write(*,*)

   time = index_ct*contact_time
!
   write(130,1103) (v(k),k=1,m)

   count = count+1
   if (count>=interval_writeup) then
      count=0
      write(310,*)
      write(310,1100) time
      tot_volume = 0.
      tot_moles = 0.
      do i=l,cells_Y
         do j=l,cells_X
            tot_volume = tot_volume + cell(i,j)%volume%bulk
            tot_moles = tot_moles + cell(i,j)%xc(1)*cell(i,j)%volume%bulk
         end do
      end do
      tot_conc = tot_moles/tot_volume
      do i=cells_Y,1,-l
         write(310,1101) ((cell(i,j)%xc(k)/tot_conc,k=1,n),j=1,cells_X)
      end do
      write(311,1101) time, (cell(2,i)%xc(1),i=1,cells_X)
!
      write(311,1101) time, (cell(i+1,1)%xc(1)-cell(i,1)%xc(1),i=1,cells_Y-1)
!  do i=1,cells_Y
!  do j=1,cells_X
!   write(130,1105) i,j
!   write(130,1103) (cell(i,j)%ni(k),k=1,m)
! end do
! end do
end if

1100 format(F10.5)
1101 format(<n*cells_X>(F15.9,3X))
1103 format(<m*cells_X>(E15.7,3X))
1104 format(<n*cells_Y>(F10.4,E15.7,3X))
1105 format(F15,15)

!----------------------------------
!
end do

end subroutine network

!******************************************************************************
C.3 Interface

*************************************************************************
! FILE Interface_module.f90
!
! Solution of interfacial phenomena
! This file contains:
! - Module PBE_module: all general declarations
! - Subroutine Interface_initdata: initialise the interface
! - Subroutine Interface_element: solution of interfacial phenomena
! - Subroutine Derive: calculate partial derivatives by perturbation
! - Subroutine Integrate_int: used to calculate total mol and moments in the interface
! - Subroutine Thomas: solution of tridiagonal systems using the Thomas algorithm distributions
!
! By Stelios Rigopoulos
! 18/02/2002: initial version
! 12/04/2002: without zeroth moment implemented in bubble column
!
*************************************************************************

module interface_module
implicit none
real(B) :: contact_time, interface_depth, k_1, gas_holdup
real(8) :: Pressure, Temperature, Henry
real(8) :: alpha, pdeerr
integer :: Nx, Nt_int, Itmax_int, interval_int

end module interface_module

*************************************************************************

subroutine Interface_initdata(Dif)
*************************************************************************
!
! Reads integration-related data
! By Stelios Rigopoulos
! 18/02/2002
!
*************************************************************************

use reaction_module !for n
use interface_module
implicit none
real(8), dimension(n), intent(inout) :: Dif
integer i

do i=1,6
Fortran 90 source code

```fortran
read(71,*)
end do
read(71,5001) gas_holdup
read(71,5001) k_l
read(71,5001) interface_depth
read(71,5001) Pressure
read(71,5001) Temperature
read(71,5001) Henry
read(71,5001) alpha
read(71,5001) pdeerr
read(71,5002) Nx
read(71,5002) Nt_int
read(71,5002) Itmax_int
read(71,5002) interval_int

do i=1,3
  read(71,*)
end do

do i=1,n
  read(71,5001) Dlf(i)
end do

5001 format(29X,F20.10)
5002 format(29X,I10)

end subroutine Interface_initdata

!*****************************************************************************
!*****************************************************************************
subroutine Interface_element(Clnltlal,Interface_area,Dlfspecies,y_gas,&
& mol_lnterface)

Calculates the Interfacial element according to the penetration model
By Stelios Rigopoulos
18/02/2002
!
*****************************************************************************

use PBE_module
use Interface_module
use reaction_module
use Integrator_module

implicit none

real(8), intent(in) :: interface_area         !Interfacial area
real(8), dimension(n,2), intent(in) :: y_gas    !Molar fraction of gas species
real(8), dimension(n), intent(in) :: Cnitial    !Initial concentrations
real(8), dimension(n), intent(in) :: Dispecies !Species' diffusion coefficients
real(8), dimension(n), intent(out) :: mol_interface !Total mol and moments in the end
real(8), dimension(n) :: C0                   !Initial concentration and moments vector
real(8), dimension(n) :: Dif                   !Diffusivity vector
```
real(8), dimension(Nx,n) :: C           ! Concentrations and moments
real(8), dimension(Nx,n) :: Cprev      ! Concentrations at previous time step
real(8), dimension(Nx,n) :: Ctemp      ! Concentrations at previous iteration
                    ! (for linearisation)
real(8), dimension(Nt_int) :: SSu       ! Supersaturation
real(8) :: Area              ! Integral of concentrations over contact time
real(8), dimension(Nx,Nx)   :: Coeff     ! Coefficients matrix
real(8), dimension(Nx)      :: Constant  ! Constants array
real(8), dimension(Nx)      :: Conc      ! Array returned by Thomas algorithm
real(8), dimension(Nx,n,n)  :: pd        ! Partial derivative of equation \( m \) at point \( Nx \) with respect to component \( m \)
real(8), dimension(n)       :: deriv     ! Subroutine, calculates the partial derivatives of the source terms
real(8) :: pderv,source      ! Interphase concentration of volatile component
integer :: Converge           ! Variable used in convergence check
integer :: iter               ! Iteration index
integer :: count              !
integer :: i,j,k,t,x          ! i: reaction index, j: species index
real(8) :: dx,dt              ! length of spatial and temporal steps
real(8) :: pH                  ! Function, calculates pH
real(8) :: vol_growthrate     !
integer :: jcrash

! Not used
vol_growthrate = 0.

! Preliminary calculations

! Calculation of interfacial equilibrium concentrations for volatile components

call Henry_const(Cinitial,Henry)
write(*,*) Henry

do i=1,n
  if (y_gas(i,2)>0.) then
    A_int(i) = y_gas(i,1)*Pressure/Henry
  end if
end do

! Create the initial concentration/moment vector

do i=1,n
  CO(i) = Cinitial(i)
  Dif(i) = Dispecies(i)
end do

! Initial Conditions
Fortran 90 source code

do j=1,n
    do x=1,Nx
        C(x,j) = CO(j)
    end do
end do

count = 0
! Length of steps
dt=contact_time/Nt_int
dx = interface_depth/Nx

! Time iteration
do t=1,Nt_int
    ! Store the concentrations of the previous time step (for implicit scheme)
    do j=1,n
        do x=1,Nx
            Cprev(x,j) = C(x,j)
        end do
    end do
    ! Calculate all source terms and partial derivatives
    do i=1,n ! Loop over equations
        do x=1,Nx ! Loop over space points
            Call derive(Cprev(x,i:n),i,deriv,n) ! Partial derivatives of equation i at point x
        end do
        do j=1,n
            pd(x,i,j) = deriv(j) ! Partial derivative of equation i at point x with respect to species j
        end do
    end do
    ! Linearisation iteration ---------------------------
    iter=0
    do ! To be repeated until linearisation has converged
        ! Keep current values to evaluate convergence
        do j=1,n
            do x=1,Nx
                Ctemp(x,j) = C(x,j)
            end do
        end do
        ! Update concentrations for next iteration
        ...
    end do
end do
end do

! Generation of coefficients matrix ---------------

do i=1,n  ! Loop over equations
    do x=1,Nx
        pder = 0.0

        ! Calculate source terms and partial derivatives at spatial point
        do j=1,n  ! Loop over species
            pder = pder + ( pd(x,i,j)*(C(x,j)-Cprev(x,j)) )
        end do
        pder = pder - pd(x,i,i)*(C(x,i)-Cprev(x,i))

        ! Coefficients ---------------

        ! Middle coefficient
        Coeff(x,x) = dx + 2.0*alpha*Dif(i)*dt/dx - alpha*dx*dt*pd(x,i,i)

        ! Left coefficient (not for first point)
        if (x>i) then
            Coeff(x,x-l) = -Dif(i)*dt/dx
        end if

        ! Right coefficient (not for last point)
        if (x<Nx) then
            Coeff(x,x+l) = -Dif(i)*dt/dx
        end if

        ! Constant term

        ! source(Cprev(x,1:j),i)
        Call ReactionSource(i,Cprev(x,1:n),source)

        Constant(x) = Cprev(x,i)*dx &
                      & - 2.0*Dif(i)*(1-alpha)*Cprev(x,i)*dt/dx &
                      & + source*dx*dt &
                      & + pder*dx*dt &
                      & - alpha*dx*dt*pd(x,i,l) &
        end do  ! End of space loop

    ! Boundary conditions ---------------

    ! 1: Interface

    if (y_gas(i,2)>0.) then

        ! Volatile component - Dirichlet condition

        Coeff(1,1) = Coeff(1,1) + alpha*Dif(i)*dt/dx
        Constant(1) = Constant(1) - Dif(i)*(1-alpha)*Cprev(1,1)*dt/dx + 2.0*A_int(i)*Dif(i)*dt/dx
    end if
else

! Non-volatile components - Newmann condition

Coeff(1,1) = Coeff(1,1) - alpha*Dif(i)*dt/dx
Constant(l) = Constant(l) + Dif(i)*(1+alpha)*Cprev(l,i)*dt/dx
end if

! 2: Bulk - Dirichlet condition for all

Coeff(Nx,Nx) = Coeff(Nx,Nx) - alpha*Dif(i)*dt/dx
Constant(Nx) = Constant(Nx) - Dif(i)*(1-alpha)*Cprev(Nx,i)*dt/dx + 2.0*C0(i)*Dif(i)*dt/dx

! Solve the algebraic equations

Call Thomas(Conc,Coeff,Constant,Nx)

do x=l,Nx

C(x,i) = Conc(x)

end do

end do ! End of equations loop

!-----------------------------------------------------------------------

! Check for convergence

Converge=1

do j=1,n

do x=1,Nx

if ((abs(Ctemp(x,j))<1.D-50).or.(abs((Ctemp(x,j)-C(x,j))/Ctemp(x,j))<pdeerr)) then

converge = converge * 1
else

converge = converge * 0
jcrash = j

end if

end do

end do

if (Converge==1) then

exit

else if ((Converge==0).and.(iter>Itmax_int)) then

write(*,*) 'Outer iteration in the interface did not converge after steps ',Iter
write(*,*) 'crashed on component', jcrash
stop
else

iter = iter + 1
Fortran 90 source code

end if
end do    !End of linearisation iteration
end do    !End of time integration

!Interface write-up

rewind(115)
write(115,*) t*dt

do i=1,Nx
  write(115,1198) real(i)*dx, C(i,1),C(i,2),C(i,3),C(i,4),C(i,5)
end do

!Calculate total mol in the interfacial element

do j=1,n
  !Area under curve C(x)-x
  Call Integrate_int(Area,C(l:Nx,j),dx,Nx)
  mol_interface(j) = Area*Interface_Area
  if ((mol_interface(j)<0.).and.(abs(mol_interface(j))>1.D-50))  then
    write(*,*) ' negative flux, stop - component ', j , ' flux', mol_interface(j)
    write(200,*) ' negative flux, stop - component ', j , ' flux', mol_interface(j)
    ! stop
  end if
end do

1198 format (F10.8,3X,<n>(E15.9,3X))

end subroutine Interface_element

!********************************************************************

subroutine deriveC(index,deriv,n)

!********************************************************************

!Calculates the partial derivatives of the non-linear terms
!Numerical
!22/01/2001

********************************************************************

implicit none

integer, intent(in) :: n
real(8), dimension(n), intent(in) :: C
integer, intent(in) :: index
real(8), dimension(n), intent(out) :: deriv
real(8), dimension(n) :: dC
real(8), dimension(n) :: CNew
integer :: i, j
real(8) :: source
real(8) :: r, renew
real(8) :: vol_growthrate ! CAREFUL

! Not used
vol_growthrate = 0.

! Perturbation vector
do i = 1, n
   dC(i) = 1.0D-8
end do

! Calculate source term
Call ReactionSource(index, C, source)
r = source

! Partial derivatives
do i = 1, n
   CNew(i) = C(i)
end do

do i = 1, n
   CNew(i) = CNew(i) + dC(i) * dC(index)
   Call ReactionSource(index, Cnew, source)
   mew = source
   deriv(i) = (1/dC(i)) * (rneu - r)
   CNew(i) = C(i)
end do

end subroutine derive

!**********************************************************************
!
! subroutine Integrate_int(Area, Cone, dx, Ne)
!
! **********************************************************************
!
! Integrates the concentrations along the interface
!
! By Stelios Rigopoulos
!
! **********************************************************************

implicit none
integer, intent(in) :: Ne
real(8), intent(in) :: dx
real(8), intent(out) :: Area
real(8), dimension(Ne), intent(in) :: Conc
ingter::
Area = 0.0
do i=l,Ne
   Area = Area+Conc(i)*dx
end do
end subroutine Integrate_int

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! Solution of tridiagonal matrices using the Thomas algorithm
!
! By Stelios Rigopoulos
!

! *************************************************************************************************************************
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implicit none

integer, intent(in) :: Ne
real(8), dimension(Ne), intent(in) :: constant
real(8), dimension(Ne,Ne), intent(in) :: Coeff
real(8), dimension(Ne), intent(out) :: X
integer :: i
real(8), dimension(Ne) :: P,Q

! Forward substitution
P(1) = Coeff(1,2)/Coeff(1,1)
Q(1) = constant(1)/Coeff(1,1)

do i=2,(Ne-1)
   P(i) = Coeff(i,i+1)/(Coeff(i,i)-Coeff(i,i-1)*P(i-1))
   Q(i) = (constant(i)-Coeff(i,i-1)*Q(i-1))/(Coeff(i,i)-Coeff(i,i-1)*P(i-1))
end do

P(Ne) = 0
Q(Ne) = (constant(Ne)-Coeff(Ne,Ne-1)*Q(Ne-1))/(Coeff(Ne,Ne)-Coeff(Ne,Ne-1)*P(Ne-1))

! Backward substitution
X(Ne) = Q(Ne)

do i=(Ne-1),1,-1
   X(i) = Q(i)-P(i)*X(i+1)
end do

end subroutine Thomas

!*************************************************************************************************************************
C.4 Other programs

FILE Bulk_module.f90

Integration of the differential equations describing the bulk of a fully mixed cell with possible input and output, where chemical reactions and particulate processes take place

This file contains:
- Module integrator_module: general declarations
- Subroutine integrator_initdata: initialise values
- Subroutine RHSide: calculates the right hand side of the ODEs to be integrated
- Subroutine Bulk: Performs the temporal integration
- Subroutine Moments: Calculates the moments of the distribution

By Stelios Rigopoulos

11/01/2002
18/01/2002: first version of modular program
03/02/2002: coupling of PSD and species' concentrations
31/03/2002: modification of vol_growthrate for scaled variables

module integrator_module

implicit none

integer :: Nt, tratio
integer :: flowflag, initflag, inputflag
integer :: interval
real(8) :: tau, tres
integer :: Itmax_coupling, coupling
real(8) :: err_coupling
real(8) :: relerr_n, abserr_n, relerr_C, abserr_C, machineprec
integer :: N_of_Int, maxno_of_steps
integer :: method

real(8) :: Total_Par

end module integrator_module

subroutine integrator_initdata

! Reads integration-related data
! By Stelios Rigopoulos
! 14/02/2002

use integrator_module

implicit none
integer i

do i=1,6
    read(61,*)
end do

read(61,5001) tau
read(61,5001) tres
read(61,5002) flowflag
read(61,5001) relerr_n
read(61,5001) abserr_n
read(61,5001) relerr_C
read(61,5001) abserr_C
read(61,5002) maxno_of_steps
read(61,5001) machineprec
read(61,5002) method

5001 format(29X,F20.10)
5002 format(29X,I10)

end subroutine integrator_initdata

!********************************************************************

Subroutine F (neqn,t,y,yprime)

!********************************************************************

! Calculates the right hand side of the DDEs to be integrated
! By Stelios Rigopoulos
! 14/01/2002
!
!********************************************************************

use integratorjmodule
use PBE_module
use reaction_module

implicit none

integer, intent(in) :: neqn
real(8), intent(in) :: t
real(8), dimension(neqn), intent(in) :: y
real(8), dimension(neqn), intent(out) :: yprime
real(8) :: reaction_source

integer i,index
real(8) :: weight_in,weight_out
real(8), dimension(6) :: Cin
common weight_in,weight_out,Cin

!********************************************************************
! Write current time
! write (*,*) t

! if the variable is a species' concentration
    do index=1,neqn
        Call ReactionSource(index,y,reaction_source)
        yprime(index) = + reaction_source + weight_in*Cin(index) - weight_out*y(index)
    end do
end subroutine F

! * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * *

Subroutine Bulk(ni,nin,C,C_in,w_in,w_out,integration_time,&
    moment,vol_growthrate)

! Directs the integration of ODEs
! for the Shampine and Gordon ODE/DE/STEP integrator
! By Stelios Rigopoulos
! 28/01/2002
! 03/02/2002: Coupling algorithm
! * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * * *

use PBE_module
use integrator_module
use reaction_module

implicit none

real(8), dimension(0:m), intent(inout) :: ni
real(8), dimension(0:m), intent(in) :: nin
real(8), dimension(n), intent(inout) :: C
real(8), dimension(n), intent(in) :: C_in
real(8), intent(in) :: integration_time
real(8), intent(in) :: w_in,w_out
real(8), intent(inout) :: vol_growthrate
real(8), dimension(0:1), intent(inout) :: moment
real(8), dimension(0:m) :: y
real(8), dimension(n) :: C_temp,C_prev,C_ave
real(8), dimension(m) :: ni_prev
real(8) :: dt
real(8) :: Vmean,Dmean,MomentIprev
integer :: t,i,j, index
real(8) :: count
real(8) :: pH, agg_degree
real(8) :: tin, tout
real(8) :: weight_in, weight_out
real(8), dimension(6) :: Cin
real(8) :: rtol, atol
integer :: neq, itol, itask, istate, iopt, lrw, liw, mf

!++++++++++ For Shampine and Gordon ODE +++++++++++++

external f
integer :: iflag
real(8), dimension (100+21*(n)) :: work
integer, dimension(5) :: iwork

common weight_in, weight_out, Cin

!--------------------------------------------------------

do i=1,n
    Cin(i) = C_in(i)
end do

weight_in = w_in
weight_out = w_out

!--------------------------------------------------------

! Integration of concentrations

tin = 0.
tout = integration_time
iflag = 1

do
    call ode(f, n, C, tin, tout, relerr_C, abserr_C, iflag, work, iwork)
    if (iflag==6) then
        write(6,*) 'warning - an error with the temporal integration &' &
        'of the population density has occurred, time is ', tin, ' iflag is ', iflag
    else if (iflag==2) then
        exit
    else
        write(6,*) 'continuing integration, flag is ', iflag, ' t is ', tin
    end if
end do

!--------------------------------------------------------

1190 format ('Number: ', 12.6, ' Volume: ', 12.6, ' Mean size: ', 12.6)
1195 format (FIG.4, 3X, 12.6, 3X, 12.6, 3X, 12.6, 3X, 12.6, 3X)
1198 format (FIG.4, 3X, <n>2G.6)
! Subroutine moments(ni,moment,agg_degree)

use PBE_module
use integrator_module
use reaction_module

implicit none

real(B), dimension(0:m), intent(in) :: ni
real(S), dimension(0:1), intent(inout) :: moment
real(B), intent(inout) :: agg_degree
integer :: i
real(B) :: vm,nim

Moment(0) = 0.0
Moment(1) = 0.0

do i=1,m

! Zeroth moment is calculated by trapezoid rule
Moment(0) = Moment(0) + 5.D-1*(ni(i-1)+ni(i))*dv(i)

! Interpolate mid-point
Vm = (v(i-1)+v(i))*5.D-1
Nm = ni(i-1)*v(i)/dv(i) - ni(i-1)*Vm/dv(i) + ni(i)*Vm/dv(i) - ni(i)*v(i-1)/dv(i)

! First and second moments are calculated by Simpson's rule
Moment(1) = Moment(1) + dv(i)*((1./6.)*ni(i-1)*v(i-1) + (4./6.)*Nm*Vm + (1./6.)*ni(i)*v(i))

Agg_degree = (Total_Par - Moment(0))/Total_Par

end do

end subroutine moments

! General-purpose subroutines
! This file contains:
! - Subroutine Growth_coeff: calculates growth coefficients as functions of supersaturation
! - Subroutine Nucleation_coeff: calculates nucleation coefficients as functions of supersaturation
! - Subroutine SS: calculates supersaturation
! - Function pH: calculates pH
!
! By Stelios Rigopoulos
! 11/01/2002
! 18/01/2002: first version of modular program
subroutine Growth_coeff(C)
! Calculates growth coefficients
! This implementation is for: 
! - Volume-dependent growth due to size-independent linear growth
! - Power law function
! Set Growth_kernel_type to 1 in PBE_source
use PBE_module
use reaction_module
implicit none
real(8), dimension(n), intent(in) :: C
!real(8), intent(out) :: G_coeff1,G_Coeff2,G_coeff3
real(8) :: SS
------------------------------------------------------------------------------------------------------------------------
Call SuperSaturation(C,SS)
if (abs(SS-0.)>threshold) then
   ! Growth rate constant, as function of supersaturation
   G_coeff1 = 3.*((Kv**(1./3.)) * G_ss_coeff * (SS**G_ss_exp)
else
   G_coeff1 = 0.
end if
! Volume exponent, due to size-independent linear growth
G_coeff2 = 2./3.
G_coeff3 = 0.
!g_coeff1 = 1.
end subroutine Growth_coeff

subroutine Nucleation_rate_add(C,mass)
! Calculates nucleation rate
! This implementation is for: 
! - Power law function
use PBE_module
use reaction_module
implicit none
real(8), dimension(n), intent(in) :: C
!real(8), intent(out) :: Nuc
real(8), intent(in) :: mass
real(8) :: SS

Call SuperSaturation(C,SS)

if (abs(SS-0.0)>threshold) then
  Nuc = 5.6*D6*(SS**6.) + 1.D13*(SS**0.1)*mass**1.066
  Nuc = 1.6*D6*(SS**4.) + 5.D12*(SS**0.1)*mass**1.
  Nuc = 1.67*(SS**4.) + 6.D11*(SS**0.1)*mass**1.012
  Nuc = 1.68*(SS**4.) + 1.D15*(SS**0.1)*mass**1.013
  Nuc = 2.07*(SS**4.) + 1.D15*(SS**0.1)*mass**1.015
else
  Nuc = 0.
end if
end subroutine Nucleation_rate_add

Subroutine SuperSaturation(C,SS)
Calculates supersaturation
use reaction_module
implieit none

real(8), dimension(n), intent(in) :: C
real(8), intent(out) :: SS
real(8) iota, alpha3, alpha5, C3_cor, C5_cor

if ((C(3)<=0.0).or.(C(5)<=0.0)) then
  SS = 0.0
else
  !Ionic strength
Iota = 0.5*( C(5)/1000*2**2. + C(3)/1000*2**2. + C(4)/1000 + C(2)/1000 )

\[
\begin{align*}
\alpha_3 &= \exp( -0.5085 \times 2.**2. \times (\sqrt{\text{iota}}/(1+\sqrt{\text{iota}}))-0.3 \times \text{iota} ) \\
C_{3, \text{cor}} &= \alpha_3 \times C(3)
\end{align*}
\]

\[
\begin{align*}
\alpha_5 &= \exp( -0.5085 \times 2.**2. \times (\sqrt{\text{iota}}/(1+\sqrt{\text{iota}}))-0.3 \times \text{iota} ) \\
C_{5, \text{cor}} &= \alpha_5 \times C(5)
\end{align*}
\]

\[
\text{SS} = \sqrt{C(3) \times C(5)} - \sqrt{\text{Ksp}}
\]

\[
\text{SS} = \sqrt{C_{3, \text{cor}} \times C_{5, \text{cor}} / \text{Ksp}} - 1.
\]

end if
if (SS<0.0) then
  \text{SS} = 0.0
end if

end subroutine SuperSaturation

! ***************************************************

real(8) function pH(BB)
! Calculates pH

implicit none
real(8) :: BB

if (BB>0.0) then
  \text{pH} = 14 + \log_{10}(BB*0.001)
else
  \text{pH} = 0.0
end if

end function

! ***************************************************

subroutine Aggregation_rate(C,agg)
! Calculates aggregation rate
! This implementation is for:
! - Dirac function of pH dependence

use PBE_module
use reaction_module
Fortran 90 source code

```
implicit none

real(8), dimension(n), intent(in) :: C
real(8), intent(out) :: agg
real(8) :: SS

!-------------------------------------------

Call SuperSaturation(C,SS)
if (abs(SS-0.)>threshold) then
  agg = aggl_kernel_const * SS**2.
else
  agg = 0.
end if
end subroutine Aggregation_rate

!***********************************************************************

!***********************************************************************

subroutine Breakage_rate(C,bre)
!Calculates breakage rate
!This implementation is for:
!- Dirac function of pH dependence
use PBE_module
use reactionjmodule
implicit none

real(8), dimension(n), intent(in) :: C
real(8), intent(out) :: bre

!-------------------------------------------

write(*,*)
bre = br_kemel_const
if (C(2)<0.001) then
  bre = br_const
else
  bre = 0.
endif
end subroutine Breakage_rate

!***********************************************************************

!***********************************************************************

subroutine Henry_const(C,Henry)
!Calculates Henry’s constant
```
use reaction_module

implicit none

real(8), dimension(n), intent(in) :: C
real(8), intent(out) :: Henry
real(8) Hw,h_CaOH2,h_CaHC032,h_CaC03

!----------------------------------------------------------

Hw = 10**(9.1229-0.059004*293+0.000078857*293**2)

!CaOH2
h_CaOH2 = (0.000053 + 0.000066 - 0.000015) * (0.5*(C(2)**2.+C(2)))

!CaHCG32
h_CaHC032 = (0.000053 + 0.000021 - 0.000015) * (0.5*(C(3)**2.+C(3)))

!CaC03
h_CaC03 = (0.000053 + 0.000021 - 0.000015) * (0.5*(C(4)**2.+C(4)**2.))

Henry = 10.**(-(h_Ca0H2+h_CaHC032+h_CaCG3))*Hw
Henry = l./Henry*100.
end subroutine Henry_const

!----------------------------------------------------------

Subroutine file_open(file_flag)

!Opens and closes the necessary files

implicit none

integer, intent(in) :: file_flag

!------------------------------------------------------------------------------------------------------------------------

if (file_flag==1) then

open(31,file='input/inputPBE.in')
open(34,file='input/initdis.in')
open(35,f ile='input/inputdis.in')
open(41,file='input/inputreng.in')
open(43,file='input/initconc.in')
open(44,file='input/inputconc.in')
open(61,file='input/integration.in')
open(71,file='input/interface.in')
open(101,file='output/moments.out')
open(103,file='output/species.out')
open(105,file='output/PSD.out')
open(106,file='output/PSDlast.out')
open(111,file='output/initdis.out')
open(113,file='output/inputdis.in')
open(115,file='output/interface.out')
open(119,file='output/nucleation.out')
open(200,file='output/error.log')

else

close(31)

end subroutine file_open

!------------------------------------------------------------------------------------------------------------------------
close(34)
close(35)
close(41)
close(43)
close(44)
close(61)
close(71)
close(101)
close(103)
close(105)
close(111)
close(113)
close(115)
close(200)
end if
end subroutine file_open

**********************************************************************************************

FILE reaction_module.f90
!
! Chemical reaction engineering subroutines
! This file contains:
! - Module reaction_module: all general declarations
! - Subroutine reaction_source: source terms in species' balances due to reactions
! - Subroutine reaction_initdata: read values of reaction parameters
! - Subroutine reaction_initarrays: initialise concentrations
!
! By Stelios Rigopoulos
! 11/01/2002
! 18/01/2002: first version of modular program
!
**********************************************************************************************

module reaction_module
implicit none
integer :: n
real(8) k1,k2,Keq1,Keq2,rho,MW,Ksp,kv

end module reaction_module

**********************************************************************************************

Subroutine ReactionSource(index,C,reaction_source)

**********************************************************************************************

! Reaction kinetics
!
! This implementation is for:
! CaCO3 precipitation


! By Stelios Rigopoulos (14/01/2002)
!
******************************************************************************

use reaction_module
use PBE_module !for v_nuc

implicit none

integer, intent(in) :: index
real(8), dimension(n), intent(in) :: C
real(8), intent(in) :: vol_growthrate
real(8), intent(out) :: reaction_source
real(8) :: Growth
real(8) :: r1, r2, r3
real(8) :: temp

******************************************************************************

! Define kinetics

! Call Nucleation_rate(C)

r1 = k1*C(1)*C(2) + (k1/Keql)*C(4)

r2 = k2*C(4)*C(2) + (k2/Keq2)*C(3)

r3 = rho*(l/MW)*vol_growthrate + rho*(l/MW)*v_nuc*Nuc

******************************************************************************

! Define non-linear source terms

if (index==1) then
  reaction_source = -r1
else if (index==2) then
  reaction_source = -r1 -r2
else if (index==3) then
  reaction_source = r2 !-r3
else if (index==4) then
  reaction_source = r1 -r2
else if (index==5) then
  reaction_source = 0. !-r3
end if

end subroutine ReactionSource

******************************************************************************

******************************************************************************
subroutine reaction_initdata

!
! Reads kinetic data
! By Stelios Rigopoulos
! 09/02/2002
!
!*****************************************************************************

use reaction_module
implicit none
integer i

*****************************************************************************

do i=1,9
  read(41,*)
end do

read(41,5002) n
!
!Read kinetic data from file
!
read(41,5001) k1
read(41,5001) k2
read(41,5001) Keq1
read(41,5001) Keq2
read(41,5001) rho
read(41,5001) MW
read(41,5001) Ksp
read(41,5001) Kv

*****************************************************************************

5001 format(29X,F20.10)
5002 format(29X,I10)

end subroutine reaction_initdata

*****************************************************************************

*****************************************************************************

subroutine reaction_initarrays(C,Cin)

*****************************************************************************

!
! Initialises arrays
! By Stelios Rigopoulos
! 09/02/2002
!
!*****************************************************************************

use reaction_module
implicit none
real(8), dimension(n), intent(inout) :: C,Cin

integer i
do i=1,n
    Cin(i) = 0.
    C(i) = 0.
end do
!
Initial distribution
rewind(43)
do i=1,6
    read(43,*)
end do
doi=1,n
    read(43,5001) C(i)
end do
!
Input distribution
rewind(44)
do i=1,6
    read(44,*)
end do
doi=1,n
    read(44,5001) Cin(i)
end do

5001 format(29X,F20.10)
5002 format(29X,110)
end subroutine reaction_initarrays

program main
    call network()
end program main
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