The past, present and future of Multi-Scale Modelling applied to wave-structure interaction in Ocean Engineering

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Abstract

Concepts and evolution of multi-scale modelling from the perspective of wave-structure interaction have been discussed. In this regard, both domain and functional decomposition 10 approaches have come into being. In domain decomposition, the computational domain is 11 spatially segregated to handle the far-field using potential flow models and the near field using Navier-Stokes equations. In functional decomposition, the velocity field is separated into irrotational and rotational parts to facilitate identification of the free surface. These two approaches have been implemented alongside partitioned or monolithic schemes for 15 modelling the structure. The applicability of multi-scale modelling approaches has been established using both mesh-based and meshless schemes. Owing to said diversity in numerical techniques, massively collaborative research has emerged wherein comparative numerical studies are being carried out to identify shortcomings of developed codes and establish best-practices in numerical modelling. Machine learning is also being applied to 20 handle large-scale ocean engineering problems. This paper reports on the past, present and 21 future research consolidating the contributions made over the past 20 years. Some of these 22 past as well as future research contributions have and shall be actualized through funding 23 from the Newton International Fellowship as the next generation of researchers inherits the 24 present-day expertise in multi-scale modelling.

1 Introduction

In this paper, modelling tools and approximations that are in practice for wave-structure interactions (WSI) are discussed. Emphasis is provided for ocean engineering which encompasses offshore and coastal engineering, naval architecture as well as ocean sciences. Thus large time/spatial scale and local time/spatial scales are important. Different modelling aspects based on their level of approximation or theoretical understandings are discussed. This leads to understanding the limitations of the numerical tool, subsequently emphasising how and what to interpret from the results. In recent years, coupling of these standalone tools is being extensively implemented to resolve various levels of the physical process. This is discussed in detail after the brief explanation of the individual tools and how the development took place in each of these modelling efforts.

At present, these numerical models are available as open-source as well as commercial tools using different numerical methods. Thus, said models have varying degrees of approximations in spatio-temporal resolution, stability, accuracy and computational efficiency. Hence, one of the recent efforts in the numerical modelling community is the comparative and benchmarking exercises; this shall also be discussed in the present paper. So 41 the readers can test their own development/existing tools using any of these benchmark tests, available theory and open-source experimental data.

In this paper, apart from providing an overview of the existing tools, a proper classification of the models, their applicability range, computation and physical processes, a thorough

literature review on the history of developments are provided. It should be noted that the details of each of the presented models is beyond the scope of this paper and the reader may refer the corresponding literature cited. Through the course of this review, several numerical techniques pertaining to multi-scale modelling shall be covered. However, we refrain from making any best practices recommendations as these strongly depend on the problem at hand and, thus, could be highly subjective. Rather, the aim of this review is to provide the reader with a comprehensive listing of the available methods. This listing has been developed based on the authors' prior experience with multi-scale modelling as well as through a comprehensive review of the state-of-the-art. In the following sections, these mathematical models are discussed with their governing equations to handle physical problems, assumptions, implementation strategies and adopted numerical methods along with their applications. The existing numerical efforts carried out worldwide are provided along with a detailed discussion on numerical model development actualized by the authors' research group that has been supported in-part by the Newton International Fellowship. The remainder of the paper is structured as follows: the spatio-temporal scales associated with various physical processes in ocean engineering along with application-specific levels of approximation necessary in a given model are discussed in §2, the depth-resolving Navier-Stokes models along with numerical strategies for solution, wave/current generation and absorption, free-surface tracking as well as turbulence modelling have been discussed in detail in §3; potential flow models are introduced in §3.2, the depth-averaged Boussinesq-type models are discussed in §4, the state-of-the-art in global and regional-scale ocean science multi-scale modelling is presented in detail in §5, multi-scale modelling achieved through coupling of different models is discussed in detail from the standpoint of both domain as well as functional decomposition strategies in §6, the past and present effort of benchmarking numerical models through comparative studies is reviewed in §7 and finally the future of multi-scale modelling in WSI is discussed from the standpoint of AI/ML techniques as well as the development of hybrid models for floating renewables in §8. The reader will appreciate that significant effort has been made to cover a broad range of modelling techniques in ocean engineering in general and WSI in particular. However, this review is not all-inclusive and hence some fields of research such as hydroelasticity, metocean analysis, wind-wave interaction and phase-averaged wave action modelling could not be included.

2 Different Levels of Approximations

A single numerical tool to address all class of problems in ocean engineering is ideal. However such a model is not possible due to the following reasons: (a) a large sea area, having a large range of spatial and time scales, (b) highly nonlinear wave-structure interaction process (here not only fluid, sometimes the structure can also behave nonlinearly such as vegetation or fenders or hydro-elasticity), (c) waves co-exist with nonlinear currents of various levels, sediment transport and others, (d) viscosity, surface tension and turbulence, (e) two phase (air-sea) or multiphase processes (air-sea-oil or air-sea-sediment), (f) violent wave impacts (during cyclonic storm surges, flooding) and aeration on rubble mound structures, green water shipping and slamming. For these above phenomena, one needs to model large spatial/time scale to capture wave propagation phenomenon as well as resolve

small spatial/time scale to understand the wave-structure(-soil) interactions processes. A single mathematical model may not always be a solution for this complex problem. Hence, the researchers have developed various levels of approximations in the mathematical modelling.

The level of approximations in the mathematical modelling are decided based on two guiding principles: (a) which physical process is governing the problem at hand and (b) strive to minimize the computational effort in the resulting numerical algorithm for industrial/practical application by balancing computational efficiency and fidelity. In context to the first principle, the requirement of modelling a physical process is mapped with respect to various applications in Table 1. Table 1 lists various applications that require either large domain or local/small domain modelling.

100 Table 1. A summarization of the various physical processes and the requirement for them to 101 be modelled for various large and small-scale ocean engineering applications (the

102 information is based on the authors' experience with multi-scale modelling).

			PHYSICAL PROCESSES					
			Surface	Viscous effects and/or	Nonlinearity	Nonlinearity	Modelling the	
			tension	turbulence modelling	in fluid	in structure	air-phase	
		Current/flow-structure	N _o	It depends	Yes	N _o	N _o	
		interaction						
		Wave propagation and interaction	No	It depends	It depends	N _o	N _o	
	LARGE-SCALE	Seakeeping/motion of marine structures	No	It depends	Yes	It depends	N ₀	
		Geophysical flows	No	It depends	It depends	No	It depends	
		Sediment transport	N _o	Yes	Yes	N _o	No	
APPLICATIONS		Wave-breaking	N _o	Yes	Yes	Yes	It depends	
		Aeration dynamics	It depends	Yes	Yes	Yes	Yes	
		Wind-wave interaction	It depends	Yes	Yes	No	Yes	
		Steep wave and rigid structure	Yes	It depends	Yes	N _o	No	
	ALE LOCAL-/SMALL-SC	Extreme waves and rigid structure	It depends	It depends	Yes	No	Yes	
		Wave-structure-soil interaction	It depends	It depends	Yes	No	No	
		Wave-deformable structure-interaction	It depends	It depends	Yes	Yes	Yes	
		Current/flow-structure interaction	No	Yes	Yes	It depends	No	
		Sediment transport	N _o	Yes	Yes	No	N _o	
		Wave-breaking	It depends	Yes	Yes	It depends	Yes	
		Aeration dynamics	Yes	Yes	Yes	It depends	Yes	
		Wind-wave interaction	It depends	Yes	Yes	No	Yes	

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104 A typical example for large domain modelling in coastal engineering is wave propagation 105 from offshore to near shore and its interactions with a harbour structure to understand its 106 tranquillity, run-up or inundations. Similar examples from naval architecture and offshore engineering would be ship maneuvering under the action of waves and an offshore wind turbine farm interacting with waves or offshore platform interactions with waves, respectively. For these applications, the physical processes such as surface tension and nonlinearities in structural response are not important. Further, complete physics in modelling the air-sea process is also not required; some empirical treatment would be deemed sufficient. Thus, the full continuity and momentum equations can be simplified based on these approximations.

Similarly, consider an application of small domain modelling, wherein one is interested in quantifying the forces experienced by structures (such as ships, semi-submersible platforms, seawalls, scour around monopiles and jackets etc.) against operating or extreme sea state conditions. In this scenario, normally researchers would carry out the physical model studies in an experimental wave tank. A similar study can be done using numerical modelling based on so-called numerical wave tanks. A numerical wave tank is a numerical tool that could reproduce the experimental facility with a high degree of fidelity. Thus, a detailed physical flow process is realized by solving the continuity and momentum equations, only slightly reducing the physical approximations, nonetheless reproducing the dominating forces as close to reality as possible. For instance, in coastal engineering, surface tension, nonlinearity in structure, sediment to sediment interactions or rigid body interactions (say in a rubble-mound breakwater) can be relaxed without greatly compromising the fidelity of the numerical approximation. Thus, for large scale problems, one can employ various levels of approximations based on the wave characteristics and its applications thus leading to savings in the computational cost. This aspect of modelling is further emphasised by means of a bubble plot in Figure 1 wherein the various environmental aspects in spatial and time scale for mathematical modelling of wave-structure interaction as well as wave-propagation are illustrated.

Figure 1. Bubble plot variation to represent the spatio-temporal scales of various physical

processes in ocean engineering.

Figure 1 showcase various processes ranging from climate change, sea-level rise, morphodynamics, tides, tsunami/storm surges, wave propagation over varying bathymetry from offshore to their interactions with structures. Each of these cases has a different horizontal spatial scale from 1mm to more than 10,000 km and time scale from less than 1s to 100 years. Further, which type of modelling is dominant and ought to be carried out is also represented (in brackets) in Figure 1, along with global scale or regional scale modelling. The mathematical modelling approximations based on depth averaging can be seen as predominant for increasing large scale problems. This is based on the assumptions of the vertical flow structure. When the vertical flow motion is considered weak or insignificant, then depth averaged horizontal velocities can be adopted. Such classifications of mathematical models are called as depth averaged models and depending upon the approximations adopted in the horizontal velocities different models are available. This will be discussed in the later part of this paper. When the time scale and horizontal spatial scale are small, then the wave-structure interaction becomes dominant; in such cases depth resolving models are normally adopted. This is solved based on Navier-Stokes equations (NSE) with various simplified approximations. Depending upon the application (such as porous-structure, vegetation interactions, hydroelasticity or sediment transport) either microscopic or macroscopic modelling can be adopted within the NS framework to model the structure interaction process. On the other hand, the physical process involved in the ship manoeuvring is of the order of kilometers and minutes at the prototype-scale, however for numerical modelling the same would normally be carried out at a reduced scale using depth resolving models. Hence, for some applications, although the physical process is at a large scale, the numerical simulations are normally carried out at a smaller scale due to computational limitations.

In the past decades, one of the major reasons for resorting to the different physical approximations to model the different scales of the problem was to reduce the computational time. However, this leads to a compromise on the physics of the problem. Figure 2 shows three different broader classifications namely depth averaging, depth resolving and hybrid models. In this broader classification, different governing equations for modelling based on approximations are available, which are currently in practise within the numerical modelling community. The basic modelling task in each case is to solve the continuity and momentum equations for the fluid dynamics problem. However, the modelling complexity increases based on the physical problem being addressed, type of structure (coastal, offshore or marine) and type of sea-state under consideration.

In coastal engineering, the majority of the structures (e.g. breakwaters, sea-walls and pile structures) are fixed or stationary. Then the physical problem to represent is the wave transformation process (i.e., wave shoaling, diffraction, refraction, reflection, wave-overtopping and wave-breaking) and its interaction with the structures. In case of offshore 173 engineering, the structures may be fixed (e.g. offshore wind turbine foundations in ≤ 50 m deep water) or floating (e.g. oil production platforms, floating offshore wind turbines and floating solar arrays). In the latter case, the modelling complexity increases because the fluid flow and structure motion(s) are coupled and thus need to be solved in conjunction; failure to do so would over-predict the hydrodynamic loads. Nonetheless, the overall excursion of an offshore structure is small when compared to marine structures such as a ship or submarine.

For a marine structure, the numerical modelling needs to account for large displacements (e.g. ship maneuvering in waves) thus necessitating large domains and, if the sea-state is violent, also hydroelasticity plays a role (e.g. hull-slamming in violent sea-states). Nonetheless, these scenarios may not always necessitate the NS equations; potential-flow models are a viable alternative as long as the hydrodynamic loads and resulting body motions are properly accounted for. For instance, models based on the Boussinesq equations are quite popular for modelling wave tranquillity and recently, for ship-generated waves.

Figure 2. Different modelling strategies characterized by the level of physics approximation 188 and the resulting computational cost.

Various numerical methods are currently in practice to solve a given mathematical model. The numerical methods are broadly classified into strong and weak forms. The traditional methods such as the Finite Difference Method (FDM), the Finite Element Method (FEM), the Boundary Element Method (BEM) and the Finite Volume Method (FVM) as well as modern techniques such as particle/mesh-free methods are being employed in the ocean engineering problems. Mostly, the choice of the numerical methods depends upon the developers and one is not superior to the others as one might expect. Each of these numerical methods has their own advantages and disadvantages, and the overall goal is to reduce or minimize the disadvantages using numerical treatments/algorithms/schemes.

3 Depth Resolving Mathematical Models

In the present section, we review the depth-resolving models. These models are mostly used for the wave-structure interaction problems to estimate the wave loads, wave damping characteristics and motion/structural responses. The problems that are based on small spatial and time scale are normally handled by the depth-resolving approach which models the

203 physical process using a high (spatio-temporal) resolution thus leading to high computational 204 costs.

205 3.1 Navier Stokes Equations

The Navier-Stokes equations include the equations governing the conservation of mass (termed "equation of continuity" (EOC) for incompressible flows which is in turn a reasonable assumption for WSI) and conservation of momentum. The term "full" indicates 209 the *absence* of simplifying assumptions such as irrotationality, depth-averaging, Reynolds-averaging, two-dimensionality, axisymmetry, single-phase nature of the flow (density is spatio-temporally constant) etc.

212 3.1.1 Governing Equations

- 213 The full Navier-Stokes equations (NSE) governing fluid motion are written here in
- 214 differential form for the instantaneous velocity field \vec{V} :

$$
\frac{\partial \rho}{\partial t} + (\vec{V} \cdot \vec{\nabla}) \rho = 0
$$
\nconservation of mass\n
$$
\frac{\partial \rho \vec{V}}{\partial t} + \vec{\nabla} \cdot (\rho \vec{V} \otimes \vec{V}) = \frac{-\vec{\nabla} p}{\text{arduction}} + \vec{\nabla} \cdot \left(\mu \left\{ \vec{\nabla} \vec{V} + (\vec{\nabla} \vec{V})^T - \frac{2}{3} (\vec{\nabla} \cdot \vec{V}) \vec{\mathbb{I}} \right\} \right) + \rho \vec{g}
$$
\ntime
\ntime $\frac{\partial t}{\partial t} + \vec{V} \cdot (\rho \vec{V} \otimes \vec{V}) = \rho \vec{r} \cdot \frac{\partial \vec{V}}{\partial t} + \vec{V} \cdot (\mu \left\{ \vec{\nabla} \vec{V} + (\vec{\nabla} \vec{V})^T - \frac{2}{3} (\vec{\nabla} \cdot \vec{V}) \vec{\mathbb{I}} \right\} + \rho \vec{g}$ \n(1)

215 where, p is the total pressure, ρ and μ are the density and viscosity respectively, \bar{I} is the 216 identity tensor and \vec{q} is the gravitational acceleration vector. Equation (1) represents the 217 compressible Navier-Stokes equations, however, the compressibility of the fluid may only be 218 important during violent wave-structure interaction at large-scale. For the remainder of the applications, the conservation of mass simplifies to the equation of continuity (EOC): $\vec{\nabla} \cdot \vec{V} =$ 220 0 which holds for incompressible flow. It is also worth noting that Equation (1) is written for 221 the "instantaneous" velocity-field (Anghan *et al.*, 2019) indicating that \vec{V} is neither time-222 averaged (RANS) nor spatially-filtered (LES). The fluid properties ρ and μ can be replaced 223 with the mixture properties ρ^* and μ^* to account for the presence of multiple contiguous 224 phases in the domain. Here, advantage is derived from the fact that the phases can be 225 considered as being "individually incompressible" (Saincher and Banerjee, 2018) for most 226 applications which precludes the necessity of solving (say) N sets of the NSE for N phases. 227 This results in the so-called "single-fluid formulation" wherein the entire computational 228 domain is assumed to be filled with a single, albeit, variable-property fluid (Saincher and 229 Sriram, 2022a). It should also be noted that within the single-fluid framework, equation (1) is 230 "conservative" (Saincher and Sriram, 2023) meaning ρ is on the left-hand-side with the time 231 and advection terms. On the other hand, the formulation would be termed "non-conservative" 232 if ρ were on the right-hand-side with the pressure and diffusion terms. The positioning of ρ in 233 the governing equations is immaterial for a single-phase treatment of the NSE (for instance 234 cf. Sriram et al., 2014). The same, however, would have far-reaching consequences for a 235 multiphase framework especially for violent flows involving wave-breaking and/or slamming 236 loads; a conservative formulation is recommended in these cases (Saincher and Sriram, 237 2023). However, an important limitation of the conservative formulation is that it may lead to

238 the formation of unrealistically large velocities at the interface (Tryggyason *et al.*, 2007) and thus is deemed unnecessary for more benign wave propagation and WSI scenarios. In context 240 to equation (1), it is also worth mentioning that the total pressure p is comprised of static, 241 hydrostatic as well as dynamic contributions; \dot{v} is not the true pressure but rather a pseudo pressure which satisfies the EOC. The advantage with WSI and ocean engineering problems in general is that the simulation begins from quiescent/calm water conditions which allows for a very accurate "guess" of the initial pressure field using the hydrostatic law. This results in a dynamic pressure field that is very close to the true (say experimentally measured) dynamic pressure, once the hydrostatic contribution has been removed (Saincher and Sriram, 2022a ; 2022b).

3.1.2 Solving the Navier-Stokes Equations

For a given flow problem, the solution variables of interest include the velocity \vec{V} and

250 pressure p . It is characteristic of the incompressible Navier-Stokes equations to *not* have a

separate equation for pressure. Owing to this, a majority of incompressible NSE flow solvers

- are based on a predictor-corrector approach which was pioneered by Chorin (1967); the same
- is illustrated in Figure 3.

Figure 3. A typical predictor-corrector loop characteristic of projection methods pioneered by Alexandre Chorin in 1967.

257 At the beginning of the solution, both \vec{V}^{n+1} and p^{n+1} at the current time-level are unknown

- 258 and the momentum equations are solved for a predicted velocity field \vec{V}^* wherein either:
- 259 **•** the pressure term $\left(-\frac{1}{\rho^*}\vec{\nabla}p\right)^n$ from the previous time-level is considered (Saincher and Banerjee, 2015) or,
- 261 the pressure term is not considered at all which was the case with Chorin's original method (normally adopted in Meshfree methods ; cf. Sriram and Ma, 2021).

263 At this point, the incompressibility condition $\vec{\nabla} \cdot \vec{V}^{n+1} = 0$ is invoked at the current time-264 level and the same is split into a mass defect $\vec{\nabla} \cdot \vec{V}^*$ and divergence correction $\vec{\nabla} \cdot \vec{V}^*$ 265 contributions. This marks the end of the "predictor-step" (highlighted in red in Figure 3). Following this, the property $\vec{V} = -\frac{\Delta t}{c}$ 266 Following this, the property $\vec{V} = -\frac{\Delta t}{\rho} \vec{\nabla} p$ is invoked to establish a relationship between either 267 $\vec{\nabla} \cdot \vec{V}^*$ and p^{n+1} (Sriram and Ma, 2021) or between $\vec{\nabla} \cdot \vec{V}^*$ and the pressure correction p' 268 (Saincher and Banerjee, 2015). In either case, one ends up with a Pressure Poisson Equation 269 (PPE) which needs to be iteratively solved for p^{n+1} (or an Equation Of Pressure Correction 270 (EOPC) which needs to be iteratively solved for p'). This is oftentimes the most 271 computationally-intensive step in a flow solver. Following solution of the Poisson equation, \vec{V}^{n+1} can be obtained using $\vec{V} = -\frac{\Delta t}{\Delta t}$ 272 \vec{V}^{n+1} can be obtained using $\vec{V} = -\frac{\Delta t}{\rho} \vec{\nabla} p$ which marks the end of the "corrector-step" 273 (highlighted in green in Figure 3). The splitting of the solution into predictor and corrector 274 steps is also known as the "projection method" since the pressure is used to project \vec{V}^* onto a 275 space of divergence-free velocity-field which is essentially the Helmholtz decomposition.

Various flow solvers (or so-called "pressure-velocity coupling" schemes) such as SIMPLE, PISO, PIMPLE essentially have the same predictor-corrector constitution but differ with 278 regards to how \vec{V}^* is calculated as well as the number of predictor-corrector cycles per time-279 step. In fact, regardless of whether \vec{V}^* is computed fully-explicitly or semi-implicitly (because a fully-implicit treatment of the advection term is not possible), the solver still 281 belongs to the SIMPLE class of algorithms (Ferziger *et al.*, 2020). However, some authors 282 also call the fully-explicit category of algorithms "semi-explicit" (Dave *et al.*, 2018; Sharma, 2022) owing to the implicit nature of solution of the PPE. It is important to note that, for a given order of time-discretization, the solutions obtained from a fully-explicit or semi-implicit predictor step should be identical. Nonetheless, the semi-implicit treatment would accord further stability to the solution.

287 In context to WSI, a forward Euler time-discretization and fully-explicit evaluation of \vec{V}^* has 288 been extensively used by the authors (Saincher and Sriram, 2022a ; 2022b ; 2023). From the 289 authors' experience, explicit (forward Euler) time discretization is recommended for waves 290 owing to the hyperbolic nature of solution propagation and a fully-explicit evaluation of \vec{V}^* 291 was found to be sufficient for relatively benign WSI scenarios especially ones that did not 292 involve slamming loads. In fact, it is demonstrated in Saincher et al. (2023a ; 2023b) that a 293 fully-explicit evaluation of \vec{V}^* works even in slamming conditions for modest mesh 294 resolutions. Thus, the CFD user ought to make an informed decision whilst selecting the 295 pressure-velocity coupling scheme keeping in mind the trade-off between numerical stability 296 (better for semi-implicit treatment) and computational efficiency (better for fully-explicit 297 treatment). Unfortunately, users of commercial CFD solvers seldom have fully-explicit 298 pressure-velocity coupling available to them and thus alternatively opt for (say) the PISO solver with a Non-Iterative Time Advancement (NITA) option available in ANSYS® 299 300 FLUENT.

301 When the predictor and corrector steps are considered in conjunction, say for a 3D flow 302 problem, a single time-step would have one iterative solution loop (for p) in case of a fully-303 explicit solver and four (for U, V, W, p) in case of a semi-implicit solver. However, our 304 experience suggests that the computational effort required for solving p may sometimes be 305 greater than the three velocity components U, V, W combined. This is primarily because of 306 differences in the rate of convergence which is in turn dependent on the type of boundary 307 conditions involved. The boundary conditions are predominantly Dirichlet in case of 308 velocities which results in predominantly Neumann conditions for the pressure thus leading 309 to an increase in the computational effort for solving the PPE.

310 3.1.3 Boundary conditions – Wave/Current Generation and Absorption

A prerequisite to accurate WSI simulations in ocean engineering applications is high fidelity wave generation as well as reflection-free absorption of waves/currents in the computational domain. The task of absorption is generally more challenging for WSI simulations involving regular and irregular waves when compared to focusing waves primarily due to the larger number of wave cycles/periods involved in the former case. The task of absorption also becomes complex if currents co-exist with waves. The various methods of wave/current generation and absorption in NSE-based NWTs are mapped against their numerical characteristics in Table 2.

Wavemaker		\boldsymbol{U}	V	W	p	n	
Inflow-boundary		Dirichlet	Dirichlet	Dirichlet	Dirichlet	Dirichlet	
Mass-source function						Source-term in EOC	
Momentum-source function		Source-term in momentum equation					
Internal inlet				Dirichlet			
Relaxation zone		Dirichlet	Dirichlet	Dirichlet	Dirichlet	Dirichlet	
	Flap / Piston type	Prescribed		Prescribed			
Moving wall	Segmented type	motion	--	motion			
Wave-absorber		U	V	W	\boldsymbol{p}	η	
Outflow boundary		Orlanski / Continuity / Sommerfeld radiation boundary condition					
Sponge-layer		Sink terms in momentum equation					
Relaxation zone		Solution gradually ramped from/to wave theory to/from numerical model					
Moving wall (active absorption)		Prescribed		Prescribed			
		motion		motion			
Adaptive passive absorption		Adaptively predicted using on-board elevation		Neumann	Neumann		

319 Table 2. Type of wave/current generation and absorption strategies in NSE-based NWTs.

With reference to Table 2, the development of "numerical wavemakers" for NSE models was pioneered by Lin and Liu (1998; 1999) wherein the inflow-boundary and mass-source function techniques were proposed. As seen from Table 2, the inflow technique involves a Dirichlet prescription of the wave-induced orbital velocities (predicted from a suitable wave theory) as well as the free-surface elevation at the domain boundary. The present research group has proposed a modified inflow technique to improve the volume-conservation properties of inflow-boundaries, particularly for scenarios involving strong Stokes drift such as steep wave generation in near-shallow water (Saincher and Banerjee, 2017a).

In conjunction with inflow boundaries, the mass-source function technique was also developed which involved the modification of the EOC through the inclusion of a time-varying source term that is in turn proportional to the wave elevation. Wave generation is achieved through periodic ejection/ingestion of water-volume from/into the source region and this offers some advantages over inflow-boundaries. For instance, the only wave 333 characteristic to be input is the time-varying free-surface elevation $\eta(t)$ and thus wave-records from the field could be reproduced. Also, waves reflected from the domain boundaries would not interfere with the wave generation. Nonetheless, the source region itself has several design variables requiring parameterization and, in this context, the authors have proposed guidelines to decide the geometry, placement and strength of the source region based on the relative depth and wave-steepness (Saincher and Banerjee, 2017b). As listed in 339 Table 2, other similar methods have also been proposed such as the internal inlet (Hafsia et al., 2009) and momentum-source function (Choi and Yoon, 2009) techniques. Some researchers have also attempted to directly model piston/flap-type wave-paddle motions into their NWTs using embedded boundary treatment for the solid (cf. fast-fictitious-domain 343 (FFD) based modelling of wave-paddles in Anbarsooz *et al.* (2013)).

However, currently, the most popular technique of numerical wave generation is the so-called 345 "relaxation zones" developed by Jacobsen et al. (2012) for OpenFoam®. Here, the solution is spatio-temporally "ramped-up" from wave-theory to NSE before the structure/region of interest and again "ramped-down" from NSE to calm-water conditions after the structure/region of interest. Thus, relaxation zones not only prevent upstream reflection of waves from the far-end of the NWT but also downstream re-reflection of waves reflected off the structure. It is also worth mentioning that relaxation zones in and of itself is a more general concept that has been implemented in hybrid potential theory-NSE models (Agarwal $et al., 2022b)$ as well as in hybrid spectral theory-NSE models (Aliyar *et al.*, 2022).

Apart from relaxation zones, other methods of wave absorption have also been implemented for NSE-based NWTs. For instance, Lin and Liu (1999) employed a radiation/outflow boundary condition for wave absorption at the far-end of the NWT. Outflow boundaries 356 generally implement the Sommerfeld condition (Dave *et al.*, 2018): $\frac{\partial \phi}{\partial t} + C \frac{\partial \phi}{\partial n} = 0$ where ϕ 357 is the property to be effluxed from the boundary, t is time, C is the phase velocity and n 358 points normal to the boundary. The prescription of C is relatively straightforward for "flow problems" making outflow boundaries suitable for tsunamis, tidal flows, scour etc. which involve a dominant current component. Sommerfeld conditions are also suitable for absorbing small-amplitude waves. However, these pose a challenge for absorbing steep 362 waves particularly because $\mathcal C$ is spatio-temporally variable along the boundary. It has been 363 shown in Dave et al. (2018) that improper prescription of C leads to severe (inward) reflections even for free-shear flows.

Self-adaptive wavemaker theory has also been used popularly in both the physical and numerical wave tanks. This method utilizes wavemaker (moving wall) whose motion is specified to generate both the incident waves and an additional wave to cancel the undesirable wave (e.g. the reflected wave from somewhere within the tank). More details can 369 be found in Yan *et al.* (2016). In addition, the same concept of "adaptive absorber" was also 370 used in our recent work on developing a passive wave absorber (Yan *et al.*, 2020). This boundary behaves similarly to the inflow boundary, however the fluid velocity condition is specified by considering its relation with the wave elevation recorded at the boundary. This method does not require the use of the relaxation zone for wave absorption and thus results in a considerable improvement of the computational efficiency. A recent application of the same 375 can be found in Xiao *et al.* (2024) .

3.1.4 Free surface capturing/tracking

A majority of ocean engineering problems involve waves and/or other flows such as bores, hydraulic jumps, etc. which necessitates computing the topology of the free-surface. In reality, the free-surface marks a discontinuity between two media (say air and water) and thus acts as an interface. The numerical algorithms for computing the interfacial topology can be broadly classified into interface-tracking and interface-capturing techniques. In the former category of algorithms, the free-surface is modelled as a boundary and is tracked by updating the mesh as the solution progresses. In the latter category, the free-surface evolves spatio-temporally within a fixed domain wherein the interface is identified by an indicator function. Interface-capturing algorithms are obviously more advantageous (especially for violent flows involving complex interfacial deformation such as overturning and aeration) and thus have been extensively employed in NSE-based flow solvers; the same have been listed in Table 3.

As evidenced from Table 3, the interface-capturing algorithms can be further classified based on the technique used for interface identification ("reconstruction") and advection. The interface identification techniques differ based on the type of indicator function used (volume fraction or level-set function) as well as whether the identification itself is geometric in nature or not. The level-set method and high-resolution schemes such as the Compressive Interface Capturing Scheme for Arbitrary Meshes (CICSAM) are algebraic in nature in that they do not involve explicit geometrical computations of the placement (or advection) of the interface within the domain. In comparison, geometric methods such as the Piecewise Linear Interface Calculation-Volume Of Fluid (PLIC-VOF) and Moment Of Fluid (MOF) are higher fidelity in that the interfacial coordinates are geometrically computed subject to conservation of the primary phase volume in each cell.

Volume conservation is intrinsic for geometric VOF methods and also for single-phase meshfree methods such as the Improved Meshless Local Petrov-Galerkin method with 401 Rankine source function (IMLPG R). This is not the case for algebraic VOF schemes or the level-set method where additional numerical treatment is necessary to achieve volume conservation. This has been comprehensively demonstrated by the present authors (Saincher 404 and Sriram, 2022a) and others (Anghan et al., 2021; Arote et al., 2021) wherein a material redistribution algorithm originally developed for geometric VOF (Saincher and Banerjee,

406 2015) has been shown to dramatically improve the volume conservation properties of 407 algebraic VOF schemes.

Similarly, interfacial diffusion is intrinsic for algebraic VOF as well as level-set methods. This could be mitigated to some extent using operator-split/direction-split advection as doing so would eliminate multi-fluxing errors (Saincher and Sriram, 2022a). Whilst algebraic VOF techniques are indeed capable of capturing large-scale interfacial segregation in WSI problems (Saincher et al., 2023a), small-scale droplets and bubbles would still diffuse upon separation from the parent phase. This diffusion seldom contributes to the hydrodynamics in a WSI simulation and, in fact, provides numerical stability to the solution. Conversely, droplets/bubbles separating from the parent phase would never dissipate in geometric VOF and thus excessive interfacial fragmentation might, in fact, lead to solver instability.

417 Table 3. Various interface-capturing algorithms developed for NSE-solvers; cf. nomenclature 418 for the abbreviations.

Authors	Algorithm	Interface identification	Interface advection	Interface diffusion	Volume conservation	Mesh
O'Shea <i>et al.</i> (2014)	NFA	Geometric VOF	Unsplit Eulerian	Zero	Intrinsic	Cartesian
Sriram et al. (2014)	IMLPG R	MPNDAF	Lagrangian	Zero	Intrinsic	--
Saincher and Banerjee (2015)	Redistribution- based PLIC- VOF	Geometric PLIC-VOF	Operator- split Eulerian	Zero	Intrinsic	Cartesian
Bihs <i>et al.</i> (2016)	REEF3D	Level-set	Unsplit Eulerian	Intrinsic	Extrinsic	Cartesian
Zinjala and Banerjee (2016)	LEAS-MOF	Geometric MOF	Lagrangian- Eulerian	Zero	Intrinsic	General Polygonal
Zinjala and Banerjee (2017)	RMOF	Geometric MOF	Lagrangian- Eulerian	Zero	Intrinsic	General Polygonal
Anghan <i>et al.</i> (2021)	MSTACS	Algebraic VOF	Unsplit Eulerian	Intrinsic	Extrinsic	Cartesian
Arote <i>et al.</i> (2021)	SAISH	Algebraic VOF	Unsplit Eulerian	Intrinsic	Extrinsic	Cartesian
Saincher and Sriram (2022a)	OS-CICSAM	Algebraic VOF	Operator- split Eulerian	Intrinsic	Extrinsic	Cartesian

419

420 3.1.5 Turbulence Modelling

Ocean engineering problems involve flow of sea-water which has a kinematic viscosity of $v \sim 1e - 06$ m²/s. The corresponding Reynolds number Re = $v \cdot \mathcal{L}/v$ would typically be $O(10^6)$ even if the characteristic velocity (V) and length (*L*) are $O(1)$, that is, at model-scale. This is generally the case since the Froude-law is invoked for scaling based on the fact that gravity is the dominant restoring force in ocean engineering applications. As a consequence, most scenarios being simulated are not laminar and some form of modelling may be required to account for the additional viscous effects near the structure. Some of the typical applications necessitating turbulence modelling include:

- 429 Wave/tsunami interactions with vegetation: turbulence-induced viscous effects 430 arising from flow separation need to be accounted for to correctly estimate energy 431 attenuation.
- 432 Response of floating bodies: failing to account for viscous effects within the 433 boundary layer may result in over-prediction of the motion response.
- 434 Resistance of marine vessels in waves/calm water: failing to account for viscous 435 effects within the boundary layer may result in under-prediction of resistance.

436
437

Figure 4. An illustration of the different means to categorize various strategies to model 438 turbulence in depth-resolving models; cf. nomenclature for the abbreviations.

Several popular methods have been developed for modelling turbulence in NSE-based solvers; some have been integrated with self-developed codes by the present research group. There exist different means of classifying turbulence modelling strategies for depth-resolving methods; the same are depicted in Figure 4. In conjunction with Figure 4, the momentum equation (1) is also re-written to account for turbulence modelling:

$$
\frac{\partial \vec{V}}{\partial t} + (\vec{V} \cdot \vec{\nabla}) \vec{V} = -\frac{1}{\rho^*} \vec{\nabla} p' + \frac{1}{\rho^*} \vec{\nabla} \cdot \left((\mu^* + \mu_t) \vec{\nabla} \vec{V} \right) + \underbrace{\vec{g}}_{\text{gravity}}
$$
\n(2)

444 where, p' is the modified pressure which includes the normal components of the Reynolds or 445 Sub-Grid-Scale (SGS) stress tensor and μ_t is the turbulent viscosity; the terms 446 modified/introduced by turbulence modelling have been highlighted in bold. In context to 447 equation (2) and Figure 4, \vec{V} can be unfiltered, spatio-temporally filtered or time-averaged. 448 The filtering and time-averaging operations are essentially decompositions of the unfiltered 449 velocity and thus, once performed, information about the instantaneous velocity field is 450 invariably lost. For instance, the \vec{V} field obtained following solution to the RANS equations is 451 time-averaged and thus, (temporal) fluctuations in \vec{V} do not represent fluctuations in the 452 instantaneous field. Only the effect of the true fluctuating field on \vec{V} is modelled through the 453 eddy viscosity μ_t .

454

456 Figure 5. The vorticity field $(\vec{\nabla} \times \vec{V})$ generated by a moving cylinder interacting with a 457 focusing wave (Saincher and Sriram, 2022b); note the change in the nature of the solution 458 based on the definition of \vec{V} . The cylinder moves from bottom-right to top-left.

This important aspect is illustrated in Figure 5 wherein vortices shed by a moving cylinder interacting with focusing waves are shown (adapted from Saincher and Sriram (2022b)). The same problem has been simulated first using unfiltered NSE (a "laminar" solver) and then 462 using time-averaged NSE (a RANS solver based on standard $k - \varepsilon$ (SKE)). The aforementioned loss of information regarding the true fluctuating velocity field is readily

464 Figure 6. Time-histories of free-surface elevation $(\eta(t))$ and pressure $(p(t))$ corresponding 465 to a moving cylinder interacting with focusing waves (Saincher and Sriram (2022b)): (a,b) 466 $\eta(t)$ variation in-line with the center of the moving cylinder and $p(t)$ variation just below the 467 SWL at the (c,d) forward and (e,f) rear stagnation points.

apparent from Figure 5; the vorticity field is "instantaneous" in both cases. It should also be noted that the so-called "laminar" solver is a misnomer as it simply refers to solving the NSE without any turbulence modelling. In this regard, the laminar approach is not a Direct Numerical Simulation or DNS (since no attempt is made to resolve the Kolmogorov scales) but rather a form of Implicit Large Eddy Simulation or ILES (wherein the discretization 473 errors would mimic SGS modelling (Rodi *et al.*, 2013)). Having said that, figure 5 indicates that the laminar solver captures more "turbulence" than the actual turbulence model!

475 In addition to the above qualitative assessment, it is also important to quantify the impact of turbulence modelling (or lack thereof) on quantities of engineering importance. In order to do this, the time-variation of the free-surface elevation measured in the vicinity as well as pressure measured on the surface of the moving cylinder is reported in Figure 6 for both ILES and SKE simulations. It can be seen that ILES and SKE results are practically identical for 480 the lower towing speed. This is corroborated by the vorticity fields (for $U_{\text{cyl}} = 0.34 \text{ m/s}$) reported in Figure 5. For the higher towing speed, the SKE results show a closer agreement 482 with experiments in terms of both $\eta(t)$ variation as well as pressure at the rear stagnation point. However, the improvement gained from turbulence modelling in this case is not dramatic (even though the computed vorticity fields are dramatically different). The findings are in line with conclusions drawn from the ISOPE 2020 comparative study which was based 486 on the same experimental dataset (Agarwal et al., 2021a).

The necessity and nature of turbulence modelling depends on the nature of the problem itself and oftentimes the fidelity of the solution/simulation (against experiments) depends on the expertise of the CFD practitioner (this is later discussed at length in §7 on comparative numerical studies). One is not only required to assess the need of a turbulence model but also the impact of a particular model on the solution. Considering a wave-floating structure interaction problem as an example, a need for turbulence modelling may arise due to an over-prediction of the angular acceleration of the body by a laminar model. If RANS-based turbulence modelling is introduced to supplement the viscous damping in the near-field of the body, the same may also negatively impact the simulation through unwanted damping of the incident waves. In such a case, the unwanted damping could be mitigated by:

-
- 497 Stabilizing the unbounded growth of μ_t using limiters (Larsen and Fuhrman, 2018).
-
-
- Increasing advection using higher-order upwind schemes (Saincher and Sriram, 2022b).
	-
- Increasing advection using conservative NSE formulations (Saincher and Sriram, 2023).
- Switching to a less empirical model such as WALE (zero-equation model with a 503 single model constant) for computing μ_t (Rodi *et al.*, 2013).

The above discussion indicates that there exist multiple solutions to a given problem and there is a general consensus that the simplest models also prove to be the most robust. Taking into account the strongly empirical nature of turbulence modelling in general (RANS in particular), a modestly accurate albeit robust model applicable to several problems should be preferred over a heavily calibrated model that works perfectly albeit only for a single

problem. Further, turbulence model should only be employed for the practical problems in need and not for all scenarios.

-
- 3.1.6 Numerical Methods

In addition to the algorithms used for pressure-velocity coupling, interface capturing and turbulence modelling, the flow solver is also comprised of spatio-temporal discretization schemes as well as linear equation systems solvers. Both categories of algorithms directly

impact the accuracy and stability of the flow solver.

Some of the popular discretization schemes that have been widely implemented for ocean engineering problems are now discussed in context to the momentum equation (1) and Table 4. Discretization of the time-term can be carried out either using Linear Multi-step Methods (LMMs) or Runge-Kutta (RK) methods. These two categories of methods can be further classified into explicit and implicit schemes. Explicit LMMs are also known as the Adams-Bashforth Methods (ABMs) whilst implicit LMMs are known as Adams-Moulton Methods (AMMs). As the name suggests, LMMs build accuracy by storing the flow solution across multiple time-levels such that a first-order LMM would require an existing flow-field solution from one time-level, a second-order LMM would necessitate solutions from two 526 time-levels and so on. Owing to the requirement of an existing flow-field solution, LMMs > $O(1)$ are not "self-starting" and some complexities exist in implementing these methods for variable time-steps. Moreover, the region of stability of LMMs shrinks with increasing order of accuracy (Drikakis and Rider, 2005). Nonetheless, a key advantage of LMMs is that the per-time-step computation effort does not increase with increasing order of accuracy; only the storage requirements increase.

On the other hand, RK methods divide a single time-step into a number of intermediate steps with all intermediate velocity fields made divergence-free; only the most recently known velocity field is necessary for a given intermediate step. Given this characteristic, RK methods are self-starting and automatically account for variable time-steps. However, the fact that the predictor-corrector loop (cf. Figure 3) is executed multiple times within a time-step introduces a unique set of merits and shortcomings. The chief merit is the numerical stability which, unlike LMMs, increases with increasing order of the method. Another merit over LMMs is that storage requirements do not increase with increasing order. The chief shortcoming associated with RK methods is that each intermediate step entails a computationally expensive solution of the elliptic PPE or EOPC; per-time-step computation effort thus increases with increasing order. Referring to Table 4, it is seen that a number of NSE algorithms implement explicit time-integration (ABM or Total Variation Diminishing-RK (TVD-RK)) which is suitable given the hyperbolic nature of wave propagation. In cases where a greater amount of numerical stability is desired, say conservative NSE formulations 546 for violent WSI (Benoit et al., 2023) authors opt for AMM rather than TVD-RK. This is probably because the additional linear equation systems encountered for AMM (one system 548 for each component of \vec{V}) is parabolic and less expensive to solve than the elliptic PPE/EOPC encountered multiple times within a time-step in the case of TVD-RK. It is also possible that very high-order AMMs might lead to dispersive (phase) errors in wave-propagation.

551 Table 4. A summary of the various discretization methods and linear equation system solvers 552 implemented for NSE algorithms applied to ocean engineering problems reported in the

Authors	Time	Momentum Advection		Pressure	Diffusion	PPE/EOPC		
		Scheme	Treatment					
Sriram et al. (2014)	ABM1	Lagrangian		SFDI	SFDI	GMRES		
Bihs <i>et al.</i> (2016)	TVD-	WENO	Non-	x	x	BiCGStab		
	RK3		conservative					
Xie and Stoesser	AMM1	Second-order	Conservative	CD2	CD2	ADI /		
(2020)		TVD				BiCGStab		
Agarwal et al. (2021b)	ABM1	Lagrangian		SFDI	SFDI	BiCGStab		
Anghan et al. (2022)	ABM2	Blended FOU-	Non-	CD2	CD4	GSSOR		
		FiOU	conservative					
Sriram Saincher and	ABM1	Blended FOU-	Non-	CD2	CD2	GSSOR		
(2022b)		FiOU	conservative					
Benoit et al. (2023)	AMM1	Slope-limited	Conservative	CD2 with third-order		GMRES		
		SOU			numerical smoothing			
Saincher Sriram and	ABM1	Blended FOU-	Conservative	CD2	CD2	GSSOR		
(2023)		FiOU						

553 literature; cf. nomenclature for abbreviations. (\boldsymbol{X} : data unavailable)

554

In addition to time-integration, the numerical schemes chosen for momentum advection, pressure and diffusion terms as well as the linear systems solver chosen for solving the pressure field also play a key role in deciding the robustness and accuracy of NSE solvers. In context to the discretization of the pressure and diffusion terms, second-order central differencing (CD2) suffices for most scenarios and is thus the most widely used (cf. Table 4). However, recent studies involving DNS of marine outfalls have instead implemented fourth-order central differencing (CD4) for higher resolution treatment of the diffusion term (cf. 562 Anghan et al., 2022). It should also be noted that CD4 treatment of the pressure gradient does not dramatically improve the accuracy of a solver and should rather be avoided to save computational effort (Tafti, 1996).

Numerical formulations of the NSE inherently contain some form of numerical diffusion. In context to ocean engineering applications, this diffusion gets manifested as a gradual reduction in wave-height (Saincher and Banerjee, 2017). Whilst the numerical diffusion can be arrested through mesh refinement, a more computationally efficient way to do this (especially for mesh-based Eulerian solvers) is by increasing the order of advection discretization. However, computational efficiency does not translate to a straightforward implementation, especially for multiphase solvers. Implementation of a high order advection scheme in its "pure form" leads to severe dispersion errors in regions of sharp velocity gradients which, in case of waves, prevail at the air-water interface; the consequence is unphysical deformation of the generated waves. This can be corrected by either using 575 inherently bounded schemes such as WENO (Bihs *et al.*, 2016) or blended schemes where (say) only 50% of the advected momentum is estimated using the high-order scheme, the rest being estimated using FOU (Saincher and Sriram, 2022b). For more violent scenarios involving wave-breaking and/or wave-slamming, a higher order treatment of advection may not be sufficient and rather the correct amount of advection being attributed to each fluidphase needs to be ensured. This is where conservative NSE formulations come into picture 581 wherein ρ^* is shifted to the left-hand-side of equation (1) with the time and advection terms. It has been recently demonstrated by the authors that conservative NSE solvers are necessary for correctly capturing the topology of waves overturning over a long distance; such as solitary waves breaking over a beach/shallow water (Saincher and Sriram, 2023). It is worth mentioning that conservative NSE formulations strongly and consistently couple mass and momentum transport (cf. the discussion on mass inconsistency in Saincher and Sriram (2023)) and thus momentum advection is more strongly governed by material transport (owing to the 1: 800 density ratio between air and water) rather than the momentum 589 advection scheme itself (Bussmann et al., 2002). This makes conservative NSE a suitable alternative to high-order advection schemes for arresting wave-damping in non-violent/moderately violent WSI scenarios.

For rigid structures, one can incorporate this in the computational domain and solve the interaction problems as shown in Figure 5. For elastic and floating structures, a separate equation of motion will be solved to understand the fluid-structure interaction process, see, 595 Sriram and Ma (2012), Rijas et al. (2019) and Vineesh and Sriram (2022). In the case of modelling porous/vegetation structure interactions with waves, one can adopt microscopic or macroscopic approaches. The macroscopic approach is commonly adopted due to the computational advantages as well as in terms of requirement for physical process (see, Divya and Sriram (2020)). For modelling the porous/vegetation structure interaction with waves, additional resistance terms such as the: (a) linear drag coefficient representing the laminar flow, (b) non-linear drag coefficient representing the turbulent flow, (c) coefficient for the transitional flow and (d) virtual mass coefficient for inertia terms were incorporated in the governing equations. The numerical studies on the wave porous structure can be carried out in two different ways:

(i) Coupling of pure fluid and porous flow equations, in which the fluid flow is solved using the NSE and porous flow with different porous flow model, following which the interface was coupled by matching the flow properties. Such coupling can be explicit, implicit or iterative in nature.

(ii) Based on unified or single governing equations to model both porous structure and fluid flow. In the microscopic approach, the aim is to capture detailed flow physics, directly resolved by the NSE (see Xie and Stoesser (2023)).

Apart from mesh based approach in solving the NS, mesh-free or particle methods are quite popular and further developments are actively being carried out. These developments have 614 been the topic of many recent review papers such as Luo *et al.* (2021), Sriram and Ma (2021), Lind et al. (2020) and references therein. However, the acceptability of the mesh-free methods or particle methods for industry and practical applications in the projects are not matured compared to mesh-based methods. The consolidation of the work carried out by the authors with regards to the mesh-free method based on Meshless Local Petrov Galerkin Method (MLPG) has been reviewed in detail in Sriram and Ma (2021) and shall not be repeated here for the sake of brevity. Further, an important relation between the widely

- popular Smoothed Particle Hydrodynamics (SPH), Moving Particle Semi-Implicit Method
- (MPS) and MLPG was established. However, as this special issue concerns the Newton

fellowships, a flow chart of development has been reproduced for completeness as shown in

Figure 7.

- Figure 7. Summary of the history of the development of the Meshless Local Petrov Galerkin
- method (MLPG) and its application in Ocean Engineering (revised and updated from Sriram
- and Ma, 2021). Grey shaded boxes are development with partial or full support from the
- Newton fellowship.

3.2 Potential Flow Theory

The fully nonlinear potential flow theory (FNPT) has significantly matured in today's context and is being extensively used by both researchers as well as industry. The methodology was pioneered by Longuet-Higgins and Cokelet (1976) using a mixed Eulerian and Lagrangian approach. The simulation of nonlinear waves using FNPT can be carried out either by fully discretising the domain and then solving the Laplace equation using numerical approaches (like FEM, BEM and so on) or by obtaining the solution of the Laplace equations using spectral, Eigen function or Fourier methods. In the former case, the computational effort would be quite significant when one extends the method to 3D, however, the advantage is that one can simulate waves interacting with any arbitrarily complex structure. In the latter approach, the computational effort is lesser in comparison and such methods are largely employed for simulating the fully nonlinear waves. Dommermuth and Yue (1987) and West et al. (1987) proposed an attractive fast convergence, high accuracy and fast resolution properties-based higher order spectral (HOS) method. These fast methods of computation are very useful for calculating the long-time evolution of nonlinear waves and can be used as an input for the numerical models based on the NS equations. A detailed review of these models can be found in Kim et al. (1999) and in Ma (2008) and references therein. Normally, the FNPT-based models are quite effective in reproducing the extreme steep non-breaking waves, however, once the wave begins to overturn (the crest becomes vertical), the simulation crashes (Mohanlal, 2023). For some models, the crash may be delayed up to the point when 651 the overturning crest hits the free-surface (Grilli *et al.*, 2001). Naturally, the conventional FNPT models cannot handle wave-trains in which multiple breaking events occur in succession (over a period of time). In order to overcome these effects and carry out the simulations of overturning waves for a longer duration, researchers employ empirical 655 treatment such as eddy viscosity models to incorporate breaking effects (Tian et al. (2010), 656 Barthelemy et al. (2018), Sieffert and Ducrozet (2018), Hasan et al. (2019)). Very recently, Mohanlal (2023) has developed a FNPT model to handle multiple depth-limited two-dimensional breaking events in irregular sea-states, steepness-limited breaking of two-dimensional focusing waves as well as depth-limited breaking of regular waves over three-dimensional bathymetry. In their model, incipient breaking is detected based on whether the ratio of the orbital velocity to the wave-celerity exceeds a given threshold. Following detection, the energy of the over-turning wave is dissipated through an absorbing/damping surface pressure term introduced into the dynamic free-surface boundary condition (Grilli and Horrillo, 1997). It is worth mentioning that the wave generation techniques discussed in Table 2 in context to the NS models can also been incorporated in the FNPT models, mostly using moving wall, relaxation zone and/or prescribing inlet wave characteristics.

4 Depth-averaged Mathematical Models

The depth-averaged models are governed by the Boussinesq Equations (BSNQE), the Green-Naghdi equations, the Korteweg-De Vries (KdV) equation or the Shallow-Water Equations (SWE). These models are based on an assumption that the horizontal velocities (in the shoreward and longshore directions) are uniform or varying over the water column. A summary of the depth-averaged mathematical models is provided in Table 5; it can be appreciated that the models are being actively developed since the 1960s. In some of the models, the approach is based on the notion of a uniform horizontal velocity, which is a characteristic of long-wave-induced orbital kinematics, the baseline models are limited to shallow-water. It is evidenced from Table 5 that the research effort has been driven by the 678 need to expand the applicability of these models to deep-water $(kd \geq \pi)$. This was achieved through various means such as:

- 680 replacing the depth-averaged velocity by velocity defined at an arbitrary depth to act 681 as the velocity variable (Nwogu, 1993),
- 682 improving the dispersion characteristics through modification of the governing 683 equations (Beji and Nadaoka, 1996) and
- 684 piecewise integration of the momentum equations over multiple layers yielding 685 separate velocity profiles within each layer (Lynett and Liu, 2004a).
- 686 Table 5. A summary of various depth-averaged mathematical models developed for wave-687 propagation reported in the literature; cf. nomenclature for abbreviations.

688 These models have been employed to address larger-scale spatial $(\sim km)$ and temporal (~min) processes in ocean engineering such as ship-generated waves in bays and inland waterways (cf. Agarwal et al., 2022a). The depth-averaged models are widely used in the industry for waves and current hydrodynamics with models capable of handling interactions between waves and sheared currents being recently proposed by Yang and Liu (2020). Whilst the BSNQE-based models are based on irrotational and inviscid assumptions, turbulence and 694 wave breaking have also been treated using the empirical approaches (Lynett *et al.*, 2002; Shi et al., 2012). The BSNQE-based models have also been used for coupling with NS 696 equations-based models to minimize the computational time (cf. Agarwal *et al.*, 2022b and references therein). It is worth mentioning that the different approaches of wave-generation/absorption discussed in Table 2 in context to the NS models are also applicable to depth-averaged models except the moving wall approach. It is also worth noting that the present review only aims at providing a brief overview of the development of BSNQE-type models for the sake of completeness in context to multi-scale modelling and is by no means all-inclusive. The reader is referred to Brocchini (2013) for a detailed review.

703

704 5 Regional and global-scale modelling in Ocean Sciences

If one refers back to the spatio-temporal scale classification of physical processes and models 706 in Figure 1, it is seen that the BSNQE-type depth-averaged models belong in the \sim 1 km/1 h category. Whilst this is considerably "large-scale" compared to the depth-resolved models $({\sim}1 \text{ m}/1 \text{ min})$, the depth-averaged models are also considerably "small-scale" to regional 709 and global-scale models $({\sim}10^3 \text{ km}/1 \text{ ka})$ that are typically applied in ocean sciences. Both regional and global-scale simulations in ocean sciences have been traditionally and fundamentally based on the concept of multi-scale modelling. Having said that, popular ocean-science models are generally based on RANS-type momentum equations; multi-scale modelling is seldom achieved through the inclusion of potential theory or Laplace equations. This is primarily because such models belong to the class of Ocean General Circulation Models (OGCMs) wherein the assumptions of zero viscosity and vorticity need to be relaxed. A summarization of the state of the art in the development of regional and global-scale ocean models is provided in Table 6.

718

719 Table 6. A summary of various regional and global ocean models developed for various 720 ocean science applications reported in the literature; cf. nomenclature for abbreviations.

721

722 Referring to Table 6, the popular ocean models include POM, FVCOM, ROMS, HYCOM

723 and MOM6. The horizontal variable arrangement in these models is based on various classes

724 of the finite-difference Arakawa grids, namely:

- 725 A-grid in which scalars (eg. temperature and salinity) and vectors (eg. velocity) are 726 defined at the same point,
- 727 B-grid in which scalars are staggered from the velocity by half a grid dimension, 728 however both velocity components are defined at the same point and
- 729 C-grid in which both velocity components as well as scalars are staggered by half a 730 grid dimension from one another.
- 731 In the vertical direction, the following coordinate system(s) are implemented:
- 732 *z*-coordinates which follow the vertical direction and are suitable for resolving free-733 surface flow features,
- 734 \bullet σ -coordinates which follow the bottom topography/bathymetry and are suitable for 735 resolving the bottom boundary layer and
- 736 isopycnal or isopycnic coordinates which follow the density contours and are suitable 737 for resolving tracer (temperature, salinity etc.) transport in the open ocean.

738 As evidenced from Table 6, most ocean models employ some combination of the above 739 coordinates which accords the capability for multi-scale simulations. A hybridization of coordinates is necessary because free-surface features, topographical features and density stratification all occur at different vertical scales and moreover the individual scales vary as 742 one move from the open ocean to coastal regions (Chassignet et al., 2007). Another important aspect which differentiates OGCMs from FNPT and BSNQE approached is the need for modelling the feedback from the subgrid-scales to the inertial-scales in terms of both momentum as well as scalar transport (turbulent diffusion). Referring to Table 6, this is achieved through various turbulence closure models. It is also interesting to note that, because the horizontal scale in such problems is significantly greater than the vertical scale (cf. select domain sizes in Table 6), the horizontal and vertical directions employ different turbulence closures. Typically, the Smagorinsky model (zero-equation turbulence model 750 wherein the evaluation of μ_t is conceptually similar to Prandtl's mixing length model (Rodi *et al.*, 2013)) can be applied in the horizontal direction whilst a more comprehensive two-equation model (the MY-2.5 model is a popular choice; cf. Mellor and Yamada (1982) and 753 Chen et al., (2003) for details) can be applied in the vertical direction. This unique numerical constitution makes ocean models ideally suited for multi-scale modelling across a wide range of applications which is evidenced from Table 6. It is worth mentioning that the present review only aims at providing a brief overview of the development of OGCMs for the sake of completeness in context to multi-scale modelling and is by no means comprehensive. The reader should refer to the literature listed in Table 6 for further details.

6 Coupled models

In the previous sections, we presented several different models that are available to treat the problem at hand. However, rather than different models, it would be ideal to have one particular model to handle a wide range of problems spanning various spatio-temporal scales. One way of achieving this, as discussed in context to ocean sciences in §5, is to employ hybrid coordinate systems. Another way of achieving this is coupling different modelling tools that are developed over the period of years leading to multi-scale modelling in ocean engineering. Such coupled models are discussed in the following subsections.

6.1 Overview of coupling/decomposition strategies

With regards to coupling models, there are two approaches; one is domain decomposition and the other functional decomposition. The domain decomposition (DD) strategy divides the computational domain into parts and applies different mathematical models in each part. This is ideally done to avoid computationally expensive (and energy dissipative) NS simulations in the entire domain. Thus, multi-scale modelling is achieved by gaining the ability to model larger computational domains than would normally be allowed for a pure NS model. The functional decomposition (FD) strategy was pioneered by Dommermuth (1993) who simulated the formation of striations and scars on a free-surface due to impingement by a pair of vortex tubes shed from the tips of a submerged delta wing. In the case of FD, rather than physically decomposing the domain, the instantaneous velocity and pressure fields are decomposed into irrotational and vortical components. Another key characteristic of FD is the prescription of a constraint that the normal component of the vortical velocity is zero at 782 the free-surface. This yields a transport equation for the free-surface elevation η in terms η 783 and the velocity potential ϕ which accords an FNPT-like framework for the solution of 784 $\eta(x, y, t)$. Such a numerical treatment means that FD yields "natural and exact transition" from viscous vortical flow to inviscid vortical flow to potential flow (Dommermuth, 1993); this is not possible in a conventional primitive-variable formulation of the NSE.

6.2 Navier-Stokes coupled with potential-flow models (Domain Decomposition)

In most DD-based problems, the computational domain is decomposed into a viscous inner sub-domain and a potential outer sub-domain. The information (velocity, pressure and surface elevation) will be transferred through either relaxation zones or a sharp interface. Also, based on how the information between the solvers is being transferred, it can be either one-way coupling (weak coupling) or two-way coupling (strong coupling). In one way coupling, information is transferred only from the potential solver into the viscous solver, but in two-way coupling, the information is transferred in both ways, from depth-averaged or depth-resolving irrotational models to full NS and vice-versa. The two-way coupling is advantageous since it allows for a significantly smaller computational region for the viscous solver. However, it necessitates an iterative process or an implicit approach between the two 799 models on a shared interface, which might increase the computational costs (Sriram et al., 2014). The advantage of one-way coupling is that no such iterations are needed, but it needs a longer viscous domain to avoid the reflection from outer boundaries. This method is suitable, wherein, one needs to analyse the kinematics of the breaking waves in deep water or depth-induced breaking in the shallow water region (Saincher and Sriram, 2023).

An early implementation of the concept of weakly-coupled hybrid modelling can be seen in 805 the work of Fujima et al. (2002). They spatially nested a three-dimensional Navier-Stokes model within a two-dimensional nonlinear long-wave model to simulate tsunami-breakwater interaction at 1: 200 scale. An obvious shortcoming of the model was that spanwise vortices generated in the depth-resolving model that could not be transferred back to the depth-averaged model due to the latter's reduced dimensionality. Grilli and co-workers (1999, 2003, 2004) coupled the 2D HOBEM-FNPT with NS model based on SL-VOF. Extension of the FEM code with the NS model has also been carried out by Clauss and co-workers (2004, 2005). For the NS model they have tested with the commercial softwares such as FLUENT, CFX and COMET. They tested their coupling approach by studying the deep water wave breaking (breaking of freak waves) and comparing with experimental measurements. Yan and Ma (2009) coupled the QALE-FEM with the commercial software STAR-CD to study the 816 wind effects on breaking waves. Hildebrandt et al. (2013) coupled the FEM with the commercial software ANSYS to model the wave impacts with tripod structure. 818 Narayanaswamy et al. (2010) and Kassiotis et al. (2011) used one way coupling of the Boussinesq model with the SPH method for solitary wave simulations. Without feedback from the SPH to the Boussinesq model, a fixed overlapping zone was considered to transfer 821 the information. Recently, this was improved by Agarwal et al., (2022b) by coupling a Boussinesq model with the MLPG (Meshless Local Petrov Galerkin) method.

However, if one needs to analyse the wave structure interactions in the presence of floating bodies or fixed structure, then strong coupling of the two models is required, wherein the radiated waves will propagate from NS model to depth-averaged or depth-resolved irrotational models. In strong coupling, the computational domain is divided into two parts, in one part the generation and propagation of waves is being considered and in the other part

structure/breaking region will be present. The modelling of first part of the domain will be carried out using depth-averaged or depth-resolved irrotational models and then the boundary 830 conditions (velocity and pressure) are fed into the NS model at the same time steps, to study the remaining part of the domain. Then the velocity from NS model is again feed back to the depth-averaged or depth-resolved irrotational model domain for the next time step. Thus, in general, the strong coupling needs to couple the models both in space and time domains. For the coupling in space domain, the following four methods have been found to be employed, 835 as pointed out by Sriram *et al.* (2014): (a) fixed boundary interface, (b) moving boundary interface, (c) fixed overlapping zone and (d) moving overlapping zone.

One of the pioneering works in this regard was carried out by Grilli and co-workers (2005, 844 2010) wherein, they extended the model from weak coupling to strong coupling for studying the 3D breaking waves by coupling 3D HOBEM-VOF. Later, Grilli and co-workers (2007, 2008, 2009) coupled the NWT and NS based on Large Eddy Simulation (LES) to study the 847 forced sediment transport simulations. Later studies from Greco (2001), Colicchio et al. 848 (2006), Greco et al. (2007) and Sitanggang and Lynett (2010) further established the

849 feasibility of the DD strategy. Following these studies, a ground-breaking contribution was 850 made by Sriram et al., (2014) wherein the full capability of coupled DD modelling was 851 explored in detail.

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854 Figure 9. Simulation of directional regular waves (aligned at 30° to the x-axis) interacting 855 with a fixed cylinder using weakly coupled FEBOUSS and MLPG R using 3D cylindrical 856 coupling interfaces (Agarwal et al., 2022b).

858 859 Figure 10. Simulation of regular waves interacting with a moored floating spar using a

860 coupled model employing HOS-NWT, foamStar and MoorDyn (Aliyar et al., 2022).

861 However, until now mostly these strong coupling are realised only in the 2D problems, and, 862 to the best of the authors' knowledge, the strong coupling in 3D is yet to be attempted. 863 Typical examples of simulations performed based on one-way coupling using the codes

developed by the authors and their co-workers: IITM-RANS3D, HOS-NWT-foamstar (using depth-resolved potential and viscous models) and FEBOUSS-MLPG (using depth-averaged potential and viscous models) are reported in Figures 8-10.

One more popular hybrid model is qaleFOAM, which has been developed based on the experience from QALE-FEM. This model adopts the domain decomposition approach, which combines a two-phase Navier–Stokes (NS) model with a model based on the fully nonlinear potential theory (FNPT). In a region around the structures and/or the breaking waves (NS domain), the open-source NS solver OpenFOAM/interDyMFoam is applied. In the rest of the computational domain (FNPT domain), the FNPT-based quasi arbitrary Lagrangian–Eulerian finite element method (QALE-FEM) is adopted. The qaleFOAM was originally developed 874 for modelling the turbulent flow near offshore structures subjected to extreme waves (Li et $al., 2018$). It has now been extended and applied to model a wide range of wave-structure 876 interaction problems, such as the wave resistance (e.g. Gong *et al.*, 2020), violent wave 877 impact on sea walls (Li et al., 2023), survivability and performance of floating wind turbines 878 (Yu et al., 2023; Yuan et al., 2023) and wave energy converters (Yan et al., 2020) as well as 879 wave-driven drift of floating objects (Xiao *et al.*, 2024). Recently blind tests and numerical comparative studies have confirmed its superiority over single-model methods including the potential theory and the NS solvers. The details will be discussed below. However, one of the theoretical issues in these DD coupling is that the researchers coupled irrotational flow model with the rotational flow models. Particularly for strong coupling, there is a mathematical discontinuity in the velocity field and they overcome this with numerical approaches (Sriram and Ma, 2021). This needs to be overcome in the future modelling efforts, see Yang and Liu (2022) for the development of the multi-layer model based on rotational flow.

6.3 Navier-Stokes coupled with potential-flow models (Functional Decomposition)

As described earlier in §6.1, the fundamental concept for the functional decomposition (FD) is to use the Helmholtz decomposition to separate the velocity field into the rotational and irrotational parts to investigate the free surface flow (Dommermuth, 1993). The FD approach has also been adopted to simulate ocean engineering scenarios. In context to WSI, there are two categories under this decomposition, based on whether the structure is considered in the potential solver or not.

6.3.1 First category: structure handled by both potential and viscous solvers

In the first category, the WSI problem is split into a potential component and a viscous part. The complete problem is initially solved by a potential solver, and then rectified by adding 898 the viscous correction (Kim et al., 2005; Edmund et al., 2013; Rosemurgy et al., 2016; Robaux and Benoit, 2021). One drawback of this strategy is that the potential solver must first solve the entire problem before applying the viscosity correction. As a result, challenges such as higher-order waves, stability issues in the steep waves and breaking induced by presence of structure with complex interactions are still constraints in this classification. Recently, Robaux (2020) published a thorough description of nonlinear waves' interactions with a horizontal cylinder with a rectangular cross section employing potential solver, CFD solver, and HPC-OpenFOAM coupled DD and FD based solvers. In comparison to the full CFD simulation, both coupling approaches, in particular the FD-based approach, need a

minimal amount of computational time while providing an accurate representation of the loads and associated hydrodynamic coefficients.

909 6.3.2 Second category: structure only handled by the viscous solver

In the second category, the total unknown is decomposed into the incident part and the complementary part. Only the incident flow is modelled in the incident part (wave only), leaving all the interaction with structure calculated by the viscous solver as the complementary part. The common name among researchers for this classification is 914 SWENSE (Spectral Wave Explicit Navier Stokes Equations), proposed by (Ferrant et al., 915 2002) and actively developed by (Gentaz, 2004; Li et al., 2018; Kim, 2021). The NS equation modified into the SWENSE is solved to yield the complementary fields. The advantage of this method is that the wave models directly provide incident wave solutions, minimising the problem's complexity and cost. For a detailed derivation of single-phase and two-phase 919 SWENSE, refer to Luquet *et al.* (2007) and Li *et al.* (2018) respectively. The applications in 920 single-phase SWENSE over the years can be read in (Luquet *et al.*, 2007; Monroy *et al.*, 921 2010). Recently, the two-phase SWENSE method (Li *et al.*, 2021) has been implemented on 922 top of foamStar and is called as foamStarSWENSE, and the only difference is that in this 923 solver, the NS equations in *foamStar* are replaced by SWENSE. Recent developments of *foamStarSWENSE* such as efficient regular and irregular wave generation in the solver and higher-order forces estimation on a vertical cylinder, buoy and floating spar can be referred to 926 in Choi (2019), Kim (2021), Li et al. (2018) as well as in Aliyar et al. (2022).

6.4 Navier-Stokes coupled with geophysical fluid dynamics / ocean-science models

Recently, there has been a research effort to implement the domain decomposition (DD) strategy to (strongly) couple geophysical fluid dynamics (GFD) / ocean-sciences models with Navier-Stokes solvers. A couple of such hybrid GFD-CFD models have been listed in Table 7 wherein FVCOM (cf. Table 6 for details) has been coupled to either the overset mesh-based single-phase NS solver SIFOM or the unstructured mesh-based two-phase NS solver SIFUM. It is worth mentioning that in addition to the NS solver being based on overset meshes (SIFOM), the overset grids have also been employed to nest the SIFOM/SIFUM domain within the FVCOM domain. Referring to the domain sizes in Table 7, it should be noted that in some cases, the SIFUM domain is not necessarily entirely nested within the FVCOM 938 domain along the vertical (z) direction. This is attributable to the ability to model the air-phase within the SIFUM framework which is necessary for violent WSI scenarios.

Such hybrid GFD-CFD modelling provides the ability to perform multi-scale environmental, geological as well as FSI/WSI simulations over domains spanning several hundred or even several thousands of square kilometres. However, it is worth mentioning that some of the problems listed in Table 7 are comparatively "small-scale" and can indeed be tackled by more conventional FNPT-RANS models. For instance, Saincher and Sriram (2022b) have 945 applied IITM-RANS3D to a $0.045 \times 0.0022 \times 0.002$ km³ domain to simulate the interaction between focusing waves and a moving cylinder. In another study Saincher et al. 947 (2021) had applied IITM-RANS3D to a $0.3 \times 0.005 \times 0.007$ km³ domain to study the run-up characteristics of violently breaking long and high waves that could be generated by an 949 extreme coastal event. Thus, the lower-limit of applicability of the hybrid GFD-CFD models

- 950 can also be tackled by hybrid FNPT-RANS models.
- 951

952 Table 7. A brief overview of geophysical fluid dynamics / ocean science models recently

953 hybridized with Navier-Stokes equations models reported in the literature; cf. nomenclature

954 for abbreviations. $(X: data \text{ unavailable})$

	Models			Maximum domain extents for each model			
Authors	Small- Large-		Large $(km3)$	Small $(km3)$	Coupling	Application	
	scale	scale	$(x \times y \times z)$	$(x \times y \times z)$	strategy		
	FVCOM (weakly) 3D)		$1.8 \times 0.6 \times 0.1$	$0.4 \times 0.4 \times 0.1$		Flow over flat plate	
		SIFOM	$3.5 \times 0.4 \times 0.15$	\sim 2.0 \times 0.3 \times 0.1		Transient sill flow	
Tang et al.		(fully)	$3 \times 0.3 \times 0.009$	$\sim 0.05 \times 0.02 \times 0.009$	DD	Bridge pier (lab-scale)	
(2014)		3D, single-	$3 \times 0.6 \times X$	$\sim 0.8 \times 0.4 \times \mathbf{X}$	(strong) coupling)	Thermal effluent	
		phase)	\sim 70 \times 140 \times 0.013	x		Bridge pier (river-scale)	
			\sim 170 \times 170 \times 0.05	$\sim 0.7 \times 0.7 \times 0.05$		Flow past seamount	
	FVCOM (weakly 3D)	SIFOM	$3 \times 0.6 \times 0.012$	$0.075 \times 0.03 \times 0.012$		Thermal effluent	
Qu et al.			(fully	\sim 20 \times 15 \times 0.004	x	DD	Lagrangian tracking of
(2016)		3D,			(strong)	estuary flows	
		single-	\sim 250 \times 1 \times 0.01	\boldsymbol{x}	coupling)	Storm surge impact on	
		phase)				river bridge pier	
		SIFUM		\sim 2 \times 0.1 \times 0.05	DD (strong) coupling)	Tsunami wave runup	
Qu et al.	FVCOM	(fully	$14 \times 0.1 \times 0.2$			Tsunami wave	
(2019a)	(weakly)	3D.				impacting coastal	
	3D)	two- phase)				highway bridge	
			$3.5 \times 0.4 \times 0.15$	\sim 1 \times 0.3 \times 0.225		Transient sill flow	
			$0.2 \times 0.2 \times 0.01$	$~1.1 \times 0.12 \times 0.01$		3D dam-break flow	
	FVCOM		SIFUM				Long-wave
Qu et al.		(fully	$0.04 \times 0.001 \times 0.0004$	$0.004 \times 0.001 \times 0.0005$	DD (strong)	impingement on a	
(2019b)	(weakly	3D.				vertical cylinder	
	3D)	two-	$0.04 \times 0.03 \times 0.002$	$0.015 \times 0.03 \times 0.002$	coupling)	Hydraulic jump	
		phase)				Coastal flood impacting	
			$\sim 60 \times 60 \times 0.025$	$\sim 0.1 \times 0.1 \times X$		beachfront house	

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956 7 Benchmarking the Numerical Models through Comparative Studies

In the past researchers developed numerical models and validated with their own experimental simulations, the data sharing and comparison between different numerical models, its accuracy and performance in terms of computational efficiency are not attempted. In the field of ocean engineering, when the concept of numerical wave tank was developed inline with the numerical wind tunnels that are quite popular in those times, Clément (1999) and Tanizawa and Clément (2000) carried out such exercise for fully nonlinear potential flow 963 theory. Recently, major initiatives were undertaken by Ransley et al. (2019, 2020), Sriram et 964 al. (2021), Agarwal et al. (2021a) and Saincher et al. (2023a). These studies highlighted some of the commonly adopted guidelines by the researchers pertaining to WSI simulations:

966 The fidelity of regular/focusing wave generation deteriorates away from the 967 wavemaker irrespective of the nature of the numerical model and no single wave-968 generation method may be regarded as superior over others. Far from the wavemaker,

969 the models generally deviate by $5 - 10\%$ in terms of primary energy content which is acceptable. However, the deviation across models may be as high as 50% in terms of the sub- and super-harmonic wave components.

- The performance of a solver should be judged based on the peak values of the surface-elevation/hydrodynamic pressures/loads as well as the phase agreement captured by the model. Phase disagreement is acceptable for WSI scenarios involving regular waves (Saincher et al., 2023a) as long as the phase-shift remains constant over several wave cycles. However, for WSI scenarios involving transient waves such as focused and/or overturning waves, the phase agreement is critical as it determines the shape of the impacting wave as well as the time-varying load profile 979 on the structure (Sriram *et al.*, 2021; Agrawal *et al.*, 2021a).
- The inclusion of turbulence modeling does not necessarily improve the accuracy of a simulation. This rather depends on the problem at hand. Further, for the same problem, different turbulence models may lead to the same/similar results. The expertise of the user should also be factored-in whilst using turbulence models, especially RANS-based models which are strongly empirical. These statements are substantiated through Figure 11 wherein results from the ISOPE-2022 comparative study on WSI are reported. In this study, one of the participating institutions had employed STAR-CCM+ for the simulations and had assessed the effect of different turbulence modeling strategies (implicit LES ("laminar"), RANS and explicit LES) on the solution. It can be observed from Figure 11 that changing the modeling strategy hardly affects the pressure time-history or the value of maximum impact pressure across multiple loading cycles. This could be interpreted from two perspectives: (a) the physics of the problem under consideration is independent of turbulence modeling or (b) the employed numerical framework is independent of turbulence modeling. The first statement is (obviously) incorrect. The second interpretation, however, holds merit from the standpoint of the relation between LES 996 and RANS (Rodi *et al.*, 2013). It is worth mentioning that the STAR-CCM+ 997 simulations were two-dimensional and a rather coarse resolution of $L/300$ and $H/50$ 998 (where L is the wavelength and H is the wave-height) was chosen for the horizontal 999 and vertical directions respectively. This corresponds to a mesh-size of $\Delta x \sim 12$ cm 1000 and $\Delta z = 1.4$ cm respectively which is comparable to the $\Delta x = \Delta z = 5$ cm chosen 1001 for the RANS model IITM-RANS3D in the same study (Saincher *et al.*, 2023a). The numerical framework exhibits "independence" from turbulence modeling because a RANS resolution was applied to LES and ILES. Since the transport equations for the 1004 mean-flow are the same between unsteady RANS and LES (Rodi *et al.*, 2013), STAR-CCM+ apparently resolved the same mean-flow in all three cases. The minor differences in peak impact pressure observed in Figure 11 stem from the unresolved 1007 scales that are modeled differently across RANS, ILES and LES (Rodi *et al.*, 2013).

Figure 11. Results from the ISOPE-2022 comparative study on breaking waves impacting a seawall with a recurved parapet (Saincher et al., 2023a): (top) seaward deflection of the breaking wave, (center) time-history of the hydrodynamic pressure 1011 over the vertical wall (PP2) as well as the parapet (PP12) and (bottom) variation of **peak impact pressure over five loading cycles.**

It is also worth noting that had the opposite been done wherein an LES grid was applied to RANS, dramatically different results would have been obtained. This is because RANS would have become grid-independent on the scale of the LES grid

whilst LES itself would only become grid-independent at the Kolmogorov scale 1018 (Rodi *et al.*, 2013). Thus, one might argue that this instance constitutes a case where turbulence modeling was attempted but eventually proved to be unnecessary.

- 1020 Hybrid modeling invariably improves the computational efficiency of the solver and 1021 should be adopted for large-scale WSI problems (Agarwal *et al.*, 2021a; Saincher *et* al., 2023a).
- 1023 The state of the art in modelling large domain problems for transient waves appeared to be based on hybrid numerical modelling using weakly coupled algorithms (or one-way coupling); this strategy was adopted by most of the participants.
- In simulating the same WSI problem at different scales, no general correlation could be obtained between computational effort and the scale of the problem. For instance, amongst the ten models compared for breaking waves interacting with a recurved seawall, the hybrid codes qaleFOAM and IITM-RANS3D were simultaneously the 1030 fastest and slowest at two different scales of the problem (Saincher *et al.*, 2023a).

In these studies it was also noted that the experimental error/uncertainty should be taken into consideration during validation. The inclusion of the experimental uncertainty would make the above guidelines less stringent. However, a conservative approach is beneficial in order to maintain a reduced error margin when adopting the said guidelines in practice.

8 The Future

8.1 Application of machine learning algorithms

The machine learning (ML) techniques is becoming popular in assisting the fluid simulation, 1039 e.g. to reconstruct the fluid field from data (Raissi et al., 2020), to predict the turbulence 1040 related parameters (Ling *et al.*, 2016; Zhang *et al.*, 2015; Kutz, 2017), and to approximate time-independent flow filed governed by NS models, such as the projection-based Pressure Possion Equation (PPE, e.g. Yang et al., 2016; Xiao et al., 2018; Tompson et al., 2017; Dong 1043 et al., 2019; Ladicky et al., 2015; Wessels et al., 2020, Li et al., 2022). Recently, both the 1044 convolution neural network (CNN, Zhang et al., 2023) and graphic neural network (GNN, 1045 Zhang et al., 2024a, 2024b) have been coupled with the incompressible smoothed particle hydrodynamics (ISPH) model to accelerate the numerical simulations. In these work, high-fidelity time-domain numerical results are produced using stand-alone ISPH simulation on wave propagation and impact on fixed structure. The CNN or GNN are used to train a machine learning algorithm to predict the pressure in the future step based on the numerical results at the current time step including the velocity, velocity divergence and pressure. After the algorithm is trained, it will be used to replace the PPE solver in the classic ISPH. Both the CNN-supported and GNN-supported ISPH models have been applied to modelling wave propagation, impact on seawall and interaction with other structures. Figure 12 and Figure 13 illustrate some numerical results from the GNN-supported ISPH, which does not only show the capacity of the ML-supported ISPH but also demonstrate its promising accuracy. Further evidence on numerical accuracy and CPU speeding-up can be demonstrated in Figure 14 for the cases with solitary wave propagation. In this figure, the error is defined by the L2-norm of 1058 the time history of the wave crest; ISPH and ISPH-CQ adopt the linear and $2nd$ -order PPE solvers, respectively.

1060 Figure 12. Comparisons of the floater movement progress during green water impact

1061 between laboratory photos (Zheng et al., 2016) (left) and ISPH GNN simulations (right) at

- 1062 different instants (duplicated from Zhang et al., 2024b).
- 1063

1064

1065 Figure 13. Time histories of the impact pressure on deck at P1 (duplicated from Zhang et al., 1066 2024b).

1068 Figure 14. Averaged errors of numerical results corresponding to different particle spacing 1069 in the solitary wave propagation (a) and the CPU speeding ratio (b) against solving PPE 1070 directly (solitary wave height = $0.28*$ water depth; duplicated from Zhang et al., 2024a).

As shown in Figure 14(a), both the convergence and accuracy of the ISPH-GNN are bounded by the corresponding values of the ISPH and ISPH-CQ, implying a promising computational accuracy. Figure 14(b) illustrates excellent CPU time speeding-up ratios against directly 1074 solving the PPE using the $2nd$ order solver. For the solitary wave propagation using 80k particles, the GNN can speed up the simulation by 80 times.

The existing work related to AI and ML may be quantified as hybrid model combing a CFD 1077 solver with the ML algorithms, e.g. Zhang et al. (2024a) combining ISPH with graph neural network for simulating free surface flows. Data are needed to train the ML algorithms. Recently, researchers started solving the fluid mechanics and fluid-structure interaction problems using the AI library for discretising the required partial differential equation 1081 (AI4PDE, see, e.g. Chen et al., 2024). This work does not need to train the neural network but directly modifying the filters of the neural network. Limited benchmarking rest has demonstrated its promising computational accuracy and efficiency. The applications of the AL/ML to existing hybrid model, such as the qaleFOAM, have yet found to the best of our knowledge. Based on our preliminary work on CNN/GNN supported ISPH, its feasibility to the hybrid modelling is confirmed.

The challenges in the hybrid modelling can be fully or partially solved by the AL/ML technologies. These include: (1) replacing the NS solver by the ML-supported version; (2) intelligently decomposing the computational domain in an adaptive way, i.e. to minimising the NS domain in the run-time depends on the development of the viscosity/turbulence effect and breaking wave occurrence; (3) intelligently choosing the appropriate models, such as RANS or LES; (4) in the function-decomposition approach, using the ML algorithms for solving the compromised equations instead of solving them directly; (5) dynamic load balancing in the cases with parallel computing.

8.2 Hybrid FNPT-RANS-LES models for floating renewables

Another important aspect in the blue economy theme is the renewable energy. The development of offshore wind farms based on Floating Offshore Wind Turbine (FOWT) arrays is one of the popular, potential and realizable area. In order to reduce the CAPEX and installation costs, shared mooring systems have been proposed for FOWT arrays where anchors and a part of the mooring line are shared between turbines. This introduces challenges that manifest differently in shallow and deep water. The deep-water mooring system is susceptible to motions of the FOWT platform being amplified leading to large displacements in the mooring line and peak anchor loads. Chain catenary moorings in shallow water experience snap loads due to their susceptibility to violent wave-current-structure interactions during extreme events and individual loads superimposing nonlinearly with the structural response. In order to develop a comprehensive understanding of the mechanisms leading to snap loads and peak anchor forces in shared mooring systems of a FOWT farm, high fidelity multi-scale solver is required. To achieve this, the existing FNPT (Fully Nonlinear Potential Theory), RANS (Reynolds Averaged Navier-Stokes) and Large Eddy Simulation (LES) codes can be coupled via a zonal approach to yield a high-fidelity multi-scale solver for wave-current-structure interaction. A FEM-based structural solver will

be integrated to accurately predict the coupled fluid-structure interaction of several mooring lines and to facilitate the modelling of elastic materials. A critical aspect of the model development would be scaling-up the code for prototype-scale FOWT arrays whilst retaining computational efficiency and accuracy. This could be achieved using AI and ML-based prediction of turbulence-generation near the floating platforms, as this is expected to be the most computationally intensive aspect of the modelling (traditionally handled using hybrid RANS-LES). Thus, a continuous research efforts in the field of computational hydrodynamics is required. This is in fact supported by the Newton Fellowship (recently awarded to the second author from the authors research group in 2023) wherein the existing understanding in hybrid modelling as well as AI/ML-based prediction of turbulence shall be carried forward.

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1751 NOMENCLATURE

1752 Roman symbols

1754 Greek symbols

1755 Superscripts

1756 Abbreviations

