THE BANK OF ENGLAND INDUSTRY DATASET

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Abstract

This paper describes the sources and methods used to construct the Bank of England industry dataset. In its current form, the dataset comprises annual data on 34 industries covering the whole economy over the period 1970-2000. The main variables are gross output, value added, capital, labour and intermediate input. Capital input is built up from the services of seven assets, of which three are ICT (computers, software and communications equipment). Each industry’s intermediate input is an aggregate of purchases from all other industries and from imports. Labour input is measured by hours worked, but with an adjustment for quality change derived from aggregate data. We employ U.S. methods to measure ICT. Apart from this, the dataset is consistent with the UK national accounts both in real and nominal terms.

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1. Introduction

This paper describes the sources and methods used to construct the Bank of England industry dataset. In its current form, the dataset comprises annual data on 34 industries covering the whole economy over the period 1970-2000; in the present paper, we use only the data for 1979-2000.

The starting point was a dataset on nominal gross output, value added, and domestic and imported intermediate input, and associated price indices, for 49 industries. This was prepared for us to our specification by a private sector economic consultancy, Cambridge Econometrics (CE). To this we added our own estimates of labour and capital input, finishing up with a KLEM-type dataset for 34 industries. The reduction from 49 to 34 industries was mainly necessitated by our desire to measure ICT capital services separately. The original 49 industries (which we refer to as CE 1-CE 49) and the final 34 industries, together with their definitions in terms of the U.K.’s 1992 Standard Industrial Classification (SIC), are in Tables A.1 and A.2. Table A.3 gives the mapping between the 34 industries and the sectors they were aggregated to in Basu, Fernald, Oulton and Srinivasan (2003).

We constructed our own dataset since nothing comparable is currently available from official sources. The raw materials for our dataset are of course the series collected by the U.K.’s Office for National Statistics (ONS) and used by them to construct the national accounts. But these series are not put together in the form required for productivity analysis. For example, the ONS does not currently produce measures of capital services, only measures of capital stocks. Moreover, their estimates of stocks are only published at the aggregate level, not at the industry level. No official measures of ICT capital stocks or services have so far been produced. On the labour side, total hours worked are published only at the aggregate level, not by industry. On the output side, only real value added is published by industry, not real gross output. And there are no official measures of real intermediate input.

An important principle behind the construction of the dataset is that it should be as far as possible consistent with the national accounts, both in nominal and real terms. Since the national accounts are continually revised, the dataset can only be consistent with the accounts at a specific point in time. In the present case, this means consistent with the 2002 Blue Book, the latest available when our work began (Office for National Statistics (2002a)). For the period 1992-2000, nominal consistency is relatively easy to achieve since for this period we have the Input-Output Supply and Use Tables (I-O SUTs (ONS 2002b)) and these data are themselves consistent with the 2002 Blue Book. The Supply and Use Tables give gross output, value added, profits, the wage bill and intermediate purchases (domestic plus imports), all in nominal terms, for 123 industries and products. So we can ensure that,
for our 34 industries, these series match those of the I-O SUTs. Prior to 1992, detailed nominal consistency is harder to achieve, though we can ensure that it holds for broad sectors.

By real consistency is meant that, when we aggregate our industry estimates of real output up to the aggregate level, the growth rate of the aggregate should equal that of the official estimate of GDP. In fact, we make a number of adjustments to our output estimates (described below) so that neither real nor nominal consistency holds for the series we eventually employ to estimate TFP growth. But real consistency does hold for the output series prior to these adjustments. Actually, we do find a difference between the growth of our aggregate and the official estimate of GDP growth, but we adjust the growth of each industry to eliminate this discrepancy.

A second important principle behind the dataset is that industry output should be measured gross, so that proper account can be taken of the contribution of intermediate input. An input-output approach was therefore necessary, so something needs to be said about the availability of input-output tables in the U.K. For the period 1992-2000 we have the 2002 edition of the I-O SUTs. These are fully consistent with the 2002 national accounts and use the 1992 SIC. However, they give only total purchases by each industry, not the breakdown into domestic and imported purchases. For 1989-1991, we have earlier versions of the I-O SUTs. These are not fully consistent with the national accounts and they use the 1980 SIC. Prior to 1991 there are no annual I-O SUTs. But we do have full input-output tables for selected years. These give the breakdown into domestic and imported purchases but within our period they are only available for 1968, 1974, 1979, 1984, 1990 and 1995. Furthermore, the SIC according to which industries are defined and the conceptual basis of the tables has changed in significant ways over time.

These earlier input-output tables were converted to a common SIC and price concept. This was possible since they all break down the economy into considerably more than 49 industries. They were also made consistent with the 2002 national accounts. Intervening years were then interpolated by the RAS method using national accounts totals as controls.\(^{(1)}\)

In what follows, we first discuss our approach to measuring investment in ICT and then describe the measurement of output and of the inputs (labour, capital, and intermediate) in more detail.

\(^{(1)}\) This part of the work was performed by Cambridge Econometrics.
2. The treatment of ICT

We employ U.S., not U.K. price indices for deflating investment in computers and software. We also apply a large adjustment to the official estimate of nominal software investment: we multiply it by three. Our reasons for these decisions are set out below: for more detail, see Oulton (2001) and (2002). Broadly, we argue that this gives a more realistic picture. In addition, it facilitates comparison with the U.S., since it means that a very similar methodology is being employed for both countries.

The U.S. computer price index

There is an official Producer Price Index (PPI) for computers in the U.K. (ONS code: PQEK), but it falls much less rapidly than its U.S. counterpart. It is common to describe the U.S. index as hedonic, while the U.K. PPI certainly is not. The suggestion is then that any substantial difference between the United States and other countries’ indices arises from the use of hedonic methods.

A number of points can be made here. First, the hedonic technique has a firm basis in economic theory and has been employed in practice in United States official statistics for many years (see Triplett (1987) and (1990) and Moulton (2001)). Its application to U.S. computer prices goes back to Chow (1967) and Cole et al (1986); the latter’s work was extended by Oliner (1993) and by Berndt and Griliches (1993).\(^2\)

Second, the traditional approach of national statistical agencies is the matched models approach, under which a set of physically identical products, sold on commercially identical terms, is tracked over time. Though the U.S. computer price index is often described as a hedonic index, this is rather misleading. In fact, the index uses the normal matched model approach. Hedonic methods are employed only when an old model drops out and it is necessary to link a new model into the index: see Sinclair and Catron (1990) for an account of the U.S. methodology.

Third, the rapid rate of fall of U.S. price indices for ICT products is not due entirely to the use of hedonic techniques. Indices based purely on the matched models approach can also show rapid rates of decline. For example, a price index for semiconductors constructed at the U.S. Federal Reserve and used by Oliner and Sichel (2000) was falling at a rate of more than 40% a year between 1996-99. This index was entirely based on matched models and made no use of hedonic methods at all. Aizcorbe et al (2000) (see also Landefeld and Grimm (2000)), using a large database of computer prices gathered
by a market research firm, have shown that a matched models price index for computers can fall just as rapidly as the official U.S. index. But the models included have to be a representative sample and the data have to be sampled at relatively high frequency (quarterly in their study). It is also desirable that data on quantities as well as prices are available so that a superlative price index can be constructed. It is possible therefore that some of the difference between the U.S. computer price index and those of other statistical agencies may be due to the fact that these conditions are not always satisfied.

Fourth, the U.K. retail price index for computers (which is published as part of the Harmonised Index of Consumer Prices) is also not hedonic, but has been falling at about the same rate as its U.S. counterpart and much more rapidly than the corresponding U.K. PPI.

Fifth, in work commissioned by the ONS, Stoneman, Bosworth, Leech and McAusland constructed a hedonic index for U.K. computer prices for the years 1987 to 1992; their results are reported in Stoneman and Toivanen (1997, Table A3). They found that their index fell by 19.1% a year over this period; by contrast the official PPI for computers (ONS code PQEK) fell by only 7.2% a year.

Three criticisms are often made of the application of U.S. indices to the United Kingdom or other foreign countries:

- U.S. producers possess monopoly power so that prices charged in the United States are not representative of prices charged in the United Kingdom.
- Adjusting for the exchange rate assumes that ICT products are priced in dollars with instantaneous pass-through into sterling, which may not be true.
- The U.S. price indices are averages over different products, eg in computers they are averages over the prices of personal computers (PCs), notebooks, servers, etc. The mix of products may differ between countries.

The importance of the first point depends on whether we are concerned with the growth rates or the levels of prices. It is certainly possible that the level of prices may differ between countries because of market discrimination by suppliers who possess some monopoly power. And there is plenty of anecdotal evidence that ICT prices are higher in the U.K. If market power is constant, then U.K. and U.S. growth rates are unaffected. Even if the degree of monopoly power changes, the effect of this on

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(2) Nor are such studies confined to the United States. In a pioneering study of U.K. computer prices using hedonic methods, Stoneman found that over the period 1955-70, with quality held constant, his preferred price index fell at about 10% a year (see Stoneman, (1976), Chapter 3, Table 3.2, series (e)).
the rate of growth of U.K. prices is likely to be swamped by the huge falls observed in U.S. prices. Casual empiricism suggests that, if anything, the U.K. market for ICT has become more competitive in recent years relative to the United States. If so, U.K. prices will have fallen more rapidly than assumed here. Hence ICT stocks in the U.K. will have been growing more rapidly than on our estimates. This could affect the weight that ICT assets receive in calculating the growth of aggregate capital services.

The second and third points are valid in principle. How important they are in practice can only be resolved by direct research on prices. It is not obvious that such research would necessarily support a faster growth rate of United Kingdom prices than is assumed here.

**U.K. software investment in current prices**

Software investment has three components:

- prepackaged software, eg an office suite sold separately from the computer on which it is to be run;
- custom software, written (usually) by a software company specifically for sale to another company; and
- own account software, written in-house for a company’s own use.

There is a fourth category, bundled software, eg the operating system and other programs, which are typically sold together with a PC. This category is included under investment in computers.

Software investment was first incorporated into GDP in the U.K. in the 1998 National Accounts. Previously, all spending on software was treated as intermediate consumption (like business purchases of stationery). The procedure used by the ONS to derive a series for software investment was first to estimate a benchmark figure for 1995, based on a 1991 survey of sales of computer services companies, and then to carry this figure forwards and backwards using the growth rates of indicator series. For the earlier years, the growth of total billings by the computer services industry was used. Years after 1995 used the growth of the wage bill of full time programmers, computer engineers and managers in the computer services industry.

The growth rate of software investment in current national prices has been very similar in the United States and the United Kingdom. But there is a very large discrepancy in the levels. In the United States, software investment as a proportion of computer investment (both in current prices) began steadily climbing in 1984 and levelled off after 1991. During the 1990s it averaged 140% of computer investment. In the United Kingdom by contrast, software investment apparently averaged only 39% of
computer investment in the 1990s. Since people buy computers to run software, it seems very unlikely that there should be such a large discrepancy between the United Kingdom and the United States.

There is also a striking discrepancy in the proportion of the sales of the computer services industry that are classified as investment in the two countries. In the BEA’s 1996 input-output table, we find that 60% of total sales of products of industry 73A, ‘Computer and data processing services, including own account software’, was classified as final sales (mostly investment). The 1996 figure was based on the 1992 economic census which asked firms in this industry to distinguish between receipts from prepackaged software, from custom software and receipts from other activities, the first two of these being investment. In the United Kingdom in the same year, investment apparently accounted for only 17.5% of total sales of the corresponding product group (input-output group 107, ‘Computer and related activity’).

The United Kingdom also appears to be out of line with other European countries. Lequiller (2001) has compared France with the United States. He finds that the ratio of software investment to IT equipment investment was about the same in the two countries in 1998 (his page 25 and Chart 5). He also finds that the ratio of software investment to intermediate consumption of IT services is substantially lower in France than in the United States (pages 26-27). This ratio is exceptionally high in the United States, but equally his Chart 6 shows that it is exceptionally low in the United Kingdom. In fact, the reported U.K. ratio is substantially lower than in France, the Netherlands, Italy and Germany. Lequiller argues that in Europe software investment is based on data from purchasers while in the U.S. it is based on data from sellers, with the latter method tending to produce higher results. This however cannot explain the low U.K. level since the 1995 benchmark figure was based on sales data.

Part of the difference in software levels may be due to a different treatment of own account software in the United States. This now constitutes about a third of all U.S. software investment and is estimated from the wage bill (grossed up for other costs) of computer programmers employed throughout the economy (see Parker and Grimm (2000)). Own account software is likely to be important in the United Kingdom too. In 1995 only 27% of software engineers and computer programmers were employed in the computer services industry (see Oulton (2001), Annex B). Presumably, an important function of the other 73% was to write software.

Oulton (2001) employed U.S. methods to estimate U.K. own account software and re-considered the survey on which the 1995 benchmark was based. The result is that 1995 software investment is
estimated to be about 4.1 times the current official figure. Alternative, rougher multipliers are suggested by the two discrepancies noted above. A multiplier of 3.6 is arrived at by dividing the U.S. ratio of computer investment to software investment, averaged over 1990-98 (=1.40), by the corresponding U.K. ratio (=0.39). A factor of 3.4 is suggested by the comparison of the U.K. and U.S. input-output tables. In order to err on the conservative side, we choose a multiplier of 3. The growth rate of both nominal and real software investment is of course left unchanged by this adjustment.

**Software price indices**

In the United States, each of the three types of software has a different price index (see Parker and Grimm (2000)). In the case of prepackaged software, an index using hedonic techniques exists. For own account software, there is no hedonic index and the growth of the price index for this component is linked to the growth of wages of computer programmers. This means that the price index is assuming zero productivity growth amongst programmers. For the remaining component, custom software, the Bureau of Labor Statistics uses a weighted average of the prepackaged (25%) and own account (75%) indices. Nominal investment in each type of software is deflated by its own price index and then summed to get real software investment. The overall price index is derived as an implicit deflator: total nominal divided by total real investment. The assumption of zero productivity growth amongst computer programmers employed to write own account software is very implausible. So there is a case for saying that the index overstates inflation. But partly for reasons of compatibility, we decided to use it.

There is no official PPI for software in the United Kingdom. Expenditure on software is deflated by an index of the wages of computer programmers, with a “guesstimated” adjustment for productivity.

**Communications equipment prices**

The market for communications equipment was till recently less integrated internationally than other ICT markets. The methodological difference between the U.K. and U.S. official price indices is smaller. Hedonic methods only affect a small part of the U.S. index (Grimm et al. (2002)). And in practice the U.K. price moves in a similar fashion to the U.S. one. For these reasons, we employ the U.K. price index (ONS code: PQGT) to deflate nominal investment in communications equipment.

**Implications for the other variables in the dataset**

Our approach to ICT has implications for the other variables in the dataset. Changing the prices used for measuring real investment in computers and software means that we must also adjust the prices used to measure U.K. output of these products. The “times three” adjustment to nominal software
investment raises nominal GDP as measured from the expenditure side. To maintain consistency we must make a corresponding adjustment to the income side of the accounts. These adjustments are described more fully below.

3. Output

Nominal output
The accounting identity relating gross output and value added in nominal terms is:

\[
\text{Gross output} = \text{Value added} + \text{Domestic intermediate input} + \text{Imported intermediate input}
\]

Also,

\[
\text{Value added} = \text{Gross operating surplus (profits)} + \text{Wage bill} + \text{Taxes on production}
\]

Value added is at basic prices. Taxes on production, which include items like business rates and vehicle licences, are usually a small proportion of the total. For 1989-2000, gross output, value added and its components come from the I-O SUTs. For some earlier years, they are derived from the periodic input-output tables, when available. For years prior to 1989 when no input-output table exists, they are derived by interpolation. Aggregate value added so derived is controlled to equal GDP at current basic prices (ONS code: ABML). There is no corresponding control for aggregate gross output, which can only be carried back to 1989 using official data (ONS code: NQAF).

Real output
Given that nominal gross output is the sum of nominal value added and nominal intermediate input, a Divisia index of real gross output in industry \( i \) is:

\[
\hat{Y}_i = v_i \hat{V}_i + (1-v_i)\hat{M}_i
\]

whence

\[
\hat{V}_i = [\hat{Y}_i - (1-v_i)\hat{M}_i] / v_i
\]  \hspace{1cm} (A.1)

Here \( Y_i \) is real gross output in industry \( i \), \( V_i \) is real, double deflated, value added, \( M_i \) is real intermediate input, \( v_i \) is the share of nominal value added in nominal gross output, and a hat denotes a growth rate. This last equation serves as a definition of real value added.
Consistency of industry real output with official estimates of GDP growth

There are two ways in which real GDP may be measured, from output or from expenditure. From the output side, a Divisia index of GDP growth is:

\[
\text{GDP growth} = \sum_{i=1}^{n} w_i \tilde{V}_i
\]  

(A.2)

Here \( w_i \) is the share of nominal value added in industry \( i \) in aggregate nominal value added (current price GDP), and there are \( n \) industries. Second, from the expenditure side:

\[
\text{GDP growth} = \sum_{i=1}^{n} s_i \hat{E}_i
\]  

(A.3)

where \( E_i \) is final expenditure on the products of industry \( i \) and \( s_i \) is the share of final expenditure on \( i \) in current price GDP. We can readily show that these two measures of GDP growth are equal, in the absence of errors or omissions in the statistics.\(^3\) But note that equality is only guaranteed in principle if value added is measured by double deflation, as in equation (A.1).

In practice, of course the two estimates will differ. The ONS takes the view that for annual data the expenditure side estimate is the most reliable. In the published figures, there is no discrepancy between the two estimates (unlike in the U.S. NIPAs) but this is because the output side estimate is adjusted to conform to the expenditure side one. The reason for preferring the expenditure side estimate is twofold. First, much of the hard-to-measure part of the economy is engaged in producing intermediate products (eg business services or wholesale banking) and these activities largely drop out of GDP on the expenditure side.\(^4\) Second, in practice the ONS does not use double deflation to estimate real value added (except in electricity supply and agriculture); instead it uses real gross output as a proxy for real value added (in most cases, gross output deflated by an appropriate price index): see Office for National Statistics (1998), chapters 11 and 13, and Sharp (1998). The discrepancy between the output and expenditure side estimates is removed by adjusting output growth in the private service industries; output in the production sector (about a third of the economy) and in the government sector (about a fifth) is not adjusted.

\(^3\) This ignores the difference between market prices at which expenditure is usually measured and basic prices at which output is usually measured. But the argument can easily be extended to encompass this point.

\(^4\) Not completely, since some enter into international trade.
We accept the argument that (within its assumptions) the best available estimate of GDP comes from the expenditure side. We therefore require our estimates of industry output to be consistent with the expenditure estimate of GDP. There are two ways in which this could be implemented. In the first method (method A), we assume that the ONS’s measures of “real value added” are in fact measures of \textit{real gross output} and proceed in a number of steps:

\textit{Step 1.} Use equation (A.1) to calculate double deflated real value added in the production sector and government sectors, treating the ONS’s “real value added” as in fact “real gross output” and using data on intermediate input (see below).

\textit{Step 2.} Recalculate GDP growth by means of equation (A.2), using the \textit{new} estimates of real value added for the production sector and government sectors, but the \textit{original} real gross output estimates for the services sector, which we continue to treat as measures of real value added.

\textit{Step 3.} Step 2 will produce a different result from the original one for GDP, so we adjust the growth rates of real value added in services so that GDP growth is the same as before. The rationale for this is that overall GDP growth is given from the expenditure side, so should not be changed.

\textit{Step 4.} Given the new, adjusted, growth rates of real value added in services, we calculate new growth rates of real gross output in services from equation (A.1).

This method assumes in effect that the only reason for any difference between the output and expenditure side estimates is that gross output is erroneously used in place of value added for measuring output in the production sector. In practice, we find that this method produces a very large adjustment to the growth of output in private services: 3.6\% per annum over 1979-2000 with a high degree of year-to-year variation. This seems far higher than any plausible estimate of the difference between the growth rates of real value added and real gross output in private services. Hence we reject method A in favour of the alternative, method B.

Under method B, we assume (with the ONS) that the ONS’s output measures are the best available estimates of \textit{real value added} (the opposite to Method A). We then use these together with data on intermediate input to derive real gross output for each industry. On average across the 49 industries, the standard deviations of the growth rates of both real gross output and of real value added under
Method B are about half those found under Method A. In fact, the variability of growth under method A is often implausibly high.

*Enforcing consistency between the industry estimates and GDP*

We aggregate output in the 49 industries to the whole economy level, using method B, and compare the result with the official measure. Our industry outputs do not include private housing (the actual and imputed rentals on dwellings), so in making the comparison we exclude housing from official GDP. Our indices are in basic prices, so we use GDP at basic prices (ONS code: ABMM), after excluding housing (ONS codes for housing: nominal, QTPS; real, GDQL). We back out non-housing GDP from a Fisher index of the two components of GDP, housing output and non-housing GDP. In practice, we find that there is a discrepancy between our aggregate measure of GDP growth and the official estimate. It may seem surprising that there is any discrepancy at all, given that our real output series derive from official ones. But our estimate of GDP is built up from 49 components only, while the official estimate derives from a much lower level of aggregation. Also we use a Fisher index while the official series is Laspeyres but with weights which are updated every five years or so; however, this seems to make little difference. To enforce consistency between the micro and macro views, we adjust the growth rate of real value added in each of the 49 CE industries by the amount of the discrepancy (measured in percentage points per annum).

*Other adjustments to industry-level output*

Two other adjustments were made to the industry estimates of real value added:

1. We add back the “financial services adjustment” into nominal value added in banking
2. ICT adjustments. These lead to higher nominal value added and profits in all industries and to higher real output growth in the Electronics and Computing services industries due to the use of U.S. rather than U.K. price indices.

These adjustments all raise the estimated growth rate of GDP, particularly in 1995-2000.

*Financial services adjustment* The so-called “financial services adjustment” (FSA) is a consequence of the treatment of the banking industry in the 1968 System of National Accounts (SNA). In the latter, profits and value added in all other industries are recorded gross of interest payments, which are regarded as a transfer payment. But if the banking industry is treated in the same way, then there is double-counting since a large part of bank profits arise from net interest receipts. Hence under the 1968 SNA, interest receipts are subtracted from bank profits and value added. But then the weight
given to the banking industry in GDP is absurdly small or even conceivably negative. The 1993 SNA is an improvement conceptually since it recognises that banks perform a service of financial intermediation. So bank value added is higher and that of firms in the rest of the economy lower by the amount of this financial service (measured essentially as the difference between borrowing and lending rates times the value of loans). Intermediate consumption by non-bank firms is higher, and value added lower, by the amount of this purchased service. However, the reduction in non-bank value added does not completely offset the rise in bank value added, because banks also lend to final buyers (consumers, government and the foreign sector). This service is counted as final consumption and so raises GDP under the 1993 SNA.

Our approach is to add back in the FSA (ONS code: NSRV) to profits and value added in CE industry 40 without making any further adjustment. So we have moved half way between the 1968 and the 1993 SNAs. This gives the appropriate weight to the banking industry, but we should really make some downward adjustment to profits and value added elsewhere to be fully in accordance with the 1993 SNA. We have not done this, mainly because we do not have the necessary information to move all the way to the 1993 SNA. The ratio of NSRV to nominal GDP averaged 3.7% 1979-2000, so after allowing for final buyers, the effect on value added in other industries would probably be small, a downward adjustment averaging probably about 2%. Note that our approach only affects the weight applied to banking in calculating aggregate GDP, not banking output.

**ICT adjustments**

We use U.S. price indices (adjusted for exchange rate changes), rather than U.K ones, to deflate the output of computers and software. We also argued that the level of software investment has been underestimated in the U.K. This implies that the level of profits and value added has also been underestimated. In more detail, the ICT adjustments are as follows:

1. The growth of real gross output in CE industries 20 (Electronics) and 43 (Computing services) is adjusted by using U.S. rather than U.K price indices for (respectively) computers and software. In the case of industry CE 20, only that part of output believed to consist of computers is adjusted (about 50%).

2. We argued that in the U.K nominal software investment has been seriously underestimated, by a factor of three. There has been a corresponding overestimate of intermediate consumption of

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(5) This was estimated as the ratio of gross output of input-output industry 69 to total gross output of input-output industries 69, 73, 74 and 75; values for 1969-91 were set equal to the value for 1992. The source was the I-O SUTs.
software services. In our investment and capital estimates (see below), we therefore multiply the nominal level of software investment by three. For consistency, we must make a corresponding adjustment to profits and value added in each industry, so that the output and expenditure estimates of nominal GDP remain equal. Profits and value added are therefore increased in all industries to reflect this “times three” adjustment to nominal software investment: ie, we add twice the original level of software investment to each industry’s profits and value added.

We have calculated aggregate GDP both before and after the adjustments that we think are desirable and compared it with the ONS estimate of GDP growth. The GDP measures are as follows:

\[ gdp49: \] Fisher index of real value added in the original 49 industries, weighted together using nominal value added

\[ gdp34: \] Fisher index of real value added in our ultimate list of 34 industries, weighted together using nominal value added. Real output is adjusted for ICT and aggregated up to 34 industry level. The nominal weight for banking adds back in the “financial services adjustment”. The industry-level output series that make up \[ gdp34 \] are the ones used in our growth accounting calculations.

\[ abmmxh: \] GDP at basic prices (ONS code: ABMM), excluding housing (ONS codes: nominal, QTPS; real, GDQL ). \[ abmmxh \] is calculated as if ABMM were a Fisher index of the two components, housing output and non-housing GDP

Table A.4 and chart A.1 below shows the effect on the aggregate GDP growth rate estimated from industry data of making these adjustments. The difference between \[ abmmxh \] and \[ gdp49 \] measures the extent of the discrepancy between industry-level output and aggregate output. In general, \[ gdp49 \] grows less rapidly than the official estimate. The largest discrepancy occurs in 1990-1995, when \[ gdp49 \] falls short by 0.8% per annum. Removing the financial services adjustment raises the growth rate by 0.14 percentage points above the official level in the 1980s and by 0.03 percentage points in the 1990s. The ICT adjustment also has a substantial effect: growth in the 1980s is raised by a further 0.27 percentage points and in the 1990s by a further 0.15 percentage points. The effect is most marked in 1995-2000: 0.29 percentage points.

Input-output row 69 is Division 30 of SIC92, input-output rows 73-75 make up Division 32 of SIC92. The aggregate of SIC92 Divisions 30 and 32 is CE industry 20.
4. Capital\(^{(6)}\)

Capital input is measured by capital services from different types of assets. We distinguish between non-ICT capital and ICT capital. For each of our 34 industries we estimate the capital services flowing from stocks of the following four non-ICT\(^{(7)}\) assets:

1. Buildings
2. Equipment (excluding computers, part of software and communication equipment)
3. Vehicles
4. Intangibles (excluding rest of software)

and the following three ICT assets

5. Computers
6. Software
7. Communication equipment

While the wealth measure of capital is more firmly established and the standard measure produced by the ONS, in the context of production theory the flow of capital services is the correct measure to use.\(^{(8)}\) The measures for ICT, non-ICT and total fixed capital are calculated by weighting the growth of asset stocks in the respective categories by their rental prices. Rental prices are measured using the Hall-Jorgenson formula.

The method

The equations of our model for estimating capital services are as follows:

\[
B_{it} = I_{it} + (1 - \delta_i) \cdot B_{i,t-1}, \quad i = 1, ..., m
\]  
\[
A_{it} = (1 - \delta_i / 2) \cdot B_{it}
\]  
\[
K_{it} = \bar{A}_{it} = \left[ A_{i,t-1} \cdot A_{it} \right]^{1/2}, \quad i = 1, ..., m
\]

---

\(^{(6)}\) For a fuller discussion of the methods and the empirical issues, see Oulton and Srinivasan (2003).

\(^{(7)}\) Buildings exclude residential dwellings. The “traditional” asset classification follows that of the OECD’s System of National Accounts, 1992 and is that followed by the ONS.

\(^{(8)}\) See Oulton and Srinivasan (2003, Section 1) for a discussion of the difference between the capital wealth and capital services measure.
\[ p^K_i = T_i \left[ r_i \cdot p^K_{i,i-1} + \delta_i \cdot p^A_{i} - (p^A_{i} - p^A_{i-1}) \right], \quad i = 1, \ldots, m \]  
(A.7)

\[ \Pi_i = \sum_{t=1}^{m} p^K_i K_{i,t} = \sum_{t=1}^{m} T_i \left[ r_i \cdot p^A_{i,i-1} + \delta_i \cdot p^A_{i} - (p^A_{i} - p^A_{i-1}) \right] \cdot K_{i,t} \]  
(A.8)

\[ \ln \left( K_i / K_{i-1} \right) = \sum_{t=1}^{m} \bar{w}_{i,t} \ln \left( K_i / K_{i,t-1} \right), \]

\[ \bar{w}_{i,t} = (w_{i,t} + w_{i,t-1}) / 2, \quad w_{i,t} = \frac{p^K_i K_{i,t}}{\sum_{t=1}^{m} p^K_i K_{i,t}}, \quad i = 1, \ldots, m \]  
(A.9)

where:

- \( m \) is the number of assets
- \( A_{it} \) is the real stock of the \( i \)th type of asset at the end of period \( t \)
- \( \bar{A}_{it} \) is the real stock of the \( i \)th type of asset in the middle of period \( t \)
- \( B_{it} \) is the real stock of the \( i \)th type of asset at the end of period \( t \), if investment were assumed to be done at the end of the period, instead of being spread evenly through the period
- \( K_{it} \) is real capital services from assets of type \( i \) during period \( t \)
- \( I_{it} \) is real gross investment in assets of type \( i \) during period \( t \)
- \( \delta_i \) is the geometric rate of depreciation on assets of type \( i \)
- \( r_i \) is the nominal post-tax rate of return on capital during period \( t \)
- \( T_{it} \) is the tax-adjustment factor in the Hall-Jorgenson cost of capital formula
- \( p^K_i \) is the rental price of new assets of type \( i \), payable at the end of period \( t \)
- \( p^A_i \) is the corresponding asset price at the end of period \( t \)
- \( \Pi_i \) is profit (= nominal aggregate capital services) in period \( t \)
- \( K_i \) is real total capital services during period \( t \)

Equations (A.4) and (A.5) describe the evolution of asset stocks. They can be shown to arise from the following accumulation equation:

\[ A_{it} = (1 - (\delta / 2)) \cdot I_{it} + (1 - (\delta / 2)) \cdot (1 - \delta_i) \cdot I_{i,i-1} + (1 - (\delta / 2)) \cdot (1 - \delta)^2 \cdot I_{i,t-2} + \ldots \]  
(A.10)

The factor \((1 - \delta / 2)\) arises as investment is assumed to be spread evenly throughout the unit period, so on average it attracts depreciation at a rate equal to half the per-period rate. This assumption affects the level, but not the growth rate, of the capital stock.\(^{(9)}\)

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\(^{(9)}\) This assumption corresponds to the practice of the BEA: see U.S. Department of Commerce (1999, box on page M-5).
Equation (A.6) states that capital services during period $t$ derive from assets in place in the middle of period $t$. The capital stock in the middle of period $t$ is estimated as the geometric mean of the stocks at the beginning and end of the period. Equation (A.7) defines the rental price of assets of type $i$. Equation (A.8) says that profit equals the sum over all assets of the rental price times the asset stock. Equation (A.9) defines the growth rate of capital services.

This model can be applied to both industry and whole economy data. Given asset prices, investment, depreciation rates, the tax adjustment factors, and aggregate profits, we apply it to whole economy data in order to estimate the nominal rate of return. We assume that this rate of return applies to each industry. The sources for the whole economy estimates are described fully in Oulton and Srinivasan (2003). Armed with an estimate of the rate of return, we then apply the model to industry data to estimate capital services in each of our 34 industries. The alternative was to estimate the model for each industry separately, using industry profit to estimate a different rate of return in each industry. We rejected this alternative, as likely to lead to unrealistically volatile estimates. For reasons explained below, the aggregation (in Equation A.9) is done using a Fisher index, not a Törnqvist index.

**Real asset stocks, by industry**

Equations (A.4) and (A.5) are used to generate stocks of each asset, by industry. They require an “initial stock” value, an assumed depreciation rate and real investment for each asset, by industry. Starting stocks for buildings, plant and vehicles for end-1947 were calculated using historical data as generated in Oulton (2001). Starting stocks for intangible assets were set equal to zero in end-1947; for computers in end-1959 and software and telecommunications equipment in 1964.

We distinguish separate depreciation rates across assets, but these rates are not assumed to vary across industries. (10) A constant geometric depreciation rate is assumed for each asset. (11) These annual depreciation rates are based on Fraumeni (1997) and are given in Table A.5.

Real investment in each asset, by industry, is calculated by dividing nominal investment by the price deflator for the asset.

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(10) The only exception is vehicles where the annual depreciation rate is 5.89% for rail transport (Industry 22), 6.11% for water transport (Industry 24) and 8.25% for air transport (Industry 25). The rest of the industries have a common rate of 25%.

(11) See Oulton and Srinivasan (2003, Section 3) for a discussion on the relative merits of geometric and straight-line depreciation rates.
Nominal investment in the ICT assets for 1992-2000 is extracted from Table 6 of the Input-Output Supply and Use Tables (ONS 2002b) and for 1989-1991 from earlier Supply and Use Tables. The data is only provided for 36 purchasing industries. We exclude roads, and to match the classification in the rest of the dataset we had to merge Motor vehicles, sales and repairs (I-O SUTs Industry 19) and Wholesale trade (I-O SUTs Industry 20) thus leaving us with 34 industries. The relevant rows in the tables are 69 (“Office machinery & computers”) for computers\textsuperscript{(12)}, 107 (“Computer services”) for software and 74 (“Transmitters for TV, radio and phone”) for telecommunications equipment. For earlier years we take the 1989 industry proportions of the total and distribute the whole economy figures for those years accordingly. The whole economy nominal investment series for computers and telecommunications equipment are taken from Oulton (2001, Table B.2) and software from Oulton and Srinivasan (2003, Table C.2). Investment in I-O SUTs industries “financial intermediation” and “real estate, renting and business activities” was adjusted (in proportion to the industries’ value added) to match industry definitions for Industries 28 and 29 in our dataset.

For reasons set out in section 2 above, we employ U.S. price indices, converted to sterling terms, to deflate investment in computers and software in current prices. For communications equipment, we use the official U.K. investment deflator.\textsuperscript{(13)}

Nominal investment and associated price deflators in buildings, plant and vehicles for the 34 industries were taken from an investment dataset supplied by the ONS, and soon to be made available on their website. The data have been scaled so that the aggregate nominal investment in each asset is equal to the published total in the National Accounts 2002. Some industry specific deflators for buildings and plant were smoothed so that the rental price (Equation A.7) remained positive.

For some industries (2, 10, 16, 22, 25), for some years, nominal investment in vehicles is negative and for one year for industry 22, buildings investment is negative. This is conceptually possible since investment is measured as acquisitions less disposals. It is then arithmetically possible for the accumulation equations (A.4) and (A.5) to generate a negative stock. This is conceptually impossible and would be a sign either that our depreciation assumption is wrong or that there is an error in the data. But in fact we found that the stocks were always positive, even when investment was negative. So we have not adjusted the investment data.

\textsuperscript{(12)} Only a proportion of the investment data in row 69 is taken to be computers since the total for row 69 includes office equipment. For details on this proportion see Oulton (2001), Annex B, Section B.1.
Intangibles assets in the U.K. consist mainly of software, mineral exploration and artistic originals. The software investment series available in the ONS investment dataset is the software component of intangibles investment. However, the software series extracted from the SUTs is total software comprising software in intangibles and software subsumed in plant. Oulton and Srinivasan (2003, Appendix C) have constructed series for whole economy for each component of total software investment. Using the whole economy proportions, we divide industry level total software investment into industry levels of software in intangibles and software in plant. Except for our Industry 2 (“Oil and Gas”) and Industry 34 (“Miscellaneous Services”) we have assumed that total intangibles investment in all industries equals the “software in intangibles” investment. In other words, we assume that except for Industries 2 and 34, the only intangible investment an industry does is in software.

From the published whole economy intangibles investment series we can subtract the aggregate of software in intangibles investment to get intangibles investment in the other components: mineral exploration and artistic originals. In the absence of more detailed information, this is split equally between industry 2 and 34 i.e., we assume that industry 2 does mineral exploration and industry 34 (which includes radio, television, motion picture and video activities, and museums) is the main repository of artistic originals. Since we treat software as a separate asset, this implies that in the calculations, intangibles is really “intangibles net of part of software” or “rest of intangibles” and because of our assumptions, is zero for all industries except 2 and 34.

The U.K. National Accounts do not distinguish computers, software and communications equipment separately — they are subsumed in the plant (i.e., “other machinery and equipment”) and intangibles categories. However, because we treat computers and communications as separate assets in the calculations, plant is really “plant net of computers, part of software and communications equipment” or “rest of plant”.

In nominal terms, to calculate “rest of intangibles” and “rest of plant” is easy. Simply subtract the sub-components from the total. We use the Törnqvist formula to back out the real investment in the “rest” for each industry. For years for which real investment in one of the components (computers, part of software or communications) is zero, the Törnqvist formula breaks down. In such cases, we have assumed that the real investment in rest of plant (or intangibles) is approximately equal to real investment in total plant (or intangibles) less real investment in the non-zero components. The price

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(13) Communications equipment in the U.K. is Industry 32.2 (SIC92). In the U.S. (following the classification in Jorgenson, Ho and Stiroh (2002)) communications equipment is in Industry 366 (SIC87).
deflator for plant for each industry and intangibles (for industries 2,34) is recalculated as the ratio of nominal “rest” to real “rest”.

Thus using the real investment for each of the seven assets, starting values, and depreciation rates as given in Table A.5, we calculate real stocks for each asset by industry. Note that the depreciation rate of 13% per annum for plant in Table A.5 is applied to “rest of plant” only. ICT assets like computers and software have higher depreciation rates.

*Rental prices and shares, by asset, by industry*

Rental prices for each asset in each industry are calculated via Equation (A.7). Asset prices are given with the nominal investment data (except for intangibles and plant where they are calculated as the ratio of nominal to real investment) and depreciation rates are as provided in Table A.5. Tax rates were kindly supplied by Rod Whitaker of H.M. Treasury and are assumed to be the same across industries. We assume in addition, that the tax rate on computers, software and telecoms is the same as that for plant. As mentioned above, the nominal rate of return is calculated from whole economy data and is thus assumed to be the same across all assets and all industries.

Using Equation (A.8) we multiply the rental prices by the real asset stocks for each asset in a particular industry, sum across assets and then divide the asset specific rental price times asset stock by the sum to obtain the rental price share of the asset. The rental prices shares, mostly for buildings and plant were quite volatile in the 74-80 period and have thus been smoothed.\(^{(14)}\)

The shares of ICT capital and non-ICT capital in profits have been calculated by applying Equation (A.8) to the assets in each category (e.g., computers, software and telecoms for ICT capital). These shares are then applied to the profits data to get the value of profits originating from that category.

We calculate Fisher indices of capital services for total, ICT, non-ICT and computers & software capital using the rental price shares and the real asset stocks for each industry (as in Equation (A.9)). The reason we use Fisher for the aggregation is that for many industries telecoms stocks are zero and for some industries for some years computer and software stocks are zero.\(^{(15)}\) These indices are converted into constant price series for capital services by setting the real capital service in the base year, 1995, equal to the nominal profits in that year.

\(^{(14)}\) Some rental price shares for vehicles and plant for other years have also been smoothed.
\(^{(15)}\) The Törnqvist formula in the growth rate calculations breaks down for those years when the asset stocks are zero.
5. Labour

Labour input is best measured by hours worked but it is necessary also to adjust for labour quality. The basic principle is to break out as many different types of labour as possible — distinguished eg by age, sex and qualifications — and to measure aggregate labour input as a weighted average of hours worked by each group. The weights should be the shares of each type of labour in the aggregate wage bill. This assumes that a version of marginal productivity theory holds: each type of labour is paid in proportion to its marginal product. \(^{(16)}\) Ideally, we would like to construct a chain index of labour input for each industry. For each industry, we would need data on hours worked by age, sex and qualifications. Unfortunately, this is not possible for the U.K. at the level of industry disaggregation that we require.

We rely on the employer-based surveys for head counts of number of people employed by industry. We use the New Earnings Survey (NES), also employer-based, for hours per worker, by industry. An alternative source for hours worked and also for qualifications is the Labour Force Survey (LFS), which is a survey of households. From our point of view, this suffers from two major drawbacks: first, the LFS goes back only to 1984; second, the distribution of the labour force across industries revealed by this survey matches very poorly with that given by the employer-based surveys. The employer-based surveys are considered the more reliable in this respect. However, the LFS is generally considered to give the best estimate of aggregate employment and hours worked. The NES also provides data on sex, age and occupation and we have tried to use these data to provide estimates of quality change at the industry level. Unfortunately, we have found that the NES provides a poor basis for this purpose: when the results are aggregated up to the whole economy level, we find them to be inconsistent with what we know from other sources (the LFS). So our indices in practice are just hours worked, ie an unweighted sum of hours worked by workers of different types. But we have made two aggregate level adjustments to the industry estimates. The first is to make the growth of aggregate hours consistent with the measure derived from the LFS (ONS code: YBUS). The second is to make use of an index of quality change constructed by colleagues in the Bank of England (Burriel-Llombart and Jones (2003)). Their index is for the whole economy and is a Törnqvist one encompassing the effects of changes in the composition of the labour force by age, sex and qualifications. We add their measure of the growth of labour quality to each industry’s growth rate of labour input; we also present results without the aggregate quality adjustment.
In summary, the basic strategy is to measure total annual hours worked in each industry as the number of workers in the industry (the head count) times the annual hours per worker in this industry. The head count covers both employees and the self-employed. Total hours worked in each industry are then adjusted so that the growth of estimated total hours worked in the whole economy matches an independent estimate of this growth rate. We also apply a quality adjustment based on aggregate-level data to each industry’s growth of labour input.

Sources for head counts

The most reliable measure of the number of employees at the industry level is provided by the regular employer-based surveys, formerly the Annual Employment Survey (AES), nowadays the Annual Business Inquiry (ABI). These provide head counts of the number of employees by industry for four categories of worker: male full time (FT), male part time (PT), female FT and female PT. The most detailed level of data is for Great Britain, i.e. the U.K. excluding Northern Ireland (which has 2.5% of the U.K. population). Data is available on a consistent industrial classification (the 1992 SIC) from 1978 to the present. The detailed data is quarterly, not seasonally adjusted, though some information is available monthly. In the first instance we extract these head counts of employees for the 49 CE industries. We obtain annual average employment levels as simple averages of the quarterly series. (17)

These surveys cover employees only and exclude the self-employed, who are a growing category. In 1971 11.1% of jobs held by males were self-employed; this proportion rose to 17.8% in 1997. The comparable figures for females were 5.1% in 1971 and 7.5% in 1997. (18) We estimate the self-employed as the difference between “workforce jobs” (WFJ) and employees. This calculation can be done for each sex separately but unfortunately for a breakdown into only nine broad sectors. We assume that the self-employed proportion is the same for all industries in a given sector. The Labour Force Survey (LFS) is the only source for data on self-employment. Self-employment is measured in accordance with respondents’ perceptions, which may differ from those of employers.

Sources for hours and quality

The AES/ABI surveys have no information on wages, hours, age or skills. But this deficiency can be made up to some extent by using the New Earnings Survey (NES). This is a compulsory survey, also

(16) It is not necessary to assume that wages are equal to marginal products. But if they differ, the factor of proportionality must be the same for all types. This would be the case for example for a firm which is a price-taker in input markets and is in monopolistic competition in the product market.

(17) These data are not published at this level of detail but are provided on request by the ONS.

(18) Source: Diskette labelled “Historical supplement No. 5: Workforce data back series”, published by the ONS.
employer-based, covering 1% of all employees in Great Britain (about 250,000 employees), and based on a random draw of National Insurance numbers. This survey asks about each worker’s pay packet and hours, the worker’s age and his or her occupation. The survey does not ask about educational qualifications. The data in the NES are a snapshot of a particular week in April. We have data from the NES for the years 1975 to 1999. Nowadays the NES asks about actual weekly hours as well as usual weekly hours, but actual hours only go back to 1991. Hence we use usual weekly hours. (19)

We experimented with using the NES data on occupation as a proxy for qualifications and hence as a measure of labour quality. In principle, we can measure weekly wages and weekly hours for each of our 49 industries by age, sex and occupation. But at this level of detail the sample sizes are too small, so we drop age in the belief that changes in the age distribution are likely to have only a minor effect on labour quality. This leaves potentially (2 \times 4 = ) eight types of labour for each industry. We estimate the total weekly hours of each labour type as the average usual weekly hours of each group times the numbers in each group. The number in each group is the sample proportion of each group (from the NES) times the headcount of all workers, by sex, after grossing up for self-employment; the head counts derive from the AES/ABI, as just described. (20) The weight to be applied to each group is its share of the industry wage bill, which can be estimated from the NES. But we found that the results were quite different from those obtained from the LFS, for the years where they overlap.

We next experimented with using just the data on sex composition, ie we now use just two types of labour for each industry, males and females. As expected, this led to an index of quality which fell at a modest rate for most of our period, which simply reflects the fact that women earn less than men and the female share of total hours has been rising. But near the end of our period, in 1998 and 1999, this quality index rose sharply, by 0.96% and 1.91% respectively. This behaviour seems implausible and does not match the evidence from the LFS. As a result, we abandoned the NES as a source for quality change and decided to use the LFS to make an aggregate level adjustment for quality (see below).

**Final method**

For the reasons just discussed, we estimate simple, unweighted indices of hours worked for each industry. These are just unweighted sums over all labour types of hours worked in each industry.

(19) We are very grateful to Glenda Quintini who supplied us with the NES data, broken down into our 49 industries, for the years 1975-1999. The big advantage of her data is that it achieves consistency over such a long period in the occupational and industrial classification systems, a non-trivial task.

(20) The NES hours are averages over all workers, full time and part time. So the head count for each industry is the sum of full time and part time workers.
Prior to calculating these indices, we aggregated the industries from the original 49 to 34, to match our other data.

Usual weekly hours need to be converted to actual annual hours. We start by multiplying weekly hours by 52, but this overstates actual annual hours because it makes no allowance for eg sickness, maternity, and holidays. In fact the number of weeks worked per year likely has a downward trend due to the rise of paid holidays and the increase in the number of public holidays. Also, actual hours worked are affected by the business cycle. We have no reliable measures of these factors at the industry level. We therefore apply an adjustment derived at the aggregate level to each industry (see below).

**Implementing the method**

The solution suggested for estimating labour input at the industry level was the following:

*Step 1.* Start with employer-survey-based measures of numbers of employees by sex and industry (ie the head count data formerly gathered by the AES and nowadays by the ABI).
*Step 2.* Add the self-employed (from the LFS), to yield what we call numbers employed.
*Step 3.* Multiply numbers employed by usual weekly hours per worker. Usual weekly hours per worker, together with the wage bill, are available for each sex for 1976-1999 from data supplied by Glenda Quintini. She derived these data from the NES (actual hours are only available from 1993).
*Step 4:* Apply an aggregate adjustment to convert usual to actual hours (using the LFS-based series for total hours, YBUS).
*Step 5.* Apply an aggregate adjustment for quality based on data from the LFS (the measure derived by Burriel-Llombart and Jones (2003)).

Steps 1-3 generate for each of the 34 industries a measure of labour input which is a simple sum of hours worked by males and females, initially for 1978-1999. Our head count data from the AES/ABI go back to 1978, while our NES data go back to 1975. We backcast the head count series for 1969-77 using data for the growth rates of employees supplied by Cambridge Econometrics for these years. For the years 1969-74, we have no NES data on hours so we assume that usual hours grew at the same rate as employment.

For the year 2000, we have no NES-based data from Quintini so we assume that total usual weekly hours in each industry grew at the same rate as employment (employees plus self-employed), where employment is now from the AES survey.
Step 4: We calculate the growth of aggregate hours by aggregating up the simple industry-level indices of hours worked. We compare this with the growth of total weekly hours from the LFS (ONS code: YBUS). We then add the difference between the growth rates of YBUS and of our own estimate of aggregate hours to the growth of hours in each industry. Note that though YBUS is a measure of weekly hours, it is an average over people who actually worked and those who were absent for some reason (sickness, maternity, holidays, etc). So this procedure does capture differences between actual and usual hours.\(^{(21)}\)

Insofar as the head count varies over the business cycle, then our labour input measures already make some allowance for business cycle factors. The YBUS adjustment is the main factor adding an element of variation in hours per worker over the cycle. So it is important to note that by construction it is \textit{the same for all industries}.

Step 5: The aggregate quality adjustment is made using the index of labour quality developed by Burriel-Llombart and Jones (2003). Their Törnqvist index of labour input uses five age groups, five education groups and two sexes, ie 50 groups in all and runs from 1975-2000. It is based on qualifications and wages data from the LFS for 1985 onwards, and for the years 1975-84 on qualifications data from the General Household Survey. Quality growth is estimated as the growth of quality adjusted hours minus the growth of an unweighted index of hours. We set quality growth to zero for 1970-1975.

Table A.6 shows average annual growth rates of the aggregate measures: labour input, hours, quality and the YBUS adjustment to aggregate hours; see chart A.2 for the course of labour quality. The YBUS hours adjustment reduces the annual growth of labour input but only by 0.02 percentage points per annum, though this masks a some variation over the cycle: see chart A.3. The adjustment has been predominantly positive in the 1980s and predominantly negative in the 1990s, particularly so in 1992.

\(^{(21)}\) The published series for YBUS goes back only to 1993. For the earlier years, we rely on an internal Bank estimate extending YBUS back to 1969. This in turn was based on O’Mahony (1999).
6. Intermediate input

Nominal input
For each of our original 49 industries, we have the purchases of the products of each of the 49 industries from domestic sources and separately from imports. That is, for each year we have a 49 x 49 matrix of domestic purchases and a 49 x 49 matrix of purchases from imports.

The domestic/imports split was done using the input-output tables. For years when the I-O SUTs are not available but an input-output table exists, the latter was employed to produce estimates of intermediate input. The tables were first adjusted to conform to revised estimates of nominal GDP and nominal value added in broad sectors and aggregated up to the 49 industry level. For intervening years between input-output tables, interpolation was used.

Real input
For each product, we have a domestic price index and an import price index. These derive from the ONS’s detailed Producer Price Indices. We use them to deflate purchases. For each industry, we construct a Fisher index of aggregate domestic intermediate input and a Fisher index of aggregate imported intermediate input. Then we construct a Fisher index of these two components to arrive finally at our index of real intermediate input. We use Fisher rather than Törnqvist indices since at this level of detail many of the cells in our purchases matrices are zeroes. A Törnqvist index cannot be calculated in these cases but no such problem arises for a Fisher index.

The resulting indices turned out to be implausibly volatile. For example, over the period 1979-2000, the average across the 49 industries of the standard deviation of the growth rate of intermediate input was 20.3% per annum. This was much higher than the average volatility of real value added. This excessive volatility seems to be due not to particularly high volatility of price inflation, nor to volatility of the individual components of the domestic and import use matrices, but rather to high volatility of the total of nominal purchases by each industry from year to year, at least for the period prior to 1992. Recall that prior to 1992 we do not have the I-O SUTs to rely on. So it is possible that the volatility of intermediate input is an artefact of the estimation process. We therefore decided to smooth the growth rates of the Fisher indices of intermediate input in the following, sequential way.

1. Outliers are clipped. If a growth rate exceeds +20% pa, it is set equal to +20%. If it is less than -20% pa, it is set equal to -20%.
2. If a growth rate is outside the bounds of the mean plus or minus 2 standard deviations (calculated for the period 1979-2000), then it is set equal to the mean.

3. A two-year, moving average of the growth rates is calculated.

Finally, we obtain indices of total intermediate input at the 34 industry level as Fisher indices of the indices at the 49 industry level.

7. **Total factor productivity**

We calculate TFP growth in each industry as the growth of real gross output minus the growth of a Törnqvist index of total input. Total input comprises capital, labour and intermediate. The shares of the three inputs are the payments made to each input as a proportion of nominal gross output (excluding taxes on production). These are taken from the I-O SUTs for the years 1992-2000. For earlier years the intermediate share derives from our annual series on gross output, intermediate input and value added. To derive the share of capital in 1969-1991, we employ the 1992 share of profit in value added but adjusted so that industry profits sum to the national accounts total (this picks up any cyclical variability in the profit share at the aggregate level).

The share of ICT capital services in value added is derived as the share of ICT in total capital services times the share of capital in value added. The share of ICT in total capital services is estimated as the share of profits attributable to ICT in total profits, where profit attributable to each ICT asset is the rental price of that asset multiplied by the nominal value of the stock (see above).

The labour share in value added is derived as one minus the capital share. The labour and capital shares in value added are then converted to shares in gross output.
References


Table A.1
The original 49 industries (CE 1 - CE 49)

<table>
<thead>
<tr>
<th>Industry</th>
<th>SIC92</th>
</tr>
</thead>
<tbody>
<tr>
<td>1 Agriculture</td>
<td>01.02.05</td>
</tr>
<tr>
<td>2 Coal</td>
<td>10</td>
</tr>
<tr>
<td>3 Oil &amp; Gas etc</td>
<td>11.12</td>
</tr>
<tr>
<td>4 Other Mining</td>
<td>13.14</td>
</tr>
<tr>
<td>5 Food</td>
<td>15.1-15.8</td>
</tr>
<tr>
<td>6 Drink</td>
<td>15.9</td>
</tr>
<tr>
<td>7 Tobacco</td>
<td>16</td>
</tr>
<tr>
<td>8 Textiles</td>
<td>17</td>
</tr>
<tr>
<td>9 Clothing &amp; Leather</td>
<td>18.19</td>
</tr>
<tr>
<td>10 Wood &amp; Wood Products</td>
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<tr>
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<td>32 Distribution nes</td>
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<td>55</td>
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<td>60.1</td>
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<td>60.2, 60.3</td>
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<td>38 Other Transport Services</td>
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</tr>
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<td>39 Communications</td>
<td>64</td>
</tr>
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<td>40 Banking &amp; Finance</td>
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</tr>
<tr>
<td>41 Insurance</td>
<td>66</td>
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<td>46 Education</td>
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<tr>
<td>47 Health &amp; Social Work</td>
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<tr>
<td>49 Miscellaneous Services</td>
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</table>

Note  SIC92 is the 1992 version of the U.K.’s Standard Industrial Classification. It is identical to the European NACE system. Details on SIC92 industry codes can be found at http://www.statistics.gov.uk/methods_quality/sic/contents.asp.
<table>
<thead>
<tr>
<th>Industry</th>
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<tbody>
<tr>
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</tr>
<tr>
<td>2 Oil and gas</td>
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</tr>
<tr>
<td>3 Coal &amp; other mining</td>
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<tr>
<td>4 Manufactured fuel</td>
<td>23</td>
</tr>
<tr>
<td>5 Chemicals &amp; pharmaceuticals</td>
<td>24</td>
</tr>
<tr>
<td>6 Non-metallic mineral products</td>
<td>26</td>
</tr>
<tr>
<td>7 Basic metals &amp; metal goods</td>
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</tr>
<tr>
<td>8 Mechanical engineering</td>
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</tr>
<tr>
<td>9 Electrical engineering &amp; electronics</td>
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<tr>
<td>10 Vehicles</td>
<td>34,35</td>
</tr>
<tr>
<td>11 Food, drink &amp; tobacco</td>
<td>15,16</td>
</tr>
<tr>
<td>12 Textiles, clothing &amp; leather</td>
<td>17,18,19</td>
</tr>
<tr>
<td>13 Paper, printing and publishing</td>
<td>21,22</td>
</tr>
<tr>
<td>14 Other manufacturing</td>
<td>20,25,36,37</td>
</tr>
<tr>
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<td>40.1</td>
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<tr>
<td>16 Gas supply</td>
<td>40.2,40.3</td>
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<tr>
<td>17 Water supply</td>
<td>41</td>
</tr>
<tr>
<td>18 Construction</td>
<td>45</td>
</tr>
<tr>
<td>19 Wholesale, vehicle sales &amp; repairs</td>
<td>50,51</td>
</tr>
<tr>
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</tr>
<tr>
<td>21 Hotels &amp; catering</td>
<td>55</td>
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<tr>
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<td>27 Communications</td>
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</tr>
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<td>65, 66</td>
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<tr>
<td>29 Business Services</td>
<td>67, 70,71,72,73,74</td>
</tr>
<tr>
<td>30 Public administration and defence</td>
<td>75</td>
</tr>
<tr>
<td>31 Education</td>
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</tr>
<tr>
<td>32 Health and social work</td>
<td>85</td>
</tr>
<tr>
<td>33 Waste treatment</td>
<td>90</td>
</tr>
<tr>
<td>34 Miscellaneous services</td>
<td>91-99</td>
</tr>
</tbody>
</table>

Note  
SIC92 is the 1992 version of the U.K.'s Standard Industrial Classification. It is identical to the European NACE system. Details on SIC92 industry codes can be found at http://www.statistics.gov.uk/methods_quality/sic/contents.asp.
Table A.3
Mapping between 34 industries and sectors used in Basu, Fernald, Oulton and Srinivasan (2003)

<table>
<thead>
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<td>✓</td>
<td>✓</td>
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<td>✓</td>
<td>Manufacturing (non-durable)</td>
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<td>✓</td>
<td>✓</td>
<td>Manufacturing (non-durable)</td>
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<td>✓</td>
<td>✓</td>
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</tr>
<tr>
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<td>✓</td>
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<tr>
<td>8 Mechanical engineering</td>
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<td>✓</td>
<td>✓</td>
<td>Manufacturing (durable)</td>
</tr>
<tr>
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<td>✓</td>
<td>✓</td>
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<td>✓</td>
<td>Manufacturing (durable)</td>
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<td>✓</td>
<td>✓</td>
<td>Manufacturing (non-durable)</td>
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<td>✓</td>
<td>✓</td>
<td>Manufacturing (non-durable)</td>
</tr>
<tr>
<td>14 Other manufacturing</td>
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<td>✓</td>
<td>✓</td>
<td>Manufacturing (non-durable)</td>
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<td>✓</td>
<td>Electric/Gas/Sanitary</td>
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<td>✓</td>
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<td>✓</td>
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<td>✓</td>
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<td>✓</td>
<td>✓</td>
<td>Retail Trade</td>
</tr>
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<td>Transportation</td>
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<td>32 Health and social work</td>
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<td></td>
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</tr>
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<td>✓</td>
<td>✓</td>
<td>Electric/Gas/Sanitary</td>
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<tr>
<td>34 Miscellaneous services</td>
<td>✓</td>
<td>✓</td>
<td>✓</td>
<td>Other Services</td>
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</table>
Table A.4

Alternative measures of GDP growth, % p.a.

<table>
<thead>
<tr>
<th></th>
<th>49 industries (gdp49)</th>
<th>34 industries (gdp34)</th>
<th>Aggregate (ONS) (abmmxh)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1969-1979</td>
<td>1.80</td>
<td>2.52</td>
<td>2.15</td>
</tr>
<tr>
<td>1979-1990</td>
<td>2.26</td>
<td>2.70</td>
<td>2.29</td>
</tr>
<tr>
<td>1990-2000</td>
<td>1.98</td>
<td>2.58</td>
<td>2.40</td>
</tr>
<tr>
<td>1990-1995</td>
<td>1.06</td>
<td>1.88</td>
<td>1.86</td>
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<tr>
<td>1995-2000</td>
<td>2.91</td>
<td>3.28</td>
<td>2.93</td>
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</table>

Notes See text.

Table A.5

Asset level depreciation rates and investment price deflators

<table>
<thead>
<tr>
<th>Asset</th>
<th>Depreciation Rate (% per annum)</th>
<th>Investment Price Deflator</th>
</tr>
</thead>
<tbody>
<tr>
<td>1. Other buildings and structures</td>
<td>2.5</td>
<td>UK: Industry specific</td>
</tr>
<tr>
<td>2. Other machinery and equipment and cultivated assets (&quot;plant&quot;)</td>
<td>13.0</td>
<td>UK: Industry specific</td>
</tr>
<tr>
<td>3. Transport equipment (&quot;vehicles&quot;)</td>
<td>25.0</td>
<td>UK: Industry specific</td>
</tr>
<tr>
<td>4. Intangible fixed assets</td>
<td>13.0</td>
<td>UK: Industry specific (^{22})</td>
</tr>
<tr>
<td>5. Computers</td>
<td>31.5</td>
<td>US: common for all industries (^{23})</td>
</tr>
<tr>
<td>6. Software</td>
<td>31.5</td>
<td>US: common for all industries (^{24})</td>
</tr>
<tr>
<td>7. Telecommunications equipment</td>
<td>11.0</td>
<td>UK: common for all industries (ONS code: PQGT)</td>
</tr>
</tbody>
</table>

\(^{22}\) As explained in the text, for all industries, except Oil and gas (Industry 2) and Miscellaneous services (Industry 34), there is zero intangibles investment once software investment has been accounted for. For industries 1 and 3-33, the software deflator is used as the intangibles deflator. For the two industries that that have non-zero intangibles investment, net of software, we use industry specific deflators.

\(^{23}\) Source: US Bureau of Economic Analysis, NIPA Tables, Table 7.6, translated into pounds using the sterling exchange rate (ONS code: AJFA).

\(^{24}\) Source: US Bureau of Economic Analysis, NIPA Tables, Table 7.6, translated into pounds using the sterling exchange rate (ONS code: AJFA).
Table A.6
Mean growth rates of aggregate total hours, quality, and the YBUS adjustment to aggregate hours, % p.a.

<table>
<thead>
<tr>
<th></th>
<th>Total hours (not quality-adjusted)</th>
<th>Quality</th>
<th>Total hours (quality-adjusted)</th>
<th>YBUS adjustment</th>
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<tbody>
<tr>
<td>1979-2000</td>
<td>0.11</td>
<td>0.84</td>
<td>0.96</td>
<td>-0.02</td>
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<td>1979-1990</td>
<td>0.55</td>
<td>0.65</td>
<td>1.20</td>
<td>0.21</td>
</tr>
<tr>
<td>1990-2000</td>
<td>-0.36</td>
<td>1.05</td>
<td>0.69</td>
<td>-0.28</td>
</tr>
<tr>
<td>1990-1995</td>
<td>-1.55</td>
<td>1.33</td>
<td>-0.22</td>
<td>0.01</td>
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<tr>
<td>1995-2000</td>
<td>0.83</td>
<td>0.78</td>
<td>1.61</td>
<td>-0.56</td>
</tr>
</tbody>
</table>

Note  Total hours (both quality-adjusted and not quality-adjusted) are after making the YBUS adjustment.

Chart A.1

Alternative measures of GDP growth, % p.a.
Chart A.2

Aggregate hours adjustment, % p.a.

Chart A.3

Growth of labour quality
(Burriel-Llombart and Jones measure)