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Estimating Flexible Functional Forms using Macroeconomic Data

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#### Abstract

The paper estimates a flexible functional form for a joint cost function using US aggregate data for the years 1970-2022. There are four outputs (consumption, government, investment and exports) and six inputs (imports, labour, machinery and equipment services, structure services, other capital services and land services). Curvature conditions on the joint cost function are imposed without destroying the flexibility of the functional form. Various elasticities of supply and demand are estimated along with technical progress bias terms for each input. When using aggregate time series data based on the System of National Accounts, the paper shows that it is probably better to estimate a joint cost function rather than a gross output function or a GDP function. The paper also shows that assuming that an aggregate production function has constant elasticities of substitution is not appropriate for the US. Finally, the importance of including land as an aggregate input is stressed.

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# Estimating Flexible Functional Forms using Macroeconomic Data 

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#### Abstract

The paper estimates a flexible functional form for a joint cost function using US aggregate data for the years 1970-2022. There are four outputs (consumption, government, investment and exports) and six inputs (imports, labour, machinery and equipment services, structure services, other capital services and land services). Curvature conditions on the joint cost function are imposed without destroying the flexibility of the functional form. Various elasticities of supply and demand are estimated along with technical progress bias terms for each input. When using aggregate time series data based on the System of National Accounts, the paper shows that it is probably better to estimate a joint cost function rather than a gross output function or a GDP function. The paper also shows that assuming that an aggregate production function has constant elasticities of substitution is not appropriate for the US. Finally, the importance of including land as an aggregate input is stressed.


## Keywords

Production theory, duality theory, production functions, joint cost functions, gross output functions, GDP functions, Shephard's Lemma, Hotelling's Lemma, Samuelson's Lemma, flexible functional forms, estimation of technical progress, user costs, modeling monopolistic behavior, land as a factor of production.

JEL Classification Numbers
C01, C02, C32, C43, C51, C82, D24, D42, E01, E23, F11, O47

[^0]
## 1 Introduction

It is important to be able to estimate production functions for a large number of policy purposes. If we attempt to estimate a twice continuously differentiable production function by regressing an output of a production unit on other outputs and inputs used by the production unit over a time period, we run into multicollinearity problems if the number of other outputs and inputs is large. If we want the estimated production function to be able to provide a second order Taylor series approximation to the true production function, we require a large number of parameters and we soon run into degrees of freedom problems. A partial solution to these problems is to assume that the production unit is either minimizing the cost of producing a vector of outputs (which leads to a joint cost function) or maximizing profits subject to one or more inputs being fixed (which leads to a gross output function if all inputs are "fixed") or leads to a GDP function (if all primary inputs are "fixed"). The assumption of optimizing behavior on the part of the production unit leads to many additional estimating equations and thus helps to overcome the degrees of freedom problem. Moreover, the additional estimating equations can be obtained by simply differentiating the joint cost function or profit function which is defined by the form of optimization. Thus the search for functional forms for joint cost and gross output functions that can provide second order approximations to arbitrary differentiable joint cost and gross output functions began about 50 years ago. ${ }^{* 1}$
These dual representations of technology sets are important for policy purposes since they lead to estimates of technical change and measures of the biases in technical progress. They also lead to estimates for elasticities of output supply and for input demand. These elasticities are useful for a wide range of purposes. Index number methods are available for measuring Total Factor Productivity and technical progress, as are nonparametric methods, but these methods cannot estimate elasticities or biases in technical change.
The early literature on finding flexible functional forms for joint cost functions or gross output functions found that the estimated functional forms did not satisfy the curvature conditions that optimizing behavior imposes on the dual representations of technology. For example, a joint cost function, $C(\boldsymbol{y}, \boldsymbol{w})$ where $\boldsymbol{y}$ is a vector of outputs that the production unit produces and $\boldsymbol{w}$ is a vector of input prices that it faces, must be a concave function in input prices $\boldsymbol{w}$. Similarly, a gross output function, $G(\boldsymbol{p}, \boldsymbol{x})$ where $\boldsymbol{p}$ is vector of output prices that the production unit faces and $\boldsymbol{x}$ is a vector of input quantities, must be convex in output prices $\boldsymbol{p}$. In this paper, we will adapt the Normalized Quadratic functional form used by Diewert and Wales (1987)[23] (1992)[25] to construct functions $C(\boldsymbol{y}, \boldsymbol{w})$ and $G(\boldsymbol{p}, \boldsymbol{x})$ that are flexible and satisfy curvature conditions. ${ }^{* 2}$ In section 2 below, we will define these functions and derive alternative sets of estimating equations that can be used to estimate technologies.
In section 3, we restrict our attention to the estimation of aggregate production functions and technologies using macroeconomic data that are available from the national accounts of a country. We show that using aggregate data, it is better (in terms of fitting the data) to use a Normalized Quadratic Joint Cost Function rather than using a Normalized Quadratic Gross Output Function. Appendix B provides a proof of the flexibility of the NQ Joint Cost Function.

[^1]Section 4 estimates a Normalized Quadratic Joint Cost Function for the US using aggregate data for the years 1970-2022. We have the usual $C+G+I+X$ macro aggregates as our 4 outputs and we constructed data for 6 inputs: $M$ (imports), $L$ (quality adjusted labour), and 4 types of capital services, $K_{\mathrm{M} \mathrm{\& E}}, K_{\mathrm{S}}, K_{\mathrm{O}}$ and $K_{\mathrm{L}}$ (Machinery and Equipment, Structures, Other Capital (mainly R\&D and Inventories) and Land). The data are described more fully in Appendix A. We show that land is an important input*3 and that the aggregate US production function is very far from being a Cobb-Douglas or CES function. We also show that just using linear trends in each estimating equation to approximate the effects of technological change is a poor approximation.
Section 5 lists our econometric estimates for technical progress and for biases in technical change. Our technical progress estimates can be compared with index number methods for computing Total Factor Productivity growth. The Diewert and Morrison (1986)[22] and Kohli (1990)[39] (2003)[43] TFP growth accounting estimates for TFP are listed at the end of Appendix A. The Diewert and Fox (2018)[19] Nonparametric estimates of US TFP are explained and listed in Appendix C and compared with the econometric estimates for technical progress. ${ }^{* 4}$ Section 5 also lists the average price elasticities of input demand for our 6 aggregate inputs and the average elasticities of inverse output supply over 1970-2022.
For some purposes, it is useful to switch to estimating a GDP function or a Gross Output function or a Variable Profit function where only a few inputs are fixed and all outputs and variable inputs are free to vary. It turns out to be possible to get elasticity estimates for these alternative functions using our estimated joint cost function. We show how this can be done in Appendix D.
Section 6 lists some possible extensions (and possible problems) with our models and Section 7 concludes.

## 2 Alternative Representations for Production Possibilities Sets

Consider a production unit that can produce $M$ outputs using $N$ inputs over a period of time, which we take to be a year. Let $\boldsymbol{x} \equiv\left[x_{1}, \ldots, x_{N}\right] \geq \mathbf{0}_{N}$ be a nonnegative vector of annual inputs and let $\boldsymbol{y} \equiv\left[y_{1}, \ldots, y_{M}\right] \geq \mathbf{0}_{M}$ be a nonnegative vector of annual outputs. Define the production possibilities set $S$ to be the set of output vectors $\boldsymbol{y}$ that could be produced by the technology given the availability of the vector of inputs $\boldsymbol{x}$. Thus $(\boldsymbol{y}, \boldsymbol{x}) \in S$ if the output vector $\boldsymbol{y}$ and be produced by the input vector $\boldsymbol{x}$. We assume that $S$ is a nonempty, convex, closed cone ${ }^{* 5}$ that exhibits free disposal of inputs and outputs.
We consider alternative ways for estimating the technology set $S$ using time series data. Suppose we have information on outputs produced and inputs used for year $t,\left[\boldsymbol{y}^{t}, \boldsymbol{x}^{t}\right] \equiv$ $\left[y_{1}^{t}, \ldots, y_{M}^{t}, x_{1}^{t}, \ldots, x_{N}^{t}\right]$ for $t=1, \ldots, T$. We could attempt to estimate a production function $f$ which gives the maximum amount of output $1 y_{1}^{t}$ that could be produced conditional on having available the vector of year $t$ inputs $\boldsymbol{x}^{t}$ and conditional on producing the amounts

[^2]$y_{2}^{t}, y_{3}^{t}, \ldots, y_{M}^{t}$ of other outputs:
\[

$$
\begin{equation*}
y_{1}^{t}=f\left(y_{2}^{t}, y_{3}^{t}, \ldots, y_{M}^{t}, x_{1}^{t}, \ldots, x_{N}^{t}\right) ; \quad t=1, \ldots, T . \tag{1}
\end{equation*}
$$

\]

If $f$ is twice continuously differentiable with respect to its arguments, then we would like $f$ to be a flexible functional form; i.e., we would like $f$ to be able to provide a second order Taylor series approximation to an arbitrary twice continuously differentiable production function that satisfies the appropriate regularity conditions. ${ }^{* 6}$ In the present context where we have assumed constant returns to scale in production, a candidate function to be a flexible functional form would require at least $(M+N)(M+N-1) / 2$ free parameters. If $M+N=10$, a flexible functional form for a production function would have to have at least 45 parameters and so we would require over 45 years of data on the outputs and inputs produced by the production unit.
In order to gain more degrees of freedom to enable the econometric estimation of a technology, it is necessary to assume some form of optimizing behavior on the part of the producer. We consider some alternative assumptions below.
Suppose the production unit faces the strictly positive output price vector $\boldsymbol{p} \equiv\left[p_{1}, \ldots, p_{M}\right]$ and has the strictly positive input quantity vector $\boldsymbol{x} \equiv\left[x_{1}, \ldots, x_{N}\right]$ at its disposal. Let $\boldsymbol{y} \equiv\left[y_{1}, \ldots, y_{M}\right]$ be a nonnegative output quantity vector and let $S$ be the production unit's production possibilities set that satisfies the above regularity conditions. The gross output function, $G(\boldsymbol{p}, \boldsymbol{x})$ for this production unit is defined as follows:*7

$$
\begin{equation*}
G(\boldsymbol{p}, \boldsymbol{x}) \equiv \max _{y}\{\boldsymbol{p} \cdot \boldsymbol{y}:(\boldsymbol{y}, \boldsymbol{x}) \in S\} . .^{* 8} \tag{2}
\end{equation*}
$$

Thus $G(\boldsymbol{p}, \boldsymbol{x})$ is the maximum revenue the production unit can generate if it faces output prices $\boldsymbol{p}$ and uses the input vector $\boldsymbol{x}$ to produce the revenue maximizing output $\boldsymbol{y}(\boldsymbol{p}, \boldsymbol{x})$ which solves the constrained optimization problem (2). If intermediate inputs are included in the vector $\boldsymbol{y}$ (indexed by negative signs), then $G(\boldsymbol{p}, \boldsymbol{x})$ is a value added function or at the national level, it is a GDP function. The properties of this function were studied by Samuelson (1953)[54], McFadden (1966)[47] (1978)[48], Diewert (1973)[10] (1974a)[11] (1974b)[12] (2022)[17] and others. We assume constant returns to scale in production so that $G(\boldsymbol{p}, \boldsymbol{x})$ is linearly homogenous in $\boldsymbol{p}$ for fixed $\boldsymbol{x}$ and is linearly homogeneous in $\boldsymbol{x}$ for fixed $\boldsymbol{p}$. If $G(\boldsymbol{p}, \boldsymbol{x})$ is differentiable at a point $(\boldsymbol{p}, \boldsymbol{x})$, Hotelling's Lemma (1932; 594)[28] implies that the vector of output supply functions regarded as functions of $\boldsymbol{p}$ and $\boldsymbol{x}, \boldsymbol{y}(\boldsymbol{p}, \boldsymbol{x})$, can be obtained by differentiating $G(\boldsymbol{p}, \boldsymbol{x})$ with respect to the components of $\boldsymbol{p}$ :

$$
\begin{equation*}
\boldsymbol{y}(\boldsymbol{p}, \boldsymbol{x})=\nabla_{p} G(\boldsymbol{p}, \boldsymbol{x}) . \tag{3}
\end{equation*}
$$

Suppose further that the production unit faces the strictly positive input price vector $\boldsymbol{w} \equiv$ $\left[w_{1}, \ldots, w_{N}\right]$. Samuelson's Lemma (1953; 10) $[57]^{* 9}$ implies that the producer's system of inverse input demand functions regarded as functions of $\boldsymbol{p}$ and $\boldsymbol{x}, \boldsymbol{w}(\boldsymbol{p}, \boldsymbol{x})$, can be obtained by differentiating $G(\boldsymbol{p}, \boldsymbol{x})$ with respect to the components of $\boldsymbol{x}$ :

$$
\begin{equation*}
\boldsymbol{w}(\boldsymbol{p}, \boldsymbol{x})=\nabla_{x} G(\boldsymbol{p}, \boldsymbol{x}) . \tag{4}
\end{equation*}
$$

[^3]Instead of conditioning on output prices $\boldsymbol{p}$ and input quantities $\boldsymbol{x}$, we can condition on output quantities $\boldsymbol{y}$ and input prices $\boldsymbol{w}$. Define the production unit's joint cost function, $C(\boldsymbol{y}, \boldsymbol{w})$, as the minimum cost of producing a given output vector $\boldsymbol{y}$ :

$$
\begin{equation*}
C(\boldsymbol{y}, \boldsymbol{w}) \equiv \min _{x}\{\boldsymbol{w} \cdot \boldsymbol{x}:(\boldsymbol{y}, \boldsymbol{x}) \in S\} \tag{5}
\end{equation*}
$$

Under our strong regularity conditions on the set $S$, it can be shown that $C(\boldsymbol{y}, \boldsymbol{w})$ is linearly homogeneous in the components of $\boldsymbol{y}$ holding $\boldsymbol{w}$ constant and is linearly homogeneous in the components of $\boldsymbol{w}$ holding $\boldsymbol{y}$ constant. It is also a convex function in the components of $\boldsymbol{y}$ holding $\boldsymbol{w}$ fixed and a concave function in the components of $\boldsymbol{w}$ holding $\boldsymbol{y}$ fixed. ${ }^{* 10}$
If $C(\boldsymbol{y}, \boldsymbol{w})$ is differentiable at a point $\boldsymbol{y}, \boldsymbol{w}$, then differentiating $C(\boldsymbol{y}, \boldsymbol{w})$ with respect to the components of $\boldsymbol{y}$ will generate the vector of marginal costs. If the producer takes output prices as being fixed and there are competitive markets, then this vector of marginal costs will be equal to the vector of selling prices $\boldsymbol{p}$. Thus the production unit's system of inverse output supply functions, $\boldsymbol{p}(\boldsymbol{y}, \boldsymbol{w})$, can be obtained by differentiating $C(\boldsymbol{y}, \boldsymbol{w})$ with respect to the components of $\boldsymbol{y}$ :

$$
\begin{equation*}
\boldsymbol{p}(\boldsymbol{y}, \boldsymbol{w})=\nabla_{y} C(\boldsymbol{y}, \boldsymbol{w}) \tag{6}
\end{equation*}
$$

Shephard's Lemma (1953; 11)[57] (1970)[58] implies that the production unit's system of input demand functions regarded as functions of $\boldsymbol{y}$ and $\boldsymbol{w}, \boldsymbol{x}(\boldsymbol{y}, \boldsymbol{w})$, can be obtained by differentiating $C(\boldsymbol{y}, \boldsymbol{w})$ with respect to the components of $\boldsymbol{w}$ :

$$
\begin{equation*}
\boldsymbol{x}(\boldsymbol{y}, \boldsymbol{w})=\nabla_{w} C(\boldsymbol{y}, \boldsymbol{w}) \tag{7}
\end{equation*}
$$

Thus we have two alternative methods for estimating a representation of the technology set $S$ : the first representation assumes a functional form for the gross output function, $G(\boldsymbol{p}, \boldsymbol{x})$, and uses equations (3) and (4) as estimating equations and the second representation assumes a functional form for the joint cost function, $C(\boldsymbol{y}, \boldsymbol{w})$, and uses equations (6) and (7) as estimating equations. Note that both representations lead to $(M+N) T$ estimating equations if we have data on prices and quantities for $T$ periods. This greatly increases degrees of freedom compared to the degrees of freedom that are available when estimating a production function.
In the following section, we ask whether we can choose between (3) plus (4) or (6) plus (7) when working with macroeconomic data.

## 3 Should We Estimate Gross Output Functions or Joint Cost Functions?

Consider the following specialization of the general joint cost function $C(\boldsymbol{y}, \boldsymbol{w})$ defined by (5):

$$
\begin{equation*}
C(\boldsymbol{y}, \boldsymbol{w}) \equiv \sum_{m=1}^{M} c^{m}(\boldsymbol{w}) y_{m} . \tag{8}
\end{equation*}
$$

The function $c^{m}(\boldsymbol{w})$ is the unit cost function that is dual to the single output constant returns to scale production function for sector $m, y_{m}=f^{m}\left(\boldsymbol{x}^{m}\right)$ for $m=1, \ldots, M$ where $\boldsymbol{x}^{m}$ is the vector of inputs used by sector $m$. This is the small open country framework considered by

[^4]Samuelson (1953)[54] in his seminal paper. It is sometimes called the Nonjoint Production Model: each sector of the economy produces only a single product (or a single group of products that does not overlap with the products produced in other sectors) and it does not use the products of other sectors as intermediate inputs. Note that the aggregate input vector $\boldsymbol{x}$ is equal to $\sum_{m=1}^{M} \boldsymbol{x}^{m}$ and the $m$ th output price is equal to $p_{m}=c^{m}(\boldsymbol{w})$ for $m=1, \ldots, M$. If each unit cost function is differentiable, then by applying Shephard's Lemma to each sectoral cost function, we can deduce that $\boldsymbol{x}^{m}=\nabla_{w} c^{m}(\boldsymbol{w}) y_{m}$ for $m=1, \ldots, M$. Making use of these equalities, differentiate both sides of (8) with respect to the components of $\boldsymbol{w}$. We obtain:

$$
\begin{equation*}
\nabla_{w} C(\boldsymbol{y}, \boldsymbol{w})=\sum_{m=1}^{M} \nabla_{w} c^{m}(\boldsymbol{w}) y_{m}=\sum_{m=1}^{M} \boldsymbol{x}^{m} \equiv \boldsymbol{x} \tag{9}
\end{equation*}
$$

Now further specialize the unit cost functions $c^{m}(\boldsymbol{w})$ to be linear functions of $\boldsymbol{w}$ :

$$
\begin{equation*}
c^{m}(\boldsymbol{w}) \equiv \sum_{n=1}^{N} w_{n} d_{n m} ; \quad m=1, \ldots, M \tag{10}
\end{equation*}
$$

where the $d_{n m}$ are $N M$ constants. This means that the sectoral production functions are Leontief (no substitution) production functions. Define the $N$ by $M$ matrix of the $d_{n m}$ as $\mathbf{D} \equiv\left[d_{n m}\right]$. Substitute definitions (10) into (8) and we obtain the following expression for the overall joint cost function:

$$
\begin{equation*}
C(\boldsymbol{y}, \boldsymbol{w}) \equiv \sum_{m=1}^{M} \sum_{n=1}^{N} w_{n} d_{n m} y_{m}=\boldsymbol{w} \cdot \mathbf{D} \boldsymbol{y} \tag{11}
\end{equation*}
$$

Thus the joint cost function for this very special case of Leontief sectoral production functions turns out to be a bilinear form in the vectors of input prices $\boldsymbol{w}$ and of output quantities $\boldsymbol{y}$. Equations (6) and (7) for this special case turn out to be the following estimating equations for $t=1, \ldots, T$ :

$$
\begin{align*}
& \boldsymbol{p}^{t}=\mathbf{D}^{T} \boldsymbol{w}^{t}  \tag{12}\\
& \boldsymbol{x}^{t}=\mathbf{D} \boldsymbol{y}^{t} \tag{13}
\end{align*}
$$

where $\boldsymbol{p}^{t}, \boldsymbol{w}^{t}, \boldsymbol{y}^{t}$ and $\boldsymbol{x}^{t}$ are the period $t$ vectors of observed prices and quantities and $\mathbf{D}^{T}$ is the transpose of the matrix $\mathbf{D}$. The estimating equations defined by (12) and (13) are linear in the unknown $M N$ parameters but of course, there are cross equation equality restrictions. However, this model can readily be estimated using standard nonlinear regression packages. Consider the following specializations of the general gross output function $G(\boldsymbol{p}, \boldsymbol{x})$ defined by (2):

$$
\begin{align*}
& G(\boldsymbol{p}, \boldsymbol{x}) \equiv \sum_{m=1}^{M} g^{m}(\boldsymbol{x}) p_{m}  \tag{14}\\
& G(\boldsymbol{p}, \boldsymbol{x}) \equiv \sum_{n=1}^{N} h^{n}(\boldsymbol{p}) x_{n}  \tag{15}\\
& G(\boldsymbol{p}, \boldsymbol{x}) \equiv \sum_{m=1}^{M} \sum_{n=1}^{N} x_{n} e_{n m} p_{m}=\boldsymbol{x} \cdot \mathbf{E} \boldsymbol{p} \tag{16}
\end{align*}
$$

where $\mathbf{E} \equiv\left[e_{n m}\right]$ is an $N$ by $M$ matrix of constants. The production model defined by (14) implies that output $y_{m}$ is equal to the function $g^{m}(\boldsymbol{x})$ of aggregate input $\boldsymbol{x}$ for each output $m=1, \ldots, M$. In the context where outputs are produced by sectoral production functions, this model is not very sensible if $M>1$ : the output of sector $m$ is produced by the sector $m$ vector of inputs, $\boldsymbol{x}^{m}$; not by the aggregate input vector $\boldsymbol{x}$.
The production model defined by (15) implies that each unit of aggregate input $n, x_{n}$, produces the vector of outputs $\nabla_{p} h^{n}(\boldsymbol{p})$ independently of all other inputs. This is also not a very sensible assumption if $N>1$. Hence the bilinear model defined by (16), which is a special case of the
models defined by (14) and (15), is also not a sensible model of production in the sectoral production function context. ${ }^{* 11}$

Using (3) and (4), the estimating equations for the bilinear Gross Output function defined by (16) are as follows for $t=1, \ldots, T$ :

$$
\begin{align*}
\boldsymbol{y}^{t} & =\mathbf{E}^{T} \boldsymbol{x}^{t}  \tag{17}\\
\boldsymbol{w}^{t} & =\mathbf{E} \boldsymbol{p}^{t} \tag{18}
\end{align*}
$$

When we are estimating joint cost functions, it is convenient to start by estimating the bilinear function defined by $(11), C(\boldsymbol{y}, \boldsymbol{w})=\boldsymbol{w} \cdot \mathbf{D} \boldsymbol{y}$. When we are estimating Gross Output Functions, it proves to be convenient to start by estimating the bilinear function defined by $(16), G(\boldsymbol{p}, \boldsymbol{x})=$ $\boldsymbol{x} \cdot \mathbf{E} \boldsymbol{p}$. The analysis above suggests that the Bilinear Joint Cost Function option will fit the data better than the Bilinear Gross Output Function option when we are using aggregate national accounts data. The bilinear starting functional form defined by (11) will probably provide a much better global approximation to the national technology if we are using national macroeconomic data than will be provided by the starting bilinear functional form for a gross output function defined by (16).
We will use US data for the years 1970-2022 to investigate which bilinear functional form does better at describing the data: the Gross Output Bilinear Function or the Joint Cost Bilinear Function.

Our data are based on the Augmented Productivity Database (APDB) for the US constructed by the Asian Productivity Organization and Keio University.*12 The data on outputs for the US economy are based on the recent historical revision of the US national accounts made by the Bureau of Economic Analysis (2023)[5]. The four outputs are the usual $C, G, I$ and $X$ macroeconomic aggregates and the six inputs are Imports, Labour, Machinery and Equipment, Structures, Other Capital (mainly R\&D and Inventories) and Land ( $M, L, K_{\mathrm{M} \& \mathrm{E}}, K_{\mathrm{S}}, K_{\mathrm{O}}$ and $K_{\mathrm{L}}$ ); see Appendix A for the details of the data construction. Denote the 4 dimensional output price and quantity vectors for year $t$ by $\boldsymbol{p}^{t}$ and $\boldsymbol{y}^{t}$ and the 6 dimensional input price and quantity vectors for year $t$ by $\boldsymbol{w}^{t}$ and $\boldsymbol{x}^{t}$ for $t=1970, \ldots, 2022$.
We first estimated the parameters in the 10 equations that allow us to the Bilinear Gross Output Function. The first 6 estimating equations were equations (18) and the last 4 equations were equations (17). Although each set of equations is linear in the unknown parameters, there are cross equation restrictions on the parameters which need to be taken into account. We used the Nonlinear Estimation econometric package in Shazam*13 to do the estimation. The $R^{2}$ (between observed and predicted) were as follows for the 10 equations: 0.91590 .98450 .3454 0.90450 .87030 .52500 .99810 .47080 .96750 .9662 . The equation with the lowest $R^{2}(0.3454)$ was the third equation in (18) which regressed the user cost of Machinery and Equipment on the 4 prices of the outputs (the prices of $C, G, I$ and $X$ ). The equation with the highest $R^{2}$ (0.9981) was the first equation in (17) which regressed the quantity of consumption on the 6 aggregate quantities of inputs.

[^5]We also estimated the 10 equations that estimate the parameters in the Bilinear Joint Cost Function. The first 6 estimating equations were equations (13) and the last 4 equations were equations (12). The $R^{2}$ were as follows for the 10 estimating equations: 0.99080 .97770 .9740 0.97070 .95060 .98090 .98950 .99800 .99300 .9906 . The equation with the lowest $R^{2}(0.9506)$ was the fifth equation in equations (13) which regressed the quantity of input 5 (the quantity of Other Capital Services) on the quantities of the 4 outputs. The equation with the highest $R^{2}(0.9980)$ regressed the price of Government output on the 6 input prices.
It is clear that the Bilinear Joint Cost function fits the US data much better than the Bilinear Gross Output function. These two bilinear functional forms are building blocks for the Normalized Quadratic Joint Cost and Gross Output functional forms which are flexible functional forms as we shall see. Thus it is likely that we will end up with a functional form that fits the data better if we estimate the Normalized Quadratic Joint Cost Function rather than the Normalized Quadratic Gross Output Function (because the bilinear special case for the joint cost function provides a global fit which is much better than the fit for the bilinear special case of the Normalized Quadratic Gross Output Function). In the following sections, we will estimate several variants of the Normalized Quadratic Joint Cost Function.

## 4 Estimating the Normalized Quadratic Joint Cost Function for the US

In the previous section, we defined $\boldsymbol{p}^{t}$ and $\boldsymbol{y}^{t}$ as the year $t$ price and quantity vectors for our 4 gross outputs $(C+G+I+X)$ and $\boldsymbol{w}^{t}$ and $\boldsymbol{x}^{t}$ were defined as the year $t$ price and quantity vectors for our 6 inputs (Imports, Labour, M\&E services, Structure services, Other Capital services (mainly R\&D and Inventories) and Land). We normalized the prices and quantities of the 4 outputs so that $y_{m}^{1970}=1$ for $m=1,2,3,4$. We also normalized the prices and quantities of the 6 inputs so that $w_{n}^{1970}=1$ for $n=1, \ldots, 6$. The Normalized Quadratic Joint Cost Function requires a vector of fixed nonnegative weights for input prices, $\alpha \equiv\left[\alpha_{1}, \alpha_{2}, a_{3}\right.$, $\left.\boldsymbol{\alpha}_{4}, \mathrm{a}_{5}, \mathrm{a}_{6}\right]$. Define $\boldsymbol{x}^{*} \equiv \sum_{t=1970}^{2022} \boldsymbol{x}^{t}=\left[x_{1}^{*}, x_{2}^{*}, x_{3}^{*}, x_{4}^{*}, x_{5}^{*}, x_{6}^{*}\right]$. Thus $\boldsymbol{x}^{*}$ is simply the sample wide sum of the observed input quantity vectors. Define the sum of the $x_{n}^{*}$ as $x_{\mathrm{sum}}^{*}$. Finally define the components of the $\mathbf{a}$ vector as $\alpha_{n} \equiv x_{n}^{*} / x_{\text {sum }}^{*}$ for $n=1, \ldots, 6$. Since all input prices $w_{n}^{t}$ equal 1 when $t$ equals 1 , we have $\mathbf{\alpha} \cdot \boldsymbol{w}^{1970}=1$. Thus $\mathbf{\alpha} \cdot \boldsymbol{w}^{t}$ as $t$ varies is a fixed base index of input prices that starts out at the level 1 in 1970.
The Normalized Quadratic Joint Cost Function also requires a vector of fixed nonnegative weights for output quantities, $\boldsymbol{\beta} \equiv\left[\beta_{1}, \beta_{2}, \beta_{3}, \beta_{4}\right]$. Define $\boldsymbol{p}^{*} \equiv \sum_{t=1970}^{2022} \boldsymbol{p}^{t}=\left[p_{1}^{*}, p_{2}^{*}, p_{3}^{*}, p_{4}^{*}\right]$. Thus $\boldsymbol{p}^{*}$ is simply the sample wide sum of the observed output price vectors. Define the sum of the $p_{m}^{*}$ as $p_{\text {sum }}^{*}$. Finally define the components of the $\boldsymbol{\beta}$ vector as $\beta_{m} \equiv p_{m}^{*} / p_{\text {sum }}^{*}$ for $m=1, \ldots, 4$. Since all output quantities $y_{m}^{t}$ equal 1 when $t$ equals 1 , we have $\boldsymbol{\beta} \cdot \boldsymbol{y}^{1970}=1$. Thus $\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}$ as $t$ varies is a fixed base index of output quantities that starts out at the level 1 in 1970 .
In order to condense the notation, we now let time $t=0,1,2, \ldots, 52$ instead of $t=1970,1971, \ldots, 2022$. The Linear Time Trends Normalized Quadratic Joint Cost Function at time period $t, C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$, is defined as follows:

$$
\begin{align*}
C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) & \equiv(1 / 2)\left(\boldsymbol{w}^{t} \cdot \mathbf{A} \boldsymbol{w}^{t}\right)\left(\boldsymbol{\alpha} \cdot \boldsymbol{w}^{t}\right)^{-1}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)+(1 / 2)\left(\boldsymbol{y}^{t} \cdot \mathbf{B} \boldsymbol{y}^{t}\right)\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-1} \\
& +\boldsymbol{w}^{t} \cdot \mathbf{D} \boldsymbol{y}^{t}+\left(\boldsymbol{b} \cdot \boldsymbol{w}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right) t \tag{19}
\end{align*}
$$

where $\mathbf{A}$ is a symmetric positive semidefinite matrix that satisfies the restrictions $\mathbf{A} \boldsymbol{w}^{0}=\mathbf{0}_{6}$, $\mathbf{B}$ is a symmetric negative semidefinite matrix that satisfies the restrictions $\mathbf{B} \boldsymbol{y}^{0}=\mathbf{0}_{4}$ and
$\boldsymbol{b} \equiv\left[b_{1}, \ldots, b_{6}\right]$ is a vector of parameters which allow for biased technical change. Using the algebra in Diewert and Wales (1987)[23], it can be shown that these conditions on the $\mathbf{A}$ and $\mathbf{B}$ matrices are sufficient to imply the global convexity of $C(\boldsymbol{y}, \boldsymbol{w}, t)$ in the components of $\boldsymbol{y}$ and the global concavity of $C(\boldsymbol{y}, \boldsymbol{w}, t)$ in the components of $\boldsymbol{w}$. The year $t$ estimating equations (7) and (6) become for $t=0,1,2, \ldots, 52$ :

$$
\begin{align*}
\boldsymbol{x}^{t} & =\mathbf{D} \boldsymbol{y}^{t}+\mathbf{A} \boldsymbol{w}^{t}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)^{-1}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)+(1 / 2)\left(\boldsymbol{y}^{t} \cdot \mathbf{B} \boldsymbol{y}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-1} \mathbf{\alpha} \\
& -(1 / 2)\left(\boldsymbol{w}^{t} \cdot \mathbf{A} \boldsymbol{w}^{t}\right)\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)^{-2}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right) \mathrm{a}+\boldsymbol{b}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right) t  \tag{20}\\
\boldsymbol{p}^{t} & =\mathbf{D}^{T} \boldsymbol{w}^{t}+\mathbf{B} \boldsymbol{y}^{t}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-1}+(1 / 2)\left(\boldsymbol{w}^{t} \cdot \mathbf{A} \boldsymbol{w}^{t}\right)\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)^{-1} \boldsymbol{\beta} \\
& -(1 / 2)\left(\boldsymbol{y}^{t} \cdot \mathbf{B} \boldsymbol{y}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-2}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right) \boldsymbol{\beta}+\boldsymbol{\beta}\left(\boldsymbol{b} \cdot \boldsymbol{w}^{t}\right) t . \tag{21}
\end{align*}
$$

The unknown parameters in the above equations are the components of the 6 by 4 matrix $\mathbf{D}$, the components of the 6 by 6 and 4 by 4 symmetric matrices $\mathbf{A}$ and $\mathbf{B}$ and the 6 components of the vector of technical progress parameters $\boldsymbol{b}$ which is 51 parameters in all, taking into account the symmetry restrictions on the $\mathbf{A}$ and $\mathbf{B}$ matrices. There are 53 times 10 observations so we have 530 degrees of freedom. Note that all parameters appear in a linear fashion but there are cross equations constraints on the parameters which means that we have to use nonlinear regression techniques to estimate the unknown parameters. Note that when $t=0$, the technical progress terms involving the $\boldsymbol{b}$ vector on the right hand sides of (20) and (21) vanish.
It proved to be too difficult for the nonlinear estimation command in Shazam to estimate the model defined by equations (20) and (21). Nonlinear estimation requires good starting values for the parameters in the nonlinear regression. In the previous section, we essentially set $\mathbf{A}, \mathbf{B}$ and $\boldsymbol{b}$ equal to zero matrices and estimated the components of the $\mathbf{D}$ matrix. In this section, we used the estimates for the components of the $\mathbf{D}$ matrix as starting coefficients for the estimation of (20) and (21) with $\mathbf{A}$ and $\mathbf{B}$ set equal to zero matrices. The starting values for the components of the $\boldsymbol{b}$ vector were zeros. The log likelihood for the new model that allowed for technical progress increased by 238.85 log likelihood points for adding 6 technical progress parameters. Thus it is extremely important to allow for technical progress. The $R^{2}$ for the 10 equations were as follows:
0.98910 .95760 .98680 .99520 .99670 .82910 .99800 .99640 .97050 .9466 .

We turn our attention to the estimation of the components of the input substitution matrix $\mathbf{A}$ and the (inverse) output substitution matrix $\mathbf{B}$. We need to ensure that our estimated $\mathbf{A}$ matrix is a negative semidefinite symmetric matrix that satisfies $\mathbf{A} \boldsymbol{w}^{0}=\mathbf{A} \mathbf{1}_{6}=\mathbf{0}_{6}$ where $\mathbf{1}_{6}$ is a vector of ones of dimension 6 .

The imposition of symmetry and negative semidefiniteness on $\mathbf{A}$ can be accomplished using a technique due to Wiley, Schmidt and Bramble (1973)[60]: simply replace the matrix A by

$$
\begin{equation*}
\mathbf{A} \equiv-\mathbf{U} \mathbf{U}^{T} \tag{22}
\end{equation*}
$$

where $\mathbf{U}$ is a 6 by 6 lower triangular matrix; i.e., $u_{i j}=0$ if $i<j$.
The restrictions $\mathbf{A} \boldsymbol{w}^{0}=\mathbf{A} \mathbf{1}_{6}=\mathbf{0}_{6}$ on $\mathbf{A}$ can be imposed if we impose the following restrictions on $\mathbf{U}$ :

$$
\begin{equation*}
\mathbf{U}^{T} \mathbf{1}_{6}=\mathbf{0}_{6} \tag{23}
\end{equation*}
$$

The imposition of symmetry and positive semidefiniteness on $\mathbf{B}$ can be accomplished in a similar fashion: set $\mathbf{B}$ equal to:

$$
\begin{equation*}
\mathbf{B} \equiv \mathbf{V} \mathbf{V}^{T} \tag{24}
\end{equation*}
$$

where $\mathbf{V}$ is a 4 by 4 lower triangular matrix; i.e., $v_{i j}=0$ if $i<j$.
The restrictions $\mathbf{B} \boldsymbol{y}^{0}=\mathbf{B 1}_{4}=\mathbf{0}_{4}$ on $\mathbf{B}$ can be imposed if we impose the following restrictions on $\mathbf{V}$ :

$$
\begin{equation*}
\mathbf{V}^{T} \mathbf{1}_{4}=\mathbf{0}_{4} \tag{25}
\end{equation*}
$$

The restrictions (23) and (24) imply that the maximum rank for the $\mathbf{A}$ and $\mathbf{B}$ matrices is 5 and 3 respectively. Once the matrices $\mathbf{A}$ and $\mathbf{B}$ in the estimating equations (20) and (21) are replaced by $-\mathbf{U U}^{T}$ and $\mathbf{V} \mathbf{V}^{T}$ respectively, the resulting estimating equations are no longer linear in the unknown parameters and a nonlinear regression package must be used. For more details on how to implement this method for imposing the restrictions on the substitution matrices, see Diewert and Wales (1987)[23] (1988)[24] (1992)[25].
Setting $\mathbf{A} \equiv-\mathbf{U U}^{T}$ and $\mathbf{B} \equiv \mathbf{V} \mathbf{V}^{T}$ can lead to difficulties in estimation due to the addition of a large number of new parameters at the same time. A more fool proof method of proceeding is to introduce the columns of $\mathbf{U}$ and $\mathbf{V}$ into the regression one column at a time. Thus we set $\mathbf{A} \equiv-\boldsymbol{u} \boldsymbol{u}^{T}$ where $\boldsymbol{u}^{T} \equiv\left[u_{1}, u_{2}, \ldots, u_{6}\right]$ and $u_{1}$ is set equal to $-\sum_{n=2}^{6} u_{n}$. Similarly, we set $\mathbf{B}=\boldsymbol{v} \boldsymbol{v}^{T}$ where $\boldsymbol{v}^{T} \equiv\left[v_{1}, v_{2}, v_{3}, v_{4}\right]$ and $v_{1}$ is set equal to $-\sum_{m=2}^{4} v_{m}$. We used the final coefficients in the previous regression as starting values in this new regression along with the starting values for the $u_{n}$ and $v_{m}$ equal to 0 . Thus this new nonlinear regression estimates rank 1 substitution matrices. ${ }^{* 14}$ The log likelihood for the new model increased by 76.088 log likelihood points for adding 5 input and 3 output substitution parameters. The $R^{2}$ for the 10 equations were as follows:
0.99540 .98520 .98660 .99250 .99840 .92140 .97170 .99710 .98520 .9624 .

Note that the $R^{2}$ for equation 6 (the demand for land services) was 0.8291 in the previous nonlinear regression model but has now increased to 0.9214 when we allow for some substitution between the 6 inputs.
Looking at the components of the output substitution matrix, it turned out that the components of the estimated $\boldsymbol{v}$ vector were as follows $\left(v_{2}^{*}, v_{3}^{*}, v_{4}^{*}\right)=(-0.0114,-0.1454,0.0108)$ and $v_{1}^{*}=-\left(v_{2}^{*}+v_{3}^{*}+v_{4}^{*}\right)=0.1460$. If the products produced by the 4 sectors in our model (the $C, G, I$ and $X$ sectors) were unique to each sector (i.e., if the Nonjoint Production Model defined by (8) above held), then the estimated $\boldsymbol{v}$ vector should be a zero vector. Since the estimated $\left(v_{2}^{*}, v_{3}^{*}, v_{4}^{*}\right)$ were significantly different from zero, it is likely that there is some joint production in the US economy. ${ }^{* 15}$
The process of adding an additional column to the input and output substitution matrices $\mathbf{U}$ and $\mathbf{V}$ (and then running a new nonlinear regression using the final coefficients in the previous regression as starting values in the new regression) continued until a rank 5 input substitution matrix and a rank 3 output substitution matrix were estimated. The resulting model is the model described by equations (20) - (25). The log likelihood for this 51 parameter model was 27.867 points higher than the final $\log$ likelihood for the model that had rank 1 input and output substitution matrices. We have added 13 new parameters to the previous rank 1 model. However, 4 of the new substitution parameters converged to 0 : the data only supported the estimation of a rank 4 input substitution matrix (maximum rank is 5 ) and a rank 1 output substitution matrix (maximum rank is 3 ).
We need to develop formulae for various elasticities of substitution once we have estimates for the components of $\mathbf{A}, \mathbf{B}, \mathbf{D}$ and $\boldsymbol{b}$. Define the vector of input demand functions $\boldsymbol{x}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$ as the right hand side of equations (20). The 6 by 6 matrix of derivatives of the input demand

[^6]functions with respect to the components of the input price vector $\boldsymbol{w}$ evaluated at the year $t$ values for $\boldsymbol{y}$ and $\boldsymbol{x}, \nabla_{w} \boldsymbol{x}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$, can be obtained by differentiating the right hand side of (20) with respect to the components of $\boldsymbol{w}$ :
\[

$$
\begin{align*}
\nabla_{w} \boldsymbol{x}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) & =\mathbf{A}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)^{-1}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)-\left(\mathbf{A} \boldsymbol{w}^{t}\right) \mathbf{\alpha}^{T}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)^{-2}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right) \\
& -\mathbf{\alpha}\left(\mathbf{A} \boldsymbol{w}^{t}\right)^{T}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)^{-2}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)+\mathrm{a}^{T}\left(\boldsymbol{w}^{t} \cdot \mathbf{A} \boldsymbol{w}^{t}\right)\left(\mathrm{\alpha} \cdot \boldsymbol{w}^{t}\right)^{-3}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right) \tag{26}
\end{align*}
$$
\]

Define the year $t$ elasticity of the demand for input $n$ with respect to a change in the price of input $j, E x_{n j}^{t}$, as follows for $n=1, \ldots, 6$ and $j=1, \ldots, 6$ :

$$
\begin{equation*}
E x_{n j}^{t} \equiv \partial \ln x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial \ln w_{j} \equiv\left[\partial x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial w_{j}\right]\left[w_{j}^{t} / x_{n}^{t}\right] \tag{27}
\end{equation*}
$$

where $x_{n}^{t} \equiv x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$ is the fitted demand for input $n$ in year $t$ defined by the $n$th component on the right hand side of equation (20) for year $t$ and $\partial x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial w_{j}$ is defined in equations (26).
Recall that when $t=0, \boldsymbol{w}^{0}$ and $\mathbf{A}$ satisfy the equation $\mathbf{A} \boldsymbol{w}^{0}=\mathbf{0}_{6}$. Thus when $t=0$, the right hand side of $(26)$ when $t=0$ collapses to $\nabla_{w} \boldsymbol{x}\left(\boldsymbol{y}^{0}, \boldsymbol{w}^{0}, 0\right)=\mathbf{A}\left(\boldsymbol{\alpha} \cdot \boldsymbol{w}^{0}\right)^{-1}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{0}\right)$. As time marches on, the last 3 terms on the right hand side of (26) will tend to be fairly close to zero so the elasticities defined by (27) will be approximately equal to the following expressions for $n=1, \ldots, 6$ and $j=1, \ldots, 6$ :

$$
\begin{equation*}
E x_{n j}^{t} \approx a_{n j}\left(w_{j}^{t} / \mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right) /\left(x_{n}^{t} / \boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right) \tag{28}
\end{equation*}
$$

where $a_{n j}$ is the component of the estimated $\mathbf{A}$ matrix in row $n$ and column $j$. If there are divergent trends in either the input prices $\boldsymbol{w}^{t}$ or in the output quantities $\boldsymbol{y}^{t}$ over time, then it can be seen that the input substitution elasticities will have trends over time.
The problem of trending elasticities also arises with respect to output substitution elasticities. Define the vector of (inverse) output supply functions $\boldsymbol{p}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$ as the right hand side of equations (21). The 4 by 4 matrix of derivatives of the output supply functions with respect to the components of the output quantity vector $\boldsymbol{y}$ evaluated at the year $t$ values for $\boldsymbol{y}$ and $\boldsymbol{x}$, $\nabla_{y} \boldsymbol{p}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$, can be obtained by differentiating the right hand side of (21) with respect to the components of $\boldsymbol{y}$ :

$$
\begin{align*}
\nabla_{y} \boldsymbol{p}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) & =\mathbf{B}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-1}-\left(\mathbf{B} \boldsymbol{y}^{t}\right) \boldsymbol{\beta}^{T}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-2} \\
& -\boldsymbol{\beta}\left(\mathbf{B} \boldsymbol{y}^{t}\right)^{T}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-2}+\boldsymbol{\beta} \boldsymbol{\beta}^{T}\left(\boldsymbol{y}^{t} \cdot \mathbf{B} \boldsymbol{y}^{t}\right)\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-3} \tag{29}
\end{align*}
$$

Define the year $t$ elasticity of the (inverse) supply for output $m$ with respect to a change in the price of output $k, E p_{m k}^{t}$, as follows for $m=1, \ldots, 4$ and $k=1, \ldots, 4$ :

$$
\begin{equation*}
E p_{m k}^{t} \equiv \partial \ln p_{m}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial \ln y_{k} \equiv\left[\partial p_{m}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial y_{k}\right]\left[y_{k}^{t} / p_{m}^{t}\right] \tag{30}
\end{equation*}
$$

where $p_{m}^{t} \equiv p_{m}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$ is the fitted price for output $m$ in year $t$ defined by the $m$ th component on the right hand side of equation (21) for year $t$ and $\partial p_{m}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial y_{k}$ is defined in equations (29).
Recall that when $t=0, \boldsymbol{y}^{0}$ and $\mathbf{B}$ satisfy the equation $\mathbf{B} \boldsymbol{y}^{0}=\mathbf{0}_{4}$. Thus when $t=0$, the right hand side of $(29)$ when $t=0$ collapses to $\nabla_{y} \boldsymbol{p}\left(\boldsymbol{y}^{0}, \boldsymbol{w}^{0}, 0\right)=\mathbf{B}\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-1}$. The last 3 terms on the right hand side of (29) will tend to be fairly close to zero so the elasticities defined by (30) will be approximately equal to the following expressions for $m=1, \ldots, 4$ and $k=1, \ldots, 4$ :

$$
\begin{equation*}
E p_{m k}^{t} \approx b_{m k}\left(y_{k}^{t} / \boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right) /\left(p_{m}^{t} / \boldsymbol{\alpha} \cdot \boldsymbol{w}^{t}\right) \tag{31}
\end{equation*}
$$

where $b_{m k}$ is the component of the estimated $\mathbf{B}$ matrix in row $m$ and column $k$. If there are divergent trends in either the input prices $\boldsymbol{w}^{t}$ or in the output quantities $\boldsymbol{y}^{t}$ over time, then it can be seen that the output substitution elasticities will have trends over time.
Diewert and Lawrence (2002; 149-151)[21] noticed the above trending elasticities problem in the context of estimating a Normalized Quadratic GDP function and they suggested a method for dealing with this problem: let the components of the $\mathbf{A}$ and $\mathbf{B}$ (or $\mathbf{U}$ and $\mathbf{V}$ ) matrices have linear time trends over the sample period. We implemented their method for our US data. Thus the matrices $\mathbf{A}$ and $\mathbf{B}$ in the estimating equations (20) and (21) become functions of time $t, \mathbf{A}(t)$ and $\mathbf{B}(t) .{ }^{* 16}$ This new model has a total of 72 parameters. The log likelihood for this model increased by 78.021 points for adding 21 new parameters. The new $R^{2}$ were as follows:
0.99040 .99140 .98290 .98750 .99800 .93830 .99880 .99820 .99250 .9745 .

The lowest $R^{2}$ was for the demand for land services equation which was 0.9383 ; the $R^{2}$ for the remaining equations were all above 0.97 . Ten out of the 42 substitution matrix parameters converged to 0 . Thus the ranks of the beginning and end of period input substitution matrices were equal to 3 (maximum possible rank is 5 ) and the initial inverse output supply substitution matrix had rank 1 and the final output substitution matrix had rank 2 (maximum possible rank is 3 ).
The use of linear time trends to model technical progress will lead to smooth estimates of overall technical progress. But we know from index number estimates of technical progress that the rate varies substantially from year to year and the trends also vary substantially from decade to decade. ${ }^{* 17}$ Thus we will follow the example of Diewert and Wales (1992)[25] by replacing the linear trends in equations 1-6 by piece-wise linear spline functions. In order to determine the break points (or knots) for the spline functions, we looked at the plots of the regression residuals. These residual plots showed systematic trends that changed abruptly from time to time. We used the observations where breaks in the residual trends occurred in equations 1-6 as our break points for the piece-wise linear functions that we use to model technical progress. The break points occurred at the following observations where we now number the years 1970-2022 as 1-53:

Equation 1: 818233751
Equation 2: 81317223651
Equation 3: 101726374051
Equation 4: 132731343850
Equation 5: 51722313551
Equation 6: 19223137.
The terms $t b_{n}$ for $n=1, \ldots, 6$ appear in the estimating equations (20) and (21). We need to replace these linear functions of time $t$ by piece-wise linear functions of time where the line segments are linked together to form continuous functions of time. Thus the linear function of time, $t b_{1}$, is replaced by the function $\sum_{i=1}^{6} b_{1 i} f_{1 i}(t)$ where the $b_{1 i}$ are constants to be estimated and the functions $f_{1 i}(t)$ of time $t$ are defined below:

$$
\begin{aligned}
& f_{11}(t) \equiv t \text { for } t=0,1, \ldots, 7 ; \quad f_{11}(t) \equiv 7 \text { for } t=8,9, \ldots, 52 \\
& f_{12}(t) \equiv 0 \text { for } t=0,1, \ldots, 7 ; \quad f_{12}(t) \equiv t-7 \text { for } t=8,9, \ldots, 17 ; \\
& \quad f_{12}(t) \equiv f_{12}(17) \text { for } t=18,19, \ldots, 52 ;
\end{aligned}
$$

[^7]\[

$$
\begin{aligned}
& f_{13}(t) \equiv 0 \text { for } t=0,1, \ldots, 17 ; f_{13}(t) \equiv t-17 \text { for } t=18, \ldots, 22 ; \\
& f_{13}(t) \equiv f_{13}(22) \text { for } t=23, \ldots, 52 ; \\
& f_{14}(t) \equiv 0 \text { for } t=0,1, \ldots, 22 ; f_{14}(t) \equiv t-22 \text { for } t=23, \ldots, 36 ; \\
& f_{14}(t) \equiv f_{14}(36) \text { for } t=37, \ldots, 52 ; \\
& f_{15}(t) \equiv 0 \text { for } t=0,1, \ldots, 36 ; f_{15}(t) \equiv t-36 \text { for } t=37, \ldots, 50 ; \\
& f_{15}(t) \equiv f_{15}(50) \text { for } t=51,52 ; \\
& f_{16}(t) \equiv 0 \text { for } t=0,1, \ldots, 50 ; f_{16}(t) \equiv t-50 \text { for } t=51,52 \text {. }
\end{aligned}
$$
\]

Note how the break points for equation 1, 818233751 appear in the above definitions for the $f_{1 i}(t)$. Note also that $\sum_{i=1}^{6} f_{1 i}(t)=t$ for $t=0,1, \ldots, 52$.
The linear function of time $t b_{2}$ is replaced by the function $\sum_{i=1}^{7} b_{2 i} f_{2 i}(t)$ where the $b_{2 i}$ are constants to be estimated and the functions $f_{2 i}(t)$ of time $t$ are defined below using the break points for equation 2 :

$$
\begin{aligned}
& f_{21}(t) \equiv t \text { for } t=0,1, \ldots, 7 ; \quad f_{21}(t) \equiv 7 \text { for } t=8,9, \ldots, 52 ; \\
& f_{22}(t) \equiv 0 \text { for } t=0,1, \ldots, 7 ; \quad f_{22}(t) \equiv t-7 \text { for } t=8,9, \ldots, 12 ; \\
& f_{22}(t) \equiv f_{22}(12) \text { for } t=13, \ldots, 52 ; \\
& f_{23}(t) \equiv 0 \text { for } t=0,1, \ldots, 12 ; \quad f_{23}(t) \equiv t-12 \text { for } t=13, \ldots, 16 ; \\
& f_{23}(t) \equiv f_{23}(16) \text { for } t=17, \ldots, 52 ; \\
& f_{24}(t) \equiv 0 \text { for } t=0,1, \ldots, 16 ; \quad f_{24}(t) \equiv t-16 \text { for } t=17, \ldots, 21 ; \\
& f_{24}(t) \equiv f_{24}(21) \text { for } t=22, \ldots, 52 ; \\
& f_{25}(t) \equiv 0 \text { for } t=0,1, \ldots, 21 ; \quad f_{25}(t) \equiv t-21 \text { for } t=22, \ldots, 35 ; \\
& f_{25}(t) \equiv f_{25}(35) \text { for } t=36, \ldots, 52 \\
& f_{26}(t) \equiv 0 \text { for } t=0,1, \ldots, 35 ; \quad f_{26}(t) \equiv t-35 \text { for } t=36, \ldots, 50 ;
\end{aligned} \quad \begin{gathered}
f_{26}(t) \equiv f_{26}(36) \text { for } t=51,52 ; \\
f_{27}(t) \equiv 0 \text { for } t=0,1, \ldots, 50 ; \quad f_{27}(t) \equiv t-50 \text { for } t=51,52 .
\end{gathered}
$$

The spline functions $f_{3 i}(t), f_{4 i}(t)$ and $f_{5 i}(t)$ for $i=1, \ldots, 7$ and $f_{6 i}(t)$ for $i=1, \ldots, 5$ were defined in a similar fashion using the break points listed above. Thus replace each $t b_{n}$ term in equations (20) by the linear spline function $\sum_{i} b_{n i} f_{n i}(t)$ for $n=1, \ldots, 6$. This adds 33 additional parameters to the previous model for a total of 105 parameters. In equations (21), the scalar term $\left(\boldsymbol{b} \cdot \boldsymbol{w}^{t}\right) t$ appears on the right hand side. It needs to be replaced with the following terms: $w_{1}^{t} \sum_{i=1}^{6} b_{1 i} f_{1 i}(t)+w_{2}^{t} \sum_{i=1}^{7} b_{2 i} f_{2 i}(t)+w_{3}^{t} \sum_{i=1}^{7} b_{3 i} f_{3 i}(t)+w_{4}^{t} \sum_{i=1}^{7} b_{4 i} f_{4 i}(t)+$ $w_{5}^{t} \sum_{i=1}^{7} b_{5 i} f_{5 i}(t)+w_{6}^{t} \sum_{i=1}^{5} b_{6 i} f_{6 i}(t)$.
We used the final coefficients in the previous nonlinear regression as starting values in the new regression with 105 unknown parameters. The starting coefficient for each $b_{n i}$ was set equal to the final coefficient for $b_{n}$ in the previous regression for $n=1, \ldots, 6$. Somewhat surprisingly, Shazam had no difficulty in converging to new estimates; the nonlinear option took only 2.8 seconds with 700 iterations. The log likelihood increased by 273.11 points for adding 33 technical progress parameters. The $R^{2}$ for the 10 equations were as follows:
0.99910 .99780 .99620 .99840 .99870 .99010 .99960 .99940 .99810 .9791 .

Our final Joint Cost Function model fits the data very well indeed. However, there are a large number of parameters in our final model. In the following section, we will see if the elasticities and the estimates for technical progress that the model generates are "reasonable".*18

[^8]
## 5 Estimates for Elasticities and Biases in Technical Progress

In the previous section, we derived formula (27) (which drew on (26)) for the year $t$ elasticity of demand for input $n$ with respect to a change in the price of input $j, E x_{n j}^{t} \equiv$ $\left[\partial x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial w_{j}\right]\left[w_{j}^{t} / x_{n}^{t}\right]$. This formula was derived using linear time trends to describe technical progress but the same formula is also valid for year $t$ using our spline model for technical progress provided that we replace the constant matrix $\mathbf{A}$ by the weighted average matrix for year $t, \mathbf{A}(t)$. Thus we use our estimated coefficients to calculate these input elasticities of input demand for each year. The resulting averages of these elasticities are listed in Table 1 below. The main diagonal elasticities for each year are listed in Appendix D.

Table 1 Average Price Elasticities of Input Demand over the Period 1970-2022

| $E x_{n j}$ | $w_{1}$ | $w_{2}$ | $w_{3}$ | $w_{4}$ | $w_{5}$ | $w_{6}$ |  |
| :--- | :---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  |  | M | L | $\mathrm{K}_{\mathrm{M} \& \mathrm{E}}$ | $\mathrm{K}_{\mathrm{S}}$ | $\mathrm{K}_{\mathrm{O}}$ | $\mathrm{K}_{\mathrm{L}}$ |
| $x_{1}$ | M | -0.4352 | 0.4203 | -0.2333 | 0.0513 | 0.1564 | 0.0404 |
| $x_{2}$ | L | 0.0736 | -0.2872 | 0.1614 | 0.0584 | -0.0233 | 0.0172 |
| $x_{3}$ | $\mathrm{~K}_{\mathrm{M} \& \mathrm{E}}$ | -0.2541 | 1.0037 | -0.8031 | -0.1320 | 0.2257 | -0.0403 |
| $x_{4}$ | $\mathrm{~K}_{\mathrm{S}}$ | 0.0366 | 0.1889 | -0.0683 | -0.0997 | -0.0298 | -0.0277 |
| $x_{5}$ | $\mathrm{~K}_{\mathrm{O}}$ | 0.3426 | -0.2464 | 0.4017 | -0.0912 | -0.3389 | -0.0679 |
| $x_{6}$ | $\mathrm{~K}_{\mathrm{L}}$ | 0.0667 | 0.1337 | -0.0458 | -0.0679 | -0.0486 | -0.0381 |

Note: Imports (M), Labour (L), Machinery and Equipment ( $\mathrm{K}_{\mathrm{M} \& E}$ ), Structures ( $\mathrm{K}_{\mathrm{S}}$ ), Other Capital (mainly R\&D and Inventories: $\mathrm{K}_{\mathrm{O}}$ ) and Land ( $\mathrm{K}_{\mathrm{L}}$ ).

The main diagonal elements of the above matrix are negative, which is consistent with the concavity in input prices property of a joint cost function. A pair of inputs $n$ and $j$ are substitutes if $\partial x_{n}(\boldsymbol{y}, \boldsymbol{w}, t) / \partial w_{j}=\partial x_{j}(\boldsymbol{y}, \boldsymbol{w}, t) / \partial w_{n}$ is positive and are complements if these partial derivatives are negative. From Table 1, we see that there are 8 pairs of substitute inputs and 7 pairs of complementary inputs on average. This means that it is unlikely that CobbDouglas or CES cost functions could provide a satisfactory approximation to the US aggregate technology (because all inputs are substitutes using these functional forms). Another point to note that follows from an examination of the above Table 1 is that most of the elasticities are small in magnitude.
In the previous section, we derived a formula (30) (which drew on (29)) for the year $t$ elasticity of marginal cost for output $m$ with respect to a change in the quantity of output $k, E p_{m k}^{t} \equiv$ $\left[\partial p_{m}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial y_{k}\right]\left[y_{k}^{t} / p_{m}^{t}\right]$. This formula was derived using linear time trends to describe technical progress but the same formula is also valid using our spline model for technical progress, provided that for year $t$ elasticities, we replace the constant matrix $\mathbf{B}$ in (29) by the weighted average matrix $\mathbf{B}(t)$. Thus we use our estimated coefficients to calculate these marginal cost elasticities, which can also be interpreted as (inverse) output supply elasticities. The resulting averages of these elasticities are listed in Table 2 below. The main diagonal elasticities for each year are listed in Appendix A. ${ }^{* 19}$

The main diagonal elasticities are all positive and this follows from the fact that we have imposed convexity in $\boldsymbol{y}$ on our estimated joint cost function. However, the elasticities in Table 2 are all quite small. This indicates that the US aggregate joint cost function is close to being

[^9]Table 2 Average Elasticities of Inverse Output Supply over the Period 1970-2022

| $E p_{m k}$ | $y_{1}$ | $y_{2}$ | $y_{3}$ | $y_{4}$ |  |
| :---: | :---: | ---: | ---: | ---: | ---: |
|  |  | C | G | I | X |
| $p_{1}$ | C | 0.0787 | -0.0445 | -0.0474 | 0.0131 |
| $p_{2}$ | G | -0.1683 | 0.1195 | 0.0628 | -0.0140 |
| $p_{3}$ | I | -0.1279 | 0.0449 | 0.1225 | -0.0396 |
| $p_{4}$ | X | 0.0914 | -0.0312 | -0.0892 | 0.0914 |

Note: Consumption (C), Government (G), Investment (I), Export (X).
equal to the sum of 4 sectoral cost functions that produce non overlapping products; i.e., the assumption of no joint production almost holds.
In order to define our next set of elasticities, we need to list the functional form for the joint cost function in year $t$ :

$$
\begin{align*}
C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) & \equiv(1 / 2)\left(\boldsymbol{w}^{t} \cdot \mathbf{A}(t) \boldsymbol{w}^{t}\right)\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)^{-1}\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)+(1 / 2)\left(\boldsymbol{y}^{t} \cdot \mathbf{B}(t) \boldsymbol{y}^{t}\right)\left(\mathbf{\alpha} \cdot \boldsymbol{w}^{t}\right)\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)^{-1} \\
& +\boldsymbol{w}^{t} \cdot \mathbf{D} \boldsymbol{y}^{t}+\left(\boldsymbol{\beta} \cdot \boldsymbol{y}^{t}\right)\left[w_{1}^{t} \sum_{i=1}^{6} b_{1 i} f_{1 i}(t)+w_{2}^{t} \sum_{i=1}^{7} b_{2 i} f_{2 i}(t)+w_{3}^{t} \sum_{i=1}^{7} b_{3 i} f_{3 i}(t)\right. \\
& \left.+w_{4}^{t} \sum_{i=1}^{7} b_{4 i} f_{4 i}(t)+w_{5}^{t} \sum_{i=1}^{7} b_{5 i} f_{5 i}(t)+w_{6}^{t} \sum_{i=1}^{5} b_{6 i} f_{6 i}(t)\right] \tag{32}
\end{align*}
$$

where the year $t$ matrices $\mathbf{A}(t)$ and $\mathbf{B}(t)$ and the piece-wise linear in $t$ spline functions $f_{n i}$ were defined in the previous section. The above definition for $C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$ along with Shephard's Lemma, $x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)=\partial C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial w_{n}$, are used to define the year $t$ elasticity of input demand $n$ with respect to an increase in output $m, E x_{n m}^{t}$, as follows for $n=1, \ldots, 6, m=$ $1, \ldots, 4$ and $t=0,1, \ldots, 52$ :

$$
\begin{equation*}
E x_{n m}^{t} \equiv\left[\partial x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial y_{m}\right]\left[y_{m}^{t} / x_{n}^{t}\right]=\left[\partial^{2} C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial w_{n} \partial y_{m}\right]\left[y_{m}^{t} / x_{n}^{t}\right] . \tag{33}
\end{equation*}
$$

The averages of these elasticities are listed in Table 3 below.

Table 3 Average Elasticities of Input Demand with Respect to Output Quantities

| $E x_{n m}$ | $y_{1}$ | $y_{2}$ |  | $y_{3}$ | $y_{4}$ |
| :--- | :---: | :---: | ---: | ---: | ---: |
|  |  | C | G | I | X |
| $x_{1}$ | M | 0.0864 | 0.0600 | 0.4385 | 0.4152 |
| $x_{2}$ | L | 0.6551 | 0.1717 | 0.1771 | -0.0038 |
| $x_{3}$ | $\mathrm{~K}_{\mathrm{M} \& \mathrm{E}}$ | 0.6367 | -0.4275 | 0.5149 | 0.2760 |
| $x_{4}$ | $\mathrm{~K}_{\mathrm{S}}$ | 0.4815 | 0.2818 | 0.0655 | 0.1712 |
| $x_{5}$ | $\mathrm{~K}_{\mathrm{O}}$ | 0.8214 | 0.5049 | -0.0554 | -0.2709 |
| $x_{6}$ | $\mathrm{~K}_{\mathrm{L}}$ | 0.5110 | 0.2280 | 0.1532 | 0.1079 |

Note: Consumption (C), Government (G), Investment (I), Export (X),
Imports (M), Labour (L), Machinery and Equipment ( $\mathrm{K}_{\mathrm{M} \& E}$ ), Structures $\left(\mathrm{K}_{\mathrm{S}}\right)$, Other Capital (mainly R\&D and Inventories: $\mathrm{K}_{\mathrm{O}}$ ) and Land ( $\mathrm{K}_{\mathrm{L}}$ ).

On average, an increase in consumption $y_{1}$ leads to substantial increases in the demand for all inputs except imports where the elasticity was only 0.0864 . It is somewhat puzzling that an increase in government output $y_{2}$ should lead to a substantial drop in the demand for input 3, Machinery and Equipment. ${ }^{* 20}$

[^10]Definition (32) for the year $t$ cost function, $C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$, along with the price equals marginal cost equations, $p_{m}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)=\partial C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial y_{m}$, can be used to define the year $t$ elasticity of marginal cost $m$ with respect to an increase in the price of input $n, E p_{m n}^{t}$, as follows for $m=1, \ldots, 4, n=1, \ldots, 6$ and $t=0,1, \ldots, 52$ :

$$
\begin{equation*}
E p_{m n}^{t} \equiv\left[\partial p_{m}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial w_{n}\right]\left[w_{n}^{t} / p_{m}^{t}\right]=\left[\partial^{2} C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial y_{m} \partial w_{n}\right]\left[w_{n}^{t} / p_{m}^{t}\right] . \tag{34}
\end{equation*}
$$

The averages of these elasticities over the years 1970-2022 are listed in Table 4 below.

Table 4 Average Elasticities of Output Prices with Respect to Input Prices

| $E p_{m n}$ | $w_{1}$ | $w_{2}$ | $w_{3}$ | $w_{4}$ | $w_{5}$ | $w_{6}$ |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | M | L | $\mathrm{K}_{\mathrm{M} \mathrm{\& E}}$ | $\mathrm{~K}_{\mathrm{S}}$ | $\mathrm{K}_{\mathrm{O}}$ | $\mathrm{K}_{\mathrm{L}}$ |
| $p_{1}$ | C | 0.0222 | 0.6119 | 0.0986 | 0.1328 | 0.2931 | -0.1554 |
| $p_{2}$ | G | 0.0476 | 0.6350 | -0.2583 | 0.3060 | 0.0750 | 0.0596 |
| $p_{3}$ | I | 0.2269 | 0.4618 | 0.2183 | 0.0492 | 0.1653 | 0.1043 |
| $p_{4}$ | X | 0.5175 | -0.003 | 0.2711 | 0.2931 | -0.0056 | 0.0494 |

> Note: Consumption $(\mathrm{C})$, Government $(\mathrm{G})$, Investment (I), Export $(\mathrm{X})$, Imports $(\mathrm{M})$, Labour $(\mathrm{L})$, Machinery and Equipment $\left(\mathrm{K}_{\mathrm{M} \& \mathrm{E}}\right)$, Structures $\left(\mathrm{K}_{\mathrm{S}}\right)$, Other Capital (mainly R\&D and Inventories: $\left.\mathrm{K}_{\mathrm{O}}\right)$ and Land $\left(\mathrm{K}_{\mathrm{L}}\right)$.

For the most part, an increase in an input price flows through to output prices and increases the price of each of our four outputs. On average, a one percent increase in the price of labour, $w_{2}$, substantially increases the prices of consumption, government and investment, $p_{1}, p_{2}$ and $p_{3}$, but has a tiny negative effect on the price of exports. There are puzzles that an increase in the input price of Machinery and Equipment services $w_{3}$ leads to a substantial drop in the price of government output $p_{2}$ and that an increase in the user cost of land $w_{6}$ leads to a decline in the price of consumption $p_{1}$.
We turn now to the determination of measures of technical progress and the biases in technical progress. The use of a joint cost function to measure technical progress can be traced back to Salter (1960)[52] at least. The most comprehensive measure of technical change going from year $t$ to year $t+1$ using a joint cost function that depends on time, say $C^{t}(\boldsymbol{y}, \boldsymbol{w})$ where $\boldsymbol{y}$ and $\boldsymbol{w}$ are arbitrary reference output quantity and input price vectors, is $C^{t}(\boldsymbol{y}, \boldsymbol{w}) / C^{t+1}(\boldsymbol{y}, \boldsymbol{w}) . .^{* 21}$ Holding constant output levels $\boldsymbol{y}$ and input prices $\boldsymbol{w}$, technical progress should reduce the cost of producing $\boldsymbol{y}$ with input prices held constant so that $C^{t+1}(\boldsymbol{y}, \boldsymbol{w})<C^{t}(\boldsymbol{y}, \boldsymbol{w})$ and hence technical progress is greater than 1 . We will not use this measure of technical progress but we will use a related measure which has the advantage of giving us a decomposition of technical progress into a sum of explanatory factors.
We define year $t$ cost saving technical progress $\kappa^{t}$ as the derivative of period $t$ cost with respect to time $t$ divided by year $t$ fitted cost:

$$
\begin{equation*}
\kappa^{t} \equiv \partial \ln C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial t=\left[\partial C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial t\right] / C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) . \tag{35}
\end{equation*}
$$

The year $t$ fitted input $n$ demand function is $x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)=\partial C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial w_{n}$ for $n=1, \ldots, 6$. Define the year $t$ input $n$ cost saving bias, $\kappa_{n}^{t}$, for $n=1, \ldots, 6$ as follows:

$$
\begin{equation*}
\kappa_{n}^{t} \equiv w_{n}^{t}\left[\partial x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial t\right] / C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)=w_{n}^{t}\left[\partial^{2} C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial w_{n} \partial t\right] / C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) . \tag{36}
\end{equation*}
$$

matrix $\mathbf{D}$ in our model. The imposition of these nonnegativity constraints would lead to nonnegative elasticities for $E x_{n m}^{t}$ and $E p_{m n}^{t}$ for $t=0$.
*21 This measure is due to $\operatorname{Balk}(1998 ; 58)[2]$. For specializations and applications of this definition, see Balk (2001)[3] (2003)[4] and Diewert (2014)[16].

Thus $\kappa_{n}^{t}$ is the derivative of the year $t$ cost of input $n$ with respect to time, $w_{n}^{t} \partial x_{n}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right) / \partial t$, divided by total year $t$ fitted input cost, $C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}, t\right)$. Using the estimated cost function defined by (32), we can calculate $\kappa^{t}$ and the bias terms $\kappa_{n}^{t} .{ }^{* 22}$ Euler's Theorem on functions and the linear homogeneity of our estimated $C(\boldsymbol{y}, \boldsymbol{w}, t)$ in $\boldsymbol{w}$ implies that the following decomposition of $\kappa^{t}$ into explanatory input bias terms will hold for each year $t$ :

$$
\begin{equation*}
\kappa^{t}=\sum_{n=1}^{6} \kappa_{n}^{t} . \tag{37}
\end{equation*}
$$

Table 5 below lists the technical progress estimates $\kappa^{t}$ and the corresponding bias components.
From Table 5, we see that on average, technical progress reduced economy wide costs by 0.58 percentage points per year. For the years 1971-2022, the average cost reduction was 0.57 percentage points per year. In Appendix A, we show that the corresponding annual average Total Factor Productivity (TFP) Growth estimates using the Jorgenson and Griliches (1967)[31], Diewert and Morrison (1986)[22] and Kohli (1990)[39] (2003)[43] index number methodology was 0.59 percentage points per year. Using the Diewert and Fox (2018)[19] nonparametric methodology for measuring TFP growth explained in Appendix C led to an estimate of 0.59 percentage points per year over the period 1971-2022. Thus our econometric method for measuring technical progress is consistent with alternative methods for measuring technical progress and TFP growth, at least over longer periods. However, the econometric method smooths the year to year fluctuations technical progress and so it does not capture the annual fluctuations in technical progress that are captured by the index number and nonparametric methods, as shown in Figure 1. ${ }^{* 23}$


Figure 1 Comparison of Technical Progress Estimates based on Different Approaches

Since $\kappa_{2}^{t}, \kappa_{3}^{t}, \kappa_{4}^{t}$ and $\kappa_{6}^{t}$ were negative on average, technical progress was on average Labour saving, Machinery and Equipment saving, Structure saving and Land saving. Thus technical

[^11]Table 5 Cost Saving Technical Progress and Input Bias Components for the US

| Year | $\kappa^{t}$ | $\kappa_{1}^{t}$ | ${ }_{2}$ | $\kappa_{3}^{t}$ | $\kappa_{4}^{t}$ | $\kappa_{5}^{t}$ | $\kappa_{6}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | -0.0096 | 0.0008 | -0.0090 | -0.0010 | -0.0001 | 0.0007 | -0.0010 |
| 1971 | -0.0097 | 0.0008 | -0.0091 | -0.0009 | -0.0001 | 0.0006 | -0.0011 |
| 1972 | -0.0098 | 0.0008 | -0.0093 | -0.0006 | -0.0001 | 0.0005 | -0.0012 |
| 1973 | -0.0097 | 0.0008 | -0.0095 | -0.0002 | -0.0001 | 0.0005 | -0.0011 |
| 1974 | -0.0094 | 0.0007 | -0.0100 | 0.0001 | 0.0000 | 0.0006 | -0.0009 |
| 1975 | -0.0079 | 0.0008 | -0.0101 | 0.0000 | 0.0000 | 0.0022 | -0.0009 |
| 1976 | -0.0080 | 0.0008 | -0.0099 | -0.0001 | 0.0000 | 0.0022 | -0.0010 |
| 1977 | -0.0082 | 0.0008 | -0.0099 | -0.0001 | 0.0000 | 0.0022 | -0.0011 |
| 1978 | 0.0034 | 0.0013 | 0.0014 | 0.0000 | 0.0000 | 0.0021 | -0.0013 |
| 1979 | 0.0032 | 0.0012 | 0.0013 | 0.0001 | 0.0000 | 0.0020 | -0.0014 |
| 1980 | -0.0007 | 0.0011 | 0.0011 | -0.0036 | 0.0001 | 0.0019 | -0.0014 |
| 1981 | -0.0009 | 0.0011 | 0.0012 | -0.0035 | 0.0001 | 0.0018 | -0.0016 |
| 1982 | -0.0009 | 0.0010 | 0.0012 | -0.0033 | 0.0001 | 0.0017 | -0.0016 |
| 1983 | -0.0117 | 0.0011 | -0.0078 | -0.0035 | -0.0016 | 0.0018 | -0.0018 |
| 1984 | -0.0118 | 0.0011 | -0.0077 | -0.0036 | -0.0018 | 0.0019 | -0.0017 |
| 1985 | -0.0118 | 0.0011 | -0.0079 | -0.0035 | -0.0019 | 0.0019 | -0.0015 |
| 1986 | -0.0120 | 0.0011 | -0.0082 | -0.0034 | -0.0020 | 0.0019 | -0.0014 |
| 1987 | 0.0006 | 0.0011 | 0.0041 | -0.0049 | -0.0019 | 0.0037 | -0.0014 |
| 1988 | -0.0034 | -0.0031 | 0.0040 | -0.0045 | -0.0019 | 0.0036 | -0.0016 |
| 1989 | -0.0027 | -0.0030 | 0.0039 | -0.0043 | -0.0019 | 0.0035 | -0.0009 |
| 1990 | -0.0027 | -0.0031 | 0.0040 | -0.0040 | -0.0017 | 0.0032 | -0.0010 |
| 1991 | -0.0024 | -0.0030 | 0.0040 | -0.0038 | -0.0016 | 0.0031 | -0.0011 |
| 1992 | -0.0123 | -0.0029 | -0.0044 | -0.0038 | -0.0016 | 0.0026 | -0.0021 |
| 1993 | -0.0075 | 0.0019 | -0.0044 | -0.0039 | -0.0018 | 0.0027 | -0.0021 |
| 1994 | -0.0074 | . 0020 | -0.0043 | -0.0041 | -0.0020 | 0.0029 | -0.0019 |
| 1995 | -0.0072 | 0.0020 | -0.0043 | -0.0042 | -0.0021 | 0.0030 | -0.0016 |
| 1996 | -0.0072 | 0.0020 | -0.0044 | -0.0041 | -0.0023 | 0.0031 | -0.0014 |
| 1997 | -0.0080 | 0.0019 | -0.0045 | -0.0040 | -0.0033 | 0.0031 | -0.0012 |
| 1998 | -0.0081 | 0.0018 | -0.0047 | -0.0037 | -0.0034 | 0.0030 | -0.0011 |
| 1999 | -0.0080 | 0.0018 | -0.0049 | -0.0034 | -0.0035 | 0.0030 | -0.0010 |
| 2000 | -0.0076 | 0.0018 | -0.0052 | -0.0031 | -0.0035 | 0.0031 | -0.0008 |
| 2001 | -0.0062 | 0.0017 | -0.0054 | -0.0028 | -0.0016 | 0.0025 | -0.0005 |
| 2002 | -0.0064 | 0.0017 | -0.0055 | -0.0027 | -0.0016 | 0.0024 | -0.0007 |
| 2003 | -0.0065 | 0.0017 | -0.0055 | -0.0025 | -0.0016 | 0.0023 | -0.0009 |
| 2004 | -0.0076 | 0.0017 | -0.0056 | -0.0023 | -0.0021 | 0.0021 | -0.0014 |
| 2005 | -0.0074 | 0.0017 | -0.0056 | -0.0022 | -0.0020 | 0.0026 | -0.0019 |
| 2006 | -0.0037 | 0.0016 | -0.0015 | -0.0019 | -0.0017 | 0.0024 | -0.0026 |
| 2007 | -0.0026 | -0.0011 | -0.0016 | 0.0010 | -0.0015 | 0.0021 | -0.0015 |
| 2008 | -0.0027 | -0.0012 | -0.0017 | 0.0009 | -0.0012 | 0.0020 | -0.0015 |
| 2009 | -0.0026 | -0.0011 | -0.0017 | 0.0010 | -0.0014 | 0.0021 | -0.0014 |
| 2010 | -0.0039 | -0.0010 | -0.0017 | -0.0003 | -0.0017 | 0.0021 | -0.0012 |
| 2011 | -0.0037 | -0.0011 | -0.0018 | -0.0003 | -0.0019 | 0.0022 | -0.0010 |
| 2012 | -0.0036 | -0.0010 | -0.0017 | -0.0003 | -0.0021 | 0.0023 | -0.0008 |
| 2013 | -0.0036 | -0.0010 | -0.0017 | -0.0003 | -0.0023 | 0.0024 | -0.0007 |
| 2014 | -0.0034 | -0.0009 | -0.0018 | -0.0002 | -0.0024 | 0.0025 | -0.0005 |
| 2015 | -0.0034 | -0.0008 | -0.0017 | -0.0002 | -0.0025 | 0.0024 | -0.0005 |
| 2016 | -0.0035 | -0.0008 | -0.0018 | -0.0002 | -0.0025 | 0.0024 | -0.0006 |
| 2017 | -0.0037 | -0.0008 | -0.0018 | -0.0002 | -0.0026 | 0.0023 | -0.0006 |
| 2018 | -0.0036 | -0.0008 | -0.0018 | -0.0002 | -0.0025 | 0.0023 | -0.0007 |
| 2019 | -0.0037 | -0.0008 | -0.0018 | -0.0002 | -0.0025 | 0.0022 | -0.0008 |
| 2020 | -0.0030 | -0.0008 | -0.0019 | -0.0001 | -0.0017 | 0.0021 | -0.0006 |
| 2021 | -0.0119 | 0.0065 | -0.0167 | -0.0004 | -0.0018 | 0.0014 | -0.0007 |
| 2022 | -0.0117 | 0.0067 | -0.0165 | -0.0005 | -0.0018 | 0.0014 | -0.0009 |
| Mean | -0.0058 | 0.0005 | -0.0040 | -0.0018 | -0.0016 | 0.0022 | -0.0012 |

progress reduced the demand for these inputs over time. By far the biggest contribution to overall cost reduction due to technical progress was made by Labour. Since on average, $\kappa_{1}^{t}$ and $\kappa_{5}^{t}$ were positive, technical progress was Import augmenting and Other Capital (R\&D and Inventories) augmenting.
Table 5 shows a limitation of our econometric specification. We used piece-wise linear splines to model technical progress and so as time $t$ moves through a break point, the derivatives with respect to time of the spline functions $f_{n i}(t)$ change in a discontinuous manner and this leads to discontinuous changes in the $\kappa_{n}^{t}$. Thus $\kappa_{2}^{t}$ changes in a discontinuous manner as the spline functions $f_{2 i}(t)$ move through break points. The discontinuous movements in $\kappa_{2}^{t}$ largely determine the discontinuous movements in $\kappa^{t}$.
For additional information on the index number and nonparametric methods for measuring technical progress, see Appendices A and C.
For some purposes, it is useful to switch to estimating a GDP function or a Gross Output function or a Variable Profit function where only a few inputs are fixed and all outputs and variable inputs are free to vary. However, as we indicated in Section 3, better global fits can be obtained by estimating joint cost functions if we are using macroeconomic data. It turns out to be possible to get elasticity estimates for these alternative functions using our estimated joint cost function. We show how this can be done in Appendix D.

## 6 Extensions and Areas for Future Research

It would be of interest to estimate models that used more disaggregated data. ${ }^{* 24}$ As is explained in Appendix A, the Asian Productivity Data Base has data on more than 20 types of capital for the US for the years 1970-2022. This data base also has information on hours worked by workers classified by: education ( 2 classes), sex ( 2 classes), age ( 5 classes) and type of worker ( 2 , self employed and employee) or 40 types of labour in all.
However, as the number $N$ of inputs and outputs increases, the number of parameters required for a basic flexible functional form is $N(N-1) / 2$ and the number of degrees of freedom for a sample of $T$ periods is only $N T$ so the curse of dimensionality becomes a problem. Diewert and Wales (1988)[24] addressed this problem by limiting the number of columns that are allowed to enter into the $\mathbf{U}$ and $\mathbf{V}$ lower triangular matrices defined in Section 4. The same idea could be applied to the $\mathbf{D}$ matrix defined in Section 4; i.e., set $\mathbf{D}$ equal to a limited sum of rank one matrices. ${ }^{* 25}$
The choice of the fixed vectors $\boldsymbol{\alpha}$ and $\boldsymbol{\beta}$ also needs to be examined: are there better ways of choosing $\boldsymbol{\alpha}$ and $\boldsymbol{\beta}$ ?
Are there more efficient ways for specifying technical progress? Should we use quadratic or cubic splines in place of the linear splines that we used? Should we impose nonnegativity restrictions on the components of the $\mathbf{D}$ matrix? Answers to these questions require additional research.

[^12]Finally, an extension of the competitive model that we used is possible. Recall equations (6) which equated output prices $p_{m}^{t}$ to the marginal cost $\partial C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}\right) / \partial y_{m}$ for $m=1, \ldots, M$. These equations can be replaced by the following equations for year $t$ and $m=1, \ldots, 4$ :

$$
\begin{equation*}
p_{m}^{t}=\left(1-\mu_{m}^{t}\right)^{-1} \partial C\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}\right) / \partial y_{m} \tag{38}
\end{equation*}
$$

where $\mu_{m}^{t}$ is the year $t$ ad valorem markup for output $m .^{* 26}$ If we assume that the markup is constant over time, then this new specification for equations (6) would add an additional 4 markup parameters $\mu_{m}$ to be estimated. ${ }^{* 27}$ However, the underlying data for this more general model would have to be adjusted; i.e., we set the cost of capital equal to a rate of return that causes cost to equal the value of output. For the markup model, we would need to specify an exogenous cost of capital. This is a tricky business: if the cost of capital is set to a high level, markups could become negative. If the cost of capital is set too low, then the resulting markups will be too large.
There are many other problems with our empirical application of flexible functional form theory:

- The General Government and Owner Occupied Housing Sectors should be treated separately from the Business Sector.
- There are missing assets in our model such as Natural Resource Deposits. Environmental Assets and Monetary Holdings.
- An exogenous cost of capital probably should be used in place of our endogenously determined $r^{t}$.
- There is a general problem with how to define expected asset inflation rates and how to treat negative user costs when land and natural resource stocks are included in the list of assets.

A host of additional measurement problems are discussed in the excellent survey paper on productivity measurement by Martin and Riley (2024)[49].

## 7 Conclusion

Here are some of the important points that emerge from our paper:

- It is important to include land as a factor of production in macroeconomic models of the economy.
- It may be better to estimate joint cost functions rather than gross output functions when working with macroeconomic data.
- The Normalized Quadratic Joint Cost function can be used to model aggregate production for an economy but the estimation is quite complex. The advantage of this flexible functional form is that it can impose curvature conditions globally and it can be used to model technical progress in a way that captures longer terms trends in the more variable index number and nonparametric estimates of TFP growth.
- Elasticities of input substitution are far from being constant for aggregate production functions once we disaggregate capital services. For our 6 input model, we found that 8 pairs of inputs were substitutes and 7 pairs were complements. Macroeconomic models based on Cobb-Douglas or CES production functions are not satisfactory descriptions of reality.

[^13]- Sample wide linear trends used to model technical progress were not satisfactory. Linear spline functions were used to model technical progress with some success.
- A next step in using the methodology outlined in this paper is to include markups in the model. Instead of setting output price equal to marginal cost, set output price equal to a markup plus marginal cost. However, this extension requires exogenous estimates for the economy wide cost of capital and it is difficult to obtain a definitive estimate for this important variable.


## Appendix A: Data Construction

In this paper, we use the 1970-2022 US annual data on the prices and quantities of the macroeconomic aggregates $C+G+I+X-M$ and the data for labour and capital stocks developed by the joint project of the Asian Productivity Organization and Keio University led by Koji Nomura. The Augmented Productivity Database (APDB) for the US was produced on November 15, 2023.*28 The APDB output data are consistent with the recent data published by the Bureau of Economic Analysis (2023)[5], except that the BEA data allocates government investment into the general government sector whereas the APDB data includes government investment in the APDB investment aggregate. The constant dollar APDB investment, depreciation and reproducible capital stock data are perfectly consistent with the geometric model of depreciation.
The APDB has annual current and constant dollar estimates for 10 reproducible capital stocks plus inventory change plus 6 types of land. These 17 constant dollar capital stocks are beginning of the year estimates but the corresponding prices are midyear prices. We convert these midyear prices into beginning of the year prices by taking the arithmetic average of the current midyear price and the previous year midyear price. The APDB also has detailed information on hourly wage rates and annual hours worked for many types of labour classified by age, sex, education and type of worker (employee or self-employed). We will not make use of the detailed labour information: we simply used the resulting APDB aggregate quality adjusted price and quantity of labour for year $t, P_{L}^{t}$ and $Q_{L}^{t}$.
The price indexes for the output aggregates for year $t$ are defined as $P_{C}^{t}$ (private consumption) ${ }^{* 29}, P_{G}^{t}$ (government consumption), $P_{I}^{t}$ (gross investment), $P_{X}^{t}$ (exports of goods and services) and $P_{M}^{t}$ (imports of goods and services)*30. The corresponding quantity or volume indexes are defined as $Q_{C}^{t}, Q_{G}^{t}, Q_{I}^{t}, Q_{X}^{t}$ and $Q_{M}^{t}$. The 11 components of the APDB investment series are as follows: (1) IT hardware; (2) Communications equipment; (3) Transport equipment; (4) Other machinery and equipment; (5) Dwelling structures; (6) Non-residential buildings; (7) Other structures; (8) Research and development; (9) Computer software; (10) Other intangible assets and (17) Net increase in inventory stocks. ${ }^{* 31}$ Denote the year $t$ price index for investment good $n$ by $P_{I n}^{t}$ and the corresponding quantity or volume index by $Q_{I n}^{t}$ for $n=1, \ldots, 17$ and $t=1970, \ldots, 2022$. The APDB price series for reproducible investments are listed in Table A1 and the corresponding quantity series are listed in Table A2. The units of measurement are in trillions of 1970 constant dollars.

[^14]Table A1: Price Indexes for US Reproducible Investments

| Year | $P_{I 1}$ | $P_{I 2}$ | $P_{13}$ | P | $P_{I 5}$ | $P_{I 6}$ | $P_{I 7}$ | $P_{I 8}$ | $P_{\text {I9 }}$ | $P_{I 10}$ | $P_{I 17}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 |
| 1971 | 0.67587 | 1.01677 | 1.02154 | 1.02539 | 1.05098 | 1.09624 | 1.07348 | 1.04382 | 1.05603 | 1.08889 | 0.93258 |
| 1972 | 0.55998 | 1.06170 | 1.04238 | 1.04236 | 1.11280 | 1.17834 | 1.13004 | 1.07645 | 1.22239 | 1.12294 | 1.00000 |
| 1973 | 0.50039 | 1.08037 | 1.05074 | 1.07326 | 1.21141 | 1.26395 | 1.22238 | 1.12088 | 1.35754 | 41 | 7 |
| 19 | 0.43216 | 1. | 7 | 1. | 1.34337 | 1.40873 | 1.44461 | 1.22354 | 24 | 3 | 22 |
| 1975 | 0.39915 | 1.30770 | 0601 | 1.42353 | 21 | 2 | 0 | 1 | 6 | 8 | 00 |
| 1976 | 0.33373 | 1.37345 | 1.41440 | 1.53608 | 1.56882 | 1.62011 | 1.61517 | 1.38775 | 1.83141 | 1.42098 | 1.40164 |
| 1977 | 0.28188 | 1.32475 | 1.51750 | 1.63810 | 1.71554 | 1.73241 | 1.66631 | 1.46302 | 2.15563 | 1.48286 | 1.45752 |
| 1978 | 0.19059 | 1.37678 | 1.63484 | 1.76659 | 1.92480 | 1.90627 | 1.77852 | 1.54458 | 2.14661 | 1.59709 | 1.59586 |
| 1979 | 0.15771 | 1.4 | 1.76883 | 1.91389 | 2.13733 | 2.11003 | 1.94016 | 1.64948 | 2.67716 | 1.66546 | 1.81351 |
| 1980 | 0.12164 | 1.4959 | 1.94683 | 2.15 | 2.37790 | 2.35708 | 2.01758 | 1.80049 | 2.94414 | 4 | 4 |
| 19 | 0. | 1. | 2. | 2.3 | 2. | 2.55678 | 4 | 3 | 3.96771 | 4 | 1 |
| 1982 | 0.0 | 1.7 | 2. | 2.5 | . 6 | 2.73185 | 3 | 1 | 6 | 2 | 3 |
| 1983 | 0.0 | 1.7 | 2.3 | 2.6 | 2 | 2.83119 | 5 | 5 | 7 | 3 | 8 |
| 1984 | 0.06584 | 1.83407 | 2.44665 | 2.66571 | 2.81691 | 2.92301 | 2.26529 | 2.24344 | 6.32033 | 2.09859 | 2.35766 |
| 1985 | 0.05582 | 1.87439 | 2.50910 | 2.69479 | 2.87259 | 3.02464 | 2.37306 | 2.29201 | 6.74004 | 2.19965 | 2.41811 |
| 1986 | 0.04802 | 1.88606 | 2.60075 | 2.81758 | 3.03598 | 3.14442 | 2.65302 | 2.33250 | 6.58133 | 2.30470 | 2.20688 |
| 19 | 0.04034 | 1.87379 | 2.60770 | 2.87636 | 3.17485 | 3.24433 | 2.70729 | 2.37182 | 6.43226 | 2.38810 | 2.30719 |
| 19 | 0.03722 | 1.85554 | 2.63154 | 2.96440 | 3.29146 | 3.37893 | 2.71097 | 2.44625 | 7.04110 | 2.48258 | 2.38453 |
| 19 | 0.03474 | 1.8589 | 2.72511 | 3.05 | 3.41345 | 3.50560 | 2.84825 | 2.47653 | 7.94874 | 0 | 2.55637 |
| 19 | 0.0 | 1.857 | 2.7 | 3. | 2 | 0 | 3 | 2.50335 | 3 | 4 | 9 |
| 19 | 0.02821 | 1.86775 | 2.90313 | 3.30 | . 50 | 3.72449 | 2.95890 | 2.54501 | 2 | 90359 | 4 |
| 19 | 0.02399 | 1.84335 | 2.98052 | 3.3 | .58 | 3.7 | 3.04836 | 2.58129 | 8.27947 | 2.94880 | 84 |
| 19 | 0.02027 | 1.80267 | 3.03054 | 3.40098 | 3.72808 | 3.85506 | 2.98750 | 2.59893 | 8.98822 | 2.97831 | 2.63679 |
| 199 | 0.01783 | 1.76728 | 3.12641 | 3.43791 | 3.87175 | 3.97008 | 3.05306 | 2.65155 | 9.29016 | 3.05778 | 2.59610 |
| 1995 | 0.01477 | 1.69672 | 3.19077 | 3.47123 | 4.01020 | 4.07819 | 3.35178 | 2.75317 | 9.50958 | 3.13220 | 2.73869 |
| 199 | 0.01115 | 1.64909 | 3.21131 | 3.48756 | 4.07233 | 4.12580 | 3.57472 | 2.75073 | 9.15994 | 3.20100 | 2.78156 |
| 199 | 0.00859 | 1.63449 | 3.22029 | 3.50559 | 4.21007 | 4.26485 | 3.47910 | 2.81676 | 9.46103 | 3.25780 | 2.58470 |
| 1998 | 0.00638 | 1.53893 | 3.20878 | 3.5122 | 4.33512 | 4.43467 | 3.71418 | 2.82870 | 9.64210 | 3.24257 | 2.50680 |
| 1999 | 0.00495 | 1.43565 | 3.23021 | 3.5246 | 4.50442 | 4.62237 | 3.89519 | 2.87925 | 10.06312 | 3.35064 | 2.50096 |
| 2000 | 0.00433 | 1.38248 | 3.26157 | 3.53518 | 4.71390 | 4.81433 | 3.96958 | 2.97622 | 10.62807 | 3.45741 | 2.55675 |
| 20 | 0.00357 | 1.3 | 3.28972 | 3.5 | 73 | 4.98888 | 4.12756 | 2.99321 | 10.55350 | 4 | 2.57458 |
| 2002 | 0.00310 | 1.23452 | 3.28924 | 3.61413 | 443 | 985 | 1 | 3.03320 | 9.92951 | 3.52655 | 2.55607 |
| 2 | 0.00276 | 1.05489 | 3.30241 | 3.65456 | 5.31084 | 5.23995 | 4.69601 | 3.10133 | 9.44297 | 3.57264 | 2.20047 |
| 2004 | 0.00256 | 0.95174 | 3.33990 | 3.70865 | 5.68881 | 5.51466 | 5.02488 | 3.16943 | 9.32880 | 3.59042 | 2.41006 |
| 2005 | 0.00226 | 0.90444 | 3.41053 | 3.81705 | 6.09701 | 5.98045 | 5.74836 | 3.23837 | 9.39630 | 3.60486 | 2.80335 |
| 2006 | 0.00198 | 0.84512 | 3.44258 | 3.88153 | 6.44831 | 6.44107 | 6.60118 | 3.28987 | 9.42468 | 3.63731 | 2.46026 |
| 2007 | 0.00177 | 0.78927 | 3.45684 | 3.97596 | 6.53946 | 6.86055 | 7.19371 | 3.36838 | 9.41765 | 3.67169 | 2.77872 |
| 2008 | 0.00161 | 0.70090 | 3.48075 | 4.09216 | 6.44872 | 7.10587 | 7.64303 | 3.45528 | 9.60404 | 3.66600 | 2.80756 |
| 2009 | 0.00149 | 0.63941 | 3.55534 | 4.1 | 6.23955 | 7.28187 | 7.56937 | 3.46005 | 9.41520 | 3.67803 | 2.75002 |
| 2010 | 0.00144 | 0.57676 | 3.58793 | 4.14727 | 6.19757 | 7.12688 | 7.67974 | 3.51914 | 9.14362 | 3.61282 | 3.07708 |
| 2 | 0.00139 | 0.54953 | 3.68260 | 23631 | 6.22240 | 7.28125 | 8.01027 | 3.59899 | 9.28274 | 3.59310 | 3.23853 |
| 2012 | 0.00138 | 0.51086 | 3.77482 | 4.32437 | 6.28879 | 7.45452 | 8.45721 | 3.67948 | 9.37802 | 3.62280 | 3.19538 |
| 2013 | 0.00137 | 0.47782 | 3.79586 | 4.32164 | 6.59165 | 7.56251 | 8.51778 | 3.65614 | 9.50706 | 3.61019 | 3.16564 |
| 2014 | 0.00138 | 0.42995 | 3.84272 | 4.34573 | 6.99714 | 7.79153 | 8.94102 | 3.70782 | 9.52453 | 3.64910 | 3.09468 |
| 2015 | 0.00138 | 0.38385 | 3.90679 | 4.36940 | 7.21251 | 7.95558 | 9.12524 | 3.76212 | 9.47558 | 3.72009 | 3.25672 |
| 2016 | 0.00136 | 0.34207 | 3.93791 | 4.36464 | 7.47320 | 8.09039 | 9.10904 | 3.69651 | 9.50454 | 3.79522 | 3.63563 |
| 2017 | 0.00136 | 0.32079 | 3.95331 | 4.37696 | 7.81323 | 8.27618 | 9.39472 | 3.76160 | 9.50764 | 3.84454 | 3.10563 |
| 2018 | 0.00137 | 0.30203 | 4.03456 | 4.43593 | 8.27303 | 8.62899 | 9.26953 | 3.86586 | 9.37476 | 3.90106 | 3.22573 |
| 2019 | 0.00134 | 0.28488 | 4.08707 | 4.49939 | 8.51296 | 9.12736 | 9.48236 | 3.93387 | 9.29080 | 3.94401 | 3.14483 |
| 2020 | 0.00131 | 0.27304 | 4.13364 | 4.53565 | 8.79649 | 9.46904 | 9.60616 | 4.10597 | 9.14762 | 4.00237 | 3.90540 |
| 2021 | 0.00133 | 0.26205 | 4.27552 | 4.68426 | 9.69699 | 9.86040 | 9.99053 | 4.21436 | 8.95539 | 4.09541 | 2.90687 |
| 2022 | 0.00139 | 0.25988 | 4.68266 | 5.07353 | 11.07624 | 11.85630 | 10.78711 | 4.37195 | 8.86819 | 4.35736 | 3.81354 |

Table A2: Quantity Indexes for US Reproducible Investments

|  |  |  |  |  |  |  |  |  | $Q_{I 9}$ | $Q_{I 10}$ | $Q_{I 17}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.00325 | 0.00782 | 0.01841 | 0.05139 | 0.04397 | 0.03348 | 0.03378 | 0.02844 | 0.00313 | 0.00413 | 0.00200 |
|  | 0.00474 | 0.00742 | 0.01948 | 0.04784 | 0.05604 | 0.03290 | 0.03303 | 0.02807 | 0.00306 | 0.00408 | 0.00890 |
| 1972 | 00707 | 0.00703 | 0.02278 | 0.05272 | 0.06532 | 0.03349 | 0.03306 | 0.02918 | 0.00305 | 0.00439 | 0 |
|  | 0.00782 | 0.00838 | 0.02724 | 0.06210 | 0.06472 | 0.03594 | 0.03379 | 2 | 6 | 0 | 0 |
|  | 0.01028 | 0.00865 | 0.02483 | 0.06587 | 0.05167 | 0.03501 | 0.03358 | 0 | 0.00378 | 6 | 3 |
|  | 0.01043 | 0.0 | 0.02143 | 0.05923 | 0.0 | 0. | 0. | 0. | 0. | 0.00412 | -0.00450 |
|  | 0.01533 | 0.00897 | 0.02400 | 0.06 | 0.05532 | 0.02946 | 0.036 | 0.03120 | 0.00396 | 0.00524 | . 01220 |
| 19 | 0.02321 | 0.01187 | 0.02889 | 0.06585 | 0.06733 | 0.02886 | 0.03 | 0.03266 | 0.00355 | 0.00585 | . 01530 |
| 19 | 0.04530 | 0. | 0.0 | 0. | 0.0 | 0.03325 | 0. | 0.03483 | 0. | 0.00575 | 17 |
| 1979 | 0.07349 | 0.01647 | 0 | 0.07801 | 0. | 0. | 0 | 0.03707 | 0.00425 | 9 | 3 |
| 1980 | 0.11827 | 0.01862 | 0 | 0.07412 | 0.0 | 0. | 0.05055 | 0.03905 | 9 | 7 | 9 |
| 1981 | 0.18528 | 0.01945 | 0.02600 | 0.07538 | 0.05039 | 0.04106 | 0.05243 | 6 | 0.00422 | 2 | 3 |
|  | 0.22937 | 0.01987 | 0.02298 | 0.06863 | 0.04161 | 0.04189 | 0.04802 | 6 | 0.00413 | 7 | 8 |
|  | 0.35506 | 0.02007 | 0.02590 | 0.06697 | 0.05919 | 0.03895 | 0.04307 | 5 | 3 | 7 | 73 |
|  | . 57983 | . 02 | 0. | 0.0 | 0.06 | 0.04517 | 0.04908 | 0.04996 | 0.00 | 0.00855 | 74 |
|  | 0.74162 | 0.02425 | 0.03025 | 0.07963 | 0.06985 | 0.05127 | 0.04949 | 0.05479 | 0.00488 | 0.00851 | . 000902 |
| 19 | 0.86994 | 0.02593 | 0.02809 | 0.08089 | 0.07793 | 0.04889 | 0.04074 | 0.05710 | 0.00544 | 0.00940 | 299 |
| 1987 |  | 0.0262 | 0.02748 | 0.0 | 0.0 | 0.0 | 0.0 | 0.05950 | 0. | 0.00924 | 75 |
| 1988 | , | 0.0288 | 0.0 | 0.0 | 0.0 | 0.0 | 0. | 0.06141 | 0. | 0.00933 | 0.00776 |
| 1989 | 1.52323 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0. | 0.06403 | 0.00702 | 6 | 0.01084 |
|  | 1.52828 | 0.03022 | 0 | 0.08806 | 0 | 0.05222 | 0.04281 | 0.06691 | 0.00774 | 0 | 0.00585 |
|  | 1.66775 | 0.02920 | 0 | 0.08103 | 0 | 0.04556 | 0 | 3 | 0.00817 | 1 | 6 |
|  |  | 0.03098 |  | 0.08113 |  |  | 0.03959 | 9 | 0.00849 | 0 | 33 |
|  |  | 0.03221 |  | 0. |  |  |  | 0.06932 | 0. | 0.01128 | 89 |
|  | 3 | 0.0 | 0.03928 | 0.0 | 0. | 0.04382 | 0.0 | 0.06937 | 0.008 | 0.01188 | 58 |
|  | 5.10148 | 0.0 | 0.04068 | 0.09639 | 0.0 | 0.04900 | 0.04042 | 0.07160 | 0.0 | 0.01286 | 39 |
| 19 | 7.35439 | 0.04 | 0.04245 | 0.10002 | 0.08755 | 0.05400 | 0.04104 | 0.07639 | 0.0 | 0.01379 | 07 |
| 19 | 10.63263 | 0.05 | 0.04534 | 0.10340 | 0.08944 | 0.05811 | 0.0 | 0.08043 | . | 0.01420 | 43 |
| 1 | 15.363 | 0. | 0.0 | 0.11 | 0.0 | 0.0 | 0.0 | 0.08509 | . 0 | 0.01496 | 0.02541 |
| 1999 | 21.86683 | 0.0775 | 0.058 | 0.11 | 0.1 | 0.0 | 0.04292 | 0.09048 | 0.0 | 0.01531 | 1 |
| 20 | 26.413 | 0.1002 | 0.05526 | 0.1 | 0.1 | 0.06452 | 0.04804 | 0.09657 | 0. | 0.01622 | 32 |
|  | 2 | 0.09471 | 0.04939 |  |  |  |  |  |  | 2 |  |
|  | 2 | 0.07867 | 0 |  |  | 0.05551 |  |  | 0.01818 | 651 | 82 |
|  | 32.66864 | . 10 | 0.04237 | 0.12126 | 0.1210 | - | 0.01733 | 0.10118 | 0.01943 | 0.01775 | 41 |
| 2004 | 36.95599 | 0.10334 | 0.05071 | 0.12613 | 0.13277 | 0.05430 | 0.04606 | 0.10311 | 0.02097 | 0.01801 | 660 |
|  | 41.69228 | 0.11025 | 0.05692 | 0.13738 | 0.14135 | 0.05333 | 0.04642 | 0.10826 | 0.02231 | 0.01952 | 51 |
| 6 | 52.24049 | 0.1307 | 0.0613 | 0.14598 | 0.13063 | 0.05616 | 0.04799 | 0.11440 | 0.02359 | 0.01973 | 05 |
| 20 | 60.17714 | 0.1555 | 0.05 | 0.15 | 0.10606 | 0.06145 | 0.05067 | 0.12005 | 0.02563 | 0.01932 | 24 |
| 20 | 66.22915 | 0.1680 | 0.047 | 0.151 | 0.0 | 0.0 | 0.05306 | 0.12302 | 0.0271 | 0.01820 | -0.01040 |
| 20 | 70 | 0.16532 | 0 | 0 | 0. | 0.05258 | 0.04903 | 0.12129 | 0.02800 | 0.01753 | -0.05484 |
|  | 79 | 0.20634 | 0 | 0.13451 | 0 | 0.04178 | 0.04872 | 6 | 0.02882 | 228 | 52 |
|  | 76 | 0.21959 | 0.0 | 0.14572 | 0.0 | 0 | 0.04981 | 69 | 0.03111 | 0.01930 | 430 |
|  | 82.44852 | 0.23668 | 0.06272 | 0.15519 | 0.06852 | 0.04116 | 0.05230 | 0.12618 | 0.03346 | 0.01954 | 0.02228 |
| 2013 | 80.66736 | 0.27031 | 0.06844 | 0.15905 | 0.07709 | 0.04144 | 0.05133 | 0.13260 | 0.03507 | 0.01959 | 0.03333 |
| 20 | 79.38585 | 0.31051 | 0.07519 | 0.16693 | 0.08010 | 0.04449 | 0.05514 | 0.13725 | 0.03729 | 0.02025 | 0.02740 |
| 20 | 79.18422 | 0.3694 | 0.08348 | 0.16287 | 0.08866 | 0.05095 | 0.05012 | 0.14269 | 0.0392 | 0.02101 | 0.04302 |
| 2016 | 79.01030 | 0.42905 | 0.07814 | 0.16096 | 0.09502 | 0.05562 | 0.04422 | 0.15348 | 0.0432 | 0.02187 | 0.01075 |
| 20 | 84.12 | 0.49 | 0.07777 | 0.1680 | 0.0988 | 0.05575 | 0.04589 | 0.15864 | 0.0476 | 0.02257 | 0.01053 |
| 20 | 95.24785 | 0.50362 | 0.08402 | 0.17350 | 0.09827 | 0.05669 | 0.05029 | 0.16728 | 0.05349 | 0.02333 | . 01748 |
| 20 | 97.26011 | 0.51003 | 0.08119 | 0.17929 | 0.09750 | 0.05734 | 0.05250 | 0.17942 | 0.05712 | 0.02362 | 0.02296 |
| 2020 | 108.15529 | 0.51269 | 0.05890 | 0.16984 | 0.10451 | 0.05509 | 0.04663 | 0.18495 | 0.06204 | 0.02265 | 0.00963 |
| 2021 | 119.20516 | 0.56846 | 0.05590 | 0.17881 | 0.11672 | 0.05154 | 0.04468 | 0.20001 | 0.06999 | 0.02226 | 0.00402 |
| 2022 | 124.82850 | 0.64307 | 0.05377 | 0.18309 | 0.10620 | 0.04859 | 0.04404 | 0.21028 | 0.07921 | 0.02420 | 0.04125 |

The year $t$ overall investment price $P_{I}^{t}$ is set equal to the chained Törnqvist price index ${ }^{* 32}$ of the 11 reproducible investment components listed in the above tables and the year $t$ companion constant dollar aggregate investment is denoted by $Q_{I}^{t}$. The APDB aggregate price and quantity component series of GDP (at producer prices) are listed below in Table A3 along with the APDB quality adjusted price and quantity of labour. The units are in trillions of 1970 constant dollars.

[^15]Table A3: Prices and Quantities for Labour and the Output Components of US GDP

| Y |  | $P_{G}^{t}$ | I |  |  |  |  |  | Q |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 1.0000 | 1.0000 | 1.0000 | 1.0000 | 1.0000 | 1.0000 | 0.6035 | 0.1928 | 0.229 | 0.0597 | 0.0583 | 0.6584 |
|  | 1.0406 | 1.0 | 1.0428 | 1.0368 | 1.0659 | 1.0 | 0.6265 | 0.1930 | 0.2449 | 0.0608 | 0.0615 | 0.6552 |
| 1972 | 1.0795 | 1.1 | 1.0878 | 1.0 | 1.1307 | 1.1446 | 0.6649 | 0.1936 | 0.2655 | 0.0656 | 0.0683 | 45 |
|  | 1. | 1.2456 | 3 | 1.2260 | 1.3220 | 1.2170 | 0.6978 | 0 | 0.2886 | 0.0777 | 4 | 0.7041 |
|  | 1.2528 | 1.3523 | 1.2856 | 1.5102 | 1.8775 | 1.3229 | 0.6919 | 0.1961 | 0.2728 | 9 | 8 | 5 |
|  |  | 1.4769 |  | 1.6643 |  |  | 0.7076 |  | 0.2376 | 0.0833 | 0.0621 | 2 |
|  | 1.4 | 1.5655 |  |  |  |  |  |  |  | 0.0870 |  | 3 |
| 1977 | 1.5379 | 1.6688 | 1.6068 | 1.7889 | 2.2806 | 1.6698 | 0.7786 | 0.2045 | 0.3048 | 0.0891 | 0.0824 | 41 |
| 19 | 1.6 | 1.7 | 1.7323 | 1.8 | 2.4507 | 1.8 | 0.8 | 0. | 0.3 | 0.0985 | 0. | 69 |
| 1979 | 1.7 | 1.9300 | 1. | 2 | 2. | 1. | 0. | 0. | 0.3500 | 0.1083 | 0.0910 | 9 |
| 1980 | 1.9 | 2 | 2 | 2.3419 | 3.5426 | 2 | 0.8294 | 0 | 0.3239 | 0.1199 | 0 | 4 |
|  | 2. | 2.3446 | 2.2481 | 2.5156 | 3.7429 | 2.3630 | 0.8410 | 0.2162 | 0.3463 | 0.1213 | 0.0872 | 4 |
|  | 2 | 2.5059 |  |  |  | 2.5495 |  |  | 0.3073 |  | 0 |  |
|  | 2.3 |  |  |  |  |  |  |  | 0.3302 |  | 0 |  |
|  |  |  |  |  |  |  |  |  | 0.4089 |  |  | 29 |
|  | 2.565 | 2.8477 | 2.5 | 2.4861 | 3.3458 | 2.9 | 0.9 |  | 0.4158 | 0.1220 | 0.1283 | 539 |
| 1986 | 2.6298 | 2.9097 | 2.6166 | 2.4460 | 3.3468 | 3.0760 | 1.0397 | 0.2536 | 0.4156 | 0.1312 | 0.1394 | 660 |
| 1987 | 2.7 | 3.0042 | 2.6590 | 2.5 | 3.5 | 3 | 1.0 | 0. | 0.4313 |  | 0.1477 | 0 |
|  | 2.8 | 3. | 2 | 2.6 | 3.7 | 3 | 1.1 | 0.2637 | . | 0. |  | 2 |
|  | 2.9 | 3.2 | 2 | 2.6722 | 3.7 | 3.4647 | 1.1 | 0.2718 | , | 0.1887 | 0.1603 |  |
|  | 3. | 3.3950 |  | 2.6874 |  | 3.6381 | 1.1761 | 0.2793 |  | 4 | 0.1660 | 0.9620 |
|  | 3.1 | 3 | 2.9225 | 2.7173 | 3.8627 | 3.8273 | 1.1784 | 0.2842 | 0.4237 | 9 | 8 | 3 |
|  |  |  |  |  |  |  | 1.2219 |  |  | 0.2341 | 0.1775 | 2 |
|  |  |  |  |  |  |  |  |  | 0.4688 |  | 28 |  |
|  | 3. | 3.8 | 3.0319 |  |  |  |  |  |  | 2 | 88 | 092 |
|  | 3.4 | 3.9625 | 3. | 2.8013 | 3.9572 | . 2057 | 1.352 | 0. | 0.5256 | 0.2902 | 0.2331 | 30 |
|  | 3.5 | 4.0525 |  | 2.7639 | 3.8802 |  | 1.3992 | 0.2 | 0.5690 | 0.3139 |  |  |
| 1997 | 3.6 | 4.1409 | 3.0800 | 2.7143 | 3.7399 | 4 | 1.4 | 0.2 | 0.6241 | 0.3514 | 0.2876 | 75 |
| 1998 | 3.6 | 4.2 | 3. | 2 | 3.5 | 4.6656 | 1.5 | 0.3 | 0.6758 | 0.3596 | 0.3212 | 1.1331 |
| 1999 | 3.6 | 4.38 | 3. | 2. | 3.5 | 4. | 1. | 0.3 | 0.7266 | 0. | 0.3585 | 2 |
| 2 | 3.793 | 4.58 | 3. | 2.6 | 3.6 | 5.1102 | 1.6925 | 0. | 0.7 | 0.4089 | 0.4051 | 6 |
|  |  |  |  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  | 0.4096 | 62 |
|  |  |  |  |  | . 665 | 5.6199 |  | 0.3127 |  | 0.3865 | 0.4306 |  |
|  | 4.1030 | 5.3227 | 3.3382 | 2.7760 | 3.8371 | 5.8675 | 1.9039 | 0.3479 | 0.8290 | 0.4238 | 0.4779 | 902 |
| 2005 | 4.2213 | 5.5929 | 3.4955 | 2.8719 | 4.0612 | 6.0713 | 1.9713 | 0.3508 | 0.8720 | 0.4532 | 0.5089 | 93 |
| 2006 | 4.3 | 5.8389 | 3.6116 | 2.9 | 4.2160 | 6. | 2.0 | 0.3550 | 0.900 | 0.4962 | 0.5416 | 67 |
| 2007 | 4.453 | 6.0955 | 3.6 | 3.0 | 4.3 |  | 2.0 | 0.3607 | 0.8 | 0.5397 | 0.5555 | 48 |
| 2008 | 4.5 | 6.36 | 3.7 | 3.2 |  | 6. | 2.0 | 0.3 | 0.830 | 0.5709 | 0.5437 | 3 |
| 2 | 4.5 | 6.3175 | 3. |  |  | 6.726 | 2.0 | 0.3852 | 0.6 | 0.5232 | 0.4745 |  |
|  |  | 6. |  | 3.1515 |  | 6.8358 | 2. | 0. | 0.7 | 0.5893 | 0.5358 |  |
|  | 4.7 | 6 | 3.7436 | 3.3500 | 4.8567 | 6.9345 | 2.1291 | 0.3734 | 0.7931 | 0.6316 | 0.5616 | 0 |
| 2012 | 4. | 6.8369 | 3.8118 | . 37 | 17 | 7.0539 | 2.158 | 0.3679 | 0.8507 | 0.6570 | 0.5753 | 1.2541 |
| 2013 | 4.924 | 7.0135 | 3.8367 | 3.3807 | 4.8111 | 7.1245 | 2.1959 | 0.3608 | 0.8967 | 0.6768 | 0.5824 | 1.2791 |
| 2010 | 4.9 | 7.1605 | 3.9069 | 3.382 | 4.7756 | 7.294 | 2.2580 | 0.3578 | 0.9420 | 0.7032 | 0.6125 | 1.3072 |
|  | 5.001 | 7.1527 | 3.9502 | 3.219 | 4.3964 | 7.4733 | 2.3342 | 0.364 | 0.991 | 0.7053 | 0.6444 | 1.3379 |
| 20 | 5.058 | 7.1628 | 3.9676 | 3.15 | 4.2468 | 7.5512 | 2.391 | 0.370 | 0.990 | 0.7087 | 0.6537 | 1.3604 |
| 20 | 5.1 | 7.3 | 4. | 3.23 | 4.3 | 7.755 | 2.45 | 0.3702 | 1.031 | 0.737 | 0.6845 | 1.3842 |
| 20 | 5.24 | 7.6194 | 4.0984 | . 344 | 4.4715 | . 9942 | 2.5219 | 0.3753 | 1.087 | 0.7589 | 0.7122 | 1.4119 |
| 20 | 5.3266 | 7.7279 | 4.1648 | 3.3280 | 4.4333 | . 2065 | 2.5724 | 0.3901 | 1.1207 | 0.7628 | 0.7206 | 1.4353 |
|  | 5.4091 | 7.9177 | 4.2398 | 3.2436 | 4.3373 | 8.6934 | 2.5076 | 0.4014 | 1.0767 | 0.6629 | 0.6560 | 1.3715 |
| 20 | 5.6180 | 8.3575 | 4.4042 | 3.6184 | 4.6577 | 9.1360 | 2.7186 | 0.4028 | 1.1451 | 0.7048 | 0.7509 | 1.4144 |
| 2022 | 5.9758 | 8.9418 | 4.7843 | 3.9724 | 4.9893 | 9.4202 | 2.7874 | 0.3993 | 1.1775 | 0.7540 | 0.8154 | 1.4682 |

We turn now to the APDB data on US capital stocks. The beginning of the year $t$ capital stocks for the above 10 reproducible investment assets are also available in the APDB. Denote the asset $n$ and year $t$ stock price index by $P_{K n}^{t}$ and the corresponding quantity index by $Q_{K n}^{t}$ for $n=1, \ldots, 10$ and $t=1970,1971, \ldots, 2022$. The APDB also has current and constant dollar series for 6 types of land used by the US production sector: (11) Agricultural Land; (12) Forest Land; (13) Industrial Land; (14) Commercial Land; (15) Residential Land and (16) Other Use Land. ${ }^{* 33}$ Denote the beginning of the year $t$ price index for these land assets by $P_{K n}^{t}$ and the corresponding quantity index by $Q_{K n}^{t}$ for $n=11, \ldots, 16$ and $t=1970, \ldots, 2022$. We also constructed a beginning of the year stock of inventories by cumulating the constant dollar inventory investments, $Q_{I 17}^{t}$. Denote the beginning of the year stock of inventories for year $t$ by $Q_{K 17}^{t}$ and the corresponding beginning of the year price by $P_{K 17}^{t}$. Thus Asset 17 is Inventory Stocks. The beginning of the year prices and quantities of these 17 capital stocks are listed in Tables A4 and A5 along with the price and quantity of a Törnqvist capital stock aggregate, $P_{K}^{t}$ and $Q_{K}^{t}$.

[^16]Table A4: Beginning of the Year Prices for US Capital Stocks

| Year | $P_{K}^{t}$ | $P_{K 1}^{t}$ | $P_{K 2}^{t}$ | $P_{K 3}^{t}$ | $P_{K 4}^{t}$ | $P_{K 5}^{t}$ | $P_{K 6}^{t}$ | $P_{K 7}^{t}$ | $P_{K 8}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 |
| 1971 | 1.03507 | 0.83828 | 1.00838 | 1.01077 | 1.01270 | 1.02549 | 1.04812 | 1.03674 | 1.02191 |
| 1972 | 1.10345 | 0.61853 | 1.03921 | 1.03196 | 1.03388 | 1.08189 | 1.13729 | 1.10176 | 1.06013 |
| 1973 | 1.18431 | 0.53068 | 1.07102 | 1.04656 | 1.05781 | 1.16210 | 1.22115 | 1.17621 | 1.09867 |
| 1974 | 1.30856 | 0.46660 | 1.13276 | 1.10345 | 1.13583 | 1.27739 | 1.33634 | 1.33350 | 2 |
| 1975 | 1.48624 | 0.41584 | 1.24642 | 1.23108 | 1.31097 | 379 | . 48493 | 50441 | 27338 |
| 1976 | 1.63924 | 0.36663 | 1.34056 | 1.36020 | 1.47981 | 1.52651 | 1.59062 | 1.58969 | 1.35549 |
| 1977 | 1.75868 | 0.30796 | 1.34909 | 1.46594 | 1.58709 | 1.64218 | 1.67626 | 1.64074 | 1.42539 |
| 1978 | 1.92623 | 0.23635 | 1.35076 | 1.57616 | 1.70235 | 1.82017 | 1.81935 | 1.72242 | 1.50381 |
| 1979 | 2.15475 | 0.17425 | 1.38880 | 1.70182 | 1.84024 | 2.03106 | 2.00815 | 1.85934 | 1.59704 |
| 1980 | 2.38301 | 0.13976 | 1.44840 | 1.85782 | 2.03366 | 2.25761 | 2.23356 | 1.97887 | 9 |
| 19 | 2.57 | 0.11417 | 1.55554 | 2.04140 | 2.25634 | 2.46970 | 2.45694 | 2.15131 | 36 |
| 1982 | 2.743 | 0.10239 | 1.67315 | 2.22042 | 2.44750 | 2.62307 | 2.64432 | 38563 | 2.01448 |
| 1983 | 2.8387 | 0.08995 | 1.75790 | 2.34280 | 2.58669 | 8 | 2.78152 | 34 | 44 |
| 19 | 2.85056 | 0.07382 | 1.80933 | 2.41369 | 2.65167 | 2.78061 | 2.87710 | 2.30488 | 2.20780 |
| 1985 | 2.89506 | 0.06086 | 1.85421 | 2.47786 | 2.68025 | 2.84475 | 2.97382 | 2.31918 | 2.26773 |
| 1986 | 3.01874 | 0.05195 | 1.88021 | 2.55491 | 2.75619 | 2.95429 | 3.08453 | 2.51304 | 2.31226 |
| 1987 | 3.18727 | 0.04421 | 1.87991 | 2.60421 | 2.84697 | 3.10542 | 3.19438 | 2.68016 | 2.35217 |
| 1988 | 3.36871 | 0.03880 | 1.86465 | 2.61960 | 2.92038 | 3.23315 | 3.31164 | 2.70913 | 2.40905 |
| 19 | 3.5349 | 0.03600 | 1.85725 | 2.67831 | 3.01098 | 3.35245 | 3.44227 | 2.77961 | 2.46140 |
| 19 | . 6597 | 0.03307 | 1.85806 | 2.76132 | 3. | 3 | 3.56450 | 4 | 5 |
| 199 | 3.681 | 0.02980 | 1.86244 | 2.85033 | 3.2 | 6 | 3.67395 | 7 | 2.52419 |
| 1992 | 3.65890 | 0.02611 | 1.85553 | 2.94181 | 3.34067 | 3.54588 | 3.73739 | 00363 | 2.56316 |
| 1993 | 3.63326 | 0.02214 | 1.82299 | 3.00552 | 3.38953 | 3.65607 | 3.80267 | 3.01793 | 2.59012 |
| 19 | 3.64411 | 0.01906 | 1.78496 | 3.07846 | 3.41945 | 3.79991 | 3.91257 | 3.02028 | 2.62525 |
| 1995 | 3.72662 | 0.01631 | 1.73198 | 3.15858 | 3.45457 | 3.94097 | 4.02414 | 3.20242 | 2.70237 |
| 1996 | 3.83155 | 0.01297 | 1.67289 | 3.20103 | 3.47940 | 4.04126 | 4.10200 | 3.46325 | 2.75196 |
| 1997 | 3.92026 | 0.00987 | 1.64177 | 3.21579 | 3.49658 | 4.14120 | 4.19533 | 3.52691 | 2.78376 |
| 1998 | 4.10025 | 0.00749 | 1.58669 | 3.21452 | 3.50892 | 4.27259 | 4.34977 | 3.59664 | 2.82274 |
| 1999 | 4.35702 | 0.00567 | 1.48728 | 3.21948 | 3.51845 | 4.41977 | 4.52853 | 3.80469 | 2.85399 |
| 2000 | 4.59168 | 0.00464 | 1.40905 | 3.24588 | 3.52992 | 4.60916 | 4.71836 | 3.93239 | 2.92775 |
| 2001 | 4.85769 | 0.00395 | 1.34558 | 3.27563 | 3.55495 | 4.82931 | 4.90161 | 4.04858 | 2.98473 |
| 2002 | 5.10694 | 0.00334 | 1.27160 | 3.28947 | 3.59442 | 5.00458 | 5.05937 | 4.33019 | 3.01322 |
| 2003 | 5.37056 | 0.00293 | 1.14469 | 3.29581 | 3.63435 | 5.18763 | 5.18491 | . 61441 | 3.06728 |
| 2004 | 5.72530 | 0.00266 | 1.00331 | 3.32113 | 3.68161 | 5.49982 | 5.37731 | 4.86045 | 3.13539 |
| 2005 | 6.33815 | 0.00241 | 0.92808 | 3.37520 | 3.76285 | 5.89291 | 5.74757 | 5.38663 | 3.20391 |
| 2006 | 7.00678 | 0.00212 | 0.87477 | 3.42654 | 3.84929 | 6.27266 | 6.21077 | 6.17477 | 3.26413 |
| 2007 | 7.31432 | 0.00188 | 0.81718 | 3.44969 | 3.92875 | 6.49388 | 6.65082 | 6.89744 | 3.32914 |
| 2008 | 7.32293 | 0.00169 | 0.74507 | 3.46877 | 4.03406 | 6.49409 | 6.98322 | 7.41837 | 3.41184 |
| 2009 | 6.97863 | 0.00155 | 0.67014 | 3.51802 | 4.12079 | 6.34413 | 7.19388 | 7.60620 | 3.45768 |
| 2010 | 6.45953 | 0.00147 | 0.60807 | 3.57162 | 4.14834 | 6.21856 | 7.20438 | 7.62456 | 3.48961 |
| 2011 | 6.25588 | 0.00142 | 0.56314 | 3.63525 | 4.19179 | 6.20998 | 7.20407 | 7.84501 | 3.55908 |
| 20 | 6.40211 | 0.00139 | 0.53019 | 3.72870 | 4.28034 | 6.25559 | 7.36789 | 8.23375 | 3.63925 |
| 2013 | 6.68223 | 0.00138 | 0.49433 | 3.78532 | 4.32301 | 6.44022 | 7.50852 | 8.48750 | 3.66782 |
| 2014 | 7.06991 | 0.00138 | 0.45388 | 3.81927 | 4.33369 | 6.79439 | 7.67703 | 8.72941 | 3.68199 |
| 2015 | 7.43901 | 0.00138 | 0.40690 | 3.87474 | 4.35757 | 7.10482 | 7.87356 | 9.03314 | 3.73498 |
| 2016 | 7.72553 | 0.00137 | 0.36295 | 3.92233 | 4.36703 | 7.34285 | 8.02299 | 9.11715 | 3.72933 |
| 2017 | 8.04857 | 0.00136 | 0.33143 | 3.94559 | 4.37081 | 7.64321 | 8.18329 | 9.25189 | 3.72907 |
| 2018 | 8.33643 | 0.00137 | 0.31141 | 3.99392 | 4.40645 | 8.04312 | 8.45260 | 9.33213 | 3.81374 |
| 2019 | 8.59752 | 0.00136 | 0.29345 | 4.06080 | 4.46766 | 8.39299 | 8.87819 | 9.37595 | 3.89988 |
| 2020 | 8.94065 | 0.00133 | 0.27896 | 4.11034 | 4.51752 | 8.65472 | 9.29821 | 9.54427 | 4.01993 |
| 2021 | 9.62338 | 0.00132 | 0.26754 | 4.20456 | 4.60996 | 9.24674 | 9.66473 | 9.79835 | 4.16018 |
| 2022 | 10.55863 | 0.00136 | 0.26096 | 4.47907 | 4.87890 | 10.38661 | 10.85837 | 10.38883 | 4.29317 |
| 2023 |  | 0.00139 | 0.25987 | 4.68263 | 5.07353 | 11.07624 | 11.85632 | 10.78712 | 4.37197 |


| Year | $P_{K 9}^{t}$ |  | $P_{K 11}^{t}$ | $P_{K 12}^{t}$ | $P^{t}$ | $P^{t}$ | $P^{t}$ | $P^{t}$ | $P^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 |
| 1971 | 1.02806 | 1.04442 | 1.04569 | 1.03841 | 1.02560 | 1.02560 | 1.10341 | 1.08980 | 1.00243 |
| 19 | 1.13928 | 1.10588 | 16544 | 1.11461 | 1.09190 | . 09190 | 098 | 25088 | 43 |
| 1973 | 1.29004 | 1.14 | 1.40807 | 1.19651 | 1.16534 | 1.16534 | 1.35133 | 1.44003 | 1.00270 |
| 1974 | 1.39695 | 1.19872 | 1.68306 | 1.30862 | 1.24831 | 1.24831 | 1.55640 | 1.70914 | 1.15279 |
| 1975 | 1.57257 | 351 | . 91557 | 1.49106 | 1.40287 | 87 | 862 | 09559 | 87 |
| 1976 | 1.77016 | 1.38096 | 2.25512 | 1.68524 | 1.56443 | 1.56443 | 2.32907 | 2.49828 | 1.39290 |
| 1977 | 1.99358 | 1.45187 | 2.61475 | 1.78678 | 1.66694 | 1.66694 | 2.63734 | 2.87993 | 1.40565 |
| 19 | 2.15117 | 1.53994 | 02686 | 1.88809 | 1.79607 | 1.79607 | 3.15812 | . 43 | 89 |
| 1979 | 2.41192 | 1.63125 | 3.59217 | 1.97162 | 1.98392 | 1.98392 | 3.90813 | 4.23965 | 1.59806 |
| 1980 | 2.81071 | 1.70271 | 4.10491 | 2.07364 | 2.20040 | 2.20039 | 4.43091 | 4.90281 | 1.81456 |
| 1981 | 3.45597 | 1.79125 | 4.30774 | 2.24286 | 2.40080 | 2.40080 | 4.52062 | 5.15616 | 1.92802 |
| 1982 | 4.35807 | 1.88755 | 4.17254 | 2.36509 | 2.68945 | 2.68945 | 4.36832 | 5.06171 | 2.15954 |
| 1983 | 5.38089 | 1.97424 | 4.05451 | 2.40745 | 2.90741 | 2.90741 | 4.33089 | 5.04141 | 2.14079 |
| 1984 | 6.16690 | 2.05721 | 3.76294 | 2.39445 | 2.84557 | 2.84557 | 4.66055 | . 23408 | 065 |
| 1985 | 6.53028 | 2.14907 | 21130 | 2.38952 | 2.77969 | 2.77969 | 5.20344 | 5.51337 | 2.36233 |
| 1986 | 6.66081 | 2.25213 | 81405 | 2.39676 | 2.75306 | 2.75306 | 6.07430 | 6.14914 | 417 |
| 1987 | 6.50696 | 2.34635 | 2.73662 | 2.41339 | 2.87630 | 2.80761 | 7.24808 | . 18919 | 2.20189 |
| 1988 | 6.73684 | 2.43529 | 2.86115 | 2.50361 | 3.17995 | 2.96241 | 8.40756 | . 34143 | 2.31500 |
| 1989 | 7.49507 | 2.55403 | . 97620 | 2.67445 | 3.43780 | 3.05146 | 9.36191 | . 37472 | 2.38629 |
| 1990 | 7.97792 | 2.68176 | 3.09466 | 2.87956 | 3.65152 | 3.10456 | 9.82555 | 96964 | 2.55962 |
| 1991 | 8.08371 | 2.82076 | 3.17470 | 3.03667 | 3.82909 | 3.03975 | 9.53577 | 9.85503 | 2.49128 |
| 19 | 8.22006 | 2.92613 | 25202 | 3.15246 | 3.98418 | .75380 | 9.13590 | . 64679 | 2.46724 |
| 1993 | 8.63403 | 2.96349 | 3.43081 | 3.36844 | 4.13513 | 2.35852 | 8.87841 | 9.57953 | 2.58443 |
| 1994 | 9.13938 | 3.01798 | . 62183 | 3.60974 | 4.30336 | 2.09527 | 8.74019 | . 59139 | 2.64310 |
| 19 | 9. | 3.09 | 45 | 3.78642 | 4.51136 | 2.08314 | 247 | 4 | 2.59999 |
| 1996 | 9.33494 | 3.16654 | 3.96694 | 3.93643 | 4.78022 | 2.14650 | 8.58164 | 9.74616 | 2.74223 |
| 1997 | 9.31067 | 3.22934 | 4.16517 | 4.07082 | 5.12684 | 2.24448 | 8.79453 | 10.10489 | 2.78756 |
| 19 | 9.55177 | 3.2 | 4.37084 | 4.23043 | 5.56081 | 2.72568 | 9.38040 | 10.89924 | 2.59072 |
| 1999 | 9.85281 | 3.29653 | 4.60353 | 4.36673 | 6.08130 | 3.26726 | 10.42696 | 12.22069 | 2.51246 |
| 2000 | 10.34583 | 3.40395 | 4.93527 | 4.46973 | 6.67444 | 3.44520 | 11.90367 | 14.06072 | 2.50650 |
| 20 | 10.59102 | 3.4847 | 28 | 4.5664 | 7.31174 | 3.71645 | 22 | 16.23182 | 2.56327 |
| 2002 | 10.24173 | 3.51932 | 5.45543 | 4.57454 | 7.95061 | 3.76246 | 15.36732 | 18.71792 | 2.57780 |
| 2003 | 68 | 3.54952 | 51233 | 4.56731 | 8.53766 | 75776 | 17.67116 | 21.66917 | 2.55926 |
| 20 | 088 | 3.5 | 6.03167 | 4.6323 | 9.01516 | 3.96208 | 502 | . 4 | 2.20298 |
| 2005 | 9.36275 | 3.59756 | 7.18595 | 4.75009 | 9.33032 | 4.45552 | 24.56166 | 31.01130 | 2.41321 |
| 2006 | 9.41070 | 3.62101 | 8.36558 | 4.93447 | 9.44589 | 5.33774 | 27.71376 | 35.68362 | 2.80954 |
| 2007 | , 42137 | 3.65 | 8.87795 | 5.1386 | 9.34992 | 6.14649 | 27.344 | 6.077 | 2.46424 |
| 2008 | 9.51105 | 3.66876 | 8.94943 | 5.33659 | 9.06169 | 6.88237 | 23.64786 | 32.09902 | 2.77903 |
| 2009 | 9.50982 | 3.67194 | 8.98778 | 5.50437 | 8.63181 | 6.52851 | 19.00617 | 26.59893 | 2.81705 |
| 2010 | 27960 | 3.64534 | 9.37018 | 5.52552 | 8.13549 | 4.53200 | 16.25553 | 23.43098 | 2.75460 |
| 2011 | 9.21338 | 3.60287 | 10.27922 | 5.48306 | 7.65966 | 3.61624 | 14.27961 | 21.48031 | 3.08338 |
| 2012 | 9.33059 | 3.60787 | 11.69926 | 5.48505 | 7.28683 | 4.19682 | 13.09658 | 20.72792 | 3.24133 |
| 2013 | 9.44274 | 3.61641 | 13.18125 | 5.54368 | 7.07956 | 4.48792 | 14.74880 | 23.44144 | 3.20238 |
| 2014 | 9.51600 | 3.62956 | 14.11635 | 5.66080 | 7.06933 | 5.05679 | 17.04571 | 26.77158 | 3.17072 |
| 2015 | 9.50026 | 3.68451 | 14.45474 | 5.79841 | 7.25248 | 5.87177 | 18.56820 | 28.97868 | 3.10056 |
| 2016 | 9.49027 | 3.75757 | 14.55009 | 5.90648 | 7.59409 | 6.47944 | 20.08394 | 31.18148 | 3.26343 |
| 2017 | 9.50629 | 3.81980 | 14.91295 | 5.96136 | 8.03723 | 6.97179 | 21.78121 | 33.72247 | 3.64123 |
| 2018 | 9.44140 | 3.87272 | 15.23799 | 6.00715 | 8.51739 | 7.47011 | 23.45716 | 36.22275 | 3.10892 |
| 2019 | 9.33298 | 3.92245 | 15.42207 | 6.05945 | 8.96974 | 7.68706 | 24.79480 | 38.26933 | 3.23321 |
| 2020 | 9.21941 | 3.97310 | 15.80484 | 6.09480 | 9.34980 | 8.36316 | 26.44699 | 40.67494 | 3.14917 |
| 2021 | 9.05170 | 4.04880 | 16.95692 | 6.11076 | 9.60509 | 9.71709 | 29.35484 | 44.99057 | 3.90865 |
| 2022 | 8.91198 | 4.22629 | 18.65843 | 6.18476 | 9.75741 | 11.21533 | 33.78415 | 51.78277 | 2.90857 |
| 2023 | 8.86838 | 4.35726 | 19.49921 | 6.25259 | 9.82352 | 12.07903 | 36.36039 | 55.73730 | 3.82082 |

Table A5: Beginning of the Year US Capital Stock Quantities

| Year | $Q_{K}^{t}$ | $t$ | $Q_{K 2}^{t}$ | $Q_{K 3}^{t}$ | $Q_{K 4}^{t}$ | $Q_{K 5}^{t}$ | $Q_{K 6}^{t}$ | \% | ¢8 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 3.19 | 01349 | 0.02022 | 0.07947 | 0.23245 | 0.55234 | 0.32395 | 0.50843 | 0.07483 |
| 1971 | 3.2 | .01229 | 0.02 | 0.07847 | 0.2 | 0.56894 | 0.33003 | 0.52885 | 0.08635 |
| 1972 | 3.35665 | 0.01271 | 0.02317 | 0.07864 | 0.23906 | 0.59680 | 0.33502 | 0.54801 | 0.09535 |
| 1973 | 3.45911 | 0.01498 | 0.02362 | 0.08168 | 0.24439 | 0.63231 | 0.34004 | 0.56644 | 0.10365 |
| 1974 | 3.57489 | 0.01724 | 0.02516 | 0.08807 | 0.25743 | 0.66572 | 0.34724 | 0.58539 |  |
| 1975 | 3.65371 | 0.02092 | 0.02655 | 0.09090 | 0.27156 | 0.68438 | 0.35287 | 0.60358 | 0.11713 |
| 1 | 3.70262 | 0.02365 | 0. | 0.09009 | 0. | 0.69563 | 0.35357 | 0.62285 | 0.12162 |
| 1977 | 3.7 | 6 | 0. | 0.09176 | 0.28298 | 0.71661 | 0.35318 | 0.64301 | 0.12675 |
| 1978 | 3.8 | 0.04078 | 0. | 2 | 0. | 0.74855 | 3 | 8 | 3 |
| 1979 | 4. | 0.06739 | 0.03615 | 0.10421 | 0.30725 | 0.78349 | 0.35575 | 2 | 3 |
| 19 | 4.14166 | 16 | 0.04 | 0.11084 | 0.3235 | 0.81373 | 0.36399 | 0.71650 | 84 |
| 1981 | 4.21806 | 0.17852 | 0.04775 | 0.11058 | 0.33338 | 0.82802 | 0.37201 | 0.74820 | 0.15348 |
| 1982 | 4.30964 | 0.28388 | 0.05306 | 0.10954 | 0.34269 | 0.83747 | 0.38165 | 0.78107 | 0.16167 |
| 1983 | 4.35062 | 0.39579 | 0.05743 | 0.10603 | 0.34416 | 0.83763 | 0.39128 | 0.80861 | 0.16993 |
| 1984 | 4.4 | 0.58179 | 0.0 | 0.10587 | 0.3 | 0.85516 | 0.39706 | 0.83029 | 6 |
| 19 | 4.5 | 0.9 | 0.0 | 0.10917 | 0.3 | 0.88057 | 0.40866 | 57 | 8 |
| 1986 | 4.70776 | 1.27020 | 0.07066 | 0.11220 | 0.3 | 0.90664 | 0.42529 | 0.88439 | 64 |
| 19 | 4.82847 | . 63784 | 0.0 | 0.11267 | 0. | 82 | 9 | 0.90207 | 63 |
| 19 | 4.93892 | . 10333 | 0.08030 | 9 | 0.37 | 0.97238 | 0.44973 | 0.91931 | 33 |
| 1989 | 5.05997 | . 54907 | 0.0858 | 0.11468 | 0.3865 | 1.00245 | 0.46082 | 0.93615 | 134 |
| 1990 | 5.16438 | 3.09502 | 0.08990 | 0.11334 | 0.3998 | 1.02742 | 0.47239 | 0.95002 | 0.25344 |
| 1991 | 5.25163 | 3.48668 | 0.0942 | 0.11254 | 0.4088 | 1.04569 | 0.48474 | 0.96804 | 0.26585 |
| 19 | 5.30371 | 3.88137 | 0.0 | 0.11271 | 0.4098 | 1.05728 | 0.48931 | 0.98451 | 0.27809 |
| 19 | 5.3692 | 4.64767 | 0.09 | 0.11387 | 0.4 | 1.07502 | 0.48993 | 0.99673 | 0.28861 |
| 19 | 5.4 | 5.65727 | 0.10 | 0.11907 | 0.4 | 1.09824 | 0.49000 | 1.01292 | 0.29652 |
| 19 | 5.55905 | 88216 | 0.1 | 0.12783 | 0.4233 | 1.12626 | 0.49173 | 1.02669 | 0.30297 |
| 1996 | 5.66495 | 9.20003 | 0.12125 | 0.13601 | 0.43554 | 1.15072 | 0.49889 | 1.03962 | 0.31021 |
| 1997 | 5.80444 | 12.75841 | 0.13420 | 0.14403 | 0.44902 | 1.18118 | 0.51065 | 1.05329 | 0.32042 |
| 1998 | 5.97484 | 18.06154 | 0.1495 | 0.15285 | 0.46313 | 1.21189 | 0.52549 | 1.06836 | 0.33234 |
| 1999 | 6.15790 | 25.83353 | 0.16852 | 0.16362 | 0.48103 | 1.24845 | 0.54247 | 1.08289 | 0.34621 |
| 2000 | 6.34231 | 36.85800 | 0.19498 | 0.18008 | 0.49706 | 1.28913 | 0.55851 | 1.09714 | 0.36233 |
| 2001 | 6.55247 | 48.52102 | 0.23495 | 0.18990 | 0.51724 | 1.32886 | 0.57587 | 1.11651 | 0.38090 |
| 2002 | 6.68621 | 57.45963 | 0.26016 | 0.19229 | 0.53110 | 1.36740 | 0.59062 | 1.13839 | 0.39844 |
| 2003 | 6.81927 | 65.30110 | 0.265 | 0.19094 | 0.54187 | 1.41070 | 0.59617 | 1.15524 | 0.41239 |
| 200 | 6.94936 | 73.89978 | 0.2793 | 0.18683 | 0.55566 | 1.46157 | 0.59945 | 1.17198 | 0.42562 |
| 2005 | 7.10395 | 83.54085 | 0.3008 | 0.19085 | 0.57085 | 1.52006 | 0.60229 | 1.18576 | 0.43808 |
| 20 | 7.25761 | 94.16128 | 0 | 0.19905 | 0.59215 | 1.58040 | 2 | 1.19660 | 84 |
| 20 | 7.42491 | 110.96673 | 0.3577 | 0.21014 | 0.61969 | 63288 | 0.60770 | 1.21329 | 0.47036 |
| 20 | 7.52989 | 129.58735 | 0.4060 | . 21724 | 0.64662 | 1.65817 | 0.61780 | 1.23219 | 0.48965 |
| 2009 | 7.58682 | 147.76124 | 0.45317 | 0.21135 | 0.66883 | 1.65498 | 0.62863 | 1.25182 | 0.50797 |
| 2010 | 7.55260 | 164.57125 | 0.48662 | 0.18521 | 0.66747 | 1.63610 | 0.62806 | 1.26801 | 0.52124 |
| 2011 | 7.59068 | 183.40342 | 0.54767 | 0.18261 | 0.67091 | 1.61555 | 0.61655 | 1.28311 | 0.53269 |
| 2012 | 7.62719 | 194.40635 | 0.6051 | 0.19100 | 0.68371 | 1.59527 | 0.60393 | 1.29843 | 0.54526 |
| 2013 | 7.69440 | 207.23516 | 0.6631 | 0.20485 | 0.70259 | 1.58317 | 0.59341 | 1.31535 | 0.55588 |
| 201 | 7.77324 | 214.96042 | 0.7369 | 0.22096 | 0.72230 | 1.58182 | 0.58466 | 1.33214 | 0.57029 |
| 20 | 7.87649 | 219.34074 | 0.8279 | 0.23958 | 0.74575 | 1.58371 | 0.57976 | 1.35243 | 0.58617 |
| 2016 | 7.99470 | 222.23249 | 0.9481 | 0.26147 | 0.76124 | 1.59388 | 0.58167 | 1.36701 | 0.60396 |
| 2017 | 8.08877 | 224.06239 | 1.09075 | 0.27375 | 0.77201 | 1.60945 | 0.58792 | 1.37492 | 0.62813 |
| 2018 | 8.19136 | 229.26620 | 1.25013 | 0.28246 | 0.78574 | 1.62488 | 0.59260 | 1.38155 | 0.65237 |
| 2019 | 8.31064 | 242.85332 | 1.38366 | 0.29536 | 0.80353 | 1.64210 | 0.59897 | 1.39510 | 0.67984 |
| 2020 | 8.42130 | 254.21506 | 1.49027 | 0.30299 | 0.82364 | 1.65813 | 0.60560 | 1.41086 | 0.71308 |
| 2021 | 8.45731 | 271.41805 | 1.57238 | 0.28898 | 0.83123 | 1.67974 | 0.60920 | 1.41980 | 0.74500 |
| 2022 | 8.59638 | 292.82907 | 1.68237 | 0.27522 | 0.84523 | 1.71163 | 0.6086 | 1.4258 | 0.78449 |


|  | Q |  |  | $Q_{K 12}$ |  |  | $Q_{K 15}$ | Q | $Q_{K 17}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.00657 | 0.01482 | 0.16902 | 0.12774 | 0.22781 | 0.53839 | 0.22792 | 0.03999 | 0.04002 |
| 1971 | 0.00701 | 0.01439 | 0.16618 | 0.12802 | 0.22 | 0.5 | 0.2 | 0.0 | 0.04202 |
| 1972 | 0.00726 | 0.01404 | 0.16369 | 0.12831 | 0.22593 | 0.55832 | 0.23762 | 0.04287 | 0.05090 |
| 1973 | 0.00741 | 0.01404 | 0.16186 | 0.12858 | 0.22416 | 0.57497 | 0.24226 | 0.04378 | 0.05998 |
| 1974 | 0.00768 | 0.0 | 0.1609 | 0.128 | 0.222 | 0.59037 | 0.24698 | 0.04404 | 0.07376 |
| 197 | 0.00830 | 0.01382 | 0.16054 | 0.12915 | 0.2 | 0.59279 | 0.25066 | 0.04405 | 0.08404 |
| 76 | 0.00880 | 0.01365 | 0.16071 | 0.12915 | 0.21864 | 0.60084 | 0.25575 | 0.04378 | 0.07954 |
| 1977 | 20 | 0.01450 | 0.1 | 0.12915 | 0.2 | 0.6 | 0.25945 | 0.0 | 72 |
| 1978 | 0.00913 | 0.01564 | 0.16061 | 0.12896 | 0.21586 | 0.63 | 0.26491 | 0.04367 | 0.10699 |
| 79 | 0.00957 | 0.01639 | 0.16013 | 0.12940 | 0.21361 | 0.65524 | 0.26865 | 0.04361 | 0.12313 |
| 1980 | 0.00996 | 0.01766 | 0.15898 | 0.12926 | 0.21092 | 0.67062 | 0.2785 | 0.0 | 04 |
| 1981 | 0.01060 | 0.01806 | 0.1575 | 0.12912 | 0.20 | 0.67361 | 0.28416 | 0.0 | 0.12976 |
| 1982 | 0.01062 | 0.01942 | 0.15619 | 0.12898 | 0.20422 | 0.68327 | 0.28857 | 0.04588 | 0.14356 |
| 1983 | 0.01056 | 0.02046 | 0.15518 | 0.12798 | 0.2005 | 0.68504 | 0.2908 | 0.0470 | 59 |
| 1984 | 0.01028 | 0.02157 | 0.15448 | 0.12784 | 0.19792 | 0.70525 | 0.29585 | 0.0 | 0.13387 |
| 1985 | 0.01063 | 0.02314 | 0.15392 | 0.12 | 0.19619 | 0.72955 | 0.30129 | 0.04776 | 0.16156 |
| 1986 | 0.01120 | 0.02425 | 0.15336 | 0.12756 | 0.19473 | 0.75098 | 0.30724 | 0.04803 | 0.17056 |
| 1987 | 0.01205 | 0.02583 | 0.15265 | 0.12742 | 0.1928 | 0.77378 | 0.3113 | 0.0 | 554 |
| 1988 | 0.0 | 0.0 | 0.15192 | 0.1 | 0.19 | 0.78121 | 0.3 | 0.04808 | 0.18527 |
| 1989 | 0.01436 | 0.02766 | 0.15108 | 0.12848 | 0.18779 | 0.80666 | 0.32338 | 0.04849 | 0.19302 |
| 190 | 0.01548 | 0.028 | 0.15029 | 0.12842 | 0.182 | 0.829 | 3260 | 0.0 | 83 |
| 1991 | 0.01 | 0.02 | 0.149 | 0.12836 | 0.17562 | 0.8 | 0.3300 | 0.04927 | 0.20967 |
| 1992 | 0.01810 | 0.03046 | 0.14965 | 0.12835 | 0.16739 | 0.84467 | 0.33526 | 0.04921 | . 20951 |
| 1993 | 0.01922 | 0.03149 | 0.14968 | 0.1 | 0.16 | 0.86 | 0.33870 | 0.04889 | 83 |
| 1994 | 0.01996 | 0.03274 | 0.15008 | 0.12906 | 0.15 | 0.87799 | 0.3419 | 0.04833 | 0.22371 |
| 199 | 0.02050 | 0.03417 | 0.15059 | 0.12942 | 0.15121 | 0.89615 | 0.34868 | 0.04757 | 0.24824 |
| 1996 | 0.02 | 03606 | 0.15100 | 0.12978 | 0.15041 | 0.90 | 0.3518 | 0.04699 | 61 |
| 1997 | 0.02267 | 0.03825 | 0.15111 | 0.13014 | 0.151 | 0.93449 | 0.35717 | 0.04651 | 66 |
| 1998 | 0.02533 | 0.04020 | 0.15122 | 0.13017 | 0.15258 | 0.96423 | 0.36557 | 0.04614 | 0.29803 |
| 1999 | 0.0 | 0.04228 | 0.15089 | 0.13020 | 0.15276 | 0.99416 | 0.37357 | 0.04607 | 39 |
| 2000 | 0.0325 | 0.04410 | 0.1502 | 13 | 0.1523 | 1.0190 | 0.3800 | 0.04627 | 66 |
| 2001 | 0.03639 | 0.04621 | 0.14939 | 0.13027 | 0.15135 | 1.04771 | 0.39476 | 0.04641 | 0.36894 |
| 02 | 0.03920 | 0.04 | 0.14851 | 0.13 | 0.14987 | 1.05961 | 0.40224 | 0.04664 | 09 |
| 2003 | 0.0 | 0. | 0.1479 | 0.13 | 0.1 | 1.07 | 0.4 | 0 | 89 |
| 2004 | 0.04399 | 0.05099 | 0.14746 | 0.1306 | 0.14746 | 1.09532 | 0.41904 | 0.04670 | 0.36829 |
| 2005 | 0.04698 | 0.05279 | 0.14699 | 0.1307 | 0.14 | 1.11509 | 0.42752 | 0.04669 | . 39484 |
| 2006 | 0.05011 | 0.05 | 0.1464 | 0.13 | 0.1448 | 1.1396 | 0.435 | 0.04673 | 31 |
| 2007 | 0.05327 | 0.05751 | 0.14595 | 0.13088 | 0.14350 | 1.15652 | 0.44466 | 0.0467 | 0.44330 |
| 008 | 0.0570 | 0.0586 | 0.14526 | 0.1314 | 0.14080 | 1.16522 | 0.4456 | 0.04685 | 0.45552 |
| 09 | 0.06092 | 0.05857 | 0.14492 | 0.13196 | 0.139 | 1.16973 | 0.44541 | 0.04638 | 0.44514 |
| 2010 | 0.06420 | 0.05791 | 0.14480 | 0.13251 | 0.13768 | 1.15305 | 0.44504 | 0.04615 | 0.39041 |
| 11 | 0.06 | 0.05894 | 0.14478 | 0.13305 | 0.13717 | 1.16927 | 0.4510 | 0.04564 | 0.40789 |
| 12 | 0.07092 | 0.05971 | 0.14475 | 0.13349 | 0.13739 | 1.17963 | 0.45308 | 0.04533 | 0.42216 |
| 2013 | 0.07545 | 0.06048 | 0.14458 | 0.13361 | 0.13939 | 1.19521 | 0.45580 | 0.04527 | 40 |
| 2014 | 0.07 | 0.06108 | 0.14434 | 0.1337 | 0.14062 | 1.20240 | 0.45817 | 0.04528 | 0.47767 |
| 15 | 0.08463 | 0.06210 | 0.14400 | 0.13385 | 0.14112 | 1.22209 | 0.46235 | 0.04529 | 0.50501 |
| 16 | 0.08948 | 0.06350 | 0.14324 | 0.13397 | 0.1412 | 1.24569 | 0.46613 | 0.04559 | 0.54795 |
| 2017 | 0.09606 | 0.06526 | 0.14248 | 0.13397 | 0.14153 | 1.26444 | 0.46738 | 0.04605 | 0.55868 |
| 18 | 0.10414 | 0.06715 | 0.14231 | 0.13384 | 0.14158 | 1.28670 | 0.47155 | 0.04610 | 0.56919 |
| 2019 | 0.11443 | 0.06919 | 0.14201 | 0.13384 | 0.14071 | 1.30863 | 0.47458 | 0.04620 | 0.58664 |
| 2020 | 0.12437 | 0.07093 | 0.14201 | 0.13384 | 0.14140 | 1.32817 | 0.47419 | 0.04621 | 0.60955 |
| 2021 | 0.13513 | 0.07136 | 0.14201 | 0.13384 | 0.14210 | 1.29850 | 0.47656 | 0.04614 | 0.59994 |
| 2022 | 0.14898 | 0.07133 | 0.14201 | 0.13384 | 0.14197 | 1.34401 | 0.48259 | 0.04593 | 0.60395 |

Year $t$ nominal Gross Domestic Product (GDP), $V_{Y}^{t}$, is defined as follows:

$$
\begin{equation*}
V_{Y}^{t} \equiv P_{C}^{t} Q_{C}^{t}+P_{G}^{t} Q_{G}^{t}+P_{I}^{t} Q_{I}^{t}+P_{X}^{t} Q_{X}^{t}-P_{M}^{t} Q_{M}^{t} ; \quad t=1970, \ldots, 2022 \tag{A1}
\end{equation*}
$$

The year $t$ price index for GDP is defined as a Törnqvist price index of the 5 component price and quantity series for consumption, government output, investment, exports and (minus) imports, which we denote by $P_{Y}^{t}$. The companion quantity index for GDP is denoted by $Q_{Y}^{t}$. $V_{Y}^{t}, P_{Y}^{t}$ and $Q_{Y}^{t}$ are listed in Table A6 below.
Now that we have estimates for the US capital stock, we can calculate real and nominal capital output ratios for the US economy, $Q_{K Y}^{t}$ and $V_{K Y}^{t}$, for each year $t$. The real capital output ratio, $Q_{K Y}^{t} \equiv Q_{K}^{t} / Q_{Y}^{t}$, is an indicator of partial efficiency while the nominal capital output ratio, $V_{K Y}^{t} \equiv V_{K}^{t} / V_{Y}^{t}$, is a partial indicator of the international competitiveness of the US economy: it indicates how many dollars are required to purchase a sufficient amount of capital in order to produce one dollar of output in year $t$ (on average). The lower $V_{K Y}^{t}$ is, the easier it is to start a business in the US in year $t . V_{K Y}^{t}$ and $Q_{K Y}^{t}$ are listed in Table A6 below.
In order to indicate the importance of land as a component of the total capital stock, we construct a year $t$ Törnqvist capital stock aggregate price $P_{K R}^{t}$ that excludes the 6 land stocks. Denote the companion year $t$ reproducible capital stock quantity by $Q_{K R}^{t}$ and define the corresponding year $t$ value by $V_{K R}^{t} \equiv P_{K R}^{t} Q_{K R}^{t}$. Define the year $t$ reproducible real and nominal capital output ratios as $Q_{K R Y}^{t} \equiv Q_{K R}^{t} / Q_{Y}^{t}$ and $V_{K R Y}^{t} \equiv V_{K R}^{t} / V_{Y}^{t}$. These variables are also listed in Table A6 below.

Table A6: US GDP Aggregates and Nominal and Real Capital Output Ratios

| Year | KY | $Q_{K Y}$ |  |  | $P_{Y}$ | ${ }_{K}$ | $Q_{Y}$ | $Q_{K}$ | $V_{Y}$ | K | $V_{K R}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 3.1120 | 3.1120 | 1.8167 | 1.8167 | 1.000 | . 00 | 1.0 | 3.1975 | . 027 | , | 1.8666 |
| 1971 | 3.0 | 3.0 | 1.7 | 1.8132 | 1.0 | 1.0351 | 1.0 | 3.2 | 1.1138 | 3.3833 |  |
| 1972 | 3.0217 | 2.9958 | 1.7645 | 1.7854 | 1.0940 | 1.1035 | 1.1205 | 3.3567 | 1.2258 | 3.7039 | 2.1629 |
| 1973 | 3.0010 | 2.9 | 1.7532 | 7630 | 1.1538 | 43 | 1.1832 | 3.4591 | 1 | 4.0967 | 32 |
| 74 | 3.1643 | 3.0423 | 1.8696 | 1.8609 | 1.2581 | 1.3086 | 1.1751 | 3.5749 | 1.4784 | 80 | 640 |
| 1975 | 3.3652 | 3.1219 | 1.9853 | 1.9344 | 1.3788 | 1.4862 | 1.1704 | 3.6537 | 1.6137 | 5.4303 | . 2036 |
| 1976 | 3. | 3.004 | 1.9489 | 1.8685 | 1.4585 | 6392 | 1.23 | 3.7 | 1.7971 | 6.0695 | 3.5024 |
| 1977 | 3.3282 | 2.930 | 1.9 | 1.8354 | 1.5 | 1.7587 | 1. | 3. | 000 |  | 40 |
| 1978 | 3.3110 | 2.8523 | 1.8795 | 1.8065 | 1.6593 | 1.9262 | 1.3631 | 3.8879 | 2.2619 | 7.4891 | . 2512 |
| 79 | 3.419 | 2.8541 | 1.9216 | 1.8312 | 1.7985 | 2.1548 | 1.4066 | 4.0144 | 2.5297 | 8.6501 | 2 |
| 1980 | 3.5 | 2.950 | 2.0241 | . 9150 | 1.9561 | . 3830 | 1.4039 | 4.1417 | . 7462 | 966 | 586 |
| 81 | 3.5318 | 2.9276 | 2.0340 | . 9195 | 2.1319 | . 5719 | 1.4408 | 4.2181 | 3.0717 | 10.8486 | . 2479 |
| 82 | 3.6 | 3.0587 | 2.1731 | . 0260 | 2.2822 | 2.7436 | 1.4090 | 4.3096 | 3.2155 | 11.8241 | 77 |
| 83 | 3.531 | 2.9621 | 2.0994 | . 9737 | 2.3809 | 2.8388 | 1.468 | 4.3506 | 3.4970 | 12.3503 | 15 |
| 84 | 3.2512 | 2.8102 | 1.9417 | 1.8769 | 2.4640 | . 8506 | 1.575 | 4.4287 | 38830 | 12.6244 | 7.5396 |
| 1985 | 3.1 | 2.7911 | 1.9236 | . 8819 | 2.5491 | 2.8951 | 1.638 | 4.5718 | 4.17 | 13.235 | 19 |
| 86 | 3.219 | 2.7920 | 1.9731 | . 8979 | 2.6177 | . 0187 | 1.6862 | 4.7078 | 4139 | 14.211 | 0 |
| 1987 | 3.2 | 2.766 | 1.9 | . 89 | 2.6824 | . 18 | 1.74 | 4.8285 | . 6821 | 15.3896 | 22 |
| 1988 | 3.2 | 2.7 | 1.9 | 1.8 | 2.77 | 3.3687 | 1.8 | 4.9389 | 5.0471 | 16.6378 | 18 |
| 1989 | 3.2878 | 2.6881 | 1.9309 | 1.8649 | 2.8900 | . 5349 | 1.8824 | 5.0600 | 54402 | 17.8865 | 0.5042 |
| 1990 | 3. | 2.6906 | 1.9 | 97 | 2.9944 | 3.6597 | 1.9194 | 44 | 5.7475 | 3 | 315 |
| 199 | 3.2 | 2.7 | 1.9 | 1.9 | 3.0919 | 3.6815 | 1.9167 | 5.2516 | 9262 | 19.3339 | 25 |
| 1992 | 3.0921 | 2.6780 | 1.9095 | 1.8919 | 3.1689 | . 6589 | 1.9805 | 5.3037 | 6.2759 | 19.4057 | 11.9839 |
| 1993 | 2. | 2.6375 | 1.8747 | 1.8690 | 3.2448 | 3.6333 | 2.0358 | 5.3693 | . 6056 | 0 | 33 |
| 19 | 2. | 2.5 | 1.83 | 1.83 | 3.3066 | 3.6441 | 2.1177 | 5.4460 | 0024 | 19.8458 | 12.8801 |
| 1995 | 2.8187 | 2.5576 | 1.8556 | 1.8315 | 3.3815 | . 7266 | 2.1735 | 5.5590 | 7.3497 | 20.7165 | 13.6384 |
| 1996 | 2. | 2.5 | 1. | 1.8071 | 3.4 | 3.8316 | 2.2 | 5.6649 | .7718 | 5 | 3 |
| 1997 | 2.7 | 2.459 | 1.8298 | 1.78 | 3.4986 | . 9203 | 2.3602 | 5.80 | . 25 | 22.7549 | 15.1093 |
| 98 | 2.8072 | 2.4260 | 1.8194 | 1.7666 | 3.5435 | . 1002 | 2.4628 | 5.9748 | . 727 | 24.4983 | 5.8779 |
| 1999 | 2. | 2.3 | 1.8 | 1.7517 | 3.6 | 4.3570 | 2.5 | , | . 2835 | 1 | 6 |
| 0 | 2.9467 | 2.3676 | 1.83 | 1.7513 | 3.6894 | . 5917 | 2.67 | 6.3423 | 9.8830 | 29.1218 | 18.0958 |
| 2001 | 3.1136 | 2.4255 | 1.8936 | 1.8040 | 3.7842 | 4.8577 | 2.701 | 6.5525 | 10.2228 | 31.8299 | 19.3582 |
| 2002 | 3.2 | 2.4 | 1.9 | 1.8223 | 3.8 | 5.1069 | 2.745 | 6.6862 | 10.5445 | 1 | 64 |
| 03 | 3.3 | 2.4162 | 1.9546 | 1.8138 | 3.9155 | 706 | 2.8223 | 6.8193 | 11.0505 | 36.6233 | 21.5995 |
| 2004 | 3.3793 | 2.3712 | 1.9401 | 1.7891 | 4.0174 | 5.7253 | 2.9307 | 6.9494 | 11.7738 | 39.7872 | 22.8429 |
| 2005 | 3. | 2.3448 | 1.9 | 1.7798 | 4.1478 | 6.3382 | 3.0296 | 7.1039 | 12.566 | 5.0259 | 4 |
| 2006 | 3.8226 | 2.3298 | 2.0900 | 1.7790 | 4.2705 |  | 3.1 | 7.2576 | 13.3 | 50.8525 | 27.8036 |
| 2007 | 3.89 | 2.338 | 2.1529 | 1.7977 | 4.3949 | . 3143 | 3.1753 | 7.4249 | 13.9551 | 54.3082 | 30.0444 |
| 08 | 3.8704 | 2.367 | 2.2322 | 8363 | 4.4793 | . 3229 | 3.180 | 7.529 | 退. 24 | 5.1409 | 8022 |
| 2009 | 3. | 2.4 | 2.3120 | 100 | 4.5224 | 6.9786 | 3.096 | 7.586 | .00 | 52.9456 | 32.3805 |
| 2010 | 3.3 | 2.3746 | 2.2034 | 1.8523 | 4.5712 | . 4595 | 3.1806 | 7.5526 | 14.5389 | 48.7863 | 32.0357 |
| 2011 | 3.154 | 2.351 | 2.1634 | 1.8306 | 4.6638 | . 2559 | 3.2279 | 7.5907 | 15.0542 | 47.4864 | 2.5689 |
| 2012 | 3.1142 | 2.311 | 2.1415 | 995 | 4.7516 | . 4021 | 3.300 | 7.6272 | 15.6800 | 48.8301 | 33.5791 |
| 2013 | 3.1600 | 2.2820 | 2.1286 | 1.7780 | 4.8257 | 6.6822 | 3.3717 | 7.6944 | 16.2709 | 51.4157 | 34.6347 |
| 2014 | 3.239 | 2.249 | 2.1260 | 1.7578 | 4.9093 | 7.0699 | 3.4555 | 7.7732 | 16.9642 | 54.9561 | 6.0651 |
| 2015 | 3.3230 | 2.2168 | 2.1346 | . 7361 | 4.9627 | 7.4390 | 3.5531 | 7.8765 | 17.6328 | 58.5933 | 37.6390 |
| 2016 | 3.405 | 2.2130 | 2.1481 | 1.7380 | 5.0205 | .7255 | 3.6126 | 7.9947 | 18.1370 | 61.7633 | 38.9607 |
| 2017 | 3.4418 | 2.1857 | 2.1408 | 1.7217 | 5.1113 | 8.0486 | 3.7007 | 8.0888 | 18.9154 | 65.1030 | 40.4937 |
| 2018 | 3.4327 | 2.1503 | 2.0995 | 1.6966 | 5.2221 | 8.3364 | 3.8094 | 8.1914 | 19.8929 | 68.2866 | 41.7653 |
| 2019 | 3.4471 | 2.1291 | 2.1031 | 1.686 | 5.3104 | 8.5975 | 3.903 | 8.3106 | 20.7281 | 71.4509 | 43.5929 |
| 2020 | 3.6529 | 2.2092 | 2.2017 | 1.7587 | 5.4071 | 8.9406 | 3.8119 | 8.4213 | 20.6116 | 75.2918 | 45.3816 |
| 2021 | 3.5798 | 2.1003 | 2.1163 | 1.6840 | 5.6460 | 9.6234 | 4.0268 | 8.4573 | 22.7353 | 81.3879 | 48.1146 |
| 2022 | 3.6618 | 2.0998 | 2.1038 | 1.6812 | 6.0547 | 10.5586 | 4.0939 | 8.5964 | 24.7871 | 90.7660 | 52.1475 |
| Mean | 3.2903 | 2.5802 | 1.9913 | 1.8223 | 3.3758 | 4.6218 | 2.3854 | 5.8789 | 9.3025 | 31.238 | 19.121 |

The real capital output ratio, $Q_{K Y}^{t}$, fell from 3.12 in 1975 to 2.10 in 2022. This is probably a reflection of the growth in services relative to goods: services typically require less capital to produce a dollar of output. However, the nominal capital output ratio grew from 3.11 in 1970 to 3.95 in 1980 and then fell to 2.75 in 1997. It rose to 3.89 in 2007 (due to rising land prices) when the great recession hit the US economy. The capital output ratio fell to 3.11 in 2012 and then rose as a new property price bubble occurred, to finish at 3.66 . However, overall, the US economy has not varied too much from the average nominal capital output ratio of 3.29 so that the economy has remained competitive over the years. The average share of reproducible capital in the total capital stock was $61.2 \%$ so the average share of land was $38.8 \%$. Thus land is a huge missing input in most macroeconomic production models of the US economy.
We turn our attention to the construction of user costs for the US economy. User costs require depreciation rates for assets. Assets 11-17 are non-depreciable assets so we set the period $t$ depreciation rate for these assets, $\delta_{n}^{t}$, equal to 0 for $n=11, \ldots, 17$ and $t=1970, \ldots, 2022$. Since the APDB uses the geometric model for depreciation for reproducible assets 1-10, the APDB depreciation rates satisfy the following equations:

$$
\begin{equation*}
Q_{K n}^{t+1}=\left(1-\delta_{n}^{t}\right) Q_{K n}^{t}+Q_{I n}^{t} ; \quad n=1, \ldots, 10 ; t=1970, \ldots, 2022 . \tag{A2}
\end{equation*}
$$

We can solve equations (A1) for the depreciation rates $\delta_{n}^{t}$ for the years 1970-2022 for assets 1-10. The resulting nonzero depreciation rates are listed below in Table A7.

Table A7: Geometric Depreciation Rates for Reproducible US Capital Stocks

| Year | $\delta_{1}^{t}$ | $\delta_{2}^{t}$ | $\delta_{3}^{t}$ | $\delta_{4}^{t}$ | $\delta_{5}^{t}$ | $\delta_{6}^{t}$ | $\delta_{7}^{t}$ | $\delta_{8}^{t}$ | $\delta_{9}^{t}$ | $\delta_{10}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.32975 | 0.29376 | 0.24427 | 0.19786 | 0.04956 | 0.08459 | 0.02627 | 0.22606 | 0.40849 | 0.30782 |
| 1971 | 0.35148 | 0.28745 | 0.24605 | 0.19606 | 0.04952 | 0.08456 | 0.02624 | 0.22084 | 0.40202 | 0.30844 |
| 1972 | 0.37737 | 0.28389 | 0.25102 | 0.19823 | 0.04997 | 0.08500 | 0.02668 | 0.21903 | 0.39933 | 0.31242 |
| 19 | 0.37153 | 0.2897 | 0.25535 | 0.2 | 0.04949 | 0.08453 | 0.02621 | 0.21765 | 0.40258 | 7 |
|  | 0. | 0 | 0.24979 | 0.20099 | 0.04958 | 0.08461 | 0.02630 | 9 | 4 | 58 |
| 19 | 0.36808 | 0.28417 | 0.24463 | 0.1975 | 0.04946 | 0.08 | 0.02618 | 0.21392 | 0.40716 | 0.31043 |
| 1976 | 0.38981 | 0.28650 | 0.24788 | 0.19744 | 0.04936 | 0.08440 | 0.02608 | 0.21433 | 0.40421 | 0.32201 |
| 1977 | 0.40944 | 0.29739 | 0.25320 | 0.19873 | 0.04939 | 0.08442 | 0.02610 | 0.21443 | 0.39372 | 0.32460 |
| 1978 | 0.45837 | 0.29955 | 0.25416 | 0.20027 | 0.04936 | 0.08439 | 0.02607 | 0.21498 | 0.40475 | 0.31981 |
| 19 | 0.45576 | 0.3023 | 0.25381 | 0.20 | 0.04972 | 0.08 | 0.02643 | 0.21536 | 0.40329 | 8 |
| 19 | 0.453 | 0.3011 | 0.24557 | 0.19862 | 0.04959 | 0.08462 | 0.02631 | 0.21540 | 0.40774 | 0 |
| 19 | 0 | 0. | 0.24453 | 0.19819 | 3 | 0.08446 | 5 | 0 | 4 | 4 |
|  | 0. | 0. | 0.24183 | 0.19596 | 0.04951 | 0.08454 | 0.02622 | 6 | 9 | 0 |
| 19 | 0.4 | 0.28 | 0.24583 | 0.19569 | 0.04974 | 0.08 | 0.02645 | 0.21548 | 0.38987 | 0.32077 |
| 1984 | 0.44165 | 0.29139 | 0.24962 | 0.19793 | 0.04954 | 0.0845 | 0.02625 | 0.21648 | 0.40196 | 0.32368 |
| 1985 | 0.41571 | 0.29183 | 0.24940 | 0.19848 | 0.04972 | 0.08475 | 0.02643 | 0.21733 | 0.40580 | 0.31980 |
| 1986 | 0.39545 | 0.29111 | 0.24612 | 0.19789 | 0.04936 | 0.08439 | 0.02608 | 0.21659 | 0.41022 | 0.32248 |
| 19 | 0.3 | 0.28 | 0.24543 | 0.19739 | 0.04938 | 0.0 | 0.02609 | 0.21605 | 0.41537 | 45 |
| 19 | 0.3 | 0.290 | 0.24798 | 0.19742 | 0.04936 | 0.08 | 0.02607 | 0.21540 | 0.41128 | 2 |
| 19 | 0.3833 | 0.28 | 0.24510 | 0.19965 | 0.05020 | 0.08 | 0.02691 | 0.21516 | 0.41067 | 8 |
| 19 | 0.36724 | 0.2873 | 0. | 0.19761 | 8 | 0.08441 | 0.02609 | 0.21504 | 2 | 1 |
| 19 | 0.3651 | 0.28424 | . 24598 | 0.1 | 0.04954 | 0.08 | 0.02625 | 0.21473 | 0.41007 | 0.31667 |
| 19 | 0.38152 | 0.28 | 0.24879 | 0.197 | 0.05108 | 0.08611 | 0.02780 | 0.21387 | 0.40743 | 0.31760 |
| 1993 | 0.38386 | 0.28642 | 0.25210 | 0.19750 | 0.05015 | 0.0851 | 0.02687 | 0.21278 | 0.40285 | 0.31853 |
| 19 | 0.38456 | 0.29209 | 0.25631 | 0.19891 | 0.05085 | 0.0858 | 0.02757 | 0.21218 | 0.40057 | 0.31915 |
| 19 | 0.40447 | 0.2944 | 0.25424 | 0.19896 | 0.05006 | 0.0850 | 0.02678 | 0.21241 | 0.40135 | 0.32098 |
| 19 | 0.41261 | 0.29574 | 0.25313 | 0.19869 | 0.04962 | 0.0846 | 0.02633 | 0.21335 | 0.40962 | 0.32180 |
| 19 | 0.41772 | 0.29690 | 0.25352 | 0.19884 | 0.04971 | 0.0847 | 0.02643 | 0.21380 | 0.41837 | 0.32027 |
| 19 | 0.42032 | 0.2987 | 0.25470 | 0.19960 | 0.04976 | 0.0847 | 0.02647 | 0.21428 | 0.41986 | 0.32041 |
| 1999 | 0.41970 | 0.3029 | 0.25840 | 0.19908 | 0.04976 | 0.08479 | 0.02648 | 0.21479 | 0.42356 | 0.31905 |
| 2000 | 0.40019 | 0.3092 | 0.25236 | 0.19941 | 0.04942 | 0.08445 | 0.02613 | 0.21528 | 0.41840 | 0.31981 |
| 2001 | 0.37753 | 0.2958 | 0.24745 | 0.19829 | 0.04962 | 0. | 0.02634 | 0.21473 | 0.41047 | 0.31609 |
| 20 | 0.36918 | 0.28337 | 0.24496 | 0.19758 | 0.04956 | 0.08459 | 0.02628 | 0.21357 | 0.40653 | 0.31725 |
| 2003 | 0.36860 | 0.28847 | 0.24342 | 0.19832 | 0.04977 | 0.08481 | 0.02649 | 0.21327 | 0.40735 | 0.31924 |
| 20 | 0.36962 | 0.29293 | 0.24989 | 0.19966 | 0.05082 | 0.08586 | 0.02754 | 0.21297 | 0.40867 | 0.31786 |
| 2005 | 0.37194 | 0.29497 | 0.25528 | 0.20334 | 0.05329 | 0.08832 | 0.03000 | 0.21344 | 0.40838 | 0.32008 |
| 2006 | 0.37632 | 0.29594 | 0.25253 | 0.20002 | 0.04945 | 0.0844 | 0.02616 | 0.21396 | 0.40767 | 0.31822 |
| 2007 | 0.37450 | 0.2995 | 0.24987 | 0.1997 | 0.04946 | 0.08450 | 0.02618 | 0.21420 | 0.40940 | 0.31551 |
| 2008 | 0.37083 | 0.2979 | 0.24345 | 0.19989 | 0.05041 | 0.0854 | 0.02713 | 0.21383 | 0.40849 | 0.31204 |
| 20 | 0.36520 | 0.2 | . 23059 | 0.19536 | 0.04952 | 0.0845 | 0.02624 | 0.21264 | 0.40584 | 0.31056 |
| 20 | 0.36565 | 0.2985 | 0.24438 | 0.19637 | 0.04980 | 0.0848 | 0.02652 | 0.21220 | 0.40408 | 0.31512 |
|  | 0.35664 | 0.29609 | 0.25214 | 0.19813 | 0.05016 | 0.08519 | 0.02688 | 0.21237 | 0.40654 | 0.31436 |
| 2012 | 0.35811 | 0.29525 | 0.25582 | 0.19937 | 0.05054 | 0.08557 | 0.02725 | 0.21194 | 0.40785 | 0.31433 |
| 2013 | 0.35198 | 0.29629 | 0.25546 | 0.19833 | 0.04954 | 0.08458 | 0.02626 | 0.21262 | 0.40670 | 0.31388 |
| 2014 | 0.34893 | 0.29787 | 0.25603 | 0.19864 | 0.04944 | 0.08447 | 0.02616 | 0.21282 | 0.40708 | 0.31492 |
| 2015 | 0.34783 | 0.30106 | 0.25705 | 0.19764 | 0.04956 | 0.08460 | 0.02628 | 0.21308 | 0.40654 | 0.31585 |
| 2016 | 0.34730 | 0.30211 | 0.25191 | 0.19729 | 0.04984 | 0.08487 | 0.02656 | 0.21410 | 0.40974 | 0.31665 |
| 2017 | 0.35223 | 0.3037 | 0.25229 | 0.19984 | 0.05184 | 0.08687 | 0.02855 | 0.21395 | 0.41185 | 0.31686 |
| 2018 | 0.35618 | 0.29604 | 0.25179 | 0.19817 | 0.04988 | 0.08491 | 0.02660 | 0.21432 | 0.41475 | 0.31706 |
| 2019 | 0.35370 | 0.29156 | 0.24906 | 0.19811 | 0.04961 | 0.08464 | 0.02633 | 0.21503 | 0.41238 | 0.31626 |
| 2020 | 0.35778 | 0.28892 | 0.24064 | 0.19699 | 0.05000 | 0.08503 | 0.02672 | 0.21460 | 0.41232 | 0.31326 |
| 2021 | 0.36031 | 0.29158 | 0.24103 | 0.19828 | 0.05050 | 0.08553 | 0.02722 | 0.21546 | 0.41547 | 0.31228 |
| 2022 | 0.36031 | 0.29158 | 0.24103 | 0.19828 | 0.05050 | 0.08553 | 0.02722 | 0.21546 | 0.41547 | 0.31228 |
| Mean | 0.38427 | 0.29356 | 0.24909 | 0.19840 | 0.04988 | 0.08491 | 0.02660 | 0.21479 | 0.40769 | 0.31707 |

We now explain the user cost of capital methodology developed by Jorgenson and his coworkers. ${ }^{* 34}$ Suppose the beginning of the year $t$ price of a new unit of capital stock $n$ is $P_{K n}^{t}$ and a producer faces an annual cost of capital at the beginning of year $t$ of $r^{t}$. Thus the cost to the producer is the total of the purchase of one unit of capital stock $n$ at the beginning of year $t$ plus the interest cost of financing the purchase, which is equal to $\left(1+r^{t}\right) P_{K n}^{t}$. Suppose further that asset $n$ in year $t$ has a geometric depreciation rate equal to $\delta_{n}^{t}$. Various levels of government may charge an annual tax on the use of each unit of asset $n$ at the tax rate $\tau_{n}^{t}$ at the beginning of year $t$ so that the total tax charged on asset $n$ during year $t$ is $\tau_{n}^{t} P_{K n}^{t} Q_{K n}^{t}$. We think of firms buying their capital stocks at the beginning of year $t$, using the services of these stocks during period $t$ and then selling the used assets at the end of the period at prevailing prices at the end of the period. The year $t$ user cost $U_{n}^{t}$ of asset $n$ is the resulting per unit net cost. ${ }^{* 35}$ Using the geometric model of depreciation, we can use the price of a new unit of capital stock $n$ at the beginning of year $t$, the $P_{K n}^{t}$, listed in Table A4 as the appropriate price of the year $t$ beginning of the year capital stock quantity $Q_{K n}^{t}$ for asset $n$. Define the year $t$ ex post inflation rate for asset $n, i_{n}^{t}$, as follows where $P_{K n}^{t}$ is the beginning of year $t$ price for a unit of asset $n$ :

$$
\begin{equation*}
\left(1+i_{n}^{t}\right) \equiv P_{K n}^{t+1} / P_{K n}^{t} ; \quad n=1, \ldots, 17 ; t=1970, \ldots, 2022 \tag{A3}
\end{equation*}
$$

Thus the year $t$ user cost for asset $n, U_{n}^{t}$, times the initial year $t$ capital stock, $Q_{K n}^{t}$, is equal to the following expression:

$$
\begin{align*}
U_{n}^{t} Q_{K n}^{t} \equiv & \text { Beginning of period cost }- \text { end of period benefit } \\
& \quad n=1, \ldots, 17 ; t=1970, \ldots, 2022 \\
& =\left\{\left(1+r^{t}\right) P_{K n}^{t} Q_{K n}^{t}+\tau_{n}^{t} P_{K n}^{t} Q_{K n}^{t}\right\}-P_{K n}^{t+1}\left(1-\delta_{n}^{t}\right) Q_{K n}^{t} \\
& =\left[1+r^{t}+\tau_{n}^{t}\right] P_{K n}^{t} Q_{K n}^{t}-P_{K n}^{t}\left(1+i_{n}^{t}\right)\left(1-\delta_{n}^{t}\right) Q_{K n}^{t} \quad \text { using (A3) } \\
& =\left[r^{t}+\tau_{n}^{t}-i_{n}^{t}+\delta_{n}^{t}\left(1+i_{n}^{t}\right)\right] P_{K n}^{t} Q_{K n}^{t} . \tag{A4}
\end{align*}
$$

Thus the year $t$ Jorgensonian user cost for asset $n$ is $U_{n}^{t}=\left[r^{t}+\tau_{n}^{t}-i_{n}^{t}+\delta_{n}^{t}\left(1+i_{n}^{t}\right)\right] P_{K n}^{t} .{ }^{* 36}$ Note that the end of period benefit part of the user cost formula, $P_{K n}^{t}\left(1+i_{n}^{t}\right)\left(1-\delta_{n}^{t}\right) Q_{K n}^{t}$, decreases as the depreciation rate $\delta_{n}^{t}$ increases and increases as the asset inflation rate $i_{n}^{t}$ increases. For land assets, where the depreciation rate is zero, the annual asset inflation rate could be so great that the overall user cost becomes negative instead of being positive. In this case, the user cost becomes a user benefit; i.e., instead of the purchase of the asset, using it and then selling it at the end of the period being a cost, it becomes an addition to revenues.
In order to calculate user costs, we require information on the cost of capita $r^{t}$, asset tax rates $\tau_{n}^{t}$, asset inflation rates $i_{n}^{t}$, depreciation rates $\delta_{n}^{t}$ and beginning of the year asset prices $P_{K n}^{t}$. Alternative methods for estimating the cost of capital will be discussed below, asset tax rates will be listed in Table A8 below, ex post asset inflation rates as well as asset prices can be calculated using the information in Table A4 on beginning of the year asset prices and depreciation rates are listed in Table A7 above.

[^17]Table A8: Asset Tax Rates (Percentages)

| Year | $\tau_{1}^{t}$ | $\tau_{2}^{t}$ | $\tau_{3}^{t}$ | $\tau_{4}^{t}$ | $\tau_{5}^{t}$ | $\tau_{6}^{t}$ | $\tau_{9}^{t}$ | $\tau_{10}^{t}$ | $\tau_{11}^{t}$ | $\tau_{13}^{t}$ | $\tau_{14}$ | $\tau_{15}^{t}$ | $\tau_{16}^{t}$ | $\tau_{17}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.653 | 0.653 | 0.493 | 0.762 | 2.105 | 2.151 | 0.247 | 0.247 | 1.085 | 1.034 | 1.299 | 1.875 | 0.278 | 0.248 |
| 1971 | 0.529 | 0.661 | 0.499 | 0.774 | 2.165 | 2.257 | 0.255 | 0.259 | 1.136 | 1.064 | 1.336 | 2.058 | 0.302 | 0.268 |
| 1972 | 0.579 | 0.653 | 0.487 | 0.752 | 2.118 | 2.180 | 0.260 | 0.246 | 1.129 | 1.050 | 1.318 | 1.899 | 0.287 | 0.260 |
| 1973 | 0.580 | 0.620 | 0.465 | 0.727 | 2.065 | 2.095 | 0.245 | 0.238 | 1.143 | 1.001 | 1.257 | 1.837 | 0.283 | . 255 |
| 1974 | 0.521 | 0.588 | 0.44 | 0.692 | 1.906 | 1.951 | 0.219 | 0.218 | 0.993 | 0.926 | 1.163 | 1.768 | 0.260 | . 225 |
| 1975 | 0.502 | 0.548 | 0.418 | 0.662 | 1.768 | 1.808 | 0.215 | 0.207 | 0.925 | 0.889 | 1.116 | 1.700 | 0.247 | 190 |
| 1976 | 0.478 | 0.538 | 0.412 | 0.636 | 1.739 | 1.760 | 0.206 | 0.205 | 0.955 | 0.862 | 1.082 | 1.569 | 0.238 | 0.213 |
| 1977 | 0.469 | 0.503 | 0.400 | 0.616 | 1.724 | 1.742 | 0.210 | 0.198 | 0.898 | 0.833 | 1.046 | 1.587 | 0.234 | 0.208 |
| 1978 | 0.359 | 0.454 | 0.348 | 0.539 | 1.519 | 1.537 | 0.168 | 0.175 | 0.805 | 0.738 | 0.927 | 1.404 | 0.209 | 0.180 |
| 1979 | 0.348 | 0.38 | 0.30 | 0.467 | 1.304 | 1.330 | 0.162 | 0.149 | 0.692 | 0.641 | 0.805 | 1.228 | 0.181 | 0.137 |
| 1980 | 0.317 | 0.37 | 0.28 | 0.44 | 1.235 | 1.264 | 0.144 | 0.141 | 0.63 | 0.605 | 0.760 | 1.063 | 0.161 | 0.124 |
| 19 | 0.3 | 0.3 | 0.29 | 0.46 | 1.267 | 1.299 | 0. | 0.148 | 0. | 0.622 | 0.781 | 1.090 | 0.163 | 31 |
| 19 | 0.378 | 0.40 | . 309 | 7 | 1.301 | 1 | 0.163 | 0.153 | 0.636 | 0.671 | 0.843 | 1.090 | 0.163 | 0.130 |
| 1983 | 0.373 | 0.4 | 0.31 | . 489 | 1.339 | 1.377 | 0.174 | 0.159 | 0.684 | 0.654 | 0.822 | 1.213 | 0.179 | 0.153 |
| 1984 | 0.386 | 0.439 | 0.331 | 0.508 | 1.414 | 1.448 | 0.168 | 0.167 | 0.662 | 0.667 | 0.837 | 1.298 | 0.187 | 0.169 |
| 1985 | 0.403 | 0.444 | 0.336 | 0.515 | 1.430 | 1.472 | 0.172 | 0.170 | 0.672 | 0.699 | 0.878 | 1.343 | 0.194 | 0.148 |
| 1986 | 0.404 | 0.438 | 0.335 | 0.521 | 1.446 | 1.466 | 0.163 | 0.169 | 0.688 | 0.681 | 0.856 | 1.364 | 0.199 | 0.161 |
| 1987 | 0.39 | 0.427 | 0.324 | 0.505 | 1.413 | 1.434 | 0.161 | 0.165 | 0.730 | 0.717 | 0.881 | 1.339 | 0.198 | 0.158 |
| 1988 | 0.40 | 0.42 | 0.323 | 0.504 | 1.398 | 1.431 | 0.169 | 0.164 | 0.72 | 0.704 | 0.865 | 1.299 | 0.193 | 0.152 |
| 1989 | 0.420 | 0.43 | 0.33 | 0.515 | 1.426 | 1.457 | 0.175 | 0.169 | 0.73 | 0.712 | 0.872 | 1.305 | 0.195 | 0.150 |
| 19 | 0. | 0. | 0.3 | 0.53 | 1.466 | 1.506 | 0.171 | 0. | 0.76 | 0.7 | 0.902 | 1.295 | 0.193 | 1 |
| 19 | 0. | 0.49 | 0.377 | 0.58 | 1.586 | 1.638 | 0.188 | 0. | 0.82 | 0.793 | 0.947 | 1.361 | 0.204 | 2 |
| 19 | 0.474 | 0.51 | 0.39 | 0.609 | 1. | 1.705 | 0.197 | 0.197 | 0.873 | 0.832 | 0.953 | 1.468 | 0.220 | 0.193 |
| 1993 | 0.478 | 0.51 | 0.396 | 0.610 | 1.714 | 1.740 | 0.206 | 0.198 | 0.896 | 0.841 | 0.950 | 1.467 | 0.220 | 0.190 |
| 19 | 0.504 | 0.533 | 0.412 | 0.632 | 1.768 | 1.799 | 0.207 | 0.207 | 0.912 | 0.871 | 1.038 | 1.553 | 0.231 | 0.204 |
| 1995 | 0.478 | 0.517 | 0.402 | 0.619 | 1.731 | 1.762 | 0.202 | 0.202 | 0.899 | 0.858 | 1.076 | 1.489 | 0.224 | 0.179 |
| 1996 | 0.460 | 0.527 | 0.405 | 0.626 | 1.738 | 1.772 | 0.199 | 0.205 | 0.908 | 0.874 | 1.070 | 1.553 | 0.231 | 0.181 |
| 1997 | 0.46 | 0.528 | 0.400 | 0.620 | 1.738 | 1.775 | 0.204 | 0.203 | 0.905 | 0.872 | 1.096 | 1.542 | 0.230 | 0.197 |
| 1998 | 0.42 | 0.48 | 0.37 | 0.577 | 1.617 | 1.660 | 0.189 | 0.187 | 0.839 | 0.817 | 1.126 | 1.491 | 0.222 | 0.181 |
| 1999 | 0.41 | 0.45 | 0.35 | 0.555 | 1.560 | 1.597 | 0.184 | 0.183 | 0.814 | 0.787 | 0.987 | 1.439 | 0.214 | 0.171 |
| 2000 | 0.423 | 0.445 | 0.344 | 0.530 | 1.496 | 1.525 | 0.177 | 0.175 | 0.78 | 0.752 | 0.910 | 1.401 | 0.208 | 0.158 |
| 2001 | 0.390 | 0.419 | 0.327 | 0.506 | 1.423 | 1.445 | 0.163 | 0.165 | 0.738 | 0.713 | 0.913 | 1.303 | 0.195 | 0.145 |
| 2002 | 0.405 | 0.423 | 0.329 | 0.511 | 1.422 | 1.455 | 0.160 | 0.16 | 0.727 | 0.717 | 0.822 | 1.344 | 0.199 | 0.159 |
| 2003 | 0.397 | 0.3 | 0.31 | 0.494 | 1.389 | 1.401 | 0.155 | 0.160 | 0.70 | 0.688 | 0.880 | 1.288 | 0.192 | 0.160 |
| 2004 | 0.386 | 0.381 | 0.305 | 0.472 | 1.339 | 1.356 | 0.151 | 0.152 | 0.721 | 0.650 | 0.802 | 1.247 | 0.186 | 0.157 |
| 2005 | 0.348 | 0.362 | 0.283 | 0.439 | 1.238 | 1.272 | 0.141 | 0.141 | 0.675 | 0.595 | 0.818 | 1.169 | 0.175 | 0.132 |
| 2006 | 0.338 | 0.350 | 0.274 | 0.426 | 1.200 | 1.236 | 0.137 | 0.138 | 0.638 | 0.574 | 0.774 | 1.068 | 0.160 | 0.141 |
| 2007 | 0.331 | 0.338 | 0.271 | 0.411 | 1.250 | 1.223 | 0.139 | 0.140 | 0.637 | 0.593 | 0.773 | 1.084 | 0.163 | 0.130 |
| 2008 | 0.316 | 0.312 | 0.253 | 0.384 | 1.367 | 1.192 | 0.124 | 0.123 | 0.664 | 0.618 | 0.781 | 1.137 | 0.141 | 0.114 |
| 2009 | 0.330 | 0.327 | 0.267 | 0.382 | 1.560 | 1.342 | 0.135 | 0.136 | 0.752 | 0.723 | 0.820 | 1.358 | 0.171 | 0.127 |
| 2010 | 0.356 | 0.34 | 0.28 | 0.391 | 1.674 | 1.512 | 0.144 | 0.145 | 0.81 | 0.835 | 0.838 | 1.612 | 0.153 | 0.147 |
| 2011 | 0.335 | 0.332 | 0.26 | 0.373 | 1.782 | 1.488 | 0.133 | 0.132 | 0.788 | 0.842 | 1.222 | 1.610 | 0.158 | 0.124 |
| 20 | 0.323 | 0.313 | 0.249 | 0.356 | 1.788 | 1.485 | 0.131 | 0.131 | 0.734 | 0.832 | 1.123 | 1.768 | 0.166 | 0.125 |
| 2013 | 0.325 | 0.315 | 0.252 | 0.354 | 1.698 | 1.489 | 0.137 | 0.136 | 0.662 | 0.808 | 1.055 | 1.767 | 0.184 | 0.131 |
| 2014 | 0.327 | 0.309 | 0.256 | 0.354 | 1.656 | 1.449 | 0.142 | 0.142 | 0.645 | 0.790 | 1.047 | 1.601 | 0.173 | 0.135 |
| 2015 | 0.322 | 0.304 | 0.254 | 0.350 | 1.591 | 1.431 | 0.141 | 0.142 | 0.628 | 0.765 | 0.984 | 1.536 | 0.190 | 0.125 |
| 2016 | 0.324 | 0.307 | 0.257 | 0.338 | 1.583 | 1.384 | 0.143 | 0.144 | 0.650 | 0.763 | 0.947 | 1.505 | 0.148 | 0.108 |
| 2017 | 0.314 | 0.304 | 0.245 | 0.328 | 1.571 | 1.346 | 0.131 | 0.132 | 0.661 | 0.752 | 0.924 | 1.490 | 0.116 | 0.118 |
| 2018 | 0.30 | 0.296 | 0.242 | 0.320 | 1.532 | 1.327 | 0.129 | 0.131 | 0.650 | 0.734 | 0.893 | 1.426 | 0.116 | 0.117 |
| 2019 | 0.307 | 0.301 | 0.243 | 0.326 | 1.529 | 1.369 | 0.128 | 0.130 | 0.677 | 0.743 | 0.884 | 1.433 | 0.124 | 0.120 |
| 2020 | 0.323 | 0.319 | 0.250 | 0.341 | 1.485 | 1.304 | 0.134 | 0.136 | 0.658 | 0.702 | 0.883 | 1.391 | 0.139 | 0.096 |
| 2021 | 0.302 | 0.294 | 0.232 | 0.319 | 1.406 | 1.231 | 0.112 | 0.114 | 0.617 | 0.662 | 0.805 | 1.295 | 0.122 | 0.113 |
| 2022 | 0.293 | 0.286 | 0.231 | 0.312 | 1.301 | 1.167 | 0.114 | 0.118 | 0.585 | 0.595 | 0.721 | 1.196 | 0.102 | 0.102 |
| Mean | 0.401 | 0.426 | 0.331 | 0.501 | 1.565 | 1.532 | 0.169 | 0.168 | 0.773 | 0.764 | 0.953 | 1.434 | 0.193 | 0.160 |

In order to save space, the tax rates listed in Table A8 are in percentage terms; thus the actual 1970 tax rate for Asset 1 is 0.00653 instead of 0.653 . The tax rates for assets 7 (Other Structures), 8 (Research and Development) and 12 (Forest Land) were zero for all years.
The economy wide ex post rate of return in year $t, r_{J}^{t}$, can be calculated using Jorgensonian user costs: simply set the value of output in year $t, V_{Y}^{t}$, less the value of labour input, $V_{L}^{t}$, equal to the sum of the year $t$ user cost values defined by (A4):

$$
\begin{equation*}
V_{Y}^{t}=V_{L}^{t}+\sum_{n=1}^{17}\left[r_{J}^{t}+\tau_{n}^{t}-i_{n}^{t}+\delta_{n}^{t}\left(1+i_{n}^{t}\right)\right] P_{K n}^{t} Q_{K n}^{t} ; \quad t=1970, \ldots, 2022 \tag{A5}
\end{equation*}
$$

The $r_{J}^{t}$ are listed in Table A9 below. Once the ex post rates of return are available, Jorgensonian user cost can be calculated. However, the resulting user costs for land assets were extremely volatile and all land assets had negative Jorgensonian user costs.*37 But when we aggregated the user costs over all 17 assets using a Törnqvist quantity index to do the aggregation, we found that the resulting capital services aggregate was fairly smooth and well behaved. The aggregate user cost of capital was always positive. Thus the problem of negative user costs seems to arise in a visible manner only if we want to estimate production functions that distinguish different types of capital where land is included as an asset.
We would like user costs to approximate rental prices for assets. But rental prices are always positive so Jorgensonian user costs are not good approximations to rental prices when land assets are in scope. Owners of assets who rent them to producers and determine the rental prices at the beginning of the year are not able to accurately forecast the end of year prices for their assets. Thus they form expectations about year end asset prices when determining rental prices. This suggests that we should use expected inflation rates in the user cost formulae in place of actual ex post prices. The problem is that there are many methods for forming expectations. Our approach to this problem is to simply smooth the actual ex post asset inflation rates by a more or less arbitrary method and use these smoothed rates in place of the actual rates. We use a centered 7 year moving average of actual asset inflation rates as an approximation to expected asset inflation rates. Denote this smoothed inflation rate for asset $n$ in year $t$ as $i_{n}^{t^{*}} .{ }^{* 38}$ These smoothed inflation rates are listed in Table A9 below. Now replace the $i_{n}^{t}$ in equations (A5) by their smoothed counterparts $i_{n}^{t^{*}}$, replace $r_{J}^{t}$ by $r^{t}$ and solve the resulting equation for $r^{t}$. The $r^{t}$ are also listed in Table A9.

[^18]Table A9: Alternative Rates of Return and Smoothed Asset Inflation Rates

| Year | $r^{t}$ | $r_{J}^{t}$ | $i_{1}^{t^{*}}$ | $i_{2}^{t^{*}}$ | $i_{3}^{t^{*}}$ | $i_{4}^{t^{*}}$ | $i_{5}^{t^{*}}$ | $i_{6}^{t^{*}}$ | ${ }_{7}$ | $i_{8}^{t^{*}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.1002 | 0.0854 | -0.2119 | 0.0195 | 0.0159 | 0.0168 | 0.0402 | 0.0666 | 0.0497 | 0.0297 |
| 1971 | 0.1138 | 0.1216 | -0.1886 | 0.0232 | 0.0153 | 0.0189 | 0.0515 | 0.0690 | 0.0557 | 0.0319 |
| 1972 | 0.1396 | 0.1307 | -0.1591 | 0.0455 | 0.0432 | 0.0569 | 0.0721 | 0.0825 | 0.0858 | 0.0498 |
| 1973 | 0.1436 | 0.1633 | -0.1534 | 0.0442 | 0.0569 | 0.0694 | 0.0737 | 0.0768 | 0.0740 | 0.0521 |
| 1974 | 0.1450 | 0.1857 | -0.1635 | 0.0432 | 0.0661 | 0.0780 | 0.0856 | 0.0821 | 0.0758 | 0.0569 |
| 1975 | 0.1504 | 0.1527 | -0.1636 | 0.0428 | 0.0745 | 0.0866 | 0.0943 | 0.0848 | 0.0782 | 0.0604 |
| 1976 | 0.1569 | 0.1258 | -0.1716 | 0.0446 | 0.0856 | 0.0983 | 0.0996 | 0.0903 | 0.0778 | 0.0666 |
| 1977 | 0.1559 | 0.1497 | -0.1805 | 0.0469 | 0.0920 | 0.1034 | 0.0989 | 0.0911 | 0.0711 | 0.0694 |
| 1978 | 0.1488 | 0.1748 | -0.1797 | 0.0434 | 0.0880 | 0.0935 | 0.0925 | 0.0861 | 0.0683 | 0.0678 |
| 1979 | 0.1366 | 0.1589 | -0.1801 | 0.0398 | 0.0808 | 0.0832 | 0.0861 | 0.0833 | 0.0620 | 0.0666 |
| 1980 | 0.1198 | 0.1249 | -0.1829 | 0.0431 | 0.0741 | 0.0764 | 0.0787 | 0.0806 | 0.0509 | 0.0646 |
| 1981 | 0.1104 | 0.1154 | -0.1748 | 0.0465 | 0.0671 | 0.0676 | 0.0665 | 0.0732 | 0.0447 | 0.0606 |
| 1982 | 0.0924 | 0.0778 | -0.1581 | 0.0445 | 0.0602 | 0.0601 | 0.0555 | 0.0637 | 0.0453 | 0.0546 |
| 1983 | 0.0915 | 0.0548 | -0.1512 | 0.0383 | 0.0498 | 0.0498 | 0.0469 | 0.0527 | 0.0456 | 0.0456 |
| 1984 | 0.0980 | 0.0754 | -0.1425 | 0.0266 | 0.0366 | 0.0378 | 0.0393 | 0.0437 | 0.0347 | 0.0368 |
| 1985 | 0.0983 | 0.1042 | -0.1380 | 0.0152 | 0.0272 | 0.0301 | 0.0357 | 0.0384 | 0.0228 | 0.0291 |
| 1986 | 0.0971 | 0.1144 | -0.1323 | 0.0080 | 0.0238 | 0.0270 | 0.0349 | 0.0361 | 0.0254 | 0.0228 |
| 1987 | 0.0949 | 0.1122 | -0.1208 | 0.0042 | 0.0241 | 0.0290 | 0.0334 | 0.0355 | 0.0347 | 0.0193 |
| 1988 | 0.0914 | 0.1038 | -0.1134 | 0.0001 | 0.0249 | 0.0320 | 0.0321 | 0.0332 | 0.0379 | 0.0177 |
| 1989 | 0.0855 | 0.0923 | -0.1142 | -0.0044 | 0.0235 | 0.0300 | 0.0310 | 0.0304 | 0.0267 | 0.0164 |
| 1990 | 0.0751 | 0.0620 | -0.1128 | -0.0073 | 0.0242 | 0.0266 | 0.0293 | 0.0294 | 0.0173 | 0.0158 |
| 1991 | 0.0680 | 0.0484 | -0.1159 | -0.0104 | 0.0271 | 0.0244 | 0.0287 | 0.0283 | 0.0243 | 0.0166 |
| 1992 | 0.0695 | 0.0513 | -0.1349 | -0.0147 | 0.0258 | 0.0210 | 0.0271 | 0.0254 | 0.0323 | 0.0161 |
| 1993 | 0.0758 | 0.0675 | -0.1574 | -0.0175 | 0.0220 | 0.0167 | 0.0265 | 0.0236 | 0.0305 | 0.0161 |
| 1994 | 0.0880 | 0.0925 | -0.1778 | -0.0226 | 0.0174 | 0.0115 | 0.0290 | 0.0244 | 0.0306 | 0.0161 |
| 1995 | 0.0986 | 0.1003 | -0.1948 | -0.0310 | 0.0130 | 0.0074 | 0.0320 | 0.0278 | 0.0348 | 0.0155 |
| 1996 | 0.1086 | 0.0982 | -0.1989 | -0.0360 | 0.0111 | 0.0058 | 0.0337 | 0.0313 | 0.0389 | 0.0177 |
| 1997 | 0.1167 | 0.1218 | -0.2003 | -0.0395 | 0.0089 | 0.0056 | 0.0349 | 0.0328 | 0.0430 | 0.0185 |
| 1998 | 0.1164 | 0.1335 | -0.2020 | -0.0431 | 0.0058 | 0.0057 | 0.0348 | 0.0333 | 0.0443 | 0.0157 |
| 1999 | 0.1180 | 0.1219 | -0.1900 | -0.0525 | 0.0042 | 0.0062 | 0.0363 | 0.0341 | 0.0420 | 0.0156 |
| 2000 | 0.1188 | 0.1210 | -0.1689 | -0.0674 | 0.0046 | 0.0074 | 0.0414 | 0.0361 | 0.0470 | 0.0172 |
| 2001 | 0.1213 | 0.1095 | -0.1480 | -0.0734 | 0.0070 | 0.0100 | 0.0471 | 0.0407 | 0.0597 | 0.0183 |
| 2002 | 0.1277 | 0.1098 | -0.1307 | -0.0726 | 0.0090 | 0.0129 | 0.0514 | 0.0463 | 0.0723 | 0.0194 |
| 2003 | 0.1288 | 0.1251 | -0.1211 | -0.0745 | 0.0087 | 0.0154 | 0.0503 | 0.0505 | 0.0842 | 0.0185 |
| 2004 | 0.1219 | 0.1656 | -0.1138 | -0.0807 | 0.0082 | 0.0182 | 0.0435 | 0.0521 | 0.0908 | 0.0193 |
| 2005 | 0.1042 | 0.1615 | -0.1036 | -0.0872 | 0.0097 | 0.0197 | 0.0350 | 0.0518 | 0.0845 | 0.0199 |
| 2006 | 0.0827 | 0.0979 | -0.0941 | -0.0862 | 0.0116 | 0.0191 | 0.0269 | 0.0484 | 0.0755 | 0.0186 |
| 2007 | 0.0643 | 0.0529 | -0.0857 | -0.0791 | 0.0130 | 0.0187 | 0.0181 | 0.0431 | 0.0720 | 0.0183 |
| 2008 | 0.0532 | 0.0036 | -0.0751 | -0.0767 | 0.0144 | 0.0186 | 0.0090 | 0.0365 | 0.0636 | 0.0184 |
| 2009 | 0.0515 | -0.0205 | -0.0588 | -0.0782 | 0.0143 | 0.0167 | 0.0040 | 0.0277 | 0.0471 | 0.0168 |
| 2010 | 0.0674 | 0.0337 | -0.0426 | -0.0805 | 0.0147 | 0.0141 | 0.0068 | 0.0208 | 0.0344 | 0.0145 |
| 2011 | 0.0791 | 0.0926 | -0.0284 | -0.0827 | 0.0159 | 0.0111 | 0.0133 | 0.0174 | 0.0286 | 0.0130 |
| 2012 | 0.0901 | 0.1147 | -0.0171 | -0.0837 | 0.0157 | 0.0083 | 0.0214 | 0.0158 | 0.0263 | 0.0109 |
| 2013 | 0.1049 | 0.1301 | -0.0101 | -0.0829 | 0.0143 | 0.0075 | 0.0301 | 0.0184 | 0.0281 | 0.0096 |
| 2014 | 0.1124 | 0.1219 | -0.0050 | -0.0810 | 0.0135 | 0.0072 | 0.0378 | 0.0231 | 0.0252 | 0.0100 |
| 2015 | 0.1105 | 0.1052 | -0.0031 | -0.0808 | 0.0123 | 0.0061 | 0.0429 | 0.0270 | 0.0188 | 0.0100 |
| 2016 | 0.1079 | 0.1064 | -0.0051 | -0.0782 | 0.0118 | 0.0063 | 0.0432 | 0.0311 | 0.0170 | 0.0132 |
| 2017 | 0.1081 | 0.0979 | -0.0057 | -0.0724 | 0.0138 | 0.0089 | 0.0451 | 0.0335 | 0.0167 | 0.0177 |
| 2018 | 0.1155 | 0.0950 | -0.0018 | -0.0611 | 0.0211 | 0.0164 | 0.0562 | 0.0475 | 0.0203 | 0.0202 |
| 2019 | 0.1179 | 0.1027 | 0.0022 | -0.0463 | 0.0258 | 0.0218 | 0.0609 | 0.0579 | 0.0245 | 0.0230 |
| 2020 | 0.1147 | 0.1294 | 0.0035 | -0.0353 | 0.0325 | 0.0287 | 0.0666 | 0.0705 | 0.0296 | 0.0277 |
| 2021 | 0.1396 | 0.1549 | 0.0157 | -0.0232 | 0.0446 | 0.0396 | 0.0860 | 0.0849 | 0.0417 | 0.0284 |
| 2022 | 0.1438 | 0.1257 | 0.0259 | -0.0144 | 0.0554 | 0.0491 | 0.0948 | 0.1077 | 0.0493 | 0.0252 |
| Mean | 0.1074 | 0.1068 | -0.1139 | -0.0237 | 0.0298 | 0.0313 | 0.0470 | 0.0487 | 0.0487 | 0.0281 |


| ar | $i_{9}^{t}$ | $i^{t^{*}}$ | $i_{11}^{t^{*}}$ | $i_{12}^{t^{*}}$ | $i_{13}^{t^{*}}$ | $i_{14}^{t^{*}}$ | $i_{15}^{t^{*}}$ | $i_{16}^{t^{*}}$ | 17 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.0681 | 0.0516 | 0.0801 | 0.0559 | 0.0451 | 0.0451 | 0.1186 | 0.1188 | -0.0352 |
| 1971 | 0.0895 | 0.0469 | 0.1228 | 0.0618 | 0.0525 | 0.0525 | 0.1058 | 0.1296 | 0.0028 |
| 1972 | 954 | 0513 | 140 | 0.0837 | . 0705 | . 0705 | 0.1468 | 04 | 77 |
| 1973 | 0.1041 | 0.0548 | 0.1 | 0.0870 | 0.0762 | 0.0762 | 0.1499 | 0.1638 | 30 |
| 1974 | 0.1114 | 0.0571 | 0.1644 | 0.0896 | 0.0836 | 0.0836 | 0.1634 | 0.1788 | 0.0584 |
| 19 | 0.1133 | 057 | 17 | 0.0854 | 0.0893 | 0.0893 | - | 0.1909 | 21 |
| 1976 | 0.1180 | 0.058 | . 16 | 0.0823 | . 0 | . 0 | 0.1 | 6 | 02 |
| 1977 | 0.1390 | 0.0591 | 0.1445 | 0.0806 | 0.0981 | 0.0981 | 0.1670 | 0.1723 | 0.0778 |
| 1978 | 0.1583 | 0567 | . 120 | 0.0685 | . 0976 | 0.0976 | 0.124 | 0.13 | 0.0691 |
| 1979 | 0.1739 | 0.0524 | 0.0909 | 0.0524 | 0.0927 | 0.0927 | 0.0970 | 0.1094 | 0.0645 |
| 1980 | 0.1767 | 0.0511 | 0.0578 | 0.0430 | 0.0803 | 0.0803 | 0.0890 | 0.0930 | 0.0612 |
| 1981 | 0.1738 | 0.0488 | 0.0144 | 0.0346 | 0.0659 | 0.0659 | 0.0774 | 0.0728 | 25 |
| 82 | 0.1594 | 0472 | -0.0300 | 0.0288 | . 0496 | . 0496 | 674 | . 0561 | 0.0630 |
| 83 | 0.1325 | 0.0469 | -0.0543 | 0.0224 | 0.0404 | 0.0369 | 0.0759 | 0.0579 | 305 |
| 1984 | 0.1047 | 0.0449 | -0.0548 | 0.0160 | 0.0425 | 0.0317 | 0.0958 | 0.0734 | 0289 |
| 1985 | 0.0835 | 0442 | -0.044 | 0.0180 | . 0369 | . 018 | 0.1169 | 0.09 | . 0162 |
| 1986 | 0.0592 | 447 | -0.0 | 0. | 0.0342 | 0.0098 | 0.1252 | 0.1034 | 78 |
| 1987 | 0.0402 | 0.0461 | -0.0209 | 0.0350 | 0.0442 | 0.0098 | 0.1101 | 0.0963 | . 0260 |
| 1988 | 0.0342 | 0.0451 | 0.0035 | 0.0407 | 0.0533 | -0.0003 | 0.0875 | 0.0856 | . 0076 |
| 1989 | 0.038 | 0.0401 | . 0290 | 0.0501 | . 0601 | -0.0194 | 0.0595 | 0.0682 | 0.0106 |
| 1990 | 0.0502 | 0.0367 | 0.0409 | 0.0593 | . 0595 | -0.0382 | 0.0297 | 0.0442 | . 0270 |
| 1991 | 0.0492 | 0.0349 | 0.0410 | 0.0610 | . 0513 | -0.0469 | 0.0051 | 0.0220 | . 0173 |
| 19 | 0.0321 | 13 | 0.0420 | 0.0569 | 0.0483 | -0.0469 | -0.0120 | 0.0059 | 07 |
| 1993 | 0.0226 | 0.0270 | . 043 | 0.0508 | 0.0497 | -0.0428 | -0.0155 | 0.0021 | 0.0127 |
| 1994 | 0.0244 | 0.0205 | 0.0468 | 0.0486 | 0.0549 | -0.0092 | -0.0018 | 0.0149 | 064 |
| 1995 | 0.0 | 72 | 0.0509 | 0.0478 | 25 | 0.0326 | 0.0201 | 0.0353 | 35 |
| 1996 | 0.0264 | 0.0200 | 0.0533 | 0.0413 | 0.0710 | 0.0609 | 0.0444 | 0.0578 | -0.0037 |
| 1997 | 0.0214 | 0.0208 | 0.0554 | 0.0342 | 0.0788 | 0.0881 | 0.0659 | 0.0797 | -0.0037 |
| 1998 | 0. | 0.0186 | 0.0535 | 0.0275 | 0.0844 |  | 0.0873 | 808 | -0.0005 |
| 1999 | 0.0059 | 0.0165 | 0.0483 | 0.0216 | 0.0864 | 0.0862 | 0.1096 | 0.1218 | -0.0094 |
| 2000 | 0.0018 | 0149 | . 05 | 0.0187 | . 0841 | 0.087 | 0.1289 | 0.1416 | 316 |
| 2001 | -0. | 0.0147 | 0.0749 | 0.016 | 0.0770 | 0.0746 | 0.1477 | 0.1614 | -0.0079 |
| 2002 | -0.0060 | 0.0135 | 0.0908 | 0.0177 | 0.0654 | 0.0745 | 0.1501 | 0.1656 | 0.0199 |
| 200 | -0.0130 | 0.0102 | . 0892 | 0.0202 | 0.0500 | 0.0883 | 0.1279 | 0.1457 | 0.0027 |
| 2004 | -0.015 | 0.0074 | 0.0804 | 0.0227 | 0.0319 | 0.0942 | 3 | 0.1079 | 77 |
| 2005 | -0.0103 | 0.0061 | 0.0763 | 0.0269 | 0.0127 | 0.0851 | 0.0417 | 0.0615 | 0.0188 |
| 2006 | -0.0060 | 0.0038 | 0.0808 | 0.0277 | -0.0061 | 0.0416 | -0.0004 | 0.0220 | 0.0167 |
| 2007 | -0.002 | . 000 | 0.0812 | . 02 | -0.022 | . 00 | -0.0407 | -0.0149 | . 0536 |
| 2008 | -0.0004 | 0.0004 | 0.0736 | 0.0210 | -0.0344 | 0.0101 | -0.0808 | -0.0510 | 0.0473 |
| 2009 | 0.0006 | -0.0002 | 0.0683 | 0.0169 | -0.0402 | -0.0083 | -0.0811 | -0.0539 | 0.0221 |
| 10 | 0.0015 | -0.0010 | 0.0697 | 0.0140 | -0.0390 | -0.0118 | -0.0570 | -0.0351 | 0.0383 |
| 2011 | -0.0001 | 0.0006 | . 0719 | 0.0120 | -0.0309 | -0.0059 | -0.0249 | -0.0076 | 0.0169 |
| 2012 | -0.0002 | 0.0034 | 0.0723 | 0.0102 | -0.0174 | 0.0162 | 0.0148 | 0.0277 | 0.0224 |
| 2013 | 0.0035 | 0.0068 | 0.069 | 0.0110 | -0.0008 | 0.0707 | 0.0476 | 0.0564 | 0.0421 |
| 2014 | 0.0035 | . 0104 | . 0590 | 0.0132 | 0.0161 | 0.1098 | 0.0759 | 0.0789 | 0.0042 |
| 2015 | 0.0001 | 0.0120 | 0.0410 | 0.0143 | 0.0306 | 0.0910 | 0.0959 | 0.0919 | 0.0026 |
| 2016 | -0.0034 | 0.0135 | 0.0264 | 0.0137 | 0.0407 | 0.0937 | 0.0874 | 0.0822 | 0.0006 |
| 2017 | -0.0071 | 0.0157 | 0.0267 | 0.0110 | 0.0448 | 0.0987 | 0.0809 | 0.0771 | 0.0364 |
| 2018 | -0.0091 | 0.0198 | 0.0376 | 0.0093 | 0.0434 | 0.0977 | 0.0897 | 0.0869 | 0.0031 |
| 2019 | -0.0096 | 0.0214 | 0.0431 | 0.0082 | 0.0376 | 0.0939 | 0.0889 | 0.0869 | 0.0404 |
| 2020 | -0.0124 | 0.0239 | 0.0510 | 0.0080 | 0.0291 | 0.1020 | 0.0922 | 0.0906 | 0.0626 |
| 2021 | -0.0128 | 0.0313 | 0.0728 | 0.0086 | 0.0166 | 0.1310 | 0.1124 | 0.1111 | 0.0996 |
| 2022 | -0.0102 | 0.0374 | 0.0727 | 0.0115 | 0.0113 | 0.1156 | 0.1136 | 0.1137 | 0.0289 |
| Mean | 0.0441 | 0.0286 | 0.059 | 0.0353 | 0.0444 | 0.0531 | 0.0750 | 0.0827 | 0.0283 |

The mean ex post rate of return on US assets over the 53 year sample period is $10.68 \%$. The mean rate of return on assets when we used smoothed asset inflation rates in the user cost formulae is $10.74 \%$. Both average rates of return are quite high. Note that the ex post rate of return in 2009 was $-2.05 \%$. The smoothed rates of return $r^{t}$ are probably much closer to a realistic cost of capital for US firms. Note that the average smoothed asset inflation rates for assets 11 (Agricultural Land), 14 (Commercial Land) and 15 (Residential Land) are $5.99 \%$, $5.31 \%$ and $7.50 \%$ respectively. These rates of return on holding land are substantial but well below the overall average rate of return on US assets. ${ }^{* 39}$
The smoothed year $t$ user cost for asset $n$ is defined as $U_{n}^{t} \equiv\left[r^{t}+\tau_{n}^{t}-i_{n}^{t^{*}}+\delta_{n}^{t}\left(1+i_{n}^{t}\right)\right] P_{K n}^{t}$ and is listed in Table A10.

[^19]Table A10: User Costs Using Smoothed Asset Inflation Rates

| Year | $U_{1}^{t}$ | $U_{2}^{t}$ | $U_{3}^{t}$ | $U_{4}^{t}$ | $U_{5}^{t}$ | $U_{6}^{t}$ | $U_{7}^{t}$ | $U_{8}^{t}$ | $U_{9}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.57851 | 0.38672 | 0.33739 | 0.29218 | 0.13254 | 0.14531 | 0.07803 | 0.30329 | 0.47085 |
| 1971 | 0.49704 | 0.39464 | 0.35715 | 0.30626 | 0.13948 | 0.16541 | 0.08901 | 0.31661 | 91 |
| 1972 | 0.38463 | 0.41305 | 0.37479 | 0.30988 | 0.15392 | 0.19443 | 0.09125 | 0.33902 | 0.55168 |
| 1973 | 0.32763 | 0.43720 | 0.37807 | 0.31329 | 0.16700 | 0.21838 | 0.11507 | 0.35213 | 0.62755 |
| 1974 | 0.29572 | 0.46290 | 0.38580 | 0.33009 | 0.16903 | 0.23250 | 0.13000 | 0.37017 | 0.68831 |
| 1975 | 0.26069 | 0.51028 | 0.42217 | 0.37371 | 0.18089 | 0.26044 | 0.15107 | 0.40350 | 0.77458 |
| 1976 | 0.24059 | 0.55899 | 0.46864 | 0.41708 | 0.19689 | 0.28039 | 0.17055 | 0.43228 | 0.87249 |
| 1977 | 0.20838 | 0.57385 | 0.50493 | 0.44116 | 0.21113 | 0.29228 | 0.18504 | 0.45023 | 0.93196 |
| 1978 | 0.16735 | 0.57067 | 0.53720 | 0.47616 | 0.22829 | 0.30877 | 0.18653 | 0.46701 | 0.99168 |
| 1979 | 0.12092 | 0.57643 | 0.56693 | 0.50750 | 0.23888 | 0.31812 | 0.19092 | 0.47871 | 1.05592 |
| 1980 | 0.09448 | 0.57141 | 0.58026 | 0.53209 | 0.24131 | 0.32003 | 0.19099 | 0.49071 | 57 |
| 1981 | 0.07514 | 0.58765 | 0.62721 | 0.58452 | 0.26991 | 0.34617 | 0.20023 | 0.52148 | 1.39159 |
| 1982 | 0.06171 | 0.59762 | 0.64766 | 0.59921 | 0.26801 | 0.34923 | 0.17781 | 0.53347 | 1.70687 |
| 1983 | 0.05478 | 0.62902 | 0.70973 | 0.65213 | 0.29900 | 0.39454 | 0.17782 | 0.57707 | 2.16496 |
| 1984 | 0.04599 | 0.67838 | 0.78079 | 0.71778 | 0.34565 | 0.45191 | 0.20854 | 0.63069 | 2.70736 |
| 1985 | 0.03644 | 0.71159 | 0.81913 | 0.74447 | 0.36514 | 0.48349 | 0.23765 | 0.66402 | 2.97896 |
| 1986 | 0.02995 | 0.72736 | 0.83954 | 0.76761 | 0.37728 | 0.50304 | 0.24726 | 0.68402 | 3.15736 |
| 1987 | 0.02504 | 0.72302 | 0.84749 | 0.78024 | 0.39349 | 0.51471 | 0.23375 | 0.69583 | 3.17802 |
| 1988 | 0.02124 | 0.71902 | 0.84839 | 0.78312 | 0.40161 | 0.52869 | 0.21805 | 0.70561 | 3.26192 |
| 1989 | 0.01957 | 0.70765 | 0.84676 | 0.80166 | 0.40395 | 0.54208 | 0.24026 | 0.70839 | 3.56153 |
| 1990 | 0.01713 | 0.69145 | 0.84211 | 0.80003 | 0.38386 | 0.52612 | 0.24160 | 0.69144 | 3.66842 |
| 1991 | 0.01524 | 0.67906 | 0.84741 | 0.80997 | 0.37100 | 0.52562 | 0.20593 | 0.68079 | 3.64493 |
| 1992 | 0.01408 | 0.69063 | 0.89079 | 0.85543 | 0.39587 | 0.55845 | 0.19793 | 0.69382 | 3.77968 |
| 1993 | 0.01243 | 0.69235 | 0.94776 | 0.90163 | 0.43111 | 0.59622 | 0.22008 | 0.71458 | 4.03373 |
| 1994 | 0.01119 | 0.71640 | 1.03269 | 0.97094 | 0.48989 | 0.66312 | 0.25905 | 0.75459 | 4.35018 |
| 1995 | 0.01018 | 0.72759 | 1.09647 | 1.02870 | 0.53433 | 0.70754 | 0.29317 | 0.80747 | 4.56952 |
| 1996 | 0.00833 | 0.72771 | 1.14443 | 1.07484 | 0.58046 | 0.74787 | 0.33633 | 0.84780 | 4.71076 |
| 1997 | 0.00647 | 0.73320 | 1.18186 | 1.10928 | 0.62373 | 0.79367 | 0.35711 | 0.87940 | 4.88440 |
| 1998 | 0.00493 | 0.71427 | 1.19104 | 1.11324 | 0.63811 | 0.81503 | 0.35887 | 0.89874 | 5.07048 |
| 1999 | 0.00370 | 0.68731 | 1.21350 | 1.11767 | 0.65794 | 0.84963 | 0.39410 | 0.91485 | 5.32077 |
| 2000 | 0.00290 | 0.67515 | 1.20479 | 1.12117 | 0.66304 | 0.87510 | 0.38993 | 0.93879 | 5.56525 |
| 2001 | 0.00235 | 0.63648 | 1.20144 | 1.12559 | 0.67823 | 0.89800 | 0.36265 | 0.96024 | 5.66338 |
| 2002 | 0.00195 | 0.59428 | 1.21442 | 1.15024 | 0.71393 | 0.93311 | 0.36186 | 0.98240 | 5.52416 |
| 2003 | 0.00169 | 0.54277 | 1.21541 | 1.16185 | 0.75052 | 0.94071 | 0.33813 | 1.00449 | 5.28280 |
| 2004 | 0.00151 | 0.47722 | 1.22425 | 1.14734 | 0.79646 | 0.93404 | 0.29695 | 1.00220 | 5.07699 |
| 2005 | 0.00131 | 0.43090 | 1.19869 | 1.11472 | 0.80603 | 0.90853 | 0.28159 | 0.96774 | 4.86965 |
| 2006 | 0.00110 | 0.38738 | 1.12858 | 1.04597 | 0.74379 | 0.83990 | 0.21863 | 0.92068 | 4.66140 |
| 2007 | 0.00093 | 0.34532 | 1.05934 | 0.99442 | 0.70768 | 0.80811 | 0.14038 | 0.87919 | 4.49004 |
| 2008 | 0.00080 | 0.30409 | 1.00015 | 0.97643 | 0.70630 | 0.81809 | 0.13706 | 0.86178 | 4.40542 |
| 2009 | 0.00071 | 0.26889 | 0.96310 | 0.97766 | 0.71605 | 0.89281 | 0.24288 | 0.86767 | 4.35922 |
| 2010 | 0.00068 | 0.25895 | 1.08402 | 1.06327 | 0.79273 | 1.06831 | 0.46052 | 0.93577 | 4.38023 |
| 2011 | 0.00065 | 0.24594 | 1.17038 | 1.14059 | 0.83506 | 1.17671 | 0.61330 | 1.00100 | 4.48753 |
| 2012 | 0.00064 | 0.23724 | 1.25560 | 1.22566 | 0.86442 | 1.29759 | 0.75531 | 1.06793 | 4.65954 |
| 2013 | 0.00064 | 0.22870 | 1.33311 | 1.30003 | 0.91970 | 1.40788 | 0.88083 | 1.13689 | 4.82416 |
| 2014 | 0.00064 | 0.21340 | 1.37828 | 1.33823 | 0.96803 | 1.45993 | 0.99501 | 1.16845 | 4.93666 |
| 2015 | 0.00064 | 0.19170 | 1.39874 | 1.33658 | 0.96053 | 1.45397 | 1.07039 | 1.17932 | 4.92517 |
| 2016 | 0.00063 | 0.16975 | 1.38663 | 1.32543 | 0.97342 | 1.42949 | 1.07546 | 1.16203 | 4.94515 |
| 2017 | 0.00064 | 0.15421 | 1.39066 | 1.32904 | 1.01564 | 1.45496 | 1.11397 | 1.14899 | 4.99462 |
| 2018 | 0.00065 | 0.14248 | 1.41360 | 1.33822 | 1.02425 | 1.43878 | 1.14147 | 1.19735 | 5.06859 |
| 2019 | 0.00064 | 0.13065 | 1.42109 | 1.34802 | 1.04851 | 1.44872 | 1.12848 | 1.22772 | 5.01341 |
| 2020 | 0.00063 | 0.12050 | 1.36942 | 1.31929 | 1.00694 | 1.37881 | 1.07523 | 1.23634 | 4.93869 |
| 2021 | 0.00065 | 0.12056 | 1.46810 | 1.42627 | 1.13291 | 1.54439 | 1.23707 | 1.38457 | 5.10271 |
| 2022 | 0.00067 | 0.11701 | 1.54572 | 1.49196 | 1.21777 | 1.54713 | 1.27823 | 1.45752 | 5.04712 |
| Mean | 0.06931 | 0.48285 | 0.96780 | 0.91151 | 0.57230 | 0.76582 | 0.41470 | 0.80560 | 3.58490 |


| Year | $U_{10}^{t}$ | $U_{11}^{t}$ | $U_{12}^{t}$ | $U_{13}^{t}$ | $U_{14}^{t}$ | $U_{15}^{t}$ | $U_{16}^{t}$ | $U_{17}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.37474 | 0.03093 | 0.04429 | 0.06540 | 0.06805 | 0.00036 | -0.01584 | 0.13786 |
| 1971 | 0.40988 | 0.00250 | 0.05407 | 0.07381 | 0.07660 | 0.03158 | -0.01391 | 0.11397 |
| 1972 | 0.46366 | 0.01231 | 0.06237 | 0.08695 | 0.08988 | 0.01480 | -0.02233 | 0.06925 |
| 1973 | 0.48044 | 0.00945 | 0.06780 | 0.09029 | 0.09327 | 0.01635 | -0.02498 | 0.09340 |
| 1974 | 0.50281 | -0.01584 | 0.07253 | 0.08825 | 0.09121 | -0.00103 | -0.0532 | 0.10243 |
| 1975 | 0.54355 | -0.02876 | 0.09689 | 0.09822 | 0.10141 | -0.02119 | -0.07959 | 0.09552 |
| 1976 | 0.60982 | 0.00259 | 0.12572 | 0.10994 | 0.11339 | -0.03075 | -0.08069 | 0.09589 |
| 1977 | 0.64256 | 0.05337 | 0.13458 | 0.11024 | 0.11380 | 0.01252 | -0.04043 | 0.11277 |
| 1978 | 0.66493 | 0.11070 | 0.15165 | 0.10521 | 0.10860 | 0.12140 | 0.04643 | 0.11907 |
| 1979 | 0.69688 | 0.18914 | 0.16604 | 0.09984 | 0.10309 | 0.20287 | 36 | 2 |
| 1980 | 0.68455 | 0.28027 | 0.159 | 0.10012 | 0.10353 | 0.18356 | 0.13921 | 0.10850 |
| 1981 | 0.72202 | 0.44079 | 0.16998 | 0.12174 | 0.12556 | 0.19856 | 0.20228 | 0.07560 |
| 1982 | 0.72215 | 0.53705 | 0.15049 | 0.13303 | 0.13765 | 0.15686 | 0.19199 | 0.06636 |
| 1983 | 0.75429 | 0.61900 | 0.16657 | 0.16760 | 0.18284 | 0.12035 | 0.17859 | 0.13394 |
| 1984 | 0.80849 | 0.60003 | 0.19623 | 0.17683 | 0.21233 | 0.07051 | 0.13837 | 0.14936 |
| 1985 | 0.83758 | 0.48044 | 0.19180 | 0.18996 | 0.24513 | -0.02694 | 0.03568 | 0.19746 |
| 1986 | 0.88042 | 0.39069 | 0.16936 | 0.19175 | 0.26388 | -0.08790 | -0.02660 | 0.17185 |
| 1987 | 0.90003 | 0.33700 | 0.14471 | 0.16651 | 0.26367 | -0.01281 | 0.00454 | 0.15530 |
| 1988 | 0.92376 | 0.27201 | 0.12679 | 0.14339 | 0.29713 | 0.14196 | 0.06381 | 0.19750 |
| 1989 | 0.96443 | 0.18989 | 0.09470 | 0.11173 | 0.34675 | 0.36515 | 0.18050 | 0.18224 |
| 1990 | 0.99112 | 0.12931 | 0.04541 | 0.08356 | 0.37972 | 0.57323 | 0.32732 | 0.12730 |
| 1991 | 1.02306 | 0.11175 | 0.02134 | 0.09417 | 0.37807 | 0.72969 | 0.47322 | 0.13061 |
| 1992 | 1.07589 | 0.11779 | 0.03970 | 0.11765 | 0.34660 | 0.87822 | 0.63455 | 0.12507 |
| 1993 | 1.11989 | 0.14171 | 0.08412 | 0.14241 | 0.30211 | 0.94060 | 0.72697 | 0.16793 |
| 1994 | 1.19276 | 0.18215 | 0.14206 | 0.17980 | 0.22536 | 0.92000 | 0.72239 | 0.22092 |
| 1995 | 1.26867 | 0.21471 | 0.19249 | 0.20167 | 0.15989 | 0.80574 | 0.63182 | 0.25197 |
| 1996 | 1.32646 | 0.25530 | 0.26490 | 0.22174 | 0.12544 | 0.68444 | 0.51800 | 0.31287 |
| 1997 | 1.37193 | 0.29290 | 0.33574 | 0.23874 | 0.08875 | 0.58181 | 0.39708 | 0.34090 |
| 1998 | 1.38488 | 0.31153 | 0.37644 | 0.22362 | 0.10091 | 0.41308 | 0.19429 | 0.30770 |
| 1999 | 1.40989 | 0.35843 | 0.42123 | 0.24006 | 0.13642 | 0.23833 | -0.01962 | 0.32437 |
| 2000 | 1.46448 | 0.35550 | 0.44744 | 0.28230 | 0.13967 | 0.04678 | -0.29025 | 0.38105 |
| 2001 | 1.49513 | 0.28410 | 0.47756 | 0.37662 | 0.20777 | -0.17990 | -0.61908 | 0.33497 |
| 2002 | 1.53919 | 0.24117 | 0.50321 | 0.55270 | 0.23123 | -0.13719 | -0.67272 | 0.28203 |
| 2003 | 1.57132 | 0.25700 | 0.49580 | 0.73169 | 0.18512 | 0.24271 | -0.32488 | 0.32690 |
| 2004 | 1.56227 | 0.29378 | 0.45958 | 0.86940 | 0.14142 | 0.92288 | 0.40318 | 0.23297 |
| 2005 | 1.51665 | 0.24951 | 0.36736 | 0.90977 | 0.12179 | 1.82375 | 1.37820 | 0.20929 |
| 2006 | 1.44738 | 0.06917 | 0.27171 | 0.89318 | 0.26107 | 2.60047 | 2.22434 | 0.18952 |
| 2007 | 1.39074 | -0.09423 | 0.20414 | 0.86599 | 0.41216 | 3.16591 | 2.91600 | 0.02942 |
| 2008 | 1.34342 | -0.12346 | 0.17210 | 0.84974 | 0.35062 | 3.43781 | 3.39132 | 0.01954 |
| 2009 | 1.33501 | -0.08288 | 0.19047 | 0.85446 | 0.44425 | 2.77929 | 2.84916 | 0.08642 |
| 2010 | 1.40211 | 0.05546 | 0.29484 | 0.93338 | 0.39713 | 2.28360 | 2.43864 | 0.08426 |
| 2011 | 1.42090 | 0.15507 | 0.36809 | 0.90719 | 0.35183 | 1.71546 | 1.89764 | 0.19586 |
| 2012 | 1.45553 | 0.29440 | 0.43831 | 0.84364 | 0.35724 | 1.21747 | 1.32720 | 0.22343 |
| 2013 | 1.50256 | 0.55024 | 0.52063 | 0.80542 | 0.20051 | 1.10585 | 1.18008 | 0.20515 |
| 2014 | 1.53018 | 0.84420 | 0.56161 | 0.73655 | 0.06580 | 0.89409 | 0.94329 | 0.34728 |
| 2015 | 1.54587 | 1.09565 | 0.55762 | 0.63493 | 0.17214 | 0.55647 | 0.59346 | 0.33855 |
| 2016 | 1.56595 | 1.27985 | 0.55669 | 0.56806 | 0.15341 | 0.71418 | 0.84722 | 0.35375 |
| 2017 | 1.58709 | 1.31156 | 0.57858 | 0.56859 | 0.12963 | 0.91707 | 1.08404 | 0.26509 |
| 2018 | 1.62778 | 1.28555 | 0.63816 | 0.67656 | 0.19953 | 0.94071 | 1.07927 | 0.35321 |
| 2019 | 1.65040 | 1.25690 | 0.66464 | 0.78615 | 0.25186 | 1.07373 | 1.23153 | 0.25445 |
| 2020 | 1.64051 | 1.11048 | 0.65015 | 0.86630 | 0.18007 | 0.96465 | 1.03926 | 0.16721 |
| 2021 | 1.74721 | 1.23861 | 0.80095 | 1.24490 | 0.16182 | 1.18079 | 1.33675 | 0.16072 |
| 2022 | 1.82368 | 1.43530 | 0.81787 | 1.35050 | 0.39685 | 1.42448 | 1.61212 | 0.33714 |
| Mean | 1.14940 | 0.37721 | 0.29258 | 0.42042 | 0.20555 | 0.71533 | 0.63055 | 0.19214 |

Moving to the use of smoothed asset inflation rates in the user cost formula did not eliminate the earlier negative user costs but it did reduce the number: there are 5,8 and 14 negative user costs for assets 11 (Agricultural Land), 15 (Residential Land and 16 (Other Land) respectively. We use the data for the smoothed user costs, $U_{1}^{t}, \ldots, U_{17}^{t}$, along with the data on the beginning of the year asset stocks, $Q_{K 1}^{t}, \ldots, Q_{K 17}^{t}$, to form 4 capital services aggregates. The first subaggregate is a Machinery and Equipment aggregate consisting of assets 1-4, the second is a Structures subaggregate consisting of assets 5-7, the third is an Other subaggregate which consists of assets 8-10 and 17 (Research and Development, Computer Software, Other Intangibles and Inventory Stocks) and the fourth is a Land subaggregate consisting of assets 11-16. Denote the prices and quantities for these aggregate capital services by $P_{U M}^{t}, P_{U S}^{t}, P_{U O}^{t}$ and $P_{U L}^{t}$ and $Q_{U M}^{t}, Q_{U S}^{t}, Q_{U O}^{t}$ and $Q_{U L}^{t}$. These aggregate user costs are listed in Table A11 (with prices normalized to equal 1 in 1970). ${ }^{* 40}$ The four subaggregates were then aggregated into an overall capital services aggregate using Törnqvist direct quantity aggregation. The year $t$ value, price and quantity of the capital services aggregate are denoted by $V_{U}^{t}, P_{U}^{t}$ and $Q_{U}^{t}$.

[^20]Table A11: Prices and Quantities of Four Capital Services Subaggregates

| Year | $P_{U}^{t}$ | $P_{U M}^{t}$ | $P_{U S}^{t}$ | $P_{U O}^{t}$ | $P_{U L}^{t}$ | $Q_{U}^{t}$ | $Q_{U M}^{t}$ | $Q_{U S}^{t}$ | $Q_{U O}^{t}$ | $Q_{U L}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 1.00000 | 0.36904 | 0.11036 | 0.15996 | 0.03686 | 0.06187 |
| 1971 | 1.08313 | 1.03597 | 1.09951 | 1.01687 | 1.16595 | 0.37920 | 0.11168 | 0.16463 | 0.04069 | 0.06229 |
| 1972 | 1.16234 | 1.04721 | 1.21456 | 1.04646 | 1.30444 | 0.39040 | 0.11269 | 0.17044 | 0.04430 | 0.06317 |
| 1973 | 1.25503 | 1.05477 | 1.38372 | 1.12108 | 1.35412 | 0.40492 | 0.11631 | 0.17708 | 0.04772 | 0.06424 |
| 1974 | 1.28506 | 1.09630 | 1.46556 | 1.18792 | 1.19002 | 0.42314 | 0.12377 | 0.18379 | 0.05137 | 0.06528 |
| 1975 | 1.40903 | 1.21704 | 1.62842 | 1.27334 | 1.23802 | 0.43705 | 0.13053 | 0.18845 | 0.05437 | 0.06530 |
| 1976 | 1.56385 | 1.34589 | 1.78646 | 1.36808 | 1.48236 | 0.44421 | 0.13280 | 0.19162 | 0.05570 | 0.06572 |
| 1977 | 1.69764 | 1.41718 | 1.90740 | 1.44961 | 1.82235 | 0.45601 | 0.13666 | 0.19582 | 0.05885 | 76 |
| 1978 | 1.85720 | 1.49972 | 2.00490 | 1.51045 | 2.40931 | 0.47220 | 0.14445 | 0.20131 | 0.06220 | 0.06756 |
| 1979 | 1.98544 | 1.56750 | 2.07522 | 1.54956 | 2.92053 | 0.49305 | 0.15566 | 0.20820 | 0.06605 | 0.06866 |
| 1980 | 2.02606 | 1.60895 | 2.08786 | 1.57443 | 3.05638 | 0.51493 | 0.16824 | 0.21538 | 0.06985 | 0.06967 |
| 1981 | 2.24890 | 1.72733 | 2.27049 | 1.61651 | 3.81838 | 0.53012 | 0.17698 | 0.22113 | 0.07277 | 0.06987 |
| 1982 | 2.25221 | 1.75311 | 2.19204 | 1.66973 | 3.97121 | 0.54553 | 0.18568 | 0.22656 | 0.07661 | 0.07015 |
| 1983 | 2.49123 | 1.88233 | 2.38235 | 1.95414 | 4.56370 | 0.55411 | 0.18992 | 0.23028 | 0.07914 | 0.07004 |
| 1984 | 2.74284 | 2.0317 | 2.75739 | 2.18078 | 4.65848 | 0.56647 | 0.19563 | 0.23506 | 0.08168 | 0.07083 |
| 1985 | 2.83386 | 2.074 | 2.98456 | 2.37799 | 4.28176 | 0.58979 | 0.20784 | 0.24222 | 0.08798 | 0.07196 |
| 1986 | 2.85164 | 2.09597 | 3.09574 | 2.40715 | 3.95130 | 0.61371 | 0.22012 | 0.25024 | 0.09360 | 0.07308 |
| 1987 | 2.87525 | 2.08769 | 3.12696 | 2.41093 | 4.06416 | 0.63603 | 0.23041 | 0.25784 | 0.09922 | 0.07437 |
| 1988 | 2.97713 | 2.06138 | 3.12571 | 2.52260 | 4.88096 | 0.65663 | 0.23984 | 0.26514 | 0.10565 | 0.07482 |
| 1989 | 3.16907 | 2.06988 | 3.23373 | 2.55036 | 6.12367 | 0.67831 | 0.25007 | 0.27203 | 0.11133 | 0.07649 |
| 1990 | 3.21591 | 2.03294 | 3.13708 | 2.43889 | 7.17369 | 0.69893 | 0.26095 | 0.27811 | 0.11741 | 0.07784 |
| 199 | 3.21771 | 2.02263 | 2.96704 | 2.42620 | 7.85437 | 0.71619 | 0.26881 | 0.28381 | 0.12367 | 0.07877 |
| 1992 | 3.35499 | 2.09770 | 3.08730 | 2.47619 | 8.3365 | 0.72735 | 0.27288 | 0.28720 | 0.12936 | 0.07927 |
| 1993 | 3.51648 | 2.16494 | 3.35742 | 2.63595 | 8.34745 | 0.74080 | 0.27928 | 0.29038 | 0.13477 | 0.08029 |
| 1994 | 3.71368 | 2.28640 | 3.82189 | 2.86885 | 7.66913 | 0.75636 | 0.29028 | 0.29445 | 0.13910 | 0.08095 |
| 1995 | 3.81287 | 2.37694 | 4.17902 | 3.08175 | 6.60094 | 0.77719 | 0.30559 | 0.29928 | 0.14414 | 0.08211 |
| 1996 | 3.94083 | 2.41335 | 4.56684 | 3.30574 | 5.86780 | 0.80098 | 0.32680 | 0.30452 | 0.14872 | 0.08274 |
| 1997 | 4.01544 | 2.41650 | 4.87639 | 3.45623 | 5.16699 | 0.83132 | 0.35146 | 0.31133 | 0.15545 | 0.08388 |
| 1998 | 3.96370 | 2.35375 | 4.97204 | 3.46774 | 4.54040 | 0.86799 | 0.37957 | 0.31879 | 0.16567 | 0.08535 |
| 1999 | 4.01973 | 2.29037 | 5.22207 | 3.57517 | 4.25861 | 0.91021 | 0.41370 | 0.32728 | 0.17700 | 0.08673 |
| 2000 | 3.97127 | 2.22925 | 5.27010 | 3.77848 | 3.43361 | 0.95787 | 0.45432 | 0.33610 | 0.19015 | 0.08777 |
| 2001 | 3.91420 | 2.16638 | 5.27796 | 3.75926 | 3.14927 | 1.00696 | 0.49576 | 0.34545 | 0.20338 | 0.08878 |
| 2002 | 4.01220 | 2.13402 | 5.47148 | 3.69325 | 3.85525 | 1.04022 | 0.52053 | 0.35443 | 0.21139 | 0.08891 |
| 2003 | 4.17592 | 2.09130 | 5.54995 | 3.77135 | 5.53649 | 1.06567 | 0.53372 | 0.36220 | 0.21960 | 0.08951 |
| 2004 | 4.38270 | 2.02325 | 5.58836 | 3.57261 | 8.76703 | 1.09307 | 0.54923 | 0.37077 | 0.22822 | 0.09034 |
| 2005 | 4.64007 | 1.93015 | 5.53727 | 3.41956 | 13.08549 | 1.12583 | 0.57220 | 0.38039 | 0.23846 | 0.09153 |
| 2006 | 4.77041 | 1.78397 | 4.98457 | 3.24401 | 18.40643 | 1.15993 | 0.60135 | 0.38982 | 0.24951 | 0.09280 |
| 2007 | 4.82905 | 1.65948 | 4.54980 | 2.85301 | 22.87290 | 1.19821 | 0.64211 | 0.39935 | 0.26138 | 0.09435 |
| 2008 | 4.81727 | 1.57651 | 4.54928 | 2.77749 | 23.51953 | 1.22614 | 0.68267 | 0.40566 | 0.27406 | 0.09455 |
| 2009 | 4.80782 | 1.52587 | 5.02662 | 2.88408 | 21.39923 | 1.24361 | 0.71045 | 0.40789 | 0.28536 | 0.09447 |
| 2010 | 5.10751 | 1.62739 | 6.28166 | 3.02047 | 18.75001 | 1.24242 | 0.70563 | 0.40624 | 0.29236 | 0.09399 |
| 2011 | 5.23770 | 1.70513 | 7.10030 | 3.31386 | 15.45404 | 1.25238 | 0.72289 | 0.40301 | 0.30112 | 0.09496 |
| 2012 | 5.40376 | 1.79647 | 7.85808 | 3.51452 | 13.10178 | 1.26462 | 0.74933 | 0.40001 | 0.31128 | 0.09542 |
| 2013 | 5.57944 | 1.87690 | 8.65426 | 3.65634 | 11.00654 | 1.28291 | 0.78392 | 0.39866 | 0.32218 | 0.09617 |
| 2014 | 5.69367 | 1.90983 | 9.30205 | 3.94162 | 8.61347 | 1.30478 | 0.82042 | 0.39887 | 0.33485 | 0.09659 |
| 2015 | 5.73352 | 1.89269 | 9.51683 | 3.94753 | 8.39752 | 1.33152 | 0.86167 | 0.40053 | 0.34833 | 0.09724 |
| 2016 | 5.77818 | 1.85610 | 9.54956 | 3.94952 | 9.20232 | 1.36111 | 0.90148 | 0.40349 | 0.36399 | 0.09788 |
| 2017 | 5.89187 | 1.84398 | 9.88348 | 3.81218 | 10.05129 | 1.38832 | 0.93152 | 0.40691 | 0.38118 | 0.09823 |
| 2018 | 6.07416 | 1.84720 | 9.98701 | 4.04441 | 11.26479 | 1.41677 | 0.96266 | 0.40992 | 0.40039 | 0.09891 |
| 2019 | 6.16574 | 1.83940 | 10.05374 | 3.94408 | 12.80517 | 1.45150 | 1.00020 | 0.41416 | 0.42388 | 0.09950 |
| 2020 | 5.84925 | 1.78350 | 9.60846 | 3.81624 | 11.13304 | 1.48542 | 1.03225 | 0.41855 | 0.44916 | 0.09987 |
| 2021 | 6.50474 | 1.90310 | 10.89250 | 4.10471 | 12.98331 | 1.50862 | 1.03848 | 0.42233 | 0.47315 | 0.09970 |
| 2022 | 7.11141 | 1.97545 | 11.37361 | 4.43145 | 17.82376 | 1.54067 | 1.05234 | 0.42630 | 0.50300 | 0.10098 |

The price of aggregate capital services $P_{U}^{t}$ increased 7.11 fold over the sample period. The price of Machinery and Equipment Services $P_{U M}^{t}$ increased 1.98 fold, the price of Structures Services $P_{U S}^{t}$ increased 11.37 fold, the price of Other Capital Services $P_{U O}^{t}$ increased 4.43 fold and the price of Land Services $P_{U L}^{t}$ increased 17.82 fold.
Denote the values of the 4 subaggregates in year $t$ by $V_{U M}^{t}, V_{U S}^{t}, V_{U O}^{t}$ and $V_{U L}^{t}$. Denote the corresponding year $t$ value shares of aggregate capital services by $s_{U M}^{t}, s_{U S}^{t}, s_{U O}^{t}$ and $s_{U L}^{t}$. These variables are listed in Table A12 below.

Table A12: Values and Shares of Four Capital Services Subaggregates

| Year | $V_{U}^{t}$ | $V_{U M}^{t}$ | $V_{U S}^{t}$ | $V_{U}^{t}{ }^{\text {a }}$ | $V_{U L}^{t}$ | M | $s_{U S}^{t}$ | ${ }_{\text {U }}^{t}$ | $s_{U L}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.36904 | 0.11036 | 0.15996 | 0.03686 | 0.06187 | 0.29903 | 0.43344 | 0.09988 | 0.16764 |
| 1971 | . 41072 | 0.11570 | 0.18102 | 0.04138 | 0.07262 | 0.28171 | 0.44073 | 0.10075 | 82 |
| 1972 | 0.45378 | 0.11801 | 0.20700 | 0.04636 | 0.08240 | 0.26007 | 0.45618 | 0.10216 | 59 |
| 1973 | 0.50819 | 0.12268 | 0.24503 | 0.05349 | 0.08698 | 0.24141 | 0.48217 | 0.10526 | 0.17116 |
| 1974 | 0.54376 | 0.13569 | 0.26936 | 0.06103 | 0.07768 | 0.24955 | 0.49536 | 0.11223 | 0.14286 |
| 1975 | 0.61582 | 0.15886 | 0.30688 | 0.06923 | 0.08084 | 0.25797 | 0.49833 | 0.11242 | . 13128 |
| 1976 | 0.69468 | 0.17873 | 0.34232 | 0.07620 | 0.09742 | 0.25728 | 0.49278 | 0.10970 | 0.14024 |
| 1977 | 0.77414 | 0.19367 | 0.37351 | 0.08530 | 0.12166 | 0.25018 | 0.48248 | 0.11019 | 0.15715 |
| 1978 | 0.87697 | 0.21664 | 0.40360 | 0.09395 | 0.16278 | 0.24703 | 0.46023 | 0.10713 | 0.18562 |
| 1979 | . 97893 | 0.24400 | 0.43205 | 0.10235 | 0.20053 | 0.24925 | 0.44135 | 0.10455 | 5 |
| 1980 | 1.0 | 0.27068 | 0.44969 | 0.10998 | 0.21294 | 0.25945 | 0.43103 | 10541 | 10 |
| 19 | 1.19219 | 0.30570 | 0.50208 | 0.11763 | 0.26678 | 0.25642 | 4 | 0.09867 | 77 |
| 1982 | 1.22864 | 0.32551 | 0.49662 | 0.1279 | 0.27858 | 0.26494 | 0.40420 | 0.10412 | 0.22674 |
| 1983 | 1.38042 | 0.35750 | 0.54862 | 0.15466 | 0.31964 | 0.25898 | 0.39743 | 0.11204 | 0.23156 |
| 1984 | 1.55373 | 0.39748 | 0.64817 | 0.17813 | 0.32995 | 0.25582 | 0.41717 | 0.11465 | 0.21236 |
| 1985 | 1.67139 | 0.43111 | 0.72292 | 0.20923 | 0.30813 | 0.25794 | 0.43252 | 0.12518 | 0.18436 |
| 1986 | 1.75009 | 0.46136 | 0.77468 | 0.22531 | 0.28874 | 0.26362 | 0.44265 | 0.12874 | 0.16499 |
| 19 | 1.82875 | 0.48102 | 0.80626 | 0.23922 | 0.30225 | 0.26303 | 0.44088 | 0.13081 | 0.16528 |
| 19 | 1.95486 | 0.49440 | 0.82874 | 0.26650 | 0.36521 | 0.25291 | 0.42394 | 0.13633 | 0.18682 |
| 19 | 2.14961 | 0.51762 | 0.87967 | 0.28394 | 0.46838 | 0.24080 | 0.40922 | 0.13209 | 0.21789 |
| 1990 | 2.24770 | 0.53050 | 0.87244 | 0.28634 | 0.55842 | 0.23602 | 0.38815 | 0.12739 | 4 |
| 1991 | 2.30449 | 0.54370 | 0.84209 | 0.30004 | 0.61866 | 0.23593 | 0.36541 | 0.13020 | 0.26846 |
| 1992 | 2.44024 | 0.57241 | 0.88667 | 0.32032 | 0.66084 | 0.23457 | 0.36335 | 0.13127 | 0.27081 |
| 1993 | 2.60502 | 0.60462 | 0.97492 | 0.35525 | 0.67023 | 0.23210 | 0.37425 | 0.13637 | 0.25728 |
| 1994 | 2.80889 | 0.66371 | 1.12535 | 0.39904 | 0.62079 | 0.23629 | 0.40064 | 0.14206 | 0.22101 |
| 1995 | 2.96331 | 0.72637 | 1.25071 | 0.44420 | 0.54203 | 0.24512 | 0.42206 | 0.14990 | 0.18291 |
| 1996 | 3.15652 | 0.78869 | 1.39071 | 0.49162 | 0.48551 | 0.24986 | 0.44058 | 0.15575 | 0.15381 |
| 1997 | 3.33812 | 0.84929 | 1.51817 | 0.53726 | 0.43339 | 0.25442 | 0.45480 | 0.16095 | 0.12983 |
| 1998 | 3.44043 | 0.89342 | 1.58501 | 0.57449 | 0.38750 | 0.25968 | 0.46070 | 0.16698 | 0.11263 |
| 1999 | 3.65878 | 0.94752 | 1.70908 | 0.63282 | 0.36936 | 0.25897 | 0.46712 | 0.17296 | 0.10095 |
| 2000 | 3.80395 | 1.01280 | 1.77130 | 0.71849 | 0.30135 | 0.26625 | 0.46565 | 0.18888 | 0.07922 |
| 2001 | 3.94143 | 1.07400 | 1.82330 | 0.76454 | 0.27959 | 0.27249 | 0.46260 | 0.19398 | 0.07094 |
| 2002 | 4.17358 | 1.11083 | 1.93928 | 0.78071 | 0.34276 | 0.26616 | 0.46466 | 0.18706 | 0.08213 |
| 2003 | 4.45017 | 1.11618 | 2.01020 | 0.82821 | 0.49559 | 0.25082 | 0.45171 | 0.18611 | 0.11136 |
| 2004 | 4.79062 | 1.11124 | 2.07202 | 0.81535 | 0.79201 | 0.23196 | 0.43252 | 0.17020 | 0.16533 |
| 2005 | 5.22393 | 1.10442 | 2.10629 | 0.81544 | 1.19777 | 0.21142 | 0.40320 | 0.15610 | 0.22929 |
| 2006 | 5.53336 | 1.07279 | 1.94307 | 0.80942 | 1.70807 | 0.19388 | 0.35116 | 0.14628 | 0.30869 |
| 2007 | 5.78623 | 1.06557 | 1.81697 | 0.74573 | 2.15796 | 0.18416 | 0.31402 | 0.12888 | 0.37295 |
| 2008 | 5.90667 | 1.07623 | 1.84548 | 0.76121 | 2.22375 | 0.18221 | 0.31244 | 0.12887 | 0.37648 |
| 2009 | 5.97906 | 1.08406 | 2.05033 | 0.82299 | 2.02168 | 0.18131 | 0.34292 | 0.13765 | 0.33813 |
| 2010 | 6.34566 | 1.14833 | 2.55189 | 0.88307 | 1.76238 | 0.18096 | 0.40215 | 0.13916 | 0.27773 |
| 20 | 6.55958 | 1.23262 | 2.86152 | 0.99788 | 1.46756 | 0.18791 | 0.43624 | 0.15213 | 0.22373 |
| 2012 | 6.83371 | 1.34615 | 3.14335 | 1.09399 | 1.25022 | 0.19699 | 0.45998 | 0.16009 | 0.18295 |
| 2013 | 7.15793 | 1.47134 | 3.45008 | 1.17801 | 1.05851 | 0.20555 | 0.48199 | 0.16457 | 0.14788 |
| 2014 | 7.42901 | 1.56687 | 3.71031 | 1.31983 | 0.83200 | 0.21091 | 0.49944 | 0.17766 | 0.11199 |
| 2015 | 7.63428 | 1.63088 | 3.81180 | 1.37506 | 0.81654 | 0.21363 | 0.49930 | 0.18012 | 0.10696 |
| 2016 | 7.86474 | 1.67325 | 3.85317 | 1.43757 | 0.90075 | 0.21275 | 0.48993 | 0.18279 | 0.11453 |
| 2017 | 8.17979 | 1.71770 | 4.02164 | 1.45314 | 0.98731 | 0.20999 | 0.49166 | 0.17765 | 0.12070 |
| 2018 | 8.60569 | 1.77823 | 4.09391 | 1.61932 | 1.11423 | 0.20663 | 0.47572 | 0.18817 | 0.12948 |
| 2019 | 8.94954 | 1.83976 | 4.16382 | 1.67183 | 1.27413 | 0.20557 | 0.46526 | 0.18681 | 0.14237 |
| 2020 | 8.68861 | 1.84103 | 4.02165 | 1.71411 | 1.11181 | 0.21189 | 0.46287 | 0.19728 | 0.12796 |
| 2021 | 9.81320 | 1.97633 | 4.60023 | 1.94215 | 1.29450 | 0.20139 | 0.46878 | 0.19791 | 0.13191 |
| 2022 | 10.95630 | 2.07883 | 4.84857 | 2.22903 | 1.79988 | 0.18974 | 0.44254 | 0.20345 | 0.16428 |
| Mean | 3.87830 | 0.84691 | 1.70270 | 0.63580 | 0.69288 | 0.23556 | 0.43617 | 0.14449 | 0.18378 |

There is a considerable amount of variation in the shares of the four capital services subaggregates in total capital services. The average share of Structures was $43.6 \%$, of Machinery and Equipment was $23.6 \%$, of Land was $18.4 \%$ and of Other Capital (mostly R\&D and Inventory services) was $18.4 \%$. All of the subaggregate user costs were positive and three of them were fairly smooth but the share of Land services was quite volatile. ${ }^{* 41}$
We turn now to the construction of alternative indexes of output, input and Total Factor Productivity (TFP) for the US economy.
In Table A6, the GDP output index $Q_{Y}^{t}$ and the companion output price index $P_{Y}^{t}$ were listed. The GDP price index $P_{Y}^{t}$ is a chained Törnqvist price index that uses the prices $P_{C}^{t}, P_{G}^{t}, P_{I}^{t}, P_{X}^{t}, P_{M}^{t}$ and the quantity indexes $Q_{C}^{t}, Q_{G}^{t}, Q_{I}^{t}, Q_{X}^{t},-Q_{M}^{t}$ as inputs into the price index $P_{Y}^{t}$. The quantity index $Q_{Y}^{t}$ was defined residually by deflating the year $t$ value of GDP, $V_{Y}^{t}$ by the year $t$ GDP price index; i.e., $Q_{Y}^{t} \equiv V_{Y}^{t} / P_{Y}^{t}$. We are now in a position to define a companion index of the year $t$ quantity of inputs $Q_{Z}^{t}$ as a chained Törnqvist quantity index that uses the prices of labour and the four types of capital services $P_{L}^{t}, P_{U M}^{t}, P_{U S}^{t}, P_{U O}^{t}, P_{U L}^{t}$ and the quantity indexes $Q_{L}^{t}, Q_{U M}^{t}, Q_{U S}^{t}, Q_{U O}^{t}, Q_{U L}^{t}$ as inputs into the quantity index $Q_{Z}^{t}$ and the companion input price index $P_{Z}^{t}$ is defined residually by deflating the year $t$ value of GDP, $V_{Z}^{t}=V_{Y}^{t}$ by the year $t$ GDP input quantity index; i.e., $P_{Z}^{t} \equiv V_{Z}^{t} / Q_{Y}^{t}$. The year $t$ level of Total Factor Productivity is defined as the year $t$ output quantity $Q_{Y}^{t}$ divided by the year $t$ input quantity $Q_{Z}^{t}$; i.e.,

$$
\begin{equation*}
\mathrm{TFP}^{t} \equiv Q_{Y}^{t} / Q_{Z}^{t} ; \quad t=1970, \ldots, 2022 \tag{A6}
\end{equation*}
$$

TFP growth is defined as $\mathrm{TFPG}^{t} \equiv \mathrm{TFP}^{t} / \mathrm{TFP}^{t-1}$ for $t=1971, \ldots, 2022 .{ }^{* 42}$ The series $Q_{Y}^{t}, Q_{Z}^{t}, \mathrm{TFP}^{t}$ and $\mathrm{TFPG}^{t}$ are listed in Table A13.
Recall that Jorgensonian balancing rates of return $r_{J}^{t}$ were listed in Table A9 above. These rates of return can be used with actual ex post asset inflation rates to construct Jorgensonian user costs. We used these user costs in place of our smoothed user costs to construct Jorgensonian estimates of year $t$ Total Factor Productivity, $\mathrm{TFP}_{J}^{t}$ and year $t$ estimates of TFP growth, $\mathrm{TFPG}_{J}^{t}$. These alternative GDP based estimates of TFP are also listed in Table A13.
In the main text, we did not use GDP as our output concept. Instead, we used Gross Output (GO) as our output concept. Gross Output simply regards imports as an input instead of as a negative output. Thus the value of Gross Output in year $t, V_{G Y}^{t}$, is defined as follows:

$$
\begin{equation*}
V_{G Y}^{t} \equiv P_{C}^{t} Q_{C}^{t}+P_{G}^{t} Q_{G}^{t}+P_{I}^{t} Q_{I}^{t}+P_{X}^{t} Q_{X}^{t} ; \quad t=1970, \ldots, 2022 \tag{A7}
\end{equation*}
$$

The GO price index for year $t, P_{G Y}^{t}$, is a chained Törnqvist price index that uses the output prices $P_{C}^{t}, P_{G}^{t}, P_{I}^{t}, P_{X}^{t}$ and the corresponding quantity indexes $Q_{C}^{t}, Q_{G}^{t}, Q_{I}^{t}, Q_{X}^{t}$ as inputs into the price index $P_{G Y}^{t}$. The corresponding year $t$ GO quantity index $Q_{G Y}^{t}$ is defined residually by deflating the year $t$ value of Gross Output, $V_{G Y}^{t}$ by the year $t$ GO price index; i.e., $Q_{G Y}^{t} \equiv V_{G Y}^{t} / P_{G Y}^{t}$. The companion index of the year $t$ quantity of inputs $Q_{G Z}^{t}$ is a chained Törnqvist quantity index that uses the prices of imports, labour and the four types of capital services $P_{M}^{t}, P_{L}^{t}, P_{U M}^{t}, P_{U S}^{t}, P_{U O}^{t}, P_{U L}^{t}$ and the quantity indexes $Q_{M}^{t}, Q_{L}^{t}, Q_{U M}^{t}, Q_{U S}^{t}, Q_{U O}^{t}, Q_{U L}^{t}$ as inputs into the quantity index $Q_{G Z}^{t}$ and the companion input price index $P_{G Z}^{t}$ is defined residually by deflating the year $t$ value of GO, $V_{G Z}^{t}=V_{G Y}^{t}$ by the year $t \mathrm{GO}$ input quantity index; i.e., $P_{G Z}^{t} \equiv V_{G Z}^{t} / Q_{G Z}^{t}$. The year $t$ level of GO Total

[^21]Factor Productivity is defined as the year $t$ output quantity $Q_{G Y}^{t}$ divided by the year $t$ input quantity $Q_{G Z}^{t}$; i.e.,

$$
\begin{equation*}
\mathrm{TFP}_{G}^{t} \equiv Q_{G Y}^{t} / Q_{G Z}^{t} ; \quad t=1970, \ldots, 2022 \tag{A8}
\end{equation*}
$$

Gross Output TFP growth is defined as $\mathrm{TFPG}_{G}^{t} \equiv \mathrm{TFP}_{G}^{t} / \operatorname{TFP}_{G}^{t-1}$ for $t=1971, \ldots, 2022$. The series $Q_{G Y}^{t}, Q_{G Z}^{t}, \mathrm{TFP}_{G}^{t}$ and $\mathrm{TFPG}_{G}^{t}$ are listed in Table A13.

Table A13: Alternative GDP and Gross Output Measures of Input, Output and Productivity

| Year | $Q_{Y}^{t}$ | $Q_{G Y}^{t}$ | $Q_{Z}^{t}$ | $Q_{G Z}^{t}$ | $\mathrm{TFP}^{t}$ | $\mathrm{TFP}_{J}^{t}$ | $\mathrm{TFP}_{G}^{t}$ | TFPG ${ }^{t}$ | $\mathrm{TFPG}_{J}^{t}$ | TFPGG ${ }_{G}^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | 1.0275 | 1.0858 | 1.0275 | 1.0858 | 1.0000 | 1.0000 | 1.0000 | 1.0000 | 1.0000 | 1.0000 |
| 1971 | 1.0635 | 1.1250 | 1.0344 | 1.0959 | 1.0281 | 1.0279 | 1.0266 | 1.0281 | 1.0279 | 1.0266 |
| 1972 | 1.1205 | 1.1888 | 1.0649 | 1.1331 | 1.0522 | 1.0514 | 1.0492 | 1.0234 | 1.0228 | 1.0220 |
| 1973 | 1.1832 | 1.2546 | 1.1090 | 1.1803 | 1.0669 | 1.0662 | 1.0629 | 1.0140 | 1.0141 | 1.0131 |
| 1974 | 1.1751 | 1.2445 | 1.1296 | 1.1987 | 1.0402 | 1.0403 | 1.0382 | 0.9750 | 0.9757 | 0.9767 |
| 1975 | 1.1704 | 1.2287 | 1.1259 | 1.1842 | 1.0395 | 1.0395 | 1.0376 | 0.9993 | 0.9992 | 0.9994 |
| 1976 | 1.2322 | 1.3063 | 1.1541 | 1.2283 | 1.0677 | 1.0675 | 1.0635 | 1.0271 | 1.0270 | 1.0250 |
| 1977 | 1.2914 | 1.3755 | 1.1897 | 1.2739 | 1.0855 | 1.0848 | 1.0798 | 1.0167 | 1.0161 | 1.0153 |
| 1978 | 1.3631 | 1.4555 | 1.2386 | 1.3312 | 1.1005 | 1.0989 | 1.0934 | 1.0138 | 1.0131 | 1.0126 |
| 1979 | 1.4066 | 1.4999 | 1.2835 | 1.3771 | 1.0959 | 1.0929 | 1.0892 | 0.9958 | 0.9945 | 0.9962 |
| 1980 | 1.4039 | 1.4876 | 1.3025 | 1.3864 | 1.0779 | 1.0744 | 1.0730 | 0.9836 | 0.9830 | 0.9851 |
| 1981 | 1.4408 | 1.5266 | 1.3243 | 1.4108 | 1.0879 | 1.0842 | 1.0820 | 1.0094 | 1.0092 | 1.0084 |
| 1982 | 1.4090 | 1.4942 | 1.3223 | 1.4072 | 1.0655 | 1.0614 | 1.0618 | 0.9794 | 0.9790 | 0.9813 |
| 1983 | 1.4688 | 1.5683 | 1.3420 | 1.4414 | 1.0945 | 1.0895 | 1.0881 | 1.0272 | 1.0264 | 1.0247 |
| 1984 | 1.5759 | 1.7058 | 1.3971 | 1.5254 | 1.1280 | 1.1221 | 1.1182 | 1.0306 | 1.0299 | 1.0277 |
| 1985 | 1.6380 | 1.7771 | 1.4412 | 1.5783 | 1.1366 | 1.1309 | 1.1259 | 1.0076 | 1.0078 | 1.0069 |
| 1986 | 1.6862 | 1.8387 | 1.4767 | 1.6261 | 1.1419 | 1.1354 | 1.1307 | 1.0047 | 1.0040 | 1.0043 |
| 1987 | 1.7455 | 1.9077 | 1.5267 | 1.6853 | 1.1433 | 1.1358 | 1.1320 | 1.0012 | 1.0004 | 1.0011 |
| 1988 | 1.8182 | 1.9867 | 1.5723 | 1.7372 | 1.1564 | 1.1480 | 1.1436 | 1.0115 | 1.0107 | 1.0103 |
| 1989 | 1.8824 | 2.0586 | 1.6244 | 1.7967 | 1.1588 | 1.1502 | 1.1458 | 1.0021 | 1.0020 | 1.0019 |
| 1990 | 1.9194 | 2.1024 | 1.6565 | 1.8350 | 1.1587 | 1.1504 | 1.1457 | 0.9999 | 1.0001 | 0.9999 |
| 1991 | 1.9167 | 2.0994 | 1.6556 | 1.8339 | 1.1577 | 1.1495 | 1.1448 | 0.9991 | 0.9993 | 0.9992 |
| 1992 | 1.9805 | 2.1768 | 1.6773 | 1.8679 | 1.1808 | 1.1725 | 1.1653 | 1.0199 | 1.0200 | 1.0179 |
| 1993 | 2.0358 | 2.2500 | 1.7110 | 1.9176 | 1.1898 | 1.1810 | 1.1733 | 1.0076 | 1.0073 | 1.0069 |
| 1994 | 2.1177 | 2.3583 | 1.7587 | 1.9885 | 1.2041 | 1.1952 | 1.1860 | 1.0120 | 1.0120 | 1.0108 |
| 1995 | 2.1735 | 2.4339 | 1.8134 | 2.0607 | 1.1986 | 1.1895 | 1.1811 | 0.9954 | 0.9953 | 0.9959 |
| 1996 | 2.2587 | 2.5421 | 1.8564 | 2.1238 | 1.2167 | 1.2071 | 1.1969 | 1.0151 | 1.0148 | 1.0134 |
| 1997 | 2.3602 | 2.6814 | 1.9209 | 2.2208 | 1.2287 | 1.2184 | 1.2074 | 1.0099 | 1.0093 | 1.0087 |
| 1998 | 2.4628 | 2.8200 | 1.9921 | 2.3229 | 1.2363 | 1.2246 | 1.2140 | 1.0062 | 1.0051 | 1.0055 |
| 1999 | 2.5774 | 2.9737 | 2.0623 | 2.4261 | 1.2498 | 1.2367 | 1.2257 | 1.0109 | 1.0099 | 1.0096 |
| 2000 | 2.6788 | 3.1234 | 2.1327 | 2.5371 | 1.2561 | 1.2422 | 1.2311 | 1.0050 | 1.0044 | 1.0044 |
| 2001 | 2.7015 | 3.1365 | 2.1598 | 2.5571 | 1.2508 | 1.2366 | 1.2266 | 0.9958 | 0.9955 | 0.9963 |
| 2002 | 2.7458 | 3.1957 | 2.1866 | 2.5964 | 1.2557 | 1.2415 | 1.2308 | 1.0039 | 1.0040 | 1.0035 |
| 2003 | 2.8223 | 3.2939 | 2.2059 | 2.6327 | 1.2794 | 1.2651 | 1.2512 | 1.0189 | 1.0190 | 1.0165 |
| 2004 | 2.9307 | 3.4501 | 2.2464 | 2.7111 | 1.3046 | 1.2897 | 1.2726 | 1.0197 | 1.0194 | 1.0171 |
| 2005 | 3.0296 | 3.5811 | 2.2953 | 2.7860 | 1.3199 | 1.3043 | 1.2854 | 1.0117 | 1.0113 | 1.0101 |
| 2006 | 3.1151 | 3.7004 | 2.3546 | 2.8731 | 1.3230 | 1.3065 | 1.2880 | 1.0023 | 1.0017 | 1.0020 |
| 2007 | 3.1753 | 3.7754 | 2.4069 | 2.9383 | 1.3193 | 1.3020 | 1.2849 | 0.9972 | 0.9966 | 0.9976 |
| 2008 | 3.1806 | 3.7685 | 2.4249 | 2.9474 | 1.3117 | 1.2958 | 1.2786 | 0.9942 | 0.9952 | 0.9951 |
| 2009 | 3.0969 | 3.6137 | 2.3742 | 2.8396 | 1.3044 | 1.2898 | 1.2726 | 0.9945 | 0.9953 | 0.9953 |
| 2010 | 3.1806 | 3.7590 | 2.3793 | 2.8917 | 1.3368 | 1.3222 | 1.2999 | 1.0248 | 1.0252 | 1.0214 |
| 2011 | 3.2279 | 3.8330 | 2.4171 | 2.9512 | 1.3354 | 1.3208 | 1.2988 | 0.9990 | 0.9989 | 0.9991 |
| 2012 | 3.3000 | 3.9199 | 2.4598 | 3.0064 | 1.3416 | 1.3269 | 1.3039 | 1.0046 | 1.0046 | 1.0039 |
| 2013 | 3.3717 | 3.9996 | 2.5029 | 3.0568 | 1.3471 | 1.3316 | 1.3085 | 1.0041 | 1.0035 | 1.0035 |
| 2014 | 3.4555 | 4.1146 | 2.5525 | 3.1315 | 1.3538 | 1.3371 | 1.3139 | 1.0049 | 1.0042 | 1.0042 |
| 2015 | 3.5531 | 4.2447 | 2.6091 | 3.2141 | 1.3618 | 1.3450 | 1.3206 | 1.0059 | 1.0059 | 1.0051 |
| 2016 | 3.6126 | 4.3144 | 2.6590 | 3.2736 | 1.3586 | 1.3424 | 1.3180 | 0.9977 | 0.9981 | 0.9980 |
| 2017 | 3.7007 | 4.4328 | 2.7085 | 3.3469 | 1.3664 | 1.3498 | 1.3245 | 1.0057 | 1.0055 | 1.0049 |
| 2018 | 3.8094 | 4.5695 | 2.7632 | 3.4237 | 1.3786 | 1.3617 | 1.3347 | 1.0090 | 1.0088 | 1.0077 |
| 2019 | 3.9033 | 4.6742 | 2.8184 | 3.4883 | 1.3849 | 1.3680 | 1.3400 | 1.0046 | 1.0046 | 1.0040 |
| 2020 | 3.8119 | 4.5241 | 2.7731 | 3.3984 | 1.3746 | 1.3580 | 1.3313 | 0.9926 | 0.9926 | 0.9935 |
| 2021 | 4.0268 | 4.8282 | 2.8413 | 3.5314 | 1.4172 | 1.3995 | 1.3672 | 1.0310 | 1.0305 | 1.0270 |
| 2022 | 4.0939 | 4.9532 | 2.9284 | 3.6659 | 1.3980 | 1.3800 | 1.3512 | 0.9864 | 0.9861 | 0.9883 |
| Mean | 2.3854 | 2.7426 | 1.9055 | 2.2279 | 1.2164 | 1.2065 | 1.1946 | 1.0065 | 1.0063 | 1.0059 |

The average TFP growth rate using GDP as the output concept was $0.65 \%$ per year using smoothed user costs and $0.63 \%$ per year using Jorgensonian user costs. Smoothing the user costs did not greatly affect our measures of GDP Total Factor Productivity Growth. Using Gross Output as our output concept, the mean TFP growth rate fell to $0.59 \%$ per year. This drop was expected because regarding imports as an input rather than a negative output increases the overall input volume measure by about $10 \%$ so we could expect a $10 \%$ drop in measured TFP growth using Gross Output in place as GDP as the output measure.
We conclude this data construction Appendix by listing the year by year main diagonal elasticities that correspond to the average elasticities that were listed in Tables 2 and 1 in the main text. These elasticities were generated by differentiating our estimated Joint Cost Function over the years 1970-2022.

Table A14: Main Diagonal Elasticities of Inverse Output Supply 1970-2022

| Year | $E p_{11}$ | $E p_{22}$ | $E p_{33}$ | $E p_{44}$ |
| :---: | :---: | :---: | :---: | :---: |
| 1970 | 0.1536 | 0.1795 | 0.1001 | 0.0199 |
| 1971 | 0.1513 | 0.1715 | 0.1039 | 0.0192 |
| 1972 | 0.1466 | 0.1621 | 0.1098 | 0.0191 |
| 1973 | 0.1428 | 0.1533 | 0.1176 | 0.0206 |
| 1974 | 0.1454 | 0.1673 | 0.1136 | 0.0204 |
| 1975 | 0.1345 | 0.1881 | 0.1064 | 0.0185 |
| 1976 | 0.1332 | 0.1767 | 0.1135 | 0.0186 |
| 1977 | 0.1357 | 0.1659 | 0.1175 | 0.0184 |
| 1978 | 0.1329 | 0.1552 | 0.1198 | 0.0193 |
| 1979 | 0.1295 | 0.1523 | 0.1196 | 0.0196 |
| 1980 | 0.1237 | 0.1636 | 0.1146 | 0.0199 |
| 1981 | 0.1238 | 0.1547 | 0.1150 | 0.0200 |
| 1982 | 0.1141 | 0.1630 | 0.1081 | 0.0183 |
| 1983 | 0.1126 | 0.1569 | 0.1111 | 0.0178 |
| 1984 | 0.1179 | 0.1405 | 0.1203 | 0.0194 |
| 1985 | 0.1137 | 0.1410 | 0.1210 | 0.0195 |
| 1986 | 0.1091 | 0.1446 | 0.1200 | 0.0200 |
| 1987 | 0.1048 | 0.1421 | 0.1199 | 0.0208 |
| 1988 | 0.0976 | 0.1394 | 0.1207 | 0.0224 |
| 1989 | 0.0935 | 0.1365 | 0.1207 | 0.0236 |
| 1990 | 0.0888 | 0.1365 | 0.1185 | 0.0244 |
| 1991 | 0.0827 | 0.1386 | 0.1146 | 0.0254 |
| 1992 | 0.0795 | 0.1366 | 0.1168 | 0.0262 |
| 1993 | 0.0774 | 0.1323 | 0.1197 | 0.0263 |
| 1994 | 0.0757 | 0.1275 | 0.1239 | 0.0275 |
| 1995 | 0.0716 | 0.1264 | 0.1255 | 0.0285 |
| 1996 | 0.0695 | 0.1200 | 0.1297 | 0.0304 |
| 1997 | 0.0674 | 0.1138 | 0.1340 | 0.0335 |
| 1998 | 0.0663 | 0.1087 | 0.1364 | 0.0346 |
| 1999 | 0.0649 | 0.1041 | 0.1385 | 0.0354 |
| 2000 | 0.0618 | 0.1004 | 0.1398 | 0.0373 |
| 2001 | 0.0596 | 0.1026 | 0.1359 | 0.0353 |
| 2002 | 0.0587 | 0.1029 | 0.1340 | 0.0338 |
| 2003 | 0.0575 | 0.1001 | 0.1344 | 0.0339 |
| 2004 | 0.0561 | 0.0961 | 0.1363 | 0.0354 |
| 2005 | 0.0543 | 0.0925 | 0.1373 | 0.0358 |
| 2006 | 0.0515 | 0.0896 | 0.1372 | 0.0375 |
| 2007 | 0.0475 | 0.0890 | 0.1347 | 0.0378 |
| 2008 | 0.0440 | 0.0901 | 0.1296 | 0.0364 |
| 2009 | 0.0409 | 0.0946 | 0.1215 | 0.0335 |
| 2010 | 0.0392 | 0.0897 | 0.1252 | 0.0349 |
| 2011 | 0.0364 | 0.0854 | 0.1251 | 0.0351 |
| 2012 | 0.0351 | 0.0819 | 0.1267 | 0.0362 |
| 2013 | 0.0334 | 0.0785 | 0.1276 | 0.0368 |
| 2014 | 0.0315 | 0.0751 | 0.1277 | 0.0379 |
| 2015 | 0.0304 | 0.0735 | 0.1279 | 0.0392 |
| 2016 | 0.0286 | 0.0728 | 0.1255 | 0.0389 |
| 2017 | 0.0271 | 0.0707 | 0.1246 | 0.0392 |
| 2018 | 0.0258 | 0.0690 | 0.1232 | 0.0400 |
| 2019 | 0.0250 | 0.0695 | 0.1215 | 0.0399 |
| 2020 | 0.0246 | 0.0737 | 0.1170 | 0.0385 |
| 2021 | 0.0223 | 0.0701 | 0.1157 | 0.0375 |
| 2022 | 0.0204 | 0.0675 | 0.1149 | 0.0383 |
|  |  |  |  |  |

The above inverse elasticities are all small.*43

[^22]Table A15: Main Diagonal Elasticities of Input Demand 1970-2022

|  | $E x_{11}$ | $E x_{22}$ | $E x_{33}$ |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 | -0.51 | -0.2418 | -0. | -0. | -0.4416 | -0 |
|  | -0.5010 | -0.2464 | -0. | -0. | -0 | -0.0129 |
|  | -0.4858 | -0.2501 | -0 | -0 | -0 | 7 |
|  | -0.5021 | -0. | -0.5 | -0.0487 | -0.5352 | -0.0134 |
|  | -0. | -0 | -0 | -0 | -0.4758 | -0.0114 |
|  | -0.7562 | -0.2737 | -0. | -0. | -0.409 |  |
|  | -0.6791 | -0.2795 | -0. | -0.0509 |  |  |
|  | -0.6695 | -0.2848 | -0. | -0 | -0. |  |
| 1978 | -0. | -0.2 | -0. | -0. | -0.4269 | -0 |
| 1979 | -0 | -0 | -0 | -0 | -0 | -0.0206 |
| 1980 | -0 | -0 | -0 | -0 | -0 | -0.0210 |
| 1981 | -0 | -0 | -0 | -0.0510 | -0.3278 | -0.0244 |
|  | -0. | -0 | -0. | -0.0489 | -0.3067 |  |
|  | -0.6 | -0 | -0. | -0 | -0 | 3 |
|  | -0.5651 | -0.2920 | 0.91 | 0.05 | -0.397 | 253 |
|  | -0.5287 | -0.2942 | -0.952 | -0.059 | -0.420 | 5 |
|  | -0.5140 | -0.29 | -0 | -0 | -0. | 8 |
|  | -0 | -0 | -1 | -0.0634 | -0.3843 | -0.0216 |
|  | -0 | -0 | -0 | -0 | -0.3870 | - |
| 1989 | -0.5 | -0 | -0. | -0 | -0 | - |
|  | -0.5290 | -0.2869 | 0.9 | -0.0621 | -0.3 | 2 |
|  | -0.5225 | -0.2833 | -0.99 | -0.0595 | -0. | -0.0382 |
|  | -0.5109 | -0.2 | -1 | 0.061 | -0.3 | 88 |
|  | -0 | -0.2889 | -1 | 0.06 | -0.3 | 391 |
|  | -0 | -0.2950 | -1 | 0. | -0.3 | 51 |
|  | -0. | -0.3015 | -1 | -0 | -0.35 | 1 |
|  | -0. | -0 |  | -0 | -0.3835 |  |
| 1997 | -0. | -0 |  |  |  |  |
|  | -0.3 | -0.3 | -1 |  | -0 | -0 |
| 9 | -0.3209 | -0.3022 | -1 | -0. | -0 | -0 |
| 000 | -0.3182 | -0.2982 | 0.9 | -0. | -0. | -0 |
|  | -0.3133 | -0.2917 | -0.9349 | 0. | -0.3950 | 8 |
|  | -0 | -0 | -0.9637 | -0 | 369 | 93 |
|  | -0 | -0 | -0 | -0 | 62 | 4 |
|  | -0.2939 | -0.2888 | -0.8 | -0.1163 | -0.332 | . 0426 |
| 2005 | -0.2915 | -0.2878 | -0.78 | -0. | -0.303 | -0.0619 |
| 2006 | -0.2850 | -0.2822 |  |  |  |  |
|  | -0.2 | -0 | -0 |  | -0. |  |
|  | -0 | -0.2783 | -0 | -0 | -0 |  |
| 9 | -0 | -0.2702 | -0.56 | -0.1016 | -0.2 |  |
|  | -0 | -0.2819 | -0.592 | -0 | -0. |  |
|  | -0.3101 | -0.2876 | -0.606 | -0. | -0.236 | 0 |
|  | -0.3045 | -0.2931 | -0.62 | -0 | . 246 | -0.0607 |
|  | -0. | -0.2986 | -0.654 | -0.167 | -0.252 | -0.0508 |
|  | -0.29 | -0.2987 | -0.6418 | -0.18 | -0.2686 | -0.0399 |
|  | -0.2 | -0.2962 | -0.62 |  | -0. | -0.0393 |
|  | -0.26 | -0.2941 | -0.6 | -0. | -0.2585 | -0.0433 |
|  | -0. | -0 | -0. | -0 | -0. | -0.0471 |
| 2018 | -0.2654 | -0.2932 | -0.5750 | -0.2012 | -0.2452 | -0.0522 |
| 2019 | -0.2616 | -0.2927 | -0.5703 | -0.2032 | -0.2284 | -0.0590 |
| - | -0.2667 | -0.2772 | -0.5382 | -0.2036 | -0.206 | -0.0538 |
| 2021 | -0.2551 | -0.2916 | -0.5523 | -0.2159 | -0.2057 | -0.0583 |
| 2022 | -0.2435 | -0.3061 | -0.5588 | -0.212 | -0.212 | -0. |

The trends in the above own elasticities of input demand generated by our estimated joint cost function are quite interesting. The magnitudes of the Import, Machinery and Equipment and Other Capital Services (inputs 1, 3 and 5) own elasticities of demand trended downward over the sample period, indicating that the magnitude of substitution possibilities for these inputs declined over time. ${ }^{* 44}$ On the other hand, the magnitudes of the Structures and Land Services (inputs 4 and 6 ) own elasticities of demand trended upward over the sample period but the size of these elasticities was small over the entire sample period. During land price bubbles, the own elasticity of demand for land tended to increase in magnitude. Finally, the own elasticity of demand for labour remained relatively constant over the entire sample period.

[^23]
## Appendix B: Proof of the Flexibility of the Normalized Quadratic Joint Cost Function

A flexible functional form for a joint cost function $C(\boldsymbol{y}, \boldsymbol{w})$ has enough free parameters so that it can provide a second order Taylor series approximation to an arbitrary twice continuously differentiable joint cost function $C^{*}(\boldsymbol{y}, \boldsymbol{w})$ at an arbitrary point $\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$. We assume that both $C(\boldsymbol{y}, \boldsymbol{w})$ and $C^{*}(\boldsymbol{y}, \boldsymbol{w})$ are linearly homogeneous in the components of $\boldsymbol{y}$ holding $\boldsymbol{w}$ constant*45 and are linearly homogeneous in the components of $\boldsymbol{w}$ holding $\boldsymbol{y}$ constant. ${ }^{* 46}$ It turn out that the linear homogeneity assumptions and the assumption of twice continuous differentiability imply that $C(\boldsymbol{y}, \boldsymbol{w})$ will be a flexible functional form if it has enough parameters to satisfy the following conditions: ${ }^{* 47}$

$$
\begin{array}{lr}
\nabla_{y w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\nabla_{y w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) ; & M N \text { restrictions; } \\
\nabla_{y y}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\nabla_{y y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) ; & M(M-1) / 2 \text { independent restrictions; } \\
\nabla_{w w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\nabla_{w w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) ; & N(N-1) / 2 \text { independent restrictions. } \tag{B3}
\end{array}
$$

There are $M^{2}$ restrictions in the matrix equation (B2) but Young's Theorem in calculus implies that the upper triangle of matrix elements in the matrix of second order partial derivatives of $C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ is equal to the lower triangle; i.e., $\left[\nabla_{y y}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)\right]^{T}=\left[\nabla_{y y}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)\right]$ and similarly, $\left[\nabla_{y y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)\right]^{T}=\left[\nabla_{y y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)\right]$. Thus there are only $M(M+1) / 2$ independent restrictions on the second order partial derivatives of $C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ in the matrix equation (B2). But due to the linear homogeneity of $C(\boldsymbol{y}, \boldsymbol{w})$ in the components of $\boldsymbol{y}$, Euler's Theorem on homogeneous functions implies the following $M$ restrictions on the second order partial derivatives of $C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ and $C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ :

$$
\begin{equation*}
\nabla_{y y}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) \boldsymbol{y}^{*}=\mathbf{0}_{M} ; \nabla_{y y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) \boldsymbol{y}^{*}=\mathbf{0}_{M} . \tag{B4}
\end{equation*}
$$

Since the $M$ by $M$ matrices $\nabla_{y y}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ and $\nabla_{y y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ are symmetric, equality of the upper diagonal elements in equations (B2) plus the $2 M$ equations in (B4) will imply equality of all $M^{2}$ elements in the matrix equation $\nabla_{y y}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\nabla_{y y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$.
Similarly, due to the linear homogeneity of $C(\boldsymbol{y}, \boldsymbol{w})$ in the components of $\boldsymbol{w}$, Euler's Theorem on homogeneous functions implies the following $N$ restrictions on the second order partial

[^24]derivatives of $C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ and $C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ :
\[

$$
\begin{equation*}
\nabla_{w w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) \boldsymbol{w}^{*}=\mathbf{0}_{N} ; \nabla_{w w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) \boldsymbol{w}^{*}=\mathbf{0}_{N} \tag{B5}
\end{equation*}
$$

\]

Since the $N$ by $N$ matrices $\nabla_{w w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ and $\nabla_{w w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ are symmetric, equality of the upper diagonal elements in equations (B3) plus the $N$ equations in (B5) will imply equality of all $N^{2}$ elements in the matrix equation $\nabla_{w w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\nabla_{w w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$.
Let $\boldsymbol{y}^{*} \equiv\left[y_{1}^{*}, \ldots, y_{M}^{*}\right] \gg \mathbf{0}_{M}$ be a positive reference output vector and let $\boldsymbol{w}^{*} \equiv$ $\left[w_{1}^{*}, \ldots, w_{N}^{*}\right] \gg \mathbf{0}_{N}$ be a positive vector of reference input prices. Let $\boldsymbol{\alpha} \equiv\left[\alpha_{1}, \ldots, \alpha_{N}\right] \gg \mathbf{0}_{N}$ and $\boldsymbol{\beta} \equiv\left[\beta_{1}, \ldots, \beta_{M}\right] \gg \mathbf{0}_{M}$ be positive vector of predetermined constants and that satisfy the following linear restrictions:

$$
\begin{equation*}
\boldsymbol{\alpha}^{T} \boldsymbol{w}^{*}=1 ; \quad \boldsymbol{\beta}^{T} \boldsymbol{y}^{*}=1 \tag{B6}
\end{equation*}
$$

The basic Normalized Quadratic Joint Cost Function, $C(\boldsymbol{y}, \boldsymbol{w})$ is defined as follows:*48

$$
\begin{equation*}
C(\boldsymbol{y}, \boldsymbol{w}) \equiv(1 / 2)\left(\boldsymbol{w}^{T} \mathbf{A} \boldsymbol{w}\right)\left(\boldsymbol{\alpha}^{T} \boldsymbol{w}\right)^{-1}\left(\boldsymbol{\beta}^{T} \boldsymbol{y}\right)+(1 / 2)\left(\boldsymbol{y}^{T} \mathbf{B} \boldsymbol{y}\right)\left(\boldsymbol{\alpha}^{T} \boldsymbol{w}\right)\left(\boldsymbol{\beta}^{T} \boldsymbol{y}\right)^{-1}+\boldsymbol{w}^{T} \mathbf{D} \boldsymbol{y} \tag{B7}
\end{equation*}
$$

The $M$ by $N$ matrix $\mathbf{D}$ is unrestricted. Assume that the matrix $\mathbf{A}$ has the following properties:
A is a negative semidefinite $N$ by $N$ matrix;
$\mathbf{A}$ is symmetric so that $\mathbf{A}=\mathbf{A}^{T}$;
$\mathbf{A} \boldsymbol{w}^{*}=\mathbf{0}_{N}$.
Assume that the matrix $\mathbf{B}$ has the following properties:
$\mathbf{B}$ is a positive semidefinite $M$ by $M$ matrix;
$\mathbf{B}$ is symmetric so that $\mathbf{B}=\mathbf{B}^{T}$;
$\mathbf{B} \boldsymbol{y}^{*}=\mathbf{0}_{M}$.
Now compute the first and second order partial derivatives of $C(\boldsymbol{y}, \boldsymbol{w})$ defined by (B7) and evaluate them at the point $\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$. Using the restrictions (B6), (B10) and (B13), we obtain the following first and second order partial derivatives:

$$
\begin{align*}
& \nabla_{y} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\mathbf{D}^{T} \boldsymbol{w}^{*}  \tag{B14}\\
& \nabla_{w} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\mathbf{D} \boldsymbol{y}^{*}  \tag{B15}\\
& \nabla_{y y}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\mathbf{B}  \tag{B16}\\
& \nabla_{w w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\mathbf{A}  \tag{B17}\\
& \nabla_{y w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\mathbf{D}^{T} \tag{B18}
\end{align*}
$$

To prove the flexibility of the Normalized Quadratic Joint Cost Function defined by (B7) with the restrictions (B6)-(B13), we need to find matrices $\mathbf{A}, \mathbf{B}$ and $\mathbf{D}$ that lead to the satisfaction of equations (B1)-(B3). Using equations (B16)-(B18), this is very simple: define $\mathbf{A}, \mathbf{B}$ and $\mathbf{D}$ as follows:

$$
\begin{align*}
& \mathbf{A} \equiv \nabla_{w w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)  \tag{B19}\\
& \mathbf{B} \equiv \nabla_{y y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)  \tag{B20}\\
& \mathbf{D} \equiv\left[\nabla_{y w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)\right]^{T} . \tag{B21}
\end{align*}
$$

[^25]Under our regularity conditions on the production possibilities set $S$, it can be shown that $\nabla_{w w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ is a symmetric negative semidefinite matrix which satisfies $\nabla_{w w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) \boldsymbol{w}^{*}$ $=\mathbf{0}_{N}$ and hence, the matrix $\mathbf{A}$ will satisfy the restrictions (B8)-(B10). It also can be shown ${ }^{* 49}$ that $\nabla_{y y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ is a symmetric positive semidefinite matrix which satisfies $\nabla_{y y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) \boldsymbol{y}^{*}=\mathbf{0}_{M}$ and hence, $\mathbf{B}$ will satisfy the restrictions (B11)-(B13). This establishes the flexibility of the basic Normalized Quadratic Joint Cost function.

[^26]
## Appendix C: Nonparametric Estimates for Technical Progress

In this section, we will calculate nonparametric estimates for gross output Total Factor Productivity Growth and compare these estimates with our Appendix A index number estimates TFP growth and our main text estimates of technical progress. The analysis in this section is based on Diewert and Fox (2018)[19].*50 Their nonparametric methodology has the advantage of giving a decomposition of TFP growth into technical progress and inefficiency components. There are two key concepts that their analysis is based on:

- An approximation to the aggregate production possibilities set for an economy can be formed by using linear multiples of past output and input vectors and
- The Cost Constrained Gross Output Function can be used to form measures of efficiency, output price change, input price change, input quantity change and technology change.
Denote the year $t$ observed output and input quantity vectors by $\boldsymbol{y}^{t}$ and $\boldsymbol{x}^{t}$ and the corresponding price vectors by $\boldsymbol{p}^{t}$ and $\boldsymbol{w}^{t}$. Following Diewert and Fox (2018)[19], we assume that the year $t$ national technology set can be approximated by assuming it consists of past observed output and input vectors, $\left(\boldsymbol{y}^{s}, \boldsymbol{x}^{s}\right)$, and linear multiples of these vectors for past years and the current year $t$. Let $S^{t}$ denote the resulting year $t$ production possibilities set. Thus $S^{1} \equiv\left\{(\boldsymbol{y}, \boldsymbol{x}): \boldsymbol{y}=\lambda \boldsymbol{y}^{1}, \boldsymbol{x}=\lambda \boldsymbol{x}^{1} ; \lambda \geq 0\right\}, S^{2} \equiv\left\{(\boldsymbol{y}, \boldsymbol{x}): \boldsymbol{y}=\lambda_{1} \boldsymbol{y}^{1}, \boldsymbol{x}=\lambda_{1} \boldsymbol{x}^{1} ; \lambda_{1} \geq 0, \boldsymbol{y}=\right.$ $\left.\lambda_{2} \boldsymbol{y}^{2}, \boldsymbol{x}=\lambda_{2} \boldsymbol{x}^{2} ; \lambda_{2} \geq 0\right\}, \ldots, S^{t} \equiv\left\{(\boldsymbol{y}, \boldsymbol{x}): \boldsymbol{y}=\lambda_{s} \boldsymbol{y}^{s}, \boldsymbol{x}=\lambda_{s} \boldsymbol{x}^{s} ; \lambda_{s} \geq 0, s=1,2, \ldots, t\right\}$. These definitions for the $S^{t}$ mean that we are assuming that $S^{t}$ is a constant return to scale technology set for each period.
The year $t$ Cost Constrained Gross Output Function for the US economy, $G^{t}(\boldsymbol{p}, \boldsymbol{w}, \boldsymbol{x})$, is defined as follows for the positive price vectors $\boldsymbol{p}$ and $\boldsymbol{w}$, positive input vector $\boldsymbol{x}$ and using the period $t$ production possibilities set $S^{t}: * 51$

$$
\begin{align*}
G^{t}(\boldsymbol{p}, \boldsymbol{w}, \boldsymbol{x}) & \equiv \max _{y, z}\left\{\boldsymbol{p} \cdot \boldsymbol{y}:(\boldsymbol{y}, \boldsymbol{z}) \in S^{t} ; \boldsymbol{w} \cdot \boldsymbol{z} \leq \boldsymbol{w} \cdot \boldsymbol{x}\right\} ; \quad t=1970, \ldots, 2022 . \\
& =\max _{s}\left\{\boldsymbol{p} \cdot \boldsymbol{y}^{s} \boldsymbol{w} \cdot \boldsymbol{x} / \boldsymbol{w} \cdot \boldsymbol{x}^{s}: s=1,2, \ldots, t\right\} ; \\
& =\boldsymbol{w} \cdot \boldsymbol{x} \max _{s}\left\{\boldsymbol{p} \cdot \boldsymbol{y}^{s} / \boldsymbol{w} \cdot \boldsymbol{x}^{s}: s=1,2, \ldots, t\right\} . \tag{C1}
\end{align*}
$$

Given the period $t$ technology set $S^{t}$, output prices $\boldsymbol{p}$, input prices $\boldsymbol{w}$ and the constraint that primary input costs should not exceed cost $\boldsymbol{w} \cdot \boldsymbol{x}$, we assume that producers choose the output vector $\boldsymbol{y}$ and input vector $\boldsymbol{z}$ to maximize national gross output, $\boldsymbol{p} \cdot \boldsymbol{y}$, subject to total input cost $\boldsymbol{w} \cdot \boldsymbol{z}$ to be equal to or less than the given input cost $\boldsymbol{w} \cdot \boldsymbol{x}$.
Due to our assumptions on the year $t$ national production possibilities set $S^{t}$, the year $t$ cost constrained value added function $R^{t}(\boldsymbol{p}, \boldsymbol{w}, \boldsymbol{x})$ can be calculated for any hypothetical $\boldsymbol{p}, \boldsymbol{w}$ and $\boldsymbol{x}$ by solving the very simple maximization problem, $\max _{s}\left\{\boldsymbol{p} \cdot \boldsymbol{y}^{s} / \boldsymbol{w} \cdot \boldsymbol{x}^{s}: s=1,2, \ldots, t\right\}$., which involves taking the maximum of $t$ numbers.
The Cost Constrained Gross Output Function defined by equation (C1) can be used to decompose gross output growth from year $t-1$ to year $t, \boldsymbol{p}^{t} \cdot \boldsymbol{y}^{t} / \boldsymbol{p}^{t-1} \cdot \boldsymbol{y}^{t-1}$, into various explanatory growth factors. The explanatory factors are as follows:

- efficiency changes,
- changes in real output prices,
- changes in primary inputs,

[^27]- changes in real input prices, and
- technical progress.

We now define the above explanatory factors using the observed data and the function $G^{t}(\boldsymbol{p}, \boldsymbol{w}, \boldsymbol{x})$ defined by equation (C1). Following the example of Balk (1998; 143)[2], we define the gross output efficiency of the production unit for year $t, e^{t}$, as follows:

$$
\begin{equation*}
e^{t} \equiv \boldsymbol{p}^{t} \cdot \boldsymbol{y}^{t} / G^{t}\left(\boldsymbol{p}^{t}, \boldsymbol{w}^{t}, \boldsymbol{x}^{t}\right) \leq 1 ; \quad t=1970, \ldots, 2022 \tag{C2}
\end{equation*}
$$

where the inequality in equation (C2) follows using definition (C1). ${ }^{* 52}$ Thus if $e^{t}=1$, then production is allocatively efficient in year $t$ and if $e^{t}<1$, then production for the sector during year $t$ is allocatively inefficient. Note that the above definition of value added efficiency is a gross output counterpart to Farrell's (1957; 255)[26] cost based measure of overall efficiency. Define an index of the change in gross output efficiency $\varepsilon^{t}$ for the production sector over the years $t-1$ and $t$ as follows:

$$
\begin{array}{r}
\varepsilon^{t} \equiv e^{t} / e^{t-1}=\left[\boldsymbol{p}^{t} \cdot \boldsymbol{y}^{t} / G^{t}\left(\boldsymbol{p}^{t}, \boldsymbol{w}^{t}, \boldsymbol{x}^{t}\right)\right] /\left[\boldsymbol{p}^{t-1} \cdot \boldsymbol{y}^{t-1} / G^{t-1}\left(\boldsymbol{p}^{t-1}, \boldsymbol{w}^{t-1}, \boldsymbol{x}^{t-1}\right)\right] \\
t=1971, \ldots, 2022 \tag{C3}
\end{array}
$$

Thus if $\varepsilon^{t}>1$, then gross output efficiency has improved going from year $t-1$ to $t$ whereas it has fallen if $\varepsilon^{t}<1$.
We turn our attention to defining nonparametric measures of output price change going from year $t-1$ to $t$. Following the example of Konüs (1939)[46] in his analysis of the true cost of living index, it is natural to single out two special cases of a family of real output price indexes: one choice is $\alpha_{L}^{t}$ where we use the year $t-1$ technology and set the reference input prices and quantities equal to the year $t-1$ input prices $\boldsymbol{w}^{t-1}$ and primary input quantities $\boldsymbol{x}^{t-1}$ (which gives rise to a Laspeyres type output price index) and another choice is $\alpha_{P}^{t}$ where we use the year $t$ technology and set the reference input prices and quantities equal to the year $t$ input prices and quantities $\boldsymbol{w}^{t}$ and $\boldsymbol{x}^{t}$ (which gives rise to a Paasche type real output price index). We then define an overall measure of real output price change $\alpha^{t}$ by taking the geometric mean of these two indexes. These indexes are defined as follows:

$$
\begin{align*}
\alpha_{L}^{t} & \equiv G^{t-1}\left(\boldsymbol{p}^{t}, \boldsymbol{w}^{t-1}, \boldsymbol{x}^{t-1}\right) / G^{t-1}\left(\boldsymbol{p}^{t-1}, \boldsymbol{w}^{t-1}, \boldsymbol{x}^{t-1}\right) ; \quad t=1971, \ldots, 2022 ;  \tag{C4}\\
\alpha_{P}^{t} & \equiv G^{t}\left(\boldsymbol{p}^{t}, \boldsymbol{w}^{t}, \boldsymbol{x}^{t}\right) / G^{t}\left(\boldsymbol{p}^{t-1}, \boldsymbol{w}^{t}, \boldsymbol{x}^{t}\right) ; \quad t=1971, \ldots, 2022  \tag{C5}\\
\alpha^{t} & \equiv\left[\alpha_{L}^{t} \alpha_{P}^{t}\right]^{1 / 2} ; \quad t=1971, \ldots, 2022 \tag{C6}
\end{align*}
$$

Two natural measures of input quantity change are the Laspeyres and Paasche input quantity indexes. Denote the year $t$ indexes as $\beta_{L}^{t}$ and $\beta_{P}^{t}$. Again, it is natural to take the geometric average of these two indexes which gives rise to the Fisher ideal input quantity index, $\beta^{t}$. These indexes are defined as follows:

$$
\begin{align*}
\beta_{L}^{t} & \equiv \boldsymbol{w}^{t-1} \cdot \boldsymbol{x}^{t} / \boldsymbol{w}^{t-1} \cdot \boldsymbol{x}^{t-1} ; \quad t=1971, \ldots, 2022  \tag{C7}\\
\beta_{P}^{t} & \equiv \boldsymbol{w}^{t} \cdot \boldsymbol{x}^{t} / \boldsymbol{w}^{t} \cdot \boldsymbol{x}^{t-1} \quad t=1971, \ldots, 2022  \tag{C8}\\
\beta^{t} & \equiv\left[\beta_{L}^{t} \beta_{P}^{t}\right]^{1 / 2} ; \quad t=1971, \ldots, 2022 \tag{C9}
\end{align*}
$$

We now consider indexes which measure the effects on cost constrained gross output of a change in real input prices going from year $t-1$ to $t$. Thus we consider measures of

[^28]the change in cost constrained gross output of the form $G^{s}\left(\boldsymbol{p}, \boldsymbol{w}^{t}, \boldsymbol{x}\right) / G^{s}\left(\boldsymbol{p}, \boldsymbol{w}^{t-1}, \boldsymbol{x}\right)$. Since $G^{s}(\boldsymbol{p}, \boldsymbol{w}, \boldsymbol{x})$ is homogeneous of degree 0 in the components of $\boldsymbol{w}$, it can be seen that we cannot interpret $G^{s}\left(\boldsymbol{p}, \boldsymbol{w}^{t}, \boldsymbol{x}\right) / G^{s}\left(\boldsymbol{p}, \boldsymbol{w}^{t-1}, \boldsymbol{x}\right)$ as an input price index. If there is only one primary input, $G^{s}\left(\boldsymbol{p}, \boldsymbol{w}^{t}, \boldsymbol{x}\right) / G^{s}\left(\boldsymbol{p}, \boldsymbol{w}^{t-1}, \boldsymbol{x}\right)$ is equal to $G^{s}(\boldsymbol{p}, 1, \boldsymbol{x}) / G^{s}(\boldsymbol{p}, 1, \boldsymbol{x})=1$ and this measure of input price change will be independent of changes in the price of the single input. In the case where the number of primary inputs is greater than 1 , it is best to interpret $G^{s}\left(\boldsymbol{p}, \boldsymbol{w}^{t}, \boldsymbol{x}\right) / G^{s}\left(\boldsymbol{p}, \boldsymbol{w}^{t-1}, \boldsymbol{x}\right)$ as measuring the effects on cost constrained gross output of a change in the relative proportions of primary inputs used in production or in the mix of inputs used in production that is induced by a change in relative input prices when there is more than one primary input. We consider two special cases of this family of input mix indexes, Case 1 and Case 2. The first case index, $\gamma_{1}^{t}$, will use the year $t$ Cost Constrained Gross Output function and the year $t-1$ reference vectors $\boldsymbol{p}^{t-1}$ and $\boldsymbol{x}^{t-1}$ while the second case index, $\gamma_{2}^{t}$, will use the year $t-1$ Cost Constrained Gross Output function and the year $t$ reference vectors $\boldsymbol{p}^{t}$ and $\boldsymbol{x}^{t}$. We take the geometric mean of these two indexes $\gamma^{t}$ to provide a measure of the overall effects of a change in input prices. ${ }^{* 53}$
\[

$$
\begin{align*}
\gamma_{1}^{t} & \equiv G^{t}\left(\boldsymbol{p}^{t-1}, \boldsymbol{w}^{t}, \boldsymbol{x}^{t}\right) / G^{t}\left(\boldsymbol{p}^{t-1}, \boldsymbol{w}^{t-1}, \boldsymbol{x}^{t}\right) ; \quad t=1971, \ldots, 2022 ;  \tag{C10}\\
\gamma_{2}^{t} & \equiv G^{t-1}\left(\boldsymbol{p}^{t}, \boldsymbol{w}^{t}, \boldsymbol{x}^{t-1}\right) / G^{t-1}\left(\boldsymbol{p}^{t}, \boldsymbol{w}^{t-1}, \boldsymbol{x}^{t-1}\right) ; \quad t=1971, \ldots, 2022 ;  \tag{C11}\\
\gamma^{t} & \equiv\left[\gamma_{1}^{t} \gamma_{2}^{t}\right]^{1 / 2} ; \quad t=1971, \ldots, 2022 . \tag{C12}
\end{align*}
$$
\]

Finally, we use the Cost Constrained Gross Output function in order to define measures of technical progress going from year $t-1$ to $t$. These measures hold $\boldsymbol{p}, \boldsymbol{w}$ and $\boldsymbol{x}$ constant and only change the technology from the year $t-1$ technology to the year $t$ technology.*54 Thus, these measures are of the form $G^{t}(\boldsymbol{p}, \boldsymbol{w}, \boldsymbol{x}) / G^{t-1}(\boldsymbol{p}, \boldsymbol{w}, \boldsymbol{x})$. If there is positive technical progress going from year $t-1$ to $t$, then the production possibilities set $S^{t}$ will be larger than the period $t-1$ set, $S^{t-1}$, and thus $G^{t}(\boldsymbol{p}, \boldsymbol{w}, \boldsymbol{x})$ will be equal to or greater than $G^{t-1}(\boldsymbol{p}, \boldsymbol{w}, \boldsymbol{x})$ and our measures of technical progress will be equal to or greater than 1 . Our measures of technical progress cannot fall below 1.
We consider two measures of technical progress, a Laspeyres measure $\tau_{L}^{t}$ and a Paasche measure $\tau_{P}^{t}$. However, the Laspeyres case $\tau_{L}^{t}$ will use the year $t$ input vector $\boldsymbol{x}^{t}$ as the reference input vector and the year $t-1$ reference output price and input price vectors $\boldsymbol{p}^{t-1}$ and $\boldsymbol{w}^{t-1}$ while the Paasche case $\tau_{P}^{t}$ will use the year $t-1$ input vector $\boldsymbol{x}^{t-1}$ as the reference input and the year $t$ reference output and input price vectors $\boldsymbol{p}^{t}$ and $\boldsymbol{w}^{t} .{ }^{* 55}$ As usual, we take our overall year $t$ measure of technical change $\tau^{t}$ to be the geometric mean of the Laspeyres and Paasche measures of technical change.

$$
\begin{align*}
\tau_{L}^{t} & \equiv G^{t}\left(\boldsymbol{p}^{t-1}, \boldsymbol{w}^{t-1}, \boldsymbol{x}^{t}\right) / G^{t-1}\left(\boldsymbol{p}^{t-1}, \boldsymbol{w}^{t-1}, \boldsymbol{x}^{t}\right) ; \quad t=1971, \ldots, 2022 ;  \tag{C13}\\
\tau_{P}^{t} & \equiv G^{t}\left(\boldsymbol{p}^{t}, \boldsymbol{w}^{t}, \boldsymbol{x}^{t-1}\right) / G^{t-1}\left(\boldsymbol{p}^{t}, \boldsymbol{w}^{t}, \boldsymbol{x}^{t-1}\right) ; \quad t=1971, \ldots, 2022 ;  \tag{C14}\\
\tau^{t} & \equiv\left[\tau_{L}^{t} \tau_{P}^{t}\right]^{1 / 2} ; \quad t=1971, \ldots, 2022 . \tag{C15}
\end{align*}
$$

[^29]Using the above definitions, it can be shown that the following exact decomposition of year $t$ nominal gross output growth (relative to the 1970 level of gross output) into explanatory growth factors holds:

$$
\begin{equation*}
\mathrm{GO}_{G}^{t} \equiv \boldsymbol{p}^{t} \cdot \boldsymbol{y}^{t} / \boldsymbol{p}^{t-1} \cdot \boldsymbol{y}^{t-1}=\varepsilon^{t} \alpha^{t} \beta^{t} \gamma^{t} \tau^{t} ; \quad t=1971, \ldots, 2022 \tag{C16}
\end{equation*}
$$

Table C1 lists $\mathrm{GO}_{G}^{t}$ and the growth components on the right hand side of equation (C16).

Table C1: A Nonparametric Decomposition of Gross Output TFP Growth

| Year | $\mathrm{GO}_{G}^{t}$ | $\varepsilon^{t}$ | $\alpha^{t}$ | $\beta^{t}$ | $\gamma$ | $\tau^{t}$ | $\mathrm{NTFP}_{G}^{t}$ | $\mathrm{TFP}_{G}^{t}$ | $\kappa^{t}$ | $e^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 71 | 1.0862 | 1.0000 | 1.0483 | 1.0093 | 1.0000 | 1.0266 | 1.0266 | 1.0266 | -0.0097 | 1.0000 |
| 1972 | 1.1048 | 1.0000 | 1.0455 | 1.0340 | 1.0000 | 1.0220 | 1.0220 | 1.0220 | -0.0098 | 1.0000 |
| 1973 | 1.1201 | 1.0000 | 1.0614 | 1.0417 | 1.0000 | 1.0131 | 1.0131 | 1.0131 | -0.0097 | 1.0000 |
| 1974 | 1.1028 | 0.9775 | 1.1115 | 1.0156 | 0.9995 | 1.0000 | 0.9770 | 0.9767 | -0.0094 | 0.9775 |
| 1975 | 1.0825 | 0.9988 | 1.0971 | 0.9879 | 1.0001 | 1.0000 | 0.9989 | 0.9994 | -0.0079 | 0.9763 |
| 1976 | 1.1209 | 1.0241 | 1.0543 | 1.0373 | 1.0006 | 1.0001 | 1.0249 | 1.0250 | -0.0080 | 0.9999 |
| 1977 | 1.12 | 1.0001 | 1.0639 | 1.0371 | 1.0000 | 1.0153 | 1.0153 | 1.0153 | -0.0082 | 0 |
| 1978 | 1.1 | . 0000 | 17 | 1.0450 | 0 | 1. | . 0126 | 26 | 34 | 0 |
| 1979 | 1. | 0.99 | 1.0911 | 1.0345 | . 9998 | 1.0000 | 0.9962 | 0.9962 | 20 | 64 |
| 1980 | 1.0922 | 0.9866 | 1.1013 | 1.0067 | 0.9986 | 1.0000 | 0.9852 | 0.9851 | -0.0007 | 0.9830 |
| 1981 | 1.1151 | 1.0085 | 1.0868 | 1.0176 | 0.9997 | 1.0000 | 1.0082 | 1.0084 | -0.0009 | 0.9914 |
| 1982 | 1.0380 | 0.9815 | 1.0613 | 0.9974 | 0.9991 | 1.0000 | 0.9807 | 0.9813 | -0.0009 | 0.9731 |
| 1983 | 1.0872 | 1.0237 | 1.0354 | 1.0243 | 1.0014 | 1.0000 | 1.0251 | 1.0247 | -0.0117 | 0.9961 |
| 1984 | 1.1213 | 1.0039 | 1.0307 | 1.0583 | 1.0009 | 1.0230 | 1.0280 | 1.0277 | -0.0118 | 1.0000 |
| 19 | 1.0709 | 1.00 | 1.0279 | 1.0347 | 1.0000 | 1.0069 | 1.0069 | 1.0069 | -0.0118 | 1.0000 |
| 1986 | 1.0599 | 1.0000 | 1.0244 | 1.0303 | 1.0000 | 1.0043 | 1.0043 | 1.0043 | -0.0120 | 1.0000 |
| 1987 | 1.0668 | 1.0000 | 282 | 1.0364 | . 0000 | 1.0011 | 1.0011 | 1.0011 | 析 | 1.0000 |
| 1988 | 1.0790 | 1.0000 | 1.0361 | 1.0308 | 1.0000 | 1.0103 | 1.0103 | 1.0103 | -0.0034 | . 0000 |
| 1989 | 1.0768 | 1.0000 | 1.0392 | 1.0343 | 1.0000 | 1.0019 | 1.0019 | 1.0019 | -0.0027 | 1.0000 |
| 1990 | 1.0572 | 1.0000 | 1.0352 | 1.0214 | 0.9999 | 1.0000 | 0.9999 | 0.9999 | -0.0027 | 1.0000 |
| 1991 | 1.0269 | 0.9995 | 1.0283 | 0.9994 | 0.9997 | 1.0000 | 0.9992 | 0.9992 | -0.0024 | 0.9995 |
| 1992 | 1.0602 | 1.0005 | 1.0225 | 1.0186 | 1.0000 | 1.0175 | 1.0180 | 1.0179 | -0.0123 | 1.0000 |
| 19 | 1.055 | 1.0000 | 1.0208 | 1.0266 | 1.0000 | 1.0069 | 1.0 | 1.0069 | -0.0075 | 1.0000 |
| 19 | 1.067 | 1.0000 | 1.0179 | 1.0370 | 1.0000 | 1.0108 | 1.0108 | 1.0108 | -0.0074 | 1.0000 |
| 1995 | 1.0555 | 0.9958 | 1.0227 | 1.0363 | 1.0001 | 1.0000 | 0.9959 | 0.9959 | -0.0072 | 0.9958 |
| 1996 | 1.0584 | 1.0042 | 1.0135 | 1.0306 | 0.9999 | 1.0092 | 1.0133 | 1.0134 | -0.0072 | 1.0000 |
| 1997 | 1.0660 | 1.0000 | 1.0106 | 1.0457 | 1.0000 | 1.0087 | 1.0087 | 1.0087 | -0.0080 | 1.0000 |
| 1998 | 1.0568 | 1.0000 | 1.0048 | 1.0460 | 1.0000 | 1.0055 | 1.0055 | 1.0055 | -0.0081 | 1.0000 |
| 1999 | 1.0703 | 1.0000 | 1.0150 | 1.0445 | 1.0000 | 1.0096 | 1.0096 | 1.0096 | -0.0080 | 1.0000 |
| 2000 | 1.0783 | 1.0000 | 1.0266 | 1.0457 | 1.0000 | 1.0044 | 1.0044 | 1.0044 | -0.0076 | 1.0000 |
| 2001 | 1.0234 | 0.9968 | 1.0189 | 1.0079 | 0.9997 | 1.0000 | 0.9965 | 0.9963 | -0.0062 | 0.9968 |
| 2002 | 1.0305 | 1.0032 | 1.0113 | 1.0154 | 0.9999 | 1.0004 | 1.0036 | 1.0035 | -0.0064 | 1.0000 |
| 2003 | 1.0522 | 1.0000 | 1.0209 | 1.0140 | 1.0000 | 1.0165 | 1.0165 | 1.0165 | -0.0065 | 1.0000 |
| 2004 | 1.0775 | 1.0000 | 1.0287 | 1.0298 | 1.0000 | 1.0171 | 1.0171 | 1.0171 | -0.0076 | 1.0000 |
| 2005 | 1.0754 | 1.0000 | 1.0360 | 1.0276 | 1.0000 | 1.0101 | 1.0101 | 1.0101 | -0.0074 | 1.0000 |
| 2006 | 1.0652 | 1.0000 | 1.0308 | 1.0313 | 1.0000 | 1.0020 | 1.0020 | 1.0020 | -0.0037 | 1.0000 |
| 2007 | 1.0509 | 0.9980 | 1.0300 | 27 | 7 | 1.0000 | 0.9976 | 0.9976 | -0.0026 | 0.9980 |
| 20 | 1.0289 | 0.9961 | 1.0304 | 1.0031 | 0.9994 | 1.0000 | 0.9955 | 0.9951 | -0.0027 | 0.9940 |
| 2009 | 0.9512 | 0.9936 | 0.9924 | 0.9635 | 1.0013 | 1.0000 | 0.9949 | 0.9953 | -0.0026 | 0.9877 |
| 2010 | 1.0578 | 1.0125 | 1.0161 | 1.0184 | 1.0006 | 1.0090 | 1.0222 | 1.0214 | -0.0039 | 1.0000 |
| 2011 | 1.0486 | 0.9992 | 1.0283 | 1.0206 | 1.0000 | 1.0000 | 0.9992 | 0.9991 | -0.0037 | 0.9992 |
| 2012 | 1.0395 | 1.0008 | 1.0164 | 1.0187 | 0.9999 | 1.0032 | 1.0039 | 1.0039 | -0.0036 | 1.0000 |
| 2013 | 1.0319 | 1.0000 | 1.0114 | 1.0167 | 1.0000 | 1.0035 | 1.0035 | 1.0035 | -0.0036 | 1.0000 |
| 2014 | 1.0428 | 1.0000 | 1.0137 | 1.0244 | 1.0000 | 1.0042 | 1.0042 | 1.0042 | -0.0034 | 1.0000 |
| 2015 | 1.0290 | 1.0000 | 0.9975 | 1.0264 | 1.0000 | 1.0051 | 1.0051 | 1.0051 | -0.0034 | 1.0000 |
| 2016 | 1.0219 | 0.9981 | 1.0053 | 1.0185 | 1.0000 | 1.0000 | 0.9980 | 0.9980 | -0.0035 | 0.9980 |
| 2017 | 1.0465 | 1.0020 | 1.0186 | 1.0224 | 0.9999 | 1.0031 | 1.0049 | 1.0049 | -0.0037 | 1.0000 |
| 2018 | 1.0545 | 1.0000 | 1.0229 | 1.0230 | 1.0000 | 1.0077 | 1.0077 | 1.0077 | -0.0036 | 1.0000 |
| 2019 | 1.0366 | 1.0000 | 1.0134 | 1.0189 | 1.0000 | 1.0040 | 1.0040 | 1.0040 | -0.0037 | 1.0000 |
| 2020 | 0.9805 | 0.9943 | 1.0128 | 0.9742 | 0.9994 | 1.0000 | 0.9937 | 0.9935 | -0.0030 | 0.9943 |
| 2021 | 1.1183 | 1.0057 | 1.0482 | 1.0391 | 1.0003 | 1.0206 | 1.0267 | 1.0270 | -0.0119 | 1.0000 |
| 2022 | 1.1000 | 0.9884 | 1.0722 | 1.0381 | 0.9999 | 1.0000 | 0.9883 | 0.9883 | -0.0117 | 0.9884 |
| Mean | 1.0657 | 0.9998 | 1.0348 | 1.0238 | 0.99998 | 1.0061 | 1.0059 | 1.0059 | -0.0057 | 0.9970 |

On average, US nominal gross output grew $6.57 \%$ per year. The explanatory factors for this average growth rate are as follows: efficiency growth subtracted $0.02 \%$ per year, output price inflation contributed $3.48 \%$ per year, input quantity growth contributed $2.38 \%$ per year, the input mix effect on average was negligible, and technical progress contributed $0.61 \%$ per year. The level of efficiency $e^{t}$ was negative for 17 years: 1974-1976 (first oil shock recession), 19791983 (second oil shock recession), 1991, 1995, 2007-2009 (financial crisis), 2011, 2016, 2020 (covid shock) and 2022. For these years, production was in the interior of our estimated production possibilities set.
A new nonparametric measure of TFP growth for the economy going from year $t-1$ to $t$ can be defined (following Jorgenson and Griliches (1967)[31]) as an index of output growth divided by an index of input growth. An appropriate index of output growth is the gross output ratio $\boldsymbol{p}^{t} \cdot \boldsymbol{y}^{t} / \boldsymbol{p}^{t-1} \cdot \boldsymbol{y}^{t-1}$ divided by the gross output price index $\alpha^{t}$. An appropriate index of input growth is $\beta^{t}$. Thus define the nonparametric year $t$ Gross Output TFP growth rate, $\mathrm{NTFP}_{G}^{t}$, for the US economy as follows:

$$
\begin{equation*}
\operatorname{NTFP}_{G}^{t} \equiv\left\{\left[\boldsymbol{p}^{t} \cdot \boldsymbol{y}^{t} / \boldsymbol{p}^{t-1} \cdot \boldsymbol{y}^{t-1}\right] / \alpha^{t}\right\} / \beta^{t}=\varepsilon^{t} \gamma^{t} \tau^{t} ; \quad t=1971, \ldots, 2022 . \tag{C17}
\end{equation*}
$$

where the last equality in equation (C17) follows from equation (C16). The nonparametric year $t$ TFP growth, $\operatorname{NTFP}_{G}^{t}$, is equal to the product of period $t$ efficiency change $\varepsilon^{t}$, the period $t$ input mix index $\gamma^{t}$ (which typically will be small in magnitude) and the period $t$ measure of technical progress $\tau^{t}$. The nonparametric measure of TFP growth, $\mathrm{NTFP}_{G}^{t}$, and our old index number measure of TFP growth, $\mathrm{TFP}_{G}^{t}$ from Table A13 in Appendix A, are listed above in Table C1. Note that the index number and nonparametric estimates of TFP growth are almost identical. Note also that on average TFP growth (measured by both methods) is equal to $0.59 \%$ per year over the years 1971-2022. The average rate of technical progress is slightly higher at $0.61 \%$ per year. On average, inefficiency dragged down technical progress by $0.02 \%$ per year. Preventing recessions is important. Using the Diewert and Fox nonparametric methodology, technical progress can never be negative.
Table C1 also lists the Joint Cost Function based estimates of cost saving technical charge for each year $t, \kappa^{t}$. Taking the negative of these measures and adding one should lead to measures that are approximately equal to our measures of TFP growth. From viewing Table C1, it can be seen that this approximation is not very close. As was mentioned in the main text, the econometric estimates of technical progress are not able to capture year to year shocks to TFP. ${ }^{* 56}$ However, as was noted in the main text, the long term trends in cost saving technical change are fairly close to the long term trends in TFP growth.
We follow the example of Kohli (1990)[39] in order to obtain a levels decomposition for the observed level of nominal gross output in year $t, \boldsymbol{p}^{t} \cdot \boldsymbol{y}^{t}$, relative to its observed value in year 1 of our sample, $\boldsymbol{p}^{1970} \cdot \boldsymbol{y}^{1970}$. Define the cumulated explanatory variables as follows:

$$
\begin{equation*}
E^{1970} \equiv 1 ; A^{1970} \equiv 1 ; B^{1970} \equiv 1 ; C^{1970} \equiv 1 ; T^{1970} \equiv 1 \tag{C18}
\end{equation*}
$$

For $t=1971, \ldots, 2022$, define the above variables recursively as follows:

$$
\begin{align*}
& E^{t} \equiv \varepsilon^{t} E^{t-1} ; A^{t} \equiv \alpha^{t} A^{t-1} ; B^{t} \equiv \beta^{t} B^{t-1} ; C^{t} \equiv \gamma^{t} C^{t-1} ; T^{t} \equiv \tau^{t} T^{t-1} ; \\
& t=1971, \ldots, 2022 . \tag{C19}
\end{align*}
$$

[^30]Using the above definitions, it can be seen that we have the following levels decomposition for the level of year $t$ observed gross output relative to its level in 1970:

$$
\begin{equation*}
\mathrm{GO}^{t} \equiv \boldsymbol{p}^{t} \cdot \boldsymbol{y}^{t} / \boldsymbol{p}^{1970} \cdot \boldsymbol{y}^{1970}=A^{t} B^{t} C^{t} E^{t} T^{t} ; \quad t=1970, \ldots, 2022 \tag{C20}
\end{equation*}
$$

The components of the levels decomposition of gross output given by (C20) are listed in Table C2.

Table C2: The Levels Decomposition of Gross Output for the US 1970-2022

| Year | $\mathrm{GO}^{t}$ | $E^{t}$ | $A^{t}$ | $B^{t}$ | $C^{t}$ | $T^{t}$ | NTFP ${ }^{t}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1970 |  | 1.00 | 1.0000 | 1.0000 | 1. | 1.0000 | 1.0000 |
| 1971 | 1.0862 | 1.000 | 1.0483 | 1.0093 | 1.0000 | 1.0266 | 1.0266 |
| 1972 | 1.2000 | 1.0000 | 1.0960 | 1.0436 | 1.0000 | 1.0492 | 1.0492 |
| 1973 | 1.3442 | 1.0000 | 1.1633 | 1.0871 | 1.0000 | 1.0629 | 1.0629 |
| 1974 | 1.4824 | 0.9 | 1.2929 | 1.1040 | 0.9995 | 1.0629 | . 0385 |
| 1975 | . 6047 | 0.9763 | 1.4184 | 1.0906 | 0.9995 | 1.0629 | 3 |
| 1976 | 86 | 0.9999 | 1.4954 | 3 | 2 | 1.0631 | 1 |
| 1977 | 0 | 1.0000 | 1.5910 | 3 | 1.0001 | 3 | 4 |
| 1978 | 2.2852 | 1.0000 | . 70 | 1 | 1.0001 | 1.0929 | 930 |
| 1979 | 2.5696 | 0.996 | 1.8605 | 1.2684 | 0.9999 | 1.0929 | . 0889 |
| 1980 | 2.8066 | 0.9830 | 2.0488 | 1.2769 | 0.9985 | 1.0929 | 1.0728 |
| 1981 | 3.1296 | 0.9914 | 2.2268 | 1.2994 | 0.9982 | 1.0929 | 1.0816 |
| 1982 | 3.2486 | 0. | 2. | 1.2960 | 0. | 1.0929 | 7 |
| 1983 | 3.5319 | 0. | 2. | 1.3275 | 0. | 9 | 3 |
| 1984 | 3.9604 | 1.0000 | 2. | 1.4049 | 7 | 1.1181 | 7 |
| 1985 | 4.2410 | 1.0000 | 2. | 7 | 7 | 1.1258 | 54 |
|  | 0 | . 00 | 2. | 1.4977 | 0.9997 | 5 | 2 |
|  | 4.7951 | . 00 | 2.7 | 1.5522 | 0.9997 | 8 | 314 |
| 1988 | 5.1738 | 1.0000 | 2.8289 | 1.6000 | 0.9997 | 1.1435 | 431 |
| 1989 | 5.5709 | 1.0000 | 2.9397 | 1.6548 | 0.9997 | 1.1456 | 1.1452 |
| 19 | 5.8897 | 1.0000 | 3.0431 | 1.6901 | 0.999 | 1.1456 | 1 |
| 1991 | 6.0478 | 0.999 | 3.1292 | 1.6890 | 0.9 | 1.1456 | 3 |
| 1992 | 6.4121 | 1.000 | 3.199 | 1.7204 | 0.9 | 57 | 48 |
| 19 | 6.7652 | 1. | 3.2 | 1.7661 | 0.9993 | 1.1737 | 8 |
| 19 | 7.2182 | 1. | 3.324 | 1.8314 | 0.9993 | 3 | 5 |
| 19 | 7.6186 | 0.995 | 3.4001 | 1.8980 | 0.9993 | 1.1863 | 6 |
|  | 8.0634 | 1.0000 | 3.4459 |  | 0.9 | 2 | 963 |
| 19 | 8.5955 | 1.0000 | 3.482 | 2.0455 | 0.999 | 077 | 67 |
| 1998 | 9.0833 | 1.0000 | 3.4991 | 2.1395 | 0.9992 | 1.2143 | . 2133 |
| 19 | 9.7215 | 1.0000 | 3.5515 | 2.2346 | 0.9992 | 1.2260 | 1.2250 |
| 2 | 10.4823 | 1.0000 | 3.6459 | 2.3368 | 0.999 | 1.2313 | 1.2304 |
| 2001 | 10.7270 | 0.996 | 3.7147 | 2.3552 | 0.99 | 1.2313 | 1 |
| 20 | 11.0542 | 1.000 | 3.7567 | 2.3914 | 0.99 | 1.2318 | 1.2304 |
| 2003 | 11.6316 | 1.00 | 3.8351 | 2.4248 | 0.99 | 1.2522 | 8 |
| 20 | 12.5328 | 1. | 52 | 2.4970 | 0.9989 | 7 | 22 |
|  | 13.4772 | 1. | 4.0872 | 2.5660 | 9 | 5 | 5 |
|  | 14.3553 | 1.0000 | 4.2132 | 2.6462 | 0.9989 | 1 | 876 |
|  | 15.0854 | 0.9980 | 4.3394 | 2.7063 | 0.9985 | 1.2891 | 1.2846 |
| 2008 | 15.5213 | 0.9940 | 4.4711 | 2.7147 | 0.9980 | 1.2891 | 1.2788 |
| 2009 | 14.7643 | 0.9877 | 4.4372 | 2.6154 | 0.9992 | 1.2891 | 1.2722 |
| 2010 | 15.6176 | 1.0000 | 4.5089 | 2.6634 | 0.999 | 1.3007 | 1.3005 |
| 2011 | 16.3770 | 0.999 | 4.636 | 2.7182 | 0.999 | 1.3007 | 1.2995 |
| 2012 | 17.0230 | 1.000 | 4.7126 | 2.7691 | 0.9997 | 1.3049 | . 3045 |
| 20 | 17.5663 | 1. | 4.7 | 2.8 | 0.9997 | 1.3095 | . 3091 |
| 2014 | 18.3180 | 1.0 | 4.8312 | 2.8842 | 0.9997 | 1.3150 | 1.3146 |
|  | 18.8494 | 1.0000 | 4.8190 | 2.9604 | 0.9997 | 1.3217 | 1.3213 |
| , | 19.2614 | 0.9980 | 4.8445 | 3.0151 | 0.9996 | 1.3217 | 1.3186 |
| 2017 | 20.1568 | 1.0000 | 4.9344 | 3.0826 | 0.9995 | 1.3258 | 1.3251 |
| 2018 | 21.2546 | 1.0000 | 5.0474 | 3.1534 | 0.9995 | 1.3360 | 1.3354 |
| 2019 | 22.0333 | 1.0000 | 5.1151 | 3.2129 | 0.9995 | 1.3413 | 1.3407 |
| 2020 | 21.6040 | 0.9943 | 5.1806 | 3.1301 | 0.9989 | 1.3413 | 1.3323 |
| 2021 | 24.1606 | 1.0000 | 5.4305 | 3.2526 | 0.9993 | 1.3689 | 1.3679 |
| 2022 | 26.5764 | 0.9884 | 5.8224 | 3.3764 | 0.9992 | 1.3689 | 1.3519 |

Define the period $t$ level of Nonparametric Gross Output TFP relative to 1970, NTFP $^{t}$, as follows:

$$
\begin{equation*}
\mathrm{NTFP}^{1970} \equiv 1 ; \mathrm{NTFP}^{t} \equiv\left(\mathrm{NTFP}_{G}^{t}\right)\left(\mathrm{NTFP}^{t-1}\right) ; \quad t=1971, \ldots, 2022 \tag{C21}
\end{equation*}
$$

where $\operatorname{NTFP}_{G}^{t}$ is defined by equation (C17). Using definitions (C18)-(C21), it can be seen that we have the following levels decomposition for Nonparametric TFP using the cumulated explanatory factors defined by definitions (C18) and (C19):

$$
\begin{equation*}
\mathrm{NTFP}^{t} \equiv \mathrm{GO}^{t} /\left[A^{t} B^{t}\right]=C^{t} E^{t} T^{t} ; \quad t=1970, \ldots, 2022 \tag{C22}
\end{equation*}
$$

The series NTFP ${ }^{t}$ is listed in Table C2 above and the decomposition given by (C22) is plotted on Figure 2.


Figure 2 Decomposition of Nonparametric Gross Output TFP into Explanatory Factors

It can be seen that the oil shocks in the 1970s led to large amounts of inefficiency which led to TFP declines. The financial crisis of 2007 and the effects of Covid in 2020 and 2022 also led to inefficiency but the resulting declines in TFP were not as severe.

## Appendix D: Converting Cost Function Elasticities into Gross Output Function Elasticities

It is useful to be able to convert the elasticities of inverse output supply and input demand generated by the estimation of a joint cost function into elasticities of output supply and inverse input demand elasticities. We indicate how this can be done in this Appendix.*57
Suppose we have estimated a joint cost function $C(\boldsymbol{y}, \boldsymbol{w})$ for a particular year where $C$ is twice continuously differentiable with respect to the components of $\boldsymbol{y}$ and $\boldsymbol{w}$ at the year $t$ values for $\boldsymbol{y}$ and $\boldsymbol{w}$. Equations (6) and (7) in section 2 give us an output price vector $\boldsymbol{p}$ and an input demand vector $\boldsymbol{x}$ as functions of the derivatives of the cost function; i.e., $\boldsymbol{p}=\nabla_{y} C(\boldsymbol{y}, \boldsymbol{w})$ and $\boldsymbol{x}=\nabla_{w} C(\boldsymbol{y}, \boldsymbol{w})$. These equations implicitly define $\boldsymbol{y}$ and $\boldsymbol{w}$ as functions of $\boldsymbol{p}$ and $\boldsymbol{x}$, say $\boldsymbol{y}=\boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x})$ and $\boldsymbol{w}=\boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})$ so we have the following system of 10 equations for our particular example:

$$
\begin{align*}
& \boldsymbol{p}=\nabla_{y} C\left(\boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x}), \boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})\right) ;  \tag{D1}\\
& \boldsymbol{x}=\nabla_{w} C\left(\boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x}), \boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})\right) . \tag{D2}
\end{align*}
$$

Denote the 4 by 4 matrix of derivatives of the output supply functions $\boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x})$ with respect to the components of $\boldsymbol{p}$ as $\nabla_{p} \boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x})$ and the 4 by 6 matrix of derivatives of the functions $\boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x})$ with respect to the components of $\boldsymbol{x}$ as $\nabla_{x} \boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x})$. Denote the 6 by 4 matrix of derivatives of the inverse input demand functions $\boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})$ with respect to the components of $\boldsymbol{p}$ as $\nabla_{p} \boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})$ and the 6 by 6 matrix of derivatives of the inverse demand functions $\boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})$ with respect to the components of $\boldsymbol{x}$ as $\nabla_{x} \boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})$. If the 10 by 10 matrix of second order partial derivatives of the joint cost function $C(\boldsymbol{y}, \boldsymbol{w}), S_{G O}$, has an inverse, then the first order partial derivatives of the functions $\boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x})$ and $\boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})$ can be defined in terms of the elements of the inverse of the matrix of second order partial derivatives of the joint cost function as follows: ${ }^{* 58}$

$$
\left[\begin{array}{cc}
\nabla_{p} \boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x}) & \nabla_{x} \boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x})  \tag{D3}\\
\nabla_{p} \boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x}) & \nabla_{x} \boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})
\end{array}\right]=\left[\begin{array}{cc}
\nabla_{y y}^{2} C(\boldsymbol{y}, \boldsymbol{w}) & \nabla_{y w}^{2} C(\boldsymbol{y}, \boldsymbol{w}) \\
\nabla_{w y}^{2} C(\boldsymbol{y}, \boldsymbol{w}) & \nabla_{w w}^{2} C(\boldsymbol{y}, \boldsymbol{w})
\end{array}\right]^{-1}=\left[S_{G O}\right]^{-1} .
$$

It can be shown that the 4 by 4 matrix of Gross Output supply functions with respect to output prices, $\nabla_{p} \boldsymbol{y}^{*}(\boldsymbol{p}, \boldsymbol{x})=\left[\partial y_{m}^{*}(\boldsymbol{p}, \boldsymbol{x}) / \partial p_{k}\right]$, is a positive semidefinite matrix with rank equal to or less than 3 . For our 2022 data, this matrix has rank equal to 2. It can also be shown that the 6 by 6 matrix of derivatives of input prices with respect to input quantities, $\nabla_{x} \boldsymbol{w}^{*}(\boldsymbol{p}, \boldsymbol{x})=\left[\partial w_{n}^{*}(\boldsymbol{p}, \boldsymbol{x}) / \partial x_{j}\right]$, is a negative semidefinite matrix with rank equal to or less than 5. For our 2022 data, this matrix has rank equal to 4 . These derivatives can be converted into elasticities. Thus define $E y_{m} p_{k} \equiv\left[\partial y_{m}^{*}(\boldsymbol{p}, \boldsymbol{x}) / \partial p_{k}\right]\left[p_{k} / y_{m}\right], E y_{m} x_{n} \equiv\left[\partial y_{m}^{*}(\boldsymbol{p}, \boldsymbol{x}) / \partial x_{n}\right]\left[x_{n} / y_{m}\right], E w_{n} x_{j} \equiv$ $\left[\partial w_{n}^{*}(\boldsymbol{p}, \boldsymbol{x}) / \partial x_{j}\right]\left[x_{j} / w_{n}\right]$ and $E w_{n} p_{m} \equiv\left[\partial w_{n}^{*}(\boldsymbol{p}, \boldsymbol{x}) / \partial p_{m}\right]\left[p_{m} / w_{n}\right]$ for $n=1, \ldots, 6 ; j=$ $1, \ldots, 6 ; m=1, \ldots, 4$ and $k=1, \ldots, 4$. These estimated elasticities for the US Gross Output function for 2022 are listed in Tables D1-D4 below.

[^31]Table D1: Gross Output Elasticities with Respect to Changes in Gross Output Prices for 2022

| $E y_{m} p_{k}$ | $p_{1}$ | $p_{2}$ | $p_{3}$ | $p_{4}$ |  |
| :--- | :---: | ---: | ---: | ---: | ---: |
|  |  | C | G | I | X |
| $y_{1}$ | C | 0.778 | -0.416 | -0.517 | 0.154 |
| $y_{2}$ | G | -1.949 | 1.962 | 1.211 | -1.224 |
| $y_{3}$ | I | -1.557 | 0.779 | 1.038 | -0.260 |
| $y_{4}$ | X | 0.947 | -1.602 | -0.530 | 1.185 |

Note: Consumption (C), Government (G), Investment (I), Export (X).
Note that all main diagonal elasticities are positive; if an output price increases, output supply also increases.

Table D2: Gross Output Elasticities with Respect to Changes in Input Quantities for 2022

| $E y_{m} x_{n}$ |  | $x_{1}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ | $x_{5}$ | $x_{6}$ |
| ---: | :---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  |  | M | L | $\mathrm{K}_{\mathrm{M} \& \mathrm{E}}$ | $\mathrm{K}_{\mathrm{S}}$ | $\mathrm{K}_{\mathrm{O}}$ | $\mathrm{K}_{\mathrm{L}}$ |
| $y_{1}$ | C | -0.885 | 1.722 | 0.621 | 1.509 | -0.369 | -1.598 |
| $y_{2}$ | G | -0.640 | 1.239 | -2.719 | -1.716 | -2.433 | 7.269 |
| $y_{3}$ | I | 3.363 | -3.551 | 0.210 | -3.745 | 3.564 | 1.158 |
| $y_{4}$ | X | 0.925 | 0.004 | 0.106 | 2.388 | -1.011 | -1.412 |

Note: Consumption (C), Government (G), Investment (I), Export (X), Imports (M), Labour (L), Machinery and Equipment ( $\mathrm{K}_{\mathrm{M} \mathrm{\& E}}$ ), Structures $\left(\mathrm{K}_{\mathrm{S}}\right)$, Other Capital (mainly R\&D and Inventories: $\mathrm{K}_{\mathrm{O}}$ ) and Land ( $\mathrm{K}_{\mathrm{L}}$ ).

Table D3: Input Price Elasticities with Respect to Changes in Input Quantities for 2022

| $E w_{n} x_{j}$ | $x_{1}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ | $x_{5}$ | $x_{6}$ |  |
| :--- | :---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  |  | M | L | $\mathrm{K}_{\mathrm{M} \& \mathrm{E}}$ | $\mathrm{K}_{\mathrm{S}}$ | $\mathrm{K}_{\mathrm{O}}$ | $\mathrm{K}_{\mathrm{L}}$ |
| $w_{1}$ | M | -5.179 | 7.258 | -0.973 | 1.831 | -8.220 | 5.283 |
| $w_{2}$ | L | 2.197 | -4.035 | 0.696 | -0.734 | 4.153 | -2.278 |
| $w_{3}$ | $\mathrm{~K}_{\mathrm{M} \& \mathrm{E}}$ | -1.874 | 4.432 | -2.309 | -1.280 | -5.556 | 6.588 |
| $w_{4}$ | $\mathrm{~K}_{\mathrm{S}}$ | 1.552 | -2.056 | -0.564 | -5.341 | 0.992 | 5.417 |
| $w_{5}$ | $\mathrm{~K}_{\mathrm{O}}$ | -15.605 | 26.052 | -5.477 | 2.221 | -29.136 | 21.945 |
| $w_{6}$ | $\mathrm{~K}_{\mathrm{L}}$ | 11.884 | -16.930 | 7.693 | 14.368 | 26.001 | -43.015 |

Note: Imports (M), Labour (L), Machinery and Equipment ( $\mathrm{K}_{\mathrm{M} \mathrm{\& E}}$ ), Structures ( $\mathrm{K}_{\mathrm{S}}$ ), Other Capital (mainly R\&D and Inventories: $\mathrm{K}_{\mathrm{O}}$ ) and Land ( $\mathrm{K}_{\mathrm{L}}$ ).

Note that all main diagonal elasticities are negative; if economy wide amount of an input increases, then the demand clearing price for that increased input quantity decreases. Note that the own (inverse) elasticities of demand for Other Capital Services and Land Services (inputs 5 and 6) are very large in magnitude. This reflects the fact that the (direct) elasticities of demand for these inputs are very small in magnitude.

Table D4: Input Price Elasticities with Respect to Changes in Gross Output Prices for 2022

| $E w_{n} p_{m}$ | $p_{1}$ | $p_{2}$ | $p_{3}$ | $p_{4}$ |  |
| :--- | :---: | ---: | ---: | ---: | ---: |
|  |  | M | L | $\mathrm{K}_{\mathrm{M} \& \mathrm{E}}$ | $\mathrm{K}_{\mathrm{S}}$ |
| $w_{1}$ | M | -3.612 | -0.558 | 4.555 | 0.615 |
| $w_{2}$ | L | 2.128 | 0.327 | -1.456 | 0.001 |
| $w_{3}$ | $\mathrm{~K}_{\mathrm{M} \& \mathrm{E}}$ | 4.879 | -4.565 | 0.549 | 0.136 |
| $w_{4}$ | $\mathrm{~K}_{\mathrm{S}}$ | 5.222 | -1.269 | -4.302 | 1.348 |
| $w_{5}$ | $\mathrm{~K}_{\mathrm{O}}$ | -2.862 | -4.027 | 9.166 | -1.278 |
| $w_{6}$ | $\mathrm{~K}_{\mathrm{L}}$ | -14.668 | 14.252 | 3.529 | -2.114 |

Note: Imports (M), Labour (L), Machinery and Equipment ( $\mathrm{K}_{\mathrm{M} \mathrm{\& E}}$ ), Structures ( $\mathrm{K}_{\mathrm{S}}$ ), Other Capital (mainly R\&D and Inventories: $\mathrm{K}_{\mathrm{O}}$ ) and Land ( $\mathrm{K}_{\mathrm{L}}$ ).

Instead of working with the Gross Output function to model aggregate production in an economy, it is of interest to work with the GDP function. This model of production dates back to Samuelson (1953)[54]*59 and the model assumes that the prices $\boldsymbol{p} \equiv\left[p_{1}, p_{2}, p_{3}, p_{4}\right]$ and $w_{1}$ (the price of imports) are fixed for outputs $\boldsymbol{y} \equiv\left[y_{1}, y_{2}, y_{3}, y_{4}\right]$ and imports $x_{1}$ and the economy's endowment vector of primary inputs $\boldsymbol{x}^{\circ} \equiv\left[x_{2}, x_{3}, x_{4}, x_{5}, x_{6}\right]$ is also viewed as being fixed. ${ }^{* 60}$ Thus in this model of production, world prices are fixed and the country's endowment of labour and capital are fixed in the short run. Denote the vector of input prices that corresponds to $\boldsymbol{x}^{\circ}$ (the new vector of fixed factors of production) by $\boldsymbol{w}^{\circ} \equiv\left[w_{2}, w_{3}, w_{4}, w_{5}, w_{6}\right]$. Using our new notation explained in the previous paragraph, the 10 estimating equations for our (old) joint cost function model for the year 2022 can be written as follows:

$$
\begin{align*}
& \boldsymbol{p}=\nabla_{y} C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right)  \tag{D4}\\
& x_{1}=\partial C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right) / \partial w_{1}  \tag{D5}\\
& \boldsymbol{x}^{\circ}=\nabla_{w^{\circ}} C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right) \tag{D6}
\end{align*}
$$

where $C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right) \equiv C^{2022}\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right)=C^{2022}(\boldsymbol{y}, \boldsymbol{w})$. We now shift imports $x_{1}$ from being an input in the Gross Output model of production to imports being a negative output which leads to the GDP model of production where $\operatorname{GDP}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ is defined as the maximum of $\boldsymbol{p} \cdot \boldsymbol{y}-w_{1} x_{1}$ (with respect to $x_{1}$ and the components of $\boldsymbol{y}$ ), subject to $\left(\boldsymbol{y}, x_{1}, \boldsymbol{x}^{\circ}\right) \in S^{2022}$. The solution functions are denoted by $\boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right), x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$, The vector of equilibrium primary input prices is denoted by $\boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)=\nabla_{x^{\circ}} \operatorname{GDP}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) .{ }^{* 61}$
The first order partial derivatives of the functions $\boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) \equiv\left[y_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right), \ldots\right.$, $\left.y_{4}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)\right]^{T}$ and $\boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) \equiv\left[w_{2}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right), \ldots, w_{6}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)\right]^{T}$ can be obtained by differentiating both sides of the following 9 equations ${ }^{* 62}$ with respect to the 9 components of

[^32]$\boldsymbol{p}$ and $\boldsymbol{x}^{\circ}$ :
\[

$$
\begin{align*}
& \boldsymbol{p}=\nabla_{y} C\left(\boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right), w_{1}, \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)\right) ;  \tag{D7}\\
& \boldsymbol{x}^{\circ}=\nabla_{w^{\circ}} C\left(\boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right), w_{1}, \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)\right) . \tag{D8}
\end{align*}
$$
\]

If the 9 by 9 matrix $\mathbf{S}$ defined below has an inverse, the matrices of first order derivatives of the functions $\boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ and $\boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ with respect to the 9 components of $\boldsymbol{p}$ and $\boldsymbol{x}^{\circ}$ are defined as follows: ${ }^{* 63}$

$$
\begin{align*}
{\left[\begin{array}{cc}
\nabla_{p} \boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) & \nabla_{x^{\circ}} \boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) \\
\nabla_{p} \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) & \nabla_{x^{\circ}} \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)
\end{array}\right] } & =\left[\begin{array}{cc}
\nabla_{y y}^{2} C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right) & \nabla_{y w^{\circ}}^{2} C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right) \\
\nabla_{w^{\circ} y}^{2} C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right) & \nabla_{w^{\circ} w^{\circ}} C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right)
\end{array}\right]^{-1} \\
& \equiv[\mathbf{S}]^{-1} . \tag{D9}
\end{align*}
$$

In order to determine the first order partial derivatives of $\boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ and $\boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ with respect to $w_{1}$, differentiate both sides of (D7) and (D8) with respect to $w_{1}$ and obtain the following matrix equation:

$$
\left[\begin{array}{c}
\nabla_{w_{1}} \boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)  \tag{D10}\\
\nabla_{w_{1}} \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)
\end{array}\right]=-[\mathbf{S}]^{-1}\left[\begin{array}{c}
\nabla_{y w_{1}}^{2} C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right) \\
\nabla_{w^{\circ} w_{1}}^{2} C\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right)
\end{array}\right] .
$$

Finally, we need to calculate the first order partial derivatives of $x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ with respect to $w_{1}$ and the components of $\boldsymbol{p}$ and $\boldsymbol{x}^{\circ}$, Define $x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ as follows:

$$
\begin{equation*}
x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) \equiv \partial C\left(\boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right), w_{1}, \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)\right) / \partial w_{1} . \tag{D11}
\end{equation*}
$$

Differentiate (D11) with respect to $w_{1}$ where all cost function partial derivatives are evaluated at $\left(\boldsymbol{y}, w_{1}, \boldsymbol{w}^{\circ}\right)$ :

$$
\begin{align*}
\partial x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) / \partial w_{1} & =\nabla_{w_{1} w_{1}}^{2} C+\left[\begin{array}{ll}
\nabla_{w_{1} y}^{2} C & \nabla_{w_{1} w^{\circ}}^{2} C
\end{array}\right]\left[\begin{array}{c}
\nabla_{w_{1}} \boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) \\
\nabla_{w_{1}} \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)
\end{array}\right] \\
& =\nabla_{w_{1} w_{1}}^{2} C-\left[\begin{array}{ll}
\nabla_{w_{1} y}^{2} C & \nabla_{w_{1} w^{\circ}}^{2} C
\end{array}\right] \mathbf{S}^{-1}\left[\begin{array}{c}
\nabla_{y w_{1}}^{2} C \\
\nabla_{w^{\circ} w_{1}}^{2} C
\end{array}\right] \tag{D12}
\end{align*}
$$

where the second equality follows using (D10). Now differentiate (D11) with respect to the components of $\boldsymbol{p}$ and $\boldsymbol{x}^{\circ}$. We obtain the following formula for the partial derivatives of $x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ where $\nabla_{p}^{T} x_{1}$ denotes the transpose of the column vector $\nabla_{p} x_{1}$ :

$$
\begin{align*}
{\left[\begin{array}{ll}
\nabla_{p}^{T} x_{1} & \nabla_{x^{\circ}}^{T} x_{1}
\end{array}\right] } & =\left[\begin{array}{ll}
\nabla_{w_{1} y}^{2} C & \nabla_{w_{1} w^{\circ}}^{2} C
\end{array}\right]\left[\begin{array}{cc}
\nabla_{p} \boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) & \nabla_{x^{\circ}} \boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) \\
\nabla_{p} \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) & \nabla_{x^{\circ}} \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)
\end{array}\right] \\
& =\left[\begin{array}{ll}
\nabla_{w_{1} y}^{2} C & \nabla_{w_{1} w^{\circ}}^{2} C
\end{array}\right] \mathbf{S}^{-1} \tag{D13}
\end{align*}
$$

where the second equality follows using (D9). Comparing (D13) to the partial derivatives defined in (D10) and using the symmetry of the second order derivatives of the joint cost function, we see that:

$$
\begin{equation*}
-\nabla_{p} x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)=\nabla_{w_{1}} \boldsymbol{y}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) \text { and }-\nabla_{x^{\circ}} x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)=\nabla_{w_{1}} \boldsymbol{w}^{\circ}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) . \tag{D14}
\end{equation*}
$$

[^33]The above derivatives can be converted into elasticities. Equations (D9) are used to define the partial derivatives for the elasticities listed in Tables D5-D8. Define $E y_{m} p_{k} \equiv$ $\left[\partial y_{m}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) / \partial p_{k}\right]\left[p_{k} / y_{m}\right]$, These estimated output price elasticities for the US GDP function for 2022 are listed in Table D5 below.

Table D5: GDP Output Elasticities with Respect to Changes in Output Prices for 2022

| $E y_{m} p_{k}$ | $p_{1}$ |  | $p_{2}$ |  | $p_{3}$ |
| :--- | :---: | ---: | ---: | ---: | ---: |
| $p_{4}$ |  |  |  |  |  |
|  |  | C | G | I | X |
| $y_{1}$ | C | 1.396 | -0.321 | -1.295 | 0.049 |
| $y_{2}$ | G | -1.503 | 2.031 | 0.648 | -1.300 |
| $y_{3}$ | I | -3.902 | 0.417 | 3.996 | 0.139 |
| $y_{4}$ | X | 0.302 | -1.702 | 0.283 | 1.295 |

Note: Consumption (C), Government (G), Investment (I), Export (X).
Comparing the own elasticities in Table D5 with the corresponding own elasticities in Table D1, it can seen that the new own elasticities, $E y_{m} p_{m}$ for $m=1,2.3 .4$, are all bigger than the corresponding Table D1 entries. This follows from Samuelson's Le Chatelier Principle:*64 the quantity of imports is no longer fixed when we switch from the Gross Output Function to the GDP Function so the economy has more flexibility to respond to an increase in the price of an output and thus the output supply response to an increase in price will tend to be bigger.
Define the elasticity of output $m$ with respect to an increase in the economy's endowment of inputs 2-6 as $E y_{m} x_{n} \equiv\left[\partial y_{m}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) / \partial x_{n}\right]\left[x_{n} / y_{m}\right]$ for $m=1, \ldots, 4$ and $n=2, \ldots, 6$. These elasticities are listed in Table D6.

Table D6: GDP Output Elasticities with Respect to Changes in Input Quantities for 2022

| $y_{m} x_{n}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ | $x_{5}$ | $x_{6}$ |  |
| :--- | :---: | :---: | ---: | :---: | :---: | :---: |
|  |  | L | $\mathrm{~K}_{\mathrm{M} \& \mathrm{E}}$ | $\mathrm{K}_{\mathrm{S}}$ | $\mathrm{K}_{\mathrm{O}}$ | $\mathrm{K}_{\mathrm{L}}$ |
| $y_{1}$ | C | 0.482 | 0.787 | 1.196 | 1.035 | -2.500 |
| $y_{2}$ | G | 0.342 | -2.599 | -1.943 | -1.417 | 6.616 |
| $y_{3}$ | I | 1.162 | -0.421 | -2.556 | -1.773 | 4.589 |
| $y_{4}$ | X | 1.299 | -0.067 | 2.715 | -2.478 | -0.469 |

Note: Consumption (C), Government (G), Investment (I), Export (X), Labour (L), Machinery and Equipment ( $\mathrm{K}_{\mathrm{M} \mathrm{\& E}}$ ), Structures $\left(\mathrm{K}_{\mathrm{S}}\right)$, Other Capital (mainly R\&D and Inventories: $\mathrm{K}_{\mathrm{O}}$ ) and Land $\left(\mathrm{K}_{\mathrm{L}}\right)$.

Define the elasticity of input price $n$ with respect to an increase in the endowment of input $j$ as $E w_{n} x_{j} \equiv\left[\partial w_{n}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) / \partial x_{j}\right]\left[x_{j} / w_{n}\right]$ for $n=2, \ldots, 6$ and $j=2, \ldots, 6$. These (inverse) input demand elasticities are listed in Table D7.

[^34]Table D7: GDP Inverse Input Demand Elasticities with Respect to Changes in Input Quantities

| $E w_{n} x_{j}$ | $x_{2}$ |  | $x_{3}$ | $x_{4}$ | $x_{5}$ | $x_{6}$ |
| :--- | :---: | ---: | ---: | ---: | ---: | ---: |
|  |  | L | $\mathrm{~K}_{\mathrm{M} \& \mathrm{E}}$ | $\mathrm{K}_{\mathrm{S}}$ | $\mathrm{K}_{\mathrm{O}}$ | $\mathrm{K}_{\mathrm{L}}$ |
| $w_{2}$ | L | -0.956 | 0.205 | 0.005 | 12.970 | -0.037 |
| $w_{3}$ | $\mathrm{~K}_{\mathrm{M} \& \mathrm{E}}$ | 1.806 | -1.957 | -0.183 | -10.532 | 4.676 |
| $w_{4}$ | $\mathrm{~K}_{\mathrm{S}}$ | 0.120 | -0.855 | -4.793 | -34.581 | 7.000 |
| $w_{5}$ | $\mathrm{~K}_{\mathrm{O}}$ | 0.215 | -0.624 | -0.140 | -4.365 | 0.164 |
| $w_{6}$ | $\mathrm{~K}_{\mathrm{L}}$ | -0.276 | 5.461 | 18.569 | 262.729 | -30.891 |

Note: Labour (L), Machinery and Equipment ( $\mathrm{K}_{\mathrm{M} \& E}$ ), Structures $\left(\mathrm{K}_{\mathrm{S}}\right)$, Other Capital (mainly R\&D and Inventories: $\mathrm{K}_{\mathrm{O}}$ ) and Land ( $\mathrm{K}_{\mathrm{L}}$ ).

Note that the main diagonal elements in the above 5 by 5 matrix are less negative than the corresponding diagonal elements in the last 5 rows and columns of the matrix defined in Table D3. The GDP model of production allows for more substitution between domestic inputs and imports than the Gross Output model of production, which treats imports as a fixed factor of production. Thus if the quantity of labour or a capital service input increases in the GDP model, the corresponding price of that factor of production will decrease less than the corresponding change in the input price for the Gross Output model.
Define the elasticity of input price $n$ with respect to an increase in the price of output $m$ as $E w_{n} p_{m} \equiv\left[\partial w_{n}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) / \partial p_{m}\right]\left[p_{m} / w_{n}\right]$ for $n=1, \ldots, 6$ and $m=1, \ldots, 4$.

Table D8: Input Price Elasticities with Respect to Changes in GDP Output Prices for 2022

| $E w_{n} p_{m}$ | $p_{1}$ | $p_{2}$ | $p_{3}$ | $p_{4}$ |  |
| :--- | :---: | ---: | ---: | ---: | ---: |
|  |  | C | G | I | X |
| $w_{2}$ | L | 0.596 | 0.037 | 0.001 | 0.296 |
| $w_{3}$ | $\mathrm{~K}_{\mathrm{M} \& \mathrm{E}}$ | 6.187 | -1.699 | -0.100 | -1.145 |
| $w_{4}$ | $\mathrm{~K}_{\mathrm{S}}$ | 4.139 | -0.129 | -0.454 | -0.653 |
| $w_{5}$ | $\mathrm{~K}_{\mathrm{O}}$ | 8.023 | -0.241 | -0.034 | -0.211 |
| $w_{6}$ | $\mathrm{~K}_{\mathrm{L}}$ | -0.056 | 0.525 | 1.122 | 3.165 |

Note: Consumption (C), Government (G), Investment (I), Export (X),
Labour (L), Machinery and Equipment ( $\mathrm{K}_{\mathrm{M} \& E}$ ), Structures ( $\mathrm{K}_{\mathrm{S}}$ ),
Other Capital (mainly R\&D and Inventories: $\mathrm{K}_{\mathrm{O}}$ ) and Land ( $\mathrm{K}_{\mathrm{L}}$ ).
We use (D13) to calculate the derivatives of the import demand function, $x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ with respect to the components of $\boldsymbol{p}$ and $\boldsymbol{x}^{\circ}$. These derivatives are then transformed into the elasticities of import demand with respect to output prices, $E x_{1} p_{m}$ for $m=1, \ldots, 4$ and with respect to input endowments, $E x_{1} x_{n}$ for $n=2,3, \ldots, 6$. These elasticities for the year 2022 are listed in Table D9.

Table D9: Elasticities of Import Demand with Respect to Changes in Output Prices and Input Quantities for the GDP Model of Production

| $E x_{1} p_{1}$ | $E x_{1} p_{2}$ | $E x_{1} p_{3}$ | $E x_{1} p_{4}$ | $E x_{1} x_{2}$ | $E x_{1} x_{3}$ | $E x_{1} x_{4}$ | $E x_{1} x_{5}$ | $E x_{1} x_{6}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| C | G | I | X | L | $\mathrm{K}_{\mathrm{M} \& \mathrm{E}}$ | $\mathrm{K}_{\mathrm{S}}$ | $\mathrm{K}_{\mathrm{O}}$ | $\mathrm{K}_{\mathrm{L}}$ |
| -0.698 | -0.108 | 0.880 | 0.119 | 1.401 | -0.188 | 0.354 | -1.587 | 1.020 |

> Note: Consumption $(\mathrm{C})$, Government $(\mathrm{G})$, Investment $(\mathrm{I})$, Export $(\mathrm{X})$, Labour (L), Machinery and Equipment $\left(\mathrm{K}_{\mathrm{M} \& \mathrm{E}}\right)$, Structures $\left(\mathrm{K}_{\mathrm{S}}\right)$, Other Capital (mainly R\&D and Inventories: $\left.\mathrm{K}_{\mathrm{O}}\right)$ and Land $\left(\mathrm{K}_{\mathrm{L}}\right)$.

Using (D14), we can calculate the derivatives of $y_{m}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ for $m=1,2,3,4$ and the
derivatives of $w_{n}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right)$ for $n=2, \ldots, 6$ with respect to $w_{1}$, the price of imports. These derivatives can be turned into elasticities and they are reported in Table D10.

Table D10: Elasticities of Output Supply and Input Price with Respect to the Price of Imports

| $E y_{1} w_{1}$ | $E y_{2} w_{1}$ | $E y_{3} w_{1}$ | $E y_{4} w_{1}$ | $E w_{2} w_{1}$ | $E w_{3} w_{1}$ | $E w_{4} w_{1}$ | $E w_{5} w_{1}$ | $E w_{6} w_{1}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| C | G | I | X | L | $\mathrm{K}_{\mathrm{M} \& \mathrm{E}}$ | $\mathrm{K}_{\mathrm{S}}$ | $\mathrm{K}_{\mathrm{O}}$ | $\mathrm{K}_{\mathrm{L}}$ |
| 0.171 | 0.124 | -0.649 | -0.179 | -0.424 | 0.362 | -0.300 | 3.013 | -2.295 |

Note: Consumption (C), Government (G), Investment (I), Export (X), Labour (L), Machinery and Equipment ( $\mathrm{K}_{\mathrm{M} \& E}$ ), Structures $\left(\mathrm{K}_{\mathrm{S}}\right)$, Other Capital (mainly R\&D and Inventories: $\mathrm{K}_{\mathrm{O}}$ ) and Land ( $\mathrm{K}_{\mathrm{L}}$ ).

Finally, we can use (D12) to calculate the derivative of import demand with respect to an increase in the price of imports for $2022, \partial x_{1}\left(\boldsymbol{p}, w_{1}, \boldsymbol{x}^{\circ}\right) / \partial w_{1}$. This derivative can be multiplied by $w_{1}^{2022}$ and divided by the fitted value for imports in 2022 from our estimated cost function model to construct the GDP model elasticity $E x_{1} w_{1}=-0.119$. This elasticity can be compared to the own price elasticity of import demand which is generated by our estimated cost function for 2022. This cost function based import demand function is $x_{1}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}\right)=\partial C^{t}\left(\boldsymbol{y}^{t}, \boldsymbol{w}^{t}\right) / \partial w_{1}$ where $t=2022$. The corresponding cost function based import demand own price elasticity of demand turned out to equal -0.028 , which is less than -0.119 in magnitude. Using the Joint Cost function model of producer behavior, outputs $\boldsymbol{y}$ are held fixed as are domestic input prices $\boldsymbol{w}^{\circ}$. Using the GDP model of production, output prices $\boldsymbol{p}$ and domestic endowments $\boldsymbol{x}^{\circ}$ are held fixed. Thus when the price of imports increases, producers are free to substitute other inputs for imports using the Joint Cost function model which evidently leads to a larger reduction in import demand compared to the corresponding response using the GDP function model where the quantities of other inputs are held fixed in response to the increase in the price of imports (but outputs are free to vary). Thus the magnitude of producer responses to changes in prices and quantities can be very different depending on what is being held constant. Different models should be used to answer different questions.
A final note of caution: for this conversion of elasticities method to work successfully, the underlying joint cost function should provide a good fit to the data.

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[^1]:    *1 This approach to estimating technologies (and preferences) was developed by researchers at the University of California at Berkeley in the late 1960s and early 1970s; see McFadden (1966)[47] (1978)[48], Diewert (1971)[9] (1973)[10] (1974a)[11] (1974b)[12] and Christensen, Jorgenson and Lau (1973)[8].
    *2 In 1989, Kohli was the first person to adapt the Normalized Quadratic functional form to the GDP function. His paper was later published as Kohli (1993a)[41]. See also Kohli (1991)[40] for another early application of the NQ functional form.

[^2]:    ${ }^{* 3}$ For the importance of including land in production functions, see Kumhof, Tideman, Hudson and Goodhart (2021)[45] and Muellbauer (2024)[50].
    *4 The advantage of the nonparametric approach is that it distinguishes between technical progress and inefficiency. Of particular interest is the amount of inefficiency that occurred in the US economy during the oil shocks of the 1970s and the Covid shocks of 2020-2022.
    *5 Thus we are assuming that the technology exhibits constant returns to scale. The reason for this strong assumption is that it proves to be too difficult to distinguish technical progress from returns to scale using time series data for a production unit. Our goal in this paper is to estimate aggregate technologies using time series data.

[^3]:    ${ }^{* 6}$ See Diewert (1974a;113)[11] on the concept of a flexible functional form.
    *7 Alternative names for this function are the national product function Samuelson (1953; 10)[54], the gross profit function Gorman (1968)[27], the conditional profit function McFadden (1966)[47] (1978)[48], the variable profit function Diewert (1973)[10], the revenue function McFadden (1966)[47], Diewert (1974b)[12] and the GDP function Samuelson (1953)[54], Kohli (1978)[33] (1982)[36] (1991)[40]. These functions differ from the gross output function by taking some variable inputs (like imports or labour services) out of the list of inputs and treating them as negative outputs.
    ${ }^{* 8}$ Notation: $\boldsymbol{p} \cdot \boldsymbol{y} \equiv \sum_{m=1}^{M} p_{m} y_{m} \equiv \boldsymbol{p}^{T} \boldsymbol{y}$ is the inner product between the vectors $\boldsymbol{p}$ and $\boldsymbol{y}$.
    *9 See also Diewert (1974a; 140)[11] (2022)[17].

[^4]:    *10 See McFadden (1966)[47] (1978)[48] and Diewert (1974a)[11] (2022)[17] for references to the literature. The first flexible functional form for a joint cost function that was estimated empirically was the translog joint cost function introduced by Burgess (1974)[6]. The problem with this functional form is the difficulty in imposing curvature conditions on it without destroying its flexibility.

[^5]:    *11 There is an extensive literature on alternative specifications of joint and nonjoint production; see Samuelson (1966)[56] and Kohli (1981a)[34] (1983a)[37] (1991; 42-46)[40] (1993b)[42] (2005)[44]. Our purpose is not to test for the validity of alternative specifications of joint production: we simply wish to determine whether a joint cost function or a gross output function will fit the data better, given that a bilinear functional form is the initial functional form. Our paper does not answer this question definitively; i.e., which specification of technology is "best".
    *12 See the Asian Productivity Organization (2022)[1] for a description of the data base. The detailed data base is not available to the public. However, the US data used in this study is listed in Appendix A.
    ${ }^{* 13}$ See White (2004)[59].

[^6]:    *14 There are now 38 parameters in our model.
    *15 The standard errors for $\left(v_{2}^{*}, v_{3}^{*}, v_{4}^{*}\right)$ were ( $0.0229,0.0187,0.0052$ ).

[^7]:    ${ }^{* 16}$ Thus $\mathbf{A}(t) \equiv[1-(t / 52)] \mathbf{U}^{0} \mathbf{U}^{0 T}+[t / 52] \mathbf{U}^{1} \mathbf{U}^{1 T}$ where $\mathbf{U}^{0}$ and $\mathbf{U}^{1}$ are lower triangular matrices that satisfy $\mathbf{U}^{0 T} \mathbf{1}_{6}=\mathbf{0}_{6}$ and $\mathbf{U}^{1 T} \mathbf{1}_{6}=\mathbf{0}_{6}$ and $\mathbf{B}(t) \equiv[1-(t / 52)] \mathbf{V}^{0} \mathbf{V}^{0 T}+[t / 52] \mathbf{V}^{1} \mathbf{V}^{1 T}$ where $\mathbf{V}^{0}$ and $\mathbf{V}^{1}$ are lower triangular matrices that satisfy $\mathrm{V}^{0 T} \mathbf{1}_{4}=\mathbf{0}_{4}$ and $\mathbf{V}^{1 T} \mathbf{1}_{4}=\mathbf{0}_{4}$.
    *17 See Appendix C.

[^8]:    *18 Standard errors for our estimated coefficients and our Shazam codes are available on demand.

[^9]:    *19 We did not attempt to compute standard errors for these elasticities. The standard errors are likely to be large so our tables of elasticities are only very rough approximations to the "true" elasticities.

[^10]:    ${ }^{* 20}$ In retrospect, it may have been wiser to impose nonnegativity constraints on the components of the

[^11]:    ${ }^{* 22}$ The spline functions $f_{n i}(t)$ that were defined in the previous section are not always differentiable with respect to time $t$. However, the one sided right hand side directional derivatives exist and are straightforward to calculate. These one sided derivatives were used to calculate the various derivatives of the estimated cost function with respect to time.
    ${ }^{* 23}$ Note that $\mathrm{TFPG}^{t}$ and $\mathrm{NTFP}^{t}$ are virtually identical in Figure 1.

[^12]:    ${ }^{* 24}$ Of course, it would be preferable to estimate the technology sets for each sector. This option runs into two problems: (i) typically the price and quantity data on capital stocks and labour input are not available by sector and (ii) when estimating sectoral Joint Cost functions or Gross Output functions using sectoral data, we require data on the price and quantity of intermediate inputs, which is also not readily available. When we work with national data, intermediate input flows between sectors cancel out, except for imports.
    *25 Alternatively, we could regard a small set of inputs as fixed and treat the remaining inputs as negative outputs and estimate a variable profit function. This would fix the size of the $\mathbf{A}$ matrix at $N$ by $N$ where $N$ is small and the $\mathbf{D}$ matrix would have size $N$ by $M$ where $M$ is large. The semi-flexible idea could be applied to the $\mathbf{B}$ matrix; i.e., the rank of $\mathbf{B}$ would be limited.

[^13]:    *26 See Diewert and Fox (2008; 176-177)[18] for an example of how this model could be implemented.
    *27 More elaborate models for the markups could be estimated. For example, we could specify linear time trends for the markups or we could specify more flexible spline models.

[^14]:    *28 The methodology for constructing the labour, capital and productivity accounts for the US and Asian countries is explained in Asian Productivity Organization (2022; 165-188)[1].
    ${ }^{* 29}$ The price of consumption was adjusted downward by removing indirect taxes on outputs in order to obtain an approximation to the producer price of consumption.
    *30 The price of imports was adjusted upwards by adding tariffs to the border price of imports.
    *31 We cumulated the estimates for real inventory change to form beginning of the year inventory stocks. We set the price of inventory change equal to the end of year price of inventory stock. These conventions allowed us to apply user cost theory to the stock of inventories. In Tables A1 and A2, we labelled the year $t$ price and quantity of inventory change as $P_{I 17}^{t}$ and $Q_{I 17}^{t}$. Assets 11-16 are land stocks. We did not include investments in land stocks in this study since these investments are not recognized in the current international System of National Accounts.

[^15]:    *32 For definitions of the Törnqvist price and quantity indexes and their connection to the economic approach to index number theory, see Diewert (1976)[13] and Diewert and Morrison (1986)[22].

[^16]:    *33 This is mainly government owned land used for various purposes.

[^17]:    *34 See Jorgenson (1963)[29], Jorgenson and Griliches (1967)[31], (1972)[32], Christensen and Jorgenson (1969)[7] and Jorgenson (1989)[30].
    *35 This is the method used by Diewert (1980, 472-473)[14] (2014)[16] to derive a user cost formula.
    *36 This user cost formula was also derived by Christensen and Jorgenson (1969)[7] using a different method of derivation. Jorgenson was a firm advocate of using ex post asset inflation rates in a user cost formula so we term the user costs defined by (A4) as Jorgensonian user costs.

[^18]:    ${ }^{* 37}$ Here is a list of the land and inventory assets ( 11 to 17 ) and the number of negative user costs: 11: 10; 12: $4 ; 13: 0 ; 14: 8 ; 15: 16 ; 21 ; 17: 9$. Thus there was a total of 68 negative user costs. The user costs for assets 1-10 were all positive.
    ${ }^{* 38}$ For 1970, set $i_{n}^{1970^{*}}=(1 / 2)\left(i_{n}^{1970}+i_{n}^{1971}\right)$, for 1971, use the 3 year centered moving average $i_{n}^{1971 *}=$ $(1 / 3)\left(i_{n}^{1970}+i_{n}^{1971}+i_{n}^{1972}\right)$ and for year 3, use a 5 year centered average. Similar adjustments to the 7 year centered moving average rule were made for the final 3 years in our sample.

[^19]:    *39 However, our accounting framework does not include monetary holdings and natural resource stocks as assets and this omission means that our ex post rates of return on assets are too high.

[^20]:    *40 These capital services input subaggregates were formed using Törnqvist direct aggregation of input quantities; i.e., bilateral chained quantity indexes were formed first (quantities were always positive) and then the corresponding aggregate price indexes were calculated by deflating total subaggregate value by the quantity index.

[^21]:    ${ }^{* 41}$ When land prices increase, land inflation rates increase even more on a proportional basis, so the user cost of land tends to decrease when there is a property bubble. When the bubble ends, the user cost of land tends to increase.
    *42 This precise methodology for measuring TFP is due to Diewert and Morrison (1986)[22] and Kohli (1990)[39] but it is a variation on the methodology pioneered by Jorgenson and Griliches (1967)[31] and Christensen and Jorgenson (1969) [7].

[^22]:    ${ }^{* 43}$ If there were sectoral production functions for $C, G$ and $I$ (the no joint production hypothesis), then all of the elasticities listed in Table A14 would be equal to zero. It appears that production is "almost" Nonjoint.

[^23]:    *44 Perhaps this phenomenon is due to the growth of multinationals and the resulting internal international supply chains. Another possible explanation is the growth of services relative to goods production in most countries: services inputs may be less substitutable than goods inputs.

[^24]:    ${ }^{* 45}$ This assumption implies that the $N$ partial derivative functions, $\partial C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) / \partial w_{n}$ for $n=1, \ldots, N$, are also linearly homogeneous in the components of $\boldsymbol{y}$. Thus by Euler's Theorem on homogeneous functions, $\partial C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) / \partial w_{n}=\sum_{m=1}^{M} y_{m}^{*} \partial^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) / \partial w_{n} \partial y_{m}$ for $n=1, \ldots, N$. These equations imply that $\nabla_{w} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)^{T}=\boldsymbol{y}^{* T} \nabla_{y w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$. Similarly, $\nabla_{w} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)^{T}=\boldsymbol{y}^{* T} \nabla_{y w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$. Thus if $\nabla_{y w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\nabla_{y w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$, then $\nabla_{w} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\nabla_{w} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$.
    *46 This assumption implies that the $M$ partial derivative functions, $\partial C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) / \partial y_{m}$ for $m=1, \ldots, M$, are also linearly homogeneous in the components of $\boldsymbol{w}$. Thus by Euler's Theorem on homogeneous functions, $\partial C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) / \partial y_{m}=\sum_{n=1}^{N} w_{n}^{*} \partial^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right) / \partial y_{m} \partial w_{n}$ for $n=1, \ldots, N$. These equations imply that $\nabla_{y} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)^{T}=\boldsymbol{w}^{* T} \nabla_{w y}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$. Similarly, $\nabla_{y} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)^{T}=\boldsymbol{w}^{* T} \nabla_{w y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$. Thus if $\nabla_{y w}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\nabla_{y w}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$, then $\nabla_{w y}^{2} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\nabla_{w y}^{2} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ and $\nabla_{y} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=$ $\nabla_{y} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$. This last equality implies that $C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ since Euler's Theorem implies that $C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\boldsymbol{y}^{* T} \nabla_{y} C\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$ and $C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)=\boldsymbol{y}^{* T} \nabla_{y} C^{*}\left(\boldsymbol{y}^{*}, \boldsymbol{w}^{*}\right)$.
    *47 Note that the total number of free parameters that $C(\boldsymbol{y}, \boldsymbol{w})$ must have in order to be a flexible functional form for a joint cost function is $M N+(1 / 2) M(M-1)+(1 / 2) N(N-1)=(M+N)(M+N-1) / 2$. This is the same number of free parameters that is required in order for $f\left(y_{2}, \ldots, y_{M}, x_{1}, \ldots, x_{N}\right)$ to be a flexible functional form for a constant returns to scale production function.

[^25]:    *48 This functional form is analogous to the Normalized Quadratic Value Added Function $\Pi(\boldsymbol{p}, \boldsymbol{x})$ that was defined in Diewert and Fox (2021)[20].

[^26]:    *49 See Diewert (2022)[17].

[^27]:    *50 The only difference is that we use the Cost Constrained Gross Output Function in place of the Cost Constrained Value Added Function.
    ${ }^{* 51}$ This function is a variant of Diewert's $(1983 ; 1086)[15]$ balance of trade restricted value added function.

[^28]:    *52 Use the fact that $\left(\boldsymbol{y}^{s}, \boldsymbol{x}^{s}\right)$ is a feasible solution for the year $t$ maximization problem defined by (C1).

[^29]:    *53 These choices of the reference vectors will make our decomposition of gross output growth an exact one. Usually, these input mix growth factors are close to one for all periods.
    *54 These measures of technical progress measures were defined by Diewert and Morrison (1986; 662)[22] using the country's GDP function. A special case of the family was defined earlier by Diewert (1983; 1063)[15]. Balk (1998; 99)[2] also used this definition and Balk (1998; 58)[2], following the example of Salter (1960)[52], used the joint cost function to define a similar family of technical progress indexes.
    *55 In our case where the reference technology is subject to constant returns to scale, $\tau_{L}^{t}$ turns out to be independent of $\boldsymbol{x}^{t}$ and $\tau_{P}^{t}$ turns out to be independent of $\boldsymbol{x}^{t-1}$. These "mixed" indexes of technical progress turn out to be true Laspeyres and Paasche type indexes.

[^30]:    *56 There is another problem with our joint cost function methodology: it assumes optimizing behavior on the part of producers, which is not consistent with observed outputs and inputs being in the interior of the production possibilities set. The Diewert and Fox nonparametric methodology allows for nonoptimizing behavior on the part of producers but it has the disadvantage that the year $t$ reference technology that their methodology utilizes may not be a very close approximation to the actual year $t$ production possibilities set.

[^31]:    *57 This line of research was initiated by Kohli (1981b)[35] (1983b)[38] (1991)[40].
    *58 This follows from the Inverse Function Theorem in advanced calculus; see Rudin (1953; 177)[51].

[^32]:    *59 Kohli (1978)[33] (1982)[36] (1991)[40] (1993a)[41] also worked extensively with this model of production.
    ${ }^{* 60}$ In order to define the US GDP function for 2022, the $\boldsymbol{p}$ and $\boldsymbol{y}$ in definition (2) are replaced by $\left(p_{1}^{2022 f}, \ldots, p_{4}^{2022 f}, w_{1}^{2022}\right)$ and $\left(y_{1}^{2022}, \ldots, y_{4}^{2022},-x_{1}^{2022 f}\right)$, the $\boldsymbol{w}$ and $\boldsymbol{x}$ in definition (2) are replaced by $\left[x_{2}^{2022 f}, \ldots, x_{6}^{2022 f}\right] \equiv \boldsymbol{x}^{\circ}$ and $\left[w_{2}^{2022}, \ldots, w_{6}^{2022}\right] \equiv \boldsymbol{w}^{\circ}$ and $S$ is the year 2022 production possibilities set for the US economy. The cost function that is used in this Appendix is the estimated US Joint Cost Function for the US for the year 2022, $C(\boldsymbol{y}, \boldsymbol{w}) \equiv C^{t}(\boldsymbol{y}, \boldsymbol{w})$ where $t=2022$. The partial derivatives of $C$ are evaluated at $(\boldsymbol{y}, \boldsymbol{w})=\left(y^{2022}, w^{2022}\right)$. When we form elasticities, we use predicted values for prices and quantities that are used to convert estimated derivatives into elasticities.
    *61 For the properties of joint cost functions and the relationship between the joint cost function and the GDP function, see the section on joint cost functions in Diewert (2022)[17].
    *62 These equations follow from equations (D4) and (D6). The price of imports $w_{1}^{\circ}$ is held constant and so equation (D5) is dropped for now.

[^33]:    *63 Apply the Inverse Function Theorem to (D7) and (D8) to get this result.

[^34]:    *64 See Samuelson (1947; 36-38)[53] (1960)[55], Diewert (1974a; 146-150)[11] and Kohli (1983b)[38].

