

FURTHER MODEL-BASED ESTIMATES OF U.S. TOTAL MANUFACTURING PRODUCTION CAPITAL AND TECHNOLOGY, FOR 1949-2005

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Measuring the Nation's Economy.



OUTLINE

I. MOTIVATION

II. HOW WE PROCEED

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IV. ESTIMATION STRATEGY

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1. INTRODUCTION

- Conventional estimates of capital and technology (k , τ) in U.S. industries are based on limited theoretical and sample information.

Examples:

- ◆ k stock estimates are based mostly on depreciation schedules
 - ◆ τ stock follows exogenous processes or to be Solow residuals
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- These estimates ignore *dynamic interactions* over business cycles among (k , τ) and other variables, useful information for estimating (k , τ)

INTRODUCTION (cont.)

- We propose and illustrate an econometric method for jointly estimating (k, τ) , which has the following advantages:
 - ◆ (k, τ) are endogenous, being determined by firms' joint decisions on investment, i , research, r , and other inputs and outputs
 - ◆ Use correlations implied by the model between
 - unobserved (k, τ) and
 - observed prices and quantities of inputs and outputsto estimate unobserved (k, τ)
 - ◆ Use all theoretical restrictions and sample information to estimate (k, τ)

II. HOW WE PROCEED

1. STEPS OF THE ESTIMATION PROCESS:

- Specify a structural dynamic economic model of an industry, including the representative firm's dynamic optimization problem
- Solve the firm's dynamic optimization problem
- Set up reduced-form system and use it to obtain ML estimates of structural parameters:
 - ◆ Set up the reduced form in a state-space format
 - ◆ Use Kalman filter to "correctly ignore" missing observations on (k, τ) in estimation
- Compute KF estimates of (k, τ) .

HOW WE PROCEED (cont.)

- Thus, we use the KF in two ways:
 - ◆ Estimate the model
 - ◆ Use the estimated model to estimate unobserved k and τ

2. COMPARISON WITH CONVENTIONAL APPROACH:

a. Conventionally:

- ◆ k is estimated based on observed i flows and assumed or estimated depreciation schedules
- ◆ τ is assumed to be an exogenous process or is determined residually as the Solow residual
- ◆ k and τ are estimated separately, not under unifying principles

HOW WE PROCEED (cont.)

a. This study:

- ◆ Uses system-wide identifying restrictions, involving all variables in the model, to jointly estimate (k, τ) stocks
- ◆ Two identification conditions:
 - Parameter identification condition of determining unique values of the model's structural parameters, which requires a concave likelihood function at MLE
 - Filtering identification condition of estimating unobserved variables, using the estimated model and the data, which requires a *reconstructible* state vector

III. STRUCTURAL ECONOMIC MODEL

1. GENERAL DESCRIPTION:

- View a competitive industry in terms of a representative firm
- A sophisticated output supply side: explicitly describe and solve the representative firm's dynamic optimization problem
 - ◆ Firm uses (k, ℓ, m) to produce output, investment and research (q, i, r) , as 3 "outputs" in a production function with:
 - CET = constant elasticity of output transformation
 - CES = constant elasticity of input substitution
 - ◆ Concave-to-origin transformation curves for (q, i, r) imply convex adjustment costs on (k, τ) through (i, r) .
- A simple output demand side: static demand curve

Resulting industry equilibrium: a dynamic, stochastic, simultaneous-equations system

STRUCTURAL ECONOMIC MODEL (cont.)

2. VARIABLES IN THE MODEL:

● Endogenous Variables:

p_q = Price of output

l = Labor input

i = Investment-in-capital output

k = Capital input

q = Produced and sold output

m = Materials input

r = Research-in-technology output

τ = Technology input

● Exogenous Variables:

P_i = Price of investment

p_l = Price of labor

p_r = Price of research

p_m = Price of materials

● Disturbances:

ξ = output-demand state

ε 's = structural disturbances

Note: k , τ , ξ , and ε 's are unobserved; all variables in real terms

STRUCTURAL ECONOMIC MODEL (cont.)

3. FIRM'S DYNAMIC OPTIMIZATION PROBLEM

a. CET/CES production function:

$$[\gamma_1 \cdot q^\rho + \gamma_2 \cdot i^\rho + \gamma_3 \cdot r^\rho]^{1/\rho} = \tau \cdot [\alpha_1 \cdot k^\beta + \alpha_2 \cdot \ell^\beta + \alpha_3 \cdot m^\beta]^{1/\beta},$$

where $\gamma_i \geq 0$, $\sum_i \gamma_i = 1$, $\rho > 1$, $\alpha_i \geq 0$, $\sum_i \alpha_i = 1$, $\beta < 1$.

◆ $\rho > 1 \Rightarrow \text{CET} = (\rho - 1)^{-1} > 0 \Rightarrow$ concave-to-origin output transformation curves

◆ $\beta < 1 \Rightarrow \text{CES} = (\beta - 1)^{-1} < 0 \Rightarrow$ convex-to-origin input isoquants

STRUCTURAL ECONOMIC MODEL (cont.)

b. Restricted variable cost function:

- Dual description of the production function:

$$C(\mathbf{z}) = \min_{\{\ell, m\}} \mathbf{p}_\ell \cdot \ell + \mathbf{p}_m \cdot m,$$

subject to the CET/CES production function, for given

$$\mathbf{z} = (\mathbf{q}, \mathbf{i}, \mathbf{r}, \mathbf{k}, \tau, \mathbf{p}_\ell, \mathbf{p}_m)^\top.$$

- Use quadratic approximation $C(\mathbf{z}) \cong (1/2)\mathbf{z}^\top \cdot \nabla^2 C_0 \cdot \mathbf{z}$, where

$$\nabla^2 C_0 = \begin{bmatrix} c_{11} & \cdots & c_{17} \\ \vdots & & \vdots \\ c_{71} & \cdots & c_{77} \end{bmatrix} = \text{Hessian of } C(\mathbf{z}) \text{ at } \mathbf{z} = \mathbf{z}_0.$$

STRUCTURAL ECONOMIC MODEL (cont.)

c. Output inverse demand curve:

$$p_{qt} = -\eta \cdot q_t + d_t + \xi_t,$$

where $\eta > 0$ is the slope, $d_t \sim \text{AR}(2)$ is demand state .

d. Representative firm's REAL PROFIT FUNCTION:

$$\pi_t = -(1/2)\eta \cdot (q_t)^2 + q_t \cdot \xi_t - (1/2)(z_t)^T \cdot \nabla^2 C_0 \cdot z_t - p_{it} \cdot i_t - p_{rt} \cdot r_t.$$

e. Firm maximizes its EXPECTED PRESENT VALUE:

$$\max_{\{p_{qt}, q_t, \ell_t, m_t, i_t, r_t\}} V_t = E_t \sum_{k=0}^{\infty} \delta^k \pi_t + k_t,$$

for predetermined (k_t, τ_t) and exogenous $(p_{it}, p_{rt}, p_{\ell t}, p_{mt}, \xi_t)$ processes.

STRUCTURAL ECONOMIC MODEL (cont.)

4. SOLVE FIRM'S DYNAMIC OPTIMIZATION PROBLEM:

- Eliminate (p_{qt}, q_t, l_t, m_t) so that the problem reduces to:

$$\max_{\{K\}} V_t = E_t \sum_{k=0}^{\infty} \delta^k \pi_{t+k},$$

subject to the linear decision rule

$$u_t = K \cdot x_{t-1},$$

and to exogenous output-demand and input-price processes, where

$$u_t = (i_t, r_t)^T = \text{control vector},$$

$$x_t = (k_t, \tau_t, p_{it}, p_{rt}, p_{lt}, p_{mt}, \xi_t)^T = \text{state vector}.$$

- Compute K by solving an algebraic matrix Riccati equation
- Figure 2 illustrates optimal responses to a rise in output demand

IV. ESTIMATION STRATEGY

1. Assemble 13 equations in 8 endogenous and 5 exogenous variables from model solution as dynamic simultaneous equations system.

• 8 endogenous variable equations:

◆ Output inverse demand:

$$p_{qt} = -\eta \cdot q_t + \xi_t$$

◆ Output, labor, and materials decision rules:

$$q_t = -c_0 [c_{12} \cdot i_t + c_{13} \cdot r_t + c_{14} \cdot k_t + c_{15} \cdot \tau_t + c_{16} \cdot p_{\ell t} + c_{17} \cdot p_{mt} - \xi_t] + \varepsilon_{qt}$$

$$l_t = c_{16} \cdot q_t + c_{26} \cdot i_t + c_{36} \cdot r_t + c_{46} \cdot k_t + c_{56} \cdot \tau_t + c_{66} \cdot p_{\ell t} + c_{67} \cdot p_{mt} + \varepsilon_{\ell t}$$

$$m_t = c_{17} \cdot q_t + c_{27} \cdot i_t + c_{37} \cdot r_t + c_{47} \cdot k_t + c_{57} \cdot \tau_t + c_{67} \cdot p_{\ell t} + c_{77} \cdot p_{mt} + \varepsilon_{mt}$$

ESTIMATION STRATEGY (cont.)

◆ Investment and research decision rules:

$$i_t = K_{11} \cdot k_{t-1} + K_{12} \cdot \tau_{t-1} + K_{13} \cdot p_{i,t-1} + K_{14} \cdot p_{r,t-1} + K_{15} \cdot p_{l,t-1} + K_{16} \cdot p_{m,t-1} \\ + K_{17} \cdot \xi_{t-1} + \varepsilon_{it}$$

$$r_t = K_{21} \cdot k_{t-1} + K_{22} \cdot \tau_{t-1} + K_{23} \cdot p_{i,t-1} + K_{24} \cdot p_{r,t-1} + K_{25} \cdot p_{l,t-1} + K_{26} \cdot p_{m,t-1} \\ + K_{27} \cdot \xi_{t-1} + \varepsilon_{rt}$$

◆ Capital and technology accumulation:

$$k_t = \phi_k \cdot k_{t-1} + i_t + \varepsilon_{kt}$$

$$\tau_t = \phi_r \cdot \tau_{t-1} + r_t + \varepsilon_{\tau t}$$

ESTIMATION STRATEGY (cont.)

- 5 Exogenous AR(2) processes:

- ◆ Output-demand state:

$$\xi_t = \phi_{q1} \cdot \xi_{t-1} + \phi_{q2} \cdot \xi_{t-2} + \varepsilon_{qt}$$

- ◆ Input prices:

$$p_{it} = \phi_{i1} \cdot p_{i,t-1} + \phi_{i2} \cdot p_{i,t-2} + \varepsilon_{it}$$

$$p_{rt} = \phi_{r1} \cdot p_{r,t-1} + \phi_{r2} \cdot p_{r,t-2} + \varepsilon_{rt}$$

$$p_{\ell t} = \phi_{\ell 1} \cdot p_{\ell,t-1} + \phi_{\ell 2} \cdot p_{\ell,t-2} + \varepsilon_{\ell t}$$

$$p_{mt} = \phi_{m1} \cdot p_{m,t-1} + \phi_{m2} \cdot p_{m,t-2} + \varepsilon_{mt}$$

ESTIMATION STRATEGY (cont.)

2. Add a VAR(1) correction for each autocorrelated disturbance and restate the 13 equations as a VAR(2):

$$y_t = B_1 \cdot y_{t-1} + B_2 \cdot y_{t-2} + \zeta_t,$$

$$y = (p_q, q, l, m, i, r, k, \tau, p_i, p_r, p_l, p_m, \xi)^T.$$

= 13×1 vector of all variables in the model,

and

$$\zeta_t = (\zeta_{pqt}, \zeta_{qt}, \zeta_{lt}, \zeta_{mt}, \zeta_{it}, \zeta_{rt}, \zeta_{kt}, \zeta_{\tau t}, \zeta_{p_{it}}, \zeta_{p_{rt}}, \zeta_{p_{lt}}, \zeta_{p_{mt}}, \zeta_{dt})$$

= 13x1 vector of disturbances (6 AR(1) and 7 iid).

Use KF to "correctly ignore" missing observations on (k, τ) during MLE.

V. ESTIMATION RESULTS

- Parameters to be estimated using MLE:

$$\vartheta = (\vartheta_1, \vartheta_2)$$

Where

$$\vartheta_1 = (\delta, \alpha_1, \alpha_2, \gamma_1, \gamma_2, \sigma_{pq}^2, \sigma_{\ell}^2, \sigma_m^2)$$

$$\vartheta_2 = (\phi_{pi,1}, \phi_{pr,1}, \phi_{p\ell,1}, \phi_{pm,1}, \phi_{pi,2}, \phi_{pr,2}, \phi_{p\ell,2}, \phi_{pm,2}, \sigma_{pi}^2, \sigma_{pr}^2, \sigma_{p\ell}^2, \sigma_{pm}^2,$$

$$\theta_{pq}, \theta_q, \theta_{\ell}, \theta_m, \theta_i, \theta_r, \eta, \beta, \rho, \phi_{k1}, \phi_{\tau 1}, \phi_{q1}, \phi_{q2}, \sigma_q^2, \sigma_i^2, \sigma_r^2, \sigma_k^2, \sigma_{\tau}^2, \sigma_d^2)$$

Data

- **U.S. total manufacturing data on prices and quantities of output and inputs**
 - **Investment data and GDP deflator from BEA**
 - **Research data from NSF**
 - **All other variables from BLS**
- **Sample: 1949 – 2005**
 - **SIC DATA: 1949 – 2001**
 - **NAICS DATA: 1987 – 2005**
 - **Data from 1987 – 2001 are derived using average growth rates of the two series.**

Figure 1: U.S. manufacturing prices and quantities of output and inputs, (1947-2003)

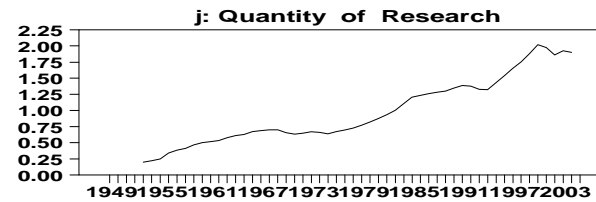
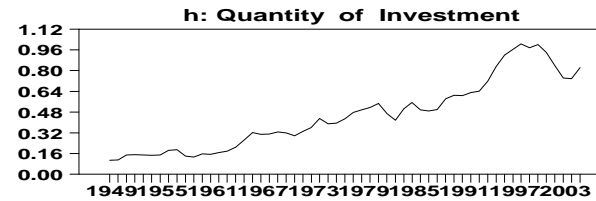
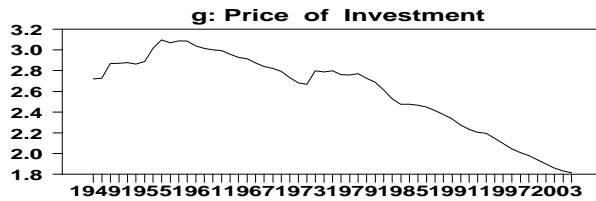
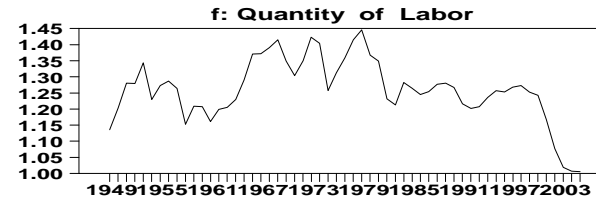
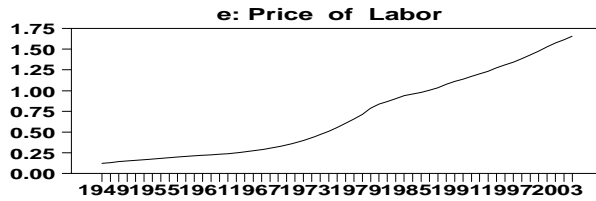
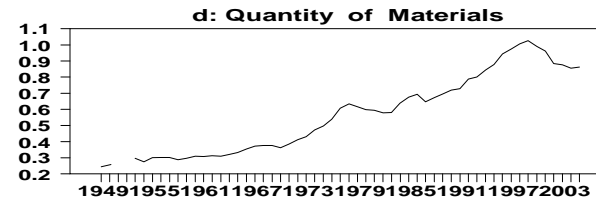
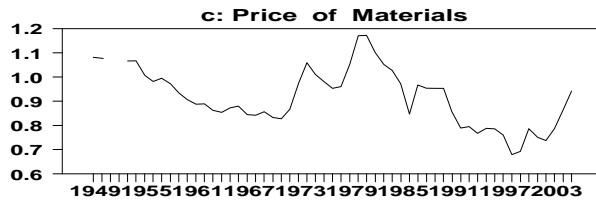
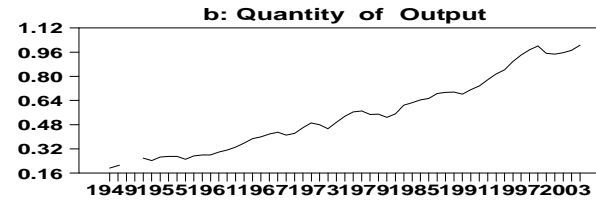
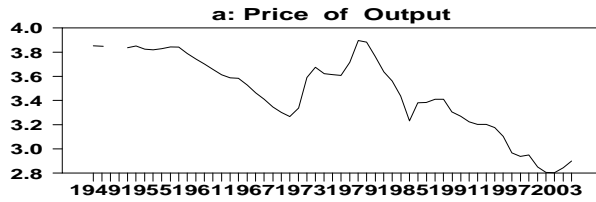


Table 1: OLS Estimates of Input-Price Process Parameters

Variables	Parameter Estimates			Fit Statistics	
	$\hat{\phi}_{.,1}$	$\hat{\phi}_{.,2}$	$ \bar{\lambda} $	R^2	Q
p_i	1.37 (10.5)	-.356 (2.61)	1.02	.988	6.76 (.563)
p_r	1.90 (28.2)	-.900 (13.1)	.999	.999	17.6 (.124)
p_ℓ	1.88 (25.6)	-.874 (11.7)	1.03	.999	13.3 (.101)
p_m	1.16 (8.34)	-.319 (2.32)	.712	.812	4.40 (.820)

Table 2: ML Estimates of Remaining Structural Parameters

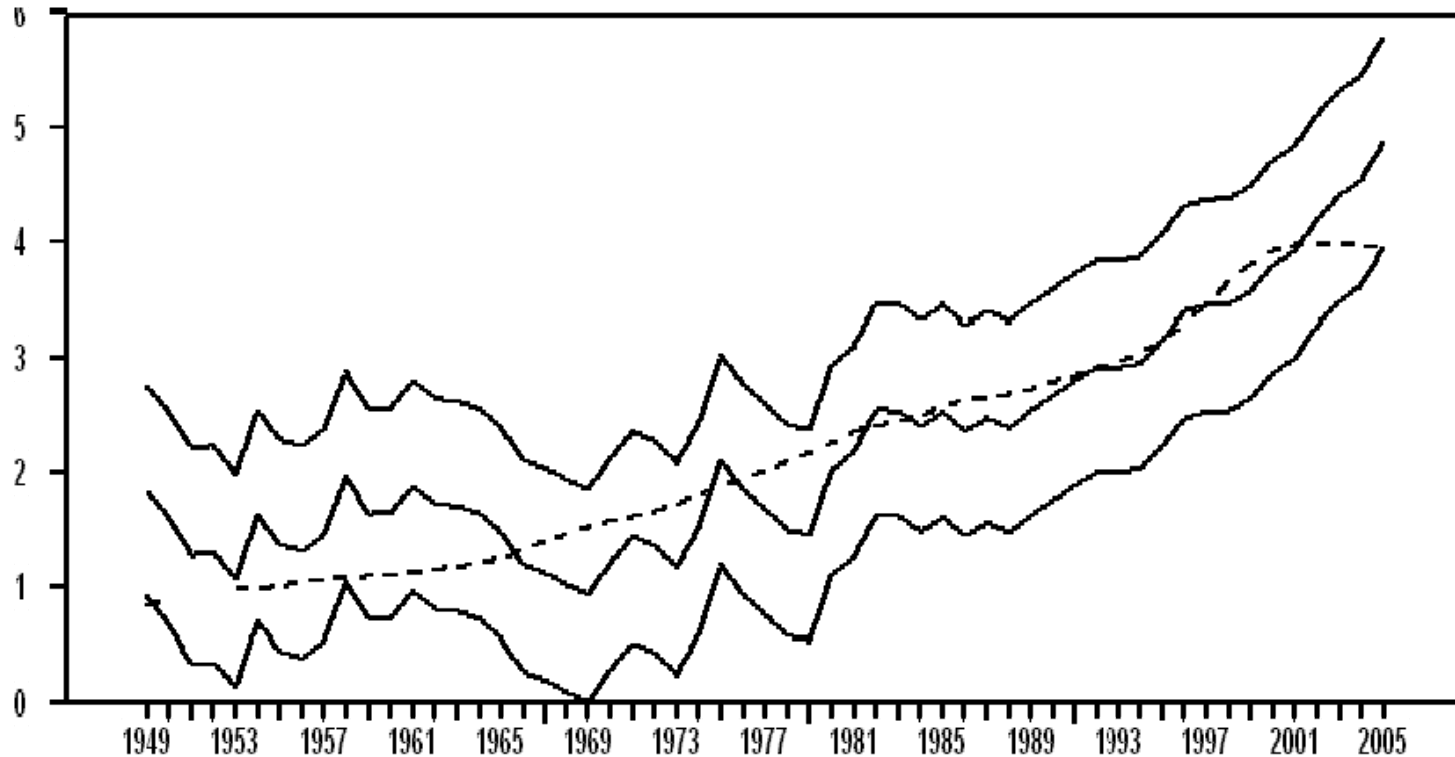
<p align="center">Production Function Parameters</p> <p align="center">$\hat{\beta} = -5.48$ (CES = -.182), $\hat{\rho} = 275$ (CET = .004)</p>
<p align="center">Output-Demand Curve Parameters</p> <p align="center">$\hat{\eta} = .869$, $\hat{\phi}_{d1} = 1.18$, $\hat{\phi}_{d2} = -.367$</p>
<p align="center">Capital and Technology Equation Coefficients</p> <p align="center">$\hat{\phi}_{k1} = .610$, $\hat{\phi}_{i0} = .789$, $\hat{\phi}_{\tau1} = .043$, $\hat{\phi}_{r0} = .304$</p>
<p align="center">Residual Autocorrelation Coefficients</p> <p align="center">$\hat{\theta}_{pq} = .999$, $\hat{\theta}_q = .675$, $\hat{\theta}_\ell = .999$, $\hat{\theta}_m = .999$, $\hat{\theta}_i = .848$, $\hat{\theta}_r = .982$</p>
<p align="center">Structural Disturbance Standard Deviations</p> <p align="center">$\hat{\sigma}_q = .144$, $\hat{\sigma}_i = .246$, $\hat{\sigma}_r = .106$, $\hat{\sigma}_k = .995$, $\hat{\sigma}_\tau = .001$, $\hat{\sigma}_d = .207$</p>
<p align="center">Reduced-Form Equation Fit Statistics</p> <p align="center">$R_{pq}^2 = .932$, $R_q^2 = .942$, $R_\ell^2 = .651$, $R_i^2 = .938$, $R_r^2 = .990$</p> <p align="center">$Q_{pq} = 2.52$, $Q_q = 2.55$, $Q_\ell = 13.1$, $Q_i = 13.6$, $Q_r = 12.9$</p> <p align="center">(.989) (.990) (.218) (.194) (.230)</p>

Table 3: Structural Variance Decomposition of the Estimated Model

	σ_q^2	σ_i^2	σ_r^2	σ_k^2	σ_τ^2	σ_{pi}^2	σ_{pr}^2	σ_{pl}^2	σ_{pm}^2	σ_d^2
$S_{10,pq,j}$	4.5	2.8	.7	.2	.0	5.2	.1	.0	3.0	83.5
$S_{10,q,j}$	19.4	12.2	3.1	.8	.2	27.5	.7	.0	15.9	20.2
$S_{10,l,j}$.9	3.9	.0	92.7	.2	.0	.0	.0	1.6	.1
$S_{10,m,j}$.9	3.9	.0	92.7	.8	.0	.0	.0	1.6	.1
$S_{10,i,j}$.0	44.5	.1	14.3	.1	17.5	.4	.0	11.5	11.6
$S_{10,r,j}$.0	.0	5.4	1.1	.2	39.3	1.0	.1	25.8	27.1
$S_{10,k,j}$.0	4.0	.0	95.3	.0	.3	.0	.0	.2	.2
$S_{10,\tau,j}$.0	.0	1.9	1.1	1.6	39.9	1.1	.1	26.5	27.8
$\bar{S}_{10,j}$	1.3	5.2	.7	69.6	.4	7.4	.2	.0	5.4	9.8

Figure 2-a: Model-Based and Standard Capital Stock Estimates for U.S. Total Manufacturing, 1949-2005

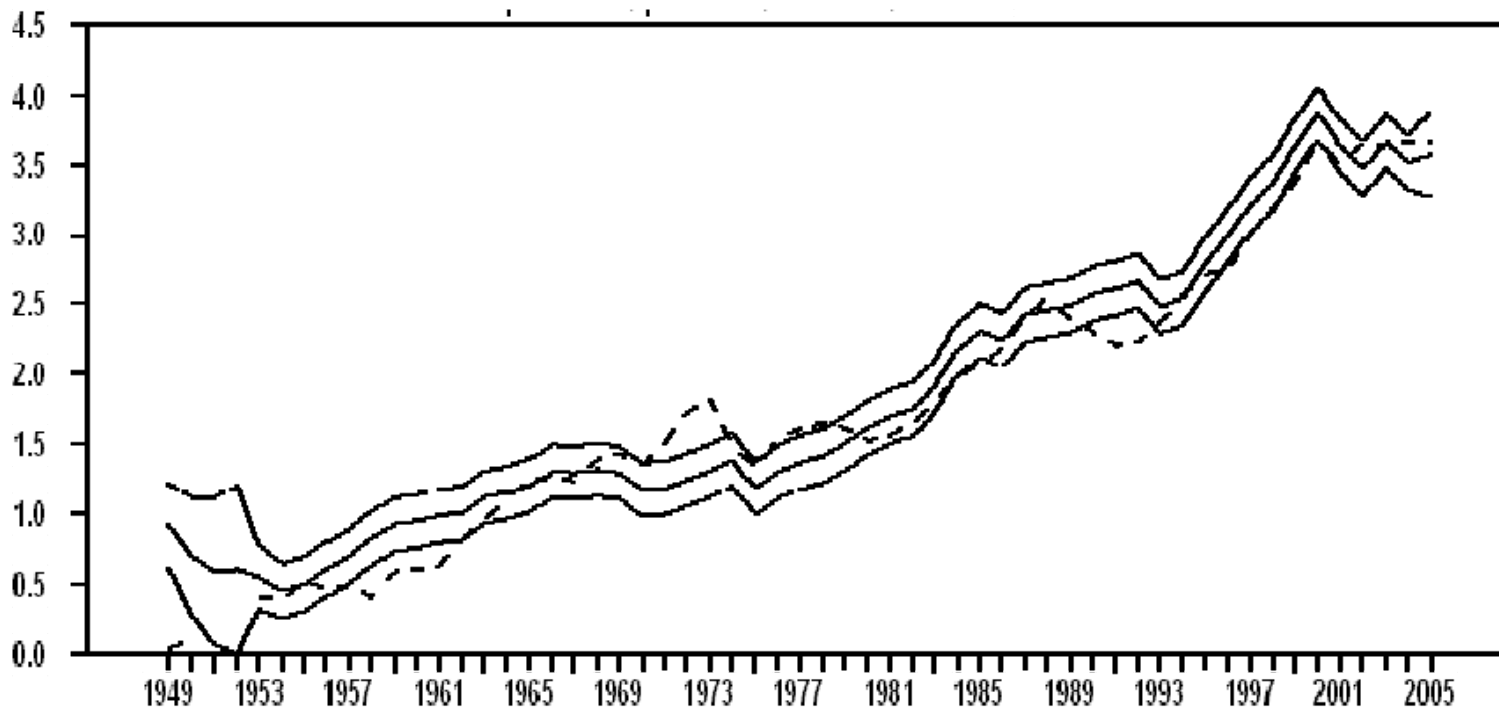
Model-Based and Standard Estimates of Capital Stock



Solid lines depict model-based capital estimates and 2-standard-error confidence bounds produced by the Kalman filter. Dashed lines depict standard capital stock estimates produced by BLS.

Figure 2-b: Model-Based and Standard Technology (TFP) Estimates for U.S. Total Manufacturing, 1949-2005

Model-Based and Standard Estimates of Technology



Solid lines depict model-based technology estimates and 2-standard-error confidence bounds produced by the Kalman filter. Dashed lines depict standard technology estimates as total factor productivity produced by BLS.

Conventional versus Filtered Estimates

1. 1949 ~ 2000

- Similar trends of model based and standard k and τ

2. 2000 ~ 2005

- Diverge paths of model-based and standard k τ
 - Model-based k continues to grow; standard k levels off
 - Model-based τ declines and levels off; standard τ continue to grow

3. 1949 ~ 2005

- Model-based k is noisy and uncertain; standard k is smooth
- Model-based τ is smooth and certain; standard τ is more noisy
- (i, k) explain growth more than (r, τ)

VI. CONCLUSION

- Proposed filtered estimates of (k, τ) are feasible:
 - ◆ Results show two identification conditions of parameter identification in MLE and state reconstructibility in k and τ estimation hold numerically.
- Proposed method uses more information:
 - ◆ Use more sample information of correlations among more observed variables;
 - ◆ Uses theoretical information of correlations among all variables in the model, as implied by the model
- Good overall fit gives (k, τ) estimates credibility
- Extension: Includes time varying variables like interest rates and taxes