

# **Divergent Transformation Paths: An Anatomy of the Baumol Cost Disease**

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# Motivation

## Baumol's Cost Disease

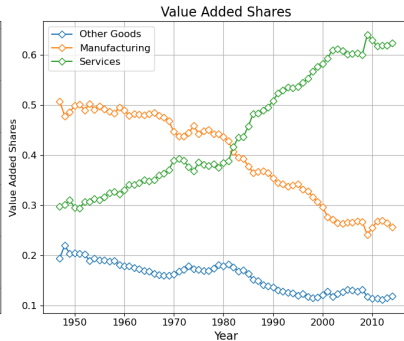
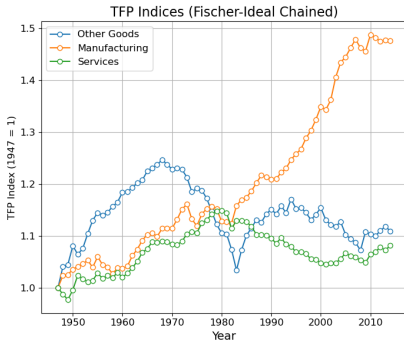
*...the very progress of the technologically progressive sectors inevitably adds to the costs of the technologically unchanging sectors of the economy...*

**William Baumol** *Macroeconomics of Unbalanced Growth: The Anatomy of Urban Crisis* - AER, 1967

# Motivation

## Sectoral TFP Growth and Structural Change in the US

- ▶ Structural Transformation  $\leftrightarrow$  Productivity Slowdown
- ▶ Reallocation of economic activity
  - ▶ from dynamic ...
  - ▶ ...into less-dynamic sectors

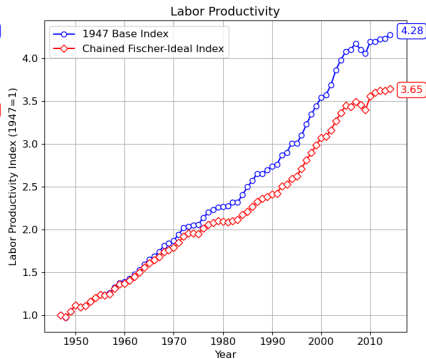
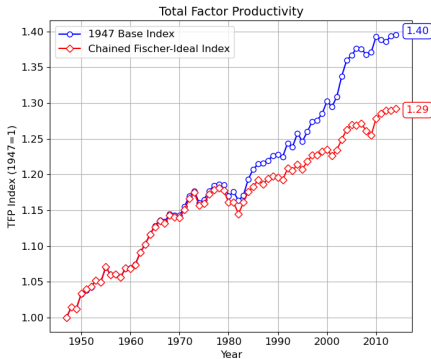


Source: Own calculations based on US KLEMS, March 2017 Release

# Motivation

## Main Implication

- Lower measured TFP growth due to compositional effects



Source: Own calculations based on US KLEMS, March 2017 Release

# Motivation

## Structural Change, Innovation, and Growth

- ▶ Sectoral productivity growth is not fixed
- ▶ The state and structure of the economy also affect
  - ▶ Incentives to innovate
  - ▶ Allocation of R&D resources
- ▶ Shaping sectoral and aggregate productivity growth
- ▶ This paper studies:
  - ▶ The **determinants** of the sectoral allocation of R&D
  - ▶ How they vary with structural change and development
  - ▶ Implications for sectoral and aggregate TFP growth

# Introduction

## Contribution and Findings

1. New Evidence on Innovation and Struct. Transformation (ST)
  - ▶ ST in **innovation** from manufacturing to services
  - ▶ Divergent Transformation Paths
    - ▶ Research activity moves to **high R&D** services
    - ▶ Production shifts to **low R&D** services
2. We build quantitative theory of ST and Innovation and study
  - ▶ Mechanisms driving the sectoral allocation of RD
  - ▶ Implications for sectoral and aggregate productivity growth
  - ▶ Potential role for policy interventions
3. Quantitative Findings; (US 1947-2014)
  - ▶ ST contributed to TFP slowdown (-2.0 p.p; vs. 11.0 p.p.)
  - ▶ Similar quant. effect as pop. growth slowdown (-2.8 p.p.)
  - ▶ Market power and markups hindered TFP growth (-7.8 p.p.)
  - ▶ Cross-sector spillovers helped sustain TFP growth (+10.2 p.p.)

# Plan

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# Related Literature

## 1. Drivers of Secular Stagnation

- ▶ **Demographics** Abiry et al. (2016); Eggertsson et al. (2019)
- ▶ **Productivity growth slowdown** Bergeaud et al. (2018); Platzer & Peruffo (2022); Liu et al. (2022)

**We study the role of SC in driving growth slowdown**

## 2. Causes & Consequences of Structural Change

- ▶ **Documenting structural change** Stigler et al. (1956); Kuznets, (1957); Fuchs et al. (1968); Herrendorf et al. (2014)
- ▶ **Causes of structural change** Ngai & Pissarides (2007); Kongsamut et al. (2001); Boppart (2014); Comin et al. (2021)
- ▶ **Heterogeneous services** Buera & Kaboski (2012); Duernecker et al. (2024)

**i) document SC in innovation ii) endogenise sectoral prod. growth**



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# Data

We merge several data sources together:

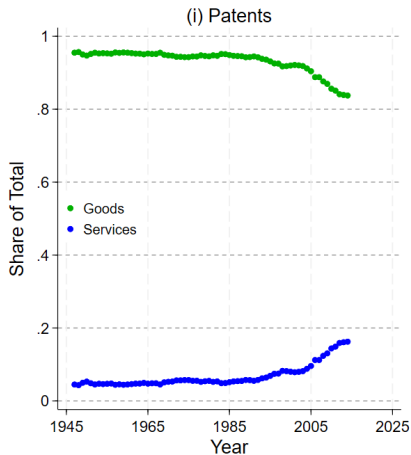
Variable	Source
Scientists	National Science Foundation (BERD survey)
Patents <sup>1</sup>	US Patent & Trademark Office (PTO)
Cross-citations	OECD Citations Database
VA, TFP, Lab. Prod.	World KLEMS (March 2017 release)

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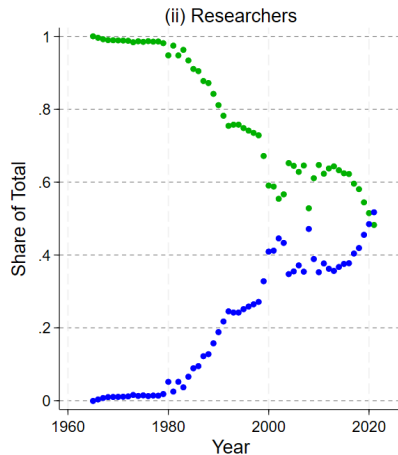
<sup>1</sup>USPC-industry crosswalk from (Goldschlag, et al., 2019)

# Structural Change in Innovation

From Manufacturing/Goods to Services



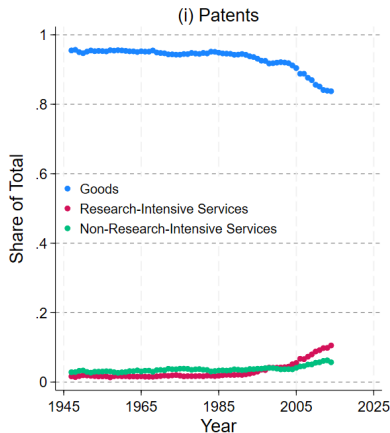
Data: Marco et al. (2015), US Patent and Trademark Office



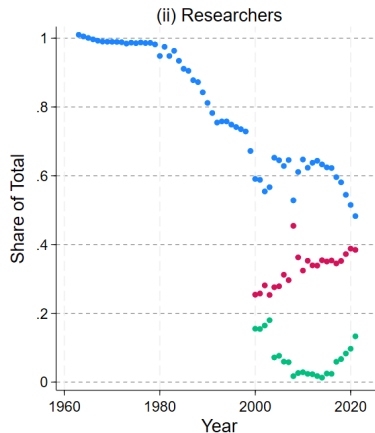
Data: US National Science Foundation

# Structural Change in Innovation

More Prominent Towards High R&D Services



Data: Marco et al. (2015), US Patent and Trademark Office

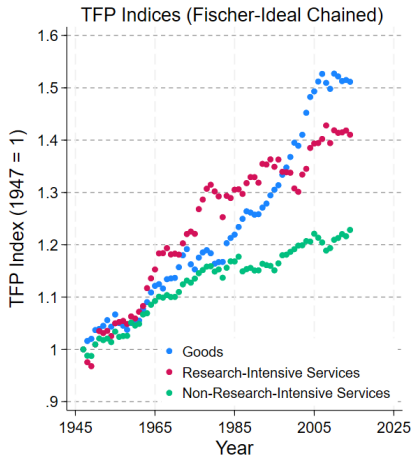


Data: US National Science Foundation

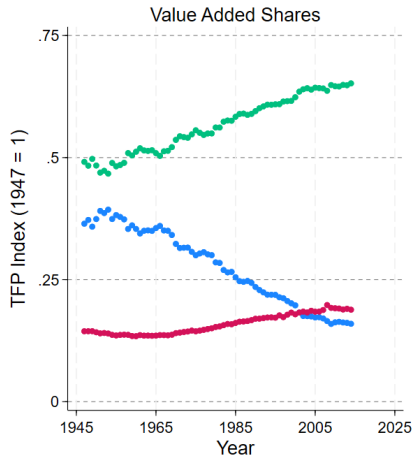
► Services Split

► OECD

# Driving Heterogeneous Productivity Paths



Data: Jorgenson, et al., 2017



Data: Jorgenson, et al., 2017

# Rich Structure of Cross-Sector Knowledge Spillovers

Manufacturing ideas have a relevant effect on services

		Citing Setor		
		MFG	RI-SERV	NRI-SERV
Cited Sector	MFG	0.93	0.62	0.82
	RI-SERV	0.03	0.32	0.05
	NRI-SERV	0.03	0.06	0.13

► Robustness

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# Key Model Features

- ▶ Traditional representative household structure
- ▶ Standard ST model via:
  1. relative price effects
  2. income effects
- ▶ Endogenous sector-level prod. growth à la Acemoglu (2002)
- ▶ With innovation incentives shaped by
  1. price effect
  2. market size effect
  3. knowledge spillovers
  4. marginal costs/markups



# Household Preferences & Demographic Structure

- ▶ Continuum of homogeneous workers of mass  $L(t)$
- ▶ Labor force grows at (time-variant) rate  $n(t)$
- ▶ Workers can work in:
  - ▶ The production sector ( $L^P(t)$ ), making wages  $w_l(t)$
  - ▶ The R&D sector ( $L^S(t)$ ), making wages  $w_s(t)$
  - ▶  $L^P(t) + L^S(t) = L(t)$
- ▶ In per-capita terms:  $s(t) = \frac{L^S(t)}{L(t)}$ ,  $l(t) = \frac{L^P(t)}{L(t)}$
- ▶ Standard CRRA preferences ( $\chi$ ), given by

$$U(c(t), L(t)) = \int_0^\infty e^{-\rho t} L(t) \left( \frac{c(t)^{1-\chi} - 1}{1-\chi} \right) dt$$

# Production Technologies

- ▶ Three-sector economy
- ▶ Nested non-homothetic CES structure
- ▶ Final good: composite of  $Y_m(t)$  and  $Y_s(t)$  output

$$\left[ \theta_m^{\frac{1}{\sigma}} (Y(t)^{-\zeta_m} Y_m(t))^{\left(\frac{\sigma-1}{\sigma}\right)} + \theta_s^{\frac{1}{\sigma}} (Y(t)^{-\zeta_s} Y_s(t))^{\left(\frac{\sigma-1}{\sigma}\right)} \right]^{\left(\frac{\sigma}{\sigma-1}\right)} = 1$$

- ▶ Services: composite of  $Y_{rs}(t)$  and  $Y_{ns}(t)$  output

$$\left[ \psi_{rs}^{\frac{1}{\epsilon}} (Y_s(t)^{-\zeta_{rs}} Y_{rs}(t))^{\left(\frac{\epsilon-1}{\epsilon}\right)} + \psi_{ns}^{\frac{1}{\epsilon}} (Y_s(t)^{-\zeta_{ns}} Y_{ns}(t))^{\left(\frac{\epsilon-1}{\epsilon}\right)} \right]^{\left(\frac{\epsilon}{\epsilon-1}\right)} = 1$$

# Sectoral Technologies

- Sectoral output produced according to

$$Y_j(t) = \left( \frac{1}{1-\beta} \right) X_j(t)^{1-\beta} (B_j(t)L_j^P(t))^\beta \quad \text{for } j \in \{m, rs, ns\}$$

- With
  - $L_j^P$ : production labor in  $j$
  - $B_j$ : labor-augmenting technology in  $j$
  - $X_j$ : aggregate machine input in  $j$
- $X_j$  is defined by the CES aggregator

$$X_j(t) = \left( \int_0^{N_j(t)} x_j(\nu, t)^{\mu_j} d\nu \right)^{(1/\mu_j)}$$

- $x_j(\nu, t)$  is the quantity of machines of type  $\nu$  in time  $t$
- $N_j(t)$  is the measure of machine varieties

# Machine Producers' Technologies

- ▶ Each variety  $\nu$  is produced by an intermediate goods producer
- ▶ Technology given by

$$x_j(\nu, t) = a_j(t)y_j^x(\nu, t)$$

- ▶  $y_j^x(\nu, t)$  are units of the final good used in  $x_j(\nu, t)$
- ▶  $a_j(t)$  is the sector-specific productivity of machine producers

# Research & Development (R&D)

- ▶ Directed innovation
  - ▶ Large number of research labs
  - ▶ Create *new* machine variety blueprints
  - ▶ R&D efforts directed to a specific sector
  - ▶ Successful innovators obtain fully-enforced perpetual patents
  - ▶ Machines fully depreciate after one period
  - ▶ Patents can be sold to machine produces at price  $V_j(t)$
- ▶ Aggregate research efforts yield the following flow of ideas

$$\dot{N}_j(t) = \eta_j \Lambda_j(t)^\gamma L_j^S(t)^{(1-\alpha)}$$

- ▶  $\eta_j$  exogenous arrival rate of ideas in  $j$
- ▶  $L^S(t)$ : R&D workers in  $j$
- ▶  $\Lambda_j(t) = N_j(t)^{\delta_{jj}} N_i(t)^{\delta_{ji}} N_k(t)^{\delta_{jk}}$  ideas aggregator

# Market Value of Patents

- ▶ Value of owning a patent of type  $\nu$  in sector  $j$  is

$$V_j(\nu, t) = \int_{s=t}^{\infty} e^{-\int_{s'=0}^s r(s') ds'} \pi_j(\nu, s) ds$$

- ▶ The corresponding Hamilton-Jacobi-Bellman equation is

$$r(t)V_j(\nu, t) = \pi_j(\nu, t) + \dot{V}_j(\nu, t)$$

- ▶  $\pi_j(\nu, t)$ : profits for a machine variety producer
- ▶  $r(t)$ : real interest rate

# Equilibrium Characterization

## Sectoral Output Demand

- Sectoral output demands are

$$Y_m(t) = \frac{e_m(t)Y(t)}{p_m(t)}$$

$$Y_k(t) = (1 - e_m(t)) \frac{e_k(t)Y(t)}{p_k(t)} \quad k \in \{rs, ns\}$$

- With sectoral output shares given by

$$e_m(t) = \theta_m Y(t)^{(\zeta_m - 1)(1 - \sigma)} p_m(t)^{1 - \sigma}$$

$$e_k(t) = \theta_k \left( \frac{p_k(t)}{p_s(t)} \right)^{1 - \epsilon} Y_s(t)^{(\zeta_k - 1)(1 - \epsilon)}$$

# Equilibrium Characterization

## Markups, Profits & Output

- ▶ Machine producers in each sector face (inverse) demand

$$p_j^x(\nu, t) = \left( \frac{p_j(t) Y_j(t)}{X_j(t)} \right) \left( \frac{X_j(t)}{x_j(\nu, t)} \right)^{(1-\mu_j)}$$

- ▶ Optimal pricing of machine producers in each sector

$$p_j^x(\nu, t) = \frac{1}{\mu_j} \frac{p(t)}{a_j(t)}$$

- ▶ Profits for each sector's machine producers are

$$\pi_j(\nu, t) = \left( \frac{1 - \mu_j}{\mu_j} \right) \frac{p(t) x_j(\nu, t)}{a_j} \quad \text{where} \quad x_j(\nu, t) = \frac{\alpha p_j(t) Y_j(t) a_j(t)}{N_j(t)}$$

- ▶ Output is

$$Y_j(t) = C_j(\beta, \mu_j) N_j(t)^{\left(\frac{1-\beta}{\beta}\right)} \left(\frac{1-\mu_j}{\mu_j}\right) (p_j(t) a_j(t))^{\left(\frac{1-\beta}{\beta}\right)} B_j(t) L_j^P(t)$$



# Equilibrium Characterization

## Drivers of Directed Technical Change

- ▶ Profits for machine producers drive incentives to innovate
- ▶ Directed technical change (Acemoglu, 2002)

$$\pi_j^{DTC}(\nu, t) = \beta p_j(t)^{1/\beta} L_j^P(t)$$

- ▶ **sector size** effect and **relative price** effect key  $\rightarrow$  ST
- ▶ Here we get

$$\pi_j(\nu, t) = (1 - \mu_j) \left( \frac{1}{1 - \beta} \right)^{\frac{1}{\beta}} N_j(t)^{\frac{1 - \mu_j - \beta}{\mu_j \beta}} (\mu_j a_j(t))^{\frac{1 - \beta}{\beta}} \frac{B_j(t) \pi_j^{DTC}}{\beta}$$

- ▶ Heterogenous sectoral trends in machine productivity  $\rightarrow$  different **marginal costs** in each sector  $\rightarrow$  influence sector profits

# Equilibrium Characterization

## Innovation Dynamics and Wages

- Free entry into the innovation sector yields

$$\dot{N}_j(t)V_j(t) = w_s(t)L_j^S(t) \quad \text{for } j \in \{m, rs, ns\}$$

- Implying that the wage for scientists is

$$\eta_j \Lambda_j(t)^\gamma L_j^S(t)^{-\alpha} V_j(t) = w_s(t)$$

- Free labor mobility between production and innovation

$$w_s(t) = w_l(t)$$

# Equilibrium Characterization

## Sectoral Allocations for Workers and Scientists

- Employment shares equal output shares

$$\frac{L_j^P(t)}{L(t)} = e_j(t) \quad j \in \{m, rs, ns\}$$

- Sectoral R&D employment is given by

$$S_m(t) = \frac{S(t)}{\left[ 1 + \left( \frac{\eta_{rs}\Lambda_{rs}(t)^\gamma V_{rs}(t)}{\eta_m\Lambda_m(t)^\gamma V_m(t)} \right)^{1/\alpha} + \left( \frac{\eta_{ns}\Lambda_{ns}(t)^\gamma V_{ns}(t)}{\eta_m\Lambda_m(t)^\gamma V_m(t)} \right)^{1/\alpha} \right]}$$

$$S_i(t) = S_m(t) \left( \frac{\eta_i\Lambda_i(t)^\gamma V_i(t)}{\eta_m\Lambda_m(t)^\gamma V_m(t)} \right)^{1/\alpha} \quad i \in \{rs, ns\}$$

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# Externally Calibrated Parameters

Parameter	Description	Value	Source
$\beta$	labor intensity	0.65	macro literature
$\chi$	intertemporal elasticity of sub.	1.50	macro literature
$\rho$	discount factor	0.05	macro literature
$\delta_{mm}$	knowledge spillover intensity MFG to MFG	0.92	authors calc.
$\delta_{rm}$	knowledge spillover intensity MFG to RI-S	0.57	authors calc.
$\delta_{nm}$	knowledge spillover intensity MFG to NRI-S	0.72	authors calc.
$\delta_{mr}$	knowledge spillover intensity RI-S to MFG	0.04	authors calc.
$\delta_{rr}$	knowledge spillover intensity RI-S to RI-S	0.27	authors calc.
$\delta_{nr}$	knowledge spillover intensity RI-S to NRI-S	0.10	authors calc.
$\delta_{mn}$	knowledge spillover intensity NRI-S to MFG	0.05	authors calc.
$\delta_{rn}$	knowledge spillover intensity NRI-S to RI-S	0.16	authors calc.
$\delta_{nn}$	knowledge spillover intensity NRI-S to NRI-S	0.18	authors calc.
$\gamma$	spillover elasticity	0.50	robustness checks
$1 - \alpha$	scientist elasticity	0.35	robustness checks

# Structurally Estimated Parameters

Via Simulated Method of Moments

Parameter	Interpretation	Value
$\sigma$	elasticity of sub. b/w MFG and SERV	0.05
$\epsilon$	elasticity of sub. b/w RI-S and NRI-S	1.07
$\zeta_m$	MFG non-homotheticity parameter	0.51
$\zeta_{rs}$	RI-S non-homotheticity parameter	1.00
$\zeta_{ns}$	NRI-S non-homotheticity parameter	1.25
$\eta_m$	MFG arrival rate of innovation	0.036
$\eta_{rs}$	RI-S arrival rate of innovation	0.005
$\eta_{ns}$	NRI-S arrival rate of innovation	0.001
$g_{b,m}$	growth in MFG labor-augmenting technology	0.0138
$g_{b,rs}$	growth in RI-S labor-augmenting technology	0.0129
$g_{b,ns}$	growth in NRI-S labor-augmenting technology	0.0103
$\theta_m$	MFG sector weight (1947)	0.596
$\theta_{rs}$	RI-S sector weight (1947)	0.236
$\theta_{ns}$	NRI-S sector weight (1947)	0.749

► Data Moments

► Income Elasticities

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# Counterfactual TFP Growth

	Aggregate TFP Growth (2010-1947, %)	Difference w.r.t Benchmark (p.p.)
Panel A. Benchmark Value	29.9	
Panel B. Drivers of Structural Change in Production		
B.1 No Income Effects	31.4	+1.5
B.2 No Income and Substitution Effects	31.9	+2.0
Panel C. Other Drivers of Sectoral Innovation		
C.2 No Change in Markups	37.7	+7.8
C.1 No Cross-Sector Spillovers	19.7	-10.2
Panel D. No Population Growth Slowdown	32.7	+2.8

►  $\gamma = 0.30$

►  $\gamma = 0.70$



# Productivity Growth in the Future

Model's Predicted Structural Change in Employment and Innovation

Variable	1947 (-67 years)	2014	2081 (+67 years)	p.p. $\Delta$ (2081-2014)
$e_m(t)$	0.598	0.338	0.197	-14.1
$e_{ns}(t)$	0.764	0.758	0.754	-0.4
$e_{rs}(t)$	0.236	0.242	0.246	+0.4
$L_m^S(t)/L^S(t)$	0.947	0.866	0.483	-38.4
$L_{rs}^S(t)/L^S(t)$	0.016	0.078	0.358	+28.0
$L_{ns}^S(t)/L^S(t)$	0.037	0.056	0.160	+10.4

- ▶ Structural change out of manufacturing predicted to continue
- ▶ Stable shares within services
  - ▶ income and substitution effects offset each other
- ▶ Scientists continue moving to RIS (markups!)

# Productivity Growth in the Future

Model's Predicted Sectoral and Aggregate TFP Growth

Variable	1947	2014	2075	$x_{2075}/x_{2010}$
$y(t)$	1.00	3.03	6.14	2.02
$N_m(t)$	1.00	1.44	2.04	1.49
$N_{ns}(t)$	1.00	1.21	1.66	1.37
$N_{rs}(t)$	1.00	1.29	2.14	1.66
$TFP(t)$	1.00	1.30	1.91	1.47

- ▶ Re-allocation hinders output growth (lab. prod)
- ▶ TFP growth speeds up in services
  - ▶ Scientists drive TFP growth in NRIS
- ▶ Scientists continue moving into RIS

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# Conclusion

1. Document divergence in structural change patterns in innovation and and production employment/VA
2. Built a theory that rationalises these trends
  - ▶ Feedback between direction of tech. change and ST
  - ▶ Sectoral allocation of R&D activity depends on ST, spillovers and markups
3. Main results
  - ▶ Markup main force behind divergent transformation paths
  - ▶ ST contributed to TFP slowdown (-2.0 p.p)
  - ▶ Market power and markups hindered TFP growth (-7.8 p.p.)
  - ▶ Cross-sector spillovers helped sustain TFP growth (+10.2 p.p.)

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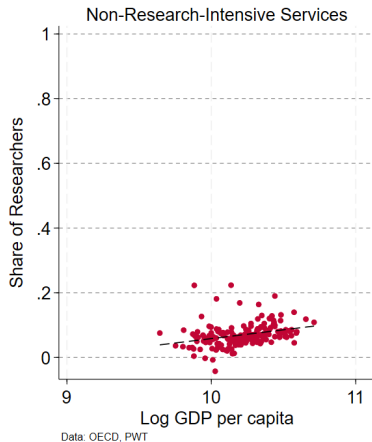
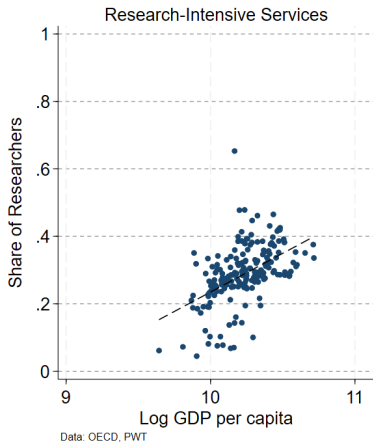
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# Services Split By Research-Intensity

Industry	Description	Patents per Hour		Sector
		Av. ('63-'14)	Initial ('63)	
51	Information	0.522	0.063	RI-Serv.
48-49	Transportation & Warehousing	0.130	0.089	RI-Serv.
54	Prof., Scientific & Technical Serv.	0.115	0.017	RI-Serv.
52	Finance & Insurance	0.104	0.062	NRI-Serv.
621-623	Healthcare and Social Assistance	0.058	0.056	NRI-Serv.
NA	Others	0.050	0.040	NRI-Serv.
55	Management of Comp. & Enterp.	0.042	0.021	NRI-Serv.
44-45	Retail Trade	0.034	0.005	NRI-Serv.
53	Real Estate, Rental & Leasing	0.002	0.002	NRI-Serv.
42	Wholesale Trade	0.001	0.000	NRI-Serv.

► Back to Facts

# Post-Industrial Economies Observe Similar Trends



**Countries included:** Australia, Austria, Belgium, China, Czechia, Spain, Estonia, France, Greece, Hungary, Iceland, Italy, Korea, Mexico, the Netherlands, Norway, Poland, Portugal, Singapore, Slovakia, Slovenia, USA [► Back to Facts](#)

# Conditional Probabilities Robust To Different Specs.

- Cited Conditional Probabilities (has X)

		Citing Sector		
		MFG	RI-SERV	NRI-SERV
Cited Sector	MFG	0.92	0.56	0.72
	RI-SERV	0.03	0.27	0.10
	NRI-SERV	0.05	0.17	0.19

- Cited Conditional Probabilities (has XY)

		Citing Sector		
		MFG	RI-SERV	NRI-SERV
Cited Sector	MFG	0.92	0.57	0.73
	RI-SERV	0.03	0.26	0.10
	NRI-SERV	0.05	0.16	0.17

- Cited Conditional Prob (has X only)

		Citing Sector		
		MFG	RI-SERV	NRI-SERV
Cited Sector	MFG	0.91	0.55	0.70
	RI-SERV	0.04	0.27	0.10
	NRI-SERV	0.05	0.17	0.19



# Conditional Probabilities Robust To Different Years

## ► Cited Probability ( $> 2000$ )

		Citing Setor		
		MFG	RI-SERV	NRI-SERV
Cited Sector	MFG	0.92	0.56	0.72
	RI-SERV	0.03	0.27	0.10
	NRI-SERV	0.05	0.17	0.18

## ► Cited Probability ( $> 2010$ )

		Citing Setor		
		MFG	RI-SERV	NRI-SERV
Cited Sector	MFG	0.92	0.55	0.73
	RI-SERV	0.03	0.28	0.10
	NRI-SERV	0.05	0.17	0.18

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# Matching Behavior of US Economy

Sectoral Data Aggregates: 1947-2014

	2014 TFP (1947 = 1)			2014 Relative Prices (1947 = 1)		
	MFG	RI-S	NRI-S	MFG	RI-S	NRI-S
<b>Moment</b>	1.48	1.32	1.23	0.93	0.97	1.08

	2014 Output Shares			2000 Scientist Shares		
	MFG	RI-S	NRI-S	MFG	RI-S	NRI-S
<b>Moment</b>	0.33	0.16	0.51	0.59	0.25	0.16

	1947 Output Shares			2014 Output p. w. (1947 = 1)
	MFG	RI-S	NRI-S	Aggregate
<b>Moment</b>	0.60	0.09	0.031	3.18

► Back to Estimation

# Income Elasticity Computations

Income elasticities according to CLM specification (2021) are

$$\Omega_i \equiv \frac{\partial \log Y_i}{\partial \log Y} = \sigma + (1 - \sigma) \frac{\zeta_i}{\bar{\zeta}} \quad \forall i = m, s$$

$$\Omega_k \equiv \frac{\partial \log Y_k}{\partial \log p_s Y_s} = \epsilon + (1 - \epsilon) \frac{\zeta_k}{\bar{\zeta}_s} \quad \forall k = rs, ns$$

where

$$\bar{\zeta} = \sum_i e_i \zeta_i \quad \text{and} \quad \bar{\zeta}_s = \sum_k \frac{e_k}{e_s} \zeta_k$$

Calibrated non-homotheticity parameters imply income elasticities

	MFG	RI-S	NRI-S
1947 shares	0.67	1.01	1.00
2014 shares	0.55	1.01	1.00

# TFP Growth Counterfactuals

Robustness ( $\gamma = 0.30$ )

	Aggregate TFP Growth (2010-1947, %)	Difference w.r.t Benchmark (p.p.)
Panel A. Benchmark Value	29.9	
Panel B. Drivers of Structural Change in Production		
B.1 No Income Effects	31.6	+1.7
B.2 No Income and Substitution Effects	32.0	+2.1
Panel C. Other Drivers of Sectoral Innovation		
C.2 No Change in Markups	38.1	+8.2
C.1 No Cross-Sector Spillovers	20.7	-9.2
Panel D. No Population Growth Slowdown	32.7	+2.8

# TFP Growth Counterfactuals

Robustness ( $\gamma = 0.70$ )

	Aggregate TFP Growth (2010-1947, %)	Difference w.r.t Benchmark (p.p.)
Panel A. Benchmark Value	29.9	
Panel B. Drivers of Structural Change in Production		
B.1 No Income Effects	31.5	+1.6
B.2 No Income and Substitution Effects	31.9	+2.0
Panel C. Other Drivers of Sectoral Innovation		
C.2 No Change in Markups	37.9	+8.0
C.1 No Cross-Sector Spillovers	20.1	-9.8
Panel D. No Population Growth Slowdown	32.8	+2.9