

Are Ideas Really Getting Harder To Find? R&D Capital and the Idea Production Function

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Stand with Ukraine 

The Idea Production Function in the Growth Literature

- Most R&D-based growth literature assumes that **R&D labor** is the only input in producing ideas (Romer, 1990; Jones, 1995, 1999; Ha and Howitt, 2007)
- The “lab equipment” specification uses the flow of **R&D spending** (Rivera-Batiz and Romer, 1991; Kruse-Andersen, 2017)
- Bloom et al. (2020) use **effective R&D employment**

$$\text{Research Output} = \underbrace{\text{Research Productivity}}_{\alpha_t} \times \text{Researchers.} \quad (1)$$

- Apparently, no role for **R&D capital!**

Introducing R&D Capital

We introduce **R&D capital** as a factor in producing ideas

- A **stock**, accumulated through R&D investment
- Arguably, R&D is an **increasingly capital-intensive activity**: experimentation, numerical computation, prototype testing, etc.
- Practicality and complexity of research equipment has undergone systematic, cumulative changes and productivity improvements over decades and centuries, from Ptolemy's astrolabe or Galileo's telescope to the LHC and VLT
- Modern R&D activity also increasingly uses AI

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Our expectation

- Predictions for **idea TFP** (Γ_t) will differ if the rate of change in R&D capital systematically differs from that of R&D labor

A First Look at the Data (USA, 1968-2019)

Variables	Symbol	Growth Rate
Patent Applications	ΔA	3.211
Patents-in-Force	A	2.410
Patent Applications Relative to Patents-in-Force	$\Delta A/A$	0.782
R&D Capital	\mathcal{K}	3.394
R&D Labor	\mathcal{R}	2.099
R&D Wage	w	0.848
R&D Expenditure (Real)	Ω	3.319
R&D Expenditure Relative to R&D Wage	Ω/w	2.450
R&D Capital Relative to Patents-in-Force	\mathcal{K}/A	0.961
R&D Labor Relative to Patents-in-Force	\mathcal{R}/A	-0.304
Patent Applications Relative to R&D Labor	$\Delta A/\mathcal{R}$	1.090
Patent Applications Relative to Ω/w	$\Delta A/(\Omega/w)$	0.743
Patent Growth Relative to R&D Labor	$(\Delta A/A)/\mathcal{R}$	-1.289
Patent Growth Relative to Ω/w	$(\Delta A/A)/(\Omega/w)$	-1.628

Source: Derived from *WIPO, IPUMS CPS*.

Contribution and Overview of Results

Contribution

- We introduce R&D capital alongside R&D labor into the idea production function (IPF)
- We estimate the IPF allowing for a **non-unitary elasticity of substitution** and **non-neutral productivity changes**

$$\Delta \tilde{A}_t = \left[\eta \left(\Gamma_t^x \tilde{\mathcal{K}}_t \right)^{\frac{\xi-1}{\xi}} + (1-\eta) \left(\Gamma_t^R \tilde{\mathcal{R}}_t \right)^{\frac{\xi-1}{\xi}} \right]^{\frac{\xi}{\xi-1}} \quad (2)$$

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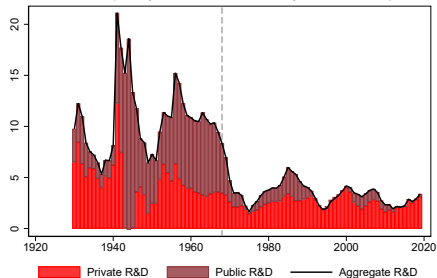
Overview of Results

- We find an elasticity of substitution $\xi \approx 0.6 - 0.8$
- We identify a systematic positive trend in R&D labor productivity at about 1% per year on average...
- ...and a cyclical dynamic in R&D capital productivity
- On average, **effective supply of R&D capital was lagging behind that of R&D labor**, constraining R&D output
- Idea TFP has not been falling but oscillating around a constant mean

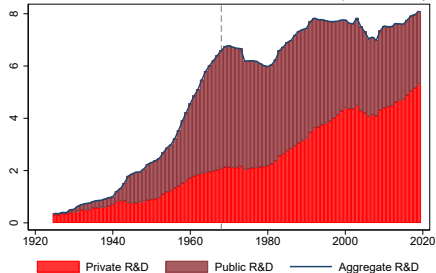
R&D Capital

R&D capital is constructed with PIM from BEA data on R&D investment (USA, 1929-2019), $\delta = 15\%$ per annum

A: Total R&D capital (Annual Growth Rate) And Its Components



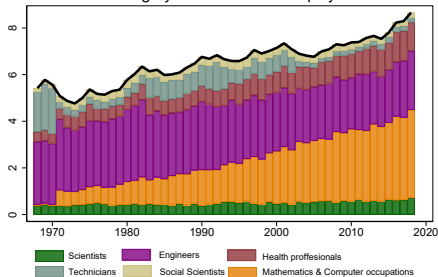
B: Share of R&D Assets in Non-Residential capital Stock (%)



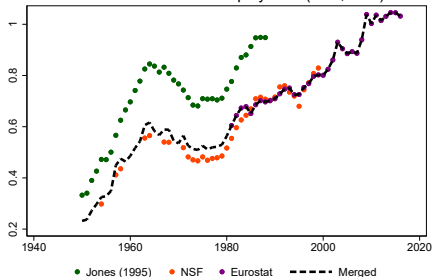
R&D Labor

R&D labor is constructed using IPUMS Current Population Survey (CPS) data. We concentrate on **Scientists; Engineers; Health Professionals; Technicians; Social Scientists; and Mathematical & Computer Occupations**.

C: Category Shares of R&D employment



D: Share of the R&D employment (FTE, in %)



R&D Output and Rental Prices

R&D output

- **Patent applications** from USPTO/WIPO (Madsen, 2008; Ang and Madsen, 2011; Venturini, 2012)
- *Robustness check*: growth rate of aggregate TFP (Bloom et al., 2020)

Rental prices in the R&D sector

- **Capital rental rate**: sum of the real interest rate (from Fred) and the R&D capital depreciation rate
- **Wage**: real hourly wage in R&D (from CPS)

Estimation

We estimate the **normalized CES specification** in a 2-equation system (Klump et al., 2007; León-Ledesma et al., 2010)

- This approach combines information from different sides of the production framework (costs and volumes) and exploits cross-equation restrictions

$$\Delta \tilde{A}_t = \left[\eta \left(\Gamma_t^{\mathcal{K}} \tilde{\mathcal{K}}_t \right)^{\frac{\xi-1}{\xi}} + (1-\eta) \left(\Gamma_t^{\mathcal{R}} \tilde{\mathcal{R}}_t \right)^{\frac{\xi-1}{\xi}} \right]^{\frac{\xi}{\xi-1}} \quad (3)$$

$$\ln \left(\frac{r_t^{\mathcal{K}} \tilde{\mathcal{K}}_t}{w_t^{\mathcal{R}} \tilde{\mathcal{R}}_t} \right) = \left(\frac{\xi-1}{\xi} \right) \ln \left(\frac{\Gamma_t^{\mathcal{K}} \tilde{\mathcal{K}}_t}{\Gamma_t^{\mathcal{R}} \tilde{\mathcal{R}}_t} \right) \quad (4)$$

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We consider three assumptions about **technical change** (growth in $\Gamma_t^{\mathcal{K}}$ and $\Gamma_t^{\mathcal{R}}$)

- Exponential trend
- Box-Cox form $\log \Gamma_t^j = \mathbf{B}(\gamma_j, \lambda_j; t)$
- Fourier form $\log \Gamma_t^j = \mathbf{F}(\gamma_j, \kappa_j^{\text{sin}}, \kappa_j^{\text{cos}}; t)$

Baseline Results

Explained variable: **normalized patent applications**

	(3)	(4)	(5)	(6)	(7)	(8)
ξ	0.844***	2.531***	0.737***	0.986***	0.793***	0.760***
$\gamma_{\mathcal{R}}$	0.012***	0.012***	0.011***	-0.013	0.011***	0.011***
$\lambda_{\mathcal{R}}$		2.890***		6.453		
$\gamma_{\mathcal{K}}$			0.004	0.060	-0.016***	-0.013***
$\lambda_{\mathcal{K}}$				5.208		
$\kappa_{\mathcal{K}}^{sin}$					0.556***	0.438***
$\kappa_{\mathcal{K}}^{cos}$					-0.427***	-0.337***
\mathcal{K} share						0.418***
LATC	Exp.	B	Exp.	B	Exp.	Exp.
KATC	no	no	Exp.	B	F	F
$\xi = 1$	[0.357]	[0.096]	[0.058]	[0.934]	[0.000]	[0.000]
$\lambda_{\mathcal{R}} = 1$		[0.014]		[0.775]		
$\lambda_{\mathcal{K}} = 1$				[0.279]		
$\gamma_{\mathcal{R}} = \gamma_{\mathcal{K}}$			[0.086]	[0.937]	[0.000]	[0.000]
$\kappa_{\mathcal{K}}^{cos} = \kappa_{\mathcal{K}}^{sin} = 0$					[0.000]	[0.006]
res_4	[0.086]	[0.066]	[0.101]	[0.020]	[0.006]	[0.008]
res_3	[0.095]	[0.237]	[0.085]	[0.051]	[0.001]	[0.000]
bic	-141.7	-174.2	-133.3	-173.9	-239.0	-237.0

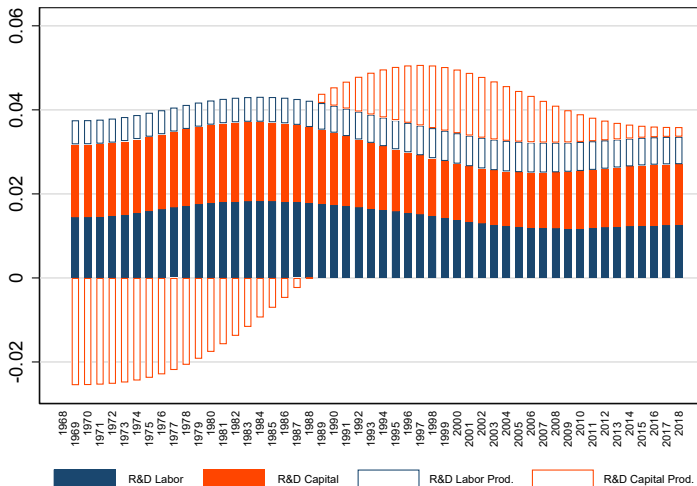
Robustness Check: Explaining TFP Growth

Explained variable: **normalized TFP growth rate**

	(2)	(3)	(4)	(5)	(6)
ξ	0.987***	1.001***	0.603***	0.609***	0.601***
$\gamma_{\mathcal{R}}$	-0.040***	0.405	-0.025***	-0.025***	-0.025***
$\lambda_{\mathcal{R}}$	0.861***		1.074***		
$\gamma_{\mathcal{K}}$		-0.892	-0.029***	-0.035***	-0.034***
$\lambda_{\mathcal{K}}$			0.423***		
$\gamma_{\mathcal{K}}^{\sin}$				0.035*	0.025
$\gamma_{\mathcal{K}}^{\cos}$				-0.126***	-0.108***
\mathcal{K} share					0.562***
LATC	B	Exp.	B	Exp.	Exp.
KATC	no	Exp.	B	F	F
$\xi = 1$	[0.611]	[0.650]	[0.000]	[0.000]	[0.000]
$\lambda_{\mathcal{R}} = 1$	[0.063]		[0.579]		
$\gamma_{\mathcal{R}} = \gamma_{\mathcal{K}}$		[0.359]	[0.132]	[0.000]	[0.000]
$\kappa_{\cos}^{\mathcal{K}} = \kappa_{\sin}^{\mathcal{K}} = 0$				[0.000]	[0.000]
$\lambda_{\mathcal{K}} = 1$			[0.000]		
res_3	[0.026]	[0.062]	[0.032]	[0.000]	[0.000]
res_4	[0.053]	[0.030]	[0.004]	[0.004]	[0.005]
<i>bic</i>	-191.2	-197.7	-206.3	-241.3	-237.5

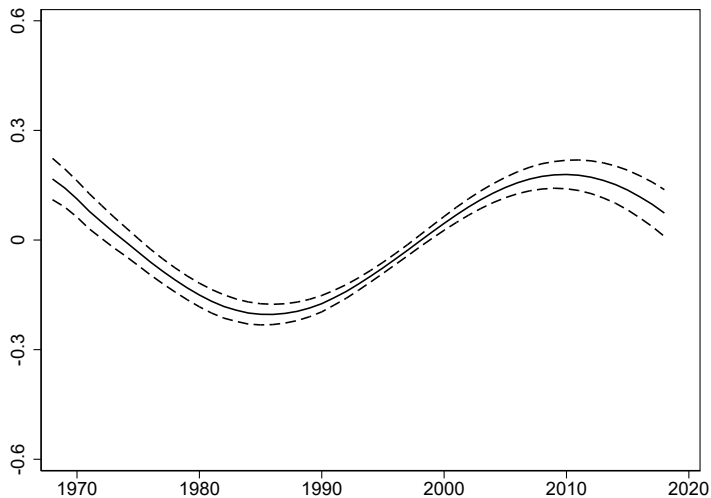
Idea Growth Decomposition (Baseline)

Annual change on HP-filtered contributions



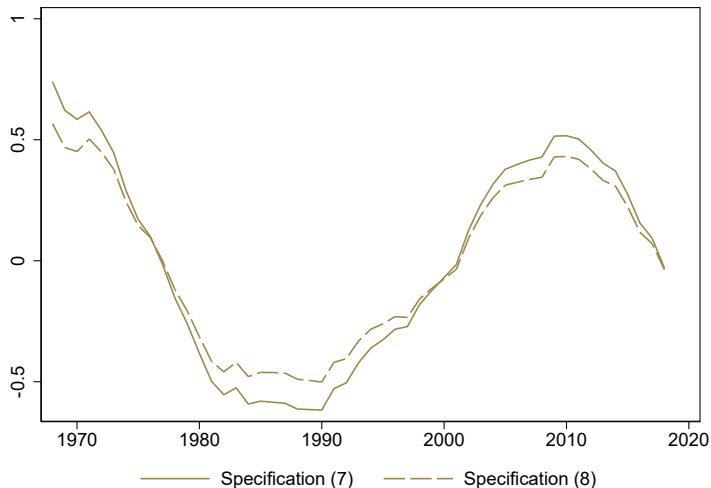
Idea TFP

Hicks-neutral idea TFP backed out from the IPF (in logs, with 95% CI)



Effective Factor Ratio in R&D

Effective R&D capital-to-labor ratio (in logs)



Discussion of Baseline Results

Are ideas getting harder to find? Not quite!

- R&D capital is an **essential, complementary** factor in R&D activity
- In R&D, as in the aggregate economy, capital accumulation outran the growth in labor supply over the long run
- **In effective terms**, average growth in R&D labor outran that of R&D capital. There is **an increasing scarcity of R&D capital required to find the new ideas**
- **Three main phases of ideas growth**
 - (i) sluggish growth in ideas (up to early 1980s)
 - (ii) sharp acceleration in ideas growth (1980s-2000s)
 - (iii) slowdown in ideas growth (since 2000s)

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Perspectives: secular stagnation vs. transition to a mature digital economy

- The current slowdown in R&D output is likely due to a relative shortage of R&D capital, not sharply falling idea TFP

Thank you for your attention.

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