

From Imitation to Innovation: Where is all that Chinese R&D Going?

König, M., Storesletten, K., Song, Z., & Zilibotti, F. (2022).

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Motivation

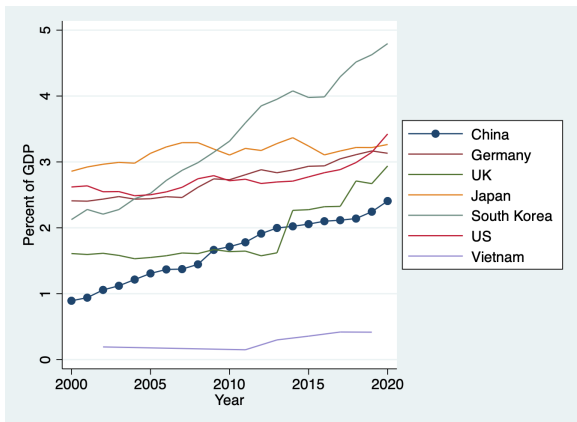
- Recently, the rapid economic growth in China has been accompanied by a boom in R&D expenditure and the government's growing emphasis on innovation
- R&D investments increased 1% of GDP in the 1990s to 2.4% of GDP by 2020.
- A common concern is that these investment decisions are distorted by policies and frictions that are pervasive in China.

⇒ This paper constructs and structurally estimates an endogenous growth model to

- study the dynamic effects of misallocation on the investments firms make to improve their productivity growth
- quantify the contribution of these investments to China's aggregate growth

Motivation

Figure 1: Gross R&D Expenditure



Outline

- 1 Theory
- 2 Data and descriptive evidence
- 3 Estimation
- 4 Results
- 5 Nontargeted moments
- 6 Counterfactual
- 7 Conclusion

Theory

- A dynamic Melitz-type (small open) economy populated by a unit measure of monopolistically competitive firms
- Produce differentiated goods using a CRS technology

$$Y_i(t) = A_i(t)K_i(t)^\alpha L_i(t)^{1-\alpha}$$

- Differentiated goods are combined into a homogeneous final good by a CES aggregator ($\eta > 1$)

$$Y = \left(\int_0^1 Y_i^{\frac{\eta-1}{\eta}} di \right)^{\frac{\eta}{\eta-1}}$$

- No entry and exit

Theory

- Firms are owned by overlapping generations of two-period-lived manager-entrepreneurs

⇒ Break down the firm's problem into two steps

- A static profit maximization
- A dynamic (intertemporal investment) problem

Firm's static problem

- The young manager chooses $K_i(t)$ and $L_i(t)$ to maximize profits subject to its residual demand, TFP $A_i(t)$ and (persistent) output wedge $\tau_i(t) \in \{\tau_l, \tau_h\}$
- Firm i 's current (period t) optimal profits:

$$\pi_i(t) = [(1 - \tau_i(t))A_i(t)]^{\eta-1} \times \tilde{\Pi}(t) \quad (1)$$

$$\tilde{\Pi}(t) \equiv (\alpha^\alpha(1 - \alpha)^{1-\alpha}(\eta - 1))^{\eta-1} \eta^{-\eta} \times \frac{Y(t)}{(r^\alpha w(t)^{1-\alpha})^{\eta-1}} \quad (2)$$

- The firm's value added

$$P_i(t)Y_i(t) = \eta\pi_i(t) \propto [A_i(t)(1 - \tau_i(t))]^{\eta-1}$$

Firm's Dynamic Problem

- The young manager decides the strategy to improve the firm's TFP in the next period
- Each successful attempt to move up the productivity ladder results in a constant accrual of log TFP

$$\log(A_{i,t+1}) = \log(A_{i,t}) + \tilde{a}, \quad (\tilde{a} > 0)$$

- The ranking in the productivity ladder: $a \equiv \log(A)/\tilde{a}$, $a \in \mathbb{N}^+$
- \mathcal{A} : the TFP distribution
- $\mathcal{A}_1, \mathcal{A}_2, \dots$: the proportion of firms at each rung of the ladder
- $F_a = \sum_{j=1}^a \mathcal{A}_j$: the associated cumulative distribution

Firm's Dynamic Problem

- Firms can increase their TFP through:
 - **Imitation (costless):** Firm A_i meets A_j .
 - $A_j > A_i$: increases productivity with prob. $q > 0$
 - $A_j \leq A_i$: retains A_i \implies Prob. of success: $q(1 - F_a)$
 - **Innovation (R&D cost):**
 - Succeed with prob. p drawn i.i.d from $G : [0, \bar{p}] \rightarrow [0, 1]$, $\bar{p} \leq 1$
 - If fail, can do passive imitation with probability $\delta q(1 - F_a) \geq 0$ \implies Total prob. of success: $p_i + (1 - p_i)\delta q(1 - F_a)$
- Firms pursue the strategy that yields the highest PV of profits (net of the R&D cost)

Equilibrium Dynamics With Costless Innovation

- No R&D cost and $\delta < 1$ (Why not $\delta \geq 1$?)
- Firm i chooses innovation \iff

$$p_i \geq Q(a, \tau; \mathcal{A}) \equiv \frac{q(1-\delta)(1-F_a)}{1-\delta q(1-F_a)} \quad (3)$$

- $\frac{\partial Q}{\partial a} < 0 \implies$ the proportion of innovating firms will be nondecreasing in the initial TFP
- the ex post TFP growth gap between innovating and imitating firms is increasing in the TFP level

Equilibrium Dynamics With Costless Innovation

- Decision function notations:

$$\chi^{im}(a, p, \tau; \mathcal{A}) = 1 - \chi^{in}(a, p, \tau; \mathcal{A}) = \begin{cases} 1 & \text{if } p \leq Q(a, \tau; \mathcal{A}) \\ 0 & \text{if } p > Q(a, \tau; \mathcal{A}) \end{cases} \quad (4)$$

- The law of motion of TFP:

$$\begin{aligned} & \mathcal{A}_a(t+1) - \mathcal{A}_a(t) \\ &= \int_0^{\bar{p}} \left[\chi^{in}(a-1, p, \tau; \mathcal{A}) \times \left(p + (1-p)\delta q(1 - F_{a-1}(t)) \right) \mathcal{A}_{a-1}(t) \right. \\ & \quad + \chi^{im}(a-1, p, \tau; \mathcal{A}) q(1 - F_{a-1}(t)) \mathcal{A}_{a-1}(t) \\ & \quad - \chi^{in}(a, p, \tau; \mathcal{A}) \times \left(p + (1-p)\delta q(1 - F_a(t)) \right) \mathcal{A}_a(t) \\ & \quad \left. - \chi^{im}(a, p, \tau; \mathcal{A}) q(1 - F_a(t)) \mathcal{A}_a(t) \right] dG(p) \end{aligned} \quad (5)$$

Equilibrium Dynamics With Costless Innovation

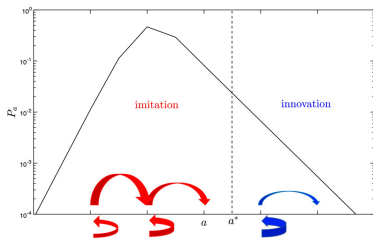
PROPOSITION 1. Under some regularity conditions, there exists a traveling wave solution of the form $\mathcal{A}_a(t) = f(a - vt)$ with velocity $v = v(q, \delta, g(\rho)) > 0$, with left and right Pareto tails. For a given t , \mathcal{A}_a is characterized as follows:

- (i) For a sufficiently large, $\mathcal{A}_a(t) = O(e^{-\rho(a-vt)})$, where the exponent ρ is the solution to the transcendental equation $\rho v = \hat{p}(e^\rho - 1)$.
- (ii) For a sufficiently small, $\mathcal{A}_a(t) = O(e^{\lambda(a-vt)})$, where the exponent λ is the solution to the transcendental equation $\lambda v = q(1 - e^{-\lambda})$.

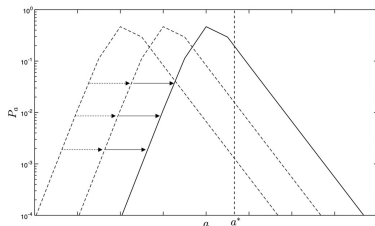
⇒ Basically, there exists a stationary equilibrium with the constant growth rate v , which is increasing in q , δ , and \bar{p}

Equilibrium Dynamics With Costless Innovation

Figure 2: Equilibrium dynamics with a stationary distribution



Panel A



Panel B

Equilibrium Dynamics With Costly Innovation

- Recall the discounted value of profits:

$$\pi_i(t) = \frac{1}{1+r} \times [(1 - \tau_i(t)) A_i(t)]^{\eta-1} \times \tilde{\Pi}(t)$$

$$\tilde{\Pi}(t) \equiv (\alpha^\alpha (1 - \alpha)^{1-\alpha} (\eta - 1))^{\eta-1} \eta^{-\eta} \times \frac{Y(t)}{(r^\alpha w(t)^{1-\alpha})^{\eta-1}}$$

$$w(t)^{1-\alpha} = \left(1 - \frac{1}{\eta}\right) \frac{\alpha^\alpha (1 - \alpha)^{1-\alpha}}{r^\alpha} \left(\int_0^1 (1 - \tau_i)^{\eta-1} A_i(t)^{\eta-1} di\right)^{\frac{1}{\eta-1}}$$

- Given $A_i(t)$, $\tau_i(t)$, profit $\pi_i(t)$ changes over time via two channels
 - + higher income \implies higher demand for all varieties
 - the progress of competitors \implies lower market share ($\eta > 1$)
- On equilibrium, wages, profits and aggregate output grow at the same rate

Equilibrium Dynamics With Costly Innovation

- The R&D cost for innovating firms

$$c_i(t) = (A_i(t)^\theta \bar{A}(t)^{1-\theta})^{\eta-1} \times \bar{c} \times \tilde{\Pi}(t) \quad (6)$$

where $\bar{c} > 0$, $\theta \in [0, 1]$, $\bar{A}(t) \equiv \int_0^1 A_i(t) di$

- Larger profits, higher costs
- The relative level of the firm TFP (but not its absolute value) matters for the R&D decision
- R&D cost is increasing in $A_i(t)$, conditional on $\bar{A}(t)$
- R&D cost follows the trend $\tilde{\Pi}(t)$

Equilibrium Dynamics With Costly Innovation

- Firm i chooses innovation $\iff p_i \geq Q(a, \tau; \mathcal{A})$

$$Q(a, \tau; \mathcal{A}) = \frac{q(1-\delta)(1-F_a)}{1-\delta q(1-F_a)} + \frac{e^{(1-\theta)(\eta-1)(\bar{a}-a)}}{(e^{\eta-1}-1)\mathbb{E}[(1-\tau')^{\eta-1} | \tau]} \frac{\bar{c}(1+r)(1+g)^{\eta-2}}{1-\delta q(1-F_a)} \quad (7)$$

- g : the (endogenous) steady-state growth rate,
 - $a \equiv \log \bar{A}$
 - $\mathbb{E}[\tau' | \tau]$: the conditional expectation of next-period wedge.
- Comparison to the costless innovation
 - Imitation becomes more attractive \implies the threshold Q is higher
 - Larger τ reduces future profit proportionally to TFP without affecting the R&D cost \implies less innovation: $\frac{\partial Q}{\partial \tau} > 0$

Equilibrium Dynamics With Costly Innovation

The law of motion of the TFP distribution:

$$\begin{aligned} & A_a(t+1) - A_a(t) \\ &= \sum_{j \in \{l, h\}} \omega_{\tau_j}(t) \int_0^P \left[\chi^{\text{in}}(a-1, \rho, \tau_j; A) (\rho + (1-\rho)\delta q(1 - F_{a-1}(t))) A_{a-1}(t) \right. \\ &\quad + \chi^{\text{in}}(a-1, \rho, \tau_j; A) q(1 - F_{a-1}(t)) A_{a-1}(t) \\ &\quad - \chi^{\text{in}}(a, \rho, \tau_j; A) (\rho + (1-\rho)\delta q(1 - F_a(t))) A_a(t) \\ &\quad \left. - \chi^{\text{in}}(a, \rho, \tau_j; A) q(1 - F_a(t)) A_a(t) \right] dG(\rho) \end{aligned} \tag{8}$$

- ω_{τ_l} and ω_{τ_h} : the proportion of low- and high-wedge firms

$$\omega_{\tau_l}(t+1) - \omega_{\tau_l}(t) = (1 - \rho_h)(1 - \omega_{\tau_l}(t)) - (1 - \rho_l)\omega_{\tau_l}(t),$$

Predictions of the theory

1. ceteris paribus, the proportion of firms engaged in R&D is increasing in TFP;
2. ceteris paribus, firms with higher wedges are less likely to engage in R&D. Then, conditional on TFP, larger firms are more likely to engage in R&D
3. expected TFP growth is falling in current TFP, especially for nonR&D firms
4. the gap in average TFP growth between R&D firms and nonR&D firms is increasing in TFP.

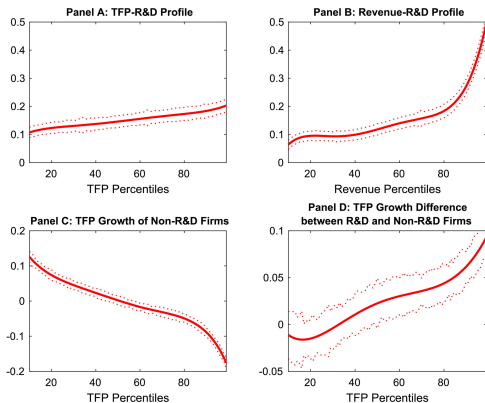
- the Annual (firm-level) Survey of Industries conducted by China's National Bureau of Statistics, 1998 – 2007 and 2011 – 2013:
 - A balanced panel for all manufacturing firms, 2007 – 2012
 - Data for R&D expenditure at the firm level are 2007
 - Innovators = positive R&D expenditure
- Plant level data collected by Taiwan's Ministry of Economic Affairs, 1999 – 2004.

Summary Statistics

Table 1: Summary Statistics

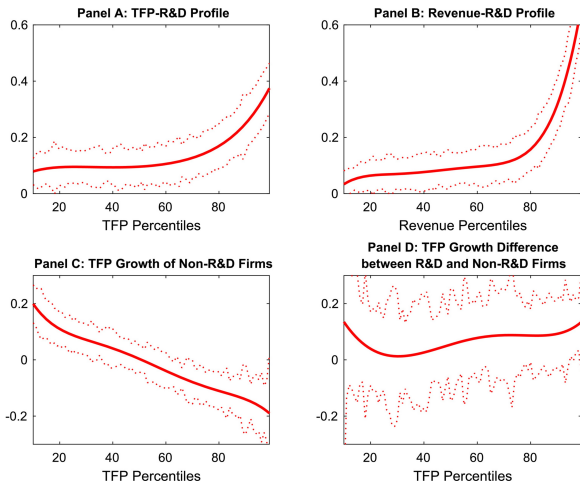
Year	Number of Firms	Number of R&D Firms	Median Value Added (Million USD)	Mean Value Added (Million USD)	Median R&D Intensity (%)	Aggregate R&D Intensity (%)
<i>Balanced Panel of Chinese Firms</i>						
2007	123,368	18,140	1.48	5.81	1.73	1.86
2012	123,368	N.A.	3.33	11.45	N.A.	N.A.
<i>Private Chinese Firms in the Balanced Panel</i>						
2007	117,983	15,828	1.43	4.67	1.65	1.54
2012	117,771	N.A.	3.26	9.57	N.A.	N.A.
<i>Balanced Panel of Taiwanese Firms</i>						
1999	11,229	1,487	0.16	2.91	8.50	3.14
2004	11,229	1,144	0.17	4.78	6.42	2.93

Figure 3: Chinese firms in the balanced panel 2007–2012



Empirical results

Figure 4: Taiwanese firms in the balanced panel 1999–2004



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Structural estimation

- Calibrate α (equal to the measured industry-specific labor income share), $\eta = 5$, θ , \tilde{a} and estimate the rests
- Target two sets of moments
 - Economics: 16 (four intervals (10th-49th, 50th-79th, 80th-94th, and the top five) of the distribution of TFP and size in each of the four panels)
 - Measurement error: 1 (the empirical variance of m.e. in TFP $\hat{v}_{\mu a}$)
- Use Simulated Method of Moments (SMM)

Measurement Error

- Classical m.e. of revenues and inputs
 - Generate an attenuation bias in the relationships between the propensity to engage in R&D and both TFP and size, and
 - Exaggerate TFP convergence
 - Identification:
 - the firm-specific wedges τ_i are constant during the 2007 – 2012 period
- ⇒ the time series variation (2007 – 2012) in value added and input measures at the firm level can be used to infer the extent of m.e.
- The paper shows how to back out (empirical) $\hat{v}_{\mu a}$ from data

Distribution of output wedges

- The estimated output wedges and TFP are correlated in the data

⇒ Assume

$$-\ln(1 - \tau_i) = b \cdot (a_{it} - a_t) + \varepsilon_{it}^{\tau} \quad , \quad \varepsilon_{it}^{\tau} \sim \mathcal{N}(0, \text{var}(\varepsilon_{it}^{\tau}))$$

- Estimate $b = 0.779$ and $\text{var}(\varepsilon_{it}^{\tau}) = 0.042$ (correcting for m.e.)

The Technology of R&D

- Cost function: estimate $\theta \approx 0.25$ by targeting the R&D cost-to-value added ratio: the top five percent and the 10th–49th percentile..

$$\frac{\mathbb{E}[\psi_j | a_j]}{\mathbb{E}[\psi_i | a_i]} = \exp\left(\frac{1-\eta}{\eta}(1-\theta+b)(a_i - a_j)\right)$$

- Step size: $\tilde{a} \approx 0.78$ implies an average cost of innovation of 3.7% of the industrial value added (about twice the ratio in the Chinese data)
- Productivity of innovation p : drawn from an i.i.d uniform probability distribution with support $[0, \bar{p}]$

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Parsimonious Model

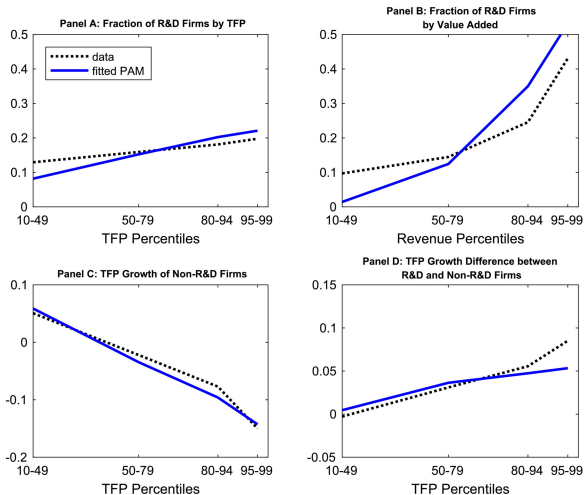
Estimate five parameters:

- $q \leftarrow$ the TFP convergence rate across imitating firms (Panel C) conditional on m.e.
- $\delta \leftarrow$ the TFP convergence rate across both imitating and innovating firms (Panels C and D).
- $\bar{p} \leftarrow$ the extent to which, conditional on TFP, innovating firms grow faster than imitating firms (Panel D).
- $\bar{c} \leftarrow$ the total share of innovating firms, given the other parameters
- $\nu_{\mu a} \leftarrow$ flatten the schedules in Panels A and B, and steepen the schedule in Panel C.

Table III. Estimation, Chinese firm balanced panel 2007–2012

	(1)	(2)	(3)	(4)	(5)	(6)
	PAM	FLM	IPM	FRM	PAM	IPM
					Higher R&D cutoff	
Imitation prob. q	0.175 (0.031)	0.271 (0.019)	0.361 (0.019)	0.275 (0.051)	0.294 (0.058)	0.546 (0.052)
Second chance δ	0.008 (0.011)	0.020 (0.021)	0.001 (0.027)	0.019 (0.024)	0.141 (0.106)	0.001 (0.080)
Innov. prod. \bar{p}	0.096 (0.008)	0.114 (0.006)	0.113 (0.006)	0.237 (0.016)	0.107 (0.013)	0.111 (0.010)
Innov. cost \bar{c}	1.627 (0.136)	3.374 (0.174)	2.318 (0.177)	10.486 (1.363)	3.601 (0.448)	9.393 (2.403)
Std. dev. m.e. $\sigma_{\mu a}$	0.549 (0.014)	0.472 (0.008)	0.431 (0.005)	0.459 (0.025)	0.476 (0.022)	0.391 (0.011)
Std. dev. innov. subs. σ_C		1.243 (0.038)	1.092 (0.037)	0.011 (0.036)		1.969 (0.213)
Policy inter. c_a			1.888 (0.159)	-10.46 (1.042)		2.499 (0.356)
Fake share Y				0.099 (0.005)		
J -statistic	1.518	0.507	0.368	0.362	2.690	0.516

Figure 5: Parsimonious model (PAM).



Heterogeneous Innovation Costs

Three cases:

FLM: innovation wedges \bar{c}_i are i.i.d. across firms

$$c_i(t) \propto \left[\bar{c} - \exp\left(\bar{\xi}_i(t) - \frac{\sigma_c}{2}\right) + 1 \right] \cdot \left(A_i(t)^\theta A(t)^{1-\theta} \right)^{\eta-1},$$

where $\bar{\xi}_i \sim \mathcal{N}(0, \sigma_c^2)$.

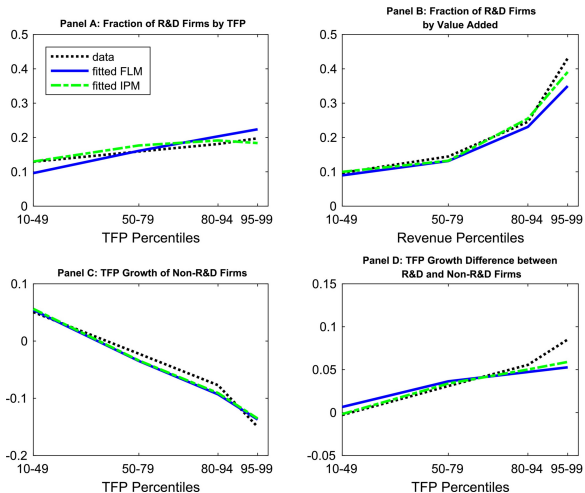
IPM: \bar{c}_i to be correlated with observable firm characteristics

$$c_i(t) \propto \left[\bar{c} - \exp\left(\bar{\xi}_i(t) - \frac{\sigma_c}{2}\right) + 1 + c_a \left(G(a_i) - \frac{1}{2} \right) \right] \left(A_i(t)^\theta \bar{A}(t)^{1-\theta} \right)^{\eta-1},$$

$c_a > 0$: industrial policy favors low-TFP firms

FRM: a share Υ of firms can collect subsidies by just *claiming* to invest in R&D

Figure 6: Flexible (FLM) and Industrial Policy (IPM) models



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Indirect inference

Table IV Indirect inference, balanced panel of Chinese firms, 2007-2012

	(1)	(2)	(3)	(4)	(5)
	Data	PAM	FLM	IPM	FRM
<i>Panel A: Dependent Variable: R&D decision in 2007</i>					
log(TFP)	0.368 (0.0284)	0.712 (0.0145)	0.287 (0.0167)	0.298 (0.0181)	0.322 (0.0167)
wedge	-0.410 (0.0357)	-0.824 (0.0177)	-0.274 (0.0202)	-0.327 (0.0214)	-0.357 (0.0203)
<i>Panel B: Dependent Variable: TFP Growth</i>					
log(TFP)	-0.062 (0.0035)	-0.094 (0.0002)	-0.104 (0.0003)	-0.112 (0.0003)	-0.081 (0.0004)
$R\&D_d$	0.036 (0.0042)	0.034 (0.0005)	0.033 (0.0005)	0.028 (0.0005)	0.053 (0.0008)

Figure 7: TFP distribution

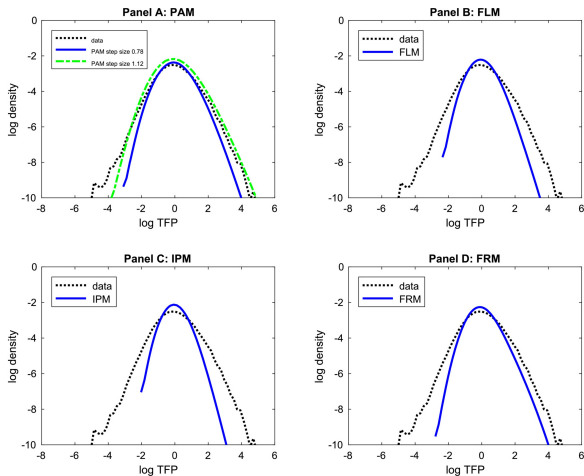
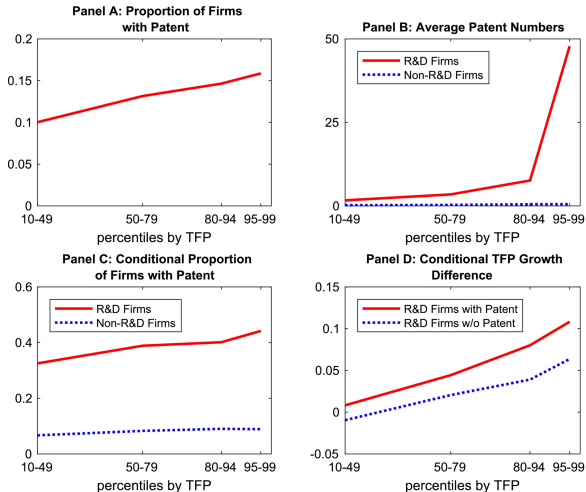


Figure 8: Patents of Chinese firms, balanced panel 2007–2012

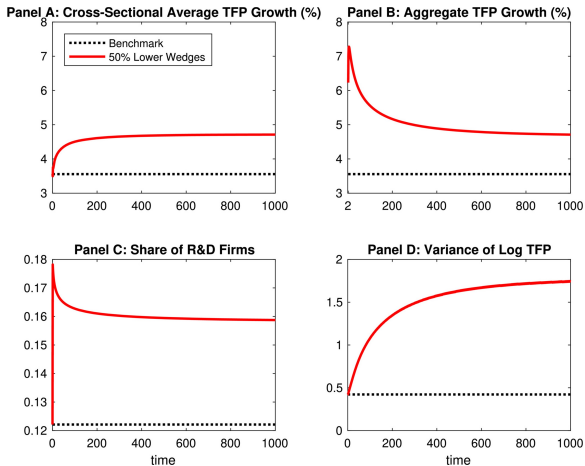


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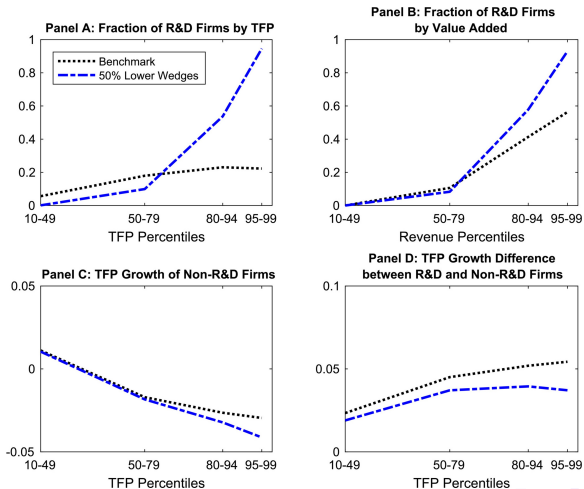
Reducing misallocation

Figure 9: Transition after lowering wedges



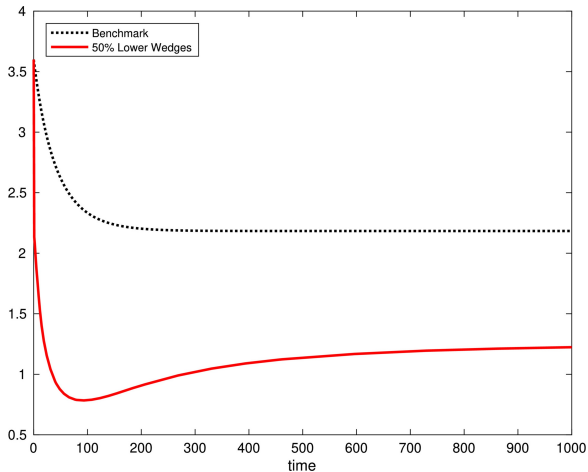
Reducing misallocation

Figure 10: Steady-state moments in the counterfactual model



International spillovers

Figure 11: TFP Gap during transition with international spillovers



The innovation technology

Table 2: Counterfactuals, parsimonious model

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	PAM estim. model	50% lower output wedges	Taiwan's q	Taiwan's \bar{p} and \bar{c}	Taiwan's \bar{p} , \bar{c} and q	Increase \bar{c} so share R&D firms is 5%	Decrease \bar{c} so share R&D firms is 20%	All firms do R&D
Fraction of R&D firms (%)	12.2	15.8	10.7	14.1	12.2	5	20	100
Steady-state TFP growth (%)	3.56	4.70	4.49	4.92	6.03	2.41	4.42	3.80

Conclusion

- Construct and structurally estimate a theory of TFP growth driven by innovation and technology diffusion through random interactions
- Estimate the model to earn new insights about the nature and effects of the R&D expenditure boom in China in recent years.
- R&D investments appear to have contributed significantly to the productivity growth of China.
- the return to productivity of R&D investments is lower in China than in Taiwan.
- Pervasive output wedges often induce the wrong firms (and, conversely, deter the right ones) from investing in R&D, thereby reducing the productivity of R&D investments.

Discussion

- Indirect approach to infer wedges \implies debate and dispute + limited policy implications (Bergquist et al. (2026))
- Abstract from entry and exit (an important margin)
- Under-explore in firm dynamics and the role of an intensive margin of R&D (the focus on innovation literature).

References I

Bergquist, Lauren F, Danial Lashkari, and Eric Verhoogen (2026), "Wedges: A microeconomic perspective on misallocation." Technical report, National Bureau of Economic Research.