## Market Power when Ideas get Harder to Find: A Theory of Directed Innovation

Julian Schärer (University of Zurich, University of Geneva)

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#### Ideas are getting harder to find (Bloom et al., 2020)

Scope	Time period	Average annual growth rate (%)	Half-life (years)	Dynamic diminishing returns, β 3.1	
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Census of Manufacturing	1992-2012	-7.8	9		

TABLE 7—SUMMARY OF THE EVIDENCE ON RESEARCH PRODUCTIVITY

#### Source: Bloom et al. (2020)

Other Macro trends since 1980s:

- Rise of market power (De Loecker et al., 2020: US markups 1)
- Declining business dynamism ( firm entry ↓, share of young firms ↓, firm size ↑)

This paper:

- Schumpeterian firm dynamics with search & directed innovation
- Prediction: in *ideas get harder to find* environment, market power should rise
- This and other predictions: all in line with data
- Key ingredient: **directed innovation**

#### Why innovation seems to be directed

Recent evidence: product market dominance deters firm entry

- Argente et al. (2021): patenting by leaders  $\rightarrow$  less innovation by competitors/entrants
- Galasso and Schankerman (2015): invalidation of focal patent leads to more citations
  - effect starts after 2 years consistent with more entry and cumulative innovation

 $\Rightarrow$  Firm entry & innovation are **directed** rather than undirected

#### Model in a nutshell

Market-specific rate of creative destruction =  $Pr(success|innovation) \times Pr(innovation)$ 

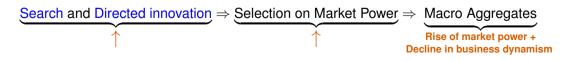
- 1. With probability > 0, each innovation cannot build on leader's technology
  - Build on follower instead
  - ► The larger the leader-follower gap, the lower the likelihood of overtaking the leader
- 2. Search and Directed innovation
  - Choose sample size of market search, then target one and do R&D
  - Firms with high leader-follower gaps are targeted less than low-gap firms

Search and Directed innovation  $\Rightarrow$  Selection on Market Power  $\Rightarrow$  Macro Aggregates

#### Ideas get harder to find

Market-specific rate of creative destruction =  $Pr(success|innovation) \times Pr(innovation)$ 

- 1. With probability > 0, each innovation cannot build on leader's technology
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- 2. Search and Directed innovation
  - Choose larger sample size of market search, then target one and do R&D
  - Firms with high leader-follower gaps are targeted even less than low-gap firms



#### **Related literature**

#### Schumpeterian growth theory

- Aghion & Howitt (1992)
- Grossman & Helpman (1991)

#### Schumpeterian firm dynamics

Klette & Kortum (2004)

#### Step-by-step innovation

• Aghion et al. (1997)

#### Endogenous markup distribution

• Peters (2020)

#### Market power, dynamism & growth: explanations

- Decline in knowledge diffusion: Akcigit & Ates (2021)
- Concentration and defensive R&D: Manera (2021)
- Role of IT: Aghion et al. (2021)
- Role of intangibles: De Ridder (2020)
- Declining interest rates: Liu et al. (2020)
- Declining population growth: Peters & Walsh (2021)

Decline in research productivity

• Bloom et al. (2020)

Rise of market power

• De Loecker et al. (2020)

## Model

#### Model environment I

• Preferences:

$$U_0 = \int_0^\infty e^{-
ho t} \ln(C_t) \, dt$$

• All output is consumed:

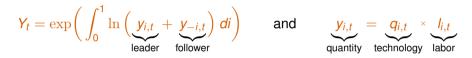
$$C_t = Y_t$$

• Labor supplied inelastically to firms *f* and entrants *e*,

$$\int_{f} \left( \underbrace{L_{P,f,t}}_{\text{production}} + \underbrace{L_{S,f,t}}_{\text{search}} + \underbrace{L_{R\&D,f,t}}_{R\&D} \right) df + \underbrace{\mathcal{M}_{0,t} \times L_{e,t}}_{\text{entrant labor}} = L$$

#### Model environment II

• Final good and intermediate products  $i \in [0, 1]$ :



- Bertrand competition  $\Rightarrow$  markup = gap,
- Only leaders produce and make profits:

$$\mu_{i,t} = rac{q_{i,t}}{q_{-i,t}} \ge 1$$
  $\Pi_t(\mu_i) = \left(1 - rac{1}{\mu_i}\right) Y_t$ 

 $\Rightarrow$  gap  $\mu_i$  is payoff-relevant for market *i* 

## Timing



Two innovation-related decisions:

- 1. Sample size s: search intensity
- 2. Arrival rate of innovations

## Timing



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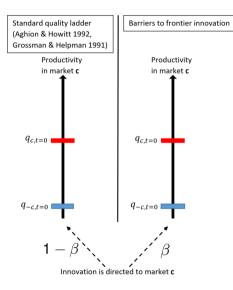
- 1. Sample size s: search intensity
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#### Innovation stage

Barriers to frontier innovation:

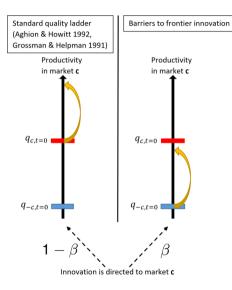
- With probability  $\beta \in (0, 1)$ , only build on follower's technology
- Microfoundations: strategic patents, trade secrets, slow technology diffusion

## Innovation stage: high-gap market



▶ low-gap market

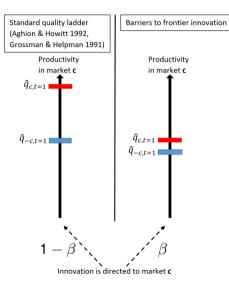
## Innovation stage: high-gap market





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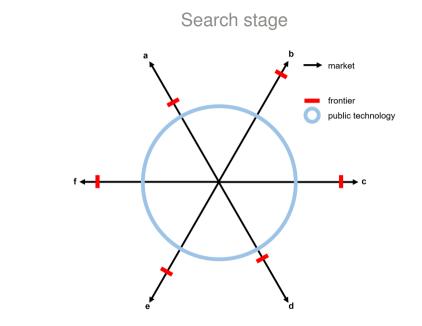
▶ low-gap market

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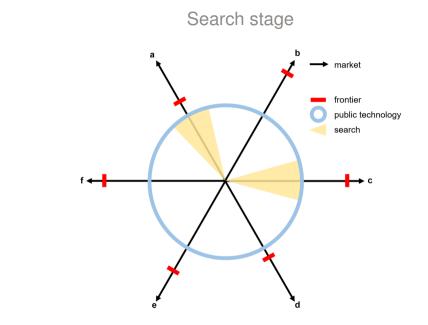
## Timing

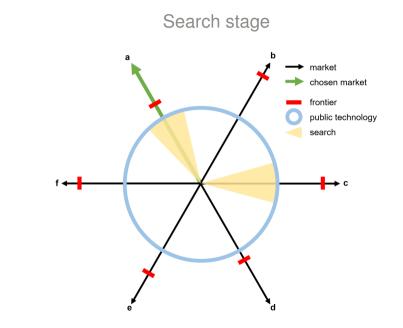


 $\Rightarrow$  search markets to target a less dominant (low-gap) leader



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Full model: Firm dynamics à la Klette & Kortum (2004)

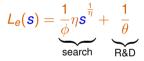
- Innovation by entrants and incumbents
- Firm size distribution

Simplified model: Standard quality ladder à la Grossman & Helpman (1991)

- Innovation by entrants
- 1-product firms

#### Entry

- Free entry
- Labor to search *s* markets and generate 1 innovation:



- $\eta$ : search elasticity
- $\phi$ : search productivity
- $\theta$ : research productivity

free entry condition

#### Steady state

Assumption: Innovation steps drawn from  $\sim Pareto(\alpha)$ 

Proposition

Stationary distribution of leader-follower gaps  $\mu$  is  $\sim \text{Pareto}(\frac{\alpha}{s^*})$ ,

graphical proof

$$\mathit{cdf}(\mu) = \mathsf{1} - \mu^{-rac{lpha}{s^*}}$$

Proposition

Optimal search intensity is

$$oldsymbol{s^*} = \left(rac{\phi/ heta}{4/eta-2-\eta}
ight)^\eta$$

 $\Rightarrow$  Decline in research productivity ( $\theta \downarrow$ ) incentivizes more search:  $s^* \uparrow$ 



#### Parametrization of pre-1980s (full model)

Assigned	Value	Description
ρ	0.01	Discount rate
eta	0.5	Barriers to frontier innovation
$\eta$	0.5	Search elasticity
$\gamma$	0.5	Incumbent R&D elasticity
δ	0.7	Entry externality
L	1	Size of labor force (normalized)

Estim	ated Value	Description	Key moment	Model	Data
$ ilde{ heta}$	0.63	Incumbent R&D productivity	TFP growth	0.0182	0.0182
heta	0.45	Entrant R&D productivity	Entrants' % TFP growth	0.25	0.25
lpha	13.2	Pareto shape of innovations	Firm entry rate	0.13	0.13
$\phi$	25.2	Search productivity	Average markup	1.1	1.1

#### What caused the macro trends since the 1980s?

Find support for 2 explanations:

- 1. **Declining research prod.**  $(\theta \downarrow, \tilde{\theta} \downarrow) \Rightarrow$  can explain sign + magnitude of effects
- 2. Barriers to frontier innovation ( $\beta \uparrow$ )  $\Rightarrow$  can explain sign of effects, but not magnitude

#### Today vs. pre-1980s: Decline in research productivity

- Optimal to search more markets ⇒ more selection, more market power
  - 1. Higher markups & dispersion, higher profit share
- R&D? Higher profit share dominates lower research productivity
  - 2. Higher % of R&D workers 🗸
- Growth? Decline in research productivity dominates increase of R&D workers
  - 3. Productivity growth slows down 🗸
  - 4. Less firm entry 🗸
- Firm size? Entrants' innovation rate drops more than incumbents'
  - 5. Larger and older firms 🗸

## Implications of research productivity $\downarrow$ by 75%

				Change		
	pre-1980 s.s.	2010 s.s.	Model	Data	Sign	Model/Data
Targeted moments						
TFP growth	0.0182	0.0096	-47%	-72%	$\checkmark$	<b>65</b> %
Entrants' % TFP growth	0.25	0.206	-18%	-	-	-
Firm entry rate	0.13	0.064	-51%	-39%	$\checkmark$	131%
Average markup	1.1	1.15	+4.5%	+7%	$\checkmark$	<b>64</b> %
Untargeted moments						
Average firm size	2.16	2.44	+13%	+15%	$\checkmark$	87%
Profit share	0.041	0.084	+105%	+75%	$\checkmark$	140%
R&D workers/labor force	0.076	0.086	+13%	+50%	$\checkmark$	<b>26</b> %

#### Conclusions

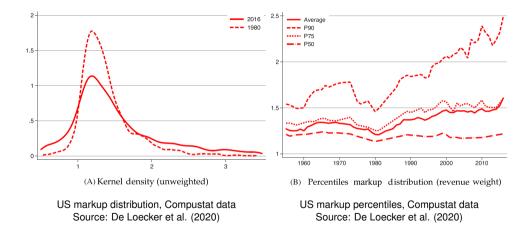
- Directed innovation matters for the macroeconomy, market power and growth
- Ideas got harder to find (Bloom et al. 2020) is a unified explanation for macro trends

## Thank you for your attention!

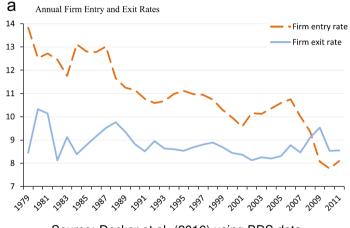
comments: julian.schaerer@econ.uzh.ch

# Appendix

#### Markups increased, especially at the top



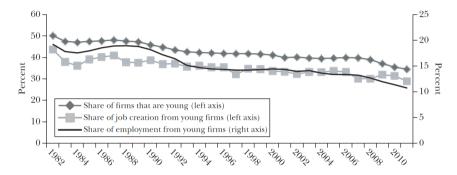
#### Firm entry rate ↓



Source: Decker et al. (2016) using BDS data

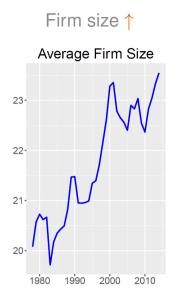
#### Share of young firms $\downarrow$

#### Declining Share of Activity from Young Firms (Firms Age 5 or Less)



Source: Author calculations from the US Census Bureau's Business Dynamics Statistics.

Source: Decker et al. (2014)



Number of workers per firm. Source: Hopenhayn et al. (2018) using BDS data



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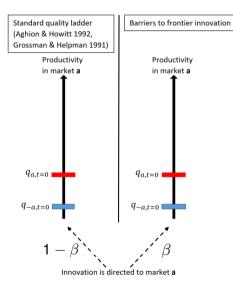
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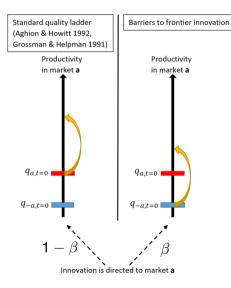
#### Innovation stage: low-gap market



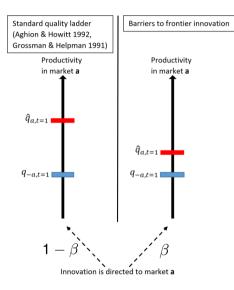
return

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#### Innovation stage: low-gap market



# $\checkmark \qquad (\checkmark)$





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#### Firms

A firm *j* is the collection  $\mathcal{I}(j)$  of markets (or product lines) in which *j* is the leader

- Payoff-relevant state variable:  $\{\mu_i\}_{i \in \mathcal{I}(j)}$
- Firm size  $n_j = |\mathcal{I}(j)|$

Firm dynamics

- Grow if innovate in a line operated by another firm
- Shrink if other firms/entrants innovate in a line  $i \in \mathcal{I}(j)$
- Exit if last product is lost
- New firms enter with 1 product

✓ return

#### Firm Problem I

To search a sample of *s* markets at rate *x*, need to employ

$$L_{\mathcal{S}}(\boldsymbol{s},x) = x \frac{1}{\phi} \eta \boldsymbol{s}^{rac{1}{\eta}}$$

- $\eta \in (0, 1)$ : search elasticity
- $\phi$ : search productivity

✓ return

#### Firm Problem II

To innovate at rate x, a size-n firm needs to employ

$$L_{R\&D}(x|n) = rac{1}{\widetilde{ heta}} \gamma x^{rac{1}{\gamma}} n^{rac{\gamma-\gamma}{\gamma}}$$

- $\gamma \in (0, 1)$ : innovation elasticity
- $\tilde{\theta}$ : incumbent research productivity
- Can show: innovation intensity per product,  $\tilde{x} \equiv x/n$ , is the same for all firms

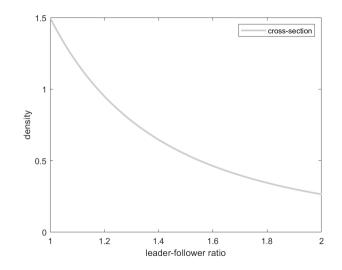
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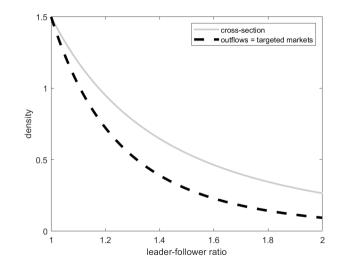
#### Free entry condition

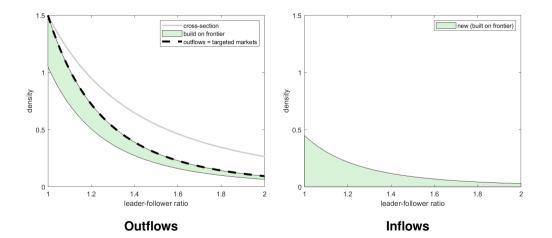
Free entry condition determines mass of entrants  $\mathcal{M}_0$ :

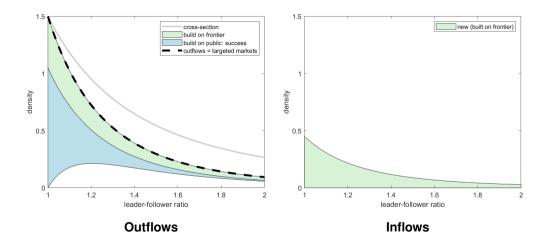
$$0 = \max_{s} \left( \underbrace{1 - \beta + \beta \operatorname{Pr}(\operatorname{innov} > \operatorname{gap}(s))}_{\operatorname{Pr}(\operatorname{success}|\operatorname{innov}), \uparrow \operatorname{in} s} \right) \times \underbrace{\mathbb{E}(V_{\operatorname{new}})}_{\downarrow \operatorname{in} \mathcal{M}_{0}} - \underbrace{L_{e}(s)}_{\uparrow \operatorname{in} s} \times wage$$

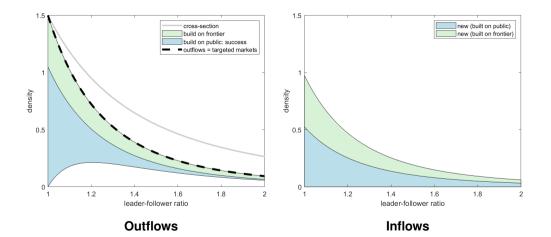
return to entry specification

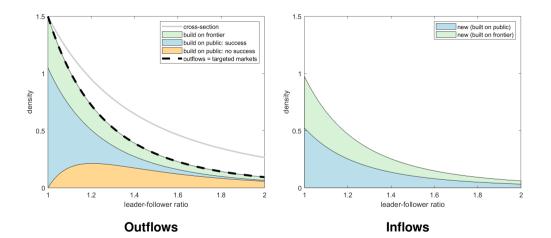


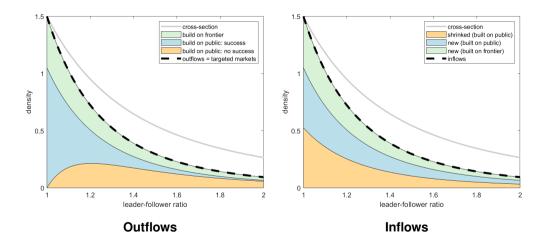


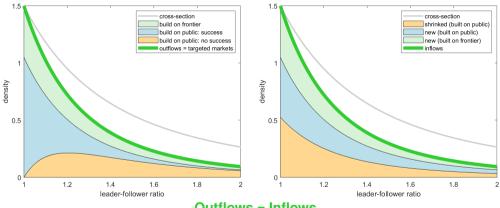






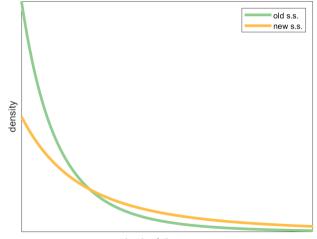






**Outflows = Inflows** 

#### Effect of more search on gap distribution



leader-follower gap

#### Productivity growth: new vs. old steady state

$$g = \underbrace{\left(\tilde{x} + \mathcal{M}_{0}\right)}_{\text{declines}} \times \underbrace{\Pr(\text{success})}_{=\left(1 - \frac{\beta}{2}\right), \text{ constant}} \times \underbrace{\mathbb{E}\left(\ln(\text{StepSize})\right)}_{=\frac{1}{\alpha}, \text{ constant}}$$

#### (where $\mathcal{M}_0$ is the mass of entrants in the full model)

✓ return

Product value function in steady state I

- Define  $v_t(\mu) \equiv \frac{V_t(\mu)}{wage_t}$  as product value normalized by the wage
- $v_t(\mu) = v(\mu)$  constant in steady state
- It holds  $\forall \mu \geq 1$ :

$$\rho \mathbf{v}(\mu) = \underbrace{\left(1 - \frac{1}{\mu}\right) \frac{\mathbf{Y}}{\mathbf{wage}} + \left(\frac{1}{\gamma} - 1\right) \frac{1}{\tilde{\theta}} \gamma(\tilde{\mathbf{x}})^{\frac{1}{\gamma}}}_{\text{instantaneous payoff:}}}_{\text{instantaneous payoff:}} - \underbrace{\left(\tilde{\mathbf{x}} + \mathcal{M}_{0}\right)}_{\text{aggregate $\#$}} \underbrace{\frac{\alpha \mu^{-\alpha - 1}}{\frac{s^{\ast} \mu}{s^{\ast} - 1}}}_{\text{gap-specific correction of hazard rate}} \left(\mathbf{v}(\mu) - \underbrace{\beta \Pr(\lambda \leq \mu) \mathbb{E}\left(\mathbf{v}(\frac{\mu}{\lambda}) | \lambda \leq \mu\right)}_{\text{leader survives but gap shrinks}}\right),$$

#### Product value function in steady state II

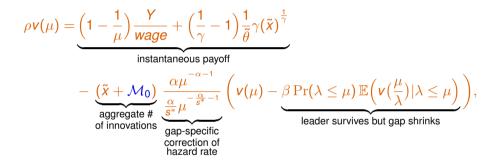
Product value for  $\mu = 1$  simplifies to

$$m{v}(1) = \Big(rac{1}{\gamma} - 1\Big) rac{1}{ ilde{ heta}} \gamma( ilde{x})^{rac{1}{\gamma}} - ig( ilde{x} + \mathcal{M}_0ig) \,m{s}^* \,m{v}(1)$$

• To solve for  $v(\mu)$  for all  $\mu > 1$ , "unravel from below"

### Solving for $\mathcal{M}_0$

Product value function  $\forall \mu \geq 1$ :



Free entry condition reads

$$\left(1-rac{eta}{2}
ight)\mathbb{E}(oldsymbol{v_{new}})=rac{1}{\phi}\etaoldsymbol{s}^{*rac{1}{\eta}}+rac{1}{ heta}(\mathcal{M}_0)^{\delta}$$

where we know the formula for  $s^* \Rightarrow \text{Easy}_{35/18}$  to solve for  $\mathcal{M}_0$  numerically