

# INTANGIBLE CAPITAL, TANGIBLE MISALLOCATION

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Stepan Gordeev

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# SUMMARY

## 1. Role of intangible assets in production is rising

- Tangible K: buildings, machines...
- Intangible K: IT, brand value, organizational structure...
- $\frac{\text{intangible K}}{\text{tangible K}}$  in the US: 0.20  $\rightarrow$  0.52 since 1948

## 2. Misallocation in the US is rising

- **THIS PAPER:** role of intangibles in generating observed misallocation
- **DATA:** US Compustat 1987-2017
- **MISALLOCATION MEASUREMENT:** Hsieh and Klenow '09 + intangibles

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## PREVIEW OF RESULTS

1. Ignoring intangibles leads to **overestimation** of both the level and growth of misallocation
  - Corrected misallocation is 35% lower
  - Corrected **misallocation** has not increased in the last two decades
2. Intan-intensive sectors have 75% **higher misallocation** than tan-intensive sectors
  - Uncertain intangible productivity + variable markups can generate one third of the difference

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## LITERATURE

- **ROLE OF INTANGIBLES:** Corrado and Hulten (2010), Haskel and Westlake (2017), McGrattan (2017), Chen (2014), Peters and Taylor (2017), Döttling and Perotti (2017)
- **MISALLOCATION:** Hsieh and Klenow (2009), Restuccia and Rogerson (2008), David and Venkateswaran (2017), Bils, Klenow, and Ruane (2018)
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# OUTLINE

1. Data
2. Estimating Misallocation
3. Potential Mechanisms
4. Variable Markup Model
5. Conclusion

DATA

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## DATA OVERVIEW

- Annual traded firm balance sheets from Compustat
  - 1987-2017
  - Sectoral output, tangible, and intangible investment deflators from BLS
- Production:  $y = ak_T^{\alpha_T} v^{1-\alpha_T}$
- Measure variable input  $v$  with cost of goods sold
- Construct tangible capital stock  $k_{T,f,t}$  from PPE investment  $x_{T,f,t-1}$

$$k_{T,f,t} = (1 - \delta_{T,f})k_{T,f,t-1} + x_{T,f,t-1}$$

- Implied firm-specific depreciation  $\delta_{T,f}$
- Construct intangible capital stock  $k_{I,f,t}$  using method from Falato, Kadyrzhanova and Sim (2013) and Peters and Taylor (2017)
  - Consists of 3 components

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# INTANGIBLE K COMPONENTS

## 1. Knowledge capital

$x_{\text{know},f,t}$  is R&D investment,  $\delta_{\text{know}} = 0.15$

$$\text{KNOWLEDGE}_{f,t} = (1 - \delta_{\text{know}})\text{KNOWLEDGE}_{f,t-1} + x_{\text{know},f,t-1}$$

$$\text{KNOWLEDGE}_{f,1} = \frac{x_{\text{know},f,1}}{\delta_{\text{know}}}$$

## 2. Organizational capital

$x_{\text{org},f,t}$  is “sales, general and administrative expense”,  $\delta_{\text{org}} = 0.2$ , weight investment by 0.3

$$\text{ORGANIZATION}_{f,t} = (1 - \delta_{\text{org}})\text{ORGANIZATION}_{f,t-1} + 0.3 \cdot x_{\text{org},f,t-1}$$

## 3. External purchases

$\text{EXTERNAL}_{f,t}$  = externally acquired intangibles: patents, trademarks, goodwill, etc

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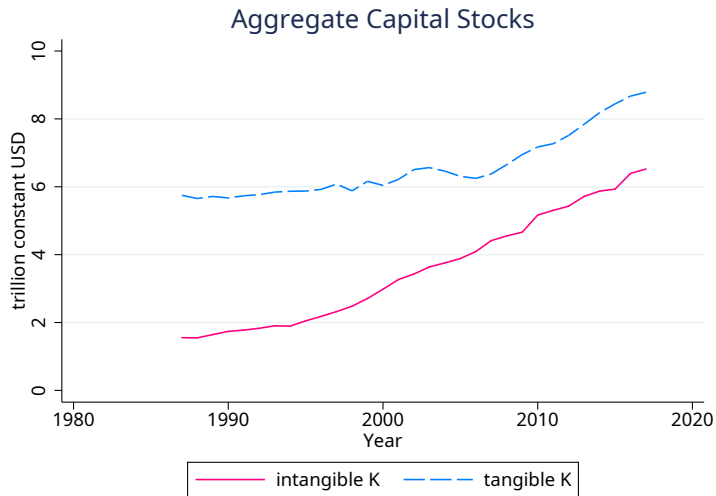
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## INTANGIBLE K

- Intangible capital  $k_{I,f,t}$  consists of 3 components

$$k_{I,f,t} = \text{KNOWLEDGE}_{f,t} + \text{ORGANIZATION}_{f,t} + \text{EXTERNAL}_{f,t}$$

# AGGREGATE INTANGIBLES



## ESTIMATING MISALLOCATION

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## OVERVIEW

- **HK+I**: Hsieh and Klenow (**HK**) with an additional input: intangible  $K_I$
- Final good  $Y$  produced competitively from  $S$  intermediates

$$Y = \prod_{s=1}^S Y_s^{\theta_s}$$

- Intermediate  $Y_s$  produced competitively from varieties of  $F_s$  firms

$$Y_s = \left( \sum_{f=1}^{F_s} y_{s,f}^{\frac{\sigma-1}{\sigma}} \right)^{\frac{\sigma}{\sigma-1}}$$

- Assume  $\sigma = 3$



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## VARIETY PRODUCER

- Each variety  $(s, f)$  produced by monop. competitive firm using
  - variable input  $v_{s,f}$  (paid  $w$ )
  - tangible capital  $k_{T,s,f}$  (paid return  $r_T$ )
  - intangible capital  $k_{I,s,f}$  (paid return  $r_I$ )
- Subject to distortions  $(1 - \tau_{Y,s,f})$ ,  $(1 + \tau_{T,s,f})$ , and  $(1 + \tau_{I,s,f})$

$$\max_{p_{s,f}, v_{s,f}, k_{T,s,f}, k_{I,s,f}} (1 - \tau_{Y,s,f}) p_{s,f} y_{s,f} - w v_{s,f} - (1 + \tau_{T,s,f}) r_T k_{T,s,f} - (1 + \tau_{I,s,f}) r_I k_{I,s,f}$$

s.t.

$$y_{s,f} = a_{s,f} k_{T,s,f}^{\alpha_{T,s}} k_{I,s,f}^{\alpha_{I,s}} v_{s,f}^{1-\alpha_{T,s}-\alpha_{I,s}}$$

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## ESTIMABLE EXPRESSIONS

- Express firm  $f$ 's productivity  $a_{s,f}$  with observables:

$$a_{s,f} \propto \frac{(p_{s,f} y_{s,f})^{\frac{\sigma}{\sigma-1}}}{k_{T,s,f}^{\alpha_{T,s}} k_{l,s,f}^{\alpha_{l,s}} V_{s,f}^{1-\alpha_{T,s}-\alpha_{l,s}}}$$

- Revenue productivity reflects distortions:

$$\text{tfpr}_{s,f} \propto \frac{(1 + \tau_{T,s,f})^{\alpha_{T,s}} (1 + \tau_{l,s,f})^{\alpha_{l,s}}}{(1 - \tau_{Y,s,f})}$$

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## REALLOCATION GAIN

- Distorted sectoral TFP:

$$TFP_s = \left( \sum_{f=1}^{F_s} \left( a_{s,f} \frac{TFPR_s}{tfpr_{s,f}} \right)^{\sigma-1} \right)^{\frac{1}{\sigma-1}}$$

- Efficient sectoral TFP if equalize distortions:

$$TFP_{s,eff} = \left( \sum_{f=1}^{F_s} a_{s,f}^{\sigma-1} \right)^{\frac{1}{\sigma-1}}$$

- Reallocation gain:

$$GAIN_s = \frac{TFP_{s,eff}}{TFP_s}$$

- Aggregate output gain from equalizing distortions:  $GAIN = \prod_{s=1}^S (GAIN_s)^{\theta_s}$

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## FACTOR SHARES

- Cannot identify distortions and factor shares separately, assume sectoral input shares not distorted
- Measure  $1 - \alpha_{T,s} - \alpha_{I,s}$  as  $\frac{\text{COGS}}{\text{SALE}}$  for each sector-year
- Pin down  $\frac{\alpha_{T,s}}{\alpha_{I,s}}$  from  $\frac{K_{T,s}r_T}{K_{I,s}r_I}$ 
  - rental rates
- Alternatively, use  $\alpha_{T,s}$  and  $\alpha_{I,s}$  from BEA
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- Alternatively, use  $\alpha_{T,s}$  and  $\alpha_{I,s}$  from BEA
- Reallocation gains not sensitive to alternative  $\alpha_x$  and  $r_x$

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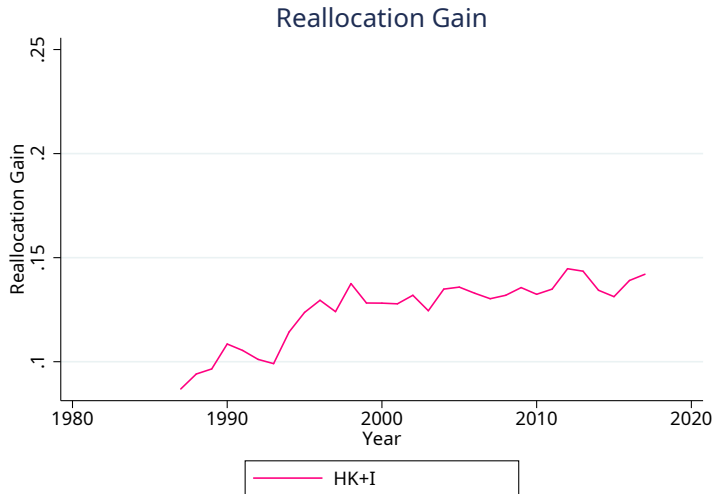
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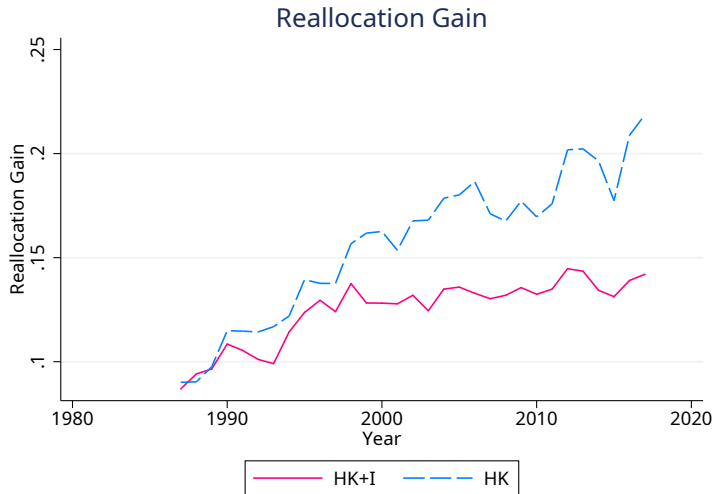
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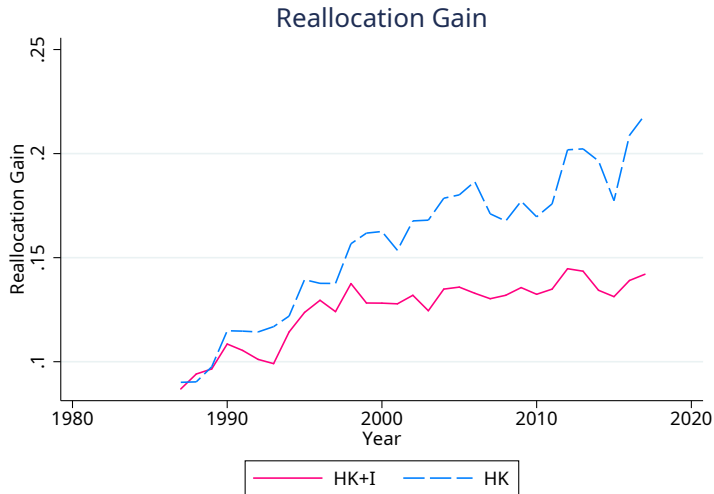
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## WHY DOES HK OVERESTIMATE?

- Over/underestimation of GAIN driven by HK's over/underestimation of  $a_{s,f}$

$$a_{s,f} = \frac{y_{s,f}}{\underbrace{k_T^{\alpha_{T,s} + \alpha_{l,s}}}_{a^{HK}\downarrow} \underbrace{k_l^{\alpha_{l,s}}}_{a^{HK}\uparrow} v^{1-\alpha_{T,s}-\alpha_{l,s}}}$$

- If  $a_{s,f}, 1 + \tau_{T,s,f}, 1 + \tau_{l,s,f}$  are jointly lognormal (means  $\mu_x$ , (co)variances  $\sigma_x$ ),

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## MISALLOCATION AND INTANGIBLE SHARE

- Are intangible-intensive sectors more distorted or less?
- Split sectors into high/low-intangible groups by their  $\frac{K_{I,s}}{K_{T,s}}$  in 2017
  - Sales Shares
- Evaluate misallocation cost in each group

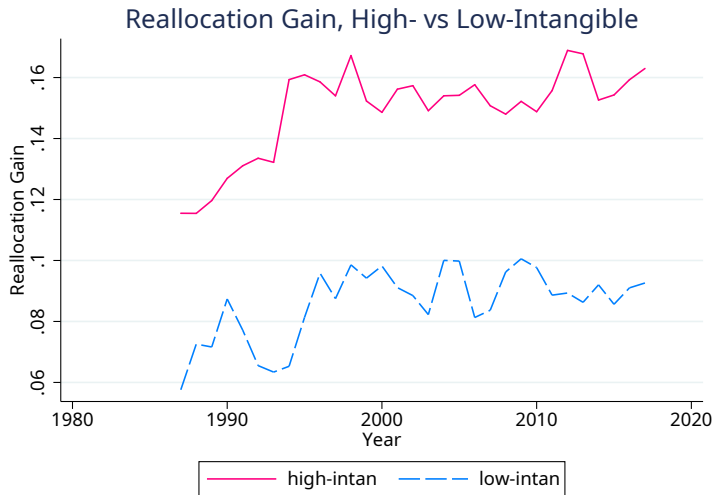
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## MISALLOCATION BY INTAN INTENSITY



- High-intan: 11.5% → 16.3%, low-intan: 5.8% → 9.3%

## POTENTIAL MECHANISMS

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## THE MOST AND THE LEAST INTAN-INTENSIVE SECTORS

$\frac{K_L}{K_T}$	Rank	Sector	$\frac{K_L}{K_T}$
	1	Motion picture and sound recording	4.85
	2	Computer systems design	4.81
	3	Publishing industries	4.47
	$\vdots$		
	26	Transportation equipment	1.14
	27	Printing	0.83
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	50	Oil and gas extraction	.05
	51	Utilities	.05
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## HIGH- VS LOW-INTAN SECTORS

	high-intan	low-intan	mechanism
$K_T$ factor share	0.083	0.230	
$K_I$ factor share	0.260	0.124	
V factor share	0.657	0.646	
SD of demeaned log sales	2.53	1.95	winner-take-most
Median firm growth rate SD	0.24	0.17	uncertainty
Median leverage ratio	0.20	0.32	fin. frictions
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## VARIABLE MARKUP MODEL

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- Uncertain productivity of intan investment + markup heterogeneity
- Add intangibles to PE version of Edmond, Midrigan and Xu (2018)
- Small sector: sectoral variables do not affect prices
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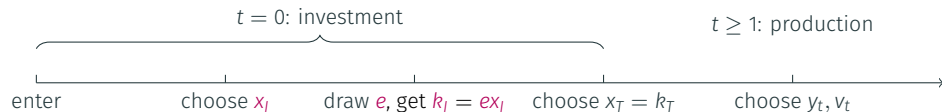
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## TIMELINE

- All firms enter, make one-time investment  $x_I$ , draw  $e$  to get  $k_I = ex_I$ , choose  $k_T$ , then produce



## DEMAND

- Sectoral composite good produced from varieties  $\omega$  using the Kimball aggregator

$$\int_0^{F_t} \Upsilon \left( \frac{y_t(\omega)}{Y_t} \right) d\omega = 1$$

- Use the functional form for  $\Upsilon$  from Klenow and Willis (2016)

$$\Upsilon \left( \frac{y_t(\omega)}{Y_t} \right) = 1 + (\tilde{\sigma} - 1) \exp \left( \frac{1}{\varepsilon} \right) \varepsilon^{\frac{\tilde{\sigma}}{\varepsilon} - 1} \left( \Gamma \left( \frac{\tilde{\sigma}}{\varepsilon}, \frac{1}{\varepsilon} \right) - \Gamma \left( \frac{\tilde{\sigma}}{\varepsilon}, \frac{(y_t(\omega)/Y)^{\frac{\varepsilon}{\tilde{\sigma}}}}{\varepsilon} \right) \right)$$

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- Produce with

$$y_t = \underbrace{k_T^{\alpha_T} k_I^{\alpha_I}}_{\equiv z} v_t^{1-\alpha_T-\alpha_I}$$

- Static production problem in  $t \geq 1$ :

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- Static production problem in  $t \geq 1$ :

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- Given  $k_I$  draw, choose tangible capital to max discounted profits less cost of investment

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# CALIBRATION

- Calibrate to the low-intangible group of sectors
- $e \sim \text{lognormal}(0, \iota^2)$

Parameter	Symbol	Value	Rationale
discount rate	$\beta$	0.96	standard
death rate	$\delta$	0.10	standard
intan inv convexity	$\phi$	1.1	ad hoc
LOW-INTAN SECTOR			
tangible share	$\alpha_{T,l}$	0.230	compensation share
intangible share	$\alpha_{I,l}$	0.124	compensation share
average elasticity	$\tilde{\sigma}$	29	aggregate markup
superelasticity	$\varepsilon$	6	sales-COGS relationship
productivity dispersion	$\iota$	2.25	sales distribution
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tangible share	$\alpha_{T,h}$	0.083	compensation share
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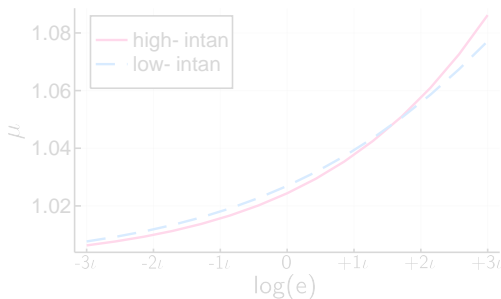
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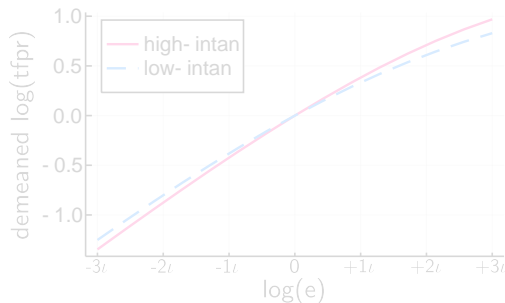
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- Intan investment productivity dispersion  $\rightarrow$  size dispersion  $\rightarrow$  markup dispersion  $\rightarrow$  measured misallocation

Markup by intan group



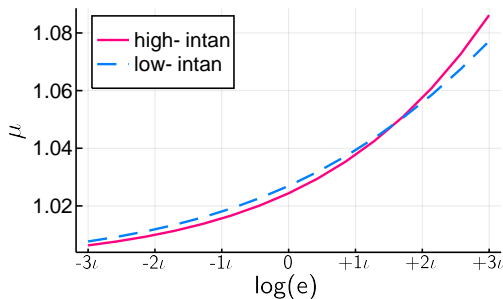
tfpr by intan group



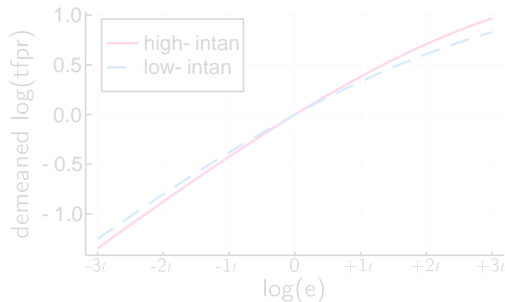
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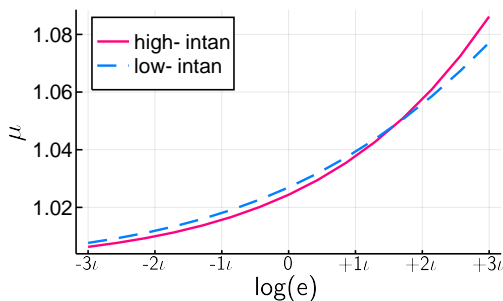
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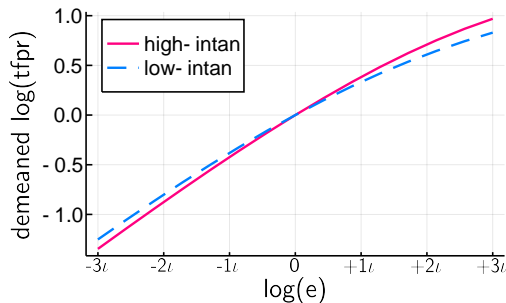
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## MEASURED MISALLOCATION

- Apply HK+I to model-generated data

	low-intan		high-intan	
	data	model	data	model
HK+I TFP GAIN	9.3%	17.0%	16.3%	19.7%
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- Ignoring intangibles **overestimates misallocation** by 50%
  - and its growth by 130%, **explaining ~all measured deterioration** in last 20 years
- High and deteriorating misallocation driven by **intangible-intensive** sectors
- Investment uncertainty + variable markups can explain  $\frac{1}{3}$  of the difference

- **FUTURE WORK**

- Figure out what mechanisms drive the high misallocation in intan-intensive sectors

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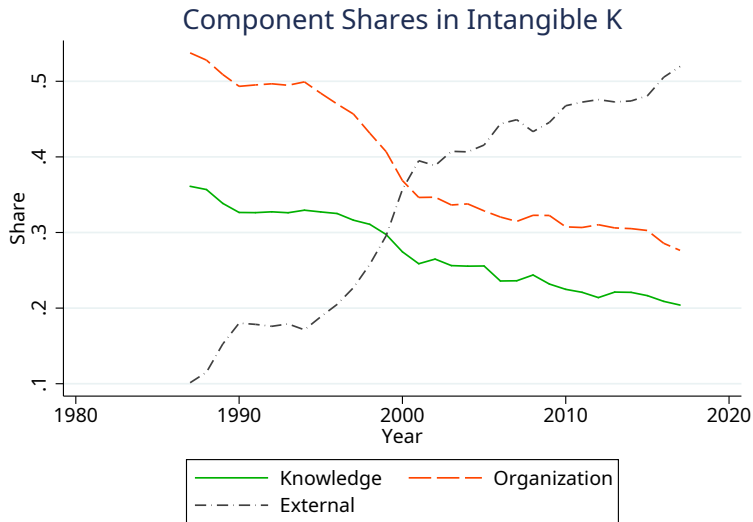
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# INTANGIBLE K COMPONENTS



## PRODUCTIVITIES

$$\text{tfpr}_{s,f} = \frac{\sigma}{\sigma - 1} \left( \frac{(1 + \tau_{T,s,f})r_T}{(1 - \tau_{Y,s,f})\alpha_{T,s}} \right)^{\alpha_{T,s}} \left( \frac{(1 + \tau_{I,s,f})r_I}{(1 - \tau_{Y,s,f})\alpha_{I,s}} \right)^{\alpha_{I,s}} \times$$

$$\left( \frac{W}{(1 - \tau_{Y,s,f})(1 - \alpha_{T,s} - \alpha_{I,s})} \right)^{1 - \alpha_{T,s} - \alpha_{I,s}}$$

$$a_{s,f} = \left( p_s y_s^{\frac{1}{\sigma}} \right)^{\frac{\sigma}{1 - \sigma}} \frac{(p_{s,f} y_{s,f})^{\frac{\sigma}{\sigma - 1}}}{k_{T,s,f}^{\alpha_{T,s}} k_{I,s,f}^{\alpha_{I,s}} v_{s,f}^{1 - \alpha_{T,s} - \alpha_{I,s}}}$$

$$\text{TFPR}_s = \left( \frac{\text{MRPK}_{T,s}}{\alpha_{T,s}} \right)^{\alpha_{T,s}} \left( \frac{\text{MRPK}_{I,s}}{\alpha_{I,s}} \right)^{\alpha_{I,s}} \left( \frac{\text{MRPV}_s}{1 - \alpha_{T,s} - \alpha_{I,s}} \right)^{1 - \alpha_{T,s} - \alpha_{I,s}}$$

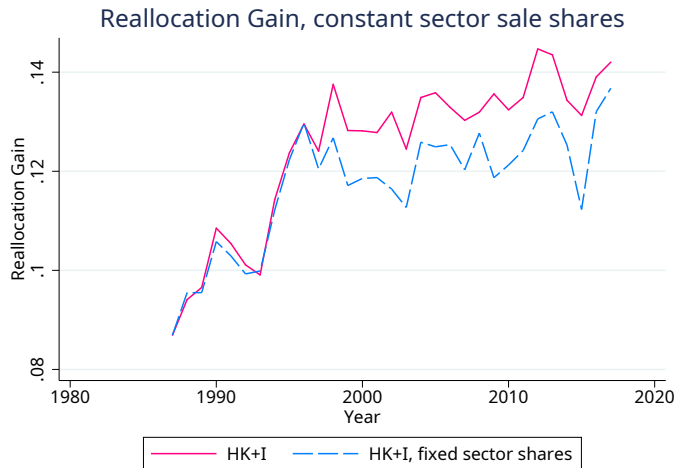
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## RENTAL RATES

- Assume  $r = 0.05 + \delta$
- $r_T = 0.12$ ,  $r_I = 0.22$
- Rates matter for factor shares, but not (directly) for reallocation gains

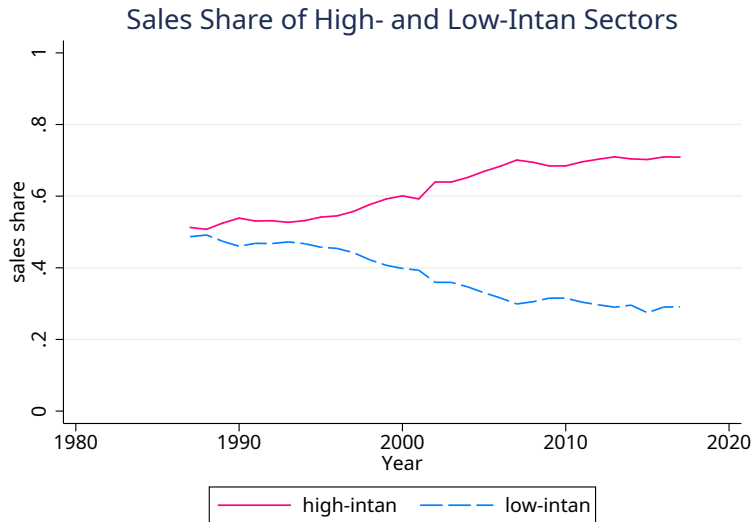
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## REALLOCATION OF SALES ACROSS SECTORS

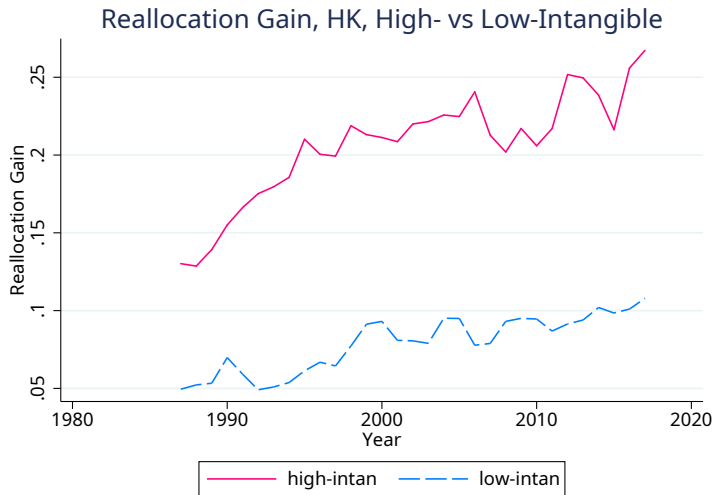


- $\frac{1}{10}$  of increase due to reallocation of revenue to more distorted sectors

## SALES SHARES BY GROUP

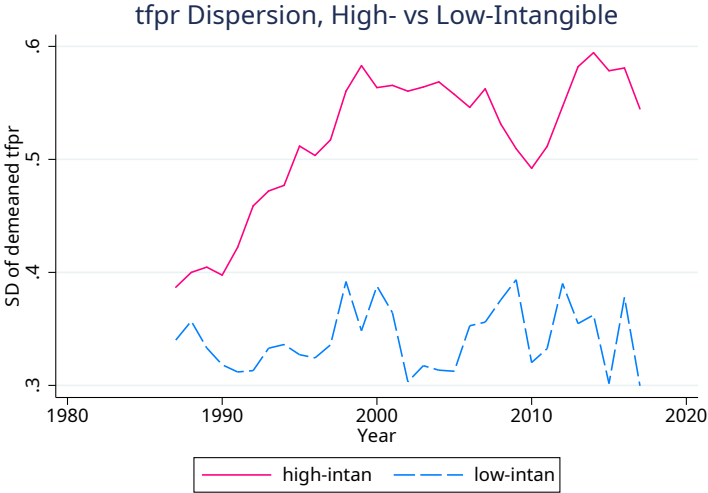


## MISALLOCATION BY INTAN INTENSITY, HK



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# DISPERSION BY INTAN INTENSITY





# TARGETED MOMENTS

	low-intan (targeted)		high-intan (non-targeted)	
	data	model	data	model
AGGR. MARKUP				
<i>M</i>	1.048	1.048	1.178	1.053
RELATIVE COGS				
decile				
1st	-4.20	-2.47	-5.17	-3.70
2nd	-3.01	-1.64	-4.14	-2.77
3rd	-2.06	-1.07	-3.52	-2.16
4th	-1.62	-0.77	-2.79	-1.67
5th	-1.22	-0.50	-2.32	-1.24
6th	-0.85	-0.21	-1.57	-0.78
7th	-0.34	0.11	-1.20	-0.39
8th	0.06	0.35	-0.47	0.05
9th	0.75	0.79	0.27	0.54
10th	1.64	1.31	2.00	1.66
RELATIVE SALES				
percentile				
20th	-3.22	-2.83	-4.23	-3.47
40th	-1.72	-1.73	-2.70	-2.25
60th	-0.77	-0.82	-1.54	-1.22
80th	0.24	0.18	-0.23	-0.06
90th	0.98	0.89	0.74	0.79
95th	1.43	1.46	1.49	1.47
99th	2.28	2.46	2.77	2.69

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