Entry Costs and the Macroeconomy

Germán Gutiérrez   Callum Jones   Thomas Philippon
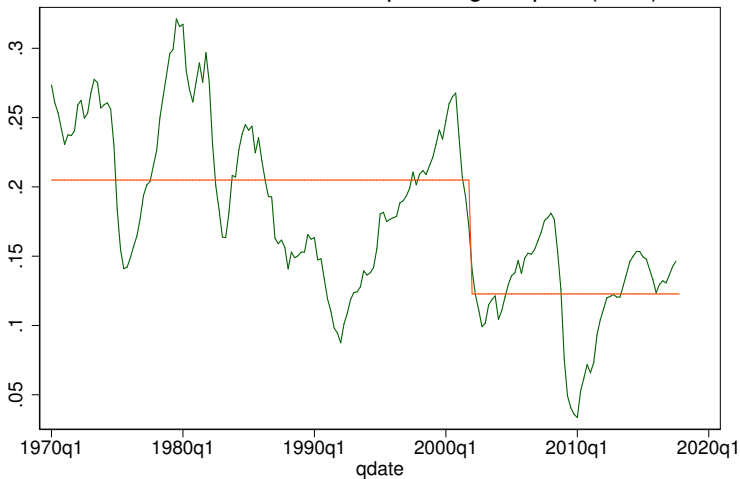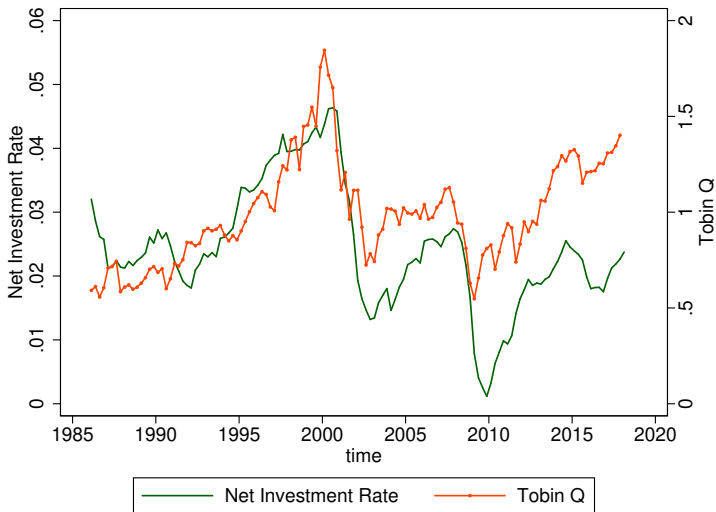
NYU, IMF[1], NBER, CEPR

July 2019, NBER

Fact 1: Investment is Low #1
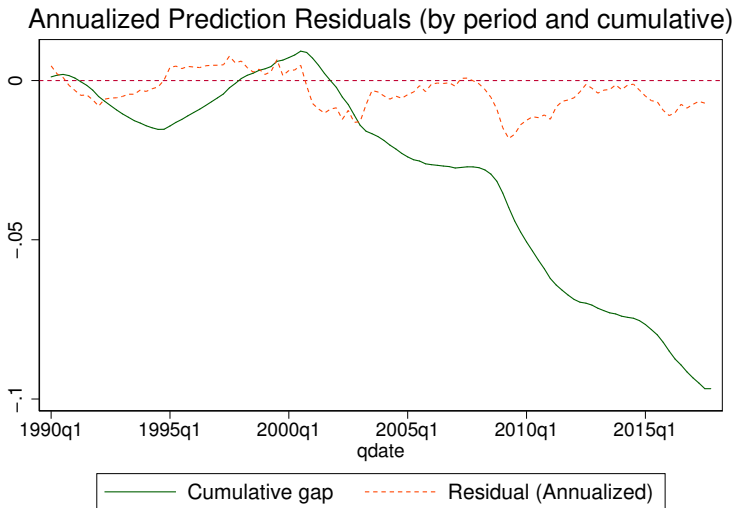

Net Investment to Net Operating Surplus (NFB)

Gutiérrez and Philippon '17, and Lee et al '16.

Fact 1: Investment is Low #2

Gutiérrez and Philippon '17, and Lee et al '16.

Fact 1: Investment is Low #3

Annualized Prediction Residuals (by period and cumulative)

Legend: Cumulative gap —— Residual (Annualized) --------

Fact 2: Concentration & Profits are High

Decker et al '14, Gomme et al '11, Furman '15
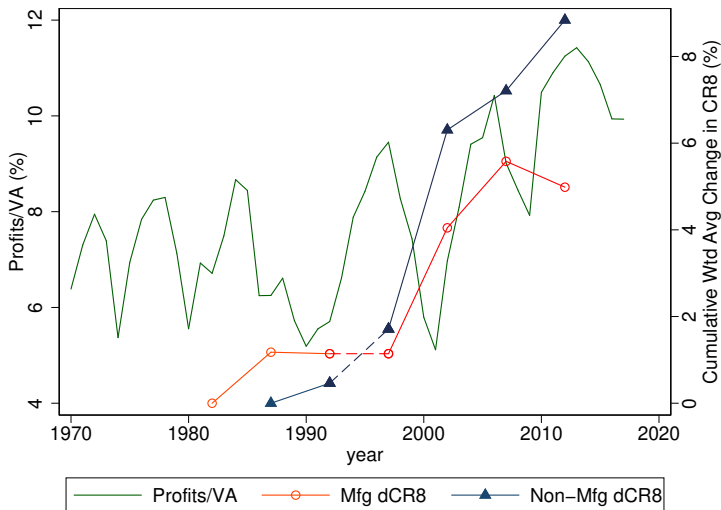
Fact 2: Concentration in Cross-Section
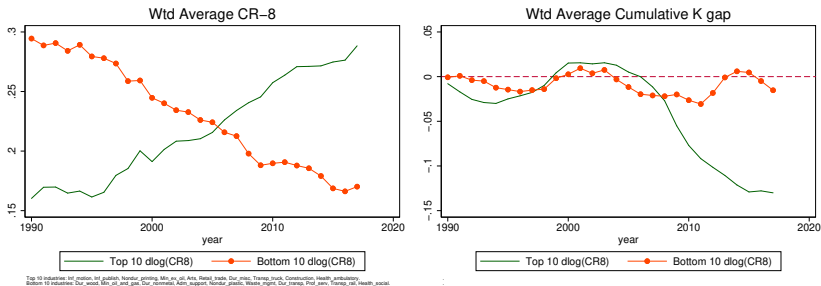
- More concentrating industries have larger capital gaps
- Grullon et al '18. Concentrating industries have higher profits

Fact #3: Entry Rates Have Fallen

Interpretation of the Evidence

- Interpretation remains controversial:

    - Endogeneity of profits/entry/concentration/investment

    - Need a model to interpret the evidence

    - Macro models so far focus on effect of (assumed) higher markups

- We use a fully specified model with competing drivers of facts

    - Use the data to derive a series for entry costs and study their importance

    - Provide direct empirical evidence of what these entry costs are

Approach and Findings

- Develop a model with monopolistic competition and firm entry
  - Entry decision based on expected profits

- Estimate model on industry/aggregate data, 1989-2015, to get:
  - shocks and relationship between entry, competition, investment
  - Identified entry costs correlate with regulation/M&A

- Use the model to study aggregate implications, 1989 to 2015
  - About 10% of entry driven by shocks to demand beliefs
  - An increase in aggregate firm entry costs from 2003
  - Absent this increase, C_t: 6.5% higher, K_t: 6% higher

Model

Goods Producers

- Continuum of j industries

- Industry output aggregated by a perfectly competitive firm

$$Y_t = \left(\int_0^1 (D_{j,t} Y_{j,t})^{\frac{\sigma-1}{\sigma}} \mathrm{d}j \right)^{\frac{\sigma}{\sigma-1}}$$

  where $D_{j,t}$ is an industry demand shock following an AR(1)

- D_j is "expected" steady-state demand

- Production function is Cobb-Douglas

$$Y_{j,t} = K_{j,t}^{\alpha} \ell_{j,t}^{1-\alpha}$$

  - Investment follows Q-theory

Goods-Producers Entry

- The number of existing firms $N_{j,t}$ evolves by

$$N_{j,t+1} = (1 - \delta^n)N_{j,t} + n_{j,t}$$

- Firms pay entry input $\kappa_{j,t}$ at price $p_{j,t}^e$ to become active in $t+1$

$$p_{j,t}^e = (\kappa_{j,t} n_{j,t})^{\phi^n}$$

- Let $V_{j,t}^e$ is the value of a firm in industry j. Firms enter until

$$\mathbb{E}_t \Lambda_{t+1} V_{j,t+1}^e = p_{j,t}^e \kappa_{j,t}$$

- Industry-level output

$$Y_{j,t} = N_{j,t-1}^{\frac{\varepsilon_j}{\varepsilon_j - 1}} y_{j,t}$$

Households and Monetary Policy

- Households choose C_t, ℓ_t and savings to

$$\max \; \mathbb{E}_t \sum_{t=0}^{\infty} \beta^t \left(\frac{C_t^{1-\gamma}}{1-\gamma} - \frac{\ell_t^{1+\psi}}{1+\psi} \right)$$

  subject to the budget constraint:

$$S_t + P_t C_t \leq \tilde{R}_t S_{t-1} + W_t \ell_t$$

- Add sticky prices and wages, and monetary policy:

$$r_t = \max \left[0, \; \phi_i r_{t-1} + \phi_p \pi_t^p + \phi_y (y_t - y_t^F) + \phi_g \ln \left(\frac{Y_t/Y_{t-1}}{Y_t^F/Y_{t-1}^F} \right) \right]$$

- r_t subject to the ZLB

Shocks

- Industry and aggregate shocks:

  - Entry-cost

  - Productivity

  - Inflation equations (Phillips curves)

  - Valuation of corporate assets

- Industry-specific shocks:

  - Industry demand

  - Regime shift in industry demand beliefs, between 1995 and 2000

- Aggregate-specific shocks:

  - Household discount factor

  - Monetary policy

Regime Shift in Demand Beliefs Before 2000

- Noisy entry: entry uncorrelated with future demand. Evidence:
  - Doms (2004): high dotcom growth rates
  - Hogendorn (2011): excess entry into Telecom sector
  - Substantial inflows into VC funds focused on dotcom

- We capture this as variations in beliefs about D_j, where

$$\log D_{j,t} = (1 - \rho_d) \log D_j + \rho_d \log D_{j,t-1} + \sigma_d \varepsilon_{j,t}^d$$

  - Before 1995 and after 2000, $D_j = 1$ for all industries
  - Between 1995 and 2000, D_j is a free parameter

Estimation

Estimation #1

- Bayesian/likelihood approach

- Data:

  - <u>Industry</u>: annual data, 1989 to 2015, for 43 industries on

    - Q Details

    - Concentration Ratio   Details

    - Nominal Output, Capital, Prices   Details

  - <u>Aggregate</u>: quarterly data, 1989Q1 to 2015Q1, on

    - Fed Funds rate / Inflation / Expected ZLB Durations

    - Consumption / Investment / Employment

    - Concentration Ratio   Details

Estimation #2

- Calibrated Parameters
  - Elasticity ε_j calibrated to NOS_j/Y_j [Distribution] [Other Parameters]

- Estimated Parameters:

  - Industry:
    - Shock processes, AR(1)
    - Expected demand regimes D_j
    - ϕ^n and σ

  - Aggregate:
    - Shock processes, AR(1)
    - Taylor Rule parameters

- Three main challenges:
  - (i) ZLB, (ii) demand expectations, (iii) industry data $\times$ 43

Challenge 1: Solution Method for ZLB

- The ZLB implies a time-varying representation for our model

$$\mathbf{A}_t x_t = \mathbf{B}_t x_{t-1} + \mathbf{D}_t \mathbb{E}_t x_{t+1} + \mathbf{F}_t \varepsilon_t$$

- This gives rise to a time-varying VAR solution

$$x_t = \mathbf{Q}_t x_{t-1} + \mathbf{G}_t \varepsilon_t$$

- Expected ZLB durations pin down $\mathbf{Q}_t$ and $\mathbf{G}_t$
    - Kulish et al '17, Jones '17, Guerreri and Iacoviello '16.

- We use survey data on durations from NYFed   (ZLB Durations)

Challenge 2: Solution for Demand Regime Shifts #1

- Suppose the regime which is driving the observables is:

$$\mathbf{A}x_t = \mathbf{B}x_{t-1} + \mathbf{D}\mathbb{E}_t x_{t+1} + \mathbf{F}\varepsilon_t$$

  - For example, steady-state demand in industry j, D_j, is low

- But agents believe that a $*$ regime is true:

$$\mathbf{A}^* x_t = \mathbf{C}^* + \mathbf{B}^* x_{t-1} + \mathbf{D}^* \mathbb{E}_t x_{t+1} + \mathbf{E}^* \varepsilon_t.$$

  - For example, steady-state demand in industry j, D_j, is high

- We seek a solution of the form (Kulish and Pagan '16)

$$x_t = \tilde{\mathbf{Q}} x_{t-1} + \tilde{\mathbf{G}} \varepsilon_t$$

Challenge 2: Solution for Demand Regime Shifts #2

- To find the solution, expectations must satisfy

$$\mathbb{E}_t x_{t+1} = \mathbf{Q}^* x_t$$

- Substitute this into $\mathbf{A} x_t = \mathbf{B} x_{t-1} + \mathbf{D}\mathbb{E}_t x_{t+1} + \mathbf{F}\varepsilon_t$ to get:

$$\tilde{\mathbf{Q}} = [\mathbf{A} - \mathbf{D}\mathbf{Q}^*]^{-1}\mathbf{B}$$

$$\tilde{\mathbf{G}} = [\mathbf{A} - \mathbf{D}\mathbf{Q}^*]^{-1}\mathbf{F}$$

- With $x_t = \tilde{\mathbf{Q}} x_{t-1} + \tilde{\mathbf{G}}\varepsilon_t$, we can form the state-space

- So beliefs about demand map to $\tilde{\mathbf{Q}}$ and $\tilde{\mathbf{G}}$ and the likelihood

Challenge 3: Likelihood Function With Industry Data #1

- We want to use industry-level data in identification

- Problem: Infeasible
  - 43 industries, 10+ state variables each, aggregate shocks, ZLB

- Recall our model is

$$x_t = \left[x_t^1,\ x_t^2, \ldots \right]' = \mathbf{Q}_t x_{t-1} + \mathbf{G}_t \varepsilon_t$$

- Solution: Write the industry j level variables as

$$x_t^j = \underbrace{\mathbf{Q} x_{t-1}^j + \mathbf{G} \varepsilon_t^j}_{\text{industry-level component}} + \underbrace{\mathbf{Q}_t^a x_{t-1}^* + \mathbf{G}_t^a \varepsilon_t^*}_{\text{aggregate component}}$$

Challenge 3: Likelihood Function With Industry Data #2

- Industry j level variables are

$$x_t^j = \underbrace{\mathbf{Q}x_{t-1}^j + \mathbf{G}\varepsilon_t^j}_{\text{industry-level component}} + \underbrace{\mathbf{Q}_t^a x_{t-1}^* + \mathbf{G}_t^a \varepsilon_t^*}_{\text{aggregate component}}$$

- Express industry-level data <u>relative</u> to average $\bar{x}_t = \int x_t^j \mathrm{d}j$

- Assuming average shock $\int \varepsilon_t^j \mathrm{d}j = 0$, then

$$x_t^j - \bar{x}_t = \mathbf{Q}\left(x_{t-1}^j - \bar{x}_{t-1}\right) + \mathbf{G}\varepsilon_t^j$$

- Shocks iid: separate the likelihood into 43 industry components

Treatment of Industry-Level Data

- Express industry-level data relative to aggregate

- For each industry series

    1. Compute a full set of time effects

    2. Subtract a industry-specific fixed effect and trend

    3. Work with the residuals

- Exploit these relative changes in identification

Industry-Level Observable: Info Publishing, 1995 to 2005
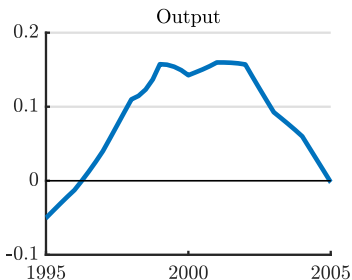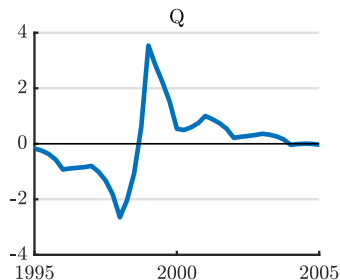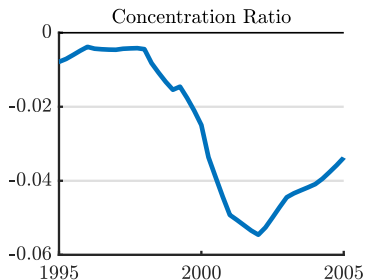
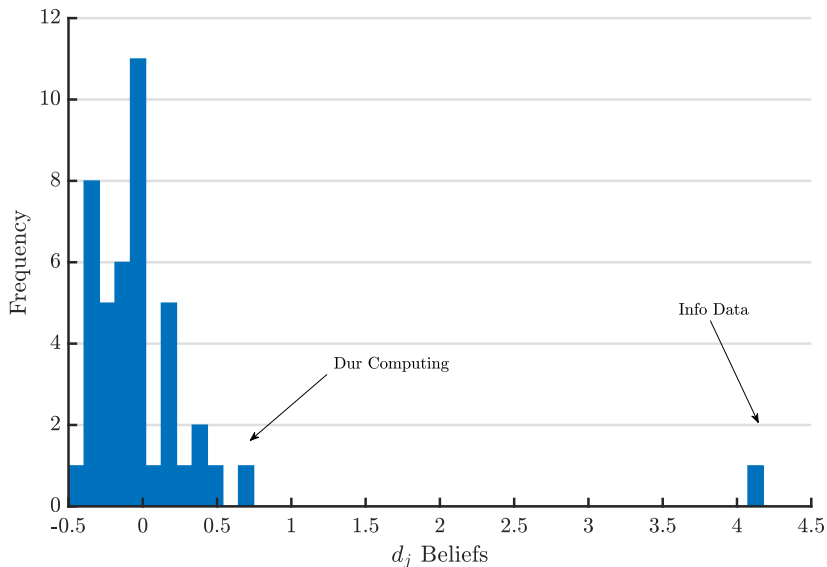# Industry and Aggregate Implications

Industry-Level Estimates

	Posterior		
Parameter	Median	10%	90%
ϕ^n	1.55	1.08	2.32
σ	0.40	0.38	0.43

Log Demand Regimes (2 of 43)			
Durable Comp.	1.6	0.2	2.9
Info Data	4.1	3.1	4.4

Full Estimates

Mode of D_j Beliefs for 43 Industries

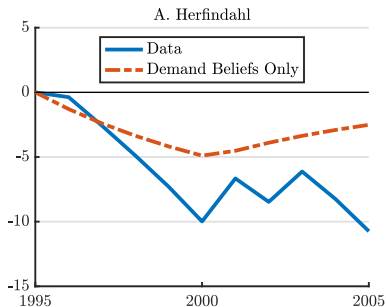## Industry Identification of Estimate of ϕ_n
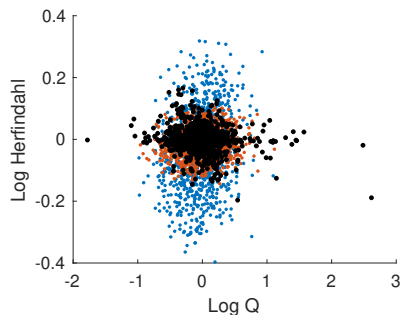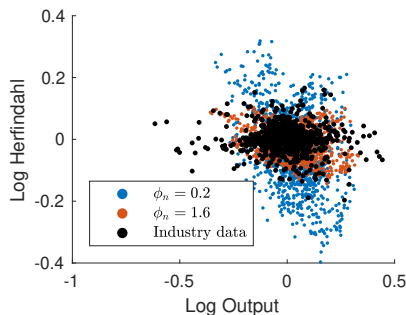
ϕ_n estimate implies:

- Following demand shock that raises Q_t to 10% after 1 year
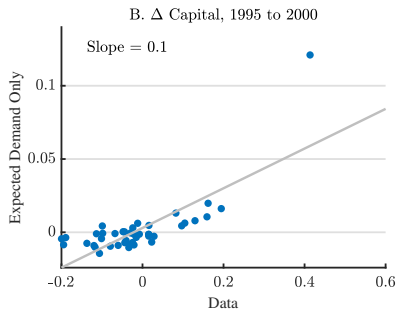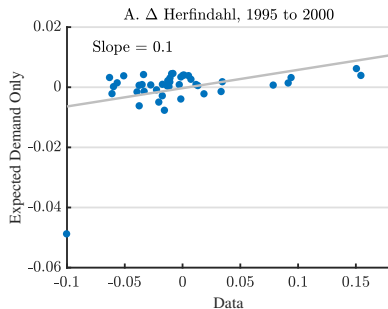- $\Rightarrow$ Number of firms increases by 1.4% after 2 years

Estimate of Demand Beliefs: Specific Industry

Example for Info Data Industry
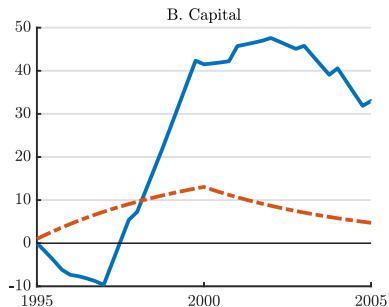
Estimate of Demand Beliefs: All Industries

- Accounts for about 10% of entry

Aggregate Implications

- What are the aggregate implications of firm entry?

- Use the estimated elasticity of firm entry to Q, ϕ_n and:

  1. Estimate the aggregate model's parameters

  2. Filter the aggregate data for the shocks

- Experiments with estimated model

  1. Interpret changes in entry cost shocks

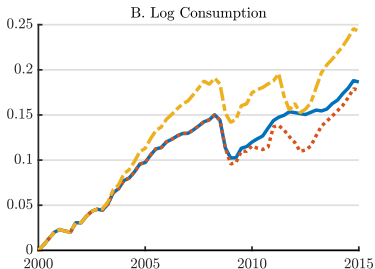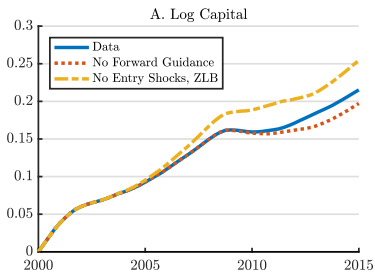 2. Turn off aggregate entry cost shock from 2003

Variance Decomposition of Aggregate Variables

Variable	Tech.	Pref.	P MU	Risk P.	Policy	Entry Cost
Fed Funds Rate	4.0	23.6	20.1	34.3	10.9	7.0
Output	50.3	6.9	0.1	13.2	0.0	29.4
Consumption	54.1	7.2	1.7	11.8	0.1	25.2
Net Investment	22.2	22.4	0.3	39.6	0.9	14.6
Employment	6.1	2.9	44.4	12.3	7.4	26.9
Inflation	2.8	20.2	11.2	37.6	15.9	12.3
Herfindahl	22.0	7.2	0.0	14.1	0.0	56.5
Natural Rate	1.4	20.7	0.0	37.5	0.0	40.4

Aggregate Implications: Less entry, lower FF rate

A. Herfindahl, Deviation

B. Fed Funds Rate
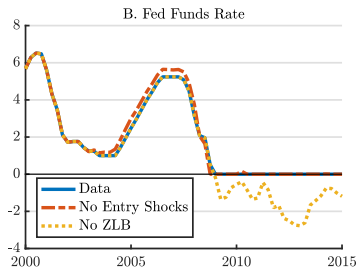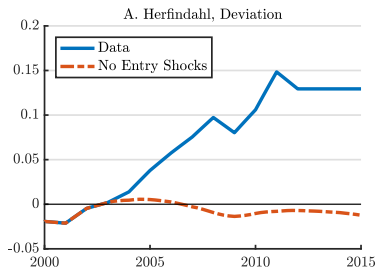
Aggregate Implications: Less investment, consumption

A. Log Capital

Data
No Forward Guidance
No Entry Shocks, ZLB

B. Log Consumption

Aggregate Implications: Lower wage income, natural rate

A. Labor Income


B. Natural Rate, Annual
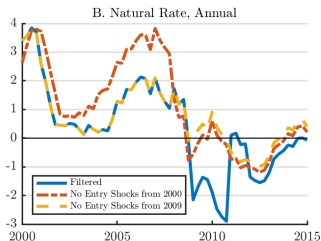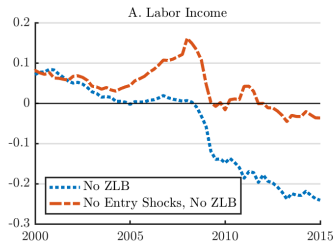
# Interpretation of Entry Cost Shocks

Empirical Proxies of Barriers to Entry

- Regulation Index:

    - Regulation based on RegData 3.1 from QuantGov

    - Machine learning / natural language processing techniques to construct measures of regulatory stringency at the industry level

    - Number of restrictive words or phrases such as 'shall', 'must' and 'may not' in each section of the Code of Federal Regulations

- Regulatory employment:

    - Regulatory employment from the Census Occupational Employment Statistics
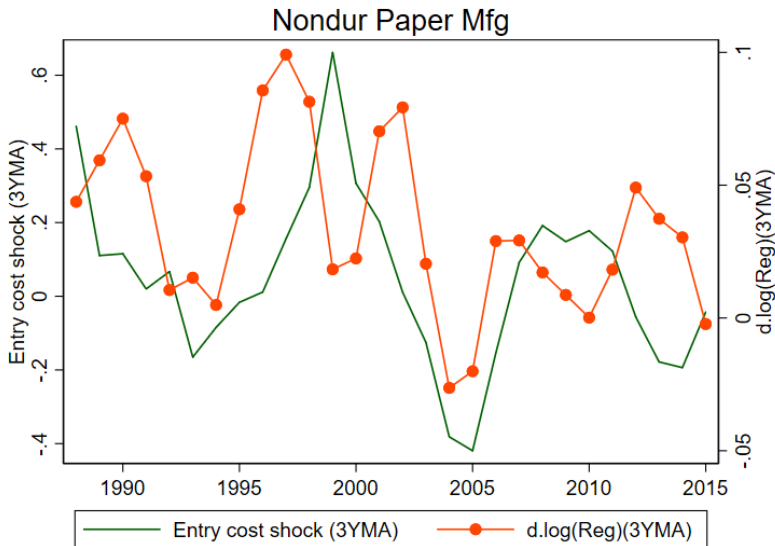
- M&A activity
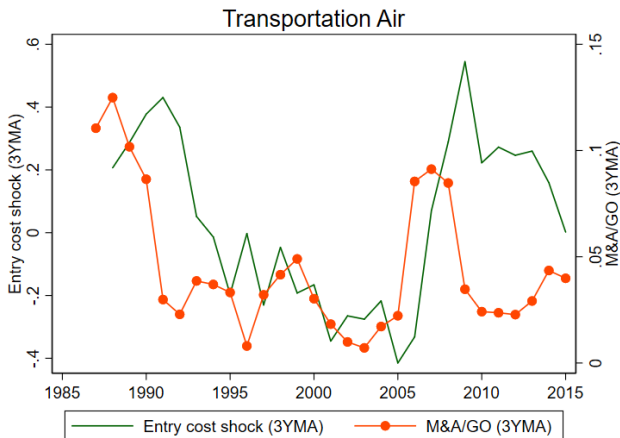
What are the Entry Cost Shocks? Aggregate
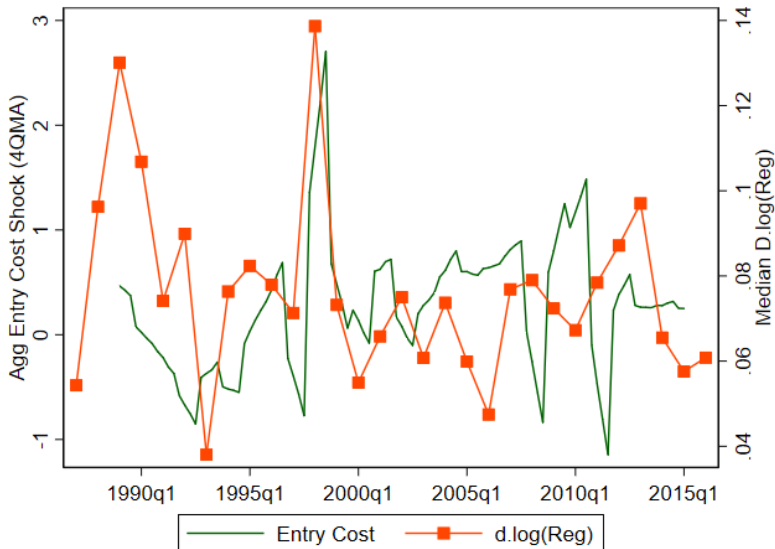
What are the Entry Cost Shocks? Aggregate
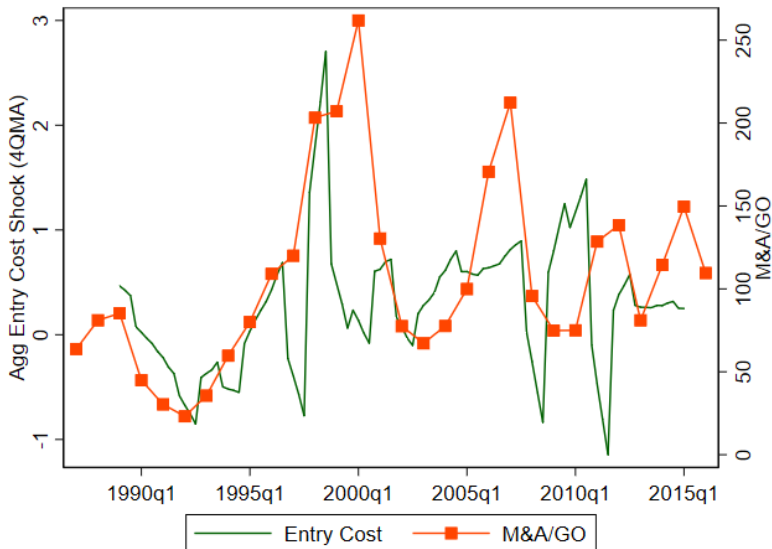
What are the Entry Cost Shocks? Non-Dur Paper Manuf

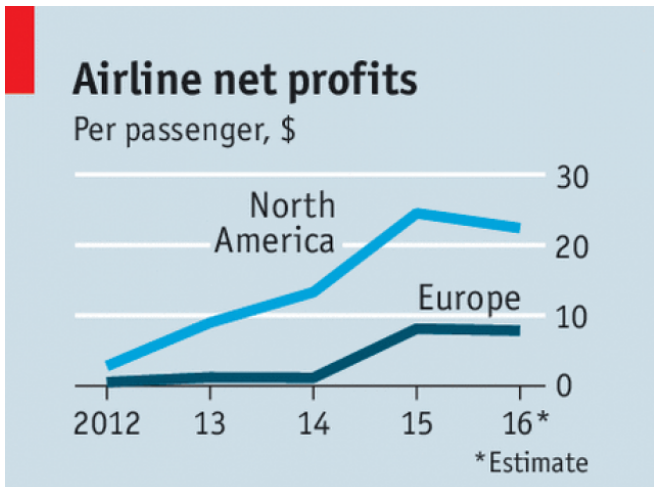

Nondur Paper Mfg

What are the Entry Cost Shocks? Air Transport

Transportation Air

- Delta-Northwest (2008), United-Continental (2010), Southwest-AirTran (2011) and American-US Airways (2014)

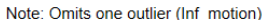## The Economist on Airlines

Economist.com

Entry Costs Across All Industries

	$\zeta_{j,t}^{\kappa}$					
	(1) All	(2) Post-02	(3) Post-02	(4) Post-02	(5) All	(6) Post-02
$\Delta \log(Reg\ Index_{t-2,t-1}^{j})$	0.045** (0.014)	0.051** (0.018)			0.045** (0.014)	
$\Delta \log(Reg\ Emp_{t,t+1}^{j})$			0.035* (0.013)			
Mean(L.dRegIndex,F.dRegEmp)				0.042** (0.010)		0.036** (0.011)
$\log(M\&A_{j,t})(2YMA)$					0.050* (0.022)	0.105* (0.042)
Ind FE	Y	Y	Y	Y	Y	Y
Year FE	Y	Y	Y	Y	Y	Y
R2	.051	.097	.09	.1	.058	.12
Observations	837	358	358	358	837	358

Conclusion

- US industries have become more concentrated
    - Lack of competition leads to lower investment
    - We find an increase in entry costs
    - We link those entry costs to regulation/M&A

- Europe has trended the other way   Comparison
    - Decreasing concentration
    - Low investment in Europe and low valuations/high risk premia
    - If true, then probably cyclical

Appendix

How to Interpret Evidence of Increasing Concentration?

- Superstar firms?
    - Leaders are investing less, when it should be more Evidence
    - TFP should be correlated with concentration. Not so from 2000
    - Concentration trends not observed in Europe Evidence

- Globalization?
    - Measurement? Use consolidated firm-level data
        - Includes investment in US and outside. Gap remains
    - External Profits? Gutiérrez and Philippon '17 BPEA
        - Profits have increased faster than foreign sales
        - So profits being shifted out of the US

- Declining Domestic Competition. Furman '15, CEA '16
    - 'Economic rents and barriers to competition'
    - Less firm entry

Aggregate Entry Cost Shock

Entry Cost Shock

What are the Entry Cost Shocks?

Note: Omits one outlier (Inf_motion)

How Should We Interpret These Facts?

- Drop in the price of capital?

- Intangible assets
    - Peters and Taylor '16, Alexander and Eberly '16

- Superstar firms?
    - Autor '17

- Globalization
    - Feenstra and Weinstein '17, Fresard and Valta '15, Hombert and Matray '15

- Decreasing Domestic Competition, DDC
    - Furman '15, Gutiérrez and Philippon '17

How Should We Interpret These Facts? #2

- Drop in the price of capital?
  - Timing is wrong, as fall in price of capital occurs before 2000

- Intangible assets
  - 25% of the gap between Q and N/K explained by intangible intensity, so not much
  - National accounts does include IP investment
  - Argument not necessarily true for industries like airlines

- Superstar firms?
  - Implies TFP correlated with concentration, but not so from 2000
  - Leaders are investing less, when it should be more (Evidence)
  - Concentration trends not observed so much in Europe

- Globalization?
  - Measurement? investment gap appears in consolidated firm-level data, which includes investment in US and outside
  - External Profits? GP (2017) BPEA show profits have increased faster than foreign sales, so profits being shifted out of the US

(Back)

Intangibles: Gutierrez and Philippon (2017)

Firm−level FE with and w/o intangibles

Productivity: Gutierrez and Philippon (2017)

Select Discussion: Super Star

- Our interpretation of the hypothesis (Autor et al.)
    - Not simply a description of skewness.
    - But an explanation for concentration: efficiency instead of market power
- What we find: some support in the 1990s, but not after 2002.

	(1)	(3)
	ΔTFP	
	97-02	02-12[†]
ΔCensus CR4	**0.481****	**0.051**
	[4.439]	**[0.301]**
Observations	469	297
R^2	4%	0%

Notes: T-stats in brackets. $+$ p<0.10, * p<0.05, ** p<.01. † TFP change to 2009 due to data availability

EXTRA: Measures of Concentration & Entry

- Traditional Herfindahl + Common ownership adjustment (Azar, et. al. (2016))

$$MHHI = \sum_j s_j^2 + \sum_j \sum_{k \neq j} s_j s_k \frac{\sum_i \beta_{ij} \beta_{ik}}{\sum_i \beta_{ij}^2}$$

$$= HHI + HHI^{adj}$$

- Other measures including entry, share of sales by top #10 firms, etc. also significant

Posteriors for Common Industry Parameters
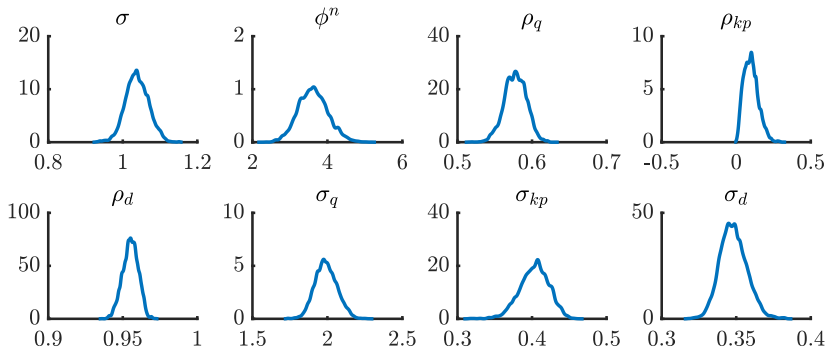
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Q Data

- Flow of funds
- Ratio of market value to the replacement cost of capital including intangibles

Back

Q

Concentration Ratio Data

- Compustat for BEA segments
- Patterns verified against aggregated firm-level census data
- Corrected for the import share, as in Feenstra and Weinstein '17
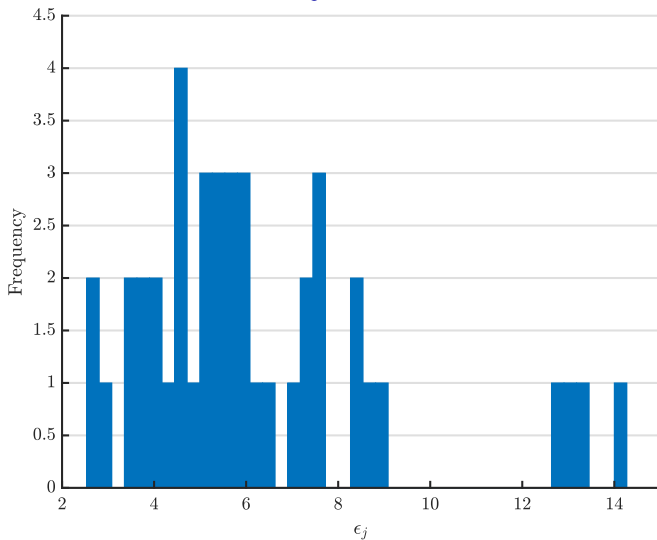- Series from Compustat/BEA have a 65-70% correlation in levels and 40-50% in 5-year changes

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Concentration Ratios

Nominal Output Data

- BEA

Back

Nominal Output
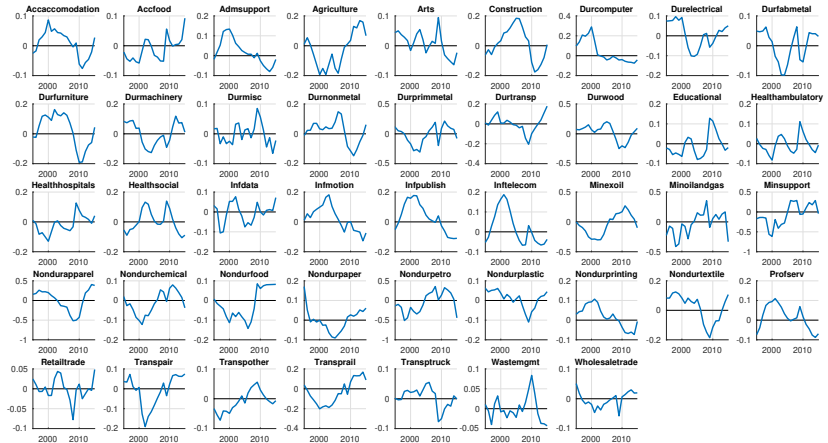
Calibrated ε_j for a given σ

ZLB Durations in Quarters

ZLB Durations, Quarters
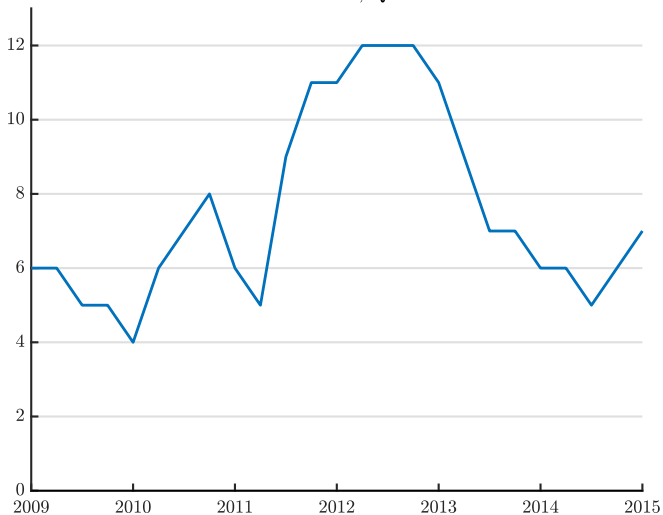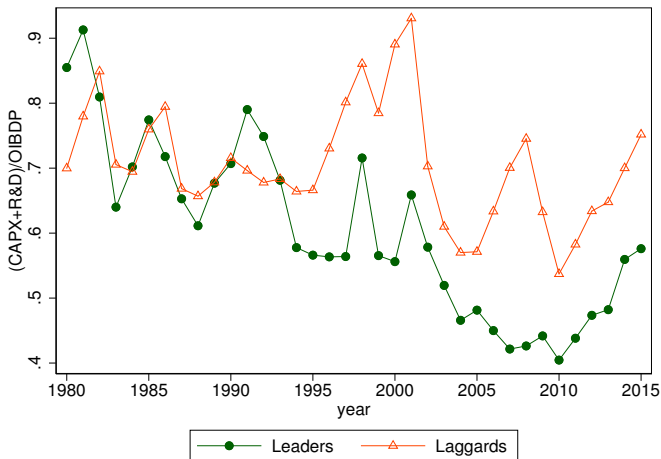
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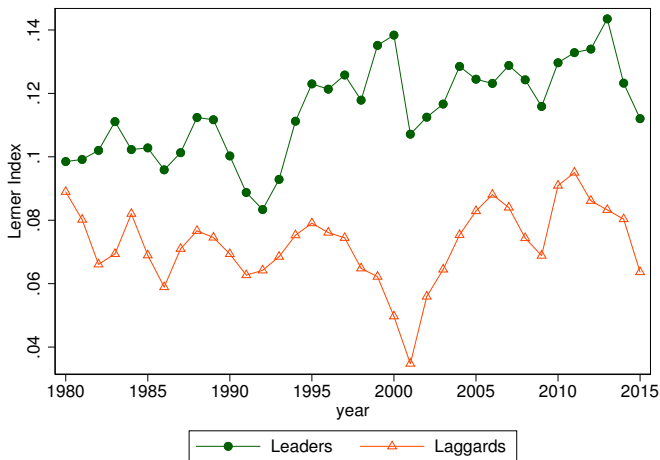# Margins Increased for Industry Leaders
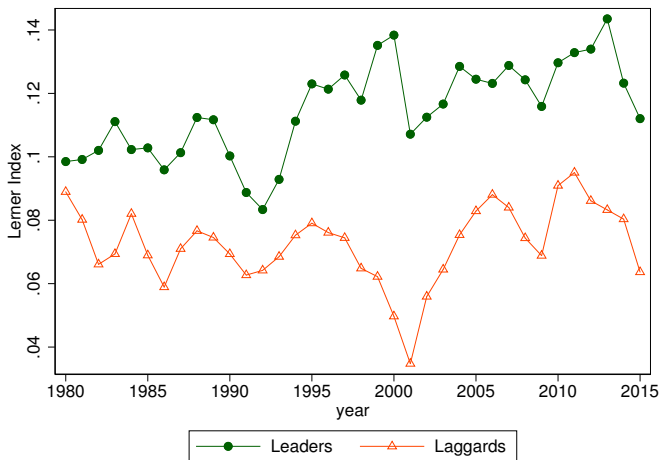
Note: Leaders includes all firms with the highest (lowest) market value (MV) that combined account for ~33% of MV within each industry and year.

Leaders Explain Investment Gap #1

Note: Leaders (laggards) include all firms with the highest (lowest) market value (MV) that combined account for ~33% of MV within each industry and year.
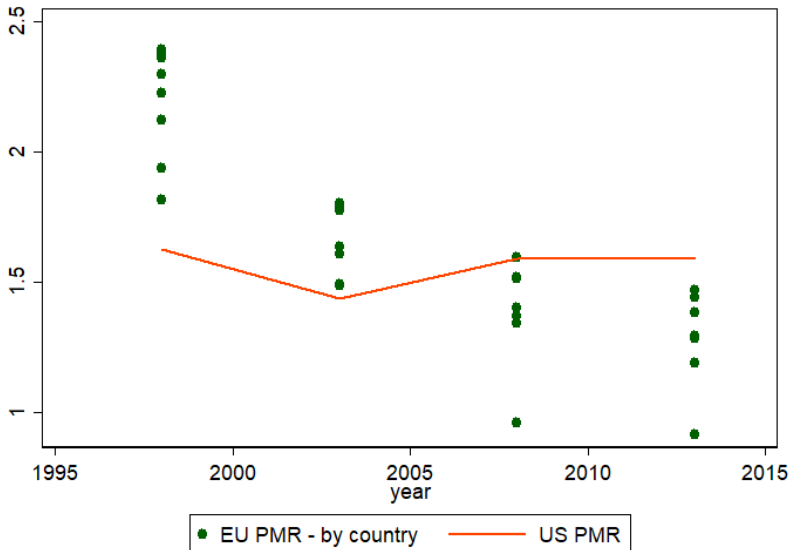
Margins Increased for Industry Leaders

Note: Leaders includes all firms with the highest (lowest) market value (MV) that combined account for ~33% of MV within each industry and year.

EU vs US
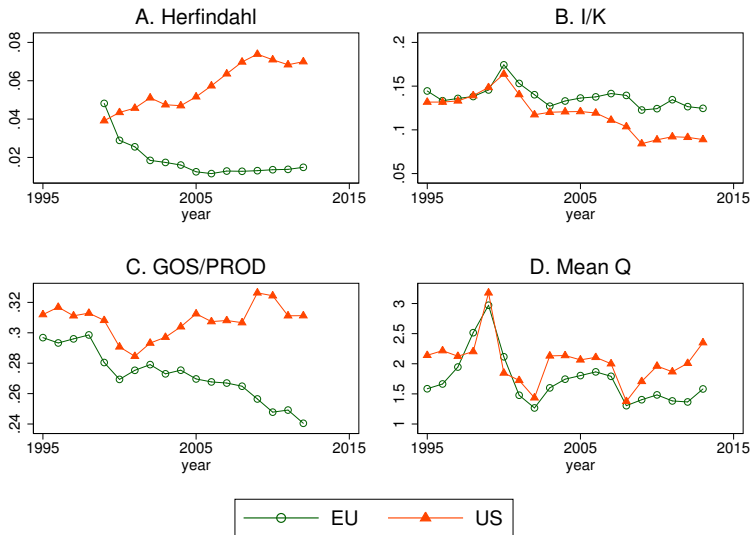
EU vs US: OECD Product Market Regulations

A Calvo-Style Entry Specification

- Staggered entry. Pay κ today, each period there is a chance λ of entry each period

Other Parameters

δ	0.025
ϕ^k	20
β	0.99
α	1/3
δ^n	0.09/4

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I/K vs. Q