

# The Macro Impact of Zero Profit Bunching

Thomas Walsh

University of Glasgow

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## My Plan for Today

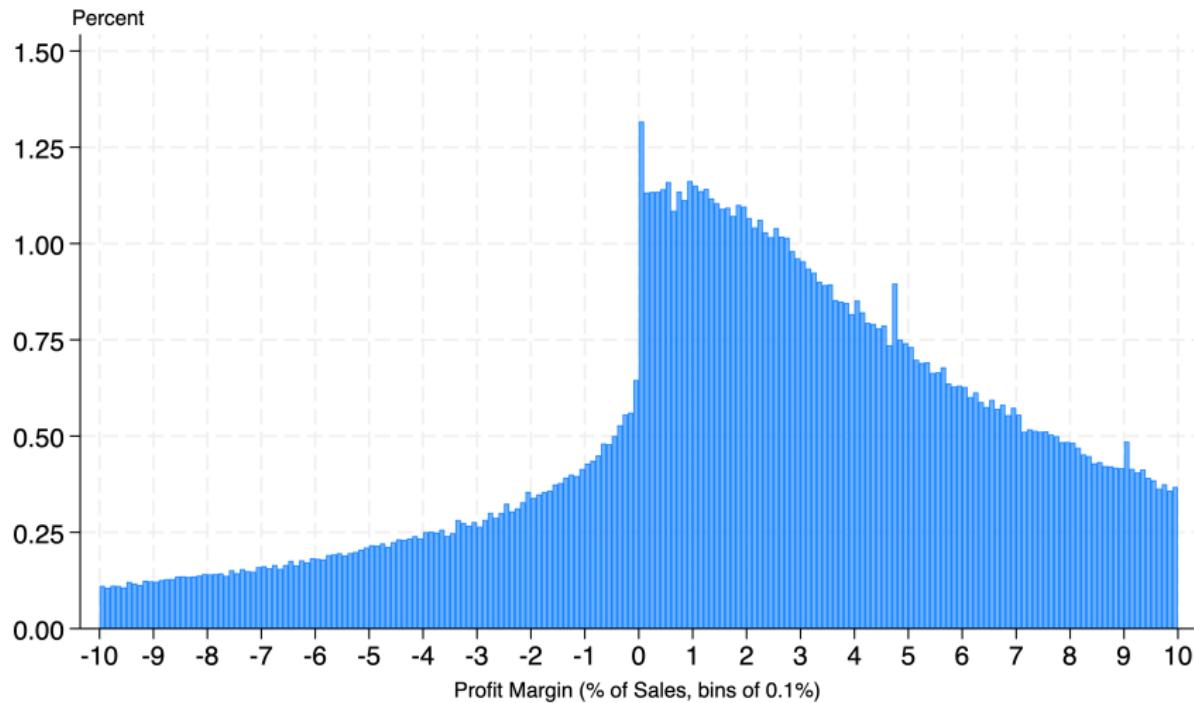
- 1a) Present some novel “stylised facts” in firm-level microdata (most work done)
- 1b) Some light empirical analysis to quantify what is going on and kick the tires
- 2) Ideas for model(s) to structure thinking about facts (least complete)

### **Zero is a special number in all ‘mainstream’ firm-level datasets (US, UK, EU):**

- very strong **bunching in the PDF** of firms over profits at  $+e\%$  profits
- spike not driven by exit of negative earners
- macro impact: firms (optimally?) cut expenditures to maintain positive earnings and cheaper credit.

### **Some quick numbers on losses/negative profits for large, listed firms:**

- around **30 percent** firms make a loss in any year
- over 5-year period, **60 percent** have at least one year with losses
- at the frequency of the business cycle, majority of firms transit across threshold



Source: Profits are pre-taxation. FAME database of UK firms, listed and unlisted. 2003-2023

**Figure 1:** Distribution of Profitability (% of Sales)

# What is the macro impact of this bunching?

If firms cut their investment to maintain positive profits...

## Firm-level:

- Looks a lot like Present-Bias or “ $\beta - \delta$  discounting” in reduced form
- Depresses firm capital accumulation, (mechanisms: cuts in inv. and bond price)
- firms slower to reach optimal size in DRTS world

## Macro-level:

- Distorts allocation of resources over active firms (agg measured TFP)
- Distorts set of active firms (shutdown-point of shocks is higher)

## GE effects could offset

- lower firm capital demand  $\downarrow R$
- lower agg TFP from misallocation  $\downarrow R$
- $K - N$  substitution

# Quantifying Bunching at Zero

## Estimate Counterfactual Distribution

**Density** is approximated by counts within narrow bins,  $b$ :

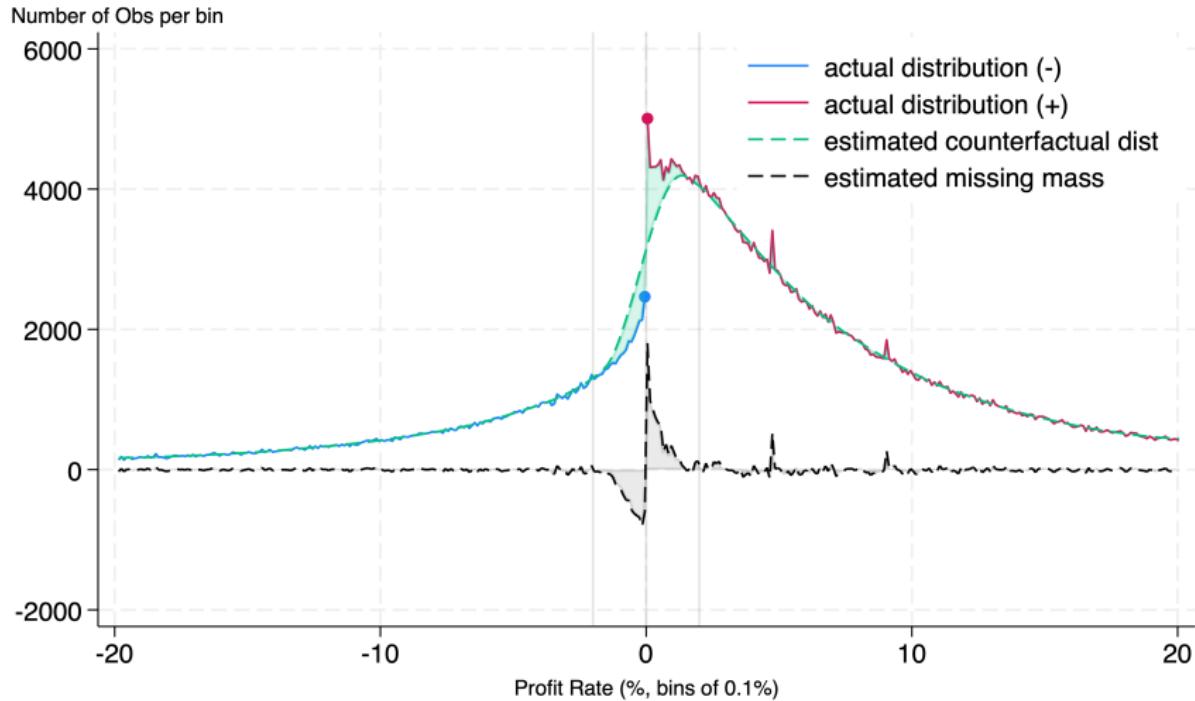
$$N_b^{data} = \sum_i 1(L_b \leq \pi_{it} < U_b) \quad (1)$$

**counterfactual**  $N_b^{cf}$  built from a **local polynomial regression**

$$N_b^{cf} = \hat{\mathcal{P}}(N_b^{data}, \mathbb{1}(\pi > 0), bw^*) \quad (2)$$

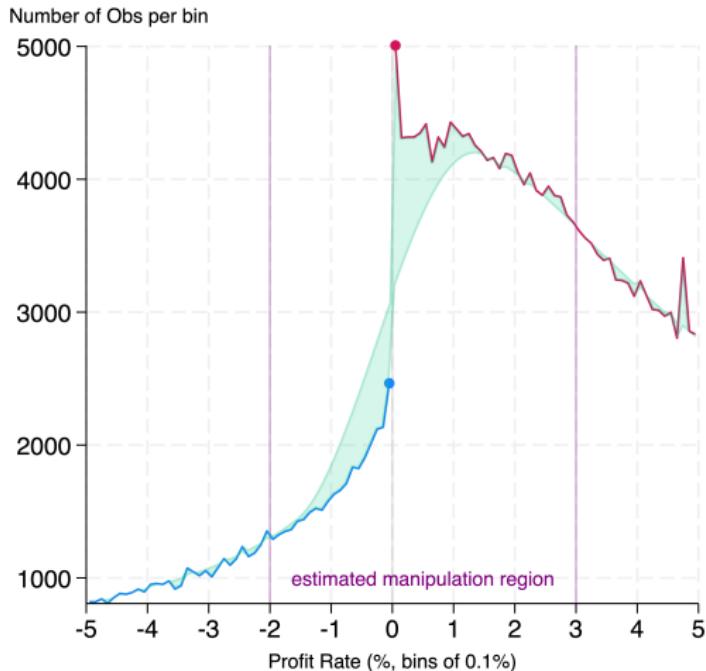
### Methodology complications

- usual fit inappropriate due to high curvature near cutoff
- uses full support to select optimal bandwidth for smoothing
- fits density away from distorted area well



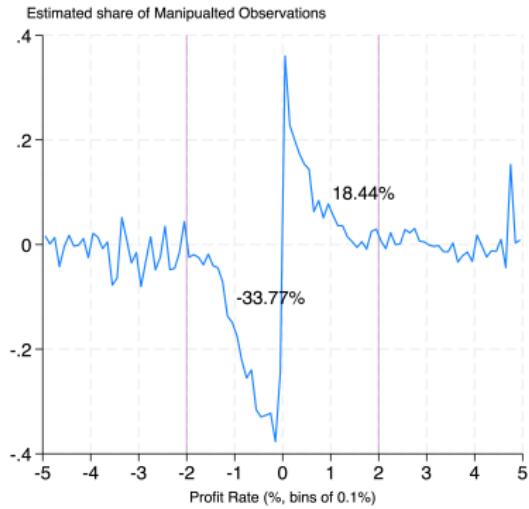
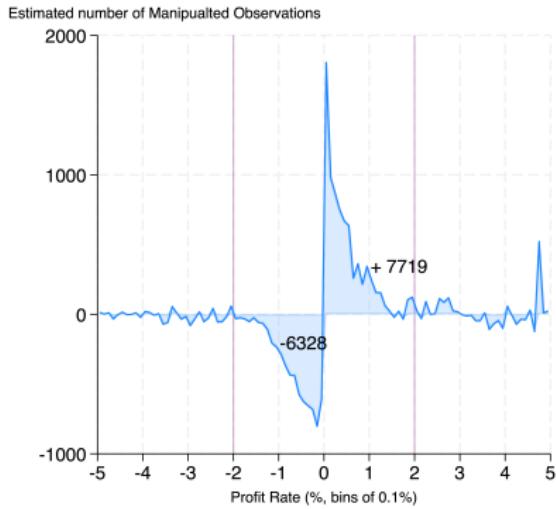
Source: FAME database of UK firms, listed and unlisted. 2003-2023

**Figure 2:** Jump in Density at  $+\varepsilon$  Profits



Source: FAME database of UK firms, listed and unlisted. 2003-2023

**Figure 3:** Jump in Density at  $+\varepsilon$  Profits

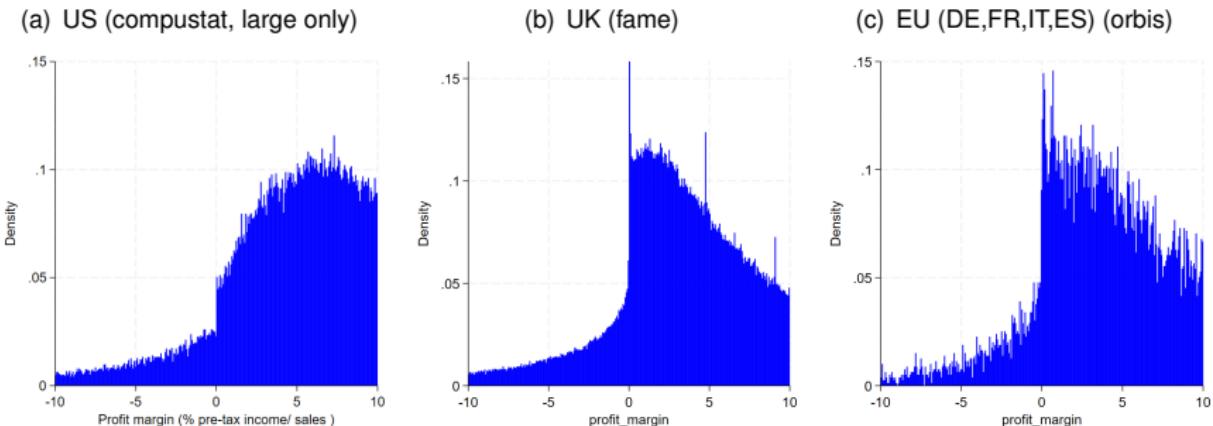


Source: FAME database of UK firms, listed and unlisted. 2003-2023

**Figure 4:** Jump in Density at  $+\varepsilon$  Profits

Will adjust CF fit such that two areas are approx. equal

# Zero the mainstream firm dynamics datasets



Note: Compustat North America database, FAME, ORBIS. Sample restricted to firms with 3 years of prior positive profitability and survive to the next year.

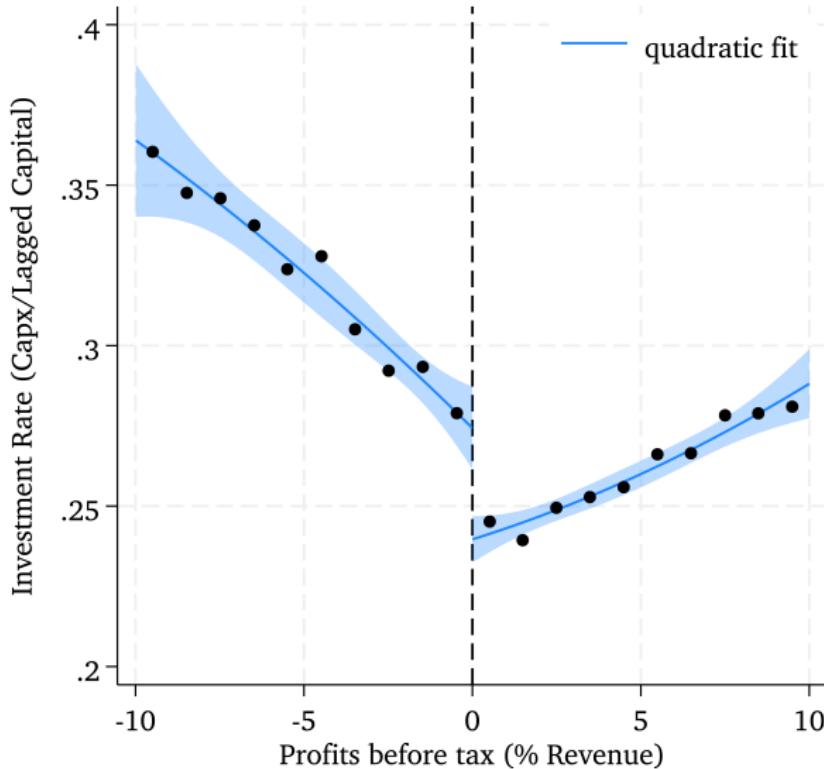
**Figure 5:** Density Jump at zero profits in US,UK,EU firms

# Firm-level Bunching and Spending

Regression Discontinuity and Donut Plots

## Capx spending rates (pps), intensive margin

CAPX Investment Rate = (*Investment in Fixed Assets<sub>t</sub>/Fixed Assets<sub>t-1</sub>*)



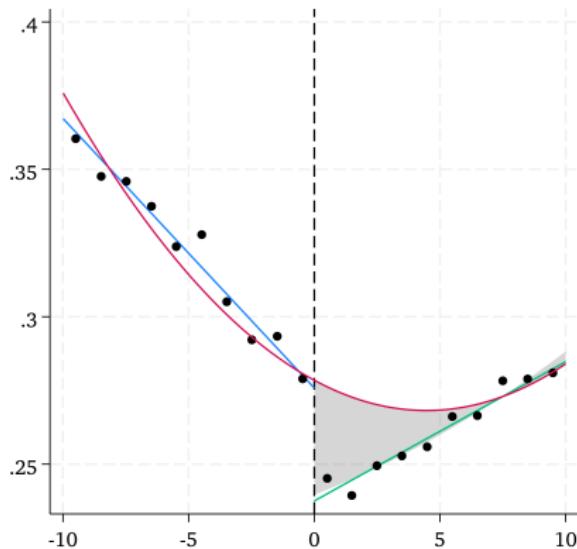
## Macro implications?

**Area measures lost investment,**  
approximated:

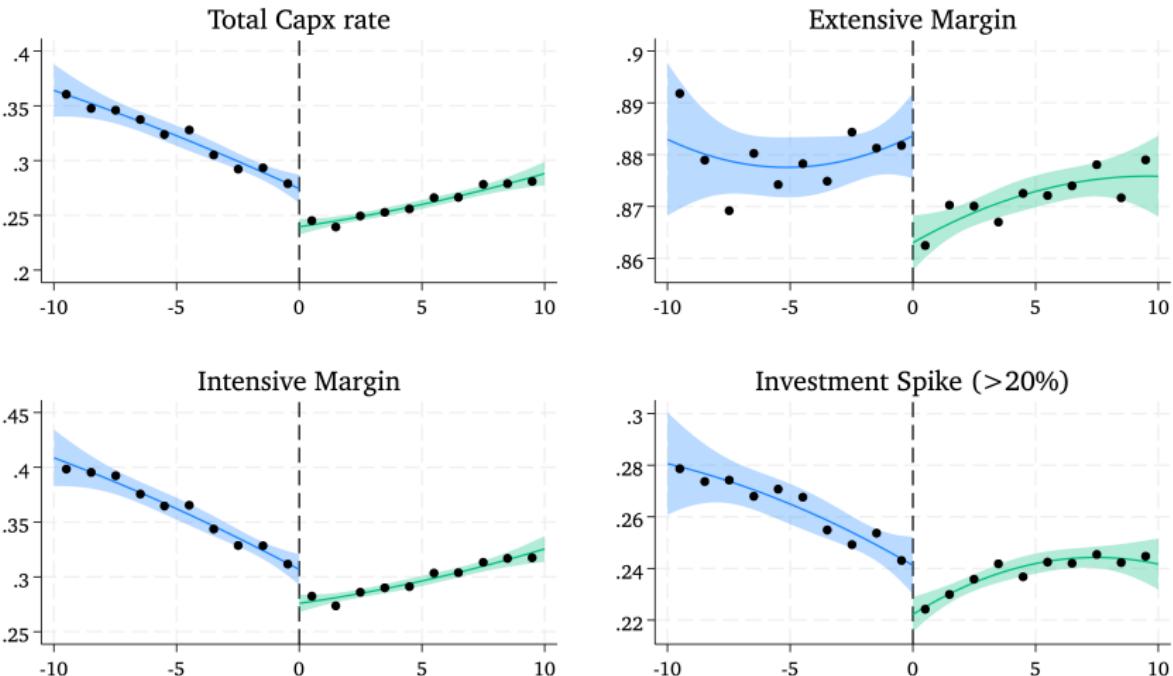
- 1 in 3 firms in the (0,5] profits range.
- 14% drop at cutoff (4/28)
- triangle area = half the “square”
- $0.33 \times 0.14 \times 0.5$

$\approx 2.3$  percent of agg investment

$\Rightarrow$  meaningful magnitudes for (potential) macro dynamics (depends on what large firms do)



**Figure 6:** Missing Investment

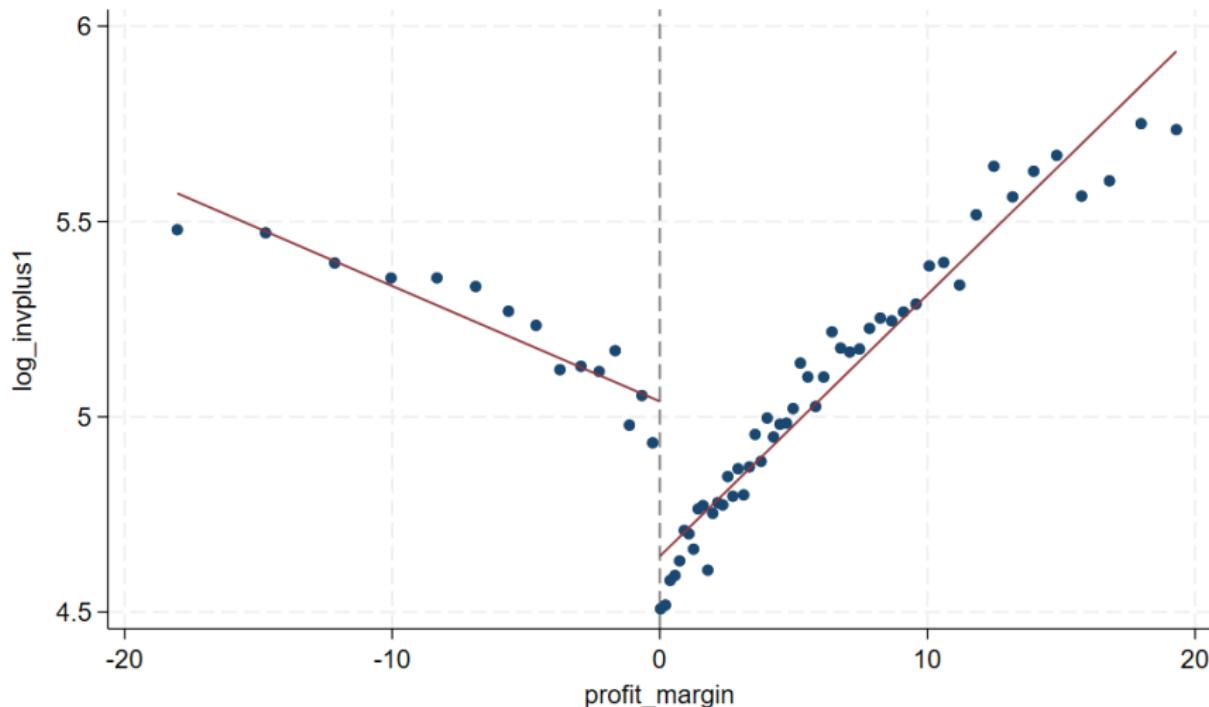


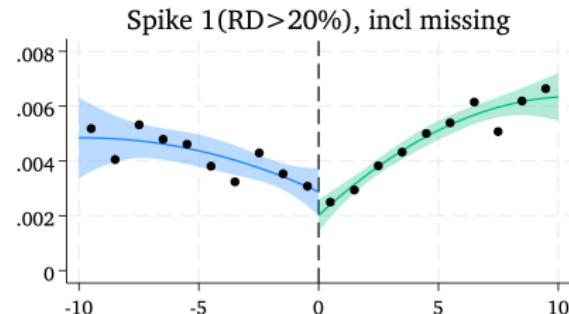
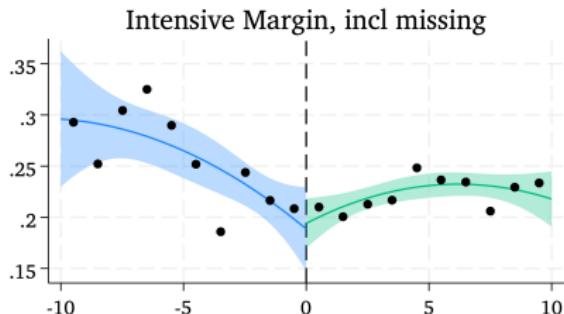
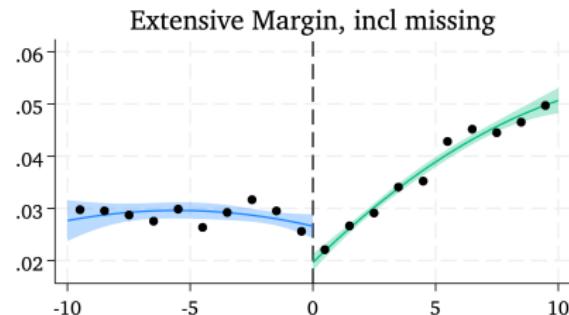
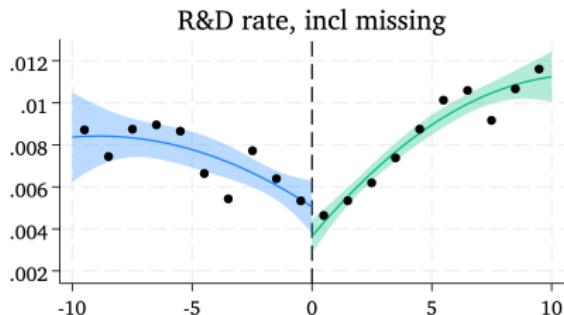
Note: Capx rate is capx over lagged capital stock. Extensive margin is the share of firms with  $\text{capx} > 0$ , intensive margin is average capx given  $\text{capx} > 0$ , spike is share of firms with  $\text{capx} > 0.2$ .

**Figure 7:** Margins of Firm-level Capital Adjustment

## Alternative measure, $\ln(1 + CAPX_t)$

combines extensive and intensive margins, harder to interpret directly, *asinh* also similar  
(applied economists look away now)





**Figure 8:** Firm-level XRD spending

## Salaries/Turnover

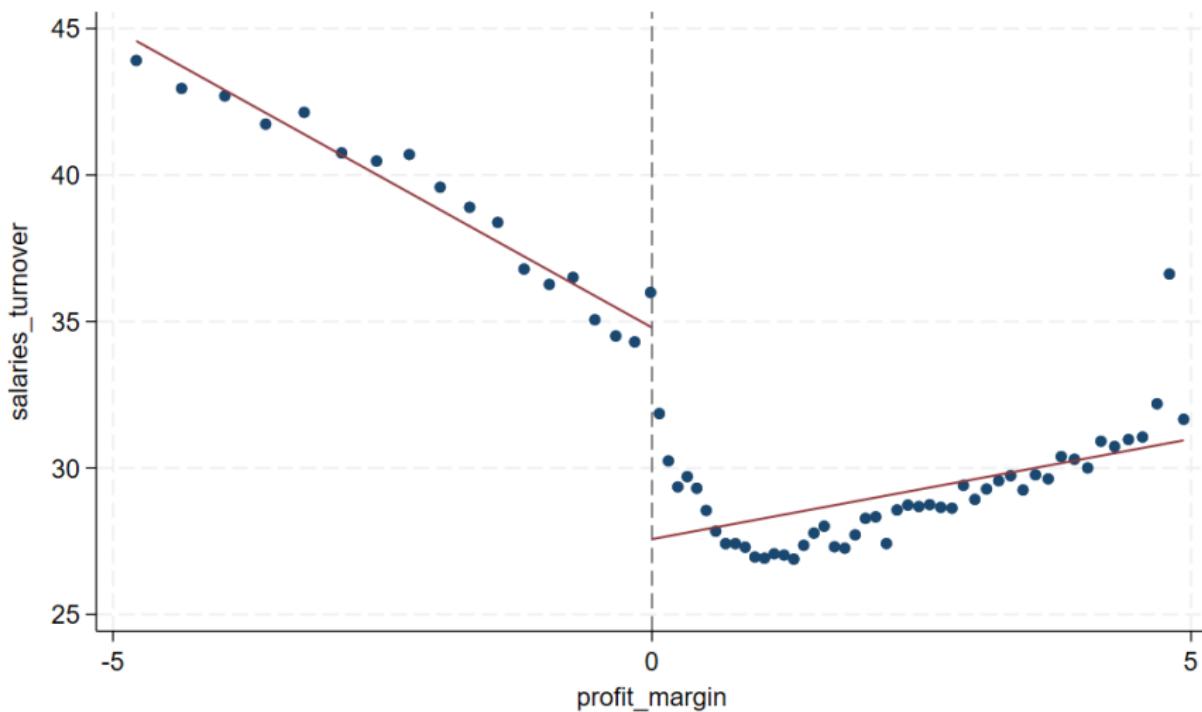
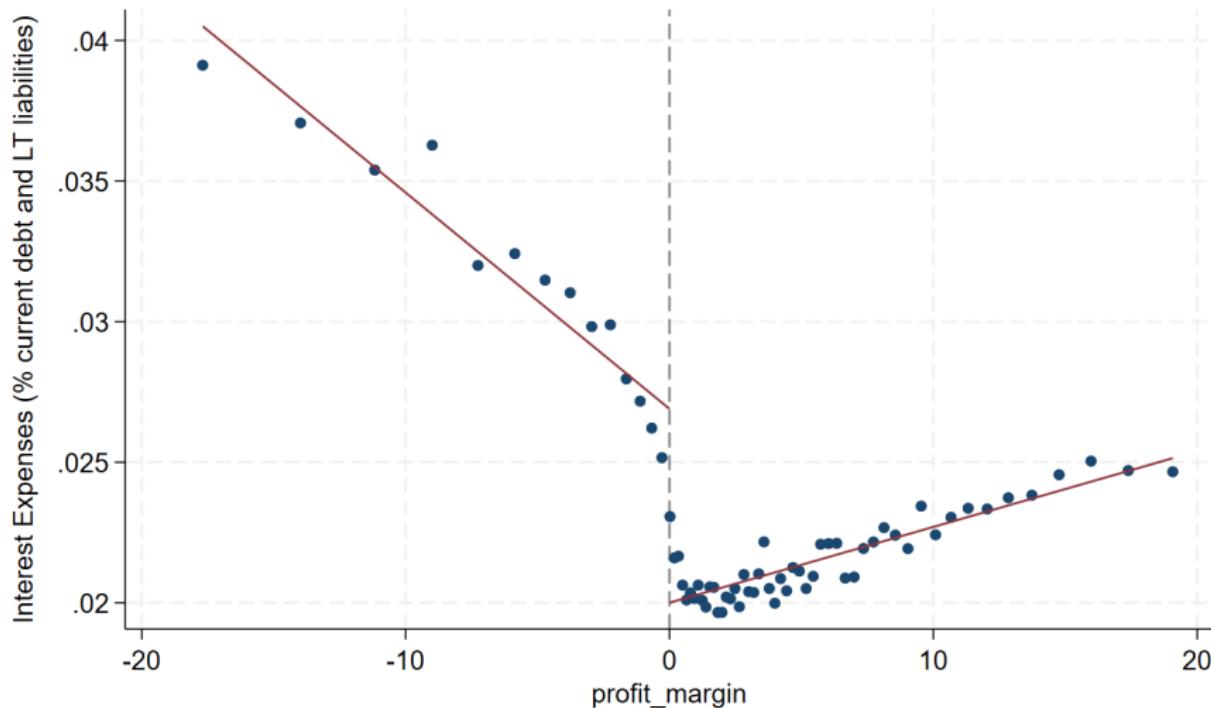


Figure 9: Caption

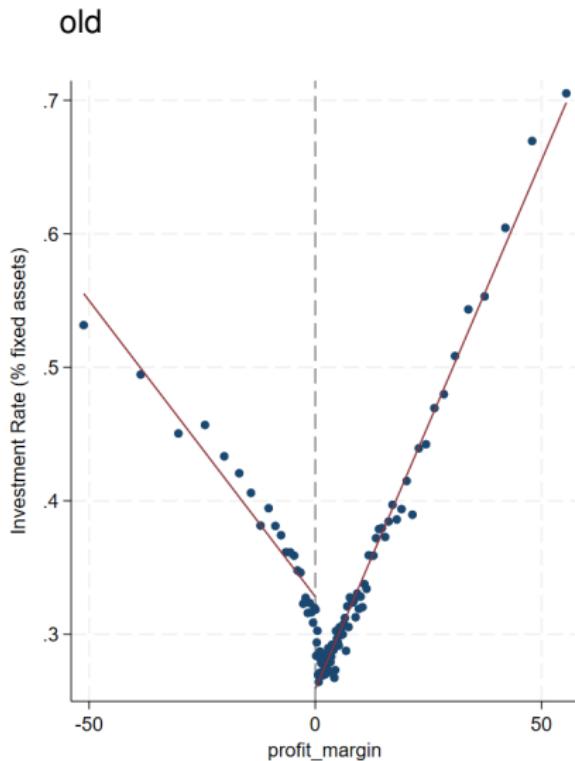
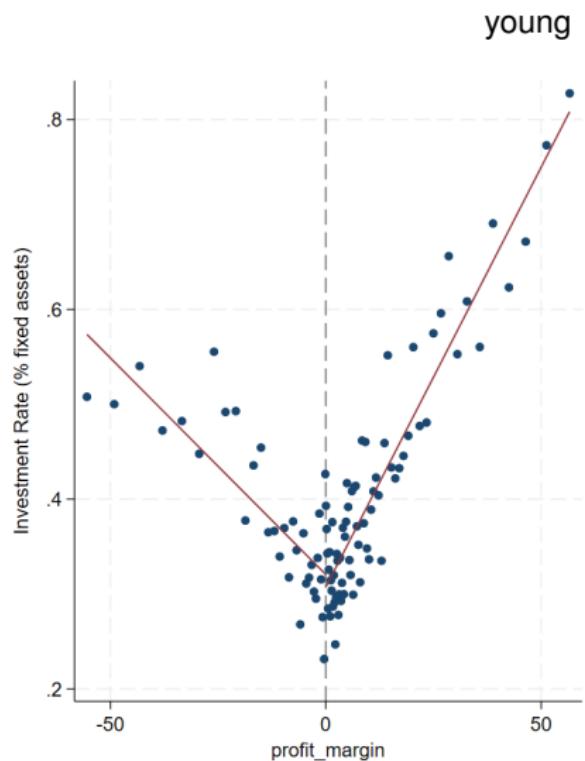
## Interest Rates on debt

XINT (interest paid on ST and LT debt as % liabilities) drops at threshold

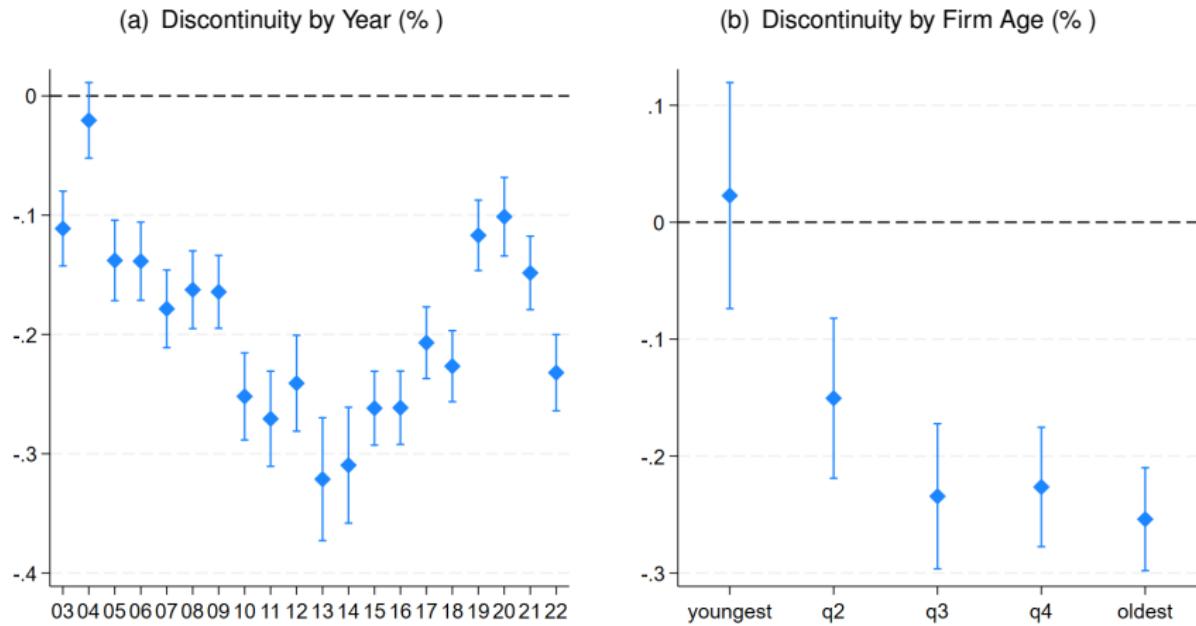


# CAPX by Young and Old Firms

no discontinuity for young (<5yrs) firms (stronger motive to grow capacity to survive?)



# Heterogeneity in Investment Discontinuity Estimates



**Figure 10:** Heterogeneity by year and age

# Productivity, Depreciation and Other Issues

## Lower Costs and Apparent Productivity

$$Y = zF(K, N) \quad (3)$$

$$\log Y = \log z + \alpha k + \beta n \quad (4)$$

Lower labour and financing costs on RHS also boosts apparent productivity

- $(rK, wN)$
- other methods will still rely on distorted stocks of capital, or distorted capx flows

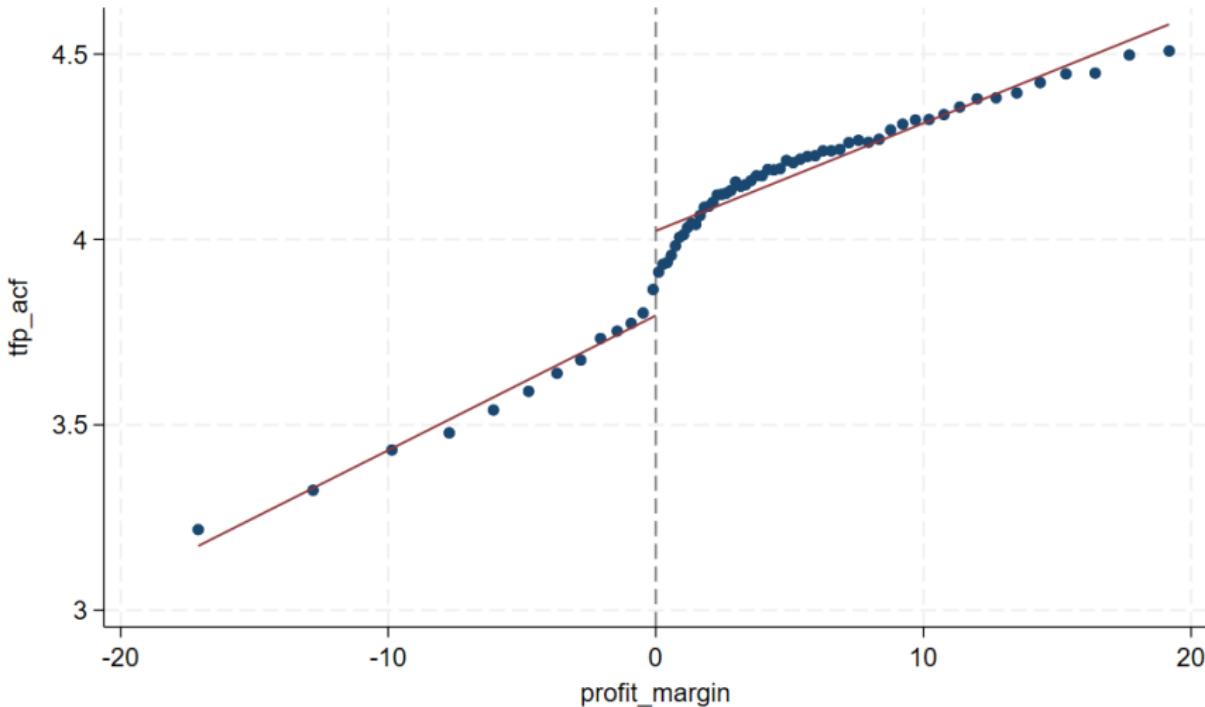


Figure 11: TFPR (AC)

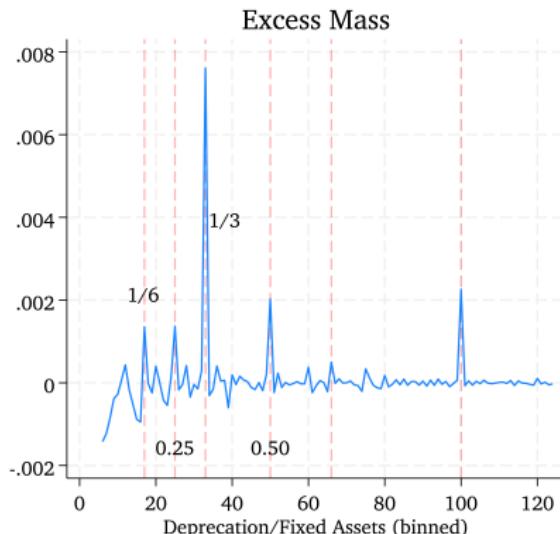
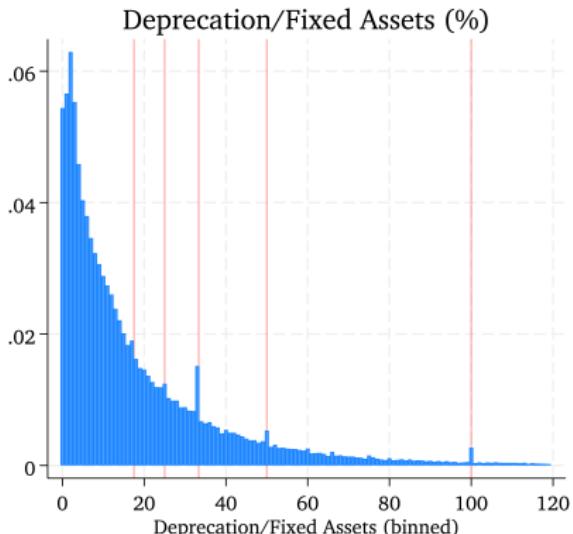
If threshold distorts choice of  $K$  it can have consequences for TFP calculations

## Other open issues

**How do you measure TFP when factor prices or optimal policies jump at  $\pi = 0$**

**Sidenote: Climbing the “capital ladder” is risky for firms?** High investment rates are correlated with large profits and large losses

## Bonus: Depreciation. More generalised distortion of choices?



**Figure 12:** Depreciation Distortions relative to polynomial fit

- **Over-represented** nice fractions:  $6^{-1}$ ,  $4^{-1}$ ,  $3^{-1}$ ,  $2^{-1}$ ,  $1$
- multiples of 5, 10 generally no more likely.
- don't know the true state of PPE, so guess? Impatient response?

# Taking Data to a Model

- Focus on **CAPX** drop
- **Interest rate** jump at zero
- ignore other stuff for now

## Modelling Choices

**Terry (ECMA 2023)** says quarterly earnings targets help control indulgent managers who make malinvestments for prestige  $[\pi_{it} - E_{t-\Delta}(\pi_{it})]$

**This work** profitability itself matters  $[\pi_{it}, \mathbb{1}_{\{\pi_{it} > 0\}}]$

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- Story B Signals of Control: not being able to make a marginal saving of  $\varepsilon$  at the threshold is a strong negative signal of management quality or decision making power:
  - lack of information on current state of firm (distance to threshold)
  - lack of precision in decision making / trembling hand choices
  - lenders infer type of firm ( $\sigma_H^2, \sigma_L^2$ ) from low negative profits

$$V(z, k, b) = \max_{k', b'} \left\{ zk^\alpha - px - b - F + q(y)b' + \beta E_{z'|z} V(z', \textcolor{blue}{k'} + \epsilon, b') \right\} \quad (5)$$

$$\epsilon \sim N(0, \sigma_j^2) \quad (6)$$

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- **Story C Bond Price:** (no explicit evidence in data)  $q(y) = [1 + r + \xi \mathbb{1}_{\{y<0\}}]^{-1}$

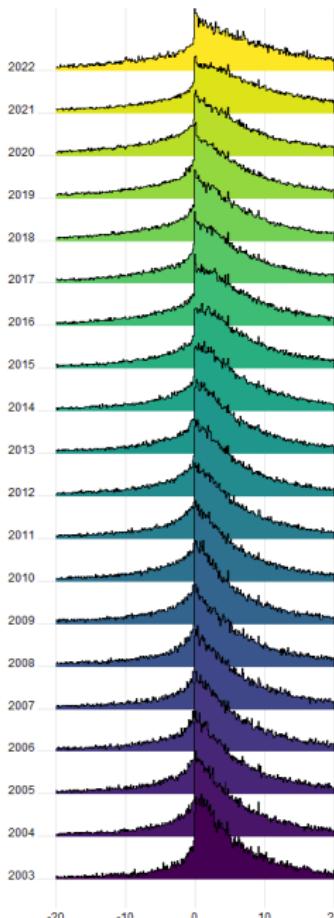
## Lender's Problem

Lender put a lot of weight on small losses – suggest managers aren't really in control and hiding worse performance if they cannot find even small savings?

is it possible to write a lenders problem which in eqm gives a discontinuous bond price?

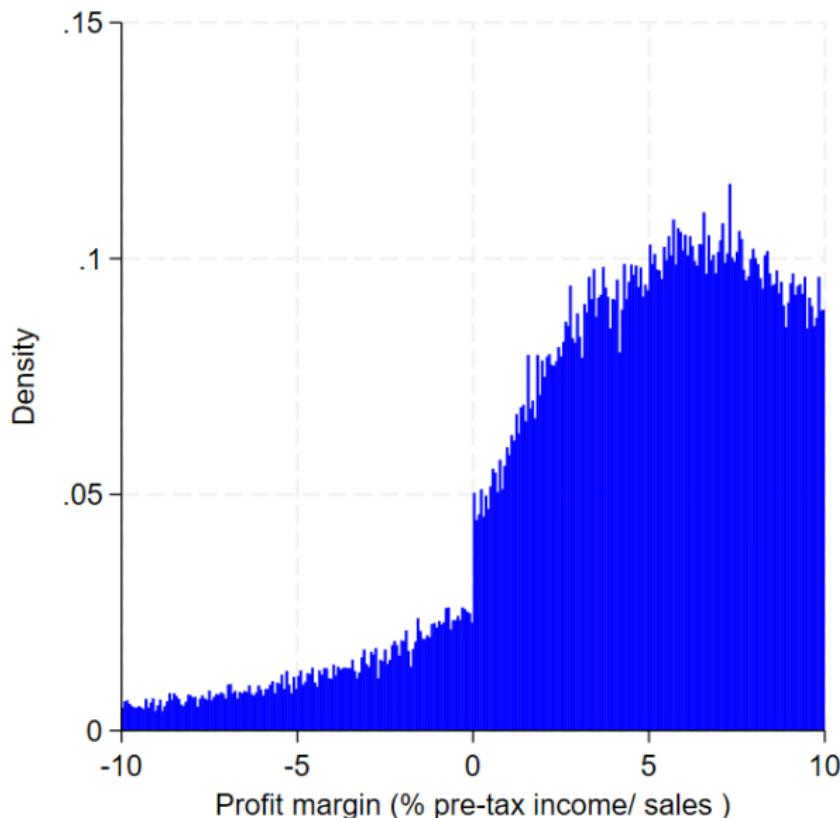
Thanks!

## Appendix: Discontinuity in Profit Margin Density by Year



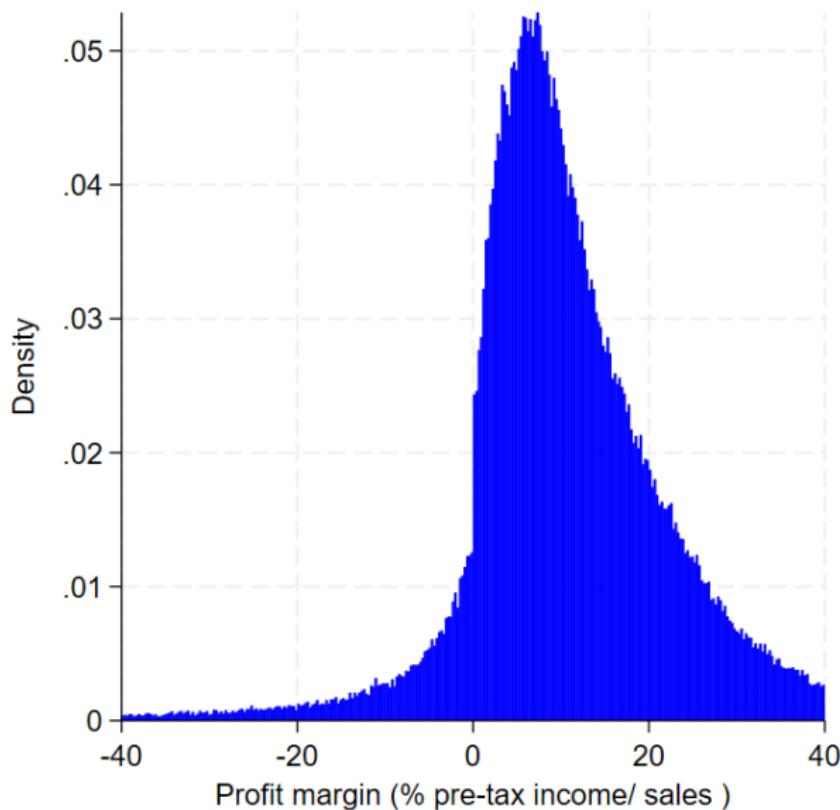
## Compustat (US listed firms only)

Compustat firms also seem to rearrange expenditures to achieve profitability



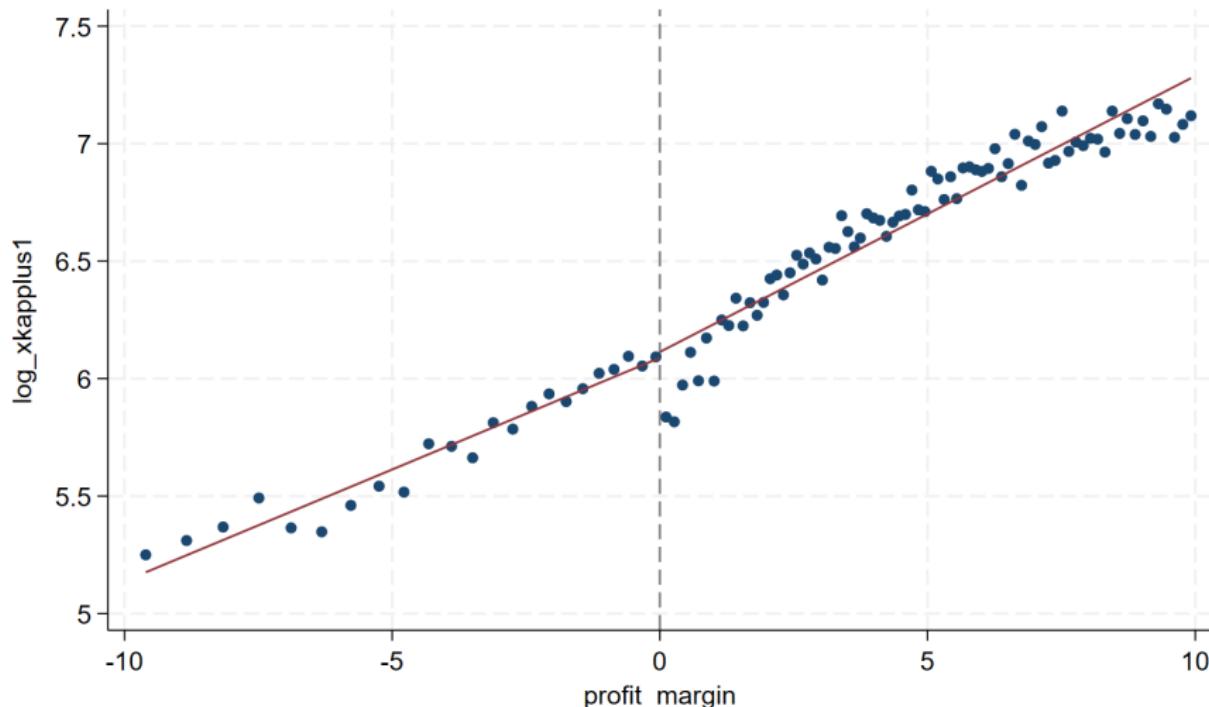
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# Capital Investment

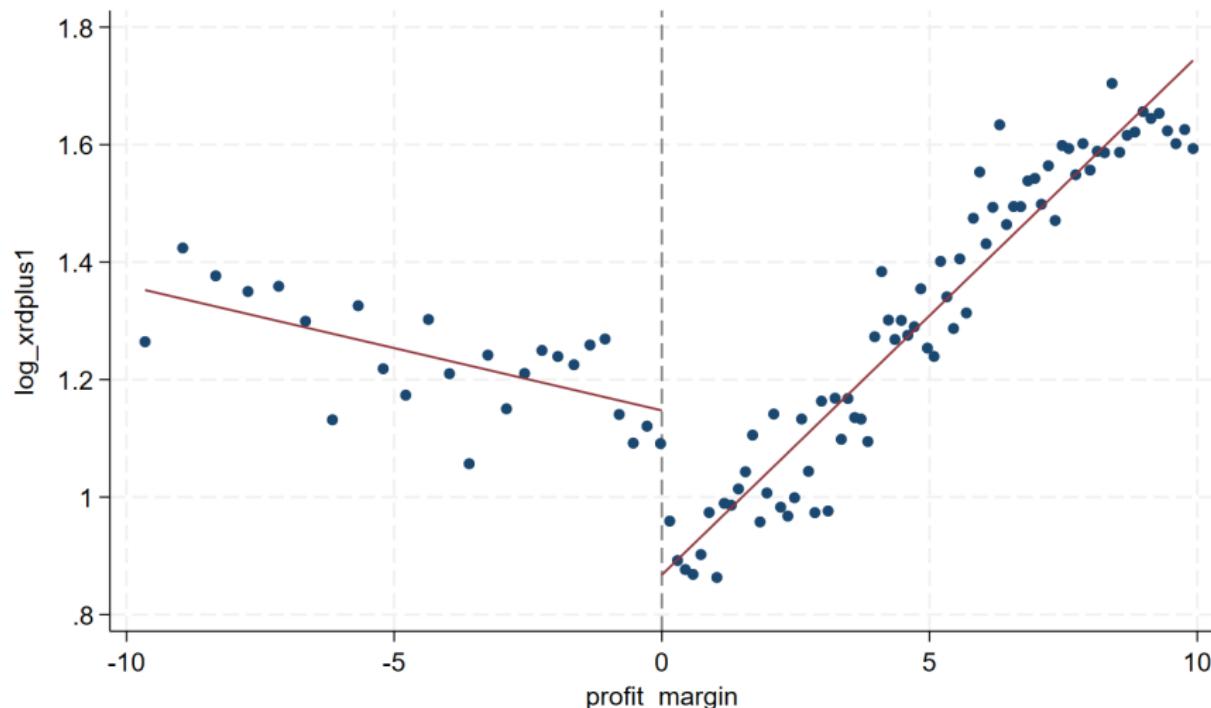
Slight drop in capital investment



**Figure 16:** Manipulators boost profits but cutting CAPX slightly

## Research Investment

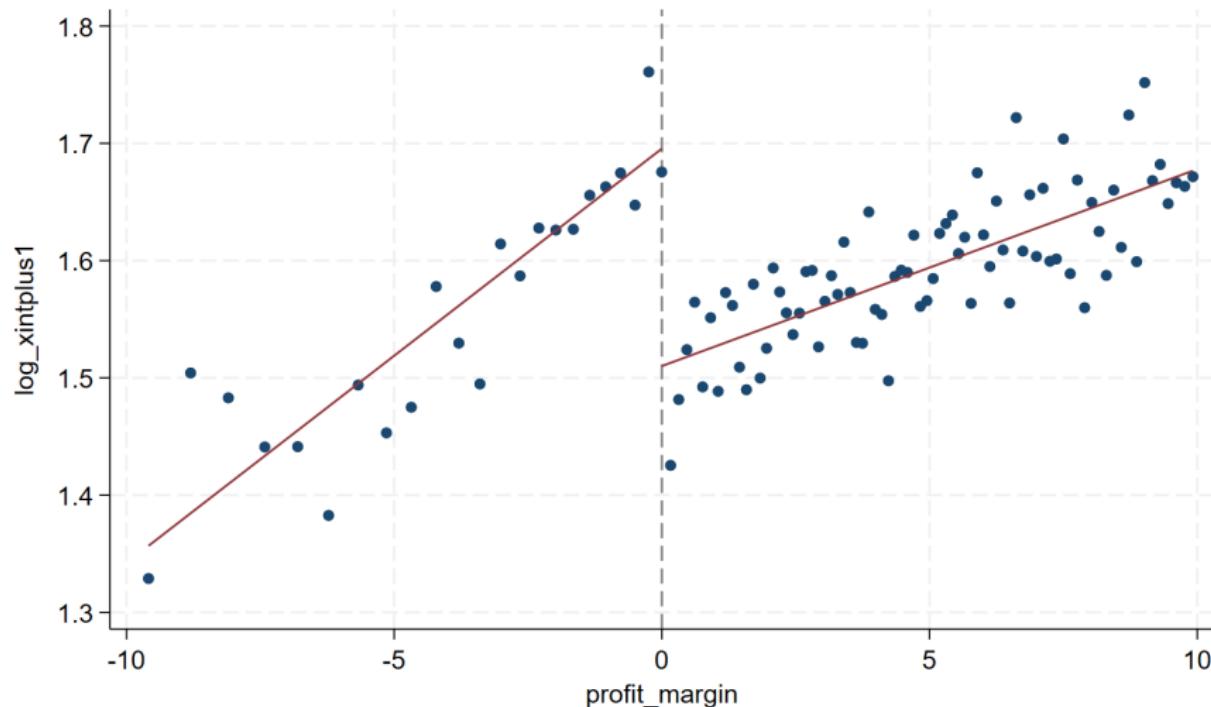
Stronger drop in research and development



**Figure 17:** Manipulators boost profits but cutting Research & Development (XRD)

## Interest Expenses

Interest expenses drop



**Figure 18:** Interest expenses more favourable on the RHS

## Employment

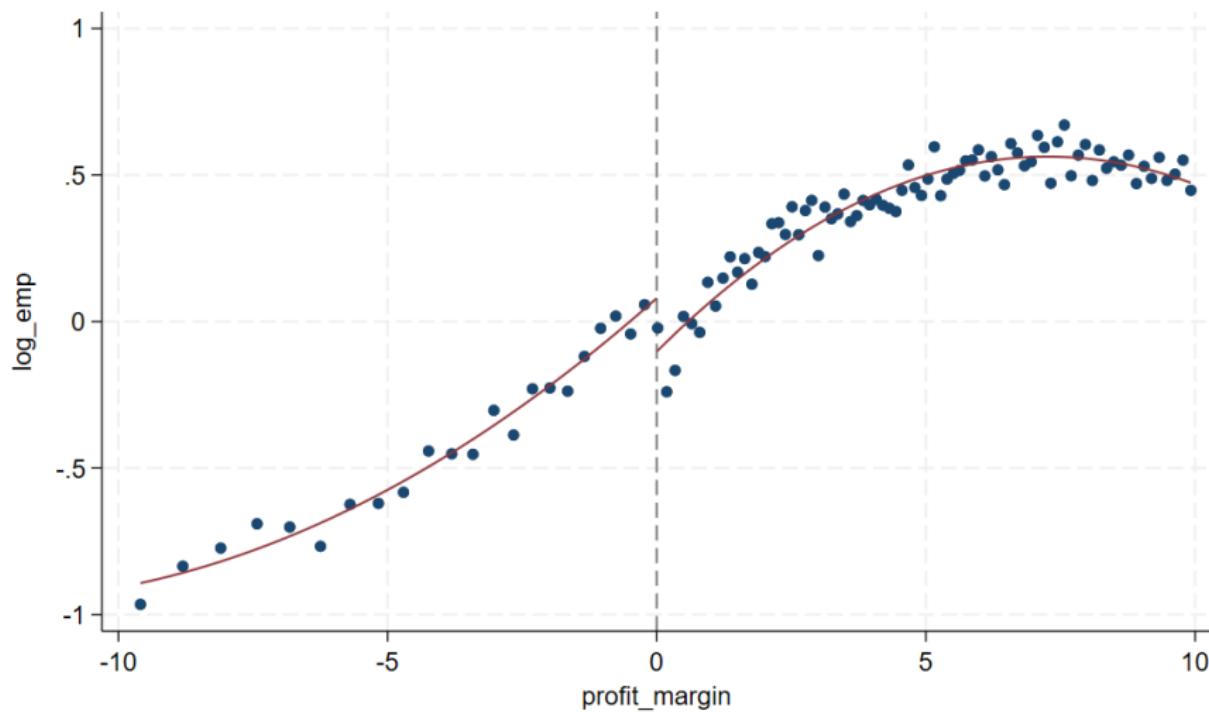
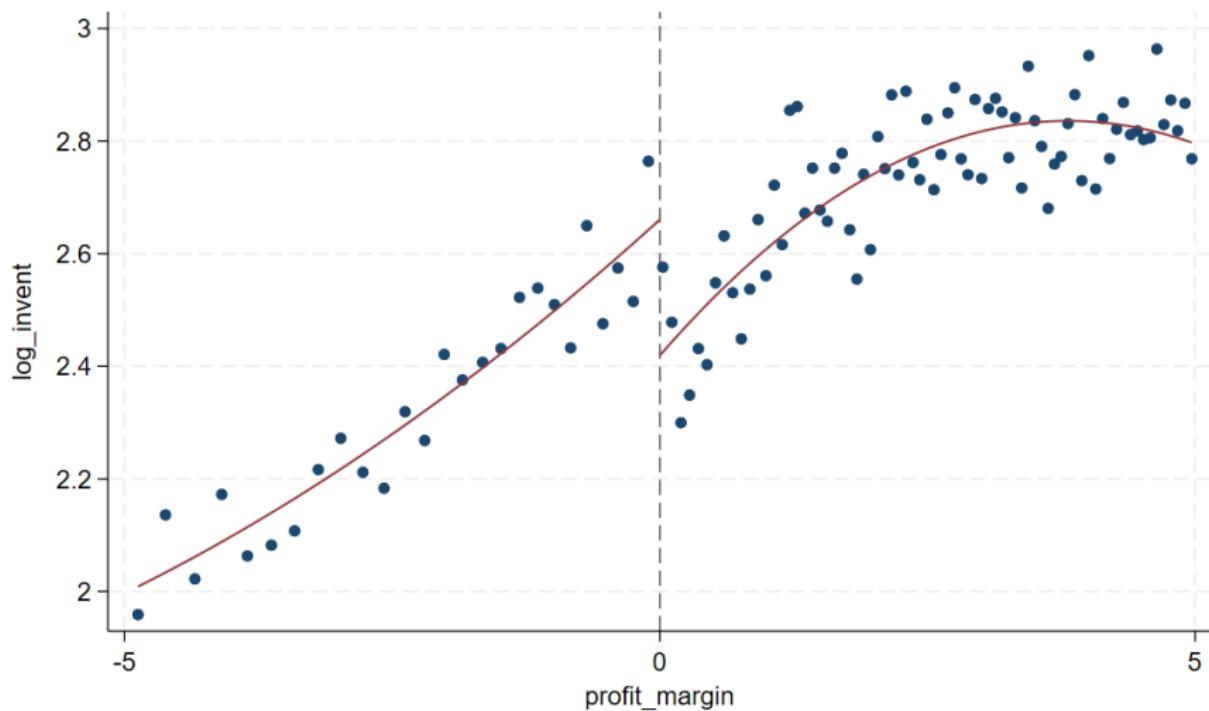


Figure 19: Employment cuts?

## Inventories



**Figure 20:** Lower stocks of inventories

	(1) log_real_capx	(2) log_xkapplus1	(3) log_xrdplus1	(4) log_xrd	(5) log_invent
posprofits	-0.242 *** (0.0580)	-0.246 *** (0.0571)	-0.139 *** (0.0405)	-0.322 *** (0.0950)	-0.238 *** (0.0501)
Observations	90122	91642	40476	27426	88490

Standard errors in parentheses

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

	(1) log_cash	(2) log_emp	(3) log_xint	(4) log_xintplus1	(5) xint_share
posprofits	-0.157 *** (0.0575)	-0.122 ** (0.0486)	-0.238 *** (0.0501)	-0.135 *** (0.0311)	-0.0845 (0.101)
Observations	80077	87038	88490	92112	84697

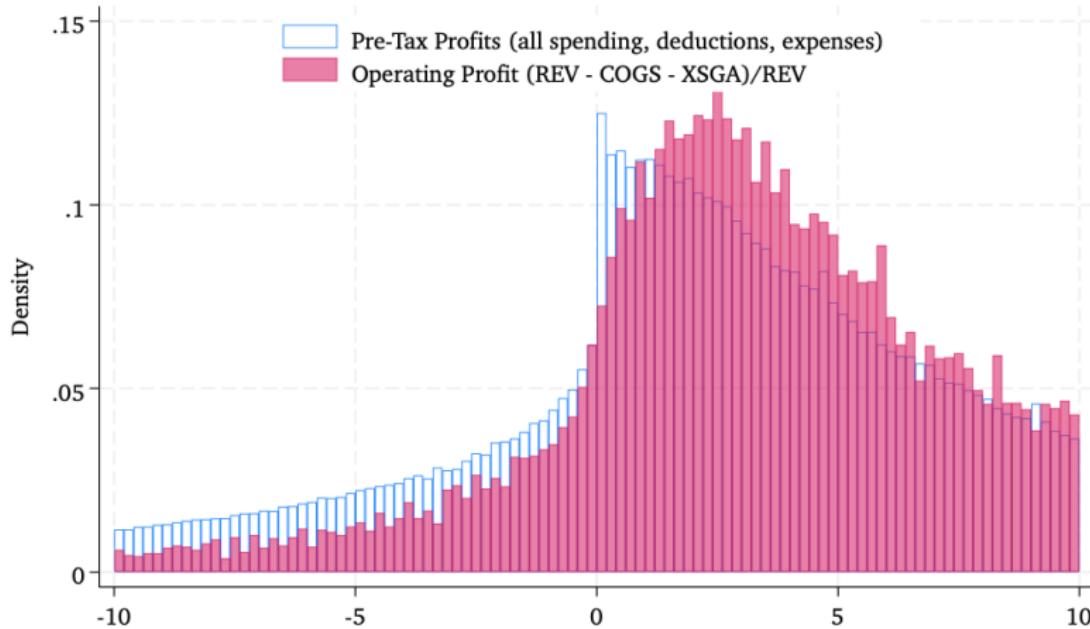
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\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

Note: RD regression, polynomial order  $p = 2$  width fixed bandwith  $h = 10$ . Controls: year, sector, age proxy. Robust standard errors clustered at firm and year levels.

**Table 1:** Simple RD regressions

## Operating Profit



**Figure 21:** Net Profit and Operating Profit

No OPEX related jump: not a story about output, driven by interest, depreciation, etc

[◀ Back to Density](#)