

Do larger firms exert more market power?

Markups and markdowns along the size distribution

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- Extensive debate on rising concentration
 - Potentially good for aggregate productivity
 - Potentially harmful to competition
- A critical Q: *Do larger firms exert more market power?*
 - Most theories assume they do (e.g. Cournot, Bertrand, linear demand)
 - Consistently with Marshall's second law of demand
 - Moving towards lower prices, the demand becomes less elastic
 - Implications for several economic outcomes
 - Resource allocation, factor shares, gains from trade, etc

Preview of the results

- Standard definitions
 - (Price) Markup: price over marginal cost ($\equiv \mu_{it}$)
 - (Wage) Markdown: MRPL over wage ($\equiv \gamma_{it}$)
- Empirical findings:
 - 1 Unexpectedly, larger firms have lower markups
 - 2 Their markdowns are instead (much) higher
 - 3 *Controlling for markdowns*, markups grow in size
 - 4 Markups and markdowns are negatively correlated
- We discuss a potential mechanism
 - Consistent with the Marshall's second law of demand

- 'Production approach' to markup (and markdown) estimation
 - Hall (1986); Dobbelaere & Mairesse, (2013); De Loecker & Warzynski, (2012)
- Application to recent trends
 - De Loecker et al. (2020); Autor et al. (2020); Mertens (2021); Yeh et al. (2022); Deb et al. (2022); Bighelli et al. (2023)
- Discussion of the related issues:
 - Unobserved monopsony (Hashemi et al., 2022)
 - Price bias (Bond et al., 2021; De Ridder et al., 2022)
 - Non-neutral technology (Demirer, 2022; Raval, 2023; Raval, 2023b)

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The 'production approach': intuition

- Ratio estimators are derived from firms' optimality conditions
 - In perfect competition, expenditure shares equal output elasticities
 - Both markups and markdowns generate a wedge between the two
 - Take a flexible input with no markdown (e.g. raw materials):
 - Markup = its output elasticity over its expenditure share
 - With monopsony, the formula above includes the markdown, too
 - Markdown = 'labour ratio' over 'materials ratio'

The 'production approach': pros and cons

- Main advantages
 - No need to estimate firm-specific product demand and labour supply
 - No need to make assumptions on the 'source' of market power
 - A convenient method to apply in a variety of countries and industries
- Main critical assumptions (Bond et al., 2021)
 - 1 No monopsony over the flexible input
 - 2 No adjustment costs in the flexible input
 - 3 Flexible input is not used to affect the demand

Estimators from cost minimization

- $\text{Min}\{r_{it}K_{it} + w_{it}L_{it} + z_{it}M_{it}\}$ s.t. $Q_{it} > Q^*$
 - FOC(M_{it}): $z_{it} = \lambda_{it} \frac{\partial Q_{it}}{\partial M_{it}}$
 - Combine with $\mu_{it} = \frac{P_{it}}{\lambda_{it}}$ and $\theta_{it}^M \equiv \frac{\partial Q_{it}}{\partial M_{it}} \frac{M_{it}}{Q_{it}}$ to get the markup:

$$\mu_{it} = \theta_{it}^M \frac{P_{it} Q_{it}}{z_{it} M_{it}}$$

- FOC(L_{it}): $w_{it} \left(1 + \frac{\partial w_{it}}{\partial L_{it}} \frac{L_{it}}{w_{it}}\right) = \lambda_{it} \frac{\partial Q_{it}}{\partial L_{it}} = \frac{P_{it}}{\mu_{it}} \theta_{it}^L \frac{Q_{it}}{L_{it}}$
- Substitute the equation from markup to get $\left(1 + \frac{\partial w_{it}}{\partial L_{it}} \frac{L_{it}}{w_{it}}\right) = \frac{\theta_{it}^L z_{it} M_{it}}{\theta_{it}^M w_{it} L_{it}}$
- Profit maximization implies $\gamma_{it} = \left(1 + \frac{\partial w_{it}}{\partial L_{it}} \frac{L_{it}}{w_{it}}\right)$, hence the markdown:

$$\gamma_{it} = \frac{\theta_{it}^L}{\theta_{it}^M} \frac{z_{it} M_{it}}{w_{it} L_{it}}$$

Estimators from 'efficient bargaining'

- Consider the following problem: $\text{Max}\{(w_{it}L_{it})^{\phi_{it}}\Pi_{it}^{1-\phi_{it}}\}$
- The estimators of μ_{it} and γ_{it} remain the same
 - Dobbelaere & Mairesse (2013), Mertens (2021), Chern Wong (2023)
- The markdown is now the outcome of two opposite forces
 - Monopsony power of firms vs bargaining power of workers
 - It lies below unity if worker power is relatively higher

Cobb-Douglas vs Translog

- What if we assume a Cobb-Douglas production function?

$$\mu_{it} = \theta^M \frac{P_{it} Q_{it}}{z_{it} M_{it}} \quad \gamma_{it} = \frac{\theta^L}{\theta^M} \frac{z_{it} M_{it}}{w_{it} L_{it}}$$

- If output elasticities are constant across firms of the same industry:
 - *Levels* of markups and markdowns need production function estimates
 - But their *variation* is fully captured by the expenditure shares alone
- What if output elasticities vary across firms?
 - The estimation of (translog) production functions becomes necessary
 - We need prices

A dual approach

- 1 Cobb-Douglas assumption for a large database
 - Market power vs firm size across 19 European countries
 - 2 Richer specification for a sub-sample with firm-level prices
 - Same analysis, but using quantity-based translog output elasticities
 - In both cases: *no price bias*
- Our procedure for #2 follows the literature:
 - Control function for productivity (Olley & Pakes, 1996)
 - Material as flexible input (Levinsohn & Petrin, 2003)
 - Control function for unobserved input prices (De Loecker et al., 2016)
 - Firm-level price indices for multi-product firms (Eslava et al., 2004)

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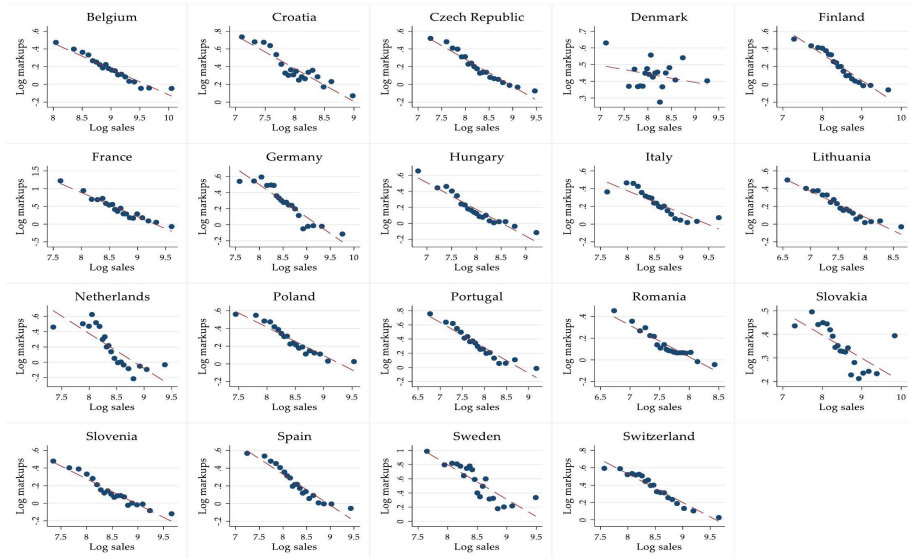
- Micro-based indicators from firm-level administrative data
 - Aggregated at the 2-digit industry level for confidentiality issues
- It includes:
 - Percentiles and dispersion measures of firm-level variables
 - Parametric indicators (TFP, markups, markdowns, MRP, OP gap)
 - Joint distributions of two variables
 - E.g. Average markups by quintile of sales
 - Or average markdown by quintile of markup
- Firm-level prices are not available
 - Our CompNet-based analysis relies on the Cobb-Douglas assumption

- Firm-level data from German administrative sources
 - Includes manufacturing firms with > 20 employees
 - Rotating sample, covers 40% of the underlying population
 - Time period observed: 1995-2016
- All necessary information for production function estimation
 - Balance-sheet data
 - Income statements
 - NACE classifications
- And, remarkably, product-level prices and quantities, up to 10-digit
 - E.g. "Workwear - Long trousers for men, cotton"

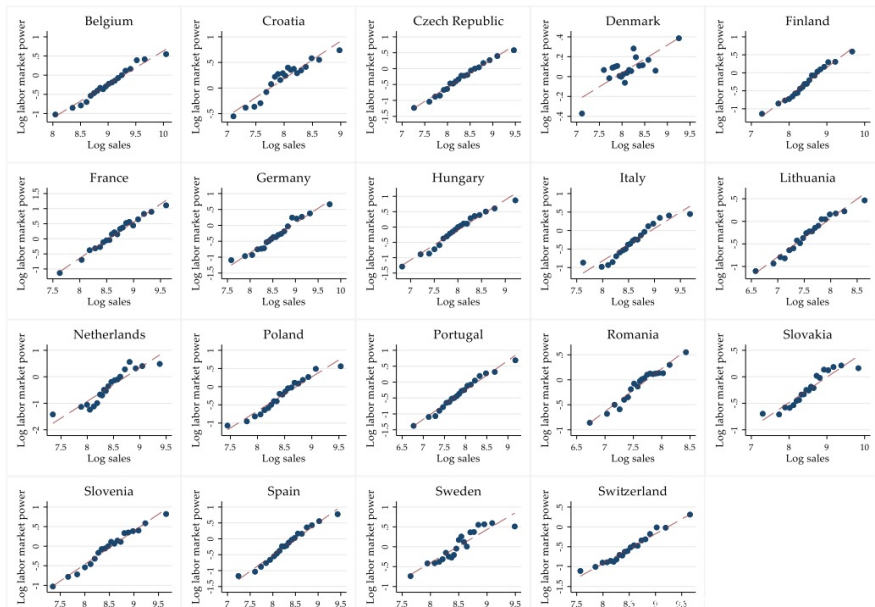
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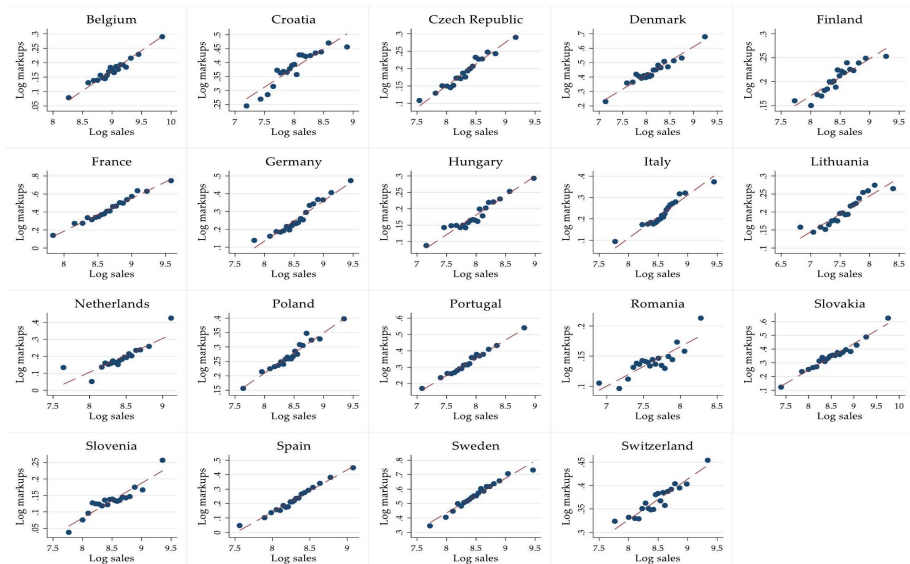
CompNet data: markups vs sales



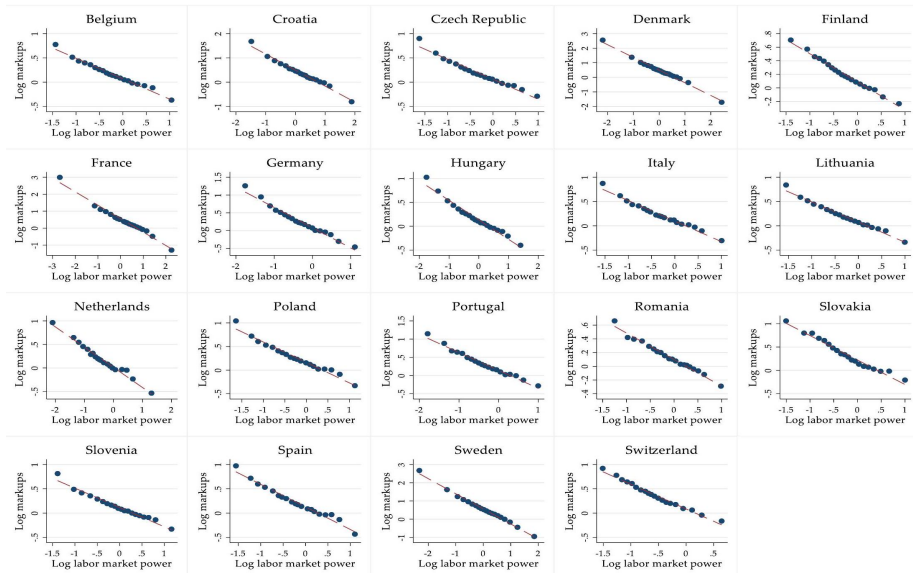
CompNet data: markdowns vs sales



CompNet data: markups vs sales controlling for markdowns



CompNet data: markups vs markdowns



AFiD data: same 4 charts

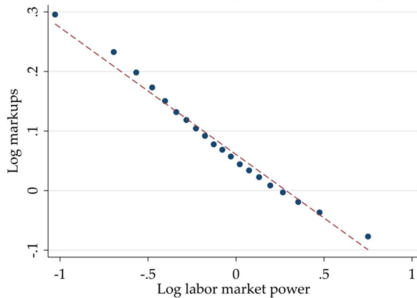
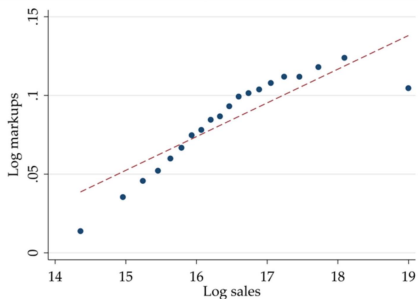
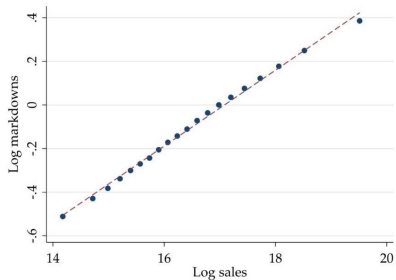
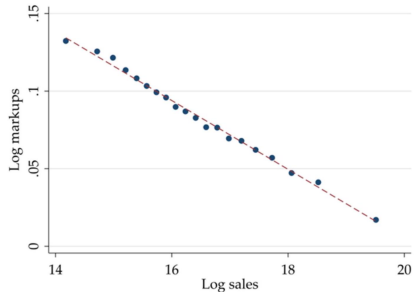


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A possible mechanism

- Efficient bargaining, with worker power depending on firm size

- $\text{Max}\{(w_{it}L_{it})^{\phi_{it}(Q_{it})}\Pi_{it}^{1-\phi_{it}(Q_{it})}\}$

- Taking the FOC wrt quantities, and rearranging, we get:

- $\left(\frac{\partial P_{it}}{\partial Q_{it}} \frac{Q_{it}}{P_{it}} + 1\right) - \frac{1}{\mu_{it}} = \frac{\phi'(Q_{it})}{1-\phi(Q_{it})} \log\left(\frac{\Pi_{it}}{w_{it}L_{it}}\right) \frac{\Pi_{it}}{P_{it}}$

$$\phi'(Q_{it}) < 0 \rightarrow \mu_{it} < \frac{1}{\frac{\partial P_{it}}{\partial Q_{it}} \frac{Q_{it}}{P_{it}} + 1}$$

- Larger inequality for large firms with high wage markdowns
- *Intuition*: profitable firms cut prices to expand and reduce rent-sharing

- ① Misallocation and optimal policies
- ② Concentration and factor shares
- ③ Gains from trade
- ④ Demand-based estimations
- ⑤ Pass-through and rent-sharing

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- Novel empirical results on firm-level markups
 - Robust to typical issues:
 - Unobserved monopsony, price bias, non-neutral technology, etc.
 - Consistent across 19 European countries (all available)
- Key findings:
 - 1 Larger firms appear to have lower markups and higher markdowns
 - 2 Evidence suggests a trade-off between product and labour market power

Thank you!

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Assumption (1/3): Unobserved monopsony

- A relevant threat for studies that use labour or COGS
 - From Hall (1986): $\mu_{it}^L = \theta_{it}^L \frac{P_{it} Q_{it}}{w_{it} L_{it}} = \mu_{it} \gamma_{it}$
 - This can explain different results on the markups-size correlation
 - Some claims might be reinterpreted in terms of *labour* market power
- Intermediate input markets are typically assumed to be competitive
 - If not, upward bias in markups, downward bias in markdowns
- Evidence from Morlacco (2020) on imported inputs:
 - Large firms have more monopsony over materials than small firms
 - Correcting this bias would reinforce our results

Assumption (2/3): Adjustment costs

- Intermediate inputs are typically assumed to be flexible
 - E.g. Hall (2004) use them to estimate adjustment costs in K_{it} and L_{it}
- Labour adjustment costs lead to an upward bias in markdowns
 - But Yeh et al. (2022) show that the impact is negligible (+3%)
 - Much smaller than the gap in markdowns between small and large firms

Assumption (3/3): Marketing

- Downward bias in markups if the flexible input is used to affect the demand
- For robustness, we test our correlations separately for firms producing:
 - 1 Consumption goods and services
 - 2 Capital goods and services
 - 3 Intermediate goods and services
- No significant difference

Non-neutral technology and expenditure shares

- Does a labour-augmenting productivity shock affect the output elasticities?
 - It depends on the elasticity of substitution
 - With Cobb-Douglas ($\sigma = 1$), no effect
 - If inputs are gross complements ($\sigma < 1$), then θ_{it}^M goes up and θ_{it}^L down
- Heterogeneous expenditure shares may depend on this:
 - $\sigma < 1$, while larger firms have higher levels of labour-augmenting tech
- Are translog elasticities flexible enough to capture this?

$$\frac{\theta_{it}^L}{\theta_{it}^M} = \frac{\beta^L + 2\beta^{LL} l_{it} + \beta^{KL} k_{it} + \beta^{ML} m_{it}}{\beta^M + 2\beta^{MM} m_{it} + \beta^{KM} k_{it} + \beta^{LM} l_{it}}$$

Translog estimates with non-neutral technology

- Raval (2023) suggests splitting firms in quintiles by the ratio $\frac{z_{it}M_{it}}{w_{it}L_{it}}$
 - Absent labour market power, the ratio is proportional to labour-augmenting productivity
- ...and then measuring output elasticities as cost shares, by quintile
- But cost shares are not informative with imperfect labour markets
- We can run separate translog estimates in the same quintiles
 - If non-neutral technology is the major force behind our results, this solution should generate sufficient heterogeneity in output elasticities

Non-neutral technology: a robustness check

