

# Increase in Turbulence and Market Power

Agnieszka Markiewicz<sup>1</sup> and Riccardo Silvestrini<sup>2</sup>

<sup>1</sup>Erasmus School of Economics and Tinbergen Institute

<sup>2</sup>Erasmus School of Economics and Tinbergen Institute

*38<sup>th</sup> Meeting of the European Economic Association (EEA 2023)*

Barcelona School of Economics

28<sup>th</sup> August - 1<sup>st</sup> September 2023

# MOTIVATION

Over the last decades, the U.S. economy was characterized by:

- Increase in **market power**: increase in profit margins, price markups and market concentration [DEU (QJE 2020), Autor et al. (QJE 2020)]
- Increase in **turbulence**: a decline in the persistence of firms' idiosyncratic productivity [Bloom et al. (ECMA 2018), Dong et al. (2022)]

# SECTORAL HETEROGENEITY

However, these trends are **heterogeneity** across sectors, with a clear positive correlation between them:

- Sectors characterized by a sharp increase in turbulence → strong(er) increase in markups and concentration.
- Sectors with flat, or even decreasing, turbulence → weak or no increase in market power.

The goal of this paper is to build a **theoretical model** that can rationalize these findings.

## PREVIEW OF RESULTS

- In our framework, a sector-specific increase in turbulence can generate the heterogeneity in sectoral market power outcomes.
- **Mechanism:** an increase in turbulence shortens leadership duration and triggers reallocation *toward high-markup firms*.
- Empirically, we confirm that high-turbulence sectors **only** are characterized by this reallocation.

# DATA

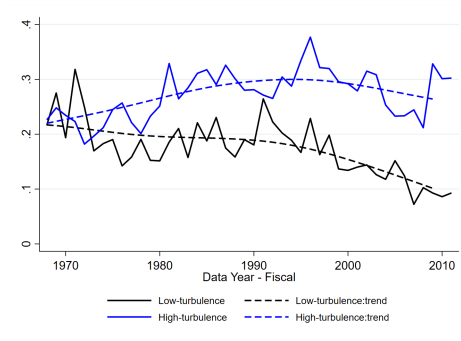
→ Compustat dataset, NAICS-3 sectors (*Appendix: CompNet*).

We measure **sectoral turbulence** as 1 – the 5-year Spearman's rank correlation of firm-level productivity.

- We compute sectoral turbulence for each pair year-sector, using a rolling window.
- We split the economy in two: low-turbulence sectors, i.e. below median turbulence growth, and high-turbulence sectors.

# HETEROGENEOUS INCREASE IN TURBULENCE

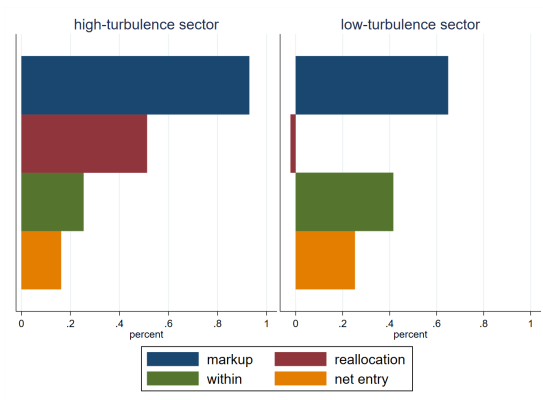
**Figure:** Turbulence trends, high vs. low-turbulence sectors





# INCREASE IN MARKUPS - DECOMPOSITION

**Figure:** Decomposition of the increase in weighted-average markups, high vs. low-turbulence sectors





# THEORETICAL FRAMEWORK

Our **dynamic** framework entails:

- ① A countable number of heterogeneous firms
  - ② Oligopolistic competition
  - ③ Idiosyncratic exit, entry and productivity shocks
- Given this competitive structure, we calibrate the model to proxy two alternative sectors, which differ in terms of **turbulence**.

# THEORETICAL FRAMEWORK - COMPETITION

- Firms compete under oligopolistic competition á la Cournot.

From the F.O.C. of a type  $x(i)$  firm, the real price  $\rho_t(i)$  is:

$$\rho_t(i) = \mu_t(i) \frac{w_t}{x(i)}$$

where  $w_t$  is the real wage. The **idiosyncratic markup**  $\mu_t(i)$  is:

$$\mu_t(i) = \left( \frac{\theta}{\theta - 1} \right) \left( \frac{1}{1 - \omega_t(i)} \right)$$

where  $\omega_t(i)$  is the type- $i$  market share.

# THEORETICAL FRAMEWORK - SHOCKS

- Productivity shocks are disciplined by a stationary Markov process.

The Markov process is crucial for our quantitative exercise, as its calibration characterizes the two sectors:

→ (changes in) its probabilities are used to capture sectoral **turbulence (shocks)**.

# METHODOLOGY

We use our framework to proxy the evolution over time of representative high and low-turbulence sectors.

- 1 We calibrate a **common** initial steady state, which replicates key features of the U.S. economy before 1980.
- 2 We shock the equilibrium by permanently changing **sector-specific primitives** to characterize the two scenarios.
- 3 We simulate the endogenous transition to the new equilibria.

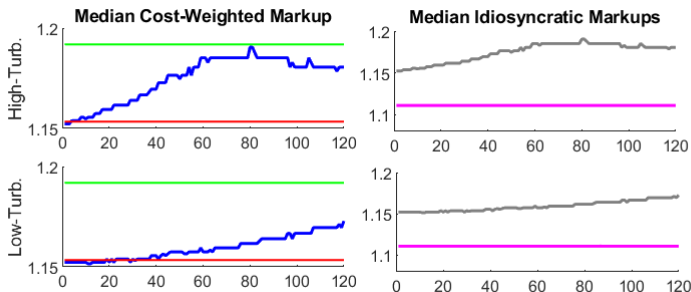
# BASELINE EXPERIMENT

We identify the **two** representative sectors as follows:

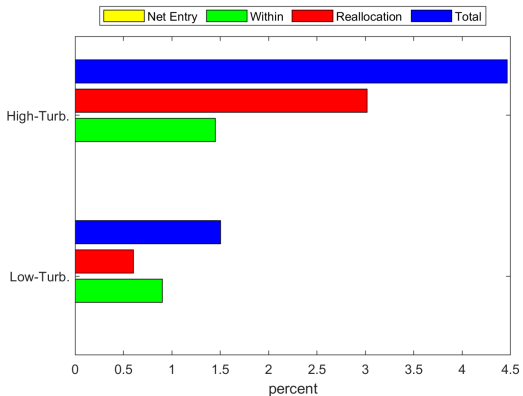
- ① A sector characterized by an increase in entry costs only, proxy for a low-turbulence industry.
  - ② A sector characterized by the **same** increase in entry costs *and* by a **sector-specific** increase in turbulence, proxy for high-turbulence.
- We add robustness checks with shocks occurring in steps, further sector-specific primitives, heterogeneous entry costs...

# SIMULATION - BENCHMARK

**Figure:** High vs. low-turbulence sectors, dynamic transition for 120 periods, median over 100 simulations. Quantities represented in levels.



# SIMULATION - MARKUP DECOMPOSITION

[◀ Robustness](#)

**Table: Model vs. data: high and low-turbulence**

	Over-time					
	High-turbulence		Low-turbulence		Ratio	
	Data	Model	Data	Model	Data	Model
$\Delta\mu_T^1$	6.13	3.50	5.56	2.60	1.10	1.35
$\Delta\mu_T^2$	8.02	3.50	5.54	2.60	1.48	1.35
$\Delta d_T$	23.98	21.06	16.19	17.03	1.48	1.24
	Cross-section					
	High-turbulence		Low-turbulence		Ratio	
	Data	Model	Data	Model	Data	Model
$\sigma_{\Delta\mu_T^1}$	0.554	0.016	0.246	0.007	2.252	2.178
$\sigma_{\Delta\mu_T^2}$	0.404	0.016	0.283	0.007	1.425	2.178
$\sigma_{\Delta d_T}$	2.094	0.298	1.461	0.248	1.431	1.200



# CONCLUSIONS

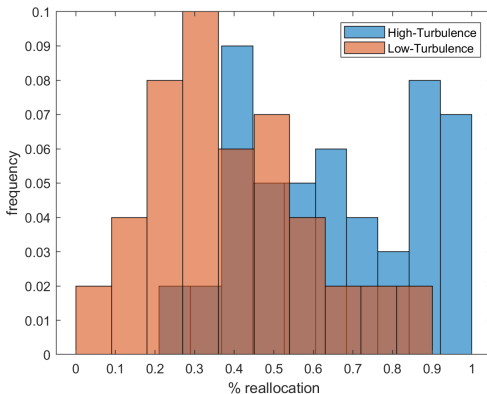
- Empirically, high-turbulence sectors present a stronger increase in markups and concentration.
- When paired to an increase in entry costs, a **sector-specific** increase in turbulence generates the sectoral heterogeneity in market power dynamics.
- The **reallocation** toward high-markup firms explains the observed trends, both in the data and the model.

## TURBULENCE AND MARKUPS

$\mu_{it} = \alpha_i + \beta\tau_{it} + \gamma_t + \epsilon_{it}$		
Markup	Cost-weighted	Rev-weighted
$\mu^1$	0.019** (0.008)	0.034** (0.012)
$\mu^2$	0.023*** (0.010)	0.037*** (0.012)
$\mu^3$	0.38*** (0.008)	0.052*** (0.012)
$\mu^4$	0.25*** (0.007)	0.036*** (0.009)
$\mu^5$	0.037*** (0.003)	0.052*** (0.012)
N. obs.	4309	4309

**Notes:** This table reports the correlation coefficients from the regressions above.  $\mu_{it}$  and  $\tau_{it}$  represent, respectively, the markup and turbulence in sector  $i$  and year  $t$ , while  $\alpha_i$  and  $\gamma_t$  are sector and time fixed effects.

## SIMULATION - REALLOCATION

[Go Back](#)