

TECHNOLOGICAL SYNERGIES, HETEROGENEOUS FIRMS, AND IDIOSYNCRATIC VOLATILITY

Jesús Fernández-Villaverde, *U Penn & NBER*

Yang Yu, *Shanghai Jiaotong U*

Francesco Zanetti, *U Oxford*

Discussion by

Stephen J. Terry, *U Michigan & NBER*

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WHAT DOES THE PAPER PROVIDE?

1. New micro facts on US firm trading relationships: sorting & separation of mismatched firms
2. New GE model: supermodular combo of firm productivity, search & matching with steady state and cycles
3. Quantitative analysis
 - ▶ Calibrated model reproduces key facts
 - ▶ Large steady state output loss from firm mismatch
 - ▶ Uncertainty shocks drive recessions via mismatch

WHY I REALLY LIKE THE PAPER

1. The facts are rich and tightly linked to the modeling
 - ▶ Assortative matching \rightarrow supermodularity
 - ▶ Separations facts \rightarrow search frictions
2. The model is novel, rich, yet still maintains transparency
 - ▶ Micro heterogeneity, sorting, search, and GE
 - ▶ Steady state and cyclical quantitative analysis
3. The quantitative implications are meaningful
 - ▶ A nice fit to untargeted facts
 - ▶ Large steady state output losses from mismatch
 - ▶ Uncertainty shocks drive a recession via a new channel

WHAT ARE MY QUESTIONS?

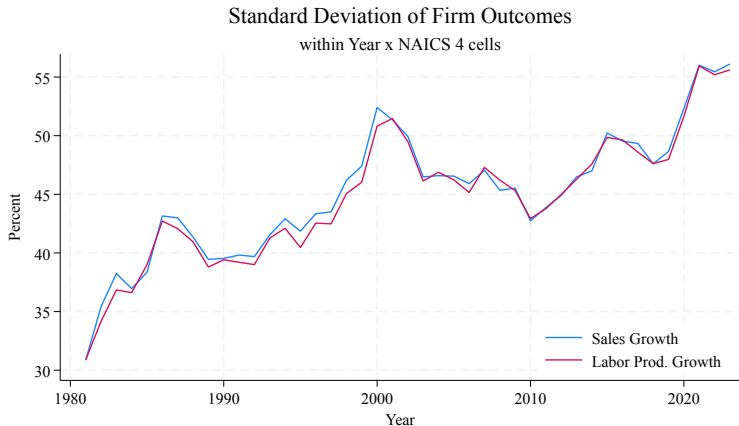
With their quantitative framework and relationship data, the authors are well placed to shed light on the following questions:

1. Where is the “big money” for quantitative analysis with this model? Volatility cycles or trends in volatility?
2. Do the details of the firm-level stochastic process matter here for quantitative analysis?
3. How many members should a match include in this quantitative model?

#1: VOLATILITY CYCLES OR TRENDS?

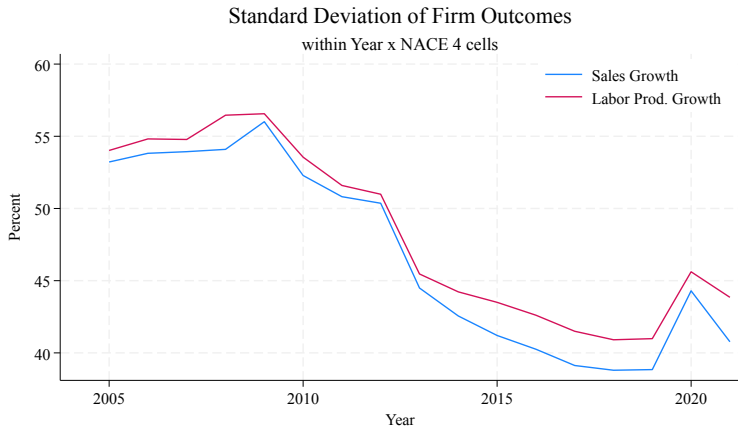
- ▶ The paper combines steady state and cyclical analysis, but the steady state magnitudes are considerably larger.
- ▶ I compute and plot multiple firm-level variance trends in the next two slides. In all cases, trends swamp cycles.
 - ▶ US Compustat: unrepresentative, upward trend
 - ▶ Spanish ORBIS: more representative, downward trend
- ▶ Exploring volatility trends seems natural. An offsetting gain from declining dynamism: more stable firm matches?

#1: VOLATILITY CYCLES OR TRENDS?



Source: Compustat 1981-2023 data from the US and my calculations. Note that Compustat volatility trends are well known to be different from privately held trends and overall US trends.

#1: VOLATILITY CYCLES OR TRENDS?



Source: Historical ORBIS 2004-2021 data from Spain and my calculations. Note that this sample is more representative and includes unlisted firms in Spain.

#2: DOES THE FIRM-LEVEL STOCHASTIC PROCESS MATTER?

- ▶ This paper allows for time-varying volatility but employs an otherwise standard Gaussian AR(1) process for firms.
- ▶ Some recent papers propose alternatives:
 - ▶ Pugsley, Sedláček, and Sterk (2021) suggests ex-ante heterogeneity matters via LBD.
 - ▶ Jaimovich, Terry, and Vincent (2023) suggests fat-tailed, or leptokurtic, shocks matter via ORBIS.
- ▶ Could firm fixed effects and/or the higher frequency of small shocks in these papers reduce mismatch?

#3: HOW MANY MEMBERS SHOULD A MATCHED SET OF FIRMS INCLUDE?

- ▶ The paper uses a pair, i.e., two firms, as its matching unit.
- ▶ One could imagine real-world scenarios with partnerships or relationships with more than two firms.
- ▶ I examined 2019 Compustat customer data:
 - ▶ Mean of 3.6 links, 45% of firms have > 1 link
 - ▶ Pairwise match labor prod. corr. = 0.20.
 - ▶ Match “mean of customers” labor prod. corr. = 0.31.
- ▶ Might mismatch be less important if there is some substitutability across partners who appear collectively better matched than individual pairs of firms?

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