

# From Macro to Micro: Heterogeneous Exporters in the Pandemic

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The views put forward are those of the author and do not represent the official views of the Banque de France or the Eurosystem.

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## Introduction

Data

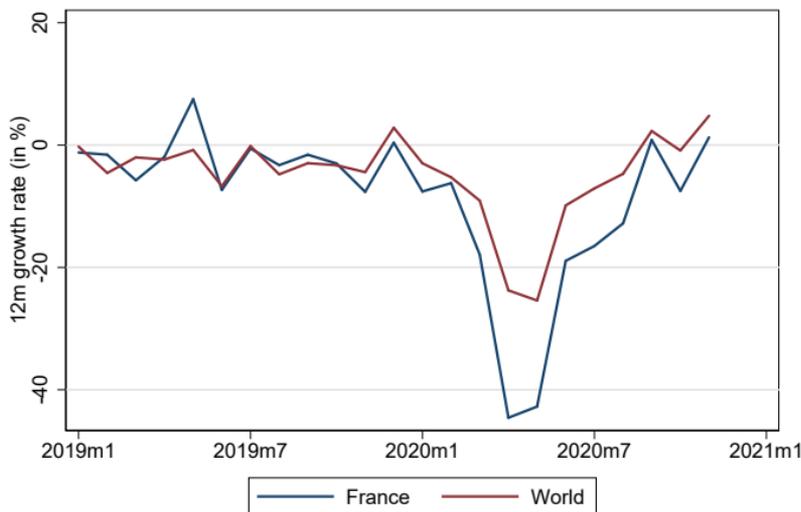
Trade margins

The collapse of top exporters

Supply versus demand

## Motivation: trade collapse in early 2020

- French exports -42% y-o-y in April 2020 (world exports: -20%)



World exports calculated from export data of 97 countries, accounting for 90% of global exports in 2018. Last data point: Nov 2020. Growth computed as 12-month midpoint growth rate

## What we do

### We study the microeconomic foundations of this collapse

- Use French transaction-level trade data Jan 2017 - Dec 2020  
→ Ex(imp)orter-by-product-by-destination-by-month.
- Decompose trade growth into margins of adjustment
- Document heterogeneity according to exporter size
- Dig deeper into the underlying drivers of such heterogeneity: supply versus demand
- Study what is behind the collapse of top exporters

## What we find

### Most of the adjustment through the firm intensive margin

- Number of exporters fell by 25%, but exporters are very small

### The largest firms drive the trade collapse

- 0.1% exporters ( $\sim 100$  exporters) account for 57% of the export collapse (pre-crisis export share: 41%)
- Top exporters do not react more to GVC disruption (intermediate good imports)
- Top exporters react more to foreign demand shocks

## Related literature

### Covid and Trade

- Impact of pandemic/lockdowns on trade flows: Antras et al. (NBER, '20), Berthou & Stumpner (mimeo '21), Espitia et al. (World Bank '21), Hayakawa et al. (JETRO WP, '20), Kejzar & Velic (Covid econ '20), de Lucio et al. (mimeo '20), Meier & Pinto (Mimeo, '20), Minondo (Applied econ '21),
- GVCs: Bonadio et al. (JIE, '21), Heise (FRBNY blog, '21), Lafrogne-Joussier, Martin, Méjean ('21)

### Large firms in international trade

- Super-star exporters: Freund & Pierola (ReStat '15)
- Aggregate effects from idiosyncratic firm shocks: Eaton et al. (NBER '12), Gabaix (Ecta'11), Gaubert & Itskhoki (JPE '21), di Giovanni & Levchenko (JPE '12), di Giovanni et al. (Ecta '14, NBER'21)

### Margins of international trade

- BJRS (AER '09), Fernandes et al. ('19)

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## Disaggregated trade data

### Monthly, firm-level export/import data, January 2017 - Dec 2020 (French Customs Office)

- Unit of observation:
  - Firm-by-CN8product-by-destination(origin)-by-month
  - For destinations inside EU: ID of the foreign partner company
- Coverage:
  - >98% of aggregate exports flows from official statistics.
  - Roughly 100k exporters per year, 45k per month.
  - Extra-EU trade: exhaustive data
  - Intra-EU trade:
    - Exporters required to file the detail of their transactions (product code, destination) if annual exports exceed 460k euros: see Bergounhon et al. (mimeo '18).
    - Importers required to file when cumulated yearly value imports exceed 460k euros

## Firm-level balance sheets / Pandemic data

### FiBEN

- Balance sheet data collected by Bank of France
- Turnover  $> 0.75$  million euros
- 200k firms / year

### Oxford stringency index

- Hale et al (2021)
- Collected daily for a sample of 180 countries
- Aggregated into indices that range from 0 to 100 and are increasing in the measures' stringency
- Country-month averages of the index normalized to lie in the interval  $[0, 1]$ .

▶ Oxford Index

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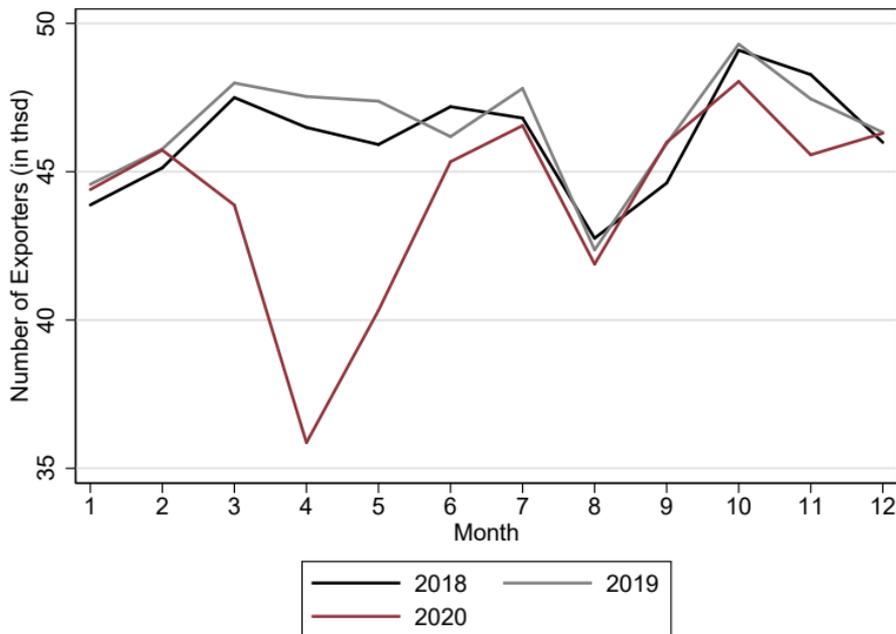
**Trade margins**

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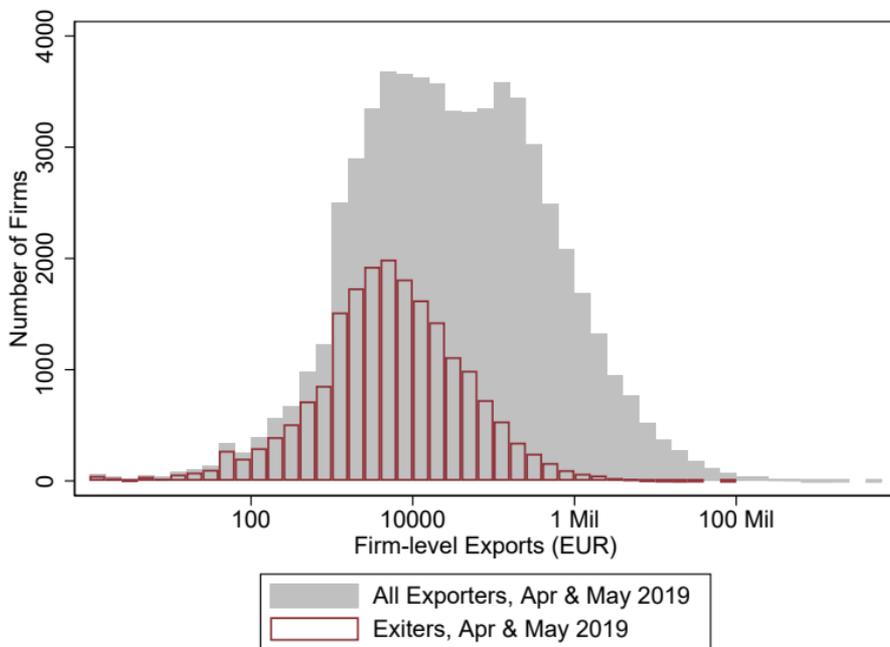
## A sharp adjustment at the Extensive Margin

- The number of exporters fell substantially (-25% in April 2020)



## Exiters hardly contribute to the aggregate collapse

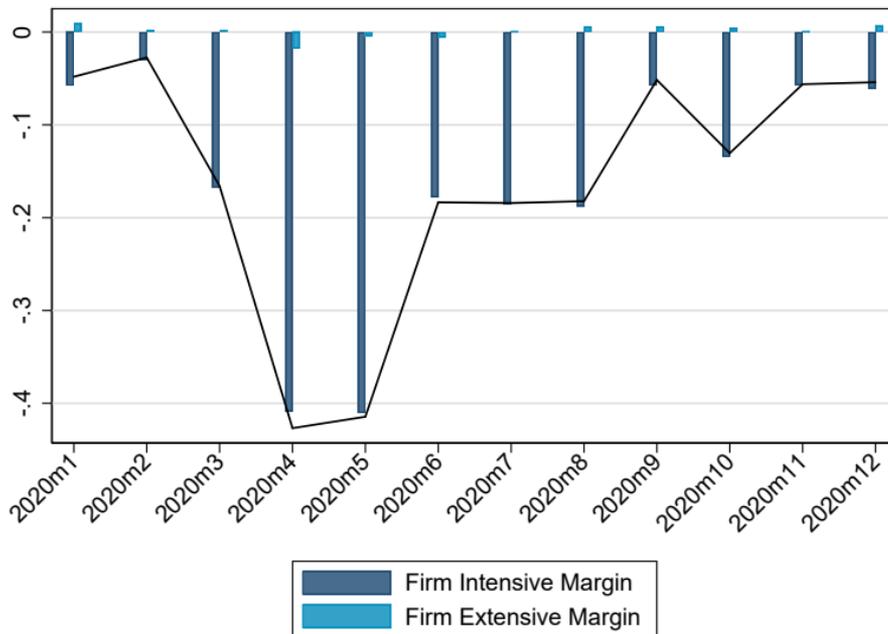
- Average exports of exiters in April & May 2019: 64k Euros → 4.5% of average of all exporters (1.4m Euros)



Note: Distribution of exporters (in grey) and distribution of exiters (in red) in April & May 2019. Exiters are defined as firms with positive exports in April & May 2019, but zero exports in April & May 2020.

## The contribution of the firm extensive margin is very small

- Decomposition: 
$$\frac{\Delta X_t}{X_{t-1}} = \frac{\sum_{f \in S} \Delta X_{ft}}{X_{t-1}} + \frac{\sum_{f \in N} X_{ft} - \sum_{f \in X} X_{ft-1}}{X_{t-1}}$$



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Data

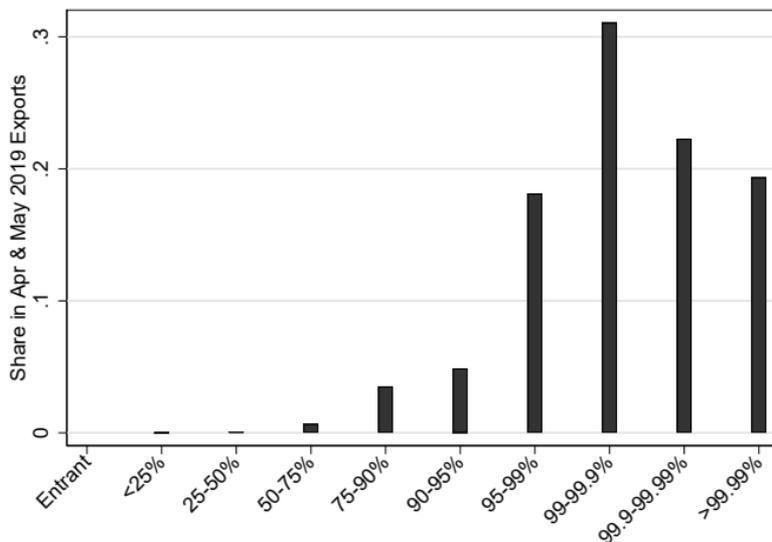
Trade margins

**The collapse of top exporters**

Supply versus demand

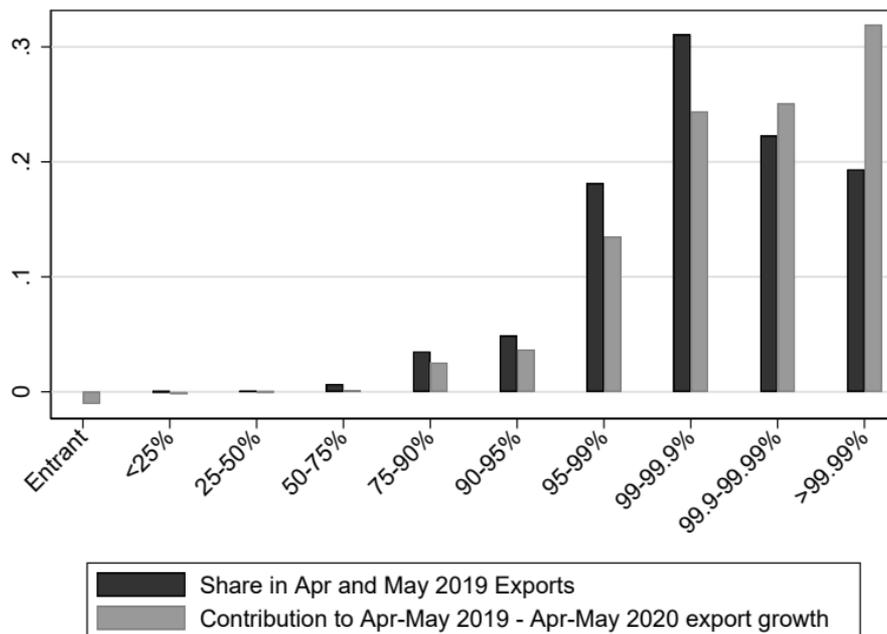
## Ten firms (legal units): one fifth of aggregate French exports

- Group exporters into bins based on their total exports in 2019:  $\approx$  100k firms
- To be in top 1% / 0.1% / 0.01%, a firm needs  $>65\text{m}$  /  $>600\text{m}$  /  $>3\text{bn}$  Euros of annual exports. Total French exports in 2019: 488bn.



## Top exporters contributed more than their share to the collapse

- Contribution of group  $i = \frac{\Delta X_i}{\Delta X}$
- Top 0.1% (~ 100 firms) contributed 57% (initial share: 41%)
- Top 0.01% (10 firms) contributed 32% (initial share: 19%)



## A flexible empirical framework: midpoint growth rates by size

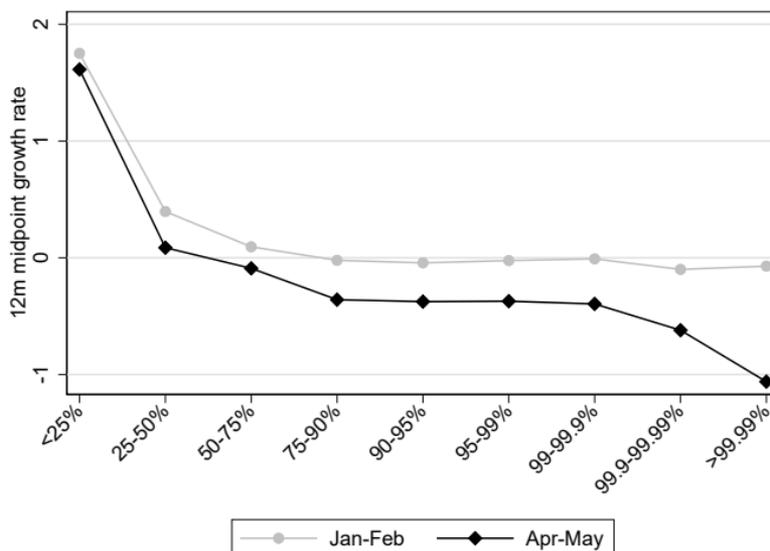
- Baseline estimation: For firm  $f$ , product  $k$ , destination (origin)  $j$  and time  $t$ :

$$g_{fjk,t} = \underbrace{\alpha_{b(f)t}}_{\text{Bins Exporter Size}} + \epsilon_{fjk,t}$$

- $g_{fjk,t}$ : year-on-year midpoint growth rate. [▶ Midpoint](#)
- Group time periods into two-months intervals
- Then add controls

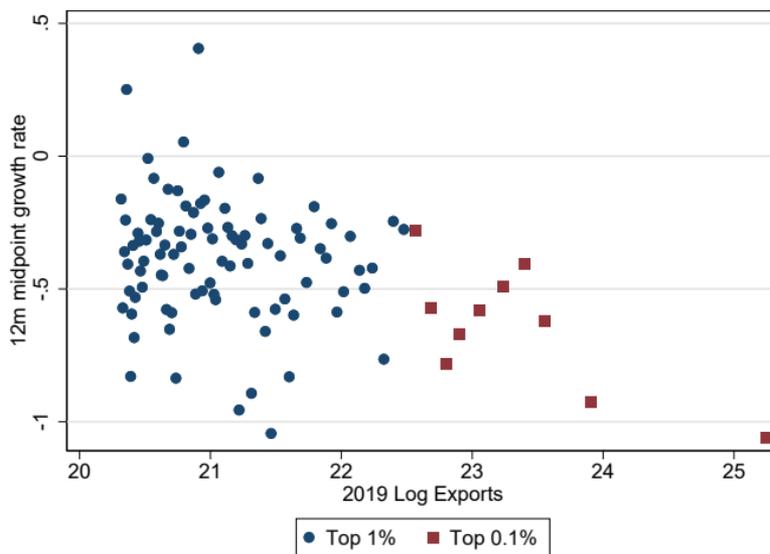
## Top exporters' growth rate fell by more in April-May

- Estimate coefficients on size dummies separately for Jan-Feb (pre-shock) and Apr-May (export collapse)



## Zooming in on the top 1,000 firms

- Placing top 1,000 firms ( $\sim$  top 1%) into 100 bins



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**Supply versus demand**

## Covid as a supply/demand shock

- Pandemic and policy responses (especially lockdowns) led to strong supply and demand responses: Closing of workplaces, work from home, shops closed, etc.
- Exporters may be affected by:
  - Foreign lockdowns through supply disruptions from intermediate inputs
  - Domestic lockdown
  - Foreign lockdowns through a demand channel
- Top exporters may have a higher elasticity with respect to these shocks

## The potential drivers of collapse of top exporters

### ■ Supply

- Higher exposure to foreign supply shocks of intermediate inputs?
- Higher elasticity to foreign supply shocks of intermediate inputs?

### ■ Demand

- Higher exposure to foreign demand shocks?
- Higher elasticity to foreign demand shocks?

## The potential drivers of collapse of top exporters

### ■ Supply

- Higher exposure to foreign supply shocks of intermediate inputs?
- Higher elasticity to foreign supply shocks of intermediate inputs?

### ■ Demand

- Higher exposure to foreign demand shocks?
- Higher elasticity to foreign demand shocks?

## Intermediate imports to sales ratio as a proxy of the dependence on value chains

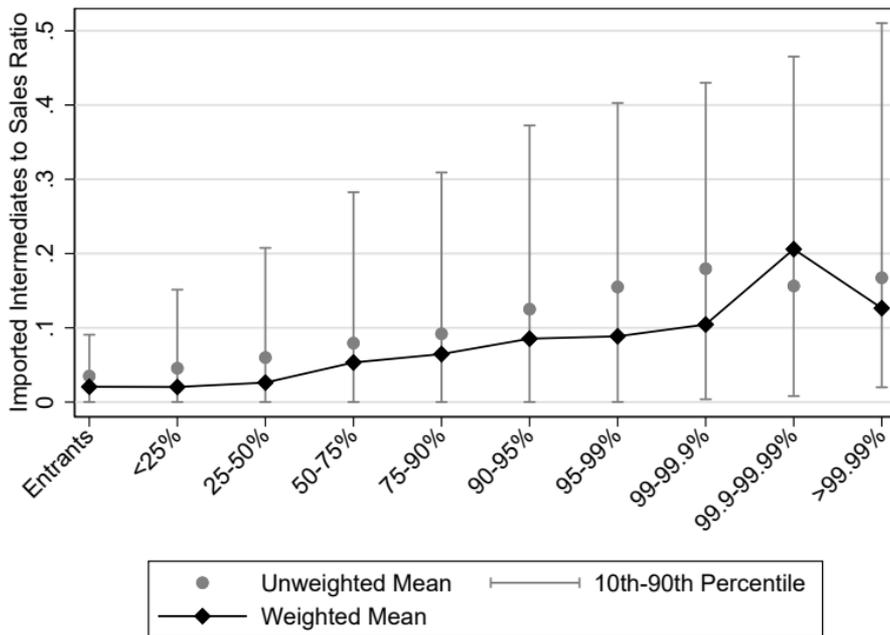
- Summarize a firm's exposure to foreign supply shocks through imported intermediate inputs by its ratio of imported intermediate inputs to sales

$$IIS_{f,2019} = \frac{X_{f,2019}^M}{Y_{f,2019}}$$

- Then control for IIS ratio in size-estimations [▶ FIBEN Subsample](#)

## IIS ratio is increasing in exporter size

- But lots of variation across exporters within a size bin
- A regression of IIS on size bin dummies only gives an R2 of 5%

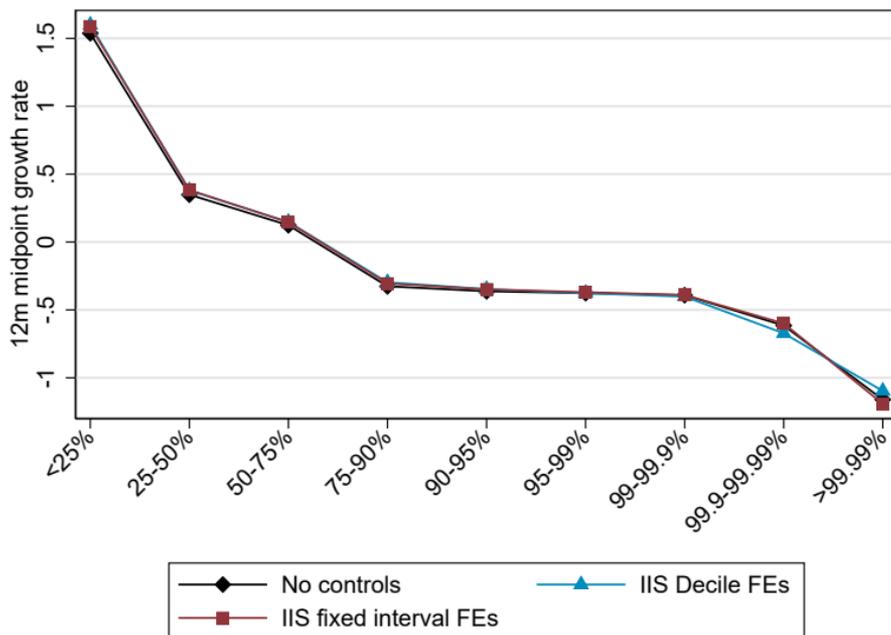


## The size effect holds when controlling for the IIS ratio

- Control for dummies of IIS ratio in size regressions:

$$g_{fkjt} = \underbrace{\alpha_{b(f)t}}_{\text{Bins Exporter Size}} + \underbrace{\gamma_{r(f)t}}_{\text{Bins IIS ratio}} + \epsilon_{fkjt}$$

- Sort IIS ratio into deciles or bins of fixed length

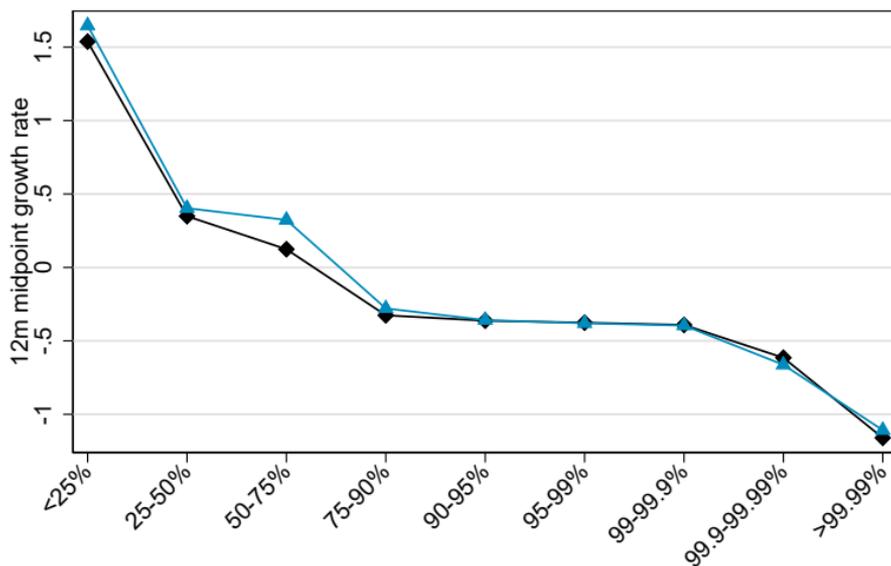


## and when controlling for origin-specific supply shocks

- Control alternatively for bins of constructed origin supply shocks:

$$\text{Input Supply Shock}_{ft} = \underbrace{\frac{M_{f,2019}^{inp}}{Y_{f,2019}}}_{\text{IIS Ratio}} \times \underbrace{\sum_j \frac{M_{fj,2019}^{inp}}{M_{f,2019}^{inp}} \text{Supply Shock}_{jt}}_{\text{Weighted Avg of Origin Supply Shocks}}$$

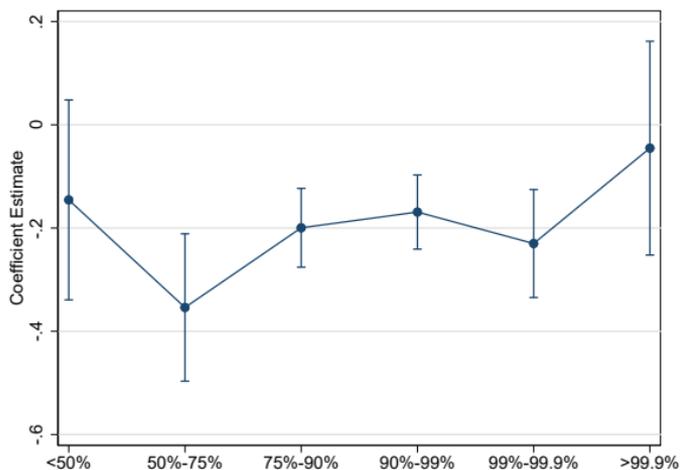
- Supply shocks in origin  $j$  taken as lockdown stringency.



## Effect of *Origin* Lockdown by Size Bin

$$g_{fjk,t} = \text{LockdownStringency}_{j,t} \times \eta_{b(f)} + \beta_{ft} + \gamma_j + \delta_{kt} + \epsilon_{fjk,t}$$

Figure : Impact of Covid at origin on imports by exporter size



Source: French customs, Author's calculation.

## The potential drivers of collapse of top exporters

### ■ Supply

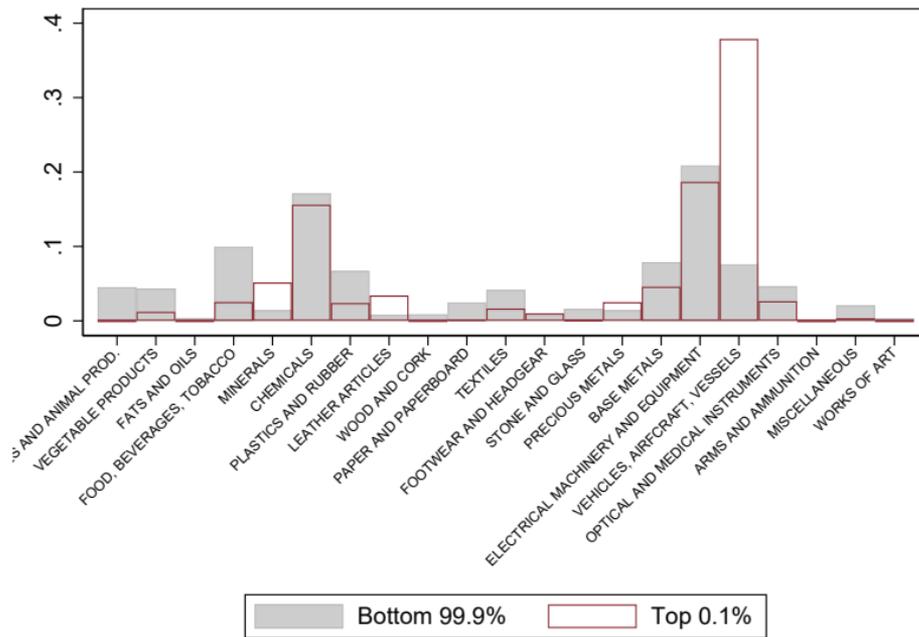
- Higher exposure to foreign supply shocks of intermediate inputs?
- Higher elasticity to foreign supply shocks of intermediate inputs?

### ■ Demand

- Higher exposure to foreign demand shocks?
- Higher elasticity to foreign demand shocks?

## Sectoral composition of bottom 99.9% vs top 0.1%

- Differences largely driven by aircrafts



## Controlling for the composition effects

- Add sector FEs to the estimation

$$g_{fjk,t} = \alpha_{b(f)t} + \beta_{st} + \epsilon_{fjk,t}$$

$\beta_{st}$ : Dummy for HS2 code of observed flow

- Add sector-by-destination FEs to the estimation

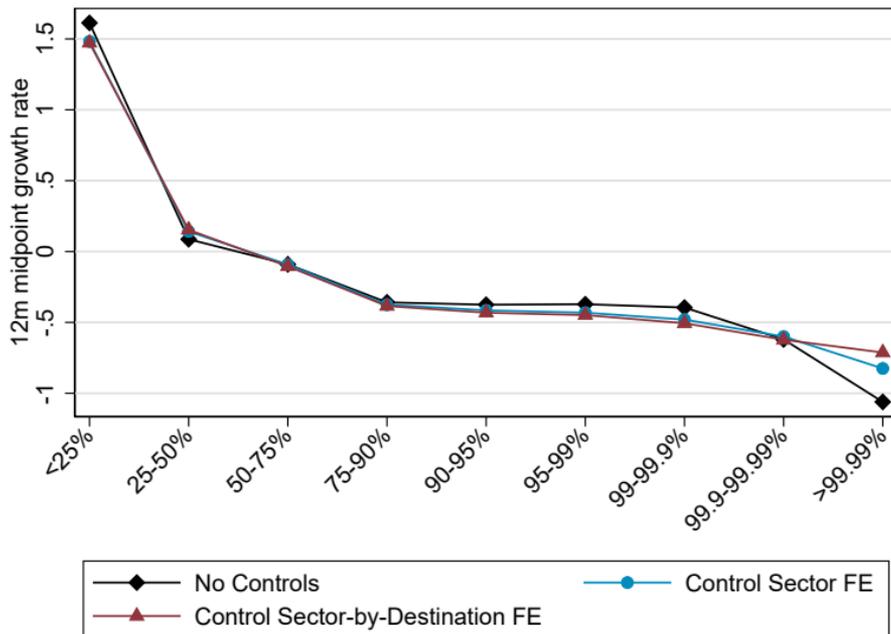
$$g_{fjk,t} = \alpha_{b(f)t} + \beta_{jst} + \epsilon_{fjk,t}$$

$\beta_{jst}$ : Dummy for the sector-by-destination cell of observed flow

## Higher exposure to foreign demand shock is only part of the story

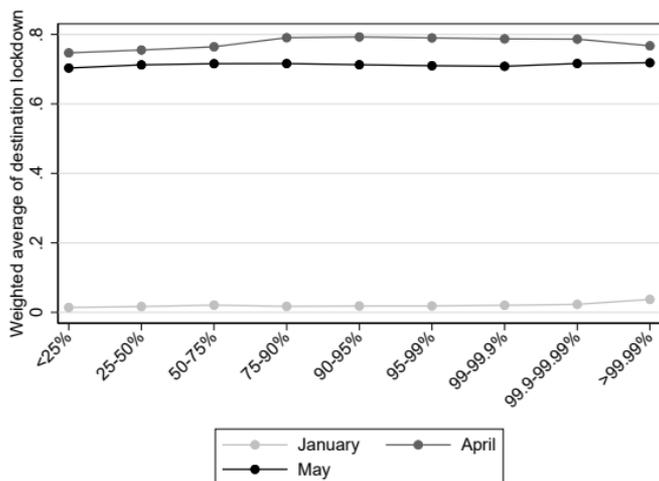
- Composition effects only partly explain the larger collapse by top exporters

→ Top exporters also fall by more **within** markets



## (Weighted) Oxford index as a direct measure of exposure to the demand shock

- Measure directly exposure to foreign demand shocks by size bin: "Oxford stringency" index from Hale et al (2021) [▶ Stringency Index](#)
- Weighted average destination lockdown stringency across size bins



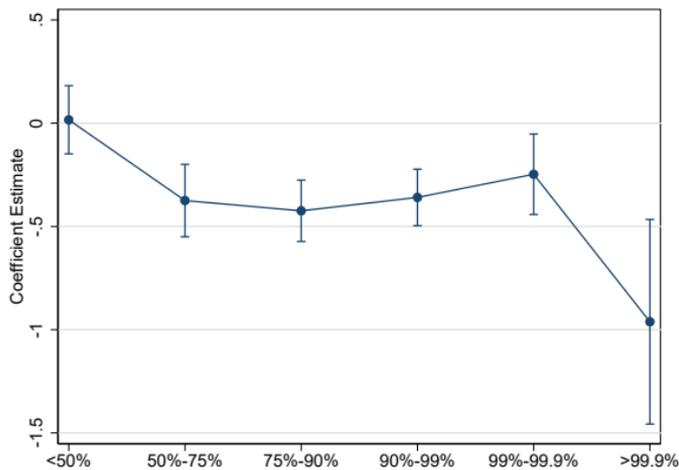
## Demand effect is larger for the top exporters

- Interact destination lockdown stringency with dummies for size bins

$$g_{fjk,t} = \text{Lockdown Stringency}_{j,t} \times \eta_{b(f)} + \beta_{ft} + \gamma_j + \delta_{kt} + \epsilon_{ijk,t}$$

- Could reflect larger absolute decline in exports or reallocation across destinations

Figure : Impact of Covid at destination by exporter size



Source: French customs, Author's calculation.

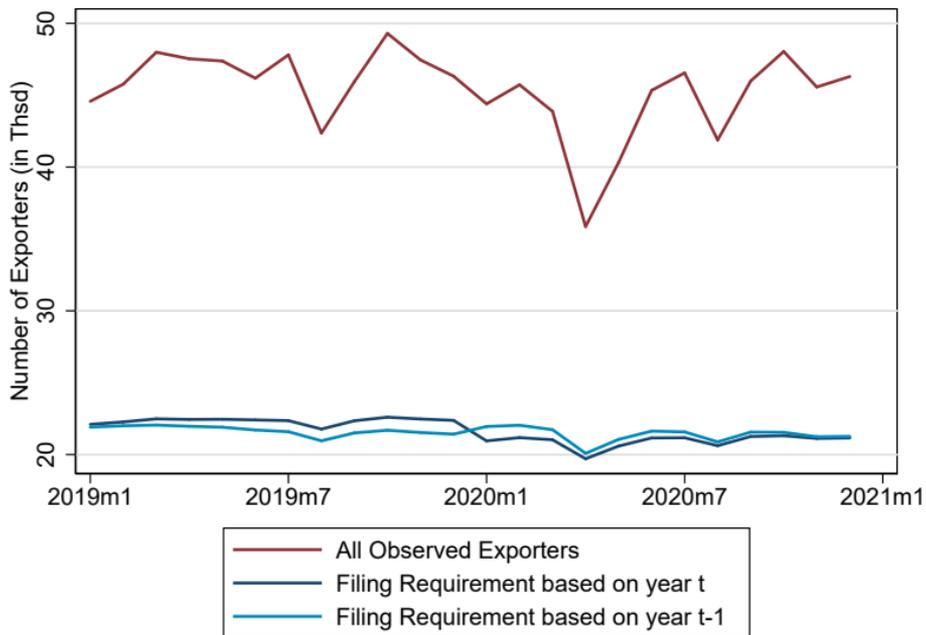
## Conclusion

- Firm intensive margin accounts for almost the entire decline of exports(imports)
- Top (top 0.1%) traders adjust procyclically
  - Differences not systematically related to GVC participation
  - Top exporters decline by more within sectors and destinations
  - Top exporters react more to foreign lockdowns during Covid
- These results point to a large reaction of top exporters to demand shocks
- Similar role of top exporters during Covid and GFC [▶ Covid vs GFC](#)

THANK YOU

# APPENDIX

## Exporters with and without filing obligation



## Midpoint growth rate accommodate churning with high frequency data

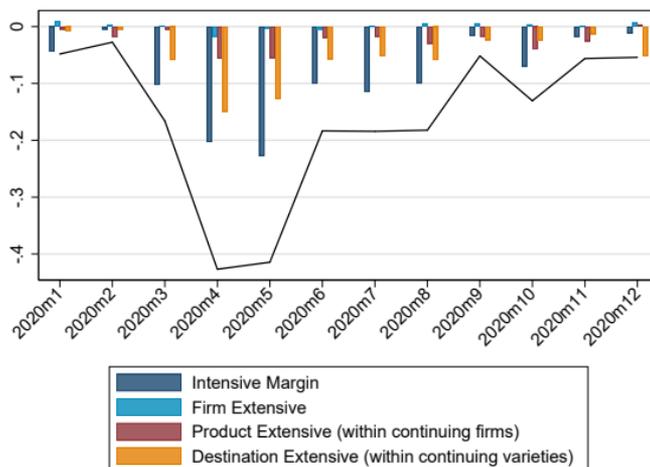
$$g_t = \frac{X_t - X_{t-12}}{\frac{1}{2}(X_t + X_{t-12})}$$

- $g_t$  varies between -2 (for exit) and +2 (for entry)
- Can accommodate entry and exit, which is important given the high level of detail (transaction = firm-by-product-by-destination) [▶ Product and Destination Margins](#)
  - see e.g. Haltiwanger et al. (Restat '13) on job creation by establishments
- For small values of  $\frac{X_t}{X_{t-12}}$ , very close to  $\log\left(\frac{X_t}{X_{t-12}}\right)$  [▶ MP vs. dlog](#)

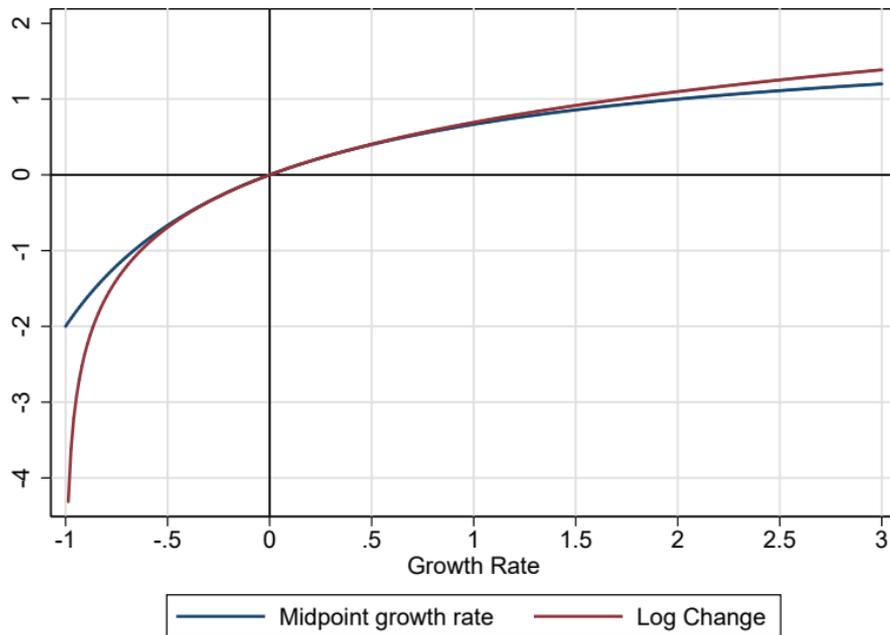
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## Decomposing the Firm Intensive Margin

- Decomposition from Bernard et al. (2009)
- *Transaction*-level intensive margin accounts for roughly half of the total export collapse

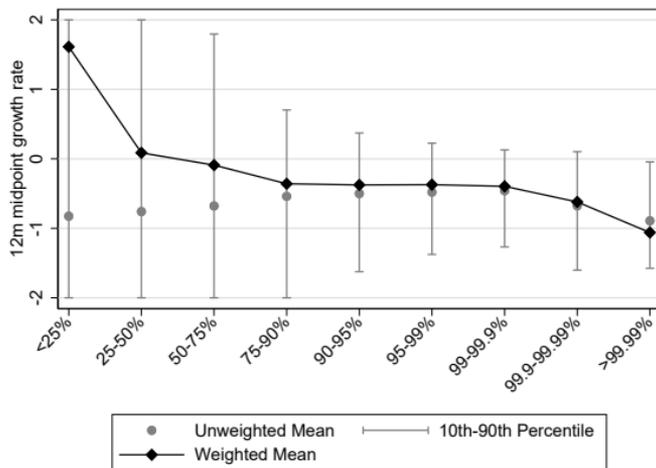


## Midpoint growth rate vs log change

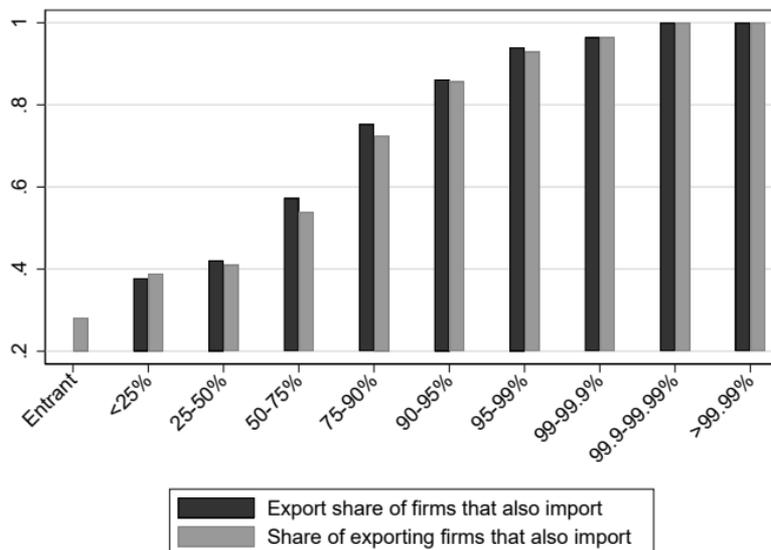


# The exceptional role of large exporters during the collapse

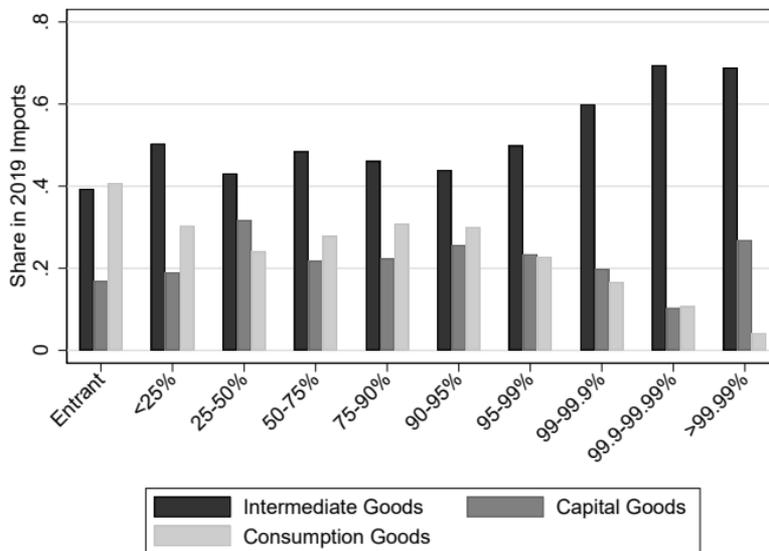
## Distribution of growth within bins



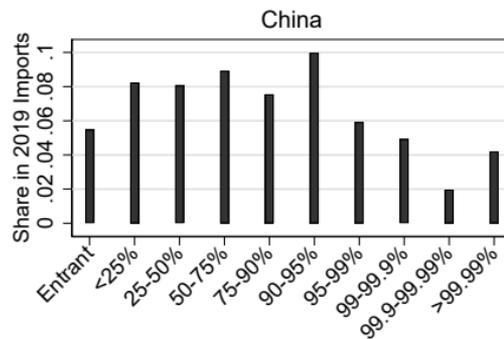
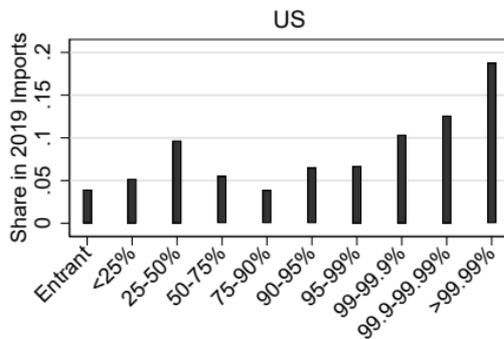
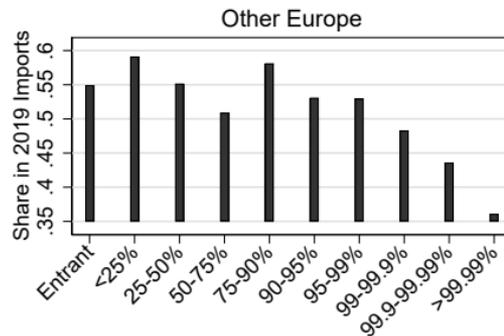
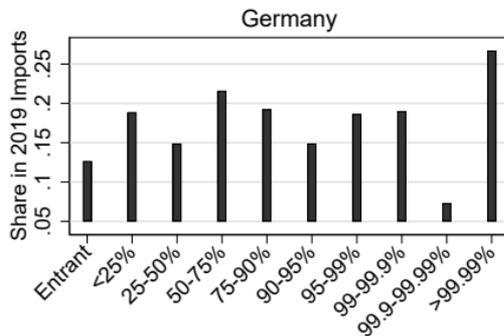
## Larger exporters more likely to import...

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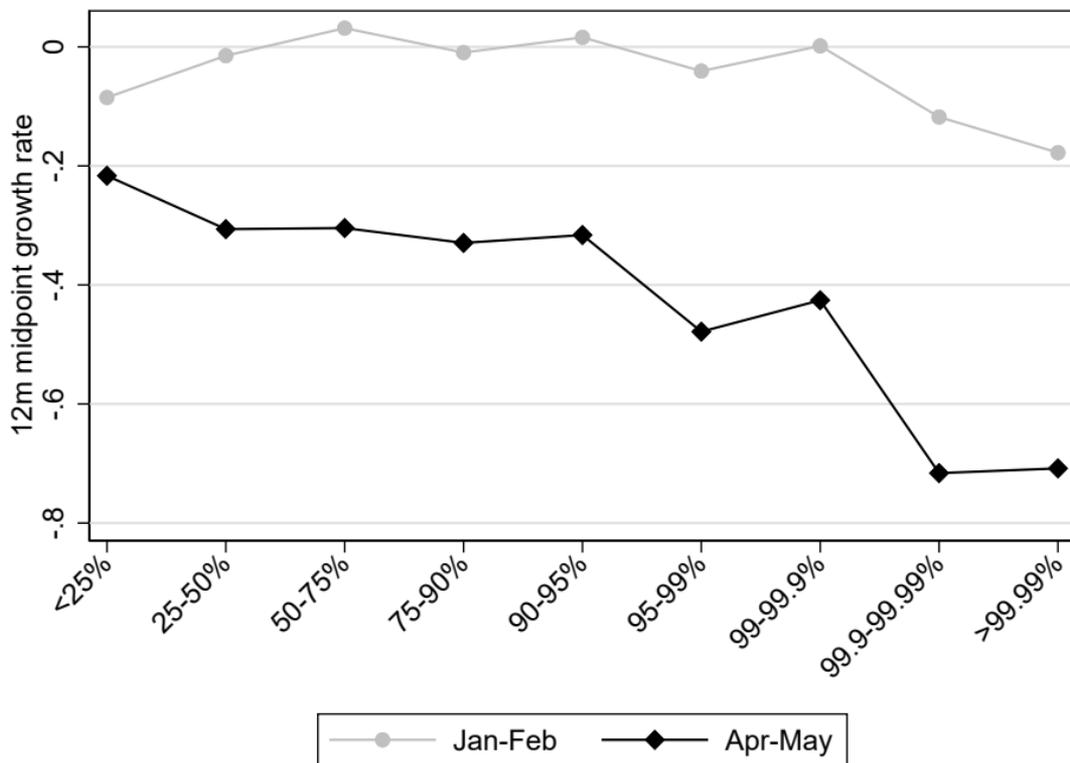
...and especially more likely to import intermediate goods



## ... sourcing concentrated in the US and Europe

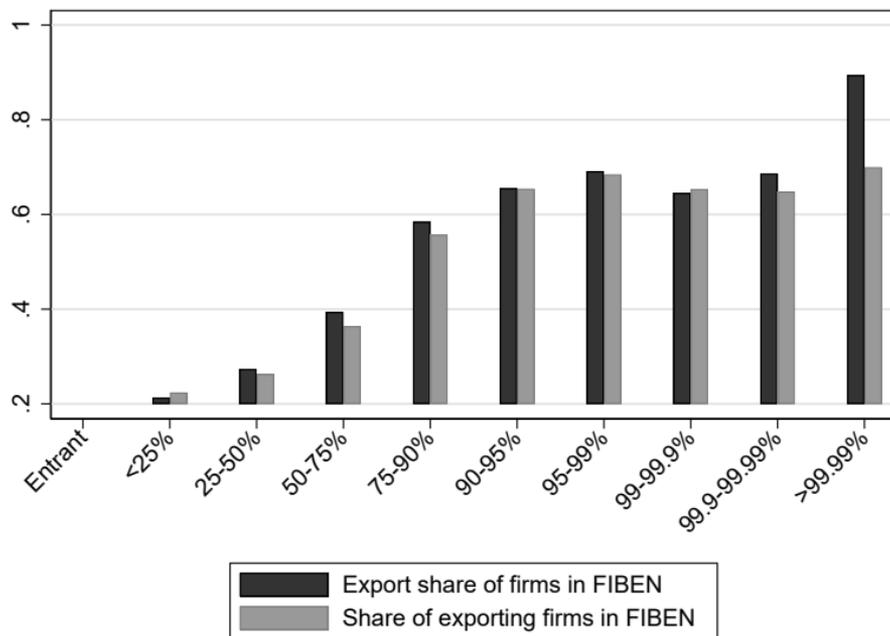


## Exporters also reduce their imports

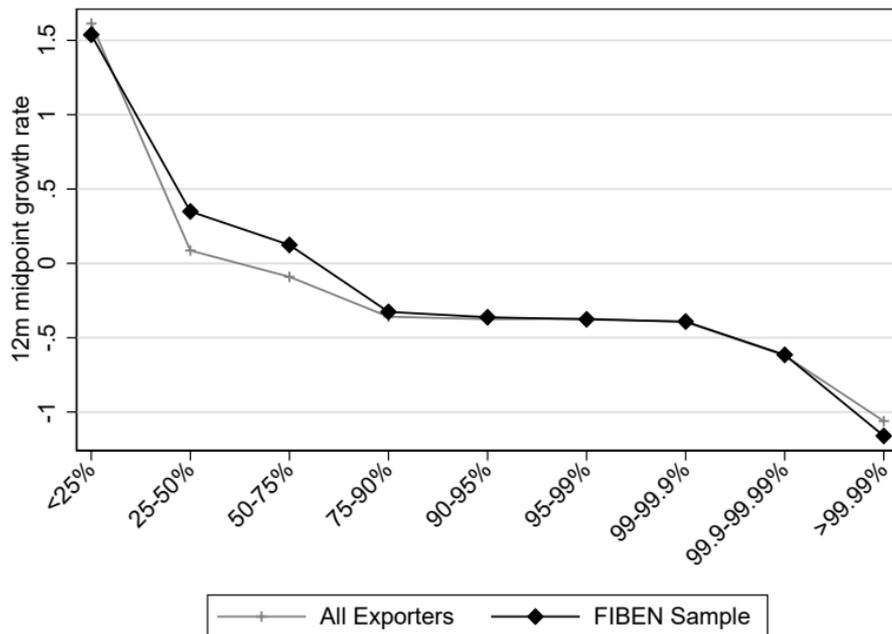


## FIBEN subsample is representative

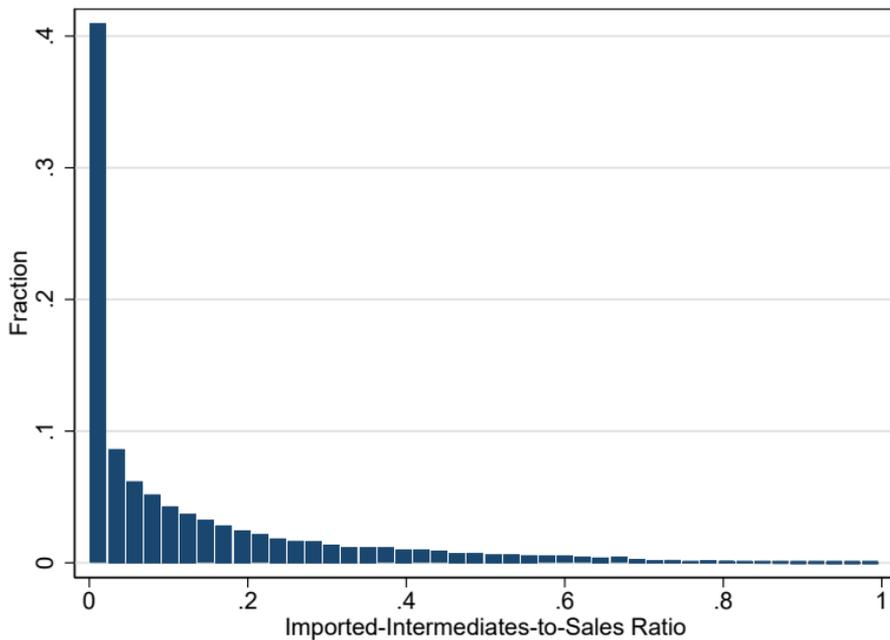
- 37% of 2019 exporters have data in FIBEN
- Mostly larger firms → 71% of 2019 export value



## Results of size estimations very similar in FIBEN subsample

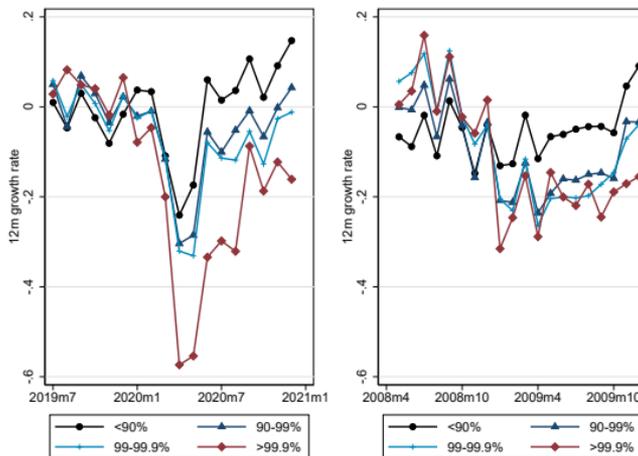


## Highly skewed distribution of IIS ratio



## The role of top exporters during Covid and GFC

- Top exporters decline by more during both Covid and GFC (but differences more pronounced during Covid)



# Data on lockdown stringency across countries and months

## Stringency

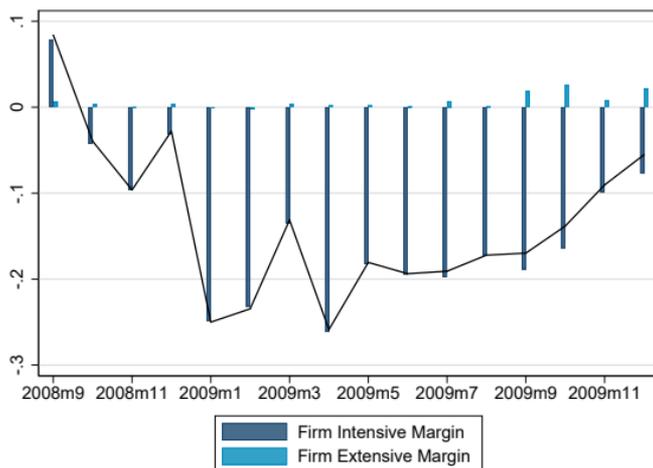
- Main indicator "Stringency index", composite indicator of:
  - School closures; workplace closures
  - Cancel public events; close public transport
  - Public information campaigns
  - Stay at home; restrictions on gatherings
  - Restrictions on internal movement; international travel controls

## Coding of stringency

- Example of workplace closures
- 0 - No measures ; 1 - recommend closing (or work from home)
- 2 - require closing (or work from home) for some sectors or categories of workers
- 3 - require closing (or work from home) all but essential workplaces (eg grocery stores, doctors)

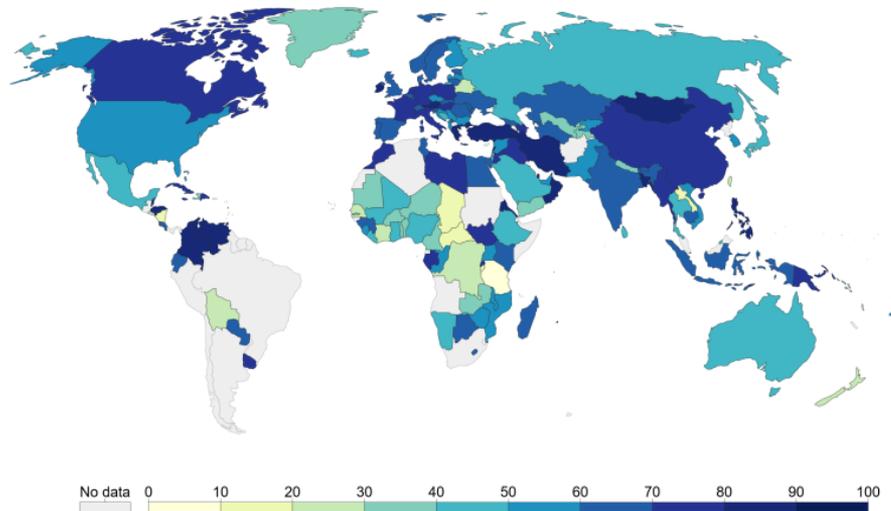
## Firm Intensive and Extensive Margins during GFC

- GFC: Export collapse in France started in Jan 2009
- Firm intensive margin explains almost the entire fall.



## COVID-19: Stringency Index, Apr 21, 2021

This is a composite measure based on nine response indicators including school closures, workplace closures, and travel bans, rescaled to a value from 0 to 100 (100 = strictest). If policies vary at the subnational level, the index is shown as the response level of the strictest sub-region.



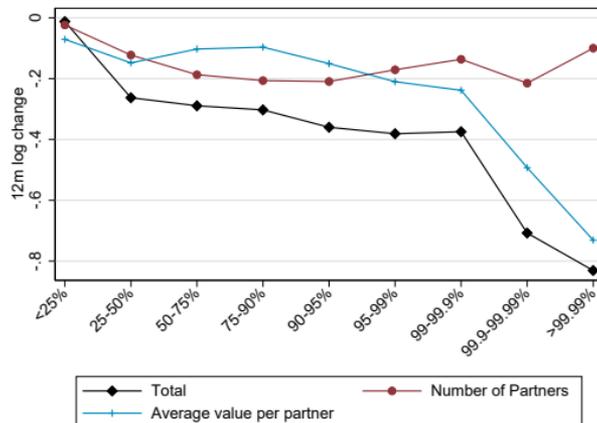
Source: Hale, Angrist, Goldszmidt, Kira, Petherick, Phillips, Webster, Cameron-Blake, Hallas, Majumdar, and Tatlow (2021). "A global panel database of pandemic policies (Oxford COVID-19 Government Response Tracker)." *Nature Human Behaviour*. – Last updated 21 April, 17:00 (London time) [OurWorldInData.org/coronavirus](https://OurWorldInData.org/coronavirus) • CC BY

**Source:** Oxford COVID-19 Government Response Tracker & Our world in data.

## Number of Partners vs Avg Value per partner

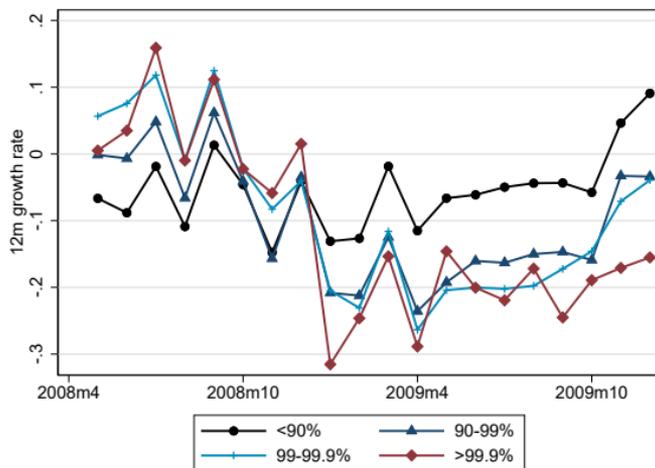
### Number of Partners vs Avg Value per partner

- Not driven by a breakup of connections with more clients.
- Instead, average exports per client fall by more for top exporters
- Focus on intra-EU exports, and intensive margin of transactions. Then decompose  $\Delta \log(X_{fsj}) = \Delta \log(N_{fsj}) + \Delta \log(\bar{X}_{fsj})$



## Top exporters overreact in crisis times

- Very similar growth rates prior to Covid
- Similar role of top exporters during the GFC



Note: Balanced Panel of exporters for each size bin. Exporters grouped into size bins based on their total exports in 2019.

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