

## The Origin of Risk

---

Alexandr Kopytov  
University of Rochester

Mathieu Taschereau-Dumouchel  
Cornell University

Zebang Xu  
Cornell University

## Where does economic risk come from?

Economists commonly assume that risk is exogenous

## Where does economic risk come from?

Economists commonly assume that risk is exogenous

But agents often have control over the risks they face

## Where does economic risk come from?

Economists commonly assume that risk is exogenous

But agents often have control over the risks they face

- Grow crops by the shore creates flood risk
- Grow crops inland creates drought risk

## Where does economic risk come from?

Economists commonly assume that risk is exogenous

But agents often have control over the risks they face

- Grow crops by the shore creates flood risk
- Grow crops inland creates drought risk

Tons of decisions affect the risk profile of a firm

- Hiring decisions, R&D projects, plant locations, investment choices, etc.

## Where does economic risk come from?

Economists commonly assume that risk is exogenous

But agents often have control over the risks they face

- Grow crops by the shore creates **flood risk**
- Grow crops inland creates **drought risk**

Tons of **decisions** affect the **risk profile of a firm**

- Hiring decisions, R&D projects, plant locations, investment choices, etc.

When aggregated, these individual decisions matter for **aggregate risk**

- If everybody grows crops by the shore, a flood can lead to mass starvation

## Where does economic risk come from?

Economists commonly assume that risk is exogenous

But agents often have control over the risks they face

- Grow crops by the shore creates **flood risk**
- Grow crops inland creates **drought risk**

Tons of **decisions** affect the **risk profile of a firm**

- Hiring decisions, R&D projects, plant locations, investment choices, etc.

When aggregated, these individual decisions matter for **aggregate risk**

- If everybody grows crops by the shore, a flood can lead to mass starvation

What drives individual risk-taking decisions and how do they affect aggregate risk?

## Approach and results

We construct a model in which risk is **endogenous** at both the **micro** and the **macro** levels



## Approach and results

We construct a model in which risk is **endogenous** at both the **micro** and the **macro** levels

- Instead of modeling each decision that matters for risk we take a holistic approach

## Approach and results

We construct a model in which risk is **endogenous** at both the **micro** and the **macro** levels

- Instead of modeling each decision that matters for risk we take a holistic approach
- Each firm can **adjust its TFP process** (mean, variance and correlation with other firms' TFP)

## Approach and results

We construct a model in which risk is **endogenous** at both the **micro** and the **macro** levels

- Instead of modeling each decision that matters for risk we take a holistic approach
- Each firm can **adjust its TFP process** (mean, variance and correlation with other firms' TFP)
  - Choosing **correlated TFPs** leads to **aggregate risk**
  - Adjusting risk is costly

## Approach and results

We construct a model in which risk is **endogenous** at both the **micro** and the **macro** levels

- Instead of modeling each decision that matters for risk we take a holistic approach
- Each firm can **adjust its TFP process** (mean, variance and correlation with other firms' TFP)
  - Choosing **correlated TFPs** leads to **aggregate risk**
  - Adjusting risk is costly

The theory predicts how firm characteristics affect their risk profile

- Since TFP multiplies the input bundle, **larger firms manage risk more aggressively**

## Approach and results

We construct a model in which risk is **endogenous** at both the **micro** and the **macro** levels

- Instead of modeling each decision that matters for risk we take a holistic approach
- Each firm can **adjust its TFP process** (mean, variance and correlation with other firms' TFP)
  - Choosing **correlated TFPs** leads to **aggregate risk**
  - Adjusting risk is costly

The theory predicts how firm characteristics affect their risk profile

- Since TFP multiplies the input bundle, **larger firms manage risk more aggressively**
- Larger firms and those with low markups are **less volatile** and **covary less with GDP**
- We find support for these predictions in detailed firm-level Spanish data

## Approach and results

We construct a model in which risk is **endogenous** at both the **micro** and the **macro** levels

- Instead of modeling each decision that matters for risk we take a holistic approach
- Each firm can **adjust its TFP process** (mean, variance and correlation with other firms' TFP)
  - Choosing **correlated TFPs** leads to **aggregate risk**
  - Adjusting risk is costly

The theory predicts how firm characteristics affect their risk profile

- Since TFP multiplies the input bundle, **larger firms manage risk more aggressively**
- Larger firms and those with low markups are **less volatile** and **covary less with GDP**
- We find support for these predictions in detailed firm-level Spanish data

The theory also has predictions for the aggregate economy

- Because of endogenous risk, **distortions** can make **GDP more volatile**

## Approach and results

We construct a model in which risk is **endogenous** at both the **micro** and the **macro** levels

- Instead of modeling each decision that matters for risk we take a holistic approach
- Each firm can **adjust its TFP process** (mean, variance and correlation with other firms' TFP)
  - Choosing **correlated TFPs** leads to **aggregate risk**
  - Adjusting risk is costly

The theory predicts how firm characteristics affect their risk profile

- Since TFP multiplies the input bundle, **larger firms manage risk more aggressively**
- Larger firms and those with low markups are **less volatile** and **covary less with GDP**
- We find support for these predictions in detailed firm-level Spanish data

The theory also has predictions for the aggregate economy

- Because of endogenous risk, **distortions** can make **GDP more volatile**

We **calibrate** the model to the Spanish economy

- Removing distortions lead to a large decline in aggregate volatility

Most of macroeconomics takes risk as **exogenous** (at the micro and/or macro level)

- In **models with individual firms**, firm-level risk is generally exogenous but macro risk can be endogenous
  - Khan and Thomas (2008), Clementi and Palazzo (2016), Bloom et al. (2018), and many others
- In **endogenous growth models**, firms influence the growth rate of TFP but not its variance
  - Romer (1990), Grossman and Helpman (1991), Aghion and Howitt (1992), Jones (1995)
- **Corporate finance** literature where managers influence how risky a project is
  - Jensen and Meckling (1976), Ross (1977)
- **Wedges in production network economies**
  - Jones (2011), Baqaee and Farhi (2019), Liu (2019) and Bigio and La'O (2020)
- **Technique choice in production networks**
  - Oberfield (2018), Acemoglu and Azar (2020), Kopytov et al. (2024)



## A model of endogenous risk

---

Static model with two types of agents

1. A **representative household** owns the firms, supplies labor and risk management resources
2.  $N$  **firms** produce differentiated goods using labor and intermediate inputs
  - Firm  $i$  has constant returns to scale **Cobb-Douglas production function**

$$F(\delta_i, L_i, X_i) = e^{a_i(\varepsilon, \delta_i)} \zeta_i L_i^{1 - \sum_{j=1}^N \alpha_{ij}} \prod_{j=1}^N X_{ij}^{\alpha_{ij}}$$

## Endogenous risk choice

Firms choose mean, variance and correlation structure of their TFP  $a_i(\epsilon, \delta_i)$

## Endogenous risk choice

Firms choose mean, variance and correlation structure of their TFP  $a_i(\boldsymbol{\varepsilon}, \delta_i)$

There are **underlying sources of risk**  $\boldsymbol{\varepsilon} = (\varepsilon_1, \dots, \varepsilon_M)$  with  $\boldsymbol{\varepsilon} \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$

- Examples: a river floods, discovery of a new drug, war between two countries, epidemic, etc.
- We don't take a stance on what  $\boldsymbol{\varepsilon}$  is. Focus on **quantity of risk** and **correlation** structure.

## Endogenous risk choice

Firms choose mean, variance and correlation structure of their TFP  $a_i(\boldsymbol{\varepsilon}, \boldsymbol{\delta}_i)$

There are **underlying sources of risk**  $\boldsymbol{\varepsilon} = (\varepsilon_1, \dots, \varepsilon_M)$  with  $\boldsymbol{\varepsilon} \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$

- Examples: a river floods, discovery of a new drug, war between two countries, epidemic, etc.
- We don't take a stance on what  $\boldsymbol{\varepsilon}$  is. Focus on **quantity of risk** and **correlation** structure.

Firms pick exposure  $\boldsymbol{\delta}_i$  to these risk factors

$$a_i(\boldsymbol{\varepsilon}, \boldsymbol{\delta}_i) = \boldsymbol{\delta}_i^\top \boldsymbol{\varepsilon}$$

## Endogenous risk choice

Firms choose mean, variance and correlation structure of their TFP  $a_i(\boldsymbol{\varepsilon}, \boldsymbol{\delta}_i)$

There are **underlying sources of risk**  $\boldsymbol{\varepsilon} = (\varepsilon_1, \dots, \varepsilon_M)$  with  $\boldsymbol{\varepsilon} \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$

- Examples: a river floods, discovery of a new drug, war between two countries, epidemic, etc.
- We don't take a stance on what  $\boldsymbol{\varepsilon}$  is. Focus on **quantity of risk** and **correlation** structure.

Firms pick exposure  $\boldsymbol{\delta}_i$  to these risk factors

$$a_i(\boldsymbol{\varepsilon}, \boldsymbol{\delta}_i) = \boldsymbol{\delta}_i^\top \boldsymbol{\varepsilon}$$

Managing risk (picking  $\boldsymbol{\delta}_i$ ) requires **risk management resources**  $R_i$  supplied by the household

$$R_i = \kappa_i(\boldsymbol{\delta}_i) = \frac{1}{2} (\boldsymbol{\delta}_i - \boldsymbol{\delta}_i^\circ)^\top H_i (\boldsymbol{\delta}_i - \boldsymbol{\delta}_i^\circ)$$

where  $\boldsymbol{\delta}_i^\circ$  is the *natural* risk exposure ( $R_i = 0$ ), and  $H_i$  is a positive definite matrix

## Representative household

Owns the firms, supplies one unit of labor inelastically, supplies risk management resources

## Representative household

Owens the firms, supplies one unit of labor inelastically, supplies risk management resources

Values the consumption bundle (GDP)

$$Y = \prod_{i=1}^N (\beta_i^{-1} C_i)^{\beta_i}$$



## Representative household

Owens the firms, supplies one unit of labor inelastically, supplies risk management resources

Values the consumption bundle (GDP)

$$Y = \prod_{i=1}^N (\beta_i^{-1} C_i)^{\beta_i}$$

Maximizes King, Plosser, Rebelo (1988) preferences

$$\mathcal{U}(Y) \mathcal{V}(R)$$

where  $\mathcal{U}$  is CRRA with risk aversion  $\rho \geq 1$ , and disutility of risk management  $\mathcal{V}(R)$  is

► Details

$$\mathcal{V}(R) = \exp(-\eta(1 - \rho)R)$$

## Representative household

Owens the firms, supplies one unit of labor inelastically, supplies risk management resources

Values the consumption bundle (GDP)

$$Y = \prod_{i=1}^N (\beta_i^{-1} C_i)^{\beta_i}$$

Maximizes King, Plosser, Rebelo (1988) preferences

$$\mathcal{U}(Y) \mathcal{V}(R)$$

where  $\mathcal{U}$  is CRRA with risk aversion  $\rho \geq 1$ , and disutility of risk management  $\mathcal{V}(R)$  is

► Details

$$\mathcal{V}(R) = \exp(-\eta(1-\rho)R)$$

Budget constraint in each state of the world (set  $W_L = 1$  from now on)

$$\sum_{i=1}^N P_i C_i \leq W_L + W_R R + \Pi$$

## Timing

1. Before  $\varepsilon$  is realized: Firms choose risk exposure  $\delta$
2. After  $\varepsilon$  is realized: All other quantities are chosen

## Timing

1. Before  $\varepsilon$  is realized: Firms choose risk exposure  $\delta$
2. After  $\varepsilon$  is realized: All other quantities are chosen

Firms set prices  $P$  at a **constant wedge**  $\tau_i$  over marginal cost  $K_i$

$$P_i = (1 + \tau_i) K_i(\delta_i, P)$$

- Example: markups, taxes, or other distortions

## Timing

1. Before  $\varepsilon$  is realized: Firms choose risk exposure  $\delta$
2. After  $\varepsilon$  is realized: All other quantities are chosen

Firms set prices  $P$  at a **constant wedge**  $\tau_i$  over marginal cost  $K_i$

$$P_i = (1 + \tau_i) K_i(\delta_i, P)$$

- Example: markups, taxes, or other distortions

Cobb-Douglas **unit cost** is

$$K_i(\delta_i, P) = \frac{1}{e^{a_i(\varepsilon, \delta_i)}} \prod_{j=1}^N P_j^{\alpha_{ij}}$$

Firm choose their risk exposure to maximize **expected discounted profits**

$$\delta_i^* \in \arg \max_{\delta_i \in \mathcal{A}_i} E[\Lambda [P_i Q_i - K_i(\delta_i, P) Q_i - \kappa_i(\delta_i) W_R]]$$

where  $Q_i$  is *equilibrium* demand and  $\Lambda$  is the **stochastic discount factor** of the household.

Firm choose their risk exposure to maximize **expected discounted profits**

$$\delta_i^* \in \arg \max_{\delta_i \in \mathcal{A}_i} E[\Lambda [P_i Q_i - K_i(\delta_i, P) Q_i - \kappa_i(\delta_i) W_R]]$$

where  $Q_i$  is *equilibrium* demand and  $\Lambda$  is the **stochastic discount factor** of the household.

Firms prefer risk exposures with

1. low risk management expenses  $\kappa_i(\delta)$
2. high expected TFP (low expected unit costs  $K_i$ )
3. low covariance with GDP

## Equilibrium definition

An *equilibrium* is a risk choice for every firm  $\delta^*$  and a stochastic tuple  $(P^*, W_R^*, C^*, L^*, R^*, X^*, Q^*)$  such that

1. (Optimal technique choice) For each  $i$ , factor demand  $L_i^*$ ,  $X_i^*$  and  $R_i^*$ , and the risk exposure decision  $\delta_i^*$  solves the firm's problem.
2. (Consumer maximization) The consumption vector  $C^*$  and the supply of risk managers  $R^*$  solve the household problem.
3. (Unit cost pricing) For each  $i$ ,  $P_i = (1 + \tau_i) K_i(\delta_i, P)$ .
4. (Market clearing) For each  $i$ ,

$$C_i^* + \sum_{j=1}^N X_{ji}^* = Q_i^* = F_i(\alpha_i^*, L_i^*, X_i^*), \quad \sum_{i=1}^N L_i^* = 1, \quad \text{and} \quad \sum_{i=1}^N \kappa_i(\delta_i^*) = R^*.$$



## Two measures of supplier importance

Cost-based Domar weight:

$$\tilde{\omega}^T = \beta^T (I - \alpha)^{-1}$$

## Two measures of supplier importance

Cost-based Domar weight:

$$\tilde{\omega}^T = \beta^T (I - \alpha)^{-1}$$

- Depends on demand from household ( $\beta$ ) and other firms ( $\mathcal{L} = (I - \alpha)^{-1} = I + \alpha + \alpha^2 + \dots$ )
- Captures firm's importance as a supplier (share of production costs)

## Two measures of supplier importance

Cost-based Domar weight:

$$\tilde{\omega}^T = \beta^T (I - \alpha)^{-1}$$

- Depends on demand from household ( $\beta$ ) and other firms ( $\mathcal{L} = (I - \alpha)^{-1} = I + \alpha + \alpha^2 + \dots$ )
- Captures firm's importance as a supplier (share of production costs)

Revenue-based Domar weight:

$$\omega^T = \beta^T \mathcal{L} = \beta^T (I - [\text{diag}(1 + \tau)]^{-1} \alpha)^{-1}$$

- Also captures importance as a supplier (share of revenues)
- Declines with **wedges**  $\tau$

Define **aggregate risk exposure**  $\Delta$  as

$$\Delta := \delta^\top \tilde{\omega}$$

- Firms with high cost-based Domar weights contribute more to aggregate risk exposure

Define **aggregate risk exposure**  $\Delta$  as

$$\Delta := \delta^\top \tilde{\omega}$$

- Firms with high cost-based Domar weights contribute more to aggregate risk exposure

## Lemma

$$\log Y = y = \Delta^\top \varepsilon - \tilde{\omega}^\top \log(1 + \tau) - \log(\text{Labor share}(\omega, \tau))$$

- Without distortions ( $\tau = 0$ ) we have Hulten's theorem:  $y = \Delta^\top \varepsilon = \omega^\top a(\varepsilon, \delta)$

$$\text{Aggregate risk: } V[y] = \Delta^T \Sigma \Delta$$

$$\text{Aggregate risk: } V[y] = \Delta^T \Sigma \Delta$$

## Impact of $\Sigma$

- A marginal increase in  $\Sigma_{mm}$  raises  $V[y]$  by  $\Delta_m^2$ 
  - Both  $\Delta_m \gg 0$  and  $\Delta_m \ll 0$  are bad for  $V[y]$
- If the economy is positively exposed to  $m$  and  $n$ , increasing  $\Sigma_{mn}$  raises  $V[y]$ .
- If  $\Delta_m > 0$  and  $\Delta_n < 0$ , the shocks offset each other. Higher  $\Sigma_{mn}$  reduces  $V[y]$ .

$$\text{Aggregate risk: } V[y] = \Delta^T \Sigma \Delta$$

## Impact of $\Sigma$

- A marginal increase in  $\Sigma_{mm}$  raises  $V[y]$  by  $\Delta_m^2$ 
  - Both  $\Delta_m \gg 0$  and  $\Delta_m \ll 0$  are bad for  $V[y]$
- If the economy is positively exposed to  $m$  and  $n$ , increasing  $\Sigma_{mn}$  raises  $V[y]$ .
- If  $\Delta_m > 0$  and  $\Delta_n < 0$ , the shocks offset each other. Higher  $\Sigma_{mn}$  reduces  $V[y]$ .

## Impact of $\Delta$

$$\frac{dV[y]}{d\Delta_m} = 2 \text{Cov}[y, \varepsilon_m] = 2 \sum_n \Delta_n \text{Cov}[\varepsilon_n, \varepsilon_m]$$

- Extra exposure to  $\varepsilon_m$  increases volatility if  $\varepsilon_m$  is positively correlated with GDP



## Lemma

The equilibrium risk exposure decision  $\delta_i$  solves

$$\mathcal{E} \underbrace{K_i Q_i}_{\text{cost of goods sold}} = \underbrace{W_R \nabla \kappa_i(\delta_i)}_{\text{marginal cost of exposure}},$$

where  $\mathcal{E}$  is the value of exposure, given by  $\mathcal{E} := E[\varepsilon] + \text{Cov}[\lambda, \varepsilon]$ .

## Lemma

The equilibrium risk exposure decision  $\delta_i$  solves

$$\mathcal{E} \underbrace{K_i Q_i}_{\text{cost of goods sold}} = \underbrace{W_R \nabla \kappa_i(\delta_i)}_{\text{marginal cost of exposure}},$$

where  $\mathcal{E}$  is the value of exposure, given by  $\mathcal{E} := E[\varepsilon] + \text{Cov}[\lambda, \varepsilon]$ .

Benefit of increasing  $\delta_i$  **grows with the size of the firm** since TFP multiplies the input bundle

- Since  $K_i Q_i = \omega_i \Gamma_L^{-1} / (1 + \tau_i)$  firms with **high  $\omega_i$**  and **low  $\tau_i$**  manage risk **more aggressively**

### Lemma

The equilibrium risk exposure decision  $\delta_i$  solves

$$\mathcal{E} \underbrace{K_i Q_i}_{\text{cost of goods sold}} = \underbrace{W_R \nabla \kappa_i(\delta_i)}_{\text{marginal cost of exposure}},$$

where  $\mathcal{E}$  is the value of exposure, given by  $\mathcal{E} := E[\varepsilon] + \text{Cov}[\lambda, \varepsilon]$ .

Benefit of increasing  $\delta_i$  **grows with the size of the firm** since TFP multiplies the input bundle

- Since  $K_i Q_i = \omega_i \Gamma_L^{-1} / (1 + \tau_i)$  firms with **high  $\omega_i$**  and **low  $\tau_i$**  manage risk **more aggressively**

Equation for  $\mathcal{E}$  implies that firms **prefer risk factors** with

- high expected value  $\mu = E[\varepsilon]$  and negative covariance with GDP ( $\text{Cov}[\lambda, \varepsilon] > 0$ )
- Risk factor is “good” if  $\mathcal{E} > 0$  and “bad” if  $\mathcal{E} < 0$

## Existence, uniqueness and efficiency

---

## Planner's problem

Define  $\bar{\kappa}_{SP}(\Delta)$  as the **smallest risk management utility cost** needed to achieve  $\Delta$ .

$$\bar{\kappa}_{SP}(\Delta) := \min_{\delta} -\log V\left(\sum_{i=1}^N \kappa_i(\delta_i)\right), \quad \text{subject to } \Delta = \delta^\top \tilde{\omega}$$

## Planner's problem

Define  $\bar{\kappa}_{SP}(\Delta)$  as the **smallest risk management utility cost** needed to achieve  $\Delta$ .

$$\bar{\kappa}_{SP}(\Delta) := \min_{\delta} -\log V\left(\sum_{i=1}^N \kappa_i(\delta_i)\right), \quad \text{subject to } \Delta = \delta^\top \tilde{\omega}$$

### Planner's problem

$$\mathcal{W}_{SP} := \max_{\Delta} \underbrace{\Delta^\top \mu}_{\mathbb{E}[y_{SP}]} - \frac{1}{2} (\rho - 1) \underbrace{\Delta^\top \Sigma \Delta}_{\mathbb{V}[y_{SP}]} - \bar{\kappa}_{SP}(\Delta)$$

The planner prefers aggregate risk exposure vectors  $\Delta$  with

- high expected GDP  $\mathbb{E}[y_{SP}]$
- low GDP volatility  $\mathbb{V}[y_{SP}]$
- low risk management cost  $\bar{\kappa}_{SP}$

## Equilibrium characterization through fictitious planner

Define  $\bar{\kappa}(\Delta)$  as the **perceived** smallest risk management utility cost needed to achieve  $\Delta$ .

$$\bar{\kappa}(\Delta) := \min_{\delta} -\log V\left(\sum_{i=1}^N g_i \kappa_i(\delta_i)\right), \quad \text{subject to } \Delta = \delta^\top \tilde{\omega}$$

where  $g_i := \frac{\tilde{\omega}_i(1+\tau_i)}{\omega_i} \geq 1$  is the **efficiency gap** of firm  $i$ .

## Equilibrium characterization through fictitious planner

Define  $\bar{\kappa}(\Delta)$  as the **perceived** smallest risk management utility cost needed to achieve  $\Delta$ .

$$\bar{\kappa}(\Delta) := \min_{\delta} -\log V\left(\sum_{i=1}^N g_i \kappa_i(\delta_i)\right), \quad \text{subject to } \Delta = \delta^\top \tilde{\omega}$$

where  $g_i := \frac{\tilde{\omega}_i(1+\tau_i)}{\omega_i} \geq 1$  is the **efficiency gap** of firm  $i$ .

Proposition (fictitious planner's problem)

There exists a **unique equilibrium**, and it solves

$$\mathcal{W}_{dist} := \max_{\Delta} \underbrace{\Delta^\top \mu - \tilde{\omega}^\top \log(1+\tau) - \log \Gamma_L}_{\mathbb{E}[y]} - \frac{1}{2}(\rho-1) \underbrace{\Delta^\top \Sigma \Delta}_{\mathbb{V}[y]} - \bar{\kappa}(\Delta).$$



## Equilibrium characterization through fictitious planner

Define  $\bar{\kappa}(\Delta)$  as the **perceived** smallest risk management utility cost needed to achieve  $\Delta$ .

$$\bar{\kappa}(\Delta) := \min_{\delta} -\log V \left( \sum_{i=1}^N g_i \kappa_i(\delta_i) \right), \quad \text{subject to } \Delta = \delta^\top \tilde{\omega}$$

where  $g_i := \frac{\tilde{\omega}_i(1+\tau_i)}{\omega_i} \geq 1$  is the **efficiency gap** of firm  $i$ .

### Proposition (fictitious planner's problem)

There exists a **unique equilibrium**, and it solves

$$\mathcal{W}_{dist} := \max_{\Delta} \underbrace{\Delta^\top \mu - \tilde{\omega}^\top \log(1+\tau) - \log \Gamma_L}_{\mathbb{E}[y]} - \frac{1}{2} (\rho - 1) \underbrace{\Delta^\top \Sigma \Delta}_{\mathbb{V}[y]} - \bar{\kappa}(\Delta).$$

The equilibrium solves a **distorted planning problem**

- Still seeks to maximize  $\mathbb{E}[y]$  and minimize  $\mathbb{V}[y]$
- But **distorted perception** of the cost of managing risk ( $\bar{\kappa}$  instead of  $\bar{\kappa}_{SP}$ )

## Determinants of equilibrium risk

---

First-order of fictitious planning problem

$$\underbrace{\mathcal{E}(\Delta)}_{\text{marginal benefit of } \Delta} = \underbrace{\nabla \bar{\kappa}(\Delta)}_{\text{marginal cost of } \Delta}$$

First-order of **fictitious planning problem**

$$\underbrace{\mathcal{E}(\Delta)}_{\text{marginal benefit of } \Delta} = \underbrace{\nabla \bar{\kappa}(\Delta)}_{\text{marginal cost of } \Delta}$$

## Proposition

Let  $\gamma$  be either  $\mu_m$  or  $\Sigma_{mn}$ . Then

$$\frac{d\Delta}{d\gamma} = \mathcal{H}^{-1} \frac{\partial \mathcal{E}}{\partial \gamma},$$

where  $\mathcal{H}^{-1} := (\nabla^2 \bar{\kappa} + (\rho - 1) \Sigma)^{-1}$  is an  $M \times M$  positive definite matrix.

- The vector  $\partial \mathcal{E} / \partial \gamma$  captures the **direct impact** of  $\gamma$  on the attractiveness of risk factors ▶  $\partial \mathcal{E} / \partial \gamma$
- The matrix  $\mathcal{H}^{-1}$  **propagates** that impact to exposure vector  $\Delta$

### Corollary

1. An increase in  $\mu_m$  raises  $\Delta_m$
2. An increase in  $\Sigma_{mm}$  reduces  $\Delta_m$  if  $\Delta_m > 0$  and increases  $\Delta_m$  if  $\Delta_m < 0$

• A marginal increase in  $\Sigma_{mm}$  raises  $V[y]$  by  $\Delta_m^2 \rightarrow$  When  $\Sigma_{mm}$  increases we want to reduce  $\Delta_m^2$

## Corollary

1. An increase in  $\mu_m$  raises  $\Delta_m$
2. An increase in  $\Sigma_{mm}$  reduces  $\Delta_m$  if  $\Delta_m > 0$  and increases  $\Delta_m$  if  $\Delta_m < 0$

- A marginal increase in  $\Sigma_{mm}$  raises  $V[y]$  by  $\Delta_m^2 \rightarrow$  When  $\Sigma_{mm}$  increases we want to reduce  $\Delta_m^2$

What is the impact of  $\mu_m$  or  $\Sigma_{mm}$  on  $\Delta_n$  with  $m \neq n$ ? Off-diagonal terms of  $\mathcal{H}^{-1}$  are important.

- If  $[\mathcal{H}^{-1}]_{mn} > 0$ ,  $m$  and  $n$  are *global complements*  $\rightarrow$  an increase in  $\mathcal{E}_m$  increases in  $\Delta_n$
- If  $[\mathcal{H}^{-1}]_{mn} < 0$ ,  $m$  and  $n$  are *global substitutes*  $\rightarrow$  an increase in  $\mathcal{E}_m$  decreases  $\Delta_n$

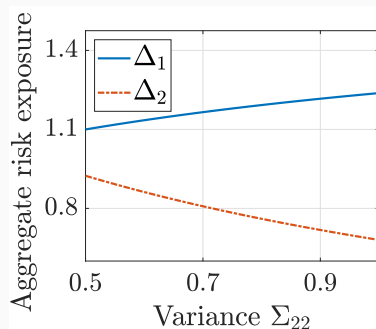
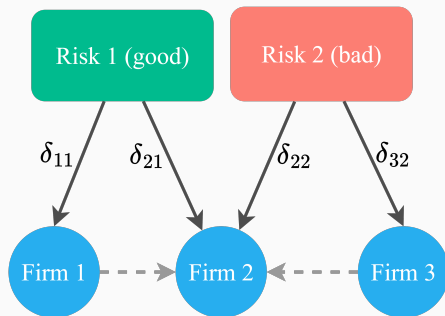
## Example of substitution patterns

There are **two regions** both with their specific shocks

- Region 1: more productive in expectation (Risk 1 – good risk)
- Region 2: bigger shocks (Risk 2 – bad risk)

Firm 2 must decide **where to locate plants**

- Challenging to manage plants in different locations → risks are substitutes



**Definition.** An economy is diagonal if  $\Sigma$  and  $H_i$  are diagonal for every  $i$



**Definition.** An economy is diagonal if  $\Sigma$  and  $H_i$  are diagonal for every  $i$

## Corollary

In a diagonal economy, a higher wedge  $\tau_i$

1. increases  $\Delta_m$  for all  $m$  such that  $\mathcal{E}_m < 0$  (bad risks)
2. reduces  $\Delta_m$  for all  $m$  such that  $\mathcal{E}_m > 0$  (good risks)

- Higher wedges make firms shrink  $\rightarrow$  manage risk less aggressively

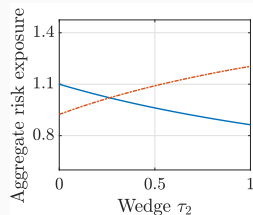
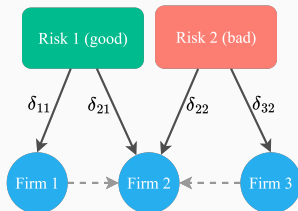
**Definition.** An economy is diagonal if  $\Sigma$  and  $H_i$  are diagonal for every  $i$

### Corollary

In a diagonal economy, a higher wedge  $\tau_i$

1. increases  $\Delta_m$  for all  $m$  such that  $\mathcal{E}_m < 0$  (bad risks)
2. reduces  $\Delta_m$  for all  $m$  such that  $\mathcal{E}_m > 0$  (good risks)

- Higher wedges make firms shrink  $\rightarrow$  manage risk less aggressively



(Blue: good risk; Red: bad risk)

## Implications for GDP and Welfare

---

In general, our **endogenous risk** mechanism is important for understanding the aggregate impact of changing risk.

Use  $\partial$  to denote changes in the economy with **exogenous risk**

### Proposition

In a diagonal economy:

$$\text{sign} \left( \frac{dE[y]}{d\mu_m} - \frac{\partial E[y]}{\partial \mu_m} \right) = \text{sign}(\mu_m) \quad \text{and} \quad \frac{dV[y]}{d\Sigma_{mm}} - \frac{\partial V[y]}{\partial \Sigma_{mm}} < 0.$$

- Increasing  $\mu_m$  raises  $\Delta_m \rightarrow$  additional increase in  $E[y]$  if  $\mu_m > 0$  compared to fixed risk
- Increasing  $\Sigma_{mm}$  decreases  $|\Delta| \rightarrow$  smaller increase in  $V[y]$  than with fixed risk

### Proposition (single risk factor)

$$\text{sign} \left( \frac{d\mathbf{E}[y]}{d\tau_i} - \frac{\partial \mathbf{E}[y]}{\partial \tau_i} \right) = -\text{sign}(\mu\mathcal{E}) \quad \text{and} \quad \text{sign} \left( \frac{d\mathbf{V}[y]}{d\tau_i} - \frac{\partial \mathbf{V}[y]}{\partial \tau_i} \right) = -\text{sign}(\Delta\mathcal{E}).$$

Suppose  $\mathcal{E} < 0$  (bad risk, e.g. business cycle): increasing  $\tau_i$  makes firms more exposed to risk factor

- if  $\mu < 0$  this leads to a decline in  $\mathbf{E}[y]$
- if  $\Delta > 0$  the economy becomes even more exposed and  $\mathbf{V}[y]$  increases
- Inefficient allocation of risk exposure: a new channel that  $\tau_i$  could lead to welfare losses.

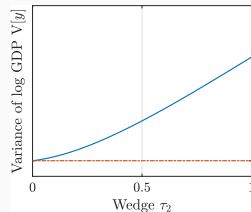
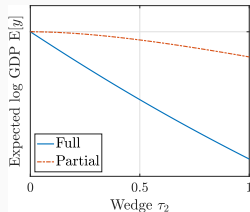
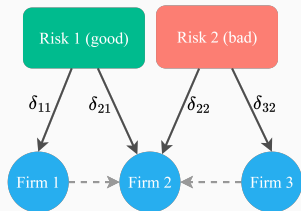
# Distortions can increase aggregate volatility

## Proposition (single risk factor)

$$\text{sign} \left( \frac{dE[y]}{d\tau_i} - \frac{\partial E[y]}{\partial \tau_i} \right) = -\text{sign}(\mu\mathcal{E}) \quad \text{and} \quad \text{sign} \left( \frac{dV[y]}{d\tau_i} - \frac{\partial V[y]}{\partial \tau_i} \right) = -\text{sign}(\Delta\mathcal{E}).$$

Suppose  $\mathcal{E} < 0$  (bad risk, e.g. business cycle): increasing  $\tau_i$  makes firms more exposed to risk factor

- if  $\mu < 0$  this leads to a decline in  $E[y]$
- if  $\Delta > 0$  the economy becomes even more exposed and  $V[y]$  increases
- Inefficient allocation of risk exposure: a new channel that  $\tau_i$  could lead to welfare losses.



## Reduced-form evidence

---

Model: firms with large Domar weights and small markups are less volatile and less corr. with GDP

▶ Details



Model: firms with **large Domar weights** and **small markups** are **less volatile** and **less corr. with GDP**

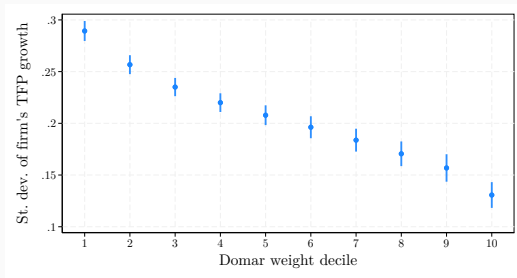
► Details

We test these predictions in the data

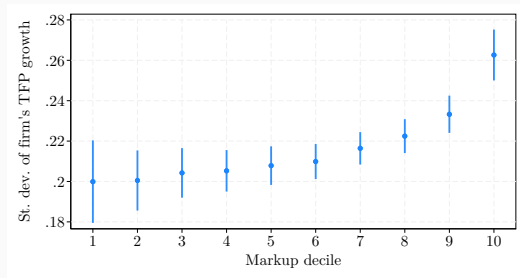
- Use detailed **micro data** from the near-universe of firms in **Spain** between 1995 and 2018 (Orbis) (7,513,081 firm-year observations)
- Compute **markups** using control function approach (De Loecker and Warzynski, 2012)
- Back out TFP growth as a residual

► Details

# TFP growth volatility



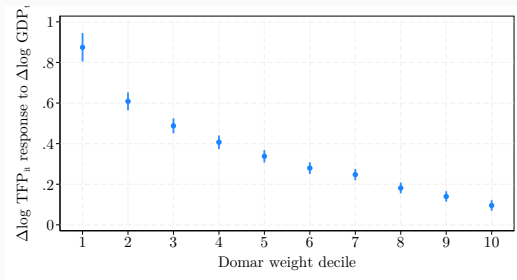
(a) TFP volatility by Domar weight decile



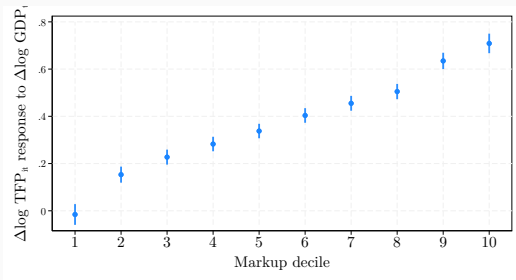
(b) TFP volatility by markup decile

► Details

# Covariance of TFP growth with GDP growth



(c) Sensitivity of firm TFP to GDP by Domar weight decile



(d) Sensitivity of firm TFP to GDP by markup decile

► Details

## Calibration

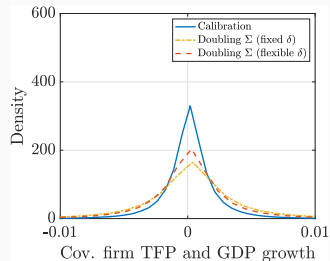
---

- Sector-level I-O linkages + Firm-level Heterogeneity in productivity and markup
- We aim at **replicating** as much of the firm-level Spanish data as possible
- Our calibrated model has 62 sectors and 492,917 individual firms
- We invert parts of the model to **exactly match some moments**
  1. Sectoral consumption shares and input/output cost shares
  2. Firm shares in sectoral sales
  3. Variance of firm TFP growth
  4. Covariance of firm TFP growth and GDP growth
  5. Variance of GDP growth

► Details

What if we double the volatility  $\Sigma$  of the risk factor?

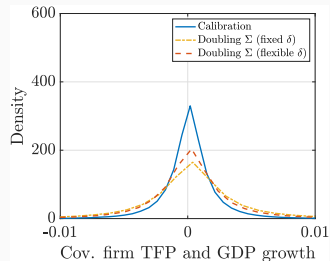
	Calibration	Doubling $\Sigma$	
		Fixed $\delta$	Flexible $\delta$
Agg. risk exposure $\Delta$	0.014	0.014	0.011
Exposure value $\mathcal{E}$	-0.06	-0.11	-0.09
Std. Dev. of GDP growth	2.4%	3.1%	2.6%



- **Fixed  $\delta$** : Large increase in **GDP variance**; exposure to  $\varepsilon_t$  becomes more harmful ( $\mathcal{E}$  declines)
- **Flexible  $\delta$** : Firms manage risk more aggressively which **limits increase in  $V[y]$**

What if we double the volatility  $\Sigma$  of the risk factor?

	Calibration	Doubling $\Sigma$	
		Fixed $\delta$	Flexible $\delta$
Agg. risk exposure $\Delta$	0.014	0.014	0.011
Exposure value $\mathcal{E}$	-0.06	-0.11	-0.09
Std. Dev. of GDP growth	2.4%	3.1%	2.6%



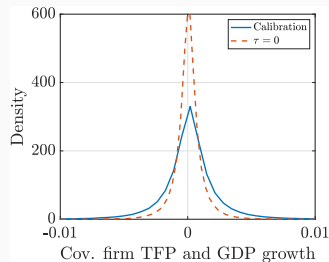
- **Fixed  $\delta$** : Large increase in **GDP variance**; exposure to  $\varepsilon_t$  becomes more harmful ( $\mathcal{E}$  declines)
- **Flexible  $\delta$** : Firms manage risk more aggressively which **limits increase in  $V[y]$**

Impact of risk can be overestimated if reaction of agents is not taken into account

## Removing distortions

What if we set wedges  $\tau$  to zero?

	Calibration	No wedges	
		Fixed $\delta$	Flexible $\delta$
Agg. risk exposure $\Delta$	0.014	0.014	0.007
Exposure value $\mathcal{E}$	-0.06	-0.06	-0.03
Std. Dev. of GDP growth	2.4%	2.4%	1.7%



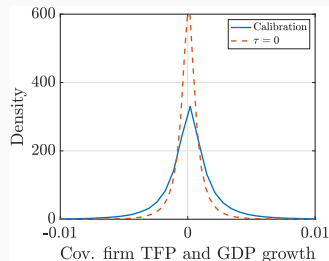
- **Fixed  $\delta$** : Since only impact of  $\tau$  is through  $\delta$ , there is no change.
- **Flexible  $\delta$** : Firms manage risk more aggressively so  $V[y]$  declines



## Removing distortions

What if we set wedges  $\tau$  to zero?

	Calibration	No wedges	
		Fixed $\delta$	Flexible $\delta$
Agg. risk exposure $\Delta$	0.014	0.014	0.007
Exposure value $\mathcal{E}$	-0.06	-0.06	-0.03
Std. Dev. of GDP growth	2.4%	2.4%	1.7%



- **Fixed  $\delta$** : Since only impact of  $\tau$  is through  $\delta$ , there is no change.
- **Flexible  $\delta$** : Firms manage risk more aggressively so  $V[y]$  declines

Distortions can make GDP more volatile

## Conclusion

---

## Main contributions

- We construct a model of **endogenous risk**, at both the micro and macro levels.
- Model predicts which firms are more volatile and covary more with business cycle.
- Distortions lead to less aggressive risk management and can increase GDP volatility.

## Future research

- What if there are entrepreneurs who cannot diversify their risk?
- Mechanisms would interact with capital/investment. Fully dynamic business cycle model.

The function  $\zeta(\alpha_i)$  is

$$\zeta(\alpha_i) = \left[ \left( 1 - \sum_{j=1}^n \alpha_{ij} \right)^{1 - \sum_{j=1}^n \alpha_{ij}} \prod_{j=1}^n \alpha_{ij}^{\alpha_{ij}} \right]^{-1}$$

This functional form allows for a simple expression for the unit cost  $K$

Given the log-normal nature of uncertainty  $\rho \leq 1$  determines whether the agent is risk-averse or not. To see this, note that when  $\log C$  normally distributed, maximizing

$$\mathbf{E} [C^{1-\rho}]$$

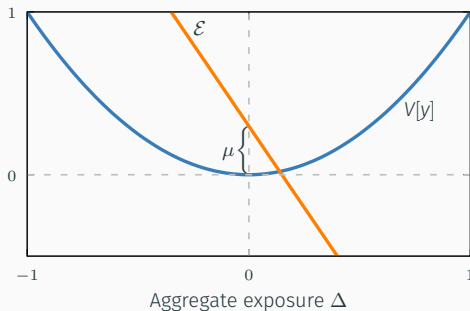
amounts to maximizing

$$\mathbf{E} [\log C] - \frac{1}{2} (\rho - 1) \mathbf{V} [\log C].$$

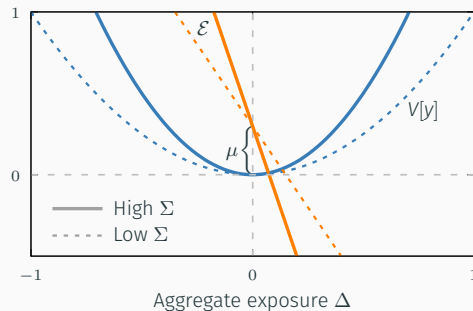
The direct impact of changes in  $(\mu, \Sigma)$  is given by

$$\frac{\partial\mathcal{E}}{\partial\mu_m} = \mathbf{1}_m \quad \text{and} \quad \frac{\partial\mathcal{E}}{\partial\Sigma_{mn}} = -\frac{1}{2}(\rho - 1)(\Delta_m\mathbf{1}_n + \Delta_n\mathbf{1}_m).$$

(a) Low  $\Sigma$



(b) High  $\Sigma$



## Proposition

The response of the equilibrium aggregate risk exposure  $\Delta$  to a change in wedge  $\tau_i$  is given by

$$\frac{d\Delta}{d\tau_i} = \mathcal{T} \left( \sum_{j=1}^N \frac{\partial [\nabla^2 \bar{\kappa}]^{-1}}{\partial g_j} \frac{dg_j}{d\tau_i} \right) \mathcal{E}, \quad (1)$$

where the impact of  $g_j$  on  $[\nabla^2 \bar{\kappa}]^{-1}$  is given by  $\frac{\partial [\nabla^2 \bar{\kappa}]^{-1}}{\partial g_j} = -\frac{1}{\eta} \frac{\tilde{\omega}_j^2}{g_j^2} H_j^{-1}$ , and where

$$\mathcal{T} := \left( I - [\nabla^2 \bar{\kappa}]^{-1} \frac{\partial \mathcal{E}}{\partial \Delta} \right)^{-1}.$$

## Proposition

Let  $\chi$  denote either  $\mu_m$ ,  $\Sigma_{mn}$ , or  $\tau_i$ . Then the impact of a change in  $\chi$  on the moments of log GDP are given by

$$\frac{dE[y]}{d\chi} - \frac{\partial E[y]}{\partial\chi} = \mu^\top \frac{d\Delta}{d\chi} \quad \text{and} \quad \frac{dV[y]}{d\chi} - \frac{\partial V[y]}{\partial\chi} = 2\Delta^\top \Sigma \frac{d\Delta}{d\chi},$$

where the use of a partial derivative indicates that  $\Delta$  is kept fixed.



## Simplified model

◀ Back

- Single risk factor  $\varepsilon_t \sim \text{iid } \mathcal{N}(0, \Sigma)$
- Firm level TFP is  $\log TFP_{it} = \delta_{it}\varepsilon_t + \gamma_i t + v_{it}$  where  $\gamma_i$  is deterministic trend and  $v_{it} \sim \text{iid } \mathcal{N}(\mu_i^v, \Sigma_i^v)$

## Simplified model

[← Back](#)

- Single risk factor  $\varepsilon_t \sim \text{iid } \mathcal{N}(0, \Sigma)$
- Firm level TFP is  $\log TFP_{it} = \delta_{it}\varepsilon_t + \gamma_i t + v_{it}$  where  $\gamma_i$  is deterministic trend and  $v_{it} \sim \text{iid } \mathcal{N}(\mu_i^v, \Sigma_i^v)$

## Variance of firm-level TFP growth

$$V[\log TFP_{it} - \log TFP_{it-1}] = 2\delta_i^2 \Sigma + 2\Sigma_i^v$$

## Simplified model

[← Back](#)

- Single risk factor  $\varepsilon_t \sim \text{iid } \mathcal{N}(0, \Sigma)$
- Firm level TFP is  $\log TFP_{it} = \delta_{it}\varepsilon_t + \gamma_i t + v_{it}$  where  $\gamma_i$  is deterministic trend and  $v_{it} \sim \text{iid } \mathcal{N}(\mu_i^v, \Sigma_i^v)$

## Variance of firm-level TFP growth

$$V[\log TFP_{it} - \log TFP_{it-1}] = 2\delta_i^2 \Sigma + 2\Sigma_i^v$$

## Covariance of firm-level TFP growth with GDP growth

$$\text{Cov}[\log TFP_{it} - \log TFP_{it-1}, y_t - y_{t-1}] = 2\Delta \Sigma \delta_i + 2\tilde{\omega}_i \Sigma_i^v.$$

## Simplified model

- Single risk factor  $\varepsilon_t \sim \text{iid } \mathcal{N}(0, \Sigma)$
- Firm level TFP is  $\log TFP_{it} = \delta_{it}\varepsilon_t + \gamma_i t + v_{it}$  where  $\gamma_i$  is deterministic trend and  $v_{it} \sim \text{iid } \mathcal{N}(\mu_i^v, \Sigma_i^v)$

## Variance of firm-level TFP growth

$$V[\log TFP_{it} - \log TFP_{it-1}] = 2\delta_i^2 \Sigma + 2\Sigma_i^v$$

## Covariance of firm-level TFP growth with GDP growth

$$\text{Cov}[\log TFP_{it} - \log TFP_{it-1}, y_t - y_{t-1}] = 2\Delta \Sigma \delta_i + 2\tilde{\omega}_i \Sigma_i^v$$

## Model-implied firm risk exposure ( $\mathcal{E} < 0$ )

$$\delta_i = \delta_i^o + \frac{1}{\eta} \frac{\omega_i}{1 + \tau_i} H_i^{-1} \mathcal{E}$$

⇒ Firms with large Domar weights and small markups are less volatile and less corr. with GDP

- Assume Cobb-Douglas production function

$$\log Q_{it} = \alpha_{L_i} \log L_{it} + \alpha_{M_i} \log M_{it} + \alpha_{K_i} \log K_{it} + \varepsilon_{it},$$

- Elasticities estimated using Levinsohn and Petrin (2003) with the Ackberg et al. (2015) correction.
  - Capital is the “state” variable, labor is the “free” variable and materials is the “proxy” variable.
- Production function estimated at NACE 2-digit sector level. As in De Loecker et al. (2020), we control for markups using firms’ sales shares in the production function estimation.
- Following De Loecker and Warzynski (2012), we compute the markup as  $1 + \tau_{it} = \hat{\alpha}_{L_i} / \left( \frac{\text{Wage Bill}_{it}}{\text{Sales}_{it}} \right)$ .
- We compute TFP growth as

$$\begin{aligned} \Delta \log \text{TFP}_{it} = & \Delta \log Q_{it} - \alpha_{L_i} \Delta \log L_{it} - \alpha_{M_i} \Delta \log M_{it} - \alpha_{K_i} \Delta \log K_{it} \\ & - \left( \Delta \log (1 + \tau_{it}) - \Delta \log (1 + \tau_{s(i)t}) \right). \end{aligned}$$

The term  $\Delta \log (1 + \tau_{it}) - \Delta \log (1 + \tau_{s(i)t})$  accounts for the firm-specific markup growth net of the sectoral markup growth. This adjustment allows us to remove the change in firm-specific nominal price that are not taken into account by the sector-level price deflator.

- We compute the standard deviation of TFP growth for each firm,  $\sigma_i (\Delta \log TFP_{it})$ , and the time-series average of its markup and Domar weight.
- We construct deciles based on average Domar weights and markups, and create dummy variables,  $FE_{ji}^{Domar}$  and  $FE_{ji}^{Markup}$ , such that  $FE_{ji}^{Domar} = 1$  if firm  $i$ 's Domar weight is in decile  $j$ , and analogously for markups.
- We run the cross-sectional regression

$$\sigma_i (\Delta \log TFP_{it}) = \alpha + \sum_{j=1}^{10} \beta_j^{Domar} FE_{ji}^{Domar} + \sum_{j=1}^{10} \beta_j^{Markup} FE_{ji}^{Markup} + \varepsilon_i,$$

and plot  $\beta_j^{Domar}$  in panel (a) and  $\beta_j^{Markup}$  in panel (b).

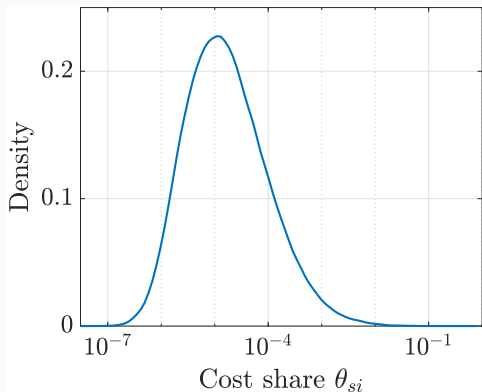
- We construct deciles based on firms' Domar weights and markups each year.
- We then construct a set of dummy variables,  $FE_{jit}^{Domar}$  and  $FE_{jit}^{Markup}$ , such that  $FE_{jit}^{Domar} = 1$  if firm  $i$ 's Domar weight is in decile  $j$  in year  $t$ , and analogously for markups.
- We then run the following panel regression,

$$\begin{aligned}\Delta \log TFP_{it} = & \sum_{j=1}^{10} \beta_j^{Domar} \left( FE_{jit}^{Domar} \times \Delta \log GDP_t \right) + \sum_{j=1}^{10} \beta_j^{Markup} \left( FE_{jit}^{Markup} \times \Delta \log GDP_t \right) \\ & + \alpha + \beta_0 \Delta \log GDP_t + \sum_{j=1}^{10} FE_{jit}^{Domar} + \sum_{j=1}^{10} FE_{jit}^{Markup} + \varepsilon_{it},\end{aligned}$$

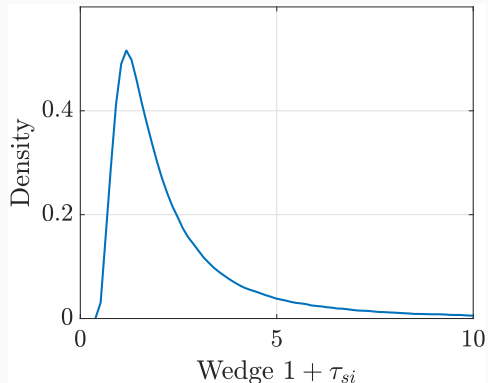
where  $\Delta \log TFP_{it}$  is the annual growth of firm  $i$ 's log TFP and  $\Delta \log GDP_t$  is the annual growth of Spanish log GDP.

- The coefficients of interest,  $\beta_j^{Domar}$  and  $\beta_j^{Markup}$ , are reported in the figure.

Figure 1: Data distributions that the calibration matches exactly



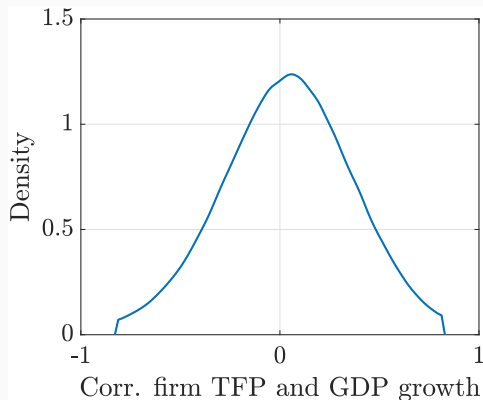
(a) Sales share  $\theta_{si}$



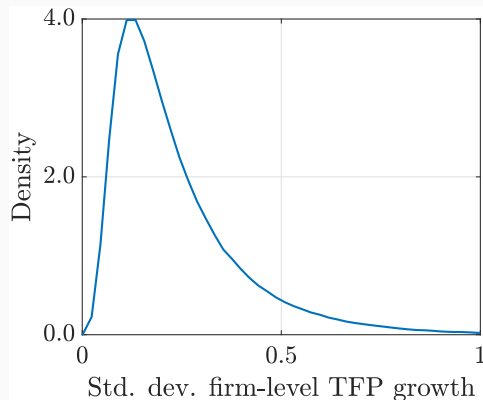
(b) Wedges  $1 + \tau_i$



Figure 2: Data distributions that the calibration matches exactly

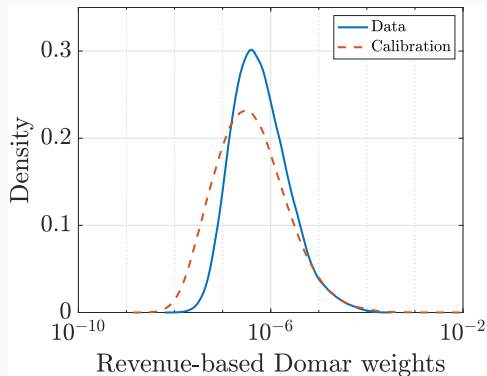


(a) Correlation firm-level TFP and GDP growth

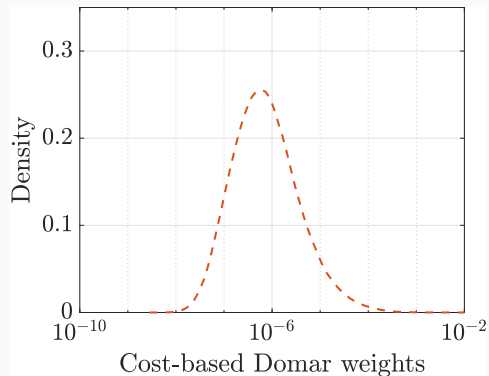


(b) Standard deviation of firm-level TFP growth

Figure 3: Domar weights of the firms in the data and in the model



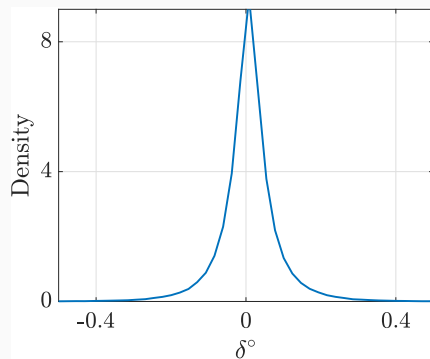
(a) Revenue-based Domar weights



(b) Cost-based Domar weights



Figure 5: Distribution of the estimated firm-level natural risk exposure  $\delta_i^\circ$



Notes. The scale of  $\delta_i^\circ$  depends on our choice of  $\rho$  and  $\Sigma$ . We set  $\rho = 5$  and  $\Sigma = 1$  for this figure.