

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 $a_i(\epsilon, \delta_i)$

There are underlying sources of risk $\epsilon = (\epsilon_1, \dots, \epsilon_M)$ with $\epsilon \sim \mathcal{N}(\mu, \Sigma)$

- Examples: a river floods, discovery of a new drug, war between two countries, epidemic, etc.
- We don't take a stance on what ϵ 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 ε is realized: Firms choose risk exposure δ
2. After ε is realized: All other quantities are chosen

Timing

1. Before ε is realized: Firms choose risk exposure δ
2. After ε is realized: All other quantities are chosen

Firms set prices P at a **constant wedge** τ_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 ε is realized: Firms choose risk exposure δ
2. After ε is realized: All other quantities are chosen

Firms set prices P at a **constant wedge** τ_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**

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

where Q_i is *equilibrium* demand and Λ 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 Λ 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 δ^* 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 δ_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:

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

Two measures of supplier importance

Cost-based Domar weight:

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

- Depends on demand from household (β) 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:

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

- Depends on demand from household (β) 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:

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

- Also captures importance as a supplier (share of revenues)
- Declines with wedges τ

Define aggregate risk exposure Δ as

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

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

Define aggregate risk exposure Δ 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: $y = \Delta^\top \varepsilon = \omega^\top a(\varepsilon, \delta)$

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

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

Impact of Σ

- A marginal increase in Σ_{mm} raises $V[y]$ by Δ_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 Σ_{mn} raises $V[y]$.
- If $\Delta_m > 0$ and $\Delta_n < 0$, the shocks offset each other. Higher Σ_{mn} reduces $V[y]$.

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

Impact of Σ

- A marginal increase in Σ_{mm} raises $V[y]$ by Δ_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 Σ_{mn} raises $V[y]$.
- If $\Delta_m > 0$ and $\Delta_n < 0$, the shocks offset each other. Higher Σ_{mn} reduces $V[y]$.

Impact of Δ

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

- Extra exposure to ε_m increases volatility if ε_m is positively correlated with GDP

Lemma

The equilibrium risk exposure decision δ_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 δ_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 δ_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 ω_i** and **low τ_i** manage risk **more aggressively**

Lemma

The equilibrium risk exposure decision δ_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 δ_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 ω_i** and **low τ_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 Δ .

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

Planner's problem

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

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

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 Δ 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 Δ .

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

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 Δ .

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

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 Δ .

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

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 γ be either μ_m or Σ_{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 γ on the attractiveness of risk factors
- The matrix \mathcal{H}^{-1} **propagates** that impact to exposure vector Δ

► $\partial \mathcal{E} / \partial \gamma$

Corollary

1. An increase in μ_m raises Δ_m
 2. An increase in Σ_{mm} reduces Δ_m if $\Delta_m > 0$ and increases Δ_m if $\Delta_m < 0$
- A marginal increase in Σ_{mm} raises $V[y]$ by $\Delta_m^2 \rightarrow$ When Σ_{mm} increases we want to reduce Δ_m^2

Corollary

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

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

What is the impact of μ_m or Σ_{mm} on Δ_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 Δ_n
- If $[\mathcal{H}^{-1}]_{mn} < 0$, m and n are *global substitutes* \rightarrow an increase in \mathcal{E}_m decreases Δ_n

Corollary

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

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

What is the impact of μ_m or Σ_{mm} on Δ_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 Δ_n
- If $[\mathcal{H}^{-1}]_{mn} < 0$, m and n are *global substitutes* \rightarrow an increase in \mathcal{E}_m decreases Δ_n

Global substitution patterns depend on $\mathcal{H}^{-1} := (\nabla^2 \bar{\kappa} + (\rho - 1) \Sigma)^{-1}$

- $\nabla^2 \bar{\kappa}$: global impact of the *local substitution patterns* embedded in $(\kappa_1, \dots, \kappa_N)$
- Σ : if $\Sigma_{mn} > 0$ an increase in Δ_m makes the planner reduce Δ_n to avoid agg. risk

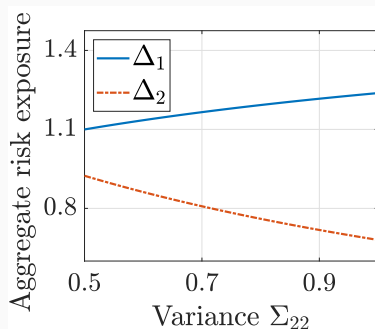
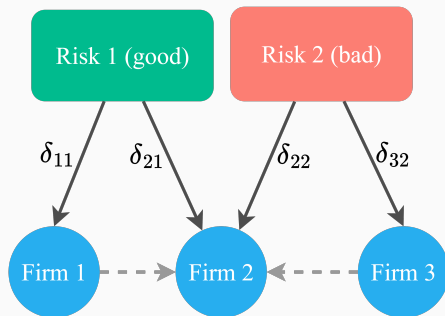
Example of substitution patterns

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 Σ and H_i are diagonal for every i

Definition. An economy is diagonal if Σ and H_i are diagonal for every i

Corollary

In a diagonal economy, a higher wedge τ_i

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

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

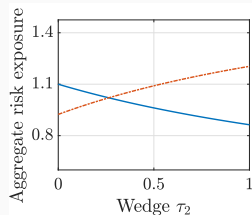
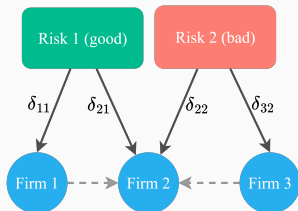
Definition. An economy is diagonal if Σ and H_i are diagonal for every i

Corollary

In a diagonal economy, a higher wedge τ_i

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

When all firms are at their **natural exposure** δ° we have $\mathcal{E}^\circ = \mu - (\rho - 1) \Sigma \Delta^\circ$

Lemma

Equilibrium risk exposure is distorted such that $(\Delta - \Delta_{SP})^\top \mathcal{E}^\circ < 0$.

When all firms are at their **natural exposure** δ° we have $\mathcal{E}^\circ = \mu - (\rho - 1) \Sigma \Delta^\circ$

Lemma

Equilibrium risk exposure is distorted such that $(\Delta - \Delta_{SP})^\top \mathcal{E}^\circ < 0$.

- Wedges make firms **inefficiently small** \rightarrow less risk management
- Eqm. is on average **overexposed to bad risks** ($\mathcal{E}^\circ < 0$) and **underexposed to good risks** ($\mathcal{E}^\circ > 0$)

Implications for GDP and Welfare

Use ∂ to denote changes in the economy with **exogenous risk**

Proposition

In a diagonal economy:

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

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

Proposition (single risk factor)

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

Suppose $\mathcal{E} < 0$ (bad risk, e.g. business cycle): increasing τ_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

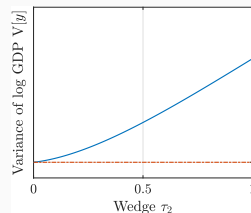
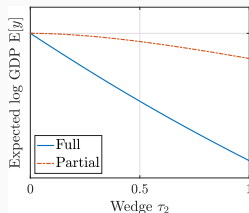
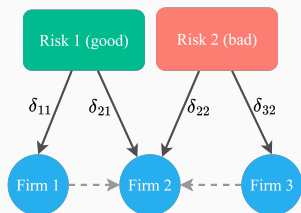
Distortions can increase aggregate volatility

Proposition (single risk factor)

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

Suppose $\mathcal{E} < 0$ (bad risk, e.g. business cycle): increasing τ_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



Proposition

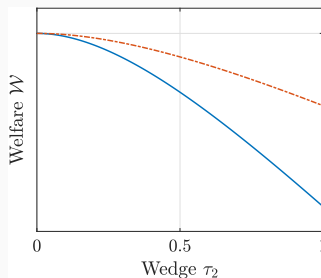
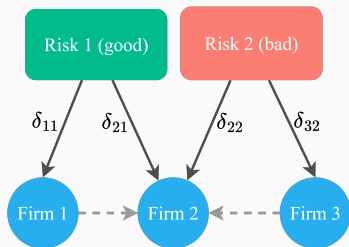
In a diagonal economy, raising τ_i hurts welfare more than under exogenous risk.

- A higher τ_i increases exposure to bad risks and lower exposure to good risks
- Additional exposure to bad risks hurts welfare, and vice-versa for good risks

Proposition

In a diagonal economy, raising τ_i hurts welfare more than under exogenous risk.

- A higher τ_i increases exposure to bad risks and lower exposure to good risks
- Additional exposure to bad risks hurts welfare, and vice-versa for good risks



(Blue: flexible risk; Red: fixed risk)

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

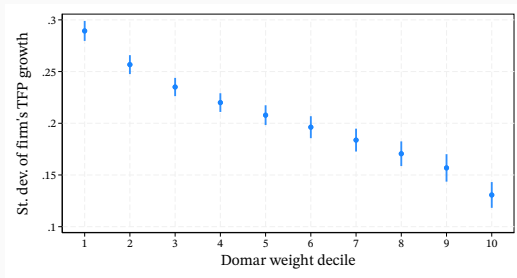
► Details

We test these predictions in the data

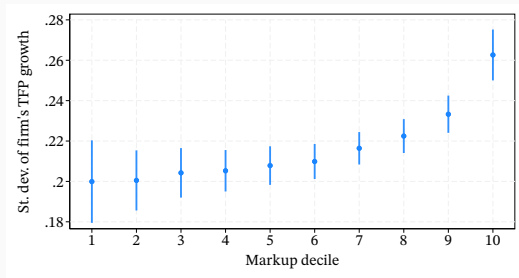
- Use detailed micro data from the near-universe of firms in Spain between 1995 and 2018 (Orbis) (7,513,081 firm-year observations)
- Compute markups using control function approach (De Loecker and Warzynski, 2012)
- Back out TFP growth as a residual

► Details

TFP growth volatility



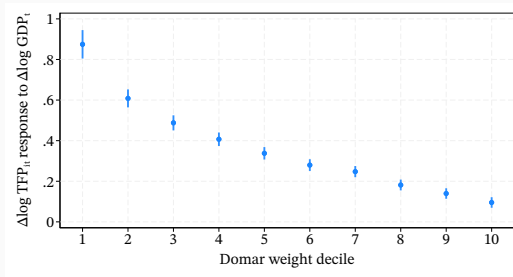
(a) TFP volatility by Domar weight decile



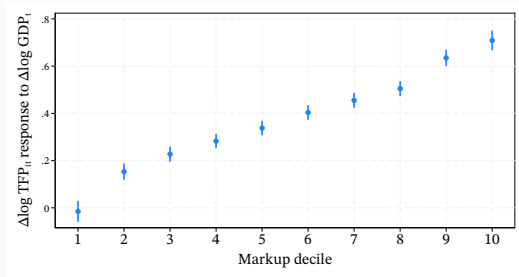
(b) TFP volatility by markup decile

► 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

A specialized model to map to the data

- **S sectors** with aggregator $Q_s = \prod_{i=1}^{N_s} e^{z_s} (\theta_{si}^{-1} Q_{si})^{\theta_{si}}$ and sectoral shocks $z_s \sim \text{iid } \mathcal{N}(\mu_s^z, \Sigma_s^z)$
- Firms have **production function**

$$Q_{si} = e^{\delta_{sit}\varepsilon_t + \gamma_{si}t + v_{sit}} \zeta_{si} L_{si}^{1 - \sum_{s'} \hat{\alpha}_{ss'}} \prod_{s'=1}^S \chi_{si,s'}^{\hat{\alpha}_{ss'}}$$

where $\hat{\alpha}_{ss'}$ are sectoral shares, $v_{sit} \sim \text{iid } \mathcal{N}(\mu_{si}^v, \Sigma_{si}^v)$ and $\varepsilon_t \sim \text{iid } \mathcal{N}(0, \Sigma)$

- Risk management **cost function** is parametrized as

$$\frac{1}{\eta} H_{si}^{-1} = a_s \tilde{\omega}_{si}^{b_s} + c_s$$

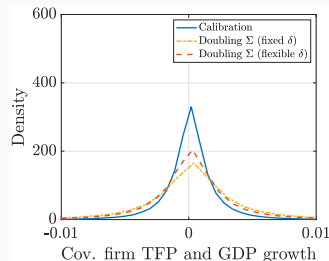
Allows for a **size effect** on risk management costs

- 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

Doubling Σ

What if we double the volatility Σ of the risk factor?

| | Calibration | Doubling Σ | |
|------------------------------|-------------|-------------------|-------------------|
| | | Fixed δ | Flexible δ |
| Agg. risk exposure Δ | 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% |

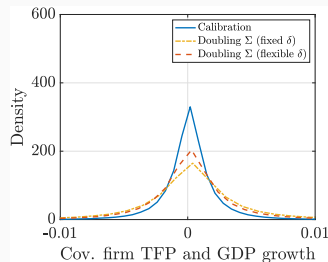


- **Fixed δ :** Large increase in **GDP variance**; exposure to ε_t becomes more harmful (\mathcal{E} declines)
- **Flexible δ :** Firms manage risk more aggressively which **limits increase in $V[y]$**

Doubling Σ

What if we double the volatility Σ of the risk factor?

| | Calibration | Doubling Σ | |
|------------------------------|-------------|-------------------|-------------------|
| | | Fixed δ | Flexible δ |
| Agg. risk exposure Δ | 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% |



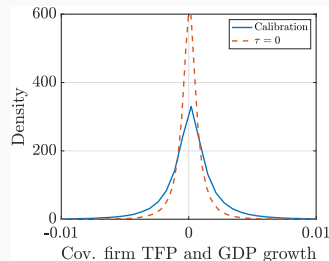
- **Fixed δ :** Large increase in **GDP variance**; exposure to ε_t becomes more harmful (\mathcal{E} declines)
- **Flexible δ :** Firms manage risk more aggressively which **limits increase in $V[y]$**

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

Removing distortions

What if we set wedges τ to zero?

| | Calibration | No wedges | |
|------------------------------|-------------|----------------|-------------------|
| | | Fixed δ | Flexible δ |
| Agg. risk exposure Δ | 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% |

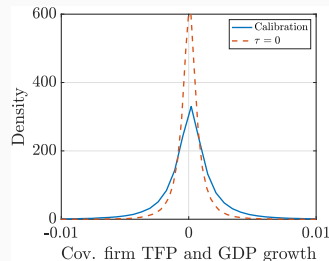


- **Fixed δ :** Since only impact of τ is through δ , there is no change.
- **Flexible δ :** Firms manage risk more aggressively so $V[y]$ declines

Removing distortions

What if we set wedges τ to zero?

| | Calibration | No wedges | |
|------------------------------|-------------|----------------|-------------------|
| | | Fixed δ | Flexible δ |
| Agg. risk exposure Δ | 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 δ :** Since only impact of τ is through δ , there is no change.
- **Flexible δ :** 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

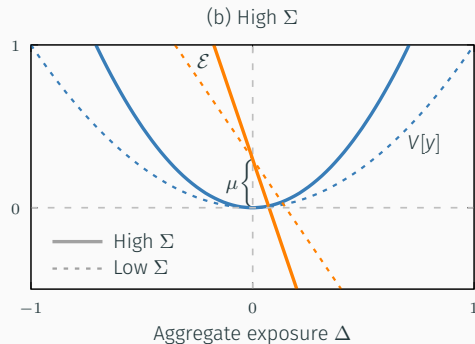
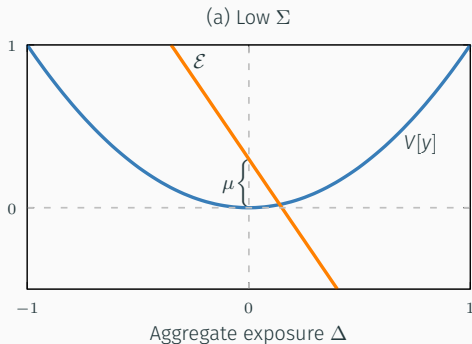
$$\mathbb{E} [C^{1-\rho}]$$

amounts to maximizing

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

The direct impact of changes in (μ, Σ) is given by

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



Proposition

The response of the equilibrium aggregate risk exposure Δ to a change in wedge τ_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 χ denote either μ_m , Σ_{mn} , or τ_i . Then the impact of a change in χ 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 Δ 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 γ_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 γ_i is deterministic trend and $v_{it} \sim \text{iid } \mathcal{N}(\mu_i^v, \Sigma_i^v)$

Variance of firm-level TFP growth

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

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 γ_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 γ_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

$$\log Q_{it} = \alpha_{Li} \log L_{it} + \alpha_{Mi} \log M_{it} + \alpha_{Ki} \log K_{it} + \varepsilon_{it},$$

- 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 $1 + \tau_{it} = \hat{\alpha}_{Li} / \left(\frac{\text{Wage Bill}_{it}}{\text{Sales}_{it}} \right)$.
- We compute TFP growth as

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

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 β_j^{Domar} in panel (a) and β_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, β_j^{Domar} and β_j^{Markup} , are reported in the figure.

- Risk exposure

$$\delta_{si} = \delta_{si}^o + \frac{\tilde{\omega}_{si}}{g_{si}} \left(\frac{1}{\eta} H_{si}^{-1} \right) \mathcal{E}$$

- The variance of GDP growth is

$$V[y_t - y_{t-1}] = 2\Sigma\Delta^2 + 2\tilde{\omega}_f^\top \Sigma^v \tilde{\omega}_f + 2\tilde{\omega}_s^\top \Sigma^z \tilde{\omega}_s.$$

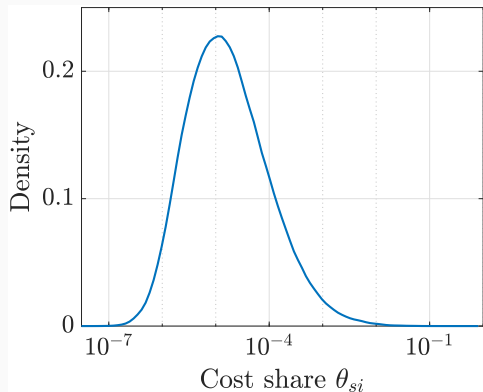
- The variance of firm-level TFP growth is

$$V[\log TFP_{si,t} - \log TFP_{si,t-1}] = 2\delta_{si}^2 \Sigma + 2\Sigma_{si}^v.$$

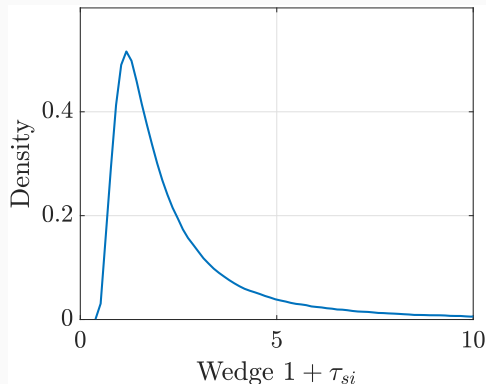
- The covariance of firm-level TFP growth with GDP growth is

$$\text{Cov}[y_t - y_{t-1}, \log TFP_{si,t} - \log TFP_{si,t-1}] = 2\Delta\Sigma\delta_{si} + 2\tilde{\omega}_{si}\Sigma_{si}^v.$$

Figure 1: Data distributions that the calibration matches exactly

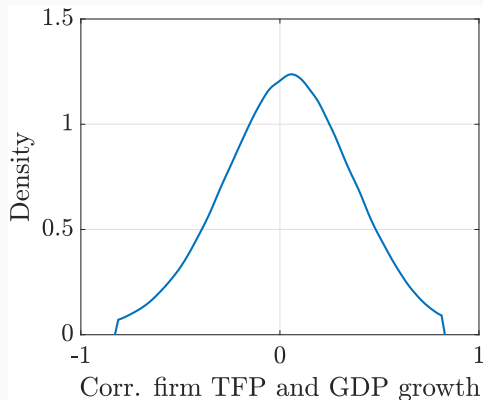


(a) Sales share θ_{si}

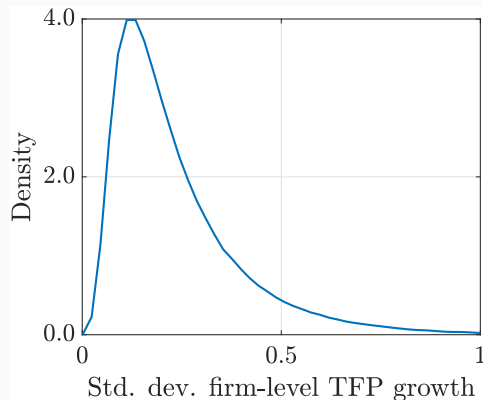


(b) Wedges $1 + \tau_i$

Figure 2: Data distributions that the calibration matches exactly

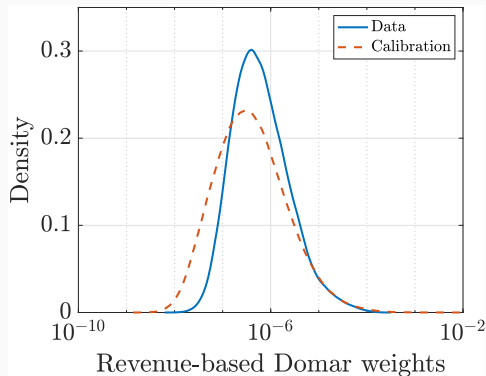


(a) Correlation firm-level TFP and GDP growth

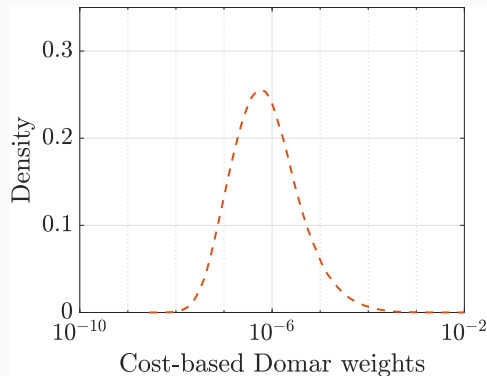


(b) Standard deviation of firm-level TFP growth

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

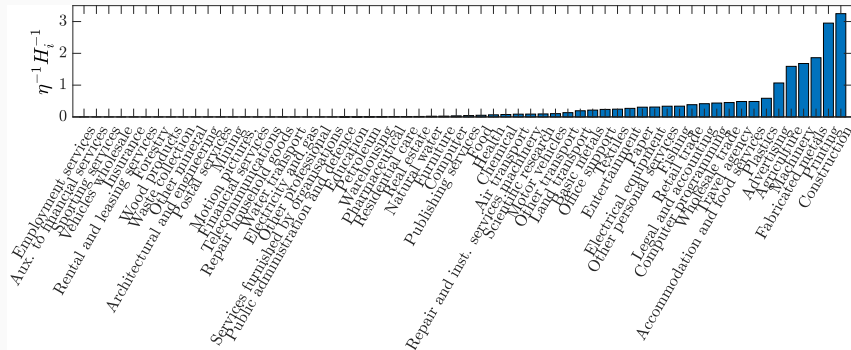


(a) Revenue-based Domar weights



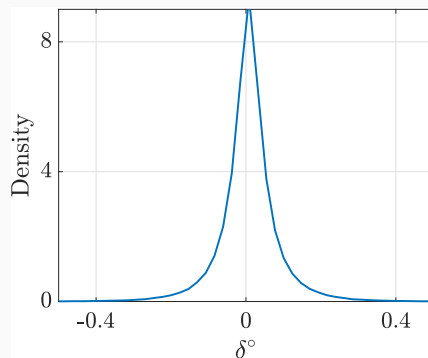
(b) Cost-based Domar weights

Figure 4: Estimated value of $\frac{1}{\eta} H_i^{-1}$ for each sector.



Notes. The scale of $\frac{1}{\eta} H_i^{-1}$ depends on our choice of ρ and Σ . We set $\rho = 5$ and $\Sigma = 1$ for this figure.

Figure 5: Distribution of the estimated firm-level natural risk exposure δ_i°



Notes. The scale of δ_i° depends on our choice of ρ and Σ . We set $\rho = 5$ and $\Sigma = 1$ for this figure.