

Incomplete markets as correlated distortions*

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Abstract

How does insurance market incompleteness affect the allocative efficiency of capital? We study this question using a model of risk-averse entrepreneurs with heterogeneous productivity who face idiosyncratic revenue shocks and make irreversible capital investments before uncertainty unfolds. We show that incomplete insurance markets operate as *correlated distortions*—that is, they are isomorphic to a progressive output tax, leading to a reallocation of capital from more to less productive entrepreneurs relative to the complete-markets benchmark. Using Portuguese administrative data, we test this hypothesis and document that capital misallocation is greater in locations and industries with higher shock volatility, consistent with the predictions of our framework. Completing financial markets would increase aggregate productivity by up to 27%.

Keywords: Insurance, volatility, misallocation, distortions, efficiency.

JEL Classification: E22, D61, L23, L26

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1 Introduction

Resource misallocation is a key driver of aggregate productivity and income losses. Yet, what ultimately drives misallocation remains an open question. In this paper, we show that insurance markets' incompleteness can act as a barrier to the efficient allocation of factors of production across firms.

We develop a general equilibrium model of consumption insurance and the capital market to study how insurance markets' incompleteness affects the allocation of capital to heterogeneous firms, as well as aggregate productivity and income. The model economy is populated by risk-averse entrepreneurs, each running a firm characterized by a decreasing return to scale production function, which uses capital as a factor of production. Entrepreneurs are heterogeneous in their productivity levels, face idiosyncratic productivity shocks, and choose how much capital to rent before uncertainty unfolds; that is, they are subject to time-to-build constraints.

Insurance markets' incompleteness ties entrepreneurs' consumption to the profits generated by their enterprises, making them willing to forgo expected profits to reduce consumption volatility. This distorts entrepreneurs' demand for capital away from expected profit maximization in an effort to avoid losses. As a result, the expected marginal products of capital are not equalized across firms; i.e., capital is misallocated.

Specifically, we show that incomplete markets operate as *correlated distortions*, introducing an endogenous wedge between the expected marginal product of capital and its price that is positively correlated with entrepreneurial productivity. Thus, capital is reallocated from more to less productive entrepreneurs relative to the complete markets benchmark. This happens because renting more capital implies a larger scale of production, which results in a higher sensitivity of expected utility to profit volatility. The wedge disproportionately impacts high-productivity entrepreneurs, as they are the ones who would employ more capital if they could fully insure their consumption, and it becomes more severe as productivity shock volatility increases.

The model also provides a novel link between entrepreneurial income risk and capital misallocation: a higher volatility of productivity shocks intensifies misallocation by

weakening the cross-sectional correlation between entrepreneurial productivity and capital holdings. We corroborate this prediction using administrative data on the quasi-universe of Portuguese firms, spanning more than a decade. We estimate production functions to recover firm-level permanent revenue productivity and idiosyncratic residual volatility in sales. We use these estimates to infer capital misallocation and the volatility of productivity shocks in each location-industry pair. The empirical evidence aligns with our model's predictions: we find a lower correlation between firm productivity and capital holdings in locations and industries with higher average idiosyncratic residual volatility.

Therefore, we leverage the structure of our model to quantify the aggregate cost of incomplete markets. To do so, we calibrate the model to reproduce the observed joint distribution of firm productivity and capital holdings from the data. Next, we conduct two counterfactual exercises. In the first one, we examine a counterfactual economy where each firm receives the expected value of the productivity shock with certainty; i.e., productivity shocks are eliminated. This exercise allows us to quantify the extent to which idiosyncratic productivity shocks impact aggregate productivity and income when entrepreneurs lack access to a full set of state-contingent claims spanning all possible realizations of productivity shocks. In the second one, we complete insurance markets while maintaining the volatility of productivity shocks to the baseline level. Our calibration implies that, in a world without productivity shocks, aggregate output per capita would be 20% higher. If insurance markets were complete, it would increase by 27%.

Finally, we check the robustness of our findings to i) incorporating firm-specific correlated distortions, meant to capture other distortions and frictions that disproportionately affect high-productivity entrepreneurs, and to ii) incorporating labor as a factor of production and allowing firms to adjust it after productivity shocks are realized. We find that the aggregate efficiency gains from completing the insurance market would remain substantial, and amount to between 24 and 46% of the benchmark value.

Our paper contributes to the literature examining the causes and consequences of factor misallocation (Hopenhayn, 2014; Restuccia and Rogerson, 2017). This body of work either explains misallocation as a result of specific policies or frictions—such as

labor regulations (Gourio and Roys, 2014; Garicano et al., 2016), borrowing constraints (Banerjee and Moll, 2010; Moll, 2014), imperfect information (David et al., 2016), and corporate income taxes (Cisneros-Acevedo and Ruggieri, 2022; Ábrahám et al., 2023)—or reduced-form wedges distorting factor allocation across firms (Guner et al., 2008; Hsieh and Klenow, 2009; Bento and Restuccia, 2017; David and Venkateswaran, 2019). Building on this literature, we show that capital misallocation arises *endogenously* among risk-averse entrepreneurs as a consequence of missing insurance markets, generating aggregate productivity losses that exist independently of other market imperfections.

Within this line of work, our paper is mostly related to the literature that studies financial frictions and uncertainty as potential sources of misallocation (Moll, 2014; Midrigan and Xu, 2014; Buera et al., 2015; Karabarbounis and Macnamara, 2021). Most related to our work is Asker et al. (2014), who develop a model of capital investment with adjustment costs and show that industries exhibiting greater time-series volatility of productivity have greater cross-sectional dispersion of the marginal revenue product of capital. Boar et al. (2022) document that the average returns to private business wealth are dispersed and persistent, and interpret this evidence as mostly reflecting uninsurable entrepreneurial risk as opposed to collateral constraints. David et al. (2022) link dispersion in the marginal product of capital across firms to exposure to aggregate shocks, showing that firm-level risk premia in capital contribute to capital misallocation and aggregate productivity. Our paper complements this literature by providing a novel mechanism behind the dispersion in the marginal revenue product of capital. Missing insurance markets weaken the cross-sectional correlation between entrepreneurial productivity and capital stock relative to the efficient benchmark, behaving as if they were *correlated distortions*, thus leading to capital misallocation.

Finally, our paper contributes to the literature examining the consequences of incomplete insurance markets for entrepreneurial activities (Quadrini, 2000; Cagetti and De Nardi, 2006; Vereshchagina and Hopenhayn, 2009). Recent contributions highlight that uninsurable idiosyncratic risk distorts the decision to become an entrepreneur (Robinson, 2021), affects investment decisions and portfolio allocations (Chen et al., 2010; Wang et al., 2012), induces over-accumulation of wealth to self-insure (Di Tella

et al., 2025), and might create inefficient recessions (Di Tella and Hall, 2022). Keeping aside the focus on aggregate risk, it remains poorly understood how the absence of insurance differentially affects entrepreneurs based on their permanent productivity. This paper fills this gap by showing that, when firms are heterogeneous, missing insurance against idiosyncratic shocks alters the allocation of capital across firms, leading to under-investment of high-productivity entrepreneurs and over-investment of low-productivity ones.

2 Model

We examine an economy with incomplete markets, populated by risk-averse heterogeneous entrepreneurs, each operating a firm and facing productivity shocks. In this section, we focus on a static model where capital is the only factor of production.

There exists a unitary measure of entrepreneurs indexed by i . Let $u(\cdot)$ be a strictly increasing, strictly concave, and twice-continuously differentiable utility function representing their preferences over consumption. Entrepreneurs differ in their productivity $z \in \mathcal{Z} \subset \mathbb{R}_{++}$ and have access to a production technology $f(z, k)$ that depends on productivity and capital k . We assume that $f(z, k)$ is strictly increasing, strictly concave, supermodular, and twice continuously differentiable. Production is subject to an idiosyncratic multiplicative shock, $s \in \mathcal{S} = [\underline{s}, \bar{s}] \subset \mathbb{R}_+$, which is distributed according to a probability density function $\phi(s)$. Each entrepreneur chooses capital before the realization of productivity shocks. Let r denote the rental rate of capital and $\bar{K}$ represent the exogenously given supply of capital.

2.1 Case 1: Complete markets

Under complete markets, entrepreneurs can trade a full set of state-contingent securities to insure against uncertainty in the realization of productivity shocks s . Hence, each entrepreneur i chooses capital k_i , state-contingent claims $\{\theta_i(s)\}_{s \in \mathcal{S}}$, and consumption in each state of the world $\{c_i(s)\}_{s \in \mathcal{S}}$ to maximize her expected utility. Let $q(s)$ denote the price of a contingent claim that pays a unit of consumption if state s realizes, and zero otherwise. Then, the problem of entrepreneur i can be written as

follows:

$$\begin{aligned} & \max_{k_i, \{\theta_i(s), c_i(s)\}_{s \in \mathcal{S}}} \int_{\mathcal{S}} u(c_i(s)) \phi(s) ds \\ \text{s.t. } & c_i(s) + \int_{\mathcal{S}} q(s') \theta_i(s') ds' \leq sf(z_i, k_i) - rk_i + T + \theta_i(s), \quad \forall s \in \mathcal{S}, \end{aligned}$$

where $T = r\bar{K}$ denotes revenues from renting capital, assumed to be rebated lump sum to each entrepreneur. In Appendix A we define a competitive equilibrium for an economy.

Dropping index i for convenience and noting that entrepreneurs exhaust their budget constraints in equilibrium, we can write an entrepreneur's problem as follows:

$$\max_{k, \{\theta(s)\}_{s \in \mathcal{S}}} \int_{s \in \mathcal{S}} u \left(sf(z, k) - rk + T - \int_{s' \in \mathcal{S}} q(s') \theta(s') ds' + \theta(s) \right) \phi(s) ds.$$

Let the subscript $*$ denote an interior solution to this problem. Letting u_c denote the marginal utility of consumption, the optimal choice of state-contingent claims under shock s , $\theta^*(s)$, satisfies the following first-order condition:

$$u_c(c^*(s)) \phi(s) = q(s) \int_{s' \in \mathcal{S}} u_c(c^*(s')) \phi(s') ds', \quad \forall s \in \mathcal{S},$$

where the marginal benefit of an extra unit of an asset that pays off if shock s realizes equates its marginal cost before uncertainty is realized. When the price for insurance is fair—i.e., $q(s) = \phi(s)$ —we can rewrite the previous first-order condition as follows:

$$u_c(c^*(s)) = \int_{s' \in \mathcal{S}} u_c(c^*(s')) \phi(s') ds', \quad \forall s \in \mathcal{S},$$

which implies that consumption is constant across output shock realizations; i.e.,

$$c^*(s) = u_c^{-1} \left(\int_{s' \in \mathcal{S}} u_c(c^*(s')) \phi(s') ds' \right), \quad \forall s \in \mathcal{S}.$$

Perfect consumption smoothing implies that the expected marginal product of capital is equalized across firms. To see this, notice that an entrepreneur's optimal choice of

capital k^* satisfies the following first-order condition:

$$\int_{s \in \mathcal{S}} u_c(c^*(s)) [sf_k(z, k^*) - r^*] \phi(s) ds = 0,$$

where f_k denotes the derivative of f with respect to k . Because $c^*(s) = c^*(s')$ for any two states $s, s' \in \mathcal{S}$, the condition above can be rewritten as:

$$\int_{s \in \mathcal{S}} [sf_k(z, k^*) - r^*] \phi(s) ds = 0. \quad (1)$$

That is, under complete markets, each entrepreneur chooses capital until the expected marginal product of capital equates its price. This leads to our first proposition.

Proposition 1. *Under complete markets and an actuarially fair price of insurance, the expected marginal product of capital is equalized across firms, and aggregate expected output is maximized.*

Proof. See Appendix A. □

With a full set of state-contingent claims ensuring perfect consumption smoothing, risk-averse entrepreneurs behave as if they were risk-neutral. As a result, the separation theorem holds: production decisions are independent of individual preferences, and each entrepreneur chooses capital to maximize expected profits.

2.2 Case 2: Incomplete markets

Consider an economy with incomplete markets. Each entrepreneur i now lacks access to state-contingent securities, and chooses capital k_i and consumption in each state of the world $\{c_i(s)\}_{s \in \mathcal{S}}$ to maximize expected utility. An entrepreneur i 's problem can be written as follows:

$$\begin{aligned} \max_{k_i, \{c_i(s)\}_{s \in \mathcal{S}}} \quad & \int_{s \in \mathcal{S}} u(c_i(s)) \phi(s) ds \\ \text{s.t.} \quad & c_i(s) \leq sf(z_i, k_i) - rk_i + T, \quad \forall s \in \mathcal{S}, \end{aligned}$$

where, again, $T = r\bar{K}$ are revenues from renting capital, assumed to be rebated lump sum to each entrepreneur. A definition of competitive equilibrium for this economy is

reported in Appendix A.

Again, let us drop subscript i . To characterize the equilibrium for this economy, we consider a CRRA utility function; i.e., $u(c) = \frac{c^{1-\gamma}}{1-\gamma}$, with $\gamma \neq 1$; and assume that $s \sim \log \mathcal{N}(\underline{s}, \sigma^2)$. Alternatively, one can assume that entrepreneurs have a strictly concave utility function and interpret our mean–variance analysis below as an approximate case, which becomes exact under specific assumptions such as CRRA utility with log-normal productivity shocks or CARA utility with normal shocks (Markowitz, 2014). Denote the expected value and variance of s as μ_s and σ_s^2 , respectively. Noting that entrepreneurs exhaust their budget constraints in equilibrium, we can write an entrepreneur’s problem as follows:

$$\max_k \int_{s \in \mathcal{S}} \frac{(sf(z, k) - rk + T)^{1-\gamma}}{1-\gamma} \phi(s) ds.$$

Given the log-normality of productivity shocks s , consumption follows a three-parameter log-normal distribution (Aitchison and Brown, 1957; Singh, 1998), with an expected value of $\bar{c} = \mu_s f(z, k) - rk + T$, variance $\sigma_c^2 = f(z, k)^2 \sigma_s^2$, and a threshold of $-rk + T$. Therefore, an entrepreneur’s problem can be rewritten in terms of consumption equivalent as follows:

$$\max_k \mu_s f(z, k) - rk + T - \frac{(\gamma - 1)}{2} f(z, k)^2 \sigma_s^2.$$

Let the subscript o denote an interior solution to this problem. Taking the first-order condition with respect to capital, we obtain:

$$f_k(z, k^o) [\mu_s - (\gamma - 1) \sigma_s^2 f(z, k^o)] - r^o = 0, \quad (2)$$

which leads to our second proposition.

Proposition 2. *Under incomplete markets, when $\gamma \neq 1$, there is misallocation along the intensive margin; i.e., the expected marginal product of capital is not equalized across entrepreneurs with different productivity.*

Proof. Obvious from Equation (1) and (2). □

When ex-post instruments for consumption smoothing fail to exist—i.e., when a full set of state-contingent claims is unavailable—risk-averse entrepreneurs choose capital to balance maximizing expected income and minimizing income volatility. As a result, capital allocation across entrepreneurs deviates from the allocation that maximizes aggregate expected output. In particular, lack of insurance introduces an endogenous *wedge* $\tau(z, k^\circ)$ between the marginal product of capital and its marginal cost; i.e., under incomplete markets, the equilibrium allocation of capital satisfies

$$\mu_s f_k(z, k^\circ)(1 - \tau(z, k^\circ)) = r^o,$$

where

$$\tau(z, k^\circ) = (\gamma - 1) \frac{\sigma_s^2}{\mu_s} f(z, k^\circ).$$

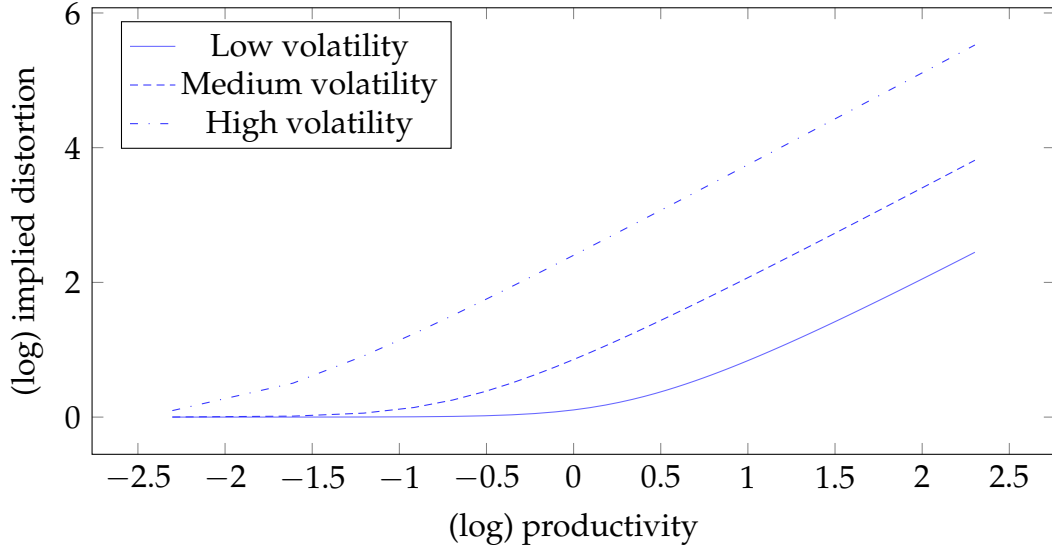
This leads to our third proposition.

Proposition 3. *When $\gamma > 1$, insurance markets' incompleteness operates as a correlated distortion, leading to a reallocation of capital from more to less productive firms relative to the complete markets benchmark.*

Proof. See Appendix A. □

When entrepreneurs are sufficiently risk-averse, market incompleteness acts as a tax on output that is positively correlated with entrepreneurial productivity. Figure 1 illustrates how the distortions implied by our model endogenously vary with entrepreneurial productivity. Under incomplete markets, all else equal, a larger scale of production makes utility more sensitive to productivity shocks. Due to complementarities in production, this distortion is particularly pronounced for high-productivity entrepreneurs, who optimally reduce their scale and forgo expected profits to secure a less volatile income. A second force operates in general equilibrium: capital being in fixed supply, the reduced demand for capital from high-productivity entrepreneurs lowers the rental price relative to the complete markets benchmark. This, in turn, incentivizes low-productivity entrepreneurs—those whose expected utility is less sensitive to productivity shocks—to rent more capital.

Figure 1: Correlated distortions and entrepreneurial productivity



NOTES: This figure reports the endogenous distortion, $\frac{1}{1-\tau(z,k^\circ)}$, implied by the estimated model as a function of the estimated entrepreneurial (log) productivity, z .

The figure also shows that this distortion becomes more pronounced when the volatility of productivity shocks is higher. This occurs because the wedge between the marginal product of capital and its price, $\tau(z, k^\circ)$ increases with the volatility of idiosyncratic productivity shocks σ . This observation leads to our fourth proposition.

Proposition 4. *Under incomplete markets, when $\gamma > 1$, a higher volatility of productivity shocks σ reduces the correlation between capital k and productivity z .*

Proof. See Appendix A. □

This is a central property of the model. Under incomplete markets, the distribution of marginal utilities of consumption across states of the world increasingly reflects the impact of productivity shocks. Amplifying production uncertainty is isomorphic to imposing larger distortions on the marginal product of capital for high-productivity entrepreneurs. As the marginal utilities of high-productivity entrepreneurs become more closely tied to their realized productivity shocks, their capital allocation decisions deviate even further from the complete markets benchmark. This result implies that, all else being equal, markets with higher volatility should also exhibit greater capital misallocation.

2.3 Discussion

Our model shows that incomplete insurance markets are sufficient to cause capital misallocation across entrepreneurs, above and beyond what is caused by other distortions in the capital market. Lack of insurance generates an endogenous wedge between the marginal product of capital and its price, which is positively correlated with firm productivity. As a result, incomplete markets lead to a reallocation of capital from high- to low-productivity firms relative to the complete markets benchmark.

In Appendix A, we show that the predictions of our model are robust to the inclusion of additional factors of production, such as labor, which can be adjusted ex-post (i.e., after the realization of productivity shocks), and a risk-free bond that entrepreneurs can use to smooth consumption over time. To the extent that insurance markets remain incomplete—that is, no set of contingent claims spans the entire set of states of the world—the equilibrium capital allocation does not maximize aggregate expected output.

Our results do not depend on the source of uncertainty or the nature of heterogeneity across entrepreneur types. Whether the former arises from supply uncertainty due to productivity shocks or demand uncertainty due to taste shocks, and whether the latter reflects differences in firm productivity or permanent taste heterogeneity for firm-specific products, the allocation of capital across entrepreneurs is still distorted by the lack of insurance, with higher uncertainty amplifying the magnitude of this distortion. In Appendix A, we also show that our mechanisms might operate alongside with financial frictions, when borrowing constraints are binding for *some firms*, eventually amplifying existing firm-level distortions, as long as entrepreneurs are enough risk averse.

Finally, none of our theoretical results are affected by allowing the production function f and the distribution ϕ to be heterogeneous across firms. Specifically, all propositions continue to hold when technology and shock volatility vary by locations and industries, and regardless of whether factor and insurance markets are integrated into a unique market or segmented into smaller local markets.¹

¹In such model, for each local market $j \in \mathcal{J}$, there exists a continuum $\mathcal{I}^j$ of entrepreneurs, with $\cup_{j=1}^{\mathcal{J}} \mathcal{I}^j = 1$. Entrepreneurs in local market j have access to a technology $f^j(z, k)$ and are subject to

3 Empirics

Our model predicts that, under incomplete markets, greater idiosyncratic production uncertainty results in increased capital misallocation. In this section, we test this prediction.

3.1 Data

We use firm-level data from Portugal’s *Sistema de Contas Integradas das Empresas* (SCIE, henceforth) for the period 2010-2021.² The dataset is an annual panel covering all firms engaged in the production of goods and services in Portugal, excluding financial and insurance companies as well as state-owned enterprises. For each firm-year pair, we observe sales, number of employees, payroll, material expenditures, current and non-current fixed assets, firm age, 4-digit industry classification, and location at the NUTS-2 level, which corresponds to one of Portugal’s seven main administrative regions.³ All nominal variables are deflated using the national CPI and expressed in 2010 constant prices. Since, in our data, labor is only recorded for firms with at least one employee, we exclude sole proprietors (“Empresas Individuais”) from our analysis.

In our analysis, we relate variation in idiosyncratic production uncertainty to capital misallocation across local markets, defined as the combination of a 2-digit industry classification and a NUTS-2 region. While the raw data includes 543 local markets, we restrict our focus to those with at least 100 firm-year observations to ensure sufficient power for estimating firm-level permanent productivity and productivity shock volatility. The resulting panel covers 397 local markets and consists of approximately 2.2 million firm-year observations, with around 190,000 firms observed annually. Table B1 in Appendix B presents selected summary statistics for our sample.

productivity shocks with pdf $\phi^j(s)$.

²In English: Integrated Business Accounts System.

³The NUTS-2 regions in Portugal are Norte, Centro, Área Metropolitana de Lisboa, Alentejo, Algarve, Região Autónoma dos Açores, and Região Autónoma da Madeira.

3.2 Volatility and misallocation

We use our data to estimate production functions and recover two key statistics. The first is firm productivity, which we use to calculate measures of capital misallocation at the local market level. The second is the volatility of productivity shocks in each local market. Firm productivity is derived from estimates of permanent unobserved heterogeneity in sales across firms, after controlling for differences in the use of factors of production and aggregate shocks. To estimate the volatility of productivity shocks, we rely on the residuals from the regression of sales on the quantity of inputs used in production, firm dummies, and year dummies.

Our empirical strategy involves estimating production functions separately for each local market. This serves two key purposes. First, it provides greater flexibility, allowing input shares in the production function to vary across local markets (Asker et al., 2014). Second, and most importantly, it enables us to account for heterogeneity in the variance of the error term across local markets. Uncovering this heterogeneity allows us to link volatility to capital misallocation across location-industry pairs.

Thus, for each location-industry pair j , consider the following revenue production function:

$$\log y_{it} = \beta^{0j} + \beta^{1j} \log k_{it} + \beta^{2j} \log \ell_{it} + \beta^{3j} \log m_{it} + \mu_i + \mu_t + \epsilon_{it}, \quad (3)$$

where y_{it} denotes revenues of firm i in year t , k_{it} is the capital stock, ℓ_{it} is the number of employees, m_{it} denotes intermediate inputs, μ_i and μ_t are firm and year fixed effects, respectively, and ϵ_{it} is an error term with mean 0 and variance σ^j .⁴

The estimated firm fixed effects $\hat{\mu}_i$ capture all firm-level unobserved factors that permanently affect revenues. In our framework, they map into firm productivity z_i ; i.e.,

$$\log z_i = \hat{\mu}_i.$$

The estimated residuals $\hat{\epsilon}_{it}$ represent instead a sample of innovations to sales that can-

⁴In equation (3) we include labor and intermediate inputs, in addition to capital, to account for the observed heterogeneity in labor and material usage across firms. In Appendix A.3, we discuss a version of the model where entrepreneurs are allowed to choose multiple factors of production.

not be explained by variations in input usage, firm-level permanent unobserved heterogeneity, or common factors affecting revenues. In our framework, they correspond to realized productivity shocks. These residuals contain information that can be used to estimate idiosyncratic uncertainty. Our preferred approach consists of first estimating the volatility of productivity shocks at the firm level, $\hat{\sigma}_i$, and then averaging these estimates across firms within the same location. Specifically, we begin by computing the within-firm standard deviation of the estimated residuals over time as:

$$\hat{\sigma}_i = \frac{1}{T} \sqrt{\sum_{t=1}^T (\hat{\epsilon}_{it} - \bar{\hat{\epsilon}}_{it})^2},$$

where $\bar{\hat{\epsilon}}_{it} = \frac{1}{T} \sum_{t=1}^T \hat{\epsilon}_{it}$ is the time-average firm-level residual. Next, we compute the within-local-market average standard deviation of the estimated residuals:

$$\hat{\sigma}^j = \sum_{i \in \mathcal{I}^j} \frac{\hat{\sigma}_i}{|\mathcal{I}^j|},$$

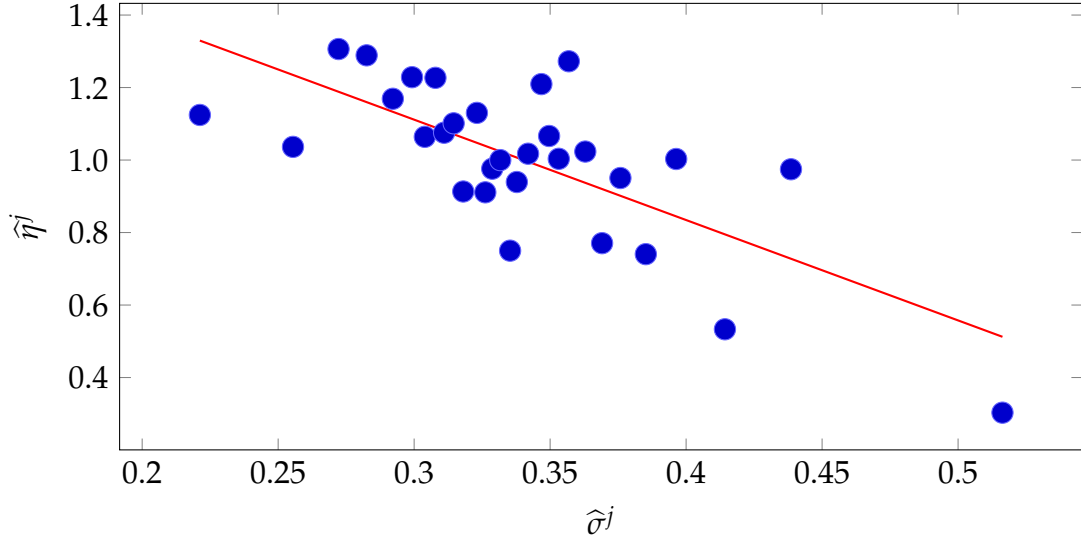
where $\mathcal{I}^j$ denotes the set of firms in local market j . Figure B.2 and Table B3 in Appendix B.1 present the distribution of estimated firm productivity and productivity shock volatility. Firm productivity exhibits substantial dispersion across firms, with the standard deviation of $\hat{\mu}_i$ approximately 0.656. The average standard deviation of productivity shocks is around 0.347.

Next, we explore the relationship between productivity shock volatility and capital misallocation across local markets. To measure capital misallocation within each location-industry pair we compute the correlation between firm productivity and average (log) capital holdings across firms within each local market. Specifically, for each local market j , we estimate the following regression:

$$\overline{\log k}_{it} = v^j + \eta^j \hat{\mu}_i + \varepsilon_i,$$

where $\overline{\log k}_{it} = \frac{1}{T} \sum_{t=1}^T \log k_{it}$. A higher η^j indicates a higher correlation between firm productivity and capital holdings, and hence a lower capital misallocation, in local market j . Figure 2 scatters our measures of capital misallocation on the average firm-

Figure 2: Volatility and capital misallocation across local markets



NOTES:: This figure presents a binscatter of the relationship between productivity shock volatility and capital misallocation across local markets. Capital misallocation is measured as the correlation between firm productivity and the average (log) capital holdings of firms within each local market. Local markets are defined as the combination of a two-digit CAE industry classification and a NUTS-2 region with at least 100 firm-year observations.

level volatility across local markets.⁵ In the figure, each dot represents an average local market within a certain bin in the distribution of volatility.

The patterns that emerge are striking: local markets with higher firm-level volatility are also those with lower correlation between firm-level capital and productivity.⁶

Figure 3 zooms into local markets with different levels of average idiosyncratic volatility. Panel A scatters capital holdings against estimated firm productivity for firms operating in low-volatility markets, while Panel B replicates this analysis for firms in high-volatility markets, where low- (high-) volatility markets are defined as location-

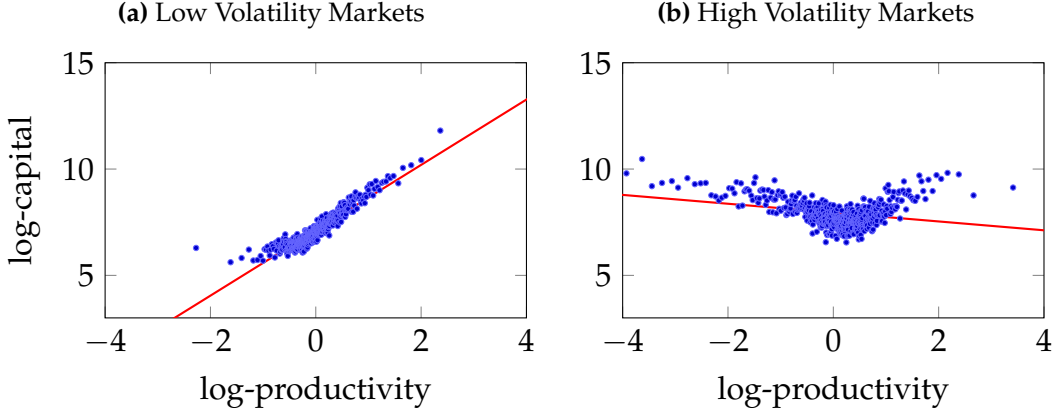
⁵In Table B4 Appendix B, we report point estimates (and standard errors) from regressing our measures of capital misallocation on the average firm-level volatility across local markets, conditional on 2-digit CAE industry and NUTS-2 region fixed effects. That is, we estimate,

$$\hat{\omega}^j = \gamma_0 + \gamma_1 \hat{\sigma}^j + \mu^{s(j)} + \mu^{\ell(j)} + \varepsilon^j,$$

where $\hat{\omega}^j$ is either $\hat{\eta}^j$ or $\hat{\delta}^j$, $\mu^{s(j)}$ are 2-digit CAE industry fixed effects and $\mu^{\ell(j)}$ are NUTS-2 region fixed effects, and ε^j is an error term.

⁶This pattern implies that the average within-firm marginal product of capital is not equalized across firms located in high-volatility markets as it would be the case in efficient allocation. Figure B.3 in Appendix B confirms this is the case in the data. Figure B.4 and Table B5 show that a similar relationship holds for labor misallocation and productivity shock volatility across local markets, though the magnitudes are approximately half as large.

Figure 3: Capital allocation in low- and high-volatility markets



NOTES: This figure presents a binscatter of firm-level (log) capital, $\log k_i$, against the estimated firm-level (log) productivity, $\hat{\mu}_i$, across firms located in low-volatility (Panel A) and high-volatility (Panel B) local markets. Low- (high-) volatility markets are defined as a tuple of 2-digit CAE industry classification and NUTS-2 region with at least 100 firm-year observations and with an estimated average volatility at the bottom (top) 1 percent of the distribution.

industry pairs that fall within the bottom (top) 1% of the estimated distribution of productivity shock volatility. Consistent with our theory, capital holdings and firm productivity show a strong positive correlation of 0.49 in low-volatility markets. This relationship breaks down in high-volatility markets, with capital allocation deviating from efficiency and becoming essentially uncorrelated with productivity, as reflected in a point estimate of -0.09.

3.3 Addressing endogeneity

Shock persistence. In our baseline specification (equation 3), the error terms represent productivity shocks, which our framework assumes to be independently and identically distributed over time. This approach aligns with recent work on the effects of uninsurable idiosyncratic risk on entrepreneurial activity (Di Tella and Hall, 2022; Di Tella et al., 2025). However, we test the robustness of our results to allowing the shocks to be persistent over time. To do it, we extend the production function in equation (3) as follows:

$$\log y_{it} = \beta^0 + \beta^1 \log k_{it} + \beta^2 \log \ell_{it} + \beta^3 \log m_{it} + \mu_i + v_{it},$$

where v_{it} follows an autoregressive process of order one; i.e.,

$$v_{it} = \rho v_{it-1} + \epsilon_{it},$$

$\rho \in (0, 1)$ captures the persistence of v_{it} , and ϵ_{it} is an i.i.d. mean-zero error term. To obtain consistent estimates of β^{1j} , β^{2j} , and β^{3j} , we follow Blundell and Bond (2000) and employ a dynamic panel data approach while treating capital as a predetermined variable. Further details are provided in Appendix B. This estimation strategy yields estimates of firm-level permanent productivity and sales residuals, which we use to construct measures of capital misallocation and productivity shock volatility at the local market level. Tables B6 and B7 report the conditional correlation between productivity shock volatility and measures of misallocation across location-industry pairs. Our results are robust to the inclusion of shock persistence. The relationship between uncertainty and capital misallocation remains significant, with magnitudes comparable to or exceeding those in our baseline specification.

Firm exit. Each year, approximately 6.5 percent of firms exit the sample. Since exiting firms are likely to be non-randomly selected from the population of operating firms — e.g., they tend to be smaller than surviving firms, ignoring this attrition could introduce bias in our estimates of the production function parameters. To address this concern, we consider the following binary selection equation for firm exit:

$$d_{it}^* = \zeta_t \zeta_i + \epsilon_{it}, \quad d_{it} = \mathbf{1}[d_{it}^* > 0] \quad (4)$$

where $\mathbf{1}[\cdot]$ is an indicator function equal to one if its argument is true, and zero otherwise, ζ_i are unobserved individual-specific effects, with a possibly time-varying effect, ζ_t , and ϵ_{it} is an unobserved disturbance term. We jointly estimate equations (3) and (4) separately for every period t following the semi-parametric approach proposed by Wooldridge (1995), and using panel-level average of time-varying covariates in equation (3) to parametrize ζ_i . We report details in Appendix B.

Tables B8 and B9 present the conditional correlation between productivity shock volatility and our measures of misallocation, constructed accounting for self-selection in firm exit. Although accounting for endogenous firm exit slightly reduces the correlation, it

remains large and statistically significant.

4 The aggregate cost of incomplete markets

In this section, we leverage the structure of our model to answer the following question: What is the aggregate cost of incomplete consumption insurance? We begin by specifying a production function in which capital is the only factor of production:

$$f(z, k) = zk^\alpha,$$

where $\alpha \in (0, 1)$.⁷ We assume that productivity z follows a gamma distribution with shape and scale parameters κ and θ , respectively:

$$z \sim \Gamma(\kappa, \theta).$$

Finally, we maintain the assumption that productivity shocks s follow a log-normal distribution. Since our focus is on aggregate outcomes, we abstract from the heterogeneity in capital share α and productivity shock volatility σ across local markets, focusing instead on their average values.

Table 1: Parameters

Parameters	Description	Value	Target/Source
<i>A. Parameters calibrated</i>			
α	Output elasticity of capital	0.416	Aggregate capital share (FRED)
σ	Volatility of shocks	0.347	Data
(κ, θ)	Distribution of permanent productivity	(1.404, 1.060)	Data
<i>B. Parameters estimated</i>			
γ	Relative risk aversion	3.361	corr. $[\log k, \log z] = 0.494$

NOTES: This table reports the values and sources of the calibrated and estimated parameters in the benchmark model.

Table 1 summarizes the model parameters and their respective sources. The output elasticity of capital, α is externally calibrated to match an aggregate capital share of 0.416, constructed as one minus the aggregate labor share for the period 2011-2019 (FRED, 2025). The aggregate volatility of productivity shocks σ is calibrated at 0.347,

⁷Later in the paper, we consider a model that includes labor as a factor of production, allowing entrepreneurs to adjust it after productivity shocks realize.

a value directly estimated from the data. The shape and scale parameters of the firm productivity distribution, κ and θ , are obtained by fitting by maximum likelihood a gamma distribution to the estimated firms' fixed effects and set to 1.404 and 1.060, respectively. Finally, we estimate the coefficient of relative risk aversion, γ , to match the observed correlation of 0.494 between (log) capital holdings and (log) permanent productivity across firms. The estimation yields $\hat{\gamma} = 3.361$, which is close to the estimate of 4.2 found in Chiappori and Paiella (2011).

Equipped with our estimates, we perform two counterfactual exercises. In the first, we eliminate production uncertainty by setting σ to zero; i.e., we simulate the aggregate behavior of a deterministic economy where entrepreneurs face no productivity shocks. In the second, we complete consumption insurance markets while maintaining productivity shock volatility at its baseline level.

Table 2 presents the results of these counterfactual exercises. Transitioning from the baseline to a deterministic economy removes the distortions in capital allocation caused by production uncertainty under incomplete markets. As a consequence, the correlation between firm productivity and capital holdings increases, leading to greater dispersion of capital holdings across firms. This reallocation enhances aggregate productivity, resulting in aggregate income gains of 20.3%.

A counterfactual economy with complete markets exhibits the same capital allocation across firms as the economy without production uncertainty, but achieves higher income gains, reaching 27.4%, respectively. This difference arises because eliminating production uncertainty, that is, setting σ to zero, affects higher moments of the productivity distribution as well as the expected value of the productivity shock. In contrast, the availability of state-contingent claims when markets are complete enables entrepreneurs to fully insure against consumption uncertainty, effectively eliminating the distortionary effects of the higher moments of the productivity distribution without reducing the expected value of productivity shocks. Intuitively, completing insurance markets eliminates misallocation without reducing the likelihood of favorable productivity shock realizations.⁸

⁸Effectively, the gains from completing insurance markets could be decomposed into the gains from efficiently allocating capital to entrepreneurs and those from allowing them to fully exploit the prospect of favorable shock realizations. In other words, completing insurance markets generates aggregate

Table 2: The aggregate cost of incomplete markets

	Counterfactual scenarios		
	Baseline	No Uncertainty	Complete markets
	(1)	(2)	(3)
Volatility, σ	0.347	0	0.347
Average distortion, $\overline{\tau(z, k)}$	0.458	0	0
corr[log z , log k]	0.485	1	1
η [log z , log k]	0.827	1.712	1.712
sd[log k]	0.776	1.594	1.594
Aggregate Income	1	1.203	1.274

NOTES: This table reports selected outcomes for the baseline economy (column 1), a counterfactual economy with no idiosyncratic production uncertainty (column 2), and a counterfactual economy with complete insurance markets (column 3).

4.1 Accounting for other sources of misallocation

The correlation between firm-level productivity and capital holding is likely to be affected by other sources of misallocation, such as size-dependent markups, regulations or corporate taxes. In Appendix C, we test the robustness of our results to the inclusion of other distortions.

To do so, we introduce generic firm-specific distortions in the form of wedges between the marginal product and the price of capital that are correlated with firm productivity. Specifically, as in Hsieh and Klenow (2009), we assume that each firm i is subject to a tax on capital expenditures, τ_i .⁹ Moreover, as in Bento and Restuccia (2017), we assume the tax is equal to

$$\tau_i = 1 - z_i^{-\varphi},$$

where $\varphi \geq 0$ governs the correlation between tax and productivity.

To quantify the gains from full insurance, we proceed by fixing the coefficient of rela-

gains that extend beyond pure allocative efficiency.

⁹In Appendix A.5 we show that there is a direct isomorphism between our framework extended to include firm-specific distortions in the form of taxes, and the exact same economy with financial frictions. See also Liu (2019) and Bigio and La'o (2020) for a similar argument.

tive risk aversion, γ , to either 1.5 or 3, and calibrate φ in both scenarios to match the correlation between (log) productivity and (log) capital holdings. We obtain values φ of 0.295 and 0.195, respectively, implying a distortion tax of up to 50% for the most productive firms, and a distortion subsidy of up to 100% for the least productive ones (Figure C.1).

Table C2 presents the counterfactual outcomes for both cases. The gains from completing the insurance markets remain large when other distortions are factored in: aggregate income increases 6.4% when $\gamma = 1.5$, and by 23.7% when $\gamma = 3$, accounting for 46 and 98% of the overall gains achievable from moving to a fully-undistorted economy.

4.2 Factor shares

In Appendix C we also assess the robustness of our findings to different values of the output elasticity of capital, α . Table C3, we report the percentage gains in aggregate productivity and income from eliminating production uncertainty or completing consumption insurance markets while varying α from 0.5 to 0.9. In each counterfactual economy, all parameters remain as in the baseline, except for the relative risk aversion coefficient, γ , which is re-calibrated to match the observed correlation between firm productivity and capital holdings. We find that increasing the output elasticity of capital raises the estimated RRA coefficient, from 3.51 when $\alpha = 0.5$ to 4.72 when $\alpha = 0.9$. This occurs because, holding all else constant, a higher α increases the expected marginal product of capital relatively more for high-productivity entrepreneurs. Consequently, a higher value of γ is required to match the correlation between firm productivity and capital holdings to the targeted moment. As before, the counterfactual gains in aggregate productivity from completing insurance markets consistently exceed those from eliminating production uncertainty. Moreover, higher values of the output elasticity of capital α amplify both types of gains. Eliminating production uncertainty increases aggregate output from 30% when $\alpha = 0.5$ up to 252% when $\alpha = 0.9$. Similarly, the gains from completing markets range from 39% when $\alpha = 0.5$ to 270% when $\alpha = 0.9$.

4.3 Technology

Finally, we assess the robustness of our findings to incorporating labor as a factor of production and allowing firms to adjust it after productivity shocks realize. To do so, we solve and estimate the model introduced in Section C.3. Table C4 reports the values of the calibrated parameters. We set the span of control parameter, η , to 0.9 as in (Erosa et al., 2023), and calibrate the output elasticity of labor, α , to 0.65 to match the aggregate labor share FRED (2025). As in the benchmark model, we estimate the relative risk aversion coefficient to match the observed correlation between firm productivity and capital holdings, obtaining a value of 5.111.

Table C5 presents the counterfactual outcomes. The gains from completing markets significantly amplify when firms produce using capital and labor: aggregate income increases by 46%, compared to 20% in the benchmark model. Although chosen ex-post, market incompleteness propagates to labor demand, introducing a further source of misallocation.

5 Conclusion

We study the allocation of capital to risk-averse entrepreneurs with heterogeneous productivity in an economy where insurance markets are incomplete and output requires time to build. We prove that insurance markets' incompleteness acts as a correlated distortion, reallocating capital from more productive to less productive entrepreneurs relative to the complete markets benchmark. Using Portuguese administrative data, we show that capital misallocation is greater in locations and industries where output is subject to higher idiosyncratic production uncertainty, consistent with our framework. By closing the model in general equilibrium, we quantify the misallocation cost of incomplete markets in general equilibrium. Our findings suggest that completing insurance markets would result in substantial aggregate efficiency gains.

Our results speak to the link between insurance markets' incompleteness and the aggregate costs of resource misallocation. In doing so, our paper introduces a complementary channel for misallocation, which can operate alongside the well-known

drivers examined in the classic misallocation literature, such as policy distortions and factor market frictions.

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Online appendix (not intended for publication)

A Theory

A.1 Definition of competitive equilibrium

Complete markets. A competitive equilibrium for an economy with complete markets is a list of capital choices $\{k_i^*\}_{i \in \mathcal{I}}$, state-contingent claims $\{\theta_i^*(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$, consumption plans $\{c_i^*(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$, a rental price of capital r^* , and a list of prices for contingent claims $\{q^*(s)\}_{s \in \mathcal{S}}$ such that

- $\{k_i^*\}_{i \in \mathcal{I}}$, $\{\theta_i^*(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$ and $\{c_i^*(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$ are the solution to the problem of each entrepreneur i ;
- the price of contingent claims is fair; i.e., $q^*(s) = \phi(s)$;
- the market for capital clears; i.e.,

$$\int_i k_i^* di = \bar{K}.$$

Incomplete markets. A competitive equilibrium for an economy with incomplete markets is a list of capital choices $\{k_i^0\}_{i \in \mathcal{I}}$, consumption plans $\{c_i^0(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$, and a rental price of capital r^0 such that

- $\{k_i^0\}_{i \in \mathcal{I}}$ and $\{c_i^0(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$ solve the problem of the entrepreneurs;
- the market for capital clears; i.e.,

$$\int_i k_i^0 di = \bar{K}.$$

A.2 Proofs

Proof of Proposition 1. Recall that an equilibrium capital allocation is such that the marginal product of capital is constant across firms and equal to the rental price of capital; i.e.,

$$\int_{s \in \mathcal{S}} [sf_k(z_i, k_i^*) - r^*] \phi(s) ds = 0, \quad \forall i \in \mathcal{I}.$$

Next, consider the problem of a planner that allocates capital to entrepreneurs to maximize aggregate expected output subject to an aggregate feasibility constraint. The problem can be stated as follows:

$$\begin{aligned} \max_{\{k_i\}_{i \in \mathcal{I}}} \quad & \int_{s \in \mathcal{S}} sf(z_i, k_i) \phi(s) ds \\ \text{s.t.} \quad & \sum_i k_i di \leq \bar{K}, \end{aligned}$$

where $\bar{K}$ is the economy-wide capital stock. A solution to this problem is characterized by the following first-order conditions:

$$\int_{s \in \mathcal{S}} sf_k(z_i, k_i^*) \phi(s) ds - \lambda = 0, \quad \forall i \in \mathcal{I},$$

where λ is the Lagrange multiplier attached to the feasibility constraint. That is, to maximize output, the planner allocates capital so that each entrepreneur's expected marginal product equals its shadow price, λ . $\square$

Proof of Proposition 3. Let $\{k_i^*\}_{i \in \mathcal{I}}$ and $\{k_i^o\}_{i \in \mathcal{I}}$ be the equilibrium capital allocations under complete and incomplete markets. Let us express both allocations as policy functions mapping entrepreneurial productivity to equilibrium capital choices; i.e., $k^*(z)$ and $k^o(z)$, respectively.

To prove that market incompleteness results in the reallocation of capital from more to less productive firms, we show that

- there exists a unique value for the entrepreneurial productivity such that the optimal capital chosen under complete markets equates the one under incom-

plete markets; i.e.,

$$\exists! \tilde{z} \in \mathcal{Z} \quad \text{s.t.} \quad k^*(\tilde{z}) = k^o(\tilde{z});$$

- at the crossing point, $\tilde{z}$, the slope of the policy function for capital under complete markets is larger than the slope of the policy function under incomplete markets; i.e.,

$$\left. \frac{\partial k^*(z)}{\partial z} \right|_{z=\tilde{z}} > \left. \frac{\partial k^o(z)}{\partial z} \right|_{z=\tilde{z}}.$$

We start by proving the second bullet point. Let $\tilde{z}$ be a point in $\mathcal{Z}$ such that $k^*(\tilde{z}) = k^o(\tilde{z}) = \tilde{k}$. We show later that this point exists and is unique.

Let $H(z, k) = \int_{s \in \mathcal{S}} s f_k(z, k) \Gamma(s) ds - r$. By the implicit function theorem, under complete market it must be the case that:

$$\left. \frac{\partial k^*(z)}{\partial z} \right|_{z=\tilde{z}} = - \frac{\left. \frac{\partial H(z, k^*(z))}{\partial z} \right|_{z=\tilde{z}}}{\left. \frac{\partial H(z, k^*(z))}{\partial k} \right|_{z=\tilde{z}}} = - \frac{f_{kz}(\tilde{z}, k^*(\tilde{z}))}{f_{kk}(\tilde{z}, k^*(\tilde{z}))} = - \frac{f_{kz}(\tilde{z}, \tilde{k})}{f_{kk}(\tilde{z}, \tilde{k})} > 0$$

Let $G(z, k) = f_k(z, k)[\mu_s - (\gamma - 1)\sigma_s^2 f(z, k)] - r$. By the implicit function theorem, under no insurance it must be the case that:

$$\begin{aligned} \left. \frac{\partial k^o(z)}{\partial z} \right|_{z=\tilde{z}} &= - \frac{\left. \frac{\partial G(z, k^o(z))}{\partial z} \right|_{z=\tilde{z}}}{\left. \frac{\partial G(z, k^o(z))}{\partial k} \right|_{z=\tilde{z}}} = - \frac{f_{kz}(\tilde{z}, k^o(\tilde{z}))[\mu_s - (\gamma - 1)\sigma_s^2 f(\tilde{z}, k^o(\tilde{z}))] - f_k(\tilde{z}, k^o(\tilde{z}))[(\gamma - 1)\sigma_s^2 f_z(\tilde{z}, k^o(\tilde{z}))]}{f_{kk}(\tilde{z}, k^o(\tilde{z}))[\mu_s - (\gamma - 1)\sigma_s^2 f(\tilde{z}, k^o(\tilde{z}))] - f_k(\tilde{z}, k^o(\tilde{z}))[(\gamma - 1)\sigma_s^2 f_k(\tilde{z}, k^o(\tilde{z}))]} \\ &= - \frac{\mu_s f_{kz}(\tilde{z}, \tilde{k}) - (\gamma - 1)\sigma_s^2 f(\tilde{z}, \tilde{k}) f_{kz}(\tilde{z}, \tilde{k}) - (\gamma - 1)\sigma_s^2 f_z(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})}{\mu_s f_{kk}(\tilde{z}, \tilde{k}) - (\gamma - 1)\sigma_s^2 f(\tilde{z}, \tilde{k}) f_{kk}(\tilde{z}, \tilde{k}) - (\gamma - 1)\sigma_s^2 f_k(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})}. \end{aligned}$$

We need to prove that:

$$\left. \frac{\partial k^*(z)}{\partial z} \right|_{z=\tilde{z}} > \left. \frac{\partial k^o(z)}{\partial z} \right|_{z=\tilde{z}}$$

or, equivalently,

$$-\frac{f_{kz}(\tilde{z}, \tilde{k})}{f_{kk}(\tilde{z}, \tilde{k})} > -\frac{\mu_s f_{kz}(\tilde{z}, \tilde{k}) - (\gamma - 1)\sigma_s^2 [f(\tilde{z}, \tilde{k}) f_{kz}(\tilde{z}, \tilde{k}) + f_z(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})]}{\mu_s f_{kk}(\tilde{z}, \tilde{k}) - (\gamma - 1)\sigma_s^2 [f(\tilde{z}, \tilde{k}) f_{kk}(\tilde{z}, \tilde{k}) + f_k(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})]}$$

This condition is equivalent to:

$$f_{kz}(\tilde{z}, \tilde{k}) \left[\mu_s f_{kk}(\tilde{z}, \tilde{k}) - (\gamma - 1) \sigma_s^2 [f(\tilde{z}, \tilde{k}) f_{kk}(\tilde{z}, \tilde{k}) + f_k(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})] \right] < \\ f_{kk}(\tilde{z}, \tilde{k}) \left[\mu_s f_{kz}(\tilde{z}, \tilde{k}) - (\gamma - 1) \sigma_s^2 [f(\tilde{z}, \tilde{k}) f_{kz}(\tilde{z}, \tilde{k}) + f_z(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})] \right]$$

which can be re-arranged as follows:

$$f_{kz}(\tilde{z}, \tilde{k}) \mu_s f_{kk}(\tilde{z}, \tilde{k}) - (\gamma - 1) \sigma_s^2 f_{kz}(\tilde{z}, \tilde{k}) [f(\tilde{z}, \tilde{k}) f_{kk}(\tilde{z}, \tilde{k}) + f_k(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})] < \\ f_{kk}(\tilde{z}, \tilde{k}) \mu_s f_{kz}(\tilde{z}, \tilde{k}) - (\gamma - 1) \sigma_s^2 f_{kk}(\tilde{z}, \tilde{k}) [f(\tilde{z}, \tilde{k}) f_{kz}(\tilde{z}, \tilde{k}) + f_z(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})]$$

Simplifying terms, we obtain:

$$-(\gamma - 1) \sigma_s^2 f_{kz}(\tilde{z}, \tilde{k}) [f(\tilde{z}, \tilde{k}) f_{kk}(\tilde{z}, \tilde{k}) + f_k(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})] < \\ -(\gamma - 1) \sigma_s^2 f_{kk}(\tilde{z}, \tilde{k}) [f(\tilde{z}, \tilde{k}) f_{kz}(\tilde{z}, \tilde{k}) + f_z(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})]$$

Because $\gamma > 1$ and $\sigma_s^2 > 0$, then we can further simplify the condition above as follows:

$$f_{kz}(\tilde{z}, \tilde{k}) [f(\tilde{z}, \tilde{k}) f_{kk}(\tilde{z}, \tilde{k}) + f_k(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})] > f_{kk}(\tilde{z}, \tilde{k}) [f(\tilde{z}, \tilde{k}) f_{kz}(\tilde{z}, \tilde{k}) + f_z(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})]$$

This condition is true if and only if

$$\underbrace{f_{kz}(\tilde{z}, \tilde{k}) f_k(\tilde{z}, \tilde{k})}_{>0} > \underbrace{f_{kk}(\tilde{z}, \tilde{k})}_{<0} \underbrace{f_z(\tilde{z}, \tilde{k})}_{>0}$$

which is always the case because $f_{kz}(z, k) > 0$, $f_k(z, k) > 0$, $f_z(z, k) > 0$ and $f_{kk}(z, k) < 0$, $\forall z$ and $\forall k$. This completes the first part of the proof.

We now move to prove existence and uniqueness of a crossing point between the capital policy function under complete market and the one under incomplete markets.

Let us start by assuming there were no crossing points. Without loss of generality, let

us assume that $k^*(z) > k^o(z), \forall z \in \mathcal{Z}$. If this is the case, then it must be true that:

$$\sum_{i \in \mathcal{I}} k_i^* di > \sum_{i \in \mathcal{I}} k_i^o di$$

On the other hand, in equilibrium the aggregate demand for capital under both scenarios (i.e., complete and incomplete markets) has to equal to an exogenously given supply $\bar{K}$; i.e.,

$$\sum_{i \in \mathcal{I}} k_i^* di = \sum_{i \in \mathcal{I}} k_i^o di = \bar{K}$$

which contradicts the previous condition. By contradiction, this implies that there must exist at least one point $\tilde{z}$ such that $k^*(\tilde{z}) = k^o(\tilde{z})$.

The crossing point must also be unique. Suppose not; i.e., $\exists \tilde{z}_1, \dots, \tilde{z}_n \in \mathcal{Z}$ such that $k^*(\tilde{z}_j) = k^o(\tilde{z}_j), \forall j = 1, \dots, n$. Following what we proved in the first part of the proof, it must be the case that

$$\left. \frac{\partial k^*(z)}{\partial z} \right|_{z=\tilde{z}_j} > \left. \frac{\partial k^o(z)}{\partial z} \right|_{z=\tilde{z}_j} \quad \forall j = 1, \dots, n.$$

On the other hand, because both policy functions $k^*(z)$ and $k^o(z)$ are continuous in $\mathcal{Z}$, and because $k^*(z)$ is monotonically increasing in z —i.e., $\frac{\partial k^*(z)}{\partial z} > 0$ —the only possibility where two (or more) crossing points exist is a scenario where

$$\left. \frac{\partial k^*(z)}{\partial z} \right|_{z=\tilde{z}_{\hat{j}}} \leq \left. \frac{\partial k^o(z)}{\partial z} \right|_{z=\tilde{z}_{\hat{j}}}$$

for some $\hat{j}$, contradicting what we proved in the first part of the proof. This completes the second part of the proof.

Proof of Proposition 4. Consider a decentralized equilibrium under incomplete markets. We need to prove that a higher volatility of productivity shocks σ_s^2 reduces the correlation between capital holdings k^o and productivity z . To do so, notice that the derivative of $\partial k^o / (\partial z)^{-1}$ with respect to σ^2 is negative if and only if

$$\left(\frac{\partial \mu_s}{\partial \sigma^2} \sigma_s^2 - \mu_s \frac{\partial \sigma_s^2}{\partial \sigma^2} \right) (\gamma - 1) f_k(z, k^o) [f_{kz}(z, k^o) f_k(z, k^o) - f_{kk}(z, k^o) f_z(z, k^o)] < 0.$$

Since $f_{kk}(z, k^0) < 0$, $f_k(z, k^0) > 0$, $f_{kz}(z, k^0) > 0$ and $f_z(z, k^0) > 0$,

$$f_k(z, k^0) [f_{kz}(z, k^0) f_k(z, k^0) - f_{kk}(z, k^0) f_z(z, k^0)] > 0.$$

Thus, if $\gamma > 1$, the derivative of $(\partial z)^{-1} / \partial k^0$ with respect to σ^2 is negative if and only if

$$\left(\frac{\partial \mu_s}{\partial \sigma^2} \sigma_s^2 - \mu_s \frac{\partial \sigma_s^2}{\partial \sigma^2} \right) < 0. \quad (5)$$

Recall that $s \sim \log \mathcal{N}(\underline{s}, \sigma^2)$; thus,

$$\mu_s = \exp \left(\underline{s} + \frac{\sigma^2}{2} \right)$$

and

$$\sigma_s^2 = (\exp(\sigma^2) - 1) \exp(2\underline{s} + \sigma^2).$$

Substituting these expressions into equation (5), we have that

$$\begin{aligned} & \left(\frac{\partial \left(\exp \left(\underline{s} + \frac{\sigma^2}{2} \right) \right)}{\partial \sigma^2} [(\exp(\sigma^2) - 1) \exp(2\underline{s} + \sigma^2)] \right. \\ & \quad \left. - \exp \left(\underline{s} + \frac{\sigma^2}{2} \right) \frac{\partial [(\exp(\sigma^2) - 1) \exp(2\underline{s} + \sigma^2)]}{\partial \sigma^2} \right) < 0; \end{aligned}$$

which is always the case because $\sigma^2 > 0$. □

A.3 An extension of the model that allows for other factors of production, monopolistic competition in the product market, and demand shocks

We consider a version of the model outlined in Section 2.2 that allows for other factors of production, monopolistic competition in the product market, and demand shocks. For simplicity, we focus on a model where labor is the only additional input besides capital; however, it is clear from the discussion that there is no conceptual difficulty in including other factors of production as well. Entrepreneurs have access to a pro-

duction technology $f(z, k, \ell)$ that depends on productivity z , capital k , and labor ℓ . As before, output produced is subject to idiosyncratic, multiplicative shocks s distributed according to a probability density function $\phi(s)$. Differently from the previous model, where entrepreneurs produce a homogeneous commodity, each entrepreneur now produces a differentiated variety of final goods or services, and faces a downward-sloping demand curve given by

$$y_i(v, \omega_i, p_i) = \Delta(v\omega_i)^{\eta-1} p_i^{-\eta},$$

where η denotes the price elasticity of demand, Δ is a parameter governing aggregate demand for final goods and services, $\omega_i \in \mathcal{W}$ is a parameter capturing permanent taste heterogeneity for firm-specific products, and $v \in \mathcal{V}$ is an idiosyncratic demand shock, distributed with a pdf $\Lambda(v)$.

Entrepreneurs choose capital before the realization of productivity and taste shocks, but decide how much labor to employ after these shocks are realized. Let w be the wage rate (normalized to 1), r be the rental rate for capital, and $\bar{K}$ be the exogenous supply of capital. For simplicity, we adopt a partial equilibrium framework for the labor market, taking wages as given.¹⁰ Entrepreneur i 's problem can be written as follows:

$$\begin{aligned} \max_{k_i, \{c_i(s, v)\}_{s \in \mathcal{S}}} \quad & \int_{s \in \mathcal{S}} \int_{v \in \mathcal{V}} u(c_i(s, v)) \phi(s) \Lambda(v) ds dv \\ \text{s.t.} \quad & c_i(s, v) \leq \pi(s, v, z_i, \omega_i, k_i) - rk_i + T \quad \forall s \in \mathcal{S}, \forall v \in \mathcal{V}, \end{aligned}$$

where $T = r\bar{K}$ are revenues from renting capital, assumed to be rebated lump-sum to each entrepreneur, while $\pi(s, v, z_i, \omega_i, k_i)$ denotes per-period maximized revenues net of labor costs (henceforth, the profit function); i.e.,

$$\begin{aligned} \pi(s, v, z_i, \omega_i, k_i) = \max_{\ell_i, p_i} \quad & y_i(s, z_i, k_i, \ell_i) p_i - \ell_i \\ \text{s.t.} \quad & y_i(s, z_i, k_i, \ell_i) = sf(z_i, k_i, \ell_i) \\ \text{and} \quad & y_i(v, \omega_i, p_i) = \Delta(v\omega_i)^{\eta-1} p_i^{-\eta}. \end{aligned} \tag{6}$$

¹⁰Alternatively, we may assume that labor is supplied at no utility cost by a unit mass of identical workers.

Substituting price p_i from the demand function into the profit function, we can rewrite the problem in equation (6) as follows:

$$\pi(s, \nu, z_i, \omega_i, k_i) = \max_{\ell_i} \Delta^{1-\zeta} (\nu \omega_i)^\zeta (sf(z_i, k_i, \ell_i))^\zeta - \ell_i,$$

where $\zeta = 1 - 1/\eta$.

For a given Δ , a competitive equilibrium for this economy is a list of capital choices, $\{k_i^o\}_{i \in \mathcal{I}}$, labor choices, $\{\ell_i^o(s, \nu)\}_{s \in \mathcal{S}, \nu \in \mathcal{V}, i \in \mathcal{I}}$, product prices $\{p_i^o(s, \nu)\}_{s \in \mathcal{S}, \nu \in \mathcal{V}, i \in \mathcal{I}}$, consumption plans, $\{c_i^o(s, \nu)\}_{s \in \mathcal{S}, \nu \in \mathcal{V}, i \in \mathcal{I}}$, and a rental price of capital r^o such that

- $\{k_i^o\}_{i \in \mathcal{I}}$, $\{\ell_i^o(s, \nu)\}_{s \in \mathcal{S}, \nu \in \mathcal{V}, i \in \mathcal{I}}$, $\{p_i^o\}_{s \in \mathcal{S}, \nu \in \mathcal{V}, i \in \mathcal{I}}$ and $\{c_i^o(s, \nu)\}_{s \in \mathcal{S}, \nu \in \mathcal{V}, i \in \mathcal{I}}$ solve entrepreneur i 's problem;
- the market for capital clears; i.e.,

$$\int_i k_i^o di = \bar{K}.$$

To characterize the competitive equilibrium, we make the following assumptions:

1. production displays constant return to scale in capital and labor; i.e.,

$$f(z, k, \ell) = zk^{1-\alpha} \ell^\alpha, \quad \alpha \in (0, 1);$$

2. the utility function exhibits constant degree of relative risk aversion; i.e.,

$$u(c) = \frac{c^{1-\gamma}}{1-\gamma}, \quad \gamma > 1;$$

3. the shocks s and ν are two independent log-normal random variables; i.e.,

$$s \sim \log \mathcal{N}(0, \sigma_s^2) \quad \text{and} \quad \nu \sim \log \mathcal{N}(0, \sigma_\nu^2).$$

We start by characterizing the solution to the problem in equation (6). Dropping index i for convenience, and taking the first-order condition with respect to labor, we

obtain

$$\alpha\zeta\Delta^{1-\zeta}(svz\omega k^{1-\alpha}\ell^\alpha)^\zeta = \ell,$$

which implies the following labor demand:

$$\ell = \left(\alpha\zeta\Delta^{1-\zeta}[svz\omega k^{1-\alpha}]^\zeta \right)^{\frac{1}{1-\zeta\alpha}}.$$

Plugging the demand for labor into the profit function, we obtain:

$$\pi(s, v, z, \omega, k) = (1 - \alpha\zeta)\Delta^{1-\zeta}[svz\omega k^{1-\alpha}]^\zeta \left(\alpha\zeta\Delta^{1-\zeta}[svz\omega k^{1-\alpha}]^\zeta \right)^{\frac{\zeta\alpha}{1-\zeta\alpha}},$$

or, equivalently,

$$\pi(s, v, z, \omega, k) = (1 - \alpha\zeta) (\alpha\zeta)^{\frac{\zeta\alpha}{1-\zeta\alpha}} \left(\Delta^{1-\zeta}[svz\omega k^{1-\alpha}]^\zeta \right)^{\frac{1}{1-\zeta\alpha}}.$$

Notice the profit function exhibits decreasing returns to scale in capital; i.e.,

$$\frac{\zeta(1-\alpha)}{1-\zeta\alpha} < 1.$$

This is because $\alpha \in (0, 1)$ and $\eta > 1$, which implies that $\zeta < 1$.

Let

$$\Omega = (1 - \alpha\zeta) (\alpha\zeta)^{\frac{\zeta\alpha}{1-\zeta\alpha}} \Delta^{\frac{1-\zeta}{1-\zeta\alpha}}.$$

Denote by x be the product of firm-idiosyncratic permanent productivity and firm-idiosyncratic permanent taste; i.e., $x = z\omega$. Let χ be the product of firm-idiosyncratic productivity shocks and firm-idiosyncratic taste shocks; i.e., $\chi = sv$. Then, we can rewrite the profit function as follows:

$$\pi(\chi, x, k) = \Omega \left(\chi^\zeta x^\zeta k^{1-\alpha} \right)^{\frac{1}{1-\zeta\alpha}}$$

Because s and v are log-normally distributed, χ is also log-normally distributed; i.e.,

$$\chi \sim \log \mathcal{N}(0, \sigma^2),$$

where $\sigma^2 = \sigma_s^2 + \sigma_v^2$. Given the log-normality of χ , consumption is a three-parameter log-normal random variable with expected value and variance equal to

$$\bar{c} = \Omega \left(x^\zeta k^{\zeta(1-\alpha)} \right)^{\frac{1}{1-\zeta\alpha}} \mathbb{E}[\chi^{\frac{\zeta}{1-\zeta\alpha}}] - rk + T$$

and

$$\sigma_c^2 = \left(\Omega \left(x^\zeta k^{\zeta(1-\alpha)} \right)^{\frac{1}{1-\zeta\alpha}} \right)^2 \text{var}[\chi^{\frac{\zeta}{1-\zeta\alpha}}],$$

respectively (and threshold equal to $-rk + T$). Since $\chi \sim \log \mathcal{N}(0, \sigma^2)$, $\log \chi \sim \mathcal{N}(0, \sigma^2)$.

Thus,

$$\frac{\zeta}{1-\zeta\alpha} \log \chi \sim \mathcal{N} \left(0, \left(\frac{\zeta}{1-\zeta\alpha} \right)^2 \sigma^2 \right),$$

and

$$\chi^{\frac{\zeta}{1-\zeta\alpha}} \sim \log \mathcal{N} \left(0, \left(\frac{\zeta}{1-\zeta\alpha} \right)^2 \sigma^2 \right),$$

which implies that

$$\mathbb{E}[\chi^{\frac{\zeta}{1-\zeta\alpha}}] = \exp \left(\frac{\left(\frac{\zeta}{1-\zeta\alpha} \right)^2 \sigma^2}{2} \right)$$

and

$$\text{var}[\chi^{\frac{\zeta}{1-\zeta\alpha}}] = \left(\exp \left(\left(\frac{\zeta}{1-\zeta\alpha} \right)^2 \sigma^2 \right) - 1 \right) \exp \left(\left(\frac{\zeta}{1-\zeta\alpha} \right)^2 \sigma^2 \right).$$

Therefore, an entrepreneur's problem can be rewritten in terms of consumption equivalent as follows:

$$\max_k \quad \Omega x^{\frac{\zeta}{1-\zeta\alpha}} k^{\frac{\zeta(1-\alpha)}{1-\zeta\alpha}} \mathbb{E}[\chi^{\frac{\zeta}{1-\zeta\alpha}}] - rk + T - \frac{(\gamma-1)}{2} \left(\Omega x^{\frac{\zeta}{1-\zeta\alpha}} k^{\frac{\zeta(1-\alpha)}{1-\zeta\alpha}} \right)^2 \text{var}[\chi^{\frac{\zeta}{1-\zeta\alpha}}].$$

Let $g(x, k) = \Omega x^{\frac{\zeta}{1-\zeta\alpha}} k^{\frac{\zeta(1-\alpha)}{1-\zeta\alpha}}$. Taking the first-order condition with respect to capital, we obtain

$$g_k(x, k^0) \left[\mathbb{E}[\chi^{\frac{\zeta}{1-\zeta\alpha}}] - (\gamma-1) \text{var}[\chi^{\frac{\zeta}{1-\zeta\alpha}}] g(x, k^0) \right] - r^0 = 0,$$

which resembles equation (2) in the main text.

A.4 A two-period model with a risk-free asset

We consider a two-period version of the model in Section 2.2, which allows entrepreneurs to self-insure through borrowing and saving in a risk-free asset. Unlike before, entrepreneurs are born with different initial wealth levels $\omega \in \mathcal{W} \subseteq \mathbb{R}_+$. Moreover, they can transfer resources over time using a risk-free asset a that costs q units of consumption. Entrepreneur i 's problem can be written as follows:

$$\begin{aligned} & \max_{k_i, a_i, c_i, \{c'_i(s)\}_{s \in \mathcal{S}}} u(c_i) + \int_{s \in \mathcal{S}} u(c'_i(s)) \phi(s) ds \\ \text{s.t. } & c_i \leq \omega_i - qa_i \\ & \text{and } c'_i(s) \leq sf(z_i, k_i) - rk_i + a_i + T, \quad \forall s \in \mathcal{S}. \end{aligned}$$

A competitive equilibrium for this economy is a list of capital choices, $\{k_i^0\}_{i \in \mathcal{I}}$, asset choices $\{a_i^0\}_{i \in \mathcal{I}}$, consumption plans, $\{c_i^0(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$, a rental price of capital r^0 , and a price of the asset q^0 such that:

- $\{k_i^0\}_{i \in \mathcal{I}}$, $\{a_i^0\}_{i \in \mathcal{I}}$ and $\{c_i^0(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$ solve entrepreneur i 's problem;
- the market for capital clears; i.e.,

$$\int_i k_i^0 di = \bar{K}.$$

- the asset market clears; i.e.,

$$\int_i a_i^0 di = 0.$$

Drop index i for convenience. Notice that, given the log-normality of s , consumption is a three-parameter log-normal random variable with expected value and variance equal to

$$\bar{c}' = \mu_s f(z, k) - rk + a + T$$

and

$$\sigma_c^2 = f(z, k)^2 \sigma_s^2,$$

respectively (and threshold equal to $-rk + a + T$). Therefore, an entrepreneur's problem can be rewritten in terms of consumption equivalent as follows:

$$\max_{k, a} u(\omega - qa) + u\left(\bar{c}' - \frac{\gamma - 1}{2} \sigma_{c'}^2\right).$$

Taking the first-order conditions with respect to k and a , we obtain

$$0 = u_c \left(\mu_s f(z, k^o) - r^o k^o + a^o + T - \frac{\gamma - 1}{2} f(z, k^o)^2 \sigma_s^2 \right) \left(f_k(z, k^o) [\mu_s - (\gamma - 1) f(z, k^o)^2 \sigma_s^2] - r^o \right),$$

and

$$q^o u_c(\omega - q^o a^o) = u_c \left(\mu_s f(z, k^o) - r^o k^o + a^o + T - \frac{\gamma - 1}{2} f(z, k^o)^2 \sigma_s^2 \right),$$

Because $u_c > 0$, the first order conditions can be rewritten as follows:

$$f_k(z, k^o) \left[\mu_s - \frac{\gamma - 1}{2} f(z, k^o)^2 \sigma_s^2 \right] = r^o$$

and

$$[1 + (q^o)^{(1-\frac{1}{\gamma})}] a^o = (q^o)^{-\frac{1}{\gamma}} \omega - \left(\mu_s f(z, k^o) - r^o k^o + T - \frac{\gamma - 1}{2} f(z, k^o)^2 \sigma_s^2 \right).$$

Notice that the first-order condition for capital allocation mirrors the one derived in a Section 2.2. While the presence of a risk-free asset enables entrepreneurs to smooth consumption over time, it is not sufficient to complete insurance markets, leaving second-period utilities uncertain.

A.5 A model with financial frictions

We consider a version of the model in Section 2.2, extended to account for financial frictions. Unlike before, the amount of capital that agents can invest is constrained by their production capability. That is, we assume that capital expenditure, rk , must be

less than or equal to a multiple $\iota \geq 1$, of expected firm-level production (Bigio and La’o, 2020). Entrepreneur i ’s problem can be written as follows:

$$\begin{aligned} & \max_{k_i, \{c_i(s)\}_{s \in \mathcal{S}}} \int_{s \in \mathcal{S}} u(c_i(s)) \phi(s) ds \\ & \text{s.t. } c_i(s) \leq s f(z_i, k_i) - r k_i + T, \quad \forall s \in \mathcal{S} \\ & \text{and } r k_i \leq \iota \int_{s \in \mathcal{S}} s f(z_i, k_i) \phi(s) ds \end{aligned}$$

where, again, $T = r\bar{K}$ are revenues from renting capital, assumed to be rebated lump sum to each entrepreneur.

A competitive equilibrium for an economy with incomplete markets and financial frictions is a list of capital choices $\{k_i^o\}_{i \in \mathcal{I}}$, consumption plans $\{c_i^o(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$, and a rental price of capital r^o such that

- $\{k_i^o\}_{i \in \mathcal{I}}$ and $\{c_i^o(s)\}_{s \in \mathcal{S}, i \in \mathcal{I}}$ solve the problem of the entrepreneurs;
- the market for capital clears; i.e.,

$$\int_i k_i^o di = \bar{K}.$$

To characterize the solution to this problem, let us drop subscript i . Given the log-normality of productivity shocks s , consumption follows a three-parameter log-normal distribution, with an expected value of $\bar{c} = \mu_s f(z, k) - rk + T$, variance $\sigma_c^2 = f(z, k)^2 \sigma_s^2$, and a threshold of $-rk + T$. Therefore, an entrepreneur’s problem can be rewritten in terms of consumption equivalent as follows:

$$\max_k \quad \mu_s f(z, k) - rk + T - \frac{(\gamma - 1)}{2} f(z, k)^2 \sigma_s^2 \quad \text{s.t.} \quad rk \leq \iota \int_{s \in \mathcal{S}} s f(z, k) \phi(s) ds.$$

If the financial constraint is binding, we have that:

$$r^o k = \iota \mu_s f(z, k)$$

whereas, if the financial constraint is not binding, firms' optimality implies:

$$r^o = f_k(z, k)[\mu_s - (\gamma - 1)\sigma_s^2 f(z, k)]$$

Let $f(z, k)$ be homogeneous of degree $\alpha \in (0, 1]$. Then, we can re-write the last equation as:

$$r^o k = \alpha f(z, k)[\mu_s - (\gamma - 1)\sigma_s^2 f(z, k)].$$

Therefore, it follows that:

$$r^o = f_k(z, k) \frac{\psi(z, k)}{\alpha}$$

where $\psi(z, k) = \min \{ \mu_s, \alpha[\mu_s - (\gamma - 1)\sigma_s^2 f(z, k)] \}$, represents the firm-specific wedge that arises between the firms' marginal cost and marginal revenue.

In this version of the model, there are two potential sources of misallocation. First, when binding, financial frictions generate a wedge between the marginal product of capital and its rental rate. By preventing entrepreneurs to reach their utility maximizing scale, they distort capital allocation across firms. Second, incomplete markets coupled with time-to-build constraint amplify the existing wedge, i) if entrepreneurs are enough risk averse, i.e., if $\gamma > 1$; and 2) the larger the volatility of the idiosyncratic productivity shock, σ_s^2 .

B Data

B.1 Summary statistics

Table B1 presents selected summary statistics for our sample. On average, firms are 15 years old and employ about 12 workers. Their average annual sales amount to roughly 1 million euros, corresponding to an average of 56 thousand euros per employee. The value of their fixed assets is approximately twice this amount. On average, firms have an annual payroll of about 230 thousand euros, implying an average salary of 14 thousand euros per employee. The vast majority of these firms are small and privately owned.

Table B1: Summary Statistics

	Mean (1)	S.d. (2)	p10 (3)	Median (4)	p90 (5)	Observations (6)
Age	15.3	13.3	2	12	32	2195287
N. of Employees	12.5	123.1	1	4	18	2195287
Fixed Assets	2188449.6	67542046.5	23074.4	168697.4	1772260.8	2195287
Sales	1032526.8	22591464.4	21683.9	115972	981823.7	2195287
Sales per employee	56227.1	780720.9	10878.5	30355.9	92847.9	2195287
Material expenditure	9462.3	194044.9	190	1816	13597	2195287
Payroll	233563.6	2615366.9	10041.9	44258.2	307049.9	2195287
Wage per employee	14031.1	12311.8	6941.6	11667.1	22786.5	2195287
Value added	1023064.5	22561743.2	20782.8	113089.6	966400.6	2195287
Value added per employee	55256.6	780490.3	10479.2	29600.5	91097.4	2195287

NOTES: Nominal variables are deflated using the national CPI and expressed in 2010 price levels. The number of employees includes both full-time and part-time workers. Fixed assets encompass both current and non-current assets.

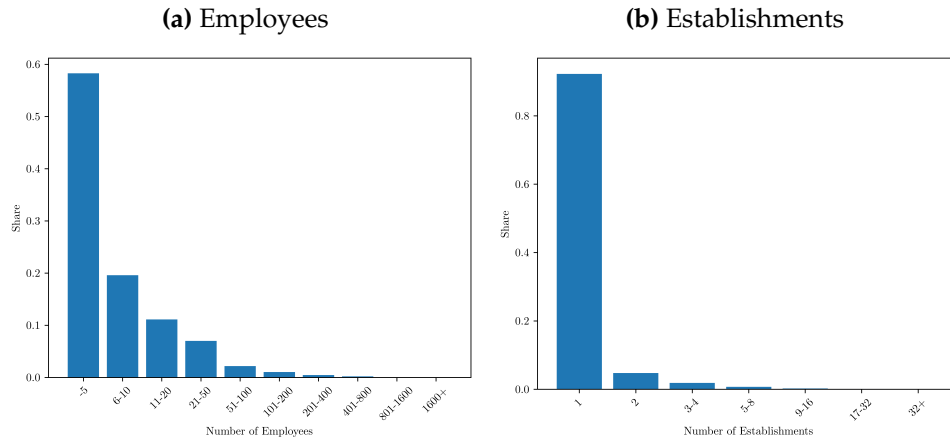
Table B2 shows the share of firms by type of ownership. Over 90% of firms are held privately help (“Sociedade por quotas”).

Table B2: Firm distribution by ownership

Type	Share
Sociedade por quotas	0.642
Sociedade unipessoal por quotas	0.289
Sociedade anonima	0.052
Other	0.016

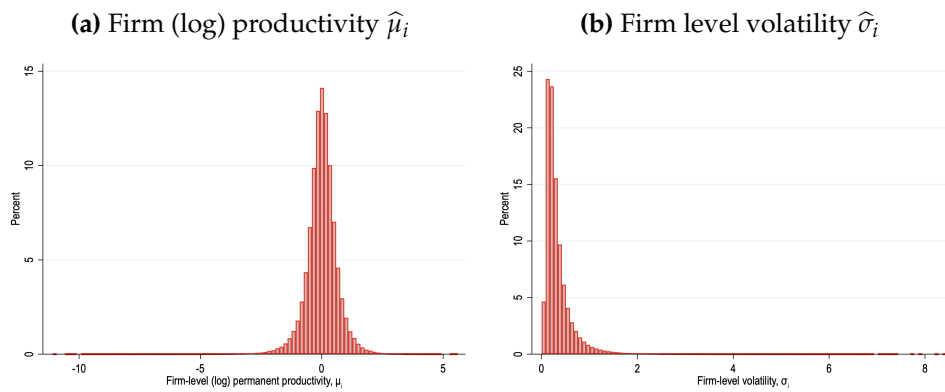
Figure B.1 shows the distributions of firm size and establishments per firm. Approximately 80% of firms have 10 or fewer employees and that 90% operate a single establishment.

Figure B.1: Firm size distributions



NOTES: This figure reports the empirical distribution of firm size by employees (Panel A), and number of establishments (Panel B).

Figure B.2: Production function estimates - Distributions



NOTES: This figure reports the empirical distribution of firm productivity estimates $\hat{\mu}_i$ (Panel A), and the estimate volatility of productivity shocks at the firm level $\hat{\sigma}_i$ (Panel B) across firms in the sample.

Figure B.2 shows the distribution of firm productivity μ_i and firm-level volatility of productivity shocks σ_i , while Table B3 reports selected statistics for both variables. (Log) permanent productivity is largely dispersed across firms: the standard deviation is 65.6%. productivity shocks are volatile: the average firm-level volatility in the sample is about 34.7%.

Table B3: Production function estimates—summary statistics

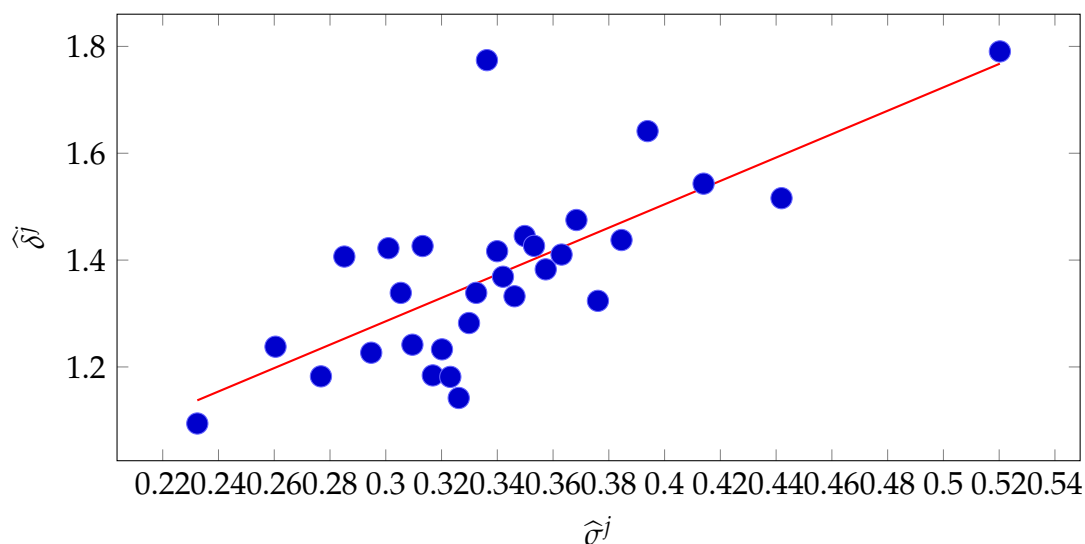
	Mean (1)	S.d. (2)	p10 (3)	Median (4)	p90 (5)	Observations (6)
Firm-level (log) permanent productivity μ_i	0	0.656	-0.691	0.011	0.706	2,170,852
Firm-level volatility σ_i	0.347	0.327	0.108	0.243	0.646	2,173,951

NOTES: Nominal variables used in the production function estimation are deflated using the national CPI and expressed in 2010 price levels. The number of employees includes both full-time and part-time workers. Fixed assets encompass both current and non-current assets.

B.2 Capital misallocation

Figure B.3 scatters the standard deviation of the average within-firm marginal product of capital across firms against the volatility of productivity shocks across local markets.

Figure B.3: Volatility and capital misallocation across local markets



NOTES:: This figure presents a binscatter of the relationship between productivity shock volatility and capital misallocation across local markets. Capital misallocation is measured as the standard deviation of the average within-firm marginal product of capital across firms in each local market. Local markets are defined as the combination of a two-digit CAE industry classification and a NUTS-2 region with at least 100 firm-year observations.

Finally, Table B4 presents the estimated correlation between productivity shock volatility and measures of capital misallocation, controlling for 2-digit CAE industry and NUTS-2 region fixed effects.

Based on estimates from columns (2) and (4), transitioning from local markets with an average volatility of 20% to those with twice the volatility reduces the correlation between firm productivity and employee count by 0.51, while increasing the dispersion of log marginal product of labor by 0.33.

Table B4: Productivity shock volatility and capital misallocation across local markets

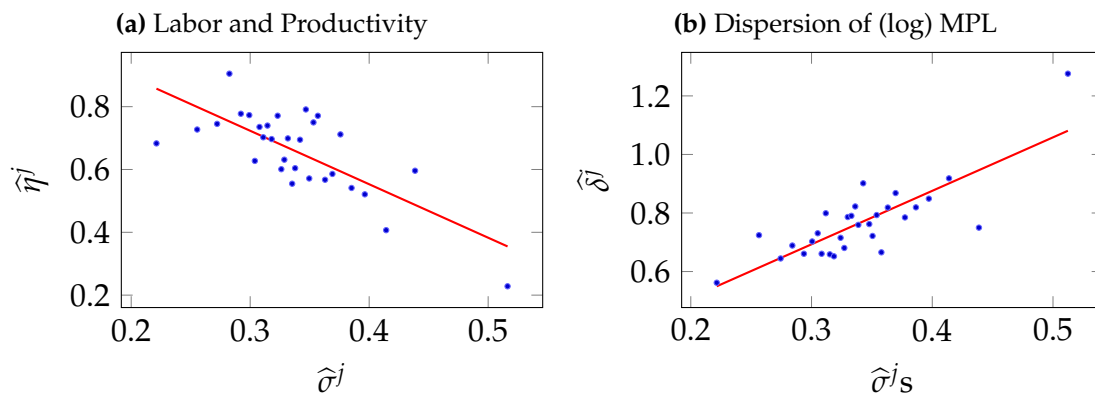
	$\hat{\eta}^j$		$\hat{\delta}^j$	
	(1)	(2)	(3)	(4)
$\hat{\sigma}^j$	-3.999*** (0.677)	-2.530** (0.730)	2.096*** (0.435)	1.649*** (0.134)
R^2	0.433	0.809	0.534	0.902
Observations	397	397	397	397
Weighted	No	Yes	No	Yes

NOTES: This table reports OLS estimates from regressions of our two measures of capital misallocation on productivity shock volatility across local markets. In Columns 1 and 2, capital misallocation is measured as the correlation between firm productivity and the average (log) capital holdings of firms within each local market. In Columns 3 and 4, capital misallocation is measured as the standard deviation of the average within-firm marginal product of capital within each local market. Local markets are defined as the combination of a two-digit CAE industry classification and a NUTS-2 region with at least 100 firm-year observations. All regressions include location and industry fixed effects. In Columns (2) and (4), observations are weighted by the number of firms in their local markets. Standard errors (in parenthesis) are clustered at the location-industry level. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

B.3 Labor misallocation

Figure B.4 scatters two measures of labor misallocation against the volatility of productivity shocks across local markets.

Figure B.4: Volatility and labor misallocation across local markets



NOTES: This figure presents a binscatter of the relationship between labor misallocation and productivity shock volatility across local markets. In Panel A, labor misallocation is measured as the correlation between firm productivity and the average (log) employment of firms within each local market. In Panel B, labor misallocation is measured as the standard deviation of the average within-firm marginal product of labor across firms in each local market. Local markets are defined as the combination of a two-digit CAE industry classification and a NUTS-2 region with at least 100 firm-year observations.

Table B5 presents the estimated correlation between the two measures of labor misallocation and the volatility of productivity shocks across local markets, controlling for 2-digit CAE industry and NUTS-2 region fixed effects.

Although firm-level volatility exhibits a stronger correlation with capital misallocation, labor allocation follows similar patterns across local markets. Based on estimates from columns (2) and (4), transitioning from local markets with an average volatility of 20% to those with twice the volatility reduces the correlation between firm productivity and employee count by 0.25, while increasing the dispersion of log marginal product of labor by 0.20.

Table B5: Productivity shock volatility and labor misallocation across local markets

	$\hat{\eta}^j$		$\hat{\delta}^j$	
	(1)	(2)	(3)	(4)
$\hat{\sigma}^j$	-2.017** (0.586)	-1.251* (0.582)	1.952*** (0.493)	1.304*** (0.244)
R^2	0.451	0.744	0.550	0.818
Observations	397	397	397	397
Weighted	No	Yes	No	Yes

NOTES: This table reports OLS estimates from regressions of our two measures of labor misallocation on productivity shock volatility across local markets. In Columns 1 and 2, labor misallocation is measured as the correlation between firm productivity and the average (log) employment of firms within each local market. In Columns 3 and 4, labor misallocation is measured as the standard deviation of the average within-firm marginal product of labor within each local market. Local markets are defined as the combination of a two-digit CAE industry classification and a NUTS-2 region with at least 100 firm-year observations. All regressions include location and industry fixed effects. In Columns (2) and (4), observations are weighted by the number of firms in their local markets. Standard errors (in parentheses) are clustered at the industry and location level. * $p < 0.1$, $p < 0.05$, ** $p < 0.01$.

B.4 Shock persistence

For each local market j , we estimate the following production function:

$$\log y_{it} = \beta^{0j} + \beta^{1j} \log k_{it} + \beta^{2j} \log \ell_{it} + \beta^{3j} \log m_{it} + \mu_i + v_{it}$$

where y_{it} represents revenues (sales) for firm i in year t , k_{it} denotes the capital stock, ℓ_{it} is the number of employees, m_{it} denotes intermediate inputs, and μ_i and μ_t represent firm and year fixed effects, respectively. Finally, v_{it} is an error term that follows an autoregressive process of order 1; i.e.,

$$v_{it} = \rho v_{it-1} + \epsilon_{it}$$

where ϵ_{it} is an i.i.d. error term. Following Blundell and Bond (2000), we can express the model using a dynamic (common factor) representation; i.e.,

$$\begin{aligned} \log y_{it} = & \beta^{0j} + \beta^{1j} \log k_{it} - \rho \beta^{1j} \log k_{it-1} \\ & + \beta^{2j} \log \ell_{it} - \rho \beta^{2j} \log(q_{it-1} \ell_{it-1}) \\ & + \beta^{3j} \log m_{it} - \rho \beta^{3j} \log m_{it-1} \\ & + \rho \log y_{it-1} + \mu_i(1 - \rho) + \epsilon_{it} \end{aligned}$$

Therefore, we can estimate the following unrestricted equation:

$$\begin{aligned} \log y_{it} = & \pi_{0j} + \pi_{1j} \log k_{it} + \pi_{2j} \log k_{it-1} \\ & + \pi_{3j} \log \ell_{it} + \pi_{4j} \log(q_{it-1} \ell_{it-1}) \\ & + \pi_{5j} \log m_{it} + \pi_{6j} \log m_{it-1} \\ & + \pi_{7j} \log y_{it-1} + \mu_i^* + \epsilon_{it}, \end{aligned}$$

using suitably lagged levels of the variables as instruments after first differences to control for firm-level permanent unobserved heterogeneity. This allows us to obtain an estimate of our coefficients of interest, $\beta^{0j}, \beta^{1j}, \beta^{2j}, \beta^{3j}$, and ρ , as given by

$$\hat{\beta}^{0j} = \hat{\pi}_{0j}, \quad \hat{\beta}^{1j} = \frac{\hat{\pi}_{1j} + \hat{\pi}_{2j}}{1 - \hat{\pi}_{7j}}, \quad \hat{\beta}^{2j} = \frac{\hat{\pi}_{3j} + \hat{\pi}_{4j}}{1 - \hat{\pi}_{7j}}, \quad \hat{\beta}^{3j} = \frac{\hat{\pi}_{5j} + \hat{\pi}_{6j}}{1 - \hat{\pi}_{7j}}, \quad \hat{\rho} = \hat{\pi}_{7j}.$$

We use $\hat{\beta}^{0j}, \hat{\beta}^{1j}, \hat{\beta}^{2j}$, and $\hat{\beta}^{3j}$ to compute the following residuals:

$$\hat{r}_{it} = \log y_{it} - \hat{\beta}^{0j} - \hat{\beta}^{1j} \log k_{it} - \hat{\beta}^{2j} \log \ell_{it} - \hat{\beta}^{3j} \log m_{it}.$$

We obtain an estimate of firm productivity $\hat{\mu}_i$ by taking the time average of $\hat{r}_{it}$ for every firm i ; i.e.,

$$\hat{\mu}_i = \frac{1}{T} \sum_{t=1}^T \hat{r}_{it}.$$

Finally, notice that $\hat{r}_{it} - \hat{\mu}_i = \hat{v}_{it}$. Hence we can estimate the volatility of productivity shocks at the firm level, $\hat{\sigma}_i$, as the standard deviation of the demeaned residual, adjusted by the estimated persistence of the error term process; i.e.,

$$\hat{\sigma}_i = \sqrt{(1 - \hat{\pi}_{\hat{v}}^2) \sum_{t=1}^T (\hat{v}_{it} - \bar{\hat{v}}_{it})^2}$$

where $\bar{\hat{v}}_{it} = \frac{1}{T} \sum_{t=1}^T \hat{v}_{it}$.

With $\hat{\beta}^{1j}, \hat{\beta}^{2j}, \hat{\mu}_i$, and $\hat{\sigma}_i$ at hand, we construct the measures of average volatility, $\hat{\sigma}^j$, and capital (labor) misallocation, $\hat{\eta}^j$ and $\hat{\delta}^j$, for each local market j , as described in the main text, and proceed to estimate the following regression:

$$\hat{\omega}^j = \gamma_0 + \gamma_1 \hat{\sigma}^j + \varepsilon^j,$$

where $\hat{\omega}^j$ is either $\hat{\gamma}^j$ or $\hat{\delta}^j$.

Tables B6 and B7 present the estimated correlation between capital misallocation and the volatility of productivity shocks across local markets, controlling for 2-digit CAE industry and NUTS-2 region fixed effects. Using estimates from columns (2) and (4), an increase in local market volatility from an average of 20% to twice that level reduces the correlation between firm productivity and capital holdings by 0.53, while the dispersion of log marginal product of capital rises by 1.00. Similarly, the correlation between firm productivity and number of employees decreases by 0.31, while the dispersion of log marginal product of labor increases by 0.35. Consistent with the main specification, variations in productivity shock volatility across local markets exhibit a

Table B6: Volatility and capital misallocation across local markets with shock persistence

	$\hat{\eta}^j$		$\hat{\delta}^j$	
	(1)	(2)	(3)	(4)
$\hat{\sigma}^j$	-2.826*** (0.728)	-2.672* (1.113)	1.974*** (0.369)	4.978** (1.717)
R^2	0.410	0.575	0.397	0.611
Observations	397	397	389	389
Weighted	No	Yes	No	Yes

NOTES: This table reports OLS estimates from regressions of our two measures of capital misallocation on productivity shock volatility across local markets. In Columns 1 and 2, capital misallocation is measured as the correlation between firm productivity and the average (log) capital holdings of firms within each local market. In Columns 3 and 4, capital misallocation is measured as the standard deviation of the average within-firm marginal product of capital within each local market. Estimates of firm productivity, the average within-firm marginal product of capital, and firm-level productivity shock volatility—used to construct our measures of capital misallocation and volatility of productivity shocks at the local market level—are obtained from regression models in which the error term follows an autoregressive process of order one. Local markets are defined as the combination of a two-digit CAE industry classification and a NUTS-2 region with at least 100 firm-year observations. All regressions in this table include location and industry fixed effects. In Columns (2) and (4), observations are weighted by the number of firms in their local markets. Standard errors (in parentheses) are clustered at the location-industry level. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

stronger correlation with capital misallocation than with labor misallocation.

Table B7: Volatility and labor misallocation across local markets with shock persistence

	$\hat{\eta}^j$		$\hat{\delta}^j$	
	(1)	(2)	(3)	(4)
$\hat{\sigma}^j$	-1.705** (0.578)	-1.530** (0.538)	2.079*** (0.380)	1.726*** (0.368)
R^2	0.443	0.675	0.627	0.812
Observations	397	397	395	395
Weighted	No	Yes	No	Yes

NOTES: This table reports OLS estimates from regressions of our two measures of labor misallocation on productivity shock volatility across local markets. In Columns 1 and 2, labor misallocation is measured as the correlation between firm productivity and the average (log) employment of firms within each local market. In Columns 3 and 4, labor misallocation is measured as the standard deviation of the average within-firm marginal product of labor within each local market. Estimates of firm productivity, the average within-firm marginal product of labor, and firm-level productivity shock volatility—used to construct our measures of labor misallocation and volatility of productivity shocks at the local market level—are obtained from regression models in which the error term follows an autoregressive process of order one. Local markets are defined as the combination of a two-digit CAE industry classification and a NUTS-2 region with at least 100 firm-year observations. All regressions in this table include location and industry fixed effects. In Columns (2) and (4), observations are weighted by the number of firms in their local markets. Standard errors (in parentheses) are clustered at the location-industry level. $*p < 0.1$, $**p < 0.05$, $***p < 0.01$.

B.5 Firm exit

We correct for firm-level exit selection following the approach proposed by Mundlak (1978). We rely on a full parametrization of the sample selection mechanism, and assume the conditional expectation of the individual effects in the selection equation, ξ_i to be a function of the firm-level average of the time-varying covariates used in equation (3), i.e.

$$\xi_i = f(\overline{\log k_{it}}, \overline{\log \ell_{it}}, \overline{\log m_{it}}) \quad (7)$$

We choose a fully interacted third-order polynomial for $f(\cdot)$. Therefore we follow Wooldridge (1995) and implement a two-step procedure. In the first step, we estimate a probit on equations (4) and (7) separately for every year t , and obtain a selection correction terms λ_{it} , which is equal to the inverse Mills ratio. In the second step we estimate equation (3) including λ_{it} as a control.

Tables B8 and B9 report the estimated correlation of firm-level volatility and capital misallocation conditional on 2-digit CAE industry and NUTS-2 region fixed effects, and accounting for self-selection in firm exit.

Using estimates from columns (2) and (4), as we move from local markets with an average volatility of 20% to those with volatility twice as large, the correlation between firm-level capital and productivity reduces by 0.45, while the dispersion of log MPK increases by 0.30.

Correcting for exit selection of firms only marginally affects the magnitude of our estimates: while it amplifies the estimated correlations in the unweighted specifications (columns 1 and 3), it hampers them in the weighted ones (columns 2 and 4), suggesting the selection introduces a larger bias as we move towards markets with a higher number of firms.

Table B8: Volatility and capital misallocation across local markets —
Heckman correction for firm exit

	$\hat{\eta}^j$		$\hat{\delta}^j$	
	(1)	(2)	(3)	(4)
$\hat{\sigma}^j$	-3.891*** (0.707)	-2.288** (0.646)	2.496*** (0.501)	1.547*** (0.103)
R^2	0.434	0.786	0.533	0.883
Observations	397	397	389	389
Weighted	No	Yes	No	Yes

NOTES: This table reports OLS estimates from regressions of our two measures of capital misallocation on productivity shock volatility across local markets. In Columns 1 and 2, capital misallocation is measured as the correlation between firm productivity and the average (log) capital holdings of firms within each local market. In Columns 3 and 4, capital misallocation is measured as the standard deviation of the average within-firm marginal product of capital within each local market. Estimates of firm productivity, the average within-firm marginal product of capital, and firm-level productivity shock volatility—used to construct our measures of capital misallocation and volatility of productivity shocks at the local market level—are obtained from regression models that account for firm exit as in Wooldridge (1995). Local markets are defined as the combination of a two-digit CAE industry classification and a NUTS-2 region with at least 100 firm-year observations. All regressions in this table include location and industry fixed effects. In Columns (2) and (4), observations are weighted by the number of firms in their local markets. Standard errors (in parentheses) are clustered at the location-industry level. $*p < 0.1$, $**p < 0.05$, $***p < 0.01$.

Table B9: Volatility and labor misallocation across local markets —
Heckman correction for exit

	$\hat{\eta}^j$		$\hat{\delta}^j$	
	(1)	(2)	(3)	(4)
$\hat{\sigma}^j$	-2.089** (0.614)	-1.279* (0.529)	2.066*** (0.490)	1.263*** (0.226)
R^2	0.455	0.721	0.557	0.822
Observations	397	397	395	395
Weighted	No	Yes	No	Yes

NOTES: This table reports OLS estimates from regressions of our two measures of labor misallocation on productivity shock volatility across local markets. In Columns 1 and 2, labor misallocation is measured as the correlation between firm productivity and the average (log) employment of firms within each local market. In Columns 3 and 4, labor misallocation is measured as the standard deviation of the average within-firm marginal product of labor within each local market. Estimates of firm productivity, the average within-firm marginal product of labor, and firm-level productivity shock volatility—used to construct our measures of labor misallocation and volatility of productivity shocks at the local market level—are obtained from regression models that account for firm exit as in Wooldridge (1995). Local markets are defined as the combination of a two-digit CAE industry classification and a NUTS-2 region with at least 100 firm-year observations. All regressions in this table include location and industry fixed effects. In Columns (2) and (4), observations are weighted by the number of firms in their local markets. Standard errors (in parentheses) are clustered at the location-industry level. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

C Estimation and counterfactual

C.1 Alternative distortions

Table C1 reports the list of model parameters, their calibrated values, and the source or target used to discipline them in the version of the model that includes exogenous correlated distortions.

Table C1: Parameters - Model with exogenous correlated distortions

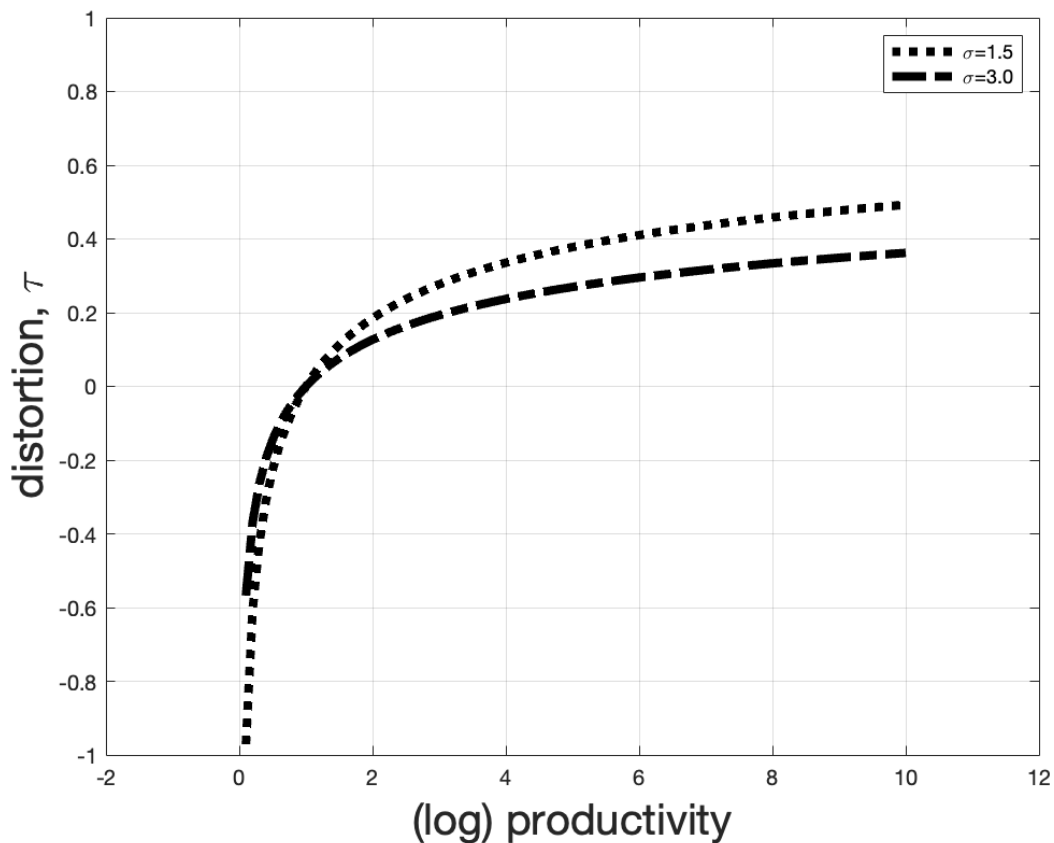
Parameters	Description	Value	Source/Target
<i>A. Parameters calibrated</i>			
α	Output elasticity of capital	0.416	Aggregate capital share (FRED)
σ	Volatility of shocks	0.347	Data
(κ, θ)	Distribution of permanent productivity	(1.404, 1.060)	Data
<i>B. Parameters estimated</i>			
<i>Case 1.</i>			
γ	Relative risk aversion	1.5	Cagetti and De Nardi (2006)
φ	Correlation distortions	0.295	corr. $[\log k, \log z] = 0.494$
<i>Case 2.</i>			
γ	Relative risk aversion	3	Fuchs et al. (2024)
φ	Correlation distortions	0.195	corr. $[\log k, \log z] = 0.494$

NOTES: This table reports the values and sources of the calibrated and estimated parameters in the version of the model that includes exogenous correlated distortions.

Case 1 refers to a version of the model with a constant relative risk aversion of 1.5. Case 2 refers to a version of the model with a constant relative risk aversion of 3. The correlation of distortions, φ is estimated to be 0.295 in Case 1 and 0.195 in Case 2.

Figure C.1 scatters the estimates of correlated distortions, τ_i against (log) firm productivity level, z_i , when the relative risk aversion coefficient, γ is either 1.5 or 3. A positive value of the wedge implies a tax on capital expenditure. A negative value implies a subsidy on capital expenditure.

Figure C.1: Exogenous correlated distortions — Estimates



NOTES: This figure reports the estimated correlated distortions τ_i for firms of different productivity level, z_i when the coefficient of relative risk aversion is either 1.5 or 3.

Table C2 shows several counterfactual results in a calibrated model that includes exogenous correlated distortions.

Table C2: The aggregate cost of incomplete markets -
Model with exogenous correlated distortions

	Counterfactual scenarios				Explained % (5)
	Baseline (1)	No uncertainty (2)	Complete markets (3)	Complete markets & no distortions (4)	
Volatility, σ	0.347	0	0.347	0.347	-
<i>Case 1. ($\gamma = 1.5$)</i>					
Correlated distortions, φ	0.295	0.295	0.295	0	-
corr[log z , log k]	0.503	0.615	0.615	1.000	22.54%
ω [log z , log k]	0.310	0.424	0.424	1.712	8.13%
sd[log k]	0.810	0.989	0.989	1.594	22.83%
Aggregate Income	1.000	1.010	1.064	1.139	46.04%
<i>Case 2. ($\gamma = 3$)</i>					
Correlated distortions, φ	0.195	0.195	0.195	0	-
corr[log z , log k]	0.489	0.996	0.996	1.000	99.22%
ω [log z , log k]	0.583	1.250	1.250	1.712	59.08%
sd[log k]	0.538	1.216	1.216	1.594	64.20%
Aggregate Income	1.000	1.169	1.237	1.242	97.93%

NOTES: This table reports selected outcomes for the baseline economy (column 1), a counterfactual economy with complete insurance markets and correlated distortions (column 2), and a counterfactual economy with complete insurance markets and no correlated distortions (column 3). Values in column (4) refer to the % of each outcomes in column (3) explained by market incompleteness, and are computed as $100 \times \frac{(3)-(1)}{(4)-(1)}$. Case 1 refers to a version of the model with a constant relative risk aversion of 1.5. Case 2 refers to a version of the model with a constant relative risk aversion of 3.

C.2 Alternative elasticity of capital

Table C3 reports the estimated coefficient of relative risk aversion (RRA) for different values of capital elasticity, along with the corresponding output per unit of capital and income gains from the counterfactual exercises under different capital share values.

Table C3: Sensitivity to alternative capital shares

Capital share α	0.5	0.6	0.7	0.8	0.9
Estimated γ	3.507	3.715	4.024	4.235	4.717
Income gains %					
from no uncertainty	29.94	47.05	85.37	126.0	251.9
from complete markets	38.86	57.22	96.85	138.3	270.6

NOTES: This table reports output per unit of capital and aggregate income gains (in %) from two counterfactual scenarios: one with no idiosyncratic production uncertainty and another with complete markets, for alternative economies that differ in the capital share α . In each of these economies, the coefficient of relative risk aversion γ is re-estimated to match the observed correlation between productivity and capital holdings across firms.

C.3 Alternative technology

Table C4 shows the calibrated parameters for the model that includes labor as a factor of production.

Table C4: Parameters - Model with capital and labor inputs

Parameters	Description	Value	Source/Target
<i>A. Parameters calibrated</i>			
α	Output elasticity of labor	0.649	Aggregate labor share (FRED)
η	Span on control	0.9	Erosa et al. (2023)
σ	Volatility of shocks	0.347	Data
(κ, θ)	Distribution of permanent productivity	(1.404, 1.060)	Data
<i>B. Parameters estimated</i>			
γ	Relative risk aversion	5.111	corr. $[\log k, \log z] = 0.494$

NOTES: This table reports the values and source of the calibrated and estimated parameters for a version of the model that includes labor as a factor of production.

Table C5 shows the counterfactual results when using the calibrated model that includes labor as a factor of production.

Table C5: The aggregate cost of incomplete markets -
Model with capital and labor inputs

	Baseline (1)	Counterfactual scenarios	
		No uncertainty (2)	Complete markets (3)
Volatility, σ	0.347	0	0.347
corr[$\log z, \log k$]	0.502	1	1
$\omega[\log z, \log k]$	1.052	4.159	4.159
sd[$\log k$]	2.534	7.010	7.010
Aggregate Income	1.000	1.461	1.632

NOTES: This table reports selected outcomes for the baseline economy (column 1), a counterfactual economy with no idiosyncratic production uncertainty (column 2), and a counterfactual economy with complete insurance markets (column 3), obtained from a version of the model that includes labor as a factor of production.