

# The Cost of Capital and Misallocation in the United States

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## Traditional approach:

1. Strong assumptions about production functions (Homogeneous Cobb-Douglas)
2. Measure heterogeneity in marginal products from cross-sectional input and production data
3. Estimate capital misallocation

## Our approach:

- Optimizing firms equate cost of capital to expected marginal product of capital
- Combine credit registry data + model to carefully measure cost of capital, and infer MPK
- Use dispersion in cost of capital to quantify welfare losses stemming from credit market frictions

# This paper

## Methodological contribution:

- Adapt a standard dynamic corporate finance model for measurement with **micro loan-level data**
- Derive a **sufficient statistic** for capital misallocation arising due to credit market frictions

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## Empirical results for the US:

- Average cost of capital tracks 5-year treasury rates, with a spread
- Measures of cost of capital correlate with traditional measures of  $ARPK_i$  at the firm level
- Credit markets efficient in normal times: losses from misallocation  $\approx 0.9\%$  of GDP
- Losses from misallocation increased to 1.8% of GDP in 2020-2021

## Related literature

- **Measuring misallocation:**

- Seminal work: Restuccia and Rogerson (2008), Hsieh and Klenow (2009)
- Challenge: Standard methods rely on strong assumptions (Haltiwanger et al., 2018).
- Recent advances: (Quasi-)Experimental methods to recover marginal products directly (Carrillo et al., 2023; Hughes and Majerovitz, 2025).
- Contribution: use **heterogeneity in funding costs** to measure **dispersion in MRPK**

- **Heterogeneity in the cost of capital:**

- Developing countries: Banerjee and Duflo (2005); Cavalcanti, Kaboski, Martins, and Santos (2024)
- US: Gilchrist, Sim, and Zakrajsek (2013); Binsbergen and Opp (2019); Gormsen and Huber (2023, 2024); Faria-e-Castro, Jordan-Wood, and Kozlowski (2024)
- Contribution:
  - Estimate firm cost of capital using **credit registry data**, controlling for loan characteristics
  - Derive and estimate **sufficient statistic** for misallocation

# Outline

1. Model
2. Welfare and misallocation
3. Measurement with credit registry data
4. Empirical results
5. Cross-country comparison

## Model Summary

- Discrete time, infinite horizon
- Continuum of firms, each matched with a lender
- No aggregate risk (for now - work in progress!)

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**Data:** We will measure heterogeneous  $\rho_i$  at the security level in credit-registry data (details later).

**Key question:** How do heterogeneity in  $\rho_i$  and financial frictions distort the allocation of capital?

# Model

**Firm value function:**

$$V_i(k_i, b_i, z_i) = \max_{k'_i, b'_i} \pi_i(k_i, b_i, z_i, k'_i, b'_i) + \beta \mathbb{E} \left[ \underbrace{\max \{ V_i(k'_i, b'_i, z'_i), 0 \}}_{\text{Limited liability}} \mid z_i \right]$$

**Firm profits:**

$$\pi_i(k_i, b_i, z_i, k'_i, b'_i) = f(k_i, z_i) + (1 - \delta) k_i - k'_i - \theta b_i + Q_i(k'_i, b'_i, z_i) [b'_i - (1 - \theta_i) b_i]$$

**Price of debt:**

$$Q_i(k'_i, b'_i, z_i) = \frac{\mathbb{E} \left\{ \underbrace{\mathcal{P}_i(k'_i, b'_i, z'_i)}_{\text{repayment prob.}} [\theta_i + (1 - \theta_i) Q_i(k''_i, b''_i, z'_i)] + (1 - \mathcal{P}_i(k'_i, b'_i, z'_i)) \underbrace{\frac{\phi_i k'_i}{b'_i}}_{\text{recovery}} \mid z_i \right\}}{\underbrace{1 + \rho_i}_{\text{lender discount rate}}}$$

- Cost of capital:

$$\underbrace{\frac{\mathbb{E} [\mathcal{P}'_i(\theta_i + (1 - \theta_i)Q'_i) | z_i]}{Q_i}}_{1+r_i^{\text{firm}}} \times \underbrace{\left[ \frac{1 - \frac{\partial Q_i}{\partial k'_i} [b'_i - (1 - \theta_i)b_i]}{1 + \frac{\partial Q_i}{\partial b'_i} \frac{[b'_i - (1 - \theta_i)b_i]}{Q_i}} \right]}_{\mathcal{M}_i}$$

- $1 + r_i^{\text{firm}}$ : implied interest rate perceived by the firm
- $\mathcal{M}_i$ : price impact term capturing how  $(k'_i, b'_i)$  affect debt price  $Q_i$

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- $1 + r_i^{\text{firm}}$ : implied interest rate perceived by the firm
- $\mathcal{M}_i$ : price impact term capturing how  $(k'_i, b'_i)$  affect debt price  $Q_i$
- **Optimality:** firm equates cost of capital to expected MRPK

$$(1 + r_i^{\text{firm}}) \cdot \mathcal{M}_i = \underbrace{\mathbb{E}[\mathcal{P}'_i(f_k(k'_i, z'_i) + 1 - \delta) | z_i]}_{\text{expected MRPK}}$$

- **Measurement idea:** measure  $r_i^{\text{firm}}$  from loan data to infer dispersion in MRPK and misallocation.

# Firm's cost of capital

## Lemma 1 (Firm's cost of capital)

The firm's cost of capital is:

$$1 + r_i^{firm} = \frac{1 + \rho_i}{1 + \Lambda_i}$$

$$\Lambda_i := \frac{\mathbb{E} [(1 - \mathcal{P}'_i) \phi_i k'_i / b'_i | k'_i, b'_i, z_i]}{\mathbb{E} [\mathcal{P}'_i (\theta + (1 - \theta_i) Q'_i) | k'_i, b'_i, z_i]}$$

▷ Proof

- Limited liability + recovery creates wedge between  $\rho$  (lender discount rate) and  $r_i^{firm}$
- Under default, firm liquidates and gets zero; liquidated capital is used to (partially) repay lender
- On the margin, firm investment raises lender's payoff in default but has no effect on firm's payoff
- In general,  $r_i^{firm} < \rho_i$  if lender expects some recoveries in default

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## Aggregate economy and welfare

- Aggregate resources available for consumption and new capital:

$$Y_{t+1} + (1 - \delta)K_{t+1} = \int_0^1 \mathbb{E}_t [\mathcal{P}_{i,t+1} (f(k_{i,t+1}, z_{i,t+1}) + (1 - \delta)k_{i,t+1}) + (1 - \mathcal{P}_{i,t+1}) \cdot \phi_i k_{i,t+1}] di$$

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- Let  $\omega_{i,t}(S^t) \in \{0, 1\}$  denote whether a firm operates or not, as a function of history  $S^t$ 
  - $S^t$  is the history of productivity draws up to  $t$
- Planner's problem:

$$U^* = \max_{\left\{ \left\{ k_{i,t}(S^{t-1}), \omega_{i,t}(S^t) \right\}_{i \in [0,1]} \right\}_{t=1}^\infty} \sum_{t=0}^{\infty} \beta^t \cdot u(C_t)$$

$$\text{s.t.} \quad K_t = \int_0^1 k_{i,t}(S^{t-1}) di$$

$$C_t + K_{t+1} = Y_t + (1 - \delta)K_t$$

$$\omega_{i,t+1}(S^{t+1}) \leq \omega_{i,t}(S^t) \quad \forall S^t \subset S^{t+1}, \forall i$$

# Aggregate economy and welfare

Planner must decide:

- Which firms operate or exit
- Aggregate capital each period
- Allocation of capital across (operating) firms

Can separate planner's problem into outer (dynamic) and inner (static) problems:

$$U^* = \underbrace{\max_{\{K_t, \{\omega_{i,t}(S^t)\}_{i \in [0,1]}\}_{t=1}^\infty}_{\text{Outer (dynamic)} \\ \text{Aggregate capital and exit}}} \sum_{t=0}^{\infty} \beta^t \cdot u \left( \underbrace{\max_{\{k_{i,t}(S^{t-1})\}_{i \in [0,1]}^\infty}_{\text{Inner (static)} \\ \text{Reallocate capital}} Y_t} - I_t \right)$$

## Aggregate economy and welfare: intensive-margin misallocation

- Inner problem: reallocate capital across firms, taking exit and aggregate capital as given
- Focus on misallocation at the intensive margin
  - As in most of the literature, e.g. Hsieh and Klenow (2009)
  - Necessary for measurement: hard to measure outcomes for counterfactual firms that don't exist
- Planner redistributes  $\{k_{i,t+1}\}_{i \in [0,1]}$  taking exit decisions  $\{\mathcal{P}_{i,t+1}^{DE}\}_{i \in [0,1]}$  and  $K_{t+1}^{DE}$  as given

$$\max_{\{k_{i,t+1}^*\}_{i \in [0,1]}} \int_0^1 \mathbb{E}_t [\mathcal{P}_{i,t+1}^{DE} (f(k_{i,t+1}^*, z_{i,t+1}) + (1 - \delta) k_{i,t+1}^*) + (1 - \mathcal{P}_{i,t+1}^{DE}) \cdot \phi_i k_{i,t+1}^*] di$$

s.t.

$$\int_0^1 k_{i,t+1}^* di = K_{t+1}^{DE}$$

## Social return on capital

Planner cares about payoffs to both lender and firm

- Total Surplus =  $\mathbb{E}_t [\mathcal{P}_{i,t+1}^{DE} (f(\mathbf{k}_{i,t+1}^*, \mathbf{z}_{i,t+1}) + (1 - \delta) \mathbf{k}_{i,t+1}^*) + (1 - \mathcal{P}_{i,t+1}^{DE}) \cdot \phi_i \mathbf{k}_{i,t+1}^*]$

Define the **social marginal product of capital at firm  $i$** ,  $r_{i,t}^{social}(\mathbf{k})$ , as:

$$1 + r_{i,t}^{social}(\mathbf{k}) := \mathbb{E} \left[ \underbrace{\mathcal{P}_{i,t+1}^{DE} (f_k(\mathbf{k}, \mathbf{z}_{i,t+1}) + 1 - \delta)}_{(1+r_{i,t}^{firm}) \times \mathcal{M}_{i,t}} + \underbrace{(1 - \mathcal{P}_{i,t+1}^{DE}) \phi_i}_{\text{Recoveries}} \right]$$

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- In the empirically relevant case,  $r_i^{firm} < r_i^{social} < \rho_i$
- Firm doesn't care about recoveries, planner cares the "right amount," lender cares too much

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Planner Optimality: Equalize  $r_{i,t}^{social}(\mathbf{k}_{i,t+1}^*)$  across firms

Inefficient Equilibrium: Dispersion in  $r_{i,t}^{social} \rightarrow$  misallocation

# Misallocation: sufficient statistic

## Proposition 1 (Misallocation)

*Misallocation can be measured with  $\mathbb{E}[r_i^{social}]$  and  $Var(r_i^{social})$  as*

$$\log(Y^*/Y^{DE}) \approx \frac{1}{2} \cdot \mathcal{E} \cdot \log \left( 1 + \frac{Var(r_i^{social})}{(\mathbb{E}[r_i^{social}] + \delta)^2} \right)$$

### ▷ Proof

- Extends Hughes and Majerovitz (2025) to a dynamic economy with default
  - Measures intensive-margin misallocation
  - Set  $\mathcal{E} = \frac{1}{2}$  (elasticity of output w.r.t.  $r_i^{social} + \delta$ ) and  $\delta = 0.06$  ▷ Calibration
- **Next:** Show how to measure  $r_i^{social}$  using credit registry data

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4. Empirical results
5. Cross-country comparison

- Quarterly loan-level panel on universe of loan facilities > \$1M
- Sample covers top 40 BHCs, 2014:Q4-2024:Q4
- 91% of C&I lending by top 25 banks; 55% of C&I undertaken by all commercial banks
- Detailed information on features of credit facilities
  - Origination date, size, maturity, interest rate/spread, probability of default, loss given default, fixed vs. floating, type of loan, etc.
- Focus on term loans at origination, issued to non-government, non-financial US companies

## Pricing term loans

For a loan  $i$  originated at  $t$ , the **break-even** condition for a lender with discount rate  $\rho_{i,t}$  is

$$1 = \sum_{s=1}^{T_{i,t}} \left[ \frac{(P_{i,t})^s \cdot \mathbb{E}_t(r_{i,t,s}) + (P_{i,t})^{s-1} \cdot (1 - P_{i,t}) \cdot (1 - LGD_{i,t})}{(1 + \rho_{i,t})^s \cdot \mathbb{E}_t(\Pi_{t,s})} \right] + \frac{(P_{i,t})^{T_{i,t}}}{(1 + \rho_{i,t})^{T_{i,t}} \cdot \mathbb{E}_t(\Pi_{t,T_{i,t}})}$$

- “Bullet” loans: pay interest each period and principal at the end
- $T_{i,t}$ : maturity
- $P_{i,t}$ : repayment probability (constant over time)
- $LGD_{i,t}$ : loss given default (constant over time)
- $\mathbb{E}_t[r_{i,t,s}]$ : fixed rate or spread over benchmark (Gürkaynak, Sack, and Wright, 07)  $\triangleright$  forward rates
- $\mathbb{E}_t(\Pi_{t,s})$ : total expected inflation between  $t$  and  $s$ , from term structure of  $\mathbb{E}_t\pi_s$  (Cleveland Fed)
- $\Rightarrow$  Solve for lender's discount rate:  $\rho_{i,t}$

# Measuring firm and social cost of capital

## Lemma 2 (Firm and social cost of capital)

We can write the firm cost of capital as:

$$1 + r_{i,t}^{\text{firm}} = (1 + \rho_{i,t}) - (1 - P_{i,t})(1 - LGD_{i,t})$$

and the social cost of capital as:

$$\begin{aligned} 1 + r_{i,t}^{\text{social}} &= (1 + r_{i,t}^{\text{firm}})\mathcal{M}_{i,t} + (1 - P_{i,t})(1 - LGD_{i,t})lev_{i,t} \\ &= \underbrace{(1 + \rho_{i,t})\mathcal{M}_{i,t}}_{\text{lender discount rate}} + \underbrace{(lev_{i,t} - \mathcal{M}_{i,t}) \cdot (1 - P_{i,t}) \cdot (1 - LGD_{i,t})}_{\text{wedge due to financial frictions}} \end{aligned}$$

▷ Proof

Where  $lev_{i,t}$  is leverage: value of firm's debt over its capital

## Sufficient statistic for misallocation

$$\log \left( Y_t^* / Y_t^{DE} \right) \approx \frac{1}{2} \cdot \mathcal{E} \cdot \log \left( 1 + \frac{\text{Var} (r_{i,t}^{social})}{(\mathbb{E} [r_{i,t}^{social}] + \delta)^2} \right) \quad (1)$$

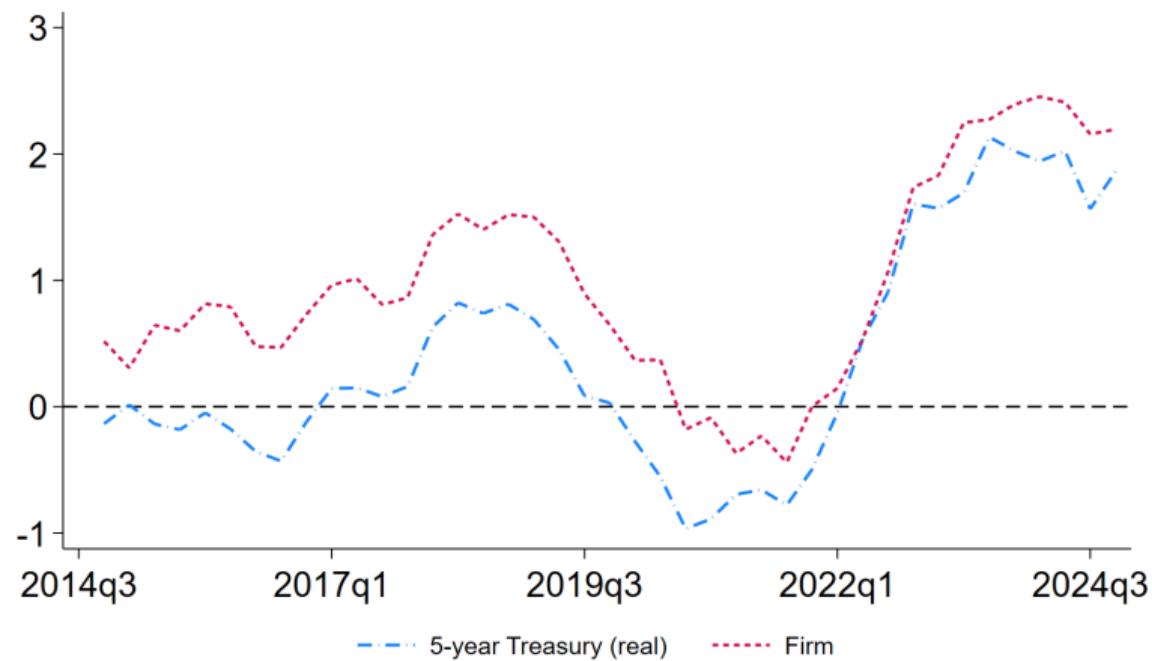
$$1 + r_{i,t}^{social} = (1 + \rho_{i,t}) \mathcal{M}_{i,t} + (lev_{i,t} - \mathcal{M}_{i,t}) \cdot (1 - P_{i,t}) \cdot (1 - LGD_{i,t}) \quad (2)$$

- Set  $\mathcal{M}_{i,t} = 1$ ; reasonable approximation given our data ▷ estimate  $\mathcal{M}$
- Can measure misallocation directly with credit registry data using (1) and (2)!
- Dispersion in  $r_{i,t}^{social}$  comes from:
  1. Dispersion in lender's discount rate,  $\rho_{i,t}$
  2. Dispersion in financial frictions wedge
  3. Covariance between  $\rho_{i,t}$  and financial frictions wedge

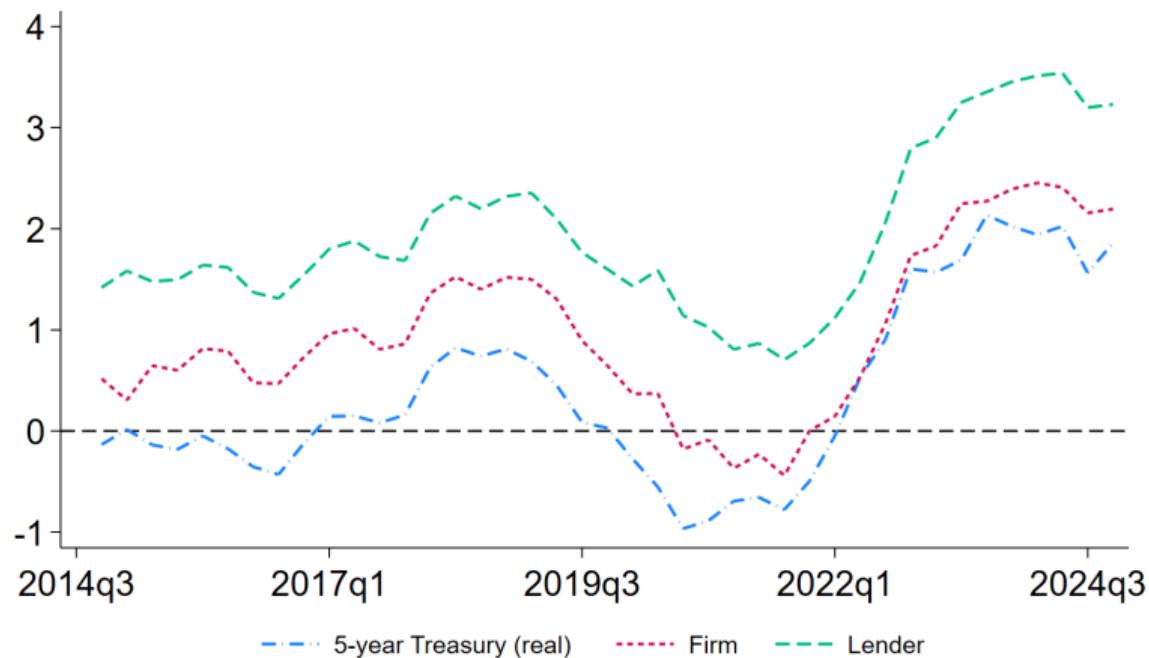
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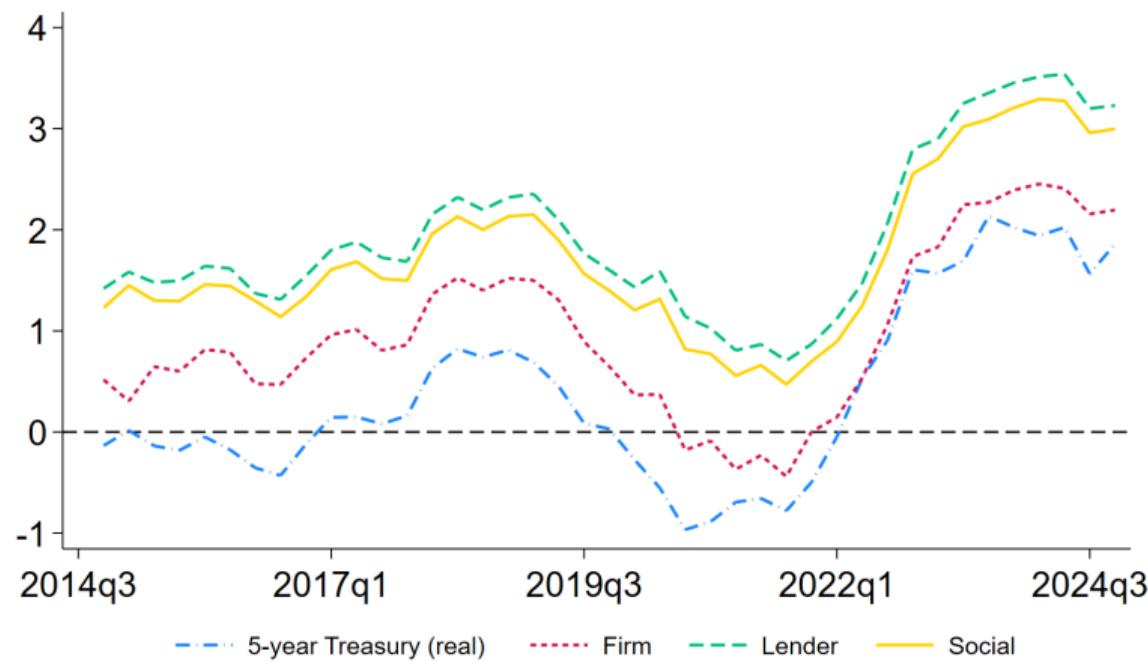
## Time series for average discount rate, firm and social cost of capital



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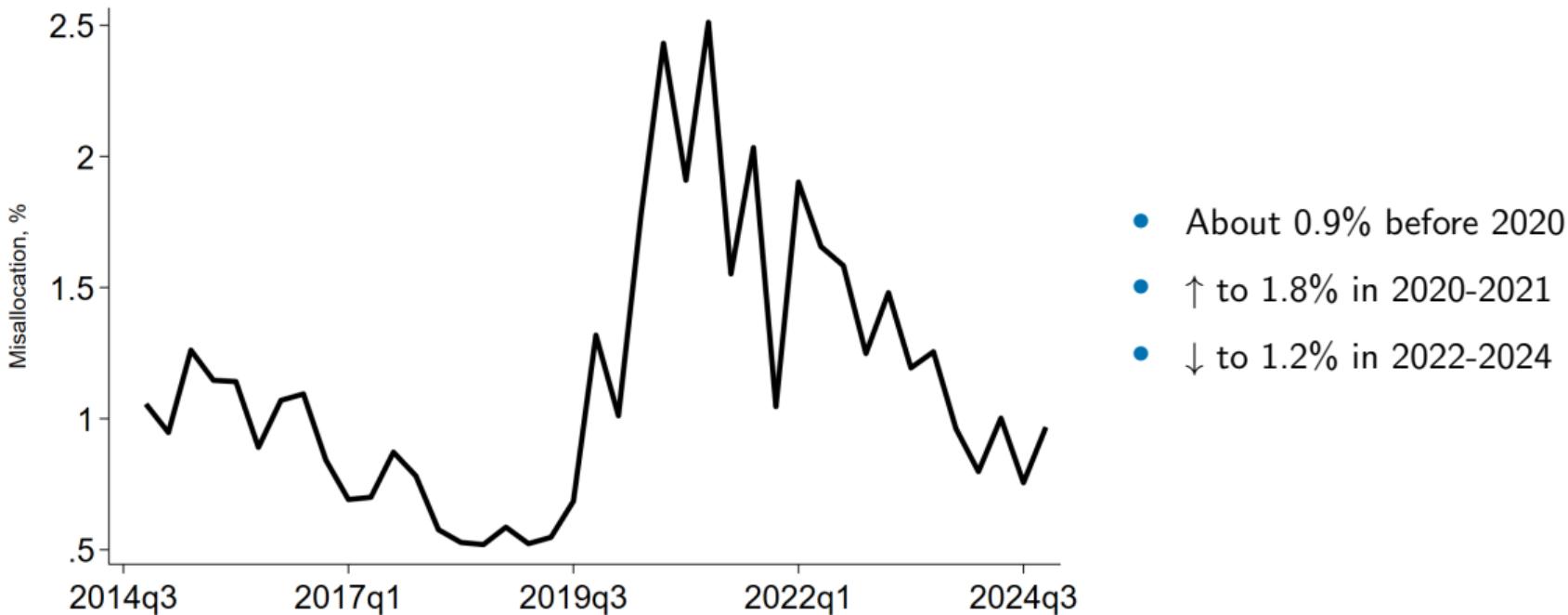
## Estimates for lender discount rate, firm and social cost of capital

	Mean	SD	p10	p50	p90
$\rho$ (%)	1.87	1.55	0.41	1.88	3.62
$r^{firm}$ (%)	0.92	2.80	-0.86	1.26	3.03
$r^{social}$ (%)	1.66	1.78	0.12	1.73	3.47

- Financial frictions/recovery:  $\mathbb{E} [r_{i,t}^{firm}] < \mathbb{E} [r_{i,t}^{social}] < \mathbb{E} [\rho_{i,t}]$
- Standard deviation of  $r^{social} = 1.78$ , What is the implication for misallocation?

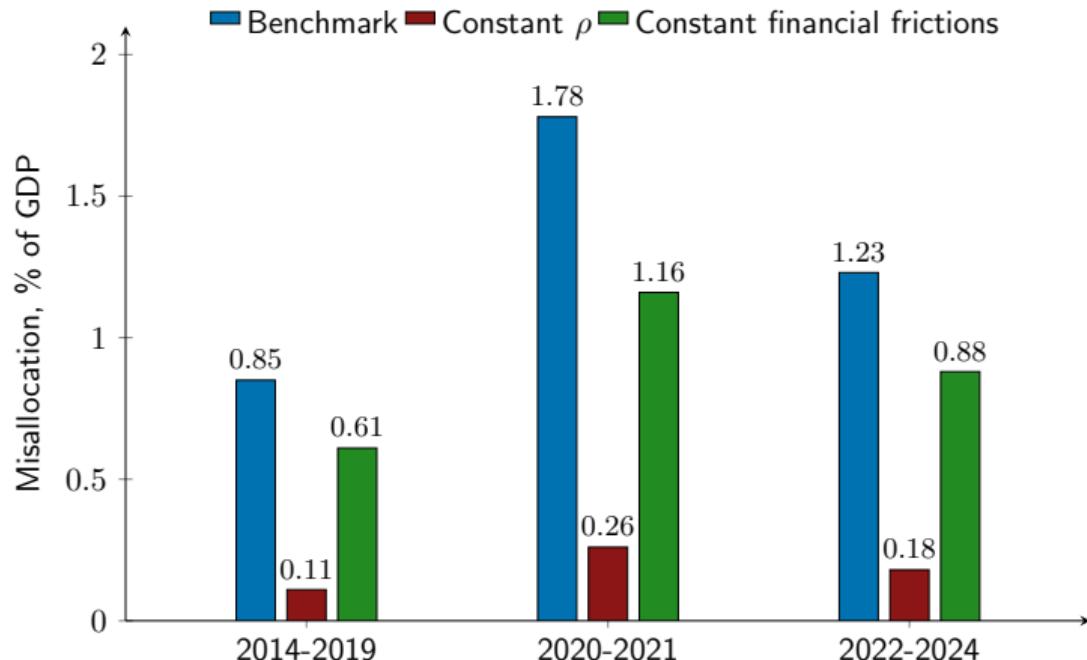
# Misallocation in the US, 2014-2024

▷ weighted



# Misallocation is driven by heterogeneity in $\rho_i$

▷ details



- Decomposition: set all  $\rho_i$  to quarterly mean (Constant  $\rho$ ); same for financial frictions term
- Main driver: dispersion in lender discount rates
- Interaction between  $\rho_i$  and financial frictions ( $0.85 > 0.11 + 0.61$ )

## Extensions & robustness

1. Estimate heterogeneous price-impact term  $\mathcal{M}$ . ▷ heterogeneous  $\mathcal{M}$
2. Variance decomposition: dispersion accounted by bank, firm, loan. ▷ variance decomposition
3. Validate  $r^{social}$  using firm-level ARPK measures. ▷ details
4. Compare to misallocation from traditional approach. ▷ details

### Work in progress:

1. Aggregate risk
2. Quantitative model

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## Are credit markets more efficient in the United States?

Overall, US credit markets seem quite efficient

- Losses of  $\approx 1\%$  from heterogeneous  $r^{social}$

How efficient are credit markets around the world?

- Widely believed that developing-country credit markets are inefficient, with high dispersion in cost of capital (Banerjee and Duflo 2005)
- Essential for understanding if this is an important driver of development

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**Ideal:** Redo **exact same analysis** in credit registry data for other countries

**For now:** Infer what we can from other papers, making some strong assumptions

## Cross-country comparison: Methodology and assumptions

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- For a fixed real interest rate  $r_{i,t}$ ,  $\rho$  has a closed-form:

$$1 + \rho_{i,t} = P_{i,t} (1 + r_{i,t}) + (1 - P_{i,t}) (1 - LGD_{i,t})$$

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  1. Obtain mean real rate by subtracting average realized inflation from mean nominal rate
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- Most papers report nominal rates, need to handle inflation
  1. Obtain mean real rate by subtracting average realized inflation from mean nominal rate
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- Assume all loans have the same  $P_{i,t}, LGD_{i,t}$ , equal to the average
- Recovery rates from the World Bank's Doing Business Database
- Approximate  $r_{i,t}^{social} \approx \rho_{i,t}$  and compute misallocation using our formula

## Cross-country comparison

	Aleem 1990 Pakistan	Khwaja & Mian 2005 Pakistan	Cavalcanti et al. 2024 Brazil	Beraldí 2025 Mexico	This paper 2025 United States
Years of data	1980–1981	1996–2002	2006–2016	2003–2022	2014–2024
Mean real rate, %	66.8	8.00	83.0	12.4	1.4
SD real rate, %	38.1	2.9	93.3	5.2	1.2
Mean def. prob., %	2.7	16.9	4.0	8.9	1.5
Mean recovery rate, %	42.8	42.8	18.2	63.9	66.6
Implied misallocation, %	6.5	13.5	21.5	2.8	0.8

- **Developing countries:** higher mean and standard deviation of real interest rates
- **U.S.:** lower mean and standard deviation of interest rates, **higher recovery**
- **Brazil:** most extreme misallocation: 21.5%.

## A plea for data

- We would like to do a follow-up project that measures the cost of capital around the world, using our methodology in each country
- Need credit data with:
  - Interest rate
  - Loan terms (e.g. fixed vs. variable rate, loan maturity)
  - Information on default probability and recovery rates
- We know this is possible in other countries (e.g. Mexico), but are looking for more data sets
- **If you know of a data set that we could use, please talk to me or send us an email!**

# Conclusion

- Framework to measure misallocation from credit registry data.
  1. Standard dynamic corporate finance model as measurement device
  2. Sufficient statistic for capital misallocation
  3. Uses standard credit registry variables ( $r, P, LGD, T, \dots$ )
- Application to U.S. credit registry data
  1. Estimate **lender discount rates**, **firm-level cost of capital** and **social cost of capital**
  2. Misallocation around 1% in normal times
  3. Cross-country comparison with other countries

Credit markets in the US are close to efficient, developing countries appear more distorted.

# **Appendices**

Firm FOCs:

$$[k'_i] : -1 + \frac{\partial Q_i(k'_i, b'_i, z_i)}{\partial k'_i} [b'_i - (1 - \theta_i)b_i] + \beta \mathbb{E} \{ \mathcal{P}_i(k'_i, b'_i, z'_i) [f_k(k'_i, z'_i) + 1 - \delta] | z_i \} = 0$$

$$[b'_i] : \frac{\partial Q_i(k'_i, b'_i, z_i)}{\partial b'_i} [b'_i - (1 - \theta_i)b_i] + Q_i(k'_i, b'_i, z_i) - \beta \mathbb{E} \{ \mathcal{P}_i(k'_i, b'_i, z'_i) [\theta_i + (1 - \theta_i)Q_i(k''_i, b''_i, z'_i)] | z_i \}$$

=0

$$\begin{aligned}
 \frac{1}{Q_t} \mathbb{E}_t [\mathcal{P}_{t+1} (\theta + (1 - \theta) Q_{t+1})] &= \frac{(1 + \rho) \mathbb{E}_t [\mathcal{P}_{t+1} (\theta + (1 - \theta) Q_{t+1})]}{\mathbb{E}_t [\mathcal{P}_{t+1} (\theta + (1 - \theta) Q_{t+1})] + \mathbb{E}_t [(1 - \mathcal{P}_{t+1}) \phi k' / b']} \\
 &= (1 + \rho) \left( 1 + \frac{\mathbb{E}_t [(1 - \mathcal{P}_{t+1}) \phi k' / b']}{\mathbb{E}_t [\mathcal{P}_{t+1} (\theta + (1 - \theta) Q_{t+1})]} \right)^{-1} \\
 &= (1 + \rho) (1 + \Lambda)^{-1}
 \end{aligned}$$

where

$$\Lambda \equiv \frac{\mathbb{E}_t [(1 - \mathcal{P}_{t+1}) \phi k' / b']}{\mathbb{E}_t [\mathcal{P}_{t+1} (\theta + (1 - \theta) Q_{t+1})]}$$

## Misallocation formula: proof

▷ back

- Formally, planner's problem is now the same as solving  $Y = \max_{\{k_i\}_i} \int_0^1 f_i(k_i) di$ , where  $f_i(k_i)$  is now expected output
- Apply Hughes and Majerovitz (2024), noting  $\frac{dY}{dk} = r^{social} + \delta$

$$\log(Y^*/Y^{DE}) \approx \frac{1}{2} \cdot \mathcal{E} \cdot \log \left( 1 + \frac{\text{Var}(r^{social})}{(\mathbb{E}[r^{social}] + \delta)^2} \right)$$

- $\mathcal{E}$  is (negative) elasticity of output w.r.t. cost of capital  $(r^{social} + \delta)$

- $\mathcal{E}_i$  is the elasticity of expected output with respect to the cost of capital
- Assume that  $f(k, z) = z \cdot k^\alpha$  and there is no default, then

$$\mathcal{E} = \frac{\alpha}{1 - \alpha}$$

- $\alpha = \frac{1}{3}$  implies  $\mathcal{E} = \frac{1}{2}$

## Summary statistics

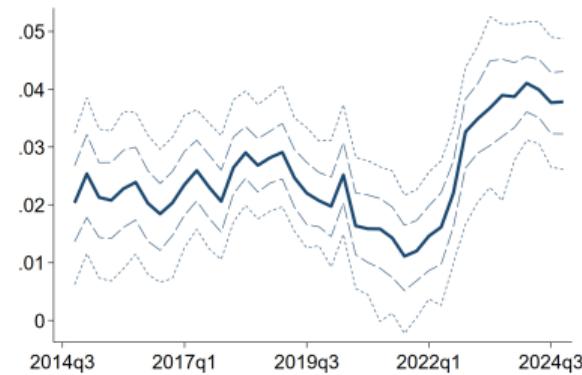
[▷ back](#)

	Mean	St. Dev.	p10	p50	p90
Interest rate	4.18	1.69	2.21	3.94	6.60
Maturity (yrs)	6.83	4.65	3.00	5.00	10.25
Real interest rate	2.39	1.24	0.88	2.33	4.00
Prob. Default (%)	1.45	2.53	0.19	0.85	2.88
LGD (%)	34.41	13.17	16.00	35.60	50.00
Loan amount (M)	10.75	67.58	1.11	2.57	22.92
Sales (M)	1,269.93	6,051.48	2.16	58.50	1,560.10
Assets (M)	1,760.37	8,894.15	1.07	35.55	1,782.22
Leverage (%)	72.17	24.68	42.68	71.29	100.00
Return on assets (%)	27.60	58.51	4.56	15.76	47.81
N Loans	65,284				
N Firms	38,751				
N Fixed Rate	32,592				
N Variable Rate	32,692				

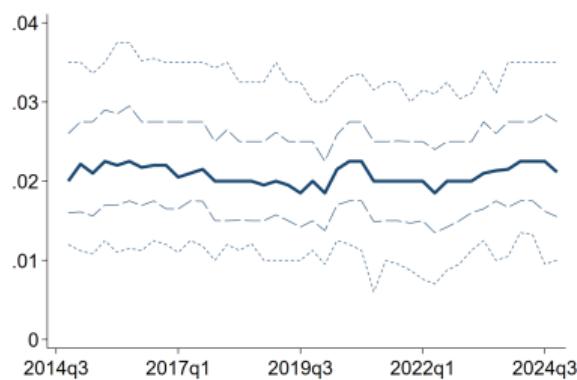
# Time series for averages and quantiles: real interest rate, PD, LGD

▷ back

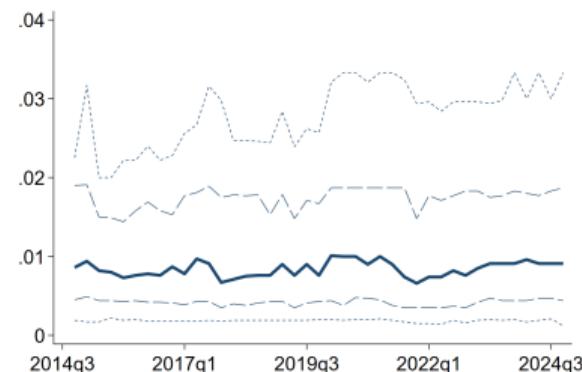
Real interest rate



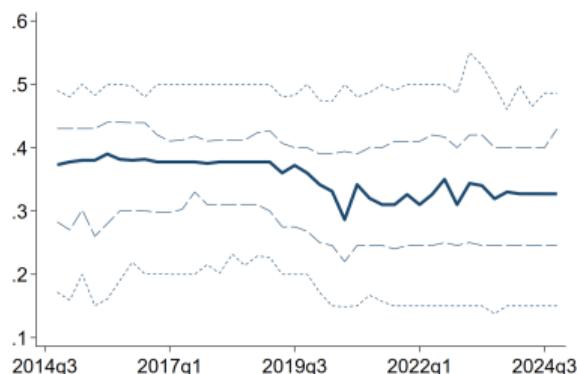
Interest rate spread (var.)



Probability of default



Loss given default



# Data cleaning and sample construction

▷ back

We use FR Y-14Q Schedule H.1 data from 2014Q4 to 2024Q4.

## Borrower Filters:

- Drop loans without a Tax ID
- Keep only Commercial & Industrial loans to nonfinancial U.S. addresses
- Drop borrowers with NAICS codes:
  - 52 (Finance and Insurance), 92 (Public Administration)
  - 5312 (Real Estate Agents), 551111 (Bank Holding Companies)

# Data cleaning and sample construction, cont'd

▷ back

## Loan Filters:

- Drop loans with:
  - Negative committed exposure
  - Utilized exposure exceeding committed exposure
  - Origination after or maturity before report date
- Keep only “vanilla” term loans (Facility type = 7)
- Drop loans with:
  - Mixed-interest rate structures
  - Maturity less than 1 year or longer than 10 years
  - Implausible interest rates or spreads (outside 1st - 99th percentile)
  - Missing or invalid PD/LGD values (outside [0, 1])
  - PD = 1 (flagged as in default)

## Forward interest rate expectations

▷ back

To estimate  $\rho_i$  for floating rate loans, need estimates of  $\mathbb{E}_0 [r_t] + s_i$

- Floating rate loans charge reference rate + spread
- Approximate LIBOR/SOFR using Treasury forward yield curve estimates (Gürkaynak, Sack, and Wright, 2007)
- Average spread between SOFR and Treasury rates 2018-2025  $\simeq 2$  basis points
- Assume expectations hypothesis: long rates reflect expected short rates
- Back out  $\mathbb{E}_0 [r_t] + s_i$  for each loan, using treasury forward rate plus loan's spread

## Firm cost of capital: model

▷ back

$$Q_t = \frac{\mathbb{E}_t [\mathcal{P}_{t+1} (\theta + (1 - \theta) Q_{t+1}) + (1 - \mathcal{P}_{t+1}) \phi k_{t+1} / b_{t+1}]}{1 + \rho}$$

Note that

$$\begin{aligned} Q_t &= Q_t^P + Q_t^D \\ Q_t^P &= \frac{\mathbb{E}_t [\mathcal{P}_{t+1} (\theta + (1 - \theta) Q_{t+1})]}{1 + \rho} \\ Q_t^D &= \frac{\mathbb{E}_t [(1 - \mathcal{P}_{t+1}) \phi k_{t+1} / b_{t+1}]}{1 + \rho} \end{aligned}$$

That is, we strip the bond into the payment in repay ( $Q_t^P$ ) and the payment in default ( $Q_t^D$ ). Then:

$$\Lambda = \frac{\mathbb{E}_t [(1 - \mathcal{P}_{t+1}) \phi k_{t+1} / b_{t+1}]}{\mathbb{E}_t [\mathcal{P}_{t+1} (\theta + (1 - \theta) Q_{t+1})]} = \frac{Q_t^D}{Q_t^P}$$

## Firm cost of capital: measurement

The firm defaults with probability  $(1 - P)$  and the lender recovers  $(1 - LGD)$ . Hence

$$Q_t^{D,data} = \frac{(1 - P)(1 - LGD)}{1 + \rho}$$

For the payment portion notice that at issuance we have the following condition

$$1 = \sum_{s=1}^T \left[ \frac{P^s \mathbb{E}_t [r_{t+s}] + P^{s-1} (1 - P) (1 - LGD)}{(1 + \rho)^s} \right] + \frac{P^T}{(1 + \rho)^T}$$

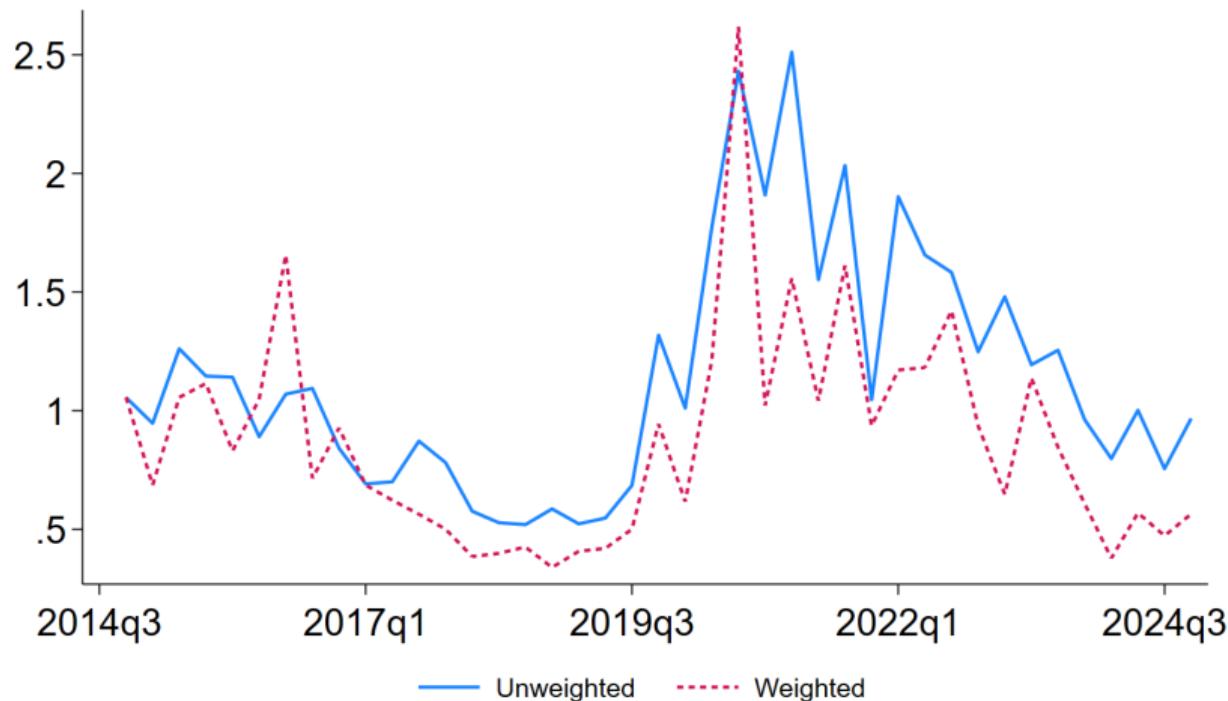
$$1 = \frac{(1 - P)(1 - LGD)}{1 + \rho} + P \frac{\mathbb{E}_t [r_{t+1}]}{1 + \rho} + \left( \sum_{s=2}^T \left[ \frac{P^s \mathbb{E}_t [r_{t+s}] + P^{s-1} (1 - P) (1 - LGD)}{(1 + \rho)^s} \right] + \frac{P^T}{(1 + \rho)^T} \right)$$

So, we can define  $Q_t^{P,data}$  as  $1 = Q_t^{P,data} + Q_t^{D,data}$  so  $Q_t^{P,data} = 1 - Q_t^{D,data}$ . Finally

$$\Lambda^{data} = \frac{Q_t^{D,data}}{Q_t^{P,data}} = \frac{(1 - P)(1 - LGD)}{1 + \rho - (1 - P)(1 - LGD)}$$

## Misallocation, weighted by loan size

▷ back



# Decomposing misallocation

▷ back

**Counterfactual I:** What if all lenders have the same  $\bar{\rho}$ ?

$$1 + r_{social}^{cf,I} = \overline{(1 + \rho)\mathcal{M}} + (lev - \mathcal{M}) \cdot PD \cdot (1 - LGD)$$

Heterogeneity in  $r_{social}^{cf}$  → Misallocation due to financial frictions

**Counterfactual II:** what if we equalize financial frictions?

$$1 + r_{social}^{cf,II} = (1 + \rho) \mathcal{M} + \overline{(lev - \mathcal{M}) \cdot PD \cdot (1 - LGD)}$$

Heterogeneity in  $r_{social}^{cf}$  → Misallocation due to heterogeneous cost of capital

## “Real yield”: decomposing $\rho$

▷ back

- The “real yield” is the implied  $\rho_{i,t}^*$  when  $P_{i,t} = 1$

$$1 = \sum_{s=1}^{T_{i,t}} \left[ \frac{\mathbb{E}_t(r_{i,t,s})}{\left(1 + \rho_{i,t}^*\right)^s \cdot \mathbb{E}_t(\Pi_{t,s})} \right] + \frac{1}{\left(1 + \rho_{i,t}^*\right)^{T_{i,t}} \cdot \mathbb{E}_t(\Pi_{t,T_{i,t}})}$$

- Real yield independent of  $P_{i,t}, LGD_{i,t}$
- Only affected by losses through the contractual rate  $r$

$$\mathcal{M} = \frac{1 - \gamma \times \frac{Qb'}{k'} \times \frac{\partial \log Q}{\partial \log k'}}{1 + \gamma \times \frac{\partial \log Q}{\partial \log b'}}$$

Need  $Q$ ,  $\gamma$ , and firm leverage  $Qb'/k'$  to compute  $\mathcal{M}$

1. To compute  $Q$ , assume that loans are perpetuities that decay at a geometric rate  $\theta$ , discounted at the loan's real interest rate  $r$ :

$$Q = \frac{\theta + (1 - \theta)Q}{1 + r} = \frac{\theta}{r + \theta}$$

$r$  is directly observed in the data, and we can approximate  $\theta = 1/T$

2. Guess a functional approximation  $Q(z, k, b, \rho)$
3. Estimate  $\log \hat{Q}(z, k, b, \rho)$  for every loan origination; compute partial derivatives
4. At steady state,  $\gamma = \theta = 1/T$

## Estimating $\mathcal{M}$ : $Q$ elasticities

▷ back

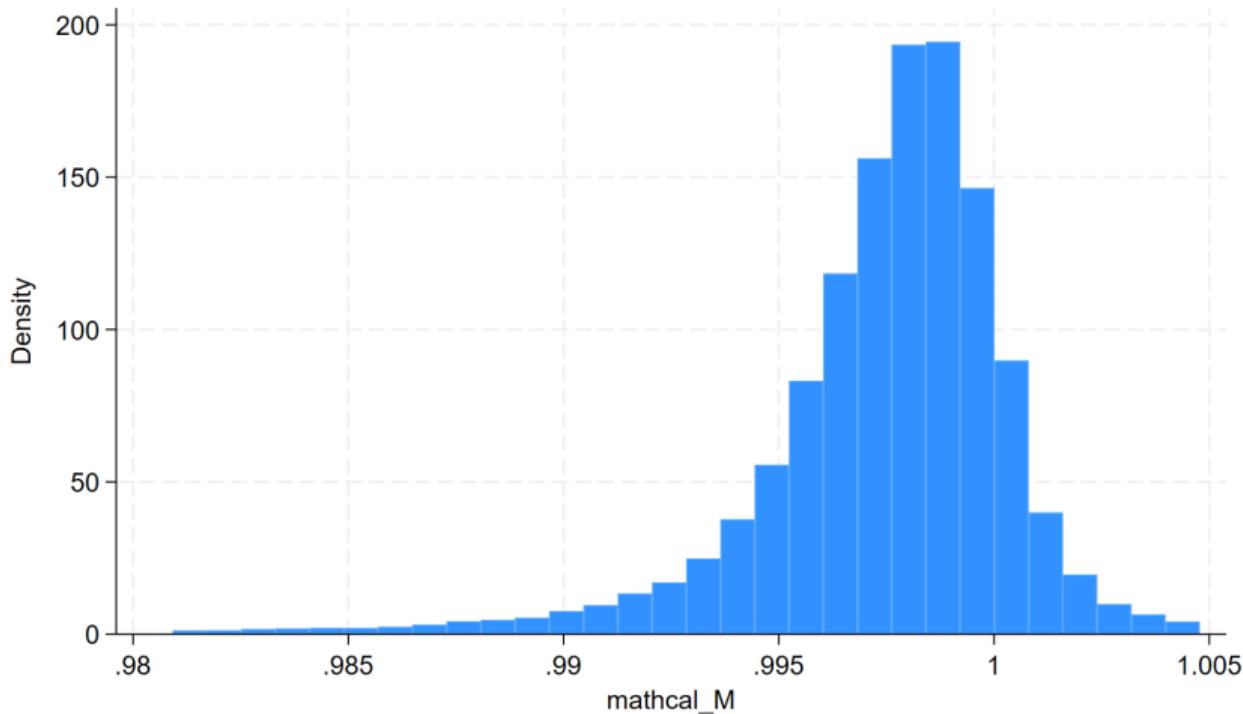
- We approximate (the log of)  $Q$  as a polynomial of firm capital, borrowing, productivity and  $\rho$

$$\begin{aligned}\log Q_i = & \alpha + \beta_k \log k_i + \beta_b \log b_i + \beta_z \log z_i + \beta_\rho \rho_i \\ & + \beta_{k,k} (\log k_i)^2 + \beta_{k,b} \log k_i \times \log b_i + \beta_{k,z} \log k_i \times \log z_i + \beta_{k,\rho} \log k_i \times \rho_i \\ & + \beta_{b,b} (\log b_i)^2 + \beta_{b,z} \log b_i \times \log z_i + \beta_{b,\rho} \log b_i \times \rho_i \\ & + \beta_{z,z} (\log z_i)^2 + \beta_{z,\rho} \log z_i \times \rho_i + \beta_{\rho,\rho} (\rho_i)^2 + \epsilon_i\end{aligned}$$

- Capital: tangible assets
- Borrowing: total debt owed by the firm at loan origination
- Productivity: sales over tangible assets
- This allows us to compute  $\frac{\partial \log Q}{\partial \log k'}$  and  $\frac{\partial \log Q}{\partial \log b'}$

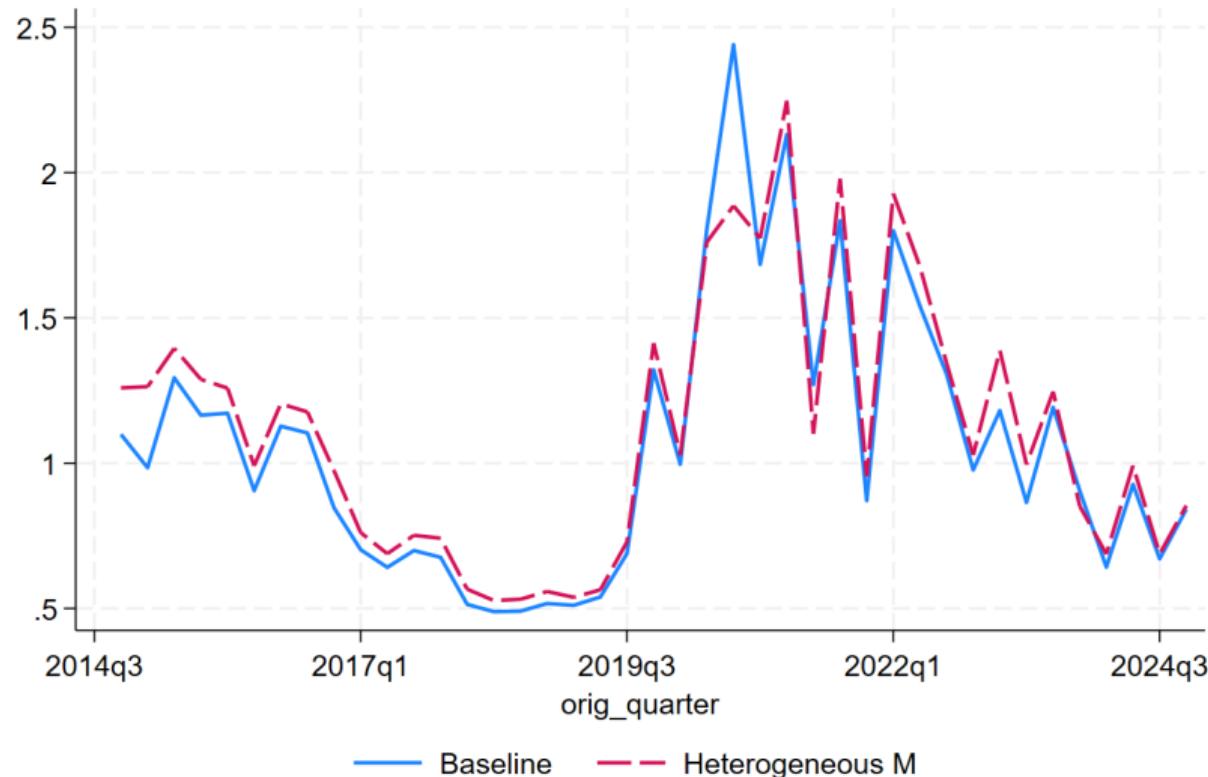
# Estimating $\mathcal{M}$ : results

▷ back



# Misallocation with heterogeneous $\mathcal{M}$

▷ back



# Variance decomposition

▷ back

	Time	Bank	Firm	Loan
Contractual rate	69	2	15	15
Real rate	49	4	25	22
$\rho$	43	4	23	30
$r^{firm}$	17	4	31	49
$r^{social}$	35	4	25	36

Notes: 1,844 firms and 16,088 loans. Sample restricted to firms with at least five securities.

Within-period dispersion of  $r^{social}$ :

- Bank 6%
- Firm 38%
- Loan 55%

Large dispersion even within a quarter-bank-firm relationship.

# Validation: $r^{social}$ correlates with standard measures of ARPK

▷ back

	(1)	(2)	(3)	(4)	(5)
	log(ARPK), Sales	log(ARPK), EBITDA	log(ARPK), Sales	log(ARPK), EBITDA	log(ARPK), VA
log( $r^{social} + \delta$ )	0.15*** (0.03)	0.24*** (0.04)	0.16** (0.07)	0.15* (0.08)	0.39*** (0.07)
Observations	59294	57334	4184	4072	3432
Adj. R2	0.27	0.22	0.68	0.52	0.61
NAICS4, Quarter FE	yes	yes	yes	yes	yes
Sample	Y-14	Y-14	Compustat	Compustat	Compustat

Robust standard errors in parentheses

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

# ARPK-based misallocation

▷ back

Focus on Compustat firms to make measures comparable

	$r^{social} + \delta$	Sales Capital	EBITDA Capital	Value Added Capital
Var(log)	0.01	0.19	0.24	0.21
Misallocation (%)	0.36	4.75	6.20	5.28

- Our measure looks only at misallocation coming from heterogeneity in the cost of capital
- ...but does not require detailed data on firm financials (i.e., value added)
- $\implies$  directly applicable to most existing credit registries

# ARPK-based misallocation measures

▷ back

	(1) log ARPK, Sales	(2) log ARPK, EBITDA	(3) log ARPK, Sales	(4) log ARPK, EBITDA	(5) log ARPK, VA
log( $r^{social} + \delta$ )	0.15*** (0.03)	0.24*** (0.04)	0.16** (0.07)	0.15* (0.08)	0.39*** (0.07)
Observations	59294	57334	4184	4072	3432
Adj. R2	0.27	0.22	0.68	0.52	0.61
NAICS4, Quarter FE	yes	yes	yes	yes	yes
Sample	Y-14	Y-14	Compustat	Compustat	Compustat
Var(log ARPK)	1.97	1.52	0.19	0.24	0.21
Misalloc., ARPK, %	63.63	46.08	4.75	6.20	5.28
Var(log( $r^{social} + \delta$ ))	0.04	0.04	0.01	0.01	0.01
Misalloc., $r^{social} + \delta$ , %	0.96	0.96	0.36	0.36	0.36

Standard errors in parentheses

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

## The 2020–2021 increase in misallocation

1. Driven by dispersion in lender discount rates  $\rho_i$ , not financial frictions.
2. Sharp rise in the coefficient of variation of  $\rho_i$ .
3. Variance of  $\rho_i$  increases due to increased dispersion of expected losses.

## 2. The CV of $\rho_i$ increased during 2020-21



- Policy rates  $\downarrow$  in 2020-21  $\Rightarrow$  mean  $\rho_i \downarrow$
- $\sigma(\rho_i) \uparrow$  during this period - why?

$\Rightarrow$  2. Coefficient of variation of  $\rho_i \uparrow$

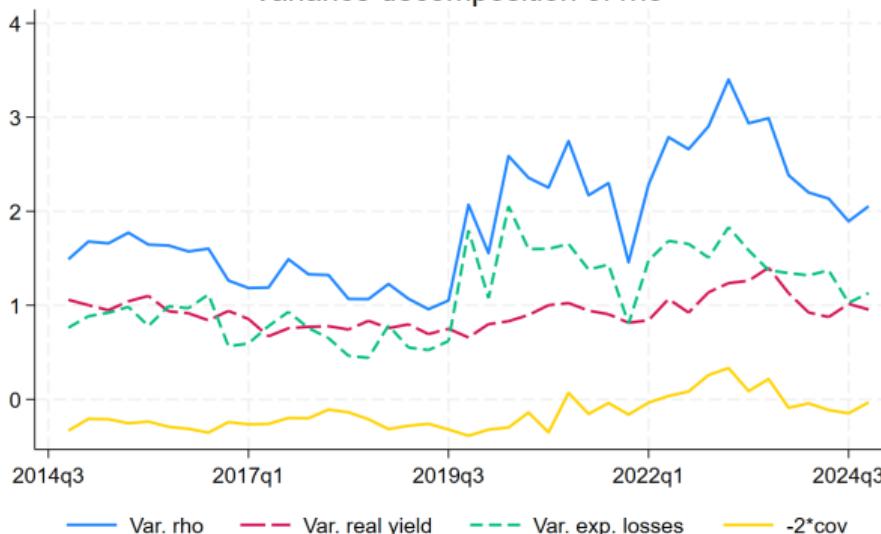
### 3. Variance of $\rho$ related to variance of expected losses

- Compute “real yield”  $\rho_{i,t}^*$ : lender discount rate if no default

▷ real yield

- Decomposition:  $\rho_i = \underbrace{\rho_i^*}_{\text{real yield}} - \underbrace{[\rho_i^* - \rho_i]}_{\text{exp. losses}}$

Variance decomposition of rho



Variance of  $\rho_i$ :

$$\mathbb{V} [\text{yield}] + \mathbb{V} [\text{exp. losses}] - 2\mathbb{C} [\text{yield}, \text{exp. losses}]$$

- Increase in variance explained by exp. losses
- Covariance falls in absolute value
- ↑ in dispersion of exp. losses without ↑ in dispersion of contractual rates