

# The poor, the rich, and the credit channel of monetary policy

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- 1. Are the poor affected differently by monetary policy compared to the wealthy?
- 2. And if so, what role does credit play in this configuration?

#### Theoretically,

- Wealth could influence lending decision if banks perceive it as collateral (Holmstrom and Tirole, 1997 QJE)
  - Explicit for full liability firms
  - Implicit for limited liability firms, if banks believe they will be able to persuade the owner during times of corporate stress
- Collateral/Wealth might matter less during expansionary monetary policy: risk-taking channel (loannidou et al., 2014 RF; Jimenez et al., 2014 ECMA)

## Findings

- 1. Monetary policy transmits to business loan approval rates, but transmission diminishes with wealth of business owner
  - ▶ Owners at 25<sup>th</sup> pctile in wealth distribution: 1 sd.  $\uparrow$  shadow rate  $\rightarrow$  4.4 pp  $\downarrow$  loan approval
  - ▶ Owners at 75<sup>th</sup> pctile in wealth distribution: 1 sd.  $\uparrow$  shadow rate  $\rightarrow$  0 pp  $\downarrow$  loan approval
- 2. Loan approval significantly affects owners' ability to generate future income & wealth

## Findings

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  - ▶ Owners at 75<sup>th</sup> pctile in wealth distribution: 1 sd.  $\uparrow$  shadow rate  $\rightarrow$  0 pp  $\downarrow$  loan approval
- 2. Loan approval significantly affects owners' ability to generate future income & wealth
- 3. 1 + 2 imply that monetary policy may have redistributive effects
- 4. The effect transmits especially via banks with low liquidity and low capital

#### Related literature

1. Monetary policy and economic inequality

- Amberg et al., 2021; Andersen et al., 2021; Auclert, 2019; Coibion et al., 2017; Holm et al., 2021; Jasova et al., 2023; Kaplan et al., 2018; Moser et al., 2024; Mumtaz and Theophilopoulou, 2017
- 2. Credit channel of monetary policy
  - Bernanke and Blinder, 1992; Ciccarelli et al., 2015; Heider et al., 2019; Hulsewig et al., 2006; Ioannidou et al., 2014; Jimenez et al., 2014; Kashyap and Stein, 2000; Kishan and Opiela, 2000; Maddaloni and Peydro, 2011
- 3. Real effects of loan approval
  - Banerjee and Duflo, 2014; Berg, 2018; Delis et al., 2023

#### Two datasets with distinct (dis)advantages:

- 1. Confidential and detailed data from a large North European bank (2002-2018)
  - + Loan applications (disentangle loan demand from supply)
  - + Private wealth observed by bank
  - + Credit Score observed (disentangle effect of business quality from private wealth)
  - Only one bank
- 2. Matched Survey on the access to finance of enterprises (SAFE) with ORBIS data (2009-2020)
  - + Loan applications (disentangle loan demand from supply)
  - + Multiple banks and countries (external validity + study bank heterogeneity)
  - Private wealth approximated through past dividends
  - Credit score not observed

#### Data set 1

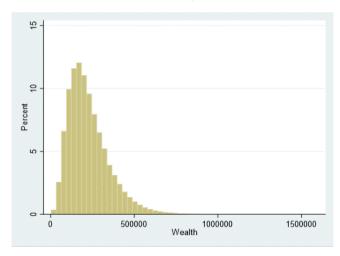
137.000 loan applications from 16.000 SMEs to a large North European bank (2002-2018).

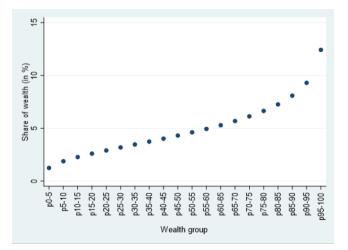
- **Loan (application) info**: e.g. granted or not, amount, spread, collateral, default status
- Private wealth of the majority owner
  - All money in savings and deposit accounts, stocks, bonds, etc. (= Holm et al. 2021 JPE)
  - Excluding: house value and mortgage debt, the (net present) value of their business

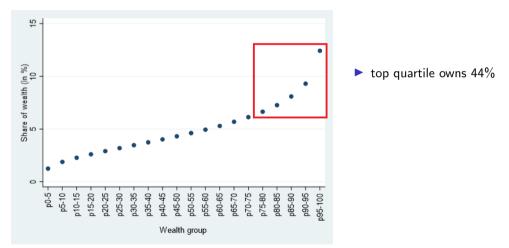
#### Credit score of the business

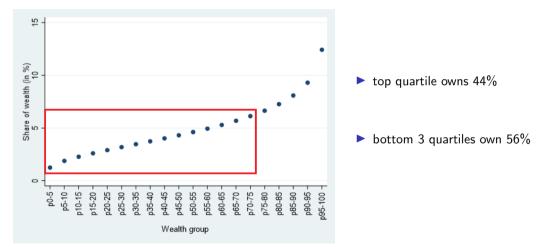
- Assessment of the firm's quality and repayment capacity, as given by the bank
- Includes both soft and hard information
- Known cutoff determines loan approval
- Other info about majority owners (e.g. education, children, age, gender) as well as the business (income statements and balance sheets)

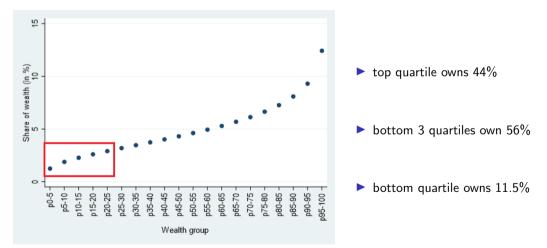
Figure: Histogram of business owners' wealth (from our large euro area bank sample)





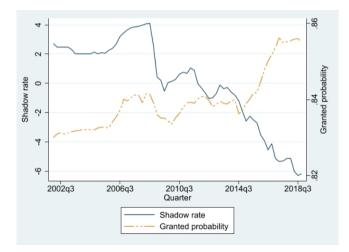






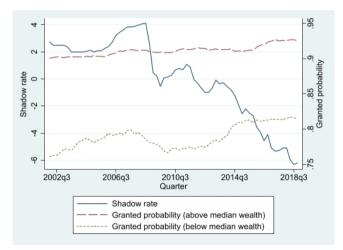
#### Summary statistics: monetary policy and loan approval

Figure: Evolution of the shadow rate and average loan approval rate during sample period



#### Summary statistics: monetary policy and loan approval

Figure: Evolution of the shadow rate and average loan approval rate during sample period



## Methodology

#### **Regression model:**

$$Granted_{ift} = \beta_1 Wealth_{it} + \beta_2 Monetary Policy_t + \beta_3 Monetary Policy_t \times Wealth_{it} + \beta_4 X'_{ift-1} (+\delta_f + \rho_t) + \epsilon_{iot}$$
(1)

- Monetary Policy<sub>t</sub> is either the shadow rate (Wu and Xia, 2016) or monetary policy shocks (Altavilla et al., 2019).
- X<sub>ift-1</sub> :i) credit score; income, education, age, firm size, ROA, cash holding, n of applications; ii) ROE, equity ratio, cash flow, size; bank liquidity and capital ratios
- Expectation:  $\beta_1 > 0$ ,  $\beta_2 < 0$ ,  $\beta_3 > 0$
- We can limit the analysis to a narrow bandwidth around the known Credit Score cutoff to ensure firms are observationally equal. Manipulation test

## Summary statistics: narrow bandwidth around cutoff

	Mean	St. dev.	Min.	Max.	Mean diff.	Std. error
<b>a</b> .				_		
Granted	0.66	0.47	0	1	1	0
Credit score	0.06	0.16	-0.30	0.30	0.277	0.073
Shadow rate	-0.19	3.28	-6.40	4.28	0.017	0.016
Wealth	11.50	0.60	7.21	13.97	0.020	0.026
Income	10.69	0.30	9.73	11.49	0.027	0.026
Education	2.13	0.99	0	5	0.033	0.021
Age	44.80	15.86	20	76	0.238	0.252
Children	1.86	1.47	0	6	0.004	0.036
Gender	0.81	0.39	0	1	0.009	0.006
Fim size	12.72	0.40	9.96	14.09	0.011	0.007
Firm leverage	0.20	0.03	0.15	0.74	0.002	0.002
Firm ROA	0.06	0.09	-0.40	0.49	0.005	0.002
Firm cash holdings	0.07	0.03	0.01	0.16	0.000	0.001

## Wealth, loan approval, and monetary policy

			Granted		
Wealth					
Shadow rate					
Shadow rate $ imes$ Wealth					
Lambda					
Credit score bandwidth	$[-\infty, +\infty]$	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.1,0.1
Observations	137,321	32,310	32,310	32,310	18,028
Observations (first stage) Adj. R-squared	0.723	0.706	0.720	414,730	0.819
Controls and Firm FE Year:quarter FE	Yes	Yes	Yes Yes	Yes	Yes

## Wealth, loan approval, and monetary policy

			Granted		
Wealth	0.016***	0.013***	0.013***	0.014***	0.012***
	(0.002)	(0.003)	(0.003)	(0.002)	(0.002)
Shadow rate	-0.239***	-0.212***		-0.206***	-0.202***
	(0.063)	(0.074)		(0.056)	(0.066)
Shadow rate $ imes$ Wealth	0.022***	0.017***	0.018***	0.020***	0.018***
	(0.005)	(0.006)	(0.004)	(0.005)	(0.005)
Lambda				-0.172	
				(0.164)	
Credit score bandwidth	$[-\infty, +\infty]$	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.1,0.1]
Observations	137,321	32,310	32,310	32,310	18,028
Observations (first stage)				414,730	
Adj. R-squared	0.723	0.706	0.720		0.819
Controls and Firm FE	Yes	Yes	Yes	Yes	Yes
Year:quarter FE			Yes		

Economic effect of a 1 standard deviation increase in the Shadow rate (3.3pp)

- ▶ Wealth at 25th percentile of the distribution is EUR 120,000 (In=11.68).
- ▶  $3.3 \times (-0.212 + 0.017 \times 11.68) = -4.4$ pp lower approval likelihood
- ▶ Wealth at 75th percentile of the distribution is EUR 270,000 (In=12.50).
- ▶  $3.3 \times (-0.212 + 0.017 \times 12.5) = 0.1$ pp no effect on approval likelihood

## Role of wealth as collateral: stronger for fully liable entrepreneurs?

		Granted	
Wealth	0.013***	0.013***	0.012***
	(0.003)	(0.004)	(0.002)
Shadow rate	-0.212***	-0.196***	-0.223***
	(0.074)	(0.113)	(0.066)
Shadow rate $ imes$ Wealth	0.017***	0.014**	0.019***
	(0.006)	(0.007)	(0.005)
Credit score bandwidth	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]
Firm type	all	limited liability	full liability
Observations Adj. R-squared	32,310 0.706	27,140 0.709	5,170 0.698
Controls and Firm FE	Yes	Yes	Yes

#### Role of wealth as collateral: does it reduce loan default?

	<b>Default</b> 3 years after origination							
Wealth	-0.032***	-0.030***	-0.047***					
	(0.009)	(0.007)	(0.010)					
Credit score	-0.051***	-0.052***	-0.051***					
	(0.020)	(0.020)	(0.017)					
Firm type	all	limited liability	full liability					
Observations	77,510	61,935	15,875					
Adj. R-squared	0.703	0.716	0.695					
Controls & Firm FE	Yes	Yes	Yes					

## Role of wealth as collateral: is it used to (re)capitalize the firm?

	(1) Ca	(2) apital increa	(3) ise		(5) apital increa rm-years wit	
		all firm-years	5	deteriorat	ing default p	probability
Wealth decrease	0.687*** (0.103)	0.489*** (0.091)	0.709*** (0.106)	0.511*** (0.152)	0.347*** (0.094)	0.655*** (0.110)
Firm type	all	limited liability	full liability	all	limited liability	full liability
Observations	32,310	27,140	5,170	16,014	13,420	2,594
Adj. R-squared	0.81	0.70	0.83	0.55	0.50	0.57
Controls Firm FE	No No	No No	No No	No No	No No	No No
Year FE	No	No	No	No	No	No

## Role of loan approval for future income and wealth generation

- We want to identify the effect of the loan decision on the loan applicant's future income and wealth.
- ▶ We follow Berg (2018) and use an RDD regression model:

$$y_{it+3} = \alpha_0 + \alpha_1 \; \text{Granted}_{it} + \alpha_2 \; (x_{it} - \overline{x}) + \alpha_3 \; \text{Granted}_{it} \times (x_{it} - \overline{x}) + \alpha_4 \; y_{it} + \alpha_5 \; x_{it-1}^{'} + \nu_{it} \tag{2}$$

- where y is ln(annual income) or ln(private wealth)
- $x_{it}$  is the Credit Score and  $\overline{x}$  is the known cutoff Manipulation test

## Does loan approval matter for future income and wealth generation?

	Income	Wealth
	3 years after loan origination	3 years after loan origination
Granted	0.072***	0.053***
	(0.015)	(0.010)
Shadow rate	-0.012**	-0.011**
Shadow fate	(0.006)	(0.005)
Credit score	0.006	0.005
	(0.004)	(0.004)
${\sf Credit\ score\ }\times\ {\sf Granted}$	-0.009	-0.006
	(0.006)	(0.005)
Income	0.036***	
	(0.007)	
Wealth		0.025***
		(0.005)
Observations	77,510	77,510
Adj. R-squared	0.629	0.703
Controls & Firm FE	Yes	Yes

#### Graphical representation of RDD

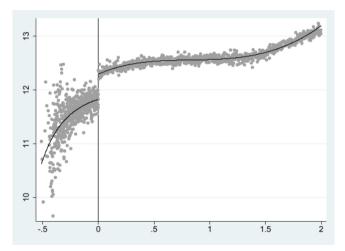


Figure: Graphical result of the RDD model: Effect on future wealth

- ▶ We use monetary policy shocks as in Altavilla et al. (2019 JME) instead of shadow rates Table
- > We look at loan amounts and loan spreads instead of the propensity to grant a loan Table

## External validity using the SAFE survey data

▶ We use the matched **SAFE-ORBIS** data from the ECB:

- 10.000 (private, profit-oriented and family-owned) SMEs
- 19 Euro Area countries
- > 2009-2020
- 16.000 loan applications (at different banks)

 $\rightarrow$  With this data, we:

- confirm the effect in a less sophisticated, but significantly broader setting
- can look into bank heterogeneity

## Wealth, loan approval and monetary policy using the SAFE survey data

		Gran	ited	
Wealth	0.07***	0.05**	0.05**	0.08
	(0.02)	(0.02)	(0.02)	(0.07)
Shadow rate	-1.51***	-1.54***	· · /	. ,
	(0.33)	(0.36)		
Shadow rate $ imes$ Wealth	2.21***	2.20***	2.24***	2.90***
	(0.52)	(0.53)	(0.53)	(0.87)
Observations	15,627	15,627	15,627	9,556
No. firms	9,158	9,158	9,158	3,087
R-squared	0.12	0.15	0.16	0.65
Control variables:	Yes	Yes	Yes	Yes
Country FE		Yes	Yes	Yes
Wave FE			Yes	Yes
Firm FE				Yes

## Wealth, loan approval and monetary policy using the SAFE survey data

Economic effect of a 1 standard deviation increase in the Shadow rate (2.4pp)

- ► At 25th percentile of distribution, the effect is 2.4 x (-0.0154 + 0.022×0) = -3.7pp lower approval likelihood
- ► At 75th percentile of distribution, the effect is 2.4 × (-0.0154 + 0.022×0.11) = -3.1pp lower approval likelihood
- ► At 95th percentile of distribution, the effect is 2.4 × (-0.0154 + 0.022×0.59) = -0.6pp lower approval likelihood

## Bank heterogeneity using the SAFE survey data

Banks with:	High Liquidity			Low Liquidity			
		Granted		Granted			
Wealth	0.19***	0.07***	-0.10	0.08	0.15**	0.19	
	(0.04)	(0.04)	(0.16)	(0.07)	(0.06)	(0.16)	
Shadow rate	-1.69***	-0.72		-3.28***	-2.46***		
	(0.50)	(0.51)		(0.40)	(0.43)		
Shadow rate $ imes$ Wealth	3.61**	2.21**	1.13	4.02**	5.39***	6.17***	
	(1.39)	(1.01)	(1.70)	(1.65)	(1.43)	(2.11)	
Observations	0.442	0.200	1 710	2 5 1 0	0.200	1 565	
0.0001.10110110	2,443	2,329	1,719	2,519	2,328	1,565	
No. firms	1,215	1,311	521	1,372	1,263	500	
R-squared	0.01	0.11	0.62	0.02	0.11	0.65	
Control variables		Yes	Yes		Yes	Yes	
Country FE, wave FE, firm FE			Yes			Yes	

## Bank heterogeneity using the SAFE survey data

Banks with:	F	ligh Capita		L	.ow Capital		
		Granted		Granted			
Wealth	-0.11	-0.11	0.18	0.21**	0.23***	-0.03	
Shadow rate	(0.10) -3.78*** (0.65)	(0.09) -2.30*** (0.60)	(0.29)	(0.09) -3.99*** (0.60)	(0.08) -3.05*** (0.60)	(0.15)	
Shadow rate $\times$ Wealth	2.11 (3.19)	1.36 (3.27)	-0.11 (7.07)	4.39** (1.86)	5.37*** (1.59)	2.91 (2.00)	
Observations No. firms	1,536 923	1,422 858	872 308	2,174 1,205	2,042 1,108	1,378 444	
R-squared Control variables	0.04	0.12 Yes	0.72 Yes	0.04	0.13 Yes	0.67 Yes Yes	
	0.01	• •	••••=				

- Monetary policy is passed through to borrowers via loan approval rates.
- ▶ The strength of this transmission depends on the wealth of the business owner.
  - ▶ Owners at 25<sup>th</sup> pctile in wealth distribution: 1 sd.  $\uparrow$  shadow rate  $\rightarrow$  4.4 pp  $\downarrow$  loan approval
  - ▶ Owners at 75<sup>th</sup> pctile in wealth distribution: 1 sd.  $\uparrow$  shadow rate  $\rightarrow$  0 pp  $\downarrow$  loan approval

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  - ▶ Owners at 25<sup>th</sup> pctile in wealth distribution: 1 sd.  $\uparrow$  shadow rate  $\rightarrow$  4.4 pp  $\downarrow$  loan approval
  - ▶ Owners at 75<sup>th</sup> pctile in wealth distribution: 1 sd.  $\uparrow$  shadow rate  $\rightarrow$  0 pp  $\downarrow$  loan approval
- This transmission happens primarily through banks with below average liquidity ratios and capital ratios.
- ▶ There are redistributive effects as loan approval matters for income & wealth generation.

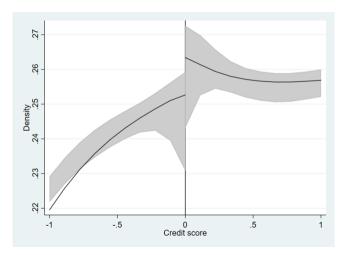
Thank you for your attention!

# Appendix

Appendix

#### Graphical test credit score manipulation (Back Methodology) (Back RDD)

Figure: Manipulation test RDD



## Summary statistics

	Mean	St. dev.	Min.	Max.	Mean diff.	Std. erro
Granted	0.66	0.47	0	1	1	0
Shadow rate	-0.19	3.28	-6.40	4.28	0.017	0.016
Monetary policy shock	0.02	2.31	-7.10	4.74	0.004	0.008
Wealth	11.50	0.60	7.21	13.97	0.020	0.026
Income	10.69	0.30	9.73	11.49	0.027	0.026
Education	2.13	0.99	0	5	0.033	0.021
Age	44.80	15.86	20	76	0.238	0.252
Dependents	1.86	1.47	0	6	0.004	0.036
Gender	0.81	0.39	0	1	0.009	0.006
Fim size	12.72	0.40	9.96	14.09	0.011	0.007
Firm leverage	0.20	0.03	0.15	0.74	0.002	0.002
Firm ROA	0.06	0.09	-0.40	0.49	0.005	0.002
Firm cash holdings	0.07	0.03	0.01	0.16	0.000	0.001
Number of applications	7.22	1.48	1	9	0.091	0.070
Credit score	0.06	0.16	-0.30	0.30	0.277	0.073
Default	0.04	0.11	0	1	0.000	0.003
Loan amount	1.98	0.54	0.71	7.01	0.099	0.008
Maturity	44.13	35.94	4	233	0.841	0.570
Loan provisions	0.46	0.50	0	1	0.023	0.036
Collateral	0.69	0.45	0	1	0.011	0.027

## Wealth, loan approval, and monetary policy shocks (Back

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	Granted	Granted	Granted	Granted	Granted	Granted	Granted	Granted	Granted
Wealth	0.012***	0.011***	0.010***	0.010***	0.012***	0.019***	0.013***	0.015***	0.010***
Mon. pol. shock	(0.003) -0.296***	(0.002) -0.257***	(0.003) -0.229***	(0.003) -0.266***	(0.004) -0.233**	(0.003)	(0.003) -0.257***	(0.005) -0.269***	(0.002) -0.237***
Mon. pol. shock $ imes$ Wealth	(0.112) $0.022^{**}$ (0.009)	(0.091) 0.020** (0.008)	(0.094) 0.016** (0.080)	(0.086) 0.023*** (0.007)	(0.087) 0.017** (0.007)	0.022*** (0.006)	(0.067) 0.021*** (0.007)	(0.074) 0.024*** (0.008)	(0.083) 0.017*** (0.006)
Mon. pol. shock $ imes$ Credit score	(0.009)	(0.008)	(0.080)	(0.007)	(0.007) 0.064*** (0.010)	(0.000)	(0.007)	(0.008)	(0.000)
Lambda					(0.010)		-0.171 (0.163)	-0.194 (0.179)	
Credit score bandwidth	$[-\infty, +\infty]$	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.1,0.1]
Firm type	all	all	limited liability	unlimited liability	all	all	all	all	all
Observations Observations (first stage)	121,540	28,750	24,150	4,600	28,750	28,750	28,750 367,988	28,750 599,214	16,101
Adj. R-squared	0.718	0.707	0.708	0.696	0.776	0.720	,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	,	0.803
Controls and Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Year:quarter FE	No	No	No	No	No	Yes	No	No	No

## Wealth, loan approval, and shadow rates

	(1) Granted	(2) Granted	(3) Granted	(4) Granted	(5) Granted	(6) Granted	(7) Granted	(8) Granted	(9) Granted
Wealth	0.016***	0.013***	0.013***	0.012***	0.005***	0.013***	0.014***	0.014***	0.012***
	(0.002)	(0.003)	(0.004)	(0.002)	(0.002)	(0.003)	(0.002)	(0.002)	(0.002)
Shadow rate	-0.239***	-0.212***	-0.196***	-0.223***	-0.116*		-0.206***	-0.226***	-0.202***
	(0.063)	(0.074)	(0.113)	(0.066)	(0.061)		(0.056)	(0.049)	(0.066)
Shadow rate $ imes$ Wealth	0.022***	0.017***	0.014**	0.019***	0.020***	0.018***	0.020***	0.020***	0.018***
	(0.005)	(0.006)	(0.007)	(0.005)	(0.007)	(0.004)	(0.005)	(0.004)	(0.005)
Shadow rate $ imes$ Credit score					0.044***				
					(0.009)				
Lambda							-0.172	-0.162	
							(0.164)	(0.135)	
Credit score bandwidth	$[-\infty, +\infty]$	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.3,0.3]	[-0.1,0.1]
Firm type	all	all	limited liability	unlimited liability	all	all	all	all	all
Observations	137,321	32,310	27,140	5,170	32,310	32,310	32.310	32.310	18,028
Observations (first stage)					. ,		414,730	675,327	
Adj. R-squared	0.723	0.706	0.709	0.698	0.935	0.720			0.819
Controls and Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Year:quarter FE	No	No	No	No	No	Yes	No	No	No

#### Wealth, loan outcomes, and monetary policy Back

	(1)	(2)	(3)	(4)
	Loan amount	Loan amount	Spread	Spread
Wealth	0.014**	0.012**	-0.055***	-0.048***
Shadow rate	(0.006) -0.319***	(0.006)	(0.012) -0.131**	(0.010)
Shadow rate $ imes$ Wealth	(0.095) 0.030***		(0.063) -0.099***	
	(0.009)		(0.017)	
Monetary policy shock		-0.428***		0.120
		(0.162)		(0.102)
Monetary policy shock $ imes$ Wealth		0.032***		-0.120***
		(0.012)		(0.021)
Observations	26,972	24,004	26,972	24,004
Adj. R-squared	0.840	0.831	0.732	0.726
Controls	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes
Year:quarter FE	Yes	Yes	Yes	Yes