Sharing the Pain? Credit Supply and Real Effects of Bank Bail-ins

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Introc	luction
	accion.

Data

Within-Firm Analysis

Motivation

- Bailouts were the preferred resolution mechanism during the crisis
 - In the euro area, accumulated financial sector assistance reached 8% of GDP between 2008-2014 (ECB, 2015)
- ► However, large costs for taxpayers (Philippon and Salord, 2017)
 - Public guarantees also lead to increased risk-taking (Gropp et al., RFS 2011; Dam and Koetter, RFS 2012)
- Regulatory response: introduction of formal bank bail-in regimes
 - Private investors and creditors (e.g., junior bondholders) bear the costs of restoring a distressed bank
 - Banks can theoretically now fail without resorting to public funding, though extraordinary public support is still possible (Avgouleas and Goodhart, JFR 2015)
- Despite the hypothetical advantages of bank bail-ins (e.g., Klimek et al., JEDC 2015), there is little empirical evidence on the effects of this new resolution tool on credit supply or the real economy ...

Research Questions

- Are bank bail-ins effective in preventing a credit crunch?
- What are the effects of this resolution tool on the real economy?

Ideal experiment:

- 1. Random bank failure(s) and resolution(s)
- 2. Firms and banks are randomly matched

In reality:

- 1. Exploit unexpected bank failure in Portugal and subsequent bail-in
- 2. Endogenous bank-firm match: credit that firm *i* obtains from bank *b* is an equilibrium outcome resulting from credit supply and demand \Rightarrow Isolate credit supply by comparing lending across banks *within* the same firm (Khwaja and Mian, AER 2008)

Introduction	Data	Within-Firm Analysis	Cross-Sectional Analysis	Conclus
Background				

- ► Unexpected collapse and subsequent resolution of a major bank in Portugal (BES) in August 2014 → "one of Europe's biggest financial failures" (FT, 2014)
 - ▶ 3rd largest bank operating in Portugal: €81bn assets in 2013:Q4
 - Considered a Significant Credit Institution by the ECB
 - Market share of 19% of credit granted to non-financial corporations
- Resolution implied that:
 - 1. Sound activities and assets were transferred to a "good bank"
 - 2. Shareholders and junior bondholders were left with the toxic assets in a "bad bank" which is in liquidation
 - €4.9bn of capital of newly-created bank fully provided by Portugal's Bank Resolution Fund → But loan to the Fund from 8 member banks (€0.7bn) and another from the Portuguese government (€3.9bn)
- Portuguese resolution regime in force was, in substance, very similar to the final European resolution framework (World Bank, 2016)

Introduction	Data	Within-Firm Analysis	Cross-Sectional Analysis
Background			

▶ Bank failure was purely idiosyncratic → due to "seriously detrimental management acts" (BoP, 2014) where managers secretly lent money to firms of the Group owned by the same family (Economist, 2014)



Other Banks: CGD, BPI, BCP. Source: Thomson Reuters Datastream

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Introduction	Data	Within-Firm Analysis	Cross-Sectional Analysis	Conclusion
Data				

- 1. <u>Portuguese Credit Register:</u> quarterly information on credit exposures above 50 euros between *all* banks and *all* non-financial firms operating in Portugal
- 2. <u>Individual Information of Interest Rates</u>: matched firm-bank interest rate information on all new loans from June 2012
- 3. <u>Firm and Bank Balance Sheet Data:</u> financial information with annual frequency for virtually *all* Portuguese firms and banks
- ▶ 40,927 firms, 98 banks
- 116,245 firm-bank lending relationships
- Period: 2013 to 2015

1st Step: Within-Firm Analysis

Are there any significant changes in the supply of credit by banks more exposed to the bail-in?

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► Following Khwaja and Mian (AER 2008) → isolate credit supply (from credit demand) by comparing the change in lending by banks more and less exposed to the shock *within* the same firm

Within-Firm Specification

 $\Delta log(Credit)_{bi} = \beta(BankExposure_b) + \delta' X_{bi} + \alpha_i + \varepsilon_{bi}$ (1)

- $\Delta log(Credit)_{bi}$: change in (log) committed credit from bank b to firm i
 - As in Khwaja and Mian (AER 2008), quarterly data for each credit exposure time-averaged into a single pre (2013:Q4-2014:Q2) and post-shock (2014:Q3-2015:Q3) period

▶ *BankExposure_b*: percentage assets of each bank exposed to the bail-in

- $1.\ \%$ assets that was effectively bailed-in for the resolved bank
- 2. Specific contribution to the ad-hoc loan granted to the Bank Resolution Fund for the 8 participating banks (as a % assets)
- 3. And 0 for all other banks
- α_i: firm FE capturing observed and unobserved firm-specific determinants of credit flows, including changes in credit demand

1.1. Credit supply – within-firm estimates: main results

		$\Delta log Tota$	$lCredit_{bi}$		$\Delta logCre$	$ditLines_{bi}$
	(1)	(2)	(3)	(4)	(5)	(6)
Bank Exposure	-0.989***	*-1.143***	*-1.520*		-2.723***	*
	(0.311)	(0.320)	(0.824)		(0.863)	
Bank Exposure $ imes$ SMEs	. ,	. ,	. ,	-1.441*	. ,	-2.659***
				(0.829)		(0.881)
Bank Exposure \times Large Firms				-3.133***		-4.048***
				(0.836)		(0.866)
No. Observations	116,245	116,245	116,245	116,245	39,573	39,573
No. Firms	40,927	40,927	40,927	40,927	14,320	14,320
Adj. R^2	0.001	0.047	0.049	0.050	0.103	0.103
Bank Controls	N	Ν	Y	Y	Y	Y
Firm FE	N	Y	Y	Y	Y	Y
No. Bank Relationships > 1	Y	Y	Y	Y	Y	Y
Credit Lines with \neq Banks	Ν	Ν	Ν	Ν	Y	Y

▶ 1 SD increase in bank exposure to the bail-in \rightarrow 3.0% decrease of total credit and 5.7% decrease of granted credit lines

1.1. Credit supply – within-firm estimates: main results



Note: each coefficient estimate in each quarter corresponds to a different KM regression - where the outcome variable is $\Delta log(Credit)_{bi}$ between that quarter and 2014:Q2 (before shock). Bank Exposure standardized with mean 0 and SD of 1

Introduction Data Within-Firm Analysis Cross-Sectional Analysis Conclusion

- 1.2. Credit supply within-firm estimates: other tests
 - Heterogeneity: ↓ reduction of credit for firms with ↓ profitability and with NPLs in the pre-period; Credit supply contraction concentrated in firms that had the bailed-in bank as their main lender trable
 - 2. Alternative Bank Exposure Measure: change in bank-specific CDS spread from 1 month before to the day of the resolution \rightarrow for the 4 Significant Institutions with available CDS data \frown Table
 - Include Firms With Single Bank Relationships → replace firm FE with a group (location-industry-size) FE to control for credit demand as in De Jonghe et al. (2018)
 - 4. Use total credit excluding credit lines as outcome variable, or alternative pre and post-shock windows rable

2nd Step: Cross-Sectional (Between-Firm) Analysis

Given the tightening of credit, were firms more exposed to the bail-in (i.e., that were borrowing from more exposed banks) able to:

- (i) substitute funding from other banks?
- (ii) maintain interest rates on credit?
- (iii) sustain level of investment and employment?

- ► Within-firm specification not appropriate to examine aggregate effects → ignores the *extensive margin of credit*
- Solution: estimate a related cross-sectional (between-firm) effect of firm exposure to the shock while still accounting for demand factors

Between-Firm Specification

$$\Delta \log(Y)_i = \beta(FirmExposure_i) + \tau' F_i + \delta' \bar{X}_i + \hat{\alpha}_i + \varepsilon_i$$
⁽²⁾

- ► FirmExposure_i: weighted average of BankExposure_b across all banks lending to a firm → using as weights the pre-period share of total credit of each bank
- ► *F_i*: firm characteristics measured in the pre period. We also include industry and district FE
- ▶ X
 _i: bank controls averaged at the firm-level according to the share of total credit granted to the firm by each bank
- Control for credit demand by including in (2) the vector of firm-level dummies estimated from the KM within-firm regression $(1) \rightarrow \hat{\alpha}_i$

e.g., Bonaccorsi Di Patti and Sette (JFI 2016), Cingano et al. (RFS 2016)

2.1. Credit supply – cross-sectional estimates

	$\Delta logTotalCredit_i$		$\Delta logCree$	$ditLines_i$
	(1)	(2)	(3)	(4)
Firm Exposure	-0.374 (0.352)		-1.785*** (0.485)	
Firm Exposure \times SMEs		-0.378 (0.355)		-1.839*** (0.572)
Firm Exposure \times Large Firms		-0.267 (0.607)		-0.526 (1.135)
No. Observations / Firms Adj. R^2	40,927 0.378	40,927 0.378	14,320 0.175	14,320 0.175
Firm and Bank Controls	Y	Y	Y	Y
Credit Demand	Y	Y	Y	Y
Industry and District FE	Y	Y	Y	Y
No. Bank Relationships > 1	Y	Y	Y	Y
Credit Lines with \neq Banks	Ν	N	Y	Y

- No relative change in overall credit for more exposed firms
- But binding contraction of credit lines to more exposed SMEs: a 2.2% decrease for a 1 SD increase in firm exposure to the resolution

2.2. Firm exposure to the bail-in and credit conditions

	$\Delta IntRate_i$ All New Credit		ΔInt Credit L	$\Delta IntRate_i$ Δ Credit Lines Only		$\Delta Maturity_i$		$\Delta Collateral_i$	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
Firm Exposure	2.335 (3.027)	1	17.97***	:	-53.29*** (12.00)		2.125** (0.793)	k I	
Firm Exposure \times SMEs	(***=*)	1.495	()	17.67***	()	-51.82***	()	2.130**	
Firm Exposure \times Large Firm	S	(3.331) 24.47*** (6.805)	¢	(2.923) 25.44** (11.61)		(4.361) -91.80** (42.29)		(0.797) 1.984** (0.762)	
No. Observations / Firms Adj. R^2	31,472 0.097	31,472 0.097	12,429 0.082	12,429 0.082	31,472 0.031	31,472 0.031	31,472 0.076	31,472 0.076	
Loan Characteristics Firm and Bank Controls	Y Y	Y Y	Y Y	Y Y	Y Y	Y Y	Y Y	Y Y	
Credit Demand Industry and District FE	Y Y	Y Y	Y Y	Y Y	Y Y	Y Y	Y Y	Y Y	
No. Bank Relationships > 1	Ŷ	Ŷ	Ŷ	Ŷ	Ŷ	Ŷ	Ý	Ŷ	

► 1 SD increase in firm exposure to the bail-in → 20bps increase in interest rates on credit lines, 1 month decrease in loan maturity, and 2.8 percentage points increase in the share of collateralized credit



Remaining Question: Did the resolution have any effects on real outcomes e.g., investment, employment?

3.1. Firm exposure to the bail-in and investment

	$\Delta logTangibleAssets_i$			Δlog	$sets_i$	
	(1)	(2)	(3)	(4)	(5)	(6)
Firm Exposure	-1.680***	-1.497***	k	-1.349***	·-1.000**	:
	(0.312)	(0.327)		(0.249)	(0.396)	
Firm Exposure $ imes$ SMEs	. ,	· /	-1.531***	· · ·	· /	-1.018**
			(0.337)			(0.394)
Firm Exposure \times Large Firms	5		-0.489			-0.460
			(1.322)			(1.242)
No. Observations / Firms	14,320	40,927	40,927	14,320	40,927	40,927
Adj. R^2	0.045	0.041	0.041	0.043	0.039	0.039
Firm and Bank Controls	Y	Y	Y	Y	Y	Y
Credit Demand	Y	Y	Y	Y	Y	Y
Industry and District FE	Y	Y	Y	Y	Y	Y
No. Bank Relationships > 1	Y	Y	Y	Y	Y	Y
Credit Lines with \neq Banks	Y	Ν	Ν	Y	Ν	Ν

 $\blacktriangleright~1~\text{SD}$ increase firm exposure $\rightarrow~2.0\%$ decrease in investment at SMEs

3.2. Firm exposure to the bail-in and employment

	$\Delta logNo.Employees_i$			$\Delta log No. Worked Hours$		
	(1)	(2)	(3)	(4)	(5)	(6)
Firm Exposure	-1.183**	-0.945***	k	-1.644***	-1.154**	*
	(0.410)	(0.182)		(0.326)	(0.163)	
Firm Exposure \times SMEs	· · ·	` ´	-0.971***	e Č	. ,	-1.182***
			(0.180)			(0.169)
Firm Exposure \times Large Firms	5		-0.190			-0.325
			(0.501)			(0.525)
No. Observations / Firms	14,320	40,927	40,927	14,320	40,927	40,927
Adj. R^2	0.080	0.041	0.041	0.054	0.047	0.047
Firm and Bank Controls	Y	Y	Y	Y	Y	Y
Credit Demand	Y	Y	Y	Y	Y	Y
Industry and District FE	Y	Y	Y	Y	Y	Y
No. Bank Relationships > 1	Y	Y	Y	Y	Y	Y
Credit Lines with \neq Banks	Υ	Ν	Ν	Y	Ν	Ν

 \blacktriangleright 1 SD increase firm exposure \rightarrow 1.3% decrease in the no. of employees and 1.5% decrease in total number of worked hours at SMEs



- ► If dampening effects of bank resolution are indeed driven by tighter credit line limits for more exposed SMEs → we should observe heterogeneous effects according to their pre-shock liquidity position
- Why? Option for firms to access liquidity from credit lines should be more valuable when internal liquidity is scarce (Campello et al., RFS 2011)
- Berg (RFS 2018): while liquid SMEs absorb credit supply shocks by using existing cash buffers, iliquid SMEs increase cash holdings when a loan application is rejected and reduce investment and employment

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3.3. The role of firms' internal liquidity

	$\frac{\Delta log}{CashHoldings_i}$	Δlog Investment _i	Δlog Employment _i
Firm Exposure x High Liquidity SMEs	· -13.579*** (3.800)	-0.093	-0.113
Firm Exposure x Low Liquidity SMEs	(3.039) 13.416*** (4.249)	-1.680*** (0.420)	-1.644*** (0.135)
No. Observations / SMEs Adj. R^2	40,234 0.022	40,234 0.040	40,234 0.067
Firm and Bank Controls Credit Demand Industry and District FE	Y Y Y	Y Y Y	Y Y Y
No. Bank Relationships > 1	Y	Y	Y

 Low liquidity SMEs more exposed to the bail-in respond by increasing cash holdings while decreasing investment and employment

Summary and Policy Implications

(1) Fears that enforcing market discipline through a bail-in would result in panic and contagion effects did not materialize

▶ Banks more exposed to the resolution significantly reduced credit supply to SMEs and large firms → but affected firms compensated the contraction in overall credit with other sources of funding

(2) However, a bank bail-in is not a silver bullet

- Resolution led to lower investment and employment at SMEs with higher exposure to the bail-in and lower pre-shock internal liquidity
- Driven by binding contraction of granted credit lines to these SMEs and increase in their cash holdings

Appendix - Within-firm estimates: Heterogeneity

(9)
-0.439 (0.830)
3.132*** (0.399)
116,253
40,927
Y
Ŷ
3

Appendix - Robustness Tests

	$\Delta log Total Credit_{bi}$					
	Alternative Bank Exposure Measure (CDS Spread Reaction)			Ind W Ban	rms Dne 1ship	
	(1)	(2)	(3)	(4)	(5)	(6)
Bank Exposure	-1.917***	-2.031***	:	-0.714***	-1.339**	
Bank Exposure \times SMEs	(0.297)	(0.345)	-1.787*** (0.350)	(0.201)	(0.049)	-1.283* (0.652)
Bank Exposure \times Large Firm	ıs		-5.956*** (1.703)			-2.915*** (0.667)
No. Observations	40,783	40,783	40,783	160,457	160,457	160,457
No. Firms Adj. R^2	17,445 0.001	17,445 0.054	17,445 0.054	85,139 0.053	85,139 0.055	85,139 0.055
No. Banks	4	4	4	98	98	98
Bank Controls	N	Y	Y	Ν	Y	Y
Firm FE	N	Y	Y	N	N	N
No. Bank Relationships > 1	N Y	N Y	N Y	N	Y N	Y N

Appendix - Robustness Tests

	Δla (Excludin	$ \begin{array}{cc} \Delta logCredit_{bi} & \Delta logCredit_{bi} \\ \text{Excluding Credit Lines)} & (2014:Q2-20) \end{array} $. <i>logCredi</i> 4:Q2-201	it _{bi} 5:Q3)	
	(1)	(2)	(3)	(4)	(5)	(6)
Bank Exposure	-0.963***	-1.108		-1.430***	-2.000**	
Bank Exposure \times SMEs	(0.500)	(0.000)	-1.063	(0.505)	(0.020)	-1.812** (0.832)
Bank Exposure $ imes$ Large Firms	5		-1.925* (0.986)			-5.460*** (0.927)
No. Observations	96,584	96,584	96,584	97,130	97,130	97,130
No. Firms Adj. R^2	35,365 0.001	35,365 0.015	35,365 0.015	34,861 0.001	34,861 0.029	34,861 0.030
Bank Controls	Ν	Y	Y	Ν	Y	Y
Firm FE No. Bank Relationships > 1	N Y	Y Y	Y Y	N Y	Y Y	Y Y

Appendix - Robustness Tests

	Δ Total	log Credit _i	Δ Cred	∆log itLines _i	Δ Total	log Credit _i	Δ Credit	log $Lines_i$
	Alterna	tive Firm Exposure Measure (Bank Exposure: CDS Spread Reaction) Alternative Firm Exposure Measure (Bank Exposure: Dummy = 1 for Bailed-in Bank)			e Measure mmy ank)			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Firm Exposure	-0.520 (0.446)		-2.747 (1.585)		-0.032 (0.021)		-0.114*** (0.034)	
Firm Exposure \times SMEs	. ,	-0.605 (0.454)	· /	-3.051* (1.522)	· · ·	-0.032 (0.022)	· · ·	-0.117*** (0.034)
Firm Exposure \times Large Firm	s	`0.888´ (1.103)		`3.291 [´] (2.275)		-0.028 (0.039)		-0.033 (0.116)
No. Observations / Firms Adj. R^2	17,444 0.299	17,444 0.299	5,420 0.162	5,420 0.162	40,927 0.378	40,927 0.378	14,320 0.175	14,320 0.175
Firm and Bank Controls	Y	Y	Y	Y	Y	Y	Y	Y
Credit Demand Industry and District FE	Y Y	Y Y	Y Y	Y Y	Y Y	Y Y	Y Y	Y Y
No. Bank Relationships > 1	Y	Y	Y	Y	Y	Y	Y	Y

Introduction

Appendix - New vs. Existing Lending Relationships

	$NewLending\\Relationship_i$		$\Delta logTot$ (except bai	$alCredit_i$ led-in bank)
	(1)	(2)	(3)	(4)
Firm Exposure	-0.659 (0.423)		4.566*** (0.558)	
Firm Exposure \times SMEs	· /	-0.674 (0.433)	· · · ·	4.540*** (0.585)
Firm Exposure \times Large Firms	S	-0.220 (0.809)		5.359*** (1.042)
No. Observations / Firms Adj. R^2	40,927 0.058	40,927 0.058	40,927 0.342	40,927 0.342
Firm and Bank Controls	Y	Y	Y	Y
Credit Demand	Y	Y	Y	Y
No. Bank Relationships > 1	Ŷ	Y	Y	Ý

- More exposed firms as likely to start new lending relationships as less exposed firms – but average firm already had 4 bank relationships
- $\blacktriangleright~1$ SD increase firm exposure $\rightarrow~5.94\%$ increase in credit from other banks