

## When Credit Dries Up: Job Losses in the Great Recession\*

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\*The paper is the sole responsibility of its authors. The views presented here do not necessarily reflect those of either the Banco de España or the Eurosystem

# Motivation



- **The Great Recession raised concerns about the economic implications of the current lack of credit.**

“Policy makers want to support [credit] markets because the decline in lending is seen as a primary factor in the slow recovery” (IMF, Global Financial Stability Report, 2013).”

- **These concerns led to massive injections of capital in banks with solvency problems (over €600 Bn in Europe).**

- **However, recent evidence on the real effects of credit supply shocks and the economic benefits of bailouts is scarce. Why?**

- It’s difficult to find good datasets on bank credit, and
- Also because there are challenging identification issues.

# This paper



- **This paper contributes to the literature that explore the real effects of credit supply shocks.**
- **In particular, we study the impact of the fall in bank lending during the Great Recession on firm-level employment in Spain.**
  - We have access to a unique dataset with confidential information about all bank loans to non-financial firms above 6,000€ and loan applications to non-current banks.
  - These data are linked to firm and bank balance sheet data.
  - The data allow us to reconstruct the entire banking relations and credit histories of over 150,000 non-financial firms.
- **To the best of our knowledge, it is the most extensive matched firm-loan-bank data set ever assembled for the analysis of credit supply shocks.**

# The Spanish experience during the Great Recession (GR)



- **The Spanish economy offers an ideal setting for exploring how shocks to credit supply affect to the real economy:**

- Spanish firms rely heavily on bank credit.

*Loans from credit institutions to non-financial corporations represents the 86% of GDP vs 62% in EU (2006).*

- Spanish firms, mostly SMEs, are highly leveraged which made them vulnerable to the reduction in credit supply during the GR.

- This credit supply shock was originated by a boom-bust cycle in housing prices with catastrophic effects on bank solvency.

*Interesting parallels with other countries, like Ireland and the US, that also suffered a boom-bust cycle in the housing market and an exceptional rise in unemployment.*

# The Basic Challenges



- **How to disentangle credit supply and credit demand shocks?**
  - A financial crisis may force banks to reduce credit supply, but it may also induce firms to reduce credit demand.
  - The economic troubles of firms may reinforce or even cause the hardship of banks, inducing problems of reversed causality.
- **The second main challenge is to control for selection effects, given that, on average, healthy banks use to work with better firms than weal banks.**

# Our Approach



- **We exploit the pronounced cross-sectional differences in lender health at the onset of the crisis.**

- We consider two types of lenders: weak banks vs healthy banks.

*Definition: a bank is considered weak if it was bailed out by the State (mostly after 2010).*

*All but one of them were cajas de ahorros.*

- We exploit that *weak banks* reduced credit more than the other banks *before* their intervention.
  - Given that during period of crisis firms cannot easily switch to new funders, the reduction of loan supply from weak banks will transmit to their borrowers
  - Thus, we compare the change in employment from 2006 to 2010 at two sets of firms with a pre-crisis exposure to either weak or healthy banks.
- **We exploit the richness of our database to face the identification problems commented.**

# Our Approach

- **We will start showing that weak banks cut more credit supply than non-weak banks**
  - First, we will work at firm-bank level where we can control for demand
  - Then, we will work at the level of the firm to show that companies were unable to completely offset this reduction of credit by substituting loans with the help of other banks.
- **Furthermore, we will analyze how this shock has an impact on firms' employment**
  - We will propose a diff-in-diff exercise to show that firms that worked more intensively with a weak bank at the beginning of the crisis reduced their employment more than similar firms that didn't.
    - *We will start with a reduced form*
    - *Then, we will show that results are driven by the credit channel previously discovered*
    - *Finally, we will test the robustness of the results changing:*
      - The definition of weak banks: using an ex-ante measure
      - The treatment variable

# Quasi-Experimental Techniques



- **The most recent literature exploits quasi-experimental techniques.**

- **Large external shocks to the banking sector.**

*Chava and Purnanandam (2011); Benmelech et al. (2012); Ongena et al. (2013).*

- **Cross-sectional differences in firms' financial vulnerability at the start of the Great Recession.**

*Almeida et al. (2011), Benmelech et al. (2011), Boeri et al. (2013) and Garicano and Steinwender (2013).*

- **Cross-sectional differences in lender health at the onset of the crisis.**

*Greenstone et al (2014), Chodorow-Reich (2012).*

- **All find sizeable effects on real variables, but none of them is able to reconstruct the complete banking relations of firms nor do they have such high-quality controls.**



# Identification issues



- **Disentangle changes in credit supply from concurrent changes in credit demand.**

- At the firm-bank level, we control for credit demand using the procedure of Khwaja and Mian (2008).

- At the firm level, we use a large set of firm covariates and even firm fixed effects

- **Selection effects (since we find significant differences between firms in the treatment and control groups).**

- We include controls for trends in large sets of municipality, industry and firm control variables.

- The potential endogeneity of banking relationship is additionally addressed in an IV model and with matching techniques*

# Underlying assumptions



- i.e. necessary conditions for the existence of real effects and their correct identification:

- A1. Weak banks reduce loan supply more than other banks during the crisis.
- A2. There are financial market frictions: firms borrowing from weak banks at the beginning of the crisis cannot easily switch to other healthier banks.
- A3. Firms must have been unable to foresee the solvency problems of weak banks when they formed their banking relationships.

→ **We will show that all conditions hold.**

# Summary of Results



- **Our most conservative estimates show that:**

- Controlling for selection, attachment to weak banks caused an extra employment reduction of **2.8 percentage points** between 2006 and 2010.

*This corresponds to 24% of job losses.*

- Credit constraints affect both intensive and extensive margins.
- The results are very robust: various estimation techniques (diff in diffs, instrumental variables, exact and propensity score matching) and many robustness checks.
- We find sizeable differences across industries and firms with different credit histories.



# Plan of the talk

- 1. Theoretical Background**
- 2. The financial crisis in Spain**
- 3. Data**
- 4. Empirical strategy**
- 5. Robustness checks**
- 6. Conclusions**

# Theoretical Background

## Financial Accelerator Mechanisms



- **Shocks in credit markets may amplify, propagate or even initiate shocks to the real economy.**
  - Bernanke and Gertler (1989, 1995).
  - Bernanke et al. (1996).
  - Gertler and Kiyotaki (2010)

# Theoretical Background

## Credit Frictions



- A causal relationship between the differences in lender health and differential employment growth at the firm level requires the existence of:
  - **Credit frictions:** Firms subject to credit restrictions from their banks must not be able to (readily) switch to other banks or alternative sources of funding (Gobbi and Settle, 2014)
  - **Asymmetric information:** Most explanations for credit friction rely on the assumption of asymmetric information between borrowers and lenders (Sharpe, 1990)

# Theoretical Background

## Relationship Banking



- **This literature explains analyzes the advantages of stable banking relationships**
  - Stable relationships help to reduce the agency cost of lending, as banks acquire soft information on their clients (Sharpe, 1990).
  - The superior information may provide better access to credit at the same bank when capital is scarce (Bolton et al., 2013).
  - While a switch to a new bank may be costly due to a lemon's problem, especially in recessions (Gobbi and Sette, 2014).
  - Ambiguous predictions for the optimal number of banking relationships.

# The Spanish experience



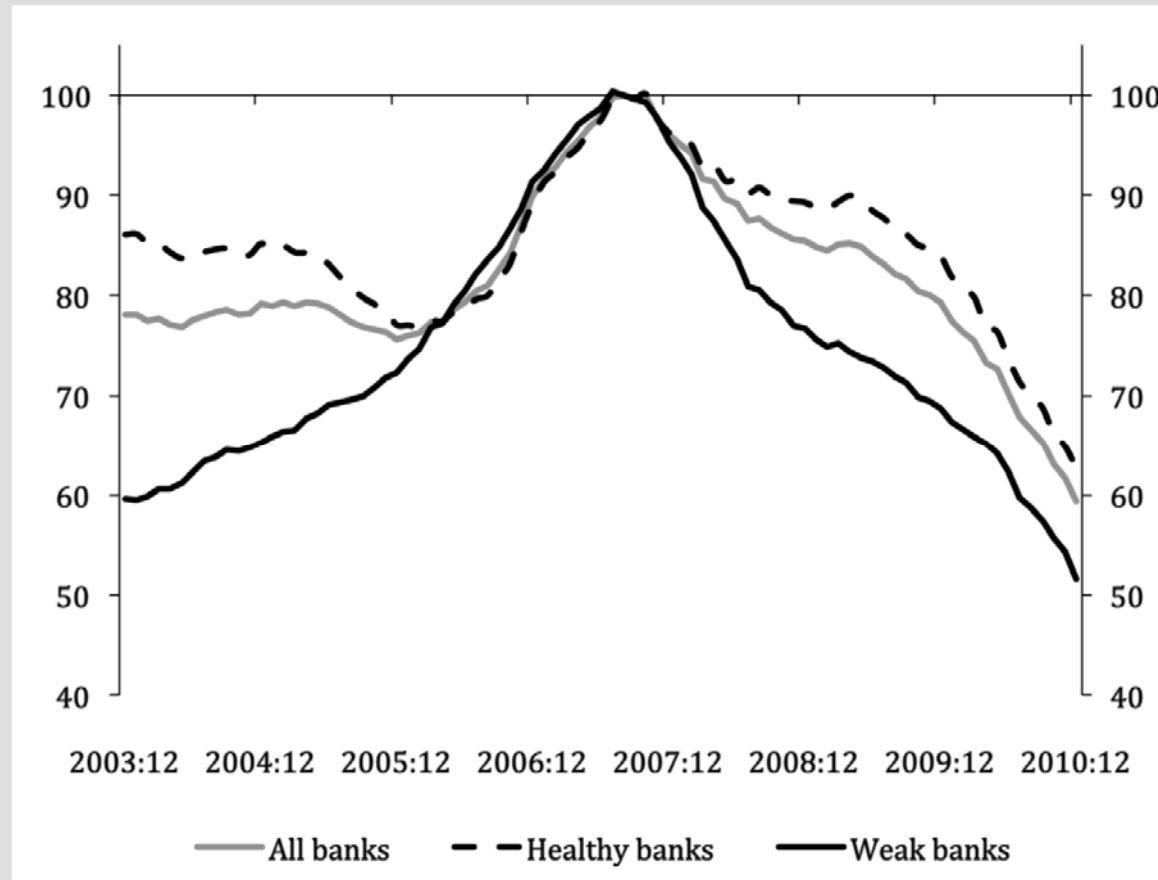
- **Spanish economy experienced a severe credit crunch during the GR.**
- **GDP and Employment**
  - Expansion, 1996-2007: GDP 3.7%; employment 4.1%.
  - Recession, 2008-2010 : GDP -1.1%; employment -3.2%.
- **A boom-bust cycle in housing priced by cheap credit.**
- **Heterogeneous exposure to real estate across banks:**
  - *Much larger share of loans to REI in weak banks (in 2006, 68% of loans to non financial firms vs 37%).*
- **Different evolution of credit flows at the two sets of banks**
  - *New credit grew more at weak banks during the boom (69% vs 12% in 2002-2007).*
  - *The fall in the slump was more pronounced in weak banks too (46% vs 35%).*
  - *This affects both intensive and extensive margins.*



# The financial crisis in Spain: The credit collapse. Testing for Hypothesis A1



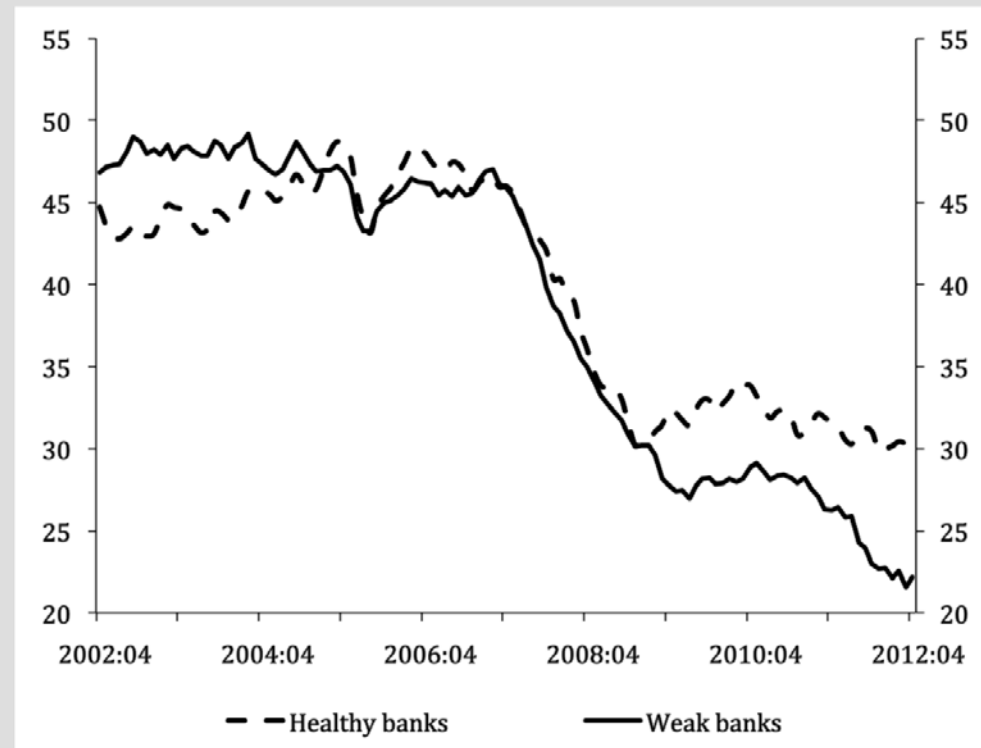
New credit to non-financial firms by bank type  
(12-month backward moving average, 2007:10=100)



# The financial crisis in Spain: The credit collapse. Testing for Hypothesis A2



Acceptance rates of loan applications by non-current clients, by bank type (%)  
[Firms applying to at least one bank of each type]



# Anticipation Effects

## Testing for Hypothesis A3



- **Could firms anticipate differences in bank insolvency risk?**
  - We look at risk premia on securitizations by Spanish banks in 2006.
  - We want to test whether risk premia was set by the market without taking into account the type of bank (weak or healthy)
  - Risk Premium = F[ *type of securitization (mortgage backed security (MBS), asset backed securities (ABS)), risk category (AAA, AA+ to BBB-, BB+ to D), collateral type, guarantor type, years to maturity, and month of issue, Weak Bank Dummy*).
  - WB Dummy=1 for bailed-out banks (p-value: 0.55).
- **We cannot reject the hypothesis that financial markets failed to recognize differential risk at weak banks, as late as 2006!**
- **We take this as evidence for claiming that private firms could not possibly have predicted them either.**

# Data



- **Firms' loan information from Central Credit Register of the Bank of Spain:**
  - All loans above €6,000: identity of bank, collateral, maturity, etc.
  - Firms' credit history: non-performing loans and potentially problematic loans.
- **Firms' loan applications at non-current banks.**
- **Firms' annual balance sheets and income statements from Spanish Mercantile Registers.**
- **Firm entry and exit from Central Business Register.**
- **Banks' balance sheets and income statements from regulatory and supervisory Bank of Spain database and branches location from the Bank of Spain Branches Registry.**

# Sample



- **We exclude financial firms**
- **To avoid the problem of reverse causality:**
  - We exclude firms in the REI
  - We also exclude those firms that sold more than 20% of their value added to the REI in 2000
- **We exclude firms with no loans in 2006, given that we are interested in shocks to bank credit**
- **Our final sample represented in 2006**
  - 19% of firms
  - 28% of value added
  - 42% of private sector employees
- **98% of firms are SMEs according to the European Commission definition**

# The Treatment Variable



## ■ Aim

- We want to measure the impact of credit constraints on employment during the GR.

## ■ How?

- In order to do that, we compare the evolution of employment in 2010 (**post**) vs. 2006 (**pre-crisis**) between firms working with weak banks vs. firms working with healthy banks.

## ■ Treatment

- $WB_i$ : Dummy variable that takes value 1 if the firm had a significant attachment with a weak bank in 2006 (the ratio of weak-bank loans/Total assets is above the first quartile of the sample distribution of firms with non-zero exposure)

*-We consider alternative specifications for the threshold:  $>0$ , median, third quartile*

*-We also consider the continuous version*

# The Treatment Variable

## Selection Bias

- **Are firms comparable in the treatment and control groups?**

**No, treated firms are on average:**

- Younger and smaller .
- Financially worse: less capitalized, liquid, and profitable, more indebted with banks.
- More defaults, more loan applications to new banks, more defaults.
- Their banks: smaller, less capitalized, liquid, higher share of loans in mortgages, and more non-performing loans.
- Their employment grew more in the boom and fell more in the recession.

**Different unconditional trends -> Need to control for firm characteristics.**

# Credit Channel

## Firm-Bank (Local) Level



- **Khwaja and Mian (2008)**

$$\Delta_4 \log(1 + Credit_{ib}) = \theta_i + \vartheta WB_b + FB_{ib}\kappa + \epsilon_{ib}$$

- *Firm controls*: set of 18 firm financial characteristics (size, age, ROA, own funds ratio, debt ratio, credit history, past loan applications...)
- *Bank Controls*: ln(total assets), leverage ratio, liquidity ratio, ROA, Doubtful ratio
- *Firm-bank controls*: length of the bank-firm relationship and past defaults
- Firm fixed effects ( $\theta_i$ ): implies restricting the sample to firms working with at least two banks in 2006
  - *Doing that we are able to **control perfectly for credit demand***



# Credit Channel

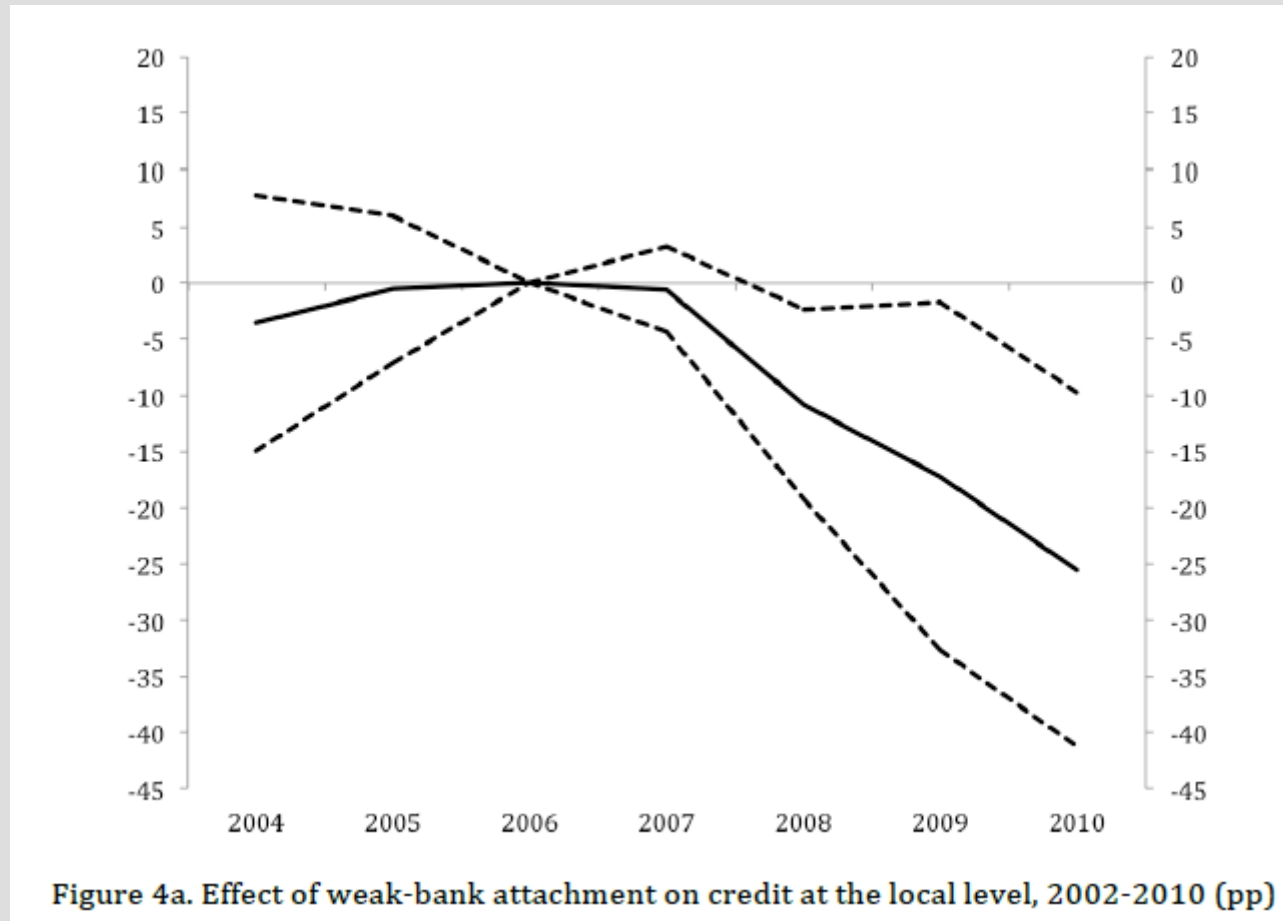
## Firm-Bank Level. Intensive Margin



Dependent variable: $\Delta_4 \log(1 + Credit_{i,jb})$				
	(1)	(2)	(3)	(5)
	All firms	Multi-bank	Fixed effects	Positive credit
$WB_b$	-0.232*** (0.088)	-0.256*** (0.094)	-0.255*** (0.008)	-0.079** (0.034)
Firm fixed effects	no	no	yes	yes
Firm controls	yes	yes	–	–
Bank fixed effects	no	no	no	yes
Bank controls	yes	yes	yes	yes
Firm-bank controls	yes	yes	yes	yes
Ind. $\times$ Prov. f.e.	yes	yes	–	–
Several banks	no	yes	yes	yes
$R^2$	0.060	0.059	0.407	0.394
No. obs.	304,089	236,691	236,691	126,863
No. firms	139,685	72,287	72,287	42,630

# Credit Channel

## Firm-Bank Level. Intensive Margin



# Credit Channel

## Firm Level (Aggregate)



Table 4. Credit rationing at the firm level  
Dependent variable:  $\Delta_4 \log(1 + Credit_{ij})$

	(1)	(2)
	All firms	Multi-bank
$WB_i$	-0.053 <sup>***</sup> (0.015)	-0.031 <sup>***</sup> (0.011)
Firm controls	yes	yes
Industry $\times$ Municipality fixed effects	yes	yes
Multiple banking relationships	no	yes
Balance-sheet data	yes	yes
$R^2$	0.215	0.246
No. obs.	149,458	74,045

- Treated firms managed to offset a substantial part of the reduction in credit supply by weak banks

# Credit Channel

## Firm. Intensive Margin

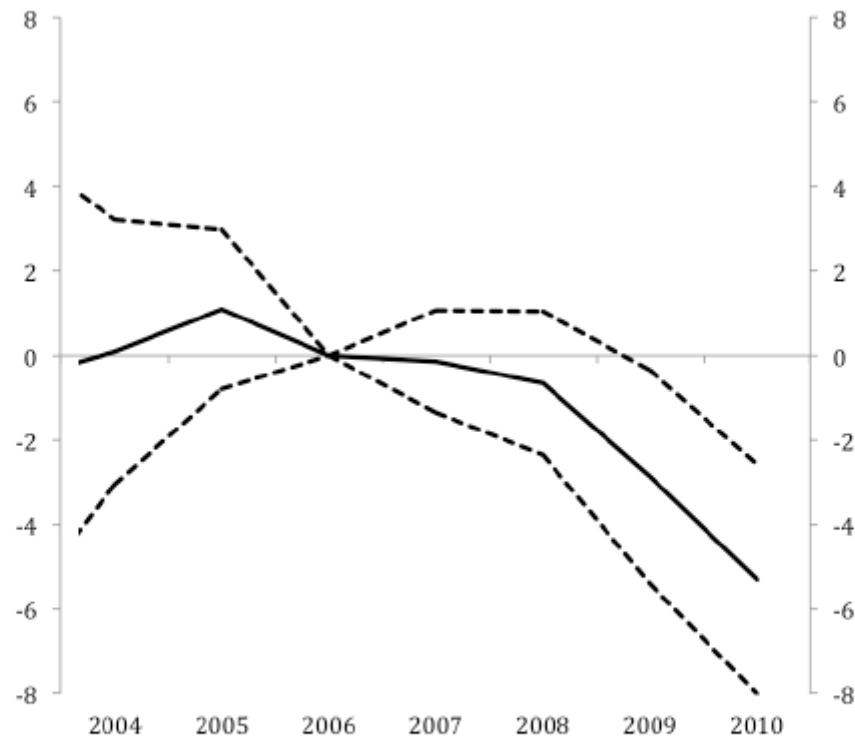


Figure 4b. Effect of weak-bank attachment at the firm level, 2002-2010 (pp)

# The Effect on Employment of Firm Level

## Diference in differences approach (DD)

$$\Delta_4 \log(1 + n_{ijk}) = \alpha + \beta WB_i + X_i \gamma + d_j \delta + d_k \lambda + u_{ijk} \quad (1)$$

- $\Delta_4$  denotes four year differences taken in 2010.
- $X_i$ : a set of firm controls (18 variables).
- $d_j$  and  $d_k$  : Municipality and industry set of dummy variables, respectively.
- $WB_i$  is a dummy variable for treated firms.
- $\beta$  measures the differential impact of credit constraints arising from attachment to weak banks.

*The estimation in first diferences implies an aggregate trend and also trends by municipality industry and firm characteristics.*

# Difference in Differences



Table 5. The employment effect of weak-bank attachment. Difference in Differences

Dependent variable: $\Delta_4 \log(1 + n_{ij})$						
	(1)	(2)	(3)	(4)	(5)	(6)
	Baseline				Placebo	
$WB_i$	-0.074*** (0.013)	-0.076*** (0.010)	-0.067*** (0.009)	-0.028*** (0.006)	-0.028** (0.013)	0.006 (0.007)
Firm controls (1)	no	no	yes	yes	yes	yes
Firm controls (2)	no	no	no	yes	yes	yes
Municipality fixed effects	yes	-	-	-	-	-
Industry fixed effects	yes	-	-	-	-	-
Industry $\times$ Municipal. f.e.	no	yes	yes	yes	yes	yes
Main bank fixed effects	no	no	no	no	yes	no
$R^2$	0.046	0.150	0.163	0.177	0.179	0.203
No. obs.	149,458	149,458	149,458	149,458	149,458	112,933

**Firm controls (1): age, size, rate of return and temporary employment rate**

**Firm controls (2) = Firm controls (1) + Other controls in 2006**

# Difference in Differences

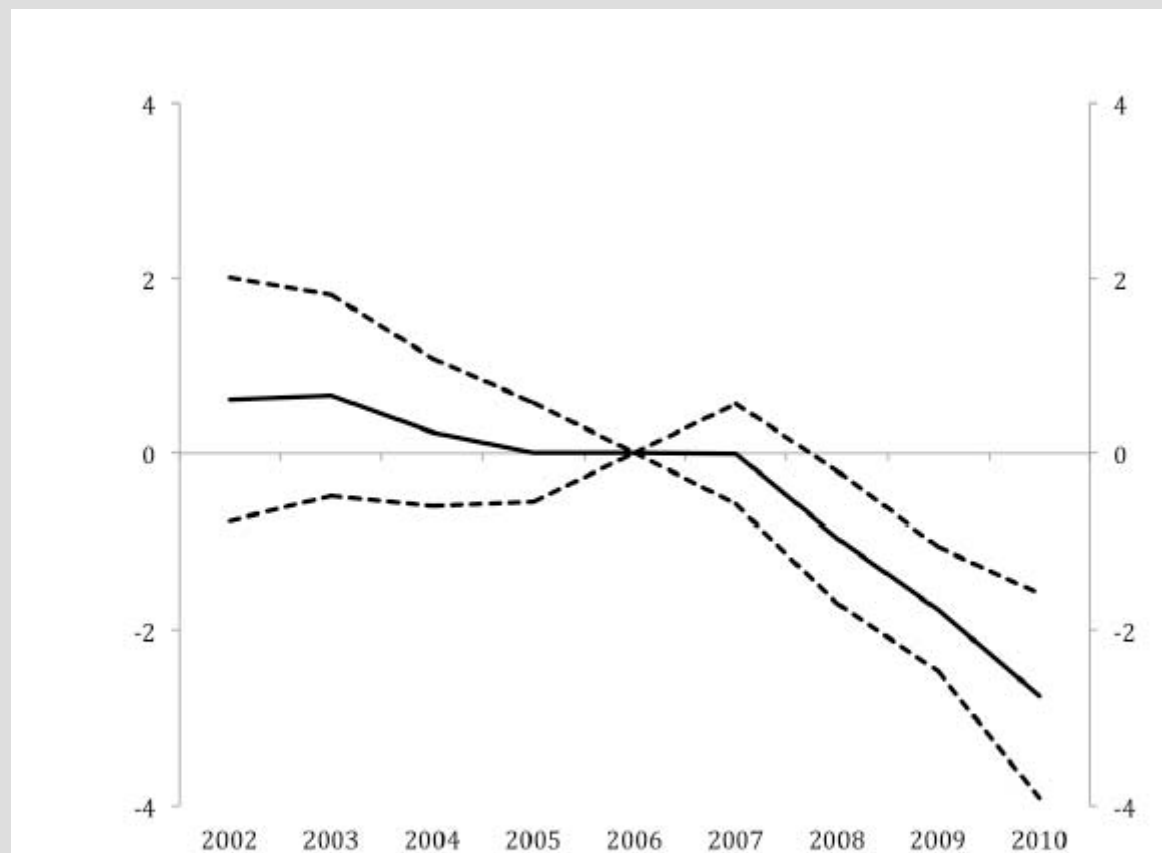


Figure 5. Employment effect of weak-bank attachment, 2002-2010 (pp)

# Alternative specifications



Table 7. The employment effect of weak-bank attachment. Difference in Differences  
Dependent variable:  $\Delta_4 \log(1 + n_{ij})$

	(1)	(2)	(3)	(4)	(5)	(7)	(8)	(9)
	WB Intensity	Median	Third quartile	Survivors	Alternat. measure	Loans to REI	2007 ex-ante	2002
$WB_i$	-0.092 <sup>***</sup> (0.020)	-0.030 <sup>***</sup> (0.008)	-0.033 <sup>***</sup> (0.008)	-0.014 <sup>***</sup> (0.004)	-0.034 <sup>***</sup> (0.004)	-0.030 <sup>***</sup> (0.008)	-0.019 <sup>***</sup> (0.006)	-0.015 <sup>***</sup> (0.006)
Firm controls	yes	yes	yes	yes	yes	yes	yes	yes
Industry × Municipality f.e.	yes	yes	yes	yes	yes	yes	yes	yes
$R^2$	0.177	0.177	0.177	0.181	0.182	0.177	0.130	0.188
No. obs.	149,458	149,458	149,458	133,122	129,246	149,458	145,322	71,703

(5): alternative dependent variable  $(n_{ijt} - n_{ijt-1}) / (0.5 * (n_{ijt} + n_{ijt-1}))$



# Robustness checks



- **To deal with the selection problem:**
  - The specifications include an exhaustive set of controls for observable firm characteristics
  - We also run a panel fixed effect model to allow for firm-specific trends:

$$\Delta \log(1 + n_{ijkt}) = \alpha'_i + W B_i d_t \beta' + X_i d_t \gamma' + d_j d_t \delta' + d_k d_t \lambda' + d_t \phi + v_{ijkt} \quad (2)$$

- $\alpha_i$  is a set of firm fixed effects.
  - $d_t$  is a set of time dummy variables.
- We use matching techniques to directly compare similar firms in the treated and in the control group

# Credit channel



- Is credit the transmission mechanism?
- Instrumental variable

$$\begin{aligned}\Delta_4 \log(1 + n_{ijk}) &= \alpha'' + \beta'' \Delta_4 \log(1 + Credit_{ijk}) + X_i \gamma'' + d_j \delta'' \\ &\quad + d_k \lambda'' + \varepsilon_{ijk} \\ \Delta_4 \log(1 + Credit_{ijk}) &= \rho + \mu WB_i + X_i \eta + d_j \sigma + d_k \psi + \omega_{ijk}\end{aligned}$$

- $\mu$  is the differential impact of weak banks on credit.
- $\beta''$  is the pass-through from credit to employment.
- Thus,  $\mu \beta''$  is equivalent to  $\beta$ .

# Credit Channel

## Instrumental Variables



Table 6. The employment effect of weak-bank attachment. Instrumental Variables

Dependent variable: $\Delta_4 \log(1 + n_{ij})$	
	(1)
	All firms
Instrumented variable	$\Delta_4 \log(1 + Credit_{ijk})$
	0.519 <sup>***</sup>
	(0.179)
First stage	
$WB_i$	-0.053 <sup>***</sup>
	(0.015)
Firm controls	yes
Industry $\times$ Municipality fixed effects	yes
Overall effect ( $\mu\phi$ )	-0.028
$F$ test / p value	13.1 / 0.00
No. obs.	149,458

# Instrumental variables



- The choice of bank is an endogenous choice.
- Here we exploit a 1988 legal change to generate exogenous variation in exposure to weak-banks.
  - The legal change allowed savings banks to expand beyond their region of origin.
  - We use the density of weak-bank branches in 1988 at the municipal level as instrument for WB.
    - *Independence assumption: proximity in 1988 to banks that required state intervention in 2008 is orthogonal to firm characteristics before the onset of the boom-bust cycle in the 2000s.*
    - *Exclusion restriction: local weak bank density only affects a firm's employment through its exposure to weak banks*
      - Not so clear because a weak bank could have a different performance in its province of origin in terms of credit during the boom.
      - It cannot be tested.

# More Results

## Financial vulnerability (DDD)



Table 11. The employment effect of weak-bank attachment. Triple Differences

Dependent variable: $\Delta_4 \log(1 + n_{ijt})$		
$WB_i$	-0.019 <sup>***</sup>	(0.007)
Rejected application <sub><i>i</i></sub>	-0.066 <sup>***</sup>	(0.008)
$WB_i \times$ Rejected application <sub><i>i</i></sub>	-0.029 <sup>**</sup>	(0.012)
Past Defaults <sub><i>i</i></sub>	-0.209 <sup>***</sup>	(0.029)
$WB_i \times$ Past Defaults <sub><i>i</i></sub>	-0.041 <sup>**</sup>	(0.020)
Short-term debt <sub><i>i</i></sub>	-0.089 <sup>***</sup>	(0.013)
$WB_i \times$ Short-term debt <sub><i>i</i></sub>	-0.036 <sup>***</sup>	(0.011)
Bank debt <sub><i>i</i></sub>	-0.096 <sup>***</sup>	(0.017)
$WB_i \times$ Bank debt <sub><i>i</i></sub>	-0.081 <sup>***</sup>	(0.022)
Own funds ratio <sub><i>i</i></sub>	0.061 <sup>***</sup>	(0.026)
$WB_i \times$ Own funds ratio <sub><i>i</i></sub>	0.134 <sup>***</sup>	(0.027)
Liquidity ratio <sub><i>i</i></sub>	0.118 <sup>***</sup>	(0.022)
$WB_i \times$ Liquidity ratio <sub><i>i</i></sub>	0.050	(0.061)
Single bank <sub><i>i</i></sub>	0.012 <sup>**</sup>	(0.005)
$WB_i \times$ Single bank <sub><i>i</i></sub>	0.019	(0.015)
$\log(\text{Total Assets}_i)$	0.009 <sup>*</sup>	(0.005)
$WB_i \times \log(\text{Total Assets}_i)$	0.003	(0.005)
$\log(1 + \text{Age}_i)$	-0.054 <sup>***</sup>	(0.006)
$WB_i \times \log(1 + \text{Age}_i)$	0.027 <sup>***</sup>	(0.008)
$I(\text{Exporter}_i)$	0.176 <sup>***</sup>	(0.011)
$WB_i \times I(\text{Exporter}_i)$	0.062 <sup>***</sup>	(0.020)
$I(\text{Temporary employees}_i)$	-0.112 <sup>***</sup>	(0.010)
$WB_i \times I(\text{Temporary employees}_i)$	-0.027 <sup>***</sup>	(0.014)
$WB_i \times I(\text{Services}_i)$	0.019	(0.017)
Industry $\times$ Municipality fixed effects		yes
Firm controls		yes
$R^2$		0.176
No. obs.		149,458

# A Deeper Analysis



- **Previously, we have quantified the additional job losses due to credit-rationing by weak banks.**
  - 1. We also are interested in the type of jobs that are more at risk (temporary employees vs. fixed employees)
  - 2. We also want to test whether the firm used alternative margins of adjustment, like changes in wages, to alleviate the impact on employment.
- **1. We analyze how weak-bank attachment affected the share of temporary workers at the firm level.**
  - We observe that temporary jobs accounted for 56% of employment adjustment in treated firms, given the lower termination costs of these contracts
- **2. We analyze how weak-bank attachment affected the evolution of the total wage bill.**
  - We observe that wage adjustments have not played a role in alleviating the impact of credit constraints

# More Results

## Firm exit

- The speed of the recovery is not the same if a large fraction of firms close down than if they downsize

Table 13. Effect of weak-bank attachment on firm exit

Dependent variable: Probability of exit

	(1)	(2)
$WB_i$	0.011 <sup>***</sup> (0.004)	
$WB_i Intensity$		0.059 <sup>***</sup> (0.014)
Firm controls	yes	yes
Municipality $\times$ Industry fixed effects	yes	yes
$R^2$	0.173	0.173
No. obs.	150,442	150,442

# Conclusions



- **Aim: We study the impact of credit constraints on employment during the Great Recession in Spain outside the real estate industry.**
- **Identification: We exploit differences in lender health at the onset of the crisis, as evidenced by the bailout of savings banks.**
- **We construct a large dataset that allows us to control almost perfectly for demand factors reducing selection bias.**
- **Finally, controlling for selection we find sizeable effects of being attached to a weak bank prior to the recession: additional job losses due to weak bank attachment around 2.8 pp in 2006-2010.**
  - **It represents around 24% of job losses at firms attached to weak banks in our sample.**





**GRACIAS POR SU ATENCIÓN**