Employment Effects of Alleviating Financing Frictions: Worker-level Evidence from a Loan Guarantee Program[†]

Jean-Noël Barrot	Thorsten Martin	Julien Sauvagnat	Boris Vallée
HEC Paris	Bocconi	Bocconi	HBS

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Abstract

We study the employment effects of loan guarantee programs aiming at mitigating financing frictions for small businesses. Exploiting worker-level panel data combined with plausibly exogenous heterogeneity in policy generosity across French regions, we find that such programs have a significant and persistent positive impact on workers' employment and earnings trajectories. The program disproportionately benefits high earnings, male and younger workers, due to differences in retention decisions by the initial employer. We estimate the gross cost to preserve a job(-year) to be around $\in 3,200$, and a negative net cost when we include the savings on unemployment benefits.

Keywords: Loan Guarantees, Financial Frictions, Labor Market, Employment Trajectory.JEL Codes: G28, G33, H81, J23, J31, J65

[†]Jean-Noël Barrot is with HEC Paris School of Management. Contact: barrot@hec.fr. Thorsten Martin is with Bocconi University. Contact: thorsten.martin@unibocconi.it. Julien Sauvagnat is with Bocconi University. Contact: julien.sauvagnat@unibocconi.it. Boris Vallée is with Harvard Business School. Contact: bvallee@hbs.edu. We thank Claire Celerier, John Earle, Debbie Lucas, Will Mullins, Antoinette Schoar, David Thesmar, and seminar participants at UCLA, UC San Diego, HEC Paris, Bocconi, Harvard Business School, MIT Sloan, University of Toronto, McGill, Laval University, Sciences Po, and HEC Montreal for comments and suggestions. Errors are ours only. We thank Bpifrance Le Lab for granting us access to their data.

1 Introduction

Numerous countries facilitate bank lending to small businesses through loan guarantee programs, in which a government agency underwrites a share of the notional of loans issued by partnering private banks to qualifying borrowers (such as the Small Business Agency programs in the U.S.; see Beck et al. (2010) for a summary of these programs around the world). As private banks retain skin-in-the game, these schemes are designed to address the mis-targeting and rent-seeking that plague credit subsidies (see for instance Khwaja and Mian (2005)). Policy-makers' interest in these programs has further increased in the wake of the financial crisis due to raising concerns that small businesses might be prevented from accessing sufficient capital for them to be resilient, grow, and create jobs (Chen et al., 2017; Bord et al., 2018).¹ Despite their large and growing importance, we know surprisingly little about the effects of these programs on employment. Whether such programs represent a cost-effective countercyclical policy also remains an open question.

While there is some evidence that these programs foster job growth at beneficiary firms, measuring their impact on employment calls for estimating their effect on workers' job-tojob mobility and their transition between employment and unemployment. In particular, a relative increase in employment at firms receiving loan guarantees is not sufficient to assess the efficiency of loan guarantee programs. For instance, the efficiency of such policy would be low if the job growth at beneficiary firms happens when the frictions associated with job-to-job transitions are low in the first place. In addition, these programs might exacerbate resource misallocation in the economy if, for instance, they foster employment in less productive firms (Gopinath et al., 2017; Cong et al., 2019; Rotemberg, 2019). However, if these programs prevent workers from experiencing costly periods of unemployment, and/or

¹In the U.S., the main Small Business Administration (SBA) 7(a) loan guarantee programs have significantly expanded with the financial crisis. The American Recovery and Reinvestment Act of 2009 (ARRA), provided the SBA an additional \$730 million, in order to temporarily subsidize the loan guaranty programs' fees and increase the program's maximum loan guaranty percentage to 90%. The Small Business Jobs Act of 2010, provided \$505 million to extend the fee subsidies and 90% loan guaranty percentage through December 31, 2010; and increased the 7(a) program's gross loan limit from \$2 million to \$5 million. The stock of SBA 7(a) loans has increased from \$46 billion in 2007 to \$92 billion in 2018. See CRS (2019) for more details.

from losing job- or industry-specific skills, the benefit of these programs could be large.

In this article, we use administrative micro data and estimate the impact of a new loan guarantee program implemented in France during the financial crisis on workers' employment and earnings trajectories. Our data tracks a representative sample of individual workers across their jobs over time, as well as their transitions between employment and unemployment. At a macro level, the data allows us to implement a cost-benefit analysis of the program that includes both the ex post cost of guarantees and the savings associated with reduced unemployment insurance, which we can benchmark against the cost of other types of employment policies.²

Launched in the midst of the financial crisis, the Recovery Loan Guarantee Program offers a public guarantee for French small and medium-sized enterprises (SMEs) to rollover and extend their short-term debt. This new program was announced in the last quarter of the year 2008 and extended until the end of 2010. As regional offices for the program screen applications in a decentralized manner, we observe plausibly exogenous variation in the intensity of the program at the regional level. We exploit this heterogeneity and interact it with a regional border discontinuity approach in order to estimate the causal impact of the program on workers at firms benefiting from a loan guarantee. The identifying assumption is that workers in firms located on each side of a regional border would have experienced similar labor market outcomes in the absence of the loan guarantee program.

We first provide strong evidence that the regional intensity of the loan guarantee program translates into a higher take-up of loan guarantees at the firm level within the regional border area. Consistent with regional differences in loan screening intensity driving this variation in treatment, we observe that beneficiary firms include more high-credit-risk firms in hightreatment-intensity regions. We then check whether higher exposure to the program is indeed associated with a relaxation of SMEs' financial constraints. To do so, we exploit balance-

 $^{^{2}}$ Having said that, it is beyond the scope of this analysis to attempt to quantify the net effect of these programs on social welfare. To undertake a welfare analysis, one would need for instance to weigh the effects of these programs during downturns against the inefficiencies that government credit policies might cause the rest of the time.

sheet data and find that firms in more exposed regions indeed increase their quantity of bank debt relative to the counterfactual. We then leverage our longitudinal worker-level data to evaluate how this program affects the employment and earnings trajectories of workers until 2015. The granularity of our data allows us to decompose worker employment spells by firm, industry, and place of work and to examine variation in the impact of the policy according to firm and worker characteristics.

We find that the program has a significant and positive impact on workers' employment and earnings trajectories. Quantitatively, when extrapolating our estimates to the average treatment at the firm level, we obtain that individual workers initially employed by a treated firm receive earnings that are 25% higher on average over the 2009-2015 period, compared to a counterfactual set of workers initially employed by non-treated SMEs. This finding reflects mostly an employment margin: workers exposed to the program are significantly less likely to separate from their initial employer, and to be unemployed over the sample period. As a result, the amount of unemployment benefits received by workers more exposed to the program are significantly lower: they represent one third of the earnings difference between the two groups. This estimate highlights the large indirect cost-saving dimension of the policy in terms of unemployment insurance.

We conduct several empirical checks to support our assumption that regional exposure to the loan guarantee program on each side of the border is uncorrelated with other shocks affecting local economic outcomes. First, we find parallel trends in workers' earnings in the years prior to the year 2009. Second, the estimates are only weakly affected when we control for firm-level observable characteristics, such as firm size and firm age, industry-fixed effects, and worker-level observable characteristics, such as age, occupation, and gender. Third, the estimates are robust to the inclusion of a set of additional controls: regional controls for public debt, taxes, state contributions, and public investment during the crisis, lending activity by local banks, and other public programs targeting employment such as subsidies for short-term work and structural funds from the European Union. By matching our worker-level data with firms' financial statements, we can also evaluate how the program differentially affects workers depending on firms' ex-ante financial constraints. Consistent with the idea that the program allows financially-constrained firms to access bank debt and avoid layoffs resulting from financial distress, we find a strong effect on workers' employment and earnings of financially-constrained firms, but virtually no effect for workers employed by unconstrained firms. Unconstrained firms' take-up does not seem to respond to regional differences in the intervention intensity.

Next, we decompose the effect between firm retention policy and labor market frictions outside of the initial firm. We find that workers more exposed to the guarantee program are more likely to stay at their initial firm, indicating a causal effect on retention rates at treated firms. For workers being more likely to be laid off in the counterfactual, moving to another firm in the same industry appears to be the main margin of adjustment, which suggests the existence of industry-specific skills. We also find that workers adjust by moving to other firms outside their original commuting zone. The frictions in the labor market appear to be high, which supports the relevance and effectiveness of the loan guarantee program.

We then turn to the cross-section of workers and estimate heterogeneous treatment effects separately for high versus low earnings workers, young versus old workers, and female versus male employees. We observe that high earnings, young, and male workers benefit more from the intervention, as the effects on both cumulative earnings and employment are more pronounced for these sub-groups. When decomposing along the adjustment margins, this heterogeneity appears to result mostly from the retention decision of the firm initially employing the worker, rather than from differences in labor market frictions outside of the firm benefiting from the loan guarantee.

Finally, we document that the program targets firms with high credit risk ex ante, while avoiding the riskiest firms. Regions with higher treatment intensity extend loan guarantees to marginally riskier borrowers. However, ex-post, we do not find evidence that default rates on the guaranteed loans are significantly larger in more generous regions, which suggests that the size of the program could have been scaled up without a significant change in its relative ex post cost. We also investigate whether an expansion of the program might tilt it towards less productive firms, or might reduce the reallocation of workers from treated firms towards more productive jobs. We find no evidence that more generous regions disproportionally target less productive firms, or that workers from the counterfactual appear to move towards more productive or larger firms, or start being employed by new firms.

We conclude the analysis by providing an aggregate cost-benefit analysis of the loan guarantee program. We find that the program had a positive impact on French aggregate employment on the order of around 210,000 jobs(-year), while the ex-ante cost was the provision of a 683 million euro fund. The ex post cost of the guarantee program can be estimated as the difference between the amount of payments to the banks of defaulting firms minus the premiums paid to BPI at origination, which equals ≤ 207 million. This corresponds to a gross cost to preserve a job(-year) of either $\leq 3,200$ or ≤ 950 (when compared to respectively the ex-ante or ex-post cost). We also estimate savings for the unemployment national fund to be around ≤ 1.3 billion, as the loan guarantee program reduced workers' unemployment spells. This translates into a negative net cost for the policy when we include the savings on unemployment benefits.

Our research contributes to the literature on government programs and small business lending (Zia, 2008; Banerjee and Duflo, 2014; Bach, 2014; Ru, 2018; Jiménez et al., 2018), and loan guarantees in particular (de Andrade and Lucas, 2009; Beck et al., 2010; Lelarge et al., 2010; Mullins and Toro, 2016; Brown and Earle, 2017; D'Acunto et al., 2017; de Blasio et al., 2018; Bachas et al., 2019; Gonzalez-Uribe and Wang, 2019), by shifting the focus from firm-level to worker-level outcomes. By estimating the difference in long-run outcomes between workers from exogenously treated firms to a relevant control group, our analysis identifies the causal effect of the loan guarantee program on the trajectories of individual workers' earnings and employment, and allows to measure the overall impact on employment from such programs. Second, our article contributes to the empirical debate on the effectiveness of public policies aiming to protect or stimulate employment in downturns, such as hiring credits (Cahuc et al., 2018a; Neumark and Grijalva, 2017), and subsidies for short-term work (Cahuc et al., 2018b; Giupponi and Landais, 2018). We show that loan guarantees have a positive impact on workers' employment and earnings, in particular for financially-constrained firms, at a relatively low cost.

Our work also complements a large body of empirical studies estimating the employment effects of credit-supply shocks. Chodorow-Reich (2013) shows that firms with pre-crisis lending relationships with weaker banks face restrictions in credit supply and reductions in employment following the collapse of Lehman Brothers in 2008. Duygan-Bump et al. (2015), Greenstone et al. (2015) and Bentolila et al. (2018) find that shocks to the supply of bank credit to (small) businesses during the Great Recession are associated with reductions in employment. Recent studies (Fonseca and Van Doornik, 2019; Barbosa et al., 2019; Caggese et al., 2019; Baghai et al., 2019; Babina, 2019) use longitudinal linked-employer-employee data that allows to estimate the heterogeneous effect of financial shocks on the cross-section of individual workers. Our findings support loan guarantee programs as effective at mitigating these employment effects.

Last, our article relates to a large literature on the long-run consequences of job loss or job market entry timing for individual workers' outcomes. Starting with the seminal study of Jacobson et al. (1993), it also includes Couch and Placzek (2010); Autor et al. (2014); Lachowska et al. (2017); Yagan (2018). The existing body of work has also shown that earnings losses after job displacement have an important cyclical component, that is longterm earnings losses due to job loss occurring in recessions are significantly larger than when it occurs in booms (Davis and Wachter, 2011; Schmieder et al., 2018). In the same vein, workers graduating in a recession earn persistently less than those graduating nearby peaks (Kahn, 2010; Oreopoulos et al., 2012). Our study expands this stream of research by fleshing out the long-term effects on worker outcomes of alleviating firms' financial frictions. Our study proceeds as follows: In section 2, we provide institutional detail on loan guarantee programs and specifically on the French one. In section 3, we describe the data we use and detail the identification strategy we implement to establish a causal effect. Section 4 provides our baseline results at the micro level while section 5 decomposes the effects into firm retention decisions and labor market frictions and examines heterogeneity in the consequences of the program by individual characteristics. Section 6 assesses direct and indirect costs of the program and develops a cost-benefit analysis at the macro level. Section 7 concludes.

2 Institutional Background

2.1 Public Loan Guarantee Programs

Numerous governments, including the US, provide loan guarantees to small firms. These programs are usually implemented through a specialized entity, such as the Small Business Administration (SBA) in the US or Bpifrance in France, which partners with banks. In 2017, the amount of new loans guaranteed respectively by the SBA and Bpifrance was around USD 25 billion in the US, and around USD 4.5 billion in France.

The economic rationale for such programs is typically threefold: mitigating financing frictions specific to small businesses, fostering economic activity that creates positive externalities, and alleviating firm behavior that can create negative externalities. Access to credit for small firms might be limited by adverse selection (Stiglitz and Weiss, 1981), moral hazard (Holmstrom and Tirole, 1997), and transaction costs. Positive externalities from small firms typically include innovation and offering job opportunities in peripheral areas. On the contrary, layoffs might generate negative externalities when frictions on the labor market prevent the efficient reallocation of the workforce.

Loan guarantees by a government-backed entity have several advantages over direct public lending. First, this public intervention design typically delegates screening and monitoring to private banks. Relying on banks' expertise and infrastructure mitigates the risk that political considerations drive the allocation of credit. As the guarantees are partial, banks retain skin-in-the-game when screening loans, which limits moral hazard on the side of the banks. A last advantage of the guarantee design is that it does not require the guarantor institution to disburse cash and raise capital, although it has to hold regulatory capital.

One potential limitation of credit guarantee schemes is that they might attract riskier borrowers and worsen the pool of firms accessing external financing. They might also deteriorate banks incentives to properly monitor borrowers in the presence of moral hazard.

2.2 The French Public Guarantor: Bpifrance

Bpifrance is the entity managing public loan guarantee programs in France. Bpifrance (previously named Sofaris, and then Oseo-Garantie) was created in 1982 as a French equivalent of the SBA. Bpifrance is a government-backed entity, whose two shareholders are the French State and the *Caisse des Depots et Consignations* - the long term investing arm of the French government - and aims at financing companies from seed phase to maturity. Bpifrance activities are therefore mostly targeted towards SMEs and encompass investing in equity (VC and Private Equity), lending, extending loan guarantees, and providing grants.³ Bpifrance does not collect deposits, but funds itself in the wholesale market.

Bpifrance works with a network of partner banks that include all major French banks, and relies on them to source loan applications. As of 2017, Bpifrance possesses 48 local branches that process the loan guarantee applications provided by the banks.

In the remainder of the paper, we focus on a new loan guarantee program created at the end of the year 2008, which specifically aims at allowing firms to rollover their short-term debt during the credit crunch.

³Bpifrance also has an activity of funds of funds to support the VC industry.

2.3 The Recovery Plan

The French recovery plan of 2009-2010 led to the creation of a large short-term credit guarantee program managed by Bpifrance (under the Oseo-Garantie name at that time). The plan guaranteed \in 5.3bn of new bank debt between 2008Q4 and 2010Q4, which represents 0.2% of the GDP of France and half of the total guarantees granted by Bpifrance over the same period. The plan targeted new lines of credit with a term between 12 and 18 months, as well as the restructuring of existing short-term debt into new loans with maturity between 2 and 7 years. 4,000 firms received guarantees on their new lines of credit for an amount of \in 1.8 bn, and 17,000 firms received guarantees on their medium-term new loans for an amount of \in 3.5 bn. A guarantee extended by Bpifrance covers between 50 and 90% of a loan notional. Bpifrance charges an average insurance premium of around 1% per annum in exchange for such a guarantee. This cost to the issuer needs to be compared to the ex post default rate: around 11% of recipients failed as of June 2011, which implies that the guarantee was heavily subsidized on average.

[INSERT FIGURE 1]

3 Empirical Strategy

3.1 Data

We use three complementary sources of data, which we obtain from Bpifrance and the French Statistical Office (INSEE): an exhaustive file of individual loan guarantees, the exhaustive firm registry, and a matched worker-firm panel covering 1/12th of the French workforce.

3.1.1 Loan Guarantees

We use proprietary data provided by Bpifrance on the whole universe of firms benefiting from loan guarantee programs since 2002. This data provides a unique firm identifier (SIREN),

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and information on the guarantee characteristics, including the date and amount of the loan, whether the guarantee was part of the recovery plan, the type of loan underlying the guarantee, and the fraction of the loan covered by the guarantee. Bpifrance data do not include information on interest rates, but includes information on default: whether the loan benefiting from the guarantee defaults over its life, and the loss amount.

3.1.2 Firm-level tax filings

We use administrative microdata extracted from tax files used by the French Ministry of Finance for corporate tax collection purposes, available until 2015. The data includes the balance sheets and profit and loss statements of the universe of French firms. The data is not publicly available, but is available for academic research through a procedure similar to accessing Census data in the US. We track firms through time using their unique identifying number ascribed by the French Statistical Office (INSEE). We retrieve industry classification using a historical four-digit industry classification code ascribed to each firm by the French Statistical Office itself, which is similar to the SIC coding system in the US. We exclude financial and real estate sectors, as well as utilities, non-profit, and regulated sectors. Unfortunately, there has been a discontinuity in the number of firm-level variables available for researchers in 2010. For the purpose of our analysis, this means that we observe bank debt only until 2009.

3.1.3 Worker-level data

Last, we rely on matched worker-firm longitudinal data ("DADS Panel"), built by the French Statistical Office (INSEE) from social security contribution declarations of firms. The sample covers all individuals born in October of each year, i.e. 1/12th of the French workforce. Each year firms declare the employment spells, the number of hours worked, and the associated wages for each worker. The DADS files cover virtually all French wage earners from 2009, except for self-employed workers, if they do not pay themselves a wage.⁴ For workers who have multiple jobs in a given year, we aggregate earnings across all jobs and retain the identifier of the employer that accounted for the largest share of the worker's earnings. Data on unemployment benefits are available since 2008.

3.2 Data Filtering

We apply the following filters at the firm and individual level. At the firm level, we first restrict the sample to non-financial SMEs (defined as firms with less than 250 employees) in the for-profit private sector. SMEs represent virtually all the beneficiaries from the recovery plan. Second, for the purpose of our identification strategy, we restrict the firm sample to firms with all their employees in the same region and located within a 10 miles distance to a regional border.

At the worker level, we restrict the sample to workers with high labor force attachment (as e.g. in Autor et al. (2014); Yagan (2018)), in our case workers with earnings above $\leq 10,000$ in each year 2006, 2007 and 2008. We then focus on workers who were born between 1957 and 1984 and study their outcomes over the period 2008-2015, during which these individuals were between 24 and 58 years old. We finally restrict our analysis to French citizens in order to minimize unobserved employment in foreign countries.

3.3 Descriptive Statistics

Table 1 presents descriptive statistics for the data obtained after filtering.

Panel A provides information on the exposure to the loan guarantee program, both at the regional and firm level. $Guarantee_{region,2009-2010}$ corresponds to the average regional ratio of loan guarantees under the recovery plan scaled by assets, computed across all eligible firms in a given region, excluding firms within 10 miles of a regional border. The generosity of

⁴Civil servants from the French central, regional and local administrations (general government), workers from the public health care sector, and workers employed by households (e.g. for house-keeping or child care) are not covered prior to 2009.

the program appears to vary significantly across regions, with firms from the least generous region having received on average 0.1% of their total assets in guarantee, while firms from the most generous region received 7.5 times more.

Turning to the treatment at the firm level, we observe that 4% of the firms in our sample received a loan guarantee. The average treatment conditional on being treated is 8% of assets (see Table A.2. in online appendix).

The worker sample consists of 38,024 individual workers employed full time in 2008 in a firm located within a 10 miles distance to a regional border. The average worker worked for 6.5 years during the 2009-2015 period, and received earnings equal to 6.5 times their initial annual earnings, including 0.2 times their initial annual earnings in unemployment benefits. The average worker is 38 years old, works 1,868 hours and earns $\in 23,630$. per year

We also present a number of firm characteristics measured in 2008. The average firm has 20 employees in 2008, is 18 years old, has assets of $\in 3.3$ million, return over assets of 10%, and bank debt representing 15% of its assets.

Table A.2 presents the same characteristics separately for firms receiving and not receiving a loan guarantee from the program.

[INSERT TABLE 1]

Our main sample by construction includes only firms located within a 10 miles distance to a regional border. One concern is that the effect of loan guarantees might be different in the universe of SMEs. In order to shed light on this potential issue, we first compare firm and worker characteristics in our sample of SMEs within a 10 miles distance to a regional border and the rest of French SMEs in Table A.1. Overall, the statistics are fairly similar. Still, we note that firms in our sample are slightly more likely to receive a loan guarantee from the program, and are slightly older. Workers' annual earnings are also significantly lower in the sample of SMEs within a 10 miles distance (on average 23,630 euros against 25,613 euros for the rest of French SMEs), probably because high-paid jobs are over-represented in large metropolitan areas, that are less likely to be located close to regional borders. Similarly, we present separately in Table A.3 the distribution of SMEs within 10 miles and the rest of French SMEs across a list of 18 industries. Overall, these distributions are similar, with some exceptions like Manufacturing which is overweighted in our sample, and Professional, Scientific and Technical Activities, which is instead underweighted in our sample. Taken together, these statistics suggest that our sample of firms located within a 10 miles distance to a regional border is reasonably representative of the universe of French SMEs. We therefore expect the results presented below to be informative for the overall impact of the program.

3.4 Empirical Design

3.4.1 Setting

Studying the effects of a loan guarantee program requires to overcome an empirical challenge: receiving a loan guarantee is most likely correlated with firm characteristics, either observables or unobservables. A naive OLS regression of worker outcomes on firm-level guarantee treatment is therefore prone to suffer from endogeneity, most likely due to the selection of treated firms on distress.

For the purpose of causal identification, we exploit plausibly exogenous variation of loan guarantee volumes at the regional level, interacted with a geographical regression discontinuity design that allows to absorb local economic conditions. Specifically, we predict firms' exposure to the loan guarantee program on each side of regional borders with the average treatment intensity of other firms in the same region.

For this purpose, we obtain the longitude and latitude coordinates of the centroid of each municipality. Using these geographic coordinates, we calculate the minimum distance from the population centroid of the municipality to the regional border. Figure 2 illustrates all the municipalities that are within 10 miles of the border, that is, the municipalities for which the minimum distance from the population centroid of the municipality to the regional border is below 10 miles. These municipalities form a strip of land on both sides of the border of fairly uniform width. Our baseline sample includes all workers working at a firm located in one of these border municipalities.

Figure 2 also displays the treatment intensity at the regional level – that is, the average ratio of loan guarantees under the recovery plan scaled by assets in 2008 computed across all eligible firms in each region, excluding firms within 10 miles of a regional border, as previously described in table 1. Our empirical strategy exploits this regional variation in treatment intensity as source of identification.

[INSERT FIGURE 2]

The geographic variation in treatment intensity likely results from differences in the level of application screening intensity for regional offices. Each regional branch of Bpifrance indeed has autonomy over the processing of the applications it receives. The Bpifrance annual report mentions that 90% of applications are screened at the regional branch level, without any intervention of the central headquarter. While data limitations prevent us from observing applications' acceptance rates at the regional Bpifrance branch level, we study whether there exists compositional differences among beneficiary firms in terms of credit risk that are supportive of this channel.

In figure 3, we plot the likelihood of receiving a loan guarantee from Bpifrance by decile of ex ante credit risk, measured by the inverse of the interest coverage ratio as of 2008. Panel A plots this quantity at an aggregate level, and confirms that the program is tilted towards high-credit risk firms, except for the riskiest ones. Panel B breaks down this probability for regions with high vs. low treatment intensity. This figure highlights how in the hightreatment-intensity regions high-credit-risk firms are significantly more likely to obtain a loan guarantee. This is consistent with a less intense screening of applications in these regions.⁵

[INSERT FIGURE 3]

⁵This difference in credit risk composition among beneficiary firms across low and high treatment intensity regions does not challenge the validity of our identification strategy. However, it matters for the interpretation of the results: our empirical setting estimates the joint impact of the shift in guarantee supply and the associated risk composition of beneficiary firms.

3.4.2 Specifications

Our empirical strategy is akin to a difference-in-difference estimation where areas are differentially exposed to the short-term loan guarantee program. The exclusion restriction relies on the regional loan guarantee exposure only affecting workers' outcomes through the subsidized access to new lines of credit and bank loans offered by the program to their employers in 2009 and 2010. In particular, regional exposure to the program needs to be orthogonal to other local shocks that would otherwise affect workers' outcomes. This motivates our regional discontinuity approach which largely mitigates the possibility that unobserved local economic shocks might confound our findings.

Our first stage boils down to the following cross-sectional regression on the set of firms located within 10 miles of a regional border:

$$Guarantee_{f,2009-2010} = \beta.Guarantee_{region,2009-2010} + \delta_1.X_f + \delta_2.X_w + \delta_3.X_r + \gamma_s + \epsilon_f, \quad (3.1)$$

where $Guarantee_{f,2009-2010}$ is the ratio of the amount of loan guarantee received by firm ffrom Bpifrance through the recovery plan over the firm total assets in 2008, $Guarantee_{region,2009-2010}$ is the average of the same ratio across all eligible firms in region r (excluding our sample firms located within 10 miles of the regional border), X_f is a vector of firm characteristics, and includes the logarithm of firms' total assets in 2008, the logarithm of firm age in 2008, as well as industry fixed effects (for 56 2-digit industries), and γ_s are department-pair fixed effects (a finer geographic division than regions). We cluster the error term, ϵ_f , at the level of regions. We run this regression both at the firm and at the worker level, to ensure both robustness and specification consistency. When running this regression at the worker level, we include X_w , a vector of worker characteristics including worker age, gender, and occupation fixed effects all measured in 2008, as additional controls.

We then estimate a similar cross-sectional specification as 3.1 with employment and earnings outcomes at the worker level as dependent variables:

$$y_{w,2009-2015} = \beta.Guarantee_{r,2009-2010} + \delta_1.X_f + \delta_2.X_w + \delta_3.X_r + \gamma_s + \epsilon_w, \qquad (3.2)$$

where y denotes an employment or related outcome over our sample period 2009-2015 for worker w employed as of 2008 in a firm located within 10 miles of a regional border. Following Autor et al. (2014) and Yagan (2018), one of our main variables of interest – cumulative earnings – are normalized by workers' initial earnings, that is, over the period 2006-2008. β , our coefficient of interest, measures the causal effect of initial regional exposure to the loan guarantee program on workers' outcomes. Importantly, we control for local economic conditions with department-pair fixed effects, which means that our identification comes from within (short) sections of the border band we study.

The main identifying assumption is that firms, and their workers, are as good as randomly assigned on one side of the border, meaning that workers in firms located on each side of the border would have experienced similar labor market outcomes in the absence of treatment. We first note that if labor markets are frictionless and workers can easily move to another region and obtain identical compensation in alternative firms, we should see no earnings or employment impact at the worker level from differences in their regional exposure to the French loan guarantee program in the period 2009-2010.

We check that workers and firms are almost indistinguishable based on observables on each side of regional borders in the year before the implementation of the loan guarantee program. For this, we run the same cross-sectional specification as 3.1 with workers' and firms' outcomes as dependent variables, all measured in 2008. We present the results in Appendix Table A.4. The differences across low and high exposure regions in workers' earnings, hours worked, unemployment benefits (Panel A), as well as firm age, firm size, and firm return on assets (Panel B), all measured in 2008, are all small and statistically insignificant.

A potential concern is that the variation in exposure we exploit might correlate with other local shocks that affect workers' employment and earnings. We address this concern in two ways. First, we show that workers' earnings prior to the intervention are uncorrelated with the subsequent regional intensity of the guarantee program, which mitigates concerns over reverse causality and omitted variable bias. Still, variation in the regional treatment intensity during the crisis years 2008Q4-2010Q4 might coincide with other regional shocks happening at the same time, for instance other regional government spending. We therefore include in all our regressions a series of controls that capture changes in public spending at the regional level, X_r . Specifically, we include the regional 2008-10 per-capita change in public debt, state contributions, local public investment, and taxes, respectively.

3.5 First-Stage Evidence

3.5.1 Predicting Firm-level Intervention using Regional Volume of Guarantees

We start by establishing the internal validity of our empirical setting. Table 2 displays the regression coefficients of the first stage as described in equation 3.1, at the firm level. In columns 1 to 3, the coefficients on $Guarantee_{region,2009-2010}$ are significant and positive, which confirms that a higher intensity of intervention in a given region translates into a higher intensity of intervention for firms close to the regional borders. We progressively introduce regional, and firm level controls, which leaves the coefficient of interest mostly unchanged. The coefficient of interest is around 0.7, which suggests that the intensity of intervention is comparable in the border area to the rest of the region, with a slight attenuation. Columns 4 to 6, where the dependent variable is an indicator variable for receiving a guarantee, illustrate that the regional intensity is associated with a significantly higher likelihood of receiving a guarantee.

[INSERT TABLE 2]

3.5.2 Balance-Sheet Evidence: Loan Guarantees and Bank Debt

To further strengthen the validity of our first stage, we study whether regional variation in the intervention is associated with the balance sheet effects targeted by the program, namely a better access to bank debt. We indeed find that a higher regional exposure to the loan guarantee program is associated with a higher growth in bank debt on firms' balance sheets relative to firms from the counterfactual.

For this, we run a specification similar to our first stage where the dependent variable is the growth rate of bank loans over 2008-2009, and the explanatory variable is the regional total amount of guarantee over total firm assets in 2009. Due to data limitations, we can only observe the debt composition of firms until the end of 2009, and therefore can only measure the effect on bank debt in the first year of the program. Table 3 displays the regression coefficients. Higher exposure to the loan guarantee program is indeed associated with an increase in bank loans on firms' balance sheets. This result is robust to using total debt growth rate over 2008-2010 as a dependent variable and $Guarantee_{region,2009-2010}$ as the explanatory variable, which covers the whole treatment period, but does not zoom in on the part of debt directly affected by the program.

[INSERT TABLE 3]

4 Impact of Loan Guarantees on Employment and Earnings

We begin by examining the impact of exposure to the loan guarantee program on workers' employment and earnings.

4.1 Baseline

We run our baseline specification to study the causal impact of this program on worker employment trajectories. Coefficients are displayed in table 4. Panel A studies cumulative effects over the period 2009-2015, whereas panel B explores the 2015 snapshot. Columns 1 and 5 include department-pair fixed effects. We progressively add regional controls in columns 2 and 6, firm-level controls in columns 3 and 7, and worker-level controls in columns 4 and 8.

We find a positive and statistically significant relation between workers' exposure to the loan guarantee program in 2009-2010, and their average cumulative employment and earnings over the period 2009-2015. First, as shown in columns 1 to 4, higher exposure to the program increases workers' employment rates over the period. Second, more exposed workers receive significantly higher cumulative earnings over 2009-2015.

The effects are economically sizable. Relative to the pre-crisis period, workers in a region with the average treatment experience a total gain in cumulative earnings of at least 6 percentage points of their initial annual earnings over the 2009-2015 period, e.g. around 1% per year, when compared to a hypothetical region with no exposure to the program.⁶ The coefficient of interest remains stable across specifications. When extrapolating this point estimate to the average treatment at the firm level *conditional* on obtaining a loan guarantee (8% of assets, see Table A.2. in online appendix), we find that exposure to the average treatment intensity is associated with additional cumulative earnings of 1.7 times workers' initial annual income. This corresponds to 25% higher earnings per year over the 2009-2015 period for workers at a firm exposed to the average treatment intensity.

In addition to their magnitude, the effects of the loan guarantee program appear to be persistent. In the 2015 snapshot displayed in Panel B, i.e. 7 years after the beginning of the program, the likelihood of being employed is still significantly higher for workers initially

 $^{^{6}}$ The average regional treatment is equal to 0.28% of total firm assets, which we multiply by the most conservative point estimate of our regression, 22%.

employed in firms more exposed to the loan guarantee program. This persistence speaks to the long-run effects that financial shocks can have on workers' outcomes when they are not mitigated.

[INSERT TABLE 4]

In table 5, we run a similar specification using an indicator variable for workers no longer being employed as of 2015 at the firm that they were working at in 2008. The likelihood of separation appears to be significantly lower for workers initially employed in firms more exposed to the loan guarantee program. Comparing the coefficient in column 4 of Table 5 with column 4 in Panel B of table 4 is indicative of the fraction of separated workers from their initial employer that are still unemployed, versus those who work for another employer as of 2015, a reallocation mechanism that we study in more detail in section 5.

4.2 Effect on Welfare Benefits

In France, earning losses due to involuntary unemployment are partly mitigated by unemployment insurance for a period up to two years. Unemployment benefits cover a fraction of the initial wage, and are subject to eligibility criteria. In our dataset, we can isolate earnings coming from unemployment benefits, which allows us to both estimate what the earning effects would have been for workers in the absence of unemployment insurance, as well as estimate the savings in unemployment benefits for the government that result from offering loan guarantees. We measure the effect of the intervention on worker unemployment benefits by using years, and the cumulated amount of unemployment benefits (scaled by initial earnings), during 2009-2015 as the dependent variables in our baseline specification. Results are displayed in table 6.

Workers from treated firms obtain unemployment insurance for a significantly shorter period of time, and collect significantly lower cumulated amounts of unemployment benefits over the period. In economic terms, the total amount of unemployment benefits received by workers in regions with average treatment intensity is lower by 2 percentage points of their initial earnings than workers in non-treated regions. This point estimate indicates that in the absence of unemployment insurance, the earnings difference between the two groups would have been one third larger. The same extrapolation exercise as in the previous subsection yields a cumulative reduction in unemployment benefits representing around 50% of the initial annual income over the 2009-2015 period for workers initially employed in a treated firm in 2008. This finding is consistent with the large effect on employment we document, and is of first order importance for the net cost of the intervention that we estimate in Section 6.

[INSERT TABLE 6]

4.3 Dynamics

We describe below the year-to-year impact of the loan guarantee program on worker outcomes. We first plot the estimated effect of exposure to the loan guarantee program for each year from 2004 to 2015 on annual worker earnings in Figure 4, and on the probability of being separated from the initial employer in Figure 5.

[INSERT FIGURE 4 AND FIGURE 5]

We then present the same point estimates in table 7, which displays the yearly effect of loan guarantees on worker earnings (Panel A), the associated cumulative effect over time for both earnings (Panel B) and the probability of separation (Panel C). As evidenced in Panel A, exposure to the loan guarantee program is associated with a large and statistically significant effect on annual earnings in the first years following the introduction of the program. The impact on annual earnings is still positive in the second part of our sample period, but smaller and not statistically significant. Reassuringly, the coefficients for the year 2004 to 2009 are all insignificant, which supports the absence of pre-trends and our interpretation of a causal impact of the guarantees on workers' earnings trajectories. As annual earnings are higher post treatment, the cumulative effect on earnings keeps growing over that period, as evidenced in panel B.

[INSERT TABLE 7]

4.4 Firm Heterogeneity and Robustness

We now turn to heterogeneity of treatment. We split our sample along proxies for firm financial constraints. We run our baseline specification on each of these sub-samples and present the regression results in table 8. In Panel A we use the number of years employed as dependent variable, while using cumulative earnings in Panel B. To robustly capture the degree of financial constraints a firm faces, we consider three proxies for financial constraints (measured in 2008) widely used in the literature to split our sample: having low cash flows in column 1 (versus high cash flows in column 2), not paying dividends in column 4 (versus paying dividends in column 5), and having a low share of tangible assets (that can be used as collateral) in column 7 (versus a high share of tangible assets in column 8).⁷ Columns 3, 6 and 9 test for the statistical significance of the difference in the coefficients on *Guarantee*_{region,2009-2010} between the sub-samples.

Consistent with the notion that the loan guarantee program mitigates SMEs financial frictions, the effects on workers employment and earnings we document are more pronounced for firms with low cash flows, low-collateral firms, and firms not paying dividends, all measured in 2008.⁸ The difference between the coefficients is economically large, and statistically significant for both workers' employment and earnings, between the firms with low versus high cash flows.

[INSERT TABLE 9]

⁷See Fazzari et al. (1988) for an early application of this methodology and Almeida et al. (2004), and Chaney et al. (2012) for recent examples.

⁸By running the first stage along the same dimensions of firm heterogeneity, we observe that the more pronounced effect for financially constrained firms is driven by their higher take-up of the program. Results are displayed in table A.7 in the online appendix.

Similarly, we split the sample of firms with respect to their credit risk measured in 2008. Reassuringly, we find virtually no effect of the program on workers initially employed in low-credit risk firms. This is in line with the evidence in Figure 3 that the share of low credit risk firms receiving a loan guarantee from BPI is negligible.

We also conduct a set of robustness tests that we report in table A.6 in the online appendix. First, we ensure that our results are robust to our definition of a regional border area. We use a cutoff of 5 miles instead of 10 miles from the regional border to define a border area, and find consistent results, even though the size of the sample substantially drops. Second, we check that our results are not picking up different economic trends between Paris and its surrounding area, and the rest of France. To do so, we exclude the region Ile - de - France, the region that includes Paris and its suburbs, from our analysis. Again, our coefficients are virtually unchanged. Third, one may be concerned that the program distorts competition on product markets in favor of firms located in regions more exposed to the guarantee program. Under this hypothesis, our coefficients would also reflect business-stealing effects between more and less exposed firms on each side of the regional borders. We address this concern by removing non-tradable industries from our sample (e.g. restaurants), where demand effects through local competition could indeed confound our estimates. Reassuringly, our baseline results are quantitatively comparable when we restrict the sample to tradable industries only.

5 Tracing Down Labor Market Frictions

Having established the causal effect of the loan guarantee program on worker employment and earnings, we next decompose the effects to assess the respective role of firms' labor retention policy, and labor market frictions outside the firm.

5.1 Adjustment Margins

We follow Autor et al. (2014) to disentangle firm retention decisions from labor market frictions by pinning down the adjustment margins of employment in table 10. We decompose the overall effect on years employed and cumulated earnings displayed in column 1, which corresponds to the results from table 4, the share coming from the firm in which the worker is initially employed as of 2008 in column 2, and the share coming from other firms in column 3. We further flesh out the adjustment coming from employment in other firms by area in columns 4 and 5, and by industry in columns 6 and 7.

The point estimate of column 2 indicates the differences in employment and earnings obtained by workers at their initial employer. The baseline coefficients of table 4, reproduced in column 1, are lower than these effects at the initial firm, and reflect the fact that employment and earnings gains at the initial firm are partially offset with workers' employment in other firms. Indeed, as shown in column 3, more-exposed workers to the loan guarantee program are less likely to work and receive earnings for other employers over the sample period. In other terms, the employment and earnings differences observed at the initial firm level are partially offset through workers mobility on the labor market. As shown in columns 6 and 7, the adjustment occurs through workers' employment within the same industry, which suggests the existence of industry-specific skills among workers which do not allow them to fully recoup the differences in earnings with the the initial employer.

We also find evidence for geographic reallocation: workers appear to adjust by moving to other firms outside their original commuting zone. In summary, the large effect on cumulative employment and earnings at the initial firm is partially undone through the mobility of workers to firms in the same industry, or to other firms in the same commuting zone. However, workers only succeed partially in offsetting the earnings differences observed at the initial employer (around half of the initial differences), which explains why the loan guarantee program has overall an impact on workers' employment and earnings trajectories.

[INSERT TABLE 10]

25

5.2 Worker-level heterogeneity

Next, we explore the heterogeneity in the main effect and in adjustment margins according to worker characteristics: age, earnings capacity (within age cohort), and gender. The sample split on earnings capacity within age cohorts allow to capture differences among workers based on ability rather than experience and seniority.

This heterogeneity analysis allows to identify which groups of workers benefit the most from the program, and whether these differences come from firms' labor retention policies or labor market frictions. In the three panels of table 11, we compare the impact of exposure to the loan guarantee program on employment and earnings separately for young and old workers in panel A, below and above median earnings workers (within each age cohort) in panel B, and men and women in panel C. We first measure the main effect for each sub-group in columns 1 and 3, and then flesh out the component coming from the initial employer of the worker in columns 2 and 4. We test for the statistical significance of the difference between the two sub-groups in columns 5 and 6.

From columns 1, 3 and 5, we observe that younger workers, workers with high earnings capacity, and male workers, seem to benefit more from the intervention overall, as the effects on years of employment and particularly on cumulative earnings are statistically higher for these sub-groups. When focusing on the effect coming from the initial employer, we find much larger effects for high earnings capacity, young and male workers, suggesting that the difference in the overall effect is mainly driven by firm's retention decisions.

[INSERT TABLE 11]

Overall, our results in the cross-section of workers provide evidence on the distributional consequences of loan guarantee programs, that primarily benefit high earnings, young and male workers. We also document that most of the cross-sectional variation stems from differences in labor retention policies rather than labor market mobility patterns.

6 Assessing the Costs of the Program

While the previous section documents the benefits of the loan guarantee program in terms of employment for the workers and savings in welfare payments for the government, these benefits need to be contrasted with the cost of the program to assess the efficiency of this public policy.

6.1 Higher Credit Risk, Similar Ex Post Default

The direct cost of the loan program is the financial cost associated with bearing the credit risk of borrowers. Panel A of Figure 3 shows that the probability of receiving a loan guarantee is higher for higher levels of credit risk, except for the top decile. Thus, the program appears to predominantly guarantee loans to risky firms, in line with the program's mandate, while avoiding the riskiest. With this in mind, we estimate how much an expansion in the recovery plan leads to a deterioration of the borrower pool's quality. For this, we regress a firm's credit risk decile in 2008 on our measure of regional treatment intensity. Table 12 shows the results for both the sample of firms receiving a guarantee under the recovery plan in columns 1 to 3 and for all firms in our sample in columns 4 to 6.

When looking at the sample of firms receiving a guarantee within the regional border zone, we find that firms located in a region with higher treatment intensity appear to extend loan guarantees to riskier borrowers, but the magnitude of this composition effect is relatively small. In contrast, we do not find statistically significant differences when we look at the entire sample of firms located in the regional border zone. Regions with higher treatment intensity (and their border zone) are not populated by more risky firms in general.

The inverse of the interest coverage ratio that we use to proxy for credit risk could instead proxy for low firm productivity. In table A.9 we investigate whether firms that receive a guarantee are of lower operating quality in regions with higher treatment intensity. We find no statistically significant difference in return on assets (ROA), in prior sales, employment and asset growth, nor in firm age. This suggests that the larger pool of guaranteed loans in high treatment regions is not associated with a deterioration of the borrower pool in terms of operating performance.

While firms in regions with higher exposure seem slightly ex ante riskier, we check in table A.10 in the online appendix whether they also have a higher likelihood of ex post default. We regress the defaulted amount normalized by the guaranteed loan amount (columns 1, 2 and 3), and a dummy for default (columns 4, 5 and 6) on our regional treatment variable, in the sample of firms that received a loan guarantee. Across specifications, coefficients are not statistically different from zero. Hence, despite of their higher riskiness ex ante, firms exposed to higher treatment intensity do not default more. This finding suggests that the scale of the program could have been at least marginally increased without significantly altering its relative cost.

[INSERT TABLE 12]

6.2 Preventing Efficient Re-allocation of Workers?

A potential indirect cost of the loan guarantee program is that it might prevent an efficient reallocation of workers, from firms in distress to more productive or new firms. Since our data allows us to track workers even when they changes job, we can observe to which type of firms workers get reallocated in our counterfactual. In table 13, we study workers' employment and earnings outcomes outside their initial employer. If anything, we find larger negative coefficients on our treatment variable for employment and earnings at low cash flow firms, which indicates that workers from the counterfactual are more likely to move to low, rather than high cash flow firms. We do not find much differences along the firm size dimension, nor on firm creation. Overall, these results suggest that workers displaced during the crisis are not particularly likely to move to highly productive or new firms. This result is hard to reconcile with the hypothesis that the loan guarantee program acts as a barrier to an efficient allocation of workers in the economy, and therefore mitigates concerns about this potential indirect cost of the program.

[INSERT TABLE 13]

6.3 Cost per Job(-year) at the Aggregate Level

Moving to the macro level, we perform an aggregate cost-benefit analysis of the loan guarantee program. As our analysis is conducted at the worker level, we multiply the average treatment of 0.28% (of total assets) by the coefficient estimated in our baseline specification (0.21) to calculate the average effect by worker. This calculation corresponds to an average gain of 0.06 years of employment per worker that we attribute to the loan guarantee program. As the full-time employee equivalent employment at SMEs in 2008 in France was 3.7 million, we obtain an estimate of 217,000 job(-years) preserved over the period 2009-15 ($3.7m \times 0.28 \times 0.21$).

This benefit needs to be compared to the cost of the intervention. The ex ante cost to the French government was the provision of a ≤ 683 M fund, which translates into an estimate for the gross cost per job(-year) of around $\leq 3200.^9$ The ex post cost of the guarantee program can be estimated as the difference between the amount of Bpifrance payments to the banks of defaulting firms, net of commissions. Banks have claimed guarantee payments for an aggregate amount of ≤ 333 M, and Bpifrance has received commissions for an aggregate amount of ≤ 126 M. The net cost is therefore ≤ 207 M, which translates into an estimate for the gross cost per job(-year) around $\leq 950.^{10}$

This cost-per-job is significantly smaller than estimates from the literature on fiscal multipliers in the US (Suárez Serrato and Wingender, 2016; Chodorow-Reich et al., 2012), which

⁹Following ?, one can alternatively value the ex ante cost of the program as a put option using derivative pricing methods. Assuming a risk-free rate of 3.5%, time to maturity of 2 years, volatility of 40%, the Black-Scholes value of a 70% guarantee on \in 5.3bn loans is \in 640M.

¹⁰These cost estimates do not account for potential distortions associated with raising the taxes used to finance the program nor do they account for potential increases in the operating cost of the Bpifrance branches due to the program.

place the cost-per-job from government spending closer to \$30,000. It is also smaller than estimates from the US loan guarantee program 7.(a) in Brown and Earle (2017), a costper-job of around \$25,000 (over three years). Finally, it is of the same order of magnitude as the gross cost per job estimated for other employment policies implemented in France in 2009: \in 2,619 for short-time work subsidies (Cahuc et al., 2018b), and \in 8,000 for hiring credits (Cahuc et al., 2018a). Overall, our analysis suggests that loan guarantee programs for short-term debt might be a cost-effective form of stimulus.

The gross cost per job(-year) we calculate ignores the savings in unemployment benefits and social benefits, as well as the avoided reduction in social contributions resulting from the loan guarantee program. We can easily adjust for the savings in unemployment benefits that we estimate in section 4.

Using a real discount rate of 10%, and the average treatment associated with a NPV of unemployment benefits of 1.6% of 2008 annual earnings, the savings amount to \in 350 per worker on average.¹¹ When applied to the existing 3.7 millions jobs in SMEs in 2008, we obtain an estimate of \in 1.3 bn savings in unemployment benefits, i.e. almost six times the non-discounted value of the expost losses on the program. This calculation yields a negative net cost for the program and the jobs it helps preserve.

7 Conclusion

In this article, we use administrative data at the worker level and examine how exposure to a new loan guarantee program implemented in France during the 2008-2009 financial crisis affects the employment and earnings trajectories of workers over the medium run. We find that exposure to the program results in a significantly higher likelihood of being employed over the next seven years, which translates into significantly higher cumulated earnings. Consistent with the idea that the program allows financially-constrained firms to rollover

¹¹We derive the NPV of unemployment benefits from discounting the yearly effects of the average treatment on unemployment benefits. The un-discounted associated value is the point estimate given in column (8) of 6 multiplied by 0.28, the average regional treatment intensity, and equals 2.2%.

their short-term debt and avoid excessive layoffs, we find a strong effect on employment and earnings trajectories of workers initially employed by financially-constrained firms, but virtually no effect for workers employed by unconstrained firms.

We then turn to the cross-section of workers, and observe that high earnings, young workers, and men, benefit more from the intervention, as the effects on cumulative earnings and employment are more pronounced for these sub-groups. However, when decomposing the effects along the adjustment margins, this heterogeneity appears to result mostly from differences in labor retention decisions by the initial employer rather than differences in labor market frictions in the cross-section of workers. Finally, we perform an aggregate cost-benefit analysis of the loan guarantee program, and estimate the gross cost to preserve a job(-year) to be around $\in 3,200$ and a negative net cost when we include the savings on unemployment benefits. Overall, our findings suggest that loan guarantees might be a cost-effective policy for sustaining employment in downturns, in particular in contexts where financial shocks hinder SMEs access to external funds.

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8 Graphs and tables

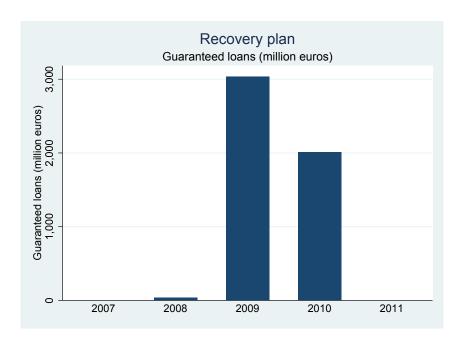


Figure 1 Yearly Volume of Guarantees of the Recovery Plan

Note: This figure displays the total volume of guarantees by Bpifrance as part of the recovery plan.

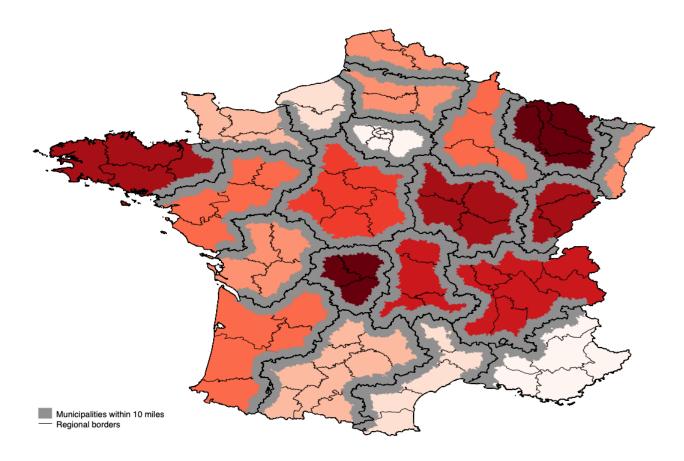
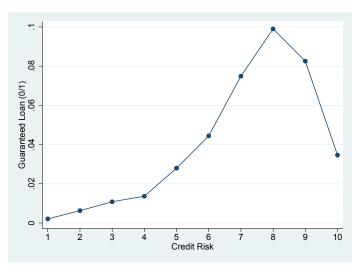


Figure 2 Regional Intensity of Loan Guarantee Intervention

Note: This figure displays the regional intensity of intervention by Bpifrance, as measured by the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. The grey area corresponds to municipalities within 10 miles of a regional border. This lines in black represent department boundaries within regions.





Panel B: High Treatment Regions vs. Low Treatment Regions

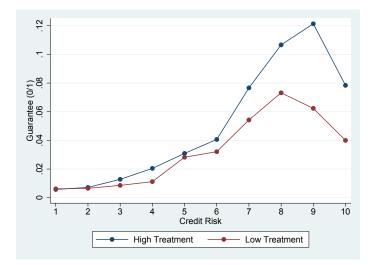


Figure 3 Probability of Loan Guarantee Intervention by Deciles of Credit Risk

Note: This graph plots the probability of receiving a loan guarantee from BPI under the recovery plan for each decile of credit risk. Credit risk is measured as the inverse of the interest coverage ratio as of 2008. The interest coverage ratio is defined as EBITDA over interest expenses. The sample consists of all firms in our sample of municipalities within 10 miles of a regional border. Panel A conducts this exercise for the whole sample, while Panel breaks it down between region with above-median treatment intensity and below-median treatment intensity.

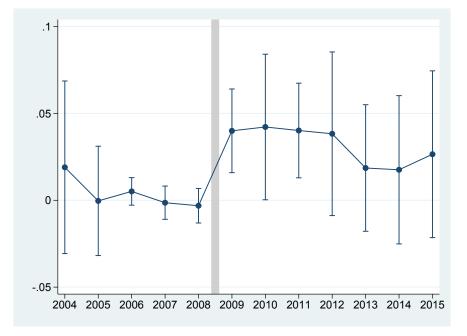


Figure 4 Dynamics: Effect on Earnings

Note: This figure plots regression coefficients and 95% confidence intervals from twelve regressions of earnings that a worker obtains in the year indicated on the x-axis, expressed in percentage points of the worker's average annual earnings in 2006-2008, on our measure of regional exposure to the 2009-2010 loan guarantee program, Guarantee_{region,09-10}. All regressions include department-pair fixed effects, the distance from the regional border, changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population), as well as firm and worker controls measured in 2008.

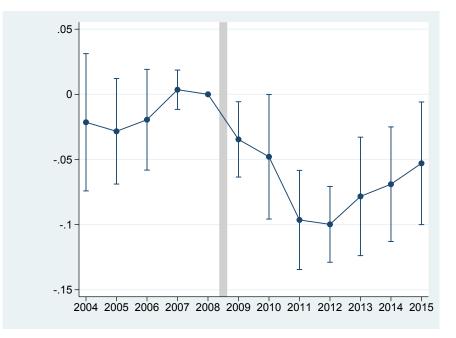


Figure 5 Dynamics: Effect on Separations

Note: This figure plots regression coefficients and 95% confidence intervals from twelve regressions of the likelihood that a worker does not work for the employer in 2008 in the year indicated on the x-axis on our measure of regional exposure to the 2009-2010 loan guarantee program, Guarantee_{region,09-10}. All regressions include department-pair fixed effects, the distance from the regional border, and changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population), as well as firm and worker controls measured in 2008.

Tab	le 1
Summary	Statistics

	(1)	(2)	(3)	(4)	(5)	(6)
	Obs.	Mean	SD	p1	p50	p99
Panel A: Loan guarantee exposure						
$\overline{\text{Guarantee}_{region,09-10} \text{ (over assets in \%)}}$	21	0.280	0.156	0.099	0.240	0.759
Guarantee _{firm,09-10} (over assets in $\%$)	$28,\!587$	0.315	1.742	0.000	0.000	12.956
Guarantee $(1/0)$	$28,\!587$	0.040	0.195	0.000	0.000	1.000
Default $\operatorname{Amount}_{firm}$ (over assets in %)	28587	0.030	0.394	0.000	0.000	0.000
Default on Guaranteed Loan $(1/0)$	28587	0.009	0.095	0.000	0.000	0.000
Panel B: Main outcome variables, 2009-2015						
Years Employed _{2009,2015}	38,024	6.520	1.284	1.000	7.000	7.000
Earnings _{2009,2015}	38,024	6.507	2.160	0.169	7.090	11.019
Separation _{2009,2015} $(1/0)$	38,024	0.485	0.500	0.000	0.000	1.000
Unemployment Benefits _{2009,2015}	38,024	0.216	0.477	0.000	0.000	2.155
Years with Unemployment Benefits _{2009,2015}	38,024	0.596	1.262	0.000	0.000	6.000
Panel C: Worker characteristics in 2008 Earnings	38,024	23,630	12,816	12,084	20,680	71,540
Hours	38,024	1,868	215	$1,\!150$	1,839	2,470
Age	38,024	38	7.7	24	39	51
Panel D: Firm characteristics in 2008 and outcomes						
$\frac{\Delta_{08-09}BankDebt}{BankDebt_{08}}$	19,103	-0.077	0.840	-1.00	-0.174	2.611
$\Delta Ln(Employment)_{08-10}$	$28,\!325$	-0.056	0.806	-3.504	0.065	1.792
$\Delta Ln(PPE)_{08-10}$	$28,\!493$	-0.174	1.378	-6.748	0.037	1.905
$\operatorname{Exit}_{08-10}$	$28,\!587$	0.044	0.205	0.000	0.000	1.000
Nb Employees	$28,\!587$	20.464	29.835	0.000	9.750	163.750
Assets $(\in 000s)$	$28,\!587$	$3,\!290$	$79,\!462$	41	731	$30,\!188$
ROA	$28,\!587$	0.104	0.192	-0.656	0.100	0.749
Firm Age	$28,\!587$	18.042	13.014	1.000	16.000	54.000
Dividend/Sales	$28,\!544$	0.016	0.037	0.000	0.000	0.222
PPE/Assets	$28,\!587$	0.461	0.333	0.000	0.386	1.000

Note: This table presents summary statistics at the regional and firm level (Panel A), at the worker level (Panel B, C), and firm level (Panel D). The sample includes 1/12th of employees who were working in SMEs located within a 10 miles distance to a regional border in 2008.

-

	(1)	(2)	(3)	(4)	(5)	(6)
		rantee $_{firm,0}$. ,		uarantee (1	
$Guarantee_{region,09-10}$	0.650^{***} (4.70)	0.707^{***} (6.03)	0.701^{***} (5.73)	0.066^{***} (4.42)	0.071^{***} (5.64)	0.069^{***} (5.40)
Distance to border	$0.000 \\ (0.01)$	-0.000 (-0.10)	$\begin{array}{c} 0.001 \\ (0.37) \end{array}$	$0.000 \\ (0.66)$	$\begin{array}{c} 0.000 \\ (0.54) \end{array}$	$\begin{array}{c} 0.000\\ (1.02) \end{array}$
Department-Pair FE Regional Controls Firm-level Controls Observations R^2	Y 28587 0.009	Y Y 28587 0.009	Y Y 28587 0.024	Y 28587 0.009	Y Y 28587 0.010	Y Y Y 28587 0.029

Table 2First Stage

Note: This table reports the results of the first stage OLS regressions. The dependent variable is the amount of guaranteed loans the firm received due to the 2009-2010 recovery plan scaled by 2008 firm assets in columns (1) to (3), and a dummy variable equal to one if the firm received any loan guarantee from the recovery plan in 2009-2010 in columns (4) to (6). The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects. Changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population) are added in columns (2) and (5). Firm-level controls added in columns (3) and (6) include log of assets, log of firm age, and two-digit industry fixed effects. Firm controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)
	Δ_0	$BankDebt_{08}$	
$Guarantee_{region,09}$	$\begin{array}{c} 0.147^{**} \\ (2.39) \end{array}$	0.172^{**} (2.48)	0.180^{**} (2.61)
Department-Pair FE Regional Controls Firm-level Controls	Y	Y Y	Y Y Y
Observations R^2	$\begin{array}{c} 19103 \\ 0.006 \end{array}$	$\begin{array}{c} 19103 \\ 0.007 \end{array}$	$19103 \\ 0.013$

Table 3Balance-Sheet Effects

Note: This table reports OLS regression results of the effect of loan guarantees on firms' bank debt. The dependent variable is the change in bank debt from 2008 to 2009, scaled by 2008 bank debt. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects and distance to the border. Column (2) adds changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population). Column (3) adds firm-level controls (log of assets, log of firm age, and two-digit industry fixed effects). Firm controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
Panel A: Cumulative effects		Years Emp	ployed _{09,15}		Earnings $_{09,15}$				
$Guarantee_{region,09-10}$	0.233^{***} (3.13)	0.256^{***} (3.14)	0.216^{***} (2.97)	0.213^{***} (2.85)	0.296^{***} (3.51)	0.329^{***} (3.54)	0.238^{**} (2.65)	0.216^{**} (2.26)	
Department-Pair FE Regional Controls Firm-level Controls Worker-level Controls	Y	Y Y	Y Y Y	Y Y Y Y	Y	Y Y	Y Y Y	Y Y Y Y	
Observations R^2	$38024 \\ 0.006$	$38024 \\ 0.006$	38024 0.028	$38024 \\ 0.035$	38024 0.007	$38024 \\ 0.007$	38024 0.042	$38024 \\ 0.054$	
Panel B: In 2015		Emplo	by ed $_{15}$		Earnings $_{15}$				
$Guarantee_{region,09-10}$	0.042^{**} (2.82)	0.044^{***} (2.87)	0.033^{**} (2.22)	0.032^{**} (2.09)	0.059^{**} (2.81)	0.055^{**} (2.48)	$\begin{array}{c} 0.033 \\ (1.49) \end{array}$	0.026 (1.15)	
Department-Pair FE Regional Controls Firm-level Controls Worker-level Controls Observations R^2	Y 38024 0.007	Y Y 38024 0.007	Y Y Y 38024 0.035	Y Y Y 38024 0.038	Y 38024 0.006	Y Y 38024 0.006	Y Y Y 38024 0.038	Y Y Y 38024 0.054	

Table 4Employment Effects: Baseline

Note: This table reports reduced-form OLS regression results of the effect of loan guarantees on worker-level outcomes. Panel A presents the cumulative effects on years employed and earnings 2009-2015. Cumulative earnings are the sum of earnings 2009-2015 scaled by average annual earnings 2006-2008. Panel B presents the effects on employment and earnings in 2015. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects and distance to the border. Columns (2) and (6) add changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population). Firm-level controls added in columns (3) and (7) include log of assets, log of firm age, and two-digit industry fixed effects. Worker-level controls added in columns (4) and (8) include worker age, gender, and occupation fixed effects. Firm and worker controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	Separ	auon		
	(1)	(2)	(3)	(4)
		Separ	ation	
$Guarantee_{region,09-10}$	-0.058** (-2.19)	-0.077*** (-3.36)	-0.050** (-2.21)	-0.056** (-2.48)
Department-Pair FE Regional Controls Firm-level Controls Worker-level Controls	Y	Y Y	Y Y Y	Y Y Y Y
Observations R^2	$\begin{array}{c} 38024 \\ 0.010 \end{array}$	$\begin{array}{c} 38024 \\ 0.011 \end{array}$	$\begin{array}{c} 38024 \\ 0.050 \end{array}$	$\begin{array}{c} 38024\\ 0.063\end{array}$

Table 5	
Separation	

Note: This table reports reduced-form OLS regression results of the effect of loan guarantees on workers' likelihood to separate from their initial employer. The dependent variable is a dummy equal to one if the worker did not work the entire period from 2009-2015 at the initial firm in 2008. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects and distance to the border. Column (2) adds changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population). Column (3) adds firm-level controls (log of assets, log of firm age, and two-digit industry fixed effects). Column (4) adds worker-level controls (worker age, gender, and occupation fixed effects). Firm and worker controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
		Years wit	th UB _{09,15}			UE	3 _{09,15}	
$Guarantee_{region,09-10}$	-0.197** (-2.13)	-0.249*** (-2.90)	-0.230*** (-2.92)	-0.239*** (-3.04)	-0.065^{*} (-1.85)	-0.085*** (-2.94)	-0.078*** (-3.12)	-0.080*** (-3.11)
Department-Pair FE	Y	Y	Y	Y	Y	Y	Y	Y
Regional Controls		Υ	Υ	Υ		Υ	Υ	Y
Firm-level Controls			Υ	Υ			Υ	Υ
Worker-level Controls				Υ				Y
Observations	38024	38024	38024	38024	38024	38024	38024	38024
\mathbb{R}^2	0.011	0.011	0.039	0.049	0.012	0.012	0.037	0.046

Table 6Unemployment Insurance

Note: This table reports reduced-form OLS regression results of the effect of loan guarantees on unemployment benefits. Columns (1) to (4) show the effects on years with positive unemployment benefits. Columns (5) to (8) show the effects on cumulative unemployment benefits. Cumulative unemployment benefits are the sum of unemployment benefits 2009-2015 scaled by average annual earnings 2006-2008. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects and distance to the border. Columns (2) and (6) add changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population). Firm-level controls added in columns (3) and (7) include log of assets, log of firm age, and two-digit industry fixed effects. Worker-level controls added in columns (4) and (8) include worker age, gender, and occupation fixed effects. Firm and worker controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

Panel A: Yearly Earnings	04	05	06	07	08	09	10	11	12	13	14	15
$Guarantee_{region,09-10}$	$\begin{array}{c} 0.019 \\ (0.80) \end{array}$	-0.000 (-0.02)	$0.005 \\ (1.34)$	-0.001 (-0.31)	-0.003 (-0.66)	0.040^{***} (3.46)	0.042^{**} (2.10)	0.040^{***} (3.08)	$0.038 \\ (1.70)$	$0.019 \\ (1.06)$	$0.018 \\ (0.86)$	$0.026 \\ (1.15)$
Department-Pair FE	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y
Full Controls	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ
Observations	38024	38024	38024	38024	38024	38024	38024	38024	38024	38024	38024	38024
R^2	0.090	0.062	0.038	0.012	0.035	0.036	0.031	0.034	0.038	0.042	0.050	0.054
Panel B: Yearly Separations	04	05	06	07	08	09	10	11	12	13	14	15
$Guarantee_{region,09-10}$	-0.021 (-0.85)	-0.028 (-1.46)	-0.019 (-1.05)	$0.004 \\ (0.49)$		-0.035^{**} (-2.50)	-0.048** (-2.09)	-0.096*** (-5.29)	-0.100^{***} (-7.15)	-0.078*** (-3.60)	-0.069*** (-3.27)	-0.053** (-2.35)
Department-Pair FE	Y	Y	Y	Y		Y	Y	Y	Y	Y	Y	Y
Full Controls	Υ	Υ	Υ	Υ		Υ	Υ	Υ	Υ	Υ	Υ	Y
Observations	38024	38024	38024	38024		38024	38024	38024	38024	38024	38024	38024
R^2	0.164	0.129	0.096	0.033		0.030	0.035	0.046	0.049	0.054	0.058	0.061
Panel C: Cum. Earnings						09	10	11	12	13	14	15
$Guarantee_{region,09-10}$						0.040^{***} (3.46)	0.080^{**} (2.74)	$\begin{array}{c} 0.118^{***} \\ (3.55) \end{array}$	0.156^{***} (3.46)	$\begin{array}{c} 0.174^{***} \\ (2.92) \end{array}$	0.192^{**} (2.52)	0.216^{**} (2.26)
Department-Pair FE						Y	Y	Y	Y	Y	Υ	Y
Full Controls						Υ	Υ	Υ	Υ	Υ	Υ	Υ
Observations						38024	38024	38024	38024	38024	38024	38024
R^2						0.036	0.036	0.039	0.042	0.046	0.050	0.054

Table 7Dynamics

Note: This table reports the effect of loan guarantees on earnings, separations, and unemployment benefits by year. Panel A reports yearly
earnings, Panel B yearly separations from the initial employer in 2008, and Panel C cumulative earnings. Earnings are scaled by average earnings
in 2006-2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
			Pan	el A: Years	Employed	2009,2015			
	(Cash-Flow	vs	Ι	Dividends		Т	angibility	τ
	Low	High	Diff	No Div	Div > 0	Diff	Low	High	Diff
$Guarantee_{region,09-10}$	0.414^{***} (4.57)	0.010 (0.11)	0.403^{***} (3.80)	0.271^{***} (3.02)	0.077 (0.81)	0.194 (1.58)	0.349^{**} (2.62)	0.120 (1.39)	0.230 (1.55)
Department-Pair FE	Y	(0.11) Y	(0.00) Y	(0.02) Y	Y	(1.00) Y	(2.02) Y	(1.55) Y	(1.00) Y
Regional Controls	Υ	Υ	Υ	Υ	Υ	Y	Υ	Υ	Υ
Firm-level Controls	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ
Worker-level Controls	Υ	Υ	Y	Υ	Υ	Υ	Υ	Υ	Υ
Observations	18885	18884	37769	24037	13981	38018	18890	18872	37762
R^2	0.037	0.049	0.043	0.038	0.037	0.042	0.042	0.042	0.042

Table 8Firm Heterogeneity: Financial Constraints

		Panel B: Cumulative Earnings 2009,2015							
	Cash-Flows			Ι	Dividends		Tangibility		
	Low	High	Diff	No Div	Div > 0	Diff	Low	High	Diff
$Guarantee_{region,09-10}$	0.527^{***} (3.90)	-0.017 (-0.10)	0.543^{**} (2.45)	0.331^{**} (2.81)	-0.023 (-0.16)	0.354^{*} (1.88)	0.314^{*} (1.99)	0.106 (0.98)	0.207 (1.10)
Department-Pair FE	(3.90) Y	(-0.10) Y	(2.45) Y	(2.81) Y	(-0.10) Y	(1.88) Y	(1.99) Y	(0.98) Y	(1.10) Y
Regional Controls	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ
Firm-level Controls	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ
Worker-level Controls	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ
Observations	18885	18884	37769	24037	13981	38018	18890	18872	37762
R^2	0.072	0.063	0.070	0.054	0.053	0.063	0.060	0.066	0.063

Note: This table reports the effect of loan guarantees on worker employment and earnings trajectories for sub-samples along proxies for financial constraints. Panel A presents the effects on years employed and Panel B on cumulative earnings 2009-2015. Cumulative earnings are the sum of earnings 2009-2015 scaled by average annual earnings 2006-2008. Column (1) and (2) show the results for sub-samples of firms below and above the median firm profitability (profit scaled by assets) in 2008, respectively. Column (3) and (4) split the full sample based on a dummy variable equal to one if the firm paid dividends in 2008. Column (5) and (6) show the results for sub-samples of firms below and above the median firm tangibility, respectively. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population), firm (log of assets, log of firm age, and two-digit industry fixed effects), and worker controls (worker age, gender, and occupation fixed effects). Firm and worker controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)
		Low Credit Risk			High Credit Risk	
	Guarantee	Years Employed	Earnings	Guarantee	Years Employed	Earnings
$Guarantee_{region,09-10}$	0.097 (0.96)	0.069 (0.74)	0.077 (0.54)	1.428^{***} (2.85)	0.401^{***} (4.22)	0.532^{***} (4.34)
Department-Pair FE	Ý	Ý	Ý	Ý	Y	Ý
Regional Controls	Υ	Υ	Υ	Υ	Υ	Υ
Firm-level Controls	Υ	Υ	Υ	Υ	Υ	Υ
Worker-level Controls	Υ	Υ	Υ	Υ	Υ	Υ
Observations	19239	19239	19239	15656	15656	15656
R^2	0.030	0.044	0.056	0.071	0.043	0.080

Table 9Firm Heterogeneity: Credit Risk

Note: This table reports the first stage results and the effect of loan guarantees on worker employment and earnings trajectories separately for firms with low and high ex-ante credit risk. Columns (1), (2) and (3) show the results for firms with low credit risk in 2008. Columns (4), (5) and (6) show the results for firms with high credit risk in 2008. Credit risk is measured as the inverse of the interest coverage ratio in 2008. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects, distance to the border, changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population), firm (log of assets, log of firm age, and two-digit industry fixed effects), and worker controls (worker age, gender, and occupation fixed effects). Firm and worker controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
(N=38,024)	all firms	initial firm	other firm	other firm same CZ	other firm other CZ	other firm same industry	other firm other industry
Years employed	$\begin{array}{c} 0.213^{***} \\ (2.85) \end{array}$	0.479^{***} (4.47)	-0.267^{*} (-2.11)	-0.020 (-0.19)	-0.246* (-1.82)	-0.315*** (-4.04)	$0.048 \\ (0.55)$
Cumulative earnings	0.216^{**} (2.26)	0.429^{***} (3.49)	-0.212 (-1.54)	$\begin{array}{c} 0.053 \ (0.50) \end{array}$	-0.266** (-2.55)	-0.185** (-2.14)	-0.027 (-0.37)

Table 10Adjustment Margins

Note: This table reports the effect of loan guarantees on employment and earnings at the initial firm and at other firms. Cumulative earnings are the sum of earnings 2009-2015 scaled by average annual earnings 2006-2008. Column (1) shows the effect across all firms. Column (2) measures employment and earnings at the initial firm (in 2008). Column (3) measures employment and earnings at other firms. Column (4) measures employment and earnings at other firms which are located in the same commuting zone (CZ) as the initial firm. Column (5) measures employment and earnings at other firms which are located in a different CZ than the initial firm. Column (6) measures employment and earnings at other firms in the same two-digit industry as the initial firm. Column (7) measures employment and earnings at other firms in different two-digit industries than the initial firm. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects, distance to the border, and changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population). Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)	
Panel A: Age							
	0	ld	Yo	Young		Diff	
	all	initial	all	initial	all	initial	
	firms	firm	firms	firm	firms		
Years employed	0.133	0.471**	0.371***	0.695***	0.238***	0.224	
1 0	(1.65)	(2.65)	(3.85)	(5.01)	(2.88)	(1.03)	
Cumulative earnings	0.067	0.286	0.524***	0.820***	0.457**	0.534**	
-	(0.76)	(1.58)	(3.43)	(5.28)	(2.73)	(2.55)	
Panel B: Earnings							
	Lo	0W	Hi	igh	Di	iff	
	all	initial	all	initial	all	initial	
	firms	firm	firms	firm	firms		
Years employed	0.201**	0.411*	0.312***	0.746***	0.112	0.335	
I J	(2.09)	(2.05)	(2.89)	(4.77)	(0.91)	(1.27)	
Cumulative earnings	0.148	0.328	0.570***	0.849***	0.422*	0.521	
Ŭ.	(1.20)	(1.53)	(3.54)	(4.09)	(1.98)	(1.66)	
Panel C: Gender							
	Wo	men	Μ	len	Di	ff	
	all	initial	all	initial	all	initial	
	firms	firm	firms	firm	firms	firm	
Years employed	0.221**	0.212	0.290***	0.699***	0.069	0.487*	
·····	(2.66)	(0.96)	(2.93)	(4.91)	(0.65)	(1.83)	
Cumulative earnings	0.036	-0.021	0.480***	0.791***	0.443**	0.813**	
0	(0.27)	(-0.08)	(3.55)	(4.64)	(2.13)	(2.46)	

Table 11Heterogeneous Effects across Workers

Note: This table reports the effect of loan guarantees on employment and earnings at all firms and at the initial firm for sub-groups of workers. Columns (1) and (3) show the effect across all firms. Columns (2) and (4) measure employment and earnings at the initial firm (in 2008). Columns (5) and (6) show the difference between sub-groups for all firms and at the initial firm respectively . Young (old) is a dummy equal to one for workers aged 22-39 (40-51) in 2008. Low (high) is a dummy equal to one for workers with below (above) median earnings in 2008, within their age cohort. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects, distance to the border, and changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population). Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)
		Fir	m Credit	$Risk_{08}$		
	1	of firms red loan guara	0	All fii	$rms \le 10$	miles
$Guarantee_{region,09-10}$	1.396^{***} (3.51)	1.208^{***} (2.96)	1.093^{**} (2.61)	$\begin{array}{c} 0.159 \\ (0.68) \end{array}$	$\begin{array}{c} 0.115 \\ (0.66) \end{array}$	$\begin{array}{c} 0.189 \\ (0.99) \end{array}$
Department-Pair FE Regional Controls Firm-level Controls	Y	Y Y	Y Y Y	Y	Y Y	Y Y Y
Observations r2	$\begin{array}{c} 1115\\ 0.095\end{array}$	$\begin{array}{c} 1115\\ 0.098\end{array}$	$\begin{array}{c} 1115\\ 0.163\end{array}$	$\begin{array}{c} 26282\\ 0.006 \end{array}$	$\begin{array}{c} 26282\\ 0.007 \end{array}$	$\begin{array}{c} 26282\\ 0.056 \end{array}$

 Table 12

 Initial Credit Risk of Firms Receiving Loan Guaranteed by BPI

Note: This table shows the results of regressing firms' initial credit risk in 2008 on the regional treatment intensity. The dependent variable is a firm's decile of credit risk, measured as the inverse of the interest coverage ratio in 2008. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects and distance to the border. Changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population) are added in columns (2) and (5). Firm-level controls added in columns (3) and (6) include log of assets, log of firm age, and two-digit industry fixed effects. Firm controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)
		Adjus		rgin: oth m type	er firms	
	Cash	-Flows	Firm	Size	Firm (Creation
	High	Low	Big	Small	New	Existing
Years employed	-0.038 (-0.31)	-0.243* (-1.89)	-0.107 (-0.84)	-0.184 (-1.56)	-0.060 (-0.59)	-0.206 (-1.69)
Cumulative earnings	-0.011 (-0.09)	-0.227* (-1.92)	-0.075 (-0.56)	-0.163 (-1.51)	-0.076 (-0.80)	-0.162 (-1.24)

Table 13A Barrier to Efficient Worker Allocation?

Note: This table reports the effect of loan guarantees on employment and earnings at other firms. Columns (1) and (2) show worker outcomes at firms with profitability above and below the initial firm in 2008. Columns (3) and (4) show worker outcomes at firms larger and smaller than the initial firm, measured by firm assets. Columns (5) and (6) show worker outcomes at firms created after 2008 and existing firms in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

Online Appendix

A DATA ACCESS

The French employment registers (DADS) and the fiscal data (FICUS-FARE), used in this paper, can be accessed by researchers. Authorization must be obtained from the *comité du secret*. The procedure is described at https://www.comite-du-secret.fr. Then researchers use a remote secure server (CASD) to work on the data. The "BPI files" that contain information on the firms receiving guarantees, is produced and owned by the Banque Publique d'Investissement.

B Tables

Table A.1Summary Statistics - Below versus Above 10 miles

	Our Sample			$SMEs \ge 10$ miles			Equality Test	
	Obs.	Mean	SD	Obs.	Mean	SD	P-value	
Panel A: Firm Sample								
$\overline{\text{Guarantee}_{firm,09-10} \text{ (over assets in \%)}}$	28587	0.315	1.742	117062	0.252	1.524	-6.063	
Default Amount _{firm,09-10} (over assets in $\%$)	28587	0.030	0.394	117062	0.027	0.385	-0.862	
Guarantee $(1/0)$	28587	0.040	0.195	117062	0.032	0.176	-6.290	
Default on Guaranteed Loan $(1/0)$	28587	0.009	0.095	117062	0.008	0.090	-1.500	
$\frac{\Delta_{08-09}BankDebt}{BankDebt_{08}}$	19103	-0.077	0.840	76169	-0.084	0.888	-0.948	
Nb Employees	28587	20.464	29.835	117062	20.035	29.668	-2.188	
Assets $(\in 0.00s)$	28587	3.290	79.462	117062	4.089	100.369	1.252	
ROA	28587	0.104	0.192	117062	0.100	0.206	-3.351	
Firm Age	28587	18.546	15.243	117062	17.616	15.589	-9.080	
Dividend/Sales	28544	0.016	0.037	116781	0.018	0.042	6.853	
PPE/Assets	28586	0.461	0.333	117046	0.393	0.327	-31.357	
Panel B: Worker Sample								
Years Employed _{2009,2015}	38024	6.520	1.284	146256	6.474	1.344	-6.058	
Earnings _{2009,2015}	38024	6.507	2.160	146256	6.510	2.287	0.265	
Unemployment Benefits 2009,2015	38024	0.216	0.477	146256	0.229	0.484	4.444	
Earnings 2008	38024	23630	12816	146256	25613	16672	21.588	
Hours 2008	38024	1868	215	146256	1861	219	-5.593	
Age 2008	38024	38.337	7.752	146256	37.959	7.686	-8.513	

Note: This table compares summary statistics at the firm (Panel A) and worker level (Panel B) for employees working in SMEs located within a 10 miles distance to a regional border in 2008 to employees working in SMEs located outside a 10 miles distance to a regional border in 2008.

Table A.2 Summary Statistics - No BPI loans vs Treated

	Ν	o BPI loa	ans		Treated	Equality Tes	
	Obs.	Mean	SD	Obs.	Mean	SD	P-value
Panel A: Firm Sample							
Guarantee _{firm,09-10} (over assets in %)	27454	0.000	0.000	1133	7.938	4.002	328.780
Default Amount _{firm,09-10} (over assets in %)	27454	0.000	0.000	1133	0.745	1.838	67.226
Default on Guaranteed Loan $(1/0)$	27454	0.000	0.000	1133	0.230	0.421	90.646
BankDebt Total Assets 08	26069	0.150	0.212	1091	0.194	0.178	6.690
$\frac{\overline{TotalAssets}_{08}}{BankDebt}$ $\overline{BankDebt_{08}}$	18227	-0.095	0.832	876	0.298	0.907	13.604
Nb Employees	27454	20.057	29.557	1133	30.319	34.489	11.371
Assets $(\in 0.00s)$	27454	$3,\!308$	81,079	1133	2,850	4,810	-0.190
ROA	27454	0.106	0.193	1133	0.048	0.140	-10.108
Firm Age	27454	18.485	15.215	1133	20.040	15.857	3.366
Dividend/Sales	27413	0.017	0.038	1131	0.006	0.018	-9.241
PPE/Assets	27453	0.460	0.334	1133	0.472	0.320	1.197
Panel B: Worker Sample							
Years Employed _{2009,2015}	36110	6.518	1.288	1914	6.556	1.210	1.254
$Earnings_{2009,2015}$	36110	6.514	2.164	1914	6.363	2.084	-2.978
Unemployment Benefits 2009,2015	36110	0.213	0.474	1914	0.280	0.532	6.036
Earnings 2008	36110	23624	12864	1914	23752	11873	0.424
Hours 2008	36110	1868	215	1914	1872	206	0.687
Age 2008	36110	38.320	7.761	1914	38.654	7.574	1.836

Note: This table compares summary statistics at the firm (Panel A) and worker level (Panel B) for SMEs that received no guarantee under the recovery plan to SMEs that received guarantees under the recovery plan. The sample includes SMEs within a 10 miles distance to a regional border in 2008.

Table A.3Industry Composition

Panel A:	Our	Sample	$\underline{\text{SMEs} \ge 10 \text{ miles}}$	
Agriculture, forestry and fishing	5	(0.0%)	18	(0.0%)
Mining and quarrying	77	(0.3%)	254	(0.2%)
Manufacturing	7574	(26.5%)	22235	(19.0%)
Electricity, gas, steam and air conditioning supply	12	(0.0%)	62	(0.1%)
Water supply; sewerage, waste management and remediation activities	162	(0.6%)	666	(0.6%)
Construction	4565	(16.0%)	17838	(15.2%)
Wholesale and retail trade; repair of motor vehicles and motorcycles	8210	(28.7%)	33512	(28.6%)
Transportation and storage	1801	(6.3%)	6056	(5.2%)
Accommodation and food service activities	1682	(5.9%)	8106	(6.9%)
Information and communication	282	(1.0%)	3263	(2.8%)
Financial and insurance activities	117	(0.4%)	654	(0.6%)
Real estate activities	427	(1.5%)	2790	(2.4%)
Professional, scientific and technical activities	1537	(5.4%)	10209	(8.7%)
Administrative and support service activities	803	(2.8%)	4766	(4.1%)
Education	197	(0.7%)	1185	(1.0%)
Human health and social work activities	426	(1.5%)	1653	(1.4%)
Arts, entertainment and recreation	212	(0.7%)	1123	(1.0%)
Other service activities	496	(1.7%)	2673	(2.3%)
	28585		117063	

anel B:		Our S	ample	mple		
	No B	PI loans	Tre	eated		
Agriculture, forestry and fishing Mining and quarrying Manufacturing Electricity, gas, steam and air conditioning supply Water supply; sewerage, waste management and remediation activ Construction Wholesale and retail trade; repair of motor vehicles and motorcycl Transportation and storage Accommodation and food service activities Information and communication Financial and insurance activities Real estate activities Professional, scientific and technical activities Administrative and support service activities Education Human health and social work activities	$\frac{5}{76}$	$(0.0\%) \\ (0.3\%)$	0 **	(0.0%) $(**%)$		
0	7038	(0.5%) (25.6%)	536	(47.3%)		
Electricity, gas, steam and air conditioning supply	12	(0.0%)	0	(0.0%)		
Water supply; sewerage, waste management and remediation activities	155	(0.6%)	7	(0.6%)		
Construction	4389	(16.0%)	176	(15.5%)		
Wholesale and retail trade; repair of motor vehicles and motorcycles	7958	(29.0%)	252	(22.2%)		
Transportation and storage	1738	(6.3%)	63	(5.6%)		
Accommodation and food service activities	1661	(6.1%)	21	(1.9%)		
Information and communication	277	(1.0%)	**	(**%)		
Financial and insurance activities	115	(0.4%)	**	(**%)		
Real estate activities	425	(1.5%)	**	(**%)		
Professional, scientific and technical activities	1500	(5.5%)	37	(3.3%)		
Administrative and support service activities	784	(2.9%)	19	(1.7%)		
Education	196	(0.7%)	**	(**%)		
Human health and social work activities	423	(1.5%)	**	(**%)		
Arts, entertainment and recreation	207	(0.8%)	**	(**%)		
Other service activities	493	(1.8%)	**	(**%)		
	27452		1133			

Note: This table presents the industry composition of SMEs. Panel A compares SMEs located within a 10 miles distance to a regional border in 2008 to SMEs located outside a 10 miles distance to a regional border in 2008. Panel B compares SMEs that received no guarantee under the recovery plan to SMEs that received guarantees under the recovery plan in our sample of SMEs within a 10 miles distance to a regional border in 2008.

Panel A : Worker Characteristics	$Ln(Wage)_{08}$	$Ln(Hours)_{08}$	$Ln(UI)_{08}$
$Guarantee_{region,09-10}$	-0.032	-0.001	0.021
0 ,	(-1.28)	(-0.16)	(0.44)
Department-Pair FE	Y	Y	Y
Regional Controls	Υ	Υ	Y
Observations	38024	38024	38024
R^2	0.045	0.008	0.005
Panel B : Firm Characteristics	$Ln(FirmAge)_{08}$	$Ln(Assets)_{08}$	$\rm EBITDA/Assets_{08}$
$Guarantee_{region,09-10}$	0.066	0.249	0.006
	(1.24)	(1.70)	(0.77)
Department-Pair FE	Y	Y	Y
Regional Controls	Υ	Υ	Υ
Observations	28587	28587	28587
R^2	0.012	0.012	0.007

	Table A.4	
Placebo Analysis:	Effects Before	the Reform?

Note: This table reports OLS regressions of worker and firm characteristics in 2008 on loan guarantees under the recovery plan in 2009-2010. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects, distance to the border, and changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population). Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)
	Years	s Employed	09,15	Ea	arnings _{09,1}	15
$Guarantee_{region,09-10}$	0.276^{***} (3.43)	$\begin{array}{c} 0.219^{***} \\ (2.93) \end{array}$	0.205^{**} (2.71)	$\begin{array}{c} 0.314^{***} \\ (3.14) \end{array}$	0.230^{**} (2.22)	0.216^{**} (2.17)
$\Delta EU \text{ funds}_{08-10}$	-0.090* (-1.99)			-0.14* (-1.99)		
Short-term work ₀₉		0.010^{**} (2.23)			0.023^{**} (2.40)	
$\Delta \text{Regional bank lending}_{08-10}$			$\begin{array}{c} 0.145 \\ (1.00) \end{array}$			$\begin{array}{c} 0.012 \\ (0.05) \end{array}$
Department-Pair FE	Y	Y	Y	Y	Y	Y
Regional Controls	Υ	Y	Y	Y	Y	Y
Firm-level Controls	Y	Y	Y	Y	Y	Y
Worker-level Controls	Υ	Y	Y	Y	Y	Y
Observations	38024	38024	38024	38024	38024	38024
R^2	0.035	0.035	0.035	0.054	0.054	0.054

 Table A.5

 Employment Effects: Controlling for EU Funds, Short-term Work, and Regional Banks

Note: This table reports reduced-form OLS regression results of the effect of loan guarantees on workerlevel outcomes, controlling for EU structural funds, subsidies for short-term work, and regional bank lending. Δ EU funds₀₈₋₁₀ is the log change of EU structural funds per capita in the region from 2008 to 2010. Short-term work₀₉ is the amount of short-term work subsidies per capita in the region in 2009. Δ Regional bank lending₀₈₋₁₀ is the average log change of loans by the four regional banks (Banque Populaire, Caisses d'Epargne, Credit Agricole, Credit Mutuel) in the region from 2008 to 2010. Cumulative earnings are the sum of earnings 2009-2015 scaled by average annual earnings 2006-2008. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects, distance to the border, changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population), firm-level controls (log of assets, log of firm age, and two-digit industry fixed effects), and worker-level controls (worker age, gender, and occupation fixed effects). Firm and worker controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

Table A.6Employment Effects: Robustness

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Panel A: Distance ≤ 5 miles	Years Employed $_{09,15}$				Earnings $_{09,15}$			
$Guarantee_{region,09-10}$	0.206^{**} (2.20)	$\begin{array}{c} 0.233^{**} \\ (2.35) \end{array}$	0.262^{***} (2.95)	0.263^{***} (2.92)	0.323^{**} (2.69)	0.248^{*} (1.79)	0.276^{**} (2.30)	0.265^{*} (2.07)
Department-Pair FE	Y	Y	Y	Y	Y	Y	Y	Y
Regional Controls		Υ	Υ	Υ		Υ	Υ	Υ
Firm-level Controls			Υ	Υ			Υ	Υ
Worker-level Controls				Υ				Υ
Observations	18680	18680	18680	18680	18680	18680	18680	18680
R ²	0.011	0.012	0.038	0.046	0.010	0.010	0.049	0.060
Panel B: Excluding Regional Pairs with Ile-de-France	de-France Years Employe		loyed $_{09,15}$		Earnings $_{09,15}$			
$Guarantee_{region,09-10}$	0.249^{**} (2.90)	$\begin{array}{c} 0.272^{***} \\ (2.99) \end{array}$	0.275^{***} (3.21)	0.268^{***} (3.04)	0.257^{**} (2.61)	$\begin{array}{c} 0.305^{**} \\ (2.83) \end{array}$	0.312^{**} (2.75)	0.306^{*} (2.40)
Department-Pair FE	Y	Y	Y	Y	Y	Y	Y	Y
Regional Controls		Υ	Υ	Υ		Υ	Υ	Υ
Firm-level Controls			Υ	Υ			Y	Υ
Worker-level Controls				Υ				Υ
Observations	24851	24851	24851	24851	24851	24851	24851	24851
R^2	0.008	0.008	0.031	0.037	0.008	0.008	0.045	0.056
Panel C: Excluding Non-Tradable Industries	Years Employed _{09,15}			Earnings _{09,15}				
$Guarantee_{region,09-10}$	$\begin{array}{c} 0.253^{***} \\ (3.33) \end{array}$	$\begin{array}{c} 0.274^{***} \\ (3.03) \end{array}$	0.227^{**} (2.37)	0.207^{**} (2.17)	0.496^{***} (4.09)	$\begin{array}{c} 0.495^{***} \\ (3.47) \end{array}$	0.388^{**} (2.44)	0.333^{*} (2.04)
Department-Pair FE	Y	Y	Y	Y	Y	Y	Y	Y
Regional Controls		Υ	Υ	Υ		Υ	Υ	Y
Firm-level Controls			Υ	Υ			Υ	Υ
Worker-level Controls				Υ				Υ
Observations	17200	17200	17200	17200	17200	17200	17200	17200
R^2	0.009	0.009	0.025	0.032	0.012	0.013	0.047	0.063

Note: This table reports robustness tests for the baseline results. See table 4 for detailed descriptions.

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	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
				Guara	antee _{firm} ,	09 - 10			
	C	ash-Flow	rs		Dividends	5	ſ	angibility	
	Low	High	Diff	No Div	$\underline{\text{Div} > 0}$	Diff	Low	High	Diff
$Guarantee_{region,09-10}$	1.115***	0.180	0.935**	1.276***	-0.095	1.372***	0.720**	0.536**	0.184
	(2.95)	(1.36)	(2.70)	(4.72)	(-0.27)	(3.30)	(2.10)	(2.59)	(0.53)
Department-Pair FE	Y	Y	Y	Y	Y	Y	Y	Y	Y
Regional Controls	Y	Y	Y	Y	Y	Y	Y	Y	Y
Firm-level Controls	Y	Υ	Y	Y	Υ	Υ	Υ	Y	Υ
Worker-level Controls	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ	Υ
Observations	18885	18884	37769	24037	13981	38018	18890	18872	37762
R^2	0.042	0.061	0.049	0.053	0.044	0.056	0.046	0.052	0.049

Table A.7Firm Heterogeneity: First Stage

Note: This table reports first stage OLS regression results for sub-samples along proxies for financial constraints. The dependent variable is the amount of loans a firm received under the recovery plan 2009-2010, scaled by firm assets in 2008. Column (1) and (2) show the results for sub-samples of firms below and above the median firm profitability (profit scaled by assets) in 2008, respectively. Column (3) and (4) split the full sample based on a dummy variable equal to one if the firm paid dividends in 2008. Column (5) and (6) show the results for sub-samples of firms below and above the median firm tangibility, respectively. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects, distance to the border, changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population), firm (log of assets, log of firm age, and two-digit industry fixed effects), and worker controls (worker age, gender, and occupation fixed effects). Firm and worker controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	ΔLn	(Employme	$(nt)_{08-10}$	ΔL	$n(PPE)_{0}$	8-10		$\operatorname{Exit}_{08-10}$	
$Guarantee_{region,09-10}$	0.056^{*} (1.90)	0.106^{***} (3.44)	0.105^{***} (3.36)	0.086 (1.21)	0.124 (1.62)	0.120 (1.59)	-0.032^{***} (-3.63)	-0.041^{***} (-4.52)	-0.039^{***} (-4.47)
Department-Pair FE Regional Controls Firm-level Controls	Y	Y Y	Y Y Y	Ŷ	Y Y	Y Y Y	Y	Y Y	Y Y Y
Observations R^2	$28325 \\ 0.005$	$28325 \\ 0.094$	$28325 \\ 0.095$	$28493 \\ 0.004$	$28493 \\ 0.012$	$28493 \\ 0.017$	$28587 \\ 0.004$	$28587 \\ 0.010$	$28587 \\ 0.014$

Table A.8Firm Employment, Investment and Exit

Note: This table reports OLS regression results of the effect of loan guarantees on firms' employment, investment, and exit. Δ Employment₀₈₋₁₀ is the log change of employment at the firm-level from 2008 to 2010. Δ PPE₀₈₋₁₀ is the log change of property, plants and equipment (PPE) from 2008 to 2010. Exit₀₈₋₁₀ is a dummy variable equal to one if the firm leaves the sample between 2008 and 2010. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects and distance to the border. Column (2) adds changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population). Column (3) adds firm-level controls (log of assets, log of firm age, and two-digit industry fixed effects). Firm controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1) <u>ROA₀₈</u>	(2) $\Delta Ln(Sales)_{05-08}$	(3) $\Delta Ln(Emp)_{05-08}$	(4) $\Delta Ln(PPE)_{05-08}$	(5) $Ln(Firmage)_{08}$
$Guarantee_{region,09-10}$	-0.030 (-1.15)	-0.144 (-0.85)	-0.086 (-0.57)	$\begin{array}{c} 0.213 \ (0.73) \end{array}$	$\begin{array}{c} 0.217 \\ (0.94) \end{array}$
Department-Pair FE	Y	Y	Y	Y	Y
Regional Controls	Υ	Υ	Υ	Υ	Υ
Observations	1119	1035	1035	1035	1119
R^2	0.090	0.104	0.128	0.092	0.085

Table A.9Zombie Lending?

Note: This table shows the results of regressing firms' pre-period characteristics on the regional treatment intensity within the sample of firms receiving a guarantee. The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects, distance to the border and changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population). Firm controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.

	(1)	(2)	(3)	(4)	(5)	(6)
	D	efault Ra	ate	De	efault (1)	(0)
$Guarantee_{region,09-10}$	$\begin{array}{c} 0.030 \\ (0.80) \end{array}$	$\begin{array}{c} 0.031 \\ (0.76) \end{array}$	$\begin{array}{c} 0.024 \\ (0.55) \end{array}$	$\begin{array}{c} 0.076 \\ (0.71) \end{array}$	$\begin{array}{c} 0.030 \\ (0.27) \end{array}$	0.027 (0.22)
Department-Pair FE Regional Controls Firm-level Controls	Y	Y Y	Y Y Y	Y	Y Y	Y Y Y
Observations R^2	$\begin{array}{c} 1119\\ 0.092 \end{array}$	$1119 \\ 0.097$	$\begin{array}{c} 1119\\ 0.164\end{array}$	$\begin{array}{c} 1119 \\ 0.096 \end{array}$	$\begin{array}{c} 1119\\ 0.102 \end{array}$	$\begin{array}{c} 1119\\ 0.192 \end{array}$

Table	A.10
Defa	ault

Note: This table shows the effect of loan guarantees on default within the sample of firms receiving a loan guarantee. The dependent variable is default amount scaled by the guaranteed loan amount in columns (1) to (3) and a default dummy in columns (4) to (6). The main explanatory variable is the average regional ratio of loans guaranteed under the recovery plan in 2009-2010 scaled by SME assets, computed excluding firms within 10 miles of a regional border. All regressions include department pair fixed effects and distance to the border. Changes in regional controls from 2008 to 2010 (local taxes, equipment expenditures, public debt, and state contribution, all scaled by population) are added in columns (2) and (5). Firm-level controls added in columns (3) and (6) include log of assets, log of firm age, and two-digit industry fixed effects. Firm controls are measured in 2008. Standard errors are clustered by region. *, **, and *** denote significance at the 10%, 5%, and 1%, respectively.