Non-Performing Loans, Financial Stability, and Banking Competition: Evidence for Listed and Non-Listed Eurozone Banks*

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Abstract

This article analyses the effects of non-performing loans (NPLs hereafter) and financial (in)stability on banking competition in the Eurozone lending markets. Our results suggest that NPLs increase marginal costs and the degree of competition in banking markets. This phenomenon is economically more significant for non-listed than for listed banks. The results are robust to any endogeneity due to the impact of the economic downturn and the sovereign debt crisis on the growth of NPLs.

Keywords: Bank competition, banking and financial crisis, listed banks, NPLs, sovereign debt crisis.

JEL Classification: G01, G21, G32, H63.

1. Introduction

Until the onset of the financial crisis of 2008, the level of non-performing loans (NPLs) remained relatively stable. From then on, the bad loan portfolios of European banks embarked on the same downward trend as the macroeconomic environment.¹ At the time of writing this paper, the NPL problem is concentrated in the so-called peripheral countries of the Eurozone. According to data published by the European Banking Authority (EBA), Eurozone bank balance sheets are still heavily weighed down by bad loans which are slowing down the lending market, and threatening financial stability.²

Banks keeping large volumes of NPLs on their balance sheets may create doubts about their viability in the event of worsening macroeconomic conditions, and increase uncertainty about

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their future profitability and asset value. Such banks are exposed to negative risk perception by market players, which could face them with costlier financing and a marginal cost increase (Chiesa and Mansilla-Fernández, 2018a,b). On the demand side, adverse macro-economic conditions, the sovereign debt crisis, and the deleveraging of the real sector may have contributed to lowering the price of loans. Analysing the market, one would expect banks affected by an increase in bad loans to reduce lending and narrow their mark-ups. In specific terms, the first question posed by this paper is: *Can it be proven that bad loans and, by extension financial (in)stability, may affect the level of competition between banks in the lending market?* To date, the financial literature has presented various conclusions as to the repercussions of the level of competition on financial stability (e.g., Boyd and Nicoló, 2005; Martínez-Miera and Repullo, 2010; Jiménez *et al.*, 2013; IJtsma *et al.*, 2017). As far as we know, this is one of the first studies to demonstrate that financial stability can affect the level of competition.

In second place, this paper investigates whether access to capital markets can mitigate the effects of the transmission mechanism described above. In specific terms, the research question could be formulated as follows: *Do bad loans have the same effect on the level of competition in listed as in non-listed banks?* The literature cited above reveals that larger banks –of the type sometimes described as “too big to fail” (TBTF)– and systemic banks tend to accumulate higher levels of risk than smaller ones (Cai *et al.*, 2018; Salas and Saurina, 2002; Stern and Feldman, 2004). This paper considers the role of capital management in assessing levels of competition.³ Listed banks have greater market access capacity and greater potential to increase their capital base than is the case for non-listed banks. Non-listed banks should therefore face a higher marginal cost increase than listed banks. In line with previous studies (e.g., Accornero *et al.*, 2017; Louzis *et al.*, 2012; Zhang *et al.*, 2016), this paper measures the NPL ratio as the volume of NPLs over total loans to customers. Graph 1 illustrates the motivation for this research.

Graph 1
EVOLUTION AND DISTRIBUTION OF THE NPL RATIO FOR
LISTED VS. NON-LISTED BANKS

(a) *Evolution of the NPL ratio*

(Continued.)

(b) *Distribution of the NPL ratio*

Source: Bankscope and Orbis Bank Focus (*Bureau van Dijk*). The NPL ratio is measured as the volume of loans to customers. Graph 1.a shows the evolution of the mean value of the NPL ratio for listed (solid line) and non-listed (broken line) banks. Graph 1.b is a box and whisker plot of the distribution of the NPL ratio for listed and non-listed banks. The whiskers show the upper and lower limits of the distribution. The line through each box indicates the median, i.e., the 50th percentile of the distribution. The upper (lower) boundaries of the box are the upper (lower) quartiles.

Graph 1.a shows that the average NPL ratios of listed banks are higher than those of non-listed banks in the period preceding the financial crisis of 2007. The slope of the time series reveals that listed banks generate NPLs at a faster rate than non-listed banks. Graph 1.b displays the distribution of the NPL ratio before (2002Q1-2007Q2) and after the onset of the financial crisis (2007Q3-2016Q4) in the form of a box and whisker plot. As the graph shows, the median listed bank holds a higher stock of NPLs than the average non-listed bank.

The empirical analysis in this paper uses a panel of high-frequency data drawn from Bankscope and Orbis Bank Focus (*Bureau van Dijk*) containing 388 Eurozone banks for the period 2002 to 2016. The panel is completed with market and macroeconomic data drawn from the Thomson Reuters Datastream. With respect to the variables used in the econometric analysis, financial stability is calculated using the *Z-score* indicator as employed in the financial literature to assess the inverse probability of bank default (Laeven and Levine, 2009; Schaeck *et al.*, 2012; Liu *et al.*, 2013; Köhler, 2015). In line with previous studies (Cruz-García *et al.*, 2017, 2018; Carbó *et al.*, 2009; Maudos and de Guevara, 2004, 2007), the degree of competition between banks is calculated using the Lerner index, which measures bank's level of market power by relating price to marginal cost. As a robustness check, this

result is tested against the Boone (2008) indicator which measures the temporary impact of marginal cost on the bank's share in the lending market. The Ordinary Least Squares (OLS) estimator with fixed effects is used for the panel data, and the Instrumental Variables (IV) estimator is used to check whether the final result is conditioned by the effect of crisis-related factors on the NPL level. This study contributes to the research by bringing to light that higher NPL ratios increase banks' marginal cost and also lending market competition. The results suggest, furthermore, that non-listed banks are more vulnerable than listed banks to changes in NPLs.

The remainder of the paper is organized in five sections. Section two discusses the related literature. Section three presents the hypotheses, defines the database and describes the empirical model and study variables. Section four provides a discussion of the results and the paper closes with a concluding section.

2. Literature review

There now follows a discussion on the theoretical basis for the questions to be addressed and the methodology to be used in this research, and an explanation of how the study fits into the existing body of research.

The first question to be addressed is whether NPLs and financial (in)stability are determining factors of the levels of competition in banking markets. Until now, the economics literature has focused on the level of banking competition as a determining factor for financial stability and the amount of risk assumed by banks (e.g., Allen and Gale, 2004; Coccoresse, 2004; Martinez-Miera and Repullo, 2010).⁴ Although these studies share the common denominator of treating banking competition and financial stability as a trade-off, their conclusions are ambiguous. On the one hand, the "competition fragility" predicts that any increase in the level of competition will reduce mark-ups and lower the franchise value of banks, causing them to increase their risk taking (IJtsma *et al.*, 2017). The intuition behind this reasoning is simple: banks operating in relatively non-competitive markets are more able to identify the more solvent borrowers; and this allows them to increase their interest margin, capital ratios and franchise value (Coccoresse, 2004). The "competition stability" view, on the other hand, predicts that banks operating in low competition settings raise loan prices to relatively uncompetitive levels, thus increasing their customers' probability of default, and thereby their own portfolio risk, while posing a threat to financial stability (Boyd and Nicoló, 2005). The empirical literature demonstrates that there is a parabolic relationship between banking competition and financial stability (Martinez-Miera and Repullo, 2010; Liu *et al.*, 2013). This study contributes to the literature by disentangling the transmission mechanism through which the opposite occurs: that is, financial (in)stability determines banking competition. The process is explained by applying standard industrial organization models to banking markets (e.g., VanHoose, 2017). The literature leaves open the question of estimating the thresholds above which NPLs become pernicious for the proper functioning of the banking sector. With respect to the first issue, in a study conducted on a sample of Chinese banks, Zhang *et al.* (2016) report a threshold of around 4.8 %, above which banks start to exhibit

moral hazard-type behaviour; sustaining future losses as a consequence (Hellmann *et al.*, 2000).⁵ Beyond this threshold, bank profitability suffers because the resources for lending are diminished (e.g., Angelini, 2018; Angelini *et al.*, 2017). Banks address this problem by laying up insolvency provisions as a safety net in case they need to write off some or all of their doubtful loans (Mohaddes *et al.*, 2017; Zhang *et al.*, 2016). The size and evolution of NPL ratios are observable data. Banks with rising NPL trends are perceived among financial market players as more being risk-taking than the rest (Accornero *et al.*, 2017). Bank borrowers and capital investors, alike, may question the viability of such banks and hence demand non-diversifiable risk compensation (e.g., Chiesa and Mansilla-Fernández, 2018 a, b). In other words, if banks find it harder to access funding (Aiyar *et al.*, 2015; Chiesa and Mansilla-Fernández, 2018 a), it can be assumed that the price inputs used for lending will increase, and there will be a vertical shift in the marginal cost curve (VanHoose, 2017; Degryse and Ongena, 2008).

The empirical literature studies whether the NPL transmission mechanism is static, i.e., measured in levels; or dynamic, i.e., measured as changes in the amount of NPLs on the balance sheet. This paper is closely related with that of Accornero *et al.* (2017) where it is shown that “change” in the NPL level rather than the NPL ratio *per se*, is the transmission mechanism through which the loan supply decreases. Meanwhile, Bending *et al.* (2014) use a dynamic approach to show how an increase in NPLs reduces the supply of loans to non-financial firms. In line with the streams of research discussed above, this study uses the dynamic approach to test whether increases in the NPL ratio work as a transmission mechanism affecting competition in the Eurozone banking markets. The results confirm that an increase in the variation of the NPL ratio causes a marginal cost increase for banks, but provide no empirical evidence that an increase (decrease) in the NPL ratio (as an indicator of financial stability) has any impact on the price of loans.⁶ In a monopolistic market setting, the size of the markup is limited by the fact that banks have less room for manoeuvre; which is a symptom of higher competition. The results of this study suggest, furthermore, that an upward trend in NPLs reduces lending market concentration.

The second question that this study will try to answer is whether this effect is more significant in the case of listed or non-listed banks. As far as we are aware, this is one of the first studies to investigate the repercussions of financial (in)stability on banking market competition, while taking into consideration whether the banks are listed in secondary banking markets.⁷ In seeking an answer to this question, this paper finds its basis in the literature highlighting the role of bank size, in particular the size of banks deemed “too-big-to-fail” (TBTF), as a transmission mechanism leading banks to engage in risky trading and increase their stock of NPLs (Cai *et al.*, 2018; Salas and Saurina, 2002; Stern and Feldman, 2004). Stern and Feldman (2004) show that, in comparison with non-listed banks, TBTF banks are relatively less disciplined by their *stakeholders*, who are confident that such entities are covered by government protection schemes –i.e., deposit guarantees or the possibility of rescue in the event of failure (Boyd and Graham, 1998; Nier and Baumann, 2006). The seminal work by Boyd and Gertler (1994) shows that a policy benefiting the big US banks caused an increase in NPL levels. In a similar vein, Louzis *et al.* (2012), studying a sample of Greek banks, show that fulfilment, or otherwise, of the TBTF hypothesis depends on each loan cat-

egory. Using the principal-agent theory, Rud *et al.* (2018) show that the bank (the agent) and the borrower (the principal) are more closely aligned in a low competition setting, where the entity has less risk exposure. However, the findings of Cai *et al.* (2018) show that syndicated loans, despite being extended by a group of banks, can increase systemic risk due to the effect of interconnections between banks during periods of economic recession. This research contributes to the economics literature by demonstrating that the effect of an NPL increase on marginal cost is economically more significant in non-listed than in listed banks. This can be explained by the fact that access to capital markets is more restricted for non-listed than for listed banks. By incorporating risk, non-listed banks would be accepting harsher financing conditions than those faced by listed banks, thus increasing their input costs. Non-listed banks would therefore have less lending capacity and suffer greater losses in terms of market share and mark-up than listed banks.

3. Hypotheses, data and methodology

3.1. Hypotheses

The purpose of this research is to analyse the effect of the NPL ratio on competition between banks, making a distinction between listed and non-listed banks. Based on the reviewed literature, the following hypotheses are proposed:

Hypothesis 1.—*The NPL ratio increases competition in markets for bank loans.*

Hypothesis 2.—*The NPL ratio has a more significant impact on the level of competition for non-listed than for listed banks.*

3.2. The data

The main sources of data for this study are Bankscope and Orbis bank Focus (*Bureau van Dijk*). The database is formed from quarterly data for Eurozone Banks for the period 2002Q1-2016Q4.⁸ The initial sample consisted of consolidated state data on commercial banks, savings banks and loan cooperatives. All the financial entities included in the sample provide data from 1st January to 31st December. Because Bankscope and Orbis Bank Focus make a distinction between listed and non-listed entities, a variable is extracted to allow the sample to be split into these two groups. Banks are selected at the highest possible level of consolidation, usually as banking groups, in order to avoid double-entry problems. Those that do not form banking groups are kept as individual entities. The final sample consists of 388 financial entities, of which 86 are listed and 302 are non-listed.

The sample years are considered representative because they include the years preceding the financial crisis, when banks were extending more loans and assuming higher levels of risk (2002Q1-2007Q2), as well as the years of the banking and sovereign debt crisis (2007Q3-2016Q4).

Table 1
VARIABLE DEFINITION

Variable	Acronym	Definition
<i>Bank variables</i>		
Bank size	Dim_{it}	Natural logarithm of bank's total assets (TA_{it}).
Risk-adjusted marginal cost	MC_{it}	This variable measures the cost to the bank of each additional loan excluding the risk premium (Jiménez <i>et al.</i> , 2013; Martín-Oliver <i>et al.</i> , 2006).
Loan price	P_{it}	Financial and non-financial earnings over TA_{it}
Banking market competition	$Lerner_{it}$	This index is defined as the difference between the bank's marginal price and cost as a percentage of the price. This variable measures the bank's mark-up.
The Boone indicator	B_{ht}	This variable measures the degree of banking market concentration. This indicator is calculated as a country aggregate.
Profitability	ROA_{it}	Earnings before tax over total assets.
Capitalization	CAR_{it}	Equity over total assets.
Standard deviation of profitability	$\sigma(ROA_{it})$	Standard deviation of ROA calculated over a three-year window.
Financial stability	$Ln(Z_{it})$	Natural logarithm of the Z-score calculated as $Z_{it} = (ROA_{it} + CAR_{it}) / \sigma(ROA_{it})$.
Non-performing loan ratio	NPL_{it}	This variable is measured as the volume of NPLs over total loans to customers.
Earnings structure	INC_{it}	Earnings other than interests (operating income minus interest margin) over gross margin.
Efficiency ratio	EFF_{it}	Operating costs over operating income.
Leverage ratio	LEV_{it}	Total assets over equity.
Banking concentration	HHI_{ht}^{CR}	The Herfindhal-Hirschman index measures the level of banking market concentration. It is defined as the square of the summed market shares of the banks of each country.
Listed bank indicator	L_i	This qualitative variable takes the value one for a listed bank and zero otherwise.
<i>Macroeconomic variables</i>		
Crisis	$Crisis_t$	This qualitative variable takes the value one after the onset of the financial crisis ($t \geq 2007Q3$), and zero for the period before.
Economic growth	GDP_{ht}	Rate of variation in gross domestic product (GDP) for each country.
Inflation	INF_{ht}	Harmonized inflation rate in each country.

Note: Subscripts i , t , and h denote bank, period and country, respectively.

The data are expressed in thousands of Euros and deflated by the harmonized consumer price index (HCPI). Inconsistent observations, such as zero total assets, negative equity values, and zero number of workers, were removed from the sample. Finally, a panel of data comprising 23,265 observations was obtained.

The macroeconomic time series data for each country were drawn from the Thomson Reuters Datastream. The banking data and macroeconomic data were then combined into a single panel.

Table 1 contains the definitions of the study variables, which are winsorized at 1 % to remove any outliers that might skew the results.

3.3. Measures of banking competition

This study uses the Lerner index and Boone's (2008) indicator as measures of the degree of banking market competition. Both these indicators use marginal cost as the main variable.

3.3.1. The Lerner Index

This study uses the risk-adjusted Lerner index as a measure of the degree of banking market competition in the Eurozone. This index measures the bank's mark-up capacity as a percentage of the price and can vary between zero and one, where $Lerner_{it} = 1$ in the case of perfect monopoly, $Lerner_{it} = 0$ in the case of perfect competition, and, exceptionally, $Lerner_{it} < 0$ in the case of non-optimal performance, in which case the bank must resort to mark-down. The Lerner index is calculated as follows:

$$Lerner_{it} = \frac{P_{it} - MC_{it}}{P_{it}} \quad (1)$$

where the subscripts i and t denote the bank and the time period, respectively. The variable P_{it} is calculated as the ratio of financial and non-financial revenue to total assets. MC_{it} is the marginal cost. Given that banks take the risk premium into account when setting loan prices, a higher mark-up does not necessarily indicate a lower degree of competition in the banking market, and may be due to an increase in the risk cost (Cruz-García *et al.*, 2018). Thus, if marginal cost were estimated in the conventional manner, the Lerner index estimates could be biased. For this reason, the risk-adjusted marginal cost is estimated as follows (see Martín-Oliver *et al.*, 2006; Jiménez *et al.*, 2013):⁹

$$MC_{it} = \frac{r_t + PD_{it} - LGD}{1 - PD_{it} - LGD} \quad (2)$$

where r_t is the risk-free interest rate measured as the quarterly average of the three-month Euribor. PD_{it} is the probability of default measured as the quotient between impairment loss on financial assets and the loan volume.¹⁰ LGD is the value of the loans the bank will be unable to recoup in the event of default. Following Jiménez *et al.* (2013), LGD is assigned the value

45 %, as recommended in the bank capital rules laid down by the Basel Banking Supervision Committee.

3.3.2. The Boone (2008) indicator

The Boone (2008) indicator is another measure of banking competition which is used as a robustness check for the Lerner index. The Boone indicator assumes that more efficient banks will gain a greater share of the market and increase their profits at the expense of less efficient entities. In other words, competition improves (worsens) the profitability of more (less) efficient Banks. It is estimated by means of the following expression:

$$\ln(MS_{it}) = \alpha_0 + \sum_{t=2002Q1}^{2016Q3} \beta_t MC_{it} + \sum_{t=2002Q1}^{2016Q3} \theta_t D_t + \varepsilon_{it} \quad (3)$$

where MS_{it} is the lending market share for each bank measured as the ratio of the i bank's volume of customer loans to total bank loans for country h in period t . The term MC_{it} denotes the marginal cost as defined in the section above. The matrix D_t includes discrete time variables to control for factors common to all banks within each specific period. Coefficient β_t is the Boone indicator. Negative values ($\beta_t < 0$) mean that more cost-efficient banks gain market share from comparatively less efficient banks. A large negative value of β_t is indicative of a relatively more competitive market, while values higher than zero ($\beta_t > 0$) mean that higher marginal costs increase the bank's market share. Heteroscedasticity can be overcome by means of a logarithmic transformation.

3.4. Financial stability indicators: the NPL ratio and the Z-score

This study uses two measures of risk exposure: the NPL ratio (NPL_{it}), and, as a robustness check, the *Z-score*. The NPL ratio, calculated as the volume of NPLs over total customer loans, is the variable of interest. The *Z-score*, widely used in the literature on financial stability (e.g., Laeven and Levine, 2009; Schaeck *et al.*, 2012; Liu *et al.*, 2013; Köhler, 2015), is calculated as follows:

$$Z_{it} = \frac{ROA_{it} + CAR_{it}}{\sigma(ROA_{it})} \quad (4)$$

where ROA_{it} is the profitability ratio measured as earnings before tax over total assets; CAR_{it} , is the capitalization ratio measured as equity capital over total assets; and $\sigma(ROA_{it})$ is the standard deviation of ROA computed for a three-year rolling time window. A moving standard deviation is used to allow for time variability of the denominator.

The *Z-score* is interpreted as the decrease in the standard deviation of its ROA that would be required to absorb the bank's equity capital. In other words, the *Z-score* is the inverse probability of bank failure.

3.5. Empirical specification

This section presents a discussion on strategies for identifying and testing hypotheses regarding the effects of NPLs on lending market structure.

The empirical methodology for this study is based on the following Ordinary Least Squares (OLS) estimator with fixed effects for panel data:

$$y_{it} = \gamma_0 + \gamma_1 \Delta NPL_{i,t-1} + X'_{i,t-1} \Phi + v_i + \epsilon_{it} \quad (5)$$

where y_{it} is the dependent variable for the following indicators. The first is the Lerner index ($Lerner_{it}$), which is computed as in expression (1). The second, used as a robustness check, is the Boone (2008) indicator, computed as shown in expression (3).

For the explanatory variables, the indicator of interest is the quarterly change in the one-period lagged NPL ratio (ΔNPL_{it}) computed as the volume of NPLs over total customer loans. Accornero *et al.* (2017) show that it is not the NPL ratio, *per se*, but its variation that determines loan demand. Taking this finding from the literature into account, a test is performed to determine whether a change in the NPL ratio has a negative impact on banking market power. This is followed with a robustness check in which this variable is replaced with quarterly change in the natural logarithm of the Z-score ($\Delta \ln(Z_{i,t-1})$), computed as indicated in expression (4). The matrix $X'_{i,t-1}$ captures the following control variables lagged by one period. Bank size ($Dim_{i,t-1}$) is measured as the natural logarithm of the bank's total assets. The income structure ratio ($INC_{i,t-1}$) is computed as the ratio of non-interest earnings (operating earnings minus the interest margin) to operating earnings and is a measure of bank revenue diversification. The efficiency ratio ($EFF_{i,t-1}$) is calculated as operating costs over operating income and is used to control for bank (in)efficiency. Financial leverage ($LEV_{i,t-1}$) is computed as the ratio of total assets to total equity.¹¹ The Herfindhal-Hirschman concentration index ($HHI_{h,t-1}^{Cr}$), which indicates the level of concentration in the bank lending market, is calculated as the squared sum of the market shares of banks in every country h . Formally, the Herfindhal-Hirschman index is calculated as $HHI_{h,t}^{Cr} = \sum_{i=1}^N (MS_{it})^2$, where MS_{it} is the lending market share of each bank in the country in which it operates (h). Standard Industrial Organization models predict equivalence between the Lerner index and the HHI, proportional to market contestability and inverse price elasticity of demand. At the empirical level, however, the HHI and the Lerner index can yield diverse results if controls for the input-output patterns of regional markets are not included (Carbó-Valverde *et al.*, 2003, 2009; Carbó *et al.*, 2009). Previous studies have used the HHI as a determinant of the Lerner index, although the ratio can become increasingly endogenous with growing market contestability (see, Carbó-Valverde and Rodríguez-Fernández, 2007; Maudos and de Guevara, 2004, 2007).

Macroeconomic variables are then included to control for business cycle effects on market structure. GDP_{ht} denotes the quarterly rate of change in the gross domestic product of

every country (h); while the discrete variable, $Crisis_t$, which is included to control for potential structural change, is assigned a value of one from the onset of the financial crisis onwards ($t \geq 2007Q3$), and zero otherwise.

Finally, to control for possible endogeneity issues in the drivers of loan default, the NPL ratio (ΔNPL_{it}) is instrumentalized with the macro-economic factors discussed in section 2:

$$\Delta \widehat{NPL}_{it} = \lambda_0 + \lambda_1 UNEM_{ht} + \lambda_2 CDS_{ht} + \varpi_{it} \quad (6)$$

where $\Delta \widehat{NPL}_{it}$ is the predicted value of the NPL ratio based on the selected instruments; $UNEM_{ht}$ is the unemployment rate for each country; CDS_{ht} denotes the 10-year sovereign CDS value for each country, which is used as the country risk proxy. Since Thomson Reuters Datastream data on this variable are only available from 2008 onwards, the second part of the exercise covers the crisis period (2008Q1-2016Q4) exclusively. ϖ_{it} is the error component of the regression equation. This second exercise is approached using the instrumental variables (IV) estimation method.

4. Results

4.1. Descriptive statistics and parametric tests

A description of the sample distribution and the values registered before and after the crisis are summarized in table 2. Panel A shows the descriptive statistics of the study variables. With respect to the competition indicators, the Lerner index ($Lerner_{it}$) shows an average of 0.31, oscillating between 0.10 and 0.50; and the Boone indicator (B_{ht}) an average value of -0.16, oscillating in the range of -0.29 to -0.05. The Herfindhal-Hirschman index ($HHI_{h,t}^{cr}$) for the lending market (HHI_{it}^{cr}) shows an average of 0.37, oscillating between approximately 0.001 and 1.00. Marginal cost (MC_{it}) shows an average of 0.03, oscillating between 0.01 and 0.09; while the average price of loans (P_{it}) is 0.04, oscillating between 0.01 and 0.14. As far as the financial stability variables are concerned, the NPL ratio (NPL_{it}) shows an average of 8.21 % for the Eurozone as a whole, oscillating approximately between 0.01 % and 54.55 %. The quarterly variation in the NPL ratio (ΔNPL_{it}) shows an average of 0.0034, oscillating between -0.05 and 0.07. The natural logarithm of the Z-score ($Ln(Z_{it})$) shows an average of 1.69, oscillating between -5.89 and 4.96, in line with previous findings. The quarterly variation in the variable ($\Delta Ln(Z_{it})$), meanwhile, shows an average of 0.007, oscillating between -5.89 and 4.96. Finally, the average value of the dummy variable L_{it} suggests that 22.18 % of the banks included in the sample are listed; which means that the mean individual is a non-listed entity.

Panel B shows the mean differences between the pre-crisis period (2002Q1-2007Q2) and the post-crisis period (2007Q3-2016Q4). These were obtained using the time variable, $Crisis_t$. The mean differences for the Lerner index ($Lerner_{it}$) reject the null hypothesis ($H_1: Crisis_t(0) - Crisis_t(1) < 0$) thus indicating a decline in competition following the onset of the crisis.

Table 2
DESCRIPTIVE STATISTICS

This table shows the distribution of the variables used in this study between 2002Q1 and 2016Q4. Panel A shows the sample distribution for the study period as a whole. Panel B shows the means of the same variables for the Pre-crisis period (2002Q1 - 2007Q2) and the Post-crisis period (2007Q3 - 2016Q4). The parametric means estimates the statistical difference between the sample means of the variables under the null hypothesis of equality between the two periods ($H_0: Crisis_t(0) - Crisis_t(1) = 0$), where $Crisis_t = 1$ if $t \geq 2007Q3$. The sample includes commercial banks, savings banks, and loan cooperatives in Eurozone countries for each period. The monetary values are deflated. All the bank variables are winsorized at 1% to remove outliers. All the las variables are defined in Table 1.

Panel A: Description of the sample (2002Q1 - 2016Q4)								Panel B: Differences in means $H_0: Crisis_t(0) - Crisis_t(1)$		
N	Mean	SD	Min.	Pc. 25	Median	Pc. 75	Max.	Pre-crisis	Post-crisis	p-value
<i>Bank variables</i>										
$Lerner_{it}$	23,265	0.3101	0.1116	0.1049	0.2108	0.3197	0.4048	0.2781	0.3254	0.0000
B_{it}	23,265	-0.1553	0.0789	-0.2926	-0.2025	-0.1649	-0.0749	-0.2474	-0.1437	0.0000
MC_{it}	23,265	0.0252	0.0163	0.0039	0.0158	0.0193	0.0318	0.0385	0.0187	0.0000
P_{it}	23,265	0.0366	0.0237	0.0055	0.0209	0.0294	0.0456	0.0545	0.0279	0.0000
NPL_{it}	23,265	0.0821	0.0923	0.0000	0.0273	0.0528	0.1000	0.0433	0.0882	0.0000
ΔNPL_{it}	23,265	0.0034	0.0154	-0.0483	-0.0016	0.0012	0.0062	0.0004	0.0037	0.0737
$Ln(Z_{it})$	23,265	1.6915	0.9214	-5.8953	1.1365	1.6624	2.2429	1.9854	1.6248	0.0000
$\Delta Ln(Z_{it})$	23,265	0.0067	1.2976	-3.6539	-0.9209	-0.5027	1.2496	-0.0342	0.0136	0.0000
Dim_{it}	23,265	23.2958	2.1284	17.3990	21.8864	23.0259	24.6151	23.0485	23.3513	0.0006
INC_{it}	23,265	0.3747	0.2822	-0.1153	0.2075	0.3099	0.4459	0.3894	0.3710	0.0102
EFF_{it}	23,265	0.3271	0.1780	0.0002	0.2286	0.2873	0.3729	0.3183	0.3291	0.0149
LEV_{it}	23,265	15.2321	7.9851	2.4545	10.0000	136.364	18.0000	16.7522	14.8900	0.0000
MS_{it}	23,265	0.2050	0.2699	0.0009	0.0090	0.0649	0.3299	0.1864	0.2875	0.0000
HHH_{it}^{Cr}	23,265	0.3711	0.3189	0.0000	0.0000	0.3581	0.5318	0.3282	0.3959	0.0000
L_{it}	23,265	0.2218	0.4155	0.0000	0.0000	0.0000	0.0000	—	—	—
<i>Macroeconomic variables</i>										
$Crisis_t$	23,265	0.6333	0.4819	0.0000	0.0000	1.0000	1.0000	—	—	—
GDP_{ht}	23,265	0.0151	0.0323	-0.1930	0.0044	0.0157	0.0297	0.0293	0.0069	0.0000
INF_{ht}	23,265	1.9705	1.8140	-3.8667	0.8667	1.9333	2.7000	2.5105	1.6577	0.0000

Note: Subscripts i , t , and h subscripts refer to bank, time and country, respectively.

Likewise, the parametric test rejects the null for the Boone indicator (Bh_t) thereby confirming a decline in competition in the Eurozone banking markets during the post-crisis period ($H_1: Crisis_t(0) - Crisis_t(1) < 0$). This is due to a significant drop in marginal cost (MC_{it}) and the price of loans ($H_1: Crisis_t(0) - Crisis_t(1) > 0$), and to an increase in the lending market concentration indices (MS_{it} and $HHI_{h,t}^{cr}$) during the post-crisis period ($H_1: Crisis_t(0) - Crisis_t(1) < 0$). The results for the financial stability variables support a relatively higher NPL ratio (NPL_{it}) for the post-crisis period. The mean difference results also enable rejection of the null hypothesis for the quarterly growth of the NPL ratio (ΔNPL_{it}) in favour of the post-crisis period ($H_1: Crisis_t(0) - Crisis_t(1) < 0$). The results for the financial stability indicator ($Ln(Z_{it})$) remain qualitatively consistent with the those reported above, enabling rejection of the null hypothesis in favour of the pre-crisis period for the level-based indicator ($\Delta Ln(Z_{it})$) ($H_1: Crisis_t(0) - Crisis_t(1) > 0$). Analysis of the overall results of the parametric hypothesis test by sub-periods shows that the crisis:

- i. Reduces competition in the bank lending markets of the Eurozone as a result of the banking integration processes carried out in several countries.¹²
- ii. Reduces financial stability because adverse macroeconomic conditions trigger an increase in NPLs, leading to a drop in bank earnings, and, in some cases, even to losses.

4.2. The baseline regression

This section discusses the econometric results which prove hypothesis 1. Table 3 gives the regression coefficients for expression (5). The hypotheses for this study are tested using OLS regression analysis with fixed effects for panel data (columns (1)-(4)). The estimation results show that a 1 % increase of variation in the NPL ratio (ΔNPL_{it}) triggers significant reductions in the Lerner index ($Lerner_{it}$) and the Boone indicator (of 0.75 % and 0.54 %, respectively). This result is robust to replacing ΔNPL_{it} with the variation in the natural logarithm of the *Z-score* ($\Delta Ln(Z_{it})$). The estimates suggest that a 1 % increase in $\Delta Ln(Z_{it})$ causes $Lerner_{it}$ and B_{ht} to grow by 0.72 % and 0.26 % ($p\text{-value} < 0,000$), respectively. The results of this study show that a growing trend in the amount of risk held on bank balance sheets, due either to an increase in NPLs or to a lack of liquidity or a drop in earnings, has a positive impact on the degree of competition in the lending markets of the Eurozone. The control variables have the expected signs and significance levels.

The results reported above are robust to the inclusion of instrumental variables (IV), as shown in expression (6), to control for the effect of the economic crisis on the growth of the NPL ratio (e. g., Balgova *et al.*, 2016, 2018; Buch *et al.*, 2016; Ghosh, 2017; Podstawski and Velinov, 2018). The estimates of the instrumental variables are given in columns (5)-(8). The sign and magnitude of the IV estimator support the notion that the degree of competition in banking markets is sensitive to increases in the NPL ratio and to financial (in)stability. Furthermore, the first-stage analysis fulfils initial expectations by confirming that the unemployment rate ($UNEM_{h,t-1}$) and 10-year sovereign CDS (CDS_{ht-1}) drive NPL growth.

4.3. Explaining the transmission mechanism

The next question to arise in the light of the above results concerns the mechanism through which variations in the NPL ratio and in financial (in)stability modify the level of banking market competition. Previous studies have shown that banks with relatively high NPL levels have opted to gamble for resurrection by giving riskier loans (e. g., Hellmann *et al.*, 2000). This study contributes to the literature by analysing the effect of variations in the NPL ratio on banks' marginal costs (MC_{it}), loan prices (P_{it}) and shares in the lending market (MS_{it}). Table 4 reports the OLS estimates for panel data with fixed effects.

Table 4
THE EFFECTS OF BANK NPLs AND FINANCIAL STABILITY ON MARGINAL COST, LOAN INTEREST RATES AND MARKET SHARE

This table shows the results when a variation in financial stability, defined as the NPL ratio (NPL_{it}) and the natural logarithm of the *Z-score* ($Ln(Z_{it})$), regressed against marginal cost (MC_{it}), loan price (P_{it}), and market share (MS_{it}) as dependent variables. Specifications (1), (2), (3), (4), (5) and (6) are estimated using the Ordinary Least Squares (OLS) estimator for panel data with fixed effects for the period 2002Q1-2016Q4. All the regressions contain individual fixed effects, and the standard errors are clustered by countries. *, **, *** denote minimum 10 %, 5 %, 1 % significance, respectively.

	MC_{it}		P_{it}		MS_{it}	
	(1)	(2)	(3)	(4)	(5)	(6)
ΔNPL_{it}	0.005** (0.002)		-0.065 (0.049)		-0.327*** (0.033)	
$\Delta Ln(Z_{it})$		-0.002** (0.001)		-0.001 (0.002)		0.315*** (0.029)
$Dim_{i,t-1}$	-0.045*** (0.007)	-0.037*** (0.009)	0.025* (0.012)	0.028* (0.014)	0.247*** (0.047)	0.293*** (0.042)
$Dim_{i,t-1}^2$	0.021*** (0.006)	0.017*** (0.005)	-0.001* (0.000)	-0.001* (0.000)	-0.534*** (0.054)	-0.523*** (0.052)
$INC_{i,t-1}$	-0.035** (0.014)	-0.024** (0.011)	0.064** (0.024)	0.060** (0.023)	0.210*** (0.055)	0.245*** (0.048)
$EFF_{i,t-1}$	0.131*** (0.026)	0.128*** (0.025)	0.026** (0.009)	0.016*** (0.005)	-0.238*** (0.075)	-0.228*** (0.073)
$LEV_{i,t-1}$	0.051*** (0.012)	0.063*** (0.013)	-0.048** (0.017)	-0.043** (0.018)	-0.082*** (0.012)	-0.081*** (0.011)
$HHI_{h,t-1}^{Cr}$	-0.248*** (0.001)	-0.214*** (0.002)	0.040** (0.012)	0.036** (0.014)		
$(HHI_{h,t-1}^{Cr})^2$	0.064*** (0.011)	0.053** (0.022)	-0.034 (0.027)	0.036 (0.025)		
GDP_{ht}	-0.211*** (0.031)	-0.227*** (0.032)	0.088** (0.029)	0.094*** (0.022)	0.142** (0.083)	0.156** (0.097)
$INF_{h,t-1}$	0.021*** (0.005)	0.023*** (0.008)	0.002** (0.001)	0.001** (0.000)	-0.085*** (0.013)	-0.087*** (0.016)
$Crisis_t$	0.037*** (0.001)	0.037*** (0.001)	-0.006*** (0.001)	-0.003*** (0.001)	-0.155*** (0.076)	-0.171** (0.075)
N	5,160	5,160	5,160	5,160	5,160	5,160
Wald test [p-value]	0.000	0.000	0.000	0.000	0.000	0.000

Columns (1)-(2) give the estimates of the regression for banks' marginal costs, which suggest that a variation in the NPL ratio (ΔNPL_{it}) increases the marginal cost by 0.005 % ($p\text{-value} < 0.005$). They also reveal that an increase in financial stability ($\Delta Ln(Z_{it})$), or, equivalently, a reduction in the bank's default risk, reduces the marginal cost by 0.002 %. It should be noted that data on the evolution of the NPL ratio and the default risk and profitability ratios are readily observable by the bank's borrowers and investors. If these perceive the bank to be gradually increasing its risk ratio, they might compromise its future profitability and asset value by demanding more compensation for the extra non-diversifiable risk (Chiesa and Mansilla-Fernández, 2018 a, b). Put another way, the accumulation of risk increases the bank's input costs, thereby increasing the marginal cost of loans. The diminishing demand trend may have put downward pressure on loan prices (BdE, 2017). The results of this study provide no empirical evidence to suggest that an increase in the NPL ratio had any effect on loan prices (P_{it}) [columns (3)-(4)]. This may be due to the influence of a reduction in the demand for loans resulting from the economic slowdown and the process of deleverage in the real sector (e. g., Balgova *et al.*, 2016, 2018; Ghosh, 2017; BdE, 2017). Finally, the estimates shown in columns (5)-(6) confirm that an increase in bank NPLs reduces banks' lending market shares (MS_{it}). An overall analysis of the results enables the conclusion that the accumulation of risk on bank balance sheets, as shown by an increase in the NPL or default risk ratio, drives lending market competition by reducing cost efficiency. As a result, some banks suffer a loss of market share to more efficient banks, and thus see a reduction in their mark-up.

4.4. Listed vs. non-listed banks

This section discusses the statistics confirming hypothesis 2. The sample is split by creating a qualitative variable (L_i) which takes the value one if the shares of bank i are listed in the secondary markets ($L_i = 1$), and zero otherwise ($L_i = 0$). Table 5 reports a parametric means test for the main study variables under the null hypothesis of equality between non-listed and listed banks ($H_0: L_i(0) - L_i(1) = 0$). The data were also split into a pre-crisis period (2002Q1-2007Q2) and a post-crisis period (2007Q3-2016Q4). The test results indicate that the differences in the Lerner index ($Lerner_{it}$) between listed and non-listed banks emerge during the pre-crisis period, but fail to reject the null hypothesis for the crisis period. It is important to note that the test rejects the null for the marginal cost (MC_{it}), thereby indicating that non-listed banks are comparatively less efficient than listed banks during the crisis period ($H_1: L_i(0) - L_i(1) < 0$).

According to the above results, the parametric test shows that listed banks are relatively larger (Dim_{it}) and hold less risk (LEV_{it}) than non-listed banks. Furthermore, the results for the NPL ratio (NPL_{it}) suggest that the differences between the two types of bank arise after the onset of crisis, the mean value of this variable being higher for listed than for non-listed banks. Meanwhile, the difference-in-means test fails to reject the null hypothesis for the quarterly rate of variation in the NPL ratio (ΔNPL_{it}), thereby indicating that NPL growth is similar in both types of banks. However, the difference-in-means test reveals a relatively higher financial stability indicator ($Ln(Z_{it})$) for the non-listed than for the listed banks ($H_1: L_i(0) - L_i(1) > 0$) during both periods, while failing to reject the null for the quarterly rate of

variation in the financial stability indicator ($\Delta \ln(Z_{it})$) (see Salas and Saurina, 2002; Louzis *et al.*, 2012; Cai *et al.*, 2018).¹³

Table 5
PARAMETRIC MEANS TEST FOR LISTED VS. NON-LISTED BANKS

This table splits the sample into two groups of individuals: listed banks ($L_i = 1$) and non-listed banks ($L_i = 0$). The parametric test is conducted under the null hypothesis $H_0 = L_i(0) - L_i(1) = 0$ for the pre-crisis period (2002Q1-2007Q2) and post-crisis period (2007Q3-2016Q4). All the variables are defined in Table 1. The coefficients are for the mean values, while the standard error is given in parentheses. The *p-value* of the *Student-t statistic* is shown in the third column of each period of analysis.

	Pre-crisis (2002Q1-2007Q2)			Post-crisis (2007Q3-2016Q4)		
	$L_i = 0$	$L_i = 1$	<i>p-value</i>	$L_i = 0$	$L_i = 1$	<i>p-value</i>
$Lerner_{it}$	0.2514 (0.0186)	0.2905 (0.0131)	0.0470	0.3176 (0.0123)	0.3307 (0.0099)	0.2043
MS_{it}	0.2744 (0.0159)	0.3137 (0.0212)	0.0736	0.1749 (0.0059)	0.2063 (0.0079)	0.0007
NPL_{it}	0.0415 (0.0037)	0.0465 (0.0043)	0.1998	0.0845 (0.0030)	0.0943 (0.0035)	0.0194
ΔNPL_{it}	0.0027 (0.0032)	-0.0019 (0.0014)	0.0944	0.0030 (0.0017)	0.0039 (0.0007)	0.2953
$\ln(Z_{it})$	2.0903 (0.0425)	1.6142 (0.1353)	0.0000	1.6491 (0.0206)	1.5454 (0.0324)	0.0061
$\Delta \ln(Z_{it})$	-0.0265 (0.1386)	-0.0589 (0.2482)	0.4549	0.0179 (0.0501)	0.0021 (0.0768)	0.4333
MC_{it}	0.0381 (0.0018)	0.0386 (0.0012)	0.4008	0.0163 (0.0004)	0.0205 (0.0017)	0.0246
P_{it}	0.0518 (0.0028)	0.0558 (0.0021)	0.1354	0.0241 (0.0007)	0.0305 (0.0024)	0.0192
Dim_{it}	22.5387 (0.0959)	24.0611 (0.1422)	0.0000	22.8102 (0.0446)	24.2896 (0.0673)	0.0000
INC_{it}	0.3897 (0.0089)	0.3884 (0.0132)	0.4689	0.3711 (0.0040)	0.3712 (0.0063)	0.4904
EFF_{it}	0.3212 (0.0045)	0.3096 (0.0076)	0.1973	0.3358 (0.0027)	0.3062 (0.0037)	0.0000
LEV_{it}	16.1694 (0.3921)	17.9097 (0.4699)	0.0036	14.2135 (0.1858)	16.0630 (0.2501)	0.0000

Table 6 gives the results of a repeat econometric test, with separate regressions for listed ($L_i = 1$) and non-listed ($L_i = 0$) banks, aimed at checking for any significant differences in the coefficients for the two groups. The results shown in Panel A suggest that the impact of the NPL ratio on the level of banking competition ($Lerner_{it}$) is economically more significant for listed than for non-listed banks. A robustness check including the natural logarithm of the *Z-score* ($\Delta \ln(Z_{it})$) as a regressor confirms these findings. The robustness of the above results is tested in eight alternative estimations, shown in Panel B, where $Lerner_{it}$ is replaced with B_{ht} as the dependent variable. The results are qualitatively similar to those obtained with the Lerner index. The results of the analysis of the transmission mechanism in Panel C confirm that bank NPLs and financial instability cause a greater increase in marginal cost (MC_{it}) in non-listed banks than in those that are listed.

Table 6

THE EFFECTS OF NPLs AND FINANCIAL STABILITY ON MARGINAL COST, THE LERNER INDEX AND THE BOONE INDICATOR FOR LISTED VS. NON-LISTED BANKS

This table shows the results when a variation in financial stability, defined as the NPL ratio (NPL_{it}) and the natural logarithm of the *Z-score* ($Ln(Z_{it})$) regressed against the dependent variables, marginal cost (MC_{it}) in Panel A, the Lerner index ($Lerner_{it}$) in Panel B, and the Boone indicator (B_{ht}) in Panel C. The regressions are estimated using high-frequency data. L_i is a qualitative variable that takes the value one if bank i is listed and zero otherwise. Specifications (1), (2), (3) and (4) are estimated using the Ordinary Least Squares (OLS) estimator for panel data with fixed effects for the period 2002Q1-2016Q4. Specifications (5), (6), (7) and (8) are estimated using the Instrumental Variables (IV) estimator for the period 2008Q1-2016Q4. The selected instruments are the NPL ratio (NPL_{it}), the unemployment rate ($UNEM_{h,t-1}$) and one-quarter-lagged 10-year sovereign CDS ($CDS_{h,t-1}$) for country h . Instruments validated by Sargan test, under the null of exogeneity, checking the lack of correlation between the instruments and the error term of the main equation. All the regressions contain control variables and individual fixed effects, and the standard errors are clustered by countries. *, **, ***, denote minimum 10 %, 5 %, 1 % significance, respectively.

	OLS with fixed effects				IV with fixed effects			
	$L_i = 1$ (1)	$L_i = 0$ (2)	$L_i = 1$ (3)	$L_i = 0$ (4)	$L_i = 1$ (5)	$L_i = 0$ (6)	$L_i = 1$ (7)	$L_i = 0$ (8)
<i>Panel A: Dependent variable: $Lerner_{it}$</i>								
ΔNPL_{it}	-0.547** (0.183)	-0.732** (0.213)			-0.321** (0.128)	-0.363** (0.139)		
$\Delta Ln(Z_{it})$			0.114** (0.013)	0.161** (0.011)			0.353*** (0.044)	0.537*** (0.065)
Sargan test [p-value]					0.292	0.239	0.368	0.364
<i>Panel B: Dependent variable: B_{ht}</i>								
ΔNPL_{it}	-0.520** (0.117)	-0.569** (0.116)			-0.534*** (0.116)	-0.572*** (0.121)		
$\Delta Ln(Z_{it})$			0.324*** (0.068)	0.347*** (0.080)			0.278*** (0.076)	0.358*** (0.082)
Sargan test [p-value]					0.211	0.284	0.302	0.287
<i>Panel C: Dependent variable: MC_{ht}</i>								
ΔNPL_{it}	0.019** (0.007)	0.024** (0.009)			0.019*** (0.004)	0.024*** (0.005)		
$\Delta Ln(Z_{it})$			-0.001** (0.000)	-0.002** (0.000)			0.001** (0.000)	0.002** (0.000)
Sargan test [p-value]					0.824	0.746	0.242	0.250
N	5.160	5.160	5.160	5.160	5.160	5.160	5.160	5.160
$X'_{i,t-1}$	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes

Discussion.—The results of this study suggest that the examined transmission mechanism is relatively more sensitive for non-listed than for listed banks. Previous research has shown that an accumulation of NPLs on a bank's balance sheet reduces its profitability and asset value (Coccorese and Girardone, 2017; Zhang *et al.*, 2016). Under current financial regulations, banks are required to hold a minimum capital buffer to fund new lending (Berger and Bouwman, 2013; Jiménez *et al.*, 2017; Den Heuvel, 2008)¹⁴. Non-listed banks have com-

paratively fewer options than listed banks do when it comes to raising capital in secondary markets, and are forced to obtain funding from deposits, securitization activities and inter-bank markets. Investors are therefore likely to harbour more doubts about the profitability and asset value and to demand more compensation for non-diversifiable risk from non-listed than from listed banks. In other words, higher input costs have a greater incremental effect on marginal cost and a greater decremental effect on lending in non-listed than in listed banks (Chiesa and Mansilla-Fernández, 2018 a, b). As a result, non-listed banks lose more lending market share than their competitors, and have to reduce their mark-ups to remain competitive.

5. Conclusions

This paper analyses the sensitivity of lending market competition to variations in the NPL ratio and the (inverse) probability of bank default, i.e., financial (in)stability. A unique database is constructed for this purpose by combining Bankscope accounting data (*Bureau van Dijk*) with Thomson Reuters Datastream macroeconomic and market data for each country in the Eurozone.

The main finding of this study is that higher NPL ratios and greater financial instability increase the level of competition in lending markets. In line with studies predicting that an increase in NPLs will increase banks' funding costs (Aiyar *et al.*, 2015; Chiesa and Mansilla-Fernández, 2018 a, b), this paper shows that banks suffer both market share losses and mark-up losses due to an increase in marginal cost. The results of this study also suggest that market access can compensate for potential capital loss due to an increase in NPLs. To be specific, non-listed banks are economically more vulnerable than listed banks to a loss of market share.

As far as can be ascertained, this is one of the first studies to analyse the repercussions of financial stability for banking market competition. The results are robust to endogeneity tests. However, because data on 10-year sovereign CDS are available only from 2008 onwards, the analysis of their impact on the NPL ratio is limited to the crisis period. The effects of the sale of NPLs on secondary markets and the introduction of sovereign guarantees are issues which remain for future research.

Notes

1. European supervisors consider a loan to be doubtful when there are signs that the borrower will be unable to repay it within ninety days, not having paid any of the agreed instalments. In the worst case scenario, the borrower is unable to repay the loan and the bank has to write the loan off the balance sheet.
2. According to EBA data, in June 2018, the Eurozone countries with the highest non-performing loans ratios are Greece (44.81 %), Cyprus (34.12 %), Portugal (12.44 %), Italy (9.73 %), Ireland (7.01 %), and Spain (4.24 %).
3. The Basel -III accord requires banks to hold a minimum capital buffer for withstanding risk.
4. See Giannoccolo and Mansilla-Fernández (2017) for a more thorough review of the literature.

5. The *moral hazard hypothesis* predicts that banks with low capital are relatively more prone to generate doubtful loans because they are able to increase the level of risk in their credit portfolios (a strategy known as *gambling to resurrection*) (see Berger and DeYoung, 1997; Bouwman and Malmendier, 2015).
6. Under the standard monopolistic competition model, a drop in credit demand can compensate for a possible price increase due to a marginal cost increase (see Kolmar, 2017; Pindyck and Rubinfeld, 2001). Accornero *et al.* (2017) show that an increase in the variation of the NPL ratio implies a reduction in the demand for loans. This result is due to economic slowdown (Balgova *et al.*, 2016, 2018; Ghosh, 2017), the sovereign debt crisis (Buch *et al.*, 2016; Podstawski and Velinov, 2018), and deleverage of the real sector during the crisis period (BdE, 2017) causing the loan demand curve to shift to the left.
7. Listed banks are usually larger in size, relatively more cost efficient, and have higher levels of market concentration than non-listed banks (e.g., Delis *et al.*, 2017; Meslier *et al.*, 2017; Phan *et al.*, 2018).
8. The Eurozone countries are Germany, Austria, Belgium, Cyprus, Slovakia, Slovenia, Spain, Estonia, Finland, France, Greece, Holland, Ireland, Italy, Latvia, Lithuania, Luxembourg, Malta, and Portugal.
9. I am grateful for an anonymous reviewer's suggestions for calculating the price and the risk-adjusted marginal cost.
10. Jiménez *et al.* (2013), adopting the approach used by Martín-Oliver *et al.* (2006), use probability of default data (PD_{it}) from the Bank of Spain's Risk Information Centre (CIR, for its acronym in). Access to the RIC not being possible, this study uses Bankscope accounting data instead.
11. Various studies calculate the leverage ratio as equity over total assets. The inverse enables interpretation of the ratio as the liabilities-to-equity ratio. The reasoning for this is as follows:

$$LEV_{it} = \frac{'Liabilities' + 'Equity'}{'Equity'} = \frac{'Liabilities'}{'Equity'} + 1$$
12. This mechanism has two component trends associated with the Banking Union: a cyclical component due to recent expectations of economic recovery and a clearing of balance sheets encouraging bankers to start planning for the future (Schoenmaker, 2015; Duijm and Schoenmaker, 2017); and a structural component due to the real sector's dependence on bank financing, and the coming Capital Markets Union (Langfield and Pagano, 2016). Banks prepare for new competition in the loans market—from mutual funds, pension funds and insurance companies, etc.—by upsizing and pursuing economies of scale through mergers (Evanoff and Ors, 2008).
13. The reader should recall that the higher the value of Z_{it} , the lower bank default risk. This result suggests that non-listed banks assumed higher levels of risk than listed banks did during the study period.
14. Under the Basel -III accord, banks are required to hold a tier 1 common capital buffer and a capital conservation buffer equivalent to at least 4.5 % and 2.5 %, respectively, of its risk-weighted assets. The following table summarizes the Basel-III reforms. Downloaded December 28, 2018: https://www.bis.org/bcbs/basel3/b3_bank_sup_reforms_es.pdf.

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Resumen

Este trabajo analiza los efectos de la morosidad crediticia y de la (in)estabilidad financiera sobre la competencia bancaria en los mercados de crédito de la zona euro. Los resultados sugieren que la morosidad bancaria incrementa el coste marginal y el grado de competencia de los mercados bancarios. Este fenómeno es económicamente más significativo para los bancos no cotizados que para los bancos cotizados. Los resultados son robustos a posibles problemas de endogeneidad derivados de la recesión económica y de la crisis de deuda soberana sobre el crecimiento de la morosidad.

Palabras clave: Bancos cotizados, competencia bancaria, crisis financiera y bancaria, deuda soberana, NPLs.

Clasificación JEL: G01, G21, G32, H63.