Salvador Climent-Serrano (Spain), Jose M. Pavía (Spain)

An analysis of loan default determinants: the Spanish case

Abstract

The Spanish financial system is experiencing a very turbulent economic period in which loan defaults have lived an unprecedented increasing period, going from being less than 1% in 2004 to levels of above 13% in 2013. The impact of this, along with other circumstances, has led to the greatest financial restructuring ever made to date in Spain, with important macroeconomic and microeconomic consequences. This paper studies the determinants of delinquency (loan default) in Spanish credit institutions for the period of 2004-2011 and introduces new variables that have been disclosed as relevant in the current financial crisis as well as others non- previously considered internal variables, such as hedging derivatives (which are having an increasingly greater importance in accounts of Spanish credit institutions). Among the most prominent variables that have had a significant impact on the increase of delinquency are, among external variables, house prices, unemployment rates and the number of companies declaring bankruptcy and, among internal variables, property investment, customer credits over active, interest rate, participated companies and solvency rates. The analysis also shows significant differences in delinquency’s behavior between savings banks and banks and between credit institutions that needed recapitalization and those that did not.

Keywords: default, delinquency, risk management, systemic risk, Spanish financial system.

JEL Classification: G21, M41.

Introduction

The Spanish financial system is facing a very turbulent economic period in which delinquency (loan defaults) in Spanish financial institutions has experienced an unprecedented increasing period, going from being less than 1% in 2004 to levels of above 13% in 2013 (Banco de España, 2013). Although Spain follows the pattern demonstrated in other countries (Swedberg, 2013), ensuing the macroeconomic consequences of losses in lending banks and reductions in the issuance of new loans, the Spanish financial crisis presents some particular characteristics that make it singular. Among them, the enormous amount of investment made by financial institutions in real estate at the heat of the housing boom, the lag registered in Spain in the bursting of the housing bubble, the high leverage of Spanish institutions with foreign investors and the deep recession in which Spain seems to have fallen, which has considerably damaged its businesses and employment.

The impact of this unprecedented increase of delinquency in the Spanish financial institutions, along with other circumstances, has led to the greatest financial restructuring ever made to date in Spain. Through an intense process of fusions and absorptions, the sector has experienced a really significant reduction in the number of its institutions, mainly focused on savings banks (Climent, 2012) – which have passed from 45 to 10 changing as well their corporate structure to become commercial banks. This process has practically resulted in the disappearance of traditional savings banks, which had previously formed part of the Spanish financial system for almost 200 years and represented a 45% share of the financial market. Further, in line with other developed countries (Weber and Schmitz, 2011; Rose and Wieladek, 2012), the Spanish credit institutions have received direct public funding to guarantee their sustainability. Funds that in the case of Spain have exceeded 6% of the Spanish GDP (FROB, 2013).

In light of the events to which have contributed in part delinquency, it is paramount that further analyses of the impact of default on Spanish credit institutions is carried out and also to further study its determinants in order to construct an econometric model from which explain aggregate default levels in the Spanish financial companies. From the perspective of banks, a proper management of risks requires the ability to monitor these numbers and, from a policy perspective, it is an important to quantify these risks in order to know the likelihood of new additional recapitalizations and when the issuance of new loans is going to recover. Indeed, one area that has received relatively little attention is the interrelation between risk, risk management and management accounting and control practices, some issues that likely should be reformulated (reconstructed) after the financial crisis (Huber and Scheytt, 2013) and will lead to more risk-related disclosures (Oliveira et al., 2013).

The rest of the document is structured as follows. Section 1 offers a background of the issue. Section 2 presents methodology and details objectives. Section 3 shows results. And, final section states the main conclusions.

1. Background

The analysis of the impact of loan default on the Spanish financial system and the study of its determinants is not new and has already been the object of interest in other periods. Fernandez de Lis et al. (2000) analyzed the determinants of delinquency in the period of 1963-1999 and pointed to economic cycle and competency increase as the most significant determinants. Freixas et al. (1994) focused on the period of 1973-1992 using variables such as economic growth, infla-
tion, expectations of economic activity and interest rate. Delgado and Saurina (2004) examined during the period 1982-2001 the relationship between the most relevant macroeconomic variables (such as GDP variation, interest rates, level of debt, financial burden and asset prices) and the risk of credit in banks and savings banks. Furthermore, Saurina (1998) tries to explain delinquency from 1985 to 1995 using both internal and external variables. Salas and Saurina (2002) use a wide range of internal and external variables to obtain the determinants of risky loans between 1985 and 1997, distinguishing between savings banks and banks. And, more recently, Gonzalez and Diez (2010) relate credit growth, delinquency and economic cycle to deduce that credit grows and delinquency decreases in expansive cycles whilst the opposite occurs during recessions. Foos et al. (2010), in a study involving 16 European countries including Spain, note that credit growth entails an increase in loss provisions during the three years following. De los Rios et al. (2012) study the difference between savings banks and banks with respect to risk management and social responsibility. And, Jiménez et al. (2013) find a negative relationship between loan market power and bank risk during the period of 1988-2003.

Regarding studies on other countries different from other than Spain, one can cite the papers published by Crook and Banasik (2012), Magri and Pico (2012), Esteban and Estrada (2013), and Ramcharan and Crowe (2013), among the latest researches. Crook and Banasik (2012) relate delinquency rate to debt levels, interest rates and house prices in the USA during the period between 1987 and 2009. Magri and Pico (2012) relate interest and delinquency rates in the Italian financial institutions in the period of 2000-2007. Esteban and Estrada (2013) found house prices, outstanding mortgage debts and, to a lesser extent, income levels to be the principal determinants of delinquency in the Columbian financial institutions from 1997 to 2004; while Ramcharan and Crowe (2013) show that house price fluctuations have a significant impact on credit markets well beyond the mortgage segment, being thus a potentially catalyst for macroeconomic shocks. In this line, Vallascas and Keasey (2012) identify which bank characteristics offer a shelter from systemic shocks and demonstrate some key determinants of a bank’s risk exposure.

There also exists a literature on the subprime mortgages in the USA, among which Demyanyk and Van Hemert (2011), Brent et al. (2011), Roll (2001), Sarmiento (2012) and Dell’Ariccia et al. (2012) can be cited. Demyanyk and Van Hemert (2011) maintain that the high rate of growth in house prices masked the delinquency of subprime mortgages between 2003 and 2005. Brent et al. (2011) find that the mortgagee’s income and the economic cycle were determinants of delinquency in the period from 2004 to 2009. Sarmiento (2012) shows unemployment increases playing a very significant role in mortgage defaults rise during the Great Recession. Roll (2011) proposes a completely alternative tale and disagrees about the most popular explanations of the financial crisis, advocating for a different US macroeconomic policy. And Dell’Ariccia et al. (2012) conclude that during the periods of rapid credit growth lending conditions were relaxed, with the relaxation increasing when competition and also house prices rise. A summary of the variables used as determinant of delinquency in all the above mentioned papers is shown in Table 1.

This paper adds to the delinquency literature a couple of new features that we think could be of interest. Firstly, it is centred on a long and homogenous period in Spain that up until now had not been studied; a period in which the delinquency in Spain went from historic minimums to historic maximums with a trend of continuous growth. Secondly, regarding the variables used, the study includes both new internal and external variables. In particular, in addition to the determinants highlighted in the overview of the delinquency literature, this study incorporates new account variables, such as hedging derivatives, the portfolio of participated companies and property investments (within the internal variables) and the increase in the number of companies in bankruptcy (within the external variables), that apparently have not been considered until now.

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Increase in loans</td>
<td>X X X X X X</td>
<td>X X X X</td>
<td>X X X</td>
<td>X X X</td>
<td>X X X</td>
<td>X X X</td>
<td>X X X</td>
<td>X X X</td>
<td>X X X</td>
<td>X X X</td>
<td>X X X</td>
<td>X X X</td>
</tr>
<tr>
<td>Increase in competence</td>
<td>X</td>
<td>X</td>
<td>X</td>
<td>X</td>
<td>X</td>
<td>X</td>
<td>X</td>
<td>X</td>
<td>X</td>
<td>X</td>
<td>X</td>
<td>X</td>
</tr>
</tbody>
</table>
2. Objectives and methodology

2.1. Objectives. The objective of this research is to construct an econometric model from which to explain the level of delinquency, using as indicators all the relevant variables that may have an impact on delinquency and on its effects. These determinants will include both microeconomic (e.g., house prices) and macroeconomic external factors (e.g., unemployment rates) as well as internal factors of the credit institutions (e.g., property assets or leverage of the institution). As such, this will create a tool that can help to manage the impact of certain decisions with respect to the control of delinquency. Likewise, the model will permit the projection of the impact on delinquency of some of the external variables — for example, two of them, house prices and the unemployment level, are included in the model with lags of order one and two, respectively.

In addition, and given the different impact the economic and financial events have had on banks and saving banks (see, Climent-Serrano and Pavia, 2014) a separate analysis has been carried out on each of the groups in order to analyze the possible differences. In this way, we can study whether the type of corporate governance has a differential impact on the development of delinquency, since as Park (2012, p. 907) points out “corruption significantly aggravates the problems with bad loans in the banking sector”. The analysis will also be extended to study (through two new regressions) the differences between institutions that have received aid and those that have not. The constructed models therefore could be used to improve our information levels and our knowledge about the mechanism that operate in this issue, making it easier to predict future trends in delinquency and consequently enabling the necessary preventative measures to be taken for banks, their shareholders, the supervisor and stakeholders in general.

2.2. Methodology. To perform the analysis, unbalanced panel data regression techniques are used combining macroeconomic variables with information from the latest financial annual accounts reports of the period of 2004-2011 of all the Spanish savings banks and the 13 largest commercial banks, which together make up 99% of the assets of Spanish banks and savings banks. The total number of institutions in 2004 was 58, while there were 20 in 2011. This difference is the result of the restructuring process (which is still ongoing) of the mergers and absorptions that occurred.
in the course of 8 years and especially in 2010 (Climent Serrano, 2013).

Financial institutions annual accounts reports are drawn up and approved by the board of directors of each entity and by their general meetings or shareholders’ meetings, undergo internal audit controls and are audited by independent companies overseen by the Bank of Spain. However, despite all these controls, these measures are not always sufficient to ensure that the accounts reflect fairly the assets of the institutions, as has occurred in those credit institutions that had to be intervened, characterized by pre-crisis weaker corporate governances, where the new administrators have reformulated the accounts having come to light significant losses in all cases (see, e.g., Climent-Serrano and Pavia, 2015a). Therefore, the accounts used have been the new reformulated accounts, which show a high level of losses.

Using this information, an econometric model is constructed based on an unbalanced panel data with 400 observations in which dependent variable is loan default, measured as a ratio between non-performing loans and the total amount of loans. As explicative variables, in addition to the ones already studied in the previous literature, four new features have been also considered: property investments, participated companies, hedging derivatives and, number of companies in (bankruptcy or) administration. Below (in subsection 2.3) the independent variables of the model are listed.

The availability of information for a sample of between 58 and 20 institutions (depending on the year) and 8 time periods, allows the use of a double temporal and cross-section dimension of the sample through an econometric model of non-balanced panel data. This provides us, compared to just cross-sectional databases, with a more informative database (with more variability, less collinearity and more degrees of freedom) that will permit us to attain more efficient estimates in an econometric linear regression model and mainly to control for endogeneity and/or individual unobserved heterogeneity (Arellano and Bond, 1991). Therefore, along with the components that could determine delinquency, it is acknowledged that each institution may have specific characteristics, \( \alpha_i = \alpha + \nu_i \), albeit non-measurable and non-observable, which can affect the relationship. Hence the model ultimately specified has been:

\[
\text{Delinquency}_{ijt} = \alpha_i + \sum_{j=1}^{J} \beta_j \text{int.var}_{ijt} + \sum_{k=1}^{K} \beta_k \text{ext.var}_{ijt} + \epsilon_{ijt},
\]

\( \beta \) being the coefficient of the \( j^{th} \) the internal variable in the regression model, \( \nu_i \) the coefficient of the \( \epsilon_{ijt} \) external variable, subindex \( j \) denotes entity, \( t \) represents time, \( I \) is the number of internal variables, \( E \) is the number of external variables and \( \epsilon_{ijt} \) is a disturbance error term.

According to the goals stated in subsection 2.1, five different variants of this model have been considered. The first model, the baseline model, is made up of all the institutions and all the time periods available, in total, 71 entities and 400 observations. The second and third models are made by dividing the complete sample into two sub-samples, savings banks and banks, with the objective of studying the possible differences between them. In models four and five, the sample is sub-divided into those entities that received help and those that did not.

2.3. Variables. The list of account features handled has been, as internal variables, investment properties (ratio of investment in properties and assets), leverage (ratio of customer deposits on loans) increased appropriations set (loan increasing less nominal GDP increasing), hedging derivatives (ratio of hedging derivatives over assets), real estate assets (ratio of secured loans to total loans), ratio client credit-active (ratio of customer loans over assets), interest rate (interest rate of the asset), participated companies (ratio of investees on participated companies over assets), solvency (equity ratio of the group over assets) and, as external variables, bankruptcy (annual variation in the number of companies in bankruptcy), unemployment rate (-2) (unemployment rate of the Spanish economy with a delay of two years) and house prices (-1) (annual change in the average price of housing in Spain, according to the Ministry of Industry, with a lag of a year). House price lag has been set in one to include information delays and owners’ psychological elasticity and unemployment rates are included with a lag of order two to account for the unemployment regulation benefits prevailing in Spain during the study period.

3. Results

The econometric software Eviews in its version 7 has been used to estimate with pooled data, fixed effects and random effects the models specified in the previous section. Homogeneity tests indicate that fixed effects models are preferable to pooled data models while Hausman tests point to the fixed effects specification as preferable to the random effects one. Hence, the fixed effects model will be the one presented in this research.

Table 2 shows the results obtained in the regressions. There are no problems with co-linearity or correlation problems among the predictors employed and as the Durbin-Watson test shows (see the bottom of Table 2) there are no remarkable concerns of residual autocorrelation. Likewise, the \( R^2 \) results obtained indicate that the regressions explain a large percentage of delinquencies in all models used. This latter result may lead to the idea that there is a problem of overparameterization affecting the model specifications. In order to dispel this doubt and following the philosophy of machine learning (Bishop, 2006), the sample was
randomly divided into two groups, one of modelization (learning group) with 80% of the observations and the other of evaluation (test group) with the remaining 20%. The test results showed an excellent predictive capability of the model, consistent with the observed $R^2$.

Although as a rule a similar core group of variables are identified as significant in the plurality of the models, with the one with the greatest difference being the unaided entities’ model, large differences between groups could be observed with respect to the impact of the determinants. Focusing first on the baseline model, we find that the variables that have been revealed as significant with a positive sign, and therefore contributing to the increase in delinquency, are property investment, interest rate, among the internal variables, and unemployment rates and number of companies declaring bankruptcy, among the external variables. On the other hand, with a negative impact, meaning their increase leads to a reduction in delinquency, we can find participated companies, solvency and the ratio of customer credit over active, as internal variables, and house prices, as external variable.

<table>
<thead>
<tr>
<th>Variables</th>
<th>Baseline model</th>
<th>Savings banks</th>
<th>Banks</th>
<th>Differences</th>
<th>Aided entities</th>
<th>Unaided entities</th>
<th>Differences</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>0.078***</td>
<td>0.085***</td>
<td>0.043*</td>
<td>0.041*</td>
<td>0.045***</td>
<td>0.008</td>
<td>0.037</td>
</tr>
<tr>
<td></td>
<td>(0.013)</td>
<td>(0.016)</td>
<td>(0.024)</td>
<td>(0.016)</td>
<td>(0.024)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Property investment</td>
<td>0.434***</td>
<td>0.419***</td>
<td>0.427**</td>
<td>-0.008*</td>
<td>0.544***</td>
<td>0.148</td>
<td>0.396**</td>
</tr>
<tr>
<td></td>
<td>(0.084)</td>
<td>(0.099)</td>
<td>(0.197)</td>
<td>(0.099)</td>
<td>(0.139)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Leverage</td>
<td>-0.007</td>
<td>-0.008</td>
<td>-0.011</td>
<td>0.003</td>
<td>0.003</td>
<td>-0.017</td>
<td>0.019**</td>
</tr>
<tr>
<td></td>
<td>(0.007)</td>
<td>(0.010)</td>
<td>(0.009)</td>
<td>(0.008)</td>
<td>(0.011)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Credit increase</td>
<td>0.001</td>
<td>-0.033**</td>
<td>0.001</td>
<td>-0.034**</td>
<td>0.018</td>
<td>-0.002</td>
<td>0.020*</td>
</tr>
<tr>
<td></td>
<td>(0.006)</td>
<td>(0.015)</td>
<td>(0.005)</td>
<td>(0.012)</td>
<td>(0.005)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Hedging derivatives</td>
<td>-0.115</td>
<td>-0.230</td>
<td>0.160</td>
<td>-0.390</td>
<td>0.251</td>
<td>-0.072</td>
<td>0.324</td>
</tr>
<tr>
<td></td>
<td>(0.186)</td>
<td>(0.227)</td>
<td>(0.331)</td>
<td>(0.224)</td>
<td>(0.261)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Real estate assets</td>
<td>0.008</td>
<td>0.007</td>
<td>0.021**</td>
<td>-0.014*</td>
<td>-0.002</td>
<td>0.012*</td>
<td>-0.015*</td>
</tr>
<tr>
<td></td>
<td>(0.005)</td>
<td>(0.006)</td>
<td>(0.008)</td>
<td>(0.006)</td>
<td>(0.006)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Customer credit over active</td>
<td>-0.054***</td>
<td>-0.075***</td>
<td>-0.011</td>
<td>-0.063**</td>
<td>-0.044**</td>
<td>0.044</td>
<td>-0.087**</td>
</tr>
<tr>
<td></td>
<td>(0.018)</td>
<td>(0.024)</td>
<td>(0.027)</td>
<td>(0.020)</td>
<td>(0.037)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Interest rates</td>
<td>0.373***</td>
<td>0.492***</td>
<td>-0.117</td>
<td>0.608***</td>
<td>0.548***</td>
<td>-0.168</td>
<td>0.716***</td>
</tr>
<tr>
<td></td>
<td>(0.10)</td>
<td>(0.147)</td>
<td>(0.199)</td>
<td>(0.140)</td>
<td>(0.198)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Participated companies</td>
<td>-0.204***</td>
<td>-0.198</td>
<td>-0.072</td>
<td>-0.126</td>
<td>-0.335**</td>
<td>0.041</td>
<td>-0.376**</td>
</tr>
<tr>
<td></td>
<td>(0.067)</td>
<td>(0.142)</td>
<td>(0.117)</td>
<td>(0.139)</td>
<td>(0.104)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Solvency</td>
<td>-0.603***</td>
<td>-0.658***</td>
<td>-0.128</td>
<td>-0.530***</td>
<td>-0.737***</td>
<td>-0.172</td>
<td>-0.565***</td>
</tr>
<tr>
<td></td>
<td>(0.011)</td>
<td>(0.078)</td>
<td>(0.149)</td>
<td>(0.083)</td>
<td>(0.106)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>House prices (-1)</td>
<td>-0.204***</td>
<td>-0.172***</td>
<td>-0.195***</td>
<td>0.023</td>
<td>-0.255***</td>
<td>-0.162***</td>
<td>-0.093***</td>
</tr>
<tr>
<td></td>
<td>(0.011)</td>
<td>(0.020)</td>
<td>(0.016)</td>
<td>(0.021)</td>
<td>(0.014)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Unemployment rate (-2)</td>
<td>0.235***</td>
<td>0.339***</td>
<td>0.079*</td>
<td>0.259***</td>
<td>0.441***</td>
<td>0.153***</td>
<td>0.288***</td>
</tr>
<tr>
<td></td>
<td>(0.033)</td>
<td>(0.044)</td>
<td>(0.045)</td>
<td>(0.017)</td>
<td>(0.040)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Bankruptcy</td>
<td>0.017***</td>
<td>0.015***</td>
<td>0.012**</td>
<td>0.003</td>
<td>0.018***</td>
<td>0.018***</td>
<td>0.000</td>
</tr>
<tr>
<td></td>
<td>(0.002)</td>
<td>(0.003)</td>
<td>(0.005)</td>
<td>(0.003)</td>
<td>(0.004)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>


The rest of the variables did not prove to be significant. Leverage, which had a negative coefficient, and credit increase, with a coefficient practically zero, cannot be considered as important determinants in delinquency. It also highlights that the hedging derivatives variable, which aims to reduce the risk, does not appear to be significant, in this case nevertheless the coefficient is negative, so keeping this sign would reduce delinquency. This meaningfulness of hedging could occur, according to Engel (2013), for the use of derivatives not only to hedge but also for speculative purposes.

Analyzing the two models that serve to distinguish between savings banks and banks, we see that the variable participated companies lost its significance in savings banks while the variable credit increase becomes significant with a negative coefficient. This last result tells us that the increase in credit, more so than nominal GDP increase, contributed to the moderation of the increase of delinquency in the savings banks.
Comparing the model of the banks to the baseline model we observe that a large number of internal variables become not significant, in particular the four variables: ratio of credit over active, participated companies, solvency and interest rate. The result for this last variable is really interesting given that it is no longer significant and its coefficient changes from positive to negative, telling us that the impact of an increase in the interest rate did not contribute to delinquency in banks, when in the case of savings banks, this variable is not only significant but it shows a coefficient greater than in the baseline model. Also remarkable is the behavior of the variable real estate assets, which after being not significant in the baseline model, becomes significant in the case of banks’ model.

Comparing the estimated coefficients between both models, significant differences could be observed between savings banks’ and banks’ models. In particular, and in addition to the differences in the impact of interest rate already pointed out, it highlights the difference between the coefficients of the unemployment level, 0.339 for savings banks and 0.079 for banks. That in a situation as the one in the current labor market, with levels of unemployment above 20%, suggests that the impact on savings banks is greater than on banks, without doubt a consequence of their different loan portfolio composition, which indeed is a reflection of the different risk credit policies they followed during the housing boom and their different target populations. In general, these differences in the regression coefficients, both internal and external, along with the different structure of the distributions in internal determinants between banks and savings banks, explain why the average level of delinquency during the whole period in the banks is 0.66% less than in the savings banks.

Analyzing the model of the institutions that have received aid, we observe the same significant variables as in the baseline model, whilst the model for non-aided institutions shows hardly any significant internal variables, with only real estate assets being significant. The significance of all the external variables, however, is maintained in both models.

Comparing the coefficients for these two models, one can see that among the internal variables there are significant differences in interest rates, participated companies and solvency and, to a lesser extent, in leverage, credit increase, real estate assets and client credit over active. Indeed, almost all the coefficients of the interval variables are significantly different in the two models. Furthermore, among the external variables, it highlights the significant differences in house prices and unemployment rates that in addition are significant in both models, something that hardly happens with internal variables. The impact of the reduction in house prices and the increase in unemployment rates is greater in those institutions that received help than in those that did not need help.

As in the models of banks and savings banks, these differences in the regression coefficients, both internal and external, along with the different structure of the distributions of the variables studied, explain why the average rate of delinquency has been 1.17% higher during the whole period in the institutions receiving help than in the unaided institutions.

Conclusions

The current socio-economic environment has contributed significantly to increase delinquency in Spanish credit institutions. In fact, according to the results of the regressions, the evolution of external variables in recent years has significantly impacted on the levels of delinquency. Both decrease in house prices and increase in unemployment jointly with the number of companies undergoing bankruptcy have contributed to rising loan defaults. This impact was also seen to be higher in savings banks than in banks and in those institutions that received aid rather than those that did not, without doubt as a consequence of their different target populations and risky policies, more soft-information based (Parnes, 2012), which are reflected in the composition of their loan portfolios.

As far as internal variables concerns, an increase in property investment and an increase in interest rate have been seen to contribute to an increase in delinquency whilst increases in the ratio of customer loans to assets, participated companies and solvency help to reduce delinquency.

The variables leverage, credit increase, real estate assets and hedging derivatives have proved to be insignificant. It is noticeable that variables such as leverage or credit increase have hardly had any repercussion on delinquency, being unimportant and with very small coefficients. Equally, it was initially surprising that expected results were not achieved for the hedging derivatives, as its impact is not significant, although its coefficient does appear with an expected negative sign. The impact of some of the variables has been very different between savings banks and banks, especially in relation to interest rate and unemployment level which helps to explain the difference of 0.66% on average in delinquency rates throughout the period.

In this sense, the results of this study add to those of other studies that also found differences between banks and savings banks in different periods and in different ways, such as Coello (1994), Pastor (1994), Azofra and Santamaria (2004), Fonseca et al. (2011), Climent (2012) and Climent and Pavia (2015b).

Finally, significant differences are also found between institutions that received aid and those that did not, the main differences being in the variables property in-
investment, interest rate, participated companies, solvency, house prices and unemployment rates. This situation has resulted in the average delinquency level being 1.17% higher in aided institutions, which eventually led them being faced with massive recapitalization that accounted for 6% of the Spanish GDP.

References


