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## Abstract

Following a common wisdom in banking, revenue diversification is likely to produce a "portfolio-effect" in bank income statement, enhancing the creation of greater and more stable profits. We test this hypothesis on a sample of 110 large commercial, saving and cooperative banks headquartered in 8 EMU countries for the period 2005-2013. Results indicate that diversification strategies have had a role in determining banks profitability only for selected subsamples (and in particular for commercial and saving banks); on the contrary, efficiency and credit portfolio quality have been the main drivers of profits in the period under examination. We contribute to previous literature using a cross-country balanced dataset that covers both the pre-crisis and the following economically troubled periods. Topics underlying this work - and empirical results - have relevant policy implications from a managerial and regulatory point of view.

Keywords: Bank performance
Revenue Diversification
JEL Codes: G21
L25

## 1. Introduction and brief literature review

The recent financial crisis led to a severe reduction in bank profitability, after a long extended period of prosperity. This effect was caused mainly by a decrease in the quality of credit portfolio and lower level of interest rates, in a context of steady operational costs. The concurrence of these elements created great concerns in the European Monetary Union (EMU), where the traditional business of borrowing and lending is still the core activity of commercial banks. In this context, other sources of revenue could have played a smoothing effect on bank profitability, counterbalancing the worsening of credit portfolio quality.

This expected "portfolio-effect" due to revenue diversification has been a common wisdom in banking management. Goddard et al. (2007), concerning the developments observed in the European banking system since the mid-eighties, underline that banks' answers to the changing competitive environment have usually included several key strategies: diversification, product differentiation and consolidation. However, the empirical literature has found mixed results about the effect of revenue diversification on banks' risks and profitability.

Adverse relationship between diversification and risk is found in Boyd and Graham (1986), Demsetz and Strahan (1997), De Young and Roland (2001) and Lepetit et al. (2008). Numerous papers (De Young and Rice 2004a, Stiroh 2004, Acharya et al. 2006, Hayden et
al. 2007, Mercieca et al. 2007, Berger et al. 2010a, Berger et al. 2010b) find that diversification fails to produce a greater performance. On the contrary, reduced insolvency risk and improved profitability are found to be the outcome of diversification strategies by Al-Obaidan (2010) and Sanya and Wolfe (2011). Similar results are found by Nguyen et al. (2012) and Mergaerts and Vander Vennet (2016). Other empirical works find an ambiguous effect of revenue diversification on banking profits and risks. De Young and Rice (2004b), analysing U.S. banks during the period 1986-2003, state that diversification gains from fee-based activities appear to be scarce; fee-income contributes to boosting bank earnings but also increases earnings volatility. Baele et al. (2007), analysing a panel data of banks over the period 1989-2004, find that a higher share of non-interest income positively affects bank' franchise values, but increases their systematic risk. Busch and Kick (2009) show that risk-adjusted returns on equity and total assets are positively associated with higher fee-income activities, using a panel of German banks during the period 1995-2007; however, a strong engagement in fee-generating activities goes along with higher risk. On the same line Calmes and Liu (2009) find that non-interest income has driven the variance of Canadian banks' aggregate operating income growth: the authors conclude that, by contributing to banking income volatility, market-oriented activities do not necessarily yield diversification benefits to banks.

Some papers underline that the effect of revenue diversification on banks' profits and risks may be influenced by the size of the bank. Chiorazzo et al. (2008), analysing Italian banks during the period 1993-2003, find that income diversification increases risk-adjusted returns. This relation is stronger at large banks; moreover the authors underline that there are limits to diversification gains as banks get larger. Studying U.S. credit unions for the period 1993-2004, Goddard et al. (2008) find that similar diversification strategies are not appropriate for large and small credit unions.

These ambiguous results in literature references suggest that revenue diversification seems to have a "dark side" and that its smoothing effect on profits and risk is not guaranteed. Stiroh and Rumble (2006) elegantly introduce this point arguing that the volatile pattern of non-interest income can offset the benefits linked to a "diversified-portfolio effect". In brief, the (adverse) variance effect may counterbalance the (positive) correlation effect. Under this assumption, the net influence of revenue diversification on bank performances is ambiguous and needs to be empirically determined. To this extent, our work explores the relationship between the degree of revenue diversification and the risk-return profile of a sample of commercial, saving and cooperative banks headquartered in 8 EMU countries for the period 2005-2013. In particular, we aim to test the attitude of revenue diversification in enhancing bank profits level and reducing their volatility. We contribute to previous literature using a
cross-country balanced dataset that covers both the pre-crisis and the following economically troubled periods. The presence of a common regulatory framework in EMU limits the heterogeneity that is typical of this kind of approach. The wide time-span under investigation allows us to explore the relationship between revenue diversification and bank profitability over the whole recession and under specific macroeconomic conditions (huge falls in GDP, very low-interest rates and inflation, etc.). Topics underlying this work have relevant policy implications from a managerial and regulatory point of view. In effect, revenue diversification largely depends on the bank business model, which is also one of the five high-level priorities of the Single Supervisory Mechanism.

## 2. Hypotheses and data

We use a cross-country dataset that includes firm specific (source Bvd Bankscope database) and macroeconomic information (sources Eurostat and BIS) for the period 2005-2013 (see Table 1) to explore the relationship between income diversification and the risk-return profile of banks activity.

Table 1 - Variables definition and sources

| Variable Name | Variable Description | Source |
| :--- | :--- | :--- |
| ROAE | Return on Average Equity | Bankscope |
| ROAA | Return on Average Assets | Bankscope |
| $\sigma$ ROAE | Return on Average Equity volatility | Bankscope |
| $\sigma$ ROAE | Return on Average Assets volatility | Bankscope |
| RAROAE | Risk Adjusted Return on Average <br> Equity (ROAE/бROAE) | Bankscope |
| RAROAA | Risk Adjusted Return on Average <br> Assets (ROAA/ $\sigma$ ROAA) | Bankscope |
| NONSH | Non-interest revenues on Operating <br> Income | Bankscope |
| HHI | 1 - Herfindahl-Hirschman index <br> (built on NONSH) | Bankscope |
| ASSETS | Natural logarithm of total assets | Bankscope |
| KA | Equity to total assets | Bankscope |
| LOANSH | Loans to total assets | Bankscope |
| DEPsH | Customer Deposits to total funding <br> (excluding derivatives) | Bankscope |
| COSTINCOME | Cost income ratio | Bankscope |
| LLP | Loan loss provisions to loans | Bankscope |
| GDP | GDP at market prices. Annual rate of <br> change | Eurostat |
| HICP | All-items HICP. Annual average rate <br> of change | Eurostat |

Our balanced panel comprises 110 commercial, saving and cooperative banks from 8 Euro area countries ${ }^{1}$; this latter choice allows us to limit the drawbacks usually linked to cross country

[^0]analysis, given the homogeneous regulatory framework that characterizes this area. We consider only banks with a total asset greater than 10 bln Euros in 2013 and with a complete set of balance sheet and income statement information along the whole selected sample period.

The time span under investigation includes several years of harsh economic and financial crisis; we explore the effect of this deep recession on the link between income diversification and bank profitability adopting different approaches in computing the values assumed by our variables.

More specifically, the analysis is conducted through two different set-ups. Following Stiroh and Rumble (2006) we start calculating the mean value of each variable over the whole period covered by our dataset, running a cross-section regression on the resulting data. Then, we employ a fixed effect regression on the raw annual data to explore the time dimension of our panel; in this case, since infraannual data are not available to calculate volatility measures, we can only explore the level of ROAE and ROAA as dependent variables. Equation 1 reports the baseline econometric model used to estimate the effects of revenue diversification on banks' economic results.

$$
\begin{gather*}
\Pi_{i}=c+\sum_{b=1}^{B} \beta_{j} X_{i}^{b}+\sum_{m=1}^{M} \beta_{m} X_{i}^{m}+\sum_{d=1}^{D} \beta_{d} X_{i}^{d} \\
+\varepsilon_{i} \tag{1}
\end{gather*}
$$

where $X_{i}^{b}$ is a vector of bank-specific information, $X_{i}^{m}$ a vector of macroeconomic indicators, $X_{i}^{d}$ identifies a set of bank type and time dummies, $c$ is the constant term and $\varepsilon_{\mathrm{i}}$ is the error term.

Since we are interested both in the level and volatility of bank profitability measures, we use six different dependent variables $\left(\Pi_{i}\right)$, defined as follows:

| ROAE | Return on average equity |
| :--- | :--- |
| ROAA | Return on average assets |
| $\sigma R O A E$ | Standard deviation of Return on average equity |
| $\sigma R O A A$ | Standard deviation of Return on average assets |
| RAROAE | Risk adjusted Return on average equity |
| (Roae/ $\sigma R o a e)$ |  |
| RAROAA | Risk adjusted Return on average assets (Roaa/סRoaa) |

These measure of bank profitability are widely used in literature. Typically ROAE is more volatile then ROAA; moreover, the former measure is more influenced by the bank leverage. For these reasons, the two variables are used to crosscheck - under different points of view - the effect of selected items on bank profitability. As previously reported, a common wisdom among banking practitioners ascribes an expected virtuous portfolio-effect to income diversification; in this sense, it is relevant to observe not only the
level of bank profitability but also its standard deviation over time. This has led to the widespread use of measures of revenue volatility and risk adjusted return as dependent variables (Stiroh and Rumble, 2006; Mercieca et al. 2007; Goddard et al., 2008).

According to these same latter references, we calculate two key explanatory variables to account for the level of income diversification: NON $_{\text {SH }}$ and HHI . NON ${ }_{\text {SH }}$ measures the share of net operating income represented by non-interest revenues; to prevent potential outliers, we drop banks that show value of this variable outside the $0-1$ range. Low levels of $\mathrm{NON}_{\mathrm{SH}}$ suggest the prevalence of a "traditional activity" carried out by the bank, through borrowing and lending money; this is the typical shape of the commercial banking business. In this sense, a greater share of non-interest income signals an income diversification strategy; however, over a certain threshold, this source of revenue tends to prevail (as occurs, for example, in corporate banking).

The second measure of income diversification, HHI, accounts for this issue. This variable is built on the Herfindahl-Hirschman Index approach and measures the level of revenue diversification in the composition of net operating income: it is calculated as shown by equation 2 :

$$
\begin{equation*}
\mathrm{HHI}=1-\left[(\mathrm{NONSH})^{2}+(1-\mathrm{NONSH})^{2}\right] \tag{2}
\end{equation*}
$$

By construction, the variable assumes values between 0 and 0.5 ; the minimum value is associated to banks that exhibit a single source of operating revenues (i.e. a maximum level of income concentration). The maximum value of the variable is reached when there is an equal contribution of interest and non-interest revenues in total operating income (i.e. a maximum level of income diversification).

Besides these two measures, our set of independent variables includes several information from banks' balance sheet and income statement. Firstly, we account for bank dimension using the natural logarithm of Total Assets (ASSETS): since the recent crisis period has severely affected banks economic results, we want to explore through this variable the existence of a "size effect" on bank profitability. To control for different levels of leverage, we include the ratio between Equity and Total Assets (KA): typically, higher levels of this variable are associated with lower ROAE, but signal a greater resilience capacity of the bank in troubled periods. Our work focuses on the traditional "commercial banking model", based on gathering and borrowing money; we observe the orientation towards this model through two measures (the ratio between Loans and Total Assets and the share of Customer Deposits to Total Funding respectively LOAN ${ }_{\text {SH }}$ and $\mathrm{DEP}_{\mathrm{SH}}$ ).

With regard to income statement items, we include two variables that account for banks efficiency and credit portfolio quality (namely Cost Income ratio - COSTINCOME - and Loan Loss Provisions to

Loans - LLP). The long-lasting period of crisis, widespread over different countries and economic sectors, has boosted the deterioration of credit portfolios: this has led to a severe reduction in bank profitability. In effect, loan loss provisions are likely to be one of the most important factors affecting the risk-return figures shown by European banks during the last years. Given this pressure on bank profitability, efficiency has renewed its importance in bankers and regulators concerns.

We account for the surrounding environment through two macroeconomic variables (this excludes country dummy variables in our econometric estimations). The advent of the crisis has led to remarkable effects on GDP dynamics and inflation levels; in particular, during the period under observation, both these variables has recurrently experimented negative values. We then include for each county the annual rate of change of GDP at market prices and the annual average rate of change of HICP as explanatory variables. Inflation plays a relevant role in influencing the level and dynamics of interest rates, but shows different cross-country patterns; this makes this variable preferable than other ones, like the Euribor or the key interest rates set by the ECB.

Finally, we introduce two dummies (COOP for cooperatives and SAV for saving banks) that control for specific bank type features and one dummy for the crisis period (years 2009-2013) ${ }^{2}$.

Tables 2-5 summarize the descriptive statistics of the variables included in our dataset; these figures (including minimum and maximum values) vary in the tables because of the different time span used to evaluate means and standard deviations.

Table 2 - Descriptive statistics (whole period, mean values)

| Variable | Observations | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| ROAE | 110 | 5.886 | 6.703 | -19.341 | 39.777 |
| ROAA | 110 | 0.418 | 0.345 | -0.677 | 1.560 |
| $\boldsymbol{\sigma R O A E}$ | 110 | 5.455 | 6.827 | 0.237 | 40.677 |
| $\boldsymbol{\sigma R O A A}$ | 110 | 0.310 | 0.331 | 0.016 | 2.219 |
| RAROAE | 110 | 2.710 | 2.586 | -0.535 | 13.666 |
| RAROAA | 110 | 3.021 | 3.250 | -0.513 | 17.923 |
| NON | 110 | 37.239 | 8.832 | 14.471 | 64.830 |
| HHI | 110 | 0.444 | 0.052 | 0.245 | 0.497 |
| ASSETS | 110 | 16.929 | 1.157 | 14.837 | 21.217 |
| KA | 110 | 7.371 | 2.690 | 1.992 | 14.436 |
| LOAN | 110 | 61.967 | 17.835 | 12.475 | 86.696 |
| DEP | 110 | 57.090 | 19.509 | 11.393 | 94.635 |
| COSTINCOME | 110 | 62.839 | 10.595 | 32.493 | 90.672 |
| LLP | 110 | 0.505 | 0.364 | -0.311 | 1.778 |

[^1]| GDP | 110 | 0.293 | 0.720 | -0.970 | 1.380 |
| :--- | ---: | ---: | ---: | ---: | ---: |
| HICP | 110 | 1.975 | 0.261 | 1.820 | 2.790 |
| COOP | 110 | 0.373 | 0.486 | 0.000 | 1.000 |
| SAV | 110 | 0.282 | 0.452 | 0.000 | 1.000 |

Table 3 - Descriptive statistics (pre-crisis period, mean values)

| Variable | Observations | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| ROAE | 110 | 9.015 | 7.504 | -8.413 | 52.495 |
| ROAA | 110 | 0.605 | 0.392 | -0.298 | 2.355 |
| $\boldsymbol{\sigma R O A E}$ | 110 | 3.430 | 4.186 | 0.196 | 23.932 |
| $\boldsymbol{\sigma R O A A}$ | 110 | 0.184 | 0.145 | 0.013 | 0.689 |
| RAROAE | 110 | 5.912 | 5.802 | -0.442 | 29.511 |
| RAROAA | 110 | 5.763 | 5.630 | -0.450 | 25.232 |
| NON | SH | 110 | 36.798 | 9.434 | 11.155 |
| HHI | 110 | 0.440 | 0.057 | 0.198 | 67.428 |
| ASSETS | 110 | 16.770 | 1.184 | 13.941 | 21.252 |
| KA | 110 | 7.115 | 2.951 | 1.473 | 14.228 |
| LOAN | SH | 110 | 61.380 | 19.146 | 12.231 |
| DEP | 87.332 |  |  |  |  |
| COSTINCOME | 110 | 55.340 | 20.459 | 8.253 | 91.885 |
| LLP | 110 | 62.720 | 11.209 | 32.535 | 81.955 |
| GDP | 110 | 0.436 | 0.361 | -0.125 | 2.383 |
| HICP | 110 | 1.270 | 0.755 | 0.250 | 2.775 |
| COOP | 110 | 2.086 | 0.403 | 0.950 | 3.225 |
| SAV | 110 | 0.373 | 0.486 | 0.000 | 1.000 |

Table 4 - Descriptive statistics (crisis period, mean values)

| Variable | Observations | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| ROAE | 110 | 3.767 | 7.991 | -38.570 | 33.140 |
| ROAA | 110 | 0.292 | 0.405 | -1.383 | 1.092 |
| бROAE | 110 | 4.453 | 7.393 | 0.125 | 43.173 |
| бROAA | 110 | 0.253 | 0.382 | 0.000 | 2.695 |
| RAROAE | 110 | 4.157 | 4.712 | -0.893 | 29.230 |
| RAROAA | 110 | 5.712 | 8.786 | -0.967 | 60.020 |
| NON | 110 | 37.545 | 9.392 | 15.156 | 64.504 |
| HHI | 110 | 0.446 | 0.056 | 0.255 | 0.499 |
| ASSETS | 110 | 17.035 | 1.151 | 15.433 | 21.286 |
| KA | 110 | 7.546 | 2.673 | 2.338 | 14.682 |
| LOAN | 110 | 62.360 | 17.728 | 12.638 | 90.598 |
| DEP | 110 | 58.251 | 19.653 | 11.200 | 96.468 |
| COSTINCOME | 110 | 62.916 | 11.314 | 32.465 | 96.660 |
| LLP | 110 | 0.554 | 0.553 | -0.848 | 2.553 |
| GDP | 110 | -0.358 | 0.813 | -1.783 | 0.683 |
| HICP | 110 | 1.901 | 0.256 | 1.667 | 2.650 |
| COOP | 110 | 0.373 | 0.486 | 0.000 | 1.000 |

Table 5 - Descriptive statistics (annual data)

| Variable | Observations | Mean | Std. Dev. | Min | Max |
| :--- | ---: | ---: | ---: | ---: | ---: |
| ROAE | 990 | 6.218 | 9.959 | -84.760 | 83.590 |
| ROAA | 990 | 0.434 | 0.536 | -5.880 | 3.120 |
| NON $_{\text {SH }}$ | 990 | 37.189 | 11.048 | 2.141 | 84.178 |
| HHI | 990 | 0.443 | 0.068 | 0.042 | 0.500 |
| ASSETS | 990 | 16.917 | 1.182 | 13.686 | 21.533 |
| KA | 990 | 7.330 | 2.911 | 0.932 | 16.830 |
| LOAN | SH | 990 | 62.007 | 18.816 | 8.293 |
| DEP | 93.155 |  |  |  |  |
| COSTINCOME | 990 | 56.528 | 20.439 | 5.270 | 100.000 |
| LLP | 990 | 62.779 | 13.555 | 26.960 | 145.120 |
| GDP | 990 | 0.499 | 0.641 | -5.741 | 4.611 |
| HICP | 990 | 0.312 | 2.325 | -8.700 | 6.600 |
| COOP | 990 | 2.055 | 0.886 | -0.900 | 4.100 |
| SAV | 990 | 0.373 | 0.484 | 0.000 | 1.000 |

## 3. Econometric estimations

Table 6 shows the empirical results for the cross-section regression over the whole period under investigation. The first evidence that emerges is the weak significance of the coefficients associated to income diversification; only the share of net operating income represented by non-interest revenues shows positive effects on the level of ROAE and ROAA. Other variables show a greater statistical significance over economic results; for example, the level of capitalization (KA) has a negative and significant impact on ROAE
and a positive one on ROAA. This mixed result is explained by the different nature of the two measures of profitability; being the equity the denominator of ROAE, higher levels of capitalization tends to reduce (coeteris paribus) ROAE. The coefficients associated to the traditional features of commercial banking model (LOAN ${ }_{\text {SH }}$ and $\left.\mathrm{DEP}_{\mathrm{SH}}\right)$ are positive, while not always significant. During the period under examination, a greater ability to gather money from retail customer (and consequently a lower debt exposures on the interbank market) may have reduced funding costs for banks, enhancing profitability; this has been probably due to a weaker perception of the risk-return profile of deposits on behalf of retail customers. Unsurprisingly, the cost-income ratio and the level of loan loss provisions have a significant and negative effect on bank profitability: high efficiency and good credit portfolio quality confirm to be key drivers of success in banking. These latter elements have a high importance also in determining bank revenue volatility. The size of the bank shows a positive (and feebly significant) effect on $\sigma$ ROAE and $\sigma$ ROAA, while LOAN $_{\text {SH }}$ and $\mathrm{DEP}_{\mathrm{SH}}$ lose their econometric relevance. Interestingly, macroeconomic indicators seem to affect more the volatility than the level of bank profitability. The negative coefficient associated to inflation indicates that the variability of banks profits has been higher in countries that have experimented lower levels of HICP; it must be noted that during the period under examination inflation has
sometimes reached negative values, leading some countries in a painful deflation.

In a medium-term perspective, the risk-adjusted return should be the most important item to be considered in assessing banks health profile. The outcomes of our regressions show that three main drivers explain RAROAE and RAROAA figures: good efficiency, sound portfolio quality and economic growth.

The coefficients associated to the dummies that capture the specific bank-type features (and in particular the one that identifies cooperative banks) show a peculiar pattern across the regressions. Cooperative banks have lower profitability levels but, at the same time, lower volatility of economic results during the period under investigation; the combined effect of these two elements is a better risk-adjusted return that, with regard to RAROAE, is also strongly significant. In order to further explore the impact of bank type over the topic of interest, the sample has been split in two subsamples (Cooperative vs. non-cooperative banks): results are reported in Tables 6a and 6b.

With regard to the effects of revenue diversification on banks profitability, these two sets of regressions show a different pattern of coefficients. Cooperative banks seem to enjoy a more diversified portfolio of activities; a higher degree of diversification increases ROAE and ROAA and reduces profit volatility. Non-cooperative banks, on the contrary, show a positive coefficient for $\mathrm{NON}_{\text {SH }}$ and a
negative one for HHI ; this indicates that an increase in alternative sources of revenue is beneficial, but also that it is fundamental the existence of a core-business activity. The sign and statistical significance of the other coefficients are mainly unchanged.

Since we aim to identify the effects of revenue diversification on bank profitability, the use of mean values during the whole period 2005-2013 may prevent us to appreciate the distinctive features of bank-specific and macroeconomic information before and during the recent crisis. For this reason, we split the time span under investigation into two sub-periods: pre-crisis period (2005-2008) and crisis period (2009-2013). Table 7 reports the results of the regressions.

Overall, previous findings are confirmed; it emerges more clearly the different impact of selected variables on bank profitability across time. In particular, during the crisis, it becomes crucial the role played by the quality of the bank credit portfolio; results confirm that this variable was not previously an element of particular concern and, hence, that its relevance in explaining bank profits pattern was widely lower. At the same time, during the crisis, a higher level of efficiency has reduced the volatility of economic results for the banks; a wise cost management helps to maintain a stable level of profitability also during a troubled period.

With regard to macroeconomic conditions, the coefficients associated with GDP growth change sign before and during the crisis
in ROAE and ROAA regressions. These outcomes suggest that, before 2009 , banks profitability was experimenting a countercyclical pattern; however, the advent of a severe financial and economic shock has seriously wounded the soundness of banks income statements, increasing the correlation between profitability and surrounding economic environment.

In order to explore the time dimension of our panel of data, we run a fixed effect panel regression on the yearly-collected observations ${ }^{3}$ (Equation 3).

$$
\begin{gather*}
\Pi_{i t}=c+\sum_{b=1}^{B} \beta_{j} X_{i t}^{b}+\sum_{m=1}^{M} \beta_{m} X_{i t}^{m}+\sum_{d=1}^{D} \beta_{d} X_{i t}^{d} \\
+\varepsilon_{i t} \tag{3}
\end{gather*}
$$

where $X_{i t}^{b}, X_{i t}^{m}, X_{i t}^{d}, c$ and $\boldsymbol{\varepsilon}_{\mathbf{i}}$ assumes the same meanings previously described.

Regarding the model specification, the advent of the crisis has produced a more scattered pattern of financial results in the banking system, reducing the serial recurrence of bank profits over time; this makes less appropriate the adoption of dynamic models recently employed by literature (Goddard et al. 2011).

[^2]Using annual data, we can't estimate standard deviation and RAR figures ${ }^{4}$; hence, the analysis is limited to ROAE and ROAA. Given the observed differences between cooperative and non-cooperative banks, we have tested our model on the whole sample and on two subsamples according with the bank type (Table 8).

Results confirm that efficiency and quality of loans are the main determinants of bank profitability during the period under examination; the crisis has naturally contributed to cut the level of ROAE and ROAA. With regard to the variables of main interest in our work $\left(\mathrm{NON}_{\text {SH }}\right.$ and HHI), we observe a pattern of results that is really different from the content of Table 6 . The level of non-interest income shows a weak statistical significance and a changing sign in the regressions focused on cooperative and non-cooperative banks, while HHI fails to obtain a sufficient p -value.

Bearing in mind that Table 6 and Table 8 show respectively a static/average and a dynamic view of the link between income diversification and economic results of the banks included in our sample, we can coordinate the results as follows.

In a long-term perspective, a greater average level of non-interest income can improve bank profitability; however, the expansion of this source of revenue - in particular during crisis period - may produce adverse effects. This results corroborate the idea that

[^3]diversification should be a strategy, and not a mere tactic, in bank management.

Table 6 - Regression results (whole period)

|  | $(1)$ | $(2)$ | $(3)$ | $(4)$ | $(5)$ | $(6)$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| VARIABLES | ROAE | ROAA | $\boldsymbol{\sigma R O A E}$ | $\boldsymbol{\sigma R O A A}$ | RAROAE | RAROAA |
|  |  |  |  |  |  |  |
| NON |  |  |  |  |  |  |
|  | $0.23^{*}$ | $0.01^{*}$ | -0.07 | -0.00 | 0.02 | 0.06 |
| HHI | $(0.125)$ | $(0.005)$ | $(0.122)$ | $(0.006)$ | $(0.050)$ | $(0.063)$ |
|  | -8.23 | -0.11 | 2.52 | -0.28 | 5.52 | 4.87 |
| ASSETS | $(20.226)$ | $(0.879)$ | $(19.652)$ | $(0.912)$ | $(8.077)$ | $(10.233)$ |
|  | -0.21 | -0.00 | 0.76 | 0.04 | $-0.36^{*}$ | -0.40 |
| KA | $(0.532)$ | $(0.023)$ | $(0.517)$ | $(0.024)$ | $(0.213)$ | $(0.269)$ |
|  | $-0.81^{* * *}$ | $0.03^{* *}$ | -0.07 | $0.03^{* * *}$ | 0.00 | $0.23^{*}$ |
| LOANSH | $(0.266)$ | $(0.012)$ | $(0.258)$ | $(0.012)$ | $(0.106)$ | $(0.135)$ |
|  | $0.07^{*}$ | 0.00 | 0.04 | 0.00 | -0.00 | 0.01 |
| DEPSH | $(0.036)$ | $(0.002)$ | $(0.035)$ | $(0.002)$ | $(0.014)$ | $(0.018)$ |
|  | $0.09^{* * *}$ | $0.00^{* *}$ | -0.00 | 0.00 | -0.01 | 0.00 |
| COSTINCOME | $(0.032)$ | $(0.001)$ | $(0.031)$ | $(0.001)$ | $(0.013)$ | $(0.016)$ |
|  | $-0.44^{* * *}$ | $-0.02^{* * *}$ | $0.13^{* *}$ | $0.01^{* *}$ | $-0.06^{* *}$ | $-0.08^{* * *}$ |
| LLP | $(0.058)$ | $(0.003)$ | $(0.056)$ | $(0.003)$ | $(0.023)$ | $(0.029)$ |
|  | $-5.38^{* * *}$ | $-0.24^{* * *}$ | $8.04^{* * *}$ | $0.41^{* * *}$ | $-2.78^{* * *}$ | $-2.03^{* *}$ |
| GDP | $(1.766)$ | $(0.077)$ | $(1.716)$ | $(0.080)$ | $(0.705)$ | $(0.894)$ |
|  | 0.15 | 0.03 | -1.05 | $-0.12^{* *}$ | $0.87^{* *}$ | $1.41^{* * *}$ |
| HICP | $(0.988)$ | $(0.043)$ | $(0.960)$ | $(0.045)$ | $(0.395)$ | $(0.500)$ |
|  | $-4.85^{*}$ | $-0.20^{*}$ | $-4.53^{*}$ | -0.15 | $2.31^{* *}$ | 0.05 |
| COOP | $(2.785)$ | $(0.121)$ | $(2.706)$ | $(0.126)$ | $(1.112)$ | $(1.409)$ |
|  | $-3.12^{* *}$ | $-0.13^{*}$ | $-6.14^{* * *}$ | $-0.29^{* * *}$ | $1.67^{* * *}$ | 0.79 |
| SAV | $(1.563)$ | $(0.068)$ | $(1.519)$ | $(0.070)$ | $(0.624)$ | $(0.791)$ |
|  | -2.74 | -0.12 | $-5.34^{* * *}$ | $-0.20^{* *}$ | 0.35 | 0.13 |
| Constant | $(1.733)$ | $(0.075)$ | $(1.684)$ | $(0.078)$ | $(0.692)$ | $(0.877)$ |
|  | $42.82^{* * *}$ | $1.66^{* * *}$ | -6.78 | -0.46 | 6.31 | 8.04 |
|  | $(14.039)$ | $(0.610)$ | $(13.641)$ | $(0.633)$ | $(5.607)$ | $(7.103)$ |
| Observations |  | $\mathbf{1 1 0}$ | $\mathbf{1 1 0}$ |  | $\mathbf{1 1 0}$ | $\mathbf{1 1 0}$ |
| R-squared | $\mathbf{0 . 5 1}$ | $\mathbf{0 . 6 5}$ | $\mathbf{0 . 5 5}$ | $\mathbf{0 . 5 9}$ | $\mathbf{1 1 0}$ | $\mathbf{0 . 4 7}$ |
| Adj. R-squared | $\mathbf{0 . 4 0}$ | $\mathbf{0 . 4 0}$ | $\mathbf{0 . 4 0}$ | $\mathbf{0 . 4 0}$ | $\mathbf{0 . 4 0}$ | $\mathbf{1 1 0}$ |
|  |  |  |  | $\mathbf{0 . 4 6}$ |  |  |

Level of statistical significance: ${ }^{* * *} \mathrm{p}<0.01,{ }^{* *} \mathrm{p}<0.05, * \mathrm{p}<0.10$. Standard errors in italics.

Table 6a - Regression results (whole period): cooperative banks

|  | (1) | (2) | (3) | (4) | (5) | (6) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| VARIABLES | ROAE | ROAA | бROAE | $\boldsymbol{\sigma R O A A}$ | RAROAE | RAROAA |
| NONSH | -0.00 | -0.00 | 0.05 | 0.00 | -0.03 | 0.05 |
|  | (0.054) | (0.005) | (0.052) | (0.004) | (0.086) | (0.138) |
| HHI | 23.47** | 2.31** | -18.25** | -0.97 | 25.11* | 17.08 |
|  | (9.075) | (0.860) | (8.893) | (0.650) | (14.596) | (23.396) |
| ASSETS | -0.64** | -0.05 | 1.28*** | 0.08*** | -1.21** | -1.03 |
|  | (0.311) | (0.029) | (0.305) | (0.022) | (0.500) | (0.802) |
| KA | -0.53*** | 0.01 | 0.11 | 0.03** | -0.41* | -0.29 |
|  | (0.149) | (0.014) | (0.146) | (0.011) | (0.240) | (0.384) |
| LOANSH | -0.03 | -0.00 | 0.09*** | 0.00*** | -0.00 | 0.04 |
|  | (0.023) | (0.002) | (0.022) | (0.002) | (0.036) | (0.058) |
| $\mathrm{DEP}_{\text {SH }}$ | 0.01 | 0.00 | -0.01 | -0.00 | -0.02 | -0.03 |
|  | (0.015) | (0.001) | (0.015) | (0.001) | (0.024) | (0.039) |
| COSTINCOME | -0.15*** | -0.01*** | 0.04 | 0.00 | 0.05 | 0.04 |
|  | (0.044) | (0.004) | (0.044) | (0.003) | (0.072) | (0.115) |
| LLP | -7.77*** | -0.52*** | 7.84*** | 0.36*** | -5.50** | -4.68 |
|  | (1.413) | (0.134) | (1.385) | (0.101) | (2.273) | (3.644) |
| GDP | -0.65 | -0.03 | 0.01 | 0.01 | 0.16 | 0.45 |
|  | (0.745) | (0.071) | (0.731) | (0.053) | (1.199) | (1.922) |
| HICP | 3.54 | 0.23 | -0.91 | 0.45* | 0.38 | -0.87 |
|  | (3.592) | (0.340) | (3.520) | (0.257) | (5.777) | (9.260) |
| Constant | 17.81 | 0.82 | -23.38** | -2.65*** | 17.68 | 14.33 |
|  | (10.903) | (1.033) | (10.685) | (0.780) | (17.537) | (28.110) |
| Observations | 41 | 41 | 41 | 41 | 41 | 41 |
| R-squared | 0.77 | 0.82 | 0.80 | 0.78 | 0.54 | 0.42 |
| Adj. R-squared | 0.39 | 0.39 | 0.39 | 0.39 | 0.39 | 0.39 |

Level of statistical significance: ${ }^{* * *} \mathrm{p}<0.01, * * \mathrm{p}<0.05, * \mathrm{p}<0.10$. Standard errors in italics.

Table 6b-Regression results (whole period): non-cooperative banks

|  | $(1)$ | $(2)$ | $(3)$ | $(4)$ | $(5)$ | $(6)$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| VARIABLES | $\mathbf{R O A E}$ | ROAA | $\boldsymbol{\sigma R O A E}$ | $\boldsymbol{\sigma R O A A}$ | RAROAE | RAROAA |
|  |  |  |  |  |  |  |
| NON |  |  |  | -0.05 | 0.02 |  |
|  | $0.70^{* * *}$ | $0.03^{* * *}$ | 0.01 | 0.00 | $(0.010)$ | $(0.056)$ |
| HHI | $(0.198)$ | $(0.008)$ | $(0.219)$ | $(0.010)$ |  |  |
|  | $-97.42^{* *}$ | $-4.40^{* * *}$ | -23.81 | -0.37 | -0.10 | 1.33 |
| ASSETS | $(37.474)$ | $(1.577)$ | $(41.478)$ | $(1.891)$ | $(10.630)$ | $(11.152)$ |
|  | 0.61 | 0.03 | 1.11 | 0.05 | -0.29 | -0.27 |
| KA | $(0.703)$ | $(0.030)$ | $(0.778)$ | $(0.035)$ | $(0.199)$ | $(0.209)$ |
|  | $-0.98^{*}$ | 0.01 | -0.86 | 0.04 | -0.21 | 0.13 |
| LOANSH | $(0.509)$ | $(0.021)$ | $(0.563)$ | $(0.026)$ | $(0.144)$ | $(0.151)$ |
|  | $0.18^{* * *}$ | $0.01^{* *}$ | 0.02 | 0.00 | $-0.03^{*}$ | -0.00 |
| DEPSH | $(0.052)$ | $(0.002)$ | $(0.058)$ | $(0.003)$ | $(0.015)$ | $(0.015)$ |
|  | $0.16^{* * *}$ | $0.01^{* *}$ | 0.02 | -0.00 | -0.02 | 0.01 |
| COSTINCOME | $(0.058)$ | $(0.002)$ | $(0.065)$ | $(0.003)$ | $(0.017)$ | $(0.017)$ |
|  | $-0.53^{* * *}$ | $-0.03^{* * *}$ | 0.13 | 0.01 | $-0.07^{* * *}$ | $-0.09^{* * *}$ |
| LLP | $(0.072)$ | $(0.003)$ | $(0.080)$ | $(0.004)$ | $(0.020)$ | $(0.021)$ |
|  | $-7.09^{* * *}$ | $-0.26^{* * *}$ | $10.61^{* * *}$ | $0.50^{* * *}$ | $-2.49^{* * *}$ | $-1.56^{* *}$ |
| GDP | $(2.290)$ | $(0.096)$ | $(2.534)$ | $(0.116)$ | $(0.650)$ | $(0.681)$ |
|  | -1.12 | -0.07 | $-3.63^{*}$ | -0.14 | 0.15 | $0.93^{*}$ |
| HICP | $(1.679)$ | $(0.071)$ | $(1.858)$ | $(0.085)$ | $(0.476)$ | $(0.500)$ |
|  | $-8.56^{* *}$ | $-0.39^{* *}$ | -4.43 | -0.11 | 1.61 | -0.62 |
| Constant | $(3.496)$ | $(0.147)$ | $(3.869)$ | $(0.176)$ | $(0.992)$ | $(1.040)$ |
|  | $54.97^{* *}$ | $2.72^{* * *}$ | -2.98 | -0.99 | $14.93^{* *}$ | $11.87^{*}$ |
|  | $(21.648)$ | $(0.911)$ | $(23.961)$ | $(1.092)$ | $(6.141)$ | $(6.442)$ |
| Observations | $\mathbf{6 9}$ | $\mathbf{6 9}$ | $\mathbf{6 9}$ | $\mathbf{6 9}$ | $\mathbf{6 9}$ | $\mathbf{6 9}$ |
| R-squared | $\mathbf{0 . 6 0}$ | $\mathbf{0 . 6 4}$ | $\mathbf{0 . 4 6}$ | $\mathbf{0 . 5 2}$ | $\mathbf{0 . 4 8}$ | $\mathbf{0 . 4 8}$ |
| Adj. R-squared | $\mathbf{0 . 3 9}$ | $\mathbf{0 . 3 9}$ | $\mathbf{0 . 3 9}$ | $\mathbf{0 . 3 9}$ | $\mathbf{0 . 3 9}$ | $\mathbf{0 . 3 9}$ |

Level of statistical significance: ${ }^{* * *} \mathrm{p}<0.01, * * \mathrm{p}<0.05, * \mathrm{p}<0.10$. Standard errors in italics.
Table 7 - Regression results (Pre-Crisis vs Crisis)

|  | (1) | (2) | (3) | (4) | (5) | (6) | (7) | (8) | (9) | (10) | (11) | (12) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | ROAE |  | ROAA |  | $\boldsymbol{\sigma R O A E}$ |  | $\boldsymbol{\sigma R O A A}$ |  | RAROAE |  | RAROAA |  |
| VARIABLES | Pre-crisis | Crisis | Pre-crisis | Crisis | Pre-crisis | Crisis | Pre-crisis | Crisis | Pre-crisis | Crisis | Pre-crisis | Crisis |
| $\mathrm{NON}_{\text {SH }}$ | 0.22* | 0.23* | 0.01** | 0.01 | 0.04 | -0.11 | -0.00 | -0.00 | 0.07 | 0.11 | 0.05 | 0.20 |
|  | (0.121) | (0.133) | (0.005) | (0.006) | (0.072) | (0.131) | (0.003) | (0.007) | (0.108) | (0.093) | (0.106) | (0.185) |
| HHI | -11.87 | -1.38 | -0.35 | 0.41 | -5.08 | 7.72 | 0.05 | 0.06 | 4.69 | -4.56 | 2.28 | -26.75 |
|  | (18.605) | (21.324) | (0.789) | (0.927) | (11.082) | (20.923) | (0.422) | (1.097) | (16.576) | (14.959) | (16.185) | (29.627) |
| ASSETS | -0.12 | 0.06 | 0.00 | 0.02 | 0.68* | -0.15 | 0.03** | -0.01 | -0.97* | -0.45 | -0.67 | -0.46 |
|  | (0.588) | (0.615) | (0.025) | (0.027) | (0.350) | (0.604) | (0.013) | (0.032) | (0.524) | (0.432) | (0.511) | (0.855) |
| KA | -1.23*** | -0.63** | 0.03** | 0.02* | -0.41** | -0.19 | 0.01 | 0.02 | 0.12 | 0.28 | 0.29 | 0.82** |
|  | (0.278) | (0.284) | (0.012) | (0.012) | (0.166) | (0.279) | (0.006) | (0.015) | (0.248) | (0.199) | (0.242) | (0.394) |
| $\mathrm{LOAN}_{\text {SH }}$ | 0.08* | 0.07* | 0.00 | 0.00* | 0.03 | -0.00 | 0.00 | -0.00 | -0.01 | 0.03 | 0.02 | 0.10* |
|  | (0.039) | (0.039) | (0.002) | (0.002) | (0.023) | (0.039) | (0.001) | (0.002) | (0.035) | (0.028) | (0.034) | (0.055) |
| $\mathrm{DEP}_{\text {SH }}$ | 0.07** | 0.10*** | 0.00** | 0.00** | 0.02 | -0.04 | 0.00 | -0.00 | -0.02 | -0.01 | 0.01 | -0.08 |
|  | (0.032) | (0.037) | (0.001) | (0.002) | (0.019) | (0.036) | (0.001) | (0.002) | (0.028) | (0.026) | (0.028) | (0.052) |
| COSTINCOME | -0.43*** | -0.48*** | -0.02*** | -0.02*** | 0.04 | 0.21*** | 0.00 | 0.01 *** | -0.12** | -0.05 | -0.15*** | -0.03 |
|  | (0.061) | (0.059) | (0.003) | (0.003) | (0.036) | (0.058) | (0.001) | (0.003) | (0.054) | (0.042) | (0.053) | (0.083) |
| LLP | -1.64 | -5.48*** | -0.09 | -0.24*** | 1.39 | 5.36*** | 0.04 | $0.28{ }^{* * *}$ | -2.56 | -1.69 | -0.23 | -2.18 |
|  | (1.744) | (1.636) | (0.074) | (0.071) | (1.039) | (1.605) | (0.040) | (0.084) | (1.554) | (1.148) | (1.517) | (2.273) |
| GDP | -2.75*** | 2.51 | -0.11*** | 0.18*** | -0.16 | -1.08 | -0.05** | -0.08 | 0.90 | 1.89* | 0.76 | 2.97 |
|  | (0.958) | (1.530) | (0.041) | (0.066) | (0.570) | (1.501) | (0.022) | (0.079) | (0.853) | (1.073) | (0.833) | (2.125) |
| HICP | -3.54** | 1.57 | -0.15** | 0.13 | -2.72*** | -2.90 | -0.10*** | -0.03 | 1.97 | 2.64 | 1.28 | 2.89 |
|  | (1.570) | (3.443) | (0.067) | (0.150) | (0.935) | (3.379) | (0.036) | (0.177) | (1.399) | (2.416) | (1.366) | (4.784) |
| COOP | -4.02** | -2.14 | -0.22*** | -0.08 | -3.59*** | -4.07** | -0.15*** | -0.23*** | 3.93** | 2.09* | 3.07** | 1.61 |
|  | (1.697) | (1.659) | (0.072) | (0.072) | (1.011) | (1.628) | (0.038) | (0.085) | (1.512) | (1.164) | (1.476) | (2.305) |
| SAV | -1.54 | -3.20* | -0.05 | -0.17** | -3.37*** | -4.68** | -0.11** | -0.22** | 0.98 | 0.14 | 0.67 | 2.37 |
|  | (1.908) | (1.853) | (0.081) | (0.081) | (1.137) | (1.818) | (0.043) | (0.095) | (1.700) | (1.300) | (1.660) | (2.574) |
| Constant | 48.82*** | 21.65 | 1.87*** | 0.51 | -2.61 | 3.12 | -0.15 | -0.16 | 19.64 | 5.16 | 14.91 | 7.34 |
|  | (13.681) | (17.003) | (0.580) | (0.739) | (8.149) | (16.684) | (0.310) | (0.875) | (12.188) | (11.928) | (11.901) | (23.624) |
|  |  |  |  |  |  |  |  |  |  |  |  |  |
| Observations | 110 | 110 | 110 | 110 | 110 | 110 | 110 | 110 | 110 | 110 | 110 | 110 |
| R-squared | 0.50 | 0.57 | 0.67 | 0.68 | 0.43 | 0.51 | 0.31 | 0.50 | 0.33 | 0.39 | 0.32 | 0.31 |
| Adj. R-squared | 0.23 | 0.23 | 0.23 | 0.23 | 0.23 | 0.23 | 0.23 | 0.23 | 0.23 | 0.23 | 0.23 | 0.23 |

Level of statistical significance: ${ }^{* * *} \mathrm{p}<0.01,{ }^{* *} \mathrm{p}<0.05,^{*} \mathrm{p}<0.10$. Standard errors in italics.

Table 8 - Regression results (Fixed effect panel model)

|  | (1) | (2) | (3) | (4) | (5) | (6) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | ROAE |  |  | ROAA |  |  |
| VARIABLES | Whole sample | Cooperative banks | Non- cooperative banks | Whole sample | Cooperative banks | Noncooperative banks |
| NONSH | -0.11* | 0.09** | -0.15** | -0.01 | 0.00* | -0.01* |
|  | (0.058) | (0.039) | (0.073) | (0.004) | (0.003) | (0.005) |
| HHI | 4.27 | 0.12 | 7.79 | 0.71 | -0.05 | 0.92 |
|  | (8.341) | (5.515) | (10.980) | (0.520) | (0.386) | (0.679) |
| ASSETS | -0.65 | -1.83* | -0.22 | -0.02 | -0.16 | 0.03 |
|  | (2.165) | (0.972) | (2.710) | (0.099) | (0.095) | (0.124) |
| KA | -0.12 | -0.28** | 0.03 | 0.05** | 0.01 | 0.07** |
|  | (0.440) | (0.107) | (0.663) | (0.021) | (0.009) | (0.030) |
| LOAN ${ }_{\text {SH }}$ | -0.01 | 0.09** | 0.01 | 0.00 | 0.00* | 0.00 |
|  | (0.079) | (0.042) | (0.109) | (0.002) | (0.002) | (0.004) |
| $\mathrm{DEP}_{\text {SH }}$ | 0.05 | 0.05 | 0.07 | 0.00** | 0.00* | 0.00* |
|  | (0.049) | (0.032) | (0.057) | (0.002) | (0.003) | (0.003) |
| COSTINCOME | $-0.45 * * *$ | $-0.24 * * *$ | $-0.48^{* * *}$ | $-0.02 * * *$ | $-0.02 * * *$ | $-0.02 * * *$ |
|  | (0.072) | (0.043) | (0.078) | (0.003) | (0.004) | (0.004) |
| LLP | $-5.31 * * *$ | $-4.83 * * *$ | $-5.06^{* * *}$ | $-0.29 * * *$ | $-0.32 * * *$ | $-0.28^{* * *}$ |
|  | (1.365) | (1.423) | (1.528) | (0.080) | (0.059) | (0.092) |
| GDP | 0.00 | 0.05 | -0.09 | -0.01 | 0.00 | -0.01 |
|  | (0.145) | (0.184) | (0.192) | (0.009) | (0.008) | (0.011) |
| HICP | -0.44 | -0.09 | -0.78* | $-0.04^{* * *}$ | -0.01 | $-0.06^{* * *}$ |
|  | (0.272) | (0.219) | (0.419) | (0.014) | (0.011) | (0.021) |
| CRISIS | $-4.23 * * *$ | $-1.77^{* * *}$ | -5.85*** | -0.31 *** | $-0.17^{* * *}$ | $-0.39 * * *$ |
|  | (0.877) | (0.449) | (1.287) | (0.047) | (0.038) | (0.069) |
| Constant | 51.74 | 43.73** | 46.32 | 1.83 | $3.83 * *$ | 0.96 |
|  | (38.438) | (18.628) | (48.098) | (1.760) | (1.641) | (2.246) |
| Observations | 990 | 369 | 621 | 990 | 369 | 621 |
| Number of banks | 110 | 41 | 69 | 110 | 41 | 69 |
| R-squared | 0.46 | 0.67 | 0.47 | 0.53 | 0.75 | 0.52 |
| Adj. R-squared | 0.50 | 0.48 | 0.48 | 0.50 | 0.48 | 0.48 |

Level of statistical significance: ${ }^{* * *} \mathrm{p}<0.01,{ }^{* *} \mathrm{p}<0.05,{ }^{*} \mathrm{p}<0.10$. Standard errors in italics.

## 4. Conclusions

Previous literature has found mixed results studying the effect of revenue diversification on banks profitability; our study confirms this ambiguous relationship for a wide sample of large European banks during the period 2005-2013. Overall, empirical results fail to support the common wisdom, also embraced by regulatory authorities, that revenue diversification may improve the level and reduce the volatility of bank profits.

Econometric estimates indicate that efficiency and high quality of loans portfolio have been the main drivers to enhance and anchor bank economic returns, while non-interest streams of revenue and a noticeable level of business concentration often fail to provide statistical significance. Using selected subsamples, we find that revenue diversification can play a role in contributing to bank profitability; however, these results change for different bank specialization models. Our investigation indicates that there exists a significant difference between cooperative and non-cooperative banks; cooperative, that are usually more focused on traditional borrowing and lending business, can improve their performance introducing other sources of revenue. Non cooperative banks, on the contrary, have a more diversified portfolio of product and services; increasing the non-interest share of operating incomes may distract the bank from its natural "commercial-banking-model", with adverse effects on profits.

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[^0]:    ${ }^{1}$ Austria, Finland, France, Germany, Italy, Netherlands, Portugal, Spain. We initially included all the first entrant countries in EMU; however, the lack of selected information in bank balance sheet and income statement has reduced the number of nations in our dataset.

[^1]:    ${ }^{2}$ We also tested a different set-up with a dummy accounting for the period 20082013; main results are unchanged, but the statistical significance of the coefficient associated to this dummy (and the overall quality of regressions, including Rsquared) is lower.

[^2]:    ${ }^{3}$ Hausman tests support the appropriateness of the fixed effect model against the random effect specification.

[^3]:    ${ }^{4}$ In a separate attempt, we tried to overcome this problem using a moving-average approach; in particular, each observation $\mathrm{x}_{\mathrm{t}}$ was equal to the average of the observation in year $\mathrm{t}-1, \mathrm{t}$ and $\mathrm{t}+1$ (the same occurs with regard to standard deviation computation). Estimation results are available on request.

