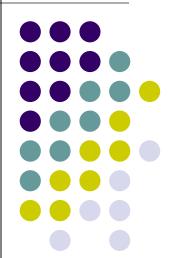
# Competition and credit procyclicality in European banking

Bank of Estonia Working Paper 9/2017

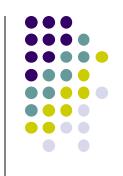
# Yannick LUCOTTE

Laboratoire d'Economie d'Orléans, France & PSB Paris School of Business

(with Aurélien LEROY, LAREFI, University of Bordeaux, France)

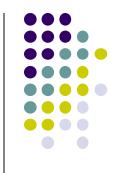


10<sup>th</sup> International Research Conference 2017 Central Bank of Sri Lanka, Colombo, 8 December, 2017

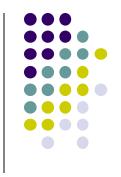


# **Agenda**

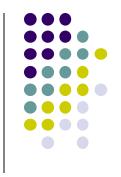
- 1) Introduction and motivation
- 2) Literature overview
- 3) Macro-level analysis
- 4) Bank-level analysis
- 5) Extensions
- 6) Concluding remarks



- Most of the existing empirical literature on the nexus between bank competition and economic outcomes focused on the link between bank competition and financial (in)stability (see, e.g., Leroy and Lucotte, 2017)
- → no consensus: "competition-fragility" vs. "competition-stability" views
- However, bank competition may also affect the real sphere by making the system more efficient, both in normal time and in response to a crisis
- → some contributions focused on the aggregate effects of bank competition on economic growth in the medium-run (see, e.g., Claessens and Laeven, 2005; de Guevara and Maudos, 2011)



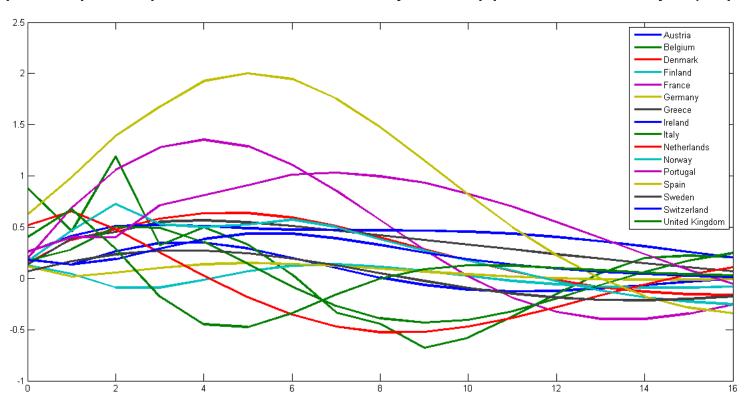
- The effects of bank competition on stability should also be considered through the global effects on macroeconomic volatility, i.e. the occurrence and intensity of economic booms and busts
- The fact that financial systems can be a source of real economic activity fluctuations is at the heart of the financial accelerator theory (Bernanke and Gertler, 1989; Kiyotaki and Moore, 1997; Bernanke et al., 1999)
- The fact that credit is more (less) expensive and has a reduced (increased) availability during a recession (expansion) tends to amplify the real economic cycle (i.e. macroeconomic volatility), due to the weakening (expanding) of investment and consumption
- No previous study has tried to empirically investigate the relationship between bank competition and credit procyclicality

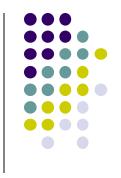


- Against this background and given the lack of empirical evidence on this issue, our study empirically investigates the relationship between bank competition and credit procyclicality for European banking
- → does bank competition reduce or exacerbate credit procyclicality?
- → does bank competition explain differences of credit procyclicality between European economies?



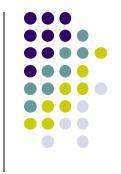
#### Country-specific impulse response functions of bank credit cycle to a 1 p.p. shock to the GDP cycle (16 quarters)





- Two levels of analysis considered:
  - Macro-level analysis: Interacted panel VAR (IPVAR), 16 European countries, 1997Q1-2014Q4
    - → impulse responses of bank credit to an activity shock are conditioned by the level of bank competition, proxied by the macro Lerner index
  - Bank-level analysis: panel framework (FE estimator), + 3600 banks, 2005-2014
  - → analyzing whether the reaction of bank loan supply to the output gap depends on the level of bank competition (i.e. the individual degree of market power): interaction term between output gap and the Lerner index
- Results suggest that credit procyclicality is higher in a low bank competitive environment:
  - → credit booms and busts are less pronounced when competition is fiercer
  - → better shock-absorbing ability of the competitive banking markets

#### Literature overview



 A large empirical literature have explored several aspects of the procyclicality of the banking sector:

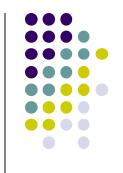
What is the impact that procyclicality has on the real economy, but also on the banking sector itself?

For example, some studies analyse the behavior of demand and supply of loans and their role in economic fluctuations (see, e.g., Lown and Morgan, 2006; Bassett et al., 2014), and the procyclical behavior of bank profitability (see, e.g., Albertazzi and Gambacorta, 2009).

# What are the factors that can contribute to strengthening or mitigating the procyclicality of the banking industry?

As discussed by Athanasoglou et al. (2014), these factors include the regulatory and supervisory framework, the monetary policy, the practices of financial firms, such as leverage and remuneration policies, and some other factors such as credit rating agencies reports or the use of automated risk management systems.

#### Literature overview



- However, only Bouvatier et al. (2012) previously investigated for a sample of OECD countries the relationship between the banking sector structure and credit procyclicality
- Bouvatier et al. (2012) proceed in two steps:
  - 1) Cluster analysis to evaluate the degree of similarity in the banking industry structures, and then split their sample of countries in different clusters
  - Estimation of a panel VAR on cyclical components for each of the clusters, and comparison of the impulse-response functions of credit to a shock in GDP
- Results that they obtain suggest that credit significantly responds to shocks to GDP, but they do not find that banking sectors with various characteristics do not exhibit differences in terms of credit procyclicality
- → Bouvatier et al. (2012) conclude that the banking sector structure is not an important cause of credit procyclicality



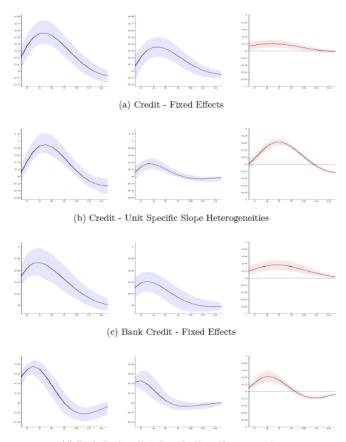
Multivariate approach: IPVAR (Towbin and Weber, 2013):

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ \alpha_{0,it}^{21} & 1 & 0 & 0 \\ \alpha_{0,it}^{31} & \alpha_{0,it}^{32} & 1 & 0 \\ \alpha_{0,it}^{41} & \alpha_{0,it}^{42} & \alpha_{0,it}^{43} & 1 \end{pmatrix} \begin{pmatrix} CPI_{i,t} \\ GDP_{i,t} \\ Cred_{i,t} \\ r_{i,t} \end{pmatrix} = \sum_{l=1}^{L} \begin{pmatrix} \alpha_{l,it}^{11} & \alpha_{l,it}^{12} & \alpha_{l,it}^{13} & \alpha_{l,it}^{14} \\ \alpha_{l,it}^{21} & \alpha_{l,it}^{22} & \alpha_{l,it}^{23} & \alpha_{l,it}^{24} \\ \alpha_{l,it}^{31} & \alpha_{l,it}^{32} & \alpha_{l,it}^{33} & \alpha_{l,it}^{34} \\ \alpha_{l,it}^{41} & \alpha_{l,it}^{42} & \alpha_{l,it}^{43} & \alpha_{l,it}^{44} \end{pmatrix} \begin{pmatrix} CPI_{i,t-l} \\ GDP_{i,t-l} \\ Cred_{i,t-l} \\ r_{i,t-l} \end{pmatrix} + \begin{pmatrix} \delta^{11}\delta^{12} \\ \delta^{11}\delta^{12} \\ \delta^{11}\delta^{12} \\ \delta^{11}\delta^{12} \end{pmatrix} \begin{pmatrix} I_{i} \\ Z_{i,t-4} \end{pmatrix} + \varepsilon_{i,t}$$

with  $\alpha_{l,it} = \beta_l + \gamma_l Z_{i,t-4}$ .

- $\beta_l$  and  $\gamma_l$  are two vectors of coefficients and  $Z_{i,t-4}$  is a cross-time-varying measure of bank competition (macro Lerner index).
- $\rightarrow \gamma_l$  is the interaction term and indicates the supplement of effect of bank competition on credit procyclicality

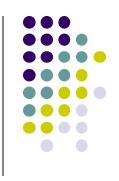
Figure 2: Impulse Response Functions of Credit to a GDP shock: Baseline model



(d) Bank Credit - Unit Specific Slope Heterogeneities

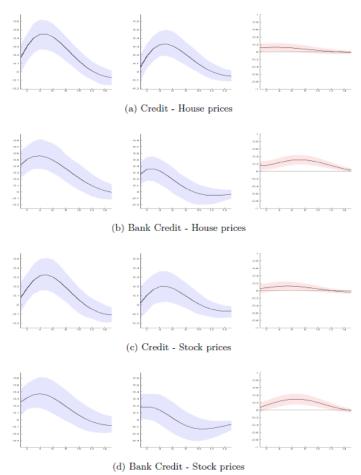
Note: The figure shows the impulse responses of credit and bank credit to a one-percentage-point shock to the output cycle evaluated (from left to right) at the  $80^{th}$  (high level) and  $20^{th}$  (low level) percentiles of the Lerner index sample distribution. The charts on the right represent the differences between the two. The colored bands represent 5% error bands (two standard deviations) generated by bootstrapping (1000 draws).





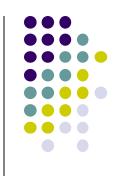
- Results robust to alternative specifications of the IPVAR model:
  - Extension of the vector of endogenous variables by including a variable reflecting the dynamics of assets prices: house price or stock price
  - Bank credit ordered before the short-term interest rate when we perform the Cholesky decomposition of the residual: the interest rate pass-through is potentially sluggish, justifying the fact that the supply and demand of credit react only with a lag to innovations in short-term interest rate
  - Different lag lengths and samples of countries: re-estimation of the model by dropping one country at a time
  - Extension of the baseline model to take into account that credit dynamics is also related to other financial characteristics. Inclusion of 3 additional interaction variables at same time: Z-score index, bank regulatory capital to risk-weighted assets ratio, and stock market capitalization to GDP ratio

Figure 4: Impulse Response Functions of Credit to a GDP shock: 5-dimensional VAR - Asset prices



Note: The figure shows the impulse responses of credit and bank credit to a one-percentage-point shock in the output cycle evaluated (from left to right) at the  $80^{th}$  (high level) and  $20^{th}$  (low level) percentiles of the Lerner index sample distribution. The charts on the right represent the difference between the two. The colored bands represent the 5% error band (two standard deviations) generated by bootstrapping (1000 draws).





 Empirical specification designed to investigate at the bank-level whether the degree of bank competition impacts the reaction of banks (in terms of supply of loans) to an output-gap shock:

$$\Delta ln(loans_{it}) = \beta_1 OG_{ct} + \beta_2 OG_{ct} * Lerner_{i,t-1/c,t-1} + \beta_3 Lerner_{i,t-1/c,t-1}$$

$$+ \sum_{j=4}^{n} \beta_j X_{j,i,t-1} + \mu_{i/c} + \lambda_t + \varepsilon_{it}$$

- Sample of more than 3600 banks: adjusted Lerner index (Koetter et al., 2012) estimated for each bank over the 2005-2014 period
- Individual control variables: log of total of assets, loans over total assets ratio, equity over total assets ratio and, also, in some specifications, the product term between the Lerner index and monetary policy shocks (Δ EONIA)



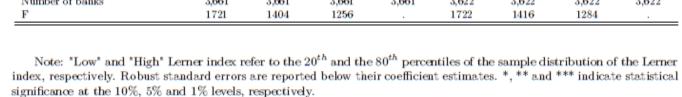


Table A1: Number of banks by country

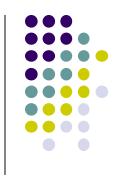
| Austria | 233 | France  | 211  | Italy    | 577 | Sweden          | 89  |
|---------|-----|---------|------|----------|-----|-----------------|-----|
| Belgium | 34  | Germany | 1711 | Norway   | 128 | Switzerland     | 356 |
| Denmark | 98  | Greece  | 16   | Portugal | 21  | the Netherlands | 23  |
| Finland | 13  | Ireland | 10   | Spain    | 126 | United Kingdom  | 90  |

Table 2: Credit procyclicality and bank competition: Bank-level measures of bank competition

|                                  | Conventionnal Lerner Index (bank level) |            |            | Efficiency-adjusted Lerner index (bank level) |           |            |            |                  |
|----------------------------------|---|------------|------------|---|-----------|------------|------------|------------------|
|                                  | (1)                                     | (2)        | (3)        | (4)   | (5)       | (6)        | (7)        | (8)              |
| Output Gap                       | 0.933***                                | 0.866***   | 0.807***   | 0.959***                                      | 1.054***  | 1.174***   | 1.100***   | 1.053**          |
|                                  | (0.128)                                 | (0.141)    | (0.138)    | (0.112)                                       | (0.110)   | (0.114)    | (0.118)    | (0.107)          |
| Lerner index                     | 5.592***                                | 6.400***   | 5.914***   | 3.751 ***                                     | 8.777***  | 7.951***   | 7.874***   | 5.941*           |
|                                  | (1.681)                                 | (1.778)    | (1.809)    | (1.281)                                       | (0.975)   | (1.037)    | (1.079)    | (0.904)          |
| Output Gap*Lemer index           | 3.493***                                | 3.527***   | 3.589***   | 3.074***                                      | 2.440***  | 1.746***   | 1.857***   | 2.204*           |
|                                  | (0.440)                                 | (0.483)    | (0.481)    | (0.385)                                       | (0.325)   | (0.308)    | (0.325)    | (0.30)           |
| Total assets                     |   | -13.660*** | -13.736*** | -0.439***                                     |           | -14.299*** | -14.369*** | $-0.250^{\circ}$ |
|                                  |   | (1.527)    | (1.516)    | (0.059)                                       |           | (1.682)    | (1.671)    | (0.065)          |
| Loans / Total assets             |   | -29.748*** | -28.617*** | -6.279***                                     |           | -28.942*** | -27.817*** | -5.785*          |
| •                                |   | (2.508)    | (2.521)    | (0.664)                                       |           | (2.649)    | (2.660)    | (0.679)          |
| Equity / Total assets            |   | -19.345**  | -20.078**  | -1.606  |           | -21.327**  | -22.413**  | -0.08            |
|                                  |   | (9.328)    | (9.336)    | (2.609)                                       |           | (10.257)   | (10.249)   | (3.019)          |
| \ MP                             |   |            | -0.618**   | -0.898**                                      |           |            | -0.931 *** | -1.217           |
|                                  |   |            | (0.292)    | (0.352)                                       |           |            | (0.350)    | (0.400           |
| ∆ MP * Lerner index              |   |            | -2.247**   | -1.902*                                       |           |            | -0.511     | -0.21            |
|                                  |   |            | (0.923)    | (1.062)                                       |           |            | (0.909)    | (0.959)          |
| Constant                         | 13.824***                               | 213.870*** | 213.664*** | 22.637***                                     | 12.159*** | 221.074*** | 220.733*** | 18.741           |
|                                  | (0.354)                                 | (21.281)   | (21.135)   | (1.042)                                       | (0.357)   | (23.646)   | (23.501)   | (1.216)          |
| Average Lerner index             | 0.209                                   | 0.209      | 0.209      | 0.209   | 0.241     | 0.241      | 0.241      | 0.24             |
| ow Lerner index                  | 0.146                                   | 0.146      | 0.146      | 0.146   | 0.163     | 0.163      | 0.163      | 0.163            |
| High Lerner index                | 0.275                                   | 0.275      | 0.275      | 0.275   | 0.313     | 0.313      | 0.313      | 0.313            |
| Procyclicality: Average          | 1.663                                   | 1.604      | 1.558      | 1.602   | 1.643     | 1.596      | 1.548      | 1.585            |
| Procyclicality: Low Lerner index | 1.443                                   | 1.381      | 1.331      | 1.408   | 1.453     | 1.459      | 1.403      | 1.413            |
| Procyclicality: High Lemer index | 1.896                                   | 1.838      | 1.797      | 1.807   | 1.819     | 1.721      | 1.682      | 1.744            |
| Difference between High and Low  | 0.453                                   | 0.457      | 0.465      | 0.399   | 0.366     | 0.262      | 0.278      | 0.330            |
| Observations                     | 24,194                                  | 24,194     | 24,194     | 24,194  | 23,765    | 23,765     | 23,765     | 23,76            |
| R-squared                        | 0.529                                   | 0.559      | 0.561      |   | 0.538     | 0.568      | 0.569      |                  |
| Number of banks                  | 3,661                                   | 3,661      | 3,661      | 3,661   | 3,622     | 3,622      | 3,622      | 3,622            |
| F                                | 1721                                    | 1404       | 1256       |   | 1722      | 1416       | 1284       |                  |



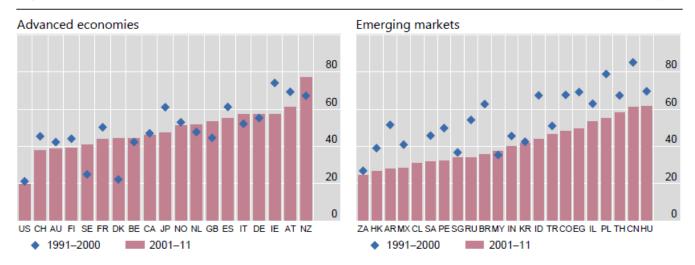




- One important extension of our analysis to the existing literature is that we do not just focus on competition among banks, but we also consider competition from direct finance as a potential driver of credit procyclicality
- → Indeed, all financial systems combine bank-based and market-based intermediation
- → But financial structure, i.e. the particular blend of the intermediation channels, varies across countries
- → Then, in line with previous results, one can expect that countries characterized by a relatively high degree of competition between banks and financial markets exhibit a lower credit procyclicality than bank-based economies

Ratio of bank credit to total private sector funding

In per cent Graph 1



AR = Argentina; AT = Austria; AU = Australia; BE = Belgium; BR = Brazil; CA = Canada; CH = Switzerland; CL = Chile; CN = China; CO = Colombia; DE = Germany; DK = Denmark; EG = Egypt; ES= Spain; FI = Finland; FR = France; GB = United Kingdom; HK = Hong Kong SAR; HU = Hungary; ID = Indonesia; IE = Ireland; IL = Israel; IN = India; IT = Italy; JP = Japan; KR = Korea; MX = Mexico; MY = Malaysia; NL = Netherlands; NO = Norway; NZ = New Zealand; PE = Peru; PL = Poland; RU = Russia; SA = Saudi Arabia; SE = Sweden; SG = Singapore; TH = Thailand; TR = Turkey; US = United States; ZA = South Africa.

The ratio of bank credit to the private sector is expressed as a percentage of the sum of bank credit plus bond and equity market capitalisation. A higher value of the indicator suggests a financial structure that is more bank-oriented. For four countries (Egypt, Israel, New Zealand and Saudi Arabia) data for outstanding bonds issued by the private sector are not available. However, the bond markets in these countries are modest in size and the ordering of the countries in the graphs does not change even in the light of proxies obtained from other sources. A blue diamond above (below) the bar indicates that a particular system became more (less) market-oriented during the 2000s than it was during the 1990s.

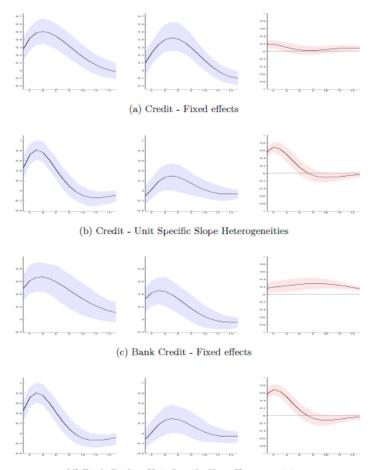
Source: Gambacorta et al. (2014)





- Moreover, investigating the link between financial structure and credit procyclicality is in line with some recent studies (Allard and Blavy, 2011; Gambacorta et al., 2014; Grjebine et al., 2014) that assessed whether a bank-based system, or on the contrary a market-based system, is more resilient to crisis, and then better to moderate business cycle fluctuations
- In line with Levine (2002) and Langfield and Pagano (2016), financial structure is proxied by considering a bank-market ratio, defined as bank credit divided by stock and private bond market capitalization: larger values of the ratio indicate a more bank-based financial system
- Results confirm previous findings: competition from direct finance reduces credit procyclicality

Figure 8: Impulse Response Functions of Credit to a GDP shock: Financial structure



(d) Bank Credit - Unit Specific Slope Heterogeneities

Note: The figure shows impulse responses of credit and bank credit to a one-percentage-point shock in the output cycle evaluated (from left to right) at the  $80^{th}$  (high level) and  $20^{th}$  (low level) percentiles of the indicator of financial structure. The charts on the right represent the differences between the two. The colored bands represent the 5% error band (two standard deviations) generated by bootstrapping (1000 draws).



# **Concluding remarks**

- First paper in the literature that empirically investigates the role of banking competition on credit procyclicality
- Panel framework for a large sample of European countries and two levels of analysis considered: macro-level and bank-level
- Bank competition matters for credit procyclicality!
- → higher credit procyclicality in a low competitive environment
- Fostering banking competition?
- → reduces credit procyclicality and macroeconomic volatility, helps monetary policy transmission, but what are the implications in terms of financial stability?
- → no consensus in the empirical literature (see, e.g., Zigraiova and Havranek, 2016) and different impact of bank competition on individual angle systemic risks (Leroy and Lucotte, 2017)



# Thank you for your attention

**Yannick LUCOTTE** 

ylucotte@gmail.com