ASSESSING THE EFFECTIVENESS OF THE PORTUGUESE BORROWER-BASED MEASURE IN THE COVID-19 CONTEXT

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Abstract
Based on the macroeconomic projections from Banco de Portugal and using an integrated micro-macro model developed by Gross and Poblacion (2017), this paper makes a first attempt at gauging the impact of the Covid-19 pandemic on Portuguese households and banks. To this end, we examine how the borrower-based measure, which has been put into place in 2018, might have been successful in dampening the negative economic effects of the pandemic on households’ debt-servicing capacities and thereby onto the banking system. We find that the borrower-based measure, defined as an LTV ratio cap of 90%, a shocked DSTI ratio cap of 50%, and a maturity cap for mortgage loans of 40 years, leads to (i) a reduction in households’ loss rate (LR), caused by both a decrease in households’ probability of default (PD) and loss given default (LGD), and (ii) an increase in the capital ratio of the banking system, compared with a scenario where these limits are not in place. We also find positive effects of introducing a shocked DSTI ratio cap, calculated according to the Portuguese borrower-based measure, as it further (i) decreases the risk parameters of the borrowers and (ii) increases the capital ratio of banks.

JEL: G01, G51, G21
Keywords: macroprudential policy, central bank, Covid-19.

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1. Introduction

The Covid-19 pandemic has deeply affected the Portuguese economy. A large number of businesses had to close down during several lockdown periods. Despite different support measures put into place by the Portuguese government (moratoria on loans, government-backed loans, lay-offs, etc.), not all businesses were able to survive the crisis, which has potentially led to a reduction in some households’ income and an increase in the unemployment rate. Once the moratoria expire, it will become clear which households that hold mortgage debt will be able or not to start servicing this debt again. This is an important question not only in terms of households’ debt sustainability but also for the banking sector as this might lead to a renewed increase in non-performing loans, also depending on further measures that may be implemented in the meantime.

Based on the macroeconomic projections from Banco de Portugal and using an integrated micro-macro model developed by Gross and Población (2017), this paper will make a first attempt at gauging the potential impact of the Covid-19 pandemic on Portuguese households and banks. To this end, we will examine if the borrower-based measure, which has been put into place in 2018, has been successful in dampening the negative economic effects of the pandemic on households’ debtservicing capacity and thereby on the banking system. We do this by comparing the respective results with a scenario where the measure is not in place.

The Portuguese borrower-based measure has been put into place in 2018 as a reaction to easing credit standards that were coupled with high levels of indebtedness and low saving rates of Portuguese households. Therefore, Banco de Portugal, as the Portuguese macroprudential authority, has adopted measures that target new loans to households, including mortgages as well as consumer credit. These measures have been issued on 1 February 2018 in the form of a Recommendation, which sets limits to some of the credit criteria that are used by financial institutions when assessing the creditworthiness of borrowers. They target the Loan-to-Value (LTV) ratio, the Debt-Service-to-Income (DSTI) ratio as well as the maturity of new loans and the regular payments of interest and capital. The main aim of this measure is to prevent credit institutions and financial companies from taking on excessive risk when granting credit to households. Ultimately, this should lead to a more resilient financial sector as well as affordable access to finance for borrowers (Leal and Lima 2018).

The Recommendation is a macroprudential measure that directly targets the borrower by potentially restricting the amount of credit available to her. This is done by tightening the borrowing constraints for certain groups of borrowers and stands in contrast to measures that are applied at the level of the bank (so-called capital-based measures), which promote an increase in capital requirements. Both macroprudential measures have the ultimate aim of improving the resilience of financial institutions. While capital-based measures raise institutions’ resilience in an immediate and direct manner, borrower-based measures improve institutions’
resilience indirectly and over the medium term by improving the risk level of new credit, which results from the enhanced resilience of borrowers.

This resilience of borrowers is especially important in challenging economic environments that are characterized by high uncertainty. The Covid-19 pandemic is an example of such a challenge. The containment measures that the great majority of countries across the globe has implemented to safeguard public health resulted in a synchronized global sudden stop in economic activity. This makes the global Covid-19 crisis unique, as it has negatively impacted both supply and demand (Boissay and Rungcharoenkitkul 2020).

In this paper, we assess the effectiveness of the aforementioned borrower-based measure in the context of the Covid-19 pandemic. To this end, we conduct a counterfactual analysis that looks at the effectiveness of the measure. In particular, we look at its effect on the loss rate (LR) of borrowers (households) as well as the LR’s respective impact, via potential credit-related losses, on the capital/resilience of credit institutions.

Borrowers’ loss rates are estimated in a scenario in which the measure is in place versus a scenario in which it is not, using an integrated micro-macro model as developed by Gross and Población (2017) and applied to Jurča et al. (2020). The effectiveness is analyzed considering each limit separately as well as all limits jointly. Additionally, we assess the importance of a DSTI ratio with interest rate and income shocks, as envisaged in the Portuguese borrower-based measure, in comparison with a DSTI ratio without shocks. The shocks refer to the way the numerator and the denominator are constructed. In the Portuguese case, the numerator of the ratio takes the impact of an interest rate increase, depending on the loan’s original maturity and the interest rate regime into account. The denominator includes a reduction in borrowers’ income of at least 20% as of the age of 70, given that a material decrease is expected in the transition from working life into retirement. These shocks are in line with the European Banking Authority’s Guidelines on loan origination and monitoring.

This paper contributes to a growing number of studies that aim to assess the effectiveness of borrower-based measures targeted at households. Overall, these studies point to a positive effect of borrower-based measures on financial stability and social welfare.

One of the most common borrower-based macroprudential policies targeting the housing sector is the adoption of LTV ratio limits to housing loans. The countercyclical nature of the LTV ratio is shown by a number of studies based on Dynamic Stochastic General Equilibrium (DSGE) models. Lambertini et al. (2013) show that countercyclical LTV ratio limits that respond to credit growth are more effective in stabilizing credit over the cycle than interest-rate rules because

1. See Appendix 2 for more details.
2. For more information, see EBA’s Final Report on the Guidelines on loan origination and monitoring.
the former do not increase inflation volatility. Additionally, by tightening leverage and domestic borrowing limits during boom periods and relaxing the limits during recessionary periods, this type of measure helps borrowers to smooth consumption over time (Mendicino and Punzi 2014). Using a DSGE model, Gelain et al. (2012) show that DSTI ratio limits can be more effective than an LTV ratio in curbing house price growth and household debt volatility.

There is a number of papers that looks at borrower-based measures from a cross-country perspective. One of the earliest is by Almeida et al. (2006) and examines the relationship between the level of the LTV ratio, house prices and demand for new mortgage lending in the presence of an income shock. Using a sample of 26 countries, they find that house prices as well as new mortgage lending are more sensitive to aggregate income shocks in countries with higher LTV ratio limits. By the same token, Lim et al. (2011), using data from a group of 49 countries to assess the effectiveness of macroprudential measures, suggest that limits to the LTV ratio and to the DSTI ratio may help in dampening the procyclicality of credit.

Other empirical cross-country studies also highlight the countercyclical nature of some borrower-based measures. Cerutti et al. (2017), using a sample of 119 countries over the 2000–2013 period, conclude that the introduction of limits to the LTV and Debt-to-Income (DTI) ratios are associated with reductions in the real growth rates of credit and house prices. However, the effectiveness is more visible when growth rates are very high, as the limits become less countercyclical during busts. Gross and Población (2017), using an integrated micro-macro model and data from the Household Finance and Consumption Survey (HFCS) of 4 European countries, find that limits to the LTV and DSTI ratios may reduce households’ risk, measured by the probability of default (PD) and loss given default (LGD), when a shock occurs, thus having a positive effect on banks’ capital. Additionally, comparing the efficacy of LTV versus DSTI, the results suggest that DSTI ratio caps are more effective in containing household risk.

Another strand of the literature uses micro-level data for country-specific studies. Asian-Pacific countries have a long experience with borrower-based measures applied to households. Therefore, they provide some interesting results from several ex-post studies. Igan and Kang (2011) use survey data on housing and mortgage decisions of households in South Korea and examine the impact of LTV ratio limits on house price dynamics, residential real estate market activity, and household leverage. They find that transaction activity decreases significantly in the short term following the tightening of LTV and DTI ratio limits. Furthermore, expected house price increases become more muted after the introduction of an LTV ratio limit, and plans to buy a house are more likely to be postponed, especially in the case of households that already own a house. The authors do not find an impact of the regulatory tightening on households’ debt levels and growth rates of mortgage loans. Wong et al. (2011) analyze the effects of the LTV ratio on Hong Kong’s property market. Using the coefficients of a cross-country model (13 economies), they simulate the impact of shocks to property prices in Hong Kong.
Their results suggest that an LTV is effective in reducing systemic risk arising from the boom-and-bust cycle of property markets. The transmission channels behind these results are developed in Wong et al. (2014). They conclude that the effect of the LTV ratio on borrowers’ leverage is responsible for strengthening the resilience of banks to property price shocks.

Ireland introduced an LTV ratio limit in 2015 in order to curb house price inflation. Duffy et al. (2016) use a structural model to simulate the implementation of the measure with respect to a scenario where the measure is not in place. They find the measure to be effective in reducing house prices. However, it also leads to a decrease in the demand for new houses and a relatively muted supply-side response due to the financial crisis of 2007/2008 and the following sovereign debt crisis. Acharya et al. (2020) assess potential side effects resulting from the implementation of the Irish limits. They do this by combining supervisory loan-level data on residential mortgages and house price data and examine the impact of the introduction of LTV and loan-to-income (LTI) ratio limits on residential mortgages in Ireland. Following the introduction of the LTV and LTI ratio limits, banks reallocated mortgage credit from low-income to high-income households and from counties where borrowers are closer to the lending limits to counties where borrowers are farther away from the lending limits. However, this led to an increase in banks’ risk-taking in both credit to firms and holdings of securities, the two largest classes not targeted by the regulation.

Gabarro et al. (2019) examine the effects of the LTV ratio limit introduced in the Netherlands in 2011 on households’ leverage, liquidity and default. Using a database that combines information from income, wealth-tax records and property ownership for the entire Dutch population, they find that the introduction of the LTV ratio limit has been effective in reducing households’ leverage and mortgage debt servicing costs, taking into account that households’ have responded to the introduction of the policy by taking out smaller loans. These results are accompanied by a reduction in mortgage default. The effects are more pronounced for households in lower percentiles of wealth and those with fewer liquid assets. de Araujo et al. (2020) use credit register and employment register data from Brazil to explore the effects of an LTV ratio limit for housing loans on contract terms and borrower behaviour. They show that borrowers affected by the regulation pay higher interest rates, borrow loans with shorter maturities and lower amounts. Moreover, these borrowers purchase cheaper houses and are therefore less likely to default.

Following the methodology by Gross and Población (2017), Jurča et al. (2020), using data from the HFCS applied to Slovakia, conclude that borrower-based measures can improve household and bank resilience to macroeconomic downturns, in particular when several limits (LTV, DTI and DSTI ratios) are in place.

Despite the overall benefits of imposing an LTV ratio, there is some evidence of negative side effects. Tzur-Ilan (2019) uses loan-level data from the Bank of Israel to explore the costs and benefits of the "strict" LTV ratio limit adopted in 2012 on housing choices and credit conditions. She finds that although reducing borrowers’ leverage, the adoption of the macroprudential measure has pushed the constrained
borrowers to purchase cheaper and lower quality assets, which are farther from the city centre and in less desirable neighbourhoods. Furthermore, the policy change has been associated with higher interest rates, which may be related to the fact that households bought riskier assets and increased unsecured credit. Therefore, the paper emphasizes that macroprudential policies that focus on the stability of the financial system can have undesirable implications at the micro level.

The main contributions of our paper to the empirical literature are twofold. Firstly, this paper focuses on the borrower-based measure applied to Portugal, which comprises country-specific LTV, DSTI and maturity limits. Secondly, to the best of our knowledge, this is the first paper to analyze the effectiveness of these types of measures in the context of the Covid-19 pandemic.

Our results point to a reduction in households’ LRs due to imposing the borrower-based measure, caused by both a decrease in households’ PDs and LGDs. The joint effect of the three caps is a reduction of the LR by 0.046 p.p.

The results also show the benefits in introducing a shocked DSTI ratio cap, as done in the Portuguese borrower-based measure. Without shocks to the DSTI ratio, the LR would decrease by 0.039 p.p., 0.005 p.p. less than with the DSTI ratio shocks.

The model also points to a positive impact of the borrower-based measure on the capital ratio of the banking system. The introduction of the macroprudential measure (with shocks to the DSTI ratio) leads to an increase in the capital ratio of the banking system by 0.74 p.p. Both the numerator and denominator of the capital ratio contribute to that. The introduction of the measure results in a reduction of credit and interest income losses, which increases capital. Moreover, it leads to a reduction in the risk weights for mortgage loans and a reduction in lending to families with higher risk profiles, given that the credit lending criteria become more restrictive, which decreases risk-weighted assets. The latter effect is particularly strong. Without the shocks to the DSTI ratio, the capital ratio of the banking system would increase by 0.63 p.p., 0.11 p.p. less than with a shocked DSTI ratio.

2. Data and Methodology

This section gives a brief overview of the data being employed in this paper. It then continues to describe the different modules of the integrated micro-macro model à la [Gross and Población (2017)].

2.1. Data

The model employed in this paper requires micro as well as macro data, as detailed in [Gross and Población (2017)]. At the micro level, we use the Eurosystem’s Household Finance and Consumption Survey (HFCS). The HFCS collects information on the finances and consumption of households in 22 (mainly Euro Area) countries. It contains information at the household level.
(e.g. assets and liabilities, savings and consumption) as well as at the level of the households’ members (e.g. employment status, labour income and sociodemographic information). The use of the different variables at their respective level of aggregation will become clear throughout the description of the different modules. Table 1 gives an overview of the variables being used in this paper.

Our paper focuses on the Portuguese part of the survey, the ISFF (Inquérito à Situação Financeira das Famílias) for the year 2017, i.e. the currently most recent available survey wave. The 2017 wave contains data collected from 5,924 valid household interviews (15,079 household members), which corresponds to a response rate of 85%, the highest of all participating countries. For more information on this survey wave, see Costa et al. (2020).

As this paper focuses on the effects of the borrower-based measure in Portugal, we restrict our sample to borrowing households only. This reduces the sample size to 2,749 households or 8,114 household members.

At the macro level, we collect data for six variables at a quarterly frequency for the period 2005Q1 - 2020Q4 that feed into the VAR model in Module 1. These variables are: unemployment rate (seasonally adjusted), nominal compensation per employee, short-term interest rate (3 months), stock price index (Dow Jones Euro Stoxx 50), nominal residential property price index and nominal domestic credit from financial institutions to the private sector. In the context of the Covid-19 pandemic, we also use the 3-year March 2021 economic projections of Banco de Portugal, from 2021Q1 to 2023Q4. These macro projections are characterized by the onset of the pandemic in Portugal in early 2021, which led to the introduction of containment measures. As a consequence, this resulted in a sharp fall in economic activity.

Furthermore, the model requires additional model parameters, the so-called metadata, which are needed for calibration purposes. These are the average duration of unemployment for the whole working population, an average implicit tax rate on labour, and a so-called net replacement rate, which is the share of previous net income that a person receives in terms of benefits when being unemployed. Table 2 gives an overview of all macro and metadata variables used in this paper.

2.2. Methodology

The integrated micro-macro model used in this analysis, which comprises six modules, estimates three risk parameters for households: Probability of Default, Loss Given Default and the Loss Rate. These can be calculated with or without imposing limits to the LTV and DSTI ratios and to maturity, thus making it...
possible to measure the impact of their implementation. In a first stage, risk parameters, calculated on the basis of micro data, are affected by developments in macroeconomic and financial variables (unemployment rate, residential property prices, stock price index, income per employee – including wages, premia, income in kind paid by employers to employees – and credit to the non-financial private sector) – the so-called 1st round effects. The developments in the macroeconomic and financial variables are based on the March 2021 economic projections of Banco de Portugal. These define the evolution of micro data from each household’s balance sheet (assets and liabilities). In a second stage, the shock to credit demand, caused by imposing limits on credit standards, influences macroeconomic and financial variables, which then in turn have an impact on the risk parameters calculated in the first stage – the so-called 2nd round effects. Figure 1 illustrates the basic structure of the model, showing the different modules and their respective interlinkages.

Module 1 estimates the evolution of macroeconomic and financial variables, such as the unemployment rate, credit granted to the non-financial private sector, the three-month interest rate, residential real estate market prices, compensation per employee and the evolution of the stock index. We start by estimating the VAR model with the six macroeconomic and financial variables previously mentioned to capture the dependencies between them. Based on the estimated model, we generate 1,000 3-year stochastic forward simulations consistent with the historical dependencies estimated in the VAR.

Then we use the 3-year March 2021 economic projections of Banco de Portugal, from 2021Q1 to 2023Q4, to re-centre the stochastic forward simulations around

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5. Only the unemployment rate is published in the economic projections of the Economic Bulletin of Banco de Portugal. The remaining variables are not publicly available. The projection for the stock price index is not calculated by Banco de Portugal. We calculate it by using the March 2021 economic projections of Banco de Portugal for the remaining variables and the historical dependencies between them obtained in the VAR.
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these economic projections (i.e., around Covid-19 pandemic macro data). For each of
the six macroeconomic and financial variables, denoted by the index i, and for
each quarter, denoted by the index t, we calculate the deviation \((shift_{i,t})\) between
the economic projections of Banco de Portugal \((economic\ projections_{i,t})\) and
the mean of the 1,000 stochastic forward simulations obtained from the VAR
\((stochastic\ forward\ simulation_{i,t,z})\), for each \(i\) and \(t\), where \(z\) denotes the
number of simulations. This allows introducing uncertainty around the Covid-19
macroeconomic scenario.:

\[
shift_{i,t} = \text{economic projections}_{i,t} - \text{mean(stochastic forward simulations}_{i,t,z})
\]  

(1)

Then, the re-centred forward simulations for each macro and financial variable
are given by:

\[
\text{recentred forward simulations}_{i,t,z} = \text{stochastic forward simulations}_{i,t,z} + shift_{i,t}
\]  

(2)

Module 2 calculates the probability of an individual being unemployed according
to his/her socio-demographic characteristics, such as age, gender, marital status,
level of education and country of birth (Portugal or foreign), using a binary logit
model. For this purpose, only employed, self-employed and unemployed household
members are included, whereas students and retirees are excluded. Therefore, the
probability of each household member being unemployed is based on the following
logistic regression:

\[
y^k_t = \alpha + \beta_1 x^k_1 + \beta_2 x^k_2 + ... + \beta_j x^k_j + \epsilon^k_t,
\]  

(3)

where \(y^k_t = 1\) if household member is unemployed at time \(t\) and \(0\) otherwise.
\(x^k_j\) stands for age, marital status, level of education, gender, and country of birth
of each household member \(k\). The results of the regression are provided in Table 3
of Appendix 1.

Module 3 calibrates the level of the probability of being unemployed based
on the results from Module 2 but adjusting them such that the aggregate
unemployment rate resulting from the micro-level estimates matches the simulated
forward path of the unemployment rate from the macro module (Module 1).
This is done by adjusting the unemployment status of randomly chosen household
members at the micro level until the unemployment rate obtained from the micro
module matches the unemployment rate from the macro module.

In module 4, households’ LRs are calculated. The criterion for a given household
to enter into default is insufficient income and assets to meet the instalments
associated with the respective level of indebtedness. To this end, we first need to
define liquid assets and how they evolve over time. Liquid assets are composed of
cash and cash equivalents (C) as well as bonds (B) and stocks (S). The rationale
for including these types of assets is that such a buffer might be drawn down in
case a household’s periodic income does not suffice to cover periodic expenses, including debt repayment.

Liquid assets can therefore be written as:

$$LiqA^k_t = C^k_t + B^k_t + S^k_t$$  \hspace{1cm} (4)

The periodic change of liquid assets is determined as follows:

$$\Delta LiqA^k_t = \Delta B^k_t + \Delta S^k_t - \min(L^k_t, EXP^k_t)$$

$$+ \begin{cases} 
INC^{G,k,n1}_t(1-r)(1-l) & \text{if employed} \\
U^{N,k,n2}_t(1-r)(1-l) & \text{if unemployed} 
\end{cases}$$  \hspace{1cm} (5)

The evolution of the micro variables is driven by the projections for the macroeconomic and financial variables. The outstanding amount of loans ($L^k_t$) decreases with the periodic debt repayment ($EXP^k_t$) until the value of the loans reaches zero, meaning all debt has been repaid.

The income $INC^{G,k,n1}_t$ as well as the unemployment benefit $U^{N,k,n2}_t$ (depending on whether a household member is employed or unemployed) are subject to a deduction regarding the expenses to cover the costs of living ($l$) and are net of tax expenses ($r$). Additionally, the model also considers the interest, dividends and valuation of assets such as bonds, stocks and investment fund units, as well as rents from real estate assets other than those intended for own and permanent residence, and interest from deposits. Income and other assets will evolve according to the projection for macroeconomic and financial variables according to the projections in module 1.

Each household defaults if liquid assets become negative ($LiqA^k_{t-1} + \Delta LiqA^k_t < 0$) in some period along the projected horizon. Once a household defaults, it is not allowed to recover and resume its debt repayment by assumption, even though we consider a recovery rate at the macro level, as explained in module 6. The probability of default for each period is therefore computed as the number of defaults over the number of households.

In parallel, LGD is computed for households entering into default, taking into account the projections for the value of the loan collateral based on the estimated market prices for residential real estate in module 1. This module takes the retirees into account that have formerly been excluded in module 2.

In module 5 the LR estimated in module 4 is being recalculated, this time under a scenario where there are limits to the LTV and DSTI ratios and to maturity, thus excluding credit that does not comply with these three limits considered individually as well as jointly (first-order effects). These limits are assumed to be imposed in 2017, i.e., before the macroprudential limits actually entered into force (July 2018). As in Gross and Población (2017), we assume that those households that breached the caps did not receive any amount of credit. In this scenario there is a reduction in credit to the non-financial private sector (NFPS), which in turn will negatively influence developments in the unemployment rate at the micro and macro level.
Assessing the effectiveness of the Portuguese borrower-based measure (modules 1 and 3), ultimately leading to a deterioration in the LR of households (second-order effects).

Therefore, the net benefit of the macroprudential measure is a result of the direct effects from the reduction of credit to higher-risk borrowers on the reduction of LRs (first-order effects), less indirect effects. These result from the potential increase in LRs, caused by the impact of the credit reduction on the other macroeconomic variables, in particular on the unemployment rate (affecting the PD) and on the evolution of prices in the residential real estate market (affecting the LGD) (second-order effects). Using the impulse-responses of the macro-financial variables to a 1 p.p. shock to NFPS credit growth, the module adjusts the simulated macro paths from module 1 based on the scaled shock to lending. It then re-computes the LRs implied by the borrower-based measure limits considered individually as well as jointly, thereby accounting for the macro-economic feedback of the policy measures.

Finally, module 6 estimates the impact of the macroprudential measure on the banking system’s average total capital ratio. This results either from the effect on the numerator (own funds), through the flow of credit losses, or from the impact on the denominator, through the new amount of credit in default net of impairments, which has a different risk weight.

The flow of provisions is computed as $\text{Stock of Provisions}_t - \text{Stock of Provisions}_{t-1}$, the stock of provisions being equal to $\text{LGD}_t \times \text{NPL}_t$. In turn, the $\text{NPL}_t$ is defined as the $\text{NPL}_{t-1}$ net of write-offs and cures plus the new defaulted credit as follows: $\text{NPL}_t = \text{NPL}_{t-1}(1 - w_t - c_t) + \text{PD} \times (L_t - \text{NPL}_{t-1})$, where $L_t$ are gross loans and $L_t - \text{NPL}_t$ is assumed to equal the exposure at default $\text{EAD}_t$. For the purpose of this paper, we assume that the write-off parameter $w$ for NPLs equals zero. $c$ stands for cures, i.e., the migration of loans from nonperforming back to performing status. Even though cures are not considered at the micro level (an assumption that is embedded in module 5), we have carried out an adjustment at the macro level based on historical values of cures regarding mortgage loans of the largest banks of the Portuguese banking system. Finally, the impact on the Total Capital ($\text{TC}$) ratio is computed as $[\text{TC}_{\text{baseline}} - \text{Flow of Provision}_t + \text{Interest Income}_t] / [\text{RWA}_{\text{baseline}} + \text{RWA}_t - \text{RWA}_{t-1}]$ where the $\text{TC}_{\text{baseline}}$ and $\text{RWA}_{\text{baseline}}$ are the total capital and the risk-weighted assets in the period $t-1$, respectively. To compute the impact on RWA, it is necessary to take the proportion in the Portuguese case of the portfolio of credit for house purchase that is subject to the internal ratings method into account and the one whose minimum capital requirements are computed using the standard method. All computations are done twice, with the measure in place as well as without the measure.

In view of the original model developed by Gross and Población (2017), some adaptations were made, namely: (i) calculation of the DSTI ratio according to the Portuguese Macroprudential Recommendation, i.e. with a shock to the income of the borrower(s), considering a reduction in income in the case of a borrower aged 70 and over, and with a shock on the debt service, considering the impact of an
interest rate increase in the case of variable and mixed interest rate agreements, and (ii) inclusion of limits to the maturity of loans granted for house purchase.

While the Portuguese borrower-based measure is very encompassing, this paper focusses on the LTV ratio cap set for mortgage credit for own and permanent residences (90%). Further limits that target mortgage credit for purposes other than own and permanent residences (80%), mortgage credit for purchasing immovable property held by credit institutions, and property financial leasing agreements (100%) are beyond the scope of this analysis. The same holds true for the original maturity limit for new consumer credit agreements. However, we calculate the DSTI ratio of households by taking all of the borrowers’ loans, including consumer credit, into account.

3. Results

We compute the individual and the combined effect of the limits of the borrower-based measure on the households’ LRs. The borrower-based measure is defined here as an LTV ratio cap of 90%, a shocked DSTI ratio cap of 50%, and a maturity cap for mortgage loans of 40 years.

Furthermore, we analyze the effect of imposing a DSTI ratio cap, calculated according to the Portuguese borrower-based measure, with shocks to the numerator (payments for households’ debt) and denominator (household income).

Finally, we link the effect on households’ LRs to the capital ratio of the banking system.

The objective of using the Covid-19 scenario is to quantify the expected increase in the resilience of households and reduction in bank mortgage portfolio losses resulting from implementing a combination of borrower-based macroprudential measures.

When we run the integrated micro-macro model, the unemployment rate path has a forecast horizon of 3 years and is based on the VAR model, re-centred around the March 2021 projections by Banco de Portugal (Figure 2).

Unemployment is expected to increase until the 2nd quarter of 2021, before entering onto a downward path until the end of the projection horizon. At the end of this horizon, the unemployment rate is expected to exceed its 2019 value, though still being considerably below the levels observed during the 2011-13 crisis.

The results, based on an average of the 1,000 macroeconomic simulations re-centred around the Covid-19 scenario, point to a reduction in households’ LRs with the borrower-based measure in place. The joint effect of the three caps, in the context of the Covid-19 pandemic, is a reduction in the LR by 0.046 p.p., after taking the 2nd round effects into account (Figure 3).

The results also show the benefits of introducing a shocked DSTI ratio cap, calculated according to the Portuguese borrower-based measure. Without shocks to the DSTI ratio, the LR would decrease by 0.039 p.p., 0.005 p.p. less than with DSTI ratio shocks (Figure 4).
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Imposing the LTV ratio, shocked DSTI ratio and maturity limits separately, after taking the 2nd round effects into account, we also find that the DSTI ratio cap with shocks is the one that reduces the borrowers’ LR the most. The LTV ratio cap decreases the LR by 0.025 p.p. The DSTI ratio cap (with shocks) reduces the LR by 0.046 p.p. The maturity cap decreases the LR by 0.011 p.p. The DSTI ratio cap (without shocks) reduces the LR by 0.03 p.p.

The model also suggests a positive impact of the borrower-based measure on the capital ratio of the banking system in the context of the Covid-19 pandemic. The introduction of the macroprudential measure leads to an increase in the capital
ratio of the banking system by 0.74 p.p. (Figure 5). Both the numerator and the denominator of the capital ratio contribute to this.

The introduction of the measure results in a reduction of credit and interest income losses, which increases banks’ capital. Moreover, it leads to a reduction in the risk weights for mortgage loans and a reduction of lending to households with higher risk, given that the credit lending criteria become more restrictive, which decreases risk-weighted assets. The latter effect is particularly strong.

The results also show that without the shocks to the DSTI ratio, the capital ratio would increase by 0.63 p.p., 0.11 p.p. less than with DSTI ratio shocks (Figure 6).

![Figure 5: 3-year cumulative change in risk weighted capital ratios | In percentage points]

### 4. Conclusions

This paper aims at assessing the effectiveness of the Portuguese borrower-based measure in the context of the Covid-19 pandemic, following the methodology by Gross and Población (2017). The Portuguese measure comprises caps to the LTV and DSTI ratios as well as a cap to maturity.

Using an integrated micro-macro approach, we conduct a counterfactual analysis that looks at the effectiveness of the Portuguese borrower-based measure during the recent pandemic. The question we wanted to answer was twofold. First, can the measure mitigate the riskiness of households in terms of their LRs? Second, what are its effects on the Portuguese banking system via potential credit-related losses?

Employing data from the HFCS, our results can be summarized as follows. First, we find evidence of a reduction in households’ riskiness, which is due to imposing
Assessing the effectiveness of the Portuguese borrower-based measure

Figure 6: 3-year cumulative change in risk weighted capital ratios | In percentage points

The joint effect of the three caps is a reduction of the LR by 0.046 p.p.

Furthermore, the integrated micro-macro model suggests a positive impact of the borrower-based measure on the capital ratio of the banking system of 0.74 p.p., in the context of the Covid-19 pandemic.

The results also show that considering a DSTI ratio with interest rate and income shocks, as envisaged in the Portuguese borrower-based measure, is effective as it further decreases households’ risk parameters and further increases the capital ratio of credit institutions.

While the pandemic is still ongoing, our analysis is a first step in assessing the effectiveness of the Portuguese borrower-based measure in the context of this severe economic shock. Though the micro-level data employed refer to the year 2017, the integrated micro-macro approach enables us to re-centre our results around the pandemic macro data, which allows for a cautious first assessment of the measure in a scenario of severe economic stress.

Summing up, our results suggest that the Portuguese borrower-based measure is indeed successful in increasing the resilience of households as well as of the banking system. However, given that our analysis is based on forecasts and simulations, due to the lack of more recent data on the Covid-19 pandemic, results should still be considered as preliminary and more research will be needed once new data becomes available.
References


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<th>Category</th>
<th>Variable</th>
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<td>Household level, asset side</td>
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<td>Current value of household’s main residence</td>
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<td></td>
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<td>DA2100</td>
<td>Total financial assets</td>
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<td>Outstanding balance of other, non-mortgage debt</td>
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<td>Monthly payments to repay outstanding mortgages</td>
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<td>Monthly payments to repay non-collateralized debt</td>
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<td>Total household gross income</td>
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Table 1. Micro variables from the HFCS used in the model
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<td>Annual stock price growth in 2017</td>
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<td>Annual compensation per employee growth in 2017</td>
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<td>Annual mortgage interest rate on OB</td>
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<td>Estimated annual expected return on mortgages</td>
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<td>Average duration of unemployment in quarters</td>
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<td>Net of tax unemployment benefit over previous income gross of tax</td>
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<td>Mortgage LGD anchor point for 2017</td>
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<td>Ratio of total household new business flows during 2015-17 (36 months)</td>
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<td>Ceiling on monthly gross unemployment benefit flow in EUR</td>
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Table 2. Macro variables used in the model
Appendix 1: Logistic model for unemployment status

The logistic model for the unemployment status of the household members considers the following explanatory variables: (i) marital status, equal to 1 if the household member is married, 0 otherwise, (ii) education, equal to 1 if the household member has a university degree, 0 otherwise, (iii) gender, equal to 1 if the household member is female, 0 if male, (iv) foreign, equal to 1 if household member is foreign, 0 otherwise, and (v) age of the household member.

Marital status enters with a negative sign, suggesting that being married decreases the probability of being unemployed. Having a university degree decreases the probability of being unemployed. Gender, foreign and age are not statistically significant.

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<th>Independent variables</th>
<th>Coefficient estimates (p-values in parenthesis)</th>
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<tr>
<td>Intercept</td>
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<td>Marital status</td>
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<td>Education</td>
<td>-0.9467 (0.00)</td>
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<td>Gender</td>
<td>0.1599 (0.25)</td>
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<td>Foreign</td>
<td>0.0315 (0.80)</td>
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<tr>
<td>Age</td>
<td>0.0026 (0.42)</td>
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Observations 7,319
AUROC 0.62
Gini 0.28

Table 3. Logistic model estimates

Notes: The table reports the coefficient estimates (p-values in parenthesis) from a logistic regression whose dependent variable equals 1 for household members that are unemployed and 0 for those that are employed. AUROC denotes the estimate of the area under the receiver operating curve.
Appendix 2: Overview of the Portuguese borrower-based measure

Banco de Portugal, as the designated Portuguese macroprudential authority, announced the macroprudential measure on 1 February 2018, which covers all new loans to households taken out from 1 July 2018 onwards. The period of five months between the announcement and the implementation was given to allow financial institutions to implement the necessary operational adaptations required to comply with the new requirements. The measure applies to all entities authorized to grant credit in Portugal, which comprise financial companies that have their head office in Portugal as well as branches on Portuguese territory from foreign financial institutions.

The macroprudential measure takes the legal form of a recommendation that follows the “comply-or-explain” principle. Despite the potential flexibility that this measure allows for, banks have broadly accepted the limits laid down in the Recommendation, given the overall consensus with respect to the benefits of this borrower-based measure for financial stability. Consequently, there was a fast convergence of the Portuguese financial system’s most relevant financial institutions to the limits defined for the LTV ratio, for the DSTI ratio and for the maturity.

When designing the scope of this measure, Banco de Portugal decided to exclude loans intended to prevent or address default situations, considering the still high level of non-performing loans on the balance sheets of Portuguese banks at that time. The Recommendation is also not applicable to credit agreements with an amount equal to or lower than the equivalent of ten times the guaranteed monthly minimum wage. Finally, the Recommendation excludes overdraft facilities and other credit with no defined repayment schedule (including credit cards and credit lines), given the difficulty of applying some of the measures envisaged in the Recommendation to these credit agreements.

The macroprudential Recommendation defines limits to the loan-to-value ratio (LTV), the debt service-to-income ratio (DSTI), the maturity of the loan, and it requires regular payments of principal and interest that should be applied to new loans secured by immovable property, credit secured by a mortgage or an equivalent guarantee, and consumer credit (Table 4).

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6. For more details concerning the convergence of the Portuguese banking system to the limits defined in the macroprudential recommendation, please see the three Progress Reports already published on Banco de Portugal’s website: (i) May 2019 progress report, (ii) March 2020 progress report, (iii) March 2021 progress report.
| **LTV limits** | • LTV $\leq 90\%$ - New credit secured by residential immovable property for the purchase or construction of own and permanent residence  
(Recommendation A)  
• LTV $\leq 80\%$ - New credit secured by residential immovable property or credit secured by a mortgage or equivalent guarantee for other purposes than own and permanent residence  
• LTV $\leq 100\%$ - New credit secured by residential immovable property and credit secured by a mortgage or equivalent guarantee for purchasing immovable property held by the institutions themselves and for property financial leasing agreements  
• The value of the property pledged as collateral is given by the minimum between the purchasing price and the appraisal value |
| **DSTI limits** | DSTI $\leq 50\%$, with the following exceptions on the total amount of credit granted by each institution in each year:  
(Recommendation B)  
• up to 10%: DSTI $\leq 60\%$; and  
• up to 5%: no DSTI limit.  
For the calculation of the DSTI, monthly instalments of new loans are assumed constant over the entire period of the loan. For variable and mixed interest rate loans, the impact of an interest rate increase should be considered. The DSTI should also take into account the impact of a reduction in the borrower’s income if the borrower’s age during the term of the loan contract is above 70, except if the borrower is already retired at the time of the creditworthiness assessment. |
| **Maturity limits** | For new loans secured by residential immovable property or credit secured by a mortgage or equivalent guarantee:  
(Recommendation C)  
• Maturity $\leq 40$ years  
• Average maturity of new loans should gradually converge to 30 years until the end of 2022  
For new consumer loans:  
• Maturity of new personal credit $\leq 7$ years  
• Maturity of new personal credit for education healthcare, renewable energy, provided that these purposes are duly evidenced, $\leq 10$ years  
• Maturity of new car credit $\leq 10$ years  
The definitions of personal credit and car credit correspond to those provided for in Instruction No 14/2013 of Banco de Portugal |

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