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# Banking sector performance during the COVID-19 crisis

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# 1. Introduction

To reduce the spread of the novel COVID-19, governments enacted mitigation strategies based on social distancing, national quarantines, and shutdown of non-essential businesses. The halt to the economy represented a large shock to the corporate sector, which had to scramble for cash to cover operating costs as a result of the revenue shortfall. The financial sector, and banks in particular, are expected to play a key role absorbing the shock, by supplying much needed funding (Acharya and Steffen, 2020; Borio, 2020).<sup>1</sup> Under these unprecedented circumstances, central banks and governments enacted a wide range of policy interventions. While some measures were aimed to reduce the sharp tightening of financial conditions in the short term, others sought to support the flow of credit to firms, either by direct intervention of credit markets (e.g., government-sponsored credit lines and liability guarantees), or by relaxing banks' constraints on the use of capital buffers.

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# ABSTRACT

This paper examines the impact of financial sector policy announcements on bank stocks around the world during the onset of the COVID-19 crisis. Overall, we find that liquidity support, borrower assistance programs and monetary easing moderated the adverse impact from the crisis, but their impact varied considerably across banks and countries. By contrast, countercyclical prudential measures led to negative abnormal returns in bank stocks, suggesting that markets price the downside risks associated with these policies.

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While credit institutions are being called to play an important countercyclical role to support the real sector, these actions also have a series of implications for the future resilience of the banking sector. For instance, as lenders exhaust their existing buffers, they might also experience deterioration of asset quality, threatening the systems' stability. As the crisis is expected to continue, even after the lockdowns are lifted and economies start to reopen, the net effect of these policy measures on the banking sector is largely unknown.

The contribution of this paper is twofold. First, we use bank stock prices around the world to assess the impact of the pandemic on the banking sector. Second, we combine bank stock prices with a global database on financial sector policy responses during the pandemic. Using an event study methodology, we examine the stock market response to announcements of different policy initiatives. To better understand the impact of the measures implemented by monetary and supervisory authorities, we exploit the cross-sectional variation in the stock price of banks. In other words, we are interested in both the aggregate response of bank stocks to a particular announcement, as well as the differential effect across banks with different characteristics, such as size, liquidity, ownership, and others.

We use bank data including stock prices, balance sheets, and ownership, for 52 countries covering 896 commercial banks. We first document a systematic underperformance of bank stocks at the onset of the COVID-19 crisis, between March and April of 2020.

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<sup>&</sup>lt;sup>1</sup> Using U.S. data, Acharya and Steffen (2020) provide evidence of the transmission mechanism, whereby firms convert commited credit lines from banks into cash, exerting significant pressure on lenders.

More precisely, for most countries, bank stocks underperform relative to other publicly traded companies in their home country, and relative to non-bank financial institutions. While bank underperformance during downturns is common given their high betas, we show that early in the pandemic, banks' stock returns were abnormally low when calculated relative to an asset pricing model with mixed factors, that is, global and domestic market returns as in Bekaert et al. (2014). In the U.S., Acharya et al. (2021) also report a risk premium on bank stocks following the COVID-19 shock. We show that negative abnormal returns among banks are ubiquoutus to developed and developing countries during our observation period and are robust to multiple model specifications (e.g., Fama and French, 2015). Furthermore, we show that banks with lower liquidity buffers experienced larger than normal price drops.

To study the stock market reaction to different policy measures, we identify financial sector initiatives by government authorities from February 2 to April 17. The data was compiled and made publicly available by the World Bank (Feyen et al., 2021). Our final sample contains 389 financial sector policy announcements in 45 countries (17 developed and 28 developing). We classify approved measures that target the banking sector into four categories. Liquidity support are measures used by monetary authorities to expand bank short-term funding in domestic and foreign currency. Prudential measures deal with the temporary relaxation of regulatory and supervisory requirements, including capital buffers. Borrower assistance include government-sponsored credit lines or liability guarantees to promote the flow of credit to households and firms. Finally, monetary policy includes policy rate cuts and quantitative easing (i.e., asset purchases). Our empirical methodology consists of estimating banks' abnormal returns around the announcement day. Our findings can be summarized as follows:

- · Borrower assistance announcements appeared to have the strongest immediate impact on bank stock prices, both on aggregate and in the cross-section. Banks experienced large abnormal returns following the announcement of these policies. Additionally, larger banks seem to benefit more compared to smaller banks. This is consistent with the observation that new government credit lines, interest rate subsidies, and liability guarantees are more likely to be used by large banks.<sup>2</sup> Borrower assistance initiatives, which typically include the introduction of government guarantees, automatically transfer risks from banks' balance sheets to the sovereign. In turn, these policies require significant fiscal commitments. Relatedly, we find that the positive association between excess stock returns and borrower assistance measures is exclusive to developed countries. In developing countries, where there is less room for fiscal expansion, announcements of borrower support had no effect on stock prices. The market response seems to suggest that the extent of borrower assistance measures is limited in such settings.
- Liquidity support initiatives have a favorable impact in the reduction of the liquidity premium – stocks of banks with less liquid assets respond more strongly to these announcements. Also, smaller banks experienced large abnormal returns when liquidity support measures were announced.
- In contrast, countercyclical prudential measures are associated with negative abnormal returns in bank stocks. Prudential policies allow banks to run down some of their buffers. They also send a strong signal of the willingness of policymakers to lessen the economic impact from the pandemic. However, the fact that bank stock prices drop following the announcements of these policies suggests that markets are also pric-

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ing the downside risk from the depletion of capital buffers, as well as the additional expansion of riskier loans in the balance sheets of banks.

• Results for monetary policy announcements are more mixed. While such announcements were not associated with aggregate bank stock price increases, policy rate cuts did reduce the liquidity premium – banks with lower liquidity displayed higher stock returns around the announcement window. This result confirms that the interest rate policy remained a key tool at the onset of the crisis.

The most important threat to our identification strategy is the endogeneity of policy actions, which could bias our estimates. While the COVID-19 was exogenous, the adoption of a policy mix by financial authorities is not random. We conduct three exercises to mitigate this concern and confirm most of our findings. First, we focus on large cross-border banks in our sample and examine the reaction of their stock prices when policies are announced in countries where they have subsidiaries. Second, we examine the extent to which external rather than domestic factors influence the announcement of financial sector policies across jurisdictions.<sup>3</sup> Third, we examine bank stock prices in Euro area countries following broad cross-country policy announcements from the European Central Bank (i.e., a setting where policies are presumably less driven by within country characteristics).<sup>4</sup>

Our paper is related to a growing literature that studies the economic impact of the pandemic using stock market data (e.g., Borio, 2020; Gormsen and Koijen, 2020; Landier and Thesmar, 2020; Ramelli and Wagner, 2020). Given the extraordinary scale and unprecedented nature of the crisis, it is difficult to quantify the effect of the shock versus the impact of the ensuing economic policies. We add to this literature by examining the effects of the pandemic on stock returns of banks based on an event study methodology.

We also contribute to a longstanding literature that examines the role of policy initiatives during financial crises (Claessens et al., 2005; Reinhart and Rogoff, 2009; Taylor and Williams, 2009). Ait-Shalia et al. (2012) examine the effects of policy announcements on liquidity risk premia and interbank credit during the Global Financial Crisis (GFC). While we also study the market response to policy initiatives, there are noticeable differences between the COVID-19 shock versus previous events of financial and economic stress that grant further analysis. First, since the shock is truly exogenous to the financial sector, the overwhelming response from policymakers has been to relax regulatory requirements and use capital buffers: During our sample, 91% of countries used prudential measures. There is, however, no prior analysis of such a coordinated type of response, and no clear evidence about the mediumand long-term effects of these policies. To the best of our knowledge, our paper represents the first global analysis examining the market response to financial policy measures during the pandemic.

#### 2. Data

We use data from Refinitiv on all publicly traded banks across 52 countries between May 2, 2018 and May 12,020. The data set

<sup>&</sup>lt;sup>3</sup> We find that the adoption of financial sector policy measures is not associated with ex-ante characteristics of the banking sector in the country, or with time-varying market factors. In contrast, policy adoption appears largely explained by the actions of peer countries, namely, by policy contagion.

<sup>&</sup>lt;sup>4</sup> One remaining concern is whether our impact estimates capture investors' reaction to information about the shock revealed through the policy announcement. However, this mechanism does not explain why bank returns are negative (positive) when prudential (borrower support) measures are announced, nor why markets only respond to borrower support measures announced in developed countries. Thus, our preferred interpretation is that our findings capture the impact of the policies rather than the information they carry.

<sup>&</sup>lt;sup>2</sup> Using detailed loan-level data Ornelas et al. (2019) find that larger banks disproportionally allocated government-sponsored credit during the GFC.

Table 1

Summary statistics of banks.

Variable	Mean	SD	P25	P50	P75	Obs
Stock returns	-0.004	0.003	-0.005	-0.003	-0.002	896
Liquidity ratio	0.07	0.06	0.02	0.06	0.11	755
Oil exposure	0.06	1.23	-0.22	0.02	0.23	894
Size	23.3	2.2	21.6	23.3	24.6	767
Public bank	0.1	0.3	0.0	0.0	0.0	880
Capital ratio	16.5	5.8	13.5	15.2	18.2	583
COVID	306.9	254.9	23.9	326.8	541.5	885

Notes: The table presents the summary statistics of the 896 banks in the sample. Stock returns for each bank are averaged over the period January 1, 2020 to May 5, 2020. Liquidity ratio is defined as the ratio of liquid assets (cash & due from banks) to total assets averaged over the 2019Q1-2019Q4 period. Oil exposure corresponds to the slope coefficient of an OLS regression of bank's stock returns on a constant, the market return where the bank is domiciled, and the rate of return of oil prices using weekly data between May 2018 and December 2019. Size is calculated as the 2019Q1-2019Q4 average total assets for each bank and is reported in logs. Public bank is an indicator variable that equals one for banks with a non-zero equity participation from the domestic government and zero otherwise. Capital ratio corresponds to the 2019Q1-2019Q4 average ratio of Tier 1 + Tier 2 capital to total assets. COVID corresponds to the average number of confirmed COVID-19 cases per million citizens in the country of banks during the sample period.

includes information on the daily stock prices, quarterly financial statements, and state ownership. We choose only stocks traded on major exchanges. Of the 1590 bank stocks in the data, we drop 242 stocks that traded less than 30% of the business days in each country throughout the sample period. We further drop 244 stocks that are not common stocks or are stocks with special features, such as depository receipts, real estate investment trusts, and preferred stocks. Finally, we drop from the dataset 208 banks that are owned by corporate groups whose core business is different from banking.

For several countries in our sample (i.e., oil-exporter countries) the contraction in the economy resulting from the COVID-19 shock was further exacerbated by a sharp contraction in oil prices resulting from the Russia-Saudi Arabia oil price war in early March. To control for the impact from the oil shock on banks, we use daily Brent crude oil prices and measure the degree to which individual banks were exposed to the oil sector *prior* to the COVID-19 pandemic. Specifically, we define oil sector exposure as the slope coefficient of an OLS regression of bank's stock returns on a constant, the market return where the bank is domiciled, and the rate of return of oil prices estimated using weekly data between May 2018 and December 2019.

We also obtained daily data on the total number of confirmed COVID-19 cases for every country since the beginning of the pandemic.<sup>5</sup> This information helps us measure the intensity of the pandemic over time and across countries.

Our final sample comprises 896 commercial banks. Table A1 in the Appendix presents the distribution of banks across the countries in our sample. Table 1 displays the summary statistics of the variables of interest. As Table 1 shows, the daily return of bank stocks from January to early May 2020 was on average -0.4%, with the median bank obtaining returns of -0.3%.

For each bank in the sample, we calculate the average liquidity ratio in the year prior to the COVID-19 pandemic. The ex-ante liquidity ratio corresponds to the ratio of liquid assets to total assets averaged over the 2019Q1-2019Q4 period. Our measure of liquid assets comprises cash on hand and due from banks, which include receivables from, or short-term loans to, other banks and/or financial institutions, which usually bear minor interest earnings. The liquidity ratio of the median bank in the sample is 7%, with banks in the 25th percentile holding as little as 2% of cash and due from banks relative to their total assets.<sup>6</sup>

Regarding the ex-ante exposure of banks to the oil sector, there is large heterogeneity around this measure, with the returns of banks in the 75th percentile experiencing a correlation with oil price returns of 0.23, compared to a correlation of -0.22 among banks in the 25th percentile. Bank size is calculated as the 2019Q1-2019Q4 average of the total assets for each bank and is reported in logs. The average log size of banks in the sample is 23.3, which corresponds to 13.5 billion dollars. In our sample, twelve percent of the banks are classified as public owned.<sup>7</sup> However, as Table A2 in the Appendix shows, there is large heterogeneity in the share of public banks across regions. Whereas state-owned banks in India represent 53% of all publicly traded banks in the country, in highincome countries they constitute just 2%. Regulatory capital ratios of banks (i.e., the sum of Tier 1 and Tier 2 capital relative to total assets) are averaged over the 2019Q1-2019Q4 period. The average bank in the sample entered the COVID-19 crisis with a regulatory capital ratio of 16.5%. There is also great variation on the number of COVID-19 cases per million citizens across countries in our sample. While in countries such as Nigeria or Vietnam there was on average less than 1 reported case per million from January to April, in countries such as Turkey or France the average reported cases per million exceeded 400.

#### 3. Banks risk premium during the crisis

Following the large crash of global equity prices at the onset of the COVID-19 pandemic, banks' share prices around the world underperformed relative to their local stock markets and relative to domestic non-financial firms (Fig. 1).<sup>8</sup> Declines of bank stock prices are expected during sharp economic downturns or in crisis periods due to their exposure to market risk, often measured by their high equity betas. In this section, we show banks' stock price underperformance is not explained by commonly used asset pricing models. Since we are using bank returns worldwide and banks with substantial differences in size, it is quite possible that the right asset pricing model differs across countries, or even across banks within a country.

To examine the performance of bank stocks, we use an international factor model with two factors following Bekaert et al. (2014) –a global factor and a domestic factor  $[R_t^G, R_{t,i}^D]$ . The two factors are value-weighted market indexes, so that the model potentially embeds a global CAPM, a domestic CAPM, and the mixed case when both global and domestic markets affect the pricing of a particular bank stock. To avoid spurrios correlation, the domestic market factor is value-weighted

<sup>&</sup>lt;sup>5</sup> The data on COVID-19 cases was retrieved from 'https://ourworldindata.org/coronavirus'.

<sup>&</sup>lt;sup>6</sup> While our liquidity measure omits other types of liquid assets held by banks, cash and due from banks are easier to observe and monitor, and their value is not subject to risk shifting (Bonner et al., 2015; Calomiris et al., 2015; Calomiris, 2012), an issue particularly relevant in less developed financial markets (Elliott, 2014). However, as a robustness check, we calculate two alternative definitions of bank liquidity. The first definition is the ratio of liquid assets to total deposits, where liquid assets includes cash and due from banks, federal funds sold, and securities purchased under resale agreements. The second measure consists of the ratio of liquid assets to total assets, where liquid assets include cash and due from banks, federal funds sold, securities purchased under resale agreements, trading account assets, other short-term investments, and total investment securities.

<sup>&</sup>lt;sup>7</sup> We define public banks as those with a non-zero equity participation from the domestic government. Our findings are robust to alternative definitions of public banks where the minimum threshold of state ownership is set to 10% and 30%.

<sup>&</sup>lt;sup>8</sup> Average returns of firms in Panel A of Fig. 1 are equally weighted across countries and are net from bank returns. To calculate these, we exclude the returns of banks from the returns of each country's stock market using the index weights of banks. The average returns of banks are weighted by the contribution of each bank to the total bank assets of each region. The regional average bank returns are then equally weighted across regions. For Panel B, we obtain data on all publicly traded non-bank financial companies across the 52 countries in our sample.

Table 2

Risk factors during the COVID-19 crisis.	
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	January-February	2020		March-April 2020		
Liquidity ratio measure:	(1) Liquid assets A/total assets	(2) Liquid assets B/total deposits	(3) Liquid assets C/total assets	(4) Liquid assets A/total assets	(5) Liquid assets B/total deposits	(6) Liquid assets C/total assets
Liquidity ratio	0.023	0.046	0.064	0.133*	0.146*	0.154***
	[0.103]	[0.073]	[0.040]	[0.062]	[0.077]	[0.053]
Oil exposure	0.176**	0.177**	0.158**	-0.141	-0.147	-0.119
	[0.074]	[0.074]	[0.075]	[0.107]	[0.108]	[0.093]
Size	-0.015	-0.010	0.040	0.198	0.177	0.146
	[0.113]	[0.113]	[0.132]	[0.235]	[0.234]	[0.237]
Public bank	0.037	0.041	0.015	0.161	0.179	0.169
	[0.345]	[0.348]	[0.309]	[0.228]	[0.232]	[0.234]
Capital ratio	-0.016	-0.016	-0.013	-0.011	-0.013	-0.016
	[0.011]	[0.012]	[0.014]	[0.015]	[0.016]	[0.017]
COVID	0.018	0.018	0.018	-0.054	-0.054	-0.053
	[0.017]	[0.017]	[0.017]	[0.039]	[0.039]	[0.040]
Constant	-0.024	-0.040	-0.096	-1.112***	-1.078***	-1.035***
	[0.191]	[0.198]	[0.243]	[0.250]	[0.250]	[0.275]
Observations	4,453	4,430	4,421	5,322	5,293	5,280
R-squared	0.043	0.043	0.040	0.090	0.089	0.090

Notes: The table presents OLS estimates of bank-week panel regressions. The dependent variable is abnormal returns, calculated as the difference between a bank stock returns during a week and the expected returns implied by a two-factor model with domestic and global factors. Returns are calculated in U.S. dollars. The definition of liquidity ratio, averaged over the 2019Q1-2019Q4 period, corresponds to: Liquid assets A/total assets, where liquid assets A/total assets, where liquid assets B include cash & due from banks (columns 1 and 4); liquid assets B/total deposits, where liquid assets B include cash & due from banks (columns 2 and 5); liquid assets C/total assets, where liquid assets C include cash & due from banks plus federal funds sold and securities purchased under resale agreements, trading account assets, other short-term investments and total investment securities (columns 3 and 6). Oil exposure corresponds to the slope coefficient of an OLS regression of bank's stock returns on a constant, the domestic markets, and the percentage change of oil prices using weekly data between May 2018 and December 2019. Size is calculated as the 2019Q1-2019Q4 average total assets for each bank and is reported in logs. Public bank is an indicator variable that equals one for banks with non-zero equity participation from the domestic government and zero otherwise. Capital ratio corresponds to the 2019Q1-2019Q4 average ratio of Tier 1 + Tier 2 capital to total assets. COVID corresponds to the percentage change in number of confirmed COVID-19 cases of a country per million citizens during each week. All control variables are standardized with mean 0 and standard deviation 1. All specifications include country and week fixed effects. Standard errors are clustered at the country level.

within the country, but excludes the returns of bank *b* itself,  $R_{t_i}^{D/b}$  (Table A1 presents the list of country indices used in the analysis). The model is:

$$R_{b,t} = \alpha_b + \beta_G R_t^G + \beta_U R_t^{D/b} + u_{b,t} \tag{1}$$

where all the returns *R* are measured in U.S. dollars in excess of the three-month U.S. T-Bill rate. We estimate Eq. (1) for each bank in the sample using weekly data between May 2018 and December 2019. We then calculate weekly abnormal returns between January and April 2020 as the difference between the realized and the model implied returns:  $ARet_{b,t} = R_{b,t} - \hat{\alpha}_b - \hat{\beta}_G R_t^G - \beta_U \hat{R}_t^{D/b}$ .

The empirical analysis, outlined in Eq. (2), consists of a panel regression that relates the weekly abnormal returns of banks to a series of bank characteristics. The set of covariates includes the bank's ex-ante ratio of liquid assets, *Liquidity ratio<sub>b</sub>*, its ex-ante exposure to the oil sector, *Oil exposure<sub>b</sub>*, its ex-ante size, *Size<sub>b</sub>*, an indicator variable that equals one for government-owned banks and zero otherwise, *Public bank<sub>b</sub>*, as well the bank's ex-ante regulatory capital ratio, *Capital ratio<sub>b</sub>*. We also control for the incidence of the pandemic using the percentage change of COVID-19 cases per million during the week *Covid<sub>c,t</sub>*.<sup>9</sup> The coefficients  $\gamma_c$  and  $\delta_t$  are country and week fixed effects respectively, and  $u_{b,c,t}$  is an error term clustered at the country level. Our final specification is as follows:

$$ARET_{b,c,t} = \alpha_0 + \alpha_1 Liquidity \ ratio_b + \alpha_2 Oil \ exposure_b + \alpha_3 Size_b + \alpha_4 Public \ bank_b + \alpha_5 Capital \ ratio_b + Covid_{c,t} + \gamma_c + \delta_t + u_{b,c,t}$$
(2)

The results are summarized in Table 2. In columns (1)–(3), we estimate abnormal returns for weeks during January and Febru-

ary of 2020, using alternative definitions of the liquidity ratio. The constant  $\alpha_0$  is economically small and statistically indistinguishable from zero, which suggests that banks' stock returns during the first two months of the year are broadly explained by the two-factor model. On the contrary, during March and April (columns 4–6), bank stock returns are abnormally low; according to the most conservative estimate, weekly abnormal returns were on average -1.04% points. We further confirm our findings after calculating abnormal returns relative to a six-factor model (Fama and French, 2015), where, in addition to the global and domestic markets, we add size, book to market, profitability, and investment risk factors from the Kenneth French website. Alternatively, we use a domestic CAPM where returns are estimated in local currency as a function of the domestic market returns (results for these alternative models are presented in Tables A3 and A4 in the Appendix).

The evidence also indicates that banks with lower liquidity ratios displayed the largest price drop during March and April. The coefficient for the liquidity ratio is positive and statistically significant for alternative definitions of liquid assets. For instance, a bank with a one-standard deviation liquidity ratio below the average underperformed other banks by 0.15% points per week. Other bank characteristics do not seem to correlate with the underperformance during the period. Overall, during March and April, markets were signaling an increased risk for the banking sector with an additional premium for less liquid banks.

Globally, banks entered the COVID-19 crisis better positioned to support the lending needs of the real economy. As documented by the BIS, the capital and liquidity buffers of banks at the onset of the crisis were substantially stronger than compared to the GFC (Borio, 2020; Lewrick et al., 2020). Through aggressive interventions in the financial markets, governments actively encouraged banks to continue providing credit, allowing for buffers to draw

<sup>&</sup>lt;sup>9</sup> All covariates are standardized with mean 0 and standard deviation 1.

Туре	Total	Examples		
Liquidity Support	183	Reserve requirements rate for deposits down to 25 from 31%	Brazil	21-Ma
		Establishment of buy/sell USD/INR Swap line with US Fed (Australia, Brazil, Denmark, Norway, Mexico, South Korea, Singapore, Sweden, New Zealand)	Several countries	20-Ma
		The Ministry of Finance to sell dollars during 2020 for up to US \$ 9.4 billion. Corporate bonds will be included as eligible collateral for all the effective liquidity operations in pesos	Chile	20-Ma
Prudential	333	loan deferment programs for 6 months for the financially vulnerable individuals	South Korea	11-Ma
		6 months forbearance (considering extending to 10 months) on all insured mortgages	Canada	20-Ma
		Banks countercyclical capital buffer cut to 0% starting April 1	Germany	21-Ma
		Italy: Moratorium on credits, for both companies and consumers. (March 21)	Italy	21-Ma
		Grant temporary regulatory flexibility so that banks can use their capital buffers	Mexico	8-Apr
		Reserve Bank lowered Pillar 2A capital buffer to zero	South Africa	6-Apr
Borrower Support	151	Package of \$15 billion for small business loans Ministry of Finance will guarantee up to 80% of the value of financing provided to SMEs. Until March 31, 2021 the MoF will subsidize 100% of interest on loans for micro and SMEs	Japan Romania	11-Ma 20-Ma
		Instrument aimed at limiting interest rates on loans to borrowers, a new instrument is introduced with a refinancing limit of UB 500 billion in order to maintain lending to SMEs	Russia	31-Ma
Asset Purchases	22	Large scale purchase of government bonds from the secondary market	Poland	31-Ma
		Authorized the purchase of government bonds in the secondary market for up to \$ 2 billion	Colombia	14-Ap
	05	The Fed announced QE purchases of USD500bn of Treasuries and USD200bn of agency MBS	United States	20-Ma
Policy Rates	95	Cut its benchmark interest rate by 25 basis point, taking it to a record low 1%	Thailand	5-Feb
		On 2/20, Bl cut its seven-day reverse repo rate by 25 basis points to 4.75%, marking the first cut in the Bl policy rate since October	Indonesia	20-Feb
		Benchmark rate cut to 9.75% from 10.75%	Turkey	21-Ma

down. In the next section, we study the immediate market response to these policy announcements.

# 4. Measuring the market response to financial sector distress policy initiatives

What impact did the different financial sector policy measures have on the banking sector? To answer this question, in this section we evaluate the market response to different financial sector policy initiatives.

#### 4.1. Policy interventions

We use data on the dates and types of major policy initiatives to support the financial sector and address the impact of the COVID-19 emergency. The data were compiled and made publicly available by the World Bank (Feyen et al., 2021). It covers policy measures in low, middle, and high-income countries. Information on each policy measure was collected from national authorities and international organizations. For each policy initiative, the dataset reports details on the announcement, and the day (and not the hour or minute) of the announcement which restricts our analysis to daily frequency. We classify approved measures that target the banking sector into the following categories: (i) liquidity support, (ii) prudential measures, (iii) borrower assistance, and (iv) monetary policy (Table 3). Liquidity support deals with potential shortages in funding, either because precautionary demand for liquidity has increased or because of lack of access to funding (e.g., wholesale funding markets have dried up). Measures included are provisions of domestic currency liquidity through broadened access to central bank refinancing, extended collateral framework, such as those in the shortterm and long-term repo market, and available foreign currency liquidity through swap agreements between central banks.

*Prudential measures* include temporary relaxation of certain key regulatory and supervisory requirements, including changes in capital requirements, limits on exposure, concentration, loan-to-value ratios, minimum reserve requirements, and cancelation of stress tests. By slowing down the decline in banks' regulatory capital ratios, these measures reduce the rate at which banks draw buffers down, spreading the recognition of losses and allowing a given amount of equity to support a larger lending volume.

*Borrower assistance* include measures to supply funds to firms and households due to the loss of revenue/income from the extended lockdowns. These include government liability guarantees for newly issued or existing wholesale financing, direct credit lines to strategic sectors, state support for interest-free loans and fully replenishable working capital financing. Other measures include enhancement of deposit protection schemes and simplified programs of loan restructuring.

Monetary policy are interest rate decisions (e.g., benchmark rate, repo rates, etc.), central bank purchases of government securities

Country	First date	Last date	Liquidity	Prudential	Borrower Support	Asset Purchases	Policy Rates
Argentina	5-Mar	5-Apr	1	1	2	0	1
Australia	1-Mar	17-Apr	2	6	0	0	2
Austria	12-Mar	16-Apr	3	7	4	2	0
Brazil	2-Mar	16-Apr	7	5	3	0	1
Bulgaria	19-Mar	15-Apr	1	2	4	0	0
Canada	4-Mar	15-Apr	0	3	0	2	3
Chile	20-Mar	20-Mar	1	0	1	0	1
China	20-Feb	15-Apr	2	1	2	0	3
Colombia	20-Mar	16-Apr	3	1	1	2	1
Croatia	20-Mar	31-Mar	1	0	1	0	0
Egypt	24-Mar	16-Apr	0	2	1	0	0
Finland	12-Mar	16-Apr	3	7	4	2	0
France	12-Mar	16-Apr	3	10	6	2	0
Germany	12-Mar	16-Apr	3	9	6	2	0
Hong-Kong	4-Mar	18-Apr	0	3	1	0	2
India	20-Mar	17-Apr	5	2	1	0	2
Indonesia	20-Feb	14-Apr	2	1	0	1	2
Ireland	12-Mar	16-Apr	3	7	4	2	0
Italy	1-Mar	16-Apr	3	9	7	2	0
Japan	20-Feb	15-Apr	0	2	2	0	1
Malaysia	3-Mar	27-Mar	2	1	1	0	1
Mexico	12-Mar	15-Apr	3	4	1	0	1
Morocco	17-Mar	27-Mar	1	1	0	0	1
Netherlands	12-Mar	16-Apr	3	7	4	2	0
Nigeria	18-Mar	20-Mar	1	1	2	0	0
Pakistan	17-Mar	16-Apr	1	1	1	0	3
Peru	19-Mar	17-Apr	2	2	3	0	1
Philippines	20-Mar	16-Apr	2	0	0	1	2
Poland	20-Mar	8-Apr	1	1	1	1	2
Portugal	12-Mar	16-Apr	3	7	4	2	0
Qatar	10-Mar	10-Mar	0	0	0	0	1
Romania	20-Mar	9-Apr	2	2	1	1	1
Russia	7-Feb	17-Apr	1	4	1	0	1
Saudi-Arabia	10-Mar	16-Mar	0	0	0	0	2
Singapore	16-Apr	16-Apr	0	1	0	0	0
South-Africa	19-Mar	14-Apr	2	5	1	1	2
South-Korea	11-Mar	13-Apr	2	6	3	0	1
Spain	12-Mar	16-Apr	3	8	5	2	0
Sweden	6-Apr	14-Apr	2	0	0	0	0
Thailand	5-Feb	20-Mar	0	1	0	0	2
Turkey	17-Mar	17-Apr	2	1	1	1	1
UAE	10-Mar	10-Mar	0	0	0	0	1
UK	11-Mar	17-Apr	2	4	6	1	2
US	15-Mar	17-Apr	5	6	2	1	1
Vietnam	17-Mar	17-Apr	0	0	2	0	1

Number of days with policy announcements.

Notes: Data obtained from the World Bank (2020) covering the period February 1 to April 17 2020.

(*Quantitative Easing*), and credit easing, which typically consist of purchases of private sector debt in secondary markets. Given the different nature of conventional and unconventional monetary policy measures, we further split this category into *Policy Rate* announcements and *Asset Purchases* (both public and private bond buying programs and purchases of asset-backed securities).

Our final sample contains policies in 17 developed and 28 developing countries, for which we have stock market and policy data (Table 4). Some European countries such as, Austria, Finland, and Netherlands, did not have domestic financial policy initiatives. We include banks of these countries in days of announcements by the European Central Bank. Other Euro countries, for example, France, Germany, Italy, and Spain have both financial support policies taken at the country level in addition to those taken by the monetary union authorities.<sup>10</sup> In total, the database includes 389 announcements between February 1 and April 17.<sup>11</sup> Prudential

measures accounted for the largest share of policy initiatives (36%), followed by borrower support, liquidity provision measures, monetary policy and asset purchases (23%, 21%, 12% and 8%, respectively). However, the type of policies adopted differed substantially across countries (see Table A3 in the Appendix). While monetary policy announcements were frequent in developing countries (representing 22% of all measures), they only accounted for 4% in developed ones. In contrast, prudential measures in developed countries constituted 43% of all announcements, compared to 26% in developing countries.

Fig. 2 presents the timing of policy announcements. Early in February, financial markets and financial authorities had little reaction to the original outbreak in China and the ensuing lockdown of Wuhan. On February 20, when Italy announced the quarantine of eleven municipalities in the northern region, and once it was apparent that the outbreak had also spread to South Korea, and Iran, stock markets declined sharply. By this date, only five countries had taken financial sector policy interventions. Among these, Russia, Thailand, China and Indonesia lowered their short-term interest rates and Japan launched a crisis loan package to small businesses.

<sup>&</sup>lt;sup>10</sup> We also include eight policy announcements that were made during weekends. For these cases, we study the stock response using the first business day when the domestic stock market opened.

<sup>&</sup>lt;sup>11</sup> In most cases and for each policy category, the first time that a country announced the adoption of a financial sector measure was prior to April 17; 94% of first-time announcements in the sample (between Jan and Sep 2020).

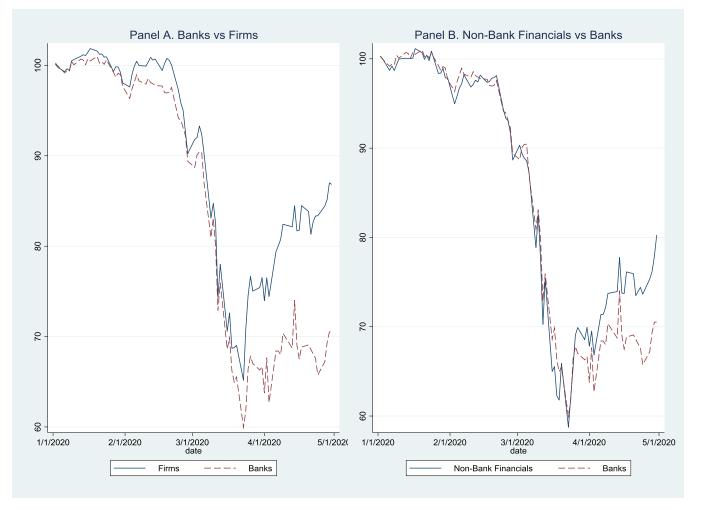


Fig.1. Average stock returns of banks vs firms and non-bank financial companies

Notes: The figures plot average daily stock market returns of banks, firms and non-bank financials in the sample normalized to January 1, 2020. The average returns of firms in the first figure are equally weighted across countries and are net from bank returns. To calculate them, we exclude the returns of banks from the returns of each country's stock market using the index weights of banks. The average returns of banks are weighted by the contribution of each bank to the total bank assets of each region. The regional average bank returns are then equally weighted across regions. The same approach is used to obtain the average return of non-bank financials.

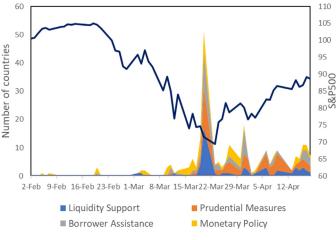
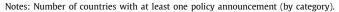


Fig.2. Timeline of policy announcements



Following a myriad of national quarantines starting with Italy on March 10, travel bans, including the U.S. decision on March 12 to severely restrict travel from the EU, stock markets around the world declined more than 30% from their peak. Interestingly, at this stage of the global crisis, most financial authorities were only taking conventional monetary policy actions, reducing short-term rates. For example, between February 20 and March 18, seventeen countries in our sample had announced policy rate cuts, and only a few had either taken measures to assist borrowers or announced prudential measures (e.g., Italy, Germany, U.K., and South Korea).

In response to the unprecedented, and rapidly evolving situation, policymakers increased the scope and number of measures to support the financial system. After March 18, most government and monetary authorities included liquidity support measures and borrower assistance. For instance, by April 1735 countries had taken some measure of direct borrower support, in most cases, dealing with SME financing. The use of countercyclical prudential rules also became ubiquitous; by the end of our sample, 40 countries had introduced prudential actions, typically through a temporary relaxation of certain key regulatory and supervisory requirements.

A large number of interventions were announced simultaneously with other financial sector policies. For example, prudential measures in developed countries were often announced on the same day of borrower support program announcements (Table A3 in the Appendix). However, as Tables A3 shows, there were also multiple single policy announcements for each of the policy categories we study. In developed countries, 42% of policies were announced on days with no other financial sector program announcements. This share is higher among developing countries, where 51% of policies were made on single announcement days.

#### 4.2. Empirical methodology

We use an event study technique (Brown and Warner, 1996; Kothari and Warner, 2007) to test the effect of policy announcements on the stock returns of banks. More specifically, we want to assess whether different policy initiatives convey new information about the ability of banks to properly operate during and after the crisis. Measuring bank-level stock returns instead of an aggregate index of bank stock prices (or interbank credit and liquidity risk premia), allows us to exploit variation across banks to better understand the market response to different types of policies.<sup>12</sup> For example, the cross-sectional analysis allows us to answer whether a specific type of government support measure favored a particular group of banks. Our focus is on the reaction of stock prices, as opposed to bond yields, because stocks are more frequently traded than bonds. Also, we do not use other bank-specific risk measures such as credit default swaps spreads, because these securities are not widely available for the majority of banks in developing countries.

To assess the stock market reaction to a policy announcement, we compute the accumulated abnormal stock returns of banks during the event. To be precise, we calculate accumulated abnormal returns over an event window *n*, which is the variation in the stock price one day prior to the announcement on day *t*, to *n* days after in excess to the expected returns implied by the mixed model with domestic and global factors ( $ARet_{ht}^n$ ).

After calculating bank-level abnormal returns, we test whether policy announcements have significant effects on bank stock returns. In our benchmark specification, we test this hypothesis by estimating the following equation:

$$ARet_{b,c,t}^{n} = \alpha_{0,n} + \alpha_{1,n}Liquidity \ ratio_{b} \\ + \alpha_{2,n}Oil \ exposure_{b} + \alpha_{3,n}Size_{b} \\ + \alpha_{4,n}Public \ bank_{b} + \alpha_{5,n}Capital \ ratio_{c,t} \\ + \alpha_{6,n}Covid_{c,t} + \gamma_{c} + \gamma_{t} + u_{b,c,t}$$
(3)

where  $ARet_{b,c,t}^n$  represents the abnormal return of bank *b* located in country *c* in day *t*, measured over the [-1, *n*] window (in days). As in the previous section, the covariates include a bank's ex-ante: liquidity ratio (measured as the ratio of cash & due from banks to total assets), oil exposure, size, and regulatory capital ratio; as well as an indicator variable for state-owned banks, and the percentage change in the number of COVID-19 cases per million on day t in each country. Finally,  $\gamma_c$  and  $\gamma_t$  are country and announcement date fixed effects. Standard errors are clustered at the country level.

We estimate Eq. (3) separately for each policy category. Since the variables are standardized (except for *Public bank*<sub>b</sub>), the constant  $\alpha_{0,n}$  captures the abnormal stock returns of the average private bank for a type of announcement in the [-1, *n*] window. The coefficients  $\alpha_{1,n}$ ,  $\alpha_{2,n}$ ,  $\alpha_{3,n}$ ,  $\alpha_{4,n}$  and  $\alpha_{5,n}$  capture cross-sectional differences in the response of bank stock prices. For example,  $\alpha_{1,n} > 0$  would suggest that banks with a larger share of liquid assets have higher accumulated abnormal returns during the event window for a specific type of announcement.

There are several challenges to our identification. First, owing to the high degree of global financial integration, there is extensive evidence of international spillovers from policy announcement by systemically important countries (Ait-Sahalia et al., 2012;

<sup>12</sup> Correa et al. (2014) also use bank stock prices to evaluate another kind of shock – the downgrade of the sovereign credit risk.

Morais et al., 2019). To avoid contaminating the analysis from the effects of financial policy initiatives taken by core countries, we exclude data from days when the U.S. or European authorities made major announcements covering the same broad policy type category. For example, on a day when the U.S. authorities announced liquidity support measures, we keep data from stock returns of U.S. banks in the sample but exclude banks in countries that issue liquidity announcement on the same day.

Second, due to the nature and scale of the crisis, in some cases, national government authorities announced multiple financial measures during the same day. In such situations, and without intraday time stamp for each announcement category, it is difficult to disentangle the effect of each policy on bank stock prices. To deal with these confounding factors, we further condition our sample to country-dates where only policies of a particular category were announced, and report results for this restricted sample.

Finally, selecting the length of the event window has important tradeoffs. Limiting the size of the window helps to avoid contaminating the analysis of a given announcement's effect with other news; particularly in a period when pandemic-related news was heightened. If the event window is too narrow, there might not be enough time for market participants to internalize the context and implications of complex policy announcements. This is likely to be more pronounced in developing countries, where trading volume in secondary markets is small and the speed of transactions is slower (e.g., the pace of business at which risk is transferred across market participants is longer). In turn, we use a five-day window, evaluating the change in stock prices from one day before the announcement to three days after, that is,  $n = \{0,1,2,3\}$ . We report accumulated abnormal returns for every day during this time window.

#### 4.3. Impact of policy announcements

As a benchmark to gauge the magnitude of the stock market effects, we first plot the average abnormal returns during the event window for each policy category, pooling banks across all countries. Abnormal returns are obtained from estimating Eq. (3) on a constant, removing all other covariates and using day and country fixed effects. The results are displayed in Fig. 3 for the full sample (Panel A) and the restricted sample (Panel B) – when announcements of a particular category do not overlap with other initiatives. During our sample period, borrower assistance initiatives were strongly associated with large increases in the abnormal returns of bank stocks during the announcement day; 146 bps and 199 bps in the full and restricted sample respectively. We find that these large excess returns are present up to three days after the announcement. Borrower assistance initiatives typically include the introduction of government guarantees, which automatically transfer risks from banks' balance sheets to the sovereign.

On days when prudential measures and policy rate reductions were announced, bank stocks display positive but small abnormal returns in the full sample (45bps and 39bps, respectively). These returns are quickly reversed within two days after the policy initiatives were announced. Notably, when we restrict the sample to single policy announcements, prudential measures seem to be accompanied by immediate price drops in bank stocks; 128 bps during the announcement day.<sup>13</sup> Prudential policies allow banks to run down some of their buffers to absorb the shock and send a strong signal about the resolve of policymakers. However, the fact that bank stock returns are negative might suggest that markets

<sup>&</sup>lt;sup>13</sup> As Table A5 shows, in developed countries, 52 of the 99 prudential policy measures were jointly announced with borrower assistance initiatives. This might explain why in the full sample the net effect of prudential policies is higher than in the restricted sample.

# Panel B. Restricted sample

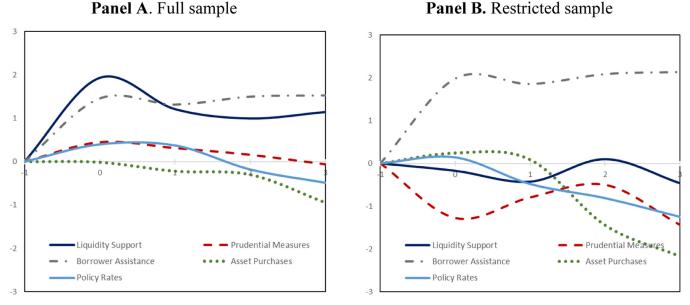


Fig.3. Abnormal returns of bank stocks around the announcement window

Notes: The variable plotted on the vertical axis shows the accumulated abnormal returns in percentage points within the event window of one day before the event and three days after the event, scaled to zero on the day before the announcement. Accumulated abnormal returns are averaged across banks for each policy category. The horizontal axis shows days within the event window, with "0" corresponding to the day of the announcement.

are also pricing the downside risk from the depletion of capital buffers.

Finally, we do not find evidence that liquidity support announcements had any aggregate short-term effects on bank stock prices. Although bank stocks seem to display positive abnormal returns when liquidity assistance measures are combined with other policies (Panel A in Fig. 4), in the restricted sample, abnormal returns are small and statistically indistinguishable from zero.

The results on the cross-sectional impact of financial sector policies are presented in Table 5 and Fig. 4.14 While we present findings for the full and restricted sample, our preferred specification and discussion focuses on single policy initiatives. The constant in each model represents the average abnormal returns of private banks in the sample. The results confirm our previous finding that borrower assistance measures are strongly associated with banks' stock price response. Other policy initiatives seem to have smaller or even negative aggregate effects on bank stocks.

#### 4.3.1. Liquidity support

Announcements of liquidity support are associated with a reduction in the liquidity premium. That is, stocks of banks with a smaller share of liquid assets overperform after policymakers announce liquidity and funding measures. The coefficient for the liquidity ratio can be read as follows: a bank with a liquidity ratio measure of one-standard deviation lower than the average, experienced an additional 210 bps of abnormal returns up to three days after the government announced liquidity support policies. The result is important because it confirms that central banks' actions helped alleviate the sharp tightening of financial conditions at the onset of the crisis. During March, due to the overwhelming volatility in securities and FX markets, many financial entities reported having difficulties in accessing funding, which is consistent with the sharp increase in the liquidity premium documented in Section 3. Overall, policies targeting funding availability (whether in domestic or foreign currency) appear to have a favorable impact in the reduction of the liquidity premium. We also see that stock prices of public banks and those of smaller banks benefit more from liquidity measures, potentially due to greater reductions in uncertainty these policies provide to such banks.

#### 4.3.2. Prudential measures

In addition to having a negative impact on aggregate bank stock prices, prudential measures do not seem to be associated with clear reductions in the liquidity premium. The use of buffers, facilitation of loan restructuring, and other measures that require regulatory forbearance (e.g., loan moratoria and different treatment of non-performing loans), might help support credit flow throughout the lockdowns, but they entail large risks in the medium term, threatening financial stability. It appears that markets are pricing these risks, since abnormal returns are negative around the announcement of countercyclical prudential measures.

#### 4.3.3. Borrower assistance

Announcements of borrower assistance were associated with large increases in abnormal returns of bank stocks. Recognizing the severity of the shock, especially to small non-essential businesses that would be unable to cover operating costs during extended lockdowns, many countries enhanced their public liability guarantees programs, with government guarantees up to 90% of loan values. Fiscal resources were also committed to subsidize interestfree loans and directly fund strategic sector during the pandemic. These initiatives generate large fiscal costs in the short term, while transferring the credit risks from banks to the government. It appears that markets internalize this information as good news for the banking industry, which could now offload part of the burden of the shock to the sovereign.

Announcements of borrower support programs had a greater impact among larger banks. For instance, a bank that is onestandard deviation above the mean in size, experiences an additional 109bps of abnormal returns up to three days after the announcement. Ornelas et al. (2019) document that during the GFC, large banks disproportionally benefited from allocating government-sponsored credit. While loans that are funded with government resources typically have interest rate ceilings, banks

<sup>&</sup>lt;sup>14</sup> To present all the policies in a single table, we only include same-day estimates (n = 0) and three days after the announcement (n = 3). However, the figures display estimates during the entire event window.

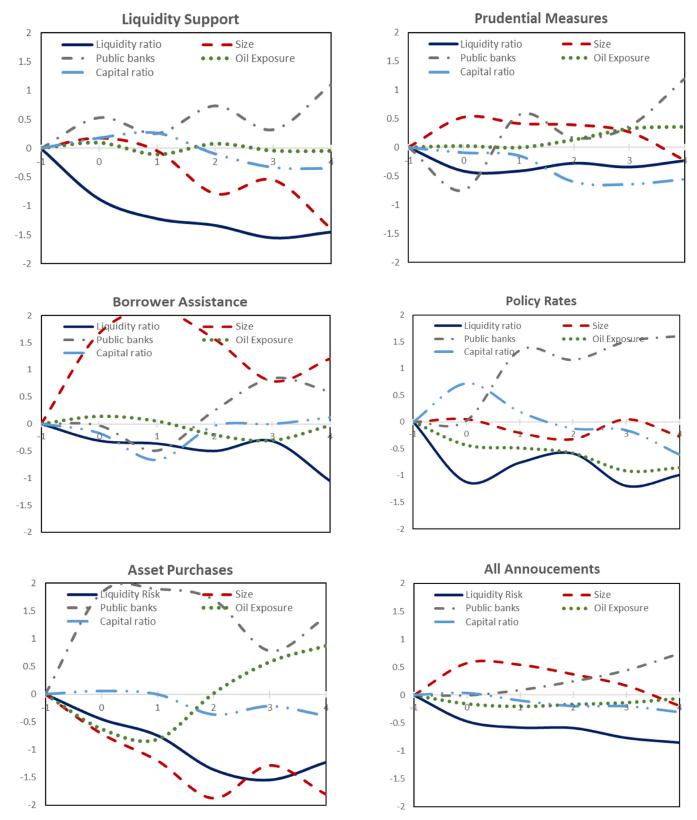


Fig. 4. Cross-sectional impact of policy announcements (in percentage points)

Notes: Estimated slope coefficients of OLS regressions estimating daily accumulated abnormal returns on bank characteristics. Liquidity ratio, Oil exposure, Size and Capital ratio are standardized. "Public banks" is a dummy variable equal to 1 if the bank is state-owned and zero otherwise. The horizontal axis shows days within the event window, with "0" corresponding to the day of the announcement.

10

Impact of policy initiatives.

	Liquidity S	upport	Prudential	Meas.	Borrower	Assist.	Asset Pur	rchases	Policy Rate	s	All announ	cements
<i>n</i> =	0	3	0	3	0	3	0	3	0	3	0	3
Panel A. Full sa	mple: days v	with multiple	e domestic p	olicy annou	ncements							
Liquidity ratio	-1.001***	-1.683***	-0.491*	-0.402	-0.386	-0.385	-0.437	-1.828*	-1.130***	-1.054	-0.534***	-0.845***
	[0.349]	[0.588]	[0.285]	[0.467]	[0.347]	[0.664]	[0.501]	[0.995]	[0.370]	[0.643]	[0.187]	[0.328]
Oil exposure	0.056	-0.084	0.001	0.281	0.123	-0.346	-0.622	0.480	-0.434**	-0.899***	-0.172	-0.157
	[0.189]	[0.323]	[0.157]	[0.262]	[0.221]	[0.443]	[0.425]	[0.847]	[0.187]	[0.318]	[0.106]	[0.189]
Size	0.135	-0.611	0.512***	0.211	1.664***	0.734	-0.720	-1.175	0.060	-0.042	0.564***	0.130
	[0.222]	[0.377]	[0.170]	[0.281]	[0.253]	[0.483]	[0.465]	[0.929]	[0.280]	[0.470]	[0.124]	[0.218]
Public bank	0.674	0.426	-0.695	0.382	-0.009	0.868	1.848*	0.509	0.026	1.496*	0.030	0.479
	[0.591]	[0.997]	[0.620]	[1.019]	[0.814]	[1.529]	[1.014]	[2.019]	[0.502]	[0.850]	[0.350]	[0.609]
Capital ratio	0.205	-0.211	-0.048	-0.540	-0.080	0.091	0.049	0.097	0.727***	-0.138	0.074	-0.069
	[0.232]	[0.401]	[0.205]	[0.343]	[0.294]	[0.656]	[0.314]	[0.628]	[0.253]	[0.439]	[0.141]	[0.255]
Constant	1.553***	0.291	0.559***	-0.332	2.107***	1.975***	-0.254	-0.382	0.597**	-1.244***	0.693***	-0.191
	[0.223]	[0.376]	[0.204]	[0.333]	[0.209]	[0.418]	[0.408]	[0.830]	[0.261]	[0.445]	[0.110]	[0.192]
Observations	717	706	1,142	1,123	367	356	100	98	333	329	1,953	1,918
R-squared	0.296	0.505	0.370	0.488	0.450	0.233	0.192	0.270	0.275	0.230	0.349	0.384
Panel B. Restric	cted sample:	days with si	ngle domest	ic policy and	nouncement	S						
Liquidity ratio	-0.816	-2.101**	0.096	-0.030	-0.274	-0.389	-0.038	0.705	-0.746*	-1.409**	-0.365*	-0.902**
	[0.494]	[1.013]	[0.366]	[0.605]	[0.465]	[0.846]	[0.892]	[1.491]	[0.380]	[0.592]	[0.207]	[0.370]
Oil exposure	-0.398	-1.283**	-0.194	0.347	0.127	0.040	-0.527	-0.585	-0.580***	-0.742**	-0.376***	-0.415*
	[0.290]	[0.591]	[0.227]	[0.380]	[0.272]	[0.534]	[0.867]	[1.458]	[0.192]	[0.288]	[0.121]	[0.219]
Size	-0.764*	-1.565*	0.662***	0.630	1.807***	1.090**	-0.255	-4.521**	0.322	0.371	0.646***	0.257
	[0.427]	[0.874]	[0.231]	[0.385]	[0.294]	[0.529]	[1.211]	[2.032]	[0.291]	[0.431]	[0.144]	[0.256]
Public bank	1.522**	0.523	-1.041	-1.275	-0.164	-1.224	0.842	3.461	0.510	1.633*	0.172	0.257
	[0.740]	[1.508]	[1.033]	[1.705]	[1.356]	[2.409]	[1.861]	[3.110]	[0.570]	[0.846]	[0.392]	[0.693]
Capital ratio	-0.014	1.254	-0.856**	-0.574	0.120	0.254	0.349	1.050	0.389	0.581	-0.088	0.247
	[0.445]	[0.910]	[0.336]	[0.558]	[0.346]	[0.758]	[1.347]	[2.279]	[0.266]	[0.413]	[0.168]	[0.314]
Constant	-0.393	0.324	-1.266***	-1.601***	3.476***	3.269***	-0.296	-0.587	-0.032	-1.826***	-0.046	-0.709**
	[0.405]	[0.830]	[0.262]	[0.431]	[0.321]	[0.614]	[0.893]	[1.529]	[0.300]	[0.452]	[0.118]	[0.211]
Observations	172	171	557	548	270	259	39	38	244	240	1296	1270
R-squared	0.277	0.385	0.289	0.425	0.341	0.188	0.218	0.469	0.332	0.243	0.378	0.333

Notes: The table presents OLS estimates of cross-sectional regressions of the impact of financial sector policies on the abnormal returns of banks on announcement days (n=0) and three days after the announcement (n=3). Panel A includes all announcements of a particular category in the sample. Panel B restricts the sample to announcements of a particular category that do not overlap with announcements of other financial sector initiatives. Abnormal returns are calculated as the difference between realized returns and the expected returns implied by a market model. Liquidity ratio is defined as the ratio of liquid assets (cash & due from banks) to total assets averaged over the 2019Q1-2019Q4 period. Oil exposure corresponds to the slope coefficient of an OLS regression of bank's stock returns on a constant, the market return where the bank is domiciled, and the rate of return of oil prices using weekly data between May 2018 and December 2019. Size is calculated as the 2019Q1-2019Q4 average total assets for each bank and is reported in logs. Public bank is an indicator variable that equals one for banks with a non-zero equity participation from the domestic government and zero otherwise. Capital ratio corresponds to the 2019Q1-2019Q4 average ratio of Tier 1 + Tier 2 capital to total assets. All control variables are standardized with mean 0 and standard deviation 1. All specifications include day and country fixed effects and control for the daily number

offset the lower revenue from these loans by extracting rents in other lending products to the recipients of government support. Smaller banks and those that operate in niche markets are less able to take advantage of cross-selling and enhanced sales in other products.

Consistent with the view that credit guarantees, and other state-funded lending programs reduce bank risk, we find that announcements of borrower assistance benefit banks with a higher share of lending portfolios prior to the COVID-19 crisis, as measured by their 2019Q1-2019Q4 average loan-to-asset ratios.<sup>15</sup>

When we split the sample into developed and developing countries, we find that the positive association between abnormal stock returns and borrower support initiatives is exclusive to developed countries (see Table 6 and Fig. 5). As noted above, borrower assistance requires potentially large fiscal commitments. In developing countries, authorities have lower fiscal space to operate, and markets might anticipate that the extent of the borrower assistance measures in such settings is limited.<sup>16</sup> On the contrary, fiscal authorities in developed countries can pledge significant resources to help distressed households and firms, which in turn reduce bank risk.

# 4.3.4. Monetary policy

Interest rate cuts were associated with significant declines in the liquidity premium. To be precise, during days of policy rate cuts, banks with lower liquidity experienced higher abnormal returns than more liquid banks (full and restricted sample in Table 5). The significant relative increase in abnormal stock returns among banks with lower liquidity ratios may have reflected markets' expectation that lower interest rates would increase liquidity in the financial system, thereby benefiting banks with larger funding risks. The result also confirms that the interest rate policy remained a key policy tool at the onset of the crisis, since markets are familiar with conventional monetary policy.

Banks with more exposure to the oil shock seem to benefit less from interest rate actions, as evidenced in both the full and restricted sample. The result highlights the key difference between liquidity and solvency risks. Banks that are heavily exposed to the oil sector are more susceptible to losses due to the likelihood of defaults in this industry. To the extent that markets internalize that the profitability of these banks will be lower, interest rate cuts would have small effects. On the other hand, banks with shortterm liquidity constraints, but with less exposed portfolios, stand to benefit more from the ease in funding conditions.

<sup>&</sup>lt;sup>15</sup> These unreported results can be shared upon request.

<sup>&</sup>lt;sup>16</sup> Fiscal expansion in developing economies is more costly, as these countries have a more difficult time raising money in the market (Kose et al., 2017) and the debt threshold at which they experience growth rate decline is lower (Reinhart and Rogoff, 2010).

Heterogeneity in the impact of policy initiatives (developed vs. developing countries).

	Liquidity	Support	Prudential	Meas.	Borrower	Assist.	Asset Purcha	ases	Policy Rate	S	All announ	cements
<i>n</i> =	0	3	0	3	0	3	0	3	0	3	0	3
Panel A. Develo	ped count	ries										
Liquidity ratio	-0.848	-1.431	0.175	-0.094	-0.232	-0.371	-0.374	0.480	-1.586	-8.186	-0.142	-0.593
	[0.523]	[1.033]	[0.418]	[0.691]	[0.486]	[0.872]	[1.002]	[1.773]	[3.846]	[5.948]	[0.271]	[0.475]
Oil exposure	-0.173	0.586	-0.154	0.409	0.102	-0.002	-1.899	-1.824	1.348	-11.363**	-0.098	0.126
	[0.833]	[1.647]	[0.236]	[0.395]	[0.278]	[0.536]	[1.173]	[2.111]	[2.785]	[4.308]	[0.173]	[0.313]
Size	-0.871	-2.755*	0.739***	0.762*	1.836***	1.155**	-1.025	-4.554*	0.357	-1.284	1.036***	0.794***
	[0.692]	[1.368]	[0.238]	[0.396]	[0.300]	[0.533]	[1.473]	[2.607]	[2.158]	[3.338]	[0.174]	[0.304]
Public bank	1.802	3.415	-1.250	-1.988	-0.171	-1.230	2.779	2.666	-2.338	24.902**	-0.567	-1.362
	[2.076]	[4.104]	[1.224]	[2.022]	[1.376]	[2.413]	[2.414]	[4.273]	[6.400]	[9.899]	[0.792]	[1.372]
Capital ratio	-0.375	0.164	-0.715**	-0.285	0.121	0.264	1.352	0.605	3.686	-3.334	-0.312	-0.062
	[0.860]	[1.700]	[0.361]	[0.600]	[0.357]	[0.775]	[1.774]	[3.204]	[5.126]	[7.929]	[0.239]	[0.452]
Constant	0.645	1.353	-1.314***	-1.633***	3.681***	3.396***	0.379	-0.318	1.362	-5.470	0.217	-0.492
	[0.710]	[1.404]	[0.314]	[0.516]	[0.335]	[0.621]	[1.235]	[2.261]	[2.216]	[3.427]	[0.189]	[0.336]
Observations	38	38	522	513	257	249	29	28	23	23	876	858
R-squared	0.356	0.410	0.280	0.428	0.331	0.187	0.249	0.472	0.610	0.694	0.439	0.390
Panel B. Develo	ping coun	tries										
Liquidity ratio	0.888	3.483**	0.175	-0.717	-0.280	-1.472	2.368***	1.963	0.639*	0.772	0.529*	1.131*
	[0.855]	[1.750]	[0.582]	[1.000]	[1.165]	[2.130]	[0.202]	[4.209]	[0.370]	[0.528]	[0.299]	[0.576]
Oil exposure	-0.420	-1.615**	-1.473	-0.330	-0.245	1.761	3.633***	4.267	-0.519***	-0.490*	-0.486***	-0.765**
	[0.325]	[0.658]	[0.926]	[1.592]	[0.885]	[1.891]	[0.101]	[2.107]	[0.188]	[0.261]	[0.159]	[0.302]
Size	-0.628	-1.057	-2.055*	-2.781	-1.373	-3.622	4.289***	1.382	-0.063	-0.618	-0.404	-1.064**
	[0.546]	[1.109]	[1.131]	[1.945]	[1.259]	[2.443]	[0.148]	[3.080]	[0.315]	[0.432]	[0.261]	[0.492]
Public bank	1.271	0.616	-0.221	1.762	1.892	14.590*	-10.345***	-4.469	0.770	2.198***	0.920**	1.350*
	[0.799]	[1.613]	[1.692]	[2.910]	[4.063]	[7.346]	[0.383]	[7.958]	[0.559]	[0.769]	[0.424]	[0.801]
Capital ratio	0.166	1.111	-0.974	-1.854	-1.060	-2.162	6.978***	8.678	0.327	0.248	0.114	0.355
	[0.519]	[1.050]	[0.773]	[1.330]	[1.054]	[5.184]	[0.228]	[4.742]	[0.250]	[0.355]	[0.217]	[0.417]
Constant	-0.488	0.342	1.755	-0.326	-1.038	-5.201	4.960***	5.592	-0.129	-1.494***	-0.186	-0.681
	[0.498]	[1.011]	[1.039]	[1.786]	[1.722]	[3.208]	[0.316]	[6.565]	[0.282]	[0.395]	[0.230]	[0.439]
Observations	134	133	35	35	33	31	10	10	221	217	418	410
R-squared	0.262	0.395	0.464	0.464	0.283	0.466	0.999	0.895	0.249	0.226	0.240	0.213

Notes: The table presents OLS estimates of cross-sectional regressions of the impact of financial sector policies on the abnormal returns of banks on announcement days (n=0) and three days after the announcement (n=3) for the restricted sample of announcements of a particular category that do not overlap with announcements of other financial sector initiatives. Panel A restricts the sample to developed countries. Panel B restricts the sample to developing countries. Abnormal returns are calculated as the difference between realized returns and the expected returns implied by a market model. Liquidity ratio is defined as the ratio of liquid assets (cash & due from banks) to total assets averaged over the 2019Q1-2019Q4 period. Oil exposure corresponds to the slope coefficient of an OLS regression of bank's stock returns on a constant, the market return where the bank is domiciled, and the rate of return of oil prices using weekly data between May 2018 and December 2019. Size is calculated as the 2019Q1-2019Q4 average total assets for each bank and is reported in logs. Public bank is an indicator variable that equals one for banks with a non-zero equity participation from the domestic government and zero otherwise. Capital ratio corresponds to the 2019Q1-2019Q4 average ratio of Tier 1 + Tier 2 capital to total assets. All control variables are standardized with mean 0 and standard deviation 1. All specifications include day and country fixed effects and control for the daily number of COVID-19 cases in a country. Standard errors are clustered at the country level.

Announcement concerning asset purchases also appear to be followed by increases in the abnormal returns of less liquid banks, albeit the coefficient is statistically indistinguishable from zero in both the full and restricted sample. We interpret this finding as complementary evidence that unconventional monetary policy also played a role in reducing the liquidity premium.

Since announcements should have a greater impact if they are unexpected than if they are expected, we further restrict our sample to days with monetary policy surprises. That is, we restrict our sample to days when the monetary authority made policy rate announcements after emergency meetings, or when the domestic policy rate cut was larger than the expected by market analysts.<sup>17</sup> We further confirm earlier findings with the sample of unexpected monetary policy announcements. Although surprises of interest rate cuts were not associated with significant increases in bank stock prices, such announcements did appear to have differential effect across banks. For instance, monetary policy surprises were associated with declines in the liquidity premium and seemed to be more beneficial for banks with lower exposure to the oil sector (Table A6 in the Appendix).

# 5. Endogeneity

Notwithstanding the robustness of the results, possible endogeneity issues remain a key challenge for our identification strategy. In particular, while the COVID-19 shock is exogenous, the policy mix adopted in a country is not random. To the extent that selected policies are precisely those that the financial authorities deemed more efficient for their jurisdictions, taking into account local economic conditions, banks' characteristics and banks' stock performance, our estimates would be biased. In this section, we present three exercises to address such endogeneity concerns. First, we examine the reaction of stock prices of large cross-border banks from policy announcements in countries where their subsidiaries make a significant share of their balance sheet. Second, we examine the extent to which domestic and external factors influence the announcement of financial sector policies. Third, we analyze the stock price response of banks in the euro area to policy announcements made by the ECB.

# 5.1. Cross-border banks

We focus on large cross-border banks headquartered outside a given jurisdiction but with exposure through a locally operating

<sup>&</sup>lt;sup>17</sup> We use the Reuters forecast based on the one-week analyst survey of benchmark rates. Some examples of monetary policy surprises include Thailand's 25bps cut in the benchmark rate (February 5), Canada's repo rate cut of 50bps (March 4), and the United Kingdom 50bps repo rate cut during the emergency meeting of March 11. The only case in which the rate cut was smaller than expected was in Thailand on March 20. We drop this date from the sample of monetary policy surprises.

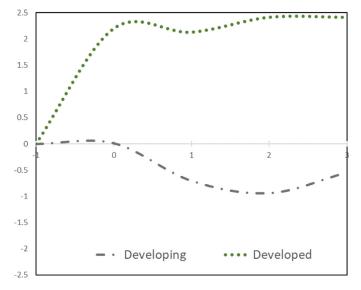


Fig. 5. Abnormal returns of bank stocks after borrower assistance announcements: developed vs. developing countries

Notes: The variable plotted on the vertical axis shows the accumulated abnormal returns in percentage points for borrower assistance policies within the event window of one day before the event and three days after the event, scaled to zero on the day before the announcement. Accumulated abnormal returns are calculated by estimating Eq. (3) on a constant removing all other covariates and including country and announcement day fixed effects. The sample is split between developed and developing countries. The horizontal axis shows days within the event window, with "0" corresponding to the day of the announcement.

subsidiary. We examine the stock price reaction of these banks on days when fiscal and regulatory authorities at the country of the subsidiary introduce financial sector measures. Since local regulators are less likely to be influenced by the performance of individual banks with operations abroad, endogeneity concerns should be reduced.

We use data from Factset Revere Geographic Revenue (GeoRev). GeoRev provides annual information of companies' revenue by geographic region and country. We use the last annual report for each bank in our sample prior to March 2021. We identify 150 banks that derived at least 1% of their total revenue from international subsidiaries. For example, according to GeoRev, Banco Santander, domiciled in Spain, generated 24% of its total revenue from Brazil, 17% from the United States, 11% from the United Kingdom, 8% from Mexico, and 5% from Chile during 2020. Table A7 presents the summary statistics of the 150 cross-border banks in our sample. As the table shows, subsidiaries abroad represent 5.9% of the total revenue of the average cross-border bank, with banks in the 25th percentile generating 1.6% of their revenues in their subsidiaries and banks in the 75th percentile making 6.8%. In terms of geographical penetration, the median cross-border bank in the sample operates in 6 countries, while banks in the 75th percentile operate in 12 countries. Of the 150 cross-border banks in the sample, 130 report revenues in subsidiaries located in developed countries, compared to 87 in developing countries.

We modify our event study methodology from Eq. (3) to examine the abnormal returns of a cross-border bank *b* in its country of domicile *c* on days when authorities in the country of its subsidiary *s* made a policy announcement,  $ARet_{b.c.s.t}^n$ . To account for contemporaneous news that affect the banking sector at the country of domicile, we control for the average abnormal returns of banks in country *c*, excluding bank *b*:  $ARet_{-b,c.t}^n$ . We also control for the incidence of the pandemic in both the location of the subsidiary and the bank's domicile using the percentage change of COVID-19 cases per million during the week,  $Covid_{s.t}$  and  $Covid_{c.t}$ .

respectively. We estimate the following equation:

$$ARet^{n}_{b,c,s,t} = \alpha_{0,n} + \alpha_{1,n}ARet^{n}_{-b,c,t} + \alpha_{2,n}Covid_{s,t} + \alpha_{3,n}Covid_{c,t} + \gamma_{b} + \gamma_{t} + u_{b,c,t}$$

$$(4)$$

where  $\gamma_b$  and  $\gamma_t$  are bank and day fixed effects and errors are clustered at the bank level. In Eq. (4), the constant  $\alpha_{0,n}$  represents the average abnormal returns of cross-border banks in the time window [-1, *n*] when a policy category is announced at the country of their subsidiary. To be precise, the abnormal returns of a bank around *t* are included in Eq. (4) anytime a policy is announced in a country where the bank operates a subsidiary. Following the analysis in Section 4, we estimate Eq. (4) separately for each policy category, for announcements in developed and developing countries and focus on days with single policy announcements at the location of the subsidiary.

Table 7 presents our results. We examine the abnormal returns of cross-border banks for announcements in countries where the subsidiary represents at least 1 and 5% of the bank's yearly revenue. The results confirm our previous findings. In developed countries, countercyclical prudential measures are followed by negative abnormal stock returns up to three days after the announcement. Conversely, when developed countries announced borrower assistance programs, foreign banks with significant exposure to the country experience large and positive abnormal returns in their stock prices. Also, consistent with previous findings, borrowing assistance measures in developing countries seem to have no effect on bank stock prices.

To verify that our findings are not driven by spurious correlations in the stock prices of global banks at the onset of the pandemic, we estimate Eq. (4) for announcements in bank-country pairs where the revenue exposure is less than 1%. In this placebo test, we confirm that for all policy categories, the estimated  $\alpha_{0,n}$  is small and indistinguishable from zero. That is, the stock returns of large cross-border banks do not seem to be driven by announcements in countries where the subsidiary only represents a small fraction of total revenue.

# 5.2. Drivers of policy adoption

As an additional exercise, we examine the drivers behind the announcement of financial sector policies across jurisdictions. More precisely, we examine the extent to which domestic factors, such as the banking sector characteristics of a jurisdiction, and external factors, such as the adoption of similar policies in neighboring countries, predict the adoption of certain policies in a jurisdiction. Concerns about the endogeneity of policy measures would be mitigated if policy adoption is driven by external factors, for instance, due to policy contagion, rather than by domestic factors. Given the pace and uncertainty from the unprecedented global shock, local authorities might implement policies following the announcements in other countries. Indeed, Mistur et al. (2020) show that the actions of geographical neighbors and peer countries significantly influenced the decision to implement social distancing rules, close non-essential businesses and institute travel bans. Earlier work has also documented financial sector policy diffusion across borders in the adoption of international banking standards (Jones and Zeitz, 2019) and in the implementation of deposit insurance (Demirguc-Kunt et al., 2006).

To investigate the factors associated with the adoption of financial sector measures in a jurisdiction between the period of February to April, we run a series of OLS regressions at the country-day

Impact of policy initiatives on cross-border banks.

	Liquidity		Prudential		Borrower	Support	Policy Ra	tes	Asset Pur	chases	All annou	ncements
<i>n</i> =	0	3	0	3	0	3	0	3	0	3	0	3
Panel A. Deve	loped cour	ıtries										
Revenue thres	hold = 1%											
Constant	-0.380	0.627	-1.101***	-1.765***	9.474**	12.79*	0.895	0.111	-0.0650	-2.238*	0.0255	-0.800**
	(2.498)	(4.308)	(0.335)	(0.480)	(4.270)	(7.536)	(0.596)	(1.138)	(0.843)	(1.346)	(0.190)	(0.357)
Observations	194	194	511	506	183	182	112	112	130	130	1,236	1,231
R-squared	0.602	0.684	0.463	0.449	0.466	0.444	0.687	0.729	0.465	0.627	0.270	0.301
Revenue thres	hold = 5%											
Constant	-0.372	1.680	-1.671***	-2.254***	11.12*	18.50**	1.193**	0.474	0.837	-0.691	0.129	-0.908**
	(3.349)	(5.736)	(0.516)	(0.689)	(5.583)	(8.806)	(0.556)	(1.372)	(1.158)	(2.093)	(0.244)	(0.452)
Observations	98	98	202	202	62	62	31	31	55	55	529	529
R-squared	0.646	0.650	0.462	0.489	0.709	0.705	0.929	0.944	0.601	0.599	0.275	0.323
Panel B. Devel	oping cou	ntries										
Revenue thres	hold = 1%											
Constant	-0.676	-1.956**	-0.834	-2.689**	0.358	-0.497	0.129	-0.483	-	-	-0.101	-1.010**
	(0.604)	(0.875)	(0.995)	(1.161)	(1.186)	(2.322)	(0.257)	(0.497)	-	-	(0.139)	(0.258)
Observations	94	94	58	58	60	60	182	181	-	-	495	494
R-squared	0.604	0.695	0.733	0.888	0.623	0.764	0.474	0.556	-	-	0.381	0.508
Revenue thres	hold = 5%											
Constant	1.126	0.0624	-0.667	-2.509*	1.126	-0.852	0.286	-0.321	-	-	-0.0199	-0.943**
	(1.131)	(1.951)	(1.358)	(1.267)	(1.353)	(2.810)	(0.283)	(0.492)	-	-	(0.204)	(0.367)
Observations	21	21	21	21	21	21	84	83	-	-	216	215
R-squared	0.906	0.975	0.900	0.950	0.712	0.982	0.350	0.635	-	-	0.460	0.639

Notes: The table presents impact estimates of financial sector policies on the abnormal returns of cross-border banks with subsidiaries generating at least 1% of the banks' revenues and located in the country of the policy announcement. The estimates correspond to the abnormal returns of cross-border banks on the announcement day (n = 0) and three days after the announcement (n = 3). The sample of announcements is restricted to announcements of a particular category that do not overlap with announcements of other financial sector initiatives. Panel A restricts the sample to announcements in developed countries. Panel B restricts the sample to announcements in developing countries. Abnormal returns are calculated as the difference between realized returns and the expected returns implied by a market model. Controls include the average abnormal returns of banks in the country where the cross-border bank is domiciled. All specifications include day and bank fixed effects. Standard errors are clustered at the bank level.

level that are summarized in Eq. (4).<sup>18</sup>

$$y_{c,t}^{p} = \alpha_{0} + \alpha_{1}INT_{c} + \alpha_{2}Covid_{c,t-1} + \alpha_{3}Market_{c,t-1} + \alpha_{4}EXT_{c,t-1}^{p} + \gamma_{w} + u_{c,t}.$$
(4)

The dependent variable of Eq. (4) equals one if country *c* announced a policy intervention type *p* on day *t* and zero otherwise. Since policies in our sample are present from the announcement date to the end of our observation period, we drop all post-announcement days.<sup>19</sup> The vector of internal factors  $INT_c$  includes the set of covariates that capture ex-ante banking sector characteristics of a jurisdiction: liquidity ratio, oil exposure, size, state ownership and capital ratio, averaged for all publicly traded banks in each country.<sup>20</sup> *Covid*<sub>c,t-1</sub> corresponds to the percentage change in the number of COVID-19 cases per million on day *t*-1 in country *c*. Time varying market factors include the accumulated returns of the domestic and global equity markets, as well as the banks' stock returns in the country.

The vector of external factors  $EXT^{p}_{c,t-1}$  includes two variables that capture policy contagion: (i) The fraction of peer countries that by day *t*-1 have adopted the same type of policy p ( $Same_{t-1,p}$ ), and (ii) the fraction of peer countries that by day t-1 have announced financial sector policies other than p ( $Other_{t,p}$ ). To evalu-

ate contagion, we define peers as countries in the same geographical region within a particular income group as classified by the World Bank and calculate the share of countries within a region announcing a policy type at each day (World Bank, 2019).<sup>21,22</sup> Appendix Table A8 presents summary statistics for the variables used in this section.<sup>23</sup>

In Eq. (4),  $\gamma_w$  corresponds to week-fixed effects to control for unobserved common elements that make a particular financial policy initiative more desirable in countries over time. In some specifications, we further include fixed effects at the *country\*week* level to isolate the role of policy contagion from all other time-varying developments within a country that may influence policy decisions of regulators. Errors are clustered at the country level and all covariates are standardized with mean 0 and standard deviation 1.

Table 8 and 8a presents our main findings. Estimates for each policy announcement are displayed across three columns. In the first column of each policy measure, our results indicate that observable bank sector characteristics of a jurisdiction prior to the COVID-19 shock had little and in most cases no effect on the type

<sup>&</sup>lt;sup>18</sup> While OLS regressions ease the interpretation of the coefficients, we also include the results from a Probit model.

<sup>&</sup>lt;sup>19</sup> The exercise resembles a hazard model of policy adoption.

<sup>&</sup>lt;sup>20</sup> To compute  $X_c$ , we average the characteristics of banks in our sample using equal weights for each bank. As a robustness check, we compute  $X_c$  using weighted averages, where the weight of a given bank is calculated using the assets (in logs) of the bank in 2019. We confirm that our results remain unchanged with this alternative measure of  $X_c$ .

<sup>&</sup>lt;sup>21</sup> High-income countries are classified in two groups: (i) High-income OECD countries, (ii) High-income non-OECD countries. All other countries are classified in six regions: (i) East Asia & Pacific, (ii) Europe and Central Asia, (iii) Latin America & the Caribbean, (iv) Middle East & North Africa, (v) South Asia, (vi) Sub-Saharan Africa.

 $<sup>^{\</sup>rm 22}$  Career concerns among regulators might also promote policy contagion (Gadinis, 2015).

<sup>&</sup>lt;sup>23</sup> In results not reported, we alternatively calculate contagion using the fraction of countries in the world that have announced a policy at each point in time. This definition deals with the possibility that some policy diffusion might occur from core to periphery countries, or even in different networks that form through cross-border financial linkages (e.g., Basel committee membership). The role of policy contagion remains similar when using this definition of peer countries.

Impact of policy contagion on policy announcements.

	Liquidity			Prudential			Borrower Su	ipport	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Liquidity ratio	0.003	0.003		0.000	0.001		0.003	0.006	
	[0.002]	[0.004]		[0.003]	[0.003]		[0.003]	[0.004]	
Oil exposure	0.000	-0.003		0.003	0.001		-0.005	-0.004	
	[0.005]	[0.005]		[0.004]	[0.005]		[0.005]	[0.005]	
Size	-0.005	-0.005		0.006	0.006		-0.000	-0.002	
	[0.003]	[0.003]		[0.004]	[0.004]		[0.004]	[0.004]	
Public bank	0.003	0.012**		0.009**	0.012***		0.005	0.011**	
	[0.005]	[0.005]		[0.004]	[0.004]		[0.004]	[0.005]	
Capital ratio	-0.005	-0.005		0.001	0.001		-0.005	-0.006	
-	[0.005]	[0.005]		[0.004]	[0.005]		[0.004]	[0.005]	
COVID	0.003	0.002	0.001	0.001	0.000	0.000	-0.001	-0.000	0.001
	[0.002]	[0.002]	[0.002]	[0.002]	[0.002]	[0.002]	[0.001]	[0.001]	[0.002]
Domestic returns	-0.001	0.003	0.008	0.018	0.019	0.021	0.003	0.011	0.019
	[0.010]	[0.011]	[0.017]	[0.011]	[0.012]	[0.024]	[0.009]	[0.009]	[0.023]
Bank returns	-0.006	-0.002	-0.028	-0.003	0.001	-0.017	-0.012	-0.010	0.002
	[0.008]	[0.009]	[0.021]	[0.008]	[0.008]	[0.029]	[0.007]	[0.008]	[0.027]
World returns	0.011*	0.013**	0.002	0.012	0.015*	0.009	0.010	0.013	-0.004
	[0.006]	[0.006]	[0.005]	[0.008]	[0.008]	[0.008]	[0.010]	[0.010]	[0.009]
Same		0.115***	0.239***		0.055*	0.234***		0.046**	0.131***
		[0.026]	[0.055]		[0.029]	[0.053]		[0.021]	[0.040]
Other		-0.038*	-0.023		-0.039	-0.034		-0.000	0.044
		[0.019]	[0.049]		[0.027]	[0.045]		[0.021]	[0.033]
Constant	0.014***	0.036***	0.059***	0.019***	0.027***	0.064***	0.017***	0.026***	0.036***
	[0.003]	[0.006]	[0.009]	[0.003]	[0.007]	[0.009]	[0.003]	[0.005]	[0.005]
Country*Week FE	No	No	Yes	No	No	Yes	No	No	Yes
Observations	1,610	1,610	1,856	1,499	1,499	1,806	1,589	1,589	1,907
R-squared	0.056	0.098	0.374	0.071	0.079	0.374	0.032	0.054	0.341

Notes: The table presents the estimates of OLS regressions at the country-day level on the probability of announcing a financial sector policy from February 1 to April 17, 2020. The dependent variable corresponds to an indicator variable that equals 1 on days that a country announces the policy in each panel and zero otherwise. Post-announcement observations are dropped from the sample. For each country, Liquidity ratio (cash & due from banks/total assets), Oil exposure, Size (log of assets) and Capital ratio (Tier 1 + Tier 2 capital over total assets) are averaged over the period 2019Q1-2019Q4 across banks in the sample. Public bank is the country share of state-owned banks in the sample. COVID corresponds to the lagged daily percentage change of confirmed COVID-19 cases per million citizens. Domestic returns correspond to the one-day lagged accumulated domestic market returns. Bank returns measures the one-day lagged accumulated bank returns averaged for each country. World returns are the one-day lagged accumulated global market returns. Same corresponds to the regional one-day-lagged share of countries that announce other financial sector policies different from the policy in each panel. All control variables are standardized with mean 0 and standard deviation 1. All specifications include week fixed effects. Standard errors are clustered at the country level.

of policy that was implemented in each country. One exception is the share of public banks, which is positively associated with an increased likelihood of announcing different measures. According to our results, a one-standard-deviation increase in a country's share of state-owned banks is linked to an increase in the likelihood of announcing prudential measures of 0.9% points and of 1.7% points in the likelihood of monetary policy announcements. Similarly, we find that the adoption of financial sector policies of a country is not driven by domestic market factors, measured by the lagged stock returns of banks in the country and the lagged accumulated returns of the domestic equity market. While global market factors have a statistically significant association with the adoption of certain measures, their economic magnitude is small. More precisely, a one-standard-deviation increase in the lagged accumulated returns of global equity markets leads to a 1.1% point increase in the probability of a liquidity measure announcement. The lack of explanatory power of ex-ante banking sector characteristics is not driven by the inclusion of lagged domestic and global factors in the regressions. That is, the coefficients and standard errors of the banking sector variables remain unaffected if we drop from the regressions the lagged accumulated returns of domestic and global equity markets and the lagged bank stock returns.

The next two columns of each panel show the role of country contagion on the likelihood of policy announcements. We find that past announcements of a specific policy type by regional peer countries have a statistically significant and economically meaningful effect on the policy adoption decision (the coefficient of  $Same_{t-1,p}$  is positive). After controlling for time-varying factors

within a country (captured by the country\*week fixed effects), we find that a one-standard deviation increase in the share of countries announcing prudential measures in the previous day, is associated with a 23.4% point increase in the probability that a local authority implements a similar measure. This is a sizable increase. from 1.2% – the average likelihood of introducing a prudential measure on any given day - to 24.6% after peer countries implement prudential policies. Notably, past announcements of other policy types by regional peer countries do not increase the likelihood that a jurisdiction announces a specific policy (with the coefficient  $Other_{t-1,p}$  being either negative or in most cases indistinguishable from zero). We find similar results when using a Probit model instead of an OLS (reported in Appendix Table A9), although with the inclusion of week fixed effects we lose about half of the observations in most specifications and the inclusion of week\*country fixed effects is prohibitive in this case.

Our evidence thus suggests that even after controlling for timevarying unobserved domestic factors, external factors, and in particular, policy contagion, are major drivers of policy adoption. That is, as more countries adopt a specific financial sector policy initiative, other countries become more likely to follow their lead. We interpret these findings as evidence that conditional on other countries' actions, regulators randomly employed financial sector policies in their jurisdictions. Moreover, given that our analysis throughout the paper includes country-fixed effects, we effectively are comparing how the announcement of a particular policy impacted banks with different characteristics *within* the country.

#### Table 8a

Impact of policy contagion on policy announcements.

	Policy Rates			Asset Purchases		
	(1)	(2)	(3)	(4)	(5)	(6)
Liquidity ratio	-0.001	-0.005		-0.001	-0.000	
	[0.003]	[0.004]		[0.001]	[0.001]	
Oil exposure	0.003	0.003		0.001	0.002	
-	[0.005]	[0.006]		[0.003]	[0.003]	
Size	-0.002	-0.000		0.000	-0.001	
	[0.004]	[0.006]		[0.002]	[0.002]	
Public bank	0.017***	0.005		-0.001	0.000	
	[0.005]	[0.007]		[0.001]	[0.001]	
Capital ratio	-0.007	0.007		-0.001	-0.002	
	[0.005]	[0.007]		[0.002]	[0.003]	
COVID	0.009	0.010	0.009	0.001	0.001	0.001
	[0.010]	[0.009]	[0.008]	[0.001]	[0.001]	[0.001]
Domestic returns	0.012	0.021	0.001	-0.001	-0.001	-0.013
	[0.014]	[0.016]	[0.024]	[0.004]	[0.005]	[0.014]
Bank returns	-0.008	-0.011	-0.031	0.000	-0.001	0.020
	[0.009]	[0.010]	[0.029]	[0.003]	[0.003]	[0.018]
World returns	0.018	0.014	0.005	0.005	0.005	0.003
	[0.016]	[0.016]	[0.013]	[0.004]	[0.005]	[0.003]
Same		0.126***	0.182***		0.004	0.074***
		[0.034]	[0.045]		[0.003]	[0.024]
Other		-0.049**	0.012		-0.003	-0.012
		[0.023]	[0.047]		[0.005]	[0.014]
Constant	0.023***	0.041***	0.047***	0.006***	0.006***	0.003*
	[0.004]	[0.010]	[0.007]	[0.002]	[0.002]	[0.002]
Country*Week FE	No	No	Yes	No	No	Yes
Observations	1318	1318	1558	1945	1945	2326
R-squared	0.040	0.090	0.340	0.012	0.014	0.319

Notes: The table presents the estimates of OLS regressions at the country-day level on the probability of announcing a financial sector policy from February 1 to April 17, 2020. The dependent variable corresponds to an indicator variable that equals 1 on days that a country announces the policy in each panel and zero otherwise. Postannouncement observations are dropped from the sample. For each country, Liquidity ratio (cash & due from banks/total assets), Oil exposure, Size (log of assets) and Capital ratio (Tier 1 + Tier 2 capital over total assets) are averaged over the period 2019Q1-2019Q4 across banks in the sample. Public bank is the country share of stateowned banks in the sample. COVID corresponds to the lagged daily percentage change of confirmed COVID-19 cases per million citizens. Domestic returns correspond to the one-day lagged accumulated domestic market returns. Bank returns measures the one-day lagged accumulated bank returns averaged for each country. World returns are the one-day lagged accumulated global market returns. Same corresponds to the regional one-day-lagged share of countries that announce other financial sector policies different from the policy in each panel. All control variables are standardized with mean 0 and standard deviation 1. All specifications include week fixed effects. Standard errors are clustered at the country level.

#### 5.3. Cross-country announcements

As a final exercise to explore whether our results are biased by the endogeneity of policy actions, we examine broad cross-country policy measures in the euro area. Arguably, interventions from the European Central Bank are less likely to be driven by within country considerations. We estimate abnormal returns in bank stocks around policy announcements in Euro area countries.<sup>24</sup> We focus on countercyclical prudential measures because these offer the widest set of announcements to perform multiple robustness tests. For instance, prudential measures are announced both by local authorities, as is the case in France, Germany, Italy, and Spain, and at the ECB level. There are also days when only prudential measures were undertaken (liquidity, borrower support, and asset purchases were exclusively announced with other initiatives for these set of countries).<sup>25</sup>

Columns (1) and (2) of Table A10 present the estimates of the accumulated abnormal returns from the announcement day to three days after the announcement. These results include days when prudential measures are introduced by domestic authorities or by the ECB. In line with previous findings, bank stocks display negative returns after prudential measures are announced. Larger banks within the EU display the lowest returns following prudential measures, although the coefficient is not statistically significant three days after the announcement. The results are robust to excluding domestic announcements (columns 3 and 4) which suggests that supranational prudential measures result in declines in bank stock prices. Our findings provide further credence to the interpretation that markets are pricing the additional risk to banks from the adoption of countercyclical measures. Finally, we restrict our sample to days when only prudential measures were announced (excluding days with multiple policy announcements in columns 5 and 6). Although the constant is not statistically significant in this case due to the reduced sample size, it appears that on average, bank stock prices have large declines up to three days following the announcement.

# 6. Conclusions

The spread of COVID-19 represents an unpresented global shock, with the disease itself and mitigation efforts –such as social distancing measures and partial and national lockdown measures both having a significant impact on the economy. In the immediate aftermath, the financial sector, particularly banks, were expected to play an important role in absorbing the shock by supplying vital credit to the corporate sector and households. In an effort to facilitate this, central banks and governments around the world enacted a wide range of policy measures to provide greater liquidity and support the flow of credit. An important policy question is the potential impact of these countercyclical lending policies on the future stability of the banking systems and to what extent their strengthened capital positions since the global financial crisis will allow them to absorb this shock without undermining their resilience.

 $<sup>^{\</sup>rm 24}$  In our sample, these are: Austria, Finland, France, Germany, Italy, the Netherlands, Portugal, and Spain.

<sup>&</sup>lt;sup>25</sup> There is no interest rate policy announcement in the sample of EU countries.

In this paper, we use daily stock prices and other balance sheet information for a sample of banks in 53 countries to take a first look at this issue. Our contribution is twofold. We first assess the impact of the pandemic on the banking sector and investigate whether the shock had a differential impact on banks versus corporates, as well as those banks with different characteristics. Second, using a global database of financial sector policy responses and an event study methodology, we investigate the role of different policy initiatives on addressing bank stress as perceived by markets, in the aggregate, as well as across different banks.

Our results suggest that the adverse impact of the COVID-19 shock on banks was much more pronounced and long-lasting than on the corporates as well as other non-bank financial institutions, revealing the expectation that banks are to absorb at least part of the shock to the corporate sector. Furthermore, larger banks and public banks suffered greater reductions in their stock returns, reflecting their greater anticipated role in dealing with the crisis. Banks with lower pre-crisis liquidity also suffered greater reduction in returns, consistent with their greater vulnerability to such a shock.

Investigating close to 400 policy announcements between February and April 2020, we next evaluate the impact of liquidity support, prudential measures, borrower assistance and monetary policy measures on bank abnormal returns. Our results suggest liquidity support and borrower assistance measures had the greatest positive impact on bank abnormal returns. Less liquid banks benefited most from liquidity support, whereas larger banks saw increased abnormal returns with the announcement of borrower assistance policies. However, since they rely on fiscal expenditures, these policies did not result in a positive impact on bank stock prices in developing countries where there is less room for fiscal expansion. Prudential measures appeared to have a negative impact on bank returns, suggesting that markets price the downside risk from depletion of capital buffers. Finally, policy rate cuts mostly benefited less liquid banks, confirming that monetary policy again played a key tool during this crisis.

Overall, our results suggest that the crisis and the countercyclical lending role they are expected to play has put banking systems around the world under stress, having a differential impact depending on their characteristics and pre-crisis vulnerabilities. While some policy measures such as liquidity support, borrower assistance and monetary easing moderated this adverse impact for some banks, this is not true for all banks or in all circumstances. For example, borrower assistance measures exacerbated the stress for banks that operate in countries with little fiscal space. These vulnerabilities will need to be carefully monitored as the pandemic continues to take its toll on the world economies.

# **Declaration of Competing Interest**

None.

# Appendix

Table A1-A10

Table A1	
Number of banks	by country.

Region	Country	Number of banks	Weights	Domestic Index	Region	Country	Number of banks	Weights	Domestic Index
EAP	Indonesia	39	0.22	JKKM100	Africa	Nigeria	10	0.47	NGSEINDEX
	Malaysia	10	0.30	KLSE		South-Africa	3	0.53	JALSH
	Philippines	13	0.11	PSI	China	China	33	0.75	CSI300
	Thailand	9	0.25	SET50		Hong-Kong	6	0.25	HSI
	Vietnam	13	0.13	VNI	India	India	32	1.00	BSESN
ECA	Bulgaria	2	0.00	SOFIX	High Income	Australia	7	0.05	AXJO
	Croatia	5	0.03	CRBEX		Austria	5	0.03	ATX
	Poland	12	0.31	WIG20		Canada	9	0.04	GSPTSE
	Romania	3	0.03	BETI		Denmark	20	0.01	OMXC25CAP
	Russia	10	0.25	IMOEX		Finland	3	0.00	OMXH25
	Turkey	10	0.38	XU030		France	12	0.05	FCHI
LAC	Argentina	6	0.04	MERVAL		Germany	5	0.00	GDAXI
	Brazil	11	0.36	IBOV		Ireland	2	0.00	ISEQ
	Chile	5	0.21	IPSA		Italy	14	0.02	FTMIB
	Colombia	2	0.12	COLCAP		Japan	70	0.04	TOPX
	Mexico	4	0.17	MEXBOL		Netherlands	2	0.01	AEX
	Peru	4	0.10	SPBL25PT		Norway	34	0.00	OBX
MENA	Egypt	9	0.02	EGX30		Portugal	1	0.00	PSI20
	Israel	8	0.18	TA35		Singapore	2	0.01	STI
	Kuwait	10	0.10	BKA		South-Korea	3	0.00	KS11
	Morocco	6	0.03	MASI		Spain	8	0.03	IBEX
	Pakistan	21	0.05	KSE		Sweden	5	0.01	OMXS30
	Qatar	9	0.16	QSI		Switzerland	13	0.01	SSMI
	Saudi Arabia	11	0.23	TASI		UK	11	0.10	FTSE
	UAE	10	0.23	ADI		US	344	0.59	SPX

Notes: The table lists the number of banks by country along with the sum of their weights (based on assets) within each region and their domestic market index.

Table	A2
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Summary statistics of banks by region.

Region	Variable	Mean	SD	P25	P50	P75	Obs
AFRICA	Stock returns	-0.002	0.003	-0.004	-0.003	-0.001	13
	Liquidity ratio	0.16	0.09	0.09	0.18	0.23	13
	Oil exposure	-0.21	0.43	-0.44	-0.39	0.13	13
	Size	22.2	1.9	21.1	22.4	23.5	13
	Public bank	0.1	0.3	0.0	0.0	0.0	13
	Capital ratio	25.9	7.5	23.3	24.0	26.2	6
	COVID	7.0	11.8	0.8	0.8	0.8	13
CHINA	Stock returns	-0.002	0.001	-0.002	-0.002	-0.001	39
	Liquidity ratio	0.12	0.02	0.10	0.11	0.14	38
	Oil exposure	-0.11	0.36	-0.27	-0.08	0.04	39
	Size	26.2	1.7	24.6	26.2	27.6	38
	Public bank	0.1	0.3	0.0	0.0	0.0	39
	Capital ratio	14.1	1.4	12.7	14.0	15.4	32
	COVID	41.1	0.0	41.1	41.1	41.1	33
EAP	Stock returns	-0.003	0.005	-0.005	-0.003	-0.002	84
2	Liquidity ratio	0.10	0.05	0.06	0.09	0.13	84
	Oil exposure	-0.02	0.62	-0.41	-0.13	0.13	84
	Size	22.9	1.8	21.6	23.3	24.4	84
	Public bank	0.3	0.5	0.0	0.0	1.0	84
	Capital ratio	21.6	9.3	16.6	18.9	22.7	63
	COVID	11.7	9.5 13.0	5.6	5.7	13.5	84
ECA	Stock returns	-0.003	0.003	-0.004	-0.003	-0.002	42
ECA	Liquidity ratio	-0.005	0.003	-0.004 0.06	-0.005	0.15	42 37
	1 0	0.09					42
	Oil exposure		0.68	-0.10	0.11	0.31	
	Size	23.4	1.5	22.5	23.7	24.7	37
	Public bank	0.2	0.4	0.0	0.0	0.0	41
	Capital ratio	15.8	4.0	14.6	16.2	17.5	27
DIDIA	COVID	195.5	171.6	66.3	113.8	471.8	37
INDIA	Stock returns	-0.006	0.003	-0.008	-0.006	-0.004	32
	Liquidity ratio	0.07	0.03	0.06	0.07	0.08	32
	Oil exposure	-0.04	0.43	-0.35	-0.07	0.28	32
	Size	24.1	1.3	23.2	24.2	25.1	32
	Public bank	0.5	0.5	0.0	1.0	1.0	32
	Capital ratio	14.3	3.5	12.0	13.8	16.0	32
	COVID	2.6	0.0	2.6	2.6	2.6	32
LAC	Stock returns	-0.004	0.003	-0.006	-0.004	-0.002	32
	Liquidity ratio	0.11	0.08	0.05	0.11	0.15	24
	Oil exposure	0.15	0.48	-0.14	0.26	0.46	32
	Size	23.7	1.3	22.7	23.6	24.7	24
	Public bank	0.1	0.3	0.0	0.0	0.0	24
	Capital ratio	15.6	3.2	13.1	14.6	18.8	14
	COVID	86.0	89.3	35.6	39.0	152.7	32
MENA	Stock returns	-0.002	0.002	-0.004	-0.003	-0.001	84
	Liquidity ratio	0.08	0.05	0.05	0.07	0.10	79
	Oil exposure	0.03	0.58	-0.15	0.11	0.24	84
	Size	23.4	1.4	22.3	23.5	24.5	79
	Public bank	0.4	0.5	0.0	0.0	1.0	84
	Capital ratio	17.1	3.0	14.7	17.3	18.8	61
	COVID	112.3	103.8	9.6	121.1	158.6	84
High-Income	Stock returns	-0.004	0.003	-0.006	-0.004	-0.002	570
-	Liquidity ratio	0.06	0.06	0.01	0.03	0.09	448
	Oil exposure	0.10	1.49	-0.18	0.03	0.22	568
	Size	23.1	2.4	21.1	22.6	24.3	460
	Public bank	0.0	0.2	0.0	0.0	0.0	563
	Capital ratio	15.8	5.2	13.4	14.6	17.1	348
	COVID	438.0	215.6	342.4	529.9	541.5	570

Notes: The table presents the summary statistics of the 896 banks in the sample. Stock returns for each bank are averaged over the period January 1, 2020 to May 5, 2020. Liquidity ratio is defined as the ratio of liquid assets (cash & due from banks) to total assets averaged over the 2019Q1-2019Q4 period. Oil exposure corresponds to the slope coefficient of an OLS regression of bank's stock returns on a constant, the market return where the bank is domiciled, and the rate of return of oil prices using weekly data between May 2018 and December 2019. Size is calculated as the 2019Q1-2019Q4 average total assets for each bank and is reported in logs. Public bank is an indicator variable that equals one for banks with a non-zero equity participation from the domestic government and zero otherwise. Capital ratio corresponds to the 2019Q1-2019Q4 average ratio of Tier 1 + Tier 2 capital to total assets. COVID corresponds to the average number of confirmed COVID-19 cases per million citizens in the country of banks during the sample period.

#### Table A3

Risk factors during the COVID-19 crisis (implied by six-factor model).

	January-February 2	2020		March-April 2020		
Liquidity ratio measure:	(1) Liquid assets A/total assets	(2) Liquid assets B/total deposits	(3) Liquid assets C/total assets	(4) Liquid assets A/total assets	(5) Liquid assets B/total deposits	(6) Liquid assets C/total assets
Liquidity ratio	0.035	0.056	0.068	0.071	0.126	0.146**
	[0.108]	[0.078]	[0.043]	[0.091]	[0.096]	[0.066]
Oil exposure	0.230**	0.231**	0.213**	-0.104	-0.109	-0.082
	[0.096]	[0.095]	[0.096]	[0.109]	[0.111]	[0.096]
Size	0.027	0.032	0.082	0.209	0.194	0.162
	[0.116]	[0.115]	[0.129]	[0.238]	[0.241]	[0.248]
Public bank	0.010	0.014	-0.006	0.144	0.157	0.151
	[0.354]	[0.356]	[0.318]	[0.228]	[0.233]	[0.236]
Capital ratio	-0.028*	-0.027*	-0.026	-0.016	-0.019	-0.023
	[0.014]	[0.014]	[0.016]	[0.016]	[0.018]	[0.018]
COVID	0.003	0.003	0.006	-0.054*	-0.055*	-0.054*
	[0.019]	[0.019]	[0.019]	[0.030]	[0.030]	[0.031]
Constant	0.351	0.333	0.290	-0.873***	-0.832***	-0.764**
	[0.233]	[0.239]	[0.272]	[0.265]	[0.279]	[0.291]
Observations	4453	4430	4421	5322	5293	5280
R-squared	0.043	0.043	0.039	0.068	0.067	0.067

Notes: The table presents OLS estimates of bank-week panel regressions. The dependent variable is abnormal returns, calculated as the difference between a bank stock returns during a week and the expected returns implied by a six-factor model: domestic market, global market, size, book to market, profitability, and investment risk. Returns are calculated in local U.S. dollars. The definition of liquidity ratio, averaged over the 2019Q1-2019Q4 period, corresponds to: Liquid assets A/total assets, where liquid assets A include cash & due from banks (columns 1 and 4); liquid assets B/total deposits, where liquid assets B include cash & due from banks plus federal funds sold and securities purchased under resale agreements (columns 2 and 5); liquid assets C/total assets, where liquid assets C include cash & due from banks, federal funds sold and securities purchased under resale agreements, trading account assets, other short-term investments and total investment securities (columns 3 and 6). Oil exposure corresponds to the slope coefficient of an OLS regression of bank's stock returns on a constant, the domestic markets, and the percentage change of oil prices using weekly data between May 2018 and December 2019. Size is calculated as the 2019Q1-2019Q4 average total assets for each bank and is reported in logs. Public bank is an indicator variable that equals one for banks with non-zero equity participation from the domestic government and zero otherwise. Capital ratio corresponds to the 2019Q1-2019Q4 average ratio of Tier 1 + Tier 2 capital to total assets. COVID corresponds to the percentage change in number of confirmed COVID-19 cases of a country per million citizens are clustered at the country level.

#### Table A4

Risk factors during the COVID-19 crisis (implied by domestic market model).

	January-February 202	0		March-April 2020		
Liquidity ratio measure:	(1) Liquid assets A/total assets	(2) Liquid assets B/total deposits	(3) Liquid assets C/total assets	(4) Liquid assets A/total assets	(5) Liquid assets B/total deposits	(6) Liquid assets C/total assets
Liquidity ratio	0.026	0.032	0.054	0.119	0.146*	0.159**
	[0.087]	[0.071]	[0.041]	[0.063]	[0.076]	[0.061]
Oil exposure	0.175**	0.175**	0.160**	-0.151	-0.159	-0.130
	[0.068]	[0.068]	[0.067]	[0.097]	[0.097]	[0.082]
Size	-0.087	-0.083	-0.033	0.139	0.120	0.087
	[0.119]	[0.118]	[0.143]	[0.222]	[0.222]	[0.220]
Public bank	0.045	0.047	0.027	0.090	0.105	0.091
	[0.331]	[0.334]	[0.298]	[0.242]	[0.247]	[0.251]
Capital ratio	-0.019	-0.019	-0.016	-0.007	-0.010	-0.014
-	[0.013]	[0.013]	[0.016]	[0.016]	[0.017]	[0.017]
COVID	0.019*	0.019*	0.018	-0.059	-0.060	-0.059
	[0.010]	[0.011]	[0.012]	[0.038]	[0.039]	[0.039]
Constant	-0.054	-0.056	-0.118	-1.269***	-1.223***	-1.168***
	[0.213]	[0.221]	[0.261]	[0.248]	[0.250]	[0.265]
Observations	4453	4430	4421	5322	5293	5280
R-squared	0.047	0.047	0.042	0.101	0.101	0.103

Notes: The table presents OLS estimates of bank-week panel regressions. The dependent variable is abnormal returns, calculated as the difference between a bank stock returns during a week and the expected returns implied by a domestic market model. Returns are calculated in local currency. The definition of liquidity ratio, averaged over the 2019Q1-2019Q4 period, corresponds to: Liquid assets A/total assets, where liquid assets A include cash & due from banks (columns 1 and 4); liquid assets B/total deposits, where liquid assets B include cash & due from banks (columns 1 and 4); liquid assets C/total assets, where liquid assets c include cash & due from banks, federal funds sold and securities purchased under resale agreements, trading account assets, other short-term investments and total investment securities (columns 3 and 6). Oil exposure corresponds to the slope coefficient of an OLS regression of bank's stock returns on a constant, the domestic markets, and the percentage change of oil prices using weekly data between May 2018 and December 2019. Size is calculated as the 2019Q1-2019Q4 average total assets for each bank and is reported in logs. Public bank is an indicator variable that equals one for banks with non-zero equity participation from the domestic government and zero otherwise. Capital ratio corresponds to the 2019Q1-2019Q4 average ratio of Tier 1 + Tier 2 capital to total assets. COVID corresponds to the percentage change in number of confirmed COVID-19 cases of a country per million citizens during each week. All control variables are standardized with mean 0 and standard deviation 1. All specifications include country and week fixed effects. Standard errors are clustered at the country level.

#### Table A5

Policy announcements in developed and developing countries.

Policy category Liquidity 40	lays with policy annound nnouncements by ategory	cements = 160) Single policy	(days with policy announ	cements = 115)			
Policy category       Liquidity     40		Single policy	(days with policy announcements = 115)				
Liquidity 40		announcements	Announcements by category	Single policy announcements			
	D	14	43	18			
Prudential 99	Ð	47	42	18			
Borrower Support 57	7	11	32	16			
Asset Purchases 22	2	17	8	2			
Policy Rates 10	0	6	36	28			
Total policies 22	28	95	161	82			
Policy categories							
announced in the same							
day							
Liquidity + Prudential 13	3		7				
Liquidity + Borrower 9			2				
Support							
1			5				
Liquidity + Prudential + Borrower							
Support							
Liquidity + Policy Rates 2			3				
Liquidity + Asset Purchases 0			3				
Prudential + Borrower 35			6				
Support							
Prudential + Policy Rates 0			2				
1			-				
Prudential + Liquidity + Asset			•				
Purchases							
Borrower 0			1				
Support + Liquidity + Policy			•				
Rates							
Borrower Support + Asset 1			0				
Purchases + Policy Rates			0				
Policy Rates + Asset 1			0				
Purchases			0				
Policy 0			1				
Rates + Liquidity + Prudential			1				
Asset 2			0				
Purchases + Prudential			0				

Notes: Data obtained from the World Bank (Feyen et al., 2021) covering the period February 1 to April 17 2020.

# Table A6

Impact of unexpected monetary policy rate announcements.

	(1)	(2)	(3)	(4)
<i>n</i> =	0	3	0	3
Liquidity ratio	-0.581	-1.639*	-0.849	-0.968
	[0.399]	[0.869]	[0.538]	[1.011]
Oil exposure	-0.747*	-1.877**	-0.742	-1.616**
	[0.378]	[0.798]	[0.456]	[0.792]
Size	0.444	1.059	-0.114	0.799
	[0.373]	[0.783]	[0.512]	[0.889]
Public bank			1.729*	2.872*
			[0.929]	[1.637]
Capital ratio			0.379	0.245
			[0.303]	[0.553]
Constant	0.546*	-0.839	-0.084	-0.911
	[0.312]	[0.656]	[0.539]	[0.788]
Observations	89	88	61	60
R-squared	0.266	0.395	0.296	0.366

Notes: The table presents OLS estimates of cross-sectional regressions of the impact of financial sector policies on the abnormal returns of banks on announcement days (n=0) and three days after the announcement (n=3). Abnormal returns are calculated as the difference between realized returns and the expected returns implied by a market model. Liquidity ratio is defined as the ratio of liquid assets (cash & due from banks) to total assets averaged over the 2019Q1-2019Q4 period. Oil exposure corresponds to the slope coefficient of an OLS regression of bank's stock returns on a constant, the market return where the bank is domiciled, and the rate of return of oil prices using weekly data between May 2018 and December 2019. Size is calculated as the 2019Q1-2019Q4 average total assets for each bank and is reported in logs. Public bank is an indicator variable that equals one for banks with a non-zero equity participation from the domestic government and zero otherwise. Capital ratio corresponds to the 2019Q1-2019Q4 average ratio of Tier 1 + Tier 2 capital to total assets. All control variables are standardized with mean 0 and standard deviation 1. All specifications include day and country fixed effects and control for the daily number of COVID-19 cases in a country. Standard errors are clustered at the country level.

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#### Table A7

Summary statistics of cross-border banks.

Subsidiary Location	Mean	SD	Min	p25	p50	p75	Max
All countries (150 banks)							
Revenue (%)	5.87	8.44	1.00	1.55	2.61	6.84	95.42
Number of countries with subsidiaries per bank Developed countries (130 banks)	7.17	4.90	1	3	6	12	17
Revenue (%)	5.88	7.75	1.00	1.59	2.67	7.19	67.56
Number of countries with subsidiaries per bank Developing countries (87 banks)	4.99	3.68	1	2	5	8	13
Revenue (%)	5.85	9.47	1.02	1.50	2.48	6.22	95.42
Number of countries with subsidiaries per bank	4.36	3.27	1	1	4	6	11

Notes: The table presents summary statistics for our sample of 150 cross-border banks that derived at least 1% of their total revenue from international subsidiaries. The first panel present statistics for all cross-border banks. The second and third panels present statistics for cross-border banks with subsidiaries in developed and developing countries, respectively. Revenue (%) corresponds to the share of total revenue generated by the subsidiaries of cross-border banks. Number of countries with subsidiaries of per bank show the number of countries with subsidiaries of cross-border banks representing at least 1% of their total revenue. **Table A8** 

## Summary statistics of country-day level variables.

	Liquidity		Prudent	rudential		Borrower Support		Policy Rates			Asset Purchases				
	Mean	SD	obs	Mean	SD	obs	Mean	SD	Obs	Mean	SD	obs	Mean	SD	obs
Liquidity ratio	0.10	0.06	2,988	0.10	0.06	2,914	0.10	0.06	3,039	0.11	0.06	2,657	0.11	0.06	3,559
Oil exposure	0.00	0.00	2,988	0.00	0.00	2,914	0.00	0.00	3,039	0.00	0.00	2,657	0.00	0.00	3,559
Size	25.0	1.5	2,988	24.8	1.5	2,914	24.9	1.5	3,039	24.9	1.5	2,657	24.9	1.5	3,559
Public bank	0.20	0.27	2,988	0.20	0.28	2,914	0.20	0.28	3,039	0.19	0.28	2,657	0.21	0.29	3,559
Capital ratio	17.3	2.7	2,361	17.2	2.8	2,235	17.3	2.7	2,330	17.3	2.8	2,072	17.3	2.8	2,774
COVID	0.20	0.51	2,031	0.21	0.52	1,978	0.21	0.51	2,081	0.20	0.47	1,727	0.19	0.46	2,573
Same	0.08	0.13	2,988	0.09	0.16	2,914	0.07	0.11	3,039	0.09	0.11	2,657	0.02	0.03	3,559
Other	0.15	0.19	2,988	0.13	0.17	2,914	0.15	0.19	3,039	0.12	0.19	2,657	0.20	0.21	3,559

Notes: The table presents country-day level summary statistics from February 1 to April 17, 2020 for the variables used in the analysis of drivers of policy announcements. Post-announcement observations of the policy in each panel are dropped from the sample. For each country, Liquidity ratio (cash & due from banks/total assets), Oil exposure, Size (log of assets) and Capital ratio (Tier 1 + Tier 2 capital over total assets) are averaged over the period 2019Q1-2019Q4 across banks in the sample. Public bank is the country share of state-owned banks in the sample. COVID corresponds to the daily percentage change of confirmed COVID-19 cases per million citizens. Same corresponds to the regional one-day-lagged share of countries that announce the policy in each panel. Other corresponds to the regional one-day-lagged share of countries that announce other financial sector policies different from the policy in each panel. **Table A9** 

#### Impact of policy contagion on policy announcements (Probit model).

	Liquidity		Prudential		Borrower Su	ipport	Monetary P	olicy	Asset Purch	ases
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Liquidity ratio	0.138	0.017	0.012	0.006	0.176	0.213	-0.081	-0.341*	-0.107	-0.017
	[0.111]	[0.134]	[0.111]	[0.109]	[0.129]	[0.139]	[0.119]	[0.201]	[0.134]	[0.136]
Oil exposure	-0.005	-0.209	0.007	-0.125	-0.169	-0.187	0.010	-0.032	0.095	0.176
	[0.156]	[0.178]	[0.102]	[0.121]	[0.126]	[0.143]	[0.082]	[0.135]	[0.185]	[0.187]
Size	-0.188	-0.094	0.135	0.140	0.020	0.038	-0.062	0.234	0.039	-0.052
	[0.127]	[0.142]	[0.114]	[0.127]	[0.134]	[0.135]	[0.123]	[0.142]	[0.153]	[0.188]
Public bank	0.149	0.297*	0.210**	0.258***	0.189*	0.303***	0.325***	0.056	-0.099	0.022
	[0.161]	[0.155]	[0.093]	[0.092]	[0.108]	[0.118]	[0.092]	[0.142]	[0.131]	[0.155]
Capital ratio	-0.170	0.020	0.020	0.073	-0.218	-0.172	-0.245*	-0.067	-0.092	-0.194
	[0.145]	[0.187]	[0.102]	[0.120]	[0.144]	[0.176]	[0.148]	[0.197]	[0.184]	[0.218]
COVID	0.049	0.065	0.029	0.009	-0.237	-0.298	0.062	0.083	0.064	0.079
	[0.045]	[0.050]	[0.062]	[0.097]	[0.175]	[0.229]	[0.055]	[0.052]	[0.099]	[0.097]
Domestic returns	0.070	0.015	0.248	0.287	0.028	0.197	0.138	0.010	-0.043	0.025
	[0.255]	[0.307]	[0.194]	[0.226]	[0.196]	[0.203]	[0.241]	[0.268]	[0.238]	[0.244]
Bank returns	-0.204	-0.035	-0.015	0.056	-0.264**	-0.245	-0.123	-0.034	0.016	-0.063
	[0.180]	[0.192]	[0.128]	[0.145]	[0.132]	[0.157]	[0.170]	[0.231]	[0.139]	[0.151]
World returns	0.153	0.346	0.339*	0.603***	0.200	0.426	0.265	0.378	0.330	0.366
	[0.142]	[0.270]	[0.174]	[0.206]	[0.241]	[0.297]	[0.273]	[0.298]	[0.265]	[0.314]
Same		2.389***		0.700**		0.582***		2.239***		0.248*
		[0.534]		[0.355]		[0.223]		[0.355]		[0.148]
Other		-1.104**		-0.415		0.247		-0.994**		-0.208
		[0.498]		[0.501]		[0.268]		[0.388]		[0.274]
Constant	-2.306***	-2.502***	-1.361***	-1.440**	-1.941***	-2.774***	-1.800***	-3.681***	-2.086***	-2.094***
	[0.356]	[0.573]	[0.459]	[0.559]	[0.486]	[0.626]	[0.390]	[0.539]	[0.409]	[0.438]
Observations	870	870	771	771	920	920	1019	1019	662	662

Notes: The table presents the estimates of Probit regressions at the country-date level on the probability of announcing a financial sector policy from February 1 to April 17, 2020. The dependent variable corresponds to an indicator variable that equals 1 on dates that a country announces the policy in each panel and zero otherwise. Post-announcement observations are dropped from the sample. For each country, Liquidity ratio (cash & due from banks/total assets), Oil exposure, Size (log of assets) and Capital ratio (Tier 1 + Tier 2 capital over total assets) are averaged over the period 2019Q1-2019Q4 across banks in the sample. Public bank is the country share of state-owned banks in the sample. COVID corresponds to the lagged daily percentage change confirmed COVID-19 cases per million citizens. Same corresponds to the regional one-day-lagged share of countries that announce the policy in each panel. Other corresponds to the regional one-day-lagged share of countries that announce the policy in each panel. Additional controls include one-day lagged accumulated bank returns averaged for each country as well as the one-day lagged accumulated domestic and world market returns. All control variables are standardized with mean 0 and standard deviation 1. All specifications include week fixed effects. Standard errors are clustered at the country level.

#### Table A10

Impact of ECB policy announcements across Euro Area countries.

	Domestic + ECB		ECB Only		Single Announce	ement
<i>n</i> =	(1) 0	(2) 3	(3) 0	(4) 3	(5) 0	(6) 3
Liquidity ratio	1.199	1.735	1.237	1.691	2.616	2.592
	[0.986]	[1.453]	[1.015]	[1.529]	[2.021]	[3.050]
Oil exposure	-0.102	0.068	-0.140	0.086	-0.859	0.621
	[0.381]	[0.561]	[0.386]	[0.581]	[1.071]	[1.617]
Size	-1.030*	-1.303	-1.016	-1.267	-1.050	-2.027
	[0.603]	[0.890]	[0.618]	[0.932]	[1.621]	[2.447]
Public bank	0.566	-0.065	0.683	-0.149	1.896	1.115
	[1.136]	[1.675]	[1.149]	[1.733]	[2.751]	[4.152]
Capital ratio	-0.305	0.092	-0.310	0.109	-0.792	1.547
	[0.353]	[0.521]	[0.354]	[0.535]	[0.958]	[1.446]
Constant	-0.525*	-1.141**	-0.585*	-1.045**	-0.595	-1.008
	[0.330]	[0.488]	[0.336]	[0.508]	[1.140]	[1.720]
Observations	315	316	251	251	50	50
R-squared	0.095	0.097	0.117	0.152	0.414	0.342

Notes: The table presents OLS estimates of cross-sectional regressions of the impact of financial sector policies on the abnormal returns of banks on announcement days (n=0) and three days after the announcement (n=3). Columns 1 and 2 include days when prudential measures are introduced by domestic authorities or by the ECB. Columns 3 and 4 exclude domestic policy announcements. Columns 5 and 6 exclude days with multiple policy announcements. Abnormal returns are calculated as the difference between realized returns and the expected returns implied by a market model. liquidity ratio is defined as the liquidity ratio of a bank averaged over the 2019Q1-2019Q4 period, calculated as the ratio of cash & due from banks to total assets. Oil exposure corresponds to the slope coefficient of an OLS regression of bank's stock returns on a constant, the market return where the bank is domiciled, and the rate of return of oil prices using weekly data between May 2018 and December 2019. Size is calculated as the 2019Q1-2019Q4 average total assets for each bank and is reported in logs. Public bank is an indicator variable that equals one for banks with a non-zero equity participation from the domestic government and zero otherwise. Capital ratio corresponds to the 2019Q1-2019Q4 average ratio of Tier 1 + Tier 2 capital to total assets. All control variables are standardized with mean 0 and standard deviation 1. All specifications include day and country fixed effects and control for the daily number of COVID-19 cases in a country. Standard errors are clustered at the country level.

# **CRediT** authorship contribution statement

**Asli Demirgüç-Kunt:** Conceptualization, Supervision, Writing – original draft, Writing – review & editing. **Alvaro Pedraza:** Methodology, Investigation, Writing – original draft, Writing – review & editing. **Claudia Ruiz-Ortega:** Data curation, Investigation, Writing – original draft, Writing – review & editing.

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