The Impact of the Deposit Channel on the International

Transmission of Monetary Shocks\*

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Abstract

What role do bank deposits play in the international transmission of US monetary policy shocks? Using a panel of US commercial banks, we find that the deposit channel impacts the transmission of monetary shocks through global banks and may account for \$460 billion of new cross-border flows into the US following 1 b.p. monetary shock. Specifically, we document that after a 1 b.p. unexpected increase to the Fed Funds rate, global banks increase deposit spreads by 1.4 b.p. and suffer a 10 p.p. decline in deposit growth. As a result, global banks increase net transfers from foreign branches to finance lending. Global banks reduce lending growth by 40 b.p. less than local banks. This corresponds to a 100% increase in the amount of net transfers from foreign branches.

As global economies have become more integrated, there has been a large discussion both in the academic and in policy circles on the outsized role that the Federal Reserve plays on the global stage. The empirical literature has argued that US monetary policy has international implications and that shocks to US monetary policy reverberate to other countries. In particular, several papers, including Cetorelli and Goldberg (2012) and Temesvary et al. (2018), argue that global banks (banks with foreign offices) actively allocate funds across borders and that much of the transmission of monetary policy and global liquidity shocks act through the internal markets of global banks. In response to a contractionary US monetary policy shock or global liquidity shock, Cetorelli and Goldberg (2012) show that global banks increase cross-border flows into the US and reduce foreign lending, propagating the shock internationally.

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At the same time, Drechsler et al. (2017) and Wang et al. (2020) recently argue that most of the transmission of US monetary policy domestically can be attributed to the bank deposit channel. In their framework, when the US Federal Funds rate increases, because banks have market power, they increase deposit spreads, defined as the spread between the Federal Funds rate and the deposit rates, resulting in households withdrawing bank deposits, and banks, as a result, reduce lending.

Interestingly, several of the existing empirical papers on the international transmission of monetary policy are either agnostic on the domestic channel of monetary policy transmission or lean heavily on the capital reserve channel of banks, rather than the deposit channel. At the same time, global banks hold a majority of US deposits and loans (Figures 1 and 2). Given the evidence of the large role of the bank deposit channel in domestic transmission, this paper addresses the open research question of how much of the international transmission of US monetary policy can be attributed to the deposit channel.

To address this question we start by considering a static model of a global bank which raises deposits and makes lending decisions in two countries, subject to policy rates in the two countries and convex adjustment costs of moving funds across borders. As in Drechsler et al. (2017), households view deposits as a composite good, and banks hold market power in the deposit market. Like domestic banks, when the US raises the Federal Funds rates, the global bank raises its US deposit spreads, resulting in an outflow of US deposits. However, unlike a domestic bank, the global bank has access to funds raised through foreign branches. Therefore the model predicts that global banks will offset the decrease in deposits by increasing cross-border funds into the US, mitigating the reduction in US lending, relative to domestic US banks. At the same time, as foreign funds flow into the US, the bank reduces its foreign lending.

We can see a few pieces of evidence looking at aggregate graphs. Figure 1 shows that deposits were rising for both groups of banks and especially in times of declining interest rates. It is also clear from Figure 2 that global banks' loans are less volatile. When interest rates declined after the Great Recession, loans of global banks responded less aggressively than total loans. Finally, Figure 3 shows that aggregate flows from foreign branches declined substantially after 2008 and increased in 2014 when rates started to soar.

Next, we evaluate the predictions of this model by empirically estimating the impact of US

<sup>&</sup>lt;sup>1</sup>In these figures, we define banks to be global if they report fund flow between domestic and foreign branches.

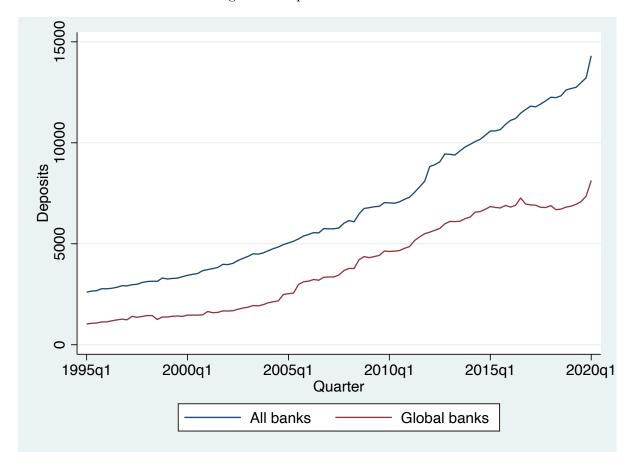


Figure 1: Deposits of the Banks

*Note:* This figure plots total deposits for US banks. Blue line corresponds to aggregate total deposits of all banks and red line — to global banks. Deposits are measured in billion of dollars.

monetary policy shocks on changes to US bank deposit rates, US bank deposits, and lending for global and domestic banks. We use high-frequency changes to Fed Funds futures around FOMC announcements to measure shocks to US monetary policy. We use weekly Ratewatch data for branch-level deposit rates, FDIC Summary of Deposits for annual branch-level deposits and quarterly US Call Report data for bank-level deposits and net transfers from foreign branches. Specifically, we use the Call Report items "Net Due to Own Related Offices in Other Countries" and "Net Due from Own Related Offices in Other Countries", the difference of which reflects the net claims from foreign offices on the domestic office. We define this difference as NetDue. As the bank increases cross-border flows into the US, NetDue increases. Our data spans years from 1994 to 2020 and covers a universe of 12,126 subsidiary banks of which 170 are global, defined as a bank for which net transfers between domestic and foreign branches are non-zero.

We start by showing that with a 1 b.p. unexpected increase to the Fed Funds rate, deposit spreads of global banks increase by 1.4 b.p. and global banks' deposit growth decreases from

1995q1 2000q1 2005q1 2010q1 2015q1 2020q1
Quarter

All banks — Global banks

Figure 2: Total Loans of the US Banks

*Note:* This figure plots total loans net of unearned income for US banks. Blue line corresponds to aggregate total loans net of unearned income of all banks and red line — to global banks. Loans are measured in billion of dollars.

2% to -8%, suggesting that the deposit channel is relevant to global banks. We also show that deposits flow out more for banks with higher spread betas, i.e. with higher sensitivity of spreads to shocks. Using the predicted changes to deposits from the first regression, we regress changes to bank lending on changes to deposits at the bank level, controlling for bank fixed effects and aggregate macroeconomic variables, including GDP and inflation. As predicted by our mechanism, we show that in response to a 1 b.p. shock to the Fed Funds rate, global banks reduce lending growth by 40 b.p. less than domestic banks. Next, we show that NetDue increases with predicted decreases to deposits. Specifically, a 1% increase in deposit growth corresponds to a decrease in cross-border flows growth of 15.28%. Finally, we run a 2SLS of changes in lending on changes to net due, instrumenting on the predicted deposit growth to show that global banks need to double their foreign transfers to finance lending in the US. Provided the recent aggregate net transfers, this could account for almost half trillion dollars

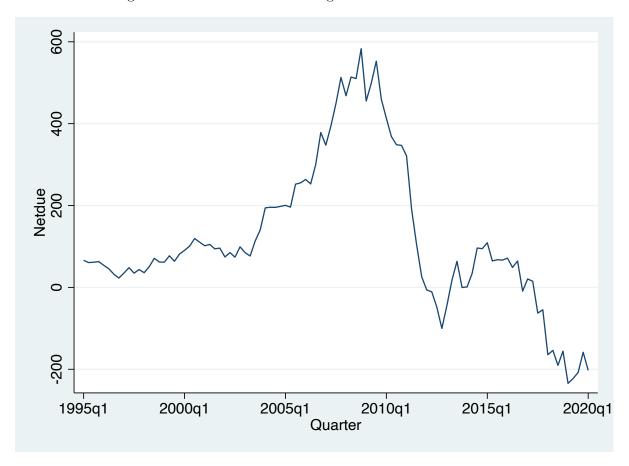


Figure 3: Net Trasfers from Foreign Branches of the US Banks

Note: This figure plots total net transfers from foreign branches for US banks. Deposits are measured in billion of dollars. We use the Call Report items "Net Due to Own Related Offices in Other Countries" and "Net Due from Own Related Offices in Other Countries". Their difference reflects the net claims from foreign offices on the domestic office and denoted by NetDue. NetDue is measured in billions of dollars.

of new cross-border flows into the US. Our results are graphically summarized in Figure 4.

We conduct branch-level analysis where deposits flow because of the market power to show that our results are not driven by alternative channels. We also show that our findings are robust to the measure of monetary policy shock (e.g. shocks or level changes), standard errors, time and banks samples, controls, and fixed effects. We refute concerns that our results are driven by a fact that global banks are large, by directly accounting for assets in regressions.

Our results indicate that the deposit channel and the market power of global banks are relevant to the transmission of US monetary policy internationally. Specifically, we show that a large portion of the international propagation of US monetary policy shock can be attributed to the deposit channel and that in response to an unexpected increase in US Fed Funds rates, when banks experience more outflow of deposits (corresponding to a larger flow beta and more market power), they increase cross border flows into the US, resulting in the greater

**United States** Foreign countries Households Households -10.4 p.p +1.4 b.p -3.1 p.p +1.4 b.p shock Global bank +1 b.p. shock +100 p.p. +100 p.p. Fed **Domestic bank** Foreign US offices offices -1 p.p. +40 b.p -1 p.p. Borrowers Borrowers

Figure 4: Graphical Summary of Results

Note: This diagram plots the summary of our main results for average banks. The Fed unexpectedly increases rates by 1 b.p. It makes both local and global bank increase their deposit spreads by 1.4 b.p. Households respond by withdrawing deposits. Banks then contract lending growth. However, global bank operates in foreign countries too, so it transfers funds from foreign offices and partly offsets lending risk.

international transmission of US monetary policy shocks.

We contribute to several strands of financial and economic literature. First, our results shed new light on the transmission of monetary policy. There are three main channels that economists considered before. The first is a reserve channel, where interest rate decisions affect required reserves and hence, lending (Bernanke and Blinder (1988, 1992); Kashyap and Stein (2000)). The second one is a capital channel — interest rate movements tight banks' capital and therefore, affect their decisions (Bolton and Freixas (2000); Brunnermeier and Sannikov (2014); Elenev et al. (2021)). Finally, recent papers argue that shocks are transmitted through banks' deposits because banks have market power in the deposit market (Drechsler et al. (2017, 2021); Wang et al. (2020)). These papers show that the deposit channel is important in the transmission of monetary policy. Wang et al. (2020) estimate a structural model and prove that the deposit channel accounts for most of the transmission. We are the first to quantify the effects of the deposit channel on international transmission. Moreover, unlike previous papers, we show that shocks are also transmitted through the deposit channel.

We also contribute to the growing literature on the international transmission of monetary

and liquidity shocks (Cetorelli and Goldberg (2012); Schnabl (2012); Acharya et al. (2014); Temesvary et al. (2018); Hale et al. (2020)). The papers not only show that shocks are transmitted across borders but also claim that global banks play a crucial role in the transmission. However, they either focus on the reserve channel or do not mention the channel at all. To our knowledge, we are the first to document the role of the deposit market on the international transmission of shocks.

Finally, we contribute to the literature on global banking. Multiple theoretical and empirical papers have shown that global banks are systematically important (Kashyap and Stein (2000); Bolton and Oehmke (2019); Bräuning and Ivashina (2020)). They focus on banks with foreign operations rather than the real effects of any domestic banks as most papers in the literature did before (Diamond (1984); Holmstrom and Tirole (1997)). We are first to ask if global banks' market power and deposits amplify the transmission of shocks abroad.

Our results may have useful policy implications. Banks are known to accelerate and even cause panics and crises (Diamond and Dybvig (1983); Goldstein and Pauzner (2005); Gennaioli et al. (2012); He and Krishnamurthy (2013); Gertler and Kiyotaki (2015); Bernanke (2018)). Global banks are usually large and their decisions and financial health have an impact on the entire economy. It is therefore very important to understand global banks' actions and impact on their assets when conducting monetary policy.

The rest of the paper proceeds as follows. Section 1 describes the stylized theoretical models where global banks choose deposits and lending in two countries. Section 3 provides information on our econometric strategy and data. Section 4 contains our main bank-level and branch-level findings. Section 5 provides robustness tests. Section 6 concludes.

### 1 Model

We consider a framework of deposits in each country which follows Drechsler et al. (2017). In each country, households have preferences over final wealth W and liquidity services l, given initial wealth  $W_0$ 

$$u\left(W_{0}\right) = \max_{W,l} \left(W^{\frac{\rho-1}{\rho}} + \lambda l^{\frac{\rho-1}{\rho}}\right)^{\frac{\rho}{\rho-1}} \tag{1}$$

where money, M, and deposits, D, are imperfect substitutes for liquidity services

$$l(M,D) = \left(M^{\frac{\epsilon-1}{\epsilon}} + \delta D^{\frac{\epsilon-1}{\epsilon}}\right)^{\frac{\epsilon}{\epsilon-1}}$$
(2)

Importantly, in Drechsler et al. (2017) (henceforth, DSS) deposits themselves are composite goods, with deposits by each bank being imperfect substitutes, where

$$D = \left(\frac{1}{N} \sum_{i=1}^{N} D_i^{\frac{\eta - 1}{\eta}}\right)^{\frac{\eta}{\eta - 1}} \tag{3}$$

To understand the behavior of a global bank we consider the static decision of a bank which demands deposits and makes lending decisions in two countries, US and UK, subject to The bank faces policy rates  $\{f_{UK}, f_{US}\}$  in each country. Following DSS and Kashyap and Stein (2000), demand for loans in each country is downward sloping,  $\ell_{UK_1}, \ell_{US_1} > 0$ . The reflects the notion that as the bank increases lending in each country, returns on loans in each country are decreasing, either due to competition for loans or because the bank reduces the quality of loans. The bank pays deposit rates  $f_{UK} - s_{UK}$  and  $f_{US} - s_{US}$ . Recall the spreads,  $s_{UK}$  and  $s_{US}$ , are the spreads between the policy rate and the rate on deposits. Finally, the bank faces convex adjustment costs to funds which it moves across borders,  $\frac{\alpha}{2}T^2$  where

$$T = L_{US} - D_{US} = D_{UK} - L_{UK} \tag{4}$$

is the amount the bank transfers from the UK to the US and  $\alpha > 0$ . This may reflect currency risk and regulatory costs and suggests that it is increasingly expensive to fund lending in one country through foreign deposits. Note that lending in each country can be expressed as a function of deposits and lending.

$$L_{US} = D_{US} + T \tag{5}$$

$$L_{UK} = D_{UK} - T \tag{6}$$

Thus, given policy rates  $\{f_{UK}, f_{US}\}$  and spreads  $\{s_{UK}, s_{US}\}$ , the bank's problem, can be expressed as a choice of deposits in each country  $D_{US}$  and  $D_{UK}$ , and transfers between each country, T.

$$\Pi = \max_{D_{US}, D_{UK}, T} \left[ f_{UK} - \left( \ell_{UK_0} + \frac{\ell_{UK_1}}{2} L_{UK} \right) \right] L_{UK} - (f_{UK} - s_{UK}) D_{UK} + \left[ f_{US} - \left( \ell_{US_0} + \frac{\ell_{US_1}}{2} L_{US} \right) \right] L_{US} - (f_{US} - s_{US}) D_{US} - \frac{\alpha}{2} T^2 \right]$$
(7)

Solving the bank's problem, the bank chooses optimal transfers

$$T = \frac{f_{US} - f_{UK} - (\ell_{US_0} - \ell_{UK_0})}{\ell_{US_1} + \ell_{UK_1} + \alpha} + \frac{\ell_{UK_1}}{\ell_{UK_1} + \ell_{US_1}} D_{UK} - \frac{\ell_{US_1}}{\ell_{UK_1} + \ell_{US_1}} D_{US}$$
(8)

We can evaluate how transfers, T, which reflect cashflows from the UK to the US, vary with policy rate  $f_{US}$ .

$$\frac{\partial T}{\partial f_{US}} = \frac{1}{\ell_{US_1} + \ell_{UK_1} + \alpha} + \frac{\ell_{UK_1}}{\ell_{UK_1} + \ell_{US_1}} \frac{\partial D_{UK}}{\partial f_{US}} - \frac{\ell_{US_1}}{\ell_{UK_1} + \ell_{US_1}} \frac{\partial D_{US}}{\partial f_{US}}$$
(9)

As  $f_{US}$  increases, the first term, reflects the increase in return from lending in the US. Since  $\ell_{US_1}, \ell_{UK_1}, \alpha > 0$ , this term is positive

$$\frac{1}{\ell_{US_1} + \ell_{UK_1} + \alpha} > 0 \tag{10}$$

Recall that in each country, i, as the policy rate  $f_i$  increases, deposit spreads increase,  $\frac{\partial f_i}{\partial s_i} > 0$ , and households withdraw deposits,  $\frac{\partial D_i}{\partial s_i} < 0$ . The rate with which households withdraw deposits depends on the bank's market power. Thus deposits in country i are decreasing in the policy rate in the same country,  $\frac{\partial D_i}{\partial f_i} < 0$ . Therefore

$$\frac{\ell_{US_1}}{\ell_{UK_1} + \ell_{US_1}} \frac{\partial D_{US}}{\partial f_{US}} < 0 \tag{11}$$

If we assume monetary policy is independent  $\frac{\partial f_{UK}}{\partial f_{US}} = 0$ , then

$$\frac{\partial D_{UK}}{\partial f_{US}} = \frac{\partial D_{UK}}{\partial f_{UK}} \frac{\partial f_{UK}}{\partial f_{US}} = 0 \tag{12}$$

so

$$\frac{\partial T}{\partial f_{US}} = \frac{1}{\ell_{US_1} + \ell_{UK_1} + \alpha} + \frac{\ell_{UK_1}}{\ell_{UK_1} + \ell_{US_1}} \frac{\partial D_{UK}}{\partial f_{US}} > 0 \tag{13}$$

and transfers are always increasing in  $f_{US}$ .

If monetary policy is positively correlated, then  $\frac{\partial D_{UK}}{\partial f_{US}} < 0$  and transfers T is increasing in

 $<sup>^{2}</sup>$ Note that if one focuses on exogenous shocks, rather than levels, policies are independent by definition of the shock.

the policy rate  $f_{US}$  so long as

$$\frac{1}{\ell_{US_1} + \ell_{UK_1} + \alpha} - \frac{\ell_{US_1}}{\ell_{UK_1} + \ell_{US_1}} \frac{\partial D_{US}}{\partial f_{US}} > -\frac{\ell_{UK_1}}{\ell_{UK_1} + \ell_{US_1}} \frac{\partial D_{UK}}{\partial f_{US}}$$
(14)

If monetary policy is negatively correlated, then  $\frac{\partial D_{UK}}{\partial f_{US}} > 0$  and thus transfers T is always increasing in policy rate  $f_{US}$ .

Now we evaluate the impact of the increase in US policy rate  $f_{US}$  on US lending and UK lending under this framework. Recall that US lending is the sum of US deposits plus transfers.

$$L_{US} = D_{US} + T \tag{15}$$

Thus as the US policy rate  $f_{US}$  increases, recall that US deposits decrease  $\frac{\partial D_{US}}{\partial f_{US}} < 0$ . If US and UK monetary policies are independent then, transfers are increasing,  $\frac{\partial T}{\partial f_{US}} > 0$ , and transfers act to mitigate the decrease in lending.

Similarly, note that UK lending is equal to UK deposits less transfers.

$$L_{UK} = D_{UK} - T \tag{16}$$

If US and UK monetary policies are independent, then UK deposits are constant, but transfers increase, decreasing UK lending.

# 2 Empirical strategy and data

Our model has three main predictions:

- Global banks increase deposit spreads and lose deposits after the contractionary monetary policy shock.
- 2. Global banks are able to use their foreign funds to finance US lending. We expect to see an increase in net transfers from foreign branches.
- 3. As a result, global banks do not contract their lending as much as domestic banks. We expect to see that the deposit channel is weaker for global banks.

We next propose an empirical strategy and data that we use to test predictions.

## 3 Empirical strategy and data

Our model has three main predictions:

- Global banks increase deposit spreads and lose deposits after a contractionary monetary policy shock.
- 2. Global banks are able to use their foreign funds to finance US lending. We expect to see an increase in net transfers from foreign branches.
- 3. As a result, global banks do not contract their lending as much as domestic banks. We expect to see that domestic transmission is weaker for global banks.

We next propose an empirical strategy to test these predictions.

#### 3.1 Empirical strategy

Our empirical strategy can be divided into two steps. First, we test if global banks are susceptible to the deposit channel, i.e. if they increase spreads and lose deposits after the contractionary monetary policy shock. While doing that, we also aim to verify classic deposit channel results for all banks. Specifically, for each bank i we run the following time-series regression:

$$y_{it} = \beta_i M S_t + \gamma_i X_{it-1} + u_{it} \tag{17}$$

where  $y_{it}$  is either a change in deposit spreads<sup>3</sup> or a log change in deposit amounts (henceforth, deposit growth),  $MS_t$  is a monetary shock, and  $X_{it-1}$  is a vector of controls that includes the growth rate of assets and macro indicators such as inflation and GDP growth. We lag controls to avoid simultaneity bias.

(17) gives us  $\beta_i$  for each bank i, or each bank's elasticity of deposit spreads and deposit growth to monetary policy shocks. We refer to  $\beta_i$  from the first set of regressions as *spread* betas (or deposit betas) and to  $\beta_i$  from the second set of regressions as flow betas. Our model predicts that spread betas should be on average positive for both domestic and global banks, while flow betas should be negative.

We also obtain fitted values from the regression of monetary shocks on deposit growth, and denote deposit growth as predicted by the deposit channel as  $\widehat{DepGrowth}_{it}$ . Importantly, we

<sup>&</sup>lt;sup>3</sup>We define deposit spreads as FFR minus deposit rate.

use fitted values of deposits rather than actual deposits because we want to evaluate the impact of monetary policy and the deposit channel, in particular, on lending and net flows. In this way we abstract from changes to deposits that are unrelated to the deposit channel but that impact lending and net flows.

In the second step of our analysis, we test if deposit outflow due to a contractionary monetary shock leads to a contraction in lending and, for global banks, to an increase in net transfers from foreign branches. To answer the first question we run the following regression:

$$\Delta \log L_{it} = \theta De\widehat{pGrowth_{it}} + \nu Global_{it} \cdot De\widehat{pGrowth_{it}} + \xi X_{it-1} + \alpha_h + \varepsilon_{iht}$$
 (18)

where the left hand side variable is the log change in lending (henceforth, lending growth),  $Global_{it}$  is a dummy that is equal to 1 if the bank i reports to have foreign branches at time t,  $X_{it-1}$  is a set of controls, and  $\alpha_h$  is a holding company fixed effects.

We are interested in 2 coefficients in (18) for which our model has direct predictions. The first is  $\theta$ , the percentage change in lending growth after 1 p.p. change in deposit growth due to the expansionary monetary shock. Our model predicts that  $\theta > 0$ , i.e. a deposit outflow leads to a contraction in lending. The second coefficient is  $\nu$ , which measures how global banks differ from domestic banks in their response to the deposit growth changes. Our model predicts  $\nu < 0$ , that is global banks' will contract lending less. To recall, given the model, global banks' lending should react less to the deposit outflow since they can use foreign funds to finance loans. Thus, we should expect to see an increase in net transfers from foreign branches. To test this, we run the following regression:

$$NetDueGr_{it} = \eta De\widehat{pGrowth_{it}} + \mu Y_{it-1} + \alpha_i + v_{it}$$
(19)

where the left hand side variable is the change in log net transfers from foreign branches (henceforth, netdue growth)<sup>4</sup>,  $Y_{it-1}$  is a set of controls, and  $\alpha_i$  is bank fixed effects. Following following Cetorelli and Goldberg (2012) we control for lagged netdue growth, contemporary and lagged GDP growth, inflation, and asset growth.

Next,  $\eta$  measures the sensitivity of netdue growth to deposit growth. Our model predicts that  $\eta < 0$ , i.e. banks increase netdue growth after the deposit outflow. We test this empirically.

<sup>&</sup>lt;sup>4</sup>Note that net transfers can be negative. That's why, we first take logs of the absolute value, then add sign, and only after that compute changes.

Moreover, we want to show that netdue growth changes due to the deposit channel lead to lending growth changes. We will address this in detail in Section 4.

#### 3.2 Data

Our data cover the universe of 12,126 banks<sup>5</sup> from 1994 to 2018. 170 banks report to have foreign branches — we define them as *global*. We next describe the data sources and variables that we use.

- 1. Bank-level quarterly data. We use Consolidated Reports of Condition and Income (US Call Reports), which are maintained by the St. Louis Federal Reserve bank, to get quarterly bank-level balance sheet data. Sample contains 24,039 banks. Notice that those are not bank-holding companies that we will refer to as BHC. One BHC may have multiple subsidiaries which we call banks. The opposite is also true one subsidiary can be owned by multiple holding companies. To be clear, we will analyze banks. Call Reports assign a unique identifier to the banks RSSD ID.
- 2. Foreign flows. We get foreign flows from Call Reports RCON series. RCON 2941 is net due to own foreign offices, Edge and Agreement subsidiaries, and IBFs and RCON 2163 is net due from own foreign offices, Edge and Agreement subsidiaries, and IBFs. The difference is what we call NetDue. Its positive value means that the bank borrowed funds from the foreign branches. Only banks that have foreign branches file RCON 2941 and RCON 2163. Hence, we define a bank i to be global at time t if its NetDue<sub>it</sub> is non-zero.<sup>8</sup> We merge RCON series with bank-level data using RSSD ID.
- 3. Monetary policy surprises. We use tick-by-tick CME Globex Fed Fund futures data to construct monetary policy surprises. They are defined as changed in futures 15 minutes before and 45 minutes after the FOMC meeting.<sup>9</sup> FOMC meetings take place 8 times

<sup>&</sup>lt;sup>5</sup>Even uninsured banks file Call Reports. Also, any depository financial institution that files Call Reports is included. Those are mainly but not only commercial banks. We will use the term 'bank' throughout the paper. <sup>6</sup>We thank Philipp Schnabl for posting and regularly updating parts of the US Call Reports.

<sup>&</sup>lt;sup>7</sup>As an additional robustness test, we restrict our sample to banks with more than \$500 million in assets to leave only medium and large commercial banks in the sample. Our results hold.

<sup>&</sup>lt;sup>8</sup>An alternative would be to define global banks as banks with non-missing figures for RCON series, hence including zero. We do not do it for two reasons. First, zero in the report can mean either that the bank has foreign branches and do not transfer funds, or that the bank do not have foreign branches. Second, even if the bank has foreign branches but do not transfer funds to or from them, that bank does not really operate globally. That's why we exclude such banks from our sample.

<sup>&</sup>lt;sup>9</sup>We thank Pascal Paul for making his data from Paul (2020) available.

per year. We convert the data into quarterly observations to make them compatible with the rest of the sources. We will refer to these surprises as monetary *shocks* as opposed to *levels*. Monetary shocks are the unexpected part of monetary policy changes. These shocks have been used in multiple papers including Bernanke and Kuttner (2005), Gertler and Karadi (2015), Gorodnichenko and Weber (2016), and Paul (2020). We also collect other shocks that have been extensively used in the literature, i.e. actual changes in FFR from FRED, Romer and Romer (2017) and Gertler and Karadi (2015) shocks from Valerie Ramey's website.

- 4. Branch-level deposits. We collect annual branch-level data on banks' deposits and assets from the FDIC Summary of Deposits. Each bank has many branches. Households and firms usually open deposit accounts within a branch that is why it is possible to observe deposits not only for parent banks but also for branches. This will let us control for county unobservables and exploit branch-level analysis. We merge SoD with Call Reports using a table that links FDIC certificate numbers with RSSD ID. The table is provided by the New York Federal Reserve bank.
- 5. Branch-level deposit rates. Weekly deposit rates by branches are available in S&P Global RateWatch. They cover almost all global banks in our sample and more than 50% of the entire sample. The data report deposit rates on new accounts. We follow Drechsler et al. (2017) and restrict our sample to 12-month certificates of deposit with an account size of \$10,000 or more, and money market deposit accounts with an account size of \$25,000. We aggregate data on the quarterly level to make it compatible with the rest of the data. We merge RateWatch data with SoD using RSSD ID and branch identifier.
- 6. County and country variables. We collect data on counties employment, wages, and population from US Census. We need these to compute the Herfindahl-Hirschman index for each county. We identify counties using fips. We can then merge county data with branch-level data using zip-fips crosswalk. Country data come from BIS, OECD, FRED, and the World Bank. We collect annual real and potential GDP, quarterly CPI inflation, FFR, GDP growth, and employment rates.

<sup>&</sup>lt;sup>10</sup>The products represent time and saving deposits, respectively.

#### 3.3 Summary statistics

Table 1 contains summary statistics of our data. Panel A represents bank characteristics. We define commercial banks as *subsidiary* banks that file Call Reports. Their main identifier is RSSD ID. We also observe their holding companies (their identifier is RSSD HCR), but we will focus on subsidiary banks in this paper unless mentioned otherwise. Our sample of banks has 685,839 bank×quarter observations. The time period is from the first quarter of 1994 to the last quarter of 2017. Only 5,185 of these observations are global banks.

We can see that global banks have on average more assets, deposits, and loans. If the bank can afford to have foreign branches, that bank is likely to be large. This fact poses potential identification concerns to our empirical strategy. That is why we control for asset growth in all our regressions. Moreover, in the robustness tests, we show that our results are robust to the inclusion of the level of assets. Hence, we are able to control for the size and make sure that our results are not driven by the simple fact that global banks are large. Another way of coping with the problem is to analyze changes rather than levels. We see from Table 1 that log changes are not significantly different across groups.

As a main measure of lending, we use loans net of unearned income for two reasons. First, we want to separate effects on lending from potential effects on unearned income that are possible if interest rates are rising. The second reason is data-driven — banks stopped reporting total loans after 2010. From 2011 banks only have to report loans net of unearned income and breakdown of loans by categories. We analyze three main loan types — commercial and industrial loans (C&I), personal loans, and real estate loans. Global banks dominate all three markets.

The table also shows net transfers from foreign branches. As mentioned above, net transfers can be negative in rare circumstances.<sup>11</sup> That is why, we first take logs from the absolute value of netdue, then add sign, and only after that take the differences. The measure is comparable to other log changes. Hence, we call it  $Netdue\ Growth$  in the paper. Of course, only global banks report netdue. Banks report non-zero netdue only if they have foreign branches. We then define the bank to be global at time t if it reports non-zero netdue at time t.

Finally, Panel A shows deposit spreads. Banks do not report their deposit rates because each branch can have its own deposit rate. However, banks report interest expenses. We thus define deposit rate in basis points as follows:

 $<sup>^{11}</sup>$ The United States remains the most important player in the financial markets. Most funds still flow *into* the US rather than out.

$$DepRate_{it} = 100 \cdot \frac{IntExp_{it}}{IntBearDep_{it}}$$
(20)

where  $IntExp_{it}$  is interest expenses and  $IntBearDep_{it}$  is the amount of interest-bearing deposits. Most deposits are interest-bearing. They only exclude checking accounts. All interest paid by banks is included in interest expenses. We can also use total deposits as a denominator. In the robustness tests, we show that results are not sensitive to the denominator in (20). We multiply the ratio by 100 to interpret rates in basis points. To convert rates into spreads, we simply subtract  $DepReate_{it}$  from the respective Fed Funds rate.

Panels B and C represent branch-level data. We focus on two deposit products — CDs and MMs. In the main branch-level analysis we leave only CDs as they are more sensitive to interest rate movements. When analyzing deposit amounts, we use FDIC data rather than Ratewatch because of the much broader sample size.

Branch-level data have two main advantages. First, we can observe deposit rates, so we do not have to use a potentially noisy measure given by (20). Second, we know branch addresses. That allows us to compute county HHIs and identify an exact effect of market power. We are unable to do it in our main bank-level analysis. HHI is computed as a squared share of deposits in the county. We then divide the number by 1000. The average HHI in our sample is 0.19. It means that the deposit market is **not** perfectly competitive. This is the crucial feature of our analysis — the deposit channel would not work in a perfectly competitive market. Imperfect competition allows branches to increase spreads without losing an entire clientele. Also, HHI for global banks is close to HHI for domestic banks, meaning that their markets are not significantly different.

#### 3.4 Measure of monetary shock

We use monetary surprises as our main measure of monetary shock. In Section 4 we also check if our results hold with actual changes in FFR. We choose surprises as our main measure for exogeneity reasons. FFR changes are driven by observable and unobservable factors that might be related to banking. That is one of the most important macroeconomic identification concerns (Nakamura and Steinsson (2018)). We successfully overcome the problem by using

<sup>&</sup>lt;sup>12</sup>The US Department of Justice considers a market with an HHI of less than 0.15 to be a competitive market, an HHI of 0.15 to 0.25 to be a moderately concentrated market, and an HHI of 0.25 or greater to be a highly concentrated market.

Table 1: Summary Statistics

|   | A      | All    | Glo    | obal    | Domestic |        |
|---|--------|--------|--------|---------|----------|--------|
|   | Mean   | Std.   | Mean   | Std.    | Mean     | Std.   |
|   |        | dev.   |        | dev.    |          | dev.   |
| Panel A: Bank characteristics (Call Rep | ports) |        |        |         |          |        |
| Total assets (mill. \$)                 | 826    | 18,549 | 73,019 | 239,307 | 460      | 5,420  |
| Total deposits (mill. \$)               | 375    | 16,126 | 40,696 | 137,319 | 165      | 12,433 |
| Interest expenses (mill. \$)            | 4.14   | 70.08  | 214.42 | 695.62  | 2.65     | 34.46  |
| Total loans (mill. \$)                  | 289    | 5,245  | 24,628 | 72,315  | 184      | 1,630  |
| Loans net of unearned income (mill. \$) | 420    | 8,626  | 37,626 | 111,923 | 232      | 2077   |
| C&I loans (mill. \$)                    | 83     | 1,368  | 6,404  | 17,647  | 57       | 647    |
| Personal loans (mill. \$)               | 43     | 1,012  | 4,075  | 13,480  | 25       | 382    |
| Real estate loans (mill. \$)            | 139    | 3,141  | 12,133 | 41,439  | 77       | 584    |
| Net transfers from abroad (mill. \$)    | 1,885  | 8,591  | 1,885  | 8,591   |          |        |
| Log deposit growth ( $\times 10^3$ )    | 16.5   | 57.4   | 20     | 86.6    | 16.5     | 57.1   |
| Log loan growth ( $\times 10^3$ )       | 20.6   | 50.7   | 21.2   | 58.2    | 20.6     | 50.6   |
| Change in log netdue ( $\times 10^3$ )  | 303    | 10,955 | 303    | 10,955  |          |        |
| Deposit spread (b.p.)                   | 2.19   | 1.98   | 1.89   | 2.12    | 2.19     | 1.98   |
| Observations (bank $\times$ quarter)    | 685    | ,839   | 5,1    | 185     | 680      | ,654   |
| Panel B: Branch characteristics (Ratew  | atch)  |        |        |         |          |        |
| Deposits (mill. \$)                     | 379    | 6,010  | 2,206  | 18,1246 | 177      | 1,695  |
| CD deposit rate (b.p.)                  | 1.58   | 1.40   | 1.24   | 1.42    | 1.65     | 1.40   |
| MM deposit rate (b.p.)                  | 0.78   | 0.87   | 0.58   | 0.83    | 0.82     | 0.87   |
| CD deposit spread (b.p.)                | -0.28  | 0.92   | 0.06   | 0.81    | -0.32    | 0.92   |
| MM deposit spread (b.p.)                | 0.66   | 1.43   | 0.91   | 1.46    | 0.64     | 1.43   |
| Branches                                | 304    | 911    | 2,339  | 1,606   | 59       | 251    |
| Observations (branch $\times$ quarter)  | 669    | ,659   | 68,    | 931     | 600      | ,728   |
| Panel C: Branch characteristics (FDIC)  | l      |        |        |         |          |        |
| Deposits (mill. \$)                     | 81     | 1,438  | 121    | 2,126   | 56       | 713    |
| _ ,                                     | 0.19   | 0.12   | 0.17   | 0.10    | 0.20     | 0.12   |
| Branch-HHI                              | U. 19  |        |        |         |          |        |
| Branch-HHI                              | 0.19   | 0.12   | 0.11   | 0.10    | 0.20     | 0.12   |

Note: This table provides descriptive statistics for banks and branches in our sample. All panels provide a breakdown into global and domestic banks. Global banks are banks that report to have foreign branches. Panel A contains statistics of bank-quarter level variables. It includes balance sheet variables and log growth (change) for deposits, loans, and netdue. Bank-level deposit spreads are computed as interest expenses divided by interest-bearing deposits. Panel B representes statistics on deposits, deposit rates, and deposit spreads for branches. It contains both CD (time) deposit rates and spreads and MM (saving) deposit rates and spreads. Panel C provides statistics on deposits of branches as reported to FDIC. Panel C also depicts Herfindahl-Hirschman indices for respective counties.

truly unexpected changes — deviations in 1-month FF futures around FOMC meetings. They include only the part of the change that was not priced by the market.

### 4 Results

We first show our main bank-level results. Specifically, we estimate equation (17) for spreads and deposits. Recall that the if the deposit channel is relevant to global banks, they will react to the contractionary monetary shocks by increasing spreads and thereby losing deposits. We then run (18) and (19) to show that global banks do not contract lending as much as domestic banks and that they transfer funds from abroad. One of the model predictions is that global banks can afford not to cut lending because they can use foreign funds. We use 2SLS to quantify the prediction. Finally, we turn to the branch-level analysis to refute the endogeneity concern—the fact that banks can shift their spreads for other reasons, not necessarily because they have market power. We address other concerns in Section 5.

#### 4.1 Bank-level results

### 4.1.1 Deposit growth and deposit spreads

We estimate equation (17) for each bank in our sample using OLS. We do not need to use VAR to mitigate reverse causality concerns here for two reasons. First, our measure of monetary policy shock is completely exogenous, hence, our LHS variables cannot impact shocks. Second, we lag controls. Omitted variable bias is unlikely to be a problem here too because, by our assumption, nothing impacts  $MS_t$ , including any unobservables that at the same time affect deposits.

Table 2 provides means and medians of estimates separately for domestic, global, and all banks. We follow Drechsler et al. (2017) and denote bank sensitivities of deposit growth to monetary shock as flow betas and sensitivity of deposit spreads as spread betas. Column 3 suggests that deposit growth declines for all banks, including global. 1 b.p. contractionary shock leads to a 3.1 units decline in log deposit growth of the average bank and 2.8 units decline in log deposit growth of the median bank. That corresponds to the decline in the growth rate from 1.7% to -1.5%. Thus, 1 b.p. shock leads to a decline in deposit amounts. Column 4 suggests that deposit spreads increase by 1.4 b.p. for the average bank and by 1.5 b.p. for the median bank following a 1 b.p. contractionary shock in the sample of all banks.

Average global bank increases its deposit spread by 1.4 b.p. and median bank increases by 1.5 b.p. Nearly the same numbers hold for domestic banks. Hence on average global banks and domestic banks are not significantly different from each other in their decision to increase spreads. It suggests that global banks have identical sources of market power as domestic banks. We explore that statement more in Subsection 4.3.

Flow betas differ across samples. Log deposit growth of the average global bank declines by 10.4 and the median bank — by 6.5. That corresponds to the decline in deposit growth from 2% to -8.6%. For domestic banks, figures are 3.1 and 2.8, respectively. For both subsets, average and median banks suffer a decline in deposit amounts. Although global banks' deposits are more sensitive to monetary shocks, we do not think that it is related to market power. First, results on deposit spreads suggest that banks' responses are nearly the same. Banks increase spreads because they have market power, and global banks increase spreads by the same amount as domestic banks. Second, global banks initially have a higher mean deposit growth rate. Finally, this can be related to the size — global banks are generally larger and grow faster.

We also find that banks with higher spread betas (i.e. with more market power) lose more deposits. We run the following regression:

$$\Delta \log D_{it} = \alpha SpreadBeta_i \cdot MS_t + \gamma X_{it-1} + u_{it}$$
(21)

where  $SpreadBeta_i$  is a spread beta for bank i. Our estimate for  $\alpha$  is equal to -0.018 and it is significant at 1% level of confidence for both domestic and global banks.<sup>13</sup> That is, banks with more market power lose more deposits, suggesting our deposit results are not driven by the deposit demand. We elaborate on this more in Section 4.3 where we account for branch-level market power using county-level HHI.

Overall, our results on deposit spreads and deposit growth suggest that contractionary monetary shock induces monopolistic banks to increase deposit spreads, and hence, they suffer an outflow of deposits. That holds for both domestic and global banks. (In contrast, in perfectly competitive markets, banks would not change their spreads in response to monetary policy, and deposits would not change. We interpret deposit flows predicted by equation (17) as a deposit growth predicted by the deposit channel. We use these fitted values in further analysis to separate the effect of monetary policy on deposits from other supply and demand-side movements.

<sup>&</sup>lt;sup>13</sup>We do not display a table with results here to save space.

Table 2: Sensitivity of Deposit Spreads and Deposit Amounts to Monetary Shocks

| $y_{it} = \beta_i M S_t + \gamma_i X_{it-1} + u_{it}$ |
|---|
|---|

| Subset   | Statistics | Flow beta      | Spread beta |
|----------|------------|----------------|-------------|
| Domestic | Mean       | $-0.031^{***}$ | 1.445***    |
|          | Median     | -0.028         | 1.513       |
| Global   | Mean       | -0.104***      | 1.353***    |
|          | Median     | -0.065         | 1.478       |
| All      | Mean       | -0.031***      | 1.444***    |
|          | Median     | -0.028         | 1.513       |

Note: This table provides mean and median flow and spread beta for domestic banks, global banks, and all banks in our sample as measured by equation (17). Column 3 depicts flow betas, i.e. estimates of (17) with log deposit growth as a LHS variable. Column 4 represents spread betas, i.e. estimates of (17) with changes in spreads as a LHS variable. Outliers are dropped at 1% level. We also conduct t-tests for means. \*\*\* above the estimate mean that we reject the hypothesis that the mean is zero at 1% level of confidence.

#### 4.1.2 Lending and net transfers

By our hypothesis, global banks respond to the deposit outflow by transferring funds from their foreign branches allowing them to reduce lending less. To test these hypotheses we estimate (18) and (19) using OLS with fixed effects. As before, we lag controls to avoid reverse causality concerns. Additionally, our independent variable is only a predicted part of deposit growth, and hence, should not be affected by the LHS variable.

We first estimate (18) for a total lending net of unearned income and separately for the three largest loan categories — C&I, personal, and real estate. Columns 1-4 of Table 3 presents results of the regression estimation. The first column represents total lending. Significant positive coefficient at  $\widehat{DepGrowth}$  implies that an increase in deposit growth as predicted by the deposit channel leads to an increase in lending growth. Recall that contractionary monetary policy cause deposits to  $flow\ out$ . Lending growth, thus, declines. The coefficients at the  $Global \cdot \widehat{DepGrowth}$  are negative and significant. It means that global banks contract lending growth less after the contractionary monetary shock. Results are in line with our predictions.

Recall that 1 b.p. shock leads to a 3.1% decrease in deposit growth of an average bank. 1% change in deposit growth leads to 0.316% change in lending growth according to Table 3. Hence, lending growth declines by 0.97% after the contractionary monetary shock. It corresponds to a decline in lending growth from 2% to 1%. Global banks contract lending growth too, but for the same deposit outflow, they would contract lending growth from 2% to 1.4%. They offset their lending losses by 40 basis points.

Table 3: Bank-level Results on Lending and Net Transfers

$$\Delta \log L_{it} = \theta De\widehat{pGrowth}_{it} + \nu Global_{it} \cdot De\widehat{pGrowth}_{it} + \xi X_{it-1} + \alpha_h + \varepsilon_{iht}$$

$$NetDueGr_{it} = \eta De\widehat{pGrowth}_{it} + \mu Y_{it-1} + \alpha_i + v_{it}$$

|                                    |             | $Dependent\ variable:$ |          |             |           |  |  |
|------------------------------------|-------------|------------------------|----------|-------------|-----------|--|--|
|                                    | Total loans | C&I loans              | Personal | Real Estate | Netdue    |  |  |
|                                    |             |                        | loans    | loans       |           |  |  |
|                                    | (1)         | (2)                    | (3)      | (4)         | (5)       |  |  |
| $\widehat{DepGrowth}$              | 0.316***    | 0.311***               | 0.312*** | 0.318***    | -15.276** |  |  |
|                                    | (0.006)     | (0.011)                | (0.014)  | (0.007)     | (7.476)   |  |  |
| $Global \cdot \widehat{DepGrowth}$ | -0.117***   | -0.155*                | -0.131   | -0.084*     |           |  |  |
|                                    | (0.040)     | (0.082)                | (0.132)  | (0.049)     |           |  |  |
| Fixed effects                      | Yes         | Yes                    | Yes      | Yes         | Yes       |  |  |
| Controls                           | Yes         | Yes                    | Yes      | Yes         | Yes       |  |  |
| Observations                       | 685,696     | 685,696                | 530,424  | 531,784     | 2,443     |  |  |
| $\mathbb{R}^2$                     | 0.156       | 0.029                  | 0.044    | 0.117       | 0.103     |  |  |

Note: This table provides results of estimation of equations (18) and (19). The first column corresponds to total lending net of unearned income. Columns 2-4 correspond to C&I, personal, and real estate loans, respectively. Column 5 shows results of Netdue regression. Independent variables are log deposit growth predicted by the deposit channel and an interaction of that variable with the indicator which is equal to 1 for global banks. There is only one independent variable in the Netdue regression because all banks in that regression are global. Standard errors for lending regressions are clustered at the bank level and displayed in the parentheses. Holding company fixed effects are included. Standard errors for netdue regression are robust. Bank fixed effects are included. \*,\*\* and \*\*\* correspond to 10-, 5-, and 1% significance level, respectively.

Banks reduce lending growth after the deposit outflow in all categories. Global banks offset commercial and real estate lending risks. However, global banks are not statistically different from domestic banks in personal loan decisions, suggesting changes in lending are supply-driven, since, if they were driven by demand, we would see differences for all types of loans, not only C&I and real estate. We elaborate more on this when we discuss alternative explanations and robustness.

We then estimate regression (19). Column 5 of Table 3 present results. The coefficient is negative and significant, suggesting that outflow of deposits leads to an increase in net transfers from foreign branches. 1% decline in deposit growth leads to 15.3% increase in netdue growth. This is both statistically and economically significant.

<sup>&</sup>lt;sup>14</sup>Magnitudes for any separate loan category seem less significant than magnitudes for total loans. We find that it is because banks reduce lending everywhere but not substantially. If we compute the sum of C&I, personal, and RE loans and estimate regressions for the sum, we find significant coefficients, analogous to the main total loans results.

Recall that, our results on lending and netdue indicate that global banks also contract lending but not to the extent of domestic banks. Global banks save 40 basis points of lending growth. At the same time, global banks transfer funds from abroad and use them as an additional source of funding which is unavailable to domestic banks. We next aim to quantify the exact amount of netdue needed to compensate for these 40 basis points of lending growth. For that we run the following regression:

$$\Delta \log L_{it} = \lambda N \widehat{etDueGr}_{it} + \kappa X_{it-1} + \varepsilon_{it}$$
(22)

We would like to test the bank's ability to able to offset some part of lending risk because they transfer funds from foreign branches. Given a contractionary monetary policy shock, under the model, banks increase their deposit spreads and suffer a decline in deposit growth. While all banks have fewer deposits and thus, less funding of assets, global banks have an additional source of funding, foreign funds, which they transfer to finance loans.

We utilize 2SLS to implement this logic. Specifically, we use predicted deposit growth  $\widehat{DepGrowth}$  as an instrumental variable for netdue growth. The first stage of that regression is equation (19). Results in Table 3 suggest that the instrument is relevant. However, the exclusion restriction here is satisfied only under the timing assumption — netdue moves first and then loans are financed. Note the exclusion restriction is not crucial to our central question — how much netdue growth do we need to offset lending risk? We can answer this question even without claiming causality.

Second-stage results are shown in Table 4. The coefficient is significant and positive indicating that, as predicted, an increase in netdue growth leads to an increase in lending growth. Recall that global banks reduce lending growth by 40 basis points less than domestic banks after a 1 b.p. monetary shock. Our results suggest that global banks need to increase their netdue growth from 0.5% to 100% to reflect this 40 basis points. The magnitude is smaller than in the first-stage regression indicating banks transfer enough funds to finance lending. The effect is economically significant. Banks have to double their net transfers to finance the rest of lending.

Table 4: 2SLS Results

|                      | Dependen              | nt variable:          |
|----------------------|-----------------------|-----------------------|
|                      | Loans                 | Loans                 |
|                      | (1)                   | (2)                   |
| $\widehat{NetDue}Gr$ | 0.003***              | 0.009***              |
|                      | (0.001)               | (0.001)               |
| Instrument           | $\widehat{DepGrowth}$ | $\widehat{DepGrowth}$ |
| Macro controls       | Yes                   | No                    |
| Bank controls        | Yes                   | Yes                   |
| Observations         | 2,450                 | 3,791                 |
| $\mathbb{R}^2$       | 0.106                 | 0.105                 |

*Note:* This table provides second-stage results of 2SLS estimation of (22). Dependent variable is log lending growth. Column 2 excludes macro variables from the list of controls. Standard errors in parentheses are clustered at the bank level. \*,\*\*, and \*\*\* correspond to 10-, 5-, and 1% significance level, respectively.

### 4.2 Alternative explanations

In the previous subsection, we presented evidence that the deposit channel impacts the international transmission of monetary shocks. Specifically, we have shown that 1 b.p. contractionary shock can lead to a 100% increase in net transfers from foreign offices. For the average bank, it is equivalent to an additional \$1.8 billion in net transfers. Provided that we have 170 global banks in the sample, total transfers can add up to \$306 billion transfer. For post-2010, the number reaches \$460 billions, and economically significant amount which may affect lending supply abroad.

Nevertheless, there are several concerns. First, our results are based on the fact that deposits flow out because banks have market power. We provide evidence that deposit spreads move, and they can only move in an imperfectly competitive market. However, deposits can flow out because of changes in lending. In other words, deposit flows might be demand-driven rather than supply-driven. If banks lend less, they do not need as many deposits. This is a lending channel of monetary policy as opposed to the deposit channel (?). We exploit our branch-level data to refute this concern in the next subsection.

Second, we focus on shocks rather than levels of monetary policy rate for exogeneity reasons. However, some deposit channel papers specifically use levels and discuss the transmission of the *policy*. We address this concern in Section 5 by repeating our analysis with monetary

<sup>&</sup>lt;sup>15</sup>For details see Drechsler et al. (2017, 2019).

policy levels rather than shocks. We expect to find no statistical difference because shocks are part of the levels. However, magnitudes are expected to be smaller.

Finally, our lending results suggest that global banks shrink lending growth less than other banks. The average bank contracts lending growth by 1 p.p. while global banks contract only by 60 b.p. They effectively offset 40 b.p. by transferring funds from foreign offices. The assumption here is that lending changes are supply-driven. An alternative explanation is that demand drives results. We believe it to be unlikely because of the following findings. Table 3 provides evidence that the growth rates of personal loans equally shrink for both domestic and global banks. However, global banks offset parts of real estate loans shrinkage. Households demand both real estate loans and personal loans. The gap that we find cannot be explained by demand but can be explained by supply. An alternative way to prove that our findings are not demand-driven is to focus on newly originated loans within counties. Community reinvestment act (CRA) provides the data on small business lending. The data have many limitations in comparison to Call Reports. Even though CRA can help us to refute demand vs. supply concerns, it will likely bring up other serious issues.

#### 4.3 Branch-level results

One of the concerns outlined above is that our results might be driven by lending channel rather than deposit channel. We address this concern by conducting branch-level analysis. We first show that branches of global banks are not different from branches of domestic banks when they change deposit spreads in response to a contractionary monetary policy shock. We interact shocks with county market power to make sure that spreads change because branches have power. We then show that branches of global banks lose the same amount of deposits as branches of domestic banks. That evidence suggests that deposit outflows are supply-driven. Finally, we aggregate predicted deposit flows at the bank level and repeat analysis with loans and netdue.

### 4.3.1 Deposit growth and deposit spreads

We run the following panel regression to find out if branches change spreads and lose deposits:

$$y_{it} = \beta M S_t \cdot Branch H H I_c + \gamma Global_{it} \cdot M S_t \cdot Branch H H I_c + \alpha_i + \theta_c + \eta_{bt} + \zeta_{st} + u_{it}$$
 (23)

where  $y_{it}$  is either a change in deposit spread or log deposit growth of branch i at time  $t^{16}$ ,  $BranchHHI_c$  is a HHI index of county c where branch i is located,  $Global_{it}$  is an indicator which is equal to 1 if branch i is a branch of the global bank at time t,  $\alpha_i$  is branch FE,  $\theta_c$  is county FE,  $\eta_{bt}$  is bank-time FE, and  $\zeta_{st}$  is state-time FE.<sup>17</sup> By our hypothesis,  $\gamma$  should be statistically indifferent from zero.

Table 5 provides estimates. The first column suggests that the coefficient at an interaction between shock and market power is negative. It implies that banks with more market power lose more deposits. This finding is in line with Drechsler et al. (2017) — deposits flow out because of the supply, not demand. If it was demand, banks with market power would never want to deliberately lose deposits. The second coefficient is insignificant. It implies that branches of global banks are not significantly different from branches of domestic banks.

Column 2 shows that banks with more market power increase CD spreads more following a contractionary monetary policy shock. Global banks are not significantly different. Finally, column 3 implies that MM spreads do not change with monetary surprises. This can be related to the fact that MM spreads are very low. Their movements are usually explained by fundamental changes in the FF rate, not by surprises. The most important fact is that global banks again are not different from domestic banks.

We next aggregate predicted deposit flows from regression (23) up to the bank level to analyze lending and netdue. We use deposit shares of branches as weights. Notice that, unlike the previous section, predicted deposit growth in this section cannot be explained by lending. All lending opportunities are included in county and state fixed effects. Market segmentations are captured by the HHI index, and deposits move because banks have market power. The branch-level analysis makes it possible to directly account for the market power. That is why all following results can be interpreted as pure deposit channel findings, not lending channel ones.

### 4.3.2 Lending and net transfers

We repeat the analysis of Section 4.1 here with predicted deposits from branch-level regressions. We denote them by  $\widehat{DepGrowthBr}$ . Notice that  $R^2$  of deposit regressions are high enough, so we fit the data well. We first ask if global banks contract lending less than domestic banks.

<sup>&</sup>lt;sup>16</sup>Regressions with deposit flows are annual because we use FDIC data.

<sup>&</sup>lt;sup>17</sup>We follow Drechsler et al. (2017) in choosing fixed effects.

Table 5: Branch-level Results on Deposits

 $y_{it} = \beta MS_t \cdot Branch HHI_c + \gamma Global_{it} \cdot MS_t \cdot Branch HHI_c + \alpha_i + \theta_c + \eta_{bt} + \zeta_{st} + u_{it}$ 

|                                    | Dependent variable: |            |            |  |  |
|------------------------------------|---------------------|------------|------------|--|--|
|                                    | Deposits            | CD Spreads | MM spreads |  |  |
|                                    | (1)                 | (2)        | (3)        |  |  |
| $MS \cdot Branch HHI$              | -0.127**            | 0.199**    | 0.177      |  |  |
|                                    | (0.060)             | (0.092)    | (0.139)    |  |  |
| $Global \cdot MS \cdot Branch HHI$ | 0.089               | 0.204      | -0.390     |  |  |
|                                    | (0.094)             | (0.220)    | (0.279)    |  |  |
| Fixed effects                      | Yes                 | Yes        | Yes        |  |  |
| Observations                       | 1,540,054           | 89,711     | 72,682     |  |  |
| $\mathbb{R}^2$                     | 0.315               | 0.834      | 0.880      |  |  |

Note: This table provides branch-level regression results for equation (23). First column runs regression with log deposit growth as a LHS variable. Second column runs regression with time deposit spreads, and third column — with saving deposit spreads. Fixed effects are branch, county, statetime, and banktime. Standard errors in parentheses are clustered at the county level. \*,\*\* and \*\*\* correspond to 10-, 5-, and 1% significance level, respectively.

We estimate the following regression:

$$\Delta \log L_{it} = \theta Dep \widehat{GrowthBr}_{it} + \nu Global_{it} \cdot Dep \widehat{GrowthBr}_{it} + \xi X_{it-1} + \alpha_i + \varepsilon_{it}$$
 (24)

where  $\alpha_i$  are bank FE.<sup>18</sup>

Next, we quantify the effect on netdue growth to understand if global banks transfer funds from abroad after contractionary monetary shock. We run the following linear regression:

$$NetDueGr_{it} = \eta Dep\widehat{Growth}Br_{it} + \mu Y_{it-1} + \alpha_i + v_{it}$$
(25)

Controls in (24) and (25) are the same as in bank-level regressions. We include bank balance sheet variables (e.g. asset growth) and macro variables. The only difference between these regressions and bank-level ones is an explanatory variable. In the current section, the explanatory variable is entirely driven by monetary shock **and** market power. All estimates thus should be interpreted as a sensitivity to the deposit channel.

Table 6 presents results. Coefficients at the  $Dep \widehat{GrowthBr}$  in the lending regressions are positive and significant. It means that deposit outflow caused by the contractionary monetary shock leads to a contraction in lending growth. Moreover, coefficients at the interaction term

<sup>&</sup>lt;sup>18</sup>We use bank FE rather than HC FE here because aggregation was specifically at the bank level.

Table 6: Results on Lending and Net Transfers: Aggregated from Branches

$$\Delta \log L_{it} = \theta Dep \widehat{Growth} Br_{it} + \nu Global_{it} \cdot Dep \widehat{Growth}_{it} + \xi X_{it-1} + \alpha_i + \varepsilon_{iht}$$

$$Net Due Gr_{it} = \eta Dep \widehat{Growth} Br_{it} + \mu Y_{it-1} + \alpha_i + v_{it}$$

|  | Dependent variable: |               |          |          |  |  |
|--|---------------------|---------------|----------|----------|--|--|
| _                                      | Loans               | Loans         | Netdue   | Netdue   |  |  |
|  | (1)                 | (2)           | (3)      | (4)      |  |  |
| $\widehat{DepGrowth}Br$                | 0.012***            | 0.013***      | -2.925** | -2.992** |  |  |
|  | (0.001)             | (0.001)       | (1.309)  | (1.212)  |  |  |
| $Global \cdot Dep \widehat{Growth} Br$ | -0.017**            | $-0.017^{**}$ |          |          |  |  |
|  | (0.007)             | (0.007)       |          |          |  |  |
| Fixed effects                          | Yes                 | Yes           | Yes      | Yes      |  |  |
| Macro controls                         | Yes                 | No            | Yes      | No       |  |  |
| Bank controls                          | Yes                 | Yes           | Yes      | Yes      |  |  |
| Observations                           | 134,042             | 134,042       | 897      | 1,089    |  |  |
| $\mathbb{R}^2$                         | 0.246               | 0.209         | 0.280    | 0.196    |  |  |

Note: This table provides results of estimation of equations (24) and (25). First and second columns correspond to total lending net of unearned income. Columns 3 and 4 show results of Netdue regression. Independent variables are log deposit growth predicted by the branch-level deposit channel and an interaction of that variable with the indicator which is equal to 1 for global banks. There is only one independent variable in the Netdue regression because all banks in that regression are global. Columns 2 and 4 exclude macro controls. Bank fixed effects are included. Standard errors in parentheses are robust. Bank fixed effects are included. \*,\*\* correspond to 10-, 5-, and 1% significance level, respectively.

are negative and significant. Hence, global banks offset lending risk. <sup>19</sup> Columns 3 and 4 suggest that the coefficients at  $\widehat{DepGrowthBr}$  in netdue regressions are negative and significant. It means that global banks transfer funds from abroad after the contractionary monetary shock.

Results above mitigate potential identification concerns and show that the deposit channel affects an international transmission of monetary shocks. Magnitudes in Table 6 are smaller than in bank-level regressions. That is because here we focus on just one source of market power — HHI, i.e. deposit shares in the county. Of course, banks and especially global banks can have other sources of market power including marketing, state law, or government. That is why our main results are bank-level. Branch-level analysis in this section proves that our results hold even in the very constrained case with just one source of the market power but at the same time these results are purely driven by the deposit channel.

Finally, we repeat the 2SLS analysis, but now we use  $Dep\widehat{GrowthBr}$ . Table 7 presents results. The coefficients are positive and significant. It means that an increase in net transfer

<sup>&</sup>lt;sup>19</sup>In these regressions it even appears that global banks offset all of the risks.

Table 7: 2SLS Results: Aggregated from Branches

|                      | Dependent variable:     |                         |  |
|----------------------|-------------------------|-------------------------|--|
|                      | Loans                   | Loans                   |  |
|                      | (1)                     | (2)                     |  |
| $\widehat{NetDue}Gr$ | 0.009***                | 0.007**                 |  |
|                      | (0.003)                 | (0.003)                 |  |
| Instrument           | $\widehat{DepGrowthBr}$ | $\widehat{DepGrowth}Br$ |  |
| Macro controls       | Yes                     | No                      |  |
| Bank controls        | Yes                     | Yes                     |  |
| Observations         | 1,336                   | 1,405                   |  |
| $\mathbb{R}^2$       | 0.276                   | 0.276                   |  |

Note: This table provides second-stage results of 2SLS estimation of (22) with DepGrowthBr as an instrumental variable. Dependent variable is log lending growth. Column 2 excludes macro variables from the list of controls. Standard errors in parentheses are clustered at the bank level. \*,\*\* and \*\*\* correspond to 10-, 5-, and 1% significance level, respectively.

growth leads to an increase in lending growth. This is in line with bank-level findings. Magnitudes are very close to what we found then.

Overall, findings in this section shed new light on the transmission of US monetary shocks abroad. We show that global banks transfer up to \$221 million following 1 b.p. contractionary monetary shock. By doing that, global banks effectively offset lending risk — they do not have to contract lending growth rates so much as domestic banks. This is strong evidence that the deposit channel impacts the international transmission of monetary shocks.

### 5 Robustness

We have already shown that our results are robust to the level of aggregation, exclusion of certain fixed effects and macro variables. In this section we show that our results are robust to the measure of monetary policy shock and to the size effects. We have evidence that our results are robust to standard errors, various sets of controls and fixed effects, exclusion of random sets of banks or quarters, denominator in (20), and measure of lending, but we leave it all beyond the text of the paper.

### 5.1 Changes in FF level

Most deposit channel papers focus on FF levels rather than shocks (Drechsler et al. (2017); Wang et al. (2020)). We are interested in the transmission of shocks, but in this section we show that our results hold for Fed Fund rate changes too. We only show bank-level results here, however, branch-level analysis produces analogous findings. We repeat the analysis of Section 4.1. Specifically, we define  $\Delta FF_t$  as a change in Fed Funds rate (henceforth, FFR) from period t-1 to t.<sup>20</sup> We first compute spread and flow betas — we estimate the following regression for each bank i:

$$y_{it} = \beta_i \Delta F F_t + \gamma_i X_{it-1} + u_{it} \tag{26}$$

where  $y_{it}$  is either a change in deposit spreads or log deposit growth. We include the same controls here as in regressions with monetary surprises.

Table 8 presents results. After a 1 b.p. change in FFR, banks increase spreads and lose deposits. The average bank increases spreads by 0.24 b.p., the average global bank — by 0.17 b.p., and the average domestic bank — by 0.24 b.p. The average bank's growth of deposits decreases by 0.6%, for the average global bank — by 1.1%, and for the average domestic bank — for 0.6%. Medians are fairly close to means.

Results above suggest two important implications. First, our first step results are robust to whether we use FF changes or monetary surprises. Second, deposit channel works for global banks not only in transmitting shocks but also in transmitting policy itself. Magnitudes in regressions with FFR are generally smaller than in regressions with surprises. It can be explained by the fact that most part of changes in FFR is anticipated by markets.

We use deposit regressions to fit predicted deposit growth as we did before. We denote the variable by  $\widehat{DepGrowth}FFR$ . We first estimate the following regression to show that our lending results are robust to the measure of monetary shock:

$$\Delta \log L_{it} = \theta Dep \widehat{Growth} FFR_{it} + \nu Global_{it} \cdot Dep \widehat{Growth} FFR_{it} + \xi X_{it-1} + \alpha_h + \varepsilon_{iht}$$
 (27)

where  $\alpha_h$  is a HC FE.

Next, we show that netdue results are also robust to the measure of monetary shock. We

<sup>&</sup>lt;sup>20</sup>We have data on FFR up to 2020 in contrast to surprises that we have only up to 2018.

Table 8: Sensitivity of Deposit Spreads and Deposit Amounts to FF Changes

$$y_{it} = \beta_i \Delta F F_t + \gamma_i X_{it-1} + u_{it}$$

| Subset   | Statistics | Flow beta      | Spread beta |
|----------|------------|----------------|-------------|
| Domestic | Mean       | $-0.006^{***}$ | 0.240***    |
|          | Median     | -0.005         | 0.245       |
| Global   | Mean       | -0.011***      | 0.166***    |
|          | Median     | -0.008         | 0.203       |
| All      | Mean       | -0.006***      | 0.240***    |
|          | Median     | -0.005         | 0.244       |

Note: This table provides statistics of estimates of equation (26). The first irst statistic is a mean, and the second one is a median. First row provides estimates for domestic banks, second — for global banks, and third — for all banks in our sample. Column 3 depicts flow betas, i.e. estimates of (26) with log deposit growth as a LHS variable. Column 4 represents spread betas, i.e. estimates of (26) with changes in spreads as a LHS variable. Outliers are dropped at 1% level. We also conduct t-tests for means. \*\*\* above the estimate mean that we reject the hypothesis that the mean is zero at 1% level of confidence.

estimate the following regression:

$$NetDueGr_{it} = \eta Dep\widehat{Growth}FFR_{it} + \mu Y_{it-1} + \alpha_i + v_{it}$$
(28)

Table 9 presents main findings. Both global and domestic banks contract lending growth after an increase in FFR. Global banks contract total and C&I lending less than domestic banks. This is in line with our main findings. Column 5 presents results on netdue. Global banks increase net transfers after an increase in FF rate. Magnitudes are very close to ones in Table 3 but still smaller. Again, the reason is that FF changes are not completely unpredictable. They are induced by market decisions and anticipated. That is one reason why we concentrate on monetary shocks and not levels.<sup>21</sup>

Finally, we repeat 2SLS analysis to quantify an exact effect of net transfers on lending, and to show that our main results are robust to the measure of monetary shock. We estimate equation (22) and use  $Dep\widehat{Growth}FFR$  as an instrument. Column 5 of Table 9 shows first-stage regression results. Table 10 displays second-stage estimates. Coefficients are positive and significant. It means that global banks use transferred funds to finance their US loans. Magnitudes are fairly close to ones in our main results. We thus can conclude that our findings are robust to the measure of monetary shock.

<sup>&</sup>lt;sup>21</sup>See Nakamura and Steinsson (2018) for more on identification issues with FFR.

Table 9: Bank-level Results on Lending and Net Transfers with FFR

$$\Delta \log L_{it} = \theta DepGrowthFFR_{it} + \nu Global_{it} \cdot DepGrowthFFR_{it} + \xi X_{it-1} + \alpha_h + \varepsilon_{iht}$$

$$NetDueGr_{it} = \eta DepGrowthFFR_{it} + \mu Y_{it-1} + \alpha_i + v_{it}$$

|   | Total loans | C&I loans | Personal | Real     | Netdue    |
|---|-------------|-----------|----------|----------|-----------|
|   |             |           | loans    | Estate   |           |
|   |             |           |          | loans    |           |
|   | (1)         | (2)       | (3)      | (4)      | (5)       |
| $Dep\widehat{Growth}FFR$                | 0.293***    | 0.228***  | 0.231*** | 0.234*** | -12.297** |
|   | (0.005)     | (0.011)   | (0.012)  | (0.006)  | (6.167)   |
| $Global \cdot Dep \widehat{Growth} FFR$ | -0.154***   | -0.082*   | -0.025   | 0.018    |           |
|   | (0.033)     | (0.048)   | (0.104)  | (0.060)  |           |
| Fixed effects                           | Yes         | Yes       | Yes      | Yes      | Yes       |
| Controls                                | Yes         | Yes       | Yes      | Yes      | Yes       |
| Observations                            | 775,087     | 733,522   | 541,288  | 543,660  | 3,506     |
| $\mathbb{R}^2$                          | 0.133       | 0.028     | 0.040    | 0.107    | 0.077     |

Note: This table provides results of estimation of equations (27) and (28). First column corresponds to total lending net of unearned income. Columns 2-4 correspond to C&I, personal, and real estate loans, respectively. Column 5 shows results of Netdue regression. Independent variables are log deposit growth predicted by the deposit channel with FFR and an interaction of that variable with the indicator which is equal to 1 for global banks. There is only one independent variable in the Netdue regression because all banks in that regression are global. Standard errors for lending regressions are clustered at the bank level and displayed in the parentheses. Holding company fixed effects are included. Standard errors for netdue regression are robust. Bank fixed effects are included. \*,\*\* and \*\*\* correspond to 10-, 5-, and 1% significance level, respectively.

#### 5.2 Large vs. small banks

There is evidence in literature that large and small banks differently transmit monetary policy (Kashyap and Stein (2000)). Large banks are believed to use their balance sheet to smooth the transmission. In our case, it would mean that large banks will be more resistant to lending cuts. It is a concern because most global banks are indeed large. In this section we show that our results are not driven by the size.

In all previous regressions we added only log growth of assets as a control variable. The reason is to keep only growth variables in our regressions. We now include log of assets as a control variable to account for the size directly. We do it only for second step regressions (loans and netdue), because for the first step we estimated betas separately for each bank. We estimate the following regression for lending:

Table 10: 2SLS Results with FFR

|                      | $Dependent\ variable:$   |                          |  |
|----------------------|--------------------------|--------------------------|--|
|                      | Loans                    | Loans                    |  |
|                      | (1)                      | (2)                      |  |
| $\widehat{NetDue}Gr$ | 0.004***                 | 0.007***                 |  |
|                      | (0.001)                  | (0.001)                  |  |
| Instrument           | $\widehat{DepGrowth}FFR$ | $\widehat{DepGrowth}FFR$ |  |
| Macro controls       | Yes                      | No                       |  |
| Bank controls        | Yes                      | Yes                      |  |
| Observations         | 3,512                    | 5,161                    |  |
| $\mathbb{R}^2$       | 0.115                    | 0.109                    |  |

Note: This table provides second-stage results of 2SLS estimation of (22) with DepGrowthFFR as an instrumental variable. Dependent variable is log lending growth. Column 2 excludes macro variables from the list of controls. Standard errors in parentheses are clustered at the bank level. \*,\*\*, and \*\*\* correspond to 10-, 5-, and 1% significance level, respectively.

$$\Delta \log L_{it} = \theta \widehat{DepGrowth_{it}} + \nu Global_{it} \cdot \widehat{DepGrowth_{it}} + \zeta \ln A_{it} + \xi X_{it-1} + \alpha_h + \varepsilon_{iht}$$
 (29)

where  $A_{it}$  are total assets of bank i at time t. We run the following regression for netdue:

$$NetDueGr_{it} = \eta \widehat{DepGrowth_{it}} + \iota \ln A_{it} + \mu Y_{it-1} + \alpha_i + v_{it}$$
(30)

We expect  $\zeta$  to be negative consistent with Kashyap and Stein (2000). We do not have expectations regarding  $\iota$  because in regressions with netdue large banks are already self-selected, so  $\iota$  is unlikely to be significant.

Table 11 presents results. The main coefficients are robust to size. Banks, both large and small, contract lending growth after the monetary shock. Global banks offset parts of the risk. As we expected, size significantly negatively affects the transmission, i.e. large banks can smooth the pass-through. Column 5 suggests that size does not impact foreign transfers. The reason is that global banks are already self-selected to be large. For the same reason, we do not find any significant effect of assets in 2SLS regressions.<sup>22</sup> Overall, we conclude that our main findings are robust to the inclusion of size and are not driven by the fact that global banks are large.

<sup>&</sup>lt;sup>22</sup>We do not show 2SLS results with size here.

Table 11: Bank-level Results on Lending and Net Transfers with FFR

$$\Delta \log L_{it} = \theta De\widehat{pGrowth}_{it} + \nu Global_{it} \cdot De\widehat{pGrowth}_{it} + \zeta \ln A_{it} + \xi X_{it-1} + \alpha_h + \varepsilon_{iht}$$

$$NetDueGr_{it} = \eta De\widehat{pGrowth}_{it} + \iota \ln A_{it} + \mu Y_{it-1} + \alpha_i + v_{it}$$

|                                    | $Dependent\ variable:$ |           |           |             |           |  |
|------------------------------------|------------------------|-----------|-----------|-------------|-----------|--|
|                                    | Total loans            | C&I loans | Personal  | Real Estate | Netdue    |  |
|                                    |                        |           | loans     | loans       |           |  |
|                                    | (1)                    | (2)       | (3)       | (4)         | (5)       |  |
| $\widehat{DepGrowth}$              | 0.325***               | 0.303***  | 0.300***  | 0.311***    | -15.349** |  |
|                                    | (0.006)                | (0.011)   | (0.014)   | (0.007)     | (7.502)   |  |
| $Global \cdot \widehat{DepGrowth}$ | $-0.147^{***}$         | -0.041    | 0.159     | 0.077       |           |  |
|                                    | (0.043)                | (0.098)   | (0.143)   | (0.060)     |           |  |
| $\ln A$                            | -0.012***              | -0.009*** | -0.014*** | -0.007***   | -0.150    |  |
|                                    | (0.000)                | (0.000)   | (0.001)   | (0.000)     | (0.668)   |  |
| Fixed effects                      | Yes                    | Yes       | Yes       | Yes         | Yes       |  |
| Controls                           | Yes                    | Yes       | Yes       | Yes         | Yes       |  |
| Observations                       | 685,838                | 685,696   | 530,424   | 531,784     | 2,443     |  |
| $\mathbb{R}^2$                     | 0.156                  | 0.030     | 0.046     | 0.120       | 0.103     |  |

Note: This table provides results of estimation of equations (29) and (30). Assets are included as a control variable in order to account for the size. First column corresponds to total lending net of unearned income. Columns 2-4 correspond to C&I, personal, and real estate loans, respectively. Column 5 shows results of Netdue regression. Independent variables are log deposit growth predicted by the deposit channel with FFR and an interaction of that variable with the indicator which is equal to 1 for global banks. There is only one independent variable in the Netdue regression because all banks in that regression are global. Standard errors for lending regressions are clustered at the bank level and displayed in the parentheses. Holding company fixed effects are included. Standard errors for netdue regression are robust. Bank fixed effects are included. \*,\*\* and \*\*\* correspond to 10-, 5-, and 1% significance level, respectively.

### 6 Conclusion

In this paper, we contribute to the understanding of how bank deposits and market power impact the international transmission of monetary policy. Given the large academic and political interest in understanding how US monetary policy is transmitted internationally, we consider the question of whether the deposit channel, which recent literature has argued is a major channel for the transmission of monetary policy domestically, has implications to the internal decision by the global banks to allocate resources across borders, and to the transmission of monetary policy. To understand the bank decision, we suggest a static model of a global bank which operates in two countries and has market power in the deposit market. We show that under this framework, when the policy rate in one country increases, like domestic banks, global

banks increase their deposit spreads. The increase in spreads is commensurate to their market power in the deposit market, with high market power banks optimally choosing a spread which results in a larger outflow of deposits. However, unlike domestic banks, global banks optimally choose to transfer funds from foreign branches, thereby reducing domestic lending less than domestic banks but also reducing foreign lending.

We then evaluate the predictions of this framework empirically. We confirm that the deposit channel holds for both domestic and global banks and for both expected and unexpected movements in the Fed Funds rate. Moreover, the decreases in deposits as predicted by unexpected changes to the Fed Funds Rate, reduce lending for both domestic and global banks, but as predicted, global banks reduce lending less. This predicted decline in deposits also increases net flows by global banks into the US. Finally, we show that increased net flows, as instrumented by changes to predicted deposits, correspond to an increase in lending by global banks.

The results of this paper suggest that understanding bank market power is critical in understanding the transmission of US monetary policy, both domestically and abroad. Relative to a closed economy, an increase in bank market power further dampens the impact of US monetary policy domestically and amplifies the transmission internationally. This paper raises a few clarifying questions. First, we show that global banks fund domestic operations through foreign flows, and this impact US lending. A quantitative understanding of the impact on foreign lending and foreign business activity, including investment and trade, would be important in understanding the full impact of the deposit channel on foreign lending and the portion of the international transmission of US monetary policy that can be attributed to the deposit channel. Next, a clear evaluation of the relative impact of the deposit channel and the reserve channel in the international transmission of monetary and liquidity shocks would provide policymakers with a better understanding of how monetary policy is transmitted and which parties are most impacted. Finally, this framework abstracts from concerns of currency and capital controls, both of which may be relevant to bank decision-making, especially since foreign lending funded through domestic deposits introduces the risk of currency mismatch in addition to duration mismatch. We leave these questions for future research.

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