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DEPOSIT INFLOWS AND OUTFLOWS IN FAILING BANKS:
THE ROLE OF DEPOSIT INSURANCE

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Deposit Inflows and Outflows in Failing Banks: The Role of Deposit Insurance

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ABSTRACT

Using unique, daily, account-level data, we investigate deposit outflows and inflows in a distressed bank. We observe an outflow of uninsured depositors following bad regulatory news. Both regular and temporary deposit insurance reduce outflows. We provide important new evidence that, simultaneous with deposit outflows, deposit inflows are first order. Uninsured deposit outflows were largely offset with new insured deposit inflows as the bank approached failure, with the bank increasing term deposit rates. This phenomenon holds in a large sample of banks that faced regulatory action, suggesting that insured deposit inflows are an important mechanism that weakens depositor discipline.

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Bank failures are important. Many banks failed during and after the financial crisis of 2007 to 2009. The inability of financial institutions to maintain stable funding sources was central to the crisis, which resulted in the high-profile failure or near failure of many financial institutions and unprecedented emergency liquidity support by governments around the world.

Since the Great Recession, much attention has been paid to systemically important institutions. However, looking closely at the roughly 500 banks that failed during the Great Recession, most were smaller banks and, individually, very few were systemic (Granja, Matvos, and Seru, 2017). However, the large number of such bank failures during the crisis was very costly to the Federal Deposit Insurance Corporation (FDIC) Deposit Insurance Fund (DIF), which backs the insured deposit provision of banks. From 2008 to 2013, almost 500 banks failed at a cost of approximately \$73 billion to the DIF, with the DIF falling to a negative \$20.9 billion by year-end 2009 (FDIC, 2017).¹ Understanding these bank failures and the drivers of deposit flows is therefore important in its own right.

In this paper we access a unique, highly granular dataset collected by the FDIC from a failed U.S. bank² to examine the drivers of both deposit inflows and outflows. Most of the attention surrounding bank failures and distressed banks has been on deposit outflows. In theory, one of the most important factors for deposit outflows should be deposit insurance. Large deposits are uninsured. In principle, these deposits provide discipline. However, what about deposit inflows? While the outflow of large, uninsured depositors can impose some depositor discipline, if a bank is able to easily attract new insured deposits, then this suggests very little depositor discipline. The theoretical model of Egan, Hortaçsu, and Matvos (2017)

¹The DIF went negative on an accounting basis (incorporating the contingent loss reserve for expected future bank failures), but did not deplete its liquid assets.

²Throughout, unless otherwise noted, we use the term “bank” to refer to any insured depository institution, whether it be a commercial bank, thrift, or credit union. We use the term “financial institution” when needed, which includes the term “bank,” as well as institutions such as nonbank finance companies, insurance companies, hedge funds, and other entities often referred to as “shadow banks.”

suggests that distressed banks may increase deposit rates to attract insured depositors, but there has been little systematic evidence to suggest that distressed banks behave this way and that depositors respond to these rates meaningfully.³ This paper fills the gap by examining deposit inflows in depth.

To study these issues, we first use our uniquely detailed dataset. These data allow us to measure daily, account-level balances and attributes for several years. Importantly, unlike aggregated data such as Call Reports, these data allow us to separately assess inflows and outflows - a distinction that turns out to be of first-order importance. We find that simultaneous with large deposit outflows, the bank experienced large deposit inflows - just below the deposit insurance limit - with the bank raising its interest rates. To shed light on whether this phenomenon is observed more generally, we then study a large panel of U.S. banks that faced regulatory actions. We find that large insured deposits with these banks increased after enforcement actions, similar to the phenomenon that we observe in the failed bank we study. We further gather data on deposit interest rates offered by large banks before and after regulatory actions through survey data in *RateWatch*. We find that the larger banks facing regulatory action raise interest rates, suggesting that the pattern of interest rate hikes and deposit inflows that we see in the failed bank under study is indeed a broader phenomenon.

Expanding on our analysis in greater detail, we first examine deposit outflows in our daily frequency data. We find that FDIC insurance and other government guarantees, including temporary measures such as the Transaction Account Guarantee (TAG) program, significantly reduce the withdrawals of insured depositors in response to ailing bank health. These results support the view that deposit insurance - even temporary measures that one might worry are not well understood by depositors - improves funding stability. Uninsured

³The evidence has been mainly anecdotal, focusing on a single example (see, for example, the Ally bank example cited in Egan et al., 2017), but no prior study has documented a systematic response by depositors.

depositors, in contrast, are much more likely to withdraw their funds. We additionally find account characteristics that are related to deposit stability, in that checking accounts are more stable than savings accounts, and depositors receiving regular direct deposits such as payroll are less likely to leave the bank. We also find that depositors who have been with the bank longer are less likely to exit, even when faced with bad regulatory news. Finally, we find that when uninsured depositors leave the bank under stress, they typically withdraw a large share even of *insured* funds. This result is especially relevant for financial stability, as even a substantial share of banks' insured funding may flee in response to bad news.

We next study deposit inflows. Surprisingly, the failed bank that we study was able to replace about a third of its deposit base in the last year of its life, while it was publicly known to be under supervisory scrutiny for its declining condition. About half of these new deposits arrived in the last 90 days before failure, after public regulatory reports showed the bank was critically undercapitalized. The bank attracted these deposits largely from small financial institutions around the United States, most with no previous relationship, using internet deposit listing services.⁴ The new deposits were almost all term deposits paying above-market interest rates and structured to fall just under the FDIC insurance limit.

To examine whether this phenomenon of insured deposit inflows to the bank in distress holds more broadly in other U.S. banks, we use a mix of public data and confidential supervisory data. We identify over 2,000 other banks that were subject to regulatory actions similar to the one faced by our subject bank. We show that these banks substitute funding away from brokered deposits and time deposits over \$250,000, but increase their reliance on

⁴Listing services are firms that provide potential depositors with a list of deposit rates (one for each participating bank) for a number of standard deposit products, often sorted from the highest rate to the lowest and displayed on a website. For supervisory purposes, being classified as a listing service need not imply that one is a third-party deposit placement service, although some deposits may fall in both categories. For the sake of clarity we therefore use the terms “placed deposit” and “listing service deposit” as mutually exclusive terms unless otherwise noted; if a deposit could be considered both, we code it as a placed deposit. For more details on the supervisory definition and treatment of listed, placed, and brokered deposits, see the FDIC’s “Frequently Asked Questions on Identifying, Accepting and Reporting Brokered Deposits,” FIL-42-2016, June 30, 2016.

listing service deposits and time deposits below \$250,000. Banks especially increase reliance on term deposits with balances above \$100,000 but below \$250,000, the range in which term deposits structured to fall just under the insurance limit would be. Another question that arises is whether it is just small banks (which comprise the majority of failed banks) that raise interest rates, or whether larger banks do so as well. To address this question, we gather additional detailed data from *RateWatch* surveys of interest rates offered by the largest banks facing regulatory action. Consistent with the behavior of the failed bank that we study and the predictions of Egan et al. (2017), we find that large banks facing regulatory action increase the interest rates offered on insured term deposits. These results suggest that our findings from the detailed micro data hold more broadly in U.S. banks facing enforcement actions related to undercapitalization or distress.

To the best of our knowledge, we are the first to show that in the presence of government guarantees, gross funding *inflows* are of first-order impact - even in imminently failing banks. This insight has significant implications for depositor discipline. In particular, while large, uninsured deposits provide depositor discipline through outflows, if banks can easily attract new insured deposits, then in effect there is little depositor discipline. Our results suggest that inflows into insured deposits are an important mechanism that weakens depositor discipline exerted by uninsured depositors. Thus, to understand bank runs, it is not enough to look at net outflows (which is what is reported in the Call Reports) - one needs to look at gross outflows and inflows separately.

Further, the source of new deposits for the failed bank that we study suggests that depositors' knowledge of the mechanics of bank failure and deposit insurance is possibly asymmetric. Unlike in many other countries, FDIC insurance makes bank failure essentially seamless for depositors – if the bank closes on Friday, insured depositors have always been able to access their funds by or before the following Monday. However, as our subject bank approached failure, even yields on riskless insured deposits rose, possibly reflecting

an “inconvenience yield” compensating depositors for the potential inconveniences of bank failure. Sophisticated depositors (namely, other insured banks and credit unions) flow into the bank to capture this premium while ordinary depositors stay away.

Our paper relates to several strands of literature. Much of the empirical literature on bank runs employs aggregate data. This literature generally finds that banks with worse fundamentals experience greater deposit withdrawals in a crisis (Gorton (1988), Saunders and Wilson (1996), Calomiris and Mason (1997)), and that large amounts of uninsured deposits can lead to unstable banks (Egan, Hortaçsu, and Matvos (2017)). A small set of papers examines responses of individual deposit(or)s to bank runs. These papers use either snapshots of data (Davenport and McDill (2006)) or data from banks in other countries, such as India (Iyer and Puri (2012), Iyer, Puri, and Ryan (2016)), Denmark (Iyer et al. (2019)), and Switzerland (Brown, Guin, and Morkoetter (2020)). The theoretical model of Egan et al. (2017) provides a framework in which to understand our results. For example, their model predicts that uninsured depositors flee distressed banks, distressed banks pay higher rates to attract depositors, and insured depositors respond little or even flow *into* distressed banks - all of which we document empirically. The deposit flows that we find are related to the causes of bank funding instability, the rationale for deposit insurance, and the literature on panic- versus fundamentals-driven bank runs (Diamond and Dybvig (1983), Jacklin and Bhattacharya (1988), Gorton (1988), Saunders and Wilson (1996), Calomiris and Mason (1997)). Finally, several prominent scholars have recently highlighted the dearth of academic empirical research on the stability and liquidity of various funding sources, in spite of liquidity concerns in the last financial crisis (Diamond and Kashyap (2016), Allen and Gale (2017)).

The remainder of the paper is organized as follows. Section I. provides a brief history of the bank to highlight our key findings and provide context for later analysis. Section II. presents regression results on the drivers of deposit liquidation, Section III. presents results

on inflows of new depositors, and Section IV. generalizes our qualitative findings to a large sample of U.S. banks. Finally, Section V. concludes.

I. Data Description and Background

We construct our dataset from data collected by the FDIC shortly after the bank’s failure, building daily account balances for each deposit account. We are able to reliably construct daily, account-level deposit balances from early 2006 until the bank’s failure. We observe all account transactions over this period at a granular level. We check our constructed account balances against Call Reports on total deposits and deposits by account category, and deposits by branch against the FDIC Summary of Deposits data.⁵ General summary statistics about these bank accounts are shown in Table I, broken out to highlight periods of special interest, as we discuss further in Section A. below.

[Place Table I Approximately Here]

Until mid-2007, this bank appeared relatively healthy. The bank had approximately \$2 billion in assets and primarily made residential real estate loans, but it also offered wealth management services. The bank sourced deposits from physical branches as well as internet depositors with a focus on savings accounts. The balances in accounts with some uninsured funds, both transaction and term deposits,⁶ were steadily rising (see Figure 1).⁷

[Place Figure 1 Approximately Here]

⁵We also, for example, check individual accounts to ensure that accounts had a zero balance before account opening and after closing, which ensures no transactions were missed.

⁶Throughout this paper, we use the phrase “transaction account” to refer to all non-maturity accounts, that is, both checking and savings accounts. We acknowledge that the phrase “transaction account” has a more precise meaning in certain contexts, such as in the Federal Reserve’s Regulation D. Additionally, we use “term deposit” and “certificate of deposit” (CD) as synonyms. We classify accounts as insured or uninsured using an approximation of the deposit insurance process, aggregating deposit accounts based on their owners and ownership categories before applying the deposit insurance limit for each category.

⁷While the bank employed Federal Home Loan Bank (FHLB) advances as funding sources, we find no evidence that it took discount window loans or Troubled Asset Relief Program (TARP) funds.

By mid-2007, there were signs of the growing financial-system-wide stress at this bank. Insured term balances fell as depositors shifted to other asset categories and the bank managed the size of its wholesale CD portfolio to purchase or sell residential real estate. Between this time and August 2008, there was net runoff in uninsured balances. Figure 1 shows that this was particularly rapid among uninsured term deposits. While less than 40% of uninsured transaction balances ran off during the period, over 50% of uninsured term deposit balances did so. There was comparatively little change in insured deposits. Perhaps because of the inherently forward-looking nature of term deposits, or perhaps because term depositors are relatively sophisticated, term deposits appear more risk sensitive than transaction deposits. While this period excludes the worst of the financial crisis, stress was building in the financial sector.⁸ Thus, depositors, particularly sophisticated depositors, began to react. During this period, the bank significantly curtailed its residential lending activities. Thereafter and up until failure, the bank continued to make commercial and personal loans, along with some residential mortgage loans.

Fall 2008 saw severe credit and liquidity risks realized across the financial system as well as significant changes in financial policy. The most important policy change for our purpose was the increase in the FDIC's deposit insurance limit from \$100,000 to \$250,000 effective October 3, 2008. In addition, the FDIC's TAG program became effective on October 14, 2008, covering all categories of checking accounts.⁹ The change in deposit insurance is evident in Figure 1, where uninsured deposits dropped and insured deposits jumped between the Pre-Crisis and Post-Crisis periods denoted with grey bars. Much of the sudden change in balances by insurance status is mechanical, as deposit accounts between \$100,000 and

⁸Bear Stearns and IndyMac failed, Northern Rock experienced a run, and some investment funds froze withdrawals.

⁹Initially, the deposit insurance limit increase was through the end of 2010, but it was made permanent by the Dodd Frank Act. TAG temporarily provided unlimited deposit insurance for negotiable order of withdrawal (NOW) accounts, non-interest-bearing demand deposit accounts, and interest on lawyer trust accounts (IOLTAs), which cover all categories of checking accounts.

\$250,000 suddenly became insured. Much of the remaining change among transaction accounts reflects the almost simultaneous application of TAG guarantees. In contrast, changes in term deposit balances were driven in part by the bank's rapid acquisition of placed deposits, as shown in Figure 2 (which we discuss at greater length in later sections). Further supporting the view that term depositors at the bank were more sophisticated and risk sensitive, uninsured term balances never increased substantially after October 2008.

[Place Figure 2 Approximately Here]

The inflow of uninsured transaction deposits suggests that the time immediately after the financial crisis was one of limited stress at this bank. The acute system-wide stress of the crisis had receded and the bank's health had not yet significantly deteriorated.¹⁰ About a year before the bank's failure, its primary federal regulator took its first publicly announced action to address the declining health of the bank through a Cease and Desist (C&D) order.¹¹ The C&D order cited many issues at the bank, including insufficient capital and poor board and management oversight, and was made public immediately, appearing in the local press within a couple of business days.¹² Unsurprisingly given the negative attention on the bank, Figure 1 shows an increase in aggregate runoff for transaction accounts - both insured and uninsured. As noted above, few uninsured term deposits remained at the bank, but the few that remained still responded to the news.

Finally, three to four months before the bank failed, the bank's public regulatory filings (including amendments to previous filings) began showing the bank to be "significantly

¹⁰Acharya and Mora (2015) document a similar inflow of deposits into the banking system after government actions in late 2008, suggesting that the actions reaffirmed markets' confidence in the financial safety net and thus the safety of the financial system (see also Pennacchi (2006) and Gatev and Strahan (2006)).

¹¹The bank had previously been subject to a non-public memorandum of understanding (MOU) as well as a subsequent troubled condition letter (TCL). These were intended to address many of the same problems that led to the bank's demise. Such confidential, informal enforcement actions are a common element of regulators' response to ailing bank health in earlier stages, when failure is relatively unlikely.

¹²The C&D order was described by a banking analyst quoted in the local press as unusually harsh and indicative of high supervisory concern about the bank. Reports in the local press also remarked on the bank's poor health as revealed by its financial ratios in a recent public regulatory report.

undercapitalized” and, within weeks, “critically undercapitalized.” Importantly, Prompt Correction Action (PCA) guidelines generally require that federal regulators place a bank into receivership or conservatorship (i.e., fail the bank) within 90 days of it becoming critically undercapitalized.¹³ With depositors therefore expecting the bank to fail soon, uninsured deposit runoff accelerated substantially, as shown in the far right of Figure 1, Panel A.

Ultimately the bank failed, and its primary federal regulator concluded that its failure was a result of heavy credit losses on the loan portfolio, especially adjustable rate mortgages. Resolution of the bank cost the FDIC approximately 10% of the bank’s assets. For context, of the 54 banks with assets between \$1 and \$10 billion that failed¹⁴ between 2007 and 2014, the average cost was 18% of bank assets with a right skew, placing this bank’s losses in the middle third of the loss rate distribution.

A. Defining Time Periods of Special Relevance

For the purpose of our empirical analysis, we separately analyze depositor behavior in each of four time periods available to us, as described below. We identified these periods using the bank’s data and macroeconomic events. In reverse chronological order, the time periods are as follows:

- *Formal Enforcement Action.* The *Formal* period was a period of significant bank-specific distress and represents the primary period of interest. This period began with the C&D order (a formal enforcement action) and ended with the failure of the bank. Unlike earlier periods, the stress arose from bank-specific adverse information, rather than from system-wide anxiety.
- *Post-Crisis.* The *Post-Crisis* period began in December 2008, shortly after the govern-

¹³The term “critically undercapitalized” is defined by law as the lowest of five ranges for bank capitalization ratios. Banks are considered critically undercapitalized if their leverage ratio falls below 2%, which is nearly insolvent in book value terms. See 12 U.S.C. §1831o for more details on PCA guidelines.

¹⁴We exclude open bank assistance (OBA) from our definition of failure in computing this statistic.

ment's emergency actions in fall 2008, and continued until the end of May 2009. The Post-Crisis period was a period of considerable distress across the financial system. Unlike in the Formal period, there were no significant revelations of bank-specific trouble. Hence, the Post-Crisis period allows us to compare depositor behavior in response to market-wide stress. We exclude a few months in fall 2008 to avoid the confounding effects of emergency actions by the U.S. government, as well as markets' expectations related to those actions.

- *Pre-Crisis.* The next period we focus on is the year-long period ending just before September 2008, that is, just before the financial crisis. As discussed above, uninsured deposits began running off during this period, particularly uninsured term deposits.
- *Placebo.* We use a period in 2006 as a placebo period, establishing baseline depositor behavior when neither the bank nor the financial system were perceived to be especially troubled.

II. Analysis of Deposit Outflows

This section presents an analysis using several regression models. As shown in equation (1), we cross-sectionally regress a dummy indicating whether an account liquidates¹⁵ on a variety of account and depositor characteristics in the context of ordinary least squares (OLS) estimated linear probability models (LPMs). Because the liquidation behavior of term deposits is quite different from that of transaction deposits, we run regressions separately on the two categories for each of the four periods described above - Placebo, Pre-Crisis,

¹⁵An account is considered to liquidate if its balance falls by 75% or more relative to the balance at the beginning of the period, and if the balance stays at or below 25% of the starting balance for at least 61 days. See Table II.

Post-Crisis, and Formal Enforcement Action:

$$\mathbb{1}(Liquidation_i) = \alpha + \beta Controls_i + \delta_b + \varepsilon_i. \quad (1)$$

Subscript i indexes deposit accounts, and δ_b denotes a set of fixed effects for the bank branch with which the bank associated each deposit account. All variables in $Controls_i$ are defined in Table II and include a variety of account and depositor characteristics including insurance coverage, account type, and historical transaction behavior. In addition, note that TAG and Dodd Frank Act (DFA) deposit guarantees were effective only after their implementation in the crisis; coverage under these temporary programs was available only in our Post-Crisis and Formal periods. That said, we include a *TAG/DFA Eligible* dummy in all regressions; estimates for the Placebo and Pre-Crisis periods are for baseline/comparison purposes only.¹⁶

[Place Table II Approximately Here]

A. Drivers of Transaction Deposit Outflows

Focusing first on transaction deposits, Table III presents our baseline LPM regressions, one for each of the four periods of interest.¹⁷ Table IV presents Probit and Cox proportional

¹⁶Note that TAG/DFA coverage was designed to effectively remove covered accounts from a depositor's exposure to the bank for the purposes of determining insurance coverage subject to the \$250,000 limit. This means that TAG/DFA could indirectly result in other accounts gaining insurance coverage, even if they were not directly guaranteed under TAG or DFA. Further, TAG ends and DFA guarantees begin midway through the Formal period, with the ultimate effect that NOW accounts are no longer covered by unlimited insurance. Given that TAG's expiration was known in advance, we may expect depositors in large NOW accounts to liquidate balances prior to the scheduled end of their deposit guarantees. This would generate a positive relationship between NOW status and liquidation at the same time that non-interest checking and IOLTA accounts (both still covered by DFA) may show a negative relationship. Thus, we only mark accounts covered by the DFA guarantees (but not TAG) with 1 for this dummy in the Formal period.

¹⁷We use asymptotically normal standard errors to measure significance, but our findings are qualitatively robust to clustering standard errors at the branch (office) level. See Tables IA.VII and IA.VIII in the Internet Appendix. The Internet Appendix is available in the online version of the article on the *Journal of Finance* website.

hazard model estimates for the Post-Crisis and Formal periods only, demonstrating that our results are robust to various model specifications.¹⁸ In the discussion below, we refer to the LPM model results (Table III) unless otherwise noted.

[Place Table III Approximately Here]

The Placebo period in 2006 establishes a baseline for “normal” depositor behavior with little financial stress. First, we find that the *Uninsured* dummy is significant, implying that uninsured accounts liquidate about 2.5 percentage points more often than the baseline hazard, meaning insured deposits are more stable than uninsured deposits. Second, the *TAG/DFA Eligible* dummy is not statistically different from zero. This result is perhaps unsurprising given that TAG and DFA were not yet in effect, and serves as a baseline against which to assess the impacts of TAG - during and after the crisis, this set of accounts was covered by the temporary, unlimited FDIC insurance provided by TAG. Third, the negative and significant coefficient on *Checking* indicates that checking accounts are a relatively stable funding source. Fourth, we find that accounts that receive direct deposits roughly every two weeks (indicative of direct-deposited paychecks or other regular transactions) are also less likely to liquidate, doing so around 10 percentage points less often than non-direct deposit accounts, as indicated by the significant *Direct Deposit* dummy. Finally, the *Trust* dummy is insignificant, meaning that accounts held by trusts liquidate at about the same rate as the baseline account.

We also control for other account and depositor characteristics. Because there is relatively little variation across time periods in our coefficient estimates for these additional controls, we discuss them mainly with respect to the Placebo period. For instance, depositors with a longer relationship with the bank are generally more stable as shown by the coefficients on *Log(Age)*. The rate at which depositors conduct transactions has a significant, non-linear

¹⁸Figure IA.3 in the Internet Appendix presents the baseline hazard rates corresponding to the Cox results, and Figure IA.2 presents non-parametric Kaplan-Meier survival curves for the same periods.

relationship with liquidation behavior, as both *Prior Transactions* coefficients are significant but with opposite signs. Accounts with very infrequent transactions (unaware or inactive depositors) and accounts with very frequent transactions (operationally important accounts from depositors' perspective) are less likely to liquidate than other accounts. Accounts in the middle, with moderate usage, are more likely to liquidate. While there are statistically significant differences in this basic result across periods, the differences are economically insignificant. Finally, transaction accounts marked by the *Institutional* dummy are not significantly more or less likely to liquidate than other accounts, with this result holding in all periods.

Moving to the Pre-Crisis period (column 2 of Table III), we see that little changes. The coefficient estimates on the *Uninsured* dummy and *TAG/DFA Eligible* dummy are statistically insignificant, and checking accounts and accounts receiving regular direct deposits remain roughly as stable as in the Placebo period. The finding of similarities between the Placebo and Pre-Crisis periods is generally consistent with the historical discussion above, where transaction deposits did not much react to building financial weaknesses before the peak of the crisis.

Next, in the Post-Crisis period, the point estimate on the *Uninsured* dummy is larger - such accounts liquidate 7 percentage points more often than other accounts at the time. Similarly, the *TAG/DFA Eligible* dummy shows that such accounts liquidate nearly 10 percentage points less and the estimate is statistically significant. This reflects the fact that the accounts were actually covered by TAG for the first time (recall that temporary insurance programs were not in place in prior periods), and TAG dampened withdrawal behavior similarly or more effectively (per the point estimate) than ordinary deposit insurance.

Finally, column (4) of Table III presents results for the Formal period. The impact of FDIC insurance is stronger than in prior periods: uninsured accounts liquidate 18 percentage points faster than other accounts according to the *Uninsured* dummy. This result is statisti-

cally stronger in the Formal period than the Placebo period at the 1% level of significance. We also find that checking accounts, accounts receiving direct deposits every other week, and accounts held by depositors with longer relationships with the bank continue to be statistically significantly stickier than other accounts. Checking accounts remain sticky following bank-specific bad news, but less so than in response to market-wide stress. In contrast, the impact of the length of depositor relationships is stronger in the Formal period than in the Placebo period, with such accounts more sticky according to the *Log(Age)* dummy. Trust accounts meanwhile reverse behavior from the Post-Crisis period and run off 6.98 percentage points *more* than other accounts, as shown by the *Trust* dummy. Similar results obtain in the probit and Cox regressions as can be seen in Table IV.

[Place Table IV Approximately Here]

We also include Table V, which incorporates interest rate spreads over market rates as well as account fees paid for the Post-Crisis and Formal periods. As shown in the first two columns analyzing transaction accounts, the addition of the rate and fee variables does not meaningfully change the other coefficients or their significance. Estimates for the rates and fees variables themselves indicate that higher net returns on accounts (from either lower fees or higher rates) are significantly associated with fewer liquidations. We note that the estimates on rates and fees are potentially subject to endogeneity bias - the left-hand side is a measure of quantity, while rates and fees measure pricing. Nonetheless, our estimated effects are consistent with the usual intuition that higher rates make depositors more likely to stay with the bank. To the extent that high-return accounts tend to attract rate-seeking or rate-sensitive depositors who generally turn over more often, our estimates might arguably be lower bounds on the causal impacts of net returns. We discuss the columns on term deposits below.

[Place Table V Approximately Here]

These results have several important implications. First, we provide clear evidence that funds covered by deposit insurance are more stable, particularly in periods of stress. More generally, the large impacts in periods of stress demonstrate that depositors were aware of the bank's declining health and the limit of deposit insurance. For outflows, absent the gross deposit inflows that we analyze in the next section, this suggests active depositor discipline. Second, in the Post-Crisis period, when TAG was in effect, the point estimates for *Uninsured* and *TAG/DFA Eligible* are not statistically different from one another. The point estimates in the first two rows of column (3) in Table III are similar in size, and a *t*-test of differences in the magnitudes of the coefficients between TAG and regular deposit insurance fails to reject the null of no difference with a *p*-value of 0.67, indicating that the magnitude of the effect of TAG is the same as that of regular deposit insurance.¹⁹ Given that TAG was new and unconventional, the program and its operational details would have been unfamiliar to depositors. The magnitude of the impact we estimate therefore bodes well for the effectiveness of such programs. We do not observe a significant impact of the DFA guarantees in the Formal period, but there are relatively few accounts covered by DFA guarantees so we have significantly lower power than in earlier periods. Third, our finding that checking accounts and accounts receiving regular direct deposits are relatively stable in all environments supports assumptions made in rules such as the Liquidity Coverage Ratio (LCR) and Net Stable Funding Ratio (NSFR).²⁰ Finally, our finding on trust accounts suggests that such depositors are more sophisticated. Trust depositors were less likely than others to flee in periods of general distress (Post-Crisis) when the bank itself was not near failure, but more likely to do so as the bank approached failure in the Formal period. This

¹⁹Given the definitions of the variables, the signs are opposite, but this implies the same effect on liquidation behavior. We take the absolute value before conducting the *t*-test.

²⁰To be considered the most stable form of funding for LCR purposes, deposit accounts must be fully insured retail deposits and either *i*) a checking account or *ii*) held by a depositor with other relationships with the bank (such as loans, other accounts, bill payment services, etc.; Basel Committee on Banking Supervision (2013)). Note that our definition of "checking account" is synonymous with the definition of "transaction account" in Federal Reserve Regulation D.

result may reflect the fact that such depositors are typically wealthier and more financially savvy, and hence better able to determine the solvency of the bank.

B. Drivers of Term Deposit Outflows

We next conduct a parallel analysis of runoff in term deposits (see Tables VI and VII; we again focus on the linear probability model estimates in Table VI).²¹ Focusing on the Placebo period (Table VI, column (1)) first, the *Uninsured* dummy is statistically insignificant, which implies that uninsured term deposits do not liquidate more than insured accounts, probably because bank solvency was not a concern. Turning to the Pre-Crisis period (column (2)), the *Uninsured* dummy is now significant with 4.3 percentage points more uninsured term deposits liquidating than insured deposits, consistent with the fact that uninsured term deposits began running off during this period (see Figure 1).

[Place Table VI Approximately Here]

Column (3) of Table VI reports responses for the Post-Crisis period. Point estimates for the impact of FDIC insurance are substantially higher than in earlier periods - uninsured accounts are 8.2 percentage points more likely to liquidate than other accounts. The results also show that brokered and placed deposits, which we expect would be particularly risk-sensitive, run off very rapidly, with the coefficient on the *Brokered/Placed* dummy indicating 28 percentage points more liquidations among these accounts. Although a longer term to maturity (*Log(Days to Maturity)*) continues to be a stabilizing feature of term deposits, relationship age (*Log(Age)*) becomes insignificant.

Finally, in the Formal period, the FDIC insurance dummy remains large and significant, and it is also larger than the point estimates from prior periods - uninsured term deposits

²¹In some time periods, some term deposit account variables are thinly populated, precluding the estimation of coefficients. When this occurs, we put a “-” in the table corresponding to the affected variable.

are 14.7 percentage points more likely to liquidate. Accounts with a longer relationship to the bank remain stickier than other deposits. Term deposits further from maturity are still less likely to liquidate, as seen in the negative *Log(Days to Maturity)* coefficient. Like trust transaction accounts in the Formal period, the *Trust* dummy demonstrates that trust term accounts are more likely to liquidate, although the estimate for term accounts is not statistically significant. Finally, the results show a large response from wholesale accounts. *Institutional - Listed/Faxed* deposits, many newly arrived at the bank, are much less likely to liquidate. The *Brokered/Placed* deposits dummy, in contrast, shows that such deposits run off 55 percentage points more than other term deposit accounts, a high response relative to other periods and deposits, showing the high risk sensitivity of these wholesale deposits.

[Place Table VII Approximately Here]

We also include Table V, above, in which the third and fourth columns add interest rate spreads and fees paid into the term deposit regressions for the Post-Crisis and Formal periods. To construct the rate spreads, we take the difference between the rate the depositor would have received had they renewed their term deposit that day and the market average rate. Much like in transaction accounts, these new variables do not much change the other coefficients. And higher interest rates are again associated with fewer liquidations. Fees for CDs are rare and only appear in the Post-Crisis period, when they are positively and significantly associated with liquidation probability. Acknowledging the same caveat regarding endogeneity as in the transaction account regressions, our results are consistent with higher returns making depositors less likely to exit the bank.

These results provide some significant insights. First and foremost, the results for term deposits again confirm the stabilizing impact of deposit insurance. In addition, we show that brokered and placed deposits exhibit a high degree of churn in all periods but respond even more under stress. This supports the view that such deposits are hot money. As in

the transaction deposit regressions, we find that longer depositor relationships help stabilize bank funding.

C. Policy Change During Crisis

Because we can observe depositors at a high level of frequency, we can study individual events with good specificity. We perform an event study on unconditional liquidation propensity around the time of the deposit insurance limit change, looking at the 36 days before and after this change.²² The liquidation probability among deposits uninsured under the old limit of \$100,000 and insured under the new limit of \$250,000 falls from 0.0512 to 0.0465, with the drop being particularly strong for CDs, going from 0.0806 to 0.0512. The effect of this change is harder to disentangle on transaction accounts, due to TAG being implemented approximately one week after the insurance limit change. These results further support the efficacy of deposit insurance in reducing withdrawals.

D. Account Liquidation and the Withdrawal of Insured Funds

Having established the increased propensity of uninsured depositors to draw down their deposits, we now investigate whether such depositors tend to draw down either to the insurance limit or well below it.

Table VIII presents our results for withdrawals from transaction accounts.²³ Results here incorporate changes in bank conditions and macroeconomic conditions over time. Each row represents one of our four periods. For each period we consider the set of accounts with balances above \$2,000 under the insurance limit at the start of the period. The columns then show balances of these accounts at the end of the period, in six account balance bins. Relative

²²We chose 36 days because this was the widest window we could use that was symmetric around the deposit insurance limit change and did not overlap with the other time periods we study.

²³This is also presented as Figure IA.1 in the Internet Appendix. We do not show a comparable table for term deposits because their behavior is simpler: generally, they remain with the bank in full or exit entirely.

to the Placebo and Pre-Crisis periods, depositors in the latter two periods, especially the Formal period, tend to draw down well below the limit: the largest groupings in the Formal period, relative to previous periods, are accounts with \$1 or less and those between \$2,000 and \$2,000 less than half the insurance limit (\$2,000 to \$123,000, in this period)²⁴, with far fewer accounts remaining above the deposit insurance limit than in other periods. Under stress, uninsured depositors tend to withdraw much more than required to achieve full insurance coverage, either drawing down to less than half of the insurance limit or liquidating all funds.

[Place Table VIII Approximately Here]

This finding has significant implications for financial stability, since even some insured funds are likely to flee banks in response to stress, and can inform banking theory models (such as Davila and Goldstein (2023)).²⁵

III. Analysis of Deposit Inflows

The previous section focuses on deposit outflows, which is traditionally the area of attention with respect to bank funding stability. In this section we demonstrate that deposit inflows are also important to funding stability, even in a bank publicly known to be at high risk of failure. After providing an overview of the deposit inflow dynamics at the bank, we use a regression framework to establish the characteristics of new depositors, we present time series evidence that these inflows were not driven solely by factors external to the bank, and we provide evidence that the deposit inflows were instead attracted by the combination of credible deposit insurance and above-market rates. We complete the section with a discussion of the policy implications of this new finding.

²⁴We chose \$2,000 as the cutoff based on where we observed depositors bunching and how much interest they may accumulate.

²⁵Iyer et al. (2019) find similar behavior among uninsured Danish bank depositors: in response to bad news about the bank during the financial crisis, they tended to follow the heuristic of splitting accounts in half between two banks to achieve full insurance coverage.

In the last year of its life, the bank attracted new, insured deposits, about half of which arrived in the last 90 days before failure - when the bank was publicly known to be critically undercapitalized. Figure 3 shows that the inflows of insured term deposits from new depositors to the bank totaled nearly \$400 million, about a third of the bank's deposit base and roughly equal to the volume of fleeing deposits. Because of these flows, the bank's total deposit balances declined little as the bank approached failure. Figure 2 above shows term deposits by various wholesale funding categories in the bank over time. This figure reveals that the inflows represented a shift in deposit composition: as placed term deposits (and transaction accounts) fled the bank, they were replaced by institutional term deposits, attracted mostly through internet deposit listing services.

[Place Figure 3 Approximately Here]

Summary statistics in Table I above provide additional details. Reflecting new deposits being structured to fall just under the insurance limit, only 2.7% of new deposit accounts in the Formal period were uninsured, down from 6.3% in the Placebo period. Moreover, the average opening balance among new depositors in the Formal period was \$186,909. Related, the share of term deposits in new deposits is increasing over time - in the Formal period, 82% of new accounts were term deposits. Finally, 81% of new deposits in the Formal period came from institutional depositors (Listed, Faxed, and Other Institutional), up from 3.1% in the Placebo period.

Additional analysis of the bank's records provides further details. The institutional deposit inflows were mostly CDs from out-of-state banks and credit unions. As noted above, 81% of the 2,600 new deposit accounts in the Formal period were from institutional deposits. Most were in term deposits - 74% of the 2,600 were institutional term depositors. Of those nearly 2,000 accounts, about 65% were from credit unions and 32% from banks, with just a few percent from other businesses. Geographically, only 4% were from the same state as the

bank (as measured by depositors' addresses in the deposit system), and the remaining 96% were spread widely across the United States. In terms of balances, about three-quarters of institutional CDs had opening balances between \$100,000 and 250,000, reflecting the fact that they were structured to fall close to the insurance limit.²⁶

Finally, internal bank records indicate that the bank began to use two deposit listing services during the Formal period, and the timing of the bank's utilization of these services corresponds closely to the surges of institutional deposit inflows shown in Figure 2. In one case, internal emails indicate that the bank began posting rates on a listing service where they apparently had not previously advertised their deposits; in the other case, we found a signed and dated contract with another listing service. These two events are each within a week of either the surge at the beginning of the Formal period or the surge that began about 90 days before the bank failed.

A. Characteristics of New Depositors

Having shown that the bank experienced substantial deposit inflows, especially late in life, we conduct a set of classification regressions to formally characterize the differences between new and extant depositors in each period (equation (2) and Table IX). We conduct a set of six regressions (one in each column of the table) in which the dependent variables are dummy variables capturing deposit(or) characteristics such as *Uninsured* or *Institutional-Listed/Faxed*. Explanatory variables are the same across all regressions and consist solely of the interaction of a dummy indicating a new versus extant depositor with the time period

²⁶Note that the particular dollar range of \$100,000 to 250,000 is consistent with Call Report variables used in the generalization exercise in Section IV. below.

dummies (including a dummy for the date of failure), plus a constant term:^{27,28}

$$\mathbb{1}(Characteristic_{i,t}) = \alpha + \sum_{t=1}^5 (\beta_t Time\ Period\ Dummy_t \times \mathbb{1}(Extant\ Depositor_{i,t})) + \varepsilon_{i,t}. \quad (2)$$

The results of these regressions (Table IX) formalize the findings discussed above and previously shown in Figure 2.²⁹ The first column shows that, relative to the extant depositors in the Placebo period, uninsured deposit accounts became much rarer in periods of stress. In the Post-Crisis and Formal periods, uninsured accounts were about five percentage points less common among existing depositors, and their prevalence among new depositors also declined. At failure, only 1.4% of accounts had any uninsured funds in them (obtained by summing the constant and Extant Depositors/Failure Date coefficient estimates), down from 6.5% in the Placebo period (read directly from the estimate on the constant term). These patterns reflect the general runoff of uninsured accounts, as well as the inflows of insured institutional CDs from listing services. Further reflecting these institutional inflows, the fourth column shows that *Institutional - Listed/Faxed* deposits were very rare at the bank until the Formal period, when they represented 71% of new depositors. Similarly, column (6) shows that while CDs had generally represented 15% to 25% of existing depositors and a little less than half of new depositors, they jumped to 82% of new depositors in the Formal period.

[Place Table IX Approximately Here]

²⁷Throughout the paper, unless otherwise noted, a depositor is considered new in any given period if they open an account within the period and have never previously appeared in the bank's deposit records. Depositors already in the bank's deposit records are referred to as extant depositors.

²⁸The observational unit is account \times time period, and with five time periods it is possible for an individual account to appear up to five times in this regression. The omitted category is extant depositors in the Placebo period. Thus, the coefficient estimate on the constant term represents the share of extant depositors in the Placebo period exhibiting that characteristic. Coefficient estimates on the other controls represent differences relative to extant depositors in the Placebo period, in percentage points. Adding any coefficient to the coefficient on the constant yields the absolute share of that characteristic among the relevant depositor type (new or extant) in the relevant time period. Note that we also include a Failure Date dummy to summarize extant depositors at the bank when it failed.

²⁹See also Table I for related summary statistics.

Finally, note also that the bank's heavy reliance on brokered and placed deposits following the crisis is reflected in the table, notably among new depositors in the Post-Crisis period. Although total brokered/placed deposit balances were high and relatively flat throughout the Post-Crisis period (see Figure 2), the high rate of churn among these accounts generated many new depositors.

B. Drivers of Gross Depositor Inflows

While the previous section illustrates the prevalence of certain depositor characteristics among extant and new depositors, it does not explain the overall prevalence of new compared to old depositors. This section builds upon the last by analyzing the daily time series of new deposits using regression equation (3):

$$y_t = \alpha + \beta \text{Time Period Dummy}_t + \gamma X_t + \varepsilon_t. \quad (3)$$

The dependent variable, y_t , is either the proportion of deposits that are new as of that day (in Table X) or the log of the dollar volume of deposits that are new that day (in Table XI). The explanatory variables consist of a constant, a set of time period dummies, and macroeconomic controls, X_t . Among the time dummies, the omitted period is the Placebo period, with dummies for the remaining three periods - Pre-Crisis, Post-Crisis, and Formal - as well as dummies for the spans of time between those periods, plus a prior-to-the-placebo dummy.

[Place Table X Approximately Here]

[Place Table XI Approximately Here]

Focusing first on the macroeconomic controls in Table X, higher current-quarter *GDP Growth* is associated with higher deposit inflows, consistent with wealth effects. Other

macro variables such as *Housing Starts*, *Daily S&P 500 Return*, stock market volatility ($\text{Log}(VIX)$), the *OFR Financial Stress Index*, or the bank's growth profile are not statistically significant. The time series of new depositors' share of deposits is strongly persistent at the daily frequency, as shown by the positive, significant $AR(1)$ term.³⁰ In contrast, Table XI shows that the log level of new deposits correlates with $\text{Log}(VIX)$, *OFR Financial Stress Index*, and the autoregressive term.

Now consider the time period dummies in Table X. Even after the inclusion of macroeconomic controls, some time dummies remain significant. The dummies for the two early time periods are significant, and we note that this was a time in which the bank was growing strategically. Depending upon how one defines a "de novo" bank, the bank that we study could be considered de novo during periods before the Pre-Crisis period.³¹ The dummies in the last three periods before failure also remain significant, and the point estimate for the Formal period is especially large relative to all others. Thus, the Formal period stands out as a period of especially high inflows.³² Table XI shows the uniqueness of the Formal period even more clearly, as the Formal period dummy is the only significant time dummy when all controls are included. Given our set of controls, these results suggest that the inflows late in the bank's life are explained by bank-specific conditions.

Figure 4 suggests that the bank-specific cause was the high deposit rates offered by the bank in the Formal period. The figure compares the rates the bank paid on newly issued 12-month CDs, a common deposit product that is representative of the bank's product portfolio and relative market position, showing both the actual rate and its relative ranking

³⁰Note that this is not a mechanical result of constructing the series with overlapping measurement periods, as we define "new depositors" at the daily frequency.

³¹The regulatory definition of de novo status has changed several times over the years, especially with regard to how many years after opening a bank remains de novo.

³²Because the Placebo period is omitted, all estimates are relative to the Placebo period. Nonetheless, the Formal period's uniqueness is not driven by the choice of omitted time period. The point estimate is much larger than in any other period, and the Formal period estimate is statistically significantly different from both the Pre-Placebo and Placebo-to-Pre-Crisis periods, as well.

along with market percentiles.³³ The figure shows that the bank paid relatively high deposit rates throughout its life. However, as deposits flowed in over the last year of its life, the bank consistently paid rates above the 95th percentile of the industry distribution, often approaching the top 1%. Our evidence of the distressed bank raising deposit rates to attract insured deposits is in line with predictions of Egan et al. (2017). Including these interest rate measures in the regressions in columns (3) and (4) of Tables X and XI shows that the bank attracts more deposits on days when it pays higher interest, although depositors do not respond to variation in account fees. Due to potential endogeneity concerns, we view these results on deposit pricing as merely suggestive.

[Place Figure 4 Approximately Here]

In addition, we use the regressions of Tables X and XI to calculate the interest elasticities of demand, obtaining an answer very close to other estimates in the literature. Specifically, we calculate an elasticity of 0.61.³⁴ Artavanis et al. (2022) find an interest rate elasticity of demand of 0.48, while Egan et al. (2017) find an elasticity of 0.56 for insured depositors; both of these estimates are very close to our calculated elasticity of 0.61.

³³Rather than taking the average deposit rate paid on all 12-month CDs at each date, we construct the series as the 31-day centered moving average of rates offered on newly issued term deposits. In this way, the rate series better reflects the rate a hypothetical depositor would have faced had they approached the bank on that date, and there are also some days in which no new 12-month CDs are issued.

³⁴To obtain this elasticity from the regression tables, we first calculate the elasticities of the share of new depositors and the quantity of new deposits. These can then be leveraged to calculate the elasticity of existing deposits. The mean share of new deposits as a share of total deposits is 4.29 basis points (bps), the product-weighted spread for these new products over the market interest rate on a given day averages 1.17 percentage points, and the coefficient on this interest rate in column (3) of Table X is 5.55. Thus, raising rates by 1 percentage point (an 85.38% increase) raises the share of new deposits by 5.55 bps (a 129% increase). Taking their ratio, the interest rate elasticity of the new deposit share is 1.52. Similarly, the coefficient on the interest rate in Table XI is 2.61; given that the dependent variable is in logs, this coefficient implies that the same 1 percentage point rise in interest rates (an 85.38% increase) drives a 261% increase in new deposits. Thus, the interest elasticity of the quantity of new deposits is 3.05. Using these figures, we can calculate the elasticity of existing deposits. The average quantity of new deposits on any given day is \$558,000 (4.29 bps of the \$1.30 billion daily average total deposits). Using the elasticity of 3.05 from above, a 100% increase in rates therefore increases new deposit balances to \$2.26 million (305%). Using the elasticity of 1.52, we know that this also increases the new deposit share to 10.79 bps. For the balances and shares of new deposits to match, total existing depositors' balances must increase to \$2.09 billion, an increase of 61% relative to the average such balance. Thus, the interest elasticity of the quantity of deposits from existing depositors is 0.61.

C. Policy Implications of Deposit Inflows

The deposit inflow phenomenon that we document has several important policy implications. First, while some depositors enforced discipline on the bank by leaving, and new depositors were able to demand somewhat higher rates, new depositors offset the disciplining effect by opening new accounts. This finding is important given that the Basel framework considers market (in this context, depositor) discipline of banks to be the third of three “pillars” of financial stability. Our finding suggests that insured deposit inflows are an important, hitherto undocumented, phenomenon that undermines any such disciplining power.^{35,36}

Second, and closely related, our finding emphasizes the importance of studying *gross* deposit flows rather than net deposit flows in understanding the implications of deposit insurance. By making depositors less sensitive to bank risk, deposit insurance stabilizes deposit funding; the inflows from new deposits makes this effect stronger.

Third, the large inflow of new deposits suggests that deposit rate restrictions placed on troubled banks are insufficient to prevent rapid insured deposit acquisition, certainly in a low rate environment. Motivated by the concerns above, U.S. law and regulation prohibits less than well capitalized institutions from paying deposit rates more than 75 bps above the national average deposit rate on deposits solicited nationally. Egan et al. (2017) present a theoretical model that suggests such rate caps can rule out particularly bad banking market outcomes where failing banks attract large shares of insured deposit funds. The bank we study was subject to these restrictions during the Formal period and yet was able to attract

³⁵Clearly, many forms of moral hazard can emanate from deposit insurance, for example, excessive risk-taking on the lending side. Deposit insurance can also inhibit the reallocation of deposit funding away from an insolvent bank. Our objective is not to look at all possible mechanisms - nor do we have the data to do so - but rather to clearly illustrate one possible effect of deposit insurance that has been hitherto undocumented, namely, its effect on deposit inflows that our results suggest can be large and a first-order effect.

³⁶We generalize the finding on inflows empirically in the next section, but we note here that anecdotal evidence supporting these findings was given by our discussant at the Chicago Financial Institutions Conference, who stated that he had witnessed such deposit interest rate increases at Indymac Bank, a \$30 billion dollar bank, in 2007 to 2008, just before its failure, and had himself moved deposits in response to these higher interest rates. He also noted observing that such interest rate increases were not limited to this one bank.

deposits equal to around one-third of its deposit base. Panel A of Table I shows that the bank complied with the rate restrictions; the spread to the market average on new accounts in the Formal period was around 69 bps.³⁷ Because the bank was able to attract so many new deposits while under the restrictions, we conclude that the rate restrictions were at best a minimally binding constraint on the bank’s behavior. However, it remains unclear whether the rate caps would be more effective if short term, riskless rates were substantially above zero; the fed funds rate was at the zero lower bound for the entire period in which the rate caps were in place for this bank. To the extent that the dispersion of deposit rates is reduced when average rates are low (as suggested in Figure 4), the 75 bp cap on the deposit rate spread would be less binding while at the zero lower bound compared to periods when rates are above it. Of course, in the absence of any rate restrictions, the bank may well have attracted even more insured funding than it actually did.

These high rates may be a sign of an “inconvenience yield” of deposits issued by a bank close to failure. FDIC resolution of failed banks is close to seamless - if the bank closes on Friday, insured depositors have access to their funds by or before the following Monday. Less sophisticated depositors may be unaware of the process and have concerns about speed and frictions of repayment, and hence refrain from putting insured deposits in a failing bank. Meanwhile, sophisticated, informed depositors (such as other insured banks) flow into the bank to capture this premium. Alternatively, the bank may be willing to pay higher rates for insured deposits because the value of the bank’s “insurance put” embedded in insured deposits is higher (Egan et al. (2017)). As the risk of bank failure increases, so does the likelihood that the FDIC, rather than the bank, will end up bearing the costs of these new,

³⁷Note that the spreads reported in the table are relative to our computed national average rate rather than that defined by the FDIC. We calculate our own national average series using a method identical to that used for the FDIC series. We use our computed series as the official data do not cover our entire sample period, and we wish to keep series consistency across our sample. The source data underlying the official average data change with vintage, and we have not been able to recover the vintages used to construct the FDIC series. As a result, our averages tend to differ slightly from the official data. Using the official data over the supported period gives the same qualitative conclusions.

insured deposits. As a result, the bank might be willing to pay more for insured funds.

The final reason that the shift in deposit composition matters to policymakers is that it transfers risk to the FDIC. In addition to fleeing insured deposits, about \$150 million of uninsured transaction deposits also left. Because the bank replaced these fleeing uninsured deposits with insured term deposits, the share of the bank’s deposits covered by insurance and the credit risk exposure of the FDIC increased as the bank approached failure.

IV. Generalization to Other Banks

The unusually granular data for this bank allow us to identify changes in deposit composition as it approached failure, but we unfortunately have such data for only a single bank despite considerable effort and investment to try and secure such data for additional banks.³⁸

To generalize our results, we conduct two additional tests. First, we use Call Report data for all U.S. banks to investigate whether banks that face a regulatory action see similar changes in deposit composition. In particular, we investigate the impact of “treatment” with a regulatory action on five funding measures. The five funding measures ($y_{j,t}$ for bank j in quarter t , below) are the share of each bank’s total deposit funding in the form of brokered deposits, listing service deposits, small term deposits (those under \$100,000), medium term deposits (those over \$100,000 but under \$250,000), and large term deposits (those over

³⁸When we initiated this project, we asked for data for a sample of 10 banks to be recovered and put on a secure server on which we could access the data. We then examined deposit data in each bank. Unfortunately, for many of the banks, the deposit data are incomplete. For some of the banks, the data between systems cannot be appropriately linked; for example, databases sometimes lack identifiers to link transactions (from one database, necessary for calculating account balances) with deposit accounts (from another database, necessary for many permanent account characteristics) or customer systems (containing separate customer characteristics.) Other banks keep only the most recent three months of detailed transaction history. Each bank has different storage and retention policies. Compounding these data challenges is the fact that the data contain large volumes of sensitive and personally identifiable information, such that access to the data is highly restricted, and uploading the data to secure IT environments with statistical analysis software is time consuming. Due to these factors, it took well over a year for us to obtain access to data for a sample of just 10 failed banks, and after analyzing their data, we determined that only the single bank we study had suitable deposit data. As reported earlier, for the bank we study, we are confident that the deposit data are complete by matching to the Call Reports, but this is not true for the other banks.

\$250,000).³⁹ We conduct this analysis under three different model specifications. Two of the specifications are simple OLS models and differ only in the specification of the treatment dummy. In one specification (equation (4)), we have a single treatment dummy that is equal to one in any bank-quarter in which regulatory action was in place:

$$y_{j,t} = \alpha + \beta \text{Under Reg. Action}_{j,t} + \gamma X_{j,t} + \delta_j + \zeta_t + \varepsilon_{j,t}. \quad (4)$$

Each series is individually regressed on a dummy (“treatment”) variable indicating whether the bank was facing regulatory action similar to the studied bank, $\text{Under Reg. Action}_{j,t}$. We also include bank-specific controls ($X_{j,t}$), bank and time fixed effects (δ_j and ζ_t , respectively), and a constant. In this case, the untreated (“control”) group is all banks not contemporaneously under a regulatory action. In another, similar specification (equation (5)), we use separate dummy variables for each quarter from four quarters before the imposition of treatment to four quarters after, plus an additional dummy for five or more quarters of continuous treatment:

$$y_{j,t} = \alpha + \sum_{i=-4}^{\geq 5} \beta_i \text{Under Reg. Action}_{j,t=\tau+i} + \gamma X_{j,t} + \delta_j + \zeta_t + \varepsilon_{j,t}. \quad (5)$$

Here, τ denotes the quarter in which regulatory action is first imposed, and the control group is banks that will not face such action over at least the next four quarters. Finally, we estimate a third specification on propensity-score-matched banks. Banks’ propensity to be treated is determined by a logistic regression using the same covariates as in the above regressions, plus contemporaneous term deposit size shares. Then, banks with similar propensity scores - where one was treated and the other was not - are compared after four quarters of continuous treatment to observe the effects of that treatment. The results are consistent across all specifications.

³⁹For the average bank in our sample, 3.1% of its deposits are classified as brokered, 1.6% as listed, 20.2% as small term deposits, 13.1% as medium term deposits, and 5.5% as large term deposits.

We define a bank as facing regulatory action if it is under a formal enforcement action that includes capital-related provisions or if it is less than well capitalized⁴⁰ without specific written permission from the regulator to continue taking brokered deposits, usually referred to as a brokered deposit waiver. Less than well capitalized institutions are forbidden from taking new brokered deposits or rolling over old ones unless they have a brokered deposit waiver. Note that any bank meeting our definition of regulatory treatment will also be subject to deposit rate restrictions. The treatment variable is defined using data from public Call Reports and confidential FDIC data, which provide details of the enforcement actions and brokered deposit waivers. The advantage of this approach is that it gives us a large panel of around 10,000 banks, of which 2,358 faced regulatory action.

The control variables, $X_{j,t}$, are derived from Call Reports. Control variables are non-performing assets as a share of assets, to capture bank health; one-year asset growth rate, to capture the growth and current risk profile of the bank; the natural logarithm of assets, to capture size; deposits as a share of assets, to capture the banks' reliance on deposits generally; and term deposits as a share of deposits, to capture their reliance on term deposits in particular.⁴¹ We use quarterly data for all U.S. banks from 2000 to 2016, with 2,358 banks facing regulatory action and 9,158 not facing such action at some point during this time. However, because we rely on the Call Report taxonomy of deposit accounts and because this taxonomy has changed over time, some regressions use shorter samples.⁴² Note that

⁴⁰See 12 U.S.C. §1831o for capital category definitions.

⁴¹We altered the raw Call Report data by correcting for apparent reporting errors and by winsorizing. Specifically, in a handful of bank-quarters, banks appear to have reported brokered deposits in dollars, inconsistent with the Call Report standard of thousands of dollars, requiring us to divide by 1,000. When funding shares calculated from Call Reports were a fraction of a percent above 100% or below 0%, we assume this is due to rounding error, and we round to 100% or 0%, respectively. Finally, we bounded one-year asset growth rates between -50% and 100%, affecting about 2% of bank-quarters, with large asset shifts usually due to new small banks growing rapidly in the first few quarters. Importantly, none of these changes materially affects the point estimates of the treatment variable.

⁴²Listing service deposits were not separately identified or reported on Call Reports before the first quarter of 2011, and data necessary to disaggregate term deposits by size is available beginning in 2010. This limited our listing service deposit sample to 559 treated and 7,020 untreated banks and our term deposit sample to 807 treated and 7,141 untreated banks.

regressions also include bank and quarter fixed effects.

Table XII shows the results of regressions with a single regulatory action dummy variable (as in equation (4)). Consistent with our earlier findings, banks under regulatory action reduce their reliance on brokered deposits while increasing listing service deposits. These two compositional shifts do not completely offset one another, but banks may also be seeking other classes of deposits. In addition, there is an increase in the reliance on term deposits below \$100,000 as well as between \$100,000 and \$250,000, with a decrease in deposits above \$250,000.

[Place Table XII Approximately Here]

Table XIII and the accompanying Figure 5 demonstrate the time path of the effects of regulatory action on the same deposit categories. Relative to banks that are more than four quarters away from regulatory action, banks one to four quarters before an action have statistically significantly more brokered deposits. These banks become much *less* likely to source such deposits, reflecting the concurrent application of brokered deposit restrictions. The exact opposite pattern appears with listing service deposits, with banks prior to such events having fewer and those after having more. Presumably, this pattern reflects inflows of listing service deposits conceptually similar to those documented previously. Small time deposits are relatively more prevalent in banks in the last few quarters before regulatory action than in the control group, and the difference becomes even more stark after regulatory action. Medium time deposits make up a smaller share of funding at banks prior to regulatory action compared to the control group, but also become more common following regulatory action. Finally, large time deposits are equally as common among banks far from regulatory action as well as those within only a few quarters of action, but they become much less common among banks following regulatory action. Table XIV shows that our generalization results are robust to using a propensity-score-matched specification.

[Place Table XIII Approximately Here]

[Place Figure 5 Approximately Here]

[Place Table XIV Approximately Here]

This analysis generalizes our earlier findings along several dimensions. The regressions consistently find that banks under regulatory action reduce reliance on brokered deposits (due to concurrently applied brokered deposit restrictions) and deposits above the insurance limit (reflecting the flight of uninsured depositors from the ailing bank). These regressions show that banks under regulatory action increase reliance on listing service deposits, much as the bank featured earlier in this paper did. Treated banks also increase reliance on term deposits under the insurance limit, especially those between \$100,000 and \$250,000. Recall that the bank featured earlier in this paper structured most of its new term deposits during the Formal period to fall just under the insurance limit - within this range.

Another question is whether these patterns hold only for small banks or also for larger banks. The approximately 500 banks that failed during the Great Recession led to significant depletion of the DIF and substantial costs to the FDIC and hence this finding is important in its own right even if it applies only to smaller banks. However, we also provide evidence that this phenomenon is common even to larger banks. To show that this behavior generalizes to larger banks, we collected deposit pricing data from *RateWatch* for large banks that we know faced regulatory enforcement actions similar to the bank we study. For the purposes of this analysis, we define large banks as those with assets over \$5 billion; to avoid sample selection issues, we use total banks assets as of 2008Q4 for this size determination. We chose a relatively low threshold for “large banks” in order to include a sufficient number of observations of banks that faced enforcement action; beyond the substantial right skew in the distribution of all banks in the United States by asset size, the enforcement actions that

we study are particularly uncommon against very large banks.⁴³

We use rates offered on the 12-month retail CD with a \$100,000 balance as the representative deposit rate in this analysis, and we construct the spread on these accounts relative to the FDIC national average rate data. To measure large banks' deposit interest rates, we use survey data collected by *RateWatch*. The raw survey data collected by *RateWatch* is at the weekly frequency, at the branch level, and includes rates for a large variety of deposit products. We drop promotional rate offers, relationship-based special offers, and business-focused offers, leaving us with regular retail deposit offerings. We then exclude nonbank institutions. Next, we adjust the correspondence between banks and branches to reflect branch sales; raw survey data take the branch as the unit of observation, and we use a file containing changes in branch ownership to reconstruct the history of banks to which those branches (and their survey responses) belonged. Having generated an accurate correspondence between branches' weekly deposit rates and the bank that owns the branch, we take an unweighted average across all branches within the bank at each point in time and for each given product; this yields bank-level average rates by deposit product. Finally, we keep the last weekly survey report for each bank within the quarter, yielding quarter-end deposit rates by bank and product.

Focusing only on banks facing enforcement actions, we conduct a pooled OLS regression of these large banks' 12-month CD rate spreads on (*i*) a dummy variable indicating whether they are currently under an enforcement action and (*ii*) a set of quarterly (time) dummies,

$$spread_{j,t} = \alpha + \beta Under\ Reg.\ Action_{j,t} + \zeta_t + \varepsilon_{j,t}. \quad (6)$$

Here, the coefficient on the enforcement action dummy represents the difference in rate

⁴³We considered higher asset size thresholds to define large banks but cannot disclose the results, as the sample sizes are sufficiently small that releasing those results may reveal agency sensitive information and would violate best practices for disclosure review.

spreads offered by banks under enforcement action versus banks that are about to be under enforcement action (within four quarters). As shown in Table XV, this regression reveals that these larger banks raise their rates by a statistically significant 41 bps around the imposition of enforcement actions. Adding this result to the constant, we find that large banks under enforcement action pay on average almost exactly 75 bps above the FDIC national average – that is, at or near the limit for such banks.⁴⁴

[Place Table XV Approximately Here]

V. Conclusion

In this paper we use a highly granular and unique dataset to identify important new findings related to deposit inflows and outflows in a failing bank. Our most important finding is that gross deposit *inflows* have a first-order impact on failing banks' balance sheets, despite the banks' elevated default risk and supervisory actions meant to prevent costly and rapid deposit acquisition. Even though there are substantial deposit outflows in the bank under study - the bank loses one-third of its existing deposits - these are made up by substantial deposit inflows, with new deposits typically just below the deposit insurance limits. The new deposit inflows arrive as the bank increases its interest rates - in line with the theoretical predictions of Egan et al. (2017). Although our focus is on one bank, we demonstrate that our main findings qualitatively generalize to other banks by using a combination of confidential supervisory data and publicly available but less granular data on other U.S. banks, and by collecting additional data on interest rates.

We also identify a number of policy-relevant findings related to gross deposit *outflows*. Most importantly, we provide further evidence that deposit insurance improves funding

⁴⁴The results above are based on a quarterly frequency regression with data coming from a total of 19 unique banks, and the panel data are imbalanced/censored.

stability. We also provide the first empirical evidence that temporary, crisis-era expansions of deposit insurance, such as FDIC's TAG, are as effective as ordinary deposit insurance in preventing deposit outflows. Many of our findings on outflows support the intuition reflected in regulations, such as that checking accounts and accounts receiving regular direct deposits are more stable. Ultimately, our results suggest that focusing on deposit outflows alone in times of stress or bank failures is inadequate - deposit inflows are also of first-order importance, worthy of further attention by regulators and academicians alike.

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Table I
Summary Statistics, by Period

Panel A reports summary statistics for new depositors arriving within the four event periods while Panel B reports statistics for depositors extant at the bank at the beginning of each period and at the date of failure. See Table II for precise variable definitions.

Panel A. New Depositors				
	<i>Placebo</i>	<i>Pre-Crisis</i>	<i>Post-Crisis</i>	<i>Formal</i>
	(1)	(2)	(3)	(4)
Number of New Accounts	3,028	2,146	927	2,600
Uninsured at Start of Account	0.063	0.087	0.044	0.027
TAG/DFA Eligible	0.015	0.038	0.016	0.008
↔ (<i>Covered in Post-Crisis & Formal</i>)				
Starting Balance	43,520	90,840	141,000	186,900
Term (count weighted)	0.461	0.478	0.417	0.824
Savings (count weighted)	0.487	0.379	0.408	0.084
Checking (count weighted)	0.052	0.144	0.175	0.092
Starting Interest Spread to Market	2.561	1.656	0.796	0.671
Institutional - Listed/Faxed	0	0.006	0.013	0.707
Institutional - Other	0.031	0.252	0.245	0.100
Brokered/Placed	0.002	0.035	0.221	0
Trust	0.038	0.037	0.094	0.041

Panel B. Extant Depositors					
	<i>Placebo</i>	<i>Pre-Crisis</i>	<i>Post-Crisis</i>	<i>Formal</i>	<i>Failure</i>
	(1)	(2)	(3)	(4)	(5)
Number of Accounts	41,366	44,910	38,374	30,342	25,141
Uninsured at Start of Period	0.065	0.087	0.009	0.021	0.014
TAG/DFA Eligible	0.006	0.008	0.004	0.004	0.005
↔ (<i>Covered in Post-Crisis & Formal</i>)					
Starting Balance	28,310	27,710	31,450	43,970	49,750
Term (count weighted)	0.190	0.242	0.222	0.123	0.191
Savings (count weighted)	0.734	0.690	0.700	0.768	0.678
Checking (count weighted)	0.076	0.068	0.079	0.109	0.130
Starting Interest Rate	4.121	4.367	2.486	0.941	0.882
Starting Interest Spread to Market	2.604	2.576	1.492	0.673	0.632
Institutional - Listed/Faxed	0	0	0	0.002	0.084
Institutional - Other	0.013	0.015	0.025	0.047	0.065
Brokered/Placed	0.013	0.012	0.045	0.037	0.005
Trust	0.015	0.017	0.016	0.023	0.028
Direct Deposit	0.028	0.030	0.023	0.035	0.032
Past Month Fees	-0.237	-0.061	0.084	0.174	0.171
Age of Relationship in Years	2.214	3.076	4.031	5.587	6.032
Years Since Start of Previous Period	-	1.25	1.25	1.78	0.92

Table II
Variable Definitions for the Regressions of Section II.

Liquidation: A dummy variable equal to 1 if the deposit account balance falls by 75% or more relative to the balance as measured at the beginning of the period, and the balance stays at or below 25% of the start-of-period balance for at least 61 days, which includes the account closing. This notion of account liquidation is generally consistent with related studies (for example, Iyer and Puri (2012)). Results in this paper are robust to different thresholds of 50% and 80%.

Uninsured: A dummy variable equal to 1 if there are *any* uninsured balances in the account at the start of the period. Deposit insurance limits apply separately to different ownership types, so we account separately for individual, corporate, municipal, joint, IRA, employee benefit plans, revocable trust, and irrevocable trust ownership categories. An exact insurance determination can sometimes be difficult, as joint and trust accounts have complex ownership structures that are often incompletely documented, and thus we construct this variable conservatively. Accounts that we flag as insured have all funds insured. Accounts that we flag as uninsured should have some uninsured funds in them, but it is possible that they are occasionally covered because of complex joint ownership. As a result, estimates in our regressions are lower bounds on the effects of being over the FDIC insurance limit.

TAG/DFA Eligible: A dummy variable equal to 1 if the account is eligible for additional insurance coverage from temporary guarantee schemes at the start of the period and 0 otherwise. The additional guarantee programs were in effect only in our Post-Crisis and Formal periods; in earlier periods, the dummy is assigned based on whether the account would qualify for coverage had the same programs been introduced earlier. The dummy is used in the Placebo and Pre-Crisis periods simply to establish a basis for comparison of these accounts' behavior; they did not have additional coverage before the financial crisis. The two additional guarantee schemes were the Transaction Account Guarantee (TAG) program and guarantees from the Dodd Frank Act (DFA). TAG placed temporary but unlimited (in dollar terms) guarantees on all categories of checking accounts at this bank from October 14, 2008 until December 31, 2010. The DFA guarantees similarly provided unlimited insurance for non-interest-bearing demand deposit accounts and IOLTA accounts, but not NOW accounts from December 31, 2010 until December 31, 2012. In calculating a depositor's insurance coverage, accounts covered by these programs (while effective) did not count toward a depositor's \$250,000 limit.

Direct Deposit: A dummy variable equal to 1 if the account is receiving an Automated Clearing House (ACH) deposit roughly every two weeks as of the start of the regression period and 0 otherwise. It is always 0 for term deposits.

Log(Age): The natural log of the years since the primary account holder first appears in the bank's deposit records, as of the start of the period. If an individual was a secondary depositor on an account before becoming a primary account holder on another account, or closed an older account, we use the date at which the original account was opened. Relationship age serves as a measure of the depth of the depositor relationship. The age of the account is dated differently in the case of placed deposits; see placed deposits below for details.

Log(Days to Maturity): The natural log of the number of days until the maturity of the account, as of the start of the regression period. This is defined only for term accounts.

Prior Transactions: The proportion of days in the past year, at the start of the measurement period, in which the account holder performed at least one transaction in the account. A value of 0 thus implies no activity while a value of 100 implies activity every day. We exclude transactions exogenous to the depositor such as monthly interest credits or fees. This serves as another measure of depositors' relationship depth. This variable is always 0 for term accounts.

Institutional-Listed/Faxed: A dummy variable equal to 1 if the deposit is owned by a bank, savings association, credit union, financial corporate, municipality, or non-financial corporation, or if it is a "business" product type as marked in the bank's records and 0 otherwise. Additionally, the deposit must have been opened via an internet listing service or facsimile as identified by the bank's records. We group faxed deposits with listed because internal bank documentation, depositor behavior, and depositor types (namely, small depository institutions making up a large portion of these deposits towards the end of the bank's life) all indicate that faxed deposits were almost exclusively obtained from depositors who saw the rates on listing services and then faxed their deposit request to the bank. This excludes third-party deposit placement services (such as deposit brokers), as we capture these entities with a separate dummy variable.

Institutional- Other: A dummy variable defined as *Institutional - Listed/Faxed* above, but we have no evidence that the deposits were received from a listing service or facsimile order.

Brokered/Placed: A dummy variable equal to 1 if the deposit was placed by a fiduciary or deposit broker and 0 otherwise. It is always 0 for transaction accounts. Many term deposits at the bank are not held by individuals but instead held by institutions acting as fiduciaries for others, and these fiduciaries do not consistently reveal the identity of the underlying holders of the account to the bank. These deposits reflect a less personal connection with the bank. For these accounts, the age of the account variable is dated to the start of the individual account, not the first relationship of the reported holder of the account, as the reported holder is only a fiduciary that may not make final renewal and withdrawal decisions. Note that we assume all placed deposits are insured; these services advertise that they structure their deposits so as to achieve full insurance coverage.

Trust: A dummy variable equal to 1 if the account is held by a trust and 0 otherwise. This includes Payable on Death (POD) arrangements. Trust accounts require effort to establish and are a useful legal device for wealthier or more complex depositors. As such, we expect accounts held in trust to represent more sophisticated customers.

Interest Rate Spread: The spread between an account's interest rate and the market average for that deposit product. The spread is intended to reflect the pricing offered to depositors opening a new account or rolling over a CD. For transaction (non-maturity) accounts, which re-price continuously and thus reflect current offered rates, we calculate the spread as the difference between the rate on an account and the account-type-level rate average constructed from *RateWatch*. Term deposits reprice only periodically, so the current rate on an existing term deposit is not generally the current rate on offer for new accounts or funds being rolled over from a maturing CD. To construct the spread for a given term account, we take the difference between the average rate paid on newly opened term deposits of the same maturity within a symmetric, centered, rolling 31-day window (i.e., 15 days before to 15 days after the relevant date) and the market average rate for that product calculated from *RateWatch* data. Particularly since depositors tend to liquidate CDs at maturity, even in periods of stress, this is the relevant measure of pricing for the decisions of both extant depositors considering leaving the bank and newly arriving depositors considering opening an account.

Past-Month Fees: The total dollar value in the past month of all fees paid on or to the account, including reversal of old fees and signup bonuses. For some regressions, this variable is divided by account balance to obtain a "fee rate."

Table III
Who Withdraws? Transaction Deposits; Linear Probability Model

This table presents estimates from linear probability models of the probability of account liquidation during each of the four periods of interest. Liquidation is defined as withdrawing 75% or more of the start-of-period account balance and staying at or below 25% of the start-of-period balance for at least 61 days. See Table II for additional variable definitions. *T*-statistics are in parentheses, calculated using asymptotically normal standard errors. ***, **, and * indicate significance at 99%, 95%, and 90%, respectively. † TAG/DFA coverage was effective only in the Post-Crisis and Formal periods; eligible accounts were fully covered/insured in those periods. In earlier periods, eligibility does not imply coverage and is shown for comparison purposes only.

	<i>Placebo</i> (1)	<i>Pre-Crisis</i> (2)	<i>Post-Crisis</i> (3)	<i>Formal</i> (4)
Uninsured	0.0249* (1.94)	-0.00117 (-0.09)	0.0693* (1.92)	0.183*** (8.45)
TAG/DFA Eligible [†]	-0.0446 (-1.50)	-0.000624 (-0.02)	-0.0944** (-2.00)	-0.0412 (-0.95)
Checking	-0.0901*** (-8.44)	-0.105*** (-8.10)	-0.0277*** (-2.64)	-0.0252** (-2.15)
Direct Deposit	-0.104*** (-7.08)	-0.123*** (-7.47)	-0.0864*** (-5.88)	-0.0459*** (-2.77)
Log(Age)	-0.0194*** (-7.47)	-0.0180*** (-4.34)	-0.00915** (-2.40)	-0.0247*** (-4.65)
Prior Transactions	0.00793*** (14.13)	0.00555*** (7.62)	0.00263*** (4.37)	-0.00310*** (-4.19)
Prior Transactions ²	-0.0000983*** (-12.71)	-0.0000746*** (-7.19)	-0.0000419*** (-4.88)	0.0000235** (2.34)
Institutional - Any	-0.0273 (-1.16)	0.0242 (1.01)	0.0218 (1.29)	0.0215 (1.33)
Trust	0.0255 (0.93)	0.0248 (0.83)	-0.0418* (-1.90)	0.0698*** (3.38)
Constant	0.217*** (13.12)	0.313*** (17.35)	0.159*** (11.62)	0.326*** (21.96)
Branch Controls	Yes	Yes	Yes	Yes
No. of Accounts	33,498	34,029	29,865	26,616
No. of Liquidations	6,920	10,795	4,740	6,218
<i>R</i> ²	1.3%	0.8%	0.4%	1.8%
Model <i>p</i> -value	< 0.001	< 0.001	< 0.001	< 0.001

Table IV
Who Withdraws? Transaction Deposits; Alternative Specifications

This table presents estimates from Probit (columns (1) and (2)) and Cox proportional hazard models (columns (3) and (4)) of the probability of account liquidation during the Post-Crisis and Formal periods. Liquidation is defined as withdrawing 75% or more of the start-of-period account balance and staying at or below 25% of the start-of-period balance for at least 61 days. See Table II for additional variable definitions. Probit estimates are expressed as marginal effects (with a coefficient of 0 reflecting a variable having no effect on the probability of liquidation), and Cox estimates are expressed as hazard ratios (with 1 indicating no effect). *t*-statistics are in parentheses, calculated using asymptotically normal standard errors. ***, **, and * indicate significance at 99%, 95%, and 90%, respectively.

	Probit		Cox P.H.	
	Post-Crisis		Post-Crisis	
	(1)	(2)	(3)	(4)
Uninsured	0.0655 (1.61)	0.170*** (6.69)	1.503** (1.99)	1.862*** (8.01)
TAG/DFA Eligible	-0.0691** (-2.24)	-0.0164 (-0.42)	0.567* (-1.83)	0.958 (-0.26)
Checking	-0.0262*** (-2.71)	-0.0280** (-2.51)	0.814*** (-2.72)	0.887** (-2.15)
Direct Deposit	-0.0875*** (-8.55)	-0.0679*** (-4.37)	0.430*** (-6.04)	0.631*** (-4.32)
Log(Age)	-0.00798** (-2.18)	-0.0231*** (-4.49)	0.936*** (-2.63)	0.896*** (-4.72)
Prior Transactions	0.00290*** (4.61)	-0.00337*** (-4.33)	1.025*** (5.43)	0.985*** (-3.75)
Prior Transactions ²	-0.00000518*** (-5.15)	0.0000259** (2.36)	1.000*** (-5.56)	1.000* (1.68)
Institutional - Any	0.0226 (1.24)	0.0192 (1.18)	1.156 (1.27)	1.090 (1.24)
Trust	-0.0444** (-2.30)	0.0601*** (2.73)	0.711* (-1.92)	1.250*** (2.71)
Branch Controls	Yes	Yes	Yes	Yes
No. of Accounts	29,865	26,616	29,865	26,616
No. of Liquidations	4,740	6,218	4,740	6,218
Log Likelihood	-13001.0	-14236.2	-48378.0	-62351.7
Model <i>p</i> -value	< 0.001	< 0.001	< 0.001	< 0.001

Table V
Who Withdraws? Linear Probability Model; Including Interest Rates

This table presents estimates from linear probability models of the probability of account liquidation for both transaction and term deposits, focusing only on the Post-Crisis and Formal periods. Relative to the main specification discussed elsewhere, this table adds interest rates and fees, at the account level, as regressors. Liquidation is defined as withdrawing 75% or more of the start-of-period account balance and staying at or below 25% of the start-of-period balance for at least 61 days. See Table II for additional variable definitions. *t*-statistics are in parentheses, calculated using asymptotically normal standard errors. ***, **, and * indicate significance at 99%, 95%, and 90%, respectively. ‡ Note that although TAG/DFA did not directly insure term deposits, it could do so indirectly. Accounts covered by TAG/DFA were excluded in calculating a depositor's total exposure to a bank when applying standard, limited-value deposit insurance, effectively freeing up insurance coverage to be applied to other, non-TAG/DFA-covered accounts.

	<i>Transaction</i>		<i>Term</i>	
	<i>Post-Crisis</i>		<i>Formal</i>	
	(1)	(2)	(3)	(4)
Uninsured	0.0723** (2.00)	0.190*** (8.72)	0.0731* (1.75)	0.129** (2.57)
TAG/DFA Eligible‡	-0.0909* (-1.93)	-0.0520 (-1.19)	0.00470 (0.03)	- -
Interest Rate Spread	-0.0135*** (-3.05)	-0.0780*** (-3.18)	-0.723*** (-23.17)	-0.507*** (-5.71)
Past-Month Fees	0.00136** (2.22)	0.00137*** (2.60)	0.0382*** (2.87)	- -
Checking	-0.0410*** (-3.64)	-0.0487*** (-3.53)		
Direct Deposit	-0.0877*** (-5.96)	-0.0443*** (-2.67)		
Prior Transactions	0.00284*** (4.69)	-0.00297*** (-4.01)		
Prior Transactions ²	-0.00000439*** (-5.10)	0.00000217** (2.16)		
Log(Age)	-0.0101*** (-2.63)	-0.0243*** (-4.57)	0.00953** (2.40)	-0.0109 (-1.56)
Log(Days to Maturity)			-0.142*** (-28.75)	-0.0849*** (-15.54)
Brokered/Placed			0.303*** (22.04)	0.537*** (25.09)
Institutional - Listed/Faxed			-0.0978 (-0.64)	-0.176*** (-3.28)
Institutional - Other	0.0116 (0.68)	-0.00527 (-0.29)	-0.00492 (-0.14)	-0.0429 (-1.38)
Trust	-0.0407* (-1.85)	0.0701*** (3.39)	0.0184 (0.77)	0.000492 (0.01)
Constant	0.182*** (11.78)	0.382*** (16.91)	1.706*** (38.40)	1.048*** (13.20)
Branch Controls	Yes	Yes	Yes	Yes
No. of Accounts	29,865	26,616	7,800	3,717
No. of Liquidations	4,740	6,218	2,506	1,734
<i>R</i> ²	0.4%	1.9%	35.2%	34.7%
Model <i>p</i> -value	< 0.001	< 0.001	< 0.001	< 0.001

Table VI
Who Withdraws? Term Deposits; Linear Probability Model

This table presents estimates from linear probability models of the probability of account liquidation during the each of the four periods of interest. Liquidation is defined as withdrawing 75% or more of the start-of-period account balance and staying at or below 25% of the start-of-period balance for at least 61 days. See Table II for additional variable definitions. *t*-statistics are in parentheses, calculated using asymptotically normal standard errors. ***, **, and * indicate significance at 99%, 95%, and 90%, respectively. † TAG/DFA coverage was effective only in the Post-Crisis and Formal periods; eligible accounts were fully covered/insured in those periods. In earlier periods, eligibility does not imply coverage, and is shown for comparison purposes only. ‡ Note that although TAG/DFA did not directly insure term deposits, it could do so indirectly. Accounts covered by TAG/DFA were excluded in calculating a depositor's total exposure to a bank when applying standard, limited-value deposit insurance, effectively freeing up insurance coverage to be applied to other, non-TAG/DFA-covered accounts.

	<i>Placebo</i> (1)	<i>Pre-Crisis</i> (2)	<i>Post-Crisis</i> (3)	<i>Formal</i> (4)
Uninsured	0.00501 (0.39)	0.0432*** (3.78)	0.0820** (2.03)	0.147*** (2.92)
TAG/DFA Eligible ^{†‡}	-0.333 (-1.12)	-0.573 (-1.24)	-0.00332 (-0.02)	- (-)
Log(Age)	-0.00664* (-1.81)	-0.0391*** (-10.60)	-0.00379 (-1.01)	-0.0124* (-1.78)
Log(Days to Maturity)	-0.144*** (-28.94)	-0.0695*** (-21.01)	-0.200*** (-54.37)	-0.0866*** (-15.90)
Brokered/Placed	0.254*** (13.08)	0.162*** (7.64)	0.279*** (21.79)	0.550*** (25.76)
Institutional - Listed/Faxed	-0.159 (-0.65)	-0.244 (-0.91)	0.0105 (0.10)	-0.166*** (-3.08)
Institutional - Other	0.0764* (1.80)	0.0880* (1.72)	0.0180 (0.55)	-0.0358 (-1.15)
Trust	-0.0745*** (-3.33)	0.00318 (0.15)	-0.00600 (-0.26)	0.00651 (0.18)
Constant	0.891*** (20.92)	0.815*** (17.79)	1.185*** (32.15)	0.683*** (13.75)
Branch Controls	Yes	Yes	Yes	Yes
No. of Accounts	7,868	10,881	8,509	3,726
No. of Liquidations	2,193	7,153	2,559	1,736
<i>R</i> ²	11.4%	5.4%	31.6%	34.0%
Model <i>p</i> -value	< 0.001	< 0.001	< 0.001	< 0.001

Table VII
Who Withdraws? Term Deposits; Alternative Specifications

This table presents estimates from Probit (columns (1) and (2)) and Cox proportional hazard models (columns (3) and (4)) of the probability of account liquidation during the Post-Crisis and Formal periods. Liquidation is defined as withdrawing 75% or more of the start-of-period account balance and staying at or below 25% of the start-of-period balance for at least 61 days. See Table II for additional variable definitions. Probit estimates are expressed as marginal effects (with a coefficient of 0 reflecting a variable having no effect on the probability of liquidation), and Cox estimates are expressed as hazard ratios (with 1 indicating no effect). $***$, $**$, and $*$ indicate significance at 99%, 95%, and 90%, respectively. \dagger Note that although TAG/DFA did not directly insure term deposits, it could do so indirectly. Accounts covered by TAG/DFA were excluded in calculating a depositor's total exposure to a bank when applying standard, limited-value deposit insurance, effectively freeing up insurance coverage to be applied to other, non-TAG/DFA-covered accounts.

	<i>Probit</i>		<i>Cox P.H.</i>	
	<i>Post-Crisis</i>	<i>Formal</i>	<i>Post-Crisis</i>	<i>Formal</i>
	(1)	(2)	(3)	(4)
Uninsured	0.110*	0.167***	1.637**	1.746***
	(1.89)	(2.80)	(2.55)	(2.99)
TAG/DFA Eligible \dagger	0.00446	-	0.915	-
	(0.03)	-	(-0.12)	-
Log(Age)	-0.00498	-0.0156	1.017	1.053*
	(-0.98)	(-1.52)	(0.81)	(1.92)
Log(Days to Maturity)	-0.214***	-0.115***	0.545***	0.664***
	(-43.56)	(-15.28)	(-56.75)	(-28.43)
Brokered/Placed	0.337***	0.586***	3.678***	8.565***
	(17.44)	(29.93)	(21.11)	(26.58)
Institutional - Listed/Faxed	-0.169	-0.233***	0.788	0.465**
	(-1.18)	(-3.35)	(-0.24)	(-2.14)
Institutional - Other	-0.0135	-0.0498	0.914	1.239
	(-0.31)	(-1.15)	(-0.46)	(1.50)
Trust	-0.00735	0.0147	0.985	1.220
	(-0.24)	(0.31)	(-0.11)	(1.31)
Branch Controls	Yes	Yes	Yes	Yes
No. of Accounts	8,509	3,726	8,509	3,726
No. of Liquidations	2,559	1,736	2,559	1,736
Log Likelihood	-3767.9	-1856.9	-21333.4	-12883.1
Model <i>p</i> -value	< 0.001	< 0.001	< 0.001	< 0.001

Table VIII
Uninsured Transaction Account Migration

For all transaction accounts that had a balance of \$2,000 less than the current deposit insurance limit or higher at the beginning of each period, this table shows their distribution into various account-size bins at the end of the period. Several bins are suppressed for disclosure reasons.

Panel A. Deposit Insurance Limit = \$100,000						
Bin Range	\$2,000- \$48,000- \$98,000- 102,000 >\$102,000					
	<\$1	\$1 - 2,000	48,000	98,000	102,000	>\$102,000
Placebo	4.98%	9.08%	11.42%	10.32%	11.71%	52.59%
Pre-Crisis	7.71%	9.59%	9.88%	15.59%	16.35%	40.88%
Panel B. Deposit Insurance Limit = \$250,000						
Bin Range	\$2,000- \$123,000- \$248,000- 252,000 >\$252,000					
	<\$1	\$1 - 2,000	123,000	248,000	252,000	>\$252,000
Post-Crisis	—	6.37%	14.22%	13.73%	—	65.69%
Formal	20.79%	6.6%	22.3%	14.04%	8.05%	27.72%

Table IX
Comparison of Depositors Across Periods

Each column corresponds to a separate account-level OLS regression of an account characteristic on the interaction of time period dummies and an indicator for extant or new depositors. The observational unit is account-time period, and with five time periods it is possible for an individual account to appear up to five times in this regression. Extant depositors are those with an existing deposit relationship at the beginning of each time period, and new depositors are those who arrive during the period. The omitted category is extant depositors in the Placebo period, so all estimates in other rows reflect differences relative to that category. *t*-statistics are in parentheses, calculated using asymptotically normal standard errors. ***, **, and * indicate significance at 99%, 95%, and 90%, respectively. † TAG/DFA coverage was only effective in the Post-Crisis and Formal periods; eligible accounts were fully covered/insured in those periods. In earlier periods, eligibility does not imply coverage and is shown for comparison purposes only.

	<i>Uninsured</i>	<i>TAG/DFA Eligible[†]</i>	<i>Brokered/Placed</i>	<i>Institutional: Listed/Faxed</i>	<i>Checking</i>	<i>CD</i>
Var. Type						
Panel A. Extant Depositors						
Placebo					<i>Omitted Dummy</i>	
Pre-Crisis	0.0212*** (15.29)	0.00205*** (3.82)	-0.000476 (-0.47)	-0.00000572 (-0.01)	-0.00803*** (-4.17)	0.0521*** (19.08)
Post-Crisis	-0.0558*** (-38.71)	-0.00226*** (-4.04)	0.0325*** (30.92)	0.000240 (0.29)	0.00254 (1.27)	0.0315*** (11.11)
Formal	-0.0445*** (-28.91)	-0.00188*** (-3.15)	0.0242*** (21.64)	0.00197** (2.23)	0.0328*** (15.37)	-0.0674*** (-22.27)
Failure Date	-0.0509*** (-31.27)	-0.00155** (-2.46)	-0.00707*** (-5.97)	0.0843*** (90.07)	0.0543*** (24.02)	0.00108 (0.34)
Panel B. New Depositors						
Placebo	-0.00160 (-0.42)	0.00903*** (6.08)	-0.0105*** (-3.78)	-0.0000725 (-0.03)	-0.0242*** (-4.55)	0.271*** (35.92)
Pre-Crisis	0.0221*** (4.91)	0.0316*** (18.09)	0.0220*** (6.70)	0.00552** (2.13)	0.0674*** (10.77)	0.287*** (32.42)
Post-Crisis	-0.0208*** (-3.07)	0.0100*** (3.83)	0.209*** (42.42)	0.0129*** (3.31)	0.0987*** (10.51)	0.227*** (17.09)
Formal	-0.0377*** (-9.16)	0.00230 (1.44)	-0.0125*** (-4.18)	0.707*** (298.88)	0.0158*** (2.77)	0.634*** (78.30)
Constant	0.0650*** (64.97)	0.00616*** (15.90)	0.0125*** (17.20)	0.0000725 (0.13)	0.0761*** (54.73)	0.190*** (96.59)
No. of Obs.	188,834	188,834	188,834	188,834	188,834	188,834
<i>R</i> ²	2.4%	0.3%	1.9%	35.0%	0.7%	5.2%
Model <i>p</i> -value	< 0.001	< 0.001	< 0.001	< 0.001	< 0.001	< 0.001
<i>F</i> -statistic	572.9	61.22	463.5	12708.7	161.3	1299.1

Table X
What Explains the Share of Deposits Which Are New?

This table presents estimates from daily frequency regressions of the share of deposit dollars at the bank that are new (as of that day) on various controls. All models are OLS with Newey-West standard errors of lag length 9, chosen by the Newey-West rule of thumb ($0.75\sqrt[3]{T}$) (Stock and Watson, 2019). ***, **, and * indicate significance at 99%, 95%, and 90%, respectively.

	(1)	(2)	(3)	(4)
Time Period Dummies:				
<i>Pre-Placebo</i>	5.439*** (2.90)	1.337 (1.13)	1.925* (1.65)	1.883* (1.65)
<i>Placebo to Pre-Crisis</i>	2.027** (2.01)	2.054** (2.21)	3.079*** (3.05)	2.889*** (2.84)
Pre-Crisis	-1.442*** (-2.72)	0.378 (0.25)	2.907 (1.58)	2.735 (1.49)
<i>Crisis</i>	0.188 (0.25)	2.069 (0.92)	3.417 (1.40)	3.157 (1.30)
Post-Crisis	-0.393 (-0.60)	3.044 (0.98)	8.875** (2.20)	8.685** (2.15)
<i>Post-Crisis to Formal</i>	-1.254** (-2.18)	2.230 (0.74)	9.264** (2.24)	8.919** (2.16)
Formal	5.821*** (2.79)	7.495** (2.14)	14.91*** (2.99)	14.48*** (2.90)
Economic Controls:				
Log(VIX)	1.137 (0.63)	-1.263 (-0.51)	-0.806 (-0.32)	
GDP Growth	0.339** (2.21)	0.543*** (2.60)	0.514** (2.46)	
Housing Starts	0.00442 (1.57)	0.00415 (1.46)	0.00430 (1.47)	
Daily S&P 500 Return	18.19 (1.34)	13.93 (1.02)	15.04 (1.10)	
AR(1)	0.390*** (7.05)	0.379*** (7.04)	0.378*** (7.02)	
OFR Financial Stress Index	0.139 (0.95)	0.380* (1.74)	0.328 (1.48)	
Rate Spread to Market (Dollar Weighted Average)		5.549*** (2.71)	5.742*** (2.77)	
Past-Month Fees/Deposit Balances (in Aggregate)			1083.8 (0.96)	
Constant	3.218*** (6.62)	-8.348 (-1.62)	-12.01** (-2.55)	-13.45*** (-2.62)
<i>N</i>	2,079	2,078	2,078	2,078
Model <i>p</i> -value	< 0.001	< 0.001	< 0.001	< 0.001

Table XI
What Explains the Log-Level of New Deposits?

This table presents estimates from daily frequency regressions of the logarithm of new deposit dollars at the bank (as of that day) on various controls. All models are OLS with Newey-West standard errors of lag length 9, chosen by the Newey-West rule of thumb ($0.75\sqrt[3]{T}$) (Stock and Watson, 2019). ***, **, and * indicate significance at 99%, 95%, and 90%, respectively.

	(1)	(2)	(3)	(4)
Time Period Dummies:				
<i>Pre-Placebo</i>	1.152 (1.57)	0.410 (0.54)	0.673 (0.90)	0.662 (0.89)
<i>Placebo to Pre-Crisis</i>	0.991 (1.46)	0.616 (0.88)	1.084 (1.51)	1.039 (1.39)
Pre-Crisis	-1.822** (-2.57)	-0.123 (-0.11)	1.064 (0.95)	1.023 (0.90)
<i>Crisis</i>	-1.167 (-1.30)	1.337 (0.78)	1.963 (1.15)	1.902 (1.09)
Post-Crisis	-1.066 (-1.17)	0.718 (0.38)	3.440* (1.67)	3.394 (1.63)
<i>Post-Crisis to Formal</i>	-2.225*** (-3.60)	-0.840 (-0.46)	2.447 (1.16)	2.364 (1.09)
Formal	0.671 (0.95)	1.798 (0.98)	5.229** (2.44)	5.124** (2.32)
Economic Controls:				
Log(VIX)		4.182*** (4.20)	3.055*** (2.72)	3.160*** (2.62)
GDP Growth		-0.0225 (-0.25)	0.0706 (0.74)	0.0639 (0.64)
Housing Starts		0.00178 (1.11)	0.00163 (1.01)	0.00167 (1.02)
Daily S&P 500 Return		-8.577 (-0.65)	-10.40 (-0.78)	-10.14 (-0.75)
AR(1)		0.276*** (19.61)	0.274*** (19.48)	0.274*** (19.42)
OFR Financial Stress Index		-0.266*** (-2.81)	-0.155 (-1.39)	-0.167 (-1.38)
Rate Spread to Market (Dollar Weighted Average)			2.606*** (3.15)	2.649*** (3.18)
Past-Month Fees/Deposit Balances (in Aggregate)				251.3 (0.30)
Constant	11.09*** (21.17)	-6.523** (-1.97)	-8.184** (-2.41)	-8.515** (-2.38)
<i>N</i>	2,079	2,078	2,078	2,078
Model <i>p</i> -value	< 0.001	< 0.001	< 0.001	< 0.001

Table XII
Impact of Regulatory Action on Funding Shares For U.S. Banks

This table presents estimates from regressing funding shares on a regulatory action dummy and bank-level controls for all U.S. banks from 2000 to 2016. The regulatory action dummy is based on public and confidential supervisory data; other variables are from the regulatory filings of all U.S. banks. Dependent variables are expressed in percentage points. *NPL* refers to non-performing loans and *A* is total assets. Observational units are bank-quarters, with brokered deposit data available from 2000 to 2016, listing service deposits from 2011 to 2016, and all other term deposit data from 2010 to 2016. Small term deposits are those with balances of \$100,000 or less, medium term deposits are between \$100,001 and \$250,000, and large term deposits are those over \$250,000. *t*-statistics are in parentheses, calculated using asymptotically normal standard errors. ***, **, and * indicate significance at 99%, 95%, and 90%, respectively.

	<i>Brokered</i>	<i>Listed</i>	<i>Sm. Term</i>	<i>Med. Term</i>	<i>Lg. Term</i>
	(1)	(2)	(3)	(4)	(5)
Under Reg. Action	-1.24*** (-22.72)	0.826*** (11.40)	0.364*** (5.21)	0.466*** (6.25)	-0.830*** (-12.10)
<i>NPL</i> _{<i>t</i>} / <i>A</i> _{<i>t</i>}	0.0102** (2.38)	0.00224 (0.42)	0.117*** (22.32)	-0.00903 (-1.62)	-0.108*** (-20.97)
% Δ <i>A</i> _{<i>t</i>}	0.0238*** (54.53)	0.0127*** (18.82)	-0.0282*** (-41.97)	0.0215*** (29.95)	0.00668*** (10.13)
<i>ln</i> (<i>A</i> _{<i>t</i>})	2.83*** (137.87)	0.354*** (8.31)	1.27*** (31.87)	-0.837*** (-19.66)	-0.432*** (-11.05)
Deposits _{<i>t</i>} / <i>A</i> _{<i>t</i>}	4.29*** (35.62)	1.62*** (7.44)	1.79*** (8.33)	1.81*** (7.92)	-3.60*** (-17.11)
Term Dep _{<i>t</i>} / <i>A</i> _{<i>t</i>}	14.4*** (173.14)	16.2*** (112.02)	29.1*** (211.50)	39.3*** (267.68)	31.6*** (234.54)
Constant	-41.9*** (-176.09)	-11.6*** (-22.44)	-3.58*** (-7.40)	4.15*** (8.04)	-0.575 (-1.21)
Bank & Quarter FE	Yes	Yes	Yes	Yes	Yes
<i>N</i>	554,180	162,123	193,306	193,306	193,306
<i>R</i> ²	13.6%	8.3%	68.8%	33.8%	24.1%
Model <i>p</i> -value	< 0.001	< 0.001	< 0.001	< 0.001	< 0.001

Table XIII
Impact of Regulatory Action on Funding Shares For U.S. Banks, Quarterly Dummies

This table presents estimates from regressing funding shares on quarterly regulatory action dummies and bank-level controls for all U.S. banks from 2000 to 2016. The regulatory action dummies are based on public and confidential supervisory data; other variables are from the regulatory filings of U.S. banks. Dependent variables are expressed in percentage points. Observational units are bank-quarters, with brokered deposit data available from 2000 to 2016, listing service deposits from 2011 to 2016, and all other term deposit data from 2010 to 2016. t -statistics are in parentheses, calculated using asymptotically normal standard errors. Small term deposits are those with balances of \$100,000 or less, medium term deposits are between \$100,001 and \$250,000, and large term deposits are those over \$250,000. Time period t is the quarter in which the regulatory action began. ***, **, and * indicate significance at 99%, 95%, and 90%, respectively.

	<i>Brokered</i>	<i>Listed</i>	<i>Sm. Term</i>	<i>Med. Term</i>	<i>Lg. Term</i>
	(1)	(2)	(3)	(4)	(5)
$(\tau - 4)$	1.17*** (11.71)	-0.958*** (-5.00)	0.443** (2.49)	-0.857*** (-4.51)	0.414** (2.37)
$(\tau - 3)$	1.07*** (11.12)	-0.951*** (-5.63)	0.725*** (4.68)	-0.863*** (-5.22)	0.138 (0.91)
$(\tau - 2)$	0.951*** (10.20)	-0.633*** (-4.01)	0.759*** (5.29)	-0.838*** (-5.46)	0.0785 (0.56)
$(\tau - 1)$	0.704*** (8.03)	-0.535*** (-3.78)	0.653*** (5.03)	-0.663*** (-4.77)	0.00987 (0.08)
(τ)	0.729*** (8.33)	-0.260* (-1.91)	0.639*** (5.14)	-0.559*** (-4.21)	-0.0797 (-0.65)
$(\tau + 1)$	0.512*** (4.03)	0.00124 (0.01)	1.22*** (7.90)	-0.743*** (-4.51)	-0.476*** (-3.14)
$(\tau + 2)$	-0.667*** (-4.50)	0.477*** (2.72)	1.40*** (9.10)	-0.457*** (-2.77)	-0.945*** (-6.25)
$(\tau + 3)$	-1.48*** (-9.13)	0.594*** (3.54)	1.14*** (7.63)	-0.0210 (-0.13)	-1.12*** (-7.62)
$(\tau + 4)$	-2.13*** (-12.53)	0.669*** (4.10)	0.759*** (5.06)	0.284* (1.77)	-1.04*** (-7.08)
$(\tau + i), i \geq 5$	-3.85*** (-41.34)	1.21*** (12.85)	-0.000122 (-0.00)	1.14*** (11.19)	-1.14*** (-12.18)
NPL_t/A_t	0.0192*** (4.43)	0.00499 (0.94)	0.114*** (21.74)	-0.00804 (-1.43)	-0.106*** (-20.58)
$\% \Delta A_t$	0.0233*** (53.55)	0.0126*** (18.66)	-0.0280*** (-41.63)	0.0213*** (29.65)	0.00668*** (10.12)
$ln(A_t)$	2.79*** (135.80)	0.383*** (8.98)	1.22*** (30.38)	-0.761*** (-17.78)	-0.455*** (-11.58)
$Deposits_t/A_t$	4.33*** (36.05)	1.55*** (7.13)	1.86*** (8.69)	1.65*** (7.21)	-3.52*** (-16.70)
$Term Dep_t/A_t$	14.4*** (173.11)	16.2*** (112.00)	29.0*** (211.38)	39.3*** (267.89)	31.6*** (234.61)
Constant	-41.4*** (-174.03)	-11.8*** (-22.96)	-3.01*** (-6.22)	3.40*** (6.56)	-0.384 (-0.81)
Bank & Quarter FE	Yes	Yes	Yes	Yes	Yes
N	554,180	162,123	193,306	193,306	193,306
R^2	13.9%	8.5%	68.8%	33.9%	24.1%
Model p -value	< 0.001	< 0.001	< 0.001	< 0.001	< 0.001

Table XIV
Matched Effects of Regulatory Action on Funding Shares for All U.S. Banks,
Four Quarters after Treatment

This table presents estimates from propensity score matching for a regulatory treatment based on bank covariates to measure the effects on certain deposit products as a share of total deposits for all U.S. banks. The regulatory action dummy is based on a combination of public and confidential supervisory data; all other variables are from the regulatory filings of all U.S. banks. Dependent variables are expressed in percentage points. Observation units are bank-quarters, with brokered deposit data available from 2000 to 2016, listing service deposits from 2011 to 2016, and all other term deposit data from 2010 to 2016. Only banks that at some time come under an enforcement action between 2000 and 2016 are used for matching. *t*-statistics are calculated using asymptotically normal standard errors. Matching is done using a logistic model to generate a propensity score for being treated with regulatory action based on the covariates used in the above regressions (1 year asset growth, natural log of assets, nonperforming loans as share of assets, deposits as a share of assets) as well as all outcome variables (brokered deposits, listing service deposits, and all three categories of term deposits as shares of deposits). Treated banks in one quarter are then matched to untreated banks in the same quarter at the time of treatment based on this score. Matches are done based on normal kernel weighting, so that close matches are weighted proportionally more. We then observe the difference between treated and untreated four quarters after treatment.

	<i>Brokered</i> (1)	<i>Listed</i> (2)	<i>Sm. Term</i> (3)	<i>Med. Term</i> (4)	<i>Lg. Term</i> (5)
Untreated	4.07	3.12	17.54	14.21	5.27
Matched Untreated	3.65	4.37	19.96	16.77	5.17
Treated	1.95	9.98	23.41	22.78	4.30
T-Stat on Differences	-3.46	3.90	3.88	4.73	-2.63
<i>N</i> Untreated	29,611	29,597	29,597	29,597	29,597
<i>N</i> Treated	142	142	142	142	142

Table XV
Large Bank Deposit Rate Responses to Enforcement Actions

<i>\$100K, 12m CD spread</i>	
Under Reg. Action	0.4121*** (4.48)
Constant	0.3547*** (2.99)
Quarter FEs	Yes
<i>N</i>	65
<i>R</i> ²	6.9%
Model <i>p</i> -value	< 0.001

This table presents estimates from pooled OLS regression of 12-month, \$100,000 CD rate spreads (measured against published FDIC national average rates) on a dummy indicating the bank is currently under regulatory enforcement action, time fixed effects, and a constant. Units are in percentage points: a value of 0.01 corresponds to 1 basis point (0.01%). The sample of banks is constrained to those over \$5 billion in assets (as of 2008Q4) that are under enforcement action or will be within four quarters. The coefficient on *Under Reg. Action* can therefore be interpreted as large banks' average change in deposit rates around/following the imposition of enforcement actions. *t*-statistics are in parentheses, calculated using asymptotically normal standard errors. ***, **, and * indicate significance at 99%, 95%, and 90%, respectively.

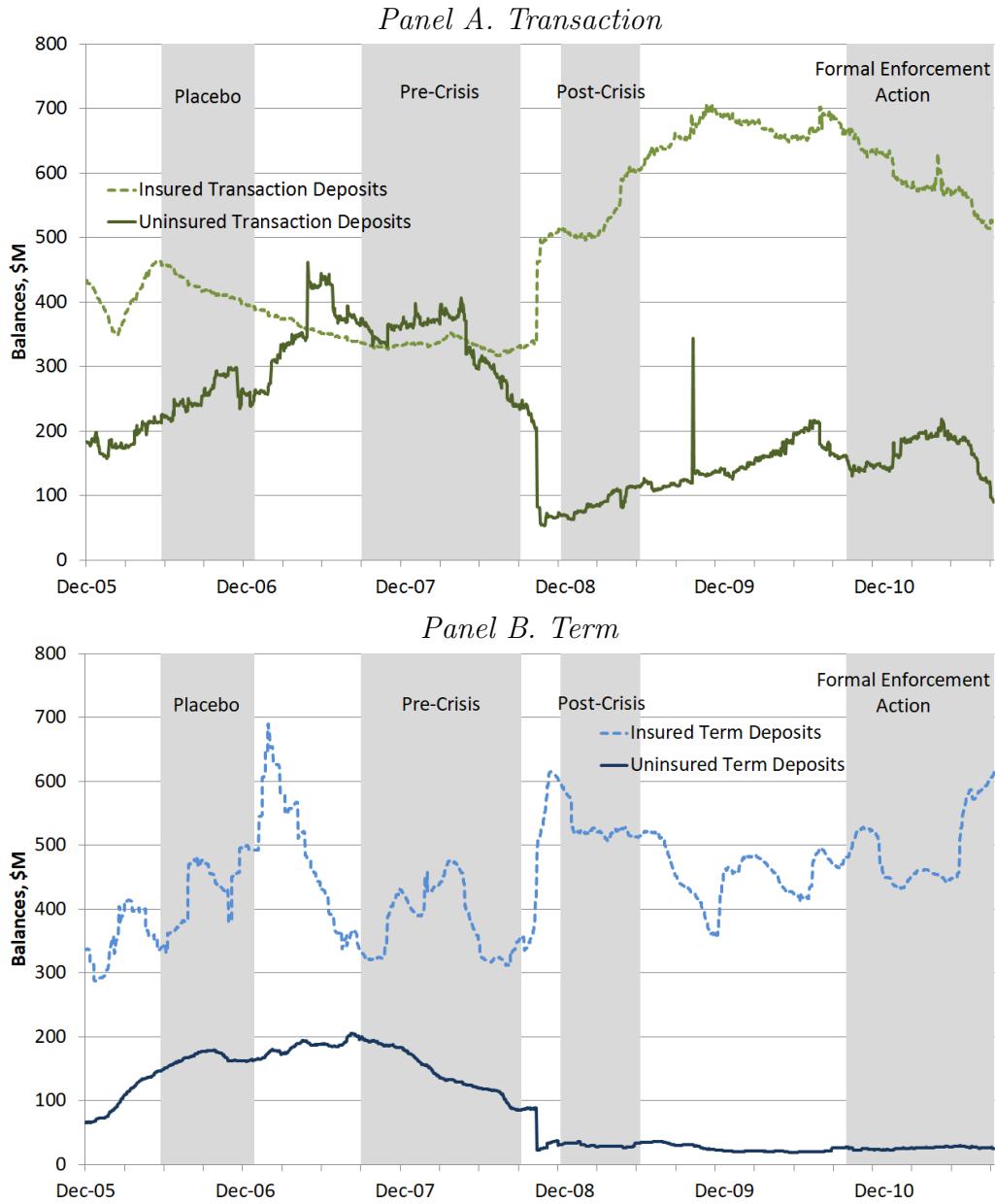


Figure 1: Deposit balances. This figure shows total balances in transaction (Panel A) and term (Panel B) deposit accounts. Grey bars denote the time periods analyzed in the regressions of Section II., and overlaid text identifies the name of each period. Note that the dramatic, brief spike in uninsured transaction deposits between the Post-Crisis and Formal periods reflects a single transaction in which another subsidiary of the bank's holding company passed funds through the bank in such a manner that they remained within the bank for a few days.

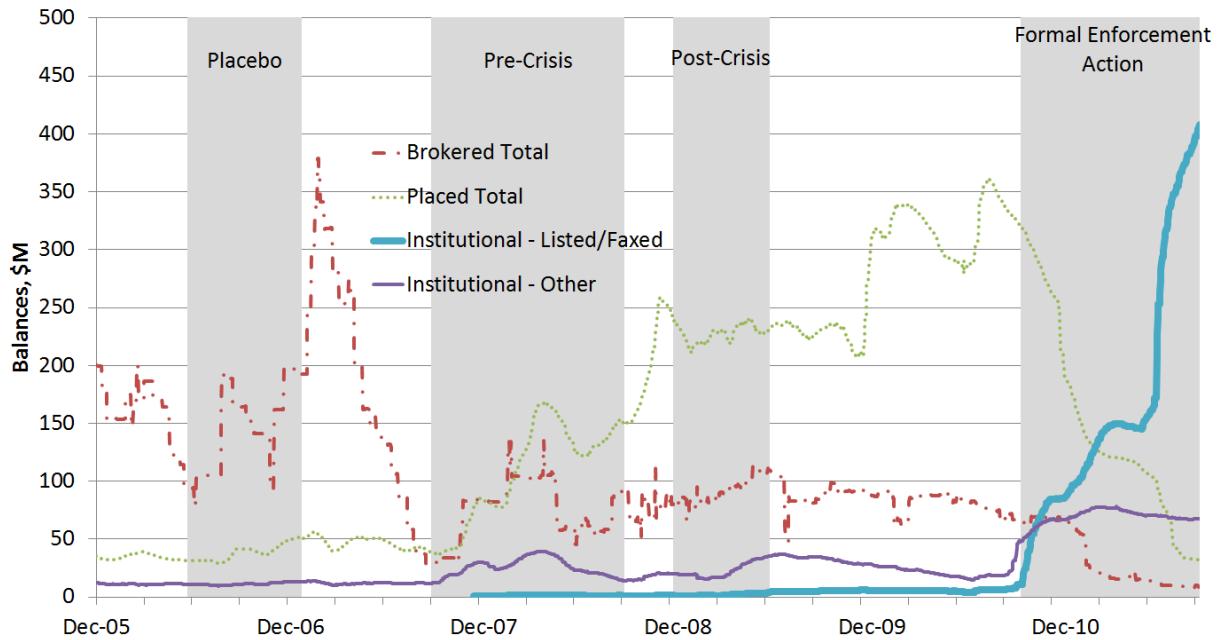


Figure 2: Term deposit balances in brokered, placed, and institutional accounts. This figure shows term deposit account balances in brokered accounts (dash-dotted red line), placed accounts (dotted green line), institutional deposits obtained via rate listing services and faxes (bold blue line), and other institutional deposits (solid, thin purple line). Placed deposits are those placed by a third party on behalf of the underlying depositor, where that third party does not meet the definition of deposit broker. Note that this is a different notion of placed deposits relative to that used in the regressions; here, we split placed and brokered deposits into two categories, whereas both were grouped as “placed” in the regressions. Among placed and brokered deposits, the underlying depositors are often not identified to the bank accepting the deposits. Grey bars denote the time periods analyzed in the regressions of Section II., and overlaid text identifies the name of each period.

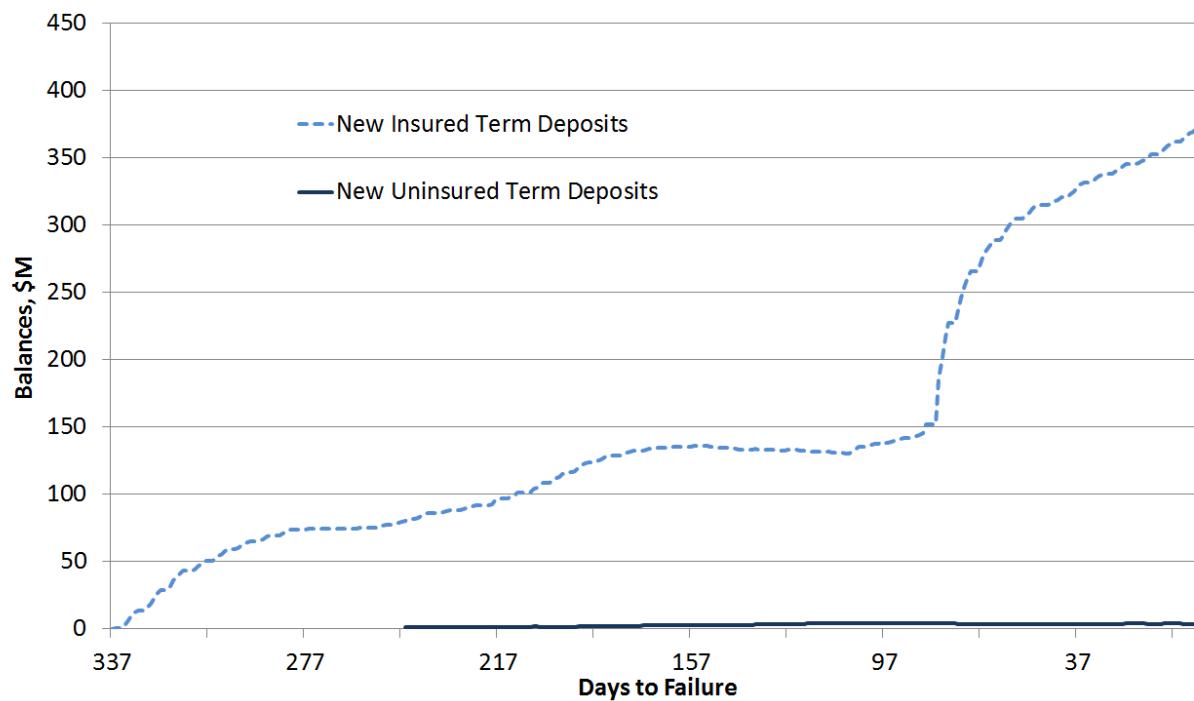


Figure 3: Term deposit balances from new depositors. This figure shows balances in term deposit accounts from depositors who opened their first deposit account with the bank after the formal enforcement action — new depositors.

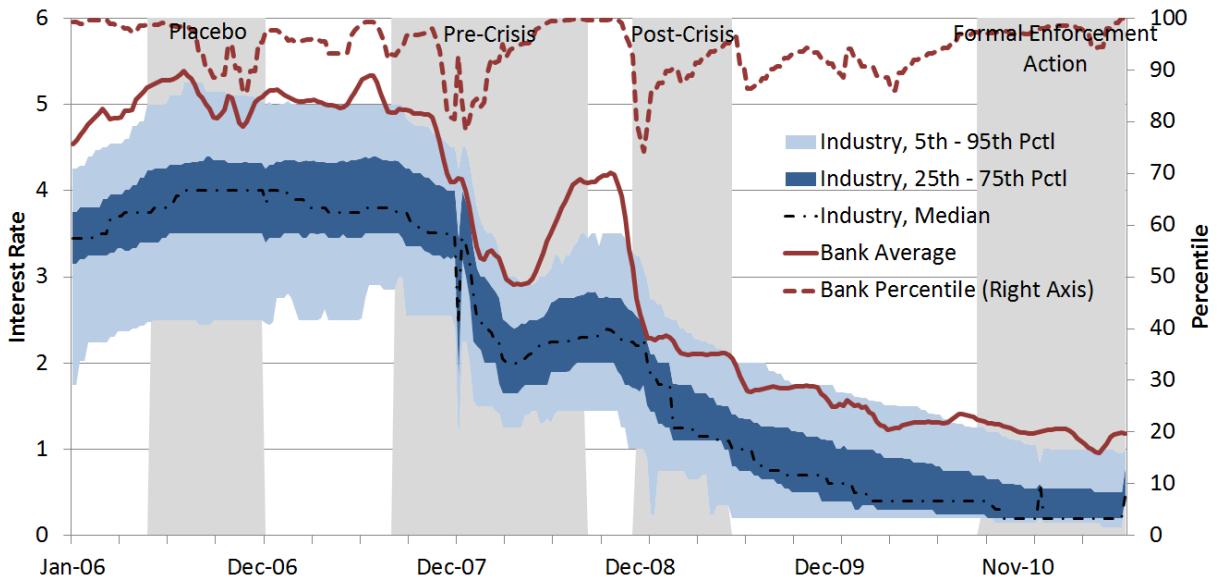


Figure 4: 12-month CD rates relative to the market distribution. This figure shows the 31-day moving average (fifteen days before, day of, and fifteen days after) of all rates offered by the bank on newly issued 12-month term deposits with balances below \$100,000 (“Bank Average;” solid red line) on the left axis, while the percentile relative to the distribution of banking industry rates (from *RateWatch*) for the same product is shown on the right axis. The bank average series is a measure of the rate that would have been faced by a depositor considering depositing funds at the bank that day. “Newly issued” term deposits include newly established term deposit accounts as well as rollovers of existing term deposits upon the expiration of the previous product.

Panel A. Brokered and listed deposits



Panel B. Term deposits, various sizes

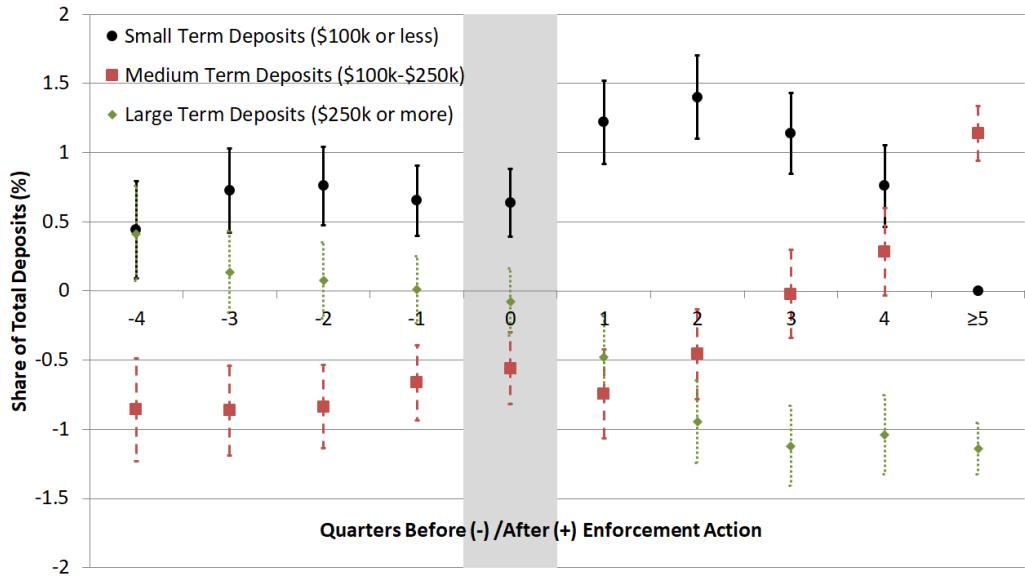


Figure 5: Impact of regulatory action on funding shares. This figure shows the impact of regulatory action on funding shares, using the estimates from Table XIII. Small term deposits are those with balances of \$100,000 or less, medium term deposits are between \$100,001 and \$250,000, and large term deposits are those over \$250,000. Time period t is the quarter in which the enforcement action was issued. Error bands represent 95% asymptotic confidence intervals.