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Leverage limits and bank risk: New evidence on an old question

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Federal Reserve Bank of New York Staff Reports

Leverage Limits and Bank Risk: New Evidence on an Old Question

Dong Beom Choi Michael R. Holcomb Donald P. Morgan

Staff Report No. 856 June 2018



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Leverage Limits and Bank Risk: New Evidence on an Old Question

Dong Beom Choi, Michael R. Holcomb, and Donald P. Morgan Federal Reserve Bank of New York Staff Reports, no. 856 June 2018

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Abstract

The supplementary leverage ratio (SLR) rule recently imposed on the very largest U.S. banks has revived the question of whether banks sidestep such rules by shifting toward riskier, higher-yielding assets. Using difference-in-difference analysis, we find that, after the SLR was finalized in 2014, covered banks shifted their portfolio toward riskier (risk-weighted) assets and higher-yielding securities compared to other large banks not subject to the rule. The shifts are sizable and tend to be larger at banks more constrained ex ante by the leverage limit. Despite increased asset risk, overall bank risk (book and market measures) did not increase, suggesting the higher capital required under the new rule offset the risk-shifting. Taken together, our findings reinforce regulators' long-standing concerns about risk-shifting around leverage limits and suggest that the recent recalibration will curb those incentives without necessarily increasing bank risk.

Key words: Basel III regulation, bank risk, leverage limit, regulatory arbitrage, reaching for yield

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Because our existing capital standards treat all bank assets alike, they have had the effect of encouraging some institutions to scale back their holdings of relatively liquid, low-risk assets (Paul Volker, 1987)

...a leverage requirement that is too high favors high-risk activities and disincentivizes low-risk activities (Randal K. Quarles, 2018)

I. Introduction

The quotes above illustrate the *déjà vu* aspect of our question. The first is Fed Chairman Volker explaining to Congress why bank regulators were moving from the leverage rule imposed in 1981 to more risk-based capital rules. The second is the current Fed Vice Chairman for Supervision explaining why the new leverage rule that took effect this year was being recalibrated to curb risk-shifting incentives. Bankers, for their part, have also warned the new rule could have unintended effects:

... the proposal would have discouraged banks from holding low-yielding, high-quality assets...in preference for riskier assets which would produce a higher relative return of capital. (J.P. Morgan (2014))

Despite these long-standing and widely held concerns, published evidence for the risk-shifting conjecture is relatively scarce. The leverage rule imposed in 1981 motivated several theoretical papers predicting potentially perverse consequences (Koehn and Santomero (1980), Kim and Santomero (1988)), but Furlong (1988) seems to be the only empirical test and finds evidence that contradicts the theoretical conjectures. The paucity of evidence from that first experiment might reflect the fact that the first leverage rule covered all banks and was rarely binding, making a plausible control group difficult to identify. By contrast, the Supplementary Leverage Ratio (SLR) covers only the very largest U.S. banks, and for some, may bind more than their risk-based capital limit. Thus, this latest leverage rule "experiment" provides a better setting to test the risk-shifting conjecture.

We look for evidence of risk-shifting around the new leverage rule using difference-indifference analysis. Our treated group comprises the 15 bank holding companies ("banks") subject to the new rule, the so-called "advanced approach" firms that use internally generated risk estimates for setting their risk-based capital requirements. Concerns that these firms might be underestimating risk and overstating their risk-based capital strength motivated the new leverage ratio as a supplement or back stop (Basel Committee on Banking Supervision 2009).² Advance approach firms, by definition, are very large; at least \$250 billion in total assets or \$10 billion in foreign exposures. Our control group comprises the 18 banks with assets between \$50 and \$250 billion not covered by the new rule. Though smaller, these banks are officially large per the Dodd-Frank Act of 2010 and thus face similar regulatory environments apart from the SLR. Using standard difference-in-difference regressions with time and institution fixed effects and a parsimonious set of controls (including lagged risk-based capital ratios), we compare various risk measures before and after the SLR was finalized in September 2014.

We study risk three ways. We first look directly at banks' risk-weighted asset shares using Basel risk-weighting classifications. The validity of that measure hinges, of course, on the accuracy of banks' risk weights and it was doubts surrounding those that motivated the new rule in the first place. Our second measure, security yields, is independent of reported risks. We gathered banks' individual security holdings from the confidential regulatory filings then matched with yields from a variety of sources. If leverage-constrained banks are shedding safer assets and "reaching for yield" on riskier ones, it should register in these data even if obscured by errors in reported, risk-weighted assets. Lastly, we look at measures of overall bank risk. Even if covered banks are shifting to riskier assets, they may not be riskier overall if they are less levered than they would have been otherwise. The overall risk measures we examine, both book and market, capture the combined effect of any risk-shifting and reduced leverage.

Our findings are largely consistent with the risk-shifting conjecture. Risk weighted assets at covered banks have increased particularly at banks for which the leverage constraint is tighter. We detect the shift in securities and trading assets, but not loans, which seems sensible as loans are arguably the least adjustable asset class. Consistent with "reach for yield" concerns, we find a sizable, about 30 basis point, increase in the average yield on securities held by covered banks; we also find suggestive evidence that this gain reflects banks *actively* adding riskier securities to their portfolio, not merely shedding safer ones.

Despite the evident risk-shifting, we find no increase in the battery of overall bank risk measures we examine, including Z scores, CDS spreads, equity volatility, and others, not even

² See Santos and Plosser (2016) for evidence and references.

for the most constrained banks. This suggests the higher capital required by the new rule offsets the risk-shifting, or vice-versa. In fact, we show that leverage ratio at the most constrained banks rose sharply in relative and absolute terms post-implementation.

While others have already "leveraged" the new leverage rule for research purposes, we are the first to study its effects at the overall portfolio and bank risk level for U.S. banks. The other recent literature focuses on narrower segments, the repurchase (repo) market in particular.³ Allahrakha et al. (2016) find that U.S. banks decreased repo borrowing in response the SLR rule introduced and, consistent with our finding, shifted toward more volatile collateral. Bicu et al (2017) find reduced liquidity in the U.K. repo after a related leverage limit was announced there. By contrast, Bucalossi and Scalia (2016) find that leverage-constrained European banks did not reduce their repo volumes between 2013 and 2014.⁴ The paper closest to ours, Acosta Smith et al. (2017), detects a shift toward riskier assets by European banks in response to a leverage rule proposal by the Basel committee, but no increase in overall risk, consistent with our findings. On the policy front, our findings suggest that that recalibrating the new leverage ratio to minimize risk-shifting incentives will not increase bank risk.⁵

Section II reviews the history of leverage limits in the U.S., from the first in 1981 to the latest. Section III discusses our empirical strategy and potential confounding regulatory changes. Section IV describes our data and results for risk-weighted asset shares, securities yields, and overall risk measures. Section V concludes.

II. From Leverage Limits and Back Again

We trace the circle of capital regulation over the last 40 years from leverage limits to risk-based capital rules, then back, partly, to leverage limits.

Concerned about rising failures and falling capital levels across the banking system, U.S. bank regulators announced explicit, uniform capital ratio requirements for the first time in 1981

³ Becker and Ivashina (2015) find evidence of somewhat different form regulatory arbitrage by insurance companies; within credit rating categories, insurance firms tend to shift toward riskier securities. Ellul et al. (2014) study how accounting rules, through the interaction with capital regulations, affect risk-taking of insurance companies.

⁴ Boyarchenko et al. (2017) attribute recent illiquidity in the corporate bond and credit default swap markets to a binding leverage limit.

⁵ https://www.federalreserve.gov/newsevents/pressreleases/bcreg20180411a.htm

(Volker 1987).⁶ Capital adequacy before that was assessed bank-by-bank per supervisors so the shift to formal standards marked an important shift in regulatory policy. The rules required at least 5.5 percent primary capital and 6 percent total capital relative to total, on balance sheet assets. While the capital requirements were conditional on capital quality, they were invariant to asset risk so they were in effect leverage limits.

That first leverage rule spurred a debate on whether, as a theoretical matter, it might actually increase bank risk via asset substitution (Koehn and Santomero (1980), Kim and Santomero (1988), Furlong and Keeley (1989)). The only empirical test of that question at the time appears to be Furlong (1988). Starting with 98 large, public banks, he compares changes in three risk measures between 1981 and 1986 for 24 "capital deficient" banks in 1981 per the new standards and 75 that were "capital sufficient" banks. Contrary to the risk-shifting hypothesis, he finds no significant differential changes in the market-based asset risk measure he constructs or in bank default risk (Z) scores. He does document a relative decline in the ratio of low-risk, liquid assets at the capital deficient banks but that trend began, as he notes, well before the capital rules were introduced.

Lingering concerns about unintended leverage rule effects led the Federal Reserve in 1986 to propose replacing it with risk-based capital requirements based on total assets, including off-balance sheet assets (Volker 1987, Wall 1989). That proposal, in cooperation with international bank regulators, evolved into the Basel I capital accord in 1990. Basel I defined standard risk weights for broad asset classes and set required capital minimums relative to risk-weighted assets. That change curbed incentives under the leverage rule to shift toward riskier categories, though not necessarily within the fairly broad asset classes defined under Basel I. Basel II in 2004 elaborated more risk-sensitive capital requirements and allowed very large, "advanced approach" firms the option of using internal models to estimate asset risk (subject to supervisory review).

Concerns that advanced approach firms might be underestimating risk and exaggerating their capital strength, as well as the 2007-2008 financial crisis, led to the (partial) return of leverage ratios (see Figure 1 below). The Basel Committee recommended a leverage rule (the Basel III leverage ratio) in 2010 and U.S. regulators proposed their version—the Supplementary

⁶ Regulators could still require higher capital at banks with substantial off-balance-sheet exposures or assets considered particularly risky (Gilbert et al. 1985, p. 16)

Leverage Ratio (SLR)—in 2012. The SLR rule requires advanced approach firms to maintain a minimum ratio of Tier 1 capital per total leverage exposures (including off-balance sheet assets) of 3 percent. The "enhanced" SLR rule finalized in 2014 required advanced approach firms that were also designated as global systemically important bank holding companies (G-SIBs) to hold a minimum of 5 percent. The denominator of the SLR was finalized, after much discussion and public comment, in September 2014—a key date in determining how binding the rule would be. Covered banks were required to begin disclosing their SLR on their Investor Relations webpages beginning January, 2015.

Figure 1: SLR Timeline

12/2010:	International (Basel) regulators propose text for LR
6/2012:	U.S regulators propose SLR
7/2013:	U.S. regulators finalize SLR rule; propose eSLR
4/2014:	Finalize eSLR; propose revisions to SLR denominator
9/2014:	SLR denominator finalized
1/2015:	Mandatory disclosures SLR by covered banks
1/ 2018:	SLR and eSLR compliance date

Under the new rule, the SLR is defined by

$$SLR = \frac{Tier\ 1\ Capital}{Total\ leverage\ exposure}$$
,

where the denominator includes both on-balance sheet assets and many off-balance sheet exposures. The risk-invariant aspect of the SLR is obvious; two banks with the same total assets (on- and off-balance sheet) face the same limit, even if one has much riskier assets than the other. By contrast, under the risk-based capital (RBC) requirement, the capital ratio is defined by

$$RBC\ ratio = \frac{Tier\ 1\ Capital}{Risk-weighted\ assets}$$
,

where assets are classified into risk classes with different associated risk weights for each. If two banks have the same total assets, but one has more assets in the riskier classes, its minimum required capital will be higher.

It was widely reported that the new leverage limits incorporating off-balance sheet exposures were more binding than the RBC limits for many covered banks (J.P Morgan 2014). Figure 2 suggests this was in fact the case; among the SLR banks in 2013:Q4, even the most constrained bank had two percentage points of slack relative to its risk-based capital requirement, while eight had slack of less than two percentage points relative to their leverage requirement and six had negative slack.⁷

Banks bound by the SLR limit have two options: increase tier 1 capital or decrease the total leverage exposures. If a bank chooses to raise more (costly) capital, one way to offset the increased costs is by shifting from safer, lower-yielding assets to riskier, higher-yielding ones. If it instead chooses to reduce its assets, the least costly way to do so would be by shedding assets with low yields, such as reserves. In both cases, the bank's share of risky assets relative to safe assets, and its average yield on assets, should rise.

III. Empirical Strategy

We test for risk shifting by SLR banks using difference-in-difference (DD) regressions:

$$\sigma_{it} = \alpha_i + \alpha_t + \beta * SLR_i + \gamma * post_t + \delta * SLR_i * post_t + \gamma * C_{it-1} + \varepsilon_{it}. \tag{1}$$

 σ_{it} is one of several risk measures for bank i at t (described below). The firm fixed effect (α_i) controls for constant risk differences across banks while the year-quarter fixed effect (α_t) controls for time-varying aggregate factors (macro, financial, or monetary) that might affect bank risk. SLR_i equals 1 for banks subject to the rule and 0 otherwise; $post_t$ equals 1 on and after the SLR treatment date (2014:Q3) and 0 before. The coefficient on SLR*post measures the DD effect, i.e. how much the difference in risk (SLR banks minus non-SLR banks) differed before and after the SLR treatment. The risk-shifting hypothesis implies $\delta > 0$.

 C_{it-1} denotes two lagged controls. The first is a proxy for exposure to the new liquidity coverage rule (LCR) imposed at the same time and on the same banks (to different degrees) as

⁷ We calculate the SLR for dates before public disclosure using the "total exposures" field of the FR Y-15. We use 2013:Q4 (instead of 2014:Q2) because the data are only available year-end.

the SLR.⁸ The liquidity rule requires banks to hold a minimum share of "high quality liquid assets" to cover unexpected cash outflows. Since such assets are usually safer as well, the liquidity rule could prevent banks from shifting toward risker assets in order to circumvent the leverage rule, thereby attenuating our estimates of the leverage rule effect.⁹ As banks do not disclose LCR ratios in their public regulatory filings, as a proxy we use the inverse of the liquidity stress ratio calculated at the Federal Reserve Bank of New York.¹⁰

The second control is the risk-based capital ratio for bank *i*. Risk-shifting around the leverage ratio will tend to reduce banks' risk-based capital ratio (if the shift is detected), hence all else equal, banks with higher risk-based capital ratios will have more room to maneuver around the leverage limit.

Identifying the appropriate treatment date for the SLR is not obvious given its roughly six year gestation period. The compliance date (January 2018) might seem like the natural treatment date, but banks almost certainly began adjusting before then to ensure a smooth "glide path" toward public disclosure in 2015 and compliance in 2018. The issuance of the final rule in July 2013 is another potential treatment date, except the rule was not literally final; it was followed by

$$LSR = \frac{\text{potential liquidity inflow}}{\text{potential liquidity outflow}} = \frac{\text{liquidity adjusted assets}}{\text{liquidity adjusted liabilities and off balance sheet exposures}}.$$

The liquidity adjustments reflect the estimated run-risk of each liability type or off balance sheet exposure. Adjusted liabilities grow when the expected funding outflow is higher, as in stressful times, while adjusted assets shrink when the expected price decline is greater in a market illiquidity event. Thus, a bank is more exposed to liquidity risks if its LSR is higher. The liquidity adjustment used in the LSR follow LCR assumptions when possible, but since the LSR is based on only public information, some differences are possible. We use the inverse of LSR to approximate historical LCR. For more discussion, see Choi and Zhou (2016).

⁸ The LCR was finalized by U.S. regulators in 2014:Q3 with phase in starting 2015:Q1. The rule covers all banks in our sample, but the SLR banks face a stricter rule. https://www.federalreserve.gov/newsevents/pressreleases/bcreg20140903a.htm.

⁹ Another confounding regulation is the "accumulated other comprehensive income" (AOCI) rule which may also, like the LCR, induce covered banks to shift toward safer assets. The AOCI rule mostly affects our treatment group of SLR-covered banks, with similar treatment dates as well. Again, the direction of its impact on bank risk-taking is the opposite of the SLR, thus attenuating our estimates.

¹⁰ The liquidity stress ratio (LSR), our proxy for the LCR calculated by the NY Fed Research, is defined as

¹¹ Anecdotally, banks were already adjusting in anticipation of the public disclosures in January 2015: Bank of New York Mellon (https://www.bnymellon.com/_global-assets/pdf/our-thinking/arriving-at-new-capital-ratios.pdf) reports that the "...banking organizations are already making changes to comply with the SLR given that the final rules require public disclosures beginning January 1, 2015." A recent Bloomberg article (https://www.bloomberg.com/news/articles/2018-02-01/deposit-shunning-banks-get-big-break-as-u-s-eases-leverage-rule) reports that "The U.S. gold-plated the Basel leverage requirement for the eight largest lenders when it implemented its own version in 2014, ... The next year, BNY Mellon was still below the 5 percent minimum and State Street barely above. That prompted them to push away deposits with measures that included additional fees on certain types of accounts. They also switched how they swept institutional clients' extra cash, moving them into money-market funds instead of regular deposits."

consultations with the banking industry which offered many suggestions for refining, i.e. shrinking, the denominator. Regulators incorporated some, but not others (e.g. excluding cash and central bank reserves), leading to the revised denominator in September 2014. Only after that date would banks know how the denominator would be calculated, and thus, how constraining the SLR would be.

IV. Data and Findings

We look first at risk-weighted asset shares, then securities yields, then overall risk measures.

a. Risk-Weighted Assets Shares

Risk-weighted assets are reported by bank holding companies in their quarterly reports to bank regulators (FR Y-9C, Schedule HC-R Part II). Reporting standards changed in 2015:Q1, when banks commenced reporting risk-weighted assets according to Basel III definitions. To ensure a consistent series, we adjusted the risk weightings after 2015:Q1 to approximate those before (see Appendix for details). As a precaution, we report results using adjusted and unadjusted weightings; our results are similar for both.

Table 1 reports sample statistics for risk-weighted (RW) asset shares for the SLR banks and non-SLR banks both pre-treatment (2011:Q1-2014:Q2) and post-treatment (2014:Q3-2016:Q2). We look at RW assets overall, and by broad asset class: securities, trading assets, and loans. We expect any risk shifting to be more apparent for securities and trading assets as they are typically more liquid and thus easier to adjust than loans. He unconditional difference-in-difference (DD), reported in Column 5, is positive and significant for RW total assets, securities, and trading assets. The unconditional DD for loans is much smaller and insignificant. Those patterns are evident in adjusted and unadjusted data.

¹² We excluded the non-bank firms Charles Schwab and General Electric.

 $^{^{13}}$ RW assets equals the weighted sum of assets across the regulatory risk classes, where the weights are the weights reported in the FR Y-9C. The RW asset $share = (RW assets/assets) \times 100$. The other RW shares are defined analogously, e.g. RW security share = $(RW securities / securities) \times 100$

¹⁴ "Trading assets" includes securities bought and held "principally for the purpose of selling in the near term." "Securities" includes debt securities banks have the "positive intent" to hold to maturity and also securities that that the bank may retain for long periods but that may also be sold. See <u>Financial Accounting Standards Board</u>, Summary of Statement No. 115

Figure 3 plots the difference (SLR minus non-SLR) in the mean of each variable over our sample period using adjusted and unadjusted risk weights. The trends are largely parallel before 2014:Q3 with some apparent differences after for total assets, securities, and trading assets.

Table 2 reports the conditional DD estimates of model (1). Using the adjusted risk weights (Panel B), the DD for total assets and loans are positive but insignificant. By contrast, the estimates for securities and trading assets are both positive and significant at the ten percent level. Both estimates are sizable relative to the standard deviation in each variable; about two-thirds of a standard deviation for securities and one third for trading assets. The estimates using the unadjusted risk weights differ somewhat (Panel A); the DD for total assets is 3.6 and for securities 5.5, both significant at the 5 percent level. The estimate for trading assets is positive but smaller and insignificant with adjusted data. Overall, we find some evidence of risk-shifting in either case, but where the shifting registers (which asset class) depends on the weighting.

The new leverage limit was much more binding for some SLR banks than others (Figure 2). Although we view the SLR vs non-SLR comparison above as a more exogenous treatment assignment, we also estimated a more flexible version of (1) that allows the SLR effect to vary by "tightness:"

$$\sigma_{it} = \alpha_i + \alpha_t + \beta_1 * SLR \ Tighter_i * post_t + \beta_2 * SLR \ Looser_i * post_t + \gamma * C_{it-1} + \varepsilon_{it}, \ (2)$$

where $SLR\ Tighter_i$ equals 1 for SLR banks with leverage slack at 2013:Q4 below the median for all SLR banks and 0 otherwise. $SLR\ Looser_i$ is defined conversely. Their coefficients measure the DD in for each group relative to non-SLR banks. We predict $\beta_1 > \beta_2 \geq 0$.

The results are reported in Table 3. The estimates for securities are essentially equal, between 5 and 6, contrary to expectations. The estimates for trading assets are arguably more consistent, as only SLR tighter is ever significantly different from zero. ¹⁵ The results for total assets, however, are entirely as predicted; the SLR tighter estimate is about 6 (significant at one to five percent), versus 1.6 or less for SLR looser.

 $^{^{15}}$ We cannot reject the equivalence of β_1 and β_2 using a Wald test.

b. Reaching for Yield?

This section investigates how yields on SLR banks' portfolios of securities changed after the SLR treatment date relative to non-SLR banks. This provides an independent test of the riskshifting hypothesis that is immune to any doubts surrounding risk-weighting classifications.

Bank-level securities yields are not easily obtainable; our collection took several steps. We first gathered the dollar amount of *every* security held by banks in our sample every quarter from FR Y-14Q Schedule B, a confidential report used for stress testing and supervisory purposes. We then collected yields for each security from a number of other data sources, depending on the security class, e.g. Treasuries, municipal bonds, corporate bonds, agency MBS, etc. ¹⁶ Of roughly 209,000 securities identified in step one, we found yields for 43 percent for SLR and 40 percent for non-SLR banks. ¹⁷ Next, we converted the security level yields into weighted average yields by asset class, weighted by each security's share of a bank's holdings of that class. Then, we calculate the overall average weighted by the share of each asset class in the full Y-14Q data set. ¹⁸

Sample statistics, winsorized at the 1 and 99 percentiles, are reported in Table 4. This sample includes the same banks but starts in 2011:Q3 when complete Y-14Q data became available. Mean weighted average yields ("yields" henceforth) were slightly higher at non-SLR banks over the full sample period.

Trends in the mean for both sets of banks are plotted in Figure 4. Pre-treatment, the paths appear roughly similar, with the mean for SLR banks substantially lower than that for non-SLR banks. The trends diverged in 2014:Q3, with yields at the SLR banks rising toward non-SLR levels.

Table 5 reports the DD in yields; the estimate is positive, significant at the 5 percent level and large, around 36 basis points. The mean yield for SLR banks pre-SLR was 2.08. Panel B reports the separate estimates for more or less SLR-constrained banks. As predicted, the estimate

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¹⁶ Yields on Treasury products are from CRSP; municipal bond yields are reported within the Y-14Q and gathered from the MSRB; and yields on corporate bond and structured products (e.g. AMBS, CDO) are from OneTick One Market Data. For mortgage-backed securities, yields are calculated from prices using the Securities Industry and Financial Markets Association (SIFMA) standard formulas.

¹⁷ The match rates were also comparable pre and post-treatment: 38 percent and 49 percent for SLR banks; 42 percent and 37 percent for non-SLR banks.

¹⁸ See the Appendix for details on construction.

for more constrained banks is larger (49 basis points, significant at the 10 percent level) than the estimate for less constrained banks (25 basis points and non-significant).

c. Active Reaching for Yield?

Our results so far are consistent with Volker's (1987) concerns that leverage rules might lead banks to shed safer assets, but do not necessarily imply they are actively replacing them with riskier ones (Quarles 2018). To investigate the latter effect, we study yields at the extensive, investment margin, i.e. securities purchased for the first time over our sample period. Specifically, we estimate the following equation:

$$y_{ist} = \alpha_i + \alpha_t + \beta * SLR_i + \gamma * post_t + \delta * SLR_i * post_t + \gamma * C_{it-1} + \varepsilon_{it}$$
(3)

where y_{ist} is the yield on a newly purchased security s (by CUSIP) purchased at time t by bank i, and all other controls are the same as before. The coefficient on SLR*post measures the DD, i.e. the deviation in the difference of purchased securities yields between SLR banks and non-SLR banks after the treatment. If the SLR banks actively reached for yield by purchasing higher yield securities, we would expect $\delta > 0$.

Figure 5 plot the mean (unweighted) yield on new securities purchases by SLR banks and non-SLR banks. ¹⁹ The pre-treatment yields are largely parallel until 2014:Q3, when they begin to diverge.

Table 6 reports the DD estimates. Column 1 reports weighted least squares (WLS) estimates, where the weights equal the value of the new security purchased divided by the value of all new security purchases by the bank that quarter. Column 2 reports OLS estimates treating all purchased securities equally. The estimate in column 1 suggests that the DD in yields for the newly purchased securities is positive, significant (5 percent level) and large, around 38 basis points.

Columns 3 and 4 reports estimate when we exclude two "custodian" banks: Bank of New York Mellon and State Street Corporation. While the custodial business model for these banks may make the leverage limit especially tight (since they typically hold relatively more safe assets), investing in certain riskier securities may be outside their mission or may be more

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¹⁹ Trends in median yields are similar.

proscribed (relative to other big banks) by regulators. In fact, the estimates are larger (50 to 60 basis point) and more significant with these banks excluded, implying more active reaching for yield by the banks more at liberty to do so.

The results for SLR tighter and SLR looser banks in panel B are more mixed. The point estimates are larger for SLR tighter banks most cases, as expected, though they not always significant. When we exclude custodian banks (columns 3 and 4), however, only the SLR looser is significant using WLS, contrary to expectations. The OLS estimate of SLR tighter is larger and significant as expected.

d. Higher Overall Risk?

If covered banks are risk-shifting around the leverage rule, has it made them riskier overall? Theoretically, the answer is ambiguous; while leverage-constrained banks may have tilted toward riskier assets due to the SLR, those banks were also less levered than they would have been without the constraint. In the Acosta Smith et al (2017) model, the latter effect—greater loss absorbency—dominates the risk shifting, so the leverage limit makes banks more stable on net.

This section tests how overall measures of bank risk changed at SLR banks. We study three book measures: ROA, ROA volatility, and Z scores; and four market measures: equity volatility, CDS spreads, implied volatility, and put option delta. The market measures and Z scores (an inverse risk measure) should reflect both asset risk and leverage.

Summary statistics are reported in Table 7 along with details on the calculation.²⁰ The unconditional DD are mixed, with significantly positive estimates for ROA, ROA volatility, and CDS spreads and significantly negative estimates for Z scores and equity volatility.

Table 8, panel A reports the conditional DD estimates. The only significant estimate is for equity volatility, but the negative sign suggests lower risk for SLR banks. Panel B reports estimates where we divide SLR banks by tightness of the leverage limit. Equity volatility actually declined more for the more constrained SLR banks (-0.30 and significant vs. -0.03 and

Banks' 5-year CDS spreads are from Markit.

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²⁰ The book risk measures (ROA, ROA volatility, and Z score) are calculated from banks' Y-9C reports. Equity volatility equals the quarterly standard deviation of the log of daily difference in stock price. Implied volatilities (on a 50% out-of-the-money option that expires in 1 year) for each entity-quarter are pulled from Bloomberg; together with Treasury rates data from FRED these are also used to calculate implied deltas using the Black-Scholes formula.

insignificant). The only other significant result in Panel B is the positive estimate for CDS for less constrained SLR banks. Since the corresponding estimate for more constrained SLR banks is less than half as large (0.48 vs. 0.2) and insignificant, that single result does not suggest higher overall risk.

If SLR banks, particularly those more bound by the leverage rule, shifted to riskier assets in a reach-for-yield, why were they not riskier overall? Presumably this is because the rule also required more capital. Figure 6 plots mean leverage capital for non-SLR banks and (separately) SLR banks that were more or less bound by the leverage limit. The trends for all three were essentially parallel before SLR, or more precisely, 2015:Q1, when SLR banks were required to publicly disclose their SLR. At that time, the more tightly constrained SLR banks began substantially increasing their ratio both in relative and absolute terms.

V. Conclusion

Leverage rules are supposed to limit bank risk but could, perversely, increase it if banks sidestep the rule by shifting to riskier assets. Though as old as leverage limits themselves, that conjecture is relatively untested as previous U.S. leverage rules applied universally to all banks and were rarely binding. The SLR requirement, by contrast, applies only to the very largest U.S. banks and is tightly binding for some, affording an opportunity to revisit that question.

Our evidence is consistent with the risk-shifting conjecture but not the perverse consequences. Banks subject to the new rule, particularly those most bound by it, appear to have rebalanced their portfolios toward riskier (self-reported) assets overall. Consistent with that finding and reaching for yield concerns, we also find a substantial rise in yields on securities held by the covered banks and a larger yield increase at the most constrained banks. Overall bank risk measures, book or market, have *not* increased, not even at the most constrained banks, suggesting that the higher capital required under the new rule offsets the shift to riskier assets, or vice-versa.

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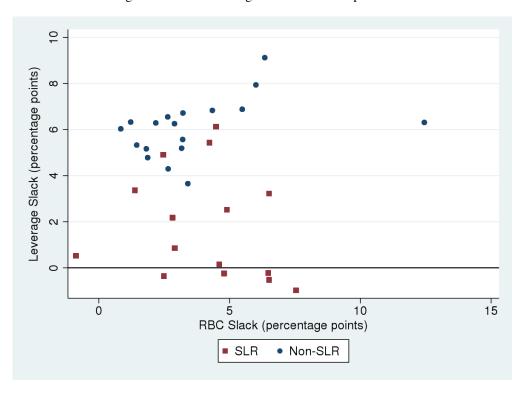
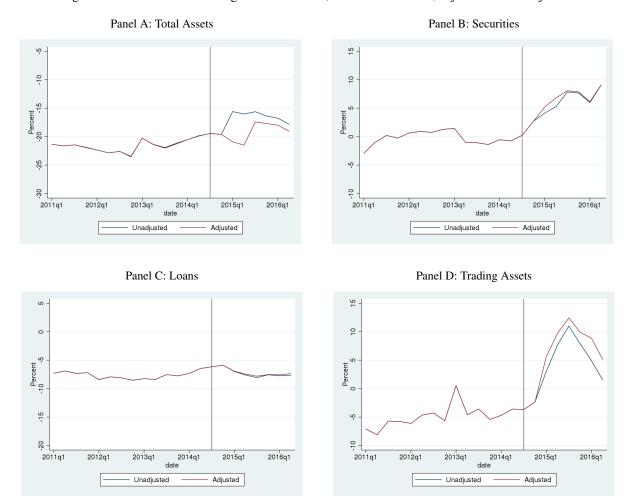


Figure 2: Slack in leverage and risk-based capital ratios.

Note: Leverage slack = leverage ratio at 2013:Q4 - required minimum. For non-SLR banks, leverage requirement is tier 1 capital/total consolidated assets \geq 4 percent. For SLR banks, requirement is tier 1 capital/total leverage exposures \geq 3 percent. https://www.occ.gov/newsissuances/ news-releases/2013/2013-110a.pdf

Figure 3: Difference in risk-weighted asset shares, SLR less non-SLR, adjusted and un-adjusted.



Note: Adjustment is made for the 2015:Q1 implementation of Basel III risk-weighting categories. In order to create a consistent time series, we adjust risk-weights post-2015:Q1 to be consistent with those before. See Appendix for details. Vertical line indicates the treatment date (2014:Q3).

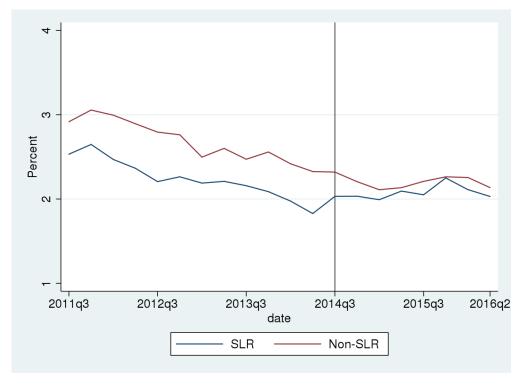


Figure 4: Mean average securities yields for SLR and non-SLR banks.

Note: The weighted average yield is calculated in two steps. First, for each bank-quarter, within each asset class we calculate the value-weighted average yield, i.e. weighted by the market value of each security in the class. We then weight the weighted average yield for each class by the value share the asset class represents within the bank-quarter in the overall Y-14 data. See Appendix for details. Vertical line indicates the treatment date (2014:Q3).

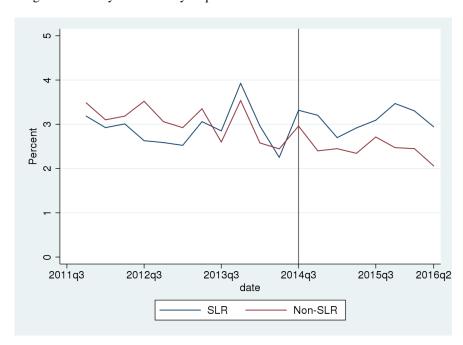


Figure 5: Mean yield on newly acquired securities for SLR and non-SLR banks.

Note: A security is denoted as newly acquired upon its first observation in the data. Vertical line indicates the treatment date (2014:Q3).

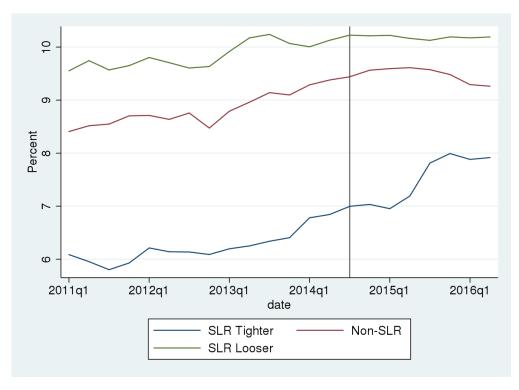


Figure 6: Mean Leverage Ratio for SLR Tighter, SLR Looser, and non-SLR Banks

Note: Leverage ratio = tier 1 capital divided by total assets for SLR and non-SLR banks. We decompose SLR banks into those who are more or less bound by the SLR. Vertical line indicates the treatment date (2014:Q3).

Table 1: Means (Standard Deviations) [Observations] of Risk-Weighted Asset Shares, By Subsample

		Panel A: Unadjusted				
	SLR I	Banks	Non-SL	R Banks	Diff-in-diff (unconditional)	
	(1) Post	(2) Pre	(3) Post	(4) Pre	$ \begin{array}{c} (5) \\ [(1) - (2)] - [(3) - (4)] \end{array} $	
Total Assets	63.36 (13.85) [120]	59.12 (16.00) [210]	80.50 (12.17) [144]	80.39 (9.57) [252]	4.14***	
Securities	21.34 (8.38) [120]	23.79 (10.78) [210]	15.92 (5.44) [144]	23.84 (11.79) [252]	5.46***	
Trading Assets	16.85 (28.01) [120]	9.38 (24.90) [210]	12.98 (15.30) [128]	11.34 (14.03) [238]	5.31***	
Loans	83.24 (9.38) [120]	80.91 (11.60) [210]	90.45 (4.94) [144]	88.54 (5.37) [252]	0.42	

Panel	B:	Adi	usted

		Tallet B. Hajustea				
	SLR Banks		Non-SLR Banks		Diff-in-diff (unconditional)	
	(1)	(2) Pre	(3) Post	(4) Pre	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	
	Post	rie	rost	rie	[(1) - (2)] - [(3) - (4)]	
Total Assets	58.19	59.08	77.38	80.39	2.12**	
	(14.15)	(16.05)	(13.02)	(9.57)		
	[120]	[210]	[144]	[252]		
Securities	21.42	23.79	15.60	23.83	5.86***	
	(8.58)	(10.77)	(5.62)	(11.79)		
	[120]	[210]	[144]	[252]		
Trading Assets	16.49	9.38	10.70	11.34	7.22***	
	(28.12)	(24.90)	(12.40)	(14.03)		
	[120]	[210]	[128]	[238]		
Loans	82.57	80.91	89.67	88.54	0.54	
	(9.52)	(11.62)	(4.92)	(5.37)		
	[120]	[210]	[144]	[252]		

Note: Risk-weighted share equals risk-weighted x per total x for each asset class (x) indicated. All shares are winsorized at the 1% and 99% levels. The unadjusted asset shares data are gathered from the Y-9C. Adjusted shares reflect authors' adjustments to minimize the effect of a reporting change beginning in 2015:Q1. Pre indicates 2011:Q1 to 2014:Q2, and Post indicates 2014:Q3 to 2016:Q2. *, **, and *** in column 5 indicate statistical significance at the 10%, 5%, and 1% level, respectively, per t-test.

Table 2: Difference-in-Differences in Risk-weighted Asset Shares

	Panel A: Unadjusted				
	(1)	(2)	(3)	(4)	
	Total Assets	Securities	Trading Assets	Loans	
SLR Bank × Post	3.64**	5.45**	5.56	0.37	
	(1.43)	(2.59)	(4.47)	(1.72)	
Observations	684	684	634	684	
R-Squared	0.97	0.66	0.78	0.91	

	Panel B: Adjusted				
	(1)	(2)	(3)	(4)	
	Total Assets	Securities	Trading Assets	Loans	
SLR Bank × Post	1.92	5.88**	7.51*	0.46	
	(3.12)	(2.63)	(4.18)	(1.64)	
Observations	684	684	634	684	
R-Squared	0.83	0.67	0.80	0.92	

Note: Reported are OLS estimates of δ from model (1) using panel data on all banks with at least \$50 billion in assets over 2011:Q1 to 2016:Q2. Robust, clustered (by bank) standard errors are in parenthesis. *Post* indicates post-SLR treatment (2014:Q3); *SLR* indicates treated banks (firms with at least \$250 billion in assets or over \$10 billion in foreign exposures). The regression model includes the risk-based capital ratio and a proxy for the liquidity coverage rule exposure (both lagged one quarter) and fixed effects (bank and year-quarter). The unadjusted asset shares data are gathered from the Y-9C. Adjusted shares reflect authors' adjustments to minimize the effect of a reporting change beginning in 2015:Q1. All dependent variables are winsorized at the 1% and 99% levels.*, **, and *** denote statistical significance at the 10%, 5%, and 1% level, respectively.

Table 3: Difference-in-differences in Risk-Weighted Asset Shares by SLR "Tightness"

	Pa	Panel A: Unadjusted			
	(1) Total Assets	(2) Securities	(3) Trading Assets	(4) Loans	
SLR Tighter × Post	5.96***	5.94*	2.87	0.48	
	(1.74)	(2.98)	(3.29)	(3.56)	
SLR Looser \times Post	1.63	5.03*	7.76	0.27	
	(1.59)	(2.59)	(6.98)	(0.89)	
Observations	684	684	634	684	
R-Squared	0.97	0.66	0.78	0.91	

	Panel B: Adjusted				
	(1) Total Assets	(2) Securities	(3) Trading Assets	(4) Loans	
SLR Tighter × Post	6.19** (2.49)	5.92* (3.02)	4.72* (2.70)	0.62 (3.38)	
SLR Looser \times Post	-1.80 (4.22)	5.84** (2.66)	9.79 (6.83)	0.32 (0.90)	
Observations R-Squared	684 0.83	684 0.67	634 0.80	684 0.92	

Note: Reported are OLS estimates of β_1 and β_2 from model (2) using panel data on all banks with at least \$50 billion in assets over 2011:Q1 to 2016:Q2. Robust, clustered (by bank) standard errors are in parenthesis. *Post* indicates post-SLR treatment (2014:Q3); *SLR* indicates treated banks (firms with at least \$250 billion in assets or over \$10 billion in foreign exposures). *SLR Tighter* denotes that the SLR was below median in 2013:Q4; *SLR Looser* is above median. The regression model includes the risk-based capital ratio and a proxy for the liquidity coverage rule exposure (both lagged one quarter) and fixed effects (bank and year-quarter). The unadjusted asset shares data are gathered from the Y-9C. Adjusted shares reflect authors' adjustments to minimize the effect of a reporting change beginning in 2015:Q1. All dependent variables are winsorized at the 1% and 99% levels.*, **, and *** denote statistical significance at the 10%, 5%, and 1% level, respectively.

Table 4: Means (Standard Deviations) of Average Security Yield, By Subsample

	SLR Banks		Non-SLR Banks		Diff-in-diff (unconditional)
	(1) Post	(2) Pre	(3) Post	(4) Pre	(5) [(1) - (2)] - [(3) - (4)]
Average Securities Yield	2.08 (0.79)	2.24 (0.84)	2.20 (0.46)	2.61 (0.61)	0.28***
Observations	108	145	140	114	507

Note: The weighted average yield is calculated by first obtaining (for each asset class) the average yield weighted by the amount of the bank's holding in that class. Then the overall average is obtained by weighting by the amount of holdings in each asset class in the original Y-14Q. Yield is winsorized at the 1% and 99% levels. *Pre* indicates 2011:Q1 to 2014:Q2, and *Post* indicates 2014:Q3 to 2016:Q2. *, **, and *** in column 5 indicate statistical significance at the 10%, 5%, and 1% level, respectively, per t-test.

Table 5: Difference-in-differences in Average Yield of the Securities Portfolio

	(1)	(2)
	Average Securities Yield	Average Securities Yield
SLR Bank × Post	0.36**	
	(0.16)	
SLR Tighter \times Post		0.49*
		(0.25)
SLR Looser \times Post		0.25
		(0.18)
Observations	469	469
R-Squared	0.85	0.85

Note: Reported in Panel A are OLS estimates of δ from model (1) using panel data on all banks with at least \$50 billion in assets over 2011:Q3 to 2016:Q2. Reported in Panel B are OLS estimates of β_1 and β_2 from model (2) using the same panel. Robust, clustered (by bank) standard errors are in parenthesis. *Post* indicates post-SLR treatment (2014:Q3); *SLR* indicates treated banks (firms with at least \$250 billion in assets or over \$10 billion in foreign exposures). *SLR Tighter* denotes that the SLR was below median in 2013:Q4; *SLR Looser* is above median. The weighted average yield is calculated by first obtaining (for each asset class) the average yield weighted by the amount of the bank's holding in that class. Then the overall average is obtained by weighting by the amount of holdings in each asset class in the original Y-14Q. The regression model includes the risk-based capital ratio and a proxy for the liquidity coverage rule exposure (both lagged one quarter) and fixed effects (bank and year-quarter). All dependent variables are winsorized at the 1% and 99% levels.*, **, and *** denote statistical significance at the 10%, 5%, and 1% level, respectively.

Table 6: Extensive Margin, Purchased Securities

	Panel A: Difference-in-differences in Yield					
	(1)	(2)	(3)	(4)		
	All Banks, WLS	All Banks, OLS	Non-Custody Banks, WLS	Non-Custody Banks, OLS		
Post × SLR Bank	0.38**	0.39	0.50**	0.62**		
	(0.18)	(0.27)	(0.19)	(0.24)		
Observations	312620	313004	269510	269868		
R-Squared	0.20	0.14	0.20	0.13		

Panel B: Difference-	-in-differences in	Yield, By	v SLR "Tightness"

	Tunot 2. 2 interested in universities in Tierre, 2, 5211 Tightwess						
	(1)	(2)	(3)	(4)			
	All Banks, WLS	All Banks, OLS	Non-Custody Banks, WLS	Non-Custody Banks, OLS			
SLR Tighter × Post	0.33	0.47	0.61	1.01**			
	(0.28)	(0.38)	(0.37)	(0.38)			
SLR Looser \times Post	0.42**	0.28	0.44**	0.28			
	(0.17)	(0.21)	(0.17)	(0.19)			
Observations	312620	313004	269510	269868			
R-Squared	0.20	0.14	0.20	0.14			

Note: Reported are OLS estimates of of δ from model (3) using panel data on all banks with at least \$50 billion in assets over 2011:Q3 to 2016:Q2. Robust, clustered (by bank) standard errors are in parenthesis. *Post* indicates post-SLR treatment (2014:Q3); *SLR* indicates treated banks (firms with at least \$250 billion in assets or over \$10 billion in foreign exposures). *SLR Tighter* denotes that the SLR was below median in 2013:Q4; *SLR Looser* is above median. The regression model includes the risk-based capital ratio and a proxy for the liquidity coverage rule exposure (both lagged one quarter) and fixed effects (bank and year-quarter). All dependent variables are winsorized at the 1% and 99% levels.*, **, and *** denote statistical significance at the 10%, 5%, and 1% level, respectively.

Table 7: Means (Standard Deviations) [Observations] of Overall Risk Measures, By Subsample

	SLR Banks		Non-SLR Banks		Diff-in-diff (unconditional)	
	(1) Post	(2) Pre	(3) Post	(4) Pre	(5) [(1) - (2)] - [(3) - (4)]	
ROA	1.02	0.95	0.82	0.83	0.07*	
	(0.79)	(0.79)	(0.56)	(0.76)		
	[120]	[210]	[144]	[252]		
ROA Volatility	0.20	0.28	0.27	0.55	0.20***	
	(0.16)	(0.21)	(0.61)	(0.89)		
	[120]	[210]	[144]	[252]		
Z Score	139.76	94.90	160.53	92.11	-23.55***	
	(55.94)	(40.94)	(53.42)	(30.28)		
	[120]	[210]	[144]	[252]		
Equity volatility	1.49	1.68	1.62	1.65	-0.17***	
	(0.16)	(0.34)	(0.15)	(0.31)		
	[104]	[182]	[96]	[154]		
5-year CDS spread	0.71	1.22	1.11	1.83	0.12***	
	(0.18)	(0.47)	(0.57)	(0.84)		
	[104]	[196]	[88]	[154]		
Implied vol.	38.54	45.87	38.08	45.47	-0.60	
_	(2.54)	(4.00)	(3.22)	(7.59)		
	[104]	[182]	[96]	[154]		
Put option Delta	-0.02	-0.07	-0.02	-0.07	-0.00	
•	(0.01)	(0.01)	(0.01)	(0.02)		
	[104]	[182]	[96]	[154]		

Note: ROA volatility is calculated as the rolling 4-quarter standard deviation, excluding periods where treatment and control dates overlap. Equity volatility is calculated as the quarterly standard deviation of the log difference in daily stock price for public firms. Implied volatilites are on a 50% out-of-the-money option that expires in 1 year. Implied deltas are calculated using the Black-Scholes formula. All risk measures are winsorized at the 1% and 99% levels. *Pre* indicates 2011:Q1 to 2014:Q2, and *Post* indicates 2014:Q3 to 2016:Q2. *, **, and *** in column 5 indicate statistical significance at the 10%, 5%, and 1% level, respectively, per t-test.

Table 8: Difference-in-Differences in Overall Risk Measures

	Panel A: Difference-in-differences							
	Book Risk			Market Risk				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	
	ROA	ROA Volatility	Zscore	Equity Volatility	5-year CDS Spread	Implied Vol.	Put Option Delta	
SLR Bank × Post	0.08	0.23	-24.14	-0.15*	0.35	0.85	-0.00	
	(0.15)	(0.22)	(34.01)	(0.09)	(0.21)	(1.57)	(0.00)	
Observations	684	550	550	500	492	487	487	
R-Squared	0.44	0.28	0.46	0.89	0.76	0.88	0.95	

Panel B: Difference-in-differences by SLR "Tightness"

	Book Risk			Market Risk			
	(1) ROA	(2) ROA Volatility	(3) Zscore	(4) Equity Volatility	(5) 5-year CDS Spread	(6) Implied Vol.	(7) Put Option Delta
SLR Tighter × Post	0.10	0.26	-45.86	-0.30***	0.20	-0.16	-0.00
SLR Looser × Post	(0.16) 0.06	(0.22) 0.21	(29.83) -5.06	(0.09) -0.03	(0.25) 0.48**	(1.65) 1.63	(0.00) -0.01*
	(0.16)	(0.23)	(43.55)	(0.09)	(0.22)	(1.66)	(0.00)
Observations	684	550	550	500	492	487	487
R-Squared	0.44	0.28	0.46	0.89	0.76	0.88	0.95

Note: Reported in Panel A are OLS estimates of δ from model (1) using panel data on all banks with at least \$50 billion in assets over 2011:Q1 to 2016:Q2. Reported in Panel B are OLS estimates of β_1 and β_2 from model (2) using the same panel. Robust, clustered (by bank) standard errors are in parenthesis. *Post* indicates post-SLR treatment (2014:Q3); *SLR* indicates treated banks (firms with at least \$250 billion in assets or over \$10 billion in foreign exposures). *SLR Tighter* denotes that the SLR was below median in 2013:Q4; *SLR Looser* is above median. The regression model includes the risk-based capital ratio and a proxy for the liquidity coverage rule exposure (both lagged one quarter) and fixed effects (bank and year-quarter). All dependent variables are winsorized at the 1% and 99% levels.*, **, and *** denote statistical significance at the 10%, 5%, and 1% level, respectively.

Appendix

Adjusted risk-weights

Starting in 2015:Q1, banks commenced reporting regulatory capital based on the new capital rule adopted by the federal banking agencies. The changes include components of regulatory capital (Schedule HC-R, Part I) and reporting of risk-weighted assets (Schedule HC-R, Part II). The rule also revised certain risk-weights, so risk-weighted assets under the new rule are not necessarily comparable to those under the old rule.

To enable such comparisons, we adjusted the data by (in brief) (1) applying previous risk-weights to assets using the revised risk-weights; (2) excluding off-balance sheet exposures that weren't previously included in total risk-weighted assets. Note that these adjustments are approximations, as for certain exposures banks simply report "risk-weighted asset amount" for which we can't infer the underlying risk-weights applied.

Below is a detailed summary of the adjustments:

- (1) Replace revised risk-weights with the closest risk-weight used prior to 2015Q1.
 - Columns D (2%) and E (4%) are merged into column C (0%)
 - Column J (150%) is merged into column I (100%)
 - Columns L,M,N (300, 400, 600%) are assumed to be with 200% risk weights
- (2) Exclude off-balance sheet exposures that weren't previously included in total risk-weighted assets
 - Unused commitment with maturity less than 1 year (item 18a)
 - Unsettled transactions (item 22)

 $^{^{1}}$ See https://www.ffiec.gov/pdf/FFIEC_forms/FFIEC031_FFIEC041_20150225_Presentation.pdf for the summary of changes

Constructing Average Securities Yield

Calculating the weighted average yield by combining all asset types may skew the average yield due to differences in asset class composition between our matched sample and the full Y-14 data. To mitigate this concern, we construct the weighted average yield in a two-step process. First, for each bank-quarter, within each asset class we calculate the value-weighted average yield, i.e. weighted by the market value of each security in the class. We then weight the weighted average yield for each class by the value share the asset class represents within the bank-quarter in the overall Y-14 data.

For example, suppose a bank's securities portfolio in a given quarter, as reported in the Y-14, is composed of 10% corporate bonds and 90% Treasuries by value. Suppose the value-weighted average yield of corporate bonds in our sample is 4% and that of Treasuries is 2%. Then our final yield would be:

$$(0.1 \times 4\%) + (0.9 \times 2\%) = 2.2\%.$$

The following notation formalizes this calculation. For each security s belonging to asset class c held by bank i in quarter t, we first calculate the average yield of the asset class c held by the bank:

$$\sum_{s \text{ in class } c} r_{isct} \times v_{isct} = w_{ict}$$

where r is the security's yield and v is the ratio of the security's market value to the total market value of all securities in asset class c held by bank i. Thus, w_{ict} is the value-weighted average yield for each asset class c. Then, we calculate the average across the asset classes:

$$\sum_{c} w_{ict} \times k_{ict} = y_{it}$$

where k is the share of the bank's portfolio belonging to class c as reported in the Y-14. Thus, y_{it} is our final bank-quarter *average securities yield* observation.