



Article

Dynamics of Funding Liquidity and Risk-Taking: Evidence from Commercial Banks

Faisal Abbas ¹, Shoaib Ali ², Imran Yousaf ² and Wing-Keung Wong ^{3,4,5,*}

- ¹ UCP Business School, University of Central Punjab, Lahore 54000, Pakistan; faisalabbaspcc@gmail.com
² Air University School of Management, Air University, Islamabad 44000, Pakistan; shoaib0819@yahoo.com (S.A.); imranyousaf.fin@gmail.com (I.Y.)
³ Department of Finance, Fintech & Blockchain Research Center, and Big Data Research Center, Asia University, Taichung 41354, Taiwan
⁴ Department of Medical Research, China Medical University Hospital, Taichung 40402, Taiwan
⁵ Department of Economics and Finance, The Hang Seng University of Hong Kong, Hong Kong 999077, China
* Correspondence: wong@asia.edu.tw; Tel.: +886-4-2332-3456

Abstract: The purpose of this study is to investigate the impact of funding liquidity risk on the banks' risk-taking behavior. To test the hypotheses, we apply the two-step system GMM technique on US commercial banks data from 2002 to 2018. We find that funding liquidity increases the banks' risk-taking of US commercial banks. Furthermore, banks with higher deposits are less likely to face a funding shortage, and bank managers' aggressive risk-taking activity is less likely to be monitored. Our findings infer that increases in bank funding liquidity increase both risk-weighted assets and liquidity creation, and deposit insurance creates a moral risk issue for banks taking excessive risks in response to deposit rises. The relationship between funding liquidity and the banks' risk-taking varies with their capitalization and market conditions; the impact of funding liquidity on risk-taking is pronounced for well-capitalized banks and the Global Financial Crisis 2007. Our tests are robust for the usage of alternate proxy of funding liquidity and by controlling economic conditions. The findings of this study have implications for regulators to develop guidelines for the level of liquidity and risk-taking of commercial banks.

Keywords: funding liquidity; risk-weighted assets; liquidity creations; Z-scores

JEL Classification: G21; G01; G18



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1. Introduction

After the world financial crisis 2007–2008, rapidly growing literature has been investigating numerous essential elements of Basel III¹, such as bank regulations for higher capital requirements (Bitar et al. 2018; Cohen and Scatigna 2016; Deli and Hasan 2017), and its impact on the bank performance (Abbas et al. 2019; Asongu and Odhiambo 2019; Berger and Bouwman 2013; Ding et al. 2017), and bank risk-taking (Balla and Rose 2019; Bougatef and Mgdmi 2016). However, the studies are limited to investigating the impact of funding liquidity on bank risk-taking after the promulgation of Basel III (Dahir et al. 2018; DeYoung et al. 2018; Kim and Sohn 2017). It is important to examine the relationship between funding liquidity risk and banks' risk-taking in the specific context of the US, because funding liquidity risk plays a significant role in all historical financial crisis and it is defined as the possibility that over a specific horizon, the bank becomes unable to settle obligation with immediacy (Drehmann and Nikolaou 2013).

The first stream of research is to examine the effects of funding liquidity on the risk of assets and the overall risk-taking of US commercial banks; the next line examines the actions that well-capitalized, undercapitalized, too-big-to-fail banks conduct to take risks due to liquidity funding. It is unclear if the current focus on bank liquidity conditions laid out in

the recent Basel III rules internationally and in the US Dodd–Frank Act² would render banks less volatile and the whole financial sector more resilient in the future (Khan et al. 2017; Tran 2020). Banks are usually advised to create a liquidity buffer to reduce liquidity losses and protect against minor liquidity shocks (Rokhim and Min 2020). We are interested in addressing the following questions on how funding liquidity influences the risk-taking of US large commercial banks. Is the impact of funding liquidity similar for well-capitalized, undercapitalized, and too-big-to-fail commercial banks?

Theoretically, with higher or stable funding liquidity, especially for deposits associated with higher risk-taking (Acharya and Naqvi 2012), in such conditions, a bank may lower lending standards by reducing interest rates and ignore the scrutiny of credible borrowers to enhance the volume of loans. The losses due to counterparty defaults after a specific time may cause a reduction in the wealth of shareholders. The counter-argument is also very appealing means higher risk-taking may become beneficial for managers and shareholders. It is possible due to the moral hazard theory, where the manager exploits the deposit insurance schemes for shareholder and their benefits. The following studies favor the positive relationship between funding liquidity and risk-taking (Acharya and Naqvi 2012; Khan et al. 2017; Lucchetta 2007; Tran 2020). Corporate finance supports higher risk-taking due to free cash flow, as argued by (Jensen 1986). The negative relationship is also not less appealing, as demonstrated by (Dahir et al. 2018; Drehmann and Nikolaou 2013; Jeitschko and Jeung 2005; Rokhim and Min 2020).

Our debate is advocating the contribution and development of funding liquidity significance. Most of the studies are related to the current decade, particularly from the last five to six years. Our results provide that the commercial banks hold higher liquidity in the post-crisis period than before-crisis. Said findings contradict the results of Khan et al. (2017) for the USA banking industry. We document that the difference between the findings may be due to the following possibilities: The sample is not similar; the period of 2010 to 2018 is entirely different from the period before the 2010 era due to regulatory changes and economic conditions. The next simulation uses a two-step system GMM method, which may provide robust and efficient estimators compared to panel regression estimation. We explain this dissimilarity in such a way that the approach of panel regression cannot integrate the effect of simultaneity and endogeneity, which is possible in the presence of capital ratios, risk ratios, liquidity ratios, and profitability ratios.

In this study, we extend the work of Khan et al. (2017) by using similar assets risk measures risk-weighted assets to total assets, loan loss provision to total loans, and overall risk measure as liquidity creation and Z-scores and funding liquidity measure as deposits to total assets. However, our study deviates from the following aspects: First, the study used a comprehensive data set of commercial banks ranging from 2002 to 2018. Second, our estimates are more reliable than other studies in terms of methodological perspective. Most of the recent studies (Khan et al. 2017; Rokhim and Min 2020) use panel OLS³ with fixed and random effects to investigate the relationship between funding liquidity and risk-taking while ignoring the aspects of endogeneity and simultaneity. After integrating the question of cross-section heterogeneity, endogeneity, and simultaneity, we add to the current studies in terms of sign and significance for their robustness. Third, the study provides a relationship between funding liquidity and risk-taking in a comparative way for more clarity and a deeper understanding by highlighting the results of the during-crisis period, well-capitalized, undercapitalized, and too-big-to-fail banks. The findings of the study enable regulators to redesign the banking regulator framework to better discipline and regulate the perverse incentives of bank managers to take too much risk in the future when bank deposits change.

The rest of the paper is structured as follows: The second section reviews the literature on funding liquidity and risk-taking. The third section provides data collection sources, sample selection, definitions, and measurement of variables used in the analysis. The fourth part explains the empirical findings. The fifth section of this research contains conclusions, limitations, and future recommendations.

2. Literature Review and Hypothesis Development

Many studies, for example, [Rokhim and Min \(2020\)](#), in the literature examine the impact of funding liquidity (ratio of customers' deposits to total assets) on risk. In developing economies, banks prefer to hold higher liquid assets as compared to developed countries. They find that banks with higher liquidity in terms of deposits tend to take lower risks. Similarly, testing the relationship between the availability of liquid assets and bank risk-taking, [Dahir et al. \(2018\)](#) document that there is an inverse relationship between liquidity and risk-taking, inferring that the availability of highly liquid assets causes the risk to decrease.

On the other hand, [Jeitschko and Jeung \(2005\)](#) argue that the relationship between higher capital and risk could be either positive or negative and find that the connection is based on the forces of deposit insurers, bank managers, and shareholders. [Drehmann and Nikolaou \(2013\)](#) find that funding liquidity leads to a decrease in market liquidity. Discovering the impact of funding uncertainty on bank lending, [Ritz and Walther \(2015\)](#) conclude that banks reduce their credit due to the difficulty of funding, which leads to lower profitability. Investigating the funding liquidity shocks impact on lending, [De Haan and van den End \(2013\)](#) reveal that during the period of liquidity shocks, the banks cut their wholesale lending and sell some securities, especially selling equity. Some of the studies, for example, ([Boyd and Runkle 1993](#); [Mercieca et al. 2007](#); [Stiroh 2004](#)), report that size and risk are negatively related, advocating that an increase in the total volume of assets decreases bank risk. In light of the above literature, the study develops the following hypothesis.

Hypothesis 1 (H1a). *The impact of funding liquidity risk on banks' assets risk is negative and significant.*

[Lucchetta \(2007\)](#) indicates that due to an increased liquidity supply through interbank lending, banks take more risk. In line with the positive relationship between funding liquidity and bank risk-taking, [Acharya and Naqvi \(2012\)](#) argue that banks that have excessive deposits to total assets ratio usually face lower liquidity shortage. Due to a more moderate liquidity shortage or higher deposits in the short run, managers take higher risks. In support of this argument, they document that under an excessive deposits ratio, banks use aggressive policies to lend at easy terms, which leads to an increase in bank risk in the end. Analogously, [Khan et al. \(2017\)](#) submit that banks with higher liquidity funding in terms of deposits take on a higher risk, as evidenced by risk-weighted assets, loan loss provisions, liquidity creation, and Z-scores in the USA. In light of empirical findings, they were advocating regulators for new packages for higher liquidity and regulatory capital. In corporate finance other than banking, [Jensen \(1986\)](#) concludes that the availability of free cash flows encourages managers to invest in risky investments. [Balasubramanyan and VanHoose \(2013\)](#) use a dynamic banking method to test the influence of liquidity coverage ratio restrictions on bank lagged deposits based on balance sheet structure and find ambiguous results. [Imbierowicz and Rauch \(2014\)](#) contribute to providing that liquidity risk and credit risk had no contemporaneous association. However, both are significant factors in bankruptcy. In light of the mixed literature and the present focus of scholars ([Khan et al. 2017](#); [Rokhim and Min 2020](#); [Tran 2020](#)), we revisit the issue by testing the following hypothesis:

Hypothesis 1 (H1b). *The impact of funding liquidity risk on banks' assets risk is positive and significant.*

In the context of a bank, size based on capitalization, the literature is mixed; for example, [Bertay et al. \(2013\)](#) find an insignificant relationship between size and Z-scores. [Abbas and Ali \(2020\)](#) recently explore that well-capitalized banks behave differently than undercapitalized banks. The study provides evidence that high-liquidity banks behave

differently than low-liquidity banks. [Chen et al. \(2018\)](#) explore that external funding and macroeconomic indicators are the significant factor of liquidity funding for commercial banks by using the data for 12 developed economies of the world. [Adusei \(2015\)](#) analyzes the rural banks of Ghana. The study concludes that the impact of liquidity risk varies with bank size; however, the relationship between liquidity risk and bank stability is positive and significant in Ghana. [Acharya and Mora \(2015\)](#) investigate the role of liquidity providers to influence the lending of commercial banks. The study demonstrates that a higher ratio of deposit withdrawal influences the lending risk of financial institutions. [Abdul-Rahman et al. \(2018\)](#) examine the impact of liquidity to influence Islamic and conventional banks. The study opines that liquidity risk has equal importance for Islamic and conventional banks in the long run. ([Abbas et al. 2020a, 2021](#)) conclude that the behavior of banks is heterogeneous across well-capitalized, adequately capitalized, undercapitalized, high- and low-liquidity banks in the USA. In light of the above-discussed literature, the study develops the following hypothesis:

Hypothesis 2. *The impact of funding liquidity to influence the risk-taking across well-capitalized, undercapitalized, and too-big-to-fail banks is similar.*

[Wagner \(2007\)](#) develops a theoretical model to explain the phenomenon that higher liquidity reduces stability in a crisis period. Following the results, [Tran \(2020\)](#) analyzes the impact of funding liquidity on bank lending using the sample of the USA banks. The study submits that the banks rely on deposits that have lower lending growth. The study concludes that there was lower loan growth in the pre-crisis period and found no relationship between loan growth and funding liquidity in the post-crisis period. [Ivashina and Scharfstein \(2010\)](#) opine that the higher availability of funding in terms of deposits in crisis causes an increase in lending. [Abbas et al. \(2021\)](#) find that banks behave differently in different economic conditions concerning risk-taking and holding of capital. [Acharya and Mora \(2015\)](#) explore that the availability of funds to provide credits during the financial crisis is only possible because of government support. [Munteanu \(2012\)](#) examines the commercial banks of Romania and concludes that liquidity remains a critical indicator that influences banks during the financial crisis 2007–2008. In light of the above-discussed literature, we developed the following hypothesis:

Hypothesis 3. *The impact of liquidity funding to influence the risk-taking of banks remains similar across pro, amid, and post-crisis eras.*

3. Data and Methodology

3.1. Data

The study uses unique and recent data for 945 U.S.-chartered commercial banks that have consolidated assets of \$300 million or above, as of 31 December 2018⁴. The data are retrieved over the period ranging from 2002 to 2018, which also covers the world financial crisis 2007–2008 period⁵. All banks have to submit form Y-9C to the Federal Reserve Bank quarterly since 1986. The reporting information contains details in the form of an income statement and balance sheet for regulatory purposes. There are several reasons to study commercial banks in current conditions: first, due to the importance and significant share of commercial banks in the USA banking industry; second, the commercial banks remain an excellent laboratory for regulators to test the results of their recommendations; third, due to the dominating role in the expansion and development of the economy; and fourth, commercial banks always remain the base for future decision-making in the USA.

To deal with outliers, we winsorized bank-specific variables at 1% and 99% by following the recent studies by ([Berger and Bouwman 2013](#); [Khan et al. 2017](#)). The data for gross domestic product and inflation rate are collected from the world development indicators database⁶.

3.2. Empirical Model

We use a dynamic panel model to explore the influence of funding liquidity on the assets risk and overall risk-taking behavior of commercials in the USA. There are several reasons to apply GMM: (1) GMM controls the endogeneity of the lagged reliant variable, (2) GMM incorporates the issue of first-order autocorrelations, (3) GMM controls the measurement errors problem, (4) GMM reduces omitted bias issue, (5) GMM controls the unobserved heterogeneity problem in panels, and (6) GMM is the best choice in case of more cross-sections than many periods. [Arellano and Bond \(1991\)](#) develop the GMM method to make the estimators consistent in the presence of a lagged dependent variable. Various studies have applied GMM in the field of banking, including ([Lee and Hsieh 2013](#); [Tran 2020](#)). The standard form of a dynamic regression model is the basic equation for the following linear regression model with endogenous variables:

$$Risk_{i,t} = \alpha_0 + \beta_1 Risk_{i,t-1} + \beta_2 Funding\ liquidity_{i,t} + \beta_3 control\ variables_{i,t} + u_{i,t} \quad (1)$$

where $Risk_{i,t}$ (loan loss provisions, risk-weighted assets, liquidity creation, and Z-score) is the dependent variable; $Funding\ liquidity_{i,t}$ is the independent variable; the sign of “i” indicates cross-sections; and the notation “t” is time. $Risk_{i,t-1}$ represents the lag of risk-taking, and “ $u_{i,t}$ ” is an error term. The control variables include loan ratio, size, capital, profitability, economic growth, and inflation rate.

Model Specifications of Difference and System GMM

Difference GMM simulation of econometrics corrects the bias of endogeneity by transforming the included regressors through differencing them, and this method controls and removes the fixed effects of the model. The first difference transformation has a weakness of subtracts the last year’s observation from the present one, which significantly magnifies the interval in the case of an unbalanced panel. However, [Arellano and Bover \(1995\)](#) and [Blundell and Bond \(1998\)](#) provide the solution to avoid the bias of first differencing to use system GMM instead of difference GMM. This simulation deals with the problem of endogeneity by introducing the use of more instruments that significantly magnify the efficiency of the model. This method has a feature to transform the instruments to make them unconnected with panel fixed effects, which validate the assumption of exogenous regressors. These methods use a system of two equations that include original linear regression and transformed equations at a time to provide consistent parameters. The next feature of the system GMM method is that it uses the average of all future available observations of the series instead of the first differencing method. There is no problem with gaps in the series, and it saves and minimizes the data loss.

The difference GMM model contains the following standard form:

$$\ln Risk_{i,t} = \phi Risk_{i,t-1} + \beta_2 Funding\ liquidity'_{i,t} + (\eta_i + \varepsilon_{i,t}) \quad (2)$$

According to the theory of difference, GMM, the model is transferred to the first difference to remove the fixed effects from panels. The standard form for the above model would be:

$$\Delta \ln Risk_{i,t} = \phi \Delta Risk_{i,t-1} + \beta_2 \Delta Funding\ liquidity'_{i,t} + \Delta \varepsilon_{i,t} \quad (3)$$

The above form of the transferred model equation represents that by transforming the regressors, the fixed effects are removed. No doubt, there is no problem with time, but the issue of endogeneity persists. The modified form of Equation (2) becomes

$$\Delta u_{i,t} = \Delta \eta_i + \Delta \varepsilon_{i,t} \quad (4)$$

$$u_{i,t} - u_{i,t-1} = (\eta_i - \eta_i) + (\varepsilon_{i,t} - \varepsilon_{i,t-1}) = (\varepsilon_{i,t} - \varepsilon_{i,t-1}) \quad (5)$$

The above producer removes the unknown fixed effects from the panel and remains constant according to the specified assumption. In this case, the lagged value of the depen-

dent variable becomes the instrument of the past level; this specification is represented in Equation (2).

The basic model of the system GMM approach is the following form:

$$\ln Risk_{i,t} = \phi Risk_{i,t-1} + \beta_2 Funding\ liquidity'_{i,t} + (\eta_i + \varepsilon_{i,t}) \quad (6)$$

It is assumed that the above specification is a random walk equation and dependent variables are persistent. Due to this reason, the results of difference GMM produce inefficient estimates for parameters and bias particularly in finite samples. This means when a period remains limited, cross-sections contain large numbers. The empirical literature provides that the high bias and poor performance of difference GMM are due to poor instruments. The system GMM uses the one equation in level forms with first differences as instruments, whereas the second equation in the first differenced form with level as instruments (Blundell and Bond 1998). The system GMM approach implicates a greater number of instruments. Still, Monte Carlo evidence recommends that where the period is limited and dependent variable are found to be persistent, the use of system GMM reduces the bias of small samples. There is another aspect of the GMM method; if data contain autocorrelation and heteroscedasticity, a two-step GMM method should be used to create a weighting matrix using residuals from the first phase. It is also argued that in the limited sample, the standard errors are found to be downward biased. In this situation, the solution is recommended by researchers to apply the robust standard error approach developed by Windmeijer (2005), which corrects the sample bias.

4. Results and Discussion

4.1. Descriptive Statistics

Table 1 presents the descriptive statistics for the proxies used in this analysis. The details include mean, standard deviation, and 5th, 25th, 75th, and 95th percentiles. The means value of loan loss provision is 0.0003 with a standard deviation of 0.0002. The average value of risk-weighted assets is 0.723 with a standard deviation to be 0.11, the average value of liquidity creation is 0.204 with a standard deviation to be 0.069, and the deposit ratio average value is 0.8459 with a standard deviation to be 0.135. Table 2 reports correlation estimates between explanatory variables. The statistics confirm that there is no high value of correlation between any pair of explanatory variables, inferring that there is no issue of multicollinearity. The signs of correlation coefficients are in line with the economic theory. The values of the descriptive and correlation matrix are in line with the following studies (Abbas and Ali 2020; Berger and Bouwman 2009).

Table 1. Variable measurement and descriptive statistics.

Variable	Measurements	Mean	Std.Dev.	p ⁵	p ²⁵	p ⁷⁵	p ⁹⁵
Dependent Variables (Risk)							
Loan loss provisions (LLP)	Loan loss provisions/total loans (Ali et al. 2019)	0.0003	0.0002	0.0003	0.0006	0.0039	0.0041
Risk-weighted assets (RRWA)	Risk-weighted assets/total assets	0.723	0.11	0.4972	0.65	0.8052	0.9001
Liquidity Creations	Liquidity Creation = 0.5 (Illiquid Assets) + 0.5 (Liquid Liabilities) – 0.5 (Liquid Assets) – 0.5 (Illiquid Liabilities) – 0.5 (Equity) (Khan et al. 2017)	0.204	0.069	0.0598	0.1582	0.2575	0.3619
Z-Score	Return on assets plus equity to total assets scaled by the standard deviation of return on assets (Khan et al. 2017)	8.172	1.97	2.5664	12.308	38.8	49.2
Independent Variables							
Funding liquidity	Ratio of customer deposits to total assets (Khan et al. 2017)	0.8459	0.135	0.4756	0.6824	0.878	0.9010
Control Variables							
Loan Ratio	Loans/total assets ratio (Abbas and Ali 2020)	0.01	0.122	0.0069	0.0221	0.1095	0.2310
Size	Natural Log of total assets (Ali et al. 2020)	13.554	0.95	12.259	12.854	14.134	0.1820
Capital	Total Equity/total Asset (Yousaf et al. 2019a)	0.102	0.018	0.0785	0.0871	0.1146	0.1661
Return on Assets (ROA)	Net income/total assets (Yousaf et al. 2019b)	0.01	0.005	–0.0009	0.0063	0.0129	0.0799
Economic growth	Annual growth rate of gross domestic product (Abbas et al. 2020b)	2.084	1.038	–0.2916	1.6773	2.8067	2.9010
Inflation rate	Consumer price index (Abbas and Masood 2020; Yousaf and Ali 2020)	1.92	0.665	0.759	1.535	2.065	2.1080

Table 2. Matrix of correlations.

Variables	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Risk	1							
Funding liquidity	0.037 *	1						
Loan Ratio	0.008 *	0.001	1					
Size	0.011 *	0.033 *	−0.002 *	1				
Capital	−0.040 *	0.031 *	0.003 *	0.124 *	1			
ROA	−0.279 *	0.036 *	−0.002 *	−0.003 *	0.090 *	1		
Economic growth	−0.341 *	−0.001 *	−0.010	−0.021 *	0.019 *	0.162 *	1	
Inflation rate	−0.200 *	−0.004 *	−0.014 *	−0.141 *	−0.110	0.166 *	0.460 *	1

This table reports the pair-wise correlations for risk and liquidity measures and other selected variables over the period from 2002 to 2019. * represents the significance at 5%.

4.2. Funding Liquidity and Risk-Taking Overall Sample

Table 3 contains the results to investigate the role of funding liquidity to influence the risk-weighted assets, loan loss provisions, liquidity creations, and Z-Score of full sample banks. Liquidity creation and Z-Score represent the overall riskiness of commercial banks. The results explore that the relationship between deposits to total assets ratio and Z-score is statistically significant and negative. The relationship between funding liquidity and liquidity creation is statistically significant and positive. Based on the results, an increase in deposits leads to a rise in the overall risk-taking of commercial banks, inferring that an increase in stable deposits reduces the problem of liquidity, which may encourage bank managers to create more credits. Our predictions are in line with the findings of (Acharya and Naqvi 2012; Khan et al. 2017). The risk-weighted assets to total assets and provision for loan losses to total loans represent the asset risk. The outcomes provide that there is a positive and statistically significant relationship between both measures of assets risk and deposits to total assets ratio. A 1% increase in deposit ratio cause to increase of 0.02% in loan loss provisions in the short run; other factors held constant. The results are consistent with Khan et al. (2017) in the case of risk-weighted assets to have a positive but insignificant relationship between liquidity funding and loan loss provisions. The model also includes bank-specific control variables. The relationship between risk-taking and bank capital ratio is positive (negative) when measuring as risk-weighted assets (liquidity creation). The profitability and risk have consisted of portfolio theory, which means the higher the risk, the higher the return. The negative sign of size represents the economies of scale or better utilization and management of assets consistent with (Khan et al. 2017).

Table 3. Impact of funding liquidity ratio on risk.

VARIABLES	Bank’s Asset Risk		Bank’s Overall Risk	
	(1) RRWA	(2) LLP	(3) LC	(4) Z-SCORE
Lagged risk	0.883 *** (0.020)	0.835 *** (0.015)	0.867 *** (0.024)	0.918 * (0.018)
Funding liquidity	0.001 *** (0.000)	0.002 *** (0.000)	0.002 *** (0.000)	0.003 *** (0.001)
Loan Ratio	0.001 *** (0.000)	−0.002 (0.000)	0.002 ** (0.000)	−0.297 *** (0.219)
Size	−0.002 (0.000)	−0.002 *** (0.000)	−0.001 ** (0.000)	−0.178 ** (0.033)

Table 3. Cont.

VARIABLES	Bank's Asset Risk		Bank's Overall Risk	
	(1) RRWA	(2) LLP	(3) LC	(4) Z-SCORE
Capital	0.207 *** (0.033)	−0.001 (0.001)	−0.350 *** (0.034)	0.550 *** (0.009)
ROA	0.717 *** (0.101)	−0.0509 *** (0.004)	−0.183 *** (0.065)	0.126 *** (0.012)
Constant	0.072 *** (0.013)	0.002 *** (0.000)	0.076 *** (0.009)	−4.019 * (0.016)
Observations	15,071	15,007	14,991	14,917
AR (1)	0.000	0.000	0.000	0.000
AR (2)	0.114	0.442	0.987	0.730
Hansen value	0.678	0.090	0.151	0.430

This table used two-step system GMM method to measure the impact of funding liquidity on banks' assets risk (risk-weighted assets ratio and loan loss provision ratio) and overall risk (liquidity creation and Z-score). Robust standard errors are reported in parentheses. ***, **, * represent statistical significance at 1%, 5%, and 10% respectively.

4.3. The Impacts of Funding Liquidity on Risk-Taking across Banks Capitalization

Columns 1, 2, 3, and 4 of Table 4 report the results for well-capitalized banks. Here, we split the sample based on risk-based capital ratios, and the banks in which risk-based capital ratios are equal to or higher than 10% classified as well-capitalized. The outcomes reveal that the increase in funding liquidity as measured by the deposits to total assets ratio causes an increase in the riskiness of well-capitalized banks as evidenced by risk-weighted assets, loan loss provisions, liquidity creation, and Z-scores in the short run when other factors hold similar. The findings are consistent with (Acharya and Naqvi 2012; Khan et al. 2017). The influence of funding liquidity of well-capitalized banks on risk-taking is economically significant with a 1% increase in liquidity funding ratio increased risk-weighted assets, loan loss provisions, and liquidity creations by 9.4%, 0.1%, and 4.5%, respectively, whereas it decreased by Z-scores 4.0%. The findings hold after incorporating the bank size, profitability, loan growth, and equity ratios as provided by loan loss provision, liquidity creation, and Z-scores similar results.

Columns 5, 6, 7, and 8 of Table 4 provide findings for undercapitalized banks' liquidity funding and risk-taking behavior. The outcomes explore that an increase in funding liquidity in terms of deposits causes an increase in the assets and overall risk of undercapitalized banks, as evidenced by risk-weighted assets, loan loss provisions, liquidity creation, and Z-scores. The results document that the impact of the increase in the deposits is positive and statistically significant on risk-weighted assets, loan loss provision, and liquidity creations and Z-scores. The findings are in line with the studies of (Acharya and Naqvi 2012; Khan et al. 2017). The impact of the funding liquidity ratio of undercapitalized banks on risk-taking is economically significant as a 1% increase in liquidity funding increased risk-weighted assets, loan loss provisions, and liquidity creations by 2.38%, 0.028%, and 0.2%, respectively. It is observed that the behavior of an increase in riskiness is higher in the case of undercapitalized banks than well-capitalized banks. Theoretically, the conservative behavior of well-capitalized banks to take lower risk is due to the stricter regulations and vigilant monitoring of significant important financial intuitions. The sign of coefficient remains consistent even after controlling for equity to total assets, loan growth, natural log of total assets, and profitability.

Table 4. Funding liquidity ratio and risk varies across banks capitalization.

VARIABLES	Well-Capitalized				Undercapitalized				Too Big to Fail			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
	TR	LLR	LC	Z-SCORE	TR	LLR	LC	Z-SCORE	TR	LLR	LC	Z-SCORE
Lagged risk	0.789 *** (0.069)	0.851 *** (0.047)	0.894 *** (0.068)	0.111 *** (0.086)	0.720 *** (0.037)	0.962 *** (0.017)	0.503 *** (0.055)	0.770 *** (0.026)	0.884 *** (0.020)	0.836 *** (0.015)	0.868 *** (0.024)	0.099 *** (0.000)
Funding liquidity	0.094 *** (0.025)	0.002 *** (0.000)	0.045 *** (0.013)	−0.040 ** (0.020)	0.024 *** (0.022)	0.003 *** (0.000)	0.020 *** (0.020)	0.002 *** (0.000)	0.001 *** (0.000)	0.002 *** (0.000)	0.005 *** (0.000)	−0.821 *** (0.000)
Funding liquidity *TB									−0.006 *** (0.002)	−0.002 *** (0.001)	−0.002 ** (0.001)	−0.012 ** (0.000)
Loan Ratio	0.069 (0.148)	0.018 *** (0.006)	0.076 (0.095)	−1.985 (1.479)	0.001 (0.000)	−0.006 *** (0.000)	0.001 (0.000)	−0.498 *** (0.345)	0.001 ** (0.000)	−0.003 (0.000)	0.001 ** (0.000)	−0.412 (0.047)
Size	−0.002 (0.002)	−0.008 (0.005)	−0.004 *** (0.001)	−0.429 *** (0.162)	−0.001 (0.000)	−0.001 *** (0.000)	−0.002 *** (0.000)	−0.303 *** (0.041)	−0.001 ** (0.000)	−0.009 *** (0.000)	−0.008 ** (0.000)	0.574 ** (0.000)
Capital	0.239 ** (0.120)	0.004 (0.003)	−0.308 *** (0.096)	0.304 *** (0.012)	0.089 ** (0.037)	−0.007 *** (0.001)	−0.729 *** (0.055)	0.851 *** (0.284)	0.206 *** (0.033)	−0.001 (0.001)	−0.349 *** (0.034)	0.601 *** (0.004)
ROA	0.231 (0.265)	−0.020 ** (0.009)	−0.288 * (0.174)	0.121 *** (0.150)	0.583 *** (0.117)	−0.064 *** (0.005)	−0.538 *** (0.085)	0.148 *** (0.741)	0.710 *** (0.101)	−0.050 *** (0.004)	−0.183 *** (0.065)	0.221 *** (0.040)
Constant	0.072 * (0.036)	0.001 (0.001)	0.070 *** (0.025)	−0.904 ** (1.424)	0.024 * (0.013)	0.001 (0.110)	0.055 *** (0.010)	3.823 *** (0.741)	0.076 *** (0.013)	0.002 *** (0.901)	0.077 *** (0.009)	−2.790 ** (2.612)
Observations	2310	2294	2230	2281	4265	4265	4265	4265	15,071	15,007	14,991	900
AR (1)	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
AR (2)	0.538	0.08	0.7	0.417	0.8	0.89	0.341	0.565	0.114	0.44	0.988	0.255
Hansen value	0.111	0.764	0.381	0.904	0.313	0.47	0.203	0.227	0.54	0.074	0.128	0.213

This table used the two-step system GMM method to measure the impact of funding liquidity on banks' assets risk (risk-weighted assets ratio and loan loss provision ratio) and overall risk (liquidity creation and Z-score) for well, adequately and undercapitalized Banks. Robust standard errors are reported in parentheses. ***, **, * represent statistical significance at 1%, 5%, and 10%, respectively.

Columns 9, 10, 11, and 12 of Table 4 present the results for the relationship of liquidity funding and bank risk-taking of too-big-to-fail banks. We follow the procedure of Federal Reserve Banks to separate the too-big-to-fail banks. The study used TB as an interactive term for too-big-to-fail banks. The results provide that there is an inverse relationship between liquidity funding and bank’s assets and overall riskiness at 1% and 5% level of significance for too-big-to-fail banks, respectively. The findings are in line with the previous studies by Khan et al. (2017) and contradicting the results of (De Haan and Poghosyan 2012). The findings support that larger banks are inclined to decrease in the assets and overall risk in response to stable funding liquidity as a comparison of smaller banks, which is consistent with (De Haan and Poghosyan 2012; Khan et al. 2017). The argument may have an explanation due to the higher importance of big banks and due to the stricter regulations and vigilant monitoring within the banking industry. The larger banks take a lower risk and keep their status more solvent and sounder to secure a high rating among counterparts. The results show a positive relationship between deposits to total assets and risk-weighted assets, liquidity creation, and Z-scores for too-big-to-fail banks in the short run—other things held unchanged.

4.4. Funding Liquidity and Risk-Taking during Financial Crisis

Table 5 reports the results for the effect of liquidity funding on bank assets and overall risk. The study uses a dummy variable to predict the impact of liquidity funding on risk-taking in crisis conditions. The study used a financial crisis dummy (FCD) to incorporate a conclusion similar to (Khan et al. 2017; Yousaf et al. 2018). The coefficient shows that the loan loss provision, liquidity creation, and Z-scores are statistical and significant relationships with liquidity funding during the crisis period consistent with (Khan et al. 2017). This indicates that liquidity funding remains significant for commercial banks to take the risk during the crisis period. In the case of the Z-score the findings reveal that when a deposit increases in the crisis period, the commercial banks become more stable, the results are in line with (Khan et al. 2017). The justification for the increase in bank deposits during the crisis period is due to lower return on securities, due that depositors prefer to deposit the money instead of investing in risky securities, as argued by (Acharya and Mora 2015). The coefficient of the intercept dummy is significant for risk-weighted assets, liquidity creations, and Z-scores. The results show that commercial banks were taking a risk in crisis as evidenced by risk-weighted assets, liquidity creation, and Z-scores.

Table 5. Relationship between funding liquidity ratio and banks risk during crisis.

VARIABLES	Bank’s Asset Risk		Bank’s Overall Risk	
	(1) RRWA	(2) LLP	(3) LC	(4) Z-SCORE
Lagged risk	0.888 *** (0.022)	0.845 *** (0.015)	0.868 *** (0.024)	0.112 *** (0.000)
Funding liquidity	0.001 *** (0.000)	0.002 *** (0.000)	0.002 *** (0.000)	−0.016 *** (0.012)
Funding liquidity *FCD	0.001 (0.000)	0.001 (0.000)	0.005 (0.000)	0.009 (0.013)
FCD	0.003 ** (0.001)	0.004 (0.000)	−0.001 ** (0.000)	0.071 *** (0.021)
Loan Ratio	0.001 ** (0.000)	−0.002 (0.000)	0.006 ** (0.000)	−0.765 (2.070)
Size	−0.001 * (0.000)	−0.009 *** (0.000)	−0.008 ** (0.000)	0.439 ** (0.001)
Capital	0.203 *** (0.033)	0.003 (0.001)	−0.350 *** (0.034)	0.901 *** (0.003)
ROA	0.700 *** (0.101)	−0.046 *** (0.004)	−0.189 *** (0.065)	0.023 *** (0.000)

Table 5. Cont.

VARIABLES	Bank's Asset Risk		Bank's Overall Risk	
	(1) RRWA	(2) LLP	(3) LC	(4) Z-SCORE
Constant	0.072 *** (0.013)	0.002 *** (0.000)	0.077 *** (0.009)	−0.015 *** (2.979)
Observations	15,071	15,007	14,991	2882
AR (1)	0.000	0.000	0.000	0.000
AR (2)	0.131	0.410	0.939	0.495
Hansen value	0.754	0.100	0.161	0.824

This table used two-step system GMM method to measure the impact of funding liquidity on banks' assets risk (Risk-weighted assets ratio and Loan loss provision ratio) and overall risk (Liquidity creation and Z-score) by using the data of US commercial banks from 2002–2009. Robust standard errors are reported in parentheses. ***, **, * represent statistical significance at 1%, 5% and 10% respectively.

4.5. Robustness Check

Table 6 shows the impact of liquidity funding on bank risk-taking after controlling the macroeconomic factors. We follow the suggestion of Khan et al. (2017) and test the robustness of our baseline findings by controlling the macroeconomic factors. We added the gross domestic product and inflation rate to the baseline equations. These factors are commonly employed as the representative of economic conditions in the literature (Buch et al. 2014; Khan et al. 2017). The findings are similar to the baseline results after adding macroeconomic factors. The results show that there is a positive and statistically significant relationship between funding liquidity and risk-weighted assets, loan loss provisions, and liquidity creations as the base results and consistent with (Acharya and Mora 2015; Khan et al. 2017). The results show that an increase in funding liquidity increases the risk-taking of commercial banks, as evidenced by risk-weighted assets, loan loss provisions, liquidity creations, and Z-scores. However, in our results, the impact of economic growth on bank risk-taking is positive and significant, which is contradicting the findings of (Khan et al. 2017). Theoretically, an increase in economic growth leads to an increase in the demand for funds, leading to an increase in banks' riskiness in terms of higher risk-weighted assets and liquidity creations in the short run when other things remain constant.

Table 6. Impact of funding liquidity on risk after including macroeconomic factors.

VARIABLES	Bank's Asset Risk		Bank's Overall Risk	
	(1) RRWA	(2) LLP	(3) LC	(4) Z-SCORE
Lagged risk	0.924 *** (0.021)	0.959 *** (0.016)	0.865 *** (0.024)	:
Funding liquidity	0.007 *** (0.000)	0.002 *** (0.000)	0.002 *** (0.000)	−0.003 *** (0.001)
Loan Ratio	0.001 *** (0.000)	−0.007 *** (0.001)	0.006 ** (0.002)	−0.268 (1.210)
Size	−0.008 * (0.000)	−0.001 *** (0.000)	−0.008 ** (0.000)	0.089 *** (0.047)
Capital	0.210 *** (0.032)	−0.002 ** (0.001)	−0.354 *** (0.034)	0.594 *** (0.270)
ROA	0.572 *** (0.105)	−0.034 *** (0.004)	−0.191 *** (0.068)	0.131 *** (0.500)
Economic growth	0.002 *** (0.000)	−0.002 *** (0.000)	0.004 (0.000)	0.025 (0.021)
Inflation rate	0.004 *** (0.000)	−0.002 *** (0.000)	−0.001 (0.000)	−0.1034 *** (0.447)

Table 6. Cont.

VARIABLES	Bank's Asset Risk		Bank's Overall Risk	
	(1) RRWA	(2) LLP	(3) LC	(4) Z-SCORE
Constant	0.030 ** (0.013)	0.002 *** (0.000)	0.076 *** (0.009)	−4.254 *** (0.413)
Observations	15,071	15,007	14,991	14,909
AR (1)	0.000	0.000	0.000	0.000
AR (2)	0.126	0.838	0.959	0.782
Hansen value	0.133	0.204	0.155	0.988

This table used two-step system GMM method to measure the impact of funding liquidity on banks' assets risk (Risk-weighted assets ratio and Loan loss provision ratio) and overall risk (Liquidity creation and Z-score) by using the data of US commercial banks from 2002–2009. Robust standard errors are reported in parentheses. ***, **, * represent statistical significance at 1%, 5% and 10% respectively.

5. Conclusions

Both funding liquidity and risk-taking behavior are important indicators to measure the stability of commercial banks, especially in the period of the post-Global financial crisis. It is, therefore, important to investigate the relationship between funding liquidity and risk-taking behavior for banks because the findings could be useful for both regulators and bank managers to make better decisions for the stability of commercial banks. Thus, numerous researchers explore the risk-taking behavior of commercial banks in the recent decade. However, most, if not all, researchers examine the relationship for some specific banks, but not for any well-capitalized, under-capitalized, and too-big-to-fail commercial bank. The lack of insights in the context of well-capitalized, under-capitalized, and too-big-to-fail commercial banks in the USA motives us to bridge the gap in the literature to investigate the impacts of funding liquidity on risk-taking behavior of commercial banks.

We use several financial and economic theories to explore the possible relationships between funding liquidity and commercial banks' risk-taking behavior in the USA, including the theory of regulatory hypothesis, moral hazard hypothesis, and agency theory. Based on the theories, we set four hypotheses to state the relationships between funding liquidity and risk-taking behavior for commercial banks in the USA. To test the hypotheses, we use the GMM technique because it helps to control the issues of endogeneity, autocorrelation, and cross-sectional heteroscedasticity.

Our findings confirm that banks with stable funding liquidity are inclined toward higher risk-taking. We assume that banks with higher deposits have a lower risk attitude on funding liquidity because deposits reduce run risks from banks in the presence of deposit insurance. Banks with higher deposits are less likely to face a funding shortage, and bank managers' aggressive risk-taking activity is less likely to be monitored. Our results demonstrate that increases in bank funding liquidity increase both risk-weighted assets and the creation of liquidity, which is consistent with the findings in some studies such as Acharya and Naqvi (2012) that banks aggressively lend at lower loan rates in response to higher deposits. Our findings are also consistent with other findings in the literature, for example, Keeley (1990) and Khan et al. (2017), in which deposit insurance creates the issue of moral risk for banks taking excessive risks in response to deposit rises. Furthermore, we find that deposits are negatively related to banks' Z-score for both well-capitalized and too-big-to-fail banks. The relationship between funding liquidity and the risk-taking behavior of the bank varies with their capitalization. The findings indicate that well-capitalized banks hold lower liquidity and lend more to secure higher profits. The second cause may be the role of deposit insurance, and the third reason is the availability of bailout packages. The undercapitalized banks hold higher liquidity and take a lower risk, but they still face difficulty in raising capital on short notice. Our analysis also finds that the impact of funding liquidity is even stronger during the period of the Global Financial Crisis 2007, inferring that banks would take a higher risk during the crisis. The results are consistent after adding other macroeconomic variables to the model.

Both academics and practitioners could benefit from the following inferences drawn from our findings. First, regulators should not only focus on improving bank capital for the stability of commercial banks but also consider the role of funding liquidity to influence the risk-taking behavior of commercial banks. Second, the heterogeneity in our findings has implications for policymakers in commercial banks for the improvement of commercial banks' stability. Third, our findings help commercial bank regulators to consider set regulations not only for well-capitalized and too-big-to-fail banks but also for under-capitalized banks. Moreover, our findings also help commercial bank regulators formulate guidelines for banks' risk-taking activities. Our study provides a thorough understanding of the relationships between the risk of funding liquidity and banks' risk-taking behavior, which could help regulators to redesign the banking regulatory framework to get better discipline and regulate the perverse incentives for bank managers to take higher risk in the future when bank deposits change, especially in the period of crisis.

In this paper, we apply the GMM technique to examine the relationship between funding liquidity and commercial banks' risk-taking behavior. An extension of our paper could apply the GMM technique to examine other issues, for example, between stocks and REITs (Chiang et al. 2008). Readers may refer to Chang et al. (2018) for other important issues that one could use the GMM technique to examine. For the limitations, the first limitation of our study is that our paper conducts analysis for banks listed only at Bankscope. The second limitation of our study is that the data are not for a very long period. Thus, extensions of our study include collecting data for a longer period and obtaining data of banks listed not only at Bankscope but also in other cities/countries. Future research could also be conducted to study the interrelationship between risk-based capital, risk-taking, and profitability of US commercial banks by incorporating both mediating and moderating roles of other economic variables and bank regulations to get better in-depth insights.

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Notes

¹ For detail see (<http://www.bis.org/bcbs/basel3.htm>, accessed on 21 December 2019).

² Dodd Frank Wall Street Reform and Consumer Protection Act 2010.

³ Ordinary least squares.

⁴ <https://www.federalreserve.gov/releases/lbr/current/>, accessed on 21 December 2019.

⁵ For more details of crisis: (Umar et al. 2021a, 2021b, 2021c) also use the world financial crisis as subsample.

⁶ <https://data.worldbank.org/indicator>, accessed on 21 December 2019.

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