LawFin Working Paper No. 44

Capital Regulation, Market-Making, and Liquidity

Rainer Haselmann | Thomas Kick | Shikhar Singla | Vikrant Vig
Capital Regulation, Market-Making, and Liquidity

Rainer Haselmann*  Thomas Kick†  Shikhar Singla‡  Vikrant Vig§

January 2022

Abstract

We employ a proprietary transaction-level dataset in Germany to examine how capital requirements affect the liquidity of corporate bonds. Using the 2011 European Banking Authority capital exercise that mandated certain banks to increase regulatory capital, we find that affected banks reduce their inventory holdings, pre-arrange more trades, and have smaller average trade size. While non-bank affiliated dealers increase their market-making activity, they are unable to bridge this gap – aggregate liquidity declines. Our results are stronger for banks with a higher capital shortfall, for non-investment grade bonds, and for bonds where the affected banks were the dominant market-makers.

Keywords: market-making, capital regulation, bond market liquidity

JEL Classification: G01, G21, G28
1 Introduction

A large fraction of financial assets (bonds, OTC derivatives, currencies, commodities, and MBS) are intermediated via banks (Duffie (2012)). Banks act as market makers and perform a valuable function of providing immediacy to various market participants. The importance of banks as market makers is underscored from the various exemptions granted to their underwriting and market-making activities around the world. For example, while the Volcker rule prohibits proprietary trading, it exempts banks’ market-making activities. European regulators took a similar approach to preserve the market-making function of banks (European Central Bank (2015)). Embedded in these actions is a view that a reduction in market-making activity would reduce liquidity in secondary markets, thereby increasing the cost of capital of firms.

Post the global financial crisis of 2008, regulators have introduced a host of new reforms aimed at increasing the stability of the financial system. Capital regulation of banks is at the core of these reforms. New regulations requiring banks to increase their capital have been introduced with the view that an increase in capital requirements increases the ‘skin in the game’ and enhances financial stability. While the overall effect of higher capital requirements on the broader economy continues to be debated by both academics and policymakers, concerns have been raised in some quarters that an increase in capital requirements may adversely impact banks’ market-making activities and subsequently bond liquidity (Bessembinder et al. (2018)). In this paper, we empirically investigate the validity of these concerns. More specifically, we examine whether the tightening of capital requirements for financial intermediaries led to a deterioration of corporate bond liquidity.

The impact of higher capital requirements on banks’ market-making activities is a priori not obvious. An increase in capital requirements directly increases the inventory holding costs as the inventory itself is subject to a higher capital charge. An increase in inventory holding cost is likely to reduce banks’ inventory. As one of the prime objectives of inventory is to provide immediacy to market participants, it is plausible that an increase in capital requirement may adversely affect the market activity of banks. This, however, is not the only possibility. It is plausible that banks higher capital requirements may potentially reduce
banks funding costs (Begenau (2020)) and increase market-making activities (Bahaj and Malherbe (2020)). Importantly, even if higher capital requirements dampen banks’ market-making activities, the impact on overall bond market liquidity is not clear since non-bank dealers might step in and take over banks’ market-making activities (Duffie (2012)).


Two obstacles hinder the evaluation of the effect of capital regulation on the market-making activities of banks. To start with, regulations are often bundled together, making it difficult to isolate the impact of a specific regulation. The empirical strategy requires to focus on a specific regulation (“events”) along with a high-frequency dataset on corporate bonds transactions at a very granular level. This is a tough ask. Causal inference is further complicated by the fact that the equilibrium market-making activity is an outcome of both supply-side and demand-side factors. On the demand-side, banks cater to clients with different preferences for liquidity and the composition of the client pool may vary over time affecting the demand for liquidity.² To isolate the effect of regulation on market-making carried by banks, one must control for demand-side forces that are simultaneously at work.

We utilize the detailed security transaction data of German banks³, and exploit the 2011 capital exercise conducted by the European Banking Authority (EBA) to examine the effect of capital regulation on the market-making activity of banks. The capital exercise, also

¹The following quote by then-Fed Chair Janet Yellen, when asked to comment on this issue, summarizes the view quite clearly: “It is not clear whether there is or is not a problem. By some metrics, there doesn’t seem to be a problem; by others, there does seem to be.”

²A shift towards institutional ownership (patient capital), for instance, could result in a lower demand for liquidity (Li (2021)).

³We employ two micro datasets provided by the German regulatory authorities: 1) The Security Holdings Statistics, and 2) The Security Transaction dataset. The first dataset provides detailed portfolio holdings of all German banks at a quarterly frequency on a security-by-security basis. The Security Transaction dataset, on the other hand, contains detailed information on all trades conducted by German banks at a daily frequency. The dataset is comprehensive and covers most German corporate bonds traded. The dataset contains date & time, price, volume, buy or sell indicator, currency and an exchange code or indicator for OTC trades. It also has security information and the parties involved (identifier for the reporting institution and, where applicable, identifiers of client, counterparty, broker or intermediaries). Please Section 3.1 for more details.
referred to as a recapitalization exercise, was a one-time policy announced on October 26, 2011, and required certain banks to increase the core tier 1 capital ratio from 7% to 9% by June 30, 2012. To trace out the causal impact of higher capital requirements, we study market-making activities of banks that were required to alter their capital ratio in the given time frame (affected banks) with those banks that were exempted from it (i.e. unaffected banks). Our primary analysis, essentially a difference-in-difference analysis, is conducted at the individual security level. The granularity afforded by our security level dataset allows us to track the same bond on the same day across different banks. By doing so, we can non-parametrically control for any time-varying bond characteristics and demand factors that have plagued previous research designs.

On examining the corporate bond inventory of banks using the Security Holding Statistic database, we find that the affected banks responded to the capital exercise event by reducing the volume of their inventory by 19.1% relative to the unaffected banks. The results are economically large and statistically significant. The results are also robust to the inclusion of bond interacted with quarter fixed effects, a saturation that allows us to examine the differential behavior of affected banks and unaffected banks for the same bond in the same quarter. Overall, these results illustrate that the patterns are driven by changes in affected banks’ behaviour and not by changes in the immediacy demanded by investors.

We next explore the Security Transactions Dataset to investigate banks’ market-making activities around the capital exercise event. Compared to the Security Holding dataset, which is at a quarterly frequency, the Security Transactions Dataset contains detailed information on all trades conducted by German banks at a daily frequency. We use three different measures that are standard in the market microstructure literature to capture the market-making activities of banks (see Bessembinder et al. (2018) and Schultz (2017)), namely: 1) absolute value of end-of-day changes in inventory, 2) average trade size, and 3) principal trading volume over total trading volume. These measures capture the degree to which market makers absorb order imbalances and provide immediacy to market participants.

A similar pattern emerges under all three measures. We find that the affected banks drastically reduced their market-making activities relative to the unaffected banks. The daily

\[4\] The policy affected 61 European banks, which included 12 German Banks.
changes in inventory levels and the average trade size are meant to capture banks’ willingness
to absorb order imbalances but could also capture investor preferences for immediacy. To
rule out demand-side factors, our main specification includes bond interacted with day fixed
effects. We find that affected banks’ changes in daily inventory levels are 38% lower relative
to the unaffected banks. Similarly, we document a 20.7% reduction in the average trade size
of affected banks relative to the unaffected banks.

Our third measure captures the extent to which banks act more like brokers rather than
dealers. Brokers do not hold inventory and only connect the two parties interested in a trade.
In contrast, bank dealers sell or buy the asset using their own inventory and, thus, have to
commit capital and bear price risk. If banks prearrange trades, they essentially act as brokers
and make the customer wait (provide lower immediacy) until the other leg of the transaction
is completed. We find that affected banks prearrange more trades after the capital exercise
event, and the percentage of principal trading by volume decreases compared to unaffected
banks. More specifically, banks reduced principal or dealer-like trading by about 7% of the
mean pre-event levels. Overall, our results support the view that the event decreased affected
banks willingness to provide immediacy to market participants.

Our analysis so far compares the behavior of affected banks with unaffected banks. How-
ever, the assignment to these categories is not random. In fact, the EBA capital exercise
explicitly targeted the larger banks. To sharpen the analysis, we test our central hypothesis
by utilizing variations within the sample of affected banks. More specifically, we compare the
set of banks with higher capital shortfall with those having lower capital shortfall in the EBA
capital exercise. The two subsets of banks are evenly balanced on size and other covariates,
assuaging concerns of biases that might be introduced by confounding variables. Our results
are both qualitatively and quantitatively similar. The net volume traded for high shortfall
banks is reduced by 22.7% compared to those not experiencing a high shortfall. Banks with
higher shortfall also pursue smaller trades (by 9.1%) compared to the other participating
banks.

We also test whether banks’ reaction has been more pronounced for those bonds with
relatively high capital requirements, i.e., non-investment grade bonds, as they have higher
risk-weights associated with them. Net volume traded for non-investment grade bonds has
reduced by 23% when compared to investment-grade bonds (low risk-weight bonds) of affected banks after the capital exercise. By incorporating bank $\times$ day fixed effects in this specification, we systematically control for any concerns related to the differences between banks that are part of the affected and unaffected group.

As a final test, we compare market-making activities in times of stress and on regular days before/after the regulatory treatment. Federal Reserve System governor Jerome Powell expressed the possibility that though the liquidity may be adequate in normal conditions, it may disappear under stress (Powell (2016)). Academics, policymakers and market participants have noted that immediacy during times of stress is an essential measure of market liquidity.\(^5\) We observe that the ability of affected banks to absorb higher inventory change on a stress day has gone down relatively by 30.7%.

In summary, the increase in capital requirements seems to have raised the trade execution costs for financial intermediaries, which reduces market-making activities in line with the concerns raised by Duffie (2012). We further investigate how persistent this effect is. During the four subsequent quarters following the announcement of the EBA capital exercise, there is a significant reduction in market-making activities by affected banks. Once we focus on non-investment grade bonds (high risk-weights) bonds only, we find that the dynamic treatment effect remains statistically significant until the end of our sample period.

The results so far suggest that the capital exercise led to the reduction in market-making by affected banks. This raises an important question: Does these banks’ reduction in market-making translate into lower bond liquidity? The impact on overall bond liquidity will depend on the extent to which non-affected banks and unregulated institutions step in and take over the activities of the affected banks. Time-series evidence of liquidity measures such as the Amihud ratio and Imputed round trip cost (IRTC) measured using transactions data (calculated at weekly level) suggests a deterioration of corporate bond market liquidity following the event. This effect is persistent for both liquidity measures until the end of our observation period (end of 2016). Presented effects are quite sizeable. The Amihud ratio is up by 1.647 percentage points per million euros after the capital exercise, which is 61.9% of the

\(^5\)Bao et al. (2018) also study liquidity during times of stress before and after the implementation of the Volcker Rule. Acharya et al. (2013) suggest the existence of time-varying liquidity risk of corporate bond returns conditional on episodes of flight to liquidity.
pre-event mean value. Similarly, the IRTC liquidity measure increases by 0.792 percentage points which is 96.4% of the pre-event mean value.

How is it possible to have lower bond market liquidity without seeing any visible movements in bid-ask spreads offered by banks? This question has puzzled market participants and policymakers. Our results provide an answer to this question. Bank-affiliated dealers execute smaller trades in response to higher capital requirements, and these trades tend to have smaller bid-ask spreads. We also find that banks take on less risk by pre-arranging more trades which would mean they earn lower profit and have smaller bid-ask spreads (Schultz (2017)). Hence, the impact on bid-ask spreads is not visible despite the reduced liquidity.

Can the illustrated deterioration in bond market liquidity be attributed to the increase in capital charges? To test for this, we suggest a differences-in-differences approach. We divide our sample bonds into two groups based on the effect capital exercise has on them, i.e. into less affected and more affected bonds based on the percentage of trading performed by affected banks in the pre-period. Liquidity has fallen only for those bonds where the affected banks were important market makers.

Another unintended consequence that we foresee is the migration of market-making activities to unregulated financial entities, i.e. shadow banks, insurance companies or hedge funds. Our findings on secondary bond market liquidity illustrate that non-regulated financial institutions could not absorb the reduction in market-making activities of affected banks even after five years of the event. We provide time-series evidence of an increase in trading activity by non-bank broker-dealers. The proportion of trading volume in the corporate bond market by non-bank brokers and dealers has increased from 12% in 2010 to 30% in 2015. Once we focus on non-investment grade bonds, the increase in the market share of non-bank dealers is more pronounced. While this observation is in line with anecdotal evidence which specifies that trading activities migrate from banks to non-regulated institutions, the inventory size of these dealers seems to be too small to restore bond market liquidity at pre-event levels (see Bessembinder et al. (2018) for a similar observation based on US data).

We provide several robustness checks that suggest that other confounding events do not drive our results during the same time. The European sovereign debt crisis and ECB’s long term refinancing operation (LTRO) in particular seem not to explain our findings. To
rule out sovereign debt crisis as an explanation of our results, we divide the affected and unaffected banks into quartiles by exposure to Greece, Ireland, Italy, Portugal and Spain (GIIPS) sovereign bonds and using GIIPS exposure \( \times \) Day fixed effects; we compare banks with similar GIIPS exposure. We also analyze how GIIPS exposure impacts banks’ market-making activities post-period by interacting the GIIPS exposure with a post-event dummy. We find that though GIIPS exposure has a negative impact on market making, it was not different for affected and unaffected groups since our diff-in-diff estimator does not change much in size or significance. In addition, ECB’s long term refinancing operation (LTRO) during our sample period does not affect our results since only one bank in our sample had an LTRO take-up. We also perform a placebo test using German government bonds, which have zero risk-weights associated with them and find that trading in these bonds was not differentially impacted for affected and unaffected banks.

Our paper contributes to the literature on financial regulation and bond market liquidity. Previous evidence on this topic is mixed. On the one hand, Mizrach (2015) and Anderson and Stulz (2017) find that corporate bond liquidity did not deteriorate, Trebbi and Xiao (2017) while testing for breakpoints in various liquidity measures do not find that the estimated structural breaks coincide with the passage of Dodd-Frank or Volcker Rule. This evidence suggests that bond market liquidity has not structurally diminished after the crisis.

On the other hand, Goldberg and Nozawa (2021) find that liquidity supply shocks were negative during the implementation of the Volcker Rule. Anderson and Stulz (2017) observe higher transaction costs and price impact for large trades. Bao et al. (2018) find that the price impact of recently downgraded corporate bonds has increased after the Volcker Rule implementation and Dick-Nielsen and Rossi (2019) show that the price of immediacy has increased after the crisis. Schultz (2017) finds that after the Volcker Rule was implemented, dealers were more reluctant to take bonds into inventory, were more likely to offset trades within a day and unwind inventory positions more quickly.\(^6\) The paper related most closely

\(^6\)Further, Choi and Huh (2019) show that dealers provide less liquidity post-crisis, but non-dealers provide more liquidity explaining why bid-ask spreads remain low. Adrian et al. (2017b) study the relationship between bond-level liquidity and financial institutions’ balance sheet constraints. Their main finding is that post-Volcker, bonds traded by institutions with lower leverage are more liquid. Theoretically, Bruche and Kuong (2019) consider a model in which dealers need to raise external financing to provide immediacy to their clients and argue that regulations that limit the leverage of bank-affiliated dealers can have adverse effects on market liquidity.
to our study is by Bessembinder et al. (2018). They conduct a comprehensive analysis of liq-
uidity in the corporate bond markets over the last two decades. Their findings are consistent 
with the reasoning that regulatory initiatives reduced dealers’ willingness to commit capital 
to market-making in bonds. We add to this literature by providing a causal identification of 
the link between regulatory capital requirements and corporate bond market liquidity.

A recent literature by He and Krishnamurthy (2013) documented the impact constraints 
on equity capital of financial intermediaries have on asset prices. According to Goldberg 
and Nozawa (2021), liquidity supply shocks to dealer inventory capacity have explanatory 
power for the cross-sectional and time-series variation in expected bond returns. Another 
strand of literature theoretically shows the impact of higher capital requirements on bank 
lending and assets. We also add to the discussion regarding the intended and unintended 
consequences of higher capital requirements. Admati et al. (2013) argue that regulatory 
capital requirements should be higher to enhance financial stability and also add that this 
would be not socially expensive in the long-run. While several papers have studied intended 
as well as unintended consequences of increasing regulatory capital requirements, the impact 
of higher capital charges on banks’ market-making activities and bond market liquidity 
remains an open question we tackle in this paper.7

Several papers highlighted that tightening regulatory financial regulation may push in-
termidiation to unregulated entities (e.g., so-called shadow banks) with potential adverse 
consequences on financial stability and overall welfare (Admati et al. (2018), Farhi and Tirole 
(2020), Martinez-Miera and Repullo (2018), Plantin (2015), for a summary see Adrian and 
Ashcraft (2016)).8 In line with Bessembinder et al. (2018) who show that non-bank dealers’

7 Gropp et al. (2018) examined how banks’ respond to higher capital requirements in terms of decreasing 
risk-weighted assets (most importantly corporate lending). Several other papers studied the implications of 
higher capital requirements on banks’ loan supply (e.g. Jimenez et al. (2017) investigate how higher dynamic 
provision rules result in a contraction in loan supply; see further Berrospide and Edge (2019) for a summary). 
Blattner et al. (ming) finds that banks not only reduce lending but reallocate credit to distressed firms with 
underreported loan losses. Behn et al. (2021) document how capital requirements that rely on model-based 
regulation have affected the reporting standard for risk parameters, as well as lending to firms with low risk, 
weighted assets.

8 Empirically, Irani et al. (2018) show that less-capitalized banks have reduced loan retention when unreg-
ulated entities step in, particularly for risky loans. With regards to mortgage lending, Buchak et al. (2018) 
develop a quantitative model of mortgage lending suggesting that regulation accounts for roughly 60% of 
shadow bank growth. Similarly, Gete and Reher (2017) document how liquidity regulation has attracted 
non-banks and originate-to-sell lenders towards the FHA market.
role has increased in the US since the financial crisis, we document that this trend has been present in Europe as well, a significant part of which can be traced to the tightening of capital regulation.

Our paper is structured as follows. Section 2 describes institutional details of the capital exercise carried out in 2011. Our data sets, as well as descriptive statistics, are presented in Section 3. The set-up of our empirical identification strategy is explained in Section 4. We present our results on banks’ market-making activities and robustness tests in Section 5. Results on corporate bond liquidity are discussed in Section 6. Concluding remarks are in Section 7.

## 2 Institutional Background

The EBA capital exercise, also referred to as recapitalization exercise, was conducted to restore investor confidence and improve the stability of the European banking sector. The capital exercise was announced on 26th October 2011 (see Figure 1 for a timeline of the event) only three months after the 2011 stress test on a nearly identical sample of banks. Given that the results of the 2011 stress test had been deemed as “meaningless” by the financial press, European regulators considered this recapitalization exercise a necessity.\(^9\)

The EBA required 61 European banks, 12 of which were German banks, to increase their core tier 1 capital ratio from 7% to 9% by June 2012.\(^10\) The core tier 1 capital comprised of the highest quality capital instruments and included common equity, ordinary shares, newly issued contingent convertibles (CoCos) and government support measures, among others.\(^11\)

While the timing of the capital exercise and the high capital requirements it mandated came as a surprise to banks and stakeholders (see, e.g. Blattner et al. (ming) and Gropp et al. (2018) on this issue), a recapitalization of the European banking sector has been discussed

---


\(^10\)The minimum CET1-to-RWA ratio under Basel II was 2%. Basel III raised this threshold to 7%. A countercyclical buffer of up to 2.5% was set, but the phasing in of the new requirements was planned to be progressive, with a first mandatory increase of the minimal CET1 ratio from 2 to 3.5% in January 2013 and a gradual implementation of the additional CET1 buffer thereafter. Eligible Core Tier 1 capital was defined in a methodological note issued on 8th December 2011.

\(^11\)For exact details please refer to EBA’s documentation which can be downloaded from: https://www.eba.europa.eu/documents/10180/26923/Sovereign-capital-shortfall_Methodology-FINAL.pdf/acac6c68-398e-4aa2-b8a1-c3dd7aa720d4
since the presentation of the stress test results in July 2011. The exercise had a significant impact and led to an increase of banks’ capital position by more than 200 Billion Euros. In Germany, where 6 of the 12 selected banks fell short of the capital requirements set by the capital exercise, the increase in the Tier 1 capital ratio was achieved mainly by shrinking assets of affected banks – a trend not observed for non-affected banks in Germany during this period.

As stated earlier, the empirical analysis focuses on corporate bonds in Germany. We briefly describe how capital charges of these bonds were determined. Under the current European regulatory framework, banks must indicate whether a specific bond is part of their trading book or the credit book. Market-making activities of bank-affiliated dealers are generally allocated to a bank’s trading book. If a bank uses internal market risk models, a daily value at risk (VaR) model is calculated for its entire trading book to determine the fundamental review expected shortfall. The higher this expected shortfall, the higher the regulatory capital requirements (European Parliament (2019)). If a bank determines its capital charges under the standard approach for market risk, it has to hold equity capital between 0% and 12% depending on the rating of the bond (see Article 336 of CRR). Importantly, irrespective of the regulatory approach chosen (SA or IRB), an increase in regulatory capital requirements increases banks’ capital commitment for a given inventory...

---


13Out of the 61 banks that were required to participate in the EBA capital exercise, 27 banks had an initial shortfall for the target ratio set by the EBA. These banks strengthened their capital position by 116 Billion Euros. The remaining banks that did not experience a shortfall found their capital ratio closer to the new minimum requirements and increased their capital position by 84 Billion Euros. Note that the EBA 2011 stress test also required banks to increase their capital requirements. However, the consequences of this event were considerably milder for the banks, i.e. the estimated shortfall for the stress test was 2.5 billion euros. Hence, we focus on the capital exercise as the event which increased capital requirements for a subset of German banks.

14If a bond is placed in the credit book, the risk weight depends on its external rating, which varies between 0% and 150% under the Standardized Approach (SA). Under the Internal Rating Based Approach (IRB), the probability of default (PD) and the loss given default are inputs in determining the risk weight. In any case, an increase in regulatory capital requirements increases the capital commitment for a bond inventory and, relatively more for an inventory of non-investment grade bonds.

15See Begley et al. (2017) for a more detailed description.

16The current rule for determining capital charges for market risk in the banking book can be found under: https://www.bundesbank.de/de/aufgaben/bankenaufsicht/einzelaspekte/eigenmittelanforderungen/marktrisiko

17If a bank cannot hedge against interest rate risk or duration risk, the bank has to further hold additional capital for each bond position (see Article 339 and 340 of CRR).
of corporate bonds (unless bonds have an AAA rating, in which case they get assigned a risk weight equal to zero). In the following sections, we investigate how an increase in regulatory capital requirements affects the market-making of affected banks.

3 Market Making Activities, Bond Market Liquidity and Data

3.1 Data and Bank-Level Descriptive Statistics

We use two proprietary datasets from the two institutions involved in regulating and supervising the German banking system. These are Deutsche Bundesbank and the German Federal Financial Supervisory Authority ("BaFin") for this study. The first dataset, the Security Holdings Statistics, comprises portfolio holdings of all German financial institutions on a security by security basis. Banks report all of their securities holdings, irrespective of where they are held, to Deutsche Bundesbank on a quarterly basis. The dataset provides information for different asset classes, such as bonds, equities, and mutual funds. We obtain the nominal amount for each security an individual bank holds at the end of each quarter from the database.

The principal dataset used for this study is called the Securities Transactions database, which is maintained by BaFin. The German Securities Trading Act (Wertpapierhandelsgesetz, "WpHG") mandates credit institutions, financial services institutions, branches of foreign institutions and central counterparties based in Germany, to report the following transactions: (i) instruments traded on an organized market; (ii) instruments included in the regulated market; (iii) instruments traded on the open market.\footnote{The open market refers to a segment of the German stock exchange that has lower transparency requirements for firms to be traded compared to the regular listed firms.}

The transactions dataset contains date and time, price, volume, buy or sell indicator, currency, exchange code or indicator for OTC trades of the transaction. It also has security information on the parties involved (identifier for the reporting institution and, where applicable, identifiers of client, counterparty, broker or intermediaries). Banks are further
required to report for each trading activity whether (1) it acts on behalf of a client and takes
the security on its book, (2) it acts on its own behalf, or (3) it acts like a riskless broker
on behalf of a client without taking the security on its book. As our primary interest is in
banks’ market-making activities, we concentrate on all transactions of categories (1) and (2),
given that these trades affect banks’ balance sheets.

We merge this dataset using the unique International Securities Identification Number
(ISIN) with the Centralized Securities Database (CSDB). The CSDB Database is managed
jointly by the members of the European System of Central Banks (ESCB). It includes in-
formation about the issuer of the security, e.g. issuer sector and issuer country. Finally,
we add detailed balance sheet and income statement information of our sample banks from
Bundesbank’s BAKIS database.

We focus on the market-making of corporate bonds issued by German firms with a CFI-
code starting with “DB” (D for debt instruments and B for bonds) and drop all financial
industry bonds.\footnote{We exclude convertible bonds (“DC”), other debt (“DM”), medium term notes (“DT”), bonds with warrants attached (“DW”) and money market instruments (“DY”).} We further consider all trades in Euros and the main non-Euro currencies:
“AUD” (Australian dollar), “CHF” (Swiss franc), “GBP” (pound sterling), “USD” (US
dollar), “CAD” (Canadian dollar), “JPY” (yen), “HKD” (Hong Kong dollar) and “SGD”
(Singapore dollar). These trades are converted into “EUR” (Euro) using daily exchange
rates.

Our sample period spans from 2010 to 2012. The pre-period ends on July 15th 2011
(announcement of the results of the stress test), and the post-period starts on the announce-
ment day of the capital exercise, October 26th 2011. We further focus on banks that trade
frequently. Overall, 95% of the total trading volume is handled by the 37 banks which form
our sample. The EBA selected the 12 largest German banking groups to be included in the
exercise. Since the capital exercise was conducted at the highest level of consolidation, we
identify the subsidiaries of these 12 largest banks and classify them as the affected group.
Out of the 37 banks in our sample, 11 belong to the affected banking groups, and the other
26 belong to unaffected banking groups.

Our dataset has comprehensive coverage of market-making in German corporate bonds.
Most of the market-making in German corporate bonds is done by German entities in either Germany or the UK. We are potentially missing some market-making by non-German banks. For example, we would not observe in our dataset if BNP Paribas was acting as a market-maker for some German bonds in Paris. This is a very small sample that pertains only to very large German corporates. Further, it is essential to note that non-German European banks were also part of the EBA capital exercise. Therefore, our sample is not only comprehensive, but representative of the market-making activity carried out for German corporate bonds.

Table 1, Panel A provides summary statistics for affected and unaffected banks. In this table, all summary statistics presented refer to pre-event figures. Affected banks are significantly larger with average total assets of 344 Billion Euros than unaffected banks with average total assets of 37 Billion Euros. The affected banks’ pre-event capital ratio is slightly lower than the capital ratio of unaffected banks (10.1% vs 11.6%). While the loans to asset ratio is almost identical for these two groups with about 50%, the unaffected institutions hold on average more deposits (41% compared to 47% deposit to asset ratio). The market value of sovereign bond holdings from GIIPS countries relative to total assets is relatively similar for both types of banks (1.8% and 1.3% for affected and unaffected banks, respectively). We also collect information on banks’ certificates of deposits, commercial paper and other short-term bank notes (bearer debt securities). Affected banks’ short-term funding amounts to 15.6%, while only 10.2% for unaffected banks. Table 1 further illustrates the volumes of banks’ security holdings. The market value of corporate bond holdings in balance sheets is 20.7 billion Euros for affected banks compared to about 5 billion Euros for unaffected banks.

To understand how the EBA capital exercise affected banks’ balance sheets, we compare changes of certain descriptive variables of affected and unaffected banks around the event in Table 1, Panel B. Both types of banks increased their capital in the post-event period. While affected banks increased their capital by 12%, unaffected banks increased their capital holdings by 7.8%. Statistically, this difference is not significant. Affected banks reacted to the EBA exercise by shrinking their total assets by 4.3%. During the same period, unaffected banks increased their total assets by 3%. In sum, the difference between affected and unaffected banks around the regulatory action is about 7.3%. This figure suggests that banks achieved the target core tier 1 ratio primarily by shrinking their assets and not
through raising capital.\textsuperscript{20} In the next step, we provide more direct evidence by constructing market-making measures based on the daily transaction data.

### 3.2 Measuring Market-Making Activities

In this section, we quantify banks’ market-making activities that are based on the daily transaction data set. More specifically, we use the following accepted measures:

- \( \log(\text{net volume traded})_{ijt} \) - Log of net volume traded on trading day \( t \) by bank \( i \) in bond \( j \). This is described as the change in inventory of the bank \( i \) in bond \( j \) from the previous day and can be thought of as deviations or shocks a bank can absorb in its inventory daily. If the net volume change at the end of the day was negative, we take the absolute value of the volume. For example, if a bank sold a bond for 20 euros but bought the same bond for 30 euros on a day, then the change in inventory at the end of the day is \( |30 - 20| = 10 \) euros. If the bank sold a bond for 60 euros and bought one for 40 euros, then the change in inventory at the end of the day is \( |40 - 60| = 20 \) euros. \textit{Bessembinder et al. (2018)} use this measure to provide evidence on dealers’ overnight capital commitment. A decrease in this measure implies less capital commitment by banks.

- \( \log(\text{average trade size})_{ijt} \) - Log of gross volume traded in a day divided by the number of transactions conducted in the day by a bank. Many market participants argue that banks perform smaller trades and provide less immediacy due to regulatory reforms.

- \( \frac{\text{Principal Trading Volume}}{\text{Total Volume}}_{ijt} \) - Banks can act either as an agent (broker) where they match customers for trades or as a principal where they absorb the customer requirements into their own inventory and thereby exposing themselves to risk directly. Within the principal trades, there can also be riskless principal trades (as defined by \textit{Bessembinder et al. (2018)} and \textit{Schultz (2017)}) in which the bank places the securities on its balance sheet only for a short time before it is offset by trades in the opposite direction. In such a case, banks act like a riskless dealer.

\textsuperscript{20}\textit{Admati et al. (2018)} find that banks’ existing shareholders prefer to increase their capital ratios by reducing risk-weighted assets instead of raising new capital.
Since we only focus on balance sheet changing transactions in our sample, all trades are principal trades. However, we rely on the information provided in the dataset, which divides principal trades into riskless principal trades (securities are placed on the balance sheet for a short time) and all others. Please note that instead of using the methodology followed by Bessembinder et al. (2018) of detecting offsetting trades within one minute, we depend on the information provided by the bank to the supervisory authority in the dataset.

We calculate principal trading volume (defined above) as a percentage of the total trading volume.

Panel C of Table 1 presents descriptive statistics of these three measures as well as the daily gross volume of corporate bonds traded. While affected and unaffected banks differ considerably in the daily gross trading volume (33 million Euros vs 6 million Euros), the daily net volume is around 1 million Euros for both types of banks. Interestingly, the average trade size is about twice as high for those banks that did not participate (343k Euros vs 649k Euros). Also, the fraction of transactions undertaken where banks act as a principal and not as a broker is quite similar (around 60%) for both groups of banks.

3.3 Liquidity measures

In order to evaluate whether potential changes in banks’ market-making activities impact the overall liquidity of bond markets, we need to define quantitative estimates of bond market liquidity over time. We employ liquidity measures calculated using transactions data at a weekly frequency (see Gündüz et al. (2018) for a similar approach). This allows us to include bonds that are typically not traded daily. We focus on the following liquidity measures:

- The Amihud measure as developed in Amihud (2002):

  \[
  Amihud_{jt} = \frac{1}{N_{jt}} \sum_{s=1}^{N_{jt}} \frac{|r_{sjt}|}{Vol_{sjt}}
  \]

  where \( N_{jt} \) is the number of trades of bond \( j \) in week \( t \), and \( r_{sjt} \) and \( Vol_{sjt} \) are bond \( j \)'s return and the trading volume of trade \( s \), respectively. The higher the price impact
of a trade of a given size, the higher the Amihud measure is and less liquid the bond. The Amihud measure is provided in units of percentage per million euros and obtained only for bond-week pairs with at least four transactions.

- Imputed Round Trip Cost (IRTC) as proposed in Feldhütter (2011) is a proxy for the bid-ask spread where the bank collects the difference between the highest and the lowest price in a set of transactions of identical volumes (round-trip):

\[
IRTC_{jt} = \frac{1}{R_{jt}} \sum_{s=1}^{R_{jt}} \left( \frac{P_{s_{jt}}^\text{max} - P_{s_{jt}}^\text{min}}{P_{s_{jt}}^\text{max}} \right)
\]

where \( R_{jt} \) is the number of round-trips of bond \( j \) in week \( t \), and \( P_{s_{jt}}^\text{min} \) and \( P_{s_{jt}}^\text{max} \) are the minimum and maximum prices of round trip \( s \), respectively. We need a minimum of four transactions for a bond-week pair, and at least two of them should be of the same size to calculate this measure.

Panel D of Table 1 presents descriptive statistics of the underlying bond level data and the two liquidity measures. On average, the outstanding amount of a bond is 396 million euros, and 2.2 years have passed since the issuance of a bond. The average time to maturity of bonds is 4.1 years. The Amihud measure in units of percentage per million Euros corresponds to a mean of 2.7, and the IRTC measured in percentage is 0.82.

4 Identification Strategy

4.1 Impact of the Capital Exercise on Market Making Activity

We rely on the following difference-in-difference specification to estimate the effect of higher capital requirements on banks’ market-making activities:

\[
Y_{ijt} = \alpha_{ij} + \alpha_{jt} + \delta \times \text{Capital Exercise}_i \times \text{Post}_t + \epsilon_{ijt}
\]

where \( i, j \) and \( t \) index banks, bonds and days, respectively. The main outcome variables \( Y_{ijt} \) are the three market-making measures (e.g., the net volume traded) discussed in Section 3.
The explanatory variable is the interaction of \( \text{Capital Exercise}_i \), a dummy that takes a value of one for affected banks and \( \text{Post}_t \), a dummy variable that takes the value of one for trading days after October 26, 2011 (EBA announcement day). The coefficient \( \delta \) measures the effect of the higher capital requirements on the affected banks relative to the unaffected banks’ market-making activity for the same bond on the same day. The pre-event period is from January 1, 2010, until July 15, 2011, and the post-event period is from October 26, 2011, until December 31, 2012. We exclude the period between the publication of the EBA stress test results on July 15 and the announcement of the EBA capital exercise on October 26, given that a recapitalization of the banking sector has been potentially anticipated during this period as discussed in Section 2.

The Bond \( \times \) Day fixed effects, \( \alpha_{jt} \), control for any time-varying heterogeneity across bonds. To illustrate an example, consider the case in which investors’ demand for a certain bond changes over time. In this case, changes in the trading activity of this bond might be unrelated to our event. The Bank \( \times \) Bond fixed effects, \( \alpha_{ij} \), control for relationship-specific patterns between the bank and the bond. Such relationship-specific patterns emerge if, e.g. a bank underwrites a specific bond and, therefore, trades the specific bond more actively.

We further saturate our main specification with lagged bank-level controls - \( \log(\text{assets}) \), \( \text{Loan/total assets} \) and \( \text{Deposits/total assets} \). In order to allow for potential correlation among bonds traded by the same bank or within the same bond, standard errors are double clustered at the bank and bond level.

In specification (3), unaffected banks might be indirectly impacted by the capital exercise in case they take over trading activities from affected banks. We discuss how such a substitution of activities would impact our estimates in Section 5.

### 4.2 Potential Concerns and Further Empirical Strategies

A fundamental identification assumption behind our empirical strategy is that market-making activities of affected and unaffected banks follow similar patterns in the absence of the capital exercise. One might worry that, despite focusing only on sample banks that actively trade, both types of banks may differ because of their size. Notably, our data incorporates the exact time of every transaction. We can, thus, examine the timing of banks’
reaction to the EBA capital exercise. If affected and unaffected banks’ trading follows different patterns for other reasons than the capital exercise, it is unlikely that this would occur exactly at the same date as the EBA announcement. We, therefore, define a short event window of 30 days before/after the EBA announcement and conduct our main identification strategy for this short time period (see Table 2). Without discussing the empirical results in detail at this stage, all affected banks’ market-making measures respond immediately following the announcement of the capital exercise. This evidence mitigates concerns about confounding factors.

Nevertheless, to sharpen the identification, we exploit cross-sectional variation in the capital requirements of bonds. Market-making activities of lower-rated bonds are likely to be more impacted by the capital exercise, given that these bonds tend to be more illiquid and have higher risk weights associated with them. To reach the mandated core tier 1 ratio, banks reduce inventory more in bonds that have higher risk weights and tie up more capital. Wang and Zhong (2019) provide a theoretical model which predicts that order rejection rate by banks would increase more for higher risk-weighted bonds when regulatory capital requirements increase. We use ratings data from Moody’s. Bonds rated lower than Baa are non-investment grade bonds. We add to the interaction term in specification (3) a dummy, $Non-Investment$ $Grade_j$, that takes the value of 1 if the bond is non-investment grade and zero otherwise. By doing so, we can compare trading activities of two kinds of bonds within the same bank and include Bank × Day fixed effects ($\alpha_{it}$). This systematically controls for any bank-specific shocks. The coefficient of this interaction term measures how affected banks adjust their market-making activities differently for non-investment grade bonds compared to investment-grade bonds that require a lower capital charge.

Similarly, we exploit cross-section variation based on banks’ market-making activities during times of stress. It has been widely discussed in the financial media that liquidity provision has fallen only on days when there is unusually high activity in the market. Bao et al. (2018) study bond liquidity in times of stress after the implementation of the Volcker rule.21

---

21 Governor Jerome H. Powell, Board of Governors of the Federal Reserve System: “It may be also that, even if liquidity is adequate in normal conditions, it has become more fragile, or prone to disappearing under stress.”
As the corporate bond markets tend to experience stress on a limited number of days only, the above observation has also been cited as a reason for not seeing any drop in bond liquidity. A Stress Day for a given bond occurs when the daily trading volume is in its upper 5th percentile across all banks for that month, i.e., the trading day with the highest volume in a month for a particular bond (Stress Day$_{jt}$) and zero otherwise. This indicator variable is substituted for Non-Investment Grade$_j$. The coefficient of interest measures the change in the market-making behaviour of affected banks compared to unaffected banks when customers required more liquidity than usual.

As a final test, we use variation in capital shortfall among the affected group. We compare market-making activities by affected banks in the top tercile according to the capital shortfall (3 banks) with the other 9 affected banks. Note that among the banks selected by the EBA for the capital exercise, high shortfall banks have 357.872 billion in assets and low shortfall banks have 337.638 billion in assets, which suggests that these two groups are quite similar in size. We use an interaction term between the diff-in-diff estimator and a dummy which is equal to 1 if the bank was in the top tercile according to capital shortfall (High Shortfall$_i$) and zero otherwise. The coefficient of interest estimates the relative response to market-making activities of the affected banks that did experience a high capital shortfall compared to those affected banks that did not. Next, we introduce our tests to measure whether the capital exercise impacted aggregate bond market liquidity.

### 4.3 Estimating changes in bond market liquidity

To test whether liquidity has indeed fallen due to a reduction in banks’ market making activity, we estimate the following equation:

$$
liquidity\ measure_{jt} = \alpha_j + Post_t + Controls_{jt} + \epsilon_{jt}
$$

(4)

where $j$ is bond, $t$ is week, and $Post_t$ is a dummy which is equal to one after the announcement and 0 before. We also include the log of bond age (time since issue), the log of bond maturity and the log of the outstanding amount of the bond as control variables.
Time-series regressions provide some evidence on whether the capital exercise played a role in the reduced liquidity in the corporate bond market or not. To control for time-varying economic and market conditions, we divide bonds by the capital exercise’s effect on them. We calculate the share of the trading volume performed by the affected banks in the pre-period for each bond. We classify a bond as more affected if its share of the trading volume by the affected banks is higher than the median for all bonds. We compare these two groups in a diff-in-diff setup, thereby controlling time-varying economic and market-wide conditions. We estimate the following specification:

$$\text{liquidity measure}_{jt} = \alpha_j + \alpha_t + Post_t \times \text{High Share}_j + \text{Controls}_{jt} + \epsilon_{jt}$$ \hspace{1cm} (5)$$

This specification compares liquidity measures for less and more affected bonds before and after the capital exercise.

5 Capital requirements and market making activities

5.1 Evidence from security holdings

We begin by presenting evidence from the Security Holdings Statistics database. The dataset contains detailed security by security information on the entire bank portfolio on a quarterly basis. This security-wise information about holdings of corporate bonds can be thought of as inventory levels. Further, the dataset provides complete information on German banks’ holdings, held anywhere in the world.

Regression results of our main specification are presented in Table 3. We report the diff-in-diff coefficients for different specifications for corporate bond holdings. The coefficient in Column 1 suggests that affected banks reduce their corporate bond holdings by 38.8% relative to unaffected banks. In Column 2, we saturate the specification by adding Security × Quarter fixed effects. Thus, we compare changes in inventory holdings of the same bond of affected and unaffected banks holding constant for overall bond holdings at each point in time. In columns 3 and 4, we replicate this analysis and further control for relationship-specific patterns (Bank × Security fixed effects) and add lagged bank controls.
(i.e. \(\text{Log(assets)}, \text{Loan/total assets and Deposits/total assets}\)). In our most stringent specification, the magnitude of our coefficient is 19.1%.

In Figure 2, we plot the dynamic treatment effects on corporate bond holdings around the announcement of the capital exercise at the quarterly level. The dotted vertical line depicts the quarter immediately before the EBA capital exercise (announced on October 26, 2011). The graph shows that before the announcement of the capital exercise, the corporate bond inventory of affected and unaffected banks follows very similar trends. We observe a reduction of corporate bond inventories by affected banks only after the EBA exercise has been announced.

The above results provide evidence that affected banks have reduced their inventory of corporate bonds in response to higher regulatory capital requirements. In the following subsection, we further examine whether this change resulted in the reduced provision of immediacy to buyers or sellers of corporate bonds in OTC markets.

### 5.2 Market making activity by banks in corporate bonds

In this section, we investigate the impact of the increase in capital charges on banks’ market-making activities. Results from estimating specification 3 for all three market-making measures are presented in Table 4. In Columns 1-4, the logarithm of the net volume traded serves as the left-hand side variable. Affected banks reduced their net volume traded by 9.7% compared to non-affected banks following our event (Column 1). Saturating our specification with time-varying bond specific factors (Column 2) results in a much higher magnitude of the coefficient (23.2%). Once we further include bank \(\times\) bond fixed effects (Column 3), our coefficient increases to 37.0%. In the strictest specification where we include bank controls (\(\text{Log(assets)}, \text{Loan/total assets and Deposits/total assets}\)), affected banks reduce their net volume traded during the day relative to unaffected banks by 38.0% (Column 4).\(^{22}\) The increase in the magnitude of the coefficient illustrates the importance of controlling for changes in the demand for bonds by investors and relationship-specific patterns in bond trading.

\(^{22}\)It is important to highlight that the market-making by the unaffected group did not increase in the post-period. This allays concern that the decrease in market-making by the affected group could have been picked up by the unaffected banks leading to an upward bias in our estimated coefficient.
Looking at aggregate figures instead of micro-level data would result in an underestimation of the impact of higher capital requirements on market making.

Banks commit less capital to corporate bonds after the capital exercise and their end of day inventory change has declined post the exercise. This implies that banks’ ability to absorb changes in their desired inventory levels is reduced if their minimum capital requirements is raised. The dynamic treatment effect of this coefficient is shown in Figure 3. The difference in the net volume traded for affected and not-affected banks is statistically not different from zero until our event. Following the capital exercise announcement, we observe a significant relative reduction in the net volume traded by affected banks in the subsequent four quarters.

In Columns 5-8, we repeat the previous analysis while substituting the log of the average trade size as the dependent variable in specification 3. Average trade size is thought to be an important measure of liquidity provision by the market-maker. Affected banks observe a decrease in average trade size by 3.8% compared to the control group banks (Column 5). Once we focus on relative changes in the average trade size for the same bond, which is traded at least by a single affected bank and by an unaffected bank, i.e. by including Bond × Day fixed effects, the coefficient of interest increases to 12.7% (Column 6). Further controlling for bank × bond fixed effects and adding bank controls, the coefficient further increases to 19.5% (Column 7) and to 20.7% (Column 8), respectively. Dynamic treatment effects for this measure are illustrated in Figure 4. The graph shows a very similar pattern as in Figure 3. The treatment effect is not different from zero until the event and statistically negative in the three quarters following the announcement of the capital exercise.

We next report the results for the third market-making measure - principal trading volume as a proportion of total trading volume - in Columns 9 to 12. Dealer-like trading has reduced for affected banks compared to the control group by 4.1 percentage points (Column 9). Including Bond × Day and Bank × Bond fixed effects and bank controls yields a coefficient of 4.2 percentage points (Column 12). The mean of the pre-event proportion of principal trading by affected banks is 0.596. This implies that banks selected for the capital exercise have observed a decrease in the principal or dealer-like trading by about 7% of the mean of pre-event proportion.
In sum, all three measures exhibit the same pattern. We find that affected banks reduce their net volume traded, average trade size and the proportion of trades where securities are carried on their balance sheets. Given this, the previously documented reduction in banks’ bond holdings implies a reduction not only in proprietary trading but in banks’ market-making activities. This constitutes an unintended consequence of higher capital charges.

The above results also shed light on the ongoing puzzle that liquidity has fallen, but bid-ask spreads have not increased. We find that liquidity provision by banks has declined, but banks’ inventory management tells us why bid-ask spreads have not increased. Banks execute smaller trades that tend to have smaller bid-ask spreads. Banks also pre-arrange more trades and act more like brokers. As shown by Schultz (2017), dealers who pre-arrange more trades bear less risk and earn lower profits due to the provision of less liquidity, which implies smaller bid-ask spreads. Overall, these results indicate that banks provide less liquidity to buyers and sellers of corporate bonds in OTC markets and that the profitability of affected banks’ market-making activities is lower.

5.3 Cross-sectional evidence

Results in the previous section provide strong evidence that affected bank reduced their market-making activities in response to higher capital requirements. In this subsection, we exploit cross-sectional tests that rely on variation in banks’ incentives to reduce specific trading activities in order to comply with the higher capital requirements. More specifically, we test whether, among the affected banks, those that experienced the highest capital shortfall during the exercise adjust their market-making activities more extensively. Furthermore, it is natural to expect that affected banks may refrain from providing immediacy, especially for those bonds that carry a high risk-weight. To examine this relationship, we compare banks’ trading of non-investment grade with investment-grade bonds in response to the event. Finally, we exploit variation in the timing of market-making activities. On stress days, i.e. days, when there is high trading activity, the risk to large inventory holdings at the end of the day is the highest, and, therefore, trading activities for affected banks may be restricted the most on these days. These cross-sectional tests also allow us to learn about the underlying mechanism driving our main empirical findings.
(i) ‘Capital shortfall’ At the bank level, we exploit variation in capital shortfall generated by the exercise among the affected banks. We compare top quartile banks (3 banks) with the highest capital shortfall against the other three quartiles (9 banks), which leaves these two groups very similar in size (as mentioned in Section 4.2). Table 5 provides regression results for the three market-making measures. In Column 1, we include an interaction between the high capital shortfall dummy and a dummy for affected banks. We find that this interaction term is statistically significant. The magnitude of the coefficient implies that the net volume traded by affected banks with a high capital shortfall is reduced by 22.7% compared to affected banks that did not experience a high shortfall. Similarly, banks with a higher capital shortfall reduced their average trade size by 9.1% more than the other affected banks (Column 4). While the interaction term is statistically insignificant for our third measure, the proportion of principal trading, the coefficient points in the same direction, indicating that affected banks with a capital shortfall have adjusted their market-making more drastically (Column 7). In sum, the response to the regulatory announcement is more pronounced among those banks most affected by this event. Given that this effect is identified within the group of affected banks, this test also allays concerns regarding size differences of affected and unaffected banks.

(ii) ‘Non-investment grade bonds’ For the bond-level cross-sectional test, we compare bonds rated non-investment grade by Moody’s versus investment-grade bonds. Ratings are used as a proxy for the risk-weights associated with bonds - the lower the bond’s rating, the higher its risk-weight.\textsuperscript{23} As the denominator of the capital ratio is determined by risk-weighted assets, the impact of the exercise could vary across differently rated bonds. Lower rated bonds also tend to be more illiquid. It is hard to find a trading partner for illiquid corporate bonds, and sometimes, the gap between two such trades can be a few days. Due to higher capital requirements, banks might reduce their trading and inventory of illiquid bonds since they involve added risk which banks cannot afford to take.

We find that affected banks reduced their net volume traded in non-investment grade bonds by 23.0% (see Table 5, Column 2) more than investment-grade bonds. We should

\textsuperscript{23}See Section 2 for a discussion on how banks’ regulatory capital requirements are determined for corporate bond positions.
note that the sample size is reduced since ratings are not available for all bonds in our sample. This cross-sectional test allows us to include Bank × Day fixed effects controlling for time-varying bank-specific shocks. We find similar results once we substitute the dependent variable with the log of the average trade size. The drop in average trade size is 20.0% sharper for non-investment grade bonds (Column 5).

In line with previous results, the principal trading volume as a proportion of total trading volume has decreased by 2.3 percentage points more for non-investment grade bonds than for investment-grade bonds in our strictest specification (Column 8). These findings show that our estimated effects are the strongest for those bonds where small reductions in inventory holdings translate into large reductions in risk-weighted assets.

We plot dynamic treatment effects of average trade size for non-investment grade bonds in Figure 5. Our coefficient of interest remains statistically lower than zero throughout the sample period. This pattern differs from the dynamic treatment effects of all corporate bonds that did reduce at the end of our sample period (see Figures 3 and 4). Adding these two observations suggests that affected banks reduced market-making for all bonds as an immediate response to the capital exercise and then specialized on the market-making for investment-grade bonds (i.e. low risk-weight bonds).

(iii) ‘Stress days’

Lastly, we analyze the market makers’ behavior on days with particular high trading volume, i.e. stress days, around the capital exercise. This cross-sectional test is founded on the hypothesis that liquidity provision has fallen, especially on unusually high market activity days. This also has been cited as a reason why we do not see any drop in bond liquidity since days of stress are few and far between (Bao et al. (2018)).

Stress day is the day on which customers had more demand for the bond (either selling or buying) and sought immediacy. Regression results for this test are shown in Table 5. Affected banks reduce their trading volume on a stress day by 30.7% (Column 3) compared to a normal day. This cross-sectional test also allows us to control for bank-specific time-varying shocks. The wide difference between trading activity on a normal day versus a stress day explains why many academic papers do not find any evidence of a reduction in the overall liquidity of corporate bond markets after the crisis. Firstly, these papers do not
compare a bank that faces a regulatory burden versus a bank that is not governed by the same regulation. Secondly, stress day events are infrequent. Here, we causally document that banks experiencing higher capital requirements provide less liquidity on a stress day. These results point to a possibility that analysts fear: that more regulation will cause market failures in stressful times.\textsuperscript{24}

Once we focus on the average trade size of affected banks, we obtain very similar findings. On a stress day, the average trade size drops by 16.5\% compared to a normal day (Column 6). We did not obtain statistically significant results for the share of principal trading for this test.

Overall, the results in this subsection suggest that our findings are driven by banks’ reaction to the increase in regulatory capital charges. We observe that the reduction in market-making is most pronounced when the marginal costs of the higher capital charges are the highest. The inclusion of Bank × Day interactions implies relying on identification within a bank. This addresses any concerns about time-varying omitted factors that could affect the selection of affected and unaffected banks by the regulator.

5.4 Robustness checks

5.4.1 Sovereign Debt Crisis and Other Related Events

There are two other events besides the capital exercise that occurred during our sample period that might impact banks’ trading and inventory holding. First, during the European sovereign debt crisis, several Eurozone member states (Greece, Ireland, Italy, Portugal and Spain (GIIPS)) were unable to repay or refinance their government debt or to bail out over-indebted banks since 2009. German banks with GIIPS sovereign debt might have shut down market-making activities to economize on regulatory capital to compensate for these losses. Second, German banks dependent on US money market funds for funding experienced significant funding withdrawals, which could have affected bank-affiliated market-making activities. Finally, the Basel III regulatory framework was rolled out during our sample period. We provide an analysis below to rule out the possibility of these events explaining

\textsuperscript{24}See, e.g. Financial Times ”Banks blame bond volatility on tighter regulation” by Tom Braithwaite and Vivianne Rodrigues from October 16\textsuperscript{th} 2014.
our main findings.

In Table 6 we address the concerns raised above. In Columns 1, 4, and 7, we divide our sample banks into quartiles based on their GIIPS exposure/total assets (GIIPS exposure ratio) and, based on this classification, define GIIPS exposure × day fixed effects. Hence, we compare banks with similar exposure in the affected group and unaffected group. In fact, coefficients for net volume traded and average trade size increase. As an additional test, we also use Post × GIIPS exposure as an explanatory variable along with the diff-in-diff estimator in Columns 2, 5, and 8. We find that there is not much change in the magnitude of the diff-in-diff estimator.

Based on data used by Carpinelli and Crosignani (2021), only one bank in our sample had a longer-term refinancing operations (LTRO) uptake; hence LTRO operations seem to not matter for our analysis. The Securities Market Programme was another ECB program launched around the same time, but ECB only bought securities from GIIPS countries during our sample period.

According to Ivashina et al. (2015), those European banks whose funding have some dependency on US money market funds may have experienced the greater withdrawal of funding which would affect their market-making activities. We use bearer debt securities holdings over total assets to measure banks’ dependence on wholesale funding and interact it with the Post dummy. We find that the magnitude of the diff-in-diff coefficient has not changed (see Columns 3, 6, and 9).

Finally, we also consider other potentially confounding regulatory actions that affected German banks during our sample period. The revision of the Basel market risk framework (informally called Basel 2.5) coincided with the EBA capital exercise. It introduced an incremental risk capital (IRC) charge for unsecuritized credit products and stressed the value-at-risk requirement from January 1, 2011. Thus, the only channel through which the IRC affected our sample banks was higher capital requirements. Further, the IRC affected and unaffected banks in the same way.

No other relevant regulatory reforms were implemented during our sample period. More specifically, the EU started with the gradual introduction of Basel III in 2013 (Capital Re-

---

25 Please refer to CGFS (2014) for further details.
quirements Directive IV). Liquidity regulations like the Liquidity Coverage Ratio (LCR) and Net Stable Funding Ratio (NSFR) also came into effect after 2014. Extra capital buffers for systemically important domestic banks were applied starting in 2017. The Markets in Financial Instruments Directive (MiFID) II was finalized in 2014 but has not been implemented yet.

5.4.2 Placebo Test

German government bonds had zero risk-weights at that time: that means an increase in capital requirements should not cause any reduction in banks’ market-making activities. We can use German government bonds as a placebo test by this reasoning.

We test for changes in affected banks’ market making of German government bonds for our three measures in Table 7. There was no change in any measure for affected banks relative to other banks around the event. All presented coefficients are statistically not different from zero, and the sign of the coefficient for the net traded volume and the average trade size has turned positive.

6 Capital Requirements and Corporate Bond Market Liquidity

In this section, we investigate the impact on bond market liquidity when affected banks reduce their market-making activities in response to higher capital requirements. A priori, the impact on overall bond market liquidity will depend on the extent to which unaffected banks and unregulated institutions step in and take over affected banks’ activities. Such unregulated institutions constitute entities with limited regulatory oversight, e.g. shadow banking companies, hedge funds, pension funds. The migration of these activities could potentially take time and, in the short-term, cause market liquidity to deteriorate.

We analyse bond market liquidity using the Amihud ratio and IRTC measures (calculated using transactions data) for German corporate bonds (see Figures 6 and 8). Although our sample period is until the end of 2012, we plot the liquidity measures until the end of 2016.
to assess the impact of the capital exercise for a more extended time series. Both graphs indicate a drastic decrease in liquidity around the capital exercise. Liquidity levels do not seem to revert to their original levels but remain low until the end of our sample period.

Bid-ask spreads did not experience the same trend as corporate bond market liquidity. In Figure 10 we illustrate the time series development of the EUR Markit iBoxx corporate bond index bid-ask spread for our sample period.\textsuperscript{26} While bid-ask spreads did increase somewhat around the capital exercise announcement, we observe a reduction to pre-event levels a few months following the event until the end of our sample. On the one hand, our findings are substantiated by lower corporate bond liquidity, while on the other hand, the bid-ask spreads do not increase. Results from the previous section suggest that banks tend to pre-arrange more trades and act less often as principal. Dealers who act as an agent instead of a principal earn lower profits for a given trade due to lower risk and lower liquidity provision to the customers involved with this trade (Schultz (2017)). Affected banks also decreased average trade size for on-balance sheet trades. Again smaller trades tend to be associated with smaller bid-ask spreads. These trends collectively translate into lower overall market liquidity, while we do not observe an increase in bid-ask spreads.

To formalize our observation, we present regression results of specification 4.3 for our liquidity measures in Tables 8 and 9. The IRTC liquidity measure rises by 0.792 percentage points which is 96.4\% of the mean value of the IRTC measure before the capital exercise (Table 8, Column 1). If bond fixed effects are included (Column 2) in addition to bond controls, the coefficient changes to 0.643 percentage points. The Amihud ratio also increases by 1.647 percentage points per million euros in a before-after capital exercise comparison (Table 9, Column 1). This is 61.9\% of the mean value of the pre-event Amihud ratio. Adding bond fixed effects and bond controls results in a very similar coefficient (Column 2).

Previous analyses on bond market liquidity do not allow us to control any time-varying factors like the sovereign debt crisis or other confounding events. To improve the identification, we suggest a similar diff-in-diff approach for the bond liquidity analysis by dividing our sample bonds into two groups. We calculate the share of trading volume handled by

\textsuperscript{26}The underlying data is taken from the ESMA report (De Renzis et al. (2018)). A similar pattern is shown in Fund (2015), Figure 2.2 (chart 5), for nonfinancial corporate bonds in major European countries.
affected banks averaged over the pre-period for each bond. We classify bonds as ‘more affected’ that had an above median share of trading volume performed by affected banks in the pre-period. We compare these two groups in a diff-in-diff setup, thereby controlling time-varying economic and market-wide conditions.

As a starting point, the IRTC and Amihud ratios are plotted for these two groups in Figures 7 and 9. As mentioned above, bonds with a below median share of trading by affected banks should be affected less if our hypothesis is true. Liquidity of bonds in which the affected group had a higher share of trading volume than the median were affected more. This is true for both the Amihud ratio and the IRTC liquidity measures.

We test this more formally by estimating specification 5 in Columns 3 and 4 of Tables 8 and 9. Corporate bond liquidity decreased following the event since the coefficient of the post dummy is positive but not significant for both Amihud and IRTC measures. Bond liquidity has declined only in bonds where the affected banks played an important role as market makers before the event (Column 3) since the coefficient of dummy Post \times \text{Higher Share} is positive and statistically significant.

This setup allows us to control for time-varying economic and market-wide conditions (Column 4) by including week fixed effects (liquidity measures calculated at a weekly level). Bonds from the affected banks with a higher share of trading exhibit a 0.891 percentage point increase in the IRTC liquidity measure (Table 8, Column 4), which is about a 108% of the pre-event mean value of the IRTC liquidity measure. Similarly, the Amihud ratio (Table 9, Column 4) has increased by 1.895 percentage points per million euros for bonds with a higher pre-event share of affected banks in market-making. This is 71.2% of the pre-event mean value of the Amihud liquidity measure. Large and statistically significant increases in liquidity measures (decrease in liquidity) for the more affected group of bonds indicate that an increase in capital requirements for banks leads to a reduction in corporate bond liquidity due to reduced market making by banks.

Next, we investigate whether non-bank broker-dealers step in and increase their market share as a response to the capital exercise. We analyse the proportion of traded volume performed by non-bank broker-dealers during our sample period. The percentage of trading volume performed by non-bank broker-dealers has increased considerably from 11.6% in
2010 to 28.7% in 2016 (Table 10). The capital exercise was announced on October 26, 2011, and hence, we should see an increase in non-bank activities from 2011 to 2012 and 2013. There is a modest increase from 12.8% in 2011 to 13.7% and 17.3% in 2012 and 2013, respectively. Once we focus on non-investment grade bonds, we observe a stronger pattern. While unregulated institutions had a market share of 9.3% in 2011, this figure increased to 19.5% in 2012. This observation is in line with our previous results that suggest stronger magnitudes for non-investment grade bonds. By the end of 2016, 35.5% of trading in riskier bonds is being conducted by non-bank broker-dealers.

These findings document an unintended consequence of higher regulatory capital requirements in the form of lower bond market liquidity. Even though only the largest banks were affected by the regulatory action, neither the remaining banks nor unregulated institutions could substitute for the reduced market-making activities of affected banks.\textsuperscript{27} The effect on bond market liquidity seems to last at least until the end of our sample period. Lower bond market liquidity is likely to impose significant costs on investors and, hence, the cost of capital of firms. Furthermore, non-regulated institutions also play a more prominent role in the new regime. Given this, the tightening of regulatory capital requirements did contribute to a shift of market-making activities to shadow banking institutions.

It is worth noting that dealers rely on being able to trade with other dealers to be able to provide immediacy. Inter-dealer networks and specialization in certain bonds are difficult to build without prior interactions with other dealers (Hollifield et al. (2012), Di Maggio et al. (2017), Di Maggio (2017), Hendershott et al. (2020)). These mechanisms explain the decrease in liquidity in the more affected bonds and non-bank broker-dealers not being able to fill the gap.

7 Conclusion

In this paper, we examine how the tightening of capital requirements affected corporate bond liquidity. We exploit the EBA conducted capital exercise event and utilized proprietary data

\textsuperscript{27}Similarly, Bessembinder et al. (2018) argue for the US that non-banks fill in the void created by banks only partially due to their small size.
sets from the Deutsche Bundesbank to investigate this question. We find that an increase in capital requirements reduced the market-making activity of banks. We find that an increase in capital requirements not only led to a reduction in the inventory holding of banks but also led to banks pre-arranging more trades (act as broker) and executing smaller size trades. This reduction in the average size of trade at least partially explains why bid-ask spreads have failed to detect a decline in corporate bond liquidity. Furthermore, as expected, we find that the effects are more pronounced for non-investment grade bonds and around stress days (days of unusually high trading demand).

We next examine how non-bank affiliated dealers respond to this event. Interestingly, we find that while the non-bank affiliated dealers increase the share of the market-making activity, they are not able to offset the reduction in market-making of affected banks. Consequently, overall corporate bond liquidity declines.

While the overall effect of a decrease in corporate bond liquidity should translate into an increase in the overall cost of capital of firms, we refrain from making any welfare claims here. Perhaps non-bank affiliated dealers are best suited to do the job of market-making, and migration of this activity to the unregulated sector could be welfare improving. That being said, inter-dealer networks are hard to build and non-bank brokers might not be able to fill the gap due to this friction (see Di Maggio (2017)). Furthermore, migration of market-making to the financial institutions not under the purview of a national regulator could have unpredictable and potentially important adverse consequences for financial stability (see Duffie (2012)). Understanding the boundaries of the bank – what tasks should be carried out inside the bank and what should be done outside it – is an important area for future research.

References


Electronic copy available at: https://ssrn.com/abstract=4250896


Electronic copy available at: https://ssrn.com/abstract=4250896
Figure 1: Timeline of the EBA capital exercise

- 2010
- 26 Oct 2011 Announcement
- minimum CET1 ratio increased to 9% for 12 German banks
- 8 Dec 2011 Publication of bank-by-bank shortfall
- 30 June 2012 Compliance date
- 2012

Figure 2: Changes in corporate bond holdings due to higher capital requirements

This figure plots dynamic treatment effects on net volume traded using Equation 3 around the quarter prior to the announcement of the capital exercise. The vertical line at 2011Q3 marks the quarter immediately before the EBA 2011 Capital Exercise.
This figure plots dynamic treatment effects on net volume traded using Equation 3 around the quarter prior to the announcement of the capital exercise. The vertical line at 2011Q3 marks the quarter immediately before the EBA 2011 Capital Exercise.
This figure plots dynamic treatment effects on average traded size using Equation 3 around the quarter prior to the announcement of the capital exercise. The vertical line at 2011Q3 marks the quarter immediately before the EBA 2011 Capital Exercise.
Figure 5: Impact of higher capital requirements on average trade size for non-investment grade bonds

This figure plots dynamic treatment effects on average traded size for non-investment grade bonds using Equation 3 around the quarter prior to the announcement of the capital exercise. The vertical line at 2011Q3 marks the quarter immediately before the EBA 2011 Capital Exercise.
Figure 6: Time series of IRTC liquidity measure of corporate bonds

This graph shows the time series of mean IRTC plotted weekly. The two dotted vertical lines at July 8, 2011, and October 21, 2011, mark the weeks immediately before the EBA 2011 stress test and the EBA 2011 Capital Exercise, respectively.
This graph shows the time series of mean IRTC plotted weekly of bonds affected differently by the capital exercise due to the share of market making conducted by affected banks before the event. We divide the bonds into low and high share using the median of trading performed by affected banks in the pre-period. The two dotted vertical lines at July 8, 2011, and October 21, 2011, mark the weeks immediately before the EBA 2011 stress test and the EBA 2011 Capital Exercise, respectively.
Figure 8: Time series of Amihud liquidity measure of corporate bonds

This graph shows the time series of the mean Amihud ratio plotted weekly. The two dotted vertical lines at July 8, 2011, and October 21, 2011, mark the weeks immediately before the EBA 2011 stress test and the EBA 2011 capital exercise, respectively.
This graph shows the time series of the mean Amihud ratio plotted weekly of bonds affected differently by the capital exercise due to the share of market making conducted by affected banks. We divide the bonds into low and high share using the median of trading performed by affected banks in the pre-period. The two dotted vertical lines at July 8, 2011, and October 21, 2011, mark the weeks immediately before the EBA 2011 stress test and the EBA 2011 capital exercise, respectively.
The graph plots EUR Markit iBoxx corporate bond index bid-ask spread at a monthly level. The two dotted vertical lines at 2011m6 and 2011m9 mark the months immediately before the EBA 2011 stress test and the EBA 2011 capital exercise, respectively. Source: Chart 20 from De Renzis et al. (2018).
Table 1: Summary statistics

Descriptives are provided before the announcement of the capital exercise. An affected bank is one that took part in the EBA capital exercise. Panel A shows descriptives from security holdings data and bank balance sheet statistics. Panel B provides a means comparison test from before and after the announcement of the Capital Exercise for both sets of banks and Diff-in-Diff coefficient for each variable using the means t-test. Panel C shows descriptives for two sets of banks from transactions data. Panel D presents Bond descriptives from the Centralized Securities Database (CSDB) and liquidity descriptives from transactions data. Note: ** indicates statistical significance at the 5% level and *** at the 1% level.

Panel A: Bank balance sheet and security holdings descriptives

<table>
<thead>
<tr>
<th></th>
<th>Affected banks</th>
<th>Unaffected banks</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>S.D.</td>
</tr>
<tr>
<td>Total assets (billions)</td>
<td>343.807</td>
<td>372.787</td>
</tr>
<tr>
<td>Capital ratio</td>
<td>0.101</td>
<td>0.108</td>
</tr>
<tr>
<td>Loans/TA</td>
<td>0.497</td>
<td>0.188</td>
</tr>
<tr>
<td>Deposits/TA</td>
<td>0.406</td>
<td>0.236</td>
</tr>
<tr>
<td>GIIPS sovereign bond holdings/TA</td>
<td>0.018</td>
<td>0.023</td>
</tr>
<tr>
<td>Bearer debt securities issued/TA</td>
<td>0.156</td>
<td>0.078</td>
</tr>
<tr>
<td>Corporate bond holdings (billions)</td>
<td>20.669</td>
<td>24.014</td>
</tr>
</tbody>
</table>

Panel B: Impact of capital exercise on banks’ balance sheets

<table>
<thead>
<tr>
<th></th>
<th>After - Before</th>
<th>After - Before</th>
<th>Diff-in-Diff</th>
</tr>
</thead>
<tbody>
<tr>
<td>Log(capital)</td>
<td>0.120</td>
<td>0.078**</td>
<td>0.042</td>
</tr>
<tr>
<td>(0.104)</td>
<td>(0.026)</td>
<td>(0.112)</td>
<td></td>
</tr>
<tr>
<td>Log(total assets)</td>
<td>-0.043**</td>
<td>0.030***</td>
<td>-0.073***</td>
</tr>
<tr>
<td>(0.021)</td>
<td>(0.007)</td>
<td>(0.028)</td>
<td></td>
</tr>
</tbody>
</table>

Panel C: Market-Making measures

|                                | Mean  | S.D.     | Mean  | S.D.     |
| Daily net volume (millions)    | 0.940  | 5.828    | 1.045  | 4.239    |
| Daily gross volume (millions)  | 33.341 | 668.495  | 6.347  | 38.486   |
| Average trade size (millions)  | 0.343  | 1.481    | 0.649  | 1.069    |
| Principal volume/gross volume  | 0.596  | 0.322    | 0.614  | 0.326    |

Panel D: Bond and liquidity measures descriptives

|                                | Mean  | S.D.     |
| Amount outstanding (millions)  | 395.671 | 369.274 |
| Age (years)                    | 2.175  | 2.384    |
| Time to maturity (years)       | 4.110  | 7.466    |
| Amihud (percentage per million euros) | 2.660 | 4.711 |
| IRTC (percentage)              | 0.822  | 2.058    |
Table 2: Change in market making measures: one month before and after the announcement of capital exercise

The sample includes corporate non-financial bonds from affected banks and other banks for one month before and after the announcement of the capital exercise. The dependent variables are log(net volume traded), log(average trade size), and principal trading volume as a proportion of total trading volume. The dummy variable $\text{Capital Exercise}_i$ is one for affected banks and zero otherwise, and the dummy variable $\text{Post}_t$ takes the value of one after October 26, 2011, and 0 otherwise. Robust standard errors adjusted for double clustering at the bank and bond level are reported in parentheses. Note: * indicates statistical significance at the 10% level, ** at the 5% level and *** at the 1% level.

<table>
<thead>
<tr>
<th></th>
<th>Log(net volume traded)</th>
<th>Log(average trade size)</th>
<th>Principal trading</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
</tr>
<tr>
<td>Post $\times$ Capital Exercise</td>
<td>-0.457***</td>
<td>-0.235**</td>
<td>-0.059***</td>
</tr>
<tr>
<td></td>
<td>(0.120)</td>
<td>(0.093)</td>
<td>(0.021)</td>
</tr>
<tr>
<td>Adj. R-squared</td>
<td>0.564</td>
<td>0.566</td>
<td>0.656</td>
</tr>
<tr>
<td>Obs.</td>
<td>6,372</td>
<td>6,372</td>
<td>6,372</td>
</tr>
<tr>
<td>Bond $\times$ Day fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank $\times$ Bond fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank Controls</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
</tbody>
</table>
Table 3: Change in security holdings

The sample includes corporate non-financial bonds from affected banks and other banks from 2010 to 2012. The dependent variable in all regressions is log(corporate bond holdings). The dummy variable $\text{Capital Exercise}_i$ is one for affected banks and zero otherwise, and the dummy variable $\text{Post}_t$ takes the value of one after October 26, 2011, and 0 otherwise. Robust standard errors adjusted for double clustering at the bank and bond level are reported in parentheses. Note: * indicates statistical significance at the 10% level, ** at the 5% level and *** at the 1% level.

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Log(corporate bond holdings)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Post × Capital Exercise</td>
<td>-0.388***</td>
<td>-0.254***</td>
<td>-0.224***</td>
<td>-0.191***</td>
</tr>
<tr>
<td></td>
<td>(0.064)</td>
<td>(0.070)</td>
<td>(0.068)</td>
<td>(0.072)</td>
</tr>
<tr>
<td>Adj. R-squared</td>
<td>0.682</td>
<td>0.597</td>
<td>0.829</td>
<td>0.829</td>
</tr>
<tr>
<td>Obs.</td>
<td>20,093</td>
<td>15,376</td>
<td>15,376</td>
<td>15,376</td>
</tr>
<tr>
<td>Security fixed effects</td>
<td>Yes</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Quarter fixed effects</td>
<td>Yes</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Bank fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Security × Quarter fixed effects</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank × Security fixed effects</td>
<td>No</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank Controls</td>
<td>No</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank × Quarter fixed effects</td>
<td>No</td>
<td>No</td>
<td>No</td>
<td>No</td>
</tr>
</tbody>
</table>
Table 4: Change in market making measures

The sample includes corporate non-financial bonds from affected banks and other banks from 2010 to 2012. The dependent variable is log(net volume traded), log(average trade size), and principal trading volume as a proportion of total trading volume in columns 1-4, 5-8, and 9-12, respectively. The dummy variable $Capital\ Exercise_i$ is one for affected banks and zero otherwise, and the dummy variable $Post_t$ takes the value of one after October 26, 2011, and 0 otherwise. Robust standard errors adjusted for double clustering at the bank and bond level are reported in parentheses. Note: * indicates statistical significance at the 10% level, ** at the 5% level and *** at the 1% level.

<table>
<thead>
<tr>
<th></th>
<th>Log(net volume traded)</th>
<th>Log(average trade size)</th>
<th>Principal trading</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
</tr>
<tr>
<td>Post $\times$ Capital Exercise</td>
<td>-0.097***</td>
<td>-0.232***</td>
<td>-0.370***</td>
</tr>
<tr>
<td></td>
<td>(0.028)</td>
<td>(0.043)</td>
<td>(0.054)</td>
</tr>
<tr>
<td>Adj. R-squared</td>
<td>0.429</td>
<td>0.474</td>
<td>0.559</td>
</tr>
<tr>
<td>Obs.</td>
<td>180,108</td>
<td>100,247</td>
<td>100,247</td>
</tr>
<tr>
<td>Bond fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Day fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bond $\times$ Day fixed effects</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank $\times$ Bond fixed effects</td>
<td>No</td>
<td>No</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank Controls</td>
<td>No</td>
<td>No</td>
<td>No</td>
</tr>
</tbody>
</table>

Electronic copy available at: https://ssrn.com/abstract=4250896
Table 5: Change in market-making measures: cross-sectional variation

The sample includes corporate non-financial bonds from affected banks and other banks from 2010 to 2012. The dependent variable is log(net volume traded), log(average trade size), and principal trading volume as a proportion of total trading volume in columns 1-3, 4-6, and 7-9, respectively. The dummy variable $Capital Exercise_i$ is one for affected banks and zero otherwise, and the dummy variable $Post_t$ takes the value of one after October 26, 2011, and 0 otherwise. $High Shortfall$ is a dummy which is 1 for 3 participating banks (top quartile among participating banks) with the highest capital shortfall and 0 for other banks. $Non-Investment Grade$ is a dummy which is 1 for bonds with rating lower than Baa by Moody’s and 0 for other bonds. $Stress day$ is a dummy which is 1 for the trading day with the highest monthly volume for a particular bond. Robust standard errors adjusted for double clustering at the bank and bond level are reported in parentheses. Note: * indicates statistical significance at the 10% level, ** at the 5% level and *** at the 1% level.

<table>
<thead>
<tr>
<th></th>
<th>Log(net volume traded)</th>
<th>Log(average trade size)</th>
<th>Principal trading</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
</tr>
<tr>
<td>Post × Capital Exercise</td>
<td>-0.303***</td>
<td>-0.177***</td>
<td>-0.035***</td>
</tr>
<tr>
<td></td>
<td>(0.062)</td>
<td>(0.032)</td>
<td>(0.007)</td>
</tr>
<tr>
<td>Post × Capital Exercise × High Shortfall</td>
<td>-0.227***</td>
<td>-0.091**</td>
<td>-0.009</td>
</tr>
<tr>
<td></td>
<td>(0.071)</td>
<td>(0.046)</td>
<td>(0.017)</td>
</tr>
<tr>
<td>Post × Capital Exercise × Non-Investment grade</td>
<td>-0.230**</td>
<td>-0.200***</td>
<td>-0.023*</td>
</tr>
<tr>
<td></td>
<td>(0.108)</td>
<td>(0.055)</td>
<td>(0.013)</td>
</tr>
<tr>
<td>Post × Capital Exercise × Stress Day</td>
<td>-0.307***</td>
<td>-0.165***</td>
<td>0.010</td>
</tr>
<tr>
<td></td>
<td>(0.063)</td>
<td>(0.032)</td>
<td>(0.009)</td>
</tr>
<tr>
<td>Capital Exercise × Stress Day</td>
<td>-0.014</td>
<td>0.173</td>
<td>0.030</td>
</tr>
<tr>
<td></td>
<td>(0.240)</td>
<td>(0.137)</td>
<td>(0.032)</td>
</tr>
<tr>
<td>Adj. R-squared</td>
<td>0.559</td>
<td>0.595</td>
<td>0.579</td>
</tr>
<tr>
<td>Obs.</td>
<td>100,247</td>
<td>45,249</td>
<td>100,247</td>
</tr>
<tr>
<td>Bond × Day fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank × Bond fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank × Day fixed effects</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank Controls</td>
<td>Yes</td>
<td>-</td>
<td>-</td>
</tr>
</tbody>
</table>

Electronic copy available at: https://ssrn.com/abstract=4250896
Table 6: Robustness check: sovereign debt crisis

The table presents robustness checks concerning the European sovereign debt crisis. The dummy variable $Capital\ Exercise_i$ is one for affected banks and zero otherwise, and the dummy variable $Post_t$ takes the value of one after October 26, 2011, and 0 otherwise. $GIIPS\ exposure$ is the market value of sovereign bond holdings from GIIPS countries as a percentage of total assets. $Bearer\ Debt\ Securities\ Ratio$ is the value of bearer debt securities issued by the bank as a percentage of total assets. Robust standard errors adjusted for double clustering at the bank and bond level are reported in parentheses. Note: * indicates statistical significance at the 10% level, ** at the 5% level and *** at the 1% level.

<table>
<thead>
<tr>
<th></th>
<th>Log(net volume traded)</th>
<th></th>
<th>Log(average trade size)</th>
<th></th>
<th>Principal trading</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
<td>(5)</td>
</tr>
<tr>
<td>$Post \times Capital\ Exercise$</td>
<td>-0.375***</td>
<td>-0.401***</td>
<td>-0.424***</td>
<td>-0.244***</td>
<td>-0.217***</td>
</tr>
<tr>
<td></td>
<td>(0.056)</td>
<td>(0.055)</td>
<td>(0.059)</td>
<td>(0.031)</td>
<td>(0.028)</td>
</tr>
<tr>
<td>$Post \times GIIPS\ Exposure$</td>
<td>-7.546***</td>
<td>-7.283***</td>
<td>-0.950</td>
<td>-0.958</td>
<td>-0.793***</td>
</tr>
<tr>
<td></td>
<td>(1.243)</td>
<td>(1.236)</td>
<td>(0.600)</td>
<td>(0.602)</td>
<td>(0.146)</td>
</tr>
<tr>
<td>$Post \times Bearer\ Debt\ Securities\ Ratio$</td>
<td>-0.404**</td>
<td>-0.012</td>
<td>0.052***</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.175)</td>
<td>(0.083)</td>
<td>(0.019)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Adj. R-squared</td>
<td>0.566</td>
<td>0.560</td>
<td>0.560</td>
<td>0.583</td>
<td>0.579</td>
</tr>
<tr>
<td>Obs.</td>
<td>100,247</td>
<td>100,247</td>
<td>100,247</td>
<td>100,247</td>
<td>100,247</td>
</tr>
<tr>
<td>Bond × Day fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank × Bond fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank Controls</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>GIIPS Exposure × Day fixed effects</td>
<td>Yes</td>
<td>No</td>
<td>No</td>
<td>Yes</td>
<td>No</td>
</tr>
</tbody>
</table>
Table 7: Placebo test: German government bonds

The table presents a placebo test of the main dependent variables in the German government bond market. The dummy variable $Capital Exercise_i$ is one for affected banks and zero otherwise, and the dummy variable $Post_t$ takes the value of one after October 26, 2011, and 0 otherwise. Robust standard errors adjusted for double clustering at the bank and bond level are reported in parentheses. Note: * indicates statistical significance at the 10% level, ** at the 5% level and *** at the 1% level.

<table>
<thead>
<tr>
<th></th>
<th>Log(net volume traded)</th>
<th>Log(average trade size)</th>
<th>Principal trading</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
</tr>
<tr>
<td>Post $\times$ Capital Exercise</td>
<td>0.039</td>
<td>0.069</td>
<td>-0.018</td>
</tr>
<tr>
<td></td>
<td>(0.136)</td>
<td>(0.175)</td>
<td>(0.023)</td>
</tr>
<tr>
<td>Adj. R-squared</td>
<td>0.431</td>
<td>0.435</td>
<td>0.396</td>
</tr>
<tr>
<td>Obs.</td>
<td>156,836</td>
<td>156,836</td>
<td>156,836</td>
</tr>
<tr>
<td>Bond $\times$ Day fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank $\times$ Bond fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Bank Controls</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
</tbody>
</table>
Table 8: Changes in IRTC liquidity measure after the capital exercise

The table shows how the IRTC liquidity measure changes after the capital exercise. The dependent variable in all regressions is the Imputed Round Trip Cost (IRTC) measured in percentage. Post is a dummy which is 1 for the weeks after the announcement of the capital exercise and 0 before. High Share is a dummy which is 1 for bonds that have an above the median share of trading volume performed by affected banks in the pre-period. Standard errors adjusted for clustering at the bond level are reported in parentheses. Note: * indicates statistical significance at the 10% level, ** at the 5% level and *** at the 1% level.

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Post</td>
<td>0.792***</td>
<td>0.643***</td>
<td>0.257</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.181)</td>
<td>(0.180)</td>
<td>(0.209)</td>
<td></td>
</tr>
<tr>
<td>Post × High Share</td>
<td>0.867**</td>
<td>0.891**</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.399)</td>
<td>(0.400)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Adj. R-squared</td>
<td>0.0304</td>
<td>0.408</td>
<td>0.411</td>
<td>0.419</td>
</tr>
<tr>
<td>Obs.</td>
<td>10,051</td>
<td>10,051</td>
<td>10,051</td>
<td>10,051</td>
</tr>
<tr>
<td>Bond fixed effects</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Week fixed effects</td>
<td>No</td>
<td>No</td>
<td>No</td>
<td>Yes</td>
</tr>
<tr>
<td>Bond Controls</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
</tbody>
</table>

Table 9: Changes in the Amihud liquidity measure after the capital exercise

The table shows how the Amihud liquidity measure changes after the capital exercise. The dependent variable in all regressions is the Amihud measure in units of percentage per million euros. Post is a dummy which is 1 for the weeks after the announcement of the capital exercise and 0 before. High Share is a dummy which is 1 for bonds that have an above the median share of trading volume performed by affected banks in the pre-period. Standard errors adjusted for clustering at the bond level are reported in parentheses. Note: * indicates statistical significance at the 10% level, ** at the 5% level and *** at the 1% level.

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Post</td>
<td>1.647***</td>
<td>1.577***</td>
<td>0.754</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.393)</td>
<td>(0.357)</td>
<td>(0.469)</td>
<td></td>
</tr>
<tr>
<td>Post × High Share</td>
<td>1.835**</td>
<td>1.895**</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.878)</td>
<td>(0.886)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Adj. R-squared</td>
<td>0.0349</td>
<td>0.328</td>
<td>0.331</td>
<td>0.334</td>
</tr>
<tr>
<td>Obs.</td>
<td>11,460</td>
<td>11,460</td>
<td>11,460</td>
<td>11,460</td>
</tr>
<tr>
<td>Bond fixed effects</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Week fixed effects</td>
<td>No</td>
<td>No</td>
<td>No</td>
<td>Yes</td>
</tr>
<tr>
<td>Bond Controls</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
</tbody>
</table>
Table 10: Percentage of trading volume by non-bank broker dealers

The table shows the percentage of trading volume by non-bank broker-dealers from 2010 to 2016.

<table>
<thead>
<tr>
<th>Year</th>
<th>Proportion of overall trading volume</th>
<th>Proportion of trading volume in non-investment grade bonds</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>0.116</td>
<td>0.102</td>
</tr>
<tr>
<td>2011</td>
<td>0.128</td>
<td>0.093</td>
</tr>
<tr>
<td>2012</td>
<td>0.137</td>
<td>0.195</td>
</tr>
<tr>
<td>2013</td>
<td>0.173</td>
<td>0.246</td>
</tr>
<tr>
<td>2014</td>
<td>0.185</td>
<td>0.265</td>
</tr>
<tr>
<td>2015</td>
<td>0.297</td>
<td>0.359</td>
</tr>
<tr>
<td>2016</td>
<td>0.287</td>
<td>0.355</td>
</tr>
</tbody>
</table>