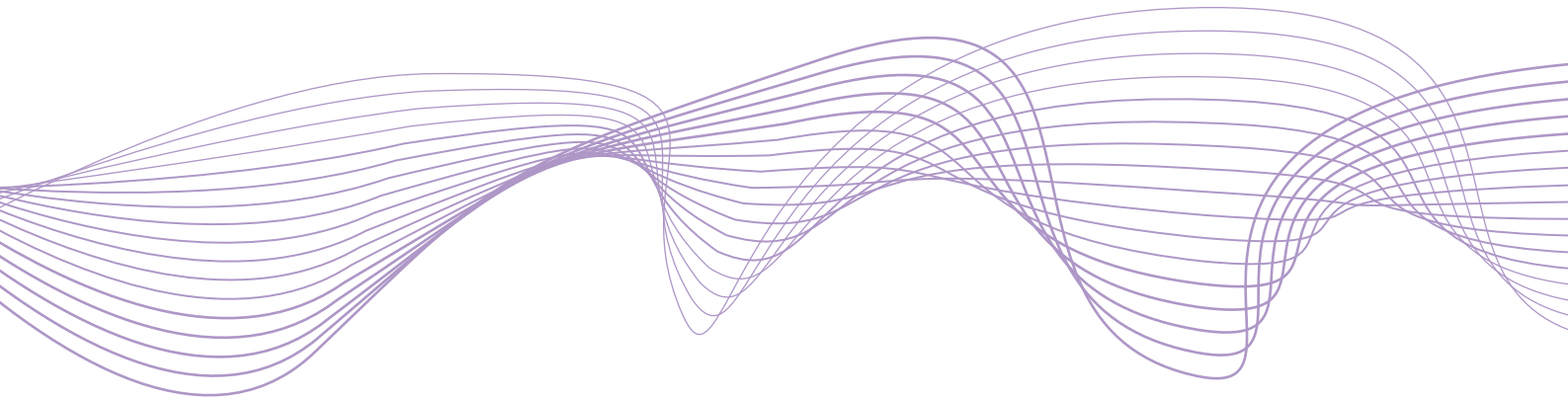


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Cyclical investment behavior across  
financial institutions

by  
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## Abstract

This paper examines the investment behavior of different financial institutions in debt securities with a particular focus on their response to price changes. For identification, we use security-level data from the German Microdatabase Securities Holdings Statistics. Our results suggest that banks and investment funds may destabilize the market by responding in a pro-cyclical manner to price changes. In contrast, insurance companies and pension funds buy securities when their prices fall and vice versa. While investment funds and banks sell securities that are trading at a discount and whose prices are falling, they buy securities that are trading at premium and whose prices are rising. The opposite is the case for insurance companies and pension funds. This counter-cyclical investment behavior of insurance companies and pension funds may stabilize markets whenever prices have been pushed away from fundamentals. Since our results suggest that institutions with impermanent balance sheet characteristics may exacerbate price dynamics, it is of crucial importance for financial stability to monitor the investor base as well as the balance sheets of both levered and non-levered investors.

**Keywords:** Cyclical, Portfolio Allocation, Financial Stability, Debt Capital Flows

**JEL classification:** F32, G11, G15, G20.

# 1 Introduction

Theory yields a variety of predictions on the buying behavior of capital market participants. The standard efficient market hypothesis claims that asset prices must reflect all available information due to the existence of arbitrageurs (Fama, 1965; Friedman, 1953). While banks may be forced to sell undervalued assets due to margin calls, non-levered institutional investors may stabilize the market by buying up fire-sold assets in order to benefit from future price gains (Shleifer and Vishny, 1992). DeLong et al. (1990b) show that it may be rational to buy when prices rise and sell when prices fall so that prices can be pushed away from fundamental values. Despite its importance for macro-prudential policy and financial stability, empirical evidence on who buys when prices are falling has been elusive due to the lack of granular data.

In order to shed more light on the question of how various institutional investors respond to price changes, security-level data is indispensable. For the identification, we use confidential security-by-security holdings data provided by the Deutsche Bundesbank (the German central bank) covering the period from 2005 Q4 until 2014 Q4. For every single security that is held in Germany we have data on the amount held by each sector. For instance, we know that banks in Germany hold an amount  $X$  of security  $Z$  in quarter  $t$ . To the best of my knowledge, this study is the first that uses the security level data of the German Microdatabase Security Holdings statistics for bank and non-bank financial institutions and their investment behavior in debt securities.<sup>1</sup> The holdings include both foreign and domestic as well as government and corporate securities. For the purpose of this study, we focus on the buying behavior of the three largest groups of institutional investors: banks; investment funds; and insurance companies and pension funds. Examining the three sectors jointly is essential because this setting allows us to compare the investment behavior for a homogenous investment horizon. In particular, we can investigate the investment behavior of banks, investment funds and insurance companies and pension funds in the same security at a given point in time. This would not have been possible if we compared the investment behavior between different datasets.

The availability of security-level data also allows us to make comparisons between securities within the same asset class and to observe idiosyncratic price movements. By including security fixed effects, we also control for time-invariant security-specific characteristics and can make judgments about the investment behavior of a specific security over time. Using security\*time fixed effects, we compare the investment behavior by insurance companies and pension funds relative to banks as well as investment funds for a given security at a given point in time. This within security comparison fully absorbs unobserved and observed time-variant security-specific characteristics such as the risk or the liquidity of the security. Hence, the estimated difference of the buying behavior can be attributed to heterogeneity in their response to price changes. Not controlling for these characteristics can lead to biases in the coefficient of the variable of interest when it is correlated with the error term. In addition, grouping securities of specific asset classes together ignores idiosyncratic security characteristics and may lead to misleading results due to compositional effects.

We find evidence that banks as well as investment funds respond pro-cyclically to price

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<sup>1</sup>Buch et al. (2013) and Abbassi et al. (2016) have investigated banks' investment behavior in debt securities. Domanski et al. (2015) use aggregate data for German insurance companies and pension funds.

changes. In contrast, insurance companies and pension funds are contrarian investors, i.e. they buy when prices fall and sell when prices rise. We also present evidence that insurance companies and pension funds have a preference for bonds that are trading at a discount and are falling. They sell bonds that are trading at a premium and are rising. Banks and investment funds buy bonds that are trading at a premium and are rising and they sell when the security is trading at a discount and is falling.

Market participants that, on average, buy when prices rise and sell when prices fall can destabilize the market (Friedman, 1953). This suggests that pro-cyclical investors may exacerbate price dynamics, while counter-cyclical investment behavior pushes prices back towards fundamentals. To the best of my knowledge, this is the first study that compares the investment behavior across sectors and shows who may stabilize the market by acting counter-cyclically.

The empirical approach is to regress the percentage change in the nominal holdings of the debt security of each sector on the lagged percent price change of these securities, controlling for time-invariant security characteristics as well as macroeconomic factors. We find that a ten percent price increase in the last quarter is associated with a 1.7 percent buildup in the nominal value held for both investment funds and banks. If the price of a security drops by ten percent in the previous quarter, insurance companies and pension funds *raise* their nominal amount held by 6.5 percent.

The heterogeneous responses may be explained by differences in their liability structure. Banks and investment funds are vulnerable to runs on their liabilities. This is even the case for mutual funds with small leverage due to the impermanent structure of their equity capital. In addition, their asset side may be relatively illiquid. The liability side of insurance companies and pension funds is more stable and movements in their balance sheets are relatively orthogonal to economic and financial conditions.

The approach brings together two literatures. First, the financial economics literature that focus on the return of these investment strategies neglects financial stability issues. Other studies that focuses predominantly on a single sector, fail to address the counterparts of pro-cyclical investors. For actual buys and sales, there needs to be someone who offsets the pro-cyclical investment behavior, as opposed to order flows, for which there can be a one-sided market of potential buyers and sellers. Easily said, for every buyer there needs to be a seller, and vice versa.

The closest paper to this one is Abbassi et al. (2016), which shows that banks with trading expertise increased their investment in debt securities with falling prices during the crisis relatively more than banks without trading expertise. In contrast to Abbassi et al. (2016), we distinguish the investment behavior of the whole banking sector to non-bank financial institutions, i.e. the investment fund and the insurance company and pension fund sector. In addition, their analysis only sheds light on the relative investment behavior of trading banks versus non-trading banks, but remains silent about whether these institutions *actually* buy or sell. In contrast to Abbassi et al. (2016), we do not only show whether certain sectors act *more* counter-cyclically than others but we also show that insurance companies and pension funds *actually* buy securities when prices fall and sell securities when prices rise. In addition, instead of concentrating only in times of stress, we aim to generalize the cyclical investment behavior across time periods, verifying that it is robust during the crisis. While periods of high stress are certainly crucial for financial stability, normal periods are important to consider because they are the times when

systemic risk builds up.

[Acharya and Steffen \(2015\)](#) show that banks in the Euro Area periphery bought government bonds of Euro Area periphery countries in the first half of 2012 when their yields were high, which indicates counter-cyclical investment behavior. However, it is not clear whether banks started buying when prices were already rising or whether they did so when prices were rock-bottom. In order to clarify this, we delve deeper into the question of whether the banking sector increases its exposure to bonds that have fallen in order to “catch the falling knife” in the hope of mean reversion or if they have jumped on the bandwagon as prices had already started to increase.

We find evidence that banks respond pro-cyclically to price changes. Banks also tend to buy securities that trade at a premium. They increase their holdings more strongly when the price has gone up in the previous quarter and the bond is trading at a premium. This indicates that they are speculating that the price will appreciate further and will sell the security aggressively once it starts decreasing in value.

There is a growing literature investigating the cyclical investment behavior of investment funds. Fund managers may act with a very short-term horizon when exposed to investor injections and redemptions ([Shek et al., 2015](#); [Goldstein et al., 2015](#)). They may also invest pro-cyclically because many are measured on monthly or quarterly performance, adding pressure to chase the market higher as it moves ([Feroli et al., 2014](#); [Shin and Morris, 2015](#); [Abreu and Brunnermeier, 2003](#)). [Brunnermeier and Nagel \(2004\)](#) confirm this finding by investigating the buying behavior of big hedge funds around the dot-com bubble. Hedge funds that were *not* riding the tech bubble underperformed and suffered significant investor redemptions. [Raddatz and Schmukler \(2012\)](#) show that mutual funds’ investment behavior tends to be pro-cyclical and thus not stabilizing; they reduce their exposure in bad times and increase it during good times. Since the pro-cyclicality seems to be existent in both upswings and downturns, delegated portfolio managers may increase market volatility and distort asset prices in general ([Guerrieri and Kondor, 2012](#)).

Our results confirm that investment funds may destabilize the market by acting in a pro-cyclical manner. Although the direction of the cyclical investment behavior has not changed during the crisis, we present evidence that pro-cyclical investment behavior is stronger during the crisis as well as when the implied stock market volatility of the S&P 500 (VIX) is high. A high VIX can be seen as a period of elevated uncertainty and illiquid markets.

There is limited evidence on the investment behavior of insurance companies and pension funds. [Becker and Ivashina \(2015\)](#) explain that insurance companies buy corporate bonds that are the highest yielding within each rating group due to their reluctance to hold more capital when they hold worse-rated bonds. Moreover, some authors have also pointed to the pro-cyclical behavior of insurance companies and pension funds ([Acharya and Morales, 2015](#); [Domanski et al., 2015](#); [Duijm and Steins Bisschop, 2015](#); [Haldane, 2014](#)).

We show that insurance companies and pension funds buy securities when their prices have dropped and sell securities when the price has risen. We also present evidence that insurance companies and pension funds have a preference for buying bonds that are trading at a discount. This supports the hypothesis that they are buy-and-hold investors and that their investment behavior can stabilize the market. Both buying at a discount and selling at a premium may push the price towards its par value as, for a given amount

of securities issued, a higher demand should push up prices.

Second, this paper also contributes to the international economics literature that studies the determinants of portfolio flows. This literature does not differentiate between the holding sector and issuing sector of the securities. In addition, the literature neglects security-specific characteristics, such as price movements but concentrates on country-specific characteristics and global factors (see for example Broner et al. (2013); Forbes and Warnock (2012)). These push and pull factors neglect the investor base of the flows.

First empirical evidence that sensitivity of capital flows can be attributed to the investor base is shown by Cerutti et al. (2015). They demonstrate that capital flows to emerging market countries that rely more on international funds and global banks are more sensitive to global factors. However, they use the correlation of debt capital flows reported by two different data providers as a proxy for the share of bank and investment funds responsible for the capital movements. In contrast, we can distinguish cleanly between the individual holding sectors of securities and thus absorbing compositional effects.

Due to the lack of bilateral data on the sectoral-level link, evidence on heterogeneous responses across investors has been rare. By presenting evidence on the cyclical behavior of different sectors, we aim to fill this gap in the literature. Our results indicate that sudden stops and surges of capital flows may be influenced by the composition of the investor base. By distinguishing between insurance companies and pension funds, banks and investment funds, we find significantly heterogeneous responses to country-specific economic and financial characteristics. Grouping all holding sectors together, the effects may neutralize each other and lead to misleading results. In addition, the granular data allows us to distinguish not only the link between the holding sector and the issuing sector, but also the link at the security level. This information enables us to make statements about security-specific characteristics, absorbing compositional effects. For instance, while it is not possible using aggregated data to establish whether bond investors respond to country-specific or security-specific characteristics, we show that both factors play a vital role. Our results indicate that monitoring the investor base of a security is key to identifying financial vulnerabilities. Relying on pro-cyclical investors such as investment funds and banks can drive prices away from fundamentals and may also lead to sudden stops and surges of capital flows.

The paper is structured as follows. In section 2, we lay out the balance sheet dynamics of the three different sectors. Section 3 describes the data. In section 4, we present some stylized facts. Section 5 shows the regression results. In section 6, we present robustness tests. Section 7 concludes.

## 2 Institutional Background

In order to understand the rationale behind the buying behavior of different financial institution, it is important to understand their balance sheet dynamics.

### 2.1 Banks

Much attention has been paid to the liability side of banks, as the high leverage of global banks can jeopardize the stability of the global financial system. Figure 1 shows different

categories of the aggregated balance sheet of German banks proportionally. The total size of the balance sheets amounted to 7.85 trillion Euros in 2014, which is around 270 per cent of Germany’s GDP (2.9 trillion Euros in 2014). The liability side consists mainly of retail and wholesale deposits. Only 382 billion Euros, approximately 5 per cent, are equity capital. Both retail and interbank borrowing are short-term liabilities that can be withdrawn without an extended period of notice.<sup>2</sup>

Figure 1: Balance Sheet of Banks in Germany

Assets	Liabilities
	Equity
Loans to Non-Banks	Retail Deposits
Loans to Banks	Interbank Borrowing
Debt Securities	Debt Securities Issued
Other	Other

Source: Author’s calculations; Data: Deutsche Bundesbank<sup>3</sup>

When creditors refuse to roll over their debt or actively withdraw their funds, the asset side needs to be reduced in order to service the liabilities. The asset side of banks mainly consists of long-term assets, such as debt securities and loans. When funding liquidity dries up, banks start reducing their most liquid assets, such as cash and excess reserves at the central bank, first. As these contribute only a small amount to the aggregate balance sheet and banks are unable to call in loans, debt securities need to be sold. If the liquidity dryup is systemic and not only specific to one bank, banks may have trouble finding a buyer for the securities, forcing them to sell them below their fundamental value, what is known as a “fire sale”. This process can be even exacerbated if banks need to write their assets down to their fair value. If security prices plunge and banks need to mark them to market, this reduces banks’ equity positions. In order to satisfy capital requirements, they

<sup>2</sup>While in the banking crisis as described in [Diamond and Dybvig \(1983\)](#) retail deposits were withdrawn, the most recent financial crisis was characterized by a withdrawal of wholesale funding and money market fund shares.

<sup>3</sup>Assets (in EUR billions, share of total assets): Loans to Non-Banks (3127, 40%), Loans to Banks (1950, 25%), Debt Securities (1176, 15%), Others (1599, 20%); Liabilities (in EUR billions, share of total liabilities): Equity (382, 5%), Retail Deposits (3299, 42%), Interbank Borrowing (1717, 22%), Debt Securities issued (1115, 14%), Other (1341, 17%); Total: EUR 7853 billion

shrink their balance sheets by selling more debt securities, which depresses their prices even further. This can lead to a spiral between lower asset prices and weaker balance sheets (Adrian and Shin, 2010).<sup>4</sup>

## 2.2 Investment Funds

The investment fund industry in Germany is a significant sector, with an aggregate balance sheet of 1.7 trillion Euros in 2014. In Germany, the sector consists almost exclusively of open-end mutual funds, such as bond and mixed funds.<sup>5</sup> The leverage of these investment funds is very limited. Figure 2 shows that only two percent of their liability side consists of loans. At first glance, the fact that investment funds are not vulnerable to runs on their debt liabilities may raise doubts about their importance to systemic risk. As their investors provide equity capital, this suggests that investment funds can be seen as benign with respect to financial stability.

Figure 2: Balance Sheet of Investment Funds in Germany

Assets	Liabilities
Debt Securities	Investment Fund Shares issued
Equity Securities	
Investment Fund Shares	
Cash and Deposits	
Other	Other

Source: Author's calculations; Data: Deutsche Bundesbank <sup>6</sup>

However, investors in open-end mutual funds can draw down their capital quickly. This changes the assets under management of the fund, which is the fund's equity capital.

<sup>4</sup>Laux and Leuz (2010) describe the mark-to-market behavior of banks in more detail. Allen and Carletti (2008) demonstrate a link between mark-to-market behavior and asset prices.

<sup>5</sup>In 2014 there have been 5,923 investment funds in Germany of which 57.2% are mixed mutual funds and 15% are bond mutual funds. Only 0.5% are hedge funds.

<sup>6</sup>Assets (in EUR billions, share of total assets): Debt Securities (825, 50%), Equity Securities (303, 18%), Investment Fund Shares (277, 17%), Cash and Deposits (70, 4%), Other (179, 11%); Liabilities (in EUR billions, share of total liabilities): Investment Fund Shares issued (1597, 97%), Other (56, 3%); Total: EUR 1653 billion



In other words, investment funds' capital is not permanent, unlike the equity capital of non-financial corporations. As investment fund shares issued make up the lion's share of investment funds' liabilities, looking at simple metrics like the total assets to equity ratio can lead to misleading conclusions when it comes to identifying financial vulnerabilities. Once investors start redeeming assets, a feedback loop between redemptions by investors and sales of portfolio managers can emerge. The redemptions of investors are usually not orthogonal to the state of the real economy. They withdraw capital in times when the economy is doing badly in order to smooth consumption. Portfolio managers' fire sales can drive down prices further, affecting both the economy and investors' balance sheets adversely. Accordingly, this may trigger more redemptions of investors.

### 2.3 Insurance Companies and Pension Funds

The total size of the insurance companies and pension funds balance sheet in Germany in 2014 was 2.4 trillion Euros. On the asset side, cash and deposit holdings are much bigger than for banks and contribute 21 per cent to total assets, while almost 60 per cent are securities (Figure 3). The leverage ratio of insurance companies is much smaller than that of banks. The lion's share of the liabilities is represented by insurance technical reserves; these are net equity of households in life insurance and pension fund reserves or prepayments of insurance premiums and reserves for outstanding claims. These long-term liabilities are mostly contingent and their payouts are relatively independent of the state of the real economy and of overall financial conditions. This predictable liability structure may give insurance companies and pension funds more autonomy in their portfolio choice during bad times compared to banks or investment funds. For instance, an accident with an insured car, a damage to an insured building or a death of a person are events that could be covered under an insurance and cause payouts. As the structure of the liability side of insurance companies and pension funds is relatively persistent, this keeps their funding and rollover risk relatively moderate and leaves them with more "skin in the game".<sup>7</sup> This enables "deep pocket investors", such as insurance companies and pension funds, to take more risk during bad times when other actors, such as banks and investment funds, may be forced to sell.

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<sup>7</sup>Acharya et al. (2011) discuss the systemic importance of insurance companies for the global economy in more detail. Manconi et al. (2016) document their selling behavior when they face a large outflow.

Figure 3: Balance Sheet of Insurance Companies and Pension Funds in Germany

Assets	Liabilities
Equity Securities and Investment Fund Shares	Equity
Cash and Deposits	Net Equity of Household in Life Insurance and Pension Funds
Debt Securities	
Loans	Unearned Premiums and Reserves for outstanding Claims
Other	Other

Source: Author's calculations; Data: Deutsche Bundesbank <sup>8</sup>

### 3 Data

The Microdatabase Securities Holding Statistics of the Deutsche Bundesbank's Research Data and Service Centre of the Deutsche Bundesbank provides data on the holdings of all sectors in Germany separately at the security-by-security level from 2005 Q4 onwards. For instance, we know that the banking sector in Germany holds a specific amount of a specific security in a given quarter. The security is identified with the International Security Identification Number (ISIN). We also have information about the currency of denomination, the security classification and the issuing sector. For a detailed data description see, [Amann et al. \(2012\)](#).<sup>9</sup>

We only consider the three holding sectors: insurance companies and pension funds; investment funds; and bank and their respective holdings of *debt* securities. The raw, nominal and market values are known for debt securities held. The raw value is the nominal value held in the currency of denomination. The nominal value is the notional amount of security holdings and does not reflect price movements.<sup>10</sup> The market value is

<sup>8</sup>Assets (in EUR billions, share of total assets): Investment Fund Shares and Equity Securities (1014, 42%), Cash and Deposits (384, 21%), Debt Securities (384, 16%), Loans (299, 12%), Other (209, 9%); Liabilities (in EUR billions, share of total liabilities): Equity (361, 15%), Net Equity of Household in Life Insurance and Pension Funds (1592, 66%), Unearned Premiums and Reserves for outstanding Claims (296, 12%), Other (90, 3%) Total: EUR 2428 billion

<sup>9</sup>Unfortunately, information on security-specific characteristics from the Centralised Security Database is not available.

<sup>10</sup>The nominal value needs to be adjusted to reflect only investment decisions (see Appendix).

the number of securities held, multiplied by the price.

The data provided by the Deutsche Bundesbank is merged with publicly available data. The country-specific 10-year generic government bond yield, the consumer price index and GDP are from the IMF. We obtain the GDP growth and the inflation rate by taking the natural log change of the GDP and the consumer price index. If the GDP is not available quarterly, we interpolate the annual value linearly. The VIX is from the Chicago Board Options Exchange and downloaded through Datastream. The EONIA is from the ECB.<sup>11</sup> The country-specific variables are merged with the first two characters of the ISIN code. This is consistent with the nationality and not the residence principle and accounts for offshore issuance of securities.<sup>12</sup>

## 4 Stylized Facts

In this section, we show summary statistics of the investment behavior across the financial institutions and their gains. [Table 1](#) shows the summary statistics of our main variables. The average value of a security held is 56 million Euros for banks, 24 million Euros for investment funds and 23 million Euros for insurance companies and pension funds. Insurance companies and pension funds, which hold a significantly smaller quantity of securities, are therefore the smallest holders of debt securities among the three sectors. Insurance companies and pension funds not only hold fewer securities, they also trade less. When they do trade, they transact larger volumes than investment funds. Investment funds are the most active traders among the three. On average, the amounts they trade are smaller than those of banks and insurance companies and pension funds. They also sell more often than they buy, but if they buy, their purchases far outstrip their sales.

[Figure 4](#) compares the holdings of debt securities of the three sectors over time. We can see that banks are the largest holder of debt securities, followed by investment funds and insurance companies and pension funds. These three sectors are the three largest holders of debt securities in Germany. While banks increased their security holdings before the beginning of the financial crisis, they reduced their security holdings significantly since then. In contrast, non-bank financial institutions, such as investment funds and insurance companies, gained more importance in the provision of market-based funding to the real economy. Against the trend, however, investment funds were selling their securities during the sovereign debt crisis. In contrast, insurance companies and pension funds have kept building up debt securities even between 2010 and 2012.<sup>13</sup>

The active selling behavior of banks and investment funds paid off in the short run, as we can see from [Figure 5](#). Their unrealized gains on their debt security portfolio were positive before they dropped into negative territory in mid-2010, but still with no big losses compared to the pre-crisis period. Insurance companies and pension funds, however, suffered severely when their bonds fell in value during the crisis, but their long-term strategy paid off when prices started to recover. Between mid-2011 and the end

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<sup>11</sup>All variables are trimmed on a 0.5 percent level.

<sup>12</sup>For instance, if Petrobras Global Finance issues a bond in the Netherlands, we assign the country-specific conditions to Brazil and not to the Netherlands, as the ultimate risk is located in Brazil.

<sup>13</sup>For the portfolio composition of the three sectors see [Table A1](#).

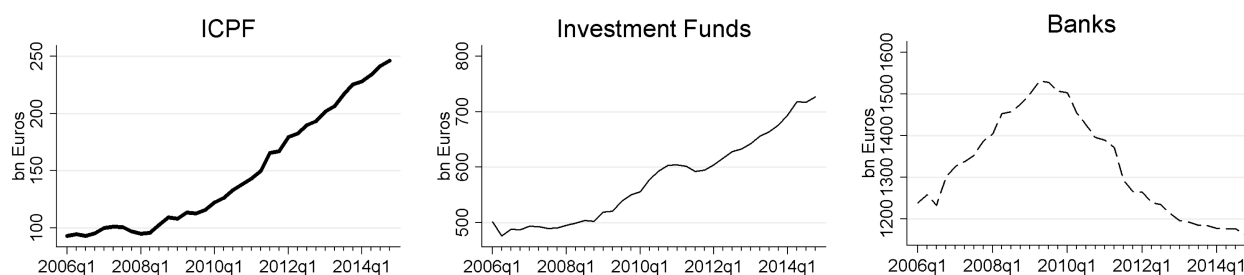
<sup>14</sup>The three panels show the nominal value held by insurance companies and pension funds (ICPF), investment funds and banks.

Table 1: Summary Statistics

Variable	Mean	Std. Dev.	N
Nominal Value Funds	24.48	102.10	562978
Nominal Value ICPF	23.22	447.00	165670
Nominal Value Banks	56.13	262.95	608263
Buy Funds	8.372	35.41	110587
Buy ICPF	12.39	35.53	19955
Buy Banks	16.32	100.89	91278
Sell Funds	7.54	30.64	125753
Sell ICPF	11.57	34.62	16868
Sell Banks	19.51	147.90	85845

Buy and sell refers to the amount bought and sold in million Euros. The nominal value is the nominal value held if a security is held in million Euros. Funds, ICPF and Banks refers to investment funds (Funds), insurance companies and pension funds (ICPF) and banks respectively. Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2004 Q4 - 2014 Q4; author's calculations.

Figure 4: Nominal Debt Security Holdings



Source: Author's calculations; Data: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4.<sup>14</sup>

of 2014 their unrealized gains on their debt securities was nearly 30 per cent. They outperformed banks and investment funds not only since mid-2010, but also since the beginning of the financial crisis. While insurance companies and pension funds kept buying securities during the crisis, temporarily suffering losses, they outperformed the other two sectors in the medium run. The investment behavior of banks, reducing their security holdings even more than investment funds, underperformed both the strategies of investment funds and insurance companies and pension funds. This is in line with the statement of Italy's prime minister, Matteo Renzi, to the Italian Senate on February 17, 2016:

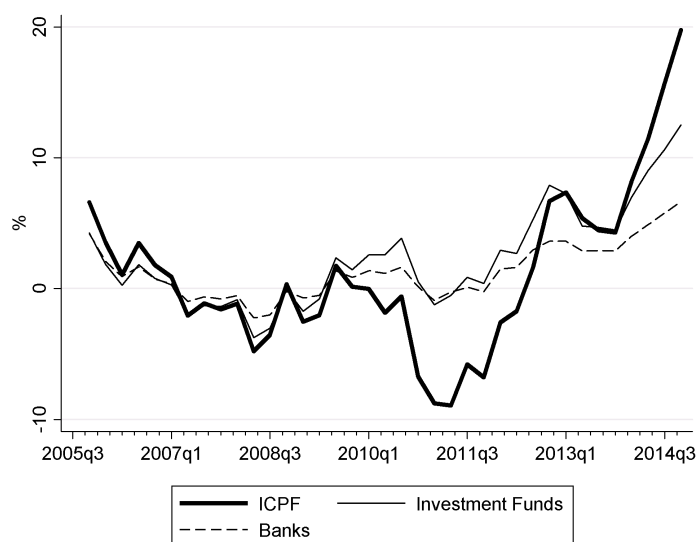
“Let me say that if some northern European lenders had kept their Italian government debt in 2011-2012, they would be earning much more.”

This raises the question whether the insurance company and pension funds sector systematically steps in when other sectors are selling their securities and prices drop.

Buying when prices have fallen is one type of search for yield, as yields rise when prices fall. Insurance companies and pension funds could have been incentivized by the minimum guarantee on their liabilities that forces them to seek yield. However, even if it is well known among potential investors that prices mean revert to their fundamentals, it is a risky decision to “catch the falling knife,” especially if they are measured on their short-term performance.

The stylized facts presented in this section only show simple aggregate numbers that can be influenced by other factors in a number of ways. In order to find out more about the systematic investment behavior of the different sectors, conditional on other characteristics, we need to regress the buying behavior on security-specific, macroeconomic and financial factors. This is done in a regression analysis in the following section.

Figure 5: Cumulative Valuation Effects of Security Holdings



Source: Author’s calculations; Data: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4.<sup>15</sup>

<sup>15</sup>The cumulative gains are calculated as the difference between the total market value of all securities and the total nominal value of all securities divided by the total nominal value of all securities.

## 5 Results

In this section we empirically investigate the cyclicity of the investment behavior of the following three sectors: insurance companies and pension funds; investment funds; and banks. We attempt to shed light on the question as to whether institutional investors respond pro-cyclically or counter-cyclically to price changes.

The regression specification is in the spirit of [Abbassi et al. \(2016\)](#) who regress the amount bought and sold on the lagged price change. This approach may enable us to identify positive feedback investors and rational arbitrageurs. Positive feedback investors “buy securities when prices rise and sell when prices fall” ([DeLong et al., 1990b](#)). “[R]ational arbitrageurs who trade against them [...] drive prices close to fundamental values” ([DeLong et al., 1990a](#)). While [Abbassi et al. \(2016\)](#) differentiate between different types of banks, we distinguish the whole banking sector from the investment fund and the insurance company and pension fund sector by estimating the following regression sector by sector:

$$Netbuy_{s,t} = \beta \Delta Price_{s,t-1} + \gamma' X_{j,t-1} + \phi' Z_{t-1} + \alpha_s + \epsilon_{s,t} \quad (1)$$

Netbuy is the log change in the nominal amount held of security  $s$  at quarter  $t$  if this amount changes.<sup>16</sup>  $\Delta Price$  is the log change in the price of the security.<sup>17</sup> The vector  $X$  includes the variables *Gov\_Bond\_Yield*,  $\pi$ , and  $\Delta GDP$ , which are the 10-year country-specific government bond yield, the quarterly inflation rate, and the log change in GDP of issuing country  $j$ , respectively. The vector  $Z$  includes the *VIX* and the *EONIA*, which do not vary by security but over time. The *VIX* is the log of the end of period implied volatility of the S&P 500 and the *EONIA* is the Euro OverNight Index Average.  $\alpha_s$  is a security fixed effect that controls for security-specific characteristics that are time invariant, like the expiration date or the coupon. This also enables us to analyze the investment behavior in one specific security over time, which circumvents the issue that the amount of securities outstanding in the economy can change.<sup>18</sup>

All variables are lagged by one quarter as the information about the independent variables may not be available when the trading decision is taken. In addition, it can help to circumvent endogeneity problems. For instance, trading decisions may have a price impact so that the variables netbuy and the change in the price may be jointly determined if they enter the regression contemporaneously. While the lagged price change is observed at the end of the quarter, the trading decision can be executed at any time during the quarter. Unless the trading decision is adjusted at the last point of the quarter, the contemporaneous independent variables may be observed only *after* the decision to transact is taken.

It is likely that price changes reflect a fundamental and a transitory component. Ac-

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<sup>16</sup>The netbuy measure reflects only buy and sell decisions and no valuation effects. The results are robust to the use of other netbuy measures. For instance, the results do not change qualitatively whether we use the log of the amount bought minus the log of the amount sold or the amount in Euros. The results are also robust when we use buy and sell separately instead of using a netbuy measure.

<sup>17</sup>The results are robust to the inclusion of higher lags of the price change as well as price changes of a lower frequency.

<sup>18</sup>See appendix for details.

According to [Cutler et al. \(1990\)](#), the fundamental component follows a random walk whereas the transitory component follows a first-order autoregressive process that is likely to be driven by a dominance of noise traders who overreact to fundamental news. In the absence of noise traders, we would expect portfolio weights to be constant ([Milgrom and Stokey, 1982](#)). However, when noise traders are active or markets are not perfectly efficient, we may be able to identify positive feedback investors with a positive demand elasticity to price changes and counter-cyclical investors with a negative demand elasticity to price changes.

The coefficient on the government bond yield shows how sensitive the three sectors are to country risk. A positive sign shows that sectors buy debt securities of countries that are riskier. Inflation and GDP growth rate control for macroeconomic conditions of the country of issue. Low inflation and high GDP growth rate may indicate that bond prices will appreciate in the future, as high inflation distorts the real return of the bond and high GDP growth suggests that countries are going to have lower corporate and government bond spreads, and thus higher prices, in the future. The EONIA shows whether investors prefer building up securities during easy funding conditions.

[Table 2](#) shows the estimation of equation (1) sector by sector. Investment funds and banks buy securities whose prices have risen and sell securities that have lost value, i.e. they have an upward sloping demand curve. In contrast, insurance companies and pension funds buy when prices have fallen and sell when prices have risen.

In terms of economic magnitude, if the price has increased by 10 percent in the previous quarter, both the banking and investment fund sector increases their nominal amount held of this security, on average, by 1.7 per cent. The insurance company and pension fund sector increases its amount held of this security, on average, by 6.5 per cent if the price has *dropped* by 10 per cent in the previous quarter. These effects are statistically and economically highly significant. A 1.7 per cent increase in the holding of the security equals, on average, a EUR 954,210 increase in the holding of this security for banks and EUR 416,160 for investment funds. For insurance companies and pension funds, a 6.5 per cent increase equals EUR 1,509,300. This counter-cyclical behavior more than offsets the pro-cyclical behavior of banks and investment funds.

The pro-cyclical investment behavior of banks and investment funds can be explained by their unstable balance sheet composition. While most investment firms are delegated by investors and have performance pressure so that they have to sell bonds when investors redeem shares, banks need to sell assets when they face a funding squeeze. The effects can be amplified through the asset side of the balance sheet, resulting in a feedback loop between lower prices and sales if creditors are inclined to withdraw their assets when prices are falling. In contrast, insurance companies and pension funds do not have to sell when prices drop, as they are not exposed to redemption pressure and benefit from a more stable balance sheet structure. In addition, insurance companies and pension funds may also not mark-to-market as frequently ([Fabozzi, 2012](#)). These “deep pocket investors” can buy securities with high yields to maturity when prices have dropped, and they can benefit from price increases when the bonds have been traded at a discount and they hold them until maturity. This form of yield-seeking behavior of insurance companies is also documented by [Becker and Ivashina \(2015\)](#). This search for yield is benign as long as it does not lead to capital misallocation.

From a financial stability perspective, these results are highly important for all types

Table 2: Baseline

	(1)	(2)	(3)
	Funds	ICPF	Banks
$\Delta$ Price	0.174*** (0.023)	-0.650*** (0.124)	0.169*** (0.063)
Gov Bond Yield	-0.144 (0.179)	4.633*** (0.857)	0.807** (0.396)
$\pi$	-0.846*** (0.227)	1.983* (1.141)	0.0347 (0.570)
VIX	0.00265 (0.004)	0.0703*** (0.015)	-0.00499 (0.008)
$\Delta$ GDP	1.440*** (0.248)	4.013*** (0.867)	0.483 (0.408)
EONIA	0.967*** (0.115)	-1.164** (0.519)	0.181 (0.225)
$R^2$	0.139	0.178	0.120
N	263612	28096	134005
Security FE	Yes	Yes	Yes

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks. All independent variables are lagged by one quarter.  $\Delta$  Price is the percentage change in the price. Gov Bond Yield is the 10-year generic government bond yield.  $\pi$  is the quarterly inflation rate. VIX is the log of the implied volatility for S&P 500 stock options.  $\Delta$  GDP is the quarterly GDP growth. EONIA is the Euro Overnight Index Average. Standard errors are in parentheses and clustered at the security level. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.



of issuers. If the investor base of the security is skewed towards short-term investors that act pro-cyclically, debtors need to be aware of sharp price drops and heightened volatility of their bonds that can worsen their funding conditions or prevent them from access to capital market funding in the first place. These stops and surges of capital flows can have severe consequences for the real economy (Forbes and Warnock, 2012; Lane and McQuade, 2014). A recent prominent example of this is Greece during its sovereign debt crisis that started in 2010, when banks held a significant amount of the debt outstanding and sold it aggressively once conditions worsened.

In addition to the lagged price change, it is also important to include global and country-specific variables, as the investment behavior is driven not only by security-specific characteristics but also by macroeconomic factors. While the VIX and the EONIA are the same for every security in a given quarter, the government bond yield, the inflation rate and the GDP growth rate are assigned according to the nationality of the issuer of the debt security. If these factors are correlated with the price change, and we exclude them from the regression, this can lead to a biased coefficient on the price change.<sup>19</sup> The VIX and the generic government bond yield of a country generally rise when prices drop and vice versa. Similarly, since inflation erodes the purchasing power of a standard debt security, we would expect prices to drop when inflation rises. A higher GDP growth is usually associated with higher prices in the bond market. As a lower EONIA intends to ease the financing conditions for the whole economy, institutional investors are expected to buy more due to easier financing conditions.

All the three sectors tend to buy more debt securities from issuers originating from countries that have recently experienced higher GDP growth. The intended effect of a low EONIA, to push down market interest rates, is taken over by insurance companies and pension funds, but not by banks and investment funds. The EONIA can also be seen as a proxy for how strong the ECB expects the Euro Area economy to be in the future, i.e. if it raises rates, it expects the economy to do well. The positive association between the EONIA and buying behavior of investment funds and banks may indicate that investment funds and banks believe the ECB's negative outlook when the EONIA is lower, and they are hesitant to buy.

In contrast to Boermans et al. (2016) we do not find that banks increase and investment funds decrease their debt security holdings in times of high stress, i.e. when the VIX is high. Insurance companies and pension funds increase their debt security holdings when the VIX is high. The reaction of investment funds with respect to government bond yields, which proxy a country risk factor, is consistent with Raddatz and Schmukler (2012) who show that mutual funds retrench from countries in bad times. We can also confirm the finding by Buch et al. (2013) who show that German banks hold more sovereign bonds in high-yield and low-inflation countries. Banks that increase their holdings of securities in risky countries could be a case of "gambling for resurrection", when investors are willing to take high risk, hoping for a good outcome (Acharya et al., 2014; Battistini et al., 2014; Bolton and Jeanne, 2011). The risk-taking behavior of insurance companies and pension funds with respect to country risk is broadly consistent with Becker and Ivashina (2015) who show that insurance companies and pension funds buy the highest-yielding assets within each rating group.

To test the sensitivity of the price change coefficient to the inclusion of further controls,

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<sup>19</sup>In Table 3 we confirm that the results hold if we do not control for these factors.

Table 3 shows a summary of the lagged price change coefficients for various specifications. Controlling for more unobserved and observed characteristics also indicates whether the sectors respond to relative price changes of the debt securities or whether the investment decision is driven by broad market valuations. Creating a more coherent sample across the sectors sheds light on the question as to whether the coefficients are driven by a sample selection bias. The coefficient is consistently positive for investment funds and banks and negative for insurance companies and pension. Row (1) is the result of a simple regression of the netbuy variable on the lagged price change excluding macroeconomic factors as well as security fixed effects.

It explains not only the time-series variation but also the cross-sectional variation. Including security fixed effects controls for all time-invariant security-specific characteristics, such as the coupon or the maturity date, but of course also for the issuing country of the security. The approach using security fixed effects focuses on one specific security and attempts to explain the buying and selling behavior over time. Both regressions indicate that, unconditional and conditional on time invariant security characteristics, banks and investment funds respond pro-cyclically to price changes, while insurance companies and pension funds act counter-cyclically.

While row (3) shows the estimates of the baseline regression, row (4) also absorbs observed and unobserved country-specific time-varying characteristics.<sup>20</sup> In order to examine how financial institutions invest in specific securities compared to other securities that were issued in the same sector of the same country, the specification is saturated with *issuing\_sector\*country\*time* fixed effects. This controls for unobserved and observed time-varying heterogeneity, such as the time-varying common component of a specific asset class. In particular, it adds the issuing sector dimension for banks, other financial corporations, non-financial corporations, and governments in their capacity as issuing sectors. Hence, for each issuing sector of a given country we control for the average amount bought or sold at a given point in time, which allows us to control for broad market valuations of this index. Even within this benchmark banks and investment funds buy securities that have increased in value. However, while for investment funds and banks the coefficients are even higher than in specification (4) the coefficient for insurance companies and pension funds is not significant anymore. This indicates that insurance companies and pension funds tend to buy securities that are included in a falling index. In contrast, banks' and investment funds' pro-cyclical investment behavior is also driven by idiosyncratic movements of the security compared to its benchmark.<sup>21</sup> These results also shed some light on the investment strategies the three sectors follow. A negative sign identifies contrarian investors that follow a value investing strategy by buying cheap and selling high, speculating on mean reversion. In contrast, momentum investors buy securities that have performed well recently (see for example Moskowitz et al. (2012) and references therein). Hence, our results suggest that banks and investment funds follow a momentum strategy. However, since there can be much heterogeneity within each sector, institutional-level data is needed to shed more light on this question.

To make the sample of securities held more comparable, row (6) restricts the security sample to all securities that have been held by insurance companies and pension funds

<sup>20</sup>The results of row (3) differ slightly from Table 2 due to a more restricted sample in Table 3.

<sup>21</sup>These results can be confirmed in Table A2, where the price change is decomposed into a broad market valuation of the issuing sector-country index and an idiosyncratic part.

at least once throughout the sample. The sample of row (7) includes securities that have been held at least once by all sectors.<sup>22</sup>

Table 3: Summary of Price Change Coefficients

Security FE	Other Characteristics	Funds	ICPF	Banks
(1) No	-	0.178*** (0.019)	-0.280** (0.118)	0.160*** (0.057)
(2) Yes	-	0.124*** (0.022)	-0.767*** (0.123)	0.155** (0.062)
(3) Yes	Macro Controls	0.175*** (0.022)	-0.680*** (0.122)	0.179*** (0.063)
(4) Yes	Country*Time FE	0.159*** (0.024)	-0.365*** (0.140)	0.154** (0.066)
(5) Yes	Country*Issuing Sector* Time FE	0.178*** (0.025)	-0.233 (0.152)	0.188*** (0.067)
(6) Yes	Sample of securities once held by ICPF	0.106** (0.041)	-0.672*** (0.099)	0.199* (0.106)
(7) Yes	Sample of securities once held by all sectors	0.084* (0.057)	-0.536*** (0.107)	0.254** (0.109)

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks. The coefficients are the estimated effect of a price change in the previous quarter. Macro Controls include the 10-year generic government bond yield, the quarterly inflation rate, the log of the VIX, the quarterly GDP growth and the EONIA. Standard errors in parentheses and clustered at the security level for specifications (2)-(7). For each sector the number of observations is the same in specifications (1)-(5). \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

Until now, we cannot rule out that our results are driven by time-varying security characteristics. For instance, a positive correlation between the error term and the change in the price leads to an overestimation of the price change coefficient. Grouping the three sectors together and comparing them against a benchmark sector allows use to control for unobserved and observed time-varying security characteristics as well as sector-specific characteristics that do and do not vary over time. While estimating the sectors separately allows us to draw inferences about whether sectors trade pro or counter-cyclically with respect to price changes, in a specification with *security \* time* fixed effects we can only make statements about whether the sectors trade more or less pro or counter-cyclically to price changes relative to other sectors. However, this specification allows to draw conclusions about the investment behavior of one specific security at a given point in time.

<sup>22</sup>While in Table 3 the number of observations are consistent for each sector in rows (1)-(5), Table A3 presents evidence that the results also hold for a balanced panel where the observations are also the same across the holding sectors.

Both banks and investment funds invest more pro-cyclically than the benchmark insurance companies and pension funds when all time-varying characteristics of the securities are taken into account (Table 4). By controlling for *issuing\_sector\*time* fixed effects, we can confirm that this is even the case if we control for the amount invested of the specific sector at a given time. This holds for the sample of all securities and only the securities that are traded by all three sectors at a given point in time. The latter indicates that insurance companies buy securities from banks and investment funds when their prices have dropped and vice versa. However, reliable conclusions about who trades with whom are not possible without bilateral trade data.

Table 4: Time-Varying Security Heterogeneity

	(1)	(2)	(3)	(4)	(5)	(6)
	Netbuy	Netbuy	Netbuy	Netbuy	Netbuy	Netbuy
$\Delta$ Price*Banks	0.0421*** (0.010)	0.0424*** (0.009)	0.0497*** (0.010)	0.684*** (0.184)	0.732*** (0.184)	0.928*** (0.199)
$\Delta$ Price*Funds	0.0924*** (0.009)	0.0937*** (0.009)	0.0779*** (0.010)	0.333*** (0.127)	0.382*** (0.127)	0.424*** (0.133)
R <sup>2</sup>	0.333	0.333	0.385	0.339	0.340	0.432
N	2437611	2437611	2437611	50751	50751	50751
Security*Time FE	Yes	Yes	Yes	Yes	Yes	Yes
Sector FE	No	Yes	No	No	Yes	No
Sector*Time FE	No	No	Yes	No	No	Yes
Sample	All	All	All	If Traded	If Traded	If Traded

The dependent variable is the log change in the nominal amount held. In specifications (1)-(3) the netbuy variable is zero for sectors that do not trade this security. Specifications (4)-(6) only includes observations when the security is traded by all sectors.  $\Delta$  Price is the percentage change in the price and is lagged by one quarter. Banks is a dummy that equals one if the holding sector is banks and zero otherwise. Funds is a dummy that equals one if the holding sector is investment funds and zero otherwise. The benchmark is insurance companies and pension funds. Standard errors are in parentheses. Standard errors are clustered at the security level. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

Since we are more interested in how financial institution respond to price changes, unconditional on benchmark indices, allowing for some macro-financial inferences, we relax the number of restrictions again and return to our baseline equation that includes a parsimonious set of macro and financial variables.

The previous analysis can shed light on which institutions stabilize prices to push them back to where they have been and which institutions may increase the volatility of prices and yields. However, before the global financial crisis some debt instruments may have traded above their fundamental values. Holding these prices artificially stable by acting counter-cyclically would not have helped to make markets more efficient.

As outlined above, we know that insurance companies and pension funds are long-term investors. Since banks and investment funds trade more frequently it might be worthwhile for them to buy securities that are trading at a premium and sell them again when prices have gone up further. They also may sell securities that trade below their fundamental value if they expect the downward trend to continue further. The typical buy and hold investor would buy securities when they trade at a discount or below their fundamental value to gain when prices revert to their par value. In order to shed more light on the role

of who buys at a premium and at a discount, we create a dummy that equals one if the security trades above its par value and 0 otherwise. When a bond trades at a premium it does not necessarily reflect that the security is trading above its fundamental value. If market interest rates are currently lower than when the bond was issued, investors may be willing to pay more for a bond to earn the additional interest. Hence, a reduction in the risk-free rate or the interest rates on alternative investments can lead to bonds trading at a premium.<sup>23</sup> However, banks may also have a preference for these bonds due to their higher collateral value compared to other bonds that do not trade at a premium.

$$Premium = \begin{cases} 1 & \text{if Price} > \text{Par Value} \\ 0 & \text{if Price} \leq \text{Par Value} \end{cases}$$

Columns (1)-(3) of [Table 5](#) show the results of a regression of the netbuy variable on the dummy *Premium*. Insurance companies and pension funds buy securities that are trading at a discount and sell them when they are trading at a premium. In contrast, banks and investment funds buy when the price of the security is above its par value and sell when it is trading at a discount.<sup>24</sup>

In specifications (4-6) we add the dummy to our baseline specification shown in equation (1). Holding constant the price change in the previous quarter and other important variables that affect the buying behavior, we can see that insurance companies and pension funds prefer securities that are trading at a discount. In contrast, banks tend to buy securities that are trading at a premium, regardless of whether they have gone up in the previous quarter or not. Adding an interaction term between the lagged price change and the dummy *premium* sheds light on the question of whether the cyclical investment behavior is stronger when the security is trading at a premium or a discount. Column (9) shows that banks act more pro-cyclically with respect to price changes when the security is trading at a premium. For investment funds, the pro-cyclicality is stronger for bonds that are trading at a discount. These results suggest that the pro-cyclicality of investment funds is stronger when bond prices are down compared to banks that act more pro-cyclically, when bond prices are up.

If this pro-cyclicality of investment funds is only driven by selling at a premium price and buying at a discount, this behavior may push prices towards par value. It is therefore important to explore whether banks and investment funds are also pushing prices away from par values by buying securities when they are at a premium and prices are rising or when the security is trading a discount and its price is falling.<sup>25</sup> In contrast, from the previous analysis we would expect insurance companies and pension funds to buy securities when prices are below par and are falling further and sell securities above par when they are rising further. If the security is trading at a premium and the price rises, the variable *Rise\_at\_Premium* takes the value one. If the price falls and the security is trading at a discount the variable *Fall\_at\_Discount* takes the value one. In both cases, the security price moves away from the par value. However, while future capital *gains* of the security increase if the variable *Fall\_at\_Discount* takes the value one, future capital

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<sup>23</sup>The inclusion of the EONIA as a control variable mitigates this.

<sup>24</sup>This heterogeneity can be confirmed for banks and insurance companies and pension funds when we control for all time-varying security characteristics ([Table A4](#)).

<sup>25</sup>Although they would not have been the ones that triggered the price reversal.

Table 5: Bonds that are Trading at a Premium

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	Funds	ICPF	Banks	Funds	ICPF	Banks	Funds	ICPF	Banks
<i>Premium</i>	0.0137*** (0.002)	-0.0465*** (0.008)	0.0225*** (0.004)	0.00389 (0.003)	-0.0638*** (0.015)	0.0280*** (0.007)	0.00468 (0.003)	-0.0646*** (0.015)	0.0242*** (0.007)
$\Delta$ Price*									
<i>Premium</i>							-0.109** (0.048)	0.122 (0.250)	0.646*** (0.157)
$\Delta$ Price				0.171*** (0.023)	-0.560*** (0.124)	0.153** (0.063)	0.208*** (0.028)	-0.605*** (0.165)	-0.00394 (0.072)
R <sup>2</sup>	0.000186	0.000963	0.000180	0.139	0.178	0.120	0.139	0.178	0.121
N	327026	37320	169472	263612	28096	134005	263612	28096	134005
Security FE	No	No	No	Yes	Yes	Yes	Yes	Yes	Yes
Macro Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks.  $\Delta$  Price is the percentage change in the price. The dummy *Premium* equals one if the security trades above its par value and zero otherwise. Macro Controls include the 10-year generic government bond yield, the quarterly inflation rate, the log of the VIX, the quarterly GDP growth and the EONIA. All independent variables are lagged by one quarter. Standard errors are in parentheses. Standard errors are clustered at the security level for specifications (4)-(9). \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

*losses* increase if the variable *Rise\_at\_Premium* takes the value one given the security is held until maturity.

Table 6 shows that banks and investment funds indeed *buy* securities that are rising and trading above its par value and will, by definition, fall in the long-run. This investment behavior will result in capital losses if the security is held until maturity. Speculating on further price rises indicates that investors attempt to ride the bubble and attempt to time the market by selling the security when the price is at the inflection point (Brunnermeier and Nagel, 2004).<sup>26</sup>

In contrast, insurance companies and pension funds *sell* assets that are trading at a premium and are rising, perhaps “riding against the bubble”. These types of investors have received rather less attention but are certainly important actors who can prevent the buildup of systemic risk during a bubble that could materialize in a crisis. Investment funds and banks also sell securities that are trading at a discount and fall further. Securities whose value is below par and falls will rise until maturity. Hence, holding on to these securities that are trading at a discount and are falling further will result in capital gains given the security is held until maturity and does not default.<sup>27</sup>

Banks and investment funds may be forced to sell these securities once short-term funding cannot be rolled over or end-investors and regulators impose restrictions on the bond holdings of investment funds and banks. This mechanism may be exacerbated if portfolio managers and traders have short-term incentives (Stein, 1989). If they believe that returns are positively auto-correlated at short-horizons, it may induce them to re-

<sup>26</sup>Shleifer and Vishny (2010) show that if banks believe that security prices will increase further, they lever up and buy securities. However, once prices start to fall, banks cannot roll over funding and may have to sell securities in order to de-lever again. This is not only the case if banks mark-to-market but can be explained by a collateral channel due to the value of the collateral and pro-cyclical margins that allow banks to borrow more when prices of securities are high, which may also explain their potential preference for securities that are high with a low coupon vs. securities that are low with a high coupon, given the same yield.

<sup>27</sup>In unreported regression analyses, we can confirm that security prices at horizons of 1 and 2 years are rising after insurance companies have bought these securities and falling after banks and investment funds have bought them. This is not the case for shorter horizons of for example one quarter.

balance their portfolio away from assets that have performed poorly (Cutler et al., 1990). On the contrary, it may induce return-oriented investors buying up troubled assets in order to benefit from valuation gains as they may see them as undervalued (Hanson and Stein, 2015). In line with these return-oriented investors, insurance companies and pension funds, who may be more risk tolerant due to their long-term liabilities and buy assets that are trading at a discount and have fallen. This behavior can act as a stabilizing force in bad times and prevent prices from falling by as much as they would otherwise.

Table 6: Away from Par Value

	(1)	(2)	(3)	(4)	(5)	(6)
	Funds	ICPF	Banks	Funds	ICPF	Banks
<i>Rise_at_Premium</i>	0.0148*** (0.002)	-0.0236** (0.010)	0.0508*** (0.006)			
<i>Fall_at_Discount</i>				-0.00863*** (0.003)	0.0794*** (0.013)	-0.0110* (0.006)
R <sup>2</sup>	0.136	0.176	0.122	0.136	0.178	0.121
N	266772	28312	137679	266772	28312	137679
Security FE	Yes	Yes	Yes	Yes	Yes	Yes
Macro Controls	Yes	Yes	Yes	Yes	Yes	Yes

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks. All independent variables are lagged by one quarter in columns (4)-(9). *Rise\_at\_Premium* is a dummy that takes the value one if the security is trading at a premium and the price rises and zero otherwise. *Fall\_at\_Discount* is a dummy that takes the value one if the price falls and the security is trading at a discount and zero otherwise. Macro Controls include the 10-year generic government bond yield, the quarterly inflation rate, the log of the VIX, the quarterly GDP growth and the EONIA. Standard errors are in parentheses. Standard errors are clustered at the security level for specifications (4)-(9). \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

In order to investigate further whether the cyclical investment behavior changes over the financial cyclical, we look at times of a high VIX in the next step. When market liquidity is low, pro-cyclical buying behavior can lead to strong market distortions and investors may be forced to sell at fire-sale prices because they have to meet margin calls or they cannot roll over their liabilities. If prices fall and investors act pro-cyclically during volatile times, their redemption can trigger a spiral of market and funding liquidity (Brunnermeier and Pedersen, 2009). In order to test whether the cyclical behavior of financial institutions intensified in volatile times, we interact the VIX with the change in the price. When the VIX is at its mean, insurance companies and pension funds still act counter-cyclically and investment funds and banks still act pro-cyclically (Table 7).

However, as soon as the VIX increases above its mean, investment funds exacerbate the pro-cyclicality, which is in favor of the hypothesis that investment funds act more pro-cyclically in times when asset prices are down. This indicates that in times of high uncertainty and illiquid markets they are reluctant to search for yield by buying bonds that have lost value. This is consistent with the theory of Amihud et al. (2006) and Amihud and Mendelson (1986), who show that short-term investors avoid illiquid securities in times of high expected volatility. The probability that illiquid assets will have to be sold at fire-sale prices increases when volatility increases. Hence, funds with daily reception notice should not hold illiquid assets in volatile times if they want to avoid selling off assets at fire-sale prices. In contrast, long-term investors can benefit from a liquidity premium as short-term investors avoid illiquid securities in times of high expected volatility.

Table 7: Interaction with the VIX

	(1) Funds	(2) ICPF	(3) Banks
$\Delta$ Price*VIX	0.166*** (0.056)	0.797*** (0.279)	-0.057 (0.152)
$\Delta$ Price	0.134*** (0.027)	-0.806*** (0.140)	0.183** (0.080)
R <sup>2</sup>	0.139	0.178	0.120
N	263612	28096	134005
Security FE	Yes	Yes	Yes
Macro Controls	Yes	Yes	Yes

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks. VIX is the demeaned log of the implied volatility for S&P 500 stock options.  $\Delta$  Price is the percentage change in the price. All independent variables are lagged by one quarter. Macro Controls include the 10-year generic government bond yield, the quarterly inflation rate, the log of the VIX, the quarterly GDP growth and the EONIA. Standard errors are in parentheses and clustered at the security level. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

## 6 Robustness

Until now we have assumed that the coefficient is the same for all kinds of bonds over the whole sample period. In the following tables, we relax this assumption by splitting by time periods and the types of bonds. [Table 8](#) shows differential effects by issuing sector. In general, we can confirm our previous findings. The highest quantitative responses to price changes are with respect to non-financial corporate bonds, which are our benchmark. A 10 percent increase in the price is associated with a 2.4 percent and 8.6 percent increase in the amount bought for investment funds and banks, respectively, but a 21 per cent increase in the amount *sold* by insurance companies and pension funds. While the sign of the coefficients is still in line with the benchmark model, the cyclicity is least pronounced for bonds issued by other-financial corporations and governments.

In [Table 9](#) we divide the sample into three subsamples: pre-crisis (2006 Q1:2007 Q4), crisis (2008 Q1:2012Q2), and post-crisis (2012 Q3:2014 Q4). The results confirm that the pro-cyclicity of investment firms has increased since the crisis. Before 2008, investment firms acted only insignificantly pro-cyclically, but in the crisis their reluctance to buy illiquid securities that have dropped in prices could have turned them into pro-cyclical investors. While the pro-cyclical investment behavior of banks and investment funds gets stronger over time, the counter-cyclical investment behavior of insurance companies and pension funds weakens over the sample. This raises the questions, who will step in as a counter-cyclical investor if this development continues. The documented patterns may indicate a healthy transfer of credit and liquidity risk towards a group of investors which is best set up to bear this risk in bad times. If the role of insurance companies and pension funds as a shock-absorber diminishes further, this may have adverse consequences for



Table 8: Issuing Sector Heterogeneity

	(1)	(2)	(3)
	Funds	ICPF	Banks
$\Delta$ Price	0.244*** (0.038)	-2.129*** (0.600)	0.861*** (0.226)
$\Delta$ Price*Banks	-0.001 (0.067)	1.868*** (0.647)	-0.649** (0.252)
$\Delta$ Price*Gov	-0.225*** (0.073)	1.452** (0.622)	-0.776*** (0.262)
$\Delta$ Price*OFC	-0.121** (0.050)	1.817*** (0.645)	-0.828*** (0.242)
R <sup>2</sup>	0.134	0.177	0.120
N	260420	27845	132621
Security FE	Yes	Yes	Yes
Macro Controls	Yes	Yes	Yes

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks. All independent variables are lagged by one quarter.  $\Delta$  Price is the percentage change in the price. Banks, Gov, OFC are dummies that equal one if the issuing sector is banks, the government or other financial corporations, respectively, and zero otherwise. The benchmark is securities issued by non-financial corporations. Macro Controls include the 10-year generic government bond yield, the quarterly inflation rate, the log of the VIX, the quarterly GDP growth and the EONIA. Standard errors are in parentheses and clustered at the security level. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

Table 9: Split by Time Periods

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	Funds	ICPF	Banks	Funds	ICPF	Banks	Funds	ICPF	Banks
$\Delta$ Price	0.090 (0.11)	-3.418*** (0.55)	0.083 (0.32)	0.146*** (0.03)	-0.352** (0.16)	0.113 (0.07)	0.254*** (0.05)	-0.334 (0.26)	0.467** (0.18)
R <sup>2</sup>	0.267	0.314	0.182	0.184	0.192	0.152	0.190	0.240	0.168
N	37545	6140	24698	115635	12861	65164	106086	8104	40603
Sample	Pre-crisis	Pre-crisis	Pre-crisis	Crisis	Crisis	Crisis	Post-crisis	Post-crisis	Post-crisis
Security FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Macro Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks. All independent variables are lagged by one quarter.  $\Delta$  Price is the percentage change in the price. Macro Controls include the 10-year generic government bond yield, the quarterly inflation rate, the log of the VIX, the quarterly GDP growth and the EONIA. Pre-crisis (2006 Q1: 2007 Q4), Crisis (2008 Q1: 2012 Q2), and Post-crisis (2012 Q3:2014 Q4). Standard errors are in parentheses and clustered at the security level. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

market liquidity.

Table 10: Foreign and Domestic Bonds

	(1)	(2)	(3)	(4)	(5)	(6)
	Funds	ICPF	Banks	Funds	ICPF	Banks
$\Delta$ Price	0.180*** (0.024)	-0.738*** (0.191)	0.142* (0.074)	0.0833 (0.083)	-0.516*** (0.148)	0.218* (0.120)
Gov Bond Yield	-0.787*** (0.189)	3.027*** (1.165)	-1.008* (0.550)	6.027*** (0.567)	7.398*** (1.315)	5.148*** (0.618)
$\pi$	-0.536** (0.234)	4.923*** (1.371)	0.612 (0.687)	-3.249*** (1.029)	-6.056*** (2.041)	-0.230 (1.058)
VIX	0.00247 (0.004)	0.121*** (0.025)	-0.0129 (0.013)	-0.00767 (0.010)	0.0211 (0.019)	0.000655 (0.010)
$\Delta$ GDP	1.566*** (0.288)	8.740*** (1.653)	2.242*** (0.702)	0.634 (0.500)	2.255** (1.098)	-1.161** (0.520)
EONIA	0.977*** (0.124)	-1.480* (0.859)	0.721** (0.304)	-0.619* (0.338)	-1.856*** (0.691)	-2.006*** (0.383)
R <sup>2</sup>	0.134	0.180	0.123	0.181	0.176	0.115
N	235025	15061	67801	28587	13035	66204
Sample	Foreign	Foreign	Foreign	Domestic	Domestic	Domestic
Security FE	Yes	Yes	Yes	Yes	Yes	Yes

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks. All independent variables are lagged by one quarter.  $\Delta$  Price is the percentage change in the price. Gov Bond Yield is the 10-year generic government bond yield.  $\pi$  is the quarterly inflation rate. VIX is the log of the implied volatility for S&P 500 stock options.  $\Delta$  GDP is the quarterly GDP growth. EONIA is the Euro Overnight Index Average. Sample refers to the issuing country of the bond. Foreign refers to foreign bonds and Domestic refers only to German bonds. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

Table 10 shows the difference between German and foreign bonds. The pattern described holds for both types of bonds. German institutions seem to consistently respond positively to increases in German bond yields. Relatively higher German bond yields are usually associated with less concern about the stability of the global financial system. Yields of riskier countries and the yield on the German Bund are usually negatively correlated as Germany can be seen a safe haven. Buying German bonds when the German Bund yield is high can serve as a hedge against losses on riskier debt securities in more volatile times. In episodes of high market turmoil the sectors can benefit from holding German bonds once a flight to safety triggers an appreciation of these bonds. The negative association with the EONIA can be interpreted as a flight to safety during low interest rates which is not the intended effect of a lower Euro Area policy rate. The EONIA may reflect a forward-looking element of the medium-term financial and economic conditions in the Euro Area and should not be interpreted causally. During periods of low interest rates, only insurance companies and pension funds increase their holdings of foreign securities. They also tend to shift their funds to countries with higher yielding government bonds, as we can see in column (2). Insurance companies and pension funds also increase their holdings of foreign debt securities when the VIX is high, indicating their higher risk

bearing capacity.

As an additional robustness test, [Table A5](#) shows results for a split between Euro and US dollar denominated debt securities. The pattern described above does not only exist for Euro-denominated securities but is even stronger for US dollar denominated debt securities.

## 7 Conclusion

This paper has analyzed the cyclical investment behavior of investment funds, banks and insurance companies and pension funds. The results suggest that investment funds and banks may exacerbate price dynamics by buying when the price of the security has gone up and selling when the price has fallen. In contrast, insurance companies and pension funds act counter-cyclically. For investment funds, this pro-cyclical investment behavior is even stronger in periods when the VIX is high. One explanation that could generate the heterogeneity in the cyclical investment behavior is based on the investors' balance sheet dynamics. Although investment funds use almost no leverage, both investment funds and banks can suffer runs on their short-term liabilities. The effect can be amplified through the asset side whenever price changes trigger a spiral between balance sheets and asset prices ([Adrian and Shin, 2010](#)). This is not only true for banks but also for investment funds as investors delegate their portfolio managers. Investors may withdraw their funds both when the portfolio managers under-perform and when economic conditions are unfavorable, which reduces the funds' equity capital. A reduction in the net asset value may cause the asset manager to sell off assets which again depresses asset prices with adverse effects on their performance and the economy.

Insurance companies and pension funds respond counter-cyclically to price changes: they buy when prices haven't fallen and sell when prices have gone up. Insurance companies and pension funds also tend to buy securities at discount prices when they are falling and sell when they are trading at a premium and are rising. This counter-cyclical buying behavior of insurance companies and pension funds may push prices back to their par value and may stabilize the market when prices are pushed away from fundamentals. In their role as contrarian investors, insurance companies and pension funds suffered severe losses in the short-run. However, since insurance companies and pension funds may be best set up to bear these losses due to their higher risk-taking capacity, they were able to keep building up securities even when prices were falling. By this strategy they outperformed pro-cyclical investors in the long-run.

These results have important implications for macro-prudential policy. While financial regulation has mainly focused on the banking sector, risk transfer to other financial institutions since the financial crisis calls also for an application of macro-prudential tools to parts of the non-banking sector. However, since the counter-cyclical investment behavior has weakened over the sample, it is important to find ways of preserving the counter-cyclical investment behavior also in the future.

The results also suggest that relying on banks and investment funds as investors can be hazardous as they may lead to higher volatility of prices and yields of a security. If a whole country relies more on these investors they may experience surges or stops of capital inflows with a higher probability ([Cerutti et al., 2015](#)). These large episodes of capital flows can pose challenges to policymakers by creating asset price booms or busts as well as unwarranted exchange rate developments.

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## Appendix

While most securities have a constant amount outstanding over time, the supply of some securities can change. The actual amount outstanding can change if the bond is callable or when for asset-backed securities a part of the amount issued is returned to investors early. The effective amount outstanding (the tradable amount) of securities can for instance be altered when securities are bought under asset-purchase programs. While if the total amount outstanding diminishes, the security is not included in the sample, the security is included when the amount outstanding is not reduced to zero. In order to make sure that the changed amount outstanding does not appear as a transaction, we adjust by the pool-factor.<sup>28</sup>

The nominal value is

$$NominalValue = RawValue * e * Poolfactor \quad (2)$$

where  $e$  is the domestic price of foreign currency. The pool factor adjusts the nominal value of the specific security by partial or special redemptions. If no redemption has occurred, the poolfactor is one. It gives the amount that is left to be distributed.

In order to obtain a nominal value that moves only when a security is actually bought or sold, the nominal value needs to be adjusted by exchange rate changes and the pool factor.

$$AdjustedNominalValue_t = \frac{NominalValue_t}{Poolfactor_t} * \frac{e_{t-1}}{e_t} \quad (3)$$

$\frac{e_{t-1}}{e_t} - 1$  is the percentage appreciation of the Euro. If the Euro appreciates and the foreign currencies depreciate, this reduces the nominal value of securities in Euros if these securities are denominated in foreign currency and these movements do not reflect buy decisions. By multiplying by the poolfactor, we adjust for partial or special redemptions. In the text, we always refer to the adjusted nominal value in order to adjust for the movements that do not reflect investment decisions. The netbuy variable is obtained by taking the natural log change of the adjusted nominal value if this amount changes.

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<sup>28</sup>This changed supply can still have effects that are not captured by the security fixed effects. However, we can control for this time variant amount outstanding by including security\*time fixed effects.



Table A1: Bond Holdings of German Investors (in %)

Variable	Funds	ICPF	Banks
Government	54.9	53.2	33.1
OFC	7.5	7.3	9.8
NFC	8.3	3.9	1.5
Banks	29.3	35.5	55.5
Euro	84.2	92.2	95.1
USD	11.8	2.4	3.4
Other Currency	4.2	5.6	1.8
Domestic	39.6	39.5	73.6
Foreign	60.7	60.7	26.7

Percentage debt securities holdings of investment funds (funds), insurance companies and pension funds (ICPF) and banks issued by the Government, Other Financial Corporations (OFC), Non-Financial Corporations (NFC), Banks, in Euros, US Dollars (USD), other currency and by domestic or foreign residents. Values are averages over the sample period. Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2004 Q4 - 2014 Q4; author's calculations.

Table A2: Broad vs. Relative Market Valuation

	(1)	(2)	(3)	(4)	(5)	(6)
	Funds	ICPF	Banks	Funds	ICPF	Banks
$\Delta$ Price <i>broad</i>	0.175*** (0.026)	-0.432*** (0.105)	0.140** (0.068)	0.146*** (0.031)	-0.764*** (0.109)	0.0875 (0.128)
$\Delta$ Price <i>relative</i>	0.201*** (0.019)	-0.238** (0.103)	0.167*** (0.054)	0.135*** (0.022)	-0.596*** (0.106)	0.166** (0.069)
R <sup>2</sup>	0.000415	0.000603	0.0000661	0.135	0.164	0.117
N	282471	32573	144323	282471	32573	144323
Security FE	No	No	No	Yes	Yes	Yes

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks.  $\Delta$  Price *broad* is the price change of the index for the issuing sector in the specific country.  $\Delta$  Price *relative* is the deviation of the security-specific price change from the price change of the country-issuing sector index. All independent variables are lagged by one quarter. Standard errors are in parentheses and clustered at the security level for specifications (4)-(6). \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

Table A3: Summary of Price Change Coefficients Including Zeros

Security FE	Other Characteristics	Funds	ICPF	Banks
(1) No	-	0.078*** (0.007)	-0.011*** (0.004)	0.029*** (0.010)
(2) Yes	-	0.057*** (0.010)	-0.019*** (0.004)	0.029*** (0.010)
(3) Yes	Macro Controls	0.070*** (0.010)	-0.019*** (0.004)	0.030*** (0.010)
(4) Yes	Country*Time FE	0.071*** (0.010)	-0.007* (0.004)	0.024** (0.011)
(5) Yes	Country*Issuing Sector* Time FE	0.077*** (0.010)	-0.004 (0.004)	0.030*** (0.010)
(6) Yes	Sample of securities once held by ICPF	0.054*** (0.019)	-0.113*** (0.019)	0.061* (0.033)
(7) Yes	Sample of securities once held by all sectors	0.066** (0.027)	-0.096*** (0.027)	0.119** (0.051)

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks. The coefficients are the estimated effect of a price change in the previous quarter. The number of observations are the same in specifications (1)-(5). For specifications (6) and (7) the observations are the same across the holding sectors. Macro Controls include the 10-year generic government bond yield, the quarterly inflation rate, the log of the VIX, the quarterly GDP growth and the EONIA. Standard errors in parentheses and clustered at the security level for specifications (2)-(7). \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

Table A4: Time-Varying Security Heterogeneity for Premium Bonds

	(1)	(2)	(3)	(4)	(5)	(6)
	Netbuy	Netbuy	Netbuy	Netbuy	Netbuy	Netbuy
$\Delta$ Price*Bank	0.0391*** (0.010)	0.0329*** (0.010)	0.0424*** (0.010)	0.686*** (0.185)	0.581*** (0.187)	0.701*** (0.201)
$\Delta$ Price*Funds	0.0976*** (0.009)	0.0924*** (0.009)	0.0782*** (0.010)	0.408*** (0.128)	0.364*** (0.129)	0.350*** (0.134)
<i>Premium</i> *Banks	0.00182*** (0.000)	0.00654*** (0.001)	0.00653*** (0.001)	-0.000642 (0.007)	0.0887*** (0.013)	0.135*** (0.019)
<i>Premium</i> *Funds	-0.00309*** (0.000)	0.000877 (0.001)	-0.000263 (0.001)	-0.0274*** (0.006)	0.0106 (0.010)	0.0438*** (0.015)
R <sup>2</sup>	0.333	0.333	0.385	0.339	0.341	0.433
N	2437611	2437611	2437611	50751	50751	50751
Security*Time FE	Yes	Yes	Yes	Yes	Yes	Yes
Sector FE	No	Yes	No	No	Yes	No
Sector*Time FE	No	No	Yes	No	No	Yes
Sample	All	All	All	If Traded	If Traded	If Traded

The dependent variable is the log change in the nominal amount held. In specifications (1)-(3) the netbuy variable is zero for sectors that do not trade this security. Specifications (4)-(6) only includes observations when the security is traded by all sectors.  $\Delta$  Price is the percentage change in the price. Banks is a dummy that equals one if the holding sector is banks and zero otherwise. Funds is a dummy that equals one if the holding sector is investment funds and zero otherwise. The benchmark is insurance companies and pension funds. The dummy *Premium* equals one if the security trades above its par value and zero otherwise. All independent variables are lagged by one quarter. Standard errors are in parentheses. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

Table A5: Split by Currency Denomination

	(1)	(2)	(3)	(4)	(5)	(6)
	Funds	ICPF	Banks	Funds	ICPF	Banks
$\Delta$ Price	0.179*** (0.048)	-0.319** (0.146)	0.093 (0.086)	0.192*** (0.025)	-1.960*** (0.354)	0.275*** (0.094)
R <sup>2</sup>	0.173	0.153	0.115	0.121	0.310	0.129
N	59865	19579	96203	187832	5104	29823
Currency	EUR	EUR	EUR	USD	USD	USD
Security FE	Yes	Yes	Yes	Yes	Yes	Yes
Macro Controls	Yes	Yes	Yes	Yes	Yes	Yes

The dependent variable is netbuy, which is the log change in the nominal amount held by investment funds (Funds), insurance companies and pension funds (ICPF) and banks. All independent variables are lagged by one quarter.  $\Delta$  Price is the percentage change in the price. Macro Controls include the 10-year generic government bond yield, the quarterly inflation rate, the log of the VIX, the quarterly GDP growth and the EONIA. Currency refers to the currency denomination of the bond. EUR refers to Euro-denominated bond. USD refers to US dollar-denominated bonds. Standard errors are in parentheses and clustered at the security level. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Source: Research Data and Service Centre of the Deutsche Bundesbank, Microdatabase Securities Holdings Statistics, 2005 Q4 - 2014 Q4; author's calculations.

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