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Menachem (Meni) Abudy, Daniel Nathan, and Avi Wohl*

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Abstract

We investigate daily flows to Israeli government bonds mutual funds, which are held primarily by retail investors. We divide the bonds into six categories: nominal/CPI-linked - short-term, intermediate-term, and long-term maturity. We find that unexpected daily net flows are contemporaneously correlated with price changes in all categories, with correlations ranging from 0.094 to 0.221 depending on the bond category. These price changes are significant, and they subsequently reverse fully or mostly within 10 trading days. The price reversal indicates that the initial price changes are due to "price pressure." We find that these price distortions affect break-even inflation—a measure of inflation expectations. Our findings indicate that even government bonds are affected by retail price pressure.

Keywords: retail investors, price pressure, government bonds, mutual fund flows, break-even inflation

JEL: G12, E43

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1. Introduction

Government bonds are traded mainly by institutional and professional investors. Therefore, intuitively, their prices are less likely to be distorted by retail investors. In this paper, we show, using Israeli data, that contrary to this intuition, retail investors distort government bond prices through their flows in and out of mutual funds. This finding is in line with papers investigating the effect of mutual fund flows on stock prices.¹ We find that daily net mutual fund flows to a bond category (nominal/CPI-linked; short-term, intermediate-term, and long-term maturity) are contemporaneously positively correlated with their price level. These price changes are subsequently reversed, indicating that mutual fund flows—which are translated to mutual funds transactions—induce transitory mispricing.

Since mutual funds are held mostly by retail investors, our findings show that these investors distort, through their flows, the prices of government bonds. In addition, we find that these flows distort the break-even inflation (i.e., the spread between the nominal yield and the real yield of a comparable maturity)—a measure that is monitored by economists, central banks, and governments.² Beyond the direct importance of mispricing of government bonds and expected inflation, government bond yields are used as the baseline for discount rates of corporate bonds and equities. Therefore, the mispricing of government bonds has important implications for other asset classes and investment decisions.³

Using Israeli data to analyze the effect of fund flows on government bond prices has several advantages. First, all the flows to mutual funds transmitted by the investors to their brokers are

¹ See among others: Edelen and Warner (2001), Frazzini and Lamont (2007), Coval and Safford (2007), Ben-Rephael, Kandel and Wohl (2011), and Ben-Rephael, Kandel and Wohl (2012).

² A recent example appears in the minutes of the December 2020 Federal Open Market Committee (FOMC) meeting, which includes the following: "The rise in longer term Treasury yields was concentrated in inflation compensation. The 5-year and 5-to-10-year measures of inflation compensation based on Treasury Inflation Protected Securities rose above their pre-pandemic levels." In addition, the minutes of the November 2020 Bank of Israel meeting include the following: "Inflation expectations for the coming year from all sources remained below the lower bound of the target range, but expectations derived from the capital market increased. Forward inflation expectations for the second year returned to within the target range, and expectations for longer terms remained anchored within the target."

³ For the usage of government bond yields as the risk-free rate in valuations, see Damodaran (2020).

immediately transferred to a centralized system. This enables us to accurately estimate daily flows for all mutual funds.⁴ Second, unlike other developed countries, Israeli CPI-linked and nominal government bonds have roughly the same market cap and liquidity. This enables us to calculate a reliable measure of expected inflation using market data not biased by the illiquidity of CPI-linked bonds (e.g., D'Amico, Kim and Wei (2018) and Ermolov, 2021). It should be mentioned that the size and the trading activity of the Israeli government bond market are comparable to other sovereign bond markets. For example, in 2020, both the trading volume and the national debt were above the median of OECD countries and larger than the national debt of Ireland, Austria, Sweden, and Finland, to name a few.

We use a proprietary database of the Bank of Israel that includes daily mutual fund inflows and outflows as well as net asset values (NAVs) and mutual funds asset holdings. The sample period is from June 12, 2008, to September 30, 2020.⁵ We focus on six government bond indices: short (0–2 years), intermediate (2–5 years), and long (5–10 years and 5+ years for CPI-linked and nominal bonds, respectively) for both nominal and CPI-linked bonds (hereafter, we refer to these six indices as bond categories). For each government bond category, we construct an aggregate measure of daily net flows using daily inflows and outflows of the mutual funds to government bonds. If, for example, nominal short-term bonds comprise 10% of the holdings of a certain fund, we associate 10% of this fund's net flows is then normalized by the previous day's aggregate value of the bonds in the category across all funds.

In Israel, flows are transmitted within 10–15 minutes to mutual funds that are required to meet their declared investment policies on a daily basis. Therefore, it is plausible to expect that the transactions that result from these flows occur during the same trading day. Hence it is likely

⁴ In the US, daily mutual fund data are reported voluntarily and many fund families, including large ones, such as PIMCO and Vanguard, do not report these data.

⁵ The starting date for our sample period is due to data availability.

⁶ This estimation is in line with Lou (2012) in the context of stocks.

for correlations to be found between fund flows and bond returns in a daily resolution. However, the main question of this paper is whether there is a reversal pattern. As argued by Baker and Wurgler (2007), among many others, a reversal of price changes is an indication that the initial price changes were caused by price pressure. This argument is in line with the microstructure models that imply price changes and reversals for uninformed transactions (see, among others, De Jong and Rindi, 2009).

We begin the analysis by examining the cross-correlation of normalized daily net flow measures. As expected, these flows are positively cross-correlated across the six categories, with correlations ranging from 0.25 to 0.87. Then, for each bond category, we examine the persistence in daily normalized net flows and find they are highly persistent: lagged flows predict flows with an average adjusted R^2 of around 70% across all bond categories.⁷ Next, we find that the contemporaneous relation between normalized net flows and bond returns is positive in all six bond categories.

To further investigate what drives the contemporaneous relation between net flows to mutual funds and government bond returns, we decompose normalized net flows into their explained and unexplained parts (using lagged flows) in a similar way to that of previous analyses of flows of equity mutual funds (e.g., Warther, 1995; Coval and Stafford, 2007; Ben-Rephael, Kandel and Wohl, 2011). We find that most of the contemporaneous relation between net flows and government bond returns stems from its unexplained part, with adjusted R^2 values of 1.40%, 6.43%, and 7.87% (9.68%, 17.36%, and 11.73%) for short-term, intermediate-term, and long-term nominal (CPI-linked) bonds, respectively.⁸

Next, to examine whether mutual fund flows cause price pressure in government bond prices, we investigate the relation between lagged unexplained net flows and government bond returns. First, we find that flows Granger cause returns. In addition, we find that lagged unexplained

⁷ The normalization of flows is a division by funds' value.

⁸ The adjusted R^2 reported in Ben-Rephael, Kandel and Wohl (2011)'s study, which analyzes the relation between mutual fund flows stock return, in the Israeli market, is higher than these figures at 23.6%.

net flows are negatively related to current government bond returns across all government bond categories. The negative relation, which indicates a reversal pattern is consistent with the notion of price pressure induced by flows. In the nominal bonds, the reversal begins in the first lag, whereas in the CPI-linked bonds, there is a continuation in the first and second lags (one lag in the intermediate maturity), followed by a negative relation between unexplained net flows and CPI-linked bond returns. We document a complete reversal in the nominal government bonds within 10 trading days, in which most of the reversal occurs within the first five trading days. There is only a partial reversal in the CPI-linked bonds: 67%, 71%, and 79% of the initial effect, depending on the bond category. In all cases, the reversal is statistically significant. For example, one standard deviation of unexplained flows in intermediate-term CPI-linked bonds is related to a 6.4 basis points return and a change of 2.0 basis points in yield to maturity. This effect is completely reversed within 10 trading days.

In addition, we investigate whether flows to mutual funds distort the market's expected inflation through their flows to nominal and CPI-linked government bonds. To examine this, we calculate the *break-even inflation* (BEI): the spread between nominal and CPI-linked government bonds with comparable maturities using zero-coupon yields.⁹ The BEI is a popular proxy for inflation expectations that is closely monitored by many market participants, such as economists, policymakers, and professional investors (D'Amico, Kim and Wei, 2018). First, we verify that net flows to nominal (CPI-linked) bonds are contemporaneously negatively (positively) correlated with the change in the BEI. The reason is that positive net flows to nominal (CPI-linked) bond prices and decrease nominal (CPI-linked) yields, and this causes a decrease (increase) in the BEI. Then, similarly to the previous analysis, we document a reversal pattern: the initial effect of nominal (CPI-linked) flows on the change in BEI is fully (partially) reversed within 10 days. The combination of

⁹ Because the real interest rate is close to zero, this measure is very close to the measure derived from the Fisher equation.

contemporaneous relation and subsequent reversal indicates that mutual fund flows distort the measure of expected inflation.

Our results correspond to the effect of price pressure via mutual fund flows on equity stock markets. The most closely related paper is Ben-Rephael, Kandel and Wohl (2011) who find that daily mutual fund flows create temporary price pressure in the Israeli equity market that is subsequently partially corrected within 10 trading days. Ben-Rephael, Kandel and Wohl (2012) find that aggregate monthly net exchanges to equity funds in the US, as a proxy for shifts between bond funds and equity funds, are positively contemporaneously correlated with aggregate stock market returns, and these price changes are subsequently reversed after four months. Edmans, Fernandez-Perez, Garel and Indriawan (2022) find that music sentiment (positivity of songs that individuals listen to) is positively related to mutual fund flows. Their main finding is that music sentiment is correlated with same-week equity market returns and negatively correlated with subsequent week returns.¹⁰ It should be mentioned that we do not take a stand on the question of what fraction of the flows is due to "sentiment" and what fraction can be justified by rational considerations.

To the best of our knowledge, our paper is the only one that shows a significant reversal pattern in government bond prices following price changes related to mutual fund flows. Huang, Jiang, Liu and Liu (2021) find a contemporaneous relation between flows and monthly government bond returns.¹¹ Their Figure 4 shows a reversal pattern in the second month after a current month's flows, but the statistical significance of this pattern is not tested. Edmans et al. (2022) find that music sentiment is contemporaneously negatively related to weekly government bond returns. As they state ,their analysis does not find a reversal effect.

¹⁰ See also Abudy, Mugerman and Shust (2022) for additional support on the effect of sentiment reflected in music on markets.

¹¹ Ma, Xiao and Zeng (2020) focus on the Covid crisis in 2020. They find that in meeting redemptions, funds first sold their liquid assets, including Treasuries and high-quality corporate bonds, which generated a price pressure in these markets.

Our paper focuses on daily fund flows, which mostly originated from retail investors. This is in contrast to papers that examine the effect of "large" players and "big" events on government bond prices. Greenwood and Vayanos (2010) find that the UK pension reform of 2004 and the US Treasury buyback program of 2000–2001 caused large changes in long-term interest. In similar vein, Ceballos and Romero (2020) find significant price pressure in government bond yields after portfolio-switching recommendations of a financial advisory firm. These recommendations caused massive pension fund portfolio reallocations of government bonds. The paper finds a significant cumulative abnormal return in government bonds five days before and 30 days after the events: between 60 to 70 bps. This effect does not exhibit mean reversion in the short horizon. D'Amico and King (2013) find that fluctuations in the supply of government bonds affect their yield by studying the effect of the Federal Reserve's large-scale asset purchases in 2009. Lou, Yan and Zhang (2013) find that US Treasury prices in the secondary market decrease significantly a few days before Treasury auctions and recover shortly afterward, even though the time and amount of each auction are known in advance. They note that the decrease in prices is consistent with the interpretation that large dealers tend to hedge their risk in the secondary market, thus exerting temporary downward price pressure. Taken together, the empirical evidence regarding price pressure in the government bond market is due to large market players, such as dealers and pension funds.¹² As far as we know, this is the first paper to show that retail investors (through their trading activity in mutual funds) cause temporary price changes in government bonds and inflation expectations.

The remainder of the paper is organized as follows. Section 2 provides background on the Israeli market. Section 3 presents the data and our flow variables. Section 4 presents summary statistics. Section 5 analyzes the relation between flows and lagged flows. Sections 6 and 7

¹² Czech, Huang, Lou and Wang (2021) study the secondary market of UK government bonds The paper sorts gilts into terciles based on the previous-day net purchases of hedge funds and finds that the tercile of gilts heavily bought outperforms the tercile heavily sold by 1.28 (2.88) bps on the following day (week). This return effect is completely reversed after two months. In contrast, mutual fund trading has insignificant returns.

investigate the contemporaneous and dynamic relation between flows and government bond returns, respectively. Section 8 examines the contemporaneous and dynamic relation between flows and the BEI. Section 9 concludes.

2. The Israeli market: Background

2.1 The Israeli government bond market

Israeli bonds—both corporate and government—have been traded historically on the Tel Aviv Stock Exchange (TASE, see Abudy and Wohl, 2018).¹³ This contrasts with the common practice used worldwide, particularly in the US, of trading bonds mostly over-the-counter (OTC). CPI-linked bonds were introduced in Israel in the early 1950s, and historically their market share has been significant. In December 2020, there were 18 nominal government bonds and 12 CPI-linked government bonds with a total market value of 351 and 234 billion NIS, respectively.¹⁴ The number of government bond series in Israel is small compared with that of the US because the government often expands current bond series instead of issuing new ones. Therefore, there are no *on-the-run* or *off-the-run effects*. The liquidity of CPI bonds is roughly similar to the liquidity of nominal bonds. The time-series daily averages nominal bond bid-ask spreads (not tabulated) are 1.6, 1.6, and 2.8 basis points (bps) for short-term, intermediate-term, and long-term, respectively. For CPI-linked bonds, the respective numbers are of a similar magnitude: 1.6, 2.0, and 3.3 bps.

In Israel, the market cap of government bonds is comparable to the stocks' market cap and twice the corporate bonds' market cap: 718 billion NIS vs. 842 billion NIS and 355 billion NIS, respectively.¹⁵ For comparison, in the US, the market cap of government bonds is half of the

¹³ For the corporate bond market in Israel, see also Gershgoren, Hadad and Kedar-Levy (2020).

¹⁴ We do not analyze government floating rate bonds because their duration is very low and therefore their price variability is low. As of December 2020, their total market cap was 46 billion NIS (\$14 billion).

¹⁵ These figures are from <u>www.tase.co.il</u>, as of the end of 2020. The value of government bonds includes 87 billion NIS of short-term bills (called Makam). In December 2020 one NIS (New Israeli Shekel) was equal to about \$0.31.

market cap of stocks and twice the market cap of corporate bonds (\$21 trillion, \$42 trillion, and \$10 trillion, respectively).¹⁶

The size and the trading activity of the Israeli government bond market are comparable to other sovereign bond markets. Taking the members of the OECD as an indicator for high-income economies with a high credit rating (Israel has been an OECD member since 2010), we note that in 2020, Israel's debt of \$285 billion is slightly above the median debt of the OECD countries of \$244 billion. As plotted in Appendix 1, Israel's national debt is larger than the national debt of Ireland, Austria, Sweden, and Finland, to name a few. In addition, the average daily trading volume in the government bond market is \$728 million—55% higher than the median OECD country, which is \$470 million.¹⁷ Furthermore, the trading volume in Israel's government bond market is higher than in countries such as Ireland, Sweden, and Finland, to name a few (see Appendix 1).¹⁸

The trading mechanism for government bonds at the TASE is similar to the trading of stocks and corporate bonds: a continuous limit order book trading with opening and closing auction trading sessions. There were minor changes in the trading hours at the TASE during the sample period. As of December 2020, the pre-opening stage for government bonds begins at 9:25, and the opening stage is conducted arbitrarily between 9:55 and 9:56, followed immediately by the continuous stage. On Mondays–Thursdays (Sunday), the pre-closing stage begins between 17:14 and 17:15 (15:39 and 15:40) and is followed by a closing stage between 17:24 and 17:25 (15:39 and 15:40).

2.2 Israeli mutual funds

¹⁶ As of the end of 2020, according to sifma.org. The estimate for government bonds includes bonds, notes, and bills.

¹⁷ According to data gathered from the IMF and the World Bank.

¹⁸ According to data gathered from the Securities Industry and Financial Markets Association (SIFMA), Bloomberg, the Trade Association of Emerging Markets (EMTA), the Tel Aviv Stock Exchange, The Investment Industry Regulatory Organization of Canada (IIROC), and the Japanese Securities Dealers Association (JSDA).

Similarly to common practices worldwide, Israeli mutual funds primarily cater to retail investors, and there are no funds targeted at institutional investors. According to Bank of Israel (BOI) data, in January 2020, retail investors held 92% of the NAV of the three mutual fund categories of government bonds (see Section 3.1). The remaining 8% is held by corporations. Mutual funds do not provide investors with any tax benefits, and therefore retirement savings are made through other entities, such as pension funds, that provide these tax benefits. The market value of the mutual fund industry has increased substantially in recent decades, from about 39 billion NIS in January 2000 to 224 billion NIS in September 2020. Mutual funds may hold various financial assets and are classified into categories (for a more detailed description of the mutual fund industry in Israel, see Mugerman, Steinberg and Wiener, 2022).

Investors can buy and sell mutual funds during most of the trading hours of the TASE. Investors transmit flows to mutual funds to their brokers. The brokers immediately transfer these flows to a centralized system.¹⁹ This system transmits the flows to the mutual funds every 10–15 minutes. The deadline for transferring flows, the *final hour*, varies between 15:00 and 16:00, according to the mutual fund investment policy and the day of the week. An order transmitted after the *final hour* is transferred to the next trading day. This allows the fund managers sufficient time to adjust their positions according to the daily flows. Each mutual fund has a declared investment policy (e.g., investing at least 90% of the assets in CPI-linked government bonds), and it needs to adhere to this policy on a daily basis. At the end of the trading day (17:25 on Mondays–Thursdays and 15:50 on Sundays), each fund calculates and transmits its NAV to the TASE for clearing. The investors' orders are executed at the NAV. Mutual funds publicly disclose information on their monthly flows and their security holdings approximately two months after the end of each month (e.g., information on monthly flows and holdings of January are published at the beginning of April).

¹⁹ This is a system operated by the TASE but it is not related to its trading system.

Lastly, we note that mutual funds' government bond holdings in Israel are of the same order of magnitude as those of other developed countries. For example, in 2019, Israeli mutual funds held about 10% of the government bond market, compared with 9.3% in the UK and 13% in the US.²⁰

3. Data and main variables

3.1 Data

The paper uses proprietary daily mutual fund data obtained from the BOI that is not publicly available. The sample period ranges from June 12, 2008, to September 30, 2020: a total of 2,887 trading days. The data include daily net flows and NAVs of all mutual funds in Israel during the sample period.

Mutual funds are classified according to their investment style. Each mutual fund can be classified into only one category. The classifications are set by the Israeli Securities Authority (which is the Israeli equivalent of the American Securities and Exchange Commission, SEC). There are 14 classifications in total, and fund classifications are reexamined on a monthly basis. We concentrate on three classifications that hold a non-negligible amount of government bonds. These classifications are:

- *Local bonds: general*—funds with principal holdings in corporate and government bonds.
- *Local bonds: nominal*—funds with principal holdings in nominal bonds, both government and corporate.
- *Local bonds: sovereign*—funds with principal holdings in government bonds, both nominal and CPI-linked bonds.

²⁰ US data is from the Securities Industry and Financial Markets Association (SIFMA). UK data is from the UK Office of National Statistics (ONS). For the Israeli data, see

<u>https://www.boi.org.il/en/DataAndStatistics/Pages/MainPage.aspx?Level=3&Sid=40&SubjectType=2</u>. In addition, according to the European Fund and Asset Management Association (EFAMA), the average holdings of European mutual funds in government bonds of European countries was 12.3% in 2018. See https://www.efama.org/Publications/AssetManagement%20in%20Europe%2026%20NOV%202020.pdf.

We find 1,420 mutual funds that belonged to one of the three abovementioned classifications for at least one month during the sample period (per day, the average number of funds in the sample is 656). The mutual funds in these three classifications held about 90% of the entire mutual fund holdings in government bonds throughout the sample period.²¹ The rest of the mutual fund holdings in government bonds are spread across the other fund classes and are in small quantities (e.g., mainly stock funds and mixed funds).

In addition to the BOI mutual fund data, we use data on monthly holdings in government bonds for each mutual fund: its end-of-the-month percentage holdings in government bonds according to the bond categories we examine—type (nominal or CPI-linked) and maturity (short, intermediate, and long). These data are obtained from Praedicta, a financial services company. To examine whether mutual fund flows generate price noises in government bonds, we obtain TASE's daily returns of government bond indices. The TASE calculates three indices grouped by maturity for each type of government bond (i.e., nominal and CPI-linked). This results in a total of six indices: short (0–2 years), intermediate (2–5 years), and long (5–10 and 5+ years for CPI-linked and nominal bonds, respectively). The return of the nominal (CPI-linked) bond index with maturity group i ($i \in \{\text{short-term}, \text{ intermediate-term}, \text{ long-term}\}$) on day t is denoted as $RET_NOMINAL_i$, (RET_REAL_i).

Finally, we obtain from the BOI a daily estimate of the term structure of nominal and real interest rates with zero-coupon yields.²² We use these estimates to calculate the BEI for the three maturity categories we use.

3.2 Construction of mutual fund flow measures

A mutual fund can invest in government bonds of all types (i.e., nominal and CPI-linked) and maturity (i.e., short, intermediate, and long) according to its declared "investment policy." For each day t and each mutual fund m, we calculate the value held of each bond category. For

²¹ Of the 90%, 54% were in "local bonds: sovereign," 29% in "local bonds: general," and 7% in "local bonds: nominal."

²² The Bank of Israel uses a cubic spline method to interpolate the synthetic zero-coupon yields.

nominal (CPI-linked) bonds with maturity i we denote it by $CBV_NOMINAL_{t,m,i}$ (CBV_REAL_{t.m.i}). This value is a multiplication of the percentage holdings in the government bond category at the end of the previous month by the daily NAV (in million NIS) of the mutual fund. The sum of all fund holdings in the government bond category *i*—denoted by $CNAV_NOMINAL_{i,t}(CNAV_REAL_{i,t})$ —is the sum over all mutual funds of $CBV_NOMINAL_{t,m,i}$ ($CBV_REAL_{t,m,i}$). Next, we calculate an estimate of daily net flows (inflows minus outflows) for each government bond category: first, we multiply the daily net flow of fund m by the fund's percentage holding in the bond category. Then, we sum this estimate across all funds. We denote the estimated nominal and CPI-linked net flows (in million NIS) of maturity i on day t as NFLOWS_NOMINAL_{i,t} and NFLOWS_REAL_{i,t}, respectively. We then define the nominal normalized net flow of maturity *i* on day *t* as:

$$NNFLOWS_NOMINAL_{i,t} = \frac{NFLOWS_NOMINAL_{i,t}}{CNAV_NOMINAL_{i,t-1}},$$

and the CPI-linked normalized flows as:

$$NNFLOWS_REAL_{i,t} = \frac{NFLOWS_REAL_{i,t}}{CNAV_REAL_{i,t-1}}.$$

For brevity, we refer to these flows as "nominal flows" and "real flows," respectively (see Appendix 4 for variable definitions).

4. Summary statistics

Table 1 presents summary statistics of the sample. The summary statistics are presented across our government bond categories—short, medium, and long maturity—for nominal (Panel A) and CPI-linked (Panel B) bonds. The first three rows in each panel refer to mutual funds, and the last two lines refer to government bonds.

According to Table 1, the mutual fund's holdings in government bonds (denoted as CNAV) are increasing with maturity in both nominal and CPI-linked bonds (values of 5,267, 7,505,

and 8,317 million NIS vs. 6,794, 7,196, and 11,042 million NIS in nominal and CPI-linked bond maturity categories, respectively). The absolute value of daily net flows is of similar magnitude in both nominal and CPI-linked bonds. Because the market cap of nominal bonds is larger than the market cap of CPI-linked bonds, their trading volume is also larger.

[INSERT TABLE 1 ABOUT HERE]

Table 2 presents the cross-correlation between the six normalized flow measures of the bond categories. The cross-correlation is positive across all six measures. We also note that some cross-correlations are high. For example, the correlation between normalized flows of short and intermediate maturity of CPI-linked government bonds is 0.87.

[INSERT TABLE 2 ABOUT HERE]

5. What drives mutual funds' flows?

In this section, we examine what explains the normalized net flows. Figure 1 plots the normalized net flows over time. Panel A (B) presents the average of these variables across three maturity categories of nominal (CPI-linked) government bonds. Two observations stand out. First, daily flows are highly persistent. Second, one can pinpoint periods of market turmoil because they coincide with periods of large outflows, such as the financial crisis of 2008 and the recent COVID-19 crisis that began in March 2020.

[INSERT FIGURE 1 ABOUT HERE]

To further analyze what explains daily flows, we rely on previous literature such as Ben-Rephael, Kandel and Wohl (2011), who focus on equities. We examine possible determinants of flows: lag flows, lag returns beginning of the month days, and market indicators. Therefore, we estimate the following regression model, conducted separately for each of the six bond categories:

$$NNFLOWS _ NOMINAL_{i,t} = \alpha_i + \sum_{k=1}^{5} \beta_{i,k} \cdot NNFLOWS _ NOMINAL_{i,t-k} + \sum_{k=1}^{5} \chi_{i,k} \cdot RET _ NOMINAL_{i,t-k} + \delta_i \Delta RET _ NOMINAL_{i,t-l-t-100} + \phi_i \Delta VIX_{t-1 \to t-100} + \phi_i \Delta CDS_{t-1 \to t-100} + \gamma_i \Delta TERM _ PREMIUM_{t-1 \to t-100} + \eta_i DUMMY_{month_beginning,t} + u_{i,t}$$

$$(1.1)$$

$$NNFLOWS_REAL_{i,t} = t_i + \sum_{k=1}^{5} \kappa_{i,k} \cdot NNFLOWS_REAL_{i,t-k} + \sum_{k=1}^{5} \lambda_{i,k} \cdot RET_REAL_{i,t-k} + \mu_i \Delta RET_REAL_{i,t-1 \to t-100} + \nu_i \Delta VIX_{t-1 \to t-100} + o_i \Delta CDS_{t-1 \to t-100} + \pi_i \Delta TERM_PREMIUM_{t-1 \to t-100} + \overline{\sigma_i DUMMY_{month_beginning,t}} + \nu_{i,t}$$

$$(1.2)$$

where $RET_NOMINAL_{i,t-k}$ ($RET_REAL_{i,t-k}$) is the nominal (real) return of government bond category *i* at day *t-k*, $\Delta RET_NOMINAL_{i,t-1 \rightarrow t-100}$ ($\Delta RET_REAL_{i,t-1 \rightarrow t-100}$) is the change in the nominal (real) return of government bond category *i* between day *t-1* and *t-100*, $\Delta VIX_{t-1 \rightarrow t-100}$ is the change in the Israeli VIX between day *t-1* and *t-100*, $\Delta CDS_{t-1 \rightarrow t-100}$ is the change in the 5-year CDS on Israel's sovereign debt between day *t-1* and *t-100*, $\Delta TERM_PREMIUM_{t-1 \rightarrow t-100}$ is the change in the 10-year zero-coupon yield minus the 2-year zero-coupon yield between day *t-1* and *t-100*. We use the nominal $\Delta TERM_PREMIUM$ as a control variable for the nominal normalized net flows and the real $\Delta TERM_PREMIUM$ as a control variable for the real normalized net flows. We use a lag of 100-day interval for the latter variables since according to our estimations (untabulated), these variables have the largest predictive power (e.g., 100 trading days amount to approximately five months). Lastly, *MONTH_BEGINING* is a dummy variable that equals one on the first and second days of the month and zero otherwise. We estimate several specifications of each equation to quantify the effect of each explanatory variable separately. Table 3 presents the estimation results. Panel A (Panel B) shows the results for nominal (CPIlinked) government bonds.²³ The *t*-statistics in the table are calculated using the Newey–West heteroskedasticity and autocorrelation (HAC) corrected *t*-statistics (Newey and West, 1987). Specifications (4) include all explanatory variables. In all specifications, the coefficients of the first lag of the flows and the first lag of return are positive and significant. This finding is consistent with Ben Rephael, Kandel and Wohl (2011) that investigate the flow return relations in the stock market in Israel. In addition, in all specifications, the coefficient of MONTH_BEGINNING is positive and significant. This is probably due to salaries paid at the beginning of the month. Regarding the other variables, we do not find clear significant relations across all categories.

Specifications (1) include only five lags of the flows. The adjusted R^2 are quite high and range from 64.59% to 74.04%. Adding all other explanatory variables to the flows does not add much to the explanatory power, and the adjusted R^2 range from 65.92% to 78.44%. The average of the six categories' adjusted R^2 increases from 71.01% to 72.83%. The reason that the variables except for lagged flows have a marginal contribution to the explanatory power is probably because they indirectly affect the prediction via lagged flows.

In the next section, we separate flows into "expected" and "unexpected" parts. To keep our model of flows parsimonious, we use specification (1), which only includes lagged flows to calculate expected and unexpected flows. Based on column (1) for each specification, we document a high and statistically significant positive correlation between normalized net flows and their respective lagged variables. A one-day lag of normalized net flows has coefficients that vary from 0.52 to 0.58 in the nominal category (Panel A) and 0.58 to 0.66 in the CPI-linked category (Panel B). The one-day lag estimates are statistically significant, with t-statistics higher than 5.52. The regressions' adjusted R^2 values vary from a low of about 64% in the short

²³ We use five lags of net flows in our specification in all six categories because this was the optimal number of lags in most of the categories according to the Akaike Information Criteria.

maturity of nominal government flows to a high of about 77% in the short maturity of the CPIlinked flows.²⁴ It should be noted that we also find the sensitivity of current flows to surprises (relative to expectations) in interest rates and inflation have an effect on flows (untabulated). However, since such surprises occur once a month at the most, this relation is based on a small number of observations, and therefore we cannot base a predictive analysis on such an estimation.

[INSERT TABLE 3 ABOUT HERE]

6. The contemporaneous relation between flows and government bond returns

Table 3 shows that daily normalized net flows are highly persistent. Therefore, we decompose normalized flows into their expected and unexpected components based on Warther (1995), Coval and Stafford (2007), Ben-Rephael, Kandel and Wohl (2011), and others. We do this by running a regression of the nominal and real normalized flows on five of their lags based on the specification that appears in column (1) of Table 3, which includes most of the explanatory power of current net flows.²⁵ For each bond category, the predicted value of the regression (denoted as *EXP_NNFLOWS_NOMINAL*_{*i*,*i*} and *EXP_NNFLOWS_REAL*_{*i*,*i*} for nominal and real flows, respectively) measures the expected normalized net flows, whereas the regression's residuals (denoted as *UNEXP_NNFLOWS_NOMINAL*_{*i*,*i*} and *UNEXP_NNFLOWS_REAL*_{*i*,*i*}

Table 4 presents the estimations of time-series regression of daily nominal index returns $RET_NOMINAL_{i,i}$ in Panel A (CPI-linked index returns $RET_REAL_{i,i}$ in Panel B) on

²⁴ For comparison, Edelen and Warner (2001) report an R^2 value of 55% for US equities and Ben-Rephael, Kandel, and Wohl (2011) report an adjusted R^2 of 10.3% for Israeli equities.

²⁵ As robustness, we also decompose flows to the "explained" and "unexplained" parts using the specification that appears in column (4) and include all the explanatory variables of the flows. The results are qualitatively similar.

$NNFLOWS_NOMINAL_{i,t}$ ($NNFLOWS_REAL_{i,t}$) in the first specification; and on $EXP_NNFLOWS_NOMINAL_{i,t}$ ($EXP_NNFLOWS_REAL_{i,t}$) with

UNEXP _NNFLOWS_NOMINAL_{i,t} (UNEXP _NNFLOWS_REAL_{i,t}) separately and together, in the (2)-(4) specifications, respectively. The table shows that the relation between bond returns and normalized net flows arises from their unexpected part in all bond categories. In all the regressions in both panels, EXP_NNFLOWS_NOMINAL_{i.t} and $EXP_NNFLOWS_REAL_{i,t}$ are insignificant. In addition, in both panels, the adjusted R^2 of column (4) is almost equal to the adjusted R^2 of column (3), which only uses UNEXP_NNFLOWS _ NOMINAL_i, and UNEXP_NNFLOWS _ REAL_i. The findings that the unexpected flows relate to contemporaneous returns are in line with previous literature (e.g., Ben-Rephael, Kandel and Wohl, 2011, 2012; Warther, 1995, Edelen and Warther, 2001; Chiang and Huang, 2017). As for the magnitude of the effects, we calculate that one standard deviation of unexpected flows is related to 0.35 bps, 2.74 bps, and 7.50 bps (2.54 bps, 6.38 bps, and 8.57 bps) return in the short, intermediate, and long maturity on nominal (CPI-linked) bonds, respectively. We also perform (untabulated) regressions where the explained variables are daily changes in yield-to-maturity (YTM). We find that one standard deviation of unexpected flows is related to 0.34 bps, 0.71 bps, and 1.10 bps (1.61 bps, 2.00 bps, and 1.41 bps) change in YTM in the short, intermediate, and long maturity on nominal (CPI-linked) bonds.²⁶

To further substantiate our results, we perform a placebo test and examine whether net flows to CPI-linked (nominal) bonds affect nominal (CPI-linked) bond prices. The results are reported in Table A1 of Appendix 2. In this examination, we add to the unexplained nominal (real) flows an additional control variable: the unexplained real (nominal) flows. We verify that

²⁶ The averages of the series of yields are 1.33%, 1.89%, and 3.15% (-0.12%, 0.15%, 0.89%) in the short, intermediate, and long maturity on nominal (CPI-linked) bonds.

real (nominal) flows are not significantly related to nominal (CPI-linked) government bonds and that the strong relationship between the unexplained nominal (real) flows and nominal (CPI-linked) bond returns continues to hold.

[INSERT TABLE 4 ABOUT HERE]

7. The relation between unexpected flows and subsequent government bond returns

In the previous section, we found a positive contemporaneous relation between mutual fund flows and government bond returns. This section analyzes the relation between unexpected flows—the driver of the contemporaneous relation between flows and subsequent government bond returns. In the spirit of Ben-Rephael, Kandel and Wohl (2011) for the equity market, a negative relation between lagged flows and government bond returns supports the hypothesis of a temporary price pressure caused by flows (that are transmitted to mutual fund transactions). Such findings will provide evidence for retail price pressure (via their flows to mutual funds) in the government bond market.

First, to establish the causality between lag flows and current government bond returns, we examine whether lag flows Granger cause future returns based on Granger's (1969) causality test. This examination is conducted for each category of government bonds separately, using an unrestricted model based on Equations (1.1) and (1.2), and the restricted models include net flow lags. The results for all categories (see Table A2) demonstrate that flows Granger cause government bond returns, and this causality is statistically significant. We also find that bond returns Granger cause flows in all bond categories. We emphasize that this finding does not contradict the Granger causality between flows and returns. In our case, past returns affect investor flows which affect current and future returns.

Second, to examine the relation between lagged unexpected net flows and government bond returns, we first need to determine the best fit for the number of lags of unexpected flows. We use the Akaike information criteria (AIC) to determine the number of lags for each government bond category. Except for the nominal short maturity category (which resulted in a best fit of nine lags), the optimal number of lags in all other bond categories is 10. Therefore, we use 10 lags in the entire analysis.

Table 5 reports the regressions of government bond returns on 10 lags of unexpected flows. For each government bond maturity, we run the following regression:

$$RET _NOMINAL_{i,t} = \alpha_i + \sum_{k=1}^{10} \beta_{i,k} UNEXP _NNFLOWS _NOMINAL_{i,t-k} + u_{i,t}. \quad (2.1)$$
$$RET _REAL_{i,t} = \chi_i + \sum_{k=1}^{10} \delta_{i,k} UNEXP _NNFLOWS _REAL_{i,t-k} + v_{i,t}. \quad (2.2)$$

Panel A (B) shows the result of the regression of the nominal (CPI-linked) government bond maturities. The results reveal that in all categories, most lags are negatively related to government bond returns with varying degrees of significance. Moreover, the *p*-value of the *F*-test of the lagged coefficients in all the regressions is statistically significant. These results indicate a negative relation between unexpected flows and subsequent returns.

[INSERT TABLE 5 ABOUT HERE]

To show the results more compactly, we sum the first five lags and the last five lags of the unexpected net flows into single variables and regress bond returns on these variables. Formally, for each nominal and CPI-linked government bond maturity, we run the following regression:

$$RET _NOMINAL_{i,t} = \alpha_i + \beta_{i,1} (\sum_{k=1}^{5} UNEXP _NNFLOWS _NOMINAL_{i,t-k}) + ...$$
$$+ \beta_{i,2} (\sum_{k=6}^{10} UNEXP _NNFLOWS _NOMINAL_{i,t-k}) + u_{i,t}. \quad (2.3)$$
$$RET _REAL_{i,t} = \chi_i + \delta_{i,1} (\sum_{k=1}^{5} UNEXP _NNFLOWS _REAL_{i,t-k}) + ...$$
$$+ \delta_{i,2} (\sum_{k=6}^{10} UNEXP _NNFLOWS _REAL_{i,t-k}) + v_{i,t}. \quad (2.4)$$

Panel C (D) shows the result of the regression of the nominal (CPI-linked) government bond categories. It can be seen that in each of the six regressions, both coefficients are negative, and

in each regression, there is at least one significant coefficient. This is a clear indication of the reversal pattern.

To estimate the magnitude of the reversal, we perform regressions of cumulative bond returns for different future horizons on unexpected standardized flows on day *t*. The unexpected standardized flows are calculated as unexpected daily flows divided by its daily standard deviation and denoted as $STD_UNEXP_NNFLOWS_NOMINAL_{i,t}$ ($STD_UNEXP_NNFLOWS_REAL_{i,t}$) for nominal (real) flows. This enables us to interpret the coefficients in standard deviation units. Formally, for the nominal and CPI-linked bond maturity *i*, we run the following regression:

$$RET _NOMINAL_{i,t+k->t+n} = \alpha_i + \beta_i \cdot STD _UNEXP _NNFLOWS _NOMINAL_{i,t} + u_{i,t}, \quad (2.5)$$

$$RET_REAL_{i,t+k->t+n} = \alpha_i + \beta_i \cdot STD_UNEXP_NNFLOWS_REAL_{i,t} + \nu_{i,t}, \qquad (2.6)$$

where $RET _NOMINAL_{i,t+k->t+n}$ ($RET _REAL_{i,t+k->t+n}$) denotes the cumulative future nominal (CPI-linked) government bond return from t + k to t + n for each bond maturity *i*.

[INSERT TABLE 6 ABOUT HERE]

Panel A (B) of Table 6 shows the results of regression 2.5 (2.6) for nominal (CPI-linked) government bond categories. For ease of reading, column (1) in each category presents the contemporaneous relation between unexpected standardized flows and government bond return (similar to column (3) in Table 4, but in this case, the explanatory variable is standardized). Panel A shows that in the nominal government bond categories, there is a complete reversal of the initial response within five to 10 days, depending on maturity. For CPI-linked government bonds (Panel B), we observe only a partial reversal within 10 trading days.

Panel C of Table 6 summarizes the reversal over five and 10 trading days. For example, the 5day reversal of the nominal long-maturity bond index is 69% because the coefficient in the regression that explains the 5-day subsequent return is -5.15, which is 69% of the coefficient that explains the contemporaneous return. It can be seen that the reversal is full and relatively quick in the nominal bonds category. There is a partial reversal in the CPI-linked bonds category (between 67% and 79%), and it is slower than the nominal case.

As robustness, in untabulated results, we also run a rolling regression of current flows on their lags to derive the estimates of the expected and unexpected flows, and then regress government bond returns on unexpected flows. The results are qualitatively similar to those presented in our primary analysis, with similar R^2 values. In addition, we verify that our results are not driven by the Covid-19 period. We examine a sample period ending in 2019, and the results are qualitatively similar to our main results.

As mentioned above, the initial effect on bond prices, followed by a reversal, indicates price pressure on government bond prices caused by mutual fund flows.

8. The relation between flows and break-even inflation

After verifying that flows to mutual funds cause price pressure in the government bond market, we investigate the effect of fund flows on the break-even inflation rate (BEI), defined as the spread between nominal and real interest rates with the same maturity using zero-coupon yields. Many market participants, including policymakers and professional investors, monitor the BEI rate as a proxy for expected inflation (D'Amico, Kim and Wei, 2018). Because we find in previous sections that net flows to nominal and CPI-linked government bonds distort their prices, one may expect that these flows also distort the BEI rate. However, this is not necessarily the case. If flows to nominal bonds and CPI-linked bonds are highly correlated, it may be that their effect on the BEI is small. We find that the correlations between nominal and CPI-linked normalized net flows are not very high: 0.25, 0.46, and 0.76 for short-term, intermediate-term, and long-term maturity, respectively. Therefore, we expect net flows to government bonds to affect the BEI rate. Indeed, in the following subsections, we show that mutual fund flows distort the BEI. In line with the previous analysis, we first verify that

relations between the BEI rate and normalized flows exist. Then, we study the dynamics of this relation and demonstrate a reversal pattern.

8.1 The contemporaneous relation between net flows and the BEI rate

We begin by investigating the contemporaneous relation between the daily changes in the BEI rate and normalized net flows. To conduct this analysis, we use a daily estimate of the BEI from the BOI. The central bank calculates a daily term structure of nominal and real zero-coupon yields, which we use to estimate the BEI (the spread between the nominal and real yields in yearly terms).

We use these daily data (which are not publicly available) and estimate an average BEI for each of our maturity groups.²⁷ That is, we calculate an estimate of short-term maturity BEI (0–2 years), intermediate-term maturity BEI (2–5 years), and long-term maturity BEI (5–10 and 5+ years for CPI-linked and nominal bonds, respectively).²⁸ Then, we run a regression for each maturity category where the explained variable is the daily change in the BEI, ΔBEI . The explanatory variables are the contemporaneous unexpected normalized net flows to nominal and CPI-linked bonds for each maturity, respectively. Formally, the regression takes the following form:

$$\Delta BEI_{i,t} = \alpha_i + \beta_i \cdot UNEXP _ NNFLOWS _ NOMINAL_{i,t} + \dots + \chi_i \cdot UNEXP _ NNFLOWS _ REAL_{i,t} + u_{i,t}.$$
(3.1)

Table 7 reveals that the ΔBEI is correlated, as expected, with both the nominal and the real unexpected normalized net flows. These relations hold for all bond maturities and are statistically significant (except for the nominal net flow in the short-term maturity, which is insignificant). The coefficients have the "correct" sign. For *UNEXP_NNFLOWS_NOMINAL (UNEXP_NNFLOWS_REAL)*, the relation with ΔBEI is

²⁷ While the daily BEI is not publicly available, a monthly estimate of the zero-coupon term structure is available on the BOI website.

 $^{^{28}}$ To calculate the BEI for short-term, intermediate-term, and long-term maturity, we average the nominal and real zero-coupon yields with 1–2, 3–5, and 6–10 years to maturity, respectively, and calculate the spread.

negative (positive). That is, a positive flow to nominal (CPI-linked) government bonds increases their prices and decreases the nominal (real) yields. Therefore, the relation with ΔBEI is negative (positive). The adjusted R^2 values of these regressions range from 5% to 9.3%.

[INSERT TABLE 7 ABOUT HERE]

8.2 The relation between unexpected net flows and subsequent changes in the BEI

After finding that mutual fund net flows are contemporaneously correlated with changes in the BEI, we turn to examining the economic magnitude of the relationship and whether there is a subsequent reversal in the BEI. Consequently, in Table 8, we show the results of regressing cumulative changes in the BEI on the standardized unexpected nominal and CPI-linked flows $STD_UNEXP_NNFLOWS_NOMINAL_{i,t}$ and $STD_UNEXP_NNFLOWS_REAL_{i,t}$.

Formally, we employ the following analysis:

$$\Delta BEI_{i,t+k->t+n} = \alpha_i + \beta_i \cdot STD_UNEXP_NNFLOWS_NOMINAL_{i,t} + \dots + \chi_i \cdot STD_UNEXP_NNFLOWS_REAL_{i,t} + u_{i,t}, \quad (3.2)$$

Where $\Delta BEI_{i,t+k->t+n}$ denotes the cumulative changes in the BEI from t+k to t+n in maturity category *i*.

[INSERT TABLE 8 ABOUT HERE]

Panel A of Table 8 presents a reversal pattern, and Panel B summarizes the reversal over five and 10 trading days. For example, the 5-day reversal of the long-maturity BEI is 67% because the coefficient in the regression that explains the 5-day Δ BEI is 0.26, which is 67% of the coefficient that explains the contemporaneous change in BEI. The reversal is full and relatively quick for nominal flows in all bond categories and the long-term CPI-linked flows. There is a partial reversal for the short- and intermediate-term CPI-linked flows (between 53% and 48%, respectively). In conclusion, Tables 7 and 8 indicate that the noise caused by mutual fund flows also distort the BEI.

9. Conclusion

Previous literature has investigated the effect of retail price pressure on stock prices and returns. We are not aware, however, of research on the effect of retail investors on government bond markets. Government bonds are traded mainly by institutional and professional investors. Therefore, intuitively, their prices are less likely to be affected by retail investing. In this paper, we show that contrary to this intuition, retail investors distort government bond prices through their flows in and out of mutual funds.

We use a proprietary database that includes daily flows of Israeli mutual funds that hold government bonds. We divide government bonds into six categories according to type (nominal and CPI-linked) and maturity (short-term, intermediate-term, and long-term). We find that net flows have a high and positive contemporaneous correlation with daily government bond returns. We also find that most of the contemporaneous relation between net flows and government bond returns stems from its unexplained component. We find that these price changes are fully or partially reversed within 10 trading days, and most of the reversal occurs within five trading days. We interpret these results as evidence of a "noise" caused by retail price pressure through their flows to mutual funds. Because flows to nominal and CPI-indexed bonds are not highly correlated, we find that mutual fund flows are related to BEI, a popular measure of inflation expectations. These effects are reversed fully or partially within 10 trading days, and most of the reversal occurs within five trading days. This reversal pattern is evidence that the BEI—a measure of expected inflation—is distorted by mutual fund flows.

The government bond market is of fundamental importance. It is considered a benchmark for estimating the risk-free rate for various maturities and the basis of determining the discount rates for a wide range of asset classes. In addition, government bond returns are used to estimate

expected inflation. Therefore, the price noises caused by retail investors through mutual fund flows may have a distortive effect beyond that of the government bond market.

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Table 1: Summary statistics

The table presents summary statistics of flows to government bonds funds and of the government bond market. The statistics are presented for the three maturities: short-term (0-2 years), intermediate-term (2-5 years), and long-term (5+ years for nominal bonds and 5-10 for CPI-linked bonds)—a total of six categories. Panels A and B present statistics for nominal and CPI-linked bonds, respectively. The construction of the flow measures is detailed in Section 3.2. The sample period ranges from June 12, 2008, to September 30, 2020 (2,887 trading days). *CNAV* is the sum of all fund holdings in a certain bond category. *Absolute value of daily net flows* is the absolute value of daily net flows of each mutual fund category (in millions NIS). *Daily return of government bonds* is the daily return of the government bond index. See Appendix 4 for other variable definitions.

Panel A: Nominal government bonds

		Short]	Intermedi	iate	Long		
	Ν	Mean	Median	STD	Mean	Median	STD	Mean	Median	STD
CNAV_NOMINAL (in NIS millions)	2887	5,267	5,293	1,371	7,505	7,005	2,387	8,317	9,938	3,596
Normalized net flows (% of CNAV)	2887	-0.026%	-0.023%	0.189%	0.010%	0.017%	0.245%	0.013%	0.015%	0.312%
Absolute value of daily net flows (in NIS millions)	2887	5.13	3.41	6.89	10.08	7.60	11.09	14.62	10.42	17.28
Daily return of government bonds (bps)	2887	0.66	0.31	3.10	1.53	0.93	12.20	2.97	2.96	30.91
Daily trading volume (in NIS millions)	2887	210	160	194	451	400	279	809	756	407

Panel B: CPI-linked government bonds

		Short]	Intermedi	ate	Long		
	Ν	Mean	Median	STD	Mean	Median	STD	Mean	Median	STD
CNAV_REAL (in NIS millions)	2887	6,794	6,831	3,595	7,196	7,213	3,527	11,042	11,189	2,706
Normalized net flows (% of CNAV)	2887	-0.050%	-0.013%	0.477%	0.007%	0.000%	0.484%	0.013%	0.015%	0.312%
Absolute value of daily net flows (in NIS millions)	2887	7.21	4.18	11.26	10.22	6.34	14.15	11.39	7.70	12.78
Daily return of government bonds (bps)	2887	0.27	0.41	8.59	0.73	1.10	16.12	1.69	2.68	26.44
Daily trading volume (in NIS millions)	2887	158	125	135	256	229	158	318	290	180

Table 2: Correlation of daily normalized net flows

The table presents the cross-correlation between the daily normalized net flows of the six government bond categories (*NNFLOWS* _ *NOMINAL*_{*i*,*t*} and *NNFLOWS* _ *REAL*_{*i*,*t*}). Sample period and maturity groups are detailed in Table 1. The construction of the normalized net flows is detailed in Section 3.2.

	Short maturity (nominal)	Intermediate maturity (nominal)	Long maturity (nominal)	Short maturity (CPI- linked)	Intermediate maturity (CPI- linked)	Long maturity (CPI- linked)
Short maturity (nominal)	1					
Intermediate maturity (nominal)	0.73	1				
Long maturity (nominal)	0.55	0.81	1			
Short maturity (CPI-linked)	0.25	0.40	0.35	1		
Intermediate maturity (CPI- linked)	0.30	0.56	0.48	0.87	1	
Long maturity (CPI-linked)	0.46	0.74	0.76	0.70	0.78	1

Table 3: Regressions of flows on lagged flows

The table presents the coefficients of time-series regressions of daily nominal and CPI-linked normalized net flows with maturity i (*NNFLOWS_NOMINAL*_{*i*,*i*}, *NNFLOWS_REAL*_{*i*,*i*}, in percent) on their lagged variables

lag return of the corresponding government bond category, on the changes of the market's volatility index (VIX), the change in the 5-year CDS on Israel's dollar debt, the change in the term premium (the 10-year yield minus the two-year yield), and on a dummy variable that denotes the beginning of the month. We use the change in the nominal term premium for the nominal normalized net flows (Panel A) and the real term premium for the real normalized net flows to nominal (CPI-linked) government bonds. Sample period and maturity groups are detailed in Table 1. The *t*-statistics (in parentheses) are the Newey–West HAC corrected *t*-statistics (Newey and West, 1987) with eight lags selected automatically (Newey and West, 1994).

				Depend	ent variable:	daily aggre	gate norm	alized nom	inal flows			
-		Short n	naturity		In	termediate	maturity			Long mat	urity	
_	[1]	[2]	[3]	[4]	[1]	[2]	[3]	[4]	[1]	[2]	[3]	[4]
Intercept	-0.0012	-0.0278	-0.0621	-0.0115	0.0011	0.0105	-0.0977	-0.0227	0.0001	0.0066	-0.1190	-0.0258
	(-0.92)	(-4.94)	(-8.76)	(-5.91)	(0.50)	(1.13)	(-9.56)	(-5.32)	(0.04)	(0.48)	(-9.15)	(-4.64)
NNFLOWS_NOMINAL _i	0.5158			0.4696	0.5745			0.4968	0.5751			0.5155
	(17.76)			(16.58)	(9.71)			(7.73)	(10.96)			(8.24)
NNFLOWS_NOMINAL _i	0.1125			0.1160	0.0963			0.1383	0.0526			0.0903
	(3.34)			(3.38)	(1.80)			(2.05)	(0.75)			(1.13)
NNFLOWS_NOMINAL _i	0.1290			0.1316	0.0739			0.0183	0.1241			0.0946
	(4.55)			(4.56)	(1.88)			(0.43)	(2.92)			(1.92)
NNFLOWS_NOMINAL _i	0.0215			0.0305	0.0347			0.0569	0.0269			0.0080
	(0.70)			(1.01)	(1.34)			(2.04)	(0.52)			(0.18)
NNFLOWS_NOMINAL _i	0.1148			0.1171	0.1044			0.0790	0.1027			0.1070
	(3.97)			(4.13)	(2.10)			(1.64)	(1.94)			(2.28)
RET_NOMINAL _{i,t-1}		0.0040		0.0030		0.0026		0.0024		0.0019		0.0014
		(4.44)		(5.35)		(4.16)		(4.43)		(5.84)		(5.00)
RET_NOMINAL _{i,t-2}		0.0022		0.0002		0.0009		-0.0006		0.0008		-0.0003
		(2.50)		(0.35)		(2.13)		(-1.03)		(3.86)		(-1.33)
RET_NOMINAL _{i,t-3}		0.0027		0.0015		0.0009		0.0002		0.0010		0.0001
		(3.18)		(2.90)		(2.20)		(0.52)		(4.40)		(0.76)
RET_NOMINAL _{i,t-4}		0.0030		0.0007		0.0017		0.0006		0.0008		0.0001
		(3.57)		(1.32)		(4.73)		(1.81)		(4.15)		(0.33)
RET_NOMINAL _{i,t-5}		0.0031		-0.0006		0.0020		0.0001		0.0012		0.0000
		(3.73)		(-1.14)		(5.05)		(0.64)		(5.49)		(-0.00)
RET_NOMINAL _{i,t-100}			0.0269	0.0025			0.0759	0.0148			0.0458	0.0067
			(8.86)	(2.49)			(17.65)	(4.01)			(13.62)	(3.38)
$\Delta VIX_{t-100 \rightarrow t-1}$			-0.0004	-0.0001			-0.0043	-0.0003			-0.0039	-0.0009
			(-0.64)	(-0.51)			(-1.82)	(-0.49)			(-1.51)	(-0.86)
$\Delta CDS_{t-100->t-1}$			0.0002	0.0001			-0.0015	-0.0004			-0.0026	-0.0006
			(0.90)	(1.23)			(-4.98)	(-2.39)			(-5.56)	(-2.68)
ΔTERM_PREMIUM _{t-100->t-1}			-0.0207	-0.0059			-0.0415	-0.0096			0.0496	0.0091
			(-1.79)	(-1.50)			(-2.18)	(-1.25)			(1.41)	(0.65)
MONTH_BEGINNING _t			0.0456	0.0486			0.0560	0.0420			0.0918	0.0790
			(5.33)	(8.29)			(4.34)	(5.58)			(4.74)	(6.10)
Adjusted R-squared (%)	64.59	4.09	16.83	65.92	69.70	7.41	32.61	71.91	67.91	8.89	30.77	70.60

Panel A: Normalized net nominal flows

Panel B: Normalized net CPI-linked flows

	Dependent variable: daily aggregate normalized nominal flows											
		Short n	naturity		Iı	ntermediate	maturity			Long ma	turity	
	[1]	[2]	[3]	[4]	[1]	[2]	[3]	[4]	[1]	[2]	[3]	[4]
Intercept	-0.0044		-0.0186	-0.0057	-0.0018	-0.0033	-0.0693	-0.0124	-0.0007	-0.0071	-0.0556	-0.0109
NNFLOWS_REALi,t-1	(-1.99) 0.6566		(-1.93)	(-2.76) 0.5615	(-0.50) 0.5990	(-0.16)	(-4.70)	(-2.56) 0.4597	(-0.41) 0.5809	(-0.91)	(-8.17)	(-3.45) 0.4759
NNFLOWS_REALi,t-2	(16.88) 0.1094			(13.41) 0.1290	(5.88) 0.2498			(3.92) 0.3211	(5.52) 0.1684			(4.57) 0.2330
NNFLOWS_REALi,t-3	(2.02) 0.0190			(2.19) 0.0002	(3.27) -0.0008			(3.43) -0.0658	(1.53) 0.0208			(1.91) -0.0642
NNFLOWS_REALi,t-4	(0.45) 0.0457			(0.00) 0.0731	(-0.01) -0.0012			(-0.96) 0.1014	(0.39) 0.0265			(-0.71) 0.1021
NNFLOWS_REALi,t-5	(1.05) 0.0846			(1.71) 0.1110	(-0.03) 0.0809			(1.41) 0.0624	(1.00) 0.1019			(2.02) 0.0886
RET_REALi,t-1	(2.70)	0.0053		(3.24) 0.0033	(2.17)	0.0053		(1.66) 0.0038	(3.09)	0.0012		(2.59) 0.0009
RET_REALi,t-2		(5.35) 0.0028		(6.42) 0.0000		(4.94) 0.0027		(4.55) -0.0005		(5.40) 0.0005		(4.53) -0.0001
RET_REALi,t-3		(3.39) 0.0030		(0.05) 0.0010		(3.31) 0.0030		(-1.17) 0.0005		(3.79) 0.0006		(-1.20) 0.0001
RET_REALi,t-4		(3.77) 0.0030		(2.48) 0.0005		(3.16) 0.0023		(1.18) 0.0003		(3.50) 0.0006		(0.74) 0.0002
RET_REALi,t-5		(3.87) 0.0033		(1.32) -0.0002		(3.29) 0.0032		(0.87) 0.0000		(4.55) 0.0007		(2.39) -0.0001
RET_REALi,t-100		(4.11)	-0.0052	(-0.45) -0.0079		(3.80)	0.0928	(0.08) 0.0061		(4.68)	0.0275	(-1.10) 0.0033
$\Delta VIX_{t\text{-}100\text{-}\text{>}t\text{-}1}$			(-0.47) -0.0049	(-2.35) -0.0006			(8.48) -0.0047	(1.15) -0.0005			(13.33) -0.0018	(2.56) 0.0000
$\Delta \text{CDS}_{t\text{-}100\text{-}\text{>}t\text{-}1}$			(-2.49) -0.0011	(-1.31) 0.0000			(-1.44) -0.0028	(-0.52) -0.0002			(-1.40) -0.0013	(0.04) -0.0002
ΔTERM_PREMIUM _{t-100->t-1}			(-2.00) 0.2052	(-0.31) 0.0225			(-3.17) 0.0979	(-0.54) 0.0149			(-4.78) -0.0112	(-1.15) -0.0007
MONTH_BEGINNING,			(8.58) 0.0783	(3.16) 0.0599			(2.90) 0.0953	(1.11) 0.0536			(-0.99) 0.0505	(-0.18) 0.0345
Adjusted R-squared (%)	77.04	9.75	(5.02) 42.24	(6.49) 78.44	75.82	11.00	(3.45) 46.58	(3.59) 77.34	70.99	10.61	(5.29) 37.01	(5.95) 72.78

Table 4: Contemporaneous regressions of returns on flows

This table presents the coefficients from time-series regressions of daily nominal government bond returns (Panel A, in bps) and CPI-linked government bond returns (Panel B, in bps) on the respective flow variables. Sample period and maturity groups are detailed in Table 1. The construction of the daily normalized net flows measures is detailed in Section 3.2. We construct the nominal and CPI-linked expected (denoted as $EXP_NNFLOWS_NOMINAL_{i,t}$ and $EXP_NNFLOWS_REAL_{i,t}$) and unexpected (denoted as $UNEXP_NNFLOWS_NOMINAL_{i,t}$ and $UNEXP_NNFLOWS_REAL_{i,t}$) daily normalized net flow measures, respectively, as follows: we auto-regress daily normalized net flows on their five lags. The residuals are the measure of the unexpected flow, while the predicted value measures the expected flow. The *t*-statistics (in parentheses) are the Newey–West HAC corrected *t*-statistics (Newey and West, 1987) with eight lags selected automatically (Newey and West, 1994).

			Depe	endent va	ariable: daily nominal government bond returns									
		Short	maturit	у	In	Intermediate maturity				Long maturity				
	[1]	[2]	[3]	[4]	[1]	[2]	[3]	[4]	[1]	[2]	[3]	[4]		
Intercept	0.698	0.662	0.653	0.666	1.454	1.541	1.507	1.511	2.703	2.947	2.955	2.902		
	(8.53)	(7.55)	(7.69)	(7.70)	(5.24)	(3.31)	(3.54)	(3.70)	(4.16)	(3.37)	(3.79)	(3.81)		
NNFLOWS_NOMINAL _{i,1}	1.763				5.331				16.206					
	(3.69)				(2.88)				(4.25)					
EXP_NNFLOWS_NOMINAL	t	-0.163		-0.244		-1.139		-1.368		-1.164		-1.907		
		(-0.20)		(-0.33)		(-0.43)		(-0.57)		(-0.35)		(-0.59)		
UNEXP_NNFLOWS_NOMINAL	t		3.806	3.806			25.483	25.483			48.137	48.137		
			(3.35)	(3.34)			(4.97)	(5.03)			(7.25)	(7.26)		
$Adj. R^2$ (%)	0.91	-0.10	1.40	1.35	0.88	-0.12	6.43	6.40	2.47	-0.14	7.87	7.83		

Panel A: Flows to nominal government bonds

Panel B: Flows to CPI-linked government bonds

				Depend	-linked b	ond ret	irns						
		Short	maturit	y	In	Intermediate maturity				Long maturity			
	[1]	[2]	[3]	[4]	[1]	[2]	[3]	[4]	[1]	[2]	[3]	[4]	
Intercept	0.420	0.311	0.258	0.326	0.628	0.715	0.710	0.684	1.690	1.678	1.677	1.676	
	(2.35)	(1.83)	(1.81)	(1.98)	(1.57)	(2.61)	(2.96)	(2.98)	(2.81)	(3.00)	(3.47)	(3.46)	
NNFLOWS _ REAL _{i,t}	5.679				7.597				34.854				
	(4.96)				(3.93)				(4.30)				
EXP_NNFLOWS _ REAL _{i,t}		0.711		0.731		-0.834		-0.930		-3.255		-3.679	
		(0.53)		(0.62)		(-0.48)		(-0.76)		(-0.57)		(-0.84)	
UNEXP_NNFLOWS_REAL,,			18.931	18.931			32.775	32.775			104.124	104.124	
			(5.79)	(5.78)			(15.37)	(15.45)			(12.09)	(11.89)	
$Adj.R^2$ (%)	3.99	-0.04	9.68	9.69	4.21	-0.05	17.36	17.37	4.87	-0.06	11.73	11.73	

Table 5: Regressions of returns on lagged unexpected flows

This table presents the coefficients from time-series regressions of daily nominal government bond returns (Panel A) and CPI-linked government bond returns (Panel B) on their respective mutual fund lagged unexpected net flows. Panels C and D present the coefficients from time-series regressions of daily nominal and CPI-linked government bond returns on lagged sums of unexpected net flows. Sample period and maturity groups are detailed in Table 1. The construction of the daily normalized net flows measures is detailed in Section 3.2. The daily unexpected nominal and CPI-linked net flow measures (denoted as $UNEXP_NNFLOWS_NOMINAL_{i,t}$ and

 $UNEXP_NNFLOWS_REAL_{i,t}$, respectively) are the residuals of an auto-regression of the daily normalized net flows on their five lags. The *t*-statistics (in parentheses) are the Newey–West HAC corrected *t*-statistics (Newey and West, 1987) with eight lags selected automatically (Newey and West, 1994).

	Dependent variable: $RET _ NOMINAL_{i,t}$						
	Short maturity	Intermediate maturity	Long maturity				
Intercept	0.66	1.83	3.73				
	(11.72)	(7.51)	(6.40)				
$UNEXP_NNFLOWS_NOMINAL_{i,t-1}$	-1.75	-7.33	-7.65				
	(-1.78)	(-1.52)	(-0.85)				
$UNEXP _ NNFLOWS _ NOMINAL_{i,t-2}$	-1.54	-5.98	-7.03				
	(-1.61)	(-1.21)	(-0.81)				
$UNEXP _ NNFLOWS _ NOMINAL_{i,t-3}$	-1.44	-13.20	-15.18				
	(-1.58)	(-2.74)	(-1.89)				
$UNEXP_NNFLOWS_NOMINAL_{i,t-4}$	-0.64	-6.61	0.13				
	(-0.70)	(-1.51)	(0.01)				
$UNEXP_NNFLOWS_NOMINAL_{i,t-5}$	0.56	-4.19	-10.10				
	(0.59)	(-0.96)	(-1.26)				
$UNEXP _ NNFLOWS _ NOMINAL_{i,t-6}$	-0.68	-6.41	-13.37				
	(-0.87)	(-1.59)	(-1.72)				
$UNEXP _ NNFLOWS _ NOMINAL_{i,t-7}$	-0.19	-9.26	-16.54				
	(-0.29)	(-2.05)	(-2.10)				
$UNEXP_NNFLOWS_NOMINAL_{i,t-8}$	-0.51	-1.19	-3.20				
	(-0.77)	(-0.30)	(-0.45)				
$UNEXP _ NNFLOWS _ NOMINAL_{i,t-9}$	-0.87	-1.93	-0.66				
	(-1.20)	(-0.55)	(-0.10)				
$UNEXP _ NNFLOWS _ NOMINAL_{i,t-10}$	0.11	-2.70	-2.21				
	(0.14)	(-0.83)	(-0.33)				
$Adj.R^2$ (%)	0.35	2.32	0.73				
<i>p</i> -value of <i>F</i> -test of regression	0.0178	< 0.01	< 0.01				

Panel A: Nominal returns on lags of unexpected flows

	Dependent variable: $RET _ REAL_{i,t}$						
-	Short maturity	Intermediate maturity	Long maturity				
Intercept	0.35	0.65	2.13				
	(3.32)	(3.67)	(4.91)				
$UNEXP _ NNFLOWS _ REAL_{i,t-1}$	2.36	0.40	4.20				
	(0.75)	(0.05)	(0.45)				
$UNEXP _ NNFLOWS _ REAL_{i,t-2}$	0.85	-0.66	5.01				
	(0.37)	(-0.13)	(0.68)				
$UNEXP _ NNFLOWS _ REAL_{i,t-3}$	-5.08	-11.33	-9.81				
	(-2.67)	(-2.46)	(-1.41)				
$UNEXP_NNFLOWS_REAL_{i,t-4}$	-4.17	-6.95	-9.07				
	(-2.42)	(-1.68)	(-1.41)				
$UNEXP _ NNFLOWS _ REAL_{i,t-5}$	-3.11	-6.86	-8.89				
	(-1.77)	(-1.69)	(-1.49)				
$UNEXP _ NNFLOWS _ REAL_{i,t-6}$	-1.80	-6.94	-16.19				
	(-0.99)	(-1.87)	(-2.66)				
$UNEXP _ NNFLOWS _ REAL_{i,t-7}$	1.21	-5.00	-14.26				
	(0.57)	(-1.18)	(-2.24)				
$UNEXP _ NNFLOWS _ REAL_{i,t-8}$	0.55	-2.54	-1.79				
	(0.24)	(-0.63)	(-0.34)				
$UNEXP _ NNFLOWS _ REAL_{i,t-9}$	3.46	2.64	0.90				
	(1.11)	(0.62)	(0.16)				
$UNEXP _ NNFLOWS _ REAL_{i,t-10}$	-1.08	-6.92	-8.42				
	(-0.35)	(-1.32)	(-1.02)				
$Adj.R^2$ (%)	1.56	1.65	1.38				
<i>p</i> -value of <i>F</i> -test of regression	< 0.01	< 0.01	< 0.01				

Panel B: CPI-linked returns on lags of unexpected flows

	Dependent variable: $RET _ NOMINAL_{i,t}$					
	Short maturity	Intermediate maturity	Long maturity			
Intercept	0.66	1.83	3.73			
	(13.17)	(8.53)	(7.16)			
$UNEXP _ NNFLOWS _ NOMINAL_{i,t-1->t-5}$	-0.97	-7.56	-8.08			
	(-1.97)	(-2.91)	(-1.88)			
$UNEXP _ NNFLOWS _ NOMINAL_{i,t-6->t-10}$	-0.38	-4.27	-7.28			
	(-1.15)	(-1.97)	(-1.99)			
$Adj.R^2$ (%)	0.29	1.98	0.55			
<i>p</i> -value of <i>F</i> -test of regression	< 0.01	0.0107	< 0.01			

Panel C: Nominal returns on lags of sums of unexpected flows

Panel D: CPI-linked returns on lags of sums of unexpected flows

	Dependent variable: $RET_REAL_{i,t}$				
-	Short maturity	Intermediate maturity	Long maturity		
Intercept	0.36	0.65	2.13		
	(2.28)	(4.10)	(5.43)		
$UNEXP _ NNFLOWS _ REAL_{i,t-1->t-5}$	-1.86	-5.14	-3.93		
	(-2.03)	(-2.05)	(-1.00)		
$UNEXP _ NNFLOWS _ REAL_{i,t-6->t-10}$	0.31	-3.96	-8.20		
	(0.20)	(-1.50)	(-2.11)		
$Adj.R^2$ (%)	0.35	0.97	0.66		
<i>p</i> -value of <i>F</i> -test of regression	< 0.01	< 0.01	< 0.01		

Table 6: Cumulative returns on lagged scaled unexpected flows

This table presents the coefficients from time-series regressions of cumulative daily nominal bond returns between time t+k and time t+n (Panel A) and CPI-linked bond returns (Panel B) on the respective lagged standardized unexpected net flow variables at time t. The standardized unexpected flow variables are the normalized net flows divided by their daily standard deviation (denoted as $STD_UNEXP_NNFLOWS_NOMINAL_{i,t}$ and

 $STD_UNEXP_NNFLOWS_REAL_{i,t}$). Sample period and maturity groups are detailed in Table 1. The

construction of the daily normalized net flows measure is detailed in Section 3.2. The daily unexpected nominal and CPI-linked net flows are the residuals of an auto-regression of the daily normalized net flows on their five lags. The *t*-statistics (in parentheses) are the Newey–West HAC corrected *t*-statistics (Newey and West, 1987) with five lags when the cumulative return horizon is five days and 10 lags when it is 10 days to correct for overlapping returns. Panel C of this table presents a summary of the reversals: the coefficients for the returns from (t + 1) to (t + 5) divided by the coefficient on the return on *t* is the 5-day reversal. The 10-day reversal is also estimated in this manner.

Panel A: Cumulative nominal government bond returns

	Short maturity $RET NOM_{i,t}$ RET NOM $RET NOM_{i,t+1->t+5}$ RET NOM $RET NOM_{i,t+1->t+10}$ RET NOM $RET NOM_{i,t+1->t+10}$					
Intercept	0.65	3.26	3.23	6.50		
	(7.69)	(22.64)	(22.55)	(28.14)		
STD_UNEXP_NNFLOWS_NOMINAL _{i,t}	0.35	-0.39	-0.05	-0.44		
	(3.35)	(-1.98)	(-0.66)	(-1.71)		
$Adj. R^2$ (%)	1.40	0.11	-0.04	0.06		

	Intermediate maturity				
	$RET _ NOM_{i,t}$	$RET _ NOM_{i,t+1->}$	$RET _NOM_{i,t+6->t+10}$	$RET _ NOM_{i,t+1->t+10}$	
Intercept	1.51	7.54	7.51	15.05	
	(3.54)	(14.65)	(10.88)	(18.87)	
STD_UNEXP_NNFLOWS_NOMINAL _{i,t}	2.74	-2.50	-0.67	-3.17	
	(4.97)	(-2.70)	(-1.39)	(-2.07)	
$Adj. R^2$ (%)	6.43	1.42	0.19	0.78	

	Long maturity					
	$RET _ NOM_{i,t}$	$RET _NOM_{i,t+1->t+5}$	$RET _NOM_{i,t+6->t+10}$	$RET _NOM_{i,t+1->t+10}$		
Intercept	2.96	14.48	14.78	29.25		
	(3.79)	(12.79)	(12.83)	(17.42)		
STD_UNEXP_NNFLOWS_NOMINAL _{i,t}	7.50	-5.15	-2.73	-7.88		
	(7.25)	(-2.14)	(-1.05)	(-1.99)		
$Adj. R^2$ (%)	7.87	0.31	0.04	0.34		

Panel B: Cumulative CPI-linked government bond returns

		Short maturity					
	$RET_REAL_{i,t}$	$RET _ REAL_{i,t+1->t+}$	5 RET _ REAL _{$i,t+6->t+10$}	$RET _ REAL_{i,t+1->t+10}$			
Intercept	0.26	1.29	1.27	2.56			
	(1.81)	(7.01)	(1.52)	(9.63)			
STD_UNEXP_NNFLOWS_REAL _{i,t}	2.54	-1.61	-0.09	-1.70			
	(5.79)	(-2.06)	(-0.10)	(-1.98)			
$Adj. R^2$ (%)	9.68	0.53	-0.06	0.25			

		Intermediate maturity					
	$RET _ REAL_{i,t} R$	$ET_REAL_{i,t+1->t+5}$	$RET_REAL_{i,t+6->t+10}$	$RET _ REAL_{i,t+1->t+10}$			
Intercept	0.71	3.55	3.53	7.08			
	(3.07)	(7.67)	(3.50)	(10.62)			
STD_UNEXP_NNFLOWS_REAL _{i,t}	6.38	-2.61	-1.88	-4.50			
	(15.37)	(-2.08)	(-1.04)	(-1.99)			
$Adj. R^2$ (%)	17.36	0.31	0.13	0.48			

	Long maturity					
	$RET_REAL_{i,t}$	$RET _ REAL_{i,t+1->t+1}$	$5 RET _REAL_{i,t+6->t+10} F$	$RET _ REAL_{i,t+1->t+10}$	_	
Intercept	1.68	8.37	8.39	16.75		
	(3.47)	(10.10)	(10.01)	(14.18)		
STD_UNEXP_NNFLOWS_REAL _{i,t}	8.57	-2.55	-4.11	-6.67		
	(12.09)	(-1.17)	(-2.20)	(-2.03)		
$Adj. R^2$ (%)	11.73	0.08	0.31	0.42		

Panel C: Summary of the reversal effects

Cat	egory	5-day reversal	10-day reversal	
short		111%	126%	
Nominal	intermediate	91%	116%	
	long	69%	105%	
	short	63%	67%	
CPI-linked	intermediate	41%	71%	
	long	30%	79%	

Table 7: Contemporaneous regressions of the change in break-even inflation on unexpected flows

This table presents the coefficients from time-series regressions of daily changes in break-even inflation on unexpected normalized net mutual fund flows to nominal and CPI-linked government bonds. Sample period and maturity groups are detailed in Table 1. The dependent variable is the daily changes in break-even inflation: ΔBEI , where BEI is expressed in basis points and in annual terms. The BEI is the difference between the nominal and the real zero-coupon interest rates for our maturity categories. The construction of the daily normalized net flows measures is detailed in Section 3.2. The daily unexpected nominal and CPI-linked net flow measures (denoted as $UNEXP_NNFLOWS_NOMINAL_{i,I}$ and $UNEXP_NNFLOWS_REAL_{i,I}$, respectively) are the residuals of

an auto-regression of the daily normalized net flows on their 5 lags. The *t*-statistics are corrected for heteroscedasticity using the Newey–West HAC corrected *t*-statistics (Newey and West, 1987) with eight lags selected automatically (Newey and West, 1994).

	Dependent variable: ΔBEI					
	Short maturity	Intermediate maturity	Long maturity			
Intercept	-0.06	-0.06	-0.07			
	(-0.58)	(-0.60)	(-1.36)			
$UNEXP _ NNFLOW _ NOMINAL_{i,t}$	-2.86	-7.34	-2.33			
	(-1.30)	(-3.41)	(-3.44)			
$UNEXP _ NNFLOW _ REAL_{i,t}$	7.16	10.10	8.94			
	(3.75)	(4.93)	(6.53)			
$Adj. R^2$ (%)	7.26	9.32	5.05			

Table 8: Cumulative changes in the break-even inflation rate and lagged standardized unexpected normalized net flows

Panel A of this table presents the coefficients from time-series regressions of cumulative daily changes in breakeven inflation on unexpected standardized normalized net mutual fund flows to nominal and to CPI-linked government bonds. The standardized unexpected flow variables are the normalized net flows divided by their daily standard deviation (denoted as *STD_UNEXP_NNFLOWS_NOMINAL*_i, and

STD_UNEXP_NNFLOWS_REAL_i,). The sample period ranges from June 12, 2008, to September 30, 2020

(2,887 trading days). Maturity categories are detailed in Table 1. The dependent variable is the daily changes in break-even inflation: ΔBEI , defined in Table 7. The BEI is the difference between the nominal and the real zerocoupon interest rates in each of the maturity categories. The construction of the daily normalized net flows measures is detailed in Section 3.2. The daily unexpected nominal and CPI-linked net flow measures are the residuals of an auto-regression of the daily normalized net flows on their five lags. The *t*-statistics are corrected for heteroscedasticity using the Newey–West HAC corrected *t*-statistics (Newey and West, 1987) with five lags when the cumulative change in the BEI horizon is five days and 10 lags when it is 10 days to correct for overlapping changes in the BEI. Panel B of this table presents a summary of the reversals: the coefficients for BEI change from (t + 1) to (t + 5) divided by the coefficient BEI change on time *t* is the 5-day reversal. The 10-day reversal is also estimated in this manner.

Panel A: The relation between lagged flows and BEI

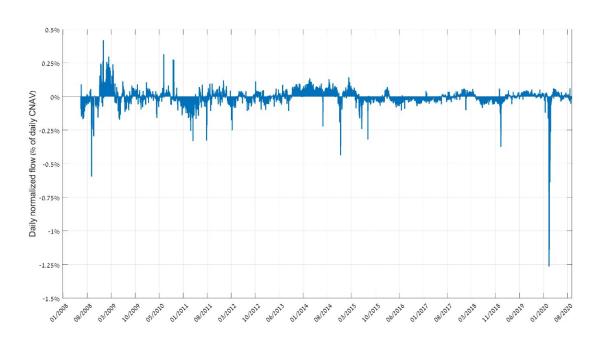
			t maturity			termediat		у		Long ma	turity	. DEV
	ΔBEI_t	$\Delta BEI_{t+1->t+5}$	$\Delta BEI_{t+6->t+10}$	$\Delta BEI_{t+1->t+10}$	ΔBEI_t	$\Delta BEI_{t+1->t+5}$	$\Delta BEI_{t+6->t+10}$	$\Delta BEI_{t+1->t+10}$	ΔBEI_t	$\Delta BEI_{t+1->t+5}$	$\Delta BEI_{t+6->t+10}$	$\Delta BEI_{t+1->t+10}$
Intercept	-0.06	0.06	-0.08	-0.02	-0.06	-0.19	-0.36	-0.56	-0.07	-0.14	-0.35	-0.49
	(-0.58)	(0.20)	(-0.13)	(-0.04)	(-0.60)	(-0.37)	(-0.64)	(-0.56)	(-1.36)	(-0.50)	(-1.22)	(-2.34)
STD_UNEXP_NNFLOWS_NOMINAL,	-0.26	0.37	-0.01	0.36	-0.77	0.49	0.31	0.80	-0.39	0.26	0.14	0.40
	(-1.30)	(1.56)	(-0.02)	(1.12)	(-3.41)	(2.91)	(1.94)	(2.63)	(-3.44)	(2.37)	(1.72)	(2.07)
STD_UNEXP_NNFLOWS_REAL	1.65	-0.82	-0.06	-0.88	2.07	-0.92	-0.07	-0.99	0.74	-0.72	-0.03	-0.74
	(3.75)	(-2.00)	(-0.08)	(-1.84)	(4.93)	(-2.03)	(-0.15)	(-1.91)	(6.53)	(-2.56)	(-0.11)	(-1.98)
$Adj. R^2$ (%)	7.26	0.46	0.13	0.10	9.32	0.58	0.23	0.52	5.05	0.74	0.19	0.43

Panel B: The reversals of BEI

Flow type	Maturity	5-day reversal	10-day reversal
	short	142%	138%
Nominal	intermediate	64%	104%
	long	67%	103%
	short	50%	53%
CPI-linked	intermediate	44%	48%
	long	97%	100%

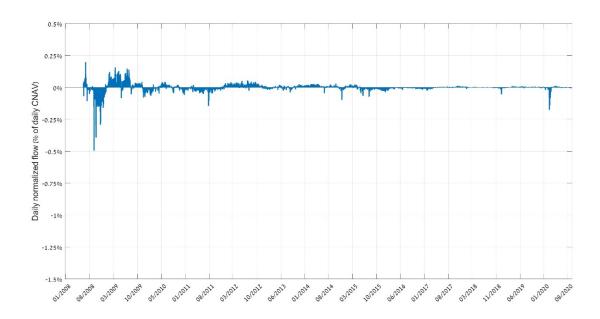
Figure 1: Normalized daily net flows

The figure plots the daily normalized net flows of mutual funds (in percent; relative to the previous daily CNAV) of government bonds. Panel A (Panel B) refers to nominal (CPI-linked) government bonds. The flows for each type of government bonds (i.e., nominal and CPI-linked) are averaged across the maturity. Sample period and maturity groups are detailed in Table 1. The construction of the normalized net flow measures is detailed in Section 3.2.



Panel A: Nominal flows

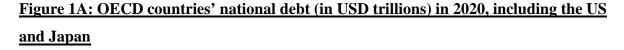
Panel B: Real flows

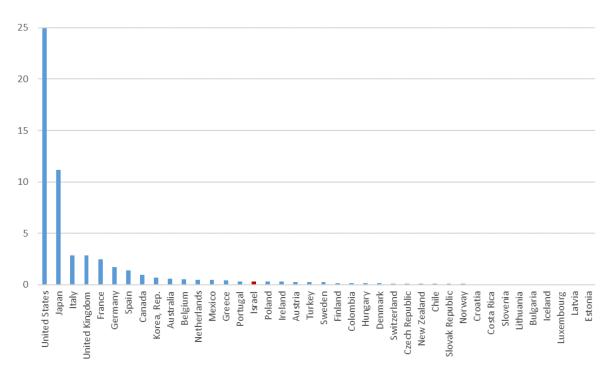


Appendix 1

Figure 1 plots the national debt (in USD trillions) of OECD countries in 2020. Since the US and Japan have exceptionally higher national debt relative to other OECD countries (\$25 and \$12 trillion, respectively), Figure 1A includes all OECD members, whereas Figure 1B excludes the US and Japan to make the rest of the OECD members' debt more comparable.

Figure 1





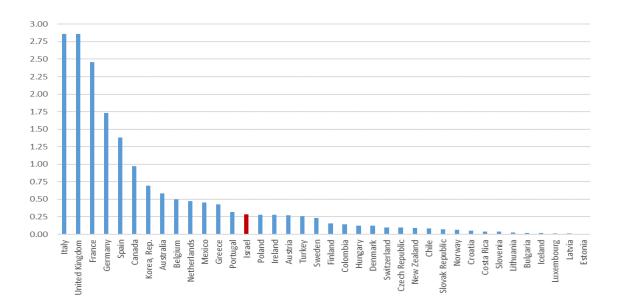


Figure 1B: OCED countries' national debt (in USD trillions) in 2020, excluding US and Japan

Note: The figure shows the national debt of OECD members in 2020 in USD trillions. Figure 1A shows it for all OECD countries, while Figure 1b shows it excluding the US and Japan. Source: The International Monetary Fund and the World Bank. The red bar shows Israel's debt.

Figure 1C plots all OECD members' daily average trading volume, excluding New Zeeland.²⁹ Figure 1D excludes the US, Japan, and Canada to make the rest of the OECD members more comparable, as their trading volume dominates all other countries. The average daily trading volume of the government bond market in the OECD countries is \$470 million. In comparison, Israel's average daily trading is \$728 million—higher than the median OECD country. Furthermore, Israel's trading volume is higher than countries such as Portugal, Ireland, Sweden, and Finland, to name a few.

²⁹ We were unable to find data regarding New Zeeland's daily trading volume.

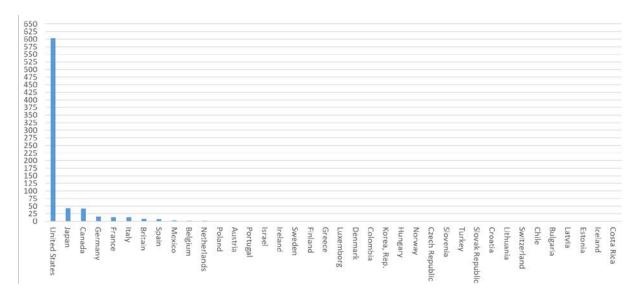
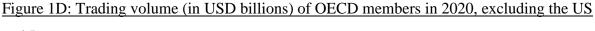
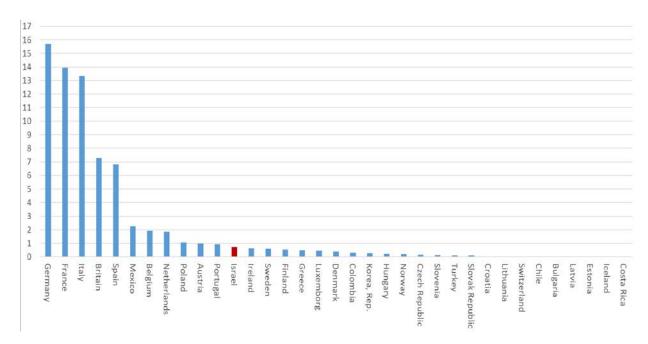


Figure 1C: Trading volume (in USD billions) of OECD members in 2020







The figure shows the average daily trading volume in government bond markets. Figure 1C plots the trading volume for all OECD countries (excluding New Zeeland), while Figure 1D plots the trading volume excluding the US, Japan, and Canada. Sources: SIFMA, Bloomberg, the Trade Association of Emerging Markets (EMTA), the Tel Aviv Stock Exchange, The Investment Industry Regulatory Organization of Canada (IIROC), and the Japanese Securities Dealers Association (JSDA).

Appendix 2

Table A1: Explaining nominal (CPI-linked) bond returns with CPI-linked (nominal) flows

This table presents the coefficients from time-series regressions of daily nominal government bond returns (Panel A, in bps) and CPI-linked government bond returns (Panel B, in bps) on the respective flow variables. Sample period and maturity groups are detailed in Table 1. The construction of the daily normalized net flows measures is detailed in Section 3.2. We construct the nominal and CPI-linked expected (denoted as and $EXP_NNFLOWS_REAL_{i,t}$) and unexpected (denoted as and $UNEXP_NNFLOWS_REAL_{i,t}$) daily normalized net flow measures, respectively, as follows: we auto-regress daily normalized net flows on their five lags. The residuals are the measure of the unexpected flow, while the predicted value measures the expected flow. The CONTROL includes the following variables: the change in the 5-year CDS on Israel's sovereign bond from *t* to *t*-100, the change in the Israeli VIX from *t* to *t*-100, the change in the term premium for the real term premium for the real return. The *t*-statistics (in parentheses) are the Newey–West HAC corrected *t*-statistics (Newey and West, 1987) with eight lags selected automatically (Newey and West, 1994).

Panel A: Nominal

	Dependent variable: daily nominal government bond returns								
	Sh	Short maturity			ediate mat	turity	Long maturity		
	[1]	[2]	[3]	[1]	[2]	[3]	[1]	[2]	[3]
Intercept	0.653	0.584	-0.009	1.507	1.342	0.994	2.955	2.956	1.605
	(7.69)	(7.90)	(-0.10)	(3.54)	(4.90)	(2.72)	(3.79)	(5.10)	(1.89)
UNEXP_NNFLOWS_NOMINAL _{it}	3.806	4.228	3.875	25.483	27.406	26.176	48.137	46.020	45.864
	(3.35)	(2.94)	(3.31)	(4.97)	(4.10)	(3.73)	(7.25)	(6.82)	
UNEXP_NNFLOWS_REAL _{it}		-1.598	-1.841		1.460	1.814		4.546	3.194
		(-1.32)	(-1.59)		(0.36)	(0.45)		(0.35)	(0.23)
CONTROLS	NO	NO	YES	NO	NO	YES	NO	NO	YES
Adjusted R-squared (%)	1.40	1.45	6.25	6.43	6.85	7.62	7.87	7.84	7.98

Panel B: CPI-linked

	Dependent variable: daily CPI-linked bond returns									
	Short maturity			Intern	nediate m	aturity	Long maturity			
	[1]	[2]	[3]	[1]	[2]	[3]	[1]	[2]	[3]	
Intercept	0.258	0.242	0.062	0.710	0.650	0.523	1.677	1.673	0.574	
	(1.81)	(1.78)	(0.33)	(2.96)	(2.95)	(1.28)	(3.47)	(3.15)	(0.67)	
UNEXP_NNFLOWS_REAL _{it}	18.931	18.514	18.811	32.775	30.837	30.615	104.124	98.887	97.830	
	(5.79)	(4.97)	(4.81)	(15.37)	(9.84)	(9.33)	(12.09)	(9.84)	(10.23)	
UNEXP_NNFLOWS_NOMIN	AL _{it}	2.179	2.898		6.268	6.664		6.087	4.597	
		(0.43)	(0.54)		(1.32)	(1.33)		(0.76)	(0.54)	
CONTROLS	NO	NO	YES	NO	NO	YES	NO	NO	YES	
Adjusted R-squared (%)	9.68	9.70	11.26	17.36	17.46	18.49	11.73	11.75	11.92	

Appendix 3

Table A2: Granger causality

This table presents the coefficients from time-series regressions of daily nominal government bond flow and returns (Panel A) and CPI-linked government bond returns (Panel B) on the respective variables. Sample period and maturity groups are detailed in Table 1. The construction of the daily normalized net flows measures is detailed in Section 3.2. We construct the nominal and CPI-linked expected (denoted as and $EXP_{NNFLOWS}_{REAL_{i,t}}$) and unexpected (denoted as and $UNEXP_{NNFLOWS}_{REAL_{i,t}}$) daily normalized net flow measures, respectively, as follows: we auto-regress daily normalized net flows on their five lags. The residuals are the measure of the unexpected flow, while the predicted value measures the expected flow. We also control for the changes of the market's volatility index (VIX) from *t-1* to *t-100*, the change in the 5-year CDS on Israel's dollar debt from *t-1* to *t-100*, the change in the term premium (the 10-year yield minus the 2-year yield) from *t-1* to *t-100*, and on a dummy variable that denotes the beginning of the month. The *t*-statistics (in parentheses) are the Newey–West HAC corrected *t*-statistics (Newey and West, 1987) with eight lags selected automatically (Newey and West, 1994). For each maturity and each dependent variable, we report the Granger *F*-test value and *p*-value comparing specification 1 and 2.

Panel A: Nominal

	Short			Intermediate				Long				
	RE	RET_it FLOW_it		RE'	T_it	FLO	W_it	RET_it		FLO	FLOW_it	
	[1]	[2]	[1]	[2]	[1]	[2]	[1]	[2]	[1]	[2]	[1]	[2]
Intercept	0.424	0.245	-0.011	-0.011	1.476	0.713	-0.023	-0.023	3.382	2.649	-0.025	-0.026
	(4.67)	(2.41)	(-5.13)	(-5.91)	(4.11)	(1.39)	(-5.21)	(-5.32)	(4.63)	(2.94)	(-4.31)	(-4.64)
NNFLOWS_NOMINAL _{i,t-1}		-2.414	0.485	0.470		-12.517	0.540	0.497		-13.458	0.560	0.516
		(-2.33)	(16.58)	(16.58)		(-2.69)	(9.19)	(7.73)		(-2.16)	(9.99)	(8.24)
NNFLOWS_NOMINAL _{i,t-2}		-0.025	0.115	0.1160		4.751	0.095	0.1383		4.920	0.045	0.0903
		(-0.02)	(3.26)	(3.38)		(0.80)	(1.71)	(2.05)		(0.81)	(0.61)	(1.13)
NNFLOWS_NOMINAL _{i,t-3}		-0.264	0.132	0.132		-8.909	0.025	0.018		-3.228	0.101	0.095
		(-0.31)	(4.50)	(4.56)		(-2.03)	(0.61)	(0.43)		(-0.49)	(2.17)	(1.92)
NNFLOWS_NOMINAL _{i,t-4}		0.135	0.017	0.031		8.028	0.044	0.057		14.009	0.001	0.008
		(0.14)	(0.55)	(1.01)		(1.77)	(1.41)	(2.04)		(2.98)	(0.02)	(0.18)
NNFLOWS_NOMINAL _{i,t-5}		0.334	0.107	0.117		1.921	0.076	0.079		-7.295	0.106	0.107
		(0.37)	(3.71)	(4.13)		(0.59)	(1.79)	(1.64)		(-1.17)	(2.28)	(2.28)
RET_NOMINAL _{i,t-1}	0.168	0.164		0.003	0.228	0.235		0.002	0.190	0.206		0.001
	(5.79)	(5.73)		(5.35)	(5.71)	(6.69)		(4.43)	(5.35)	(6.00)		(5.00)
RET_NOMINAL _{i,t-2}	-0.004	0.002		0.000	-0.062	-0.044		-0.001	-0.028	-0.014		0.000
	(-0.12)	(0.05)		(0.35)	(-1.76)	(-1.22)		(-1.03)	(-1.06)	(-0.49)		(-1.33)
RET_NOMINAL _{i,t-3}	-0.021	-0.024		0.002	-0.004	0.002		0.000	-0.019	-0.012		0.000
	(-0.75)	(-0.88)		(2.90)	(-0.10)	(0.06)		(0.52)	(-0.64)	(-0.36)		(0.76)
RET_NOMINAL _{i,t-4}	0.039	0.039		0.001	0.036	0.042		0.001	-0.010	-0.018		0.000
	(1.59)	(1.54)		(1.32)	(1.21)	(1.42)		(1.81)	(-0.32)	(-0.57)		(0.33)
RET_NOMINAL _{i,t-5}	0.061	0.067		-0.001	0.016	0.020		0.000	0.010	0.008		0.000
	(2.23)	(2.43)		(-1.14)	(0.61)	(0.73)		(0.64)	(0.41)	(0.34)		(-0.00)
RET_NOMINAL _{i.t-100}	0.089	0.166	0.004	0.002	-0.007	0.503	0.018	0.015	-0.061	0.166	0.008	0.007
· · · · · ·	(2.04)	(3.31)	(3.97)	(2.49)	(-0.03)	(1.55)	(4.93)	(4.01)	(-0.30)	(0.62)	(3.68)	(3.38)
$\Delta VIX_{t-100->t-1}$	0.0252	0.0202	0.0000	-0.0001	0.0964	0.0501	-0.0001	-0.0003	0.2066	0.1572	-0.0007	-0.0009
	(2.08)	(1.77)	(-0.21)	(-0.51)	(1.75)	(0.97)	(-0.12)	(-0.49)	(1.57)	(1.18)	(-0.73)	(-0.86)
$\Delta CDS_{t-100->t-1}$	0.0072	0.0078	0.0001	0.0001	0.0082	-0.0031	-0.0004	-0.0004	-0.0143	-0.0278	-0.0006	-0.0006
	(2.41)	(2.55)	(1.62)	(1.23)	(0.60)	(-0.20)	(-2.43)	(-2.39)	(-0.49)	(-0.82)	(-2.55)	(-2.68)
ΔTERM_PREMIUM _{t-100->t-1}	-0.314	-0.372	-0.007	-0.006	0.908	0.629	-0.012	-0.010	2.017	2.398	0.006	0.009
	(-1.60)	(-1.85)	(-1.61)	(-1.50)	(1.10)	(0.73)	(-1.43)	(-1.25)	(1.16)	(1.33)	(0.42)	(0.65)
Beginning_of_month_dummy	0.311	0.331	0.047	0.049	-0.809	-0.467	0.039	0.042	-4.372	-3.845	0.074	0.079
	(1.29)	(1.40)	(7.91)	(8.29)	(-0.94)	(-0.53)	(5.11)	(5.58)	(-1.99)	(-1.73)	(5.80)	(6.10)
Adjusted R-squared (%)	5.37	6.08	65.08	65.92	5.86	7.38	70.91	71.91	3.77	4.58	70.44	70.60
Granger F-test	14.	16	24	.98	11	.81	46.	03		4.31	31.	.40
Granger p-value	<0.	01	<0	.01	<0	.01	<0.	01		0.038	<0.	.01

Panel B: CPI-linked

	Short		Intermediate				Long					
	RE	Γ_it	FLO	FLOW_it		Γ_it	FLO	W_it	RET_it		FLO	W_it
	[1]	[2]	[1]	[2]	[1]	[2]	[1]	[2]	[1]	[2]	[1]	[2]
Intercept	0.231	0.173	-0.006	-0.006	0.615	0.343	-0.012	-0.012	1.603	1.649	-0.011	-0.011
	(1.41)	(1.04)	(-2.77)	(-2.76)	(1.94)	(1.04)	(-2.42)	(-2.56)	(2.47)	(2.49)	(-3.26)	(-3.45)
NNFLOWS_REALi,t-1		2.425	0.634	0.003		-4.371	0.579	0.004		-12.108	0.557	0.001
		(0.88)	(16.49)	(6.42)		(-1.85)	(5.82)	(4.55)		(-1.02)	(5.92)	(4.53)
NNFLOWS_REALi,t-2		3.618	0.116	0.0000		4.023	0.259	-0.0005		15.171	0.183	-0.0001
		(1.28)	(1.99)	(0.05)		(0.80)	(3.08)	(-1.17)		(1.02)	(1.56)	(-1.20)
NNFLOWS_REALi,t-3		-6.260	0.011	0.001		-6.631	-0.076	0.001		-17.098	-0.050	0.000
		(-1.92)	(0.25)	(2.48)		(-1.48)	(-1.12)	(1.18)		(-1.97)	(-0.55)	(0.74)
NNFLOWS_REALi,t-4		-1.606	0.038	0.000		2.634	0.051	0.000		7.578	0.072	0.000
		(-0.96)	(0.88)	(1.32)		(0.87)	(0.75)	(0.87)		(0.58)	(1.51)	(2.39)
NNFLOWS_REALi,t-5		0.666	0.083	0.000		1.403	0.057	0.000		-0.781	0.071	0.000
		(0.41)	(2.55)	(-0.45)		(0.49)	(1.78)	(0.08)		(-0.09)	(2.57)	(-1.10)
RET_REALi,t-1	0.140	0.127		0.562	0.208	0.229		0.460	0.215	0.214		0.476
	(4.54)	(4.10)		(13.41)	(6.15)	(8.91)		(3.92)	(6.25)	(6.28)		(4.57)
RET_REALi,t-2	-0.027	-0.045		0.129	-0.026	-0.025		0.321	-0.026	-0.026		0.233
	(-1.18)	(-1.80)		(2.19)	(-0.73)	(-0.72)		(3.43)	(-0.75)	(-0.73)		(1.91)
RET_REALi,t-3	-0.032	-0.044		0.0002	-0.016	-0.003		-0.066	-0.007	-0.007		-0.064
	(-1.17)	(-1.55)		(0.005)	(-0.63)	(-0.12)		(-0.96)	(-0.28)	(-0.28)		(-0.71)
RET_REALi,t-4	-0.032	-0.013		0.073	-0.031	-0.014		0.101	-0.007	-0.007		0.102
	(-1.66)	(-0.60)		(1.71)	(-1.32)	(-0.50)		(1.41)	(-0.22)	(-0.25)		(2.02)
RET_REALi,t-5	0.011	0.014		0.111	0.00013	-0.003		0.062	0.026	0.025		0.089
	(0.45)	(0.54)		(3.24)	(0.0048)	(-0.11)		(1.66)	(0.87)			(2.59)
RET_REALi,t-100	0.226	0.366	-0.004	-0.008	0.279		0.011	0.006	-0.062		0.004	0.003
	(1.05)	(1.81)	(-1.20)	(-2.35)	(1.26)		(1.89)	(1.15)	(-0.28)		(3.03)	(2.56)
$\Delta VIX_{t-100->t-1}$	-0.0294	-0.0337	-0.0004	-0.0006	0.0432		0.0002	-0.0005	0.0665		0.0003	0.0000
	(-0.77)	(-0.83)	(-0.86)	(-1.31)	(0.62)		(0.16)	(-0.52)	(0.66)		(0.57)	(0.04)
$\Delta CDS_{t-100->t-1}$	0.0062	0.0047	0.0000	0.0000	0.0067		-0.0002	-0.0002	-0.0022		-0.0002	-0.0002
	(0.82)	(0.58)	(-0.07)	(-0.31)	(0.44)		(-0.57)	(-0.54)	(-0.10)		(-1.15)	(-1.15)
ΔTERM_PREMIUM _{t-100->t-1}	0.312	0.395	0.022	0.023	0.141		0.015	0.015	0.601		-0.001	-0.001
	(0.72)	(0.90)	(3.04)	(3.16)	(0.17)		(1.02)	(1.11)	(0.51)		(-0.14)	(-0.18)
Beginning_of_month_dummy	0.635	0.484	0.059	0.060	0.735		0.050	0.054	-1.565		0.033	0.035
	(1.26)	(0.97)	(6.27)	(6.49)	(0.76)		(3.47)	(3.59)	(-0.97)		(5.54)	(5.95)
Adjusted R-squared (%)	2.44	3.20	77.06	78.44	4.29	5.19	75.56	77.34	3.76	5.16	70.63	72.78
Granger F-test	8.3		55.		7.2		50.			7.17		.22
Granger p-value	<0.	.01	<0.	.01	<0.	01	<0.	01		< 0.01	<0	.01

Variable	Definition
$RET_NOMINAL_{i,t}$	Daily returns of nominal government bond indices (in basis points) on day <i>t</i> in maturity <i>i</i> .
$RET_REAL_{i,t}$	Daily returns of CPI-linked government bond indices (in basis points) on day <i>t</i> in maturity <i>i</i> .
CBV_NOMINAL _{t,m,i}	The daily value of each nominal government bond maturity <i>i</i> on day <i>t</i> in fund <i>m</i> in million NIS. It is calculated as $NOMINAL - HOLDINGS_{t,m,i} \times NAV_{t,m}$, where
	<i>NOMINAL</i> _HOLDINGS _{<math>t,m,i is the percentage holdings inthe nominal government bond maturity i at the end of theprevious month of trading day t in fund m, and $NAV_{t,m}$ is the</math>}
$CBV _ REAL_{t,m,i}$	net asset value on day t in fund m in million NIS. The daily value of each CPI-linked government bond maturity i on day t in fund m in million NIS. It is calculated as $REAL _ HOLDINGS_{t,m,i} \times NAV_{t,m}$, where
	<i>REAL</i> _HOLDINGS _{t,m,i} is the percentage holdings in the CPI-linked government bond maturity <i>i</i> at the end of the previous month of trading day <i>t</i> in fund <i>m</i> , and <i>NAV</i> _{t,m} is the
	net asset value on day t in fund m in million NIS.
CNAV_NOMINAL _{i,t}	The daily sum of the value of all funds' holdings in nominal government bonds at day <i>t</i> in maturity <i>i</i> in million NIS. Formally, it equals $\sum_{m} CBV _ NOMINAL_{t,m,i}.$
CNAV_REAL _{i,t}	The daily sum of the value of all funds' holdings of CPI-linked government bonds at day <i>t</i> in maturity <i>i</i> in million NIS. Formally, it equals $\sum_{m} CBV _ REAL_{t,m,i}.$
NFLOWS_NOMINAL _{i,t}	The daily nominal government bonds net flow, in million NIS, calculated as: $NFLOW_NOMINAL_{i,t} =$ $\sum_{m} NOMINAL_HOLDINGS_{t,m,i} \times NFLOW_NOMINAL_{t,m},$ where $NFLOWS_NOMINAL_{t,m}$ is the net flow of fund m on day t in million NIS and $NOMINAL_HOLDINGS_{t,m,i}$
	is the percentage holdings in the nominal government bond in maturity i at the end of the previous month of trading day t of fund m .
NFLOWS _ REAL _{i,t}	The daily CPI-linked government bonds net flow, in million NIS, calculated as: $NFLOW_REAL_{i,t} =$
	$\sum_{m} REAL_HOLDINGS_{t,m,i} \times NFLOWS_REAL_{t,m},$
	where $NFLOWS_REAL_{t,m}$ is the net flow of fund m on
	day <i>t</i> in million NIS and $REAL_HOLDINGS_{t,m,i}$
	is the percentage holdings in the CPI-linked government bond maturity i at the end of the previous month of trading day t of fund m .

Appendix 4: Variable definitions

NNFLOWS_NOMINAL _{i,t}	The daily normalized net flow of nominal government bonds, in percent. It is obtained by dividing the aggregate net flows by the previous day's CNAV of the funds: $NNFLOWS _NOMINAL_{i,t} = \frac{NFLOWS_NOMINAL_{i,t}}{CNAV_NOMINAL_{i,t-1}}.$
NNFLOWS _ REAL _{i,t}	The daily normalized net flow of CPI-linked government bonds in percent. It is obtained by dividing the aggregate net flows by the previous day's CNAV of the funds: $NNFLOWS_REAL_{i,t} = \frac{NFLOWS_REAL_{i,t}}{CNAV_REAL_{i,t-1}}.$
EXP_NNFLOWS _ NOMINAL _{i,t}	The expected daily nominal normalized net flow in percent. It is calculated for each maturity group as the expected value of the following regression: $NNFLOWS _NOMINAL_{i,t} =$ $\alpha_i + \sum_{k=1}^{5} \beta_{i,k} NNFLOWS _NOMINAL_{i,t-k} + u_{i,t},$ where <i>t</i> is the time index, and <i>i</i> is the bond maturity group.
EXP_NNFLOWS _ REAL _{i,t}	The expected daily CPI-linked normalized net flow in percent. It is calculated for each maturity group as the expected value of the following regression: $NNFLOWS _ REAL_{i,t} =$ $\chi_i + \sum_{k=1}^{5} \delta_{i,k} NNFLOWS _ REAL_{i,t-k} + v_{i,t},$
UNEXP_NNFLOWS _ NOMINAL _{i,t}	where <i>t</i> is the time index, and <i>i</i> is the bond maturity group. The unexpected daily nominal normalized net flow in percent. It is calculated as the residual of the following regression: <i>NNFLOWS</i> _ <i>NOMINAL</i> _{<i>i</i>,<i>t</i>} = $\alpha_i + \sum_{k=1}^{5} \beta_{i,k} NNFLOWS_NOMINAL_{i,t-k} + u_{i,t},$ where <i>t</i> is the time index, and <i>i</i> is the bond maturity group.
UNEXP_NNFLOWS _ REAL _{i,t}	The unexpected daily CPI-linked normalized net flow in percent. It is calculated as the residual value of the following regression: $NNFLOWS _ REAL_{i,t} =$ $\chi_i + \sum_{k=1}^{5} \delta_{i,k} NNFLOWS _ REAL_{i,t-k} + v_{i,t},$ where <i>t</i> is the time index, and <i>i</i> is the bond maturity group.
STD_UNEXP_NNFLOWS_NOMINAL _{i,t}	The unexpected daily nominal normalized net flow ($UNEXP_NNFLOWS_NOMINAL_{i,t}$) divided by its daily standard deviation.
STD_UNEXP_NNFLOWS_REAL _{i,t}	The unexpected daily CPI-linked normalized net flow ($UNEXP_NNFLOWS_REAL_{i,t}$) divided by its daily standard deviation.
$\Delta BEI_{i,t}$	Daily change in the break-even inflation (in basis points). To calculate the BEI for the short-term, intermediate-term, and long-term maturity, we average the nominal and real zero-coupon yields with 1–2, 3–5, and 6–10 years to maturity respectively and calculate the spread.

$\Delta RET_NOMINAL_{i,t-1 \rightarrow t-100}$	The change in the return (in basis points) of nominal government bond category <i>i</i> between day <i>t</i> -1 and <i>t</i> -100.
$\Delta RET_REAL_{i,t-1 \rightarrow t-100}$	The change in the return (in basis points) of real government bond category <i>i</i> between day <i>t</i> -1 and <i>t</i> -100.
$\Delta TERM_PREMIUM_{t-1 \rightarrow t-100}$	The change in term premium, which is defined as the 10-year yield from the zero-coupon bond minus the 2-year zero-coupon yield between day $t-1$ and $t-100$. We use the nominal term premium in the analysis of the nominal flows and returns, and the real term premium in the analysis of the real flows and returns. The units of the zero-coupon yields are percent.
$\Delta CDS_{t-1 \rightarrow t-100}$	The change in the 5-year CDS on Israeli government debt between day <i>t</i> -1 and <i>t</i> -100. The units of the CDS are basis points.
$\Delta VIX_{t-1 \rightarrow t-100}$	The change in the Israeli VIX between day <i>t</i> -1 and <i>t</i> -100. The units of the VIX are percent.
MONTH_BEGINNING	A dummy variable that equals 1 on the first and second days of the month and zero otherwise.