

Monetary Policy News in the US: Effects on Emerging Market Capital Flows*

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Abstract

We examine the impact of US monetary policy news on portfolio flows to emerging markets using a Bayesian Vectorautoregression that accounts for expectations of future monetary policy. We define the US “monetary policy news shock” as one that increases monetary policy expectations while leaving the policy rate unchanged. Results suggest that the impact of this shock on portfolio flows as a share of GDP is economically small on aggregate but varies considerably across countries. The countries we identify as being the most affected also experienced larger volumes of capital in- and outflows before and after the 2013 taper tantrum episode, respectively. Also, macroeconomic performance and external vulnerabilities may matter. However, financial openness does not seem to be associated with differences in effects on capital flows over our sample period.

Keywords: Monetary policy news, Capital flows, International policy transmission, Federal funds rate expectations, Vectorautoregression

JEL: E52, E58, F42

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1. Introduction

The conduct of monetary policy in advanced economies has undergone significant changes since the global financial crisis of 2007-08, with several central banks relying increasingly on unconventional monetary policy tools in addition to the traditional approach of using short-term interest rates to stabilize the economy. Notably, with interest rates at their zero lower bound (ZLB), unconventional monetary policies (UMPs) – communication about the future evolution of interest rates as well as large scale asset purchases (LSAPs)– have been the only way for the Federal Reserve (Fed) to affect market expectations and to provide policy accommodation in the US until a few years ago. As these policy interventions are unwound, the focus among investors and policymakers has turned to the challenges that monetary policy normalization in advanced economies could pose for global financial markets. And, central to any discussion about the cross-border spillovers of monetary policy are capital flows to emerging market economies (EMEs).

One prominent example of how Fed communication can result in large swings in global financial markets is the so-called “taper tantrum” episode of May 2013 – then-Chairman Ben Bernanke’s 22 May congressional testimony hinting that the Fed would start scaling back its LSAPs.¹ Financial market participants revised their expectations as to when the Fed would begin normalizing monetary policy. These changes in policy rate expectations likely led to reductions in market participants’ tolerance for risk and a reassessment of the returns from investing in EMEs, and resulted in a sharp withdrawal of private capital flows (Nechio, 2014).

As the “taper tantrum” episode illustrates, a pressing question for emerging-market policymakers is how capital flows respond to news about future US

¹For further details on former Chairman Bernanke’s testimony before the Joint Economic Committee of the US Congress on 22 May 2013, see <http://www.federalreserve.gov/newsevents/testimony/bernanke20130522a.htm>.

monetary policy. Addressing this question is important as the past few decades have been witness to several episodes of abrupt reversals in capital flows to EMEs, followed in most cases by economic and financial crises.² In a world of
30 highly integrated capital and goods markets, US monetary policy news is bound to have some influence on capital flows to EMEs. In this paper, we provide an empirical estimate of these spillover effects.

To do so, we include expectations of the future path of the federal funds rate as well as a common factor of capital flows in a standard monetary policy
35 vector autoregressive (VAR) model. We, then, identify US monetary policy news shocks by a combination of zero and sign restrictions. Market participants receive new information (news) about the future path of the policy rate from the Fed well before these changes in the rate actually occur and, therefore, adjust their expectations about monetary policy accordingly. These news shocks
40 shift markets' expectations about future policy actions while leaving the policy rate per se unchanged. Thus, one could interpret these shocks as future or anticipated monetary policy shocks as they are observed before they materialize (in the sense of Beaudry & Portier, 2006). At the ZLB, monetary policy news shocks are related to Fed communication about the future evolution of interest
45 rates (forward guidance) but also about LSAPs, since both can shift policy rate expectations while the actual rate remains unchanged.

The idea that US monetary policy is partly anticipated is not new and has gained ground with the first release of a Federal Open Market Committee (FOMC) statement in 1994. For example, Gürkaynak et al. (2005) and
50 Campbell et al. (2012) have demonstrated that monetary policy news (from FOMC statements) affect expectations about future monetary policy decisions. At the same time, Poole (2005) shows that since February 1994 policy decisions taken at regularly scheduled FOMC meetings, whether or not involving a federal funds target change, have generated relatively little surprise in the federal funds

²See, for example, Forbes & Warnock (2012) and Reinhart & Reinhart (2008) for a detailed discussion on large capital flow movements and economic crises.

55 futures market. Such current decisions have been well anticipated by market participants.³

Finally, accounting for monetary policy expectations is of particular importance since, in the case of anticipation, VAR models with insufficient information (that is, without agent's expectations) fail to capture the true dynamics of the
60 time series (see, e.g., Hansen & Sargent, 1991 and Lippi & Reichlin, 1994).⁴

Our paper contributes to three main strands of the literature. First, a few recent studies have used an event study approach to analyze the impact of Fed announcements (such as FOMC statements) on emerging market exchange rates, interest rates, and asset prices (for example, Aizenman et al. 2016, Eichengreen
65 & Gupta 2015, and Mishra et al. 2014).⁵ In contrast to these papers, we focus directly on the impact on capital flows to emerging markets as challenges posed by large fluctuations in capital flows are often at the forefront for policymakers. In addition, we use a VAR approach which informs us about the persistence of the effects on capital flows as opposed to just the instantaneous effect of a
70 policy surprise, as is the case with an event study approach.

Second, our paper also makes an important contribution to the literature assessing spillover effects of US monetary policy to emerging markets. The notion that (expansionary) US monetary policy plays a role in driving capital flows to EMEs goes back to the seminal paper by Calvo et al. (1993). Since
75 then, a large volume of papers has examined the role of Fed's monetary policy in explaining movements in emerging markets' real activities and financial markets (see Canova, 2005, Mackowiak, 2007, Iacoviello & Navarro, 2019, and Dedola et al., 2017, among many others). Further, the relation between US interest

³Moreover, Coibion & Gorodnichenko (2012) find an increase in the ability of financial markets and professional forecasters to predict subsequent interest rate changes after 1994. Similarly, Swanson (2006) documents improved predictability of US monetary policy by both professional forecasters and Fed funds futures after Fed communications reforms (including the introduction of FOMC statements in 1994).

⁴Fiscal foresight is one prominent example, see Yang, 2005 and Leeper et al., 2013.

⁵Recent studies focusing on the impact of UMPs by the Fed at the ZLB on capital flows using different methodologies include Ahmed & Zlate (2014), Fratzscher et al. (2016), Lim et al. (2014), and Moore et al. (2013).

rates and cross-border banking flows to emerging markets is studied in, for
80 example, Bruno & Shin (2015), Avdjiev & Hale (2019), Cerutti et al. (2017),
and Goldberg (2002). The novel contribution of our paper lies in being among
the first to incorporate expectations of future US monetary policy and, thus,
allowing for monetary policy anticipation. We explicitly identify the effects of
monetary policy news on emerging-market capital flows, an aspect that has
85 largely gone unexplored.

Third, a strand of literature has isolated US monetary policy surprises by
using market-based measures to calculate the unanticipated part of the the Fed's
policy action. The dominant approach in this literature, pioneered by Kuttner
(2001), has been to measure monetary policy surprises as the change in expecta-
90 tions of the federal funds rate on the day of a Fed policy change or announcement
itself (Barakchian & Crowe 2013; Gertler & Karadi 2015; Gürkaynak 2005 and
Hamilton 2008, among others). The focus of this literature has mainly been on
the federal funds futures and Eurodollar futures contracts at very short horizons.
While we also use the federal funds futures and Eurodollar futures contracts as
95 measures of future monetary policy expectations, our approach differs in two
ways. First, in contrast to using the futures contracts at short horizons, we
use the 36-month horizon contracts in our benchmark model. This is needed
since as much as half of our sample covers the ZLB period where the short-term
expectations of the federal funds rate are essentially flat.

100 The second point of departure is the manner in which we identify monetary
policy shocks. The approach taken by recent contributions by Barakchian &
Crowe (2013) and Gertler & Karadi (2015) involves two steps: first, monetary
policy shocks are identified as changes in the expected federal funds rate on the
day of a Fed policy change or announcement itself; and then these shocks are
105 incorporated into a classic monetary policy VAR.⁶ In contrast, in our baseline

⁶Gertler & Karadi (2015) employ a set of proxy VARs (see Stock & Watson, 2012, and
Mertens & Ravn, 2013) including output, inflation and various interest rates and identify
the effects of monetary shocks on credit costs by employing the monetary policy surprises
as external instruments. Barakchian & Crowe (2013) extract the first common factor from
monetary surprises based on federal funds futures at various maturities (current month and

model we follow a more direct approach by identifying the monetary policy shock within a VAR using a combination of zero and sign restrictions. Specifically, since we are interested in the effects of news about future monetary policy decisions, we define a US monetary policy news shock as a shock that increases
110 the expectations of the future federal funds rate, while leaving the policy rate per se unchanged.

Our baseline results indicate that the effect of a US monetary policy news shock on aggregate (net) portfolio inflows to EMEs is economically small.⁷ We find that a monetary policy news shock that increases the expected federal funds
115 rate by 50 basis points (bps), indicating a future monetary policy tightening, results in aggregate portfolio flows declining by about 0.2% of GDP on impact. After a year, the cumulative decline in aggregate capital flows amounts to 0.5% of GDP. However, the effects vary considerably across countries in terms of magnitude with Hungary, South Africa, and Malaysia being potentially affected
120 most. While the estimated effects on portfolio flows are seemingly small, they can still be of relevance, as the experience from the taper tantrum episode has shown that changes in capital flows of a similar magnitude were associated with significant financial turmoil in EMEs.

Further, using the “taper tantrum” episode as an illustrative example, our
125 results show that the countries identified as being the most affected are also the ones that received greater financial flows prior to 2013 and saw greater capital outflows over May to August 2013. These results indicate that investors are likely to be better able to rebalance their portfolio allocations when recipient countries have large, liquid financial markets (Eichengreen & Gupta, 2015). The
130 estimated effects on capital flows are also found to be related to macroeconomic fundamentals such as real GDP growth and the fiscal balance/GDP. However, financial openness does not seem to be a key determinant of cross-country dif-

up to 5-month ahead). The factor is then used in a VAR containing output and inflation, which is identified using a Cholesky decomposition with the factor ordered last.

⁷Throughout the paper we refer to net inflows as flows and use the terms capital flows and portfolio flows interchangeably.

ferences in spillovers effects on capital flows. Also, we do not find any evidence supporting the view that countries with more flexible exchange rates experience
135 less spillover effects of US monetary policy news.

We also examine the robustness of our baseline results to our shock identification strategy by constructing the monetary policy news shock similar to Barakchian & Crowe (2013) and then incorporating this measure in the Bayesian VAR system. Our main results turn out to be quite robust to using this alter-
140 native approach which lends credibility to our direct method of using monetary policy expectations per se in the VAR.

The remainder of the paper is organized as follows. Section 2 motivates why expectations about future US monetary policy matter for capital flows to EMEs by looking at past Fed tightening cycles. Section 3 outlines the empirical
145 methodology, and section 4 provides a description of the data. Section 5 reports the estimation results, and section 6 discusses the key findings.

2. Capital flows and Fed monetary policy expectations: Lessons from the past

The Fed tightening cycles of previous decades provide some interesting in-
150 sights into the response of emerging-market capital flows to changes in US monetary policy expectations. The evidence from the 1994-95 and 2004-06 tightening episodes is mixed, however. After the 1990-91 recession, the Fed first increased rates in February 1994. This was largely unanticipated by markets. Over the next twelve months, the Fed raised its policy rate from 3% to 6%. Expectations
155 of the federal funds rate, as measured by the 12-quarters ahead Eurodollar futures, increased by about 200 bps over the same period (Figure 1).⁸ This, in turn, had significant spillovers on global financial markets. Portfolio flows to EMEs declined sharply after 1994.

[Insert Figure 1 here]

⁸Yields on 10-year Treasury bonds rose by around 150 bps (not shown).

160 In contrast, in the 2000s the Fed gradually increased the policy rate from
1% to 5.25% over 2004-06 by following a pre-announced schedule. The increase
in the 12-quarters ahead Eurodollar rates and the 10-yr Treasury bill rates was
small over this period, and had a limited initial impact on global financial mar-
kets compared with the 1994-95 episode. Portfolio inflows to EMEs continued
165 to be strong almost until the end of the Fed’s tightening cycle (Figure 1).

One of the key lessons learned from these episodes is the importance of
communication in influencing expectations of short-term interest rates and the
subsequent impact of monetary tightening on markets (IMF 2013). The ex-
perience from these tightening cycles also reinforces the argument of Hamilton
170 (2008) that the primary news for market participants is not what the Fed just
did but instead is the new information about what the Fed is going to do in
the future. As noted earlier, the role of private-sector expectations about future
monetary policy actions was also reflected in the significant financial market
volatility that followed the May and June 2013 tapering announcements by the
175 Fed. Thus, these episodes clearly illustrate the importance of accounting for
expectations in studying the impact of US monetary policy on portfolio flows
to EMEs. The key contribution of our paper lies in incorporating these ex-
pectations of future monetary policy and thus filling an important gap in the
literature on the linkages between US monetary policy changes and capital flows
180 to EMEs.

3. Empirical framework

To quantify the impact of US monetary policy news on net portfolio flows to
EMEs, we employ the following empirical strategy. First, we extract a common
factor from net portfolio flows to EMEs in our sample. Second, we estimate a
185 simple Bayesian VAR model containing US variables and the estimated capital
flow factor. We then identify a monetary policy news shock in this VAR frame-
work and assess its effects on portfolio flows to EMEs. Finally, we also examine
the impact on capital flows to EMEs using an alternative way to identify the

monetary policy news shock, which is based on monetary policy surprises (i.e.,
190 changes in monetary policy expectations on Fed announcement days).

3.1. Empirical model

Let \mathbf{W}_t denote a vector of N standardized capital flow series that have the following factor model representation:

$$\begin{aligned}\mathbf{W}_t &= \boldsymbol{\chi}_t + \boldsymbol{\xi}_t \\ &= \boldsymbol{\lambda}'\mathbf{F}_t + \boldsymbol{\xi}_t,\end{aligned}\tag{1}$$

where $\boldsymbol{\chi}_t$ is the common component of \mathbf{W}_t , which captures the co-movement
195 among the underlying capital flow series, while $\boldsymbol{\xi}_t$ is the idiosyncratic component, which can be interpreted as shocks affecting only individual portfolio flow series. \mathbf{F}_t is a $r \times 1$ vector of common or static factors, and $\boldsymbol{\lambda}$ is an $r \times N$ matrix of factor loadings. The factors, their loadings and the idiosyncratic errors are not observable, and have to be estimated from the data in practice. We use
200 the method of principal components to extract the first common factor of \mathbf{W}_t . We set the number of factors to one, since it is sufficient to explain most of the variation in our sample of portfolio flow series.⁹

Our baseline VAR model takes the following form:

$$\mathbf{y}_t = \boldsymbol{\alpha} + \mathbf{A}(\mathbf{L})\mathbf{y}_{t-1} + \mathbf{u}_t,\tag{2}$$

where \mathbf{y}_t is a vector of endogenous variables, $\boldsymbol{\alpha}$ a vector of constants, $\mathbf{A}(\mathbf{L})$
a matrix polynomial in the lag operator \mathbf{L} , and \mathbf{u}_t a vector of reduced-form
205 residuals, such that $\mathbf{u}_t \sim N(\mathbf{0}, \boldsymbol{\Omega})$. The reduced-form residuals can be related to the underlying structural shocks such that $\mathbf{u}_t = \mathbf{B}_0\boldsymbol{\epsilon}_t$, where \mathbf{B}_0 denotes the contemporaneous impact matrix, with $\boldsymbol{\epsilon} \sim N(\mathbf{0}, \mathbf{I})$, and $\boldsymbol{\Omega} = \mathbf{B}_0\mathbf{B}_0'$. We leave $\mathbf{A}(\mathbf{L})$ unrestricted.¹⁰

⁹The first common factor explains about 71% of the variation in the portfolio flow series, on average. See section 5.1 for details.

¹⁰One could also prevent movements in the capital flow factor from influencing US vari-

The vector of endogenous variables, \mathbf{y}_t , comprises six variables: the federal
210 funds rate, federal funds rate expectations, US inflation, US industrial pro-
duction growth, the level of the implied US stock market volatility index (or
VIX), and the common factor of portfolio flows. The choice of these variables
is motivated as follows.

The starting point for our model is a standard US monetary policy VAR
215 that includes US CPI inflation, the growth rate of US industrial production,
and the federal funds rate (see, for example, Bernanke et al. 2005 and Sims
1992). As discussed earlier and motivated in Section 2, to assess the effects of
monetary policy news, we augment the VAR with a measure of future monetary
policy expectations. To select an appropriate proxy for policy rate expectations,
220 we draw upon the growing literature focusing on the identification of monetary
policy shocks using financial market data. Our main measure of monetary policy
expectations is based on the federal funds futures and the Eurodollar futures
contracts at the 36-month horizon.¹¹ The argument for using federal funds
futures contracts to identify monetary policy shocks goes back to Rudebusch
225 (1998) and Kuttner (2001). More recent papers in this vein are Gertler &
Karadi (2015), Gürkaynak (2005), Hamilton (2008), among others. However,
we differ from these studies in using expectations at a long-run horizon since
expectations at shorter horizons (up to one year) are basically flat at the zero
lower bound (see Figure 2). Finally, we also use the 12-month-ahead as well as
230 24-month ahead futures to examine the sensitivity of our results to alternate
horizons.

[Insert Figure 2 here]

Further, we also include the implied stock market volatility in the United
States as proxied by the VIX.¹² The VIX is widely used in the literature as the

ables by imposing a zero restriction on the corresponding coefficients of $\mathbf{A}(\mathbf{L})$. Imposing this
restriction, however, does not alter our main findings.

¹¹Section 4.2 describes this measure in detail.

¹²Indices of implied stock market volatility are forward-looking measures of stock index
volatility computed based on option prices. These indices measure market expectations of

235 key indicator of risk aversion and a general proxy for financial turmoil, economic risk, and uncertainty. The literature has also found the VIX to be a significant determinant of capital flows to EMEs (for example, Ahmed & Zlate 2014; Forbes & Warnock 2012; Bruno & Shin 2014; and Lim et al. 2014).

240 Lastly, we include the portfolio flow factor in the VAR model, which is our key variable of interest. This allows us to calculate the effects of policy news on portfolio flows to individual countries in our sample, as well as on aggregate flows.

We include one lag in the VAR due to the relatively short sample size. However, our main results hold with alternative lag lengths.¹³ We estimate 245 equation (2) using standard Bayesian methods (i.e., Gibbs sampler) described in Koop & Korobilis (2010). Further details about the estimation procedure are provided in the not-for-publication online appendix. The VAR is estimated over the period January 2004 to May 2017.

3.2. Identification of a US monetary policy news shock

250 In our baseline VAR model, we define the monetary policy news shock as a shock that increases the expected federal funds rate while leaving the policy rate unchanged. Since it is a shock to the expectation of future federal funds rate, one could interpret this shock as a “future” monetary policy shock. That is, agents expect an increase in the fed funds rate by the Federal Reserve in the 255 future. The shock is identified by imposing a combination of sign restrictions and a single zero restriction on the contemporaneous impact matrix \mathbf{B}_0 underlying equation (2).

260 Table 1 summarizes the restrictions on the responses of the federal funds rate, federal funds futures, inflation, and economic activity measured by industrial production growth. The responses of the VIX and the capital flow factor are left unconstrained. The zero restriction is imposed on impact, while the sign

stock market volatility in the next 30 days.

¹³The not-for-publication online appendix shows results for alternative lag choices.

restrictions are imposed on impact and for five months.¹⁴

[Insert Table 1 here]

Specifically, we are interested in identifying monetary policy news that are
265 associated with future (anticipated) policy actions, that is, future tightenings.
Anticipated monetary policy shocks have similar economic effects as standard
unanticipated monetary policy shocks. A future monetary policy tightening by
the central bank has contractionary effects on the economy (see, e.g., Eggertsson
& Woodford, 2003, Laséen & Svensson, 2011, Milani & Treadwell, 2012). To
270 identify news about future tightening, we assume that a monetary policy news
shock, that increases federal funds rate expectations (indicating a future mon-
etary policy tightening) and leaves the federal funds rate unchanged, decreases
industrial production and prices.

In the case of informationally constrained agents (in the sense that the Fed-
275 eral Reserve has superior information about the economy), news about future
monetary policy tightening can signal either the anticipated monetary policy ac-
tion or stronger-than-anticipated economic fundamentals (see, e.g., Nakamura
& Steinsson, 2018 and Melosi, 2017, among others). The latter is the so-called
information shock where expectations of a monetary policy tightening reflect
280 expectations of stronger output growth, thus, implying positive responses of
economic activity and prices. In contrast, in the case of the anticipated mone-
etary policy shock identified in this paper, expectations of tightening reflect the
future policy action. Therefore, like Andrade & Ferroni (2018) and Jarociński &
Karadi (2020), the imposed sign restrictions on industrial production and prices
285 guarantee that our monetary policy news shock does not confound with the Fed
information shock.¹⁵

Finally, impulse-response functions that satisfy the sign and zero restrictions
are calculated using the procedure proposed by Baumeister & Benati (2013).

¹⁴Results are robust to imposing restrictions for shorter horizons.

¹⁵In a recent paper, Bauer & Swanson (2020) show that there is little role for a Fed infor-
mation effect. Our sign restrictions are consistent with the predictions of a standard model
of monetary policy without a Fed information effect channel.

They combine the method suggested by Rubio-Ramirez et al. (2010) for im-
posing sign restrictions with the imposition of a single zero restriction via a
290 deterministic rotation matrix. At each draw of the Gibbs sampler, the impact
matrix \mathbf{B}_0 is calculated and kept if the corresponding impulse responses satisfy
the sign and zero restrictions.

We test the robustness of our results to the choice of identifying restrictions
and model specification. In particular, we use policy rate expectations at the
295 12-month- and 24-month-ahead horizons and we also consider an alternative
measure of the monetary policy news shock based on high-frequency identifica-
tion, as described in the following section.

3.3. An alternative identification of the monetary policy news shock

300 One potential critique of using the federal funds rate expectations directly in
the VAR at monthly frequency is that other news shocks, distinct from monetary
policy news, such as news about total factor productivity, could influence federal
funds rate expectations. This would imply that our identified monetary policy
news shock could potentially be confounding other types of news shocks as
305 well. We address this concern by replacing the federal funds rate expectations
with a market-based measure of interest rate expectations obtained from high-
frequency data. This measure is based on federal funds rate futures surprises
calculated as the change in the expectations of the federal funds rate on the day
of a Fed announcement. It reflects the monetary policy news contained in the
310 announcement and is unlikely to be influenced by other macroeconomic news.
Following the important early contribution by Kuttner (2001), measuring the
surprise components of US monetary policy announcements by using futures
contracts has become popular in the recent academic literature (Barakchian &
Crowe 2013; Gertler & Karadi 2015; Gürkaynak 2005; Hamilton 2008, etc.).

In terms of notation, let t denote time, i.e., month. Let f_{t+j} be the settle-
ment price on the FOMC day in month t for interest rate futures (either the
federal funds or the Eurodollar futures) expiring in month $t + j$. $f_{t+j,-1}$ is the
corresponding settlement price for the day prior to the FOMC meeting. The

unexpected movement (surprise) in the target federal funds rate anticipated for month $t + j$ is denoted by $E_t \dot{i}_{t+j}$ which can be expressed as:¹⁶

$$E_t \dot{i}_{t+j} = f_{t+j} - f_{t+j,-1} \quad (3)$$

315 Table 2 lists the monetary policy announcement days we consider. Since the surprise series corresponds to announcement days, it must be converted to a monthly frequency. To do this, we assign each ‘surprise’ to the month in which the corresponding FOMC meeting occurred. If there are two meetings in a month, we sum the ‘surprise’. If there are no meetings in a month, we assume
320 the ‘surprise’ is zero for that month (see Romer & Romer, 2004).

[Insert Table 2 here]

As mentioned, we use the federal funds futures and Eurodollar futures contracts with a 36-month maturity since expectations of the federal funds rate based on contracts with shorter maturity are essentially flat at the ZLB. Section
325 4.2 describes the futures data in detail. Finally, we incorporate this constructed monetary policy surprise series in the baseline VAR (Equation 2) and exclude the federal funds futures contracts series.¹⁷ In the spirit of Barakchian & Crowe (2013), this series is included in the VAR in cumulative terms.¹⁸ We use the taper tantrum episode as an illustrative example and consider a shock of 25
330 bps to the cumulated surprise measure based on the change in this series from April to June 2013. The identifying restrictions on the responses of the other variables are the same as those in the baseline VAR (see Table 1).

¹⁶Since we do not focus on the surprise in the futures rate for the current month, we do not scale the surprises based on the number of days in the month and the timing of the FOMC meeting in that particular month. In doing so, we follow Gertler & Karadi (2015) and Barakchian & Crowe (2013).

¹⁷Thus the number of variables in the resulting VAR is still 6.

¹⁸Barakchian & Crowe (2013) and Romer & Romer (2004) argue for including the surprise measure in cumulative terms as it is the level and not the change in policy that is relevant.

4. Data

4.1. Capital flows

335 We use data on net portfolio flows from the Emerging Portfolio Fund Research (EPFR) Global database. The database contains weekly portfolio investment (net) flows by more than 14,000 (mutual and ETF) equity funds and more than 7,000 (mutual and ETF) bond funds. We use EPFR data since it is available at higher frequencies than the balance of payments data, allowing
340 us to examine movements in portfolio flows at a monthly frequency. While the database represents up to 20% of the market capitalization in equity and in bonds for most countries, it closely matches portfolio flows in the balance of payments data (as shown in, e.g., Fratzscher, 2012 and Jotikasthira et al., 2012) and is being increasingly used in academic research on capital flows (Forbes et al.,
345 2016 and Lo Duca, 2012, among many others). In addition, the EPFR data are used widely in the financial industry as a timely, high-frequency indicator of movements in portfolio flows.

Our sample covers the following 23 emerging markets: Argentina, Brazil, Bulgaria, Chile, China, Colombia, Czech Republic, Hungary, India, Indonesia,
350 s, Korea, Malaysia, Mexico, Peru, the Philippines, Poland, Romania, Russia, South Africa, Thailand, Turkey, Ukraine and Venezuela. We use monthly data over the period January 2004 to May 2017.¹⁹

4.2. US variables

We use the federal funds futures contracts as a measure of expectations of
355 the future federal funds rate. The federal funds futures contract price represents the market opinion of the average daily federal funds effective rate, as calculated and reported by the Federal Reserve Bank of New York, for a given calendar

¹⁹We also estimated our VAR model excluding the taper tantrum episode from the sample. Specifically, we excluded May 2013 as the observation corresponding to the taper tantrum episode. Results are shown in Figure 13 of the not-for-publication online appendix. The response of the capital flow factors is basically unaffected by the exclusion of the taper tantrum episode and so are the aggregate and country-specific effects on capital flows.

month. Federal funds futures have long been regarded as an effective means of tracking market expectations of Fed monetary policy actions.²⁰

360 As mentioned, in our baseline specification, we use the federal funds future contracts at the 36-month horizon as a measure of expectations about future short-term interest rates. We chose to use expectations of future monetary policy at a long-term horizon since this avoids problems with the ZLB; i.e., expectations of the federal funds rate are essentially flat at short-term horizons
365 (Figure 2). Since up to 36-month ahead futures only exist from January 2011 onward, we use the Eurodollar futures contracts for the period prior to this. The correlation between the two series is 0.99 over the period January 2011 to May 2017. Below, we briefly describe how the two series are combined.

While the federal funds futures contracts extend up to the first 36 calendar
370 months in the future, the Eurodollar futures contracts mature during the months of March, June, September, and December, extending outward 10 years into the future. Thus, in order to combine the two series we use the federal funds futures contracts expiring in March, June, September, and December. In other words, for the first month of any given quarter we use the fed funds futures contracts
375 at the 36-month horizon; for the second month of the quarter, we use the 35-month ahead futures contracts, and; for the third month of a quarter, we use the 34-month ahead futures contracts. Finally, the data on federal funds and Eurodollar futures contracts are obtained from Bloomberg. Daily futures rates are transformed into monthly frequency by taking the average over the month.
380 For robustness, we also present results based on federal funds futures at the 12-month and 24-month horizon.

US inflation is measured as the first difference of the log of the consumer price index and US industrial production growth is measured as the first difference of the log of US industrial production. These data, along with the effective federal

²⁰Futures for the federal funds rate started trading in the late 1980s but only up to a 6-month ahead horizon. Meaningful trading volumes of up to the 24-month ahead futures contracts only began in 2004, while those for the 36-month ahead futures are available since 2011.

385 funds rate and the VIX are taken from Haver Analytics. The latter data is transformed to monthly frequency by taking the monthly average.

5. Results

5.1. Common factor of capital flows to EMEs

We first examine the extent to which capital flows to emerging markets co-
390 move.²¹ Figure 3 plots the estimated first principal component for the capital flow series. The common component tracks aggregate net capital flows very well. Moreover, the common component explains about 71% of the variation in country-specific flows, on average, which lends credibility to our approach of including the common factor into the VAR model to obtain the impact of monetary policy news shocks on individual countries (see Table 3). It also suggests
395 that common (i.e., global) factors have played a much larger role than idiosyncratic factors in shaping fund flows into emerging-market bonds and equities in recent years. In some cases, however, country-specific factors matter more. For example, the common factor explains less than 40% of the variation for India, Thailand and Bulgaria .
400

[Insert Figure 3 here]

[Insert Table 3 here]

5.2. The effect of a US monetary policy news shock

As discussed in the previous section, we identify a US monetary policy news
405 shock as a shock that increases the expected federal funds rate while leaving the policy rate unchanged. We consider a shock of 50 bps, which is in line with the increase in the federal funds futures following former Chairman Bernanke's testimony on 22 May 2013.²² Figure 4 shows the impulse-response functions

²¹Recent literature has also documented co-movement of capital flows (for example, Förster et al. 2014; Fratzscher 2012).

²²The federal funds future contract at the 36-month horizon increased by 52 bps from April to June 2013. A 50 bps shock represents a relatively small-sized shock, equivalent to 0.3 times the standard deviation of the series.

for this shock. The shock leaves the federal funds rate unchanged on impact
410 as imposed by the zero restriction. Monetary policy expectations –as measured
by the federal funds futures– increase, and this effect is significant for about 10
months. The monetary policy news shock decreases US inflation and industrial
production growth by about 0.4% and 0.9% on impact, respectively. Both in-
415 flation and industrial production growth return to their pre-shock levels after
about 15 months. Stock market volatility, as measured by the VIX, increases
for about 20 months following the shock, although the response is not significant
on impact.

[Insert Figure 4 here]

Turning to the response of portfolio flows, which is the focus of this paper,
420 we find that the monetary policy news shock decreases the portfolio flow factor
significantly. Capital flows return to their pre-shock level after 9-10 months
following the shock. By combining equations (1) and (2), we obtain the effects on
portfolio flows to individual countries and then sum these effects across countries
to obtain the response of aggregate capital flows. In order to assess the economic
425 size of the effects on aggregate as well as country-specific flows, we scale these
responses by the level (nominal) of 2016 GDP for the respective countries. Table
4 reports the results. We find that the shock corresponding to an increase in
futures of 50 bps decreases aggregate capital flows to GDP by 0.15% on impact
(see column 2, Table 4). After three months, the cumulative decline in aggregate
430 capital flows amounts to 0.35% of GDP. One year after the shock, aggregate
capital flows decrease by 0.5% of GDP on a cumulative basis.

[Insert Table 4 here]

Table 4 also shows the effects on capital flows to individual countries fol-
lowing the monetary policy news shock of 50 bps. The impact effects (column
435 “h=0”) vary considerably across countries in terms of magnitude. South Africa,
Hungary and Malaysia are found to be affected most, with impact effects ranging
from about -0.6% of GDP to about -0.4% of GDP. However, the corresponding

effects on Bulgaria, China, Venezuela and Romania are relatively small, ranging from roughly -0.1% to -0.03% of GDP. The three-month cumulative effects
440 on capital flows range from -0.1% of GDP (Bulgaria) to -1.4% (South Africa). After one year, the cumulative decline in capital flows is as high as nearly 2% of GDP in the case of South Africa and 1.7% of GDP in the case of Hungary, while the corresponding figures for Bulgaria, Venezuela and Romania are quite small (between -0.3% and -0.1% of GDP).

445 Lastly, as the EPFR data base contains portfolio investment by equity and bond funds, we can decompose portfolio flows into bond and equity flows and examine the effects of monetary policy news shocks on these flows. In order to do so, we estimate two separate baseline VARs as specified in equation (2): one including the common factor extracted from equity flows as an endogenous
450 variable and the other with the common factor of bond flows. Equity flows are slightly more volatile compared to bond flows in our sample, with the common factor explaining only 55%, on average, of the variation in flows compared to an average of 62% for bond flows. Looking at the impulse-response functions for a 50 bps monetary policy news shock, the bond flows factor shows a bigger
455 drop on impact compared to the equity flows factor in response to the shock (see Figure 5). The response of bond flows also appears slightly more persistent. However, since the magnitude of equity flows is larger than that of bond flows in our sample, the cumulative 3-month response of aggregate bond flows as a share of GDP is actually smaller (-0.11%) than that of equity flows (-0.17%)
460 (see Table 4). These results are qualitatively similar with those in Fratzscher et al. (2016) and Lim et al. (2014). The latter paper concludes that bond flows appear to be affected via the portfolio balance channel while equity flows are not which, in turn, yields a bigger response in bond flows.

[Insert Figure 5 here]

465 5.3. Country characteristics and the effects of US monetary policy news

Recent studies have shown that sharp movements in capital flows are driven not only by external factors, such as changes in expectations for advanced econ-

omy monetary policies, but also by domestic macroeconomic fundamentals (see, for example, Forbes & Warnock, 2012, Edwards, 2007 and Ghosh et al., 2014).
470 To shed some light on the potential drivers for the variation in the effects on capital flows across countries, we investigate the association between the estimated effects and a country’s macroeconomic characteristics. In particular, we consider the behavior of capital flows during the 2013 taper tantrum episode, the role of financial openness and the exchange rate regime, as well as economic
475 fundamentals such as GDP growth, reserves, external debt, fiscal and current account balances, and sovereign credit default swap (CDS) spreads (see, for example, Eichengreen & Gupta, 2015 and Nechio, 2014 for a similar approach). We use the sample average of country characteristics to assess their relation to 3-month cumulative effects on capital flows.²³ Table 5 presents the sample
480 averages of the country characteristics considered, and Table 6 shows the results for cross-country regressions of the estimated effects on capital flows on each country characteristic. The online appendix presents further insights into these results in the form of scatterplots depicting the relationship between country characteristics and the estimated effects on capital flows.

485 [Insert Table 5 here]

[Insert Table 6 here]

First, we study the consistency of our estimated effects with the observed behaviour in capital markets during the taper tantrum episode in May 2013. In particular, we examine the correlation of the estimated country-specific effects
490 with capital inflows prior to 2013 and the outflows following the taper tantrum. Running a simple cross-section regression of the estimated effects on the financial inflows from 2010-12 as a share of GDP, we find a negative and significant coefficient with a R-squared of 0.2. Therefore, the countries we identify as being potentially most affected are the ones that received greater financial inflows

²³This is in line with the approach in Miniane & Rogers (2007) and Dedola et al. (2017) in which point impulse response estimates are directly regressed on average characteristics over the sample.

495 prior to 2013. These results are consistent with recent findings in Eichengreen
& Gupta (2015).

Further, we analyze the relation between the estimated effect on capital flows
and the capital outflows experienced over end-May to August 2013 following
former Chairman Bernanke’s testimony. Regressing the estimated effects on the
500 capital outflows over end-May to August 2013 shows a statistically significant
positive relation between these variables with a R-squared of 0.7 (see Table 6).
Again, there seems to be a strong association between the countries that are
identified by the model as being most affected and the ones that saw greater
outflows over May to August 2013.

505 This raises the obvious question whether the degree of capital openness po-
tentially could explain the observed cross-country differences in the estimated
effects on capital flows. More financially open and interconnected economies
may experience a larger impact of changes in global risk appetite and, hence,
US monetary policy news. We use the widely applied financial openness mea-
510 sure of Chinn & Ito (2006, 2008).²⁴ The Chinn-Ito index is normalized between
zero and one, with higher values of the index indicating greater openness to
cross-border capital transactions. We do not observe any apparent relationship
between cross-country effects and financial openness: The R^2 of the regression
is 0 and the scatterplot (Figure 4 of the not-for-publication online appendix)
515 shows a flat regression line. While this seems to be a surprising result at first,
it is in line with former findings in the literature studying spillovers of conven-
tional monetary policy shocks. For example, Miniane & Rogers (2007) show
that countries with less open capital accounts do not exhibit smaller responses
in exchange rates and interest rates following US monetary shocks.

520 Further, countries that peg their currency to the US dollar could experience
stronger spillovers. We use the standard Shambaugh (2004) classification since it
contains information on the base country which allows us to determine whether
countries pegged their currency to the US dollar at each point in time over

²⁴We obtain the index from http://web.pdx.edu/~ito/trilemma_indexes.htm.

our sample.²⁵ There are only seven countries (Argentina, China, Malaysia,
525 Philippines, Thailand, Ukraine and Venezuela) that pegged their currency to
the US dollar at some point in time over our sample (see Table 5). Countries
such as Hungary and South Africa, whose capital flows seem to be affected most
by monetary policy news shocks, did not peg their currency to the United States.
However, countries such as China or Venezuela that pegged their currency to
530 the US dollar for most of our sample seem to experience relatively small effects
on capital flows.²⁶

Further, the results in Table 6 highlight a slight positive but insignificant
relationship between effects on capital flows and the dollar peg. Therefore, we do
not find any evidence for the standard exchange rate channel, i.e., countries with
535 more flexible exchange rates experiencing less spillover effects. Nevertheless,
given that our sample covers as much as half of the ZLB period and focuses on
monetary policy news, our results seem to support the findings of Eichengreen &
Gupta (2015): the countries least able to limit the impact on their real exchange
rates during US quantitative easing were the same ones to experience large
540 and sometimes uncomfortable real exchange rate reversals (and larger capital
outflows) during the tapering episode.

We also investigate whether country-specific macroeconomic fundamentals
can be related to the estimated effects on capital flows since capital could be
more likely to flow to less risky or less vulnerable emerging market countries
545 (“flight-to-quality”). We find a positive and significant relation between effects
on capital flows and real GDP growth suggesting that countries with higher
average GDP growth tend to experience smaller negative effects on capital flows.
Further, we do not find evidence for a strong relation between cross-country
effects on capital flows and average CDS spreads (see Table 6).²⁷

²⁵The data for the Shambaugh (2004) classification is obtained from <https://www2.gwu.edu/~iiep/about/faculty/jshambaugh/data.cfm> and ends in 2014. Therefore, we use the sample average of the Dollar peg over the period 2004-2014.

²⁶Note that, at the same time, China and Venezuela have essentially closed capital accounts which reduces the volatility of capital flows to these countries.

²⁷Note that we excluded CDS spreads of Argentina, Ukraine and Venezuela as their CDS

550 Turning to indicators of external sustainability, we, first, regress the estimated effects on capital flows on external debt to GDP ratios averaged over the sample. Countries with higher external debt to GDP ratios are the ones with higher estimated impacts on capital flows (although not significantly). Similarly, the stock of reserves relative to nominal GDP appears to be positively
555 associated with the effect on capital flows. While this seems to suggest that capital flows of countries with lower reserves seem to be affected more after a monetary policy news shock, the relation is not statistically significant and the R^2 is only 0.03. In addition, we find evidence for a (marginally significant) positive relation between the fiscal balance and cross-country effects on capital
560 flows. Emerging markets with larger fiscal deficits are the ones with higher estimated effects on capital flows. Finally, we regress the estimated effects of capital flows on the current account balance (% of GDP). We find a negative, although not significant, coefficient suggesting that higher current account deficits are associated with a smaller impact on capital flows from monetary policy news
565 shocks. This counterintuitive result seems to be driven by Malaysia which, as one of the countries we identified to be the most affected, runs a large current account surplus of 9.3% of GDP on average over our sample period (see Figure 10 of the not-for-publication online appendix).

Overall, we find that the countries we identify as being the most affected also
570 experienced larger volumes of capital in- and outflows before and after the taper tantrum episode, respectively. Further, we find evidence that macroeconomic performance and external vulnerabilities may matter. However, the common candidate to explain cross-country differences in spillover effects, i.e., financial openness, does not seem to be associated with differences in effects on capital
575 flows over our sample period.

spreads data is only available since 2012 and spreads average about 1000 basis points.

5.4. Robustness

We assess the robustness of our VAR results to alternative horizons of the federal funds futures and the alternative shock identification based on federal funds future surprises described in Section 3.3.²⁸

580 First, Figure 6 shows the impulse responses for the robustness exercise in which we re-estimate the baseline VAR with the federal funds futures contracts at the 24-month and 12-month ahead horizons. As can be seen, the response of US inflation, industrial production, VIX and the capital flow factor are robust to using the futures contracts at these horizons.

585 [Insert Figure 6 here]

Figure 7 shows the impulse responses for a shock of 25 bps to the cumulated surprise series. As before, the federal funds rate is unchanged on impact, in line with the zero restriction. The response of the federal funds futures surprise series is qualitatively very similar to that of the federal funds futures in our baseline VAR. The shock decreases US inflation and industrial production growth by 590 about 0.7% and 1.3% on impact, respectively, with both variables responding slightly more than in the baseline VAR. Both inflation and industrial production growth return to their pre-shock levels after about 15 months. The response of the VIX is once again not significant on impact but qualitatively consistent with that in the baseline VAR. Finally, the response of the capital flow factor is 595 consistent with that in the baseline VAR, although it is not significant on impact. Since our results are quite consistent with those obtained in the baseline VAR, constructing monetary policy surprises around Fed announcement dates does not seem to bring any additional insights. Thus, including expectations about 600 monetary policy directly in the VAR is appealing as a straightforward and direct way of assessing the effects of US monetary policy news shocks.

²⁸In an unreported exercise, we also use a survey-based measure of federal funds rate expectations (at the five-quarters-ahead horizon) coming from the Blue Chip Financial Forecasts. VAR results based on these survey-based expectations are similar to those obtained using market-based expectations. Results are available upon request.

[Insert Figure 7 here]

6. Conclusion

This paper uses a Bayesian VAR model to examine the potential effects of
605 a US monetary policy news shock on portfolio flows to major EMEs. In our
benchmark model, we define the US monetary policy news shock as a shock
that increases expectations of the federal funds rate while leaving the policy
rate per se unchanged. Results show that the effect of a US monetary policy
news shock on aggregate portfolio flows as a share of GDP is economically small,
610 amounting to 0.5% of GDP after a year. However, the effects vary considerably
across countries with Hungary, South Africa, and Malaysia being potentially
affected the most. There is a strong association between the countries that
are identified by our analysis as being the most affected and the ones that re-
ceived greater financial flows prior to 2013 and saw greater capital outflows over
615 May to August 2013 following the Bernanke testimony. We also find evidence
of a significant relationship between the estimated country-specific effects and
macroeconomic fundamentals such as real GDP growth and the fiscal balance.
However, financial openness does not seem to be a key determinant of cross-
country differences in spillovers effects on capital flows. Our baseline results are
620 robust to alternative shock identification strategies and model specification.

The main policy implication of our findings is that the effects of US monetary
policy news shocks are likely to be asymmetric across countries. It is reasonable
to expect episodes of volatility in global financial markets, especially capital flow
reversals in EMEs, as the Fed proceeds on its eventual path of monetary policy
625 normalization. Higher bond yields will trigger portfolio rebalancing, the effects
of which could well be amplified in the presence of market imperfections. Thus,
the effects of US monetary policy news on EMEs will depend on the extent of
their vulnerabilities.

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| Variable | Restriction | |
|-----------------------|-------------|-----------------------|
| Federal funds rate | 0 | ($h = 0$) |
| Federal funds futures | + | ($h = 0, \dots, 5$) |
| Inflation | - | ($h = 0, \dots, 5$) |
| IP growth | - | ($h = 0, \dots, 5$) |
| VIX | ? | |
| Capital flow factor | ? | |

Table 1: Identification restrictions for the baseline VAR. Notes: IP denotes industrial production. “?” means that the variable is left unconstrained. h denotes the horizon (in months) of imposed restrictions.

| Year | Day | Description |
|------|---|---|
| 2004 | 1/28, 3/16, 5/4, 6/30, 8/10, 9/21, 11/10,12/14 | FOMC Meetings |
| 2005 | 2/02, 3/22, 5/03, 6/30,8/09,9/20,11/01, 12/13 | FOMC Meetings |
| 2006 | 01/31, 03/28, 05/10, 06/29, 08/08, 09/20,10/25, 12/12 | FOMC Meetings |
| 2007 | 1/31, 3/21, 5/09,6/28,8/07,8/17, 9/18, 10/31, 12/11 | FOMC Meetings |
| 2008 | 1/22, 1/30, 3/18, 4/30,6/25, 08/05, 09/16 | FOMC Meetings |
| | 10/08 | FOMC Meetings |
| | 10/29 | FOMC Meetings |
| | 11/25 | Fed announces purchases of MBS and Agency bonds |
| | 12/01 | Bernanke states Treasuries may be purchased |
| | 12/16 | FOMC Meeting |
| 2009 | 28/01, 3/18, 4/29, 6/24, 8/12, 9/23, 11/4, 12/16 | FOMC Meetings |
| 2010 | 27/01, 3/16, 4/28, 6/23, 8/10 | FOMC Meetings |
| | 8/27 | Bernanke Speech at Jackson Hole |
| | 9/21 | FOMC Meeting |
| | 10/15 | Bernanke Speech at Boston Fed |
| | 11/03, 12/14 | FOMC Meetings |
| 2011 | 1/26, 3/15, 4/27, 6/22, 8/9 | FOMC Meetings |
| | 8/26 | Bernanke Speech at Jackson Hole |
| | 9/21, 11/2, 12/13 | FOMC Meetings |
| 2012 | 01/25,3/13, 4/25, 6/20,8/1 | FOMC Meetings |
| | 8/31 | Bernanke Speech at Jackson Hole |
| | 9/13, 10/24, 12/12 | FOMC Meetings |
| 2013 | 1/30, 3/20, 5/1 | FOMC Meetings |
| | 5/22 | Bernanke Testimony |
| | 6/19, 7/31, 9/18, 10/30, 12/18 | FOMC Meetings |
| 2014 | 1/29, 3/12, 4/30, 6/18, 7/30, 9/17, 10/29, 12/17 | FOMC Meeting |
| 2015 | 1/28, 3/18, 4/29, 6/17, 7/29, 9/17, 10/28, 12/16 | FOMC Meeting |
| 2016 | 1/27, 3/16, 4/27, 6/15, 7/27, 9/12, 11/2, 12/14 | FOMC Meeting |
| 2017 | 2/1, 3/15, 5/3 | FOMC Meeting |

Table 2: Dates of U.S. monetary policy announcements.

| | | | |
|----------------|----|--------------|----|
| Argentina | 67 | Mexico | 89 |
| Brazil | 74 | Peru | 92 |
| Bulgaria | 39 | Philippines | 89 |
| Chile | 81 | Poland | 88 |
| China | 44 | Romania | 60 |
| Colombia | 66 | Russia | 70 |
| Czech Republic | 46 | South Africa | 90 |
| Hungary | 85 | Thailand | 37 |
| India | 27 | Turkey | 93 |
| Indonesia | 90 | Ukraine | 85 |
| Korea | 58 | Venezuela | 73 |
| Malaysia | 90 | | |
| Average | 71 | | |

Table 3: Variation in capital flows explained by the common component (in percent).

| | Effect | | | | | Cumulative Effect | | |
|-------------------|--------|-------|-------|-------|-------|-------------------|---------|----------|
| | h=0 | h=1 | h=2 | h=5 | h=11 | 3-month | 6-month | 12-month |
| Argentina | -0.08 | -0.07 | -0.04 | -0.01 | 0.00 | -0.19 | -0.24 | -0.27 |
| Brazil | -0.29 | -0.24 | -0.16 | -0.04 | -0.01 | -0.69 | -0.90 | -0.99 |
| Bulgaria | -0.03 | -0.03 | -0.02 | 0.00 | 0.00 | -0.08 | -0.10 | -0.11 |
| Chile | -0.21 | -0.17 | -0.11 | -0.03 | -0.01 | -0.49 | -0.64 | -0.71 |
| China | -0.06 | -0.05 | -0.03 | -0.01 | 0.00 | -0.14 | -0.18 | -0.20 |
| Colombia | -0.24 | -0.20 | -0.13 | -0.04 | -0.01 | -0.58 | -0.75 | -0.83 |
| Czech Republic | -0.09 | -0.08 | -0.05 | -0.01 | 0.00 | -0.22 | -0.28 | -0.31 |
| Hungary | -0.50 | -0.42 | -0.27 | -0.07 | -0.01 | -1.19 | -1.55 | -1.72 |
| India | -0.12 | -0.10 | -0.07 | -0.02 | 0.00 | -0.29 | -0.37 | -0.42 |
| Indonesia | -0.19 | -0.16 | -0.11 | -0.03 | -0.01 | -0.46 | -0.60 | -0.66 |
| Korea | -0.21 | -0.17 | -0.11 | -0.03 | -0.01 | -0.49 | -0.64 | -0.71 |
| Malaysia | -0.39 | -0.33 | -0.21 | -0.06 | -0.01 | -0.93 | -1.20 | -1.33 |
| Mexico | -0.27 | -0.22 | -0.14 | -0.04 | -0.01 | -0.63 | -0.82 | -0.91 |
| Peru | -0.31 | -0.26 | -0.17 | -0.05 | -0.01 | -0.73 | -0.95 | -1.05 |
| Philippines | -0.23 | -0.19 | -0.13 | -0.03 | -0.01 | -0.55 | -0.71 | -0.79 |
| Poland | -0.24 | -0.20 | -0.13 | -0.04 | -0.01 | -0.57 | -0.74 | -0.82 |
| Romania | -0.09 | -0.08 | -0.05 | -0.01 | 0.00 | -0.22 | -0.29 | -0.32 |
| Russia | -0.24 | -0.20 | -0.13 | -0.04 | -0.01 | -0.57 | -0.74 | -0.82 |
| South Africa | -0.58 | -0.49 | -0.32 | -0.09 | -0.02 | -1.39 | -1.80 | -1.99 |
| Thailand | -0.24 | -0.20 | -0.13 | -0.03 | -0.01 | -0.56 | -0.73 | -0.81 |
| Turkey | -0.18 | -0.15 | -0.10 | -0.03 | -0.01 | -0.42 | -0.55 | -0.61 |
| Ukraine | -0.29 | -0.24 | -0.16 | -0.04 | -0.01 | -0.68 | -0.88 | -0.98 |
| Venezuela | -0.09 | -0.07 | -0.05 | -0.01 | 0.00 | -0.21 | -0.27 | -0.29 |
| Aggregate | -0.15 | -0.12 | -0.08 | -0.02 | 0.00 | -0.35 | -0.45 | -0.50 |
| Agg. bond flows | -0.04 | -0.04 | -0.03 | -0.01 | 0.00 | -0.11 | -0.17 | -0.20 |
| Agg. equity flows | -0.09 | -0.06 | -0.03 | 0.00 | 0.00 | -0.17 | -0.19 | -0.20 |

Table 4: Effects of a monetary policy news shock on individual countries' capital flows (percent of 2016 GDP).

| | In | Out | Open | Peg | GDP | CDS | Debt | Reserves | Budget | CA |
|----------------|-------|-------|------|------|-------|---------|--------|----------|--------|-------|
| Argentina | 3.10 | -0.14 | 0.19 | 0.18 | 3.71 | 1046.78 | 43.02 | 0.39 | -0.88 | -0.01 |
| Brazil | 9.50 | -1.76 | 0.42 | 0.00 | 2.53 | 198.72 | 15.67 | 0.52 | -3.36 | -1.52 |
| Bulgaria | 0.27 | -0.01 | 0.89 | 0.00 | 3.34 | 167.96 | 82.10 | 1.41 | 0.00 | -6.02 |
| Chile | 15.30 | -1.31 | 0.83 | 0.00 | 3.92 | 75.07 | 44.49 | 0.56 | 1.27 | -0.33 |
| China | 1.21 | -0.35 | 0.17 | 0.73 | 9.47 | 72.75 | 13.64 | 1.45 | 5.71 | -0.61 |
| Colombia | 8.60 | -0.43 | 0.44 | 0.00 | 4.35 | 184.60 | 25.99 | 0.45 | -2.82 | -3.07 |
| Czech Republic | 7.20 | -0.23 | 1.00 | 0.00 | 2.75 | 73.45 | 52.62 | 0.98 | -2.05 | -1.59 |
| Hungary | 6.46 | -2.09 | 1.00 | 0.00 | 1.66 | 203.90 | 110.38 | 1.03 | -3.93 | -1.71 |
| India | 6.60 | -0.47 | 0.17 | 0.00 | 7.61 | 167.30 | 20.16 | 0.72 | -4.59 | -2.40 |
| Indonesia | 5.80 | -0.95 | 0.57 | 0.00 | 5.65 | 207.29 | 31.89 | 0.46 | -1.39 | -0.37 |
| Korea | 8.90 | -0.42 | 0.60 | 0.00 | 3.63 | 81.66 | 28.60 | 0.99 | 0.83 | 3.33 |
| Malaysia | 20.60 | -2.17 | 0.36 | 0.18 | 5.05 | 93.51 | 61.52 | 1.68 | -4.02 | 9.34 |
| Mexico | 15.10 | -1.27 | 0.70 | 0.00 | 2.44 | 121.84 | 25.34 | 0.44 | -1.56 | -1.34 |
| Peru | 6.20 | -1.76 | 1.00 | 0.00 | 5.57 | 131.14 | 32.87 | 1.06 | 0.00 | -1.77 |
| Philippines | 10.90 | -1.40 | 0.36 | 0.09 | 5.63 | 196.76 | 35.96 | 0.99 | -1.97 | 2.96 |
| Poland | 16.00 | -1.14 | 0.49 | 0.00 | 3.87 | 92.19 | 59.95 | 0.71 | -2.20 | -3.52 |
| Romania | 5.62 | -0.65 | 0.97 | 0.00 | 3.61 | 185.37 | 56.69 | 0.71 | -3.46 | -5.62 |
| Russia | 5.10 | -1.56 | 0.55 | 0.00 | 3.00 | 200.87 | 31.48 | 1.01 | 0.92 | 5.00 |
| South Africa | 12.20 | -2.16 | 0.17 | 0.00 | 2.79 | 160.79 | 30.45 | 0.12 | -2.59 | -0.98 |
| Thailand | 17.20 | -1.84 | 0.24 | 0.09 | 3.68 | 98.53 | 34.13 | 1.52 | -1.36 | 3.02 |
| Turkey | 18.60 | -1.05 | 0.36 | 0.00 | 5.72 | 233.91 | 39.85 | 0.46 | -1.80 | -4.95 |
| Ukraine | 18.30 | -1.95 | 0.06 | 0.64 | -0.13 | 858.54 | 76.68 | 0.68 | -2.73 | -2.20 |
| Venezuela | 7.50 | -0.60 | 0.10 | 1.00 | 3.31 | 1000.05 | 28.44 | N/A | -3.63 | 1.66 |

Table 5: Country characteristics. Notes: In is the average of (net) capital flows over the period 2010-2010 (% of GDP). Out is the average of (net) capital flows over May-Sept 2013 (% of GDP). Open is the Chinn & Ito (2006, 2008) capital openness index averaged over the sample. Peg is the U.S. dollar peg index based on Shambaugh (2004) and averaged over the sample. GDP is the sample average of real GDP growth (%). CDS is the sample average of sovereign CDS spreads (basis points). Debt is the sample average of external debt to GDP ratios. Reserves are the sample-average reserves to GDP ratios. Budget is the sample average of the budget balances (% of GDP). CA is the sample average of current account balances (% of GDP).

| Country Characteristic | Coefficient | P-value | R^2 | adj. R^2 |
|-------------------------|-------------|---------|-------|------------|
| Inflows 2010-2012 | -0.02 | 0.03 | 0.20 | 0.16 |
| Outflows May-Sept 2013 | 0.39 | 0.00 | 0.72 | 0.71 |
| Financial Openness | 0.01 | 0.96 | 0.00 | 0.00 |
| Dollar Peg | 0.31 | 0.22 | 0.07 | 0.02 |
| Real GDP Growth | 0.06 | 0.09 | 0.13 | 0.09 |
| Sovereign CDS Spreads | 0.00 | 0.33 | 0.05 | 0.00 |
| External Debt | -0.00 | 0.32 | 0.05 | 0.00 |
| Reserves/ Nominal GDP | 0.13 | 0.46 | 0.03 | -0.02 |
| Budget Balance | 0.05 | 0.12 | 0.11 | 0.07 |
| Current Account Balance | -0.02 | 0.28 | 0.06 | 0.01 |

Table 6: Cross-section regressions of effects of monetary policy news shocks on country characteristics. Notes: Cross-country regressions of the estimated 3-month cumulative effects of a 50 bps monetary policy news shock on the sample average of one country characteristic at a time.

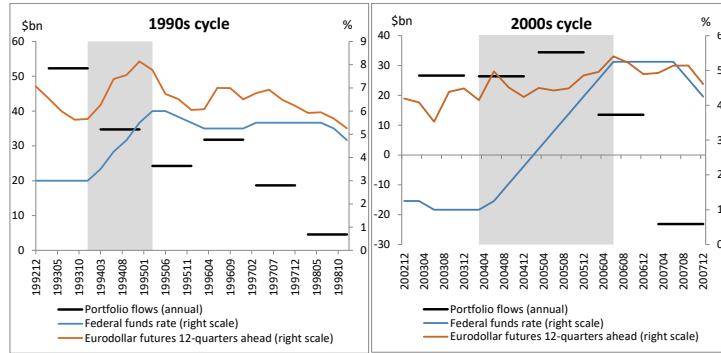


Figure 1: Portfolio flows to EMEs during previous Fed tightening cycles. Notes: The federal funds rate and the 12-quarter-ahead Eurodollar futures are shown at quarterly frequency. The capital flows data are taken from the IIF and are only available at annual frequency for the time period shown in the graph. Portfolio flows represent the sum of flows to all the EMEs in our sample. The gray shaded areas represent the period over which the Federal Reserve raised the federal funds rate.

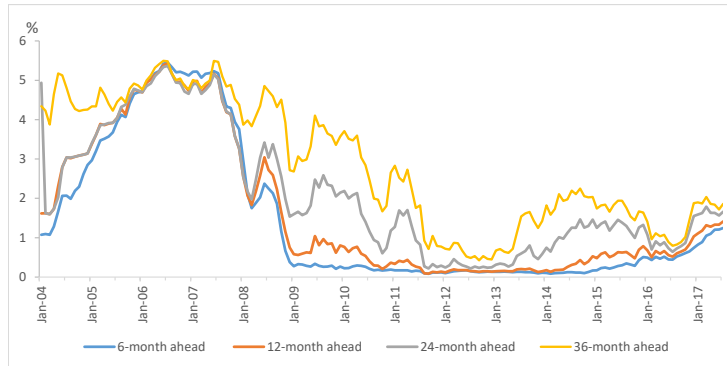


Figure 2: Market expectations of the federal funds rate at different horizons (monthly data). Notes: The figure shows the federal funds future contracts at the 6-month-, 12-month-, 24-month-ahead horizons. Expectations of the federal funds rate at the 36-month-ahead horizon are obtained by combining the data on the federal funds futures contracts and the Eurodollar futures, as described in section 4.2.

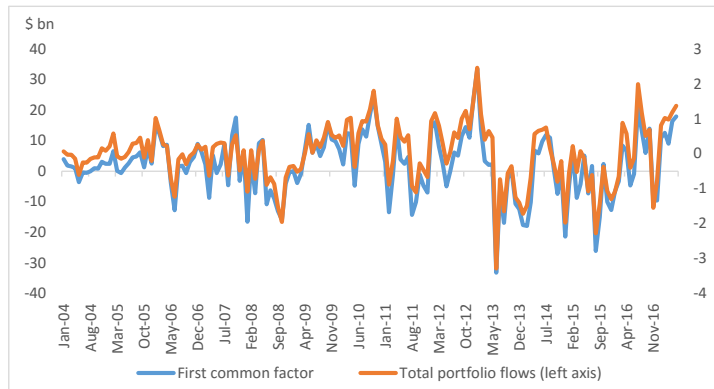


Figure 3: Common factor of portfolio flows and aggregate portfolio flows. Note: The common factor is the first principal component calculated from our panel of net portfolio flows.

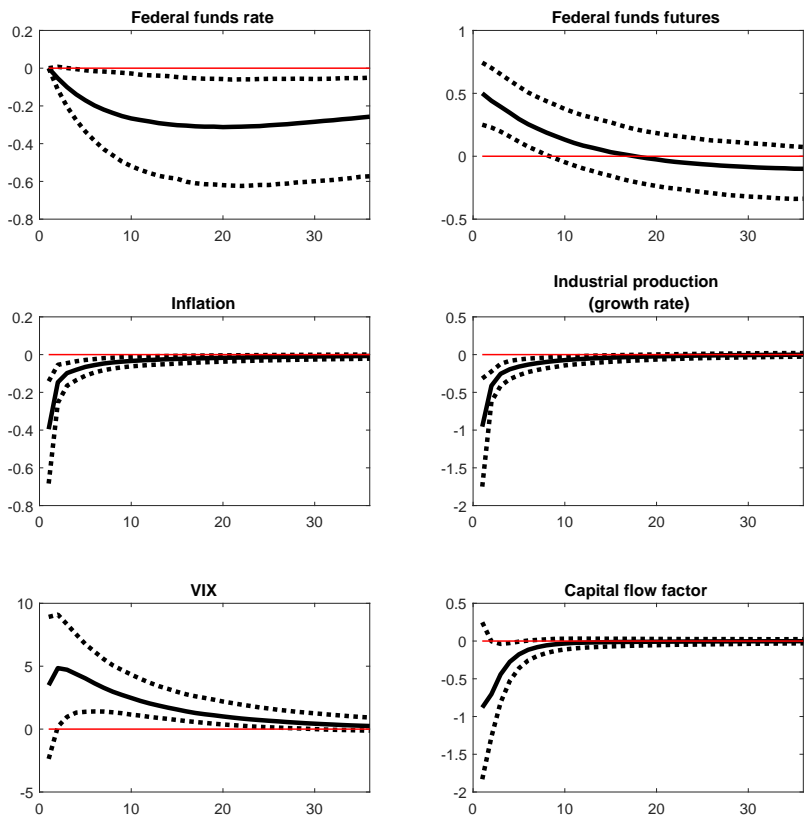


Figure 4: Impulse responses to a monetary policy news shock identified by sign restrictions. Note: Median (solid line) responses together with 68 percent credible set (dashed lines).

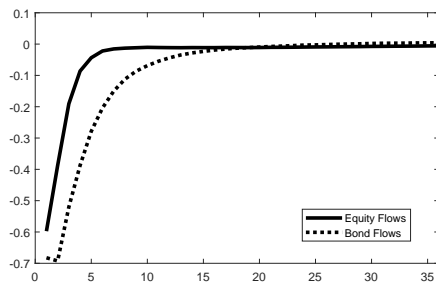


Figure 5: Impulse responses of Bond and Equity Flows. Note: Median responses of bond flows (dashed line) and equity flows (solid line).

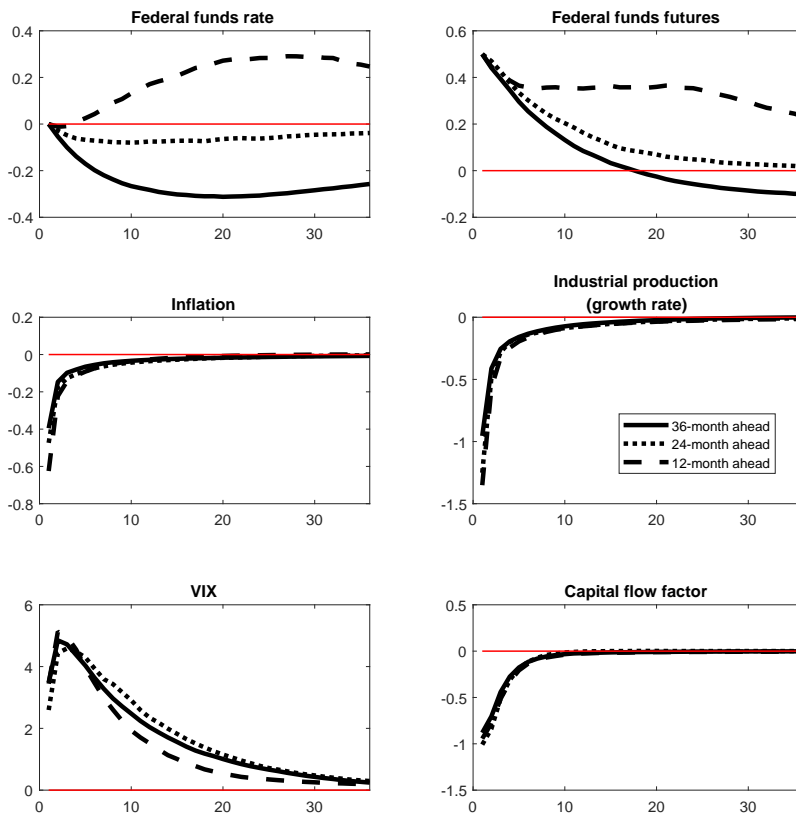


Figure 6: Impulse responses to monetary policy news shocks at alternative expectation horizons. Note: Median (solid line) responses together with 68 percent credible set (dashed lines).

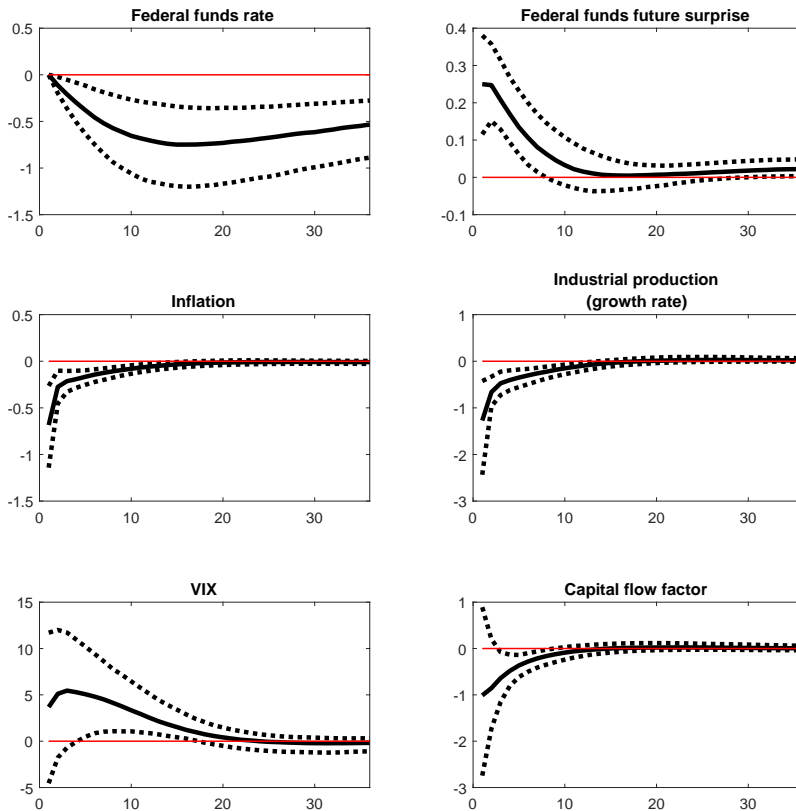


Figure 7: Impulse responses for alternative monetary policy news shock using federal funds future surprises. Note: Median (solid line) responses together with 68 percent credible set (dashed lines).