

# A model of the euro-area yield curve with discrete policy rates\*

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## Abstract

This paper presents a no-arbitrage yield-curve model that explicitly incorporates the central-bank policy rate and that is consistent with the zero lower bound. The estimation of the model, on euro-area daily data, reveals the existence of sizeable monetary-policy-related risk premia. These risk premia are typically neglected by the practice that consists in backing out market forecasts of future policy-rate moves from money-market forward rates. The model is further exploited to simulate forward-guidance measures: A central-bank commitment to keep the policy rate at 1% for 2 years would imply a decline in the 5-year rate of about 25 basis points.

**JEL codes:** E43, E44, E47, E52, G12.

**Keywords:** affine term-structure models; zero lower bound; regime switching; forward policy guidance.

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# 1 Introduction

The standard view of the monetary policy transmission mechanisms suggests that central banks' actions are mainly transmitted to the economy through their effect on market interest rates. According to this standard view, a restrictive monetary policy pushes up both short-term and long-term interest rates, leading to less spending by interest-sensitive sectors of the economy, and vice versa. While there is a strong empirical support for the assertion that monetary policy is a major driver of the yield-curve fluctuations, only a very few term-structure models explicitly incorporate the “policy rate”, that is the main central bank’s instrument. Arguably, this reflects the technical difficulties associated with accommodating the dynamics of this discrete-valued process.<sup>1</sup>

This paper proposes a novel and tractable no-arbitrage term-structure model where changes in the monetary-policy rate are explicit and central. This model is particularly adapted to depict the dynamics of the short-end of the yield curve, where the influence of monetary policy decisions is the most evident (see Cochrane and Piazzesi, 2002). The estimation, carried out on euro-area daily data covering the last 13 years, sheds light on the influence of the ECB monetary policy on the term-structure of interest rates. Notably, the results show the key effect of the monetary-policy phases –tightening, easing or status quo– on the shape of the yield curve. Besides, the analysis provides evidence of the existence of substantial risk premia at the short- to medium-end of the term structure of interest rates.<sup>2</sup> This implies in particular that the common market practice that consists in backing out market forecasts of next policy-rate moves from money-market forward rates is biased.<sup>3</sup> More

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<sup>1</sup> See e.g. Rudebusch (1995), Hamilton and Jorda (2002), Balduzzi, Bertola and Foresi (1997) and Balduzzi et al. (1998) for models of the U.S. Federal Funds rate target.

<sup>2</sup> The existence of such risk premia in the short end of the euro-area yield curve has notably been evidenced by Durré, Evjen and Pilegaard (2003).

<sup>3</sup> This common market practice implicitly assumes that the expectation hypothesis holds at the short-end of the yield curve.

precisely, the results suggest that while this practice is valid in terms of sign of the expected target moves, it tends to overestimate their size. Besides, these risk premia turn out to be the most important when the monetary policy is in a tightening phase, the deviation between the 12-month-ahead risk-neutral forecast of the policy rate (this forecast is approximately a forward rate) and its physical counterpart being of about 50 basis points.

As an additional contribution, this model is exploited to assess the potential effects of so-called *forward policy guidance* measures. These measures, that consist of commitments of the central bank regarding the future path of its policy rate, are expected to provide more accommodation at the zero lower bound (ZLB).<sup>4</sup> Indeed, the objective of these measures is to provide a stimulus to the economy by making market participants revising down their expectations of future short-term interest rates, thereby pushing down medium- to long-term interest rates. The effectiveness of such measures is the subject of substantial debate (Williams, 2011). Using new-Keynesian general equilibrium models, Eggertsson and Woodford (2003), Campbell et al. (2012) or Levin et al. (2010), among others, investigate the impacts of forward policy guidance. While the former two studies find that forward guidance can be effective in terms of macroeconomic stabilisation, the latter shows that such measures may be insufficient to deal a “Great Recession”-style shock. As in Gagnon et al. (2011), Kool and Thornton (2012), Rudebusch and Bauer (2011) or Jardet, Monfort and Pegoraro (2010), the present paper focuses on the effects of unconventional monetary policies on the term structure of interest rates. Specifically, as in the latter paper, the model is used to simulate the effects of commitments of the central bank to keep its policy rate at its current level for (at least) a deterministic period of time. In the present framework, where the policy rate is explicit, such a simulation is carried out in a straightforward and consistent manner. According to the results,

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<sup>4</sup> See Bernanke and Reinhart (2004) for a list and discussion of the potential policy options available to monetary-policy authorities when the zero bound is binding.

forward-guidance measures could lead to a substantial downward shift in the yield curve. The lower the policy rate, the larger the effect: for instance, in a context characterised by a policy rate of 1% (respectively 3.5%), the model predicts that the announcement of a commitment to keep the target rate unchanged for at least 2 years would be followed by a 25 bp (resp. a 5 bp) decline in the 5-year yield.

The papers that are the closest to the present one are those by Piazzesi (2005) and Fontaine (2009). In both papers, the authors propose term-structure models that explicitly involve the target for the policy rate. They estimate their models on U.S. data covering respectively the periods 1994-1998 (weekly) and 1994-2007 (daily). A common drawback of these frameworks is that they do not preclude negative policy rates. While this caveat may be tenable when the short-term interest rate is far enough from zero –the conditional probabilities of having negative interest in the subsequent periods being negligible–, it is more problematic in the current context of very low interest rates. More generally, many of the tractable yield-curve models are not consistent with the ZLB restriction.<sup>5</sup> This limitation is addressed in the present framework.

In my model, changes in the policy rate rate take place on pre-determined monetary-policy meeting dates and are multiples of 25 basis points (or 0.0025). The model is consistent with the fact that target-rate changes occur infrequently, on a daily time scale, and with policy inertia (i.e. that target changes are often followed by additional changes in the same direction). These appealing features stem from an original use of regime-switching techniques, each regime being characterised by a given tick of the policy rate and a given monetary-policy phase: tightening, easing or

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<sup>5</sup> See Dai and Singleton (2003) or Piazzesi (2010). Hamilton and Wu (2012) propose a way to adapt the standard Gaussian framework to account for an extended period of constant short-term rate. However, they implicitly assume that when this phase ends, (a) such a phenomenon cannot happen again and (b), the short-term rate can turn negative again. Andreasen and Meldrum (2011) or Kim and Singleton (2011) show that the quadratic Gaussian framework can be used to preclude negative interest rates. However, these latter models can not accommodate long periods of unchanged interest rates.

status quo. The definition of these phases is consistent with observed central banks' target-setting behaviour and communication (see Smaghi, 2009). The probabilities of occurrences of target moves depend on the monetary-policy phase and on the level of the target rate. In particular, the probability of a cut in the policy rate is zero when this rate is at the ZLB, thereby precluding negative rates.

The shortest-term rate of the yield curve that is considered here is the interbank overnight interest rate, which most central banks aim at stabilising to a level close to the policy (or target) rate. Therefore, after having specified the dynamics of the latter, the model is completed by specifying the dynamics of the so-called EONIA spread, that is the yield differential between the Euro Over-Night Index Average and the main policy rate. While this spread was mostly transitory before 2007, persistent deviations appeared in October 2008, following changes in the monetary-policy implementation in the euro area in response to the financial crisis. To capture that change in the behaviour of the EONIA spread, an additional two-state Markov-switching process is introduced, one of these two states corresponding to a situation in which banks' excess liquidity translates into a drop of the interbank rate with respect to the target (see Soares and Rodrigues, 2011).

Consistently with the choice of the EONIA as the shortest-term rate, the empirical exercise uses Overnight Index Swap (OIS) rates as longer-term yields.<sup>6</sup> An OIS is a fixed-for-floating interest rate swap with a floating rate leg tied to the index of daily interbank rates, that is the EONIA in the euro-area case. OIS have become especially popular hedging and positioning vehicles in euro financial markets and grew significantly in importance during the financial turmoil of the last few years.<sup>7</sup> The OIS curve is closely watched by practitioners to gauge what policy-rate changes

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<sup>6</sup> This is done only for the second part of my sample, i.e. 2005-2011. Indeed, long-term OIS are not available before then. In the first part of the sample, I use EURIBOR swaps (see Subsection 2.2).

<sup>7</sup> While the United States has a liquid Fed Funds future contract (Gurkaynak, 2005 or Gurkaynak, Sack and Swanson, 2007), markets in most other countries rely exclusively on their local-currency-denominated OIS market for hedging central bank policy (Lang, 2010).

the market has already priced in.

The model involves a lot of Markovian regimes –more than 200–, obtained by crossing the regimes describing the policy rate, the monetary policy phases and the liquidity states. This distinguishes the present framework from earlier term-structure models involving regime switching.<sup>8</sup> In spite of this unusual feature, the approach remains tractable both in terms of bond pricing and estimation. The yields of different maturities turn out to be equal to linear combinations of the factors (including the regime variable), the factors loadings being given by simple formulas involving a limited number of matrix products.<sup>9</sup> The model can generate the usual shapes of the yield curve (steep, flat, inverse, humped, inverse-humped) and accommodates heteroskedasticity in the yield dynamics. As regards the estimation, a key point is that regimes are only partially hidden: a characteristic of the regimes, namely the central-bank policy rate, is observed by the econometrician.<sup>10</sup> Therefore, the econometric model can be seen as a six-hidden-state (three monetary-policy regimes and two liquidity regimes) Markov-switching model with heterogenous probabilities of transition, the latter depending on the observed target rate.

The model is estimated by maximum likelihood techniques. The computation of the log-likelihood is based on an innovative joint use of the Kitagawa-Hamilton's filter and so-called inversion techniques introduced by Chen and Scott (1993). The fit of the model is satisfying, the standard deviations of the pricing errors being of 8 basis points (from 1 month to 4 years). An important output of the approach are the probabilities of being in the different hidden Markovian states. To that respect, this approach is an illustration of the results of Bikbov and Chernov (2008) who underline

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<sup>8</sup> See, e.g., Bansal and Zhou (2002), Dai, Singleton and Yang (2007), Ang, Bekaert and Wei (2008) or Lemke and Archontachis (2008).

<sup>9</sup> In particular, the derivation of the term-structure of yields does not rely on the recursive algorithms usually used to solve discrete-time term structure models (as in Ang and Piazzesi, 2003). This point is crucial to make the model easily amenable to estimation using high-frequency data.

<sup>10</sup> I assume that market participants observe latent regimes and factors, as in most yield-curve studies involving latent factors.

the importance of using yield-curve information to identify monetary-policy regimes.

The remainder of the paper proceeds as follows. Section 2 presents the data and emphasises stylised facts. Section 3 develops the model. Section 4 presents the estimation strategy and results. Section 5 documents the behaviour of policy-rate-related risk premia. Section 6 derives some implications of the model regarding the commitment of the central bank to keep the target rate fixed for a given period of time. Section 7 concludes.

## 2 Data and stylised facts

### 2.1 The EONIA and the Eurosystem's framework

Contrary to the Fed or the Bank of England, the ECB does not have an explicit interest-rate target. However, its aim is explicitly to *“influence money market conditions and steer short-term interest rates”* (ECB, 2011). This is done by using primarily the official interest rates: *“The (long) chain of causes and effect linking monetary policy decisions with the price level starts with a change in the official interest rates by the central bank on its own operations.”*

In order to influence short-term money-market rates, a shortage of liquidity is created by imposing mandatory reserves on banks within the euro area. Specifically, credit institutions are required to hold compulsory cash deposits on accounts with the Eurosystem. The reserve requirements are based on the amount and profile of liabilities on a bank's balance sheet as of every month end. The banks can refinance themselves through the ECB's weekly Main Refinancing Operations (MROs). In these weekly refinancing operations, the ECB returns liquidity to the market by allowing banks to tender for cash (against collateral). By abuse of language, the rate at which liquidity is supplied in the regular weekly monetary policy operations is referred to as the “policy rate” (or the “target rate”) in this paper. However,

there are two additional policy rates in the Eurosystem framework. Indeed, the latter is completed by a symmetric corridor bracketing the main policy rate.<sup>11</sup> The lower bound of the corridor, called the deposit-facility rate, is the rate at which counterparties can deposit cash overnight with the Eurosystem. The upper bound is the lending-facility rate, at which counterparties can borrow funds overnight from the Eurosystem. The target rate and the corridor is displayed in Panel A of Figure 1.

After having been fixed till June 2000, the MROs' rate then became variable.<sup>12</sup> In October 2008, in a context of worldwide financial stress, the Eurosystem adopted a fixed-rate full allotment (FRFA) tender procedure: since then, the ECB accommodates any demand for liquidity its bank counterparties might express at the policy rate –against eligible collateral– in unlimited amounts.

While the policy rate defines the rate at which banks can refinance themselves through the ECB against collateral, EONIA (Euro OverNight Index Average) fixings reflect rates at which banks refinance themselves on the interbank market on an unsecured basis.<sup>13</sup> In “normal” circumstances, EONIA rates trade in close relation to ECB marginal rates but can also include a premium related to the unsecured nature of the lending.

Panel A of Figure 1 compares the fluctuations of the target with these of the EONIA. Changes in the policy rate are decided during the first of the bimonthly meetings of the ECB's Governing Council. On a daily scale, this implies a step-like behaviour for the target rate. Over the estimation sample (January 1999 – February 2012), there were 18 rises in the target rate (16 of 25 bp and 2 of 50 bp) and 16 cuts

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<sup>11</sup> See Kahn (2010) for a comprehensive description and an international comparison of “corridor” systems.

<sup>12</sup> At that time, the target, or refi rate, acted as the minimum bid rate at the MRO.

<sup>13</sup> The EONIA is computed as a weighted average of all overnight unsecured lending transactions undertaken in the interbank market, initiated within the euro area by the contributing banks. It is computed by the ECB at the end of every TARGET day (since January, 4 1999). The banks contributing to the EONIA are the same first class market standing banks as the panel banks quoting for Euribor. See [www.euribor-ebf.eu](http://www.euribor-ebf.eu) for more details.

in target rates (7 of 25 bp, 8 of 50 bp and one of 75 bp). Panel A of Figure 1 also suggests that the EONIA is closely linked to the target rate. However, by displaying the EONIA spread –i.e. the yield differential between the EONIA and the policy rate–, Panel B highlights the break in the relationships between these two rates that occurred in 2008. This break can be related to non-standard monetary-policy measures that were taken in response to the financial crisis. A particularly important decision was the one to move from variable rate tender procedures in liquidity providing operations to FRFA. Together with the expansion of the spectrum of maturities at which liquidity was being offered to the market, this measure generated a steady excess of liquidity balances in the overnight market, as banks began supplying in the interbank market the precautionary cash buffers that they were securing at the ECB.<sup>14</sup> An excess supply of liquidity in overnight trades put downward pressure on the overnight interest rate, which drifted toward the lower limit of the monetary policy corridor (see Beirne, 2012 and Fahr et al., 2010).

**[Insert Figure 1 about here]**

## 2.2 The Overnight Index Swaps

An overnight index swap (OIS) is an interest rate swap whose floating leg is tied to an overnight rate (the EONIA in the euro-area case), compounded over a specified term. OIS contracts involve the exchange of only the interest payments, the principal amount being notional. That is, the two parties agree to exchange, on the agreed notional amount, the difference between interest accrued at the fixed rate and interest accrued through daily compounding (or geometric averaging) of the floating overnight index rate. While the tenor of these swaps was usually below 2 years before 2005, the OIS maturities were extended afterwards to more than 10 years (see Barclays,

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<sup>14</sup> A large share of the cash buffers is held with the ECB, the banks using massively the marginal deposit facility.

2008). The OIS curve is more and more seen by market participants as a proxy of the risk-free yield curve (see e.g. Joyce et al., 2011).<sup>15</sup> In spite of that, OIS have failed to attract significant attention from academics for the time being.

As an interest-rate swap, an OIS can be used to manage interest-rate risks. In particular, the OIS are structured in such a way that if a bank (a) has some money available for investment, (b) has access to the overnight interbank market and (c) can enter OIS contracts, then this bank can synthetically create a fixed-income instrument that is equivalent to a maturity- $h$  bond paying a coupon equal to the maturity- $h$  OIS rate.

An important point that is going to be investigated below relates to the use of OIS curves to back out market expectations of future policy rate's moves. Heuristically, under the expectation hypothesis, the forward rates based on the OIS term structure should reflect the market expectations of the interbank rate, that is supposed to be close to the target rate. This principle is widely used by market analysts, investors or central banks themselves.<sup>16</sup>

## 2.3 Data sources and treatments

The sample period is January 15, 1999 to February 17, 2012 (3416 dates). While the target rate and the EONIA series come from the ECB, the OIS yields are taken from Bloomberg. All yields are translated on a continuously compounded basis, and market holidays are filled with observations from the previous trading days' rates.<sup>17</sup>

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<sup>15</sup> While OIS rates reflect the credit risk of an overnight rate, this may be regarded as negligible in most situations. Besides, even during financial-markets turmoil, the counterparty risk is limited in the case of a swap contract, due to netting and credit enhancement, including call margins (see Bomfin, 2003). To that respect, one can note that German sovereign bonds, usually perceived as being the European "safest haven" both in terms of credit quality and liquidity, trade at levels that have remained close to the OIS yield curve over the last years.

<sup>16</sup> See e.g. Barclays, 2008, Joyce, Relleen and Sorensen, 2008, Joyce and Meldrum 2008, Bank of England, 2005 or Lang (2010).

<sup>17</sup> Let  $r$  denote a market-quoted interest rate (the OIS, say). Using the fact that money-market rate are based on the ACT/360 day-count basis, the corresponding continuously compounded rate is computed as  $\ln(1 + d \times r/360) \times 365/d$ , where  $d$  is the residual maturity of the instrument.

In the estimation, we consider six maturities (in addition to the overnight one): 1 month, 3 months, 6 months, 12 months, 2 years and 4 years.

As said above, OIS yields are not available for longer-than-one-year maturities before 2005. Before that date, we use EURIBOR swaps data in place of the 2-year and 4-year OIS yields. This appear to be a reasonable assumption given that the short-term EONIA swaps and maturity-matching EURIBORs had extremely close variations before 2007.<sup>18</sup> Swap yields are homogenous to coupon-bond yields. Since the pricing formula presented below (Subsection 3.2) are consistent with zero-coupon yields, zero-coupon yields are computed using classic bootstrapping methods.<sup>19</sup>

The estimation procedure involves survey-based forecasts of short-term yields (as in Kim and Orphanides, 2012; this is discussed in Section 4). Specifically, 12-month-ahead forecasts provided by the *Consensus Forecasts* are used. Forecasts of the ECB's policy rate are available since July 2009 only; before that, I use 3-month EURIBOR forecasts.<sup>20</sup> Since EURIBOR and OIS were closely linked until summer 2007, using EURIBOR forecasts instead of OIS forecasts is appropriate till then. In mid-2007 however, the widening in the EURIBOR-OIS spread is likely to induce a bias in the forecasts. This is addressed by subtracting from the EURIBOR forecasts –from August 2007 to June 2009– the 1-year-ahead forward spread between the 3-month EURIBOR and OIS rates (averaged over the same period). All these survey-based expectations are available at the monthly frequency only and are released about mid-month. Using a cubic spline, this series is converted into a daily one. The discrepancies that arise from these approximations are expected to be captured by

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<sup>18</sup> During summer 2007, credit and liquidity risks affected unsecured interbank lending rates (IBOR), leading to a sudden widening of the IBOR-OIS spreads. Before that, this spread was small and steady. For each maturity (2-year and 4-year), I subtract the 2005-2006 IBOR-OIS average spread from the EURIBOR swap series used in the estimation before 2005, which is about 10 basis points (standard deviation below 3 basis points).

<sup>19</sup> OIS rates with a maturities lower than one year are already homogenous to zero-coupon instruments. The bootstrapping methods are applied only for longer-than-one-year maturities. See Barclays (2008) for more information about EONIA swaps.

<sup>20</sup> Naturally, the fact that the nature of the forecasted rate changes in mid-2009 is taken into account in the estimation procedure.

measurement errors of the state-space model that will be presented below.<sup>21</sup>

## 2.4 Preliminary analysis of the yields

Table 1 reports descriptive statistics for the different yields used in the analysis. These statistics suggest that yields are highly persistent. While the daily autocorrelation is nearly one, the correlations between the yields and their 1-year lags is still substantial (higher than 50%). The correlation across maturities is also extremely high, with near-unit correlations for adjacent maturities. Mean and median statistics show that the term structure is positively sloped on average.

The lowest Panel in Table 1 shows the results of a principal component analysis carried out on the set of seven spreads between OIS yields –with maturities of 1 day to 4 years– versus the policy rate. The three principal components are sufficient to explain most of the fluctuations of these spreads. Notably, the first principal component explains more than 90% of the variances of the spreads associated with yields of maturities comprised between 3 months and 1 year. This is graphically illustrated in Panel D of Figure 1, that highlights the common fluctuations in some of these spreads. Half of the variance of the EONIA spread and of the spread between the 4-year rate and the target rate is accounted for by this first factor, indicating that there are important correlations between the EONIA spread and longer-term spreads. However, further investigations mitigate this finding. Specifically, the same kind of analysis has been carried out on a shorter sample, excluding the crisis period: 1999-2008 (bottom of Table 1). On that period, the EONIA spread turns out to be almost orthogonal to the first principal component. Therefore, the apparent comovement between the EONIA spread and the other spreads on the whole sample seems to be related to the fall in the EONIA spread that took place in mid-2008 (see Subsection

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<sup>21</sup> Anticipating on the estimation results presented in Section 4, the standard deviation of the measurement errors associated to the forecasts is slightly larger than 20 bps ( $\sigma_{fcst}$  in Table 2), which is of the same order of magnitude as the errors expected from the previous points.

2.1 for a description of this phenomenon).

**[Insert Table 1 about here]**

### 3 The model

This section formulates a model of the daily dynamics of the overnight interbank interest rate.<sup>22</sup> Two dynamics are considered: the historical (or physical, or real-world) one and the risk-neutral (or pricing) one. The knowledge of the risk-neutral dynamics of the interbank rate makes it possible to price financial instruments – such as the OIS contracts – whose cash flows depend on the overnight interbank rate. The simultaneous knowledge of the two dynamics allows to study term premiums’ behaviour, as will be done in Section 5. The historical ( $\mathbb{P}$ ) and the risk-neutral ( $\mathbb{Q}$ ) dynamics of the different processes are of the same kind, but their respective parameterizations differ. These differences and the implied stochastic discount factor (s.d.f.) are detailed in Subsection 3.2, that also deals with the derivation of the term-structure of OIS rates. Before that, the next subsection presents the different components of the overnight interest rate.

#### 3.1 The components of the overnight interest rate

The target rate prevailing at date  $t$  is denoted by  $\bar{r}_t$ . As is the case in most currency areas, the target rate is assumed to be a multiple of 0.25%. I proceed under the assumption that the target rate is lower than a maximal rate denoted by  $r_{max}$  and equal to  $0.25\% \times N$ , say. Therefore:

$$\bar{r}_t = \Delta' z_{r,t}$$

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<sup>22</sup> The extension to a lower frequency is straightforward.

where  $z_{r,t}$  is a selection vector, i.e. one of the column of  $I_{N+1}$ , the identity matrix of dimension  $(N+1) \times (N+1)$  and where the entries of the vector  $\Delta$  are the continuously-compounded possible policy rates. Specifically, using the money-market day-count convention, the  $i$ th entry of  $\Delta$  is given by  $\log(1 + (i - 1)0.25\%/360)$ . Note that at the daily frequency, many of the successive  $\bar{r}_t$ 's are equal. In particular,  $\bar{r}_{t-1} = \bar{r}_t$  as soon as there is no policy meeting at date  $t$ . This results in a step-like process for the policy rate (as seen in Panel A of Figure 1).

The interbank overnight interest rate is denoted by  $r_t$ . Its deviations from the target rate are accounted for by two components:  $\xi_t$  and  $s_t$ :

$$r_t = \bar{r}_t + s_t + \xi_t \quad (1)$$

I assume that  $\bar{r}_t$ ,  $s_t$  and  $\xi_t$  are independent of each others.<sup>23</sup> The variables  $s_t$  and  $\xi_t$  are unobservable but can be inferred from yields through the bond-pricing model. The historical dynamics of these factors are presented in the following. The risk-neutral dynamics are of the same kind, but their parameterizations is different from their physical counterparts. These differences are made explicit in Subsection 3.2.

### 3.1.1 The dynamics of the target rate $\bar{r}_t$

Central bankers can decide to change the target rate at their regular meetings. On these dates, the target can be raised or cut if the the tightening regime or the easing regime respectively prevail, but the target remains necessarily unchanged under the status quo regime. Formally, the monetary regime is represented by a 3-dimensional selection vector  $z_{m,t}$  that is valued in the set of the three columns of the identity matrix  $I_3$ , corresponding respectively to the tightening, the status quo and the easing regimes. Contrary to the econometrician, market participants observe the regime,

<sup>23</sup> Such independence assumptions are common in that literature (see Balduzzi et al., 1997 and 1998, or Piazzesi (2005)).

this knowledge being based on a variety of detailed policy-relevant information that is not modelled here.

The Kronecker product of the selection vectors  $z_{r,t}$  and  $z_{m,t}$ , denoted by  $\bar{z}_t$ , is also a selection vector that is valued in the set of the columns of  $I_{3(N+1)}$  (recall that  $N + 1$  is the number of possible values of the target rate, between 0% and  $r_{max} = N \times 0.25\%$ ). The dynamics of  $\bar{z}_t$  is described by a Markov chain. The matrix of transition probabilities of  $\bar{z}_t$  is denoted by  $\bar{\Pi}_t$ . These matrices are time-inhomogenous, but in a deterministic way. Indeed, the matrices  $\bar{\Pi}_t$  can take two values, one of them being specific to those days at which a monetary-policy meeting are scheduled.<sup>24</sup> The number of entries of these  $\bar{\Pi}$  matrices is considerable: for  $r_{max} = 10\%$ , there are 15.129 of them. However, owing to the following assumptions, most of these entries are zero:

1. Conditionally on being in an easing, a status quo or a tightening regime, the target moves are respectively valued in  $\{-0.50\%, -0.25\%, 0\}$ ,  $\{0\}$  and  $\{0, +0.25\%, +0.50\%\}$ .
2. Easing or tightening phases are necessarily followed by status quo phases.

Even with these restrictions, many of  $\bar{\Pi}_t$ 's entries still require to be parameterised. Eight sets of probabilities needs to be defined: two of them contain the probabilities of switching to the status quo regime (the probability of exiting the easing and the tightening regimes are respectively denoted by  $p_{ES}$  and  $p_{TS}$ ), two others are the probabilities of exiting the status quo regime ( $p_{SE}$  and  $p_{ST}$ ), two of them contain the probabilities of 25-bp changes in the target rate (rise:  $p_{r25}$ ; cut:  $p_{c25}$ ) and the last two are the probabilities of 50-bp moves (rise:  $p_{r50}$ ; cut:  $p_{c50}$ ). These probabilities may vary with the policy rate. For instance, the probability of switching from the

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<sup>24</sup> Contrary to the policy rate ( $z_{r,t}$ ), that can change only following a monetary-policy meeting, the monetary-policy regime ( $z_{m,t}$ ) can switch at any date. For instance, such changes could be triggered by ECB officials' speeches or the release of macroeconomic news or figures.

tightening to the status quo regime could be larger for higher target rates, say. In order to keep the model parsimonious, the probabilities are based on logit-based parametric functions of the target rate  $\bar{r}$ . Formally, let me define the function  $f$  by:

$$f(\bar{r}, [a_1, a_2]') = [1 + \exp(a_1 + a_2\bar{r})]^{-1}. \quad (2)$$

For  $i \in \{TS, ES, SE\}$ , the probabilities  $p_i$  are characterised by some  $2 \times 1$  vectors of parameters  $\alpha_i$  and are given by  $f(\bar{r}, \alpha_i)$ . Further, so as to have  $p_{ST} + p_{SE} < 1$ , the probabilities  $p_{ST}$  are defined by  $(1 - p_{SE}(\bar{r}))f(\bar{r}, \alpha_{ST})$ . Moreover,  $\alpha_i$  vectors are not defined for each of the four kinds of target moves, but only for two: one for the rises in the policy rate ( $\alpha_r$ ) and one for the cuts ( $\alpha_c$ ). Two additional parameters,  $k_r$  and  $k_c$ , are then introduced to share the rise and cut probabilities into those of 25-bp and 50-bp moves. Formally, the conditional probabilities of target-rate changes (i.e.  $p_{c25}$ ,  $p_{r25}$ ,  $p_{c50}$  and  $p_{r50}$ ) are defined through:

$$\begin{cases} p_{r25}(\bar{r}) = k_r f(\bar{r}, \alpha_r) & \text{and} & p_{r50}(\bar{r}) = (1 - k_r) f(\bar{r}, \alpha_r) \\ p_{c25}(\bar{r}) = k_c f(\bar{r}, \alpha_c) & \text{and} & p_{c50}(\bar{r}) = (1 - k_c) f(\bar{r}, \alpha_c) \end{cases}$$

where  $k_c$  and  $k_r$  are valued in  $[0, 1]$ .<sup>25</sup> Eventually, the 15.129 entries of matrix  $\bar{\Pi}$  are defined by 16 parameters only.

### 3.1.2 The dynamics of $\xi_t$

The factor  $\xi_t$  is aimed at capturing the volatile and short-lived (noise) fluctuations of the EONIA spread. However, as clearly appears on Panel B of Figure 1, the distribution of this component is related to the level of this spread. Typically, the noise

<sup>25</sup> Before November 2001, possible changes in the policy rate were discussed in each of the bi-weekly meetings of the ECB Governing Council. Since then, they are considered during the first of these two bi-weekly meetings only. Accordingly, for the first part of the sample (up to November 2001), the target-moves probabilities are divided by two so as to result in (approximately) the same probabilities of target moves over a month.

distribution became strongly positively skewed after the drop in the EONIA spread, in late 2008. As discussed in Subsection 2.1, this drop follows the implementation of non-standard monetary-policy measures that gave rise to a *banks' excess liquidity regime*. Hence, both the distribution of the noise component of the EONIA spread as well as its (conditional) mean have to be conditioned on the excess liquidity regime. The latter is modelled by an additional Markovian regime process  $z_{exc,t}$ . This process can take two values  $[1, 0]'$  (no excess-liquidity conditions) or  $[0, 1]'$  (excess liquidity conditions). The matrix of transition probabilities associated with this process is time-homogenous and is denoted by  $\Pi_{exc}$ .<sup>26</sup> Formally,  $\xi_t$  is given by:

$$\xi_t = \begin{bmatrix} (w_{norm} + \xi_{norm,t}) & (w_{exc} + \xi_{exc,t}) \end{bmatrix} z_{exc,t}$$

where the  $w_i$ 's are scalar parameters and the  $\xi_{i,t}$ 's follow taylor-made distributions, denoted by  $\mathcal{L}$ , that allows for non-zero skewness and fat tails. The definition and features of this distribution are detailed in Appendix A. The support of this distribution is the compact  $[-1, 1]$  (in annualised terms), which is consistent with the fact that the EONIA is bounded by the corridor set by the ECB's standing facilities.<sup>27</sup>

### 3.1.3 The dynamics of $s_t$

The variable  $s_t$  is aimed at contributing to persistent fluctuations of yields that can not be accounted for by the regime variables ( $z_{r,t}$ ,  $z_{m,t}$  and  $z_{exc,t}$ ). Combined with  $\xi_t$ , the latter are expected to account for most of EONIA's fluctuations. Therefore, the variable  $s_t$  is expected to have a far lower impact on the overnight rate than on longer-term yields. To obtain such a feature (without resorting to an explosive dynamics for  $s_t$ ),  $s_t$  is decomposed into two components denoted by  $s_{1,t}$  and  $s_{2,t}$ , that

<sup>26</sup> The columns of  $\Pi_{exc}$  sum to one. Note that, as  $z_{m,t}$ , the regime variable  $z_{exc,t}$  is assumed to be observed by market participants but not by the econometrician ( $z_{r,t}$  is observed by everybody).

<sup>27</sup> Note that the width of the corridor has changed over time (between 150 and 200 bp). However, taking into account such a variability would induce severe complexity in the framework.

is,  $s_t = s_{1,t} + s_{2,t}$ . The dynamics of  $[s_{1,t}, s_{2,t}]'$  is given by:

$$\begin{bmatrix} s_{1,t} \\ s_{2,t} \end{bmatrix} = \Phi \begin{bmatrix} s_{1,t-1} \\ s_{2,t-1} \end{bmatrix} + \Sigma \varepsilon_t, \quad \varepsilon_t \sim \text{i.i.d. } \mathcal{N}(0, I) \quad (3)$$

where  $\Phi = \begin{bmatrix} \rho_1 & \beta \\ 0 & \rho_2 \end{bmatrix}$  and  $\Sigma = \begin{bmatrix} 0 & 0 \\ 0 & \sigma \end{bmatrix}$

The smaller  $\beta$ , the less variable  $s_{1,t}$  is. In the limit, if  $\beta$  is equal to zero and if  $s_{1,t}$  was zero at some point in the past, then  $s_t = s_{2,t}$ . I assume this is the case under the physical measure, but not under the risk-neutral one. Under the latter measure, if the  $\rho_i^*$ 's –the risk-neutral counterpart of the  $\rho_i$ 's– are close to one, a shock on  $s_{2,t}$  can have a very persistent impact on  $s_t$ . In addition, if  $\beta^*$  is large enough, these effects are multiplied by feeding through  $s_{1,t}$ . Therefore,  $s_t$ 's innovations may have a far more long-lasting impact under the risk-neutral measure than under the physical measure. This implies that  $s_t$  may account for a far larger variance of long-term yields than of short-term yields.<sup>28</sup>

### 3.1.4 Definition of the single regime variable $z_t$

Defining a single regime variable will prove convenient for notational reasons in the remaining of this paper. Accordingly, I introduce the selection vector  $z_t$ , defined as the Kronecker product of  $\bar{z}_t$  and  $z_{exc,t}$ . Since  $\bar{z}_t$  is itself the Kronecker product of  $z_{r,t}$  and  $z_{m,t}$ , I have:

$$z_t = z_{r,t} \otimes z_{m,t} \otimes z_{exc,t}.$$

Hence,  $z_t$  is valued in the set of the columns of  $I_{6(N+1)}$ , each of the  $6(N+1)$  different regimes being characterised by the policy rate (there are  $N+1$  of them), a monetary-policy stance (there are three of them) and the situation of Eurosystem's liquidity

<sup>28</sup> The choice of this dynamics builds on Dubecq and Gouriou (2011).

(the situation being “normal” or “in surplus”). Recall that  $z_t$  is observed by market participants but not by the econometrician (who observes  $z_{r,t}$  but not  $z_{m,t}$  and  $z_{exc,t}$ ).

Given the assumption of independence between  $\bar{r}_t$  and  $(w_t, \xi_t)$ , the matrix of transition probabilities of  $z_t$ , denoted by  $\Pi_t$ , is equal to the Kronecker product of  $\bar{\Pi}_t$  and  $\Pi_{exc}$ .

### 3.1.5 About the seasonality of the EONIA spread

This framework do not account for potential seasonality in the EONIA spread. While this could bias the pricing of short-term yields (with maturities of one week, say), this simplification has a limited impact for longer maturities. As noted by Balduzzi et al. (1998), only little seasonal variability of the overnight interest rate should be transmitted to longer-term rates, since seasonal variability is “averaged out” in the expectation process (especially if one considers maturities that are multiple of the reserve maintenance period, which is the case in that study).

## 3.2 Pricing

### 3.2.1 The stochastic discount factor (s.d.f.)

I assume that the risk-neutral dynamics of  $z_t$  and  $s_t$  are of the same kinds as their historical counterparts except that the  $\Pi_t$ 's and  $\Phi$  are respectively replaced by  $\Pi_t^*$ 's and  $\Phi^*$  matrices, that depend on the same number of free parameters.<sup>29</sup> In this context, it can be shown that the stochastic discount factor (s.d.f.), or pricing kernel, is explicit.<sup>30</sup> Specifically, the s.d.f.  $M_{t-1,t}$  between  $t - 1$  and  $t$  is given by:

$$M_{t-1,t} = \exp \left( -\Delta'_m z_{t-1} - s_{t-1} - \xi_{t-1} - \frac{1}{2} \nu'_{t-1} \nu_{t-1} + \nu'_{t-1} \varepsilon_t + (z'_{t-1} \delta_t) z_t \right)$$

<sup>29</sup> In particular, the  $p_{ES}^*$ ,  $p_{SE}^*$ ,  $p_{ST}^*$ ,  $p_{TS}^*$ ,  $p_{r25}^*$ ,  $p_{c25}^*$ ,  $p_{r50}^*$  and  $p_{c50}^*$ , that define the  $\Pi_t^*$ 's matrices, are based on functions  $f(\bar{r}, \bullet)$ . Still using the superscript \* to denote risk-neutral parameters, these probabilities depend on some vectors  $\alpha_i^*$  (see end of Subsection 3.1.1).

<sup>30</sup> See Monfort and Renne (2012).

where  $\Delta_m$  is the concatenation of six vectors  $\Delta$ , that is  $\Delta_m = 1_{6 \times 1} \otimes \Delta$ , which reflects the fact that there are three monetary regimes ( $z_{m,t}$ ) and two Eurosystem-liquidity situations ( $z_{exc,t}$ ), and where the risk sensitivities  $\delta_t$  and  $\nu_t$ —that price respectively the risks associated to the regime shifts and to the Gaussian shocks  $\varepsilon_t$ —are defined by:

$$\begin{cases} \delta_{ij,t} &= \log(\Pi_{t,ij}^*/\Pi_{t,ij}) \\ \nu_t &= \Sigma^{-1}(\Phi^* - \Phi) \begin{bmatrix} s_{1,t} & s_{2,t} \end{bmatrix} \end{cases} \quad \forall i, j, t. \quad (4)$$

### 3.2.2 Bond prices

It is well-known that the existence of a positive stochastic discount factor is equivalent to the absence of arbitrage opportunities (see Hansen and Richard, 1987) and that the price at  $t$  of a zero-coupon bond with residual maturity  $h$ , denoted by  $P(t, h)$  is given by:

$$\begin{aligned} P(t, h) &= E_t(M_{t,t+1} \times \dots \times M_{t+h-1,t+h}) \\ &= E_t^{\mathbb{Q}}(\exp[-r_t - \dots - r_{t+h-1}]). \end{aligned} \quad (5)$$

Substituting equation (1) into equation (5) leads to:

$$P(t, h) = E_t^{\mathbb{Q}} \left( \exp \left[ - \sum_{i=0}^{h-1} (\bar{r}_{t+i} + s_{t+i} + \xi_{t+i}) \right] \right) \quad (6)$$

Under the assumption that  $\bar{r}_t$ ,  $s_t$  and  $(w_t, \xi_t)$  are independent processes, it comes:

$$\begin{aligned} P(t, h) &= \left( E_t^{\mathbb{Q}} e^{-\sum_{i=0}^{h-1} \bar{r}_{t+i}} \right) \left( E_t^{\mathbb{Q}} e^{-\sum_{i=0}^{h-1} \xi_{t+i}} \right) \left( E_t^{\mathbb{Q}} e^{-\sum_{i=0}^{h-1} s_{t+i}} \right) \\ &= P_1(t, h) \times P_2(t, h) \times P_3(t, h) \quad (\text{say}). \end{aligned}$$

The computations of  $P_1(t, h)$ ,  $P_2(t, h)$  and  $P_3(t, h)$  are detailed in Appendix C. It is important to stress that explicit formulas are available to compute each of these

three terms, each of them turning out to be exponential affine in  $(z'_t, s_t)'$ . Accordingly, the yields associated with zero-coupon bonds of maturity  $h$ , denoted by  $y(t, h)$ , are of the form:

$$y(t, h) = -\frac{1}{h} [G(t, h)z_t + A_h + B_h s_t] \quad (7)$$

Note that  $G(t, h)$  is deterministic (i.e., the only stochastic components of the yields are  $z_t$  and  $s_t$ ).

## 4 Estimation

### 4.1 The state-space form of the model

Kim and Orphanides (2012) have shown that the estimation of dynamic no-arbitrage term structure models with a flexible specification of the market price of risk is beset by a severe small-sample problem arising from the highly persistent nature of interest rates. They show that using survey-based forecasts of a short-term interest rate as an additional input to the estimation can overcome this problem. Following their approach, I enlarge the state-space model to make the estimated model consistent with 12-month-ahead forecasts of short-term rates provided by the *Consensus Forecasts*.<sup>31</sup>

Let me denote by  $R_t$  a vector of  $M$  observed yields of maturities  $h_1, \dots, h_M$ , that is  $R_t = [y(t, h_1), \dots, y(t, h_M)]'$ . Equation (7) shows that these yields are affine in  $(z_t, s_t)$ . It is straightforward to show that it is also the case for the 12-month-ahead forecasts included in the estimation. These forecasts are denoted by  $CF_t$ . Introducing some vectors of –supposedly i.i.d. normal– measurement errors

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<sup>31</sup> Other methodologies have been proposed to address this problem, see e.g. Jardet, Monfort and Pegoraro (2009).

denoted by  $\xi$ , we can write:

$$\begin{cases} R_t &= \Lambda_{z,R} z_t + \Lambda_{s,R} s_t + \xi_t^R \\ CF_t &= \Lambda_{z,C} z_t + \Lambda_{s,C} s_t + \xi_t^C \end{cases} \quad (8)$$

where the  $\Lambda$  matrices are functions of the model parameters (see Subsection 3.2). The model admits a Markov-switching state-space representation whose measurement equations are given by (8). The dynamics of the state vectors  $s_t$  and  $z_t$  are respectively defined by equation (3) and by the matrices of transition probabilities  $\Pi_t$ .

## 4.2 Computation of the log-likelihood

Whereas the Markov chain  $z_{r,t}$  is observed, the remaining state variables ( $s_t$ ,  $z_{m,t}$  and  $z_{exc,t}$ ) are not. This latency is handled by using an estimation strategy building on Monfort and Renne (2012). The approach consists in applying inversion techniques *à la* Chen and Scott (1993) together with the Kitagawa-Hamilton filter to address the hidden nature of the switching regimes. The idea of the inversion technique is the following: assuming that a combination of the yields –gathered in the vector  $R_t$ – is observed without error, one can recover the latent variable  $s_t$  as a function of  $R_t$  and  $z_t$ . Further, one can compute the likelihood function based on the specified dynamics of the latent factor as well as on the distribution of the (remaining) pricing errors. Usually, one uses trivial perfectly-priced combinations of yields: specifically, if there are  $m$  latent factors with continuous support in the model, one assumes that  $m$  yields are priced without error. However, as noted for instance by Piazzesi (2010), the choice of this maturity is arbitrary. Therefore, I resort to an original alternative approach and choose  $s_t$  in order to minimise the average squared pricing errors across

the different maturities.<sup>32</sup> In that case, the latent factor  $s_t$  (as a function of  $R_t$  and  $z_t$ ) is simply obtained by using the OLS formula:<sup>33</sup>

$$s_t = (\Lambda'_{s,R}\Lambda_{s,R})^{-1}\Lambda'_{s,R}(R_t - \Lambda_{z,R}z_t). \quad (9)$$

Details of the exact computation of the likelihood are provided in Appendix D.

### 4.3 Estimation results

Table 2 reports the maximum-likelihood parameter estimates.<sup>34</sup> The computation of the estimates' standard errors are based on the outer product of the first derivative of the likelihood function. The standard deviation of the pricing error –i.e. the deviation between modelled and observed yields) is equal to eight basis points–, which is comparable to Piazzesi's (2005) fit of the U.S. yield curve.<sup>35</sup> Panels B, C and D of Figure 2 respectively show the fit of the 3-month, the 2-year and the 4-year yields. These plots also show the part of those yields that is explained by the regime variable  $z_t$ . It appears that most of the yields' fluctuations can be accounted for by  $z_t$ : more than 95% of the sample variances of yields with maturities lower than 2 years are captured by the term  $G(t, h)z_t$  appearing in equation (7).<sup>36</sup>

**[Insert Figures 2 to 5 about here]**

Panel A of Figure 3 illustrates the ability of the model to reproduce survey-based forecasts of the target rate. Panel B and Panel C respectively present the

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<sup>32</sup> I am grateful to Simon Dubecq for providing me with this procedure. Diebold and Li (2006) also get estimates of Nelson-Siegel's  $\beta_t$  factors by regressing observed yields on factor loadings. However, they do not incorporate this procedure within a maximum-likelihood estimation.

<sup>33</sup> Note that this procedure results in one  $s_t$  conditionally to each of the different hidden regimes.

<sup>34</sup> In order to avoid that the factor  $s_t$ , thanks to its flexible Gaussian dynamics, explains too large a share of the yield fluctuations, I limit the size of its unconditional variance in the estimation. Specifically, I impose that the unconditional standard deviation of the  $s_t$ -related component of the one-year yield is lower than 10 basis points. Eventually, fifty one parameters remain to be estimated.

<sup>35</sup> Note however that the sample period used by Piazzesi (2005) is shorter (4 years against 13 years here) and the frequency is higher here (daily vs. weekly).

<sup>36</sup> 85% of the variance of the 4-year yield is accounted for by  $G(t, h)z_t$ .

estimated (smoothed) probabilities of being in the different monetary-policy regimes ( $z_{m,t}$ ) and in the liquidity-surplus regime ( $z_{exc,t}$ ) characterised by the disconnection of the EONIA from the main ECB policy rate.<sup>37</sup> According to the estimation, the first period of the liquidity-surplus regime is October 17, 2008, i.e. a few days after the announcement of the fixed-rate full-allotment procedure by the ECB. This regime was interrupted three times since then. The last interruption ended on August 2, 2011, two days before the ECB announced supplementary 6-month long-term refinancing operations (LTRO) in a context of renewed financial tensions.<sup>38</sup>

Searching for potential explanations of each change in regime is beyond the scope of this paper. For the sake of illustration, though, let me highlight an episode where monetary-policy-regime shifts can be directly related to central bankers' announcements.<sup>39</sup> During the press conference following the ECB Governing Council that took place on 5 June 2008, J.-C. Trichet said: "we could decide to move our rates [by] a small amount in our next meeting". As is shown in Figure 4, this triggered a change in the monetary-policy regime, from status quo to tightening. A rate hike was then decided by the Governing Council in the next meeting, on 3 July 2008. The latter meeting was however followed by a more dovish press conference by Trichet, which induced a return to the status-quo regime in the next few days.

Figure 5 illustrates the influence of the monetary-policy regimes on the term structure of interest rates. For three dates, the modelled yields are compared with the observed ones. For each date, three additional yield curves are displayed, each of them corresponding to one of the three monetary-policy regimes. The modelled yield curve corresponds to one of these three curves, the attribution being based on the smoothed probabilities associated with the Markov chain  $z_{m,t}$ .<sup>40</sup> The two remaining

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<sup>37</sup> Smoothing is based on Kim's (1993) algorithm.

<sup>38</sup> See the press release at [http://ecb.int/press/pr/date/2011/html/pr110804\\_1.en.html](http://ecb.int/press/pr/date/2011/html/pr110804_1.en.html).

<sup>39</sup> Naturally, central bankers' speeches are key events that are subject to indicate changes in monetary-policy regimes (see e.g. Rosa and Verga, 2008 in the euro-area case).

<sup>40</sup> In the present case, the smoothing algorithm results in a clear-cut identification of the hidden monetary-policy regime: Most of the time, the smoothed probabilities are either 1 or 0.

curves are the answers to the question: what if the monetary-policy stance were different on that date? These plots show that monetary-policy regimes are key to shape the yield curve. Furthermore, this figure illustrates the ability of the model to reproduce various shapes of the yield curve (steep, flat, humped, inverse-humped).

**[Insert Figures 6 and 7 and about here]**

Figure 6 displays the 30-day-ahead probabilities of change in the monetary-policy regime as well as in the policy rate as functions of the latter.<sup>41</sup> Both historical and risk-neutral probabilities are reported. Interestingly, all three monetary-policy regimes are more persistent under the risk-neutral measure than under the physical one, which can be seen from the fact that the risk-neutral probabilities of exiting a given monetary-policy phase are lower than their historical counterparts. The implications of the differences between the two dynamics (historical vs. risk-neutral) are explored in Section 5. Overall, the probabilities of monetary-policy changes substantially depend on the target rate: This appears on the plots of Figure 6 and is also reflected by the statistical significance of the parameters  $a_2$  (equation 2) that relate the probabilities of changes in the policy rate or in the monetary-policy regime to the level of the policy rate (see Table 2).

In this model, the volatility of the policy rate, and hence of the whole term structure of interest rates is not trivial. This is illustrated in Figure 7, that displays the standard deviation associated with the model-implied 3-month-ahead forecasts of the policy rate. The left-hand (right-hand) side plot regards the historical (risk-neutral) measure. The volatility of the policy rate turns out to strongly depend on the level of the rate itself as well as with the monetary-policy phase. Notably, these results echo those of Fontaine (2009) who finds –using U.S. data– that the uncertainty is lowest (highest) in tightening (loosening) cycles.

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<sup>41</sup> These probabilities are based on the matrix product  $\bar{\Pi}_{MP}\bar{\Pi}_{NMP}^{29}$ , where  $\bar{\Pi}_{MP}$  and  $\bar{\Pi}_{NMP}$  are the two possible matrices of transition probabilities for  $\bar{z}_t (= z_{r,t} \otimes z_{m,t})$ :  $\bar{\Pi}_{MP}$  (respectively  $\bar{\Pi}_{NMP}$ ) is the matrix that corresponds to a monetary-policy-meeting day (resp. a day without meeting).

[Insert Table 2 about here]

## 5 Term premia associated with target changes

The fact that the historical ( $\mathbb{P}$ ) and the risk-neutral ( $\mathbb{Q}$ ) dynamics of  $\bar{r}_t$  differ gives rise to target-related risk premia.<sup>42</sup> The existence of such term premia is important in several respects. Let me mention two of them. First, if these risk premia are sizeable, OIS forward rates should not be interpreted as the market perceptions of future target rates, though this is the basis of a widespread market practice (see Subsection 2.2). Second, the existence of risk premia at the short-end of the yield curve implies that excess returns associated with a long position in money-market instruments may be partially predictable or, alternatively said, that the expectation hypothesis does not hold at the short-end of the yield curve. While there is strong evidence against the expectation hypothesis for long-term yields, the evidence is weaker for short-term ones (see Longstaff, 2000).

[Insert Figure 8 about here]

In order to assess the size of target-related risk premia, policy-rate forecasts are computed under the two different measures. Conceptually, under the risk-neutral measure  $\mathbb{Q}$ , the forecasted paths of the policy rate are very close to the term structure of forward annualised rates (up to small Jensen-inequality correction terms). Here, emphasis is put on the risk premia associated with policy-rate changes, those associated with the  $s_t$  process having a straightforward and orthogonal influence.<sup>43</sup> Figure 8 displays the term structure of the policy-rate forecasts. Nine pairs of plots are reported. Each pair of plot corresponds to a given policy rate (1%, 2.5% or 4%)

<sup>42</sup> These target-related premia contribute to the total term premia, that also include risk premia associated with the  $s_t$  component of the EONIA.

<sup>43</sup> The mean reversion of  $s_t$  being far larger under the historical measure than under the risk-neutral measure, the risk premia associated with this factor are almost  $-B_h s_t/h$  (see Subsection 3.2 and equation (7) for details regarding the latter expression).

and a given monetary-policy phase (tightening, status quo or easing). For each pair of charts, an upper plot presents the forecasts of the policy rate (w.r.t. the horizon forecast, on the x axis) and a lower one displays the associated risk premia, i.e. the spread between the  $\mathbb{Q}$  and  $\mathbb{P}$  forecasts. 90% confidence intervals for the risk premia are reported in the lower charts.<sup>44</sup> These premia are discussed in the following.

First, it appears that the risk premia can be substantial, even at the short end of the yield curve. In particular, under the tightening regime (see the first column of charts in Figure 8), the risk premia are higher than 50 basis points for maturities higher than 12 months. Furthermore, for policy rates that are higher than the sample average (of about 2.5%), the risk premia associated with tightening and easing monetary-policy regimes turn out to have opposite signs at the short- to medium-end of the yield curve (see the second and third rows of pairs of charts, corresponding respectively to a 2.5% and a 4% policy rates). This stems from the fact that the probabilities of remaining in the tightening and easing regimes are higher under the risk-neutral measure than under the historical one (as shown in Figure 6), implying higher life expectancies for these regimes and, thereby, a higher probability –compared with the physical measure  $\mathbb{P}$ – of having several policy-rate moves in the next months or quarters. This translates into positive (negative) risk premia at the short end of the yield curve when the tightening (easing) regime prevails. Therefore, the estimation results suggest that under the risk-neutral measure, the central bank is more “aggressive”, in the sense that the yield curve reflects the behaviour of a central bank that tends to rise (respectively cut) the policy rate in a more rapid way than under the real-world measure when in the tightening (resp. easing) regime.<sup>45</sup> This supports the findings of Balduzzi et al. (1997), who observe that the target-change

<sup>44</sup> The confidence intervals are based on bootstrap techniques, the parameter estimates being drawn from their asymptotic distribution, see Figure 8’s caption for more details.

<sup>45</sup> Regarding the rise in rate, this is not any more the case for high target rates, since the risk-neutral probability of a rise in the target is lower than its historical counterpart when the policy rate is above 4%. However, note that the unconditional probability of being in the targeting regime when the target rate is higher than 4% is low (see lowest Panel of Figure 6).

predictions that may be obtained from the short-end of the yield curve –under the expectation hypothesis– are correct in terms of sign, but tend to overestimate the size of realised target moves.

## 6 Estimated impact of forward policy guidance

In the present framework, the behaviour of the central bank is modelled through a set of probabilities: some of them correspond to probabilities of switching from one regime to the other (tightening, easing and status quo), some of them correspond to probabilities of rises or cuts in the target rate (the latter being conditional to the monetary-policy regime). If a change in these probabilities is made public, it may have an impact on the whole yield curve because the pricing of financial assets depend in part on the entire expected future path of short-term interest rates. This expectation channel of monetary policy transmission is at the heart of the rationale for *forward policy guidance* measures. In the current context in which the zero bound is binding for the overnight nominal interest rate, these measures are aimed to provide additional stimulus to the economy by pushing down medium- to long-term interest rates and, thereby, to support other asset prices (see e.g. Bernanke and Reinhart, 2004).

My framework makes it possible to assess the impacts of such announcements in a straightforward and consistent manner. In the following, I consider a basic form of forward guidance in which a central bank commits itself to maintaining its target rate constant for (at least) a deterministic period of time. The recent decision by the U.S. Federal Reserve to release federal funds rate forecasts and to extend its pledge to keep rates near zero at least through late 2014 is of that kind.<sup>46</sup> In the past, other central banks have signalled future policy intentions through official

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<sup>46</sup> See the Fed press release at <http://www.federalreserve.gov/monetarypolicy/files/fomcprojtabl20120125.pdf>

communication. For instance, the Bank of Canada announced on April 21, 2009 its conditional commitment to “*hold current policy rate [close to zero] until the end of the second quarter of 2010.*”<sup>47</sup>

As in nearly all of the existing literature, the following simulations abstract from issues that could arise under imperfect credibility of the central bank and focus on the case where the monetary authorities benefit from a perfect commitment technology.

Let me assume that the central bank has announced at date  $t$  that it will keep its policy rate unchanged for the next  $p$  periods. Then, equation (7) can be used to compute the yields of different maturities, up to a few parameters’ adjustments: the matrix  $G(t, h)$  has simply to be replaced by  $\tilde{G}(t, h)$ , the latter being computed in the same way as the former (i.e. using the formulas presented in Appendix C) after having modified the matrices of  $\Pi_{t+i}^*$ ,  $i \leq p$  by setting the probabilities of policy-rate moves to zero.<sup>48</sup>

**[Insert Figure 9 about here]**

Figure 9 displays the results of four simulations. These simulations are based on two different target rates (1% and 3.5%) and two commitment durations (12 months and 24 months). Consistently with the fact that the policy rate is fixed for several months, the monetary-policy regime is set to the status-quo one (in the baseline as well as in the counterfactual case). The results suggest that such measures would have a statistically significant downward impact on the yield curve (90% confidence intervals of the downward effects are reported for each of the four cases presented in Figure 9). The impact appears to be far larger when the current target rate is low. For instance, a commitment to keep the target rate unchanged for the next 24

<sup>47</sup> There exist older cases of forward policy guidance: the Reserve Bank of New Zealand announced a path for its 3-month bank bill rate in 1997, it was followed by the Norges Bank and the Riskbank in 2005 and 2007, respectively.

<sup>48</sup> The fact that the probabilities of having policy-rate moves over the next  $p$  periods are equal to zero implies that the same is true under the risk-neutral measure because  $\mathbb{P}$  and  $\mathbb{Q}$  are equivalent measures. If this was not the case, it would imply the existence of infinitely large Sharpe ratios associated with policy-rate changes.

months leads to a decrease in the 5-year yield of about 25 bp when the target rate is of 1% and of about 5 bp when the target rate is of 3.5%.

## 7 Conclusion

While central banks' decisions are obvious drivers of the fluctuations of the term structure of interest rates, only few of the available yield-curve models feature a realistic modeling of the policy rate. This paper proposes a framework that captures simultaneously the dynamics of the policy rate and the yields of longer maturities. Importantly, this model is consistent with the existence of the zero-lower-bound restriction, making it appealing in the current context of extremely low interest rates.

A key ingredient of the model is an extensive and innovative use of switching-regime features. Each regime is characterised by (a) a target level, (b) a monetary-policy regime (easing, tightening or status quo) and (c) the Eurosystem aggregate liquidity situation (normal or “in surplus”). The latter is introduced so as to accommodate the recent situation in which banks resort massively to the ECB deposit facility, which has an impact on the overnight interbank rate –the shortest-maturity yield considered in the model.

In order to illustrate the flexibility of the model, it is estimated using daily data covering the last thirteen years. Consistently with the choice of the interbank rate (EONIA) as the shortest yield, the overnight index swap (OIS) curve is fitted. Being impressively tractable, the model is estimated by standard maximum likelihood techniques. In order to alleviate potential small-sample bias and, hence, to properly estimate the physical dynamics of the processes, the estimation data set includes survey-based forecasts of short-term rates.

Various by-products are available, including the estimation of the market-perceived monetary-policy regime (at the daily frequency). In addition, the model is used in or-

der to explore the size and influence of risk premia at the short end of the yield curve, the approach making it possible to exhibit monetary-policy-related risk premia. My analysis suggests that market yields reflect the behaviour of a central bank that would tend to rise (respectively cut) the target rate more rapidly than is physically observed when in a tightening (resp. easing) phase. This has implications regarding the common practice that consists in inverting the OIS yield curve to extract market-based short-term forecasts of the policy-rate path. Specifically, it means that such a practice –that assumes that the expectation hypothesis holds at the short-end of the yield curve– is valid in terms of sign of next target changes, but tend to overestimate their size.

Finally, the model is exploited to predict the potential effects of a forward policy guidance measure that consists of a commitment of the central bank to keep its rate unchanged for (at least) a given period of time. The simulations show that, in the current context of low short-term rates and with a commitment duration of 2 years, such an (unanticipated) announcement would be followed by a decrease of about 25 basis points of the 5-year rate.

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## A The $\mathcal{L}$ distribution

The  $\mathcal{L}$  distribution accommodates non-zero skewness and fat tails. A random variable follows the distribution  $\mathcal{L}(p, \alpha_P, \beta_P, \alpha_N, \beta_N)$  if it is equal to  $\mathbb{I}_{\{u=0\}}v_P - \mathbb{I}_{\{u=1\}}v_N$ , where  $u$  is Bernoulli distributed with success probability  $p$ , and where  $v_P$  and  $v_N$  follow beta distributions with respective parameters  $(\alpha_P, \beta_P)$  and  $(\alpha_N, \beta_N)$  [“ $P$ ” and “ $N$ ” respectively stand for “positive” and “negative”].

The bond-pricing formula (Subsection 3.2 and Appendix C.2) require the computation of  $E(\exp[\xi])$ , where  $\xi \sim \mathcal{L}(p, \alpha_P, \beta_P, \alpha_N, \beta_N)$ :

$$\begin{aligned} E(\exp[\xi]) &= pE(\exp \xi_P) + (1 - p)E(\exp \xi_N) \\ &= p.f(\alpha_P, \beta_P) + (1 - p).f(\alpha_N, \beta_N) \end{aligned}$$

where  $f(\alpha, \beta)$  is given by:

$$f(\alpha, \beta) = 1 + \sum_{k=0}^{\infty} \frac{1}{k!} \left( \prod_{i=0}^{k-1} \frac{\alpha + i}{\alpha + \beta + i} \right).$$

## B Multi-horizon Laplace transform of a (homogenous) Markov-switching process

In the following, I consider a  $n$ -state Markov process  $z_t$ , valued in  $\{e_1, \dots, e_n\}$ , the set of columns of  $I_n$ , the identity matrix of dimension  $n \times n$ . I assume that the matrix of transition probabilities is deterministic and denoted by  $P_t$  (the columns sum to one). We have:  $\mathbb{P}(z_{t+1} = e_i | z_t) = e_i' P_{t+1} z_t$ .

### Computation of $E_t(\exp(\alpha' z_{t+1}))$

$$\begin{aligned} E_t(\exp(\alpha' z_{t+1})) &= \sum_{i=1}^n \exp(\alpha_i) e_i' P_{t+1} z_t \\ &= \left( \sum_{i=1}^n \exp(\alpha_i) e_i' \right) P_{t+1} z_t \\ &= [1 \ \cdots \ 1] D(\exp \alpha) P_{t+1} z_t \end{aligned}$$

where  $\exp \alpha$  is the vector whose entries are the  $\exp(\alpha_i)$ 's and where  $D(x)$  is a diagonal matrix whose diagonal entries are the elements of the vector  $x$ .

### Computation of $E_t(\exp[\alpha_1' z_{t+1} + \alpha_2' z_{t+2}])$

The law of iterated expectations leads to:

$$\begin{aligned} E_t(\exp[\alpha_1' z_{t+1} + \alpha_2' z_{t+2}]) &= E_t(E_t[\exp[\alpha_1' z_{t+1} + \alpha_2' z_{t+2}] | z_{t+1}]) \\ &= E_t(\exp[\alpha_1' z_{t+1}] E_t[\exp[\alpha_2' z_{t+2}] | z_{t+1}]) \end{aligned}$$

Then, using the previous case:

$$\begin{aligned} &E_t(\exp[\alpha_1' z_{t+1} + \alpha_2' z_{t+2}]) \\ &= E_t(\exp[\alpha_1' z_{t+1}] [1 \ \cdots \ 1] D(\exp \alpha_2) P_{t+2} z_{t+1}) \\ &= E_t([1 \ \cdots \ 1] D(\exp \alpha_2) P_{t+2} z_{t+1} \exp[\alpha_1' z_{t+1}]) \\ &= E_t([1 \ \cdots \ 1] D(\exp \alpha_2) P_{t+2} z_{t+1} z_{t+1}' D(\exp \alpha_1) [1 \ \cdots \ 1]')'. \end{aligned}$$

Using the facts that  $z_{t+1} z_{t+1}'$  commutes with any matrix and that  $z_{t+1} z_{t+1}' [1 \ \cdots \ 1]' = z_{t+1}$ , we get:

$$\begin{aligned} E_t(\exp[\alpha_1' z_{t+1} + \alpha_2' z_{t+2}]) &= E_t([1 \ \cdots \ 1] D(\exp \alpha_2) P_{t+2} D(\exp \alpha_1) z_{t+1}) \\ &= [1 \ \cdots \ 1] [D(\exp \alpha_2) P_{t+2}] [D(\exp \alpha_1) P_{t+1}] z_t. \end{aligned}$$

## Generalisation

It is straightforward to generalise and to show that:

$$E_t(\exp [\alpha'_1 z_{t+1} + \dots + \alpha'_h z_{t+h}]) = \begin{bmatrix} 1 & \dots & 1 \end{bmatrix} [D(\exp \alpha_h) P_{t+h}] \times \dots \\ \dots \times [D(\exp \alpha_1) P_{t+1}] z_t.$$

## C Pricing formulas

In this appendix, I detail the computation of the three multiplicative components of  $P(t, h)$  (the price at date  $t$  of a bond with residual maturity  $h$ ), namely  $P_1(t, h)$ ,  $P_2(t, h)$  and  $P_3(t, h)$ . More precisely, this appendix propose a way to compute  $G_1(t, h)$ ,  $G_2(t, h)$ ,  $A_h$  and  $B_h$  that are such that:

$$\begin{cases} P_1(t, h) & = G_1(t, h) z_t \\ P_2(t, h) & = G_2(t, h) z_t \\ P_3(t, h) & = \exp(A_h + B_h s_t) \end{cases}$$

These formulas leads to equation (7).<sup>49</sup>

### C.1 Computation of $P_1(t, h)$

The targets  $\bar{r}_t$  are the only stochastic variables involved in the computation of  $P_1(t, h)$ . The previous Appendix shows that the expectation of an exponential-affine combination of a variable that follows a Markov-switching process is available in closed form. This leads to the following formula:

$$P_1(t, h) = E_t^{\mathbb{Q}} \left( \exp \left[ - \sum_{i=0}^{h-1} \bar{r}_{t+i} \right] \right) = G_1(t, h) z_t \quad (10)$$

$$\text{with } G_1(t, h) = \begin{bmatrix} 1 & \dots & 1 \end{bmatrix} \left[ \prod_{i=h-1}^1 D(\exp [-\Delta_m]) \Pi_{t+i}^* \right] D(\exp [-\Delta_m])$$

and where

- $D(x)$  is a diagonal matrix whose diagonal entries are those of the vector  $x$ .
- The matrices  $\Pi_t^*$ , which are of dimension  $6(N+1) \times 6(N+1)$ , contain the risk-neutral probabilities of switching from one regime –defined by a policy rate, a monetary-policy regime and a bank's liquidity situation– to another. As their physical-measure counterparts, these matrices can take two values, depending on whether a monetary-policy meeting is scheduled at date  $t$  or not.
- The product operator  $\prod$  works in a backward direction: if  $X_1$  and  $X_2$  are some square matrices,  $\prod_{i=2}^1 X_i = X_2 X_1$

<sup>49</sup> In equation (7), the  $i^{th}$  entry of  $G(t, h)$  is the logarithm of the  $i^{th}$  entry of  $G_1(t, h) + G_2(t, h)$ .

It is important to stress that this formula does not require the use of time-demanding recursive algorithms used by most alternative discrete-time affine term-structure models. Since policy meetings do not take place at a fully regular frequency, the matrices  $G_t$  should be computed for every date. As in Piazzesi (2005), I resort however to an intermediate approach where I consider only the exact number of days until the next decision meeting whereas subsequent meetings are assumed to be equally spread (every 30 days). The latter approximation, that leads to the computation of (only) 31 matrices  $G_i$  (instead of one per day), results in negligible pricing errors.

## C.2 Computation of $P_2(t, h)$

The computation of  $E^{\mathbb{Q}} \left( \exp \left[ - \sum_{i=0}^{h-1} \xi_{t+i} \right] \right)$  is very close to this of  $P_1(t, h)$ . Indeed, using the law of iterated expectations, it comes:

$$P_2(t, h) = E_t^{\mathbb{Q}} \left( E^{\mathbb{Q}} \left[ \exp \left( - \sum_{i=0}^{h-1} \xi_{t+i} \right) \middle| z_{exc,t+1}, \dots, z_{exc,t+h-1} \right] \right). \quad (11)$$

Then remark that  $w_{t+i} + \xi_{t+i} = [(w_{norm} + \xi_{norm,t+i}) (w_{exc} + \xi_{exc,t+i})] z_{exc,t+i}$  and recall that the  $\xi$ 's follow  $\mathcal{L}$  distributions based on beta distributions. Appendix A gives the Laplace transform of a variable drawn from a  $\mathcal{L}$  distribution, which provides us with  $E(\exp(-\xi_{j,t}))$  for  $j \in \{norm, exc\}$ . This leads to:

$$E_t^{\mathbb{Q}} \left( \exp \left[ - \sum_{i=0}^{h-1} \xi_{t+i} \right] \right) = E_t^{\mathbb{Q}} \left( \exp \left[ \sum_{i=0}^{h-1} [ \vartheta_{norm} \quad \vartheta_{exc} ] z_{exc,t} \right] \right)$$

where  $\exp \vartheta_j = E(\exp(-w_j - \xi_{j,t}))$ . Then, using Appendix B again, one obtains:

$$P_2(t, h) = G_2(t, h) z_t \quad (12)$$

$$\text{with } G_2(t, h) = \begin{bmatrix} 1 & 1 \end{bmatrix} \left\{ D \left( \exp \begin{bmatrix} \vartheta_{norm} \\ \vartheta_{exc} \end{bmatrix} \right) \Pi_{exc} \right\}^{h-1} D \left( \exp \begin{bmatrix} \vartheta_{norm} \\ \vartheta_{exc} \end{bmatrix} \right) H_{exc}$$

where  $H_{exc}$  is the selection matrix (whose entries are 0 or 1) that is such that  $z_{exc,t} = H_{exc} z_t$ .

## C.3 Computation of $P_3(t, h)$

We have

$$P_3(t, h) = E_t^{\mathbb{Q}} e^{-\sum_{i=0}^{h-1} s_{t+i}} = E_t^{\mathbb{Q}} e^{-\sum_{i=0}^{h-1} s_{1,t+i} + s_{2,t+i}}$$

where  $\begin{bmatrix} s_{1,t} \\ s_{2,t} \end{bmatrix} = \Phi^* \begin{bmatrix} s_{1,t-1} \\ s_{2,t-1} \end{bmatrix} + \Sigma \varepsilon_t^*$ ,  $\varepsilon_t^* \sim \text{i.i.d. } \mathcal{N}^{\mathbb{Q}}(0, I)$ .

In that Appendix, I describe an algorithm originally presented by Borgy et al. (2011). This algorithm results in the same matrices than the recursive formula given

in the seminal paper by Ang and Piazzesi (2003). However, this latter approach turns out to be time-demanding for high-frequency (weekly or daily) processes. As shown by Borgy et al. (2011), the algorithm described below is substantially quicker when  $h$  is large.

Let me denote by  $X_t$  the vector  $[s_{1,t}, s_{2,t}, s_{1,t-1}, s_{2,t-1}]'$ .  $X_t$  follows:

$$X_t = \tilde{\mu}^* + \tilde{\Phi}^* X_{t-1} + \tilde{\Sigma} \varepsilon_t^*, \quad \varepsilon_t^* \sim \mathcal{N}^{\mathbb{Q}}(0, I),$$

where  $\tilde{\mu}^*$ ,  $\tilde{\Phi}^*$  and  $\tilde{\Sigma}$  are easily deduced from  $\mu^*$ ,  $\Phi^*$  and  $\Sigma$ . In the following, I show how to compute the vectors  $A_h$  and  $C_h$  that are such that

$$P_3(t, h) = E_t^{\mathbb{Q}}(\exp(\delta' X_{t+1} + \dots + \delta' X_{t+h})) = \exp(A_h + C_h X_t)$$

where  $\delta = [0, 0, 1, 1]'$ . Denoting by  $F_{t,t+h}$  the random variable  $X_{t+1} + \dots + X_{t+h}$ , we get:

$$P_3(t, h) = E_t^{\mathbb{Q}}(\exp(\delta' F_{t,t+h}))$$

Note that  $F_{t,t+h}$  is a Gaussian random variable. We have

$$\begin{aligned} F_{t+h,h} &= (hI + (h-1)\Phi^* + \dots + \Phi^{*(h-1)}) \mu^* + \\ &\quad (\Phi^* + \Phi^{*2} + \dots + \Phi^{*h}) X_t + \\ &\quad (I + \dots + \Phi^{*(h-1)}) \varepsilon_{t+1}^* + (I + \dots + \Phi^{*(h-2)}) \varepsilon_{t+2}^* + \dots + \varepsilon_{t+h}^*. \end{aligned}$$

Therefore  $F_{t+h,h} \sim \mathcal{N}^{\mathbb{Q}}(\Lambda_{0,h} + \Lambda_h X_t, \Omega_h)$  with

$$\begin{cases} \Lambda_h &= \Phi^* (\Phi^{*h} - I) (\Phi^* - I)^{-1} \\ \Lambda_{0,h} &= [\chi_{1,h} - hI] (\Phi^* - I)^{-1} \mu^* \end{cases}$$

and with

$$\begin{aligned} \Omega_h &= \text{Var}((I + \dots + \Phi^{*(h-1)}) \varepsilon_{t+1}^* + (I + \dots + \Phi^{*(h-2)}) \varepsilon_{t+2}^* + \dots + \varepsilon_{t+h}^*) \\ &= (\Phi^* - I)^{-1} \left[ (\Phi^{*h} - I) \Sigma \Sigma' (\Phi^{*h} - I)' + \dots \right. \\ &\quad \left. + (\Phi^* - I) \Sigma \Sigma' (\Phi^* - I)' \right] (\Phi^* - I)^{-1} \\ &= (\Phi^* - I)^{-1} [(h-1)\Sigma \Sigma' - \Lambda_h \Sigma \Sigma' - \Sigma \Sigma' \Lambda_h' + \Pi(h, \Phi^*, \Sigma)] (\Phi^* - I)^{-1} \end{aligned}$$

where  $\Pi : (h, \Phi^*, \Sigma) \rightarrow (\Phi^{*h}) \Sigma \Sigma' (\Phi^{*h})' + \dots + (\Phi^*) \Sigma \Sigma' (\Phi^*)' + \Sigma \Sigma'$ . Instead of using a brute-force approach (based on  $h$  loops) to compute  $\Pi(h, \Phi^*, \Sigma)$ , we exploit the fact that  $\Pi(kp, \Phi^*, \Sigma) = \Pi(k, \Phi^{*p}, \Pi(p, \Phi^*, \Sigma) - \Sigma \Sigma') + \Sigma \Sigma'$ . This can substantially reduce the computation time to compute. It suffices to apply the latter formula a few times, based on an integer factorization of  $h$ . Finally

$$\begin{cases} A_h &= \delta' \Lambda_{0,h} + \frac{1}{2} \delta' \Omega_h \delta \\ C_h &= \delta' \Lambda_h. \end{cases}$$

Finally, since  $s_{1,t} \equiv 0$  under  $\mathbb{P}$ , denoting by  $B_h$  the second column of  $C_h$ , we get:

$$P_3(t, h) = \exp(A_h + B_h s_t). \quad (13)$$

## D Computation of the likelihood

$$\begin{cases} R_t - \Lambda_{z,R} z_t - \Lambda_{s,R} \Lambda_s (R_t - \Lambda_{z,R} z_t) & = \xi_t^R \\ CF_t - \Lambda_{z,C} z_t - \Lambda_{s,C} \Lambda_s (R_t - \Lambda_{z,R} z_t) & = +\xi_t^C \end{cases} \quad (14)$$

$$\begin{cases} (I - \Lambda_{s,R} \Lambda_s) R_t - \Lambda_{z,R} z_t + \Lambda_{s,R} \Lambda_s \Lambda_{z,R} z_t & = \xi_t^R \\ CF_t & = \Lambda_{z,C} z_t + \Lambda_{s,C} s_t + \xi_t^C \end{cases} \quad (15)$$

This Appendix complements Subsection 4.2. For a given vector of observed yields  $R_t$  and a regime vector  $z_t$ , the latent factor  $s_t$  is given by  $s_t = \Lambda_s (R_t - \Lambda_{z,R} z_t)$  with  $\Lambda_s = (\Lambda'_{s,R} \Lambda_{s,R})^{-1} \Lambda'_{s,R}$  (this is eq. 9). For a given regime vector  $z_t$ , there is the same information in  $R_t$  as in  $\{s_t, \tilde{R}_t\}$ , where  $\tilde{R}_t$  is any subvector of  $R_t$  containing  $M - 1$  yields of distinct maturities. Without loss of generality, I assume that  $\tilde{R}_t = [y(t, h_2), \dots, y(t, h_M)]'$ . As a consequence, from the econometrician point of view, the model reads:

$$\begin{cases} \tilde{\Gamma} \xi_t^R & = \tilde{\Gamma} \{(I - \Lambda_R) R_t - (I - \Lambda_R) \Lambda_{z,R} z_t\} \\ \xi_t^C & = CF_t - (\Lambda_{z,C} - \Lambda_C \Lambda_{z,R}) z_t - \Lambda_C R_t \\ \varepsilon_t & = \frac{1}{\sigma} \Lambda_s [(R_t - \rho_2 R_{t-1}) - \Lambda_{z,R} (z_t - \rho_2 z_{t-1})] \end{cases}$$

where  $\Lambda_R = \Lambda_{s,R} \Lambda_s$ ,  $\Lambda_C = \Lambda_{s,C} \Lambda_s$  and where  $\tilde{\Gamma}$  is the  $(M - 1) \times M$  matrix that selects the last  $M - 1$  entries of an  $M \times 1$  vector.

Assuming that the  $\varepsilon_t$ 's, the  $\xi_t^R$ 's and the  $\xi_t^{cf}$ 's are i.i.d. normal, the computation of the log-likelihood associated with the previous model is easily obtained by applying the Kitagawa-Hamilton filter. However, this likelihood is the one associated with the vector  $\{s_t, \tilde{R}_t, CF_t\}_{t=1, \dots, T}$ , while we need to maximise the one associated with actually observed data  $\{R_t, CF_t\}_{t=1, \dots, T}$ . The latter is obtained by multiplying the former by the determinant of the Jacobian resulting from this change in variables, that is  $\left| \partial \begin{bmatrix} s_t \\ \tilde{R}_t \end{bmatrix} / \partial R_t \right| = \frac{1}{\Lambda'_{s,R} \Lambda_{s,R}} \Lambda_{s,R,1}$  where  $\Lambda_{s,R,1}$  is the first entry of  $\Lambda_{s,R}$ .

## E Tables and Figures

Tab. 1: Descriptive statistics of yields

Notes: The table reports summary statistics for selected yields. The data are monthly and cover the period from January 1999 to February 2012. Two auto-correlations are shown (the 1-day and the 1-year auto-correlations). The yields are continuously compounded and are in percentage annual terms. Panel B presents the covariances and the correlations of the yields. The EONIA spread is the yield differential between the (annualised) EONIA and the target rate. Panel C reports some results of a principal-component analysis carried out on the spreads between the yields and the target rate. More precisely, it shows the share of the variances of the different spreads that are explained by the first three principal components. Two samples are considered: January 1999 to February 2012 and January 1999 to August 2008.

Panel A - Descriptive statistics 1999-2012								
	Target	EONIA	1-mth	3-mth	6-mth	12-mth	2-yr	4-yr
Mean	2,64	2,56	2,58	2,60	2,64	2,74	2,94	3,35
Median	2,50	2,57	2,60	2,64	2,67	2,77	2,91	3,37
Standard dev.	1,15	1,34	1,33	1,34	1,34	1,34	1,28	1,16
Skewness	0,14	0,14	0,14	0,14	0,13	0,13	0,17	0,22
Kurtosis	1,91	2,04	2,01	1,95	1,93	1,92	2,02	2,23
Auto-cor. (1 day)	1,00	1,00	1,00	1,00	1,00	1,00	1,00	1,00
Auto-cor. (1 year)	0,52	0,56	0,57	0,57	0,57	0,58	0,60	0,62

Panel B – Correlations   Covariances								
	Target	EONIA	1-mth	3-mth	6-mth	12-mth	2-yr	4-yr
Target	1,31	1,51	1,50	1,49	1,48	1,44	1,34	1,16
EONIA	0,98	1,80	1,77	1,77	1,75	1,71	1,60	1,39
1-mth OIS	0,98	0,99	1,77	1,77	1,76	1,73	1,62	1,41
3-mth OIS	0,98	0,99	1,00	1,78	1,79	1,76	1,65	1,43
6-mth OIS	0,96	0,98	0,99	1,00	1,80	1,78	1,68	1,46
12-mth OIS	0,94	0,96	0,97	0,99	1,00	1,79	1,70	1,48
2-yr OIS	0,91	0,93	0,95	0,96	0,98	0,99	1,65	1,46
4-yr OIS	0,87	0,90	0,91	0,93	0,94	0,96	0,99	1,34

Panel C – Principal component analysis of spreads vs. target								
1999-2012								
	Eonia	1-mth	3-mth	6-mth	12-mth	2-yr	4-yr	Total
1st PC	0,51	0,80	0,93	0,96	0,93	0,80	0,48	0,77
2d PC	0,86	0,96	0,97	0,96	0,96	0,99	0,91	0,94
3rd PC	0,99	0,96	0,99	1,00	0,98	0,99	0,99	0,98

1999-2008								
	Eonia	1-mth	3-mth	6-mth	12-mth	2-yr	4-yr	Total
1st PC	0,03	0,58	0,89	0,93	0,95	0,89	0,63	0,70
2d PC	0,84	0,74	0,90	0,93	0,96	0,93	0,71	0,86
3rd PC	0,99	0,85	0,97	0,96	0,96	0,99	0,95	0,95

Tab. 2: Parameter estimates

Notes: The table reports the estimates of the parameters defining the dynamics of the factor under historical and risk-neutral measures. The estimation data are daily and span the period from January 1999 to February 2012. Standard errors are reported in parenthesis. The sign “\*” (after a number) denotes significance at the 5% level. The parameters  $a_i$  relate the probabilities of changes in the policy rate or in the monetary-policy regime to the level of the policy rate (see Subsection 3.1.1 and notably equation 2). The parameters that define the risk-neutral dynamics are indicated by \*. The dynamics of the Markov chain  $z_{exc,t}$  is defined by  $p_{exc,exc}$  and  $p_{norm,norm}$  which are, respectively, the probabilities of remaining in the excess-liquidity regime and the non-excess-liquidity regime.  $\sigma_{fcst}$  and  $\sigma_{pric}$  are, respectively, the standard deviations of the measurement errors  $\xi_t^C$  and of the pricing errors  $\xi_t^R$  (see equation 8).

	$a_1^*$	$a_2^*$	$k^*$	$a_1$	$a_2$	$k$
rise in the target	-0 (0.81)	0.11* (0.036)	0.61 (0.58)	1.2* (0.23)	-0.17* (0.014)	0.82* (0.105)
cut in the target	0.41 (0.46)	0.098* (0.02)	0.38 (0.44)	-0.00011 (0.31)	0.11* (0.0104)	0.49* (0.14)
<i>ES</i>	3.8* (0.065)	0.4* (0.022)		3.9* (0.22)	0.065* (0.032)	
<i>SE</i>	9* (0.49)	-0.87* (0.091)		9* (0.32)	-1.3* (0.072)	
<i>TS</i>	4.7* (0.068)	0.25* (0.018)		4.1* (0.37)	0.19* (0.058)	
<i>ST</i>	5.3* (0.081)	0.109 (0.058)		5.4* (0.16)	-0.33* (0.09)	
$\alpha_{P,norm}$	$\alpha_{N,norm}$	$\beta_{P,norm}$	$\beta_{N,norm}$	$\mu_{norm}$	$p_{norm}$	
0.76* (0.035)	0.22* (0.0056)	8.3* (0.13)	0.86* (0.064)	0.04* (0.00012)	0.48* (0.0105)	
$\alpha_{P,exc}$	$\alpha_{N,exc}$	$\beta_{P,exc}$	$\beta_{N,exc}$	$\mu_{exc}$	$p_{exc}$	
0.75* (0.05)	0.5* (0.038)	4.6* (0.25)	9* (1.6)	-0.65* (0.0012)	0.55* (0.021)	
$\rho_1^*$	$\rho_2^*$	$\beta^*$	$\rho_2$	$\sigma$		
0.9999* (0)	0.9999* (0)	16* (6.1)	0.94* (0.0012)	0.00007* (0.00003)		
$p_{norm,norm}^*$	$p_{exc,exc}^*$	$p_{norm,norm}$	$p_{exc,exc}$	$\sigma_{fcst}$	$\sigma_{pric}$	
0.9999* (0.0001)	0.9999* (0.00014)	0.99* (0.00093)	0.9999* (0)	0.23* (0.0029)	0.085* (0.00021)	

Fig. 1: Target rate, EONIA and OIS

Notes: Panel A shows the target rate together with the overnight interbank interest rate (EONIA). Panel B displays the EONIA spread, i.e. the spread between the EONIA and the target. The four vertical bars in Panel B indicate the four following dates, respectively: 8 October 2008 (introduction of Fixed-Rate Full Allotment procedures), 3 December 2009 (announcement of the phasing out of the very long-term refinancing operations), 4 August 2011 (given the renewed financial-market tensions, announcement of supplementary 6-month LTRO), 8 December 2011 (3-year VLTRO). Panel C presents the target rate and two OIS yields, the spreads between the latter and the target being reported in Panel D.

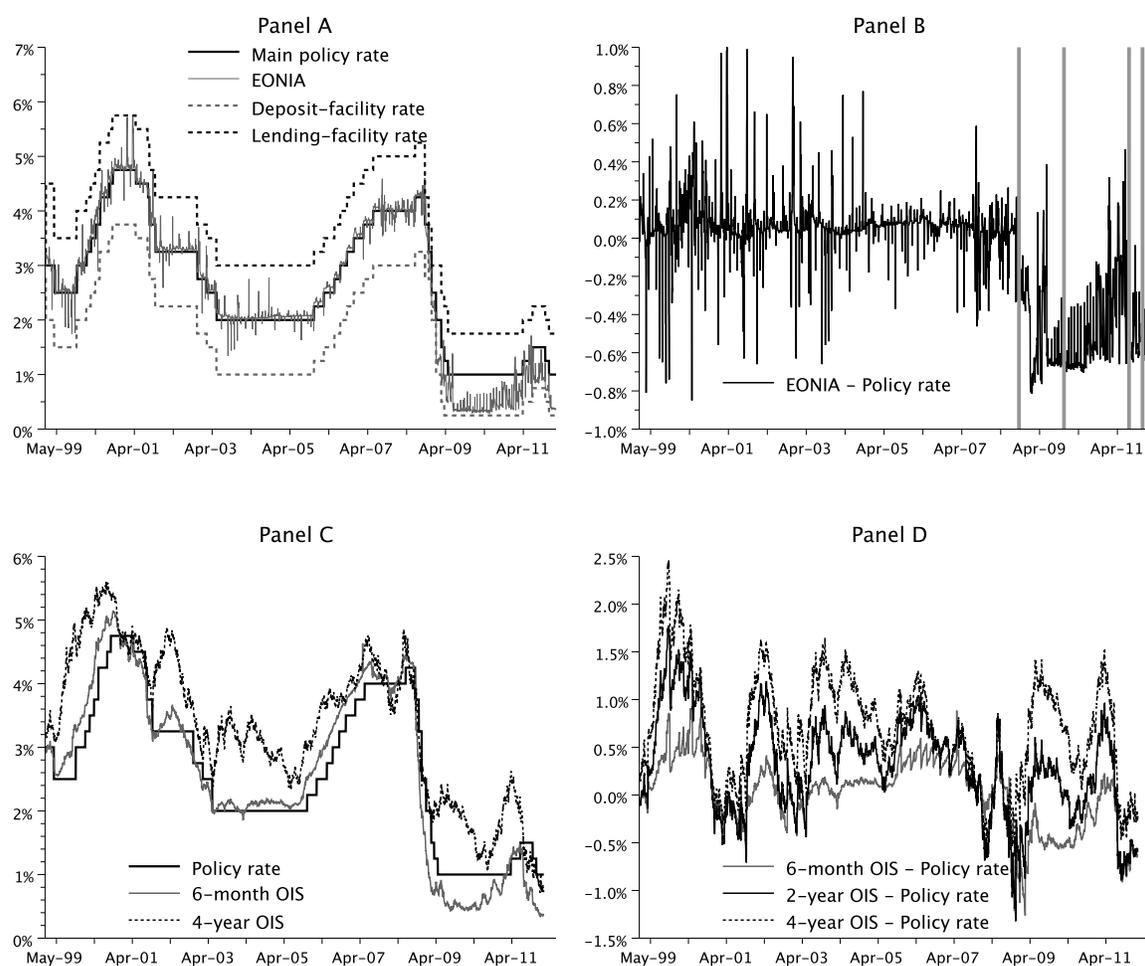


Fig. 2: Estimated  $s_t$  process and model fit

Notes: Panel A displays the estimated  $s_t$  process (see equation (9)). Panels B, C and D compare model-implied yields with their data (actual) counterparts. The latter panels also display (grey dashed line) the part of the model-implied yields that is accounted for by the regime variable  $z_t$  (that is  $-\frac{1}{h} [G(t, h)z_t + A_h]$  in equation 7).

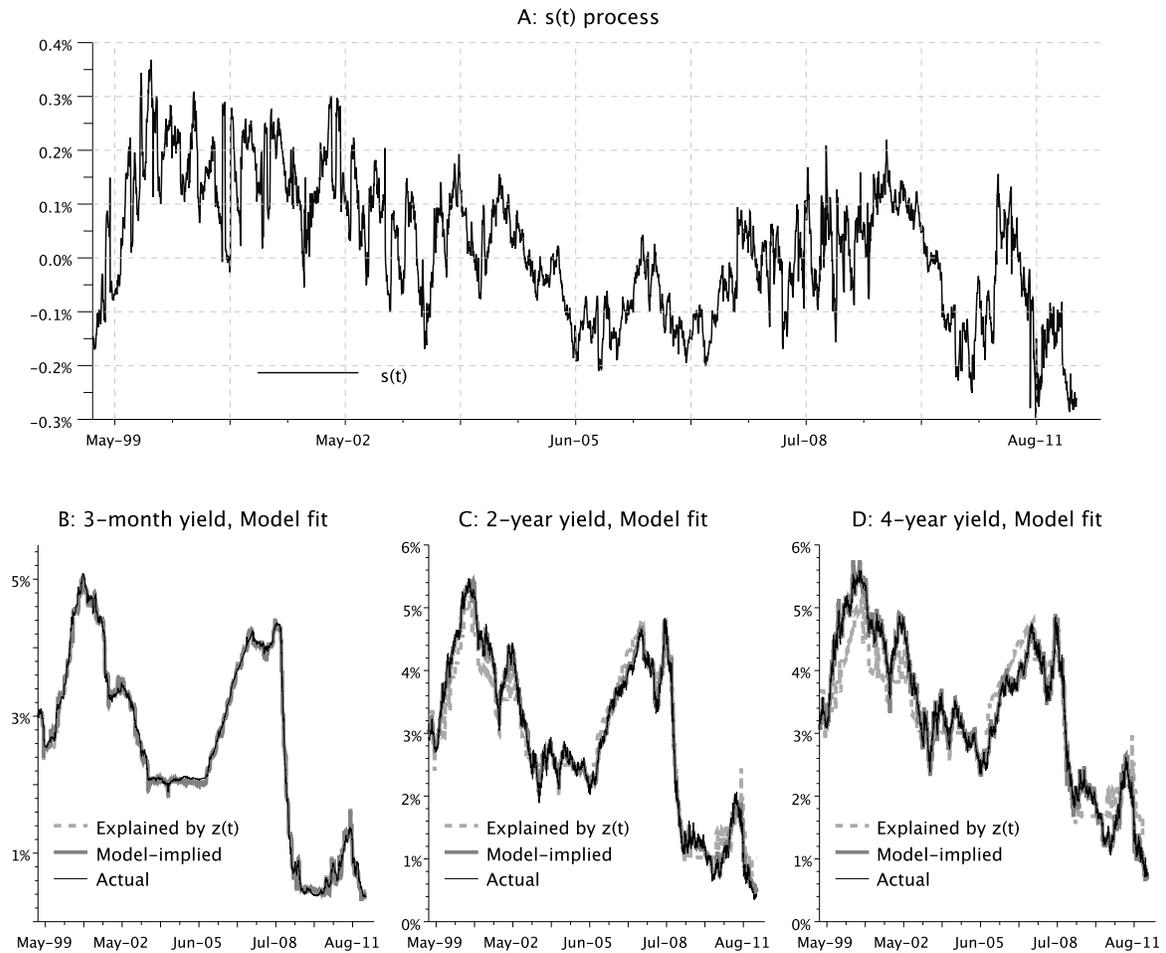


Fig. 3: Regimes' estimates

Notes: Panel A compares the model-implied forecasts with the survey-based ones (*Consensus Forecasts*). Panel B displays the (smoothed) probabilities of being in the different monetary-policy regimes. Panel C shows the smoothed probabilities of being in the excess-liquidity regime. The four vertical lines reported in Panel C indicate the following dates: 8 October 2008 (introduction of Fixed-Rate Full Allotment procedures), 3 December 2009 (announcement of the phasing out of the very long-term refinancing operations), 4 August 2011 (given the renewed financial-market tensions, announcement of supplementary 6-month LTRO), 8 December 2011 (3-year VLTRO).

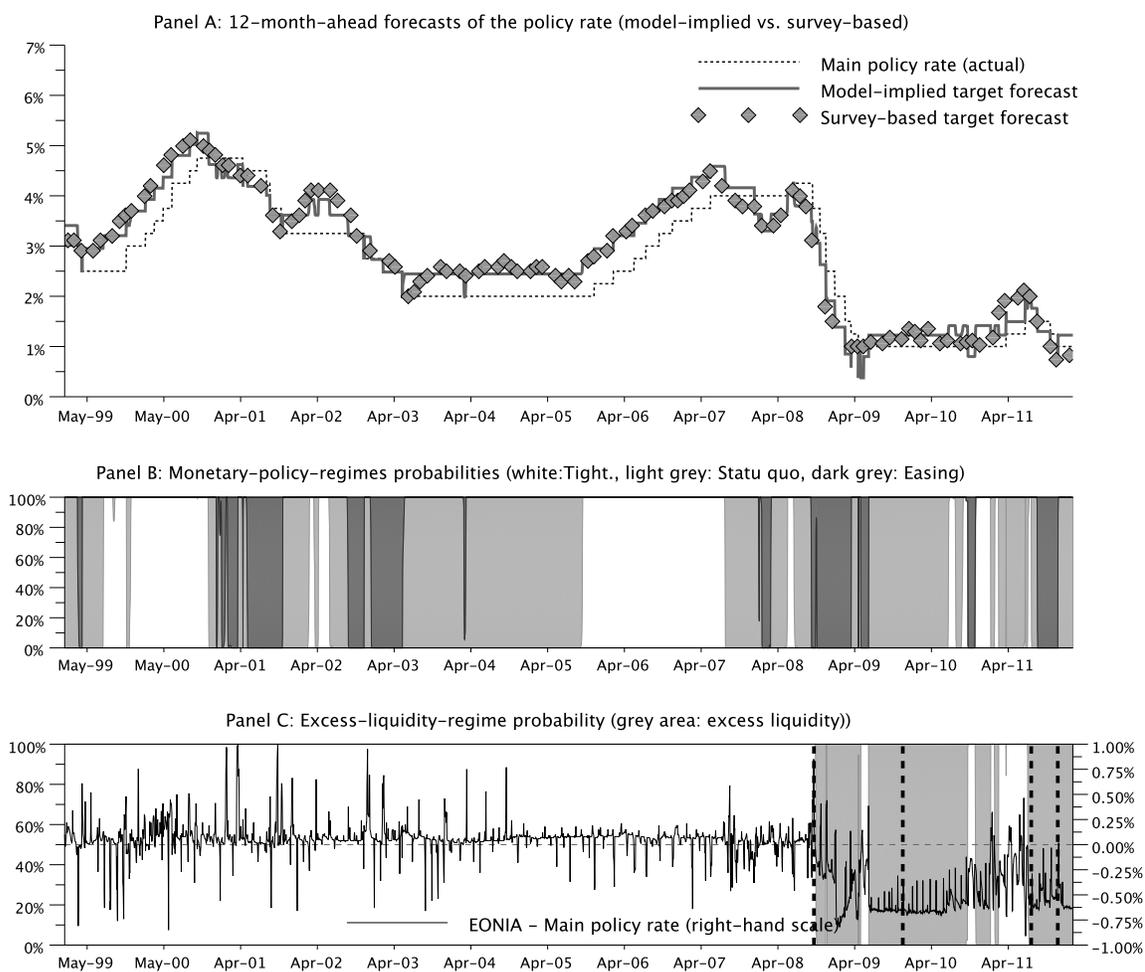


Fig. 4: Changes in monetary-policy regimes and central-bankers' announcements

Notes: This Figure relates some specific (estimated) changes in the monetary-policy regimes to specific central bankers speeches (summer 2008). The grey shaded area corresponds to the (smoothed) probability of being in the tightening monetary-policy regime. The two vertical bars indicate the dates of two subsequent ECB governing councils (5 June 2008 and 3 July 2008). The quotation from J.-C. Trichet comes from the "Questions & Answers" part of the Press Conference held at the ECB on 5 June 2008. The extract from the *Financial Times* comes from [www.ft.com](http://www.ft.com) (article entitled "ECB raises interest rates to 4.25%", by Ralph Atkins).

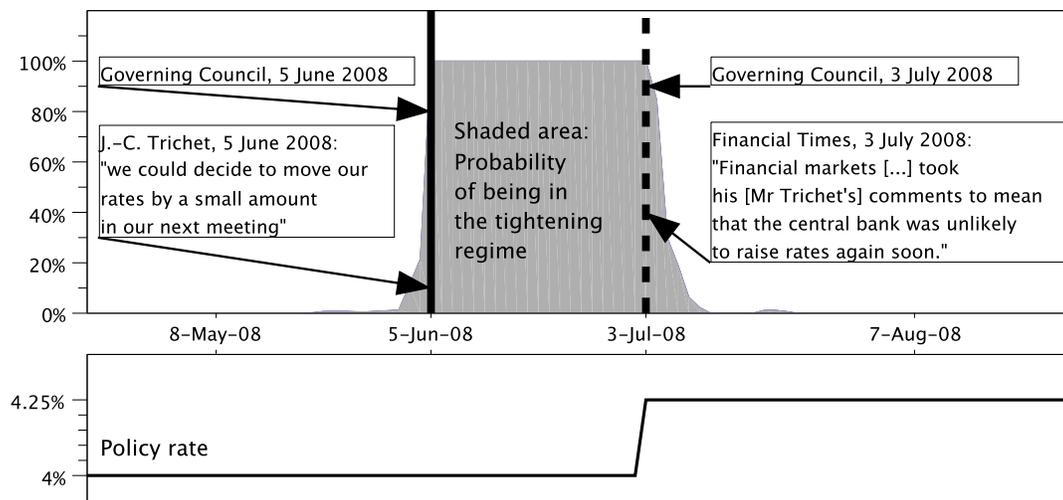


Fig. 5: Fitted yield curves and influence of monetary-policy regimes

Notes: These plots compare model-implied (diamonds) with observed (black circles) yield curves at different dates. In addition, each plot reports the (model-implied) yield curves that would have been obtained if other monetary-policy regimes had prevailed. The seven circles (and diamonds) correspond respectively to the following maturities: 1 day, 1, 3, 6 months, 1, 2 and 4 years.

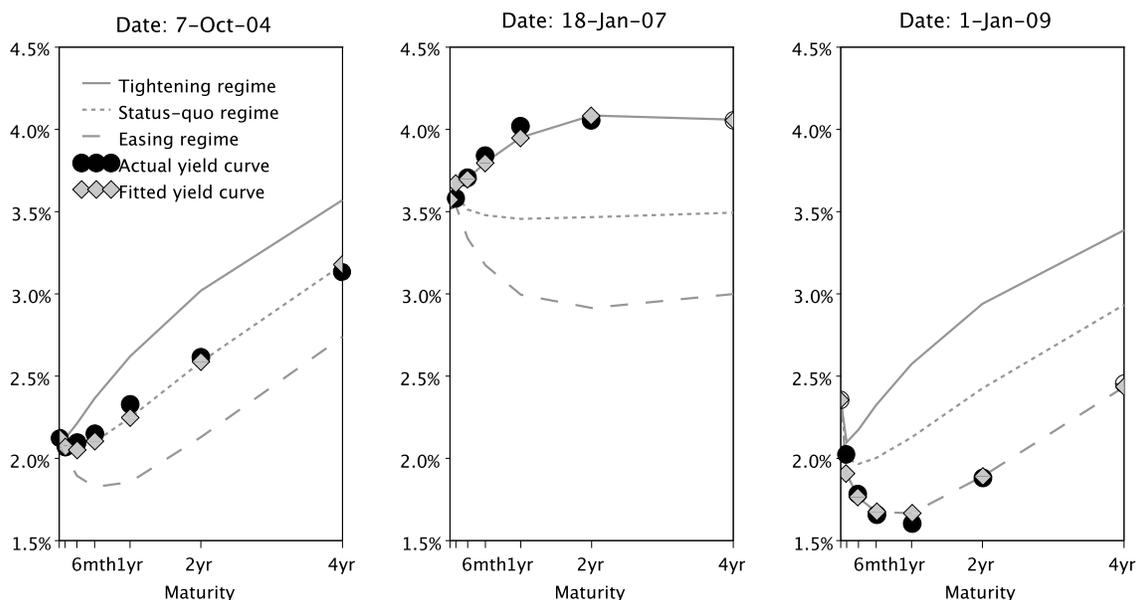
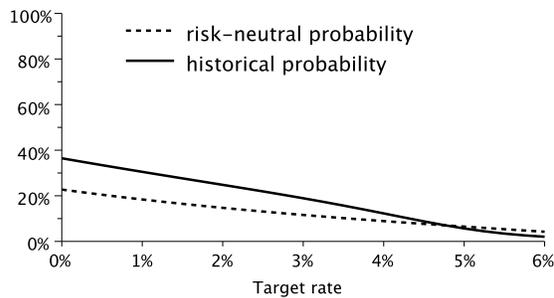


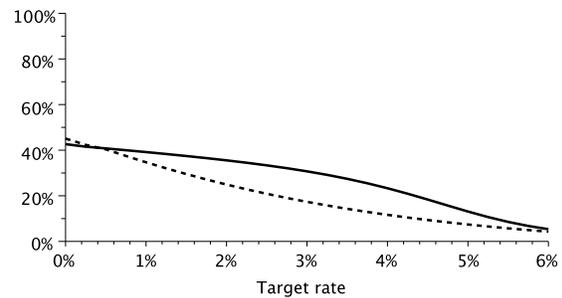
Fig. 6: Estimated probabilities of regime changes

Notes: These plots show the estimated probabilities of regime change over the next 30 days (period which includes only one monetary-policy meeting). As detailed in Subsection 3.1, these curves are based on some parametric forms of the target rate  $\bar{r}_t$ . Each plot displays the historical, or physical, probabilities as well as the risk-neutral ones. The upper four panels define the probabilities of monetary-policy-regime changes, the lower two show the target-change probabilities. Altogether, these probabilities define the matrices  $\Pi_t$  and  $\Pi_t^*$  describing respectively the dynamics of the Markov chain  $z_t$  (indicating the current target rate and the monetary-policy regime) under the physical and the risk-neutral measures.

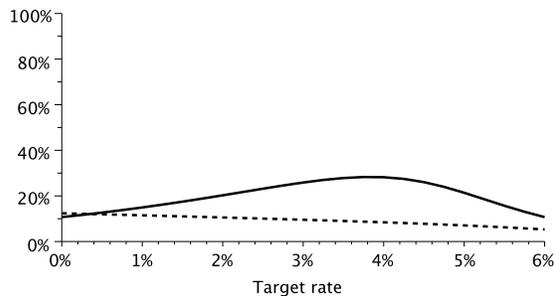
Panel A- Switch from Tightening to Status quo



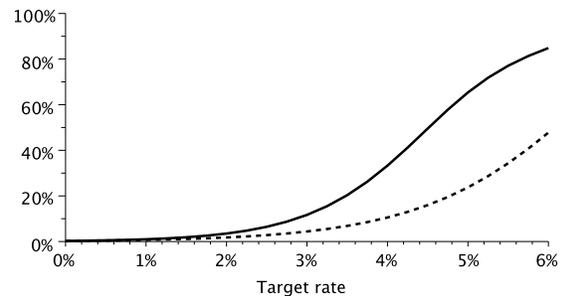
Panel B- Switch from Easing to Status quo



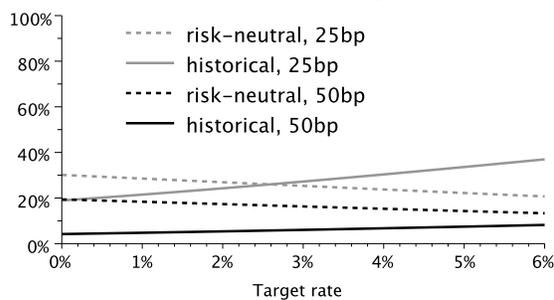
Panel C- Switch from Status quo to Tightening



Panel D- Switch from Status quo to Easing



Panel E- Rise in the target rate



Panel F- Cut in the target rate

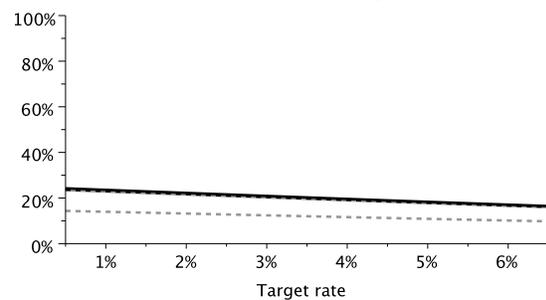




Fig. 8: Risk-neutral vs. physical policy-rate forecasts, and associated risk premia

Notes: These plots show the term structures of the forecasts of the policy rate under the physical (grey circles) and the risk-neutral (black circles) measures. Up to the Jensen inequality, these curves can be considered as forward rates of the policy rate (as regards the risk-neutral measure). The three columns of plots correspond to the current (period 0) monetary-policy regime (either tightening, status quo or easing). The three rows of plots correspond to different (current) policy rates (1%, 2.5% and 3.5%). Each of the 9 plots presenting the policy-rate forecasts is completed by a plot (placed below the first plot) of the corresponding risk premia, i.e. the spread between the two forecast curves (in basis points). 90% confidence intervals are reported. These confidence intervals are based on bootstrap techniques: the asymptotic distribution of the parameter estimates is used to draw 1000 alternative sets of parameter estimates that, in turn, are used to compute 1000 sets of alternative risk premia; the dashed lines correspond to the 5 and 95 percentiles of the obtained risk-premia distributions.

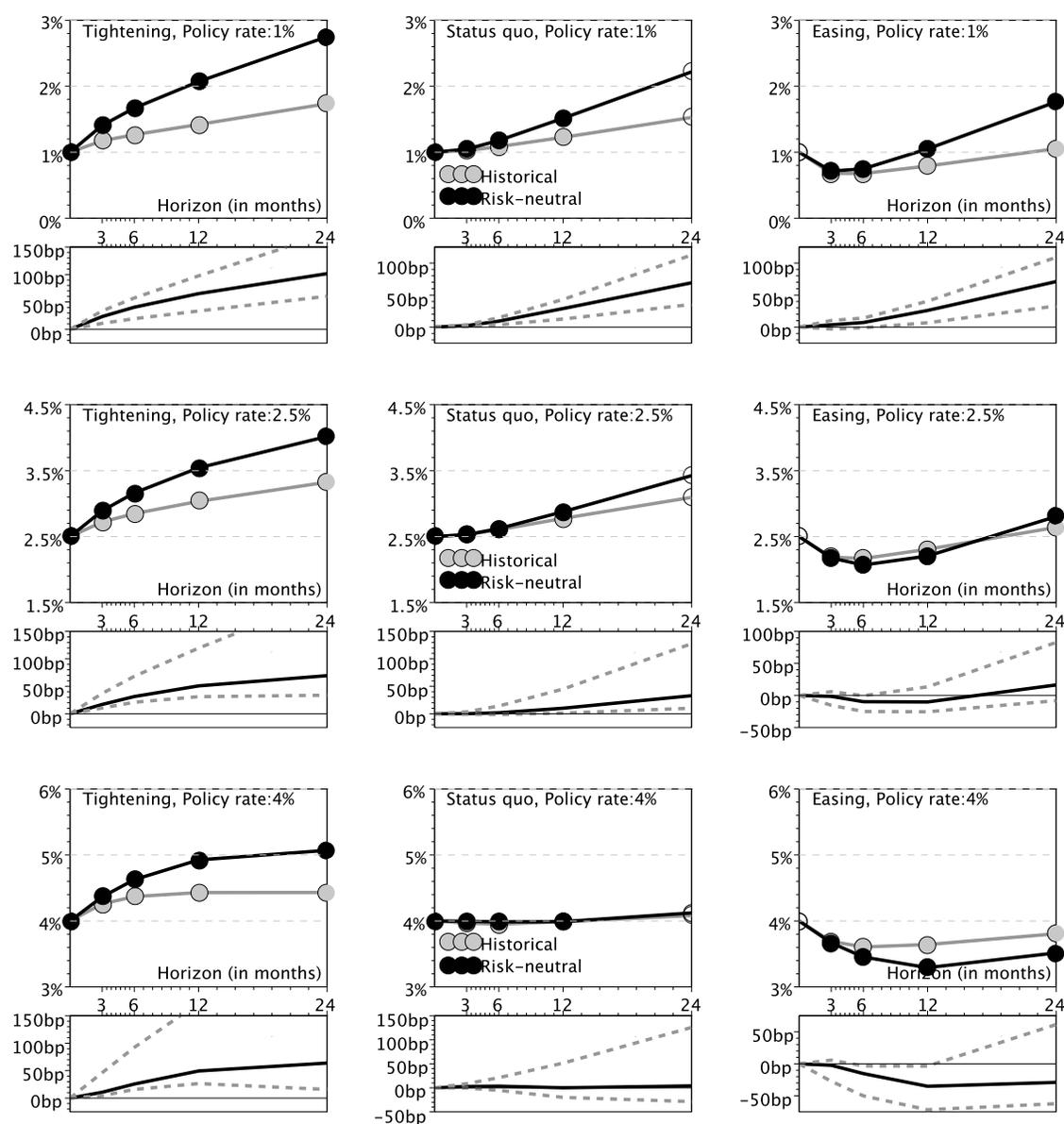


Fig. 9: Simulation of forward-guidance measures

Notes: These plots show the term-structure impact of a central bank's commitment to keep the target rate at its current level for 12 or 24 months. Two different policy rates are considered (1% and 3.5%). For each policy rate (1% or 3.5%) and each commitment durations (12 or 24 month), two plots are reported: the upper one displays yield curves with/without commitment of the central bank, the lower plot present the associated downward effect of the forward-guidance measure (that is the spread between the two curves plotted in the upper plot, in basis points). Note that here, I abstract from the effects of the excess-liquidity regime ( $w_t$ ) and  $s_t$  is set at 0, its unconditional level (the rationale behind this is that in my framework, these two latter factors are independent from the policy rate, which is the only factor affected by the measure). In the baseline as well as in the counterfactual case, the monetary-policy regime is set to the status-quo regime. Regarding the downward effect of the measure (lower plots of each pair of charts), 90% confidence intervals are reported. These confidence intervals are based on bootstrap techniques: the asymptotic distribution of the parameter estimates is used to draw 1000 alternative sets of parameter estimates that, in turn, are used to compute 1000 sets of alternative effects of the measure; the dashed lines correspond to the 5 and 95 percentiles of the obtained downward-effects distributions.

