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“What Moves the Yield Curve? Lessons From an Affine Term Structure Model for Chile”

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A mi esposa, Daniela

What Moves the Yield Curve? Lessons From an Affine Term Structure Model for Chile

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Abstract

This paper attempts to provide an economic interpretation of the factors that drive the movements of interest rates of bonds of different maturities in a continuous-time no-arbitrage term structure model. The dynamics of yields in the model are explained by two latent factors, the instantaneous short rate and its time-varying central tendency. The model estimates suggest that the short end of the yield curve is mainly driven by changes in first latent factor, while long-term interest rates are mainly explained by the second latent factor. Consequently, when thinking about movements in the term structure one should think of at least two forces that hit the economy; temporary shocks that change short-term and medium-term interest rates by much larger amounts than long-term interest rates, causing changes in the slope of the yield curve; and long-lived innovations which have persistent effects on the level of the yield curve.

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1 Introduction

Affine no-arbitrage dynamic term structure models are “the model[s] of choice in finance” (Diebold, Piazzesi, and Rudebusch 2005). These models employ a structure that consists of a small set of factors that characterize the yield curve, providing a parsimonious representation of the yield curve. At the same time they impose restrictions to ensure that yield dynamics are consistent in the cross-section and in the time series dimension (Dai and Singleton 2000, Piazzesi 2003). Even though these models provide a useful statistical description of the yield curve, they offer little insight of what the latent factors are, their relationship to macroeconomic variables, or the forces that drive its movements.

In this paper I attempt to provide an economic interpretation of the factors that drive the movements of interest rates of bonds of different maturities in a continuous-time no-arbitrage term structure model. Following Jegadeesh and Pennacchi (1996) and Balduzzi, Das, and Foresi (1998), the model I use in my analysis assumes the existence of a stochastic discount factor, which is governed by two latent factors with a proper economic interpretation. The first latent factor is the instantaneous short-term interest rate, and the second factor is the time-varying stochastic mean of the instantaneous short rate, capturing the notion that short rates display short-lived fluctuations around a time-varying rest level. This model generalizes the continuous-time term structure model presented in Vasicek (1977), but it is nevertheless solvable in closed form. Under this framework, bond yields are linear or affine functions of the latent factors, with restrictions on the cross-sectional and time-series properties of the yield curve that rule out arbitrage strategies.

While most of the related literature has focused on the analysis of the U.S. economy, little attention has been paid to emerging markets. I estimate the proposed model using monthly information on index-linked securities issued by the central bank and traded in the Chilean bond market from January 1990 through March 2006. Compared to developed economies, the Chilean fixed-income bond market is not underdeveloped, and Central Bank bonds are by far the most traded as well as the most liquid fixed-income securities (see Braun and Briones 2006). Nevertheless, the analysis of an emerging economy comes along with several estimation challenges. Most studies use zero-coupon yields that are estimated from prices of coupon bonds using one of the interpolation methods proposed by Fama and Bliss (1987) and McCulloch and Kwon (1993). Typical implementation of these methods requires

a rich set of bond prices with different maturities, which are unavailable in many emerging markets where there is a substantial number of ‘missing bond yields’. To overcome this problem, I rely on an extended version of the Kalman filter and maximum-likelihood to estimate the unobservable state variables and the model parameters using cross-sectional/time-series data of zero-coupon and coupon bond prices. Using this approach, I am able to estimate the zero-coupon term structure even for days where bonds with only few maturities are traded, while at the same time the Kalman filter allows the state variables to be handled correctly as unobservable variables.¹

The model estimates suggest that the instantaneous rate is more volatile and exhibits short-lived deviations from its time-varying mean, while the time-varying central tendency exhibits a weak reversion to its long-run value and a volatility less than half that of the short rate. According to the factor loadings implied by the estimated parameters, the very short end of the yield curve is driven by the first latent factor, but as the term to maturity increases the central tendency starts driving the movements in the yield curve, playing an important role at explaining the evolution of long-term interest rates. Consequently, a shock to the first latent factor increases short-term and medium-term interest rates by much larger amounts than the long-term interest rates, so that the yield curve becomes less steep (i.e., experiences a decrease in its slope). Thus, it is reasonable to interpret this short-lived shock as a temporary positive shock to the economy which increases temporarily production possibilities. On the other hand, a shock to the time-varying mean translates into an increase in short-term and long-term yields, changing the level of the yield curve. Therefore, a second source of shocks that moves the yield curve can be related to long-lived innovations which will induce a persistent change in the level of the yield curve, like a persistent (almost permanent) positive shock to productivity.

My work is more closely related to recent economics and finance papers exploring the macroeconomic determinants of the unobservable factors, and to macro-finance modeling which explicitly incorporates macroeconomic variables into multi-factor yield curve models. For instance, Wu (2001) and Piazzesi (2005) relate monetary policy shocks to temporary changes in the factor that influences the slope of the yield curve (slope factor), since mone-

¹Babbs and Nowan (1999) provide a generalization of the Kalman filter to estimate generalized Vasicek term structure models, and Cortázar, Naranjo, and Schwartz (2003) present an application of this methodology for a market with low-frequency transactions, namely Chile.

tary policy surprises affect short rates more than long ones. Ang and Piazzesi (2003) examine the influences of macroeconomic variables and latent factors that jointly determine the term structure of interest rates. They find that inflation and real activity have a significant impact on medium-term bond yields (up to a maturity of one year), but most of the movements in long-term yields are still accounted for by the unobservable factors. While most of these papers agree on the effects of macroeconomic variables on the slope of the yield curve, there are still conflicting results when explaining the movements of the level of the yield curve. Diebold, Rudebusch, and Aruoba (2006) obtain more favorable results using a Nelson-Siegel model combined with a VAR-model for the macroeconomy. They find that inflation is closely related to the factor that influences the slope of the yield curve, while the factor that changes the level of the yield curve (level factor) is highly correlated with real activity. Using different modeling strategies, Rudebusch and Wu (2004) and Dewatcher and Lucio (2006) interpret the slope factor as the cyclical response of the central bank to changes in the economy. While Rudebusch and Wu (2004) argue that the level factor reflects market participants' view about the inflation target of the central bank, Dewatcher and Lucio (2006) link this factor to long-run inflation effects.

The early literature investigating the term structure of interest rates in Chile has relied on the parametric model proposed by Nelson and Siegel (1987) (Herrera and Magendzo 1997, Zuñiga and Soria 1998 and Lefort and Walker 2000), however, this type of model does not acknowledge that bond yields need to be consistent giving rise to term structure estimates which allow arbitrage opportunities. The studies of Parisi (1998), Parisi (1999) and Zuñiga (1999b) go further and investigate the ability of one-factor no-arbitrage models in explaining the short-term interest rate, while Zuñiga (1999a) and Cortázar et al. (2003) provide estimates of multi-factor models. They find that no-arbitrage models outperform parametric specifications (Cortázar et al. 2003) and more simple autoregressive models (Parisi 1999). Fernández (2001), on the other hand, uses a semi-parametric specification of a one-factor no-arbitrage model to study the empirical properties of the term structure of interest rates. In contrast to the previous literature for Chile, I don't only present estimates of the term structure of interest rates, but also attempt to give an economic interpretation to the model estimates. Moreover, I characterize the way in which shocks to the factors behind the movements of the term structure change interest rates of different maturities, and provide a possible interpretation to this innovations.

The rest of the paper unfolds as follows. In Section 2 I present the continuous-time no-arbitrage term structure model. In Section 3 I present the state space formulation of the model and the estimation of the parameters by means of the Kalman filter. I also present an interpretation of the estimated parameters and discuss the reliability of the model-based yield curves obtained. In Section 4 I use impulse-response functions implied by the affine term structure model to discuss the sources of shocks that move the yield curve and their possible economic interpretation. Section 5 concludes.

2 A model of the term structure

Here I introduce the framework that I use for modeling the term structure of interest rates. First, I present several important relationships on asset pricing. Then, I present the solution to the no-arbitrage term structure model governed by two latent factors. Are two factors enough? Even though the work of Litterman and Scheinkman (1991) finds that almost all of movements in various US Treasury bond yields are captured by three unobservable factors, the study of Diebold et al. (2006) finds that two factors may suffice to capture the time-series variation in yields at a monthly frequency, since the first two principal factors account for almost 99% of the movements of yield-curve variation. The third factor is found unimportant since it is related to heteroskedasticity, and yields exhibit little heteroskedasticity at monthly frequency. The third factor is more important at daily and weekly frequencies (Ang, Piazzesi, and Wei 2006).

2.1 Background issues on asset pricing

Let $P(t, T)$ denote the price at time t of a zero-coupon bond maturing at time T . Assuming, without loss of generality, that the bond has a face value equal to one and its yield to maturity with continuous compounding is equal to $R(t, T)$, its price can be written as,

$$P(t, T) = \exp[-(T - t)R(t, T)] \tag{1}$$

and the yield to maturity on this zero-coupon bond is equal to,

$$R(t, T) = -\frac{\log P(t, T)}{T - t} \tag{2}$$

Using these results, one can define the instantaneous short rate as,

$$r_t = \lim_{T \rightarrow t} R(t, T) = -\lim_{T \rightarrow t} \frac{\log P(t, T)}{T - t} \quad (3)$$

Similarly, the price of a coupon bond paying τ coupons until the maturity date can be written as a portfolio of τ zero-coupon bonds,

$$\begin{aligned} P^c(t, T) &= \sum_{i=1}^{\tau} C_i P(t, t+i) \\ &= \sum_{i=1}^{\tau} C_i \exp[-iR(t, t+i)] \end{aligned} \quad (4)$$

where $\tau = T - t$ is also the time to maturity of the coupon bond (see Campbell, Lo, and MacKinlay 1997).

2.2 An affine term structure model

In an arbitrage free-environment and with a positive stochastic discount factor equal to Λ_t , the price at date t of a zero-coupon bond paying one unit of account at time T can be expressed as,

$$P(t, T) = \mathcal{E}_t \left(\frac{\Lambda_T}{\Lambda_t} \right) \quad (5)$$

where \mathcal{E}_t is the expectation taken conditional on time- t information. Following the standard asset-pricing theory, I assume that the stochastic discount factor is governed by the following process,

$$d\Lambda_t = -r_t \Lambda_t dt - \Lambda_t \lambda' d\mathbf{W}_t \quad (6)$$

where $\lambda = [\lambda_1, \lambda_2]'$ is a vector containing the market prices of risk which are assumed to be constant over time.

Following the work of Jegadeesh and Pennacchi (1996) and Balduzzi, Das, and Foresi (1998), I assume that the short rate r_t reverts toward a time-varying mean μ_t , whose dynamics is described by the following set of stochastic differential equations,

$$dr_t = \kappa_1(\mu_t - r_t)dt + \sigma_1 dW_{1t} \quad (7a)$$

$$d\mu_t = \kappa_2(\theta - \mu_t)dt + \sigma_2 dW_{2t} \quad (7b)$$

where κ_1 , κ_2 , σ_1 , and σ_2 are constants, and $\mathbf{W}_t = [W_{1t}, W_{2t}]'$ is a two-dimensional Brownian motion with a correlation coefficient equal to ρ . The coefficients κ_1 and κ_2 measure the speed of mean reversion of the two variables to their respective means, μ_t and θ , ρ measures the covariance between these two variables, while σ_1 and σ_2 are the volatilities of the short-term interest rate and the stochastic mean, respectively. This model generalizes the continuous-time term structure presented in Vasicek (1977) by letting the short-rate revert toward a time-varying stochastic mean, and captures the notion that short-term rates display short-lived fluctuations around a time-varying rest level, or central tendency.²

Under the above assumptions and using Itô's lemma, the price of a bond is characterized by the following partial differential equation,³

$$\begin{aligned} \frac{1}{2} \left(\frac{\partial^2 P}{\partial r^2} \sigma_1^2 + \frac{\partial^2 P}{\partial \mu^2} \sigma_2^2 \right) + \frac{\partial^2 P}{\partial r \partial \mu} \sigma_1 \sigma_2 \rho + \frac{\partial P}{\partial r} \left[\kappa_1 \left(\mu - r - \frac{\lambda_1 \sigma_1}{\kappa_1} \right) \right] \\ + \frac{\partial P}{\partial \mu} \left[\kappa_2 \left(\theta - \mu - \frac{\lambda_2 \sigma_2}{\kappa_2} \right) \right] + \frac{\partial P}{\partial t} - rP = 0 \end{aligned} \quad (8)$$

subject to the boundary condition $P(T, T) = 1$. As shown in Langetieg (1980) and Cochrane (2005), the solution for the fundamental valuation equation (8) is an exponentially linear function of the two latent variables, r_t and μ_t , taking the following exponential-affine form,

$$P(t, T) = \exp[A(\tau) + B_1(\tau)r_t + B_2(\tau)\mu_t] \quad (9)$$

where $\tau = T - t$ is the time to maturity of the bond and $A(\tau)$, $B_1(\tau)$ and $B_2(\tau)$ are functions of the maturity, the parameters of the model and satisfy the no-arbitrage condition in the bond market.

Equations (8) and (9) determine the solution for the functions $A(\tau)$, $B_1(\tau)$ and $B_2(\tau)$ in terms of a set of ordinary differential equations, which have a

² The Vasicek (1977) model is of the form,

$$dr_t = \kappa_1(\mu - r_t)dt + \sigma dW_t$$

where the instantaneous interest rate converges to a target level μ that is constant over time. Jegadeesh and Pennacchi (1996) and Balduzzi et al. (1998) show that a model with a time-varying target level outperforms the Vasicek model.

³The details of the model solution are spelled out in Appendix A.

solution equal to,

$$\begin{aligned}
B_1(\tau) &= \frac{\exp(-\kappa_1\tau) - 1}{\kappa_1} \\
B_2(\tau) &= \frac{\exp(-\kappa_2\tau) - 1}{\kappa_2} - \frac{\exp(-\kappa_1\tau) - \exp(-\kappa_2\tau)}{\kappa_1 - \kappa_2} \\
A(\tau) &= \int_0^\tau \left(\frac{1}{2}B_1^2\sigma_1^2 + B_2^2\sigma_2^2 + B_1B_2\sigma_1\sigma_2\rho - B_1\lambda_1\sigma_1 - B_2\lambda_2\sigma_2 + B_2\kappa_2\theta \right) ds
\end{aligned}$$

Using equation (2) and (9) we can write the yield of a bond maturing τ periods ahead as,

$$\begin{aligned}
R(t, T) &= -\frac{1}{\tau}[A(\tau) + B_1(\tau)r_t + B_2(\tau)\mu_t] \\
&= a(\tau) + b_1(\tau)r_t + b_2(\tau)\mu_t
\end{aligned} \tag{10}$$

where $a(\tau) = -A(\tau)/\tau$, $b_1(\tau) = -B_1(\tau)/\tau$ and $b_2(\tau) = -B_2(\tau)/\tau$. In this model, bond yields are linear functions of the state variables, r_t and μ_t . Therefore, it belongs to the family of affine term structure models, in which zero-coupon bond yields, their physical dynamics and their equivalent martingale dynamics are all affine (constant-plus-linear) functions of an underlying vector of state variables.⁴

Using equation (3), one can see that the instantaneous short rate is given by,

$$\lim_{\tau \rightarrow 0} R(t, t + \tau) = r_t \tag{11}$$

and the long-term rate is equal to the mean of the stochastic time varying long-term factor θ adjusted by risk premia,

$$\lim_{\tau \rightarrow \infty} R(t, t + \tau) = \theta - \frac{\lambda_2\sigma_2}{\kappa_2} - \frac{\lambda_1\sigma_1}{\kappa_1} - \frac{\sigma_1^2 + \sigma_2^2}{2\kappa_1^2} - \frac{\sigma_1\sigma_2\rho}{\kappa_2\kappa_1}$$

Notice that the time-varying mean of the first latent factor, μ_t , does not affect the short end of the yield curve, since its loading starts at zero for the instantaneous interest rate (i.e., $\lim_{\tau \rightarrow 0} B_2(\tau) = 0$). However, it does affect yields of longer maturities, influencing the long-end of the term structure.

⁴See Piazzesi (2003) for a review of affine models of the term structure.

3 Estimation

3.1 Data

In this paper I use three types of instruments to estimate the yield curve; pure discount bonds, which make a single fixed payment at the maturity date; coupon bonds which make coupon payments at equally spaced intervals until the maturity date, in which the face value is also paid and, coupon bonds paying both, the face value and the coupon rate, in each coupon payment.

I use end-of-month price quotes for index-linked instruments issued by the Central Bank of Chile, which have both their coupon and principal payments linked to the *Unidad de Fomento* (UF), from January 1990 through March 2006. The UF is a unit of account that varies according to past inflation, and is not perfectly correlated with current inflation. For short-maturity instruments, UF-linked yields cannot be considered real yields, but as the maturity of the instrument increases these yields approximate real interest rates more closely (see Chumacero 2002).

The database contains 4,472 observations on pure-discount bonds (Pagare Reajutable del Banco Central), and semi-annual amortizing coupon bonds (Pagare Reajutable con Cupones and Bonos del Banco Central de Chile). Over the sample period, each month contains an average of twenty seven observations. In early years, however, the number of index-linked bonds outstanding is very small, resulting on a median of 8 monthly observations over 1990 and 1991. During these first two years, the maturity of coupon bonds, as well, is confined to values that range from fifteen to twenty semesters and zero-coupon bonds which have maturities of less than one year. As we move forward in time, the maturity of traded coupon bonds diversifies ranging from 1 month to 40 semesters (see Figure 1).

3.2 State space specification

To estimate the affine term structure model presented in section 2, I first derive the discrete-time dynamics implied by the continuous-time model in order to match the observation frequency data. Then, using the discrete-time equivalent model I present the state space model formulation of the term structure model.

Note that the two stochastic differential equations (7a) and (7b) that

describe the state variables can be expressed as,

$$d \begin{bmatrix} r_t \\ \mu_t \end{bmatrix} = \left(\begin{bmatrix} -\kappa_1 & \kappa_1 \\ 0 & -\kappa_2 \end{bmatrix} \begin{bmatrix} r_t \\ \mu_t \end{bmatrix} + \begin{bmatrix} 0 \\ \kappa_2 \theta \end{bmatrix} \right) dt + \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \end{bmatrix} d\mathbf{W}_t$$

or more compactly,

$$d\mathbf{X}_t = (\mathbf{A}\mathbf{X}_t + \mathbf{b})dt + \mathbf{C}d\mathbf{B}_t \quad (12)$$

where \mathbf{B}_t is a standardized Brownian motion with a covariance matrix equal to $\mathbf{C}\mathbf{C}' = \sigma\rho\sigma'$. Using the methods presented in Langetieg (1980) and Bergstrom (1990), the equivalent discrete-time for this process is obtained from the solution to the system of stochastic differential equations (12). This solution implies the following equivalent discrete-time,

$$\mathbf{X}_{t_k} = \mathbf{\Phi}(\psi)\mathbf{X}_{t_{k-1}} + \mathbf{c}(\psi) + \varepsilon_{t_k} \quad (13)$$

where ε_{t_k} is normally distributed with variance-covariance matrix equal to $\mathbf{V}(\psi)$ and,

$$\begin{aligned} \mathbf{\Phi}(\psi) &= e^{\mathbf{A}(t_k - t_{k-1})} \\ \mathbf{c}(\psi) &= (e^{\mathbf{A}(t_k - t_{k-1})} - \mathbf{I})\mathbf{A}^{-1}\mathbf{b} \\ \mathbf{V}(\psi) &= \int_{t_{k-1}}^{t_k} e^{\mathbf{A}(t_k - s)}\mathbf{C}\mathbf{C}'e^{\mathbf{A}'(t_k - s)}ds \end{aligned}$$

The coefficients of this VAR(1) process, $\mathbf{\Phi}(\psi)$ and $\mathbf{c}(\psi)$, and the distribution of the innovation ε_{t_k} depend on the parameters describing the exponential-affine model.⁵

To complete the state space representation of the model, I present the measurement equation which relates the theoretical yields and the latent factors describing the model. At time t_k , the data is comprised by N_k bond prices of different maturity denoted by $\mathbf{P}_{t_k} = (P_{1k}, P_{2k}, \dots, P_{N_k})$ for $k = 1, \dots, T$. The set of instruments contains both, zero-coupon and coupon bonds with maturities that vary over time. I assume that there are discrepancies between observed prices and their theoretical counterparts explained by exogenous factors such as non-synchronous trading, rounding of prices, and bid-ask spreads. Therefore, in the presence of measurement errors we need

⁵A derivation of specific expressions for the matrices $\mathbf{\Phi}(\psi)$, $\mathbf{c}(\psi)$ and $\mathbf{V}(\psi)$ describing the transition equation is presented in Appendix A.

to distinguish between the theoretical term structure given by (9) and observed prices. Assuming that measurement errors are additive and normally distributed, theoretical and observed prices are related by,

$$\mathbf{P}_{t_k} = \Theta(\psi, \mathbf{X}_{t_k}) + \mathbf{v}_{t_k} \quad (14)$$

where $\mathbf{v}_{t_k} \sim \mathcal{N}(0, \mathbf{H}_k(\psi))$ is the measurement error, the vector ψ contains the unknown parameters of the model, \mathbf{X}_{t_k} is a vector containing the two state variables of the model (i.e., $\mathbf{X} = (r, \mu)'$), and the i -th row of $\Theta(\psi, \mathbf{X}_{t_k})$ is equal to,

$$\begin{aligned} \Theta_i(\psi, \mathbf{X}_{t_k}) &= \exp[A(\tau_i) + B_1(\tau_i)r_{t_k} + B_2(\tau_i)\mu_{t_k}] \\ &= \exp[A(\tau_i) + \mathbf{B}(\tau_i)' \mathbf{X}_{t_k}] \end{aligned}$$

when the i -th bond is a zero-coupon bond and equal to,

$$\Theta_i(\psi, \mathbf{X}_{t_k}) = \sum_{j=1}^{\tau_i} C_{ij} \exp[A(j) + \mathbf{B}(j)' \mathbf{X}_{t_k}]$$

when the i -th bond is a coupon bond paying τ_i coupons until the maturity.

The measurement equation (14) plus the transition system describing the state variables (13) form the non-linear state space model describing the term structure of interest rates.

3.3 Extended Kalman Filter

Even though the conventional Kalman filter cannot be used in the presence of non-linear measurement and/or transition equations, an approximate filter can be obtained by linearizing the non-linear measurement equation and then applying the extended Kalman filter presented in Harvey (1990). The original non-linear measurement equation (14) can be approximated using a Taylor expansion around the conditional mean of the state variables,

$$\mathbf{P}_{t_k} = \mathbf{d}(\psi, \mathbf{X}_{t_k|t_{k-1}}) + \mathbf{Z}(\psi, \mathbf{X}_{t_k|t_{k-1}}) \mathbf{X}_{t_k} + \mathbf{v}_k \quad (15)$$

where $\mathbf{X}_{t_k|t_{k-1}}$ is the conditional mean of the state variables given the information set available at t_k , $\mathbf{d}(\psi, \mathbf{X}_{t_k|t_{k-1}})$ is an N_k -vector with the i -th row given by

$$d_i = \Theta_i(\psi, \hat{\mathbf{X}}_{t_k|t_{k-1}}) - \frac{\partial \Theta_i(\psi, \hat{\mathbf{X}}_{t_k|t_{k-1}})}{\partial \mathbf{X}_{t_k}'} \hat{\mathbf{X}}_{t_k|t_{k-1}}$$

and $\mathbf{Z}(\psi, \mathbf{X}_{t_k|t_{k-1}})$ is an $N_k \times 2$ matrix with rows equal to

$$Z_i = \frac{\partial P(\hat{\mathbf{X}}_{t_k|t_{k-1}}, \tau)}{\partial \mathbf{X}'_{t_k}}.$$

Using the linear approximation of the measurement equation (15) and the transition equation (13), the parameters of the model can be estimated using the extended Kalman filter algorithm discussed in Harvey (1990). This algorithm consists of a sequence of two steps, a prediction and an update step. The prediction step yields the estimator of the state variables given by,

$$\hat{\mathbf{X}}_{t_k|t_{k-1}} = \mathcal{E}(\hat{\mathbf{X}}_{t_k} | \Xi_{t_{k-1}}) = \mathbf{c}(\psi) + \Phi(\psi) \hat{\mathbf{X}}_{t_{k-1}} \quad (16)$$

with a mean square error (MSE) equal to,

$$\begin{aligned} \mathbf{F}_{t_k|t_{k-1}} &= \mathcal{E} \left[(\mathbf{X}_{t_k} - \hat{\mathbf{X}}_{t_k|t_{k-1}})(\mathbf{X}_{t_k} - \hat{\mathbf{X}}_{t_k|t_{k-1}})' | \Xi_{t_{k-1}} \right] \\ &= \Phi \mathbf{F}_{t_{k-1}} \Phi' + \mathbf{V} \end{aligned}$$

where the expectation is based on the available information up to time t_{k-1} represented by $\Xi_{t_{k-1}}$.

In the update step, we use the additional information given by \mathbf{P}_{t_k} to obtain a more precise estimator of \mathbf{X}_{t_k} ,

$$\begin{aligned} \hat{\mathbf{X}}_{t_k} &= \mathcal{E}(\hat{\mathbf{X}}_{t_k} | \Xi_{t_k}) \\ &= \hat{\mathbf{X}}_{t_k|t_{k-1}} + \mathbf{F}_{t_k|t_{k-1}} \mathbf{Z}' (\mathbf{Z} \mathbf{F}_{t_k|t_{k-1}} \mathbf{Z}' + \mathbf{H})^{-1} \left[\mathbf{P}_{t_k} - \Theta(\psi, \hat{\mathbf{X}}_{t_k|t_{k-1}}) \right] \end{aligned} \quad (17)$$

with a MSE matrix equal to,

$$\begin{aligned} \mathbf{F}_{t_k} &= \mathcal{E} \left[(\mathbf{X}_{t_k} - \hat{\mathbf{X}}_{t_k})(\mathbf{X}_{t_k} - \hat{\mathbf{X}}_{t_k})' | \Xi_{t_k} \right] \\ &= \mathbf{F}_{t_k|t_{k-1}} - \mathbf{F}_{t_k|t_{k-1}} \mathbf{Z}' (\mathbf{Z} \mathbf{F}_{t_k|t_{k-1}} \mathbf{Z}' + \mathbf{H})^{-1} \mathbf{Z} \mathbf{F}_{t_k|t_{k-1}} \end{aligned}$$

Once we obtain estimates of the state variables (i.e., $\hat{\mathbf{X}}_{t_k}$) using information about the observed bond prices, we can evaluate the likelihood function using the prediction error decomposition (see Harvey (1990) for details). Then, the log-likelihood function is given by,

$$\sum_{k=1}^T \log f(\mathbf{P}_{t_k} | \psi) = - \sum_{t_k=1}^n \frac{N_k}{2} \log(2\pi) - \frac{1}{2} \sum_{k=1}^n \log |\Sigma_k| - \frac{1}{2} \sum_{k=1}^n \mathbf{v}'_{t_k} \Sigma_k^{-1} \mathbf{v}_{t_k} \quad (18)$$

where,

$$\begin{aligned}\mathbf{v}_{t_k} &= \mathbf{P}_{t_k} - \Theta(\psi, \hat{\mathbf{X}}_{t_k|t_{k-1}}) \\ \Sigma_k &= \mathbf{Z}\mathbf{F}_{t_k|t_{k-1}}\mathbf{Z}' + \mathbf{H}\end{aligned}$$

3.4 Estimation results

I estimate the term structure model using the extended Kalman filter algorithm and assuming that prices are observed with an error. I aggregate bonds into five categories to obtain a parsimonious covariance matrix of the measurement errors. Each group is assumed to have the same measurement error, therefore, the covariance matrix of the measurement errors $\mathbf{H}_k(\psi)$ is a diagonal matrix with five different elements characterizing the variance of each group, h_i . The first group contains discount bonds which have maturities less or equal to one year. The remaining four groups comprise coupon bonds with maturities up to five years, between six and ten years, between eleven and fifteen years, and above sixteen years, respectively.

Unlike the standard practice, I treat the factors r and μ as unobservables, and do not approximate the instantaneous rate using an observed short-term interest rate or the time-varying mean using observed yields as in Chan et al. (1992), Longstaff and Schwartz (1992), and Balduzzi et al. (1998). While approximating latent factors with observable yields is convenient, note that yields of any finite-maturity depend on both factors, r and μ , as well as on the model parameters we are trying to identify.

Table 1 presents the estimation results. With exception of the coefficient capturing the market price of risk of the central tendency, λ_2 , all parameter estimates are significant at conventional values. Both, the instantaneous interest rate and the time-varying central tendency present statistically significant reversion toward its central tendency. However, the mean reversion and the volatility of the instantaneous rate are considerably higher than the coefficients estimated for its time-varying mean. Therefore, the short-term rate is more volatile and returns faster to its time-varying mean, while the central tendency exhibits a weak reversion to its long-run value and a volatility less than a half that of the short-rate. The mean-reversion parameters imply a half-life of about 1 year for the short-rate, while the estimated half-life for the time-varying mean is 14.83 years (see Figure 3).

These results suggest that the time-varying mean can be interpreted as the level of interest rate that would prevail after ongoing temporary imbal-

ances in the economy –those that are expected to dissipate over the short-run– work themselves through. In contrast, fluctuations of the first latent state variable r_t around its time-varying mean reflect shorter-lived shocks. The model-based equilibrium real interest rate, which abstracts from both short-lived and long-lived shocks, is equal to $R_\infty = 2.47$ percent in semestral base or $R_\infty = 4.94$ percent in an annual base. The estimate of θ , the steady-state value of the short-rate and the central tendency, is statistically significant and equal to 2.95 percent in semestral base or 5.92 percent in annual terms.

The estimated loadings of the two factors driving the yield curve provides an insight on how each factor dynamics translates into movements of the yield curve. In order to give an interpretation of the estimated factor loadings, note that one can rewrite equation (10) as,

$$R(t, t + \tau) = a_1(\tau)R(\infty) + b_1(\tau)r_t + b_2(\tau)\mu_t$$

where $R(\infty) = \lim_{\tau \rightarrow \infty} R(t, t + \tau)$. Figure 2 depicts the loadings on the long-term interest rate $a_1(\tau)$, on the short-term interest rate $b_1(\tau)$, and on the time varying long-term rate $b_2(\tau)$ along different maturities, calculated using the estimates presented in Table 1. The very short end of the yield curve is driven by the first latent factor r_t . The loading associated to the short-term interest rate $b_1(\tau)$ starts at 1 and decays monotonically as the term to maturity increases reaching a value close to zero at long maturities. In contrast, as the term to maturity increases, the central tendency μ_t becomes a central factor behind the movements of the long end of the yield curve as well as intermediate maturities.

These characteristics of the model produce model-based yield curves that are capable to reproduce some important stylized facts. The higher volatility and lower persistence of the instantaneous short rate compared to that of the central tendency translates into a higher volatility at the short end than at the long end of the yield curve, plus more persistent long interest rates than short interest rates (see Table 2). The dynamics of the factors behind the term structure and the estimated parameters also produce yield curves with a variety of shapes over time, including downward sloping, upward sloping, and hump-shaped, and at the same time produces yields that rule out arbitrage possibilities between bonds of different maturities (see Figure 4). Even though the model imposes restrictions on the cross-sectional and time-series properties of the yield curve in order to rule out arbitrage possibilities, the

model fits quite well the observed yield-to-maturity of zero and coupon bonds (see Figure 5).

The model estimates also produce an average yield curve that is increasing and concave (see Figure 6). To understand why, note that the market price of risk of the short-term interest rate imply the following risk premia for a discount bond maturing in τ periods,⁶

$$\lambda_1 \frac{\sigma_1}{P} \frac{\partial P}{\partial r} = -\lambda_1 \sigma_1 \left(\frac{1 - \exp(-\kappa_1 \tau)}{\kappa_1} \right)$$

The sign of the risk premium es equal to minus the sign of the respective market price of risk, and the magnitude of the premium is an increasing function of bond maturity. The estimate of λ_1 , the market price of risk of the short-term interest rate, is negative and statistically significant. This implies that a bond's interest rate risk premium is positive and increasing with bond's maturity, suggesting that the yield curve is usually upward sloping.

Finally, the correlation between changes in the short rate and its time-varying mean ρ is negative and statistically significant. The correlation might be interpreted as a link between agents expectation of future economic conditions and changes in the short-rate. The intuition is straightforward. Suppose the economy is in a growth stage, and the monetary authority increases the short-term interest rate in an effort to avoid an overheating of the economy. Then, if movements in the short rate are pro-cyclical and agents believed that a hike in the short rate signals future adverse economic conditions, there is an incentive to sacrifice today's consumption to buy long-term bonds that pays off in the bad times. This increase in the demand for long-term bonds will bid up their price and lower long-term yields, resulting in a negative correlation between r and μ . The model also implies that a downward sloping yield curve not only indicates good times today, but bad times tomorrow. Therefore, when agents expect a recession, short-term rates will increase, while long rates will decrease.⁷

⁶See Pennacchi (1991) and Jegadeesh and Pennacchi (1996) for a derivation.

⁷See Harvey (1988), Estrella and Hardouvelis (1991), Plosser and Rouwenhorst (1994), Kamara (1997), Chapman (1997), Estrella and Mishkin (1998), Hamilton and Kim (2002), Berardi and Torous (2005) and Ang, Piazzesi, and Wei (2006) for a more detailed explanation of the information content of the shape of the term structure about economic activity.

4 Movements in the yield curve

The yield curve might move due to changes in announcements of unemployment or inflation, changes in market participants' risk aversion aroused from perceived changes in the prospects for continued economic growth, or due to changes in other economic variables (Bliss 1997). In the model presented above changes in the instantaneous interest rate and the its time-varying central tendency drive changes in interest rates of different maturities to varying degrees. Therefore, it is reasonable to think that these latent factors should capture the economic factors influencing interest rates and the changes in the underlying determinants of the term structure of interest rates. Here, I discuss how the term structure of interest rates changes in response to new information about r_t and μ_t using impulse-response functions implied by the affine term structure model. Additionally, I attempt to provide an economic interpretation of these shocks.⁸

Let me start by analyzing the effect of one standard innovation of the instantaneous interest rate and how it translates into movements of the yield curve. The first two panels of Figure 7 exhibit the response of the instantaneous short-rate r_t and its time-varying mean μ_t to an innovation in the first latent factor. Figure 8 presents the impulse-response functions of selected yields, the term structure at its initial value, and one month and 5 years after a one standard deviation of r_t . The results show that the instantaneous rate r_t increases immediately after the shock and decays monotonically returning to its initial value. On the other hand, the central tendency μ decreases slightly due to the negative correlation between the two state variables, but the confidence intervals indicate that this response is not statistically significant. As the loadings of each factor suggest, the shock to the instantaneous interest rate increases short-term and medium-term interest rates by much larger amounts than long-term interest rates. Consequently, the yield curve initially becomes less steep, presenting a decrease in its slope. The yield curve returns back to its initial position five years after the shock initiated (see Figure 8).

The movements in yields in response to this shock have an intuitive explanation. To better understand the following discussion, recall that the basic asset pricing model predicts that the price of a discount bond maturing in τ

⁸Appendix B contains the analytical derivation of the model-based impulse-response functions and their respective standard errors.

periods is given by,

$$P(t, T) = \beta^\tau \mathcal{E}_t \left[\frac{U'(c_{t+\tau})}{U'(c_t)} \right] \quad (19)$$

where $U'(c_t)$ is the marginal utility of consumption, $0 < \beta < 1$ denotes the subjective discount factor, and \mathcal{E}_t is the expectation conditional on information available at time t . Under this framework, interest rates reflect the rate at which people are willing to trade consumption today for consumption tomorrow (Altug and Labadie 1994).

A shock that temporarily increases short-term and medium-term yields can be interpreted as a temporary positive shock to the economy which increases production possibilities. Intuitively, after the realization of this shock agents will face an increase in their consumption, but considering that this gain will eventually die out, economic agents will save part of the output and invest into capital in order to smooth their consumption (den Haan 1995). Therefore, the expected growth rate of consumption is expected to be positive in the short-term and medium-term, implying an initial increase in the slope of the yield curve. However, as agents start to dissave, interest rates will fall back to their long-term level and consumption growth will return to its steady-state level. This result is consistent with the results of Rendu de Lint and Stolin (2003) who find that a temporary productivity shock in a simple production stochastic growth model increases the one-period interest rate more than the τ -period interest rate, increasing the slope of term structure of interest rates.

The last two panels of Figure 7 exhibit the response of the two latent factors to an innovation to the central tendency, while Figure 9 depicts the impulse-response functions of the six-month, 5-year and 20-year yields as well as the term structure of interest rates at its initial level, and one month and 5 years after the shock to μ_t . An innovation to the central tendency increases immediately the time-varying mean μ_t , which reduces monotonically and slowly as one would expect given the estimated high persistence of this factor. The instantaneous interest rate r_t increases quickly to catch up the new level the central tendency reached after the shock. After reaching a value near the central tendency, the instantaneous rate decreases slowly following the path of the central tendency.

This shock translates into an initial increase in medium-term and long-term yields, while short-term interest rates initially remain muted. As the instantaneous rate increases to catch up its time-varying mean, short-term

interest rates start to increase, while long-term interest rates start to slowly fall as the long-run time varying mean falls back to its equilibrium level. As a consequence, initially the slope of the yield curve increases, but rapidly the term structure of interest rates exhibits a change in its level. Five years after the innovation, yields of all maturities change by almost identical amounts (see Figure 9). Slowly, the term structure of interest rates will move toward its initial position as the effect of the shock dies off. However, the high persistence of the time-varying mean makes this effect look as permanent, despite the fact that this variable is stationary.

In this case, this shock can be interpreted as a persistent (almost permanent) positive shock to productivity. To understand why, suppose that the economy faces a shock that has no initial impact, but eventually grows to a permanent technology shock, whose path is perfectly anticipated by agents. A positive innovation permanently increases the level of expected future consumption, thereby the high expected levels of future consumption makes long-term bonds less attractive driving prices down and driving yields up.⁹ As the impact of the innovation materializes, agents will require a higher return as an inducement to save, leading to an increase in yields of all maturities.

5 Final remarks

I show that when thinking about movements in the term structure, one should think in changes in at least two type of forces that hit the economy. First, shocks that are short-lived, which change the slope of the yield curve. Second, long-lived shocks that influence yields of all maturities and shift the level of the term structure of interest rates.

However, two questions remain unanswered. Is it reasonable to relate the effects of the short-lived shocks to the effects of inflationary pressures or monetary policy surprises as in Piazzesi (2005) or Wu (2001)? Furthermore, can the long-lived innovations be related to changes in household consumption preferences, or perceptions about future economic prospects?. A second unanswered question is whether one can improve the understanding of the factors that lie behind the movements of the term structure by including macroeconomic variables explicitly to the model presented in the paper.

⁹Again, this results is consistent with the equilibrium condition (19)

The findings from the estimated model might also be useful for building an equilibrium model of the economy. Labadie (1994) shows that the assumption about the persistence of the shocks is very important when evaluating the asset pricing implications of the representative agent framework. She argues that the dichotomous results that arise about the behavior of the yield curve depending on whether endowment shocks are temporary or permanent, are the natural outcome of assuming that there is a single shock to the economy. As previously argued by Christiano and Eichenbaum (1990), the results obtained here suggest that instead of using one type of shock over the other, the best strategy is probably to cast a model with both types of shocks.

A Model solution and state space representation

A.1 Model solution

As shown in Cochrane (2005), the partial equation that characterizes the price of a discount bond of maturity τ is given by,

$$\begin{aligned} & \frac{1}{2} \left(\frac{\partial^2 P}{\partial r^2} \sigma_1^2 + \frac{\partial^2 P}{\partial \mu^2} \sigma_2^2 \right) + \frac{\partial^2 P}{\partial r \partial \mu} \sigma_1 \sigma_2 \rho + \frac{\partial P}{\partial r} \left[\kappa_1 \left(\mu - r - \frac{\lambda_1 \sigma_1}{\kappa_1} \right) \right] \\ & + \frac{\partial P}{\partial \mu} \left[\kappa_2 \left(\theta - \mu - \frac{\lambda_2 \sigma_2}{\kappa_2} \right) \right] + \frac{\partial P}{\partial t} - rP = 0 \end{aligned} \quad (20)$$

I assume that there is a solution for this fundamental valuation equation that is represented by,

$$P(t, T) = \exp [A(\tau) + B_1(\tau) r_t + B_2(\tau) \mu_t] \quad (21)$$

Obtaining the partial derivatives of the solution (21) and replacing them back into the partial differential equation that characterizes the price of a discount bond (20) one obtains,

$$\begin{aligned} & \frac{1}{2} B_1^2(\tau) \sigma_1^2 + \frac{1}{2} B_2^2(\tau) \sigma_2^2 + B_1(\tau) B_2(\tau) \sigma_1 \sigma_2 \rho \\ & + B_1(\tau) (\kappa_1 (\mu_t - r_t) - \lambda_1 \sigma_1) + B_2(\tau) (\kappa_2 (\theta - \mu_t) - \lambda_2 \sigma_2) \\ & - \left[\frac{dA(\tau)}{d\tau} + \frac{dB_1(\tau)}{d\tau} r_t + \frac{dB_2(\tau)}{d\tau} \mu_t \right] - r_t = 0 \end{aligned} \quad (22)$$

Collecting terms and knowing that (22) must hold for all r_t and μ_t we obtain the following system of ordinary differential equations,

$$\begin{aligned} 0 &= \frac{1}{2} B_1^2(\tau) \sigma_1^2 + \frac{1}{2} B_2^2(\tau) \sigma_2^2 + B_1(\tau) B_2(\tau) \sigma_1 \sigma_2 \rho - B_1 \lambda_1 \sigma_1 - B_2 \lambda_2 \sigma_2 + B_2 \kappa_2 \theta - \frac{dA(\tau)}{d\tau} \\ 0 &= -B_1(\tau) \kappa_1 - \frac{dB_1(\tau)}{d\tau} - 1 \\ 0 &= B_1(\tau) \kappa_1 - B_2(\tau) \kappa_2 - \frac{dB_2(\tau)}{d\tau} \end{aligned}$$

With boundary initial conditions, $\mathbf{B}(0) = \mathbf{0}$ and $A(0) = 0$, the solution to this system of ordinary differential equations is equal to,

$$\begin{aligned}
B_1(\tau) &= \frac{\exp(-\kappa_1\tau) - 1}{\kappa_1} \\
B_2(\tau) &= \frac{\exp(-\kappa_2\tau) - 1}{\kappa_2} - \frac{\exp(-\kappa_1\tau) - \exp(-\kappa_2\tau)}{\kappa_1 - \kappa_2} \\
A(\tau) &= \int_0^\tau \left(\frac{1}{2}B_1^2\sigma_1^2 + B_2^2\sigma_2^2 + B_1B_2\sigma_1\sigma_2\rho - B_1\lambda_1\sigma_1 - B_2\lambda_2\sigma_2 + B_2\kappa_2\theta \right) ds
\end{aligned}$$

which are the equations presented in the text.

A.2 State space representation

The equivalent discrete-time for the process (12) is given by,

$$\mathbf{X}_{t_k} = \mathbf{\Phi}(\psi)\mathbf{X}_{t_{k-1}} + \mathbf{c}(\psi) + \varepsilon_{t_k} \quad (23)$$

where ε_{t_k} is normally distributed with variance-covariance matrix equal to $\mathbf{V}(\psi)$ and,

$$\begin{aligned}
\mathbf{\Phi}(\psi) &= e^{\mathbf{A}(t_k - t_{k-1})} \\
&= \begin{bmatrix} \exp(-\kappa_1\Delta t) & \frac{\kappa_1}{\kappa_2 - \kappa_1} (\exp(-\kappa_1\Delta t) - \exp(-\kappa_2\Delta t)) \\ 0 & \exp(-\kappa_2\Delta t) \end{bmatrix} \\
\mathbf{c}(\psi) &= (e^{\mathbf{A}(t_k - t_{k-1})} - \mathbf{I})\mathbf{A}^{-1}\mathbf{b} \\
&= \begin{bmatrix} \theta \left(1 - \frac{\kappa_2}{\kappa_2 - \kappa_1} \exp(-\kappa_1\Delta t) + \frac{\kappa_1}{\kappa_2 - \kappa_1} \exp(-\kappa_2\Delta t) \right) \\ \theta (1 - \exp(-\kappa_2\Delta t)) \end{bmatrix} \\
\mathbf{V}(\psi) &= \int_{t_{k-1}}^{t_k} e^{\mathbf{A}(t_k - s)} \mathbf{C}\mathbf{C}' e^{\mathbf{A}'(t_k - s)} ds
\end{aligned}$$

B Model-based impulse-response functions and their standard errors

Using the discrete-time process describing r_t and μ_t given by (13), one can obtain the following VAR(1) model,¹⁰

$$(\mathbf{X}_{t_k} - \bar{\mathbf{X}}) = \Phi(\psi)(\mathbf{X}_{t_{k-1}} - \bar{\mathbf{X}}) + \varepsilon_t$$

where, $\bar{\mathbf{X}} = [\mathbf{I} - \Phi(\psi)]^{-1}\mathbf{c}$, and $\varepsilon_t \sim \mathcal{N}(0, \mathbf{V}(\psi))$.

This model can be written in vector MA(∞) form as,

$$\mathbf{X}_{t_k} = \bar{\mathbf{X}} + \varepsilon_{t_k} + \Gamma_1 \varepsilon_{t_{k-1}} + \Gamma_2 \varepsilon_{t_{k-2}} + \dots$$

where $\Gamma_1 = \Phi$, $\Gamma_2 = \Gamma_1 * \Phi$, and in general $\Gamma_s = \Gamma_{s-1} * \Phi$.

The consequence for $\mathbf{X}_{t_{k+s}}$ of new information about r_t beyond that contained in $\mathbf{X}_{t_{k-1}}$ is given by,

$$\frac{\partial E(\mathbf{X}_{t_{k+s}} | r_{t_k}, \mathbf{X}_{t_{k-1}})}{\partial r_{t_k}}$$

To calculate this magnitude note that one can write the variance-covariance matrix of ε_t as the product of a lower triangular matrix with ones along the principal diagonal, and a diagonal matrix with positive entries along the principal diagonal,

$$\mathbf{V} = \mathbf{A}\mathbf{D}\mathbf{A}'$$

with,

$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ V_{21}V_{11}^{-1} & 1 \end{bmatrix}, \quad \mathbf{D} = \begin{bmatrix} V_{11} & 0 \\ 0 & V_{22} - V_{21}V_{11}^{-1}V_{12} \end{bmatrix}$$

The orthogonalized impulse response function is given by,

$$\begin{aligned} \hat{\mathbf{h}}_{1,s} &= \frac{\partial E(\mathbf{X}_{t_{k+s}} | r_{t_k}, \mathbf{X}_{t_{k-1}})}{\partial r_{t_k}} = \Gamma_s \mathbf{A}_1 \\ \hat{\mathbf{h}}_{2,s} &= \frac{\partial E(\mathbf{X}_{t_{k+s}} | \mu_{t_k}, \mathbf{X}_{t_{k-1}})}{\partial \mu_{t_k}} = \Gamma_s \mathbf{A}_2 \end{aligned}$$

where \mathbf{A}_j is the j -th column of matrix \mathbf{A} . The Cholesky decomposition of the matrix the variance-covariance matrix of ε_t is given by $\mathbf{V} = \mathbf{P}\mathbf{P}'$. Using this expression the impulse-response function is given by,

¹⁰The results here are obtained using the methods described in Hamilton (1994).

$$\begin{aligned}\widehat{\mathbf{h}}_{1,s} &= \frac{\partial E(\mathbf{X}_{t_{k+s}}|r_{t_k}, \mathbf{X}_{t_{k-1}})}{\partial r_{t_k}} = \Gamma_s \mathbf{P}_1 = \Gamma_s \mathbf{A}_1 \sqrt{d_{11}} \\ \widehat{\mathbf{h}}_{2,s} &= \frac{\partial E(\mathbf{X}_{t_{k+s}}|\mu_{t_k}, \mathbf{X}_{t_{k-1}})}{\partial \mu_{t_k}} = \Gamma_s \mathbf{P}_2 = \Gamma_s \mathbf{A}_2 \sqrt{d_{22}}\end{aligned}$$

Since yields are an affine function of the vector of latent factors, the impulse response functions for a yield of maturity τ is given by,

$$\begin{aligned}\widehat{z}_{1,s} &= \frac{\partial E(R(t_{k+s}, T)|r_{t_k}, \mathbf{X}_{t_{k-1}})}{\partial r_{t_k}} = (-1/\tau) \mathbf{B}' \frac{\partial E(\mathbf{X}_{t_{k+s}}|r_{t_k}, \mathbf{X}_{t_{k-1}})}{\partial r_{t_k}} \\ \widehat{z}_{2,s} &= \frac{\partial E(R(t_{k+s}, T)|\mu_{t_k}, \mathbf{X}_{t_{k-1}})}{\partial \mu_{t_k}} = (-1/\tau) \mathbf{B}' \frac{\partial E(\mathbf{X}_{t_{k+s}}|\mu_{t_k}, \mathbf{X}_{t_{k-1}})}{\partial \mu_{t_k}}\end{aligned}$$

The impulse-response functions are a nonlinear function of the parameters of the model ψ , therefore the standard errors for $\mathbf{h}_{j,s}$ and $z_{j,s}$ can be calculated using the delta expansion of the asymptotic distribution of ψ , obtaining,

$$\begin{aligned}\sqrt{T}(\widehat{\mathbf{h}}_{j,s} - \mathbf{h}_{j,s}) &\rightarrow \mathbf{N}\left(\mathbf{0}, \left. \frac{\partial \mathbf{h}_{j,s}}{\partial \psi'} \right|_{\psi=\widehat{\psi}} \mathcal{J}_{\mathbf{T}} \left. \frac{\partial \mathbf{h}_{j,s}}{\partial \psi'} \right|'_{\psi=\widehat{\psi}}\right) \\ \sqrt{T}(\widehat{\mathbf{z}}_{j,s} - \mathbf{z}_{j,s}) &\rightarrow \mathbf{N}\left(\mathbf{0}, \left. \frac{\partial z_{j,s}}{\partial \psi'} \right|_{\psi=\widehat{\psi}} \mathcal{J}_{\mathbf{T}} \left. \frac{\partial z_{j,s}}{\partial \psi'} \right|'_{\psi=\widehat{\psi}}\right)\end{aligned}$$

To calculate this derivatives recall that, $\Gamma_s = \Gamma_{s-1} * \Phi$ then the derivative of the non-orthogonalized impulse-response with respect to the scalar ψ_i denoting some particular element of ψ is equal to,

$$\begin{aligned}\frac{\partial \Gamma_s}{\partial \psi_i} &= \frac{\partial \Gamma_{s-1} \Phi}{\partial \psi_i} \\ &= \frac{\partial \Gamma_{s-1}}{\partial \psi_i} \Phi + \Gamma_{s-1} \frac{\partial \Phi}{\partial \psi_i}\end{aligned}$$

Then,

$$\frac{\partial \widehat{\mathbf{h}}_{j,s}}{\partial \psi_i} = \frac{\partial \Gamma_s}{\partial \psi_i} \mathbf{A}_j + \Gamma_s \frac{\partial \mathbf{A}_j}{\partial \psi_i}$$

In the case of the Cholesky decomposition one obtains,

$$\frac{\partial \widehat{\mathbf{h}}_{j,s}}{\partial \psi_i} = \frac{\partial \Gamma_s}{\partial \psi_i} \mathbf{A}_j \sqrt{d_{jj}} + \Gamma_s \frac{\partial \mathbf{A}_j}{\partial \psi_i} \sqrt{d_{jj}} + \Gamma_s \mathbf{A}_j \frac{1}{2\sqrt{d_{jj}}} \frac{\partial d_{jj}}{\partial \psi_i}$$

Similarly, for the impulse-response functions of the yields the partial derivative is equal to,

$$\frac{\partial z_{j,s}}{\partial \psi_i} = (-1/\tau) \mathbf{B}' \frac{\partial \hat{\mathbf{h}}_{j,s}}{\partial \psi_i} + (-1/\tau) \frac{\partial \mathbf{B}'}{\partial \psi_i} \hat{\mathbf{h}}_{j,s}$$

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Table 1: Maximum Likelihood Estimates (1990:01 - 2006:03)

Parameter	Estimate	Parameter	Estimate
κ_1	0.2959 (0.0195)	h_1	0.01645 (0.0040)
σ_1	0.0090 (0.0006)	h_2	0.5808 (0.0298)
λ_1	-0.2686 (0.1566)	h_3	0.7498 (0.0254)
κ_2	0.0265 (0.0031)	h_4	0.0793 (0.0064)
σ_2	0.0040 (0.0002)	h_5	1.9466 (0.1280)
λ_2	0.0140 (0.0232)		
ρ	-0.1480 (0.0856)		
θ	0.0295 (0.0058)		
Half-life r	1.0046 years	$R(\infty)$	0.0247
Half-life μ	14.8333 years		

Note: The table reports maximum likelihood estimates of an affine term structure model described by,

$$dr_t = \kappa_1(\mu_t - r_t)dt + \sigma_1 dW_{1t}$$

$$d\mu_t = \kappa_2(\theta - \mu_t)dt + \sigma_2 dW_{2t}$$

using monthly data on zero and coupon bonds for the 1990:01-2005:03 period. The estimates of the measurement equation error covariance matrix h_i are multiplied by 10^2 . The time unit t is expressed in semesters, therefore the model-based yields are in semestral base.

Table 2: Yield curve estimates descriptive statistics (1990:01-2006:03)

Maturity	Mean	Std. dev.	Minimum	Maximum	V_{t+12}
1 month	4.863	3.026	-2.594	12.159	1.797
6 months	4.921	2.882	-2.069	11.794	1.796
1 year	5.021	2.637	-1.157	11.161	1.792
2 years	5.177	2.275	0.2367	10.193	1.812
3 years	5.293	2.030	1.227	9.512	1.839
4 years	5.382	1.858	1.740	9.349	1.862
5 years	5.453	1.731	1.838	9.346	1.887
6 years	5.509	1.633	1.937	9.325	1.904
7 years	5.556	1.555	2.034	9.291	1.916
8 years	5.594	1.490	2.118	9.249	1.930
9 years	5.625	1.434	2.203	9.200	1.937
10 years	5.651	1.384	2.286	9.147	1.949
15 years	5.723	1.193	2.666	8.848	1.968
20 years	5.738	1.053	2.977	8.539	1.979

Note: Descriptive statistics for model-based monthly yields at different maturities in annual base. The last column presents the variance ratio defined as, $V_{t+12} = \frac{1}{k} \frac{var(y_{t+12}-y_t)}{var(y_{t+1}-y_t)}$.

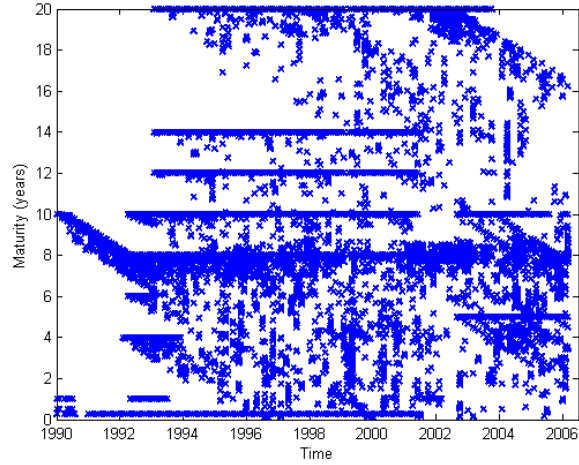


Figure 1: End-of-month available discount and coupon bonds over the sample period. A point represents a bond available for the corresponding maturity and time period.

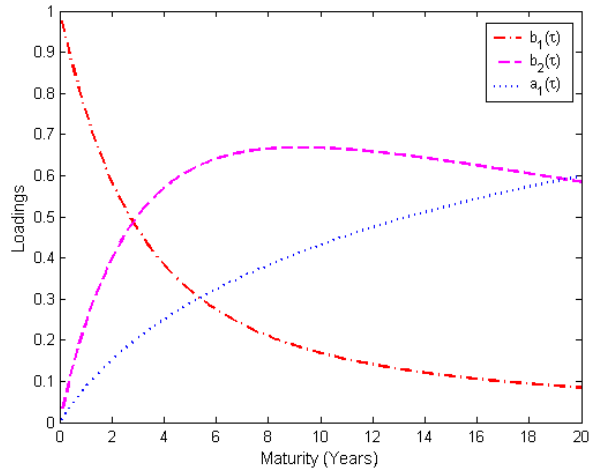


Figure 2: Estimated factor loadings in the affine term structure model, $R(t, t + \tau) = a_1(\tau)R(\infty) + b_1(\tau)r_t + b_2(\tau)\mu_t$

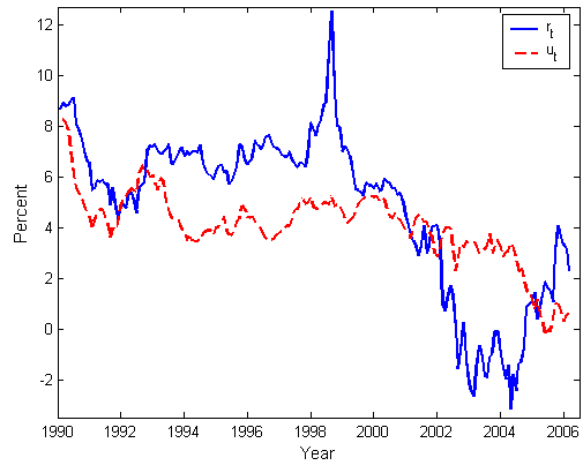


Figure 3: Estimated latent factors r_t and μ_t in annual base over 1990:01-2006:03

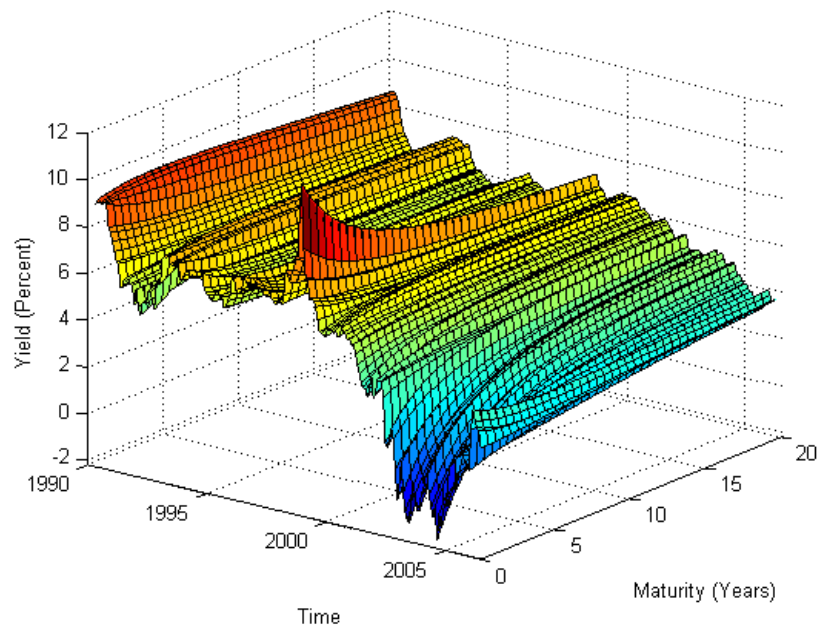
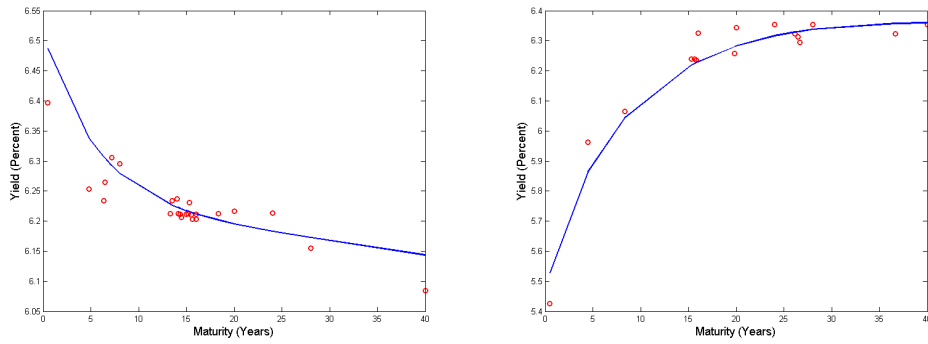


Figure 4: Yield curves in annual base estimated using the estimates of an affine term structure model for the 1990:01-2006:03 period



(a) Yield-to-maturity on September, 1993 (b) Yield-to-maturity on March, 2002

Figure 5: Model-based yield to maturity (continuous line) and the observed yields of traded instruments (circle points) in annual base

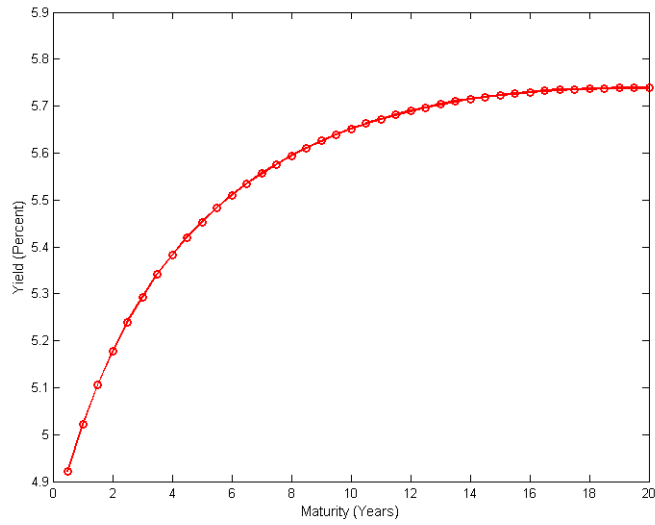
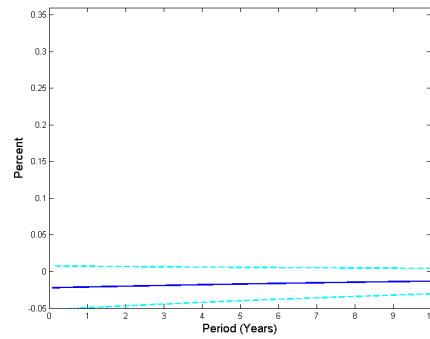
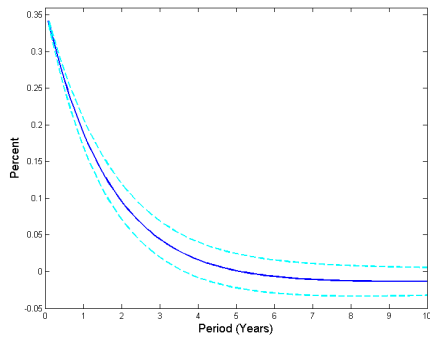
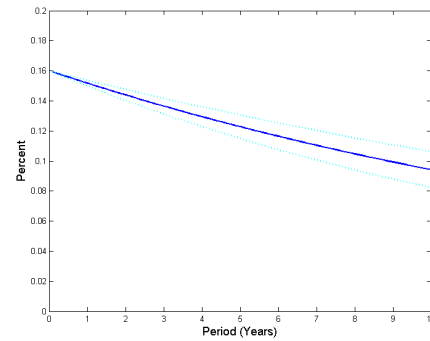
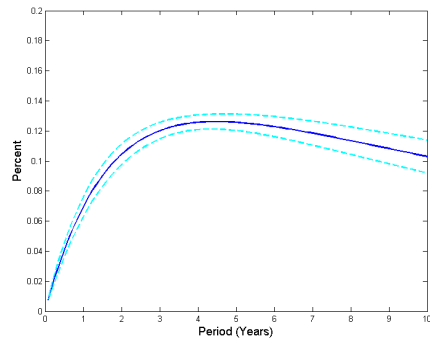


Figure 6: Model-based average yield curve in annual base for the 1990:01-2006:03 period

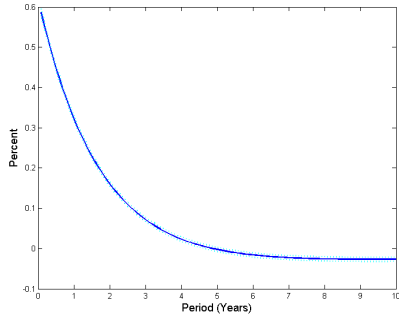


(a) Response of r_t to a one std. dev. of r_t (b) Response of μ_t to a one std. dev. of r_t

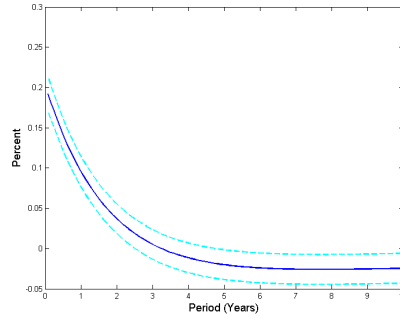


(c) Response of r_t to a one std. dev. of μ_t (d) Response of μ_t to a one std. dev. of μ_t

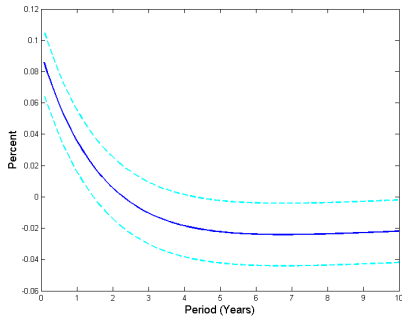
Figure 7: Estimated impulse-response functions of the instantaneous short rate and the central tendency (solid line) along 95 percent confidence bands (dotted lines)



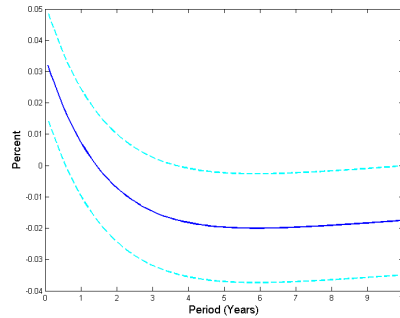
(a) Response of 6-month yield to a one standard deviation of r_t



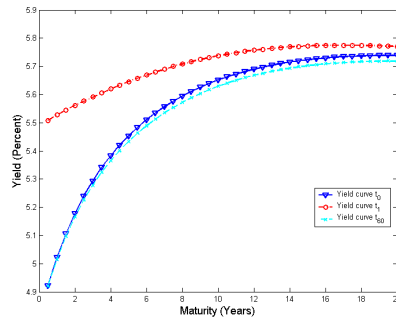
(b) Response of 5-year yield to a one standard deviation of r_t



(c) Response of 10-year yield to a one standard deviation of r_t

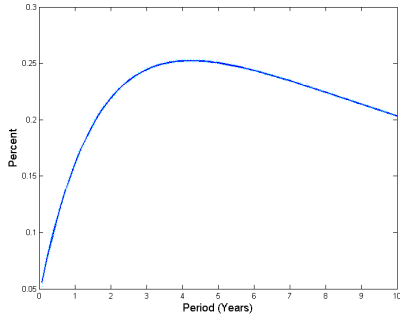


(d) Response of 20-year yield to a one standard deviation of r_t

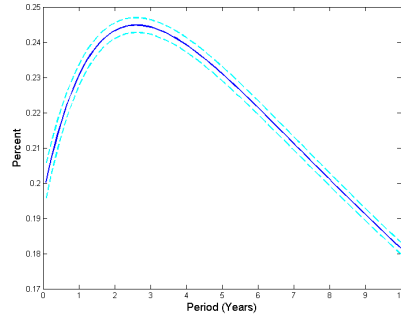


(e) Original yield curve (t_0) and the yield curve one month (t_1) and 5 years (t_{60}) after a one standard deviation of r_t

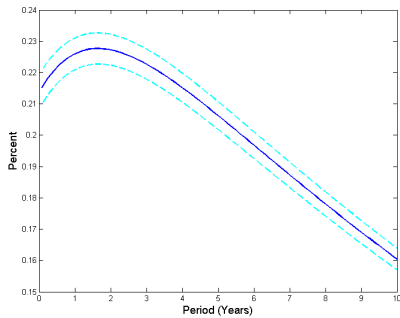
Figure 8: Estimated impulse-response functions of selected yields and the yield curve to a one standard deviation of r_t (solid line) along with 95 percent confidence bands (dotted lines)



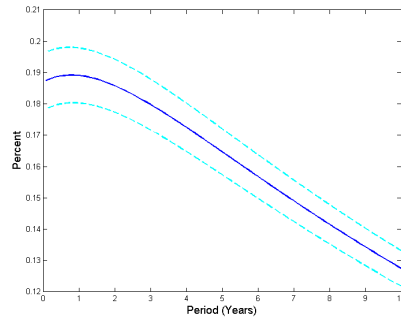
(a) Response of 6-month yield to a one standard deviation of μ_t



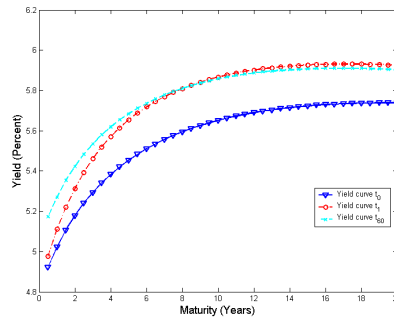
(b) Response of 5-year yield to a one standard deviation of μ_t



(c) Response of 10-year yield to a one standard deviation of μ_t



(d) Response of 20-year yield to a one standard deviation of μ_t



(e) Original yield curve (t_0) and the yield curve one month (t_1) and 5 years (t_{60}) after a one standard deviation of μ_t

Figure 9: Estimated impulse-response functions of selected yields and the yield curve to a one standard deviation of μ_t (solid line) along with 95 percent confidence bands (dotted lines)