

Multi-Lag Term Structure Models with Stochastic Risk Premia

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[Preliminary and incomplete version]

Abstract

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In this paper we propose discrete-time term structure models where we specify an autoregressive of order $p > 1$ dynamics for the factor (x_t) , which is considered as an exogenous or endogenous variable : in the second case the factor is a vector of several yields. We present the Gaussian AR(p) Factor-Based Term Structure Model in which the stochastic discount factor is specified as an exponential-affine function of the factor, with a stochastic risk-correction coefficient, and the associated yield to maturity formula is an affine function of the p most recent lagged values of (x_t) . Several characterization under the risk-neutral and the S -forward probability are considered, and the problem of matching the theoretical and the currently-observed market term structure is also presented. The multifactor generalization is proposed by means of the Gaussian VAR(p) Factor-Based Term Structure Model. An empirical application to the U.S. term structure of interest rates, observed from June 1964 to December 1995, is presented.

Keywords : Discrete-time Affine Term Structure Models, Stochastic Discount Factor, Gaussian VAR(p) processes, Stochastic risk premia, Moving Average or HJM representations, Exact Fitting of the currently-observed yield curve.

JEL number : C1, C5, G1

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1 INTRODUCTION

One of the most important Affine Term Structure Models was the one proposed by Vasicek (1977) in a famous paper where the factor driving the specification of the entire interest rate curve was the instantaneous spot interest rate $r = (r_t, t \geq 0)$. The model was defined in a continuous-time framework and the dynamics of (r_t) was described, under the historical probability, by an Ornstein-Uhlenbeck process with constant coefficients [the continuous-time equivalent of a discrete-time Gaussian AR(1) process]. The limits of this model are well-known: for each time t the rate r_t can be negative with positive probability and the term structure can show a limited number of shapes [monotone increasing, constant, monotone decreasing and humpshaped]. The first lack is compensated by the analytical tractability, induced by the Gaussian (historical and risk-neutral) dynamics of the factor, which is hardly achieved when other conditional distributions are considered for the process (r_t) : indeed, the success of this model, along with the Cox, Ingersoll and Ross (1985) model, comes from the possibility to provide explicit or quasi explicit pricing formula for zero-coupon bonds and interest rate derivatives.

With regard to the second limit, the continuous-time approach can generalize the yield to maturity formula only by the introduction of other (observable or latent) factors [see, among the others, Duffie and Kan (1996), Jagadeesh and Pennacchi (1996), Dai and Singleton (2000), De Jong (2000), Brandt and Chapman (2002) and Ang and Piazzesi (2003)]. On the contrary, the discrete-time approach we follow in this paper is characterized by an additional degree of freedom : the possibility to specify an autoregressive of order p [larger than one] dynamics for the factor driving the term structure shapes.

We propose discrete-time term structure models where the factor (x_t) follows, in the univariate case, a Gaussian AR(p) process, and, in the multivariate case, a Gaussian n -dimensional VAR(p) process. The factor (x_t) is considered as an exogenous or an endogenous variable : in the second case (x_t) is a vector of several yields. We consider an exponential-affine SDF, with a stochastic risk premium defined as an affine function of the p most recent lagged values of the factor and, consequently, the yield to maturity is an affine function of the p lagged values of (x_t) .

Compared with the continuous-time affine case, our approach proposes a more general specification of the conditional historical mean of the factor, gives the possibility to price the different sources of risk taking into account

the recent realizations of (x_t) , and not only the contemporaneous one, and proposes a more general term structure formula.

The plan of the paper is as follows. We consider in Section 2 the Gaussian AR(p) Factor-Based Term Structure Model : the (scalar) exogenous factor (x_t) driving the term structure shapes and dynamics is assumed to be a gaussian AR(p) process, the stochastic discount factor (or pricing kernel) is specified as an exponential-affine function of (x_t) , with a stochastic risk-correction coefficient, and the associated yield to maturity formula is an affine function of the p most recent lagged values of the factor. In Section 3 we study the effect of the autoregressive order p on the possible shapes of term structures that the model is able to replicate, while in Section 4 we consider, under the risk-neutral measure, the Moving Average (or Heath, Jarrow and Morton) representation of the yield and short-term forward rate processes : this representation gives the possibility to exactly replicate the currently-observed yield curve. An alternative methodology to match the theoretical and the market term structure is presented in Section 5 and consists in replacing one of the model's parameter by a deterministic function of time. In Section 6 we deal with the S -forward framework and in Section 7 we specify the endogenous setting, where the scalar factor is the (pre-determined) short rate process (r_{t+1}) : we study, in particular, the problem of propagation of short rate shocks on the yield surface, and the problem of exact replication of the observed term structure. In Section 8 we consider the multifactor VAR(p) generalization [the Gaussian VAR(p) Factor-Based Term Structure Model], while in Section 9 we present an empirical analysis of the above mentioned models using observations, of U. S. term structure of interest rates, from June 1964 to December 1995. Section 10 concludes and appendices gather the proofs.

2 GAUSSIAN AR(p) FACTOR-BASED TERM STRUCTURE MODELS

2.1 Historical Dynamics

Let us assume that the (scalar) exogenous factor x_{t+1} characterizing the specification of the term structure is an AR(p) process of the following type:

$$\begin{aligned} x_{t+1} &= \nu + \varphi_1 x_t + \dots + \varphi_p x_{t+1-p} + \sigma \varepsilon_{t+1} \\ &= \nu + \varphi' X_t + \sigma \varepsilon_{t+1}, \end{aligned} \tag{1}$$

where ε_{t+1} is a gaussian white noise with $\mathcal{N}(0, 1)$ distribution, $\varphi = [\varphi_1, \dots, \varphi_p]'$, $X_t = [x_t, \dots, x_{t+1-p}]'$, and where $\sigma > 0$, ν and φ_i , for $i \in \{1, \dots, p\}$, are scalar coefficients. This model can also be represented in the following multivariate AR(1) form :

$$X_{t+1} = \tilde{\nu} + \Phi X_t + \sigma \tilde{\varepsilon}_{t+1}, \quad (2)$$

where $\tilde{\nu} = [\nu, 0, \dots, 0]'$ and $\tilde{\varepsilon}_{t+1} = [\varepsilon_{t+1}, 0, \dots, 0]'$ are p -dimensional vectors, and where

$$\Phi = \begin{bmatrix} \varphi_1 & \varphi_2 & \dots & \varphi_p \\ 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 1 & 0 \end{bmatrix}$$

is a $(p \times p)$ -matrix.

2.2 The Stochastic Discount Factor Specification

The development of the zero-coupon bond pricing model is characterized, after the historical distribution assumption presented above, by the specification of the stochastic discount factor (SDF) $M_{t,t+1}$ for the period $(t, t+1)$. The price at t of a derivative paying $g(x_{t+H})$ at $t+H$ is:

$$\begin{aligned} C_t(g, H) &= E [M_{t,t+1} \cdot \dots \cdot M_{t+H-1,t+H} g(x_{t+H}) | I_t] \\ &= E_t [M_{t,t+H} g(x_{t+H})], \end{aligned} \quad (3)$$

where E_t denotes the expectation, under the historical probability \mathbb{P} , conditional on the information I_t given by the current and the lagged values of the state variable. We choose a SDF which is exponential-affine in the state variable x_{t+1} :

$$M_{t,t+1} = \exp \left[-\beta - \alpha' X_t + \Gamma_t \varepsilon_{t+1} - \frac{1}{2} \Gamma_t^2 \right], \quad (4)$$

where the coefficients $\alpha = [\alpha_1, \dots, \alpha_p]'$ and β are path independent, and where $\Gamma_t = \Gamma(X_t) = (\gamma_o + \gamma' X_t)$ is a stochastic risk correction coefficient which allows to well represent time variations in assets' risk premia [see sections 2.3 and 2.5]. The absence of arbitrage restriction on the discount bond

with unitary residual maturity requires $E_t(M_{t,t+1}) = \exp(-r_{t+1})$, where r_{t+1} is the (predetermined) short-term interest rate; this condition implies the relation $r_{t+1} = \beta + \alpha'X_t$. This means that, under the absence of arbitrage opportunities, the SDF can be written as:

$$M_{t,t+1} = \exp \left[-r_{t+1} + \Gamma_t \varepsilon_{t+1} - \frac{1}{2} \Gamma_t^2 \right]. \quad (5)$$

2.3 Risk Premium

In order to give an interpretation of the risk-correction coefficient Γ_t , we consider the following definition of risk premium.

Definition 1 : If we denote by P_t the price at time t of a given asset, its risk premium between t and $t + 1$ is :

$$\begin{aligned} \lambda_t &= \log E_t \left(\frac{P_{t+1}}{P_t} \right) - r_{t+1} \\ &= \log E_t \exp(y_{t+1}) - r_{t+1}, \end{aligned} \quad (6)$$

where $y_{t+1} = \log(P_{t+1}/P_t)$ denotes the one-period geometric return of the asset.

We can interpret λ_t as the excess growth rate of the expected price with respect to the present price. Now, starting from this definition of the risk premium we obtain interpretations of the function Γ_t , appearing in the SDF, by means of the following example.

Example : If we consider an asset providing the payoff $\exp(-bx_{t+1})$ at $t + 1$, its price in t is given by:

$$\begin{aligned} P_t &= E_t [M_{t,t+1} P_{t+1}] \\ &= E_t \left[\exp \left(-r_{t+1} - \frac{1}{2} \Gamma_t^2 + (\Gamma_t - b\sigma) \varepsilon_{t+1} - b(\nu + \varphi' X_t) \right) \right] \\ &= \exp \left[-r_{t+1} - b(\nu + \varphi' X_t) - b\sigma \Gamma_t + \frac{1}{2} (b\sigma)^2 \right], \end{aligned}$$

and

$$\begin{aligned} E_t P_{t+1} &= E_t [\exp(-bx_{t+1})] \\ &= \exp[-b(\nu + \varphi' X_t)] E_t \{ \exp[-b\sigma \varepsilon_{t+1}] \} \\ &= \exp \left[-b(\nu + \varphi' X_t) + \frac{1}{2} (b\sigma)^2 \right]. \end{aligned}$$

Finally, the risk premium is:

$$\lambda_t = b\sigma\Gamma_t. \quad (7)$$

Therefore, b , Γ_t and σ can be seen respectively as a risk sensitivity of the asset, a risk price and a risk measure.

2.4 The Affine Term Structure

With the specification of the SDF, and applying formula (3), we determine the price of a zero-coupon bond in the following way :

$$B(t, h) = E_t [M_{t,t+1} \cdot \dots \cdot M_{t+h-1,t+h}], \quad (8)$$

where $B(t, h)$ denoted the price at time t for a time to maturity h .

Note that, for arbitrary real constants μ_1 and μ_2 , we obtained the same SDF dynamics if we replace x_t by $\mu_1 + \mu_2 x_t$ (and therefore X_t by $\mu_1 e + \mu_2 X_t$, with $e = (1, \dots, 1)' \in \mathbb{R}^p$), γ by $\frac{\gamma}{\mu_2}$, γ_o by $\gamma_o - \frac{\mu_1}{\mu_2} \gamma' e$, α by $\frac{\alpha}{\mu_2}$ and β by $\beta - \frac{\mu_1}{\mu_2} \alpha' e$. Therefore, if x_t is not directly observed, we can assume for instance, as far as the term structure is concerned, that $\nu = 0$ and $\sigma = 1$, or $\beta = 0$ and $\alpha_1 = 1$.

Proposition 1 : The price at date t of the zero-coupon bond with time to maturity h is :

$$B(t, h) = \exp(c'_h X_t + d_h), \quad (9)$$

where c_h and d_h satisfies the recursive equations :

$$\begin{aligned} c_h &= -\alpha + \Phi' c_{h-1} + c_{1,h-1} \sigma \gamma \\ &= -\alpha + \Phi^{*'} c_{h-1}, \end{aligned} \quad (10)$$

$$d_h = -\beta + c_{1,h-1}(\nu + \gamma_o \sigma) + \frac{1}{2} c_{1,h-1}^2 \sigma^2 + d_{h-1},$$

with :

$$\Phi^* = \begin{bmatrix} \varphi_1 + \sigma \gamma_1 & \varphi_2 + \sigma \gamma_2 & \dots & \varphi_p + \sigma \gamma_p \\ 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 1 & 0 \end{bmatrix}$$

and where $h \geq 1$, with initial conditions $c_0 = 0, d_0 = 0$ (or $c_1 = -\alpha, d_1 = -\beta$); $c_{1,h}$ is the first component of the p -dimensional vector c_h . [**Proof**: see Appendix 1.]

Corollary 1 : The yields to maturity associated to formula (9) are :

$$\begin{aligned} R(t, h) &= -\frac{1}{h} \log B(t, h) \\ &= -\frac{c'_h}{h} X_t - \frac{d_h}{h}, \quad h \text{ varying,} \end{aligned} \tag{11}$$

and they are affine functions of the factor X_t , that is of the p most recent lagged values of x_{t+1} .

2.5 Excess Returns of Zero-Coupon Bonds

In our framework (with $B^*(t, T) = B(t, T - t)$), we have the following specification for the zero-coupon bond return process.

Proposition 2 : Under the absence of arbitrage opportunity, and for a fixed maturity T , the one-period geometric zero-coupon bond return process $\rho = [\rho(t, T), 0 \leq t \leq T]$, where $\rho(t + 1, T) = \log [B^*(t + 1, T)] - \log [B^*(t, T)]$, is given by:

$$\rho(t + 1, T) = r_{t+1} - \frac{1}{2} \omega(t + 1, T)^2 + \omega(t + 1, T) \Gamma_t - \omega(t + 1, T) \varepsilon_{t+1}, \tag{12}$$

where $\omega(t + 1, T) = -(\sigma c_{1, T-t-1})$ [**Proof**: see Appendix 2].

This means that the process ρ is such that, conditionally to \underline{x}_t , $\rho(t + 1, T)$ is normally distributed with mean $\mu(t + 1, T) = r_{t+1} - \frac{1}{2} \omega(t + 1, T)^2 + \omega(t + 1, T) \Gamma_t$ and variance $\omega(t + 1, T)^2 = (\sigma c_{1, T-t-1})^2$.

The associated risk premium between t and $t + 1$, denoted by $\lambda_t(T)$, is:

$$\lambda_t(T) = \log E_t \exp[\rho(t + 1, T)] - r_{t+1} = \omega(t + 1, T) \Gamma_t. \tag{13}$$

We note that, $\Gamma_t = (\gamma_o + \gamma' X_t)$ plays for any T the role of a risk premium per unit of "risk" $\omega(t + 1, T)$: in particular, the variability of $\lambda_t(T)$ is driven, for a fixed γ different from zero, by the p most recent lagged values of x_{t+1} . If we assume $\gamma = 0$ (i. e., $\Gamma_t = \gamma_o$), the risk correction coefficient and the risk premium of the zero-coupon bond become constants. Also note that, if $T = t + 2$ and $x_t = r_{t+1}$, we have $\omega(t + 1, T) = \sigma$ and we get the result of the example presented in section 2.3 for $b = 1$.

2.6 Risk-Neutral Dynamics

In the previous sections we have presented the Gaussian AR(p) Term Structure model under the historical probability \mathbb{P} . Now, it is well known from asset pricing theory that, under the absence of arbitrage opportunity, there exist equivalent (to \mathbb{P}) probability measures under which asset prices, evaluated with respect to some numeraire³ N_t , are martingales. This change of measure is important in an asset pricing perspective if it leads to convenient closed-form or numerically tractable pricing formulas. The most used choices of numeraire are the money-market account, presented in this section, and the zero-coupon bond choice, presented in the following section.

If we consider as numeraire the money-market account $N_t = \exp(r_1 + \dots + r_t) = (A_{0,t})^{-1}$, where $A_{0,t} = E_0(M_{0,1}) \cdots E_{t-1}(M_{t-1,t})$, the associated equivalent probability \mathbb{Q}_t has a one-period conditional (to I_t) density, with respect to \mathbb{P}_t , given by :

$$\frac{d\mathbb{Q}_t}{d\mathbb{P}_t} = \frac{A_{0,t}M_{t,t+1}}{A_{0,t+1}} = \frac{M_{t,t+1}}{E_t(M_{t,t+1})}.$$

In a general $(T-t)$ -period horizon, the conditional (to I_t) density of the risk-neutral probability \mathbb{Q}_t^T with respect to the historical probability \mathbb{P}_t^T is given by :

$$\frac{d\mathbb{Q}_t^T}{d\mathbb{P}_t^T} = \frac{M_{t,t+1} \cdots M_{T-1,T}}{E_t(M_{t,t+1}) \cdots E_{T-1}(M_{T-1,T})}, \quad (14)$$

and the associated risk-neutral pricing formula for a derivative paying $g(x_T)$ at T is:

$$\begin{aligned} C^*(t, T) &= E_t^{\mathbb{Q}} [E_t(M_{t,t+1}) \cdots E_{T-1}(M_{T-1,T})g(x_T)] \\ &= E_t^{\mathbb{Q}} [\exp(-r_{t+1} - \dots - r_T)g(x_T)]. \end{aligned} \quad (15)$$

The one-period transition from the historical world to the risk-neutral one is given, in our framework, by the conditional density function :

$$\frac{M_{t,t+1}}{E_t(M_{t,t+1})} = \exp \left[\Gamma_t \varepsilon_{t+1} - \frac{1}{2} \Gamma_t^2 \right]. \quad (16)$$

³A numeraire is defined as a non-dividend-paying price process $N = (N_t, t \geq 0)$ with $N_0 = 1$; in other words, N is a stochastic process such that, for every $T > t$, $N_t = E_t[M_{t,T}N_T]$ and $E_0[M_{0,T}N_T] = 1$, where $M_{t,T} = M_{t,t+1} \cdots M_{T-1,T}$. The process $N^* = (N_t M_{0,t}, t \geq 0)$ is therefore a \mathbb{P} -martingale with unitary value in $t = 0$, and if \mathbb{Q} is the probability defined by the sequence of conditional densities $\frac{N_{t+1}M_{t,t+1}}{N_t}$ with respect to \mathbb{P} , a price process S_t is such that S_t/N_t is a \mathbb{Q} -martingale.

Moreover, for any asset, the price P_t at t is equal to $\exp(-r_{t+1})E_t^{\mathbb{Q}}(P_{t+1})$ and, therefore, the risk premium λ_t presented in Definition 1 is equal to $\log E_t(P_{t+1}) - \log E_t^{\mathbb{Q}}(P_{t+1})$.

The risk-neutral Laplace transform of x_{t+1} , conditionally to \underline{x}_t , is given by:

$$\begin{aligned}
E_t^{\mathbb{Q}}[\exp(ux_{t+1})] &= E_t \left[\frac{M_{t,t+1}}{E_t(M_{t,t+1})} \exp(ux_{t+1}) \right] \\
&= E_t \left[\exp \left((\gamma_o + \gamma' X_t) \varepsilon_{t+1} - \frac{1}{2} (\gamma_o + \gamma' X_t)^2 + ux_{t+1} \right) \right] \\
&= \exp \left[u(\nu + \varphi' X_t) - \frac{1}{2} (\gamma_o + \gamma' X_t)^2 \right] \\
&\quad E_t[\exp(\gamma_o + \gamma' X_t + u\sigma)\varepsilon_{t+1}] \\
&= \exp \left[u[(\nu + \sigma\gamma_o) + (\varphi + \sigma\gamma)' X_t] + \frac{1}{2} u^2 \sigma^2 \right],
\end{aligned}$$

where $\varphi = [\varphi_1, \dots, \varphi_p]'$. Therefore, we get the following result.

Proposition 3 : Under the risk neutral probability \mathbb{Q} , x_{t+1} is an $\text{AR}(p)$ process of the following type:

$$x_{t+1} \stackrel{\mathbb{Q}}{=} \nu^* + \varphi_1^* x_t + \dots + \varphi_p^* x_{t+1-p} + \sigma^* \eta_{t+1}, \quad (17)$$

with

$$\begin{aligned}
\nu^* &= (\nu + \sigma\gamma_o) \\
\varphi_i^* &= (\varphi_i + \sigma\gamma_i) \quad \text{for } i \in \{1, \dots, p\} \\
\sigma^* &= \sigma,
\end{aligned}$$

where $\stackrel{\mathbb{Q}}{=}$ denotes the equality in distribution associated to the probability \mathbb{Q} , and where $\eta_{t+1} \stackrel{\mathbb{Q}}{\sim} \mathcal{IIN}(0, 1)$.

This model can be represented in the following vectorial form :

$$X_{t+1} \stackrel{\mathbb{Q}}{=} \tilde{\nu}^* + \Phi^* X_t + \sigma^* \tilde{\eta}_{t+1}, \quad (18)$$

where $\tilde{\nu}^* = [\nu^*, 0, \dots, 0]'$ and $\tilde{\eta}_{t+1} = [\eta_{t+1}, 0, \dots, 0]'$ are p -dimensional vectors, and where Φ^* has been introduced in section 2.4.

We observe that, given the stochastic specification of the risk-premium, in the risk-neutral world x_{t+1} has not only a different constant term, but also different autoregressive coefficients.

With regard to the zero-coupon bond return process, under the risk-neutral probability we have the following specification.

Proposition 4 : In the risk-neutral framework, for a fixed maturity T , the one-period geometric zero-coupon bond return process satisfies the relation:

$$\rho(t+1, T) \stackrel{\mathbb{Q}}{=} r_{t+1} - \frac{1}{2}\omega(t+1, T)^2 - \omega(t+1, T)\eta_{t+1}, \quad (19)$$

with a risk premium equal to :

$$\lambda_t^{\mathbb{Q}}(\rho, 1) = \log E_t^{\mathbb{Q}} \exp[\rho(t+1, T)] - r_{t+1} = 0.$$

[**Proof :** see Appendix 3].

3 TERM STRUCTURE SHAPES

3.1 General Results

The different shapes that the yield curve relation (11) is able to reproduce depend crucially on the system of difference equations (10). Taking into account the result presented in Proposition 1, the system of linear difference equations characterizing (c_h, d_h) , for $h \geq 1$, can be written as:

$$\begin{cases} c_h &= \Phi^{*'} c_{h-1} - \alpha \\ d_h &= -\beta + c_{1,h-1} \nu^* + \frac{1}{2} c_{1,h-1}^2 \sigma^2 + d_{h-1}, \end{cases} \quad (20)$$

with initial conditions $c_0 = 0$ and $d_0 = 0$; in this case, it is well known that the steady state $\bar{\mathbf{C}} = [\bar{c}_1, \dots, \bar{c}_p]'$ of the system c_h is given, I denoting the $(p \times p)$ identity matrix, by:

$$\bar{\mathbf{C}} = -(I - \Phi^{*'})^{-1} \alpha, \quad (21)$$

under the (stability) condition that the p eigenvalues $(\lambda_1, \dots, \lambda_p)$ of $\Phi^{*'}$ are all smaller than unity in modulus, or, equivalently, that the risk-neutral dynamics of x_t is stationary, or that the roots of the risk-neutral autoregressive polynomial (of degree p) $\Psi^*(L) = 1 - \varphi_1^* L - \dots - \varphi_p^* L^p$ have a modulus larger than one (given that these roots are the inverse of the eigenvalues). More

precisely, the system of equations c_h can be rewritten as:

$$\begin{cases} c_{1,h} &= \varphi_1^* c_{1,h-1} + c_{2,h-1} - \alpha_1 \\ c_{2,h} &= \varphi_2^* c_{1,h-1} + c_{3,h-1} - \alpha_2 \\ &\vdots \\ c_{p-1,h} &= \varphi_{p-1}^* c_{1,h-1} + c_{p,h-1} - \alpha_{p-1} \\ c_{p,h} &= \varphi_p^* c_{1,h-1} - \alpha_p, \end{cases}$$

and if we substitute the p^{th} equation in the $(p-1)^{th}$ for $c_{p,h-1}$, and then the $(p-1)^{th}$ equation in the $(p-2)^{th}$ for $c_{p-1,h-1}$, and so on till the first one, we find that $c_{1,h}$ is described by the following p^{th} order linear difference equation :

$$\Psi^*(L)c_{1,h} = - \sum_{i=1}^p \alpha_i,$$

where $\Psi^*(L) = 1 - \varphi_1^*L - \dots - \varphi_p^*L^p$ operates here to h . The remaining equations are given by :

$$c_{p-j,h} = - \sum_{i=0}^j \alpha_{p-i} + \sum_{i=0}^j \varphi_{p-i}^* c_{1,h-j+i-1}, \quad j \in \{0, \dots, p-2\}.$$

Given the risk-neutral stationary assumption on the x_t process, the $c_{p-j,h}$, for $j \in \{0, \dots, p-1\}$, converge at an exponential rate with possible oscillations, when $h \rightarrow \infty$. The limits are :

$$\begin{aligned} \bar{c}_1 &= - \frac{\sum_{i=1}^p \alpha_i}{\Psi^*(1)} \\ \bar{c}_{p-j} &= - \sum_{i=0}^j \alpha_{p-i} + \bar{c}_1 \sum_{i=0}^j \varphi_{p-i}^*, \quad j \in \{0, \dots, p-2\}; \end{aligned}$$

note that $\Psi^*(1) > 0$ because of the stability conditions. In the endogenous framework with $x_t = r_{t+1}$, $\alpha = e_1$ and $\beta = 0$ because of the absence of arbitrage restrictions [see Section 7.1], we have $\bar{c}_1 = -\Psi^*(1)^{-1} < 0$; in the exogenous case we can always assume $\sum_{i=1}^p \alpha_i > 0$ and therefore $\bar{c}_1 < 0$ also.

With regard to d_h , its equation gives the specification of the long-term yield $R(t, \infty)$ as a function of the steady state \bar{c}_1 ; indeed, the difference equation d_h can be written (assuming the identification condition $\beta = 0$) as:

$$d_h = \begin{cases} 0 & \text{for } h = 1, \\ \nu^* \sum_{j=1}^{h-1} c_{1,j} + \frac{1}{2} \sigma^2 \sum_{j=1}^{h-1} c_{1,j}^2, & \forall h \geq 2, \end{cases} \quad (22)$$

and, under the stability of the system c_h , we have from relation (11) that:

$$\begin{aligned} R(t, \infty) &= \lim_{h \rightarrow +\infty} R(t, h) \\ &= \lim_{h \rightarrow +\infty} -\frac{c_h'}{h} X_t - \frac{\nu^*}{h} \sum_{j=1}^{h-1} c_{1,j} - \frac{\sigma^2}{2h} \sum_{j=1}^{h-1} c_{1,j}^2 \\ &= -\bar{c}_1 \nu^* - \frac{1}{2} (\bar{c}_1 \sigma)^2, \end{aligned}$$

which is positive under the condition $[\nu^* + \frac{1}{2} \sigma^2 \bar{c}_1] > 0$.

The shape of the c_h , for h varying, depends on whether the eigenvalues $(\lambda_1, \dots, \lambda_p)$ of Φ^{*} are real or complex, single or multiple, larger or smaller than one in modulus.

The purpose of the following examples is to study the (quantitative and qualitative) properties of (10) and to represent, with some numerical examples, the associated possible shapes of the term structures : here, the values of the parameters are initially fixed on the basis of estimation results we will present in Section 9, and then, variations on each parameter are applied in order to study the richness of shapes the models are able to replicate. Now, given that for the parameters' estimation we will consider as short rate the yield with time to maturity equal to one month, the parameter values will be expressed (along with the short rate itself) on a monthly basis⁴.

Let us study more deeply the solutions of c_h and d_h in the case of $p = 1$ and $p = 2$, and the shapes the associated models are able to replicate.

3.2 Example of the Gaussian AR(1) Factor-Based Term Structure Model

When the scalar endogenous factor $x_t = r_{t+1}$ follows a Gaussian AR(1) process, the associated term structure model is the discrete-time equivalent

⁴Observe that the choice of the unit of measurement for the time to maturity h (cross-section dimension) implies the same unit of measurement for the unit time step (time series dimension).

of the Vasicek model, with a stochastic risk premium. In this case c_h satisfies the first-order difference equation:

$$c_h = -1 + (\varphi + \sigma\gamma)c_{h-1},$$

where $\sigma > 0$, γ and $|\varphi| < 1$ are scalar coefficients, and with a general solution, denoted $c(h)$, given by:

$$\begin{aligned} c(h) &= - \left[\frac{1}{1 - (\varphi + \sigma\gamma)} \right] [1 - (\varphi + \sigma\gamma)^h] \\ &= - \left[\frac{1 - \varphi^{*h}}{1 - \varphi^*} \right], \end{aligned}$$

which tends, for h increasing to infinity, to the limit:

$$\bar{c} = - \left[\frac{1}{1 - \varphi^*} \right],$$

under the condition $|\varphi^*| < 1$, where $\varphi^* = (\varphi + \sigma\gamma)$ is the unique eigenvalue of the (scalar) matrix Φ^* ; this condition implies $c(h) < 0$ for every $h > 0$. In addition, if $0 < \varphi + \sigma\gamma < 1$ (respectively, $-1 < \varphi + \sigma\gamma < 0$), the function $c(h)$ converges in decreasing (respectively, oscillating) towards \bar{c} .

With regard to d_h , it is easy to verify that :

$$\begin{aligned} d(h) &= - \left[\frac{\nu^*}{1 - \varphi^*} \right] (h - 1) + \left[\frac{\varphi^* - \varphi^{*h}}{1 - \varphi^*} \right] \left[\frac{\nu^*}{1 - \varphi^*} - \frac{\sigma^2}{(1 - \varphi^*)^2} \right] \\ &\quad + \frac{\sigma^2}{2(1 - \varphi^*)^2} \left[(h - 1) + \frac{\varphi^{*2} - \varphi^{*2h}}{1 - \varphi^{*2}} \right]; \end{aligned}$$

consequently, the yield to maturity formula (11), for $p = 1$, is given by :

$$\begin{aligned} R(t, h) &= \frac{1}{h} \left[\frac{1 - \varphi^{*h}}{1 - \varphi^*} \right] r_{t+1} + \frac{(h - 1)}{h} \left[\frac{\nu^*}{1 - \varphi^*} \right] \\ &\quad - \frac{1}{h} \left[\frac{\varphi^* - \varphi^{*h}}{1 - \varphi^*} \right] \left[\frac{\nu^*}{1 - \varphi^*} - \frac{\sigma^2}{(1 - \varphi^*)^2} \right] \\ &\quad - \frac{\sigma^2}{2h(1 - \varphi^*)^2} \left[(h - 1) + \frac{\varphi^{*2} - \varphi^{*2h}}{1 - \varphi^{*2}} \right]. \end{aligned}$$

Observe that, in the classical continuous-time Vasicek model, the market risk premium is constant ($\gamma = 0$).

Examples of the term structures are provided in Figures 1 to 4. For a value of $\varphi^* = 0.99$, $r_{t+1} = 0.003$ and $\sigma^2 = 0.00000039$, we observe in Figure 1 that a value of ν^* increasing from 0.00005 to 0.00030 induce the term structure to move from an almost flat shape to a monotone increasing one. In Figure 2 we study once more the effect of variations in ν^* on the term structure with $\varphi^* = 0.99$ and $r_{t+1} = 0.003$, but now we fix $\sigma^2 = 0.000008$ and ν^* increases from 0.00010 to 0.00015 : in this case the yield curve became humpshaped. In Figure 3 we observe the effect on the yield curve of a value of φ^* increasing from 0.87 to 0.99, with $\varphi = 0.95$, $r_{t+1} = 0.003$, $\sigma^2 = 0.00000039$ and $\nu^* = 0.00007$: starting from a monotone decreasing shape, the yield curve find the Vasicek case for $\varphi^* = \varphi = 0.95$ (dashed line) and ends with a monotone increasing shape for $\varphi^* = 0.99$. Figure 4 presents the effect on the term structure of σ^2 increasing from 0.0000004 to 0.0000024: the shape moves from a monotone increasing case to a humped one.

These numerical examples show the shapes the classical one-factor, Markovian of order one, term structure models [Vasicek (1977), Cox, Ingersoll and Ross (1985), Pearson and Sun (1994)] are able to reproduce: monotone increasing, monotone decreasing, flat and humpshaped term structures [see Figure A]. Instead, we frequently observe yield curves with different shapes like, for instance, *J*-shaped (when the yield curve has an interior minimum), *L*-shaped (when the yield curve, starting from r_{t+1} at $h = 1$, takes a decidedly lower value at the following maturities, and then it remains at an almost constant level), inverted *L*-shaped (when the yield curve, starting from r_{t+1} at $h = 1$, takes a decidedly higher value at the following maturities, and then it remains at an almost constant level) or *J*-humpshaped (when the yield curve has, first, an interior minimum, and then an interior maximum) term structures [see Figures B, C and D].

The scalar Gaussian AR(p) Factor-Based Term Structure model, for $p > 1$, is able to overcome these limits and, in particular is able to replicate the observed term structures presented in Figures B, C and D. We present here below, the Gaussian AR(2) case and we give examples of yield curves for $p \in \{2, 3, 4\}$.

3.3 Example of the Gaussian AR(2) Factor-Based Term Structure Model

If the factor $x_t = r_{t+1}$ is a Gaussian AR(2) process, the recursive equation for c_h is described by a first-order (2×2) system of difference equations of

the following type:

$$\begin{bmatrix} c_{1,h} \\ c_{2,h} \end{bmatrix} - \begin{bmatrix} \varphi_1 + \sigma\gamma_1 & 1 \\ \varphi_2 + \sigma\gamma_2 & 0 \end{bmatrix} \begin{bmatrix} c_{1,h-1} \\ c_{2,h-1} \end{bmatrix} = - \begin{bmatrix} 1 \\ 0 \end{bmatrix} ; \quad (23)$$

substituting the first equation into the second, we find for $c_{1,h+1}$ the following second-order linear difference equation:

$$c_{1,h+1} = -1 + \varphi_1^* c_{1,h} + \varphi_2^* c_{1,h-1}, \quad (24)$$

where $\varphi_1^* = (\varphi_1 + \sigma\gamma_1)$ and $\varphi_2^* = (\varphi_2 + \sigma\gamma_2)$; under the condition that the two eigenvalues (λ_1, λ_2) of Φ^* (or the inverse of the roots of $1 - \varphi_1^* L - \varphi_2^* L^2$) are not equal and less than unity in modulus, and regardless of their real or complex nature, the limit of $c_{1,h}$ is given by:

$$\bar{c}_1 = -\frac{1}{(1 - \lambda_1)(1 - \lambda_2)};$$

these conditions can equivalently be expressed in the following way : $\varphi_1^* + \varphi_2^* < 1$, $\varphi_2^* - \varphi_1^* < 1$ and $|\varphi_2^*| < 1$. If we substitute \bar{c}_1 into the second equation of system (23) we find, consequently, the limit of $c_{2,h}$:

$$\bar{c}_2 = -\varphi_2^* \frac{1}{(1 - \lambda_1)(1 - \lambda_2)}.$$

The recursive equation characterizing d_h is given by (22).

Examples of the yield curves that the Gaussian AR(2) model is able to replicate are presented in Figures 5 to 8. In Figure 5 we consider $\varphi_1^* = 0.74$, $\varphi_2^* = 0.24$, $\sigma^2 = 0.00000039$, with $r_{t+1} = 0.0036$ and $r_t = 0.0030$, and we observe what happens when ν^* increases from 0.00005 to 0.00030; the curve start from an *L*-shape and then, as ν^* increases, the long-term yield increases towards larger values with the term structure taking a *J*-shape. In Figure 6 we study what happens when φ_2^* increases from 0.04 to 0.24, with $\varphi_1^* = 0.74$, $\nu^* = 0.00008$, $\sigma^2 = 0.00000039$, and with $r_{t+1} = 0.0030$ and $r_t = 0.0036$. The term structure is initially humpshaped, with an interior maximum for a short maturity ($h = 2$ months), and with a long rate much lower than the short rate; then, as φ_2^* increases, the long rate increases till levels larger than the short rate; here, for $\varphi_2^* = 0.24$, the curve takes an inverted *L*-shape.

In Figure 7 we fix $\varphi_1^* = 0.74$, $\varphi_2^* = 0.24$, $\sigma^2 = 0.000008$, with $r_{t+1} = 0.0032$ and $r_t = 0.0036$, and we study the effect of ν^* increasing from 0.00010 to 0.00015 : the curves are humpshaped as in Figure 2 [Gaussian AR(1)

case], but now we have a strong increment in the yield levels when we move from $h = 1$ to $h = 3$. In Figure 8 we have $\varphi_1^* = 0.74$, $\varphi_2^* = 0.24$ and $\nu^* = 0.00007$, with $r_{t+1} = 0.0036$ and $r_t = 0.0032$; we study the effect on the term structure of σ^2 increasing from 0.0000004 to 0.0000024: the curve is always J -shaped, with an interior minimum for $h = 2$, but when σ^2 increases, the long rate ($h = 60$ months) moves from values larger to values lower than the r_{t+1} , and a hump forms at intermediate maturity yields.

In Figures 9 to 12 we present the yield curves associated to the Gaussian AR(3) specification : we observe that this model is able to replicate the same kind of shapes as the Gaussian AR(2) case, but with curves which are smoother in the short term part as we frequently observe [see the inverted L -shaped curves in Figures C and D].

The Gaussian AR(4) case further enrich, with respect to the previous cases, the family of term structure shapes [see Figures 13 to 16] : more precisely, this model is able to replicate term structure with two interior local maxima and an interior local minimum [see Figure 14], or two interior local minima and an two interior local maxima [see Figure 15] concentrated at short maturities, or yield curves with several changes in the slope around short maturities [see Figures 13 and 16].

The examples presented above, about the Gaussian AR(p) Factor-Based Term Structure Model, show that our (discrete-time) multi-lag approach gives the possibility to reproduce, with a scalar factor, yield curves that Markovian of order one univariate models are not able to replicate. In particular, we have see that the introduction of several lags allow the model to reproduce J -shaped, L -shaped, inverted L -shaped and J -humpshaped yield curves we observe on the data.

4 MOVING AVERAGE OR HJM REPRESENTATIONS

4.1 Risk-Neutral Moving Average Representation of the Yield Process

The purpose of this section is to derive the joint risk-neutral dynamics of the yield to maturity processes $R^*(\cdot, T) = [R^*(t, T), 0 \leq t \leq T]$ (with $R^*(t, T) = R(t, T - t)$), driven by the factor (x_t) introduced above, conditionally to the initial market yield curve $R_M^* = \{R_M^*(0, \tau), \tau \geq 0\}$.

More precisely, we start in this section with the derivation of the joint risk-neutral dynamics of the processes $\{R^*(\cdot, T)\}$, based on the representa-

tion of the log-price zero-coupon bond process in terms of the conditional variances of the bond return process $\rho(t, T)$. In the next section we will represent the process $R^*(\cdot, T)$ in terms of the forward-rate volatility structure.

Starting from the identity

$$\log[B^*(t, T)] = \sum_{j=1}^t \rho(j, T) + \log[B^*(0, T)], \quad (25)$$

we have in the risk-neutral world, using (19):

$$\begin{aligned} \log[B^*(t, T)] &\stackrel{\mathbb{Q}}{=} -\sum_{j=1}^t \omega(j, T)\eta_j + \sum_{j=1}^t r_j \\ &\quad -\frac{1}{2}\sum_{j=1}^t \omega(j, T)^2 + \log[B^*(0, T)], \end{aligned} \quad (26)$$

and, consequently, we find :

Proposition 5 : For every fixed maturity T , the zero-coupon bond price process $B^*(\cdot, T) = [B^*(t, T), 0 \leq t \leq T]$, under the risk-neutral probability \mathbb{Q} , has the following representation:

$$B^*(t, T) \stackrel{\mathbb{Q}}{=} B^*(0, T) \exp \left[\sum_{j=1}^t \left[r_j - \frac{1}{2}\omega(j, T)^2 \right] - \sum_{j=1}^t \omega(j, T)\eta_j \right], \quad (27)$$

where $\eta_j \stackrel{\mathbb{Q}}{\sim} \mathcal{ILN}(0, 1)$, with $j \in \{1, \dots, t\}$.

If we put $T = t$ in (26), we get a relation for the sum of the short-rates:

$$\sum_{j=1}^t r_j \stackrel{\mathbb{Q}}{=} \sum_{j=1}^t \omega(j, t)\eta_j + \frac{1}{2}\sum_{j=1}^t \omega(j, t)^2 - \log[B^*(0, t)] \quad (28)$$

that we can substitute in (26) to find the following alternative representation for the bond price process :

Proposition 6 : For every fixed maturity T , the zero-coupon bond price process $B^*(\cdot, T) = [B^*(t, T), 0 \leq t \leq T]$, under the risk-neutral probability \mathbb{Q} , can be written as :

$$\begin{aligned} B^*(t, T) &\stackrel{\mathbb{Q}}{=} \frac{B^*(0, T)}{B^*(0, t)} \exp \left(-\sum_{j=1}^t [\omega(j, T) - \omega(j, t)] \eta_j \right. \\ &\quad \left. -\frac{1}{2}\sum_{j=1}^t [\omega(j, T)^2 - \omega(j, t)^2] \right). \end{aligned} \quad (29)$$

Relation (29) leads to the following proposition:

Proposition 7 : For every fixed maturity T , the yield process $R^*(\cdot, T) = [R^*(t, T), 0 \leq t \leq T]$ has, under the risk-neutral probability \mathbb{Q} , the following representation:

$$R^*(t, T) \stackrel{\mathbb{Q}}{=} -\frac{1}{T-t} \log \left[\frac{B^*(0, T)}{B^*(0, t)} \right] + \frac{1}{T-t} \sum_{j=1}^t [\omega(j, T) - \omega(j, t)] \eta_j + \frac{1}{T-t} \frac{1}{2} \sum_{j=1}^t [\omega(j, T)^2 - \omega(j, t)^2]. \quad (30)$$

Conditionally to the information \underline{x}_0 , and for every maturity T , the processes $\{R^*(\cdot, T)\}$ are, under the risk-neutral probability \mathbb{Q} , joint MA processes with time-varying coefficients, driven by the same white noise η_t , and the past information appears through the term structure at date $t = 0$.

If we identify the term structure at date $t = 0$ with the market yield curve R_M^* , the yield process $R^*(\cdot, T)$ exactly replicates the term structure observed at the current time $t = 0$. We will see in section 5 a different approach able to guarantee the exact fitting of R_M^* .

4.2 The Risk-Neutral MA Representation of Short-Term Forward Rates: the HJM Framework

It is possible to translate the results we have presented above in terms of short-term forward rates as in the Heath, Jarrow and Morton (1992) (henceforth HJM) approach⁵. In particular, if we denote the one-period forward rate as $f(t, T) = \log[B^*(t, T)] - \log[B^*(t, T+1)]$, and if we use relation (29) we have :

Proposition 8 : For any fixed maturity T , the forward rate process $f(t, T)$ satisfies, under the risk-neutral measure:

$$f(t, T) \stackrel{\mathbb{Q}}{=} f(0, T) + \sum_{j=1}^t [\omega(j, T+1) - \omega(j, T)] \eta_j + \frac{1}{2} \sum_{j=1}^t [\omega(j, T+1)^2 - \omega(j, T)^2], \quad (31)$$

where $f(0, T) = \log \left[\frac{B^*(0, T)}{B^*(0, T+1)} \right]$.

⁵In their paper, which generalizes the discrete-time Ho and Lee (1986) model, HJM (1992) proposed the instantaneous forward rate as the factor to model, and, under the absence of arbitrage, they derived the stochastic evolution of the yield curve, with the forward-rates dynamics fully specified by their instantaneous volatility structures.

Proposition 9 : The one-period forward rate increment $\Delta f(t, T) = f(t + 1, T) - f(t, T)$ satisfies:

$$\Delta f(t, T) \stackrel{\mathbb{Q}}{=} \mu^{\mathbb{Q}}(t + 1, T) + \sigma^{\mathbb{Q}}(t + 1, T)\eta_{t+1}, \quad (32)$$

where

$$\begin{aligned} \sigma^{\mathbb{Q}}(t + 1, T) &= \omega(t + 1, T + 1) - \omega(t + 1, T), \\ \mu^{\mathbb{Q}}(t + 1, T) &= \frac{1}{2} [\omega(t + 1, T + 1)^2 - \omega(t + 1, T)^2] \\ &= \sigma^{\mathbb{Q}}(t + 1, T) \left[\frac{\sum_{\tau=t+1}^T \sigma^{\mathbb{Q}}(t + 1, \tau) + \sum_{\tau=t+1}^{T-1} \sigma^{\mathbb{Q}}(t + 1, \tau)}{2} \right]. \end{aligned}$$

[**Proof :** see Appendix 4].

For every maturity T , the process $f(\cdot, T) = [f(t, T), 0 \leq t \leq T]$ has, therefore, risk-neutral independent increments and a dynamics fully specified by the conditional volatilities of these increments. Relation (31) can be rewritten in the following way:

Corollary 2 : For any fixed maturity T , the forward rate $f(t, T)$ risk-neutral dynamics is given by:

$$\begin{aligned} f(t, T) &\stackrel{\mathbb{Q}}{=} f(0, T) + \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, T) \eta_j \\ &\quad + \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, T) \left[\frac{\sum_{\tau=j}^T \sigma^{\mathbb{Q}}(j, \tau) + \sum_{\tau=j}^{T-1} \sigma^{\mathbb{Q}}(j, \tau)}{2} \right]. \end{aligned} \quad (33)$$

We observe that, as for the yield processes $\{R^*(\cdot, T)\}$, conditionally to the information \underline{x}_0 , and for every maturity T , the processes $\{f(\cdot, T)\}$ are, under the risk-neutral probability \mathbb{Q} , joint MA processes driven by the same white noise η_t , and the past information appears through the forward rates at the date $t = 0$.

If we consider $f(0, T) \equiv f_M(0, T)$, where $f_M(0, T)$ is the market forward rate at $t = 0$, the process $f(\cdot, T)$ exactly fits the currently observed forward rate. With the specification of the risk-neutral dynamics of the forward-rate process in terms of its volatility structure, we can represent the short-term rate process (r_t) in the following alternative way:

Proposition 10 : Under the risk-neutral probability \mathbb{Q} the short-term

interest rate $r_{t+1} = f(t, t)$ is given by the expression:

$$r_{t+1} \stackrel{\mathbb{Q}}{=} f(0, t) + \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, t) \eta_j + \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, t) \left[\frac{\sum_{\tau=j}^t \sigma^{\mathbb{Q}}(j, \tau) + \sum_{\tau=j}^{t-1} \sigma^{\mathbb{Q}}(j, \tau)}{2} \right]. \quad (34)$$

One may observe that formulas (32), (33) and (34) presented above are discrete-time equivalent of classical HJM formulas [see chapter 13 in Musiela and Rutkowski (1997)] in which the conditional risk-neutral variance $\sigma^{\mathbb{Q}}(t+1, T)^2 = \sigma^2(c_{1,T-t} - c_{1,T-t-1})^2$ is a deterministic function of the time to maturity $(T-t)$ and of the parameters $(\sigma, \varphi', \gamma', \alpha')$.

The results presented in sections 4.1 and 4.2 (risk-neutral MA representation of the yield and forward rate processes) can be transposed in the historical probability (\mathbb{P}) setting if the risk premium is constant ($\Gamma_t = \gamma_o$) [see Appendix 5].

5 EXACT FITTING OF THE INITIAL TERM STRUCTURE : EXTENDED AR(p) APPROACH

In the Gaussian AR(p) Factor-Based Term Structure Model derived in sections 2 and 3 the theoretical term structure may produce a poor fit of the market yield curve $R_M^* = \{R_M^*(0, \tau), \tau \geq 0\}$, while the need of an exact fitting is important in order to well price derivative securities like zero-coupon bonds and coupon bonds written on options, or caps, floor and swaptions.

We have seen in the previous section that the HJM approach leads to the exact replication of R_M^* when we identify the term structure at $t = 0$ with the market yield curve. Matching the theoretical and the market term structure of interest rates, at the current time $t = 0$, is also possible by replacing one of the model parameters by a deterministic function of time⁶, and consequently, the resulting model, named Extended Gaussian AR(p) Factor-Based Term Structure model, will be time-non-homogeneous. In particular, if we consider the time-homogeneous model specification (in the exogenous setting), and if we replace the parameter β , in the SDF $M_{t,t+1}$, by a time-dependent function $\beta(t)$, the historical and risk-neutral dynamics

⁶This approach is proposed in the continuous-time literature, for instance, by Ho-Lee (1986) and Dybvig (1988) [extended Merton (1970) model], Hull-White (1990) [extended Vasicek model], Hull-White (1990) and Jamshidian (1995) [extended CIR model], and by Black-Karasinski (1991) [extended log-normal Vasicek model].

of the factor (x_t) are the same, while the short rate process, under the absence of arbitrage opportunities, is given by :

$$r_{t+1} = \beta(t) + \alpha' X_t \quad \forall t \geq 0, \quad (35)$$

and it is, consequently, characterized by a non-homogeneous historical and risk-neutral dynamics. The introduction of the function $\beta(t)$ induces also the recursive equation $d_{T-t} = d_h$ in (10) to take a non-homogeneous specification, denoted $d(t, T)$. More precisely, following the same steps as in the proof of Proposition 1, we easily obtain that, for each maturity $T > 0$, the yields to maturity associated to the extended model, are given by :

$$R_e^*(t, T) = -\frac{c(t, T)'}{T-t} X_t - \frac{d(t, T)}{T-t}, \quad (36)$$

where $(c(t, T), d(t, T))$ satisfy :

$$\begin{cases} c(t, T) &= c_{T-t} = \Phi^* c_{T-t-1} - \alpha \\ d(t, T) &= -\beta(t) + c_{1, T-t-1} \nu^* + \frac{1}{2} c_{1, T-t-1}^2 \sigma^2 + d(t+1, T), \end{cases} \quad (37)$$

with terminal conditions $c(T, T) = 0$ and $d(T, T) = 0$, for each $T > 0$. Observe that the new specification of $d(t, T)$ does not create any problem for the derivation of the solution of the system $(c(t, T), d(t, T))$: first, we solve the (time-homogeneous) difference equation for $c(t, T) = c_h$, as indicated in section 3, and then we substitute backward its solution in $d(t, T)$ starting from the terminal condition $d(T, T) = 0$ ⁷.

The function $\beta(t)$, able to guarantee the exact fitting of R_M^* , can be chosen by means of the following two propositions.

Proposition 11 : In the univariate Extended Gaussian AR(p) Factor-Based Term Structure model, the yield at date $t = 0$ of the zero-coupon bond maturing in T can be written as :

$$R_e^*(0, T) = \frac{1}{T} \sum_{t=0}^{T-1} \beta(t) + R_o^*(0, T), \quad \forall T > 0 \quad (38)$$

where $R_o^*(0, T)$ is obtained from (11) with $\beta = 0$, and for any value of the other parameters and X_0 . [**Proof :** see Appendix 6].

⁷The result presented above, about the yield to maturity formula associated to the extended model, can be generalized to the case where all the model's parameters are deterministic function of time; in this case the backward solution approach is applied to both recursive relations in (37).

Then, if we denote by $f_M = \{f_M(0, \tau), \tau \geq 0\}$ the market term structure of forward rates observed at date $t = 0$, and by $f_o(0, t) = \log B_o^*(0, t) - \log B_o^*(0, t + 1)$, where $B_o^*(0, t)$ is obtained from (9) with $\beta = 0$, we can find the function $\beta(t)$ permitting an exact fitting.

Proposition 12 : The univariate Extended Gaussian AR(p) Factor-Based Term Structure model fits the currently-observed yield curve R_M^* if and only if :

$$\begin{aligned} \beta(t) &= f_M(0, t) - f_o(0, t), \\ &= f_M(0, t) + \mu_{o,t} + \mu_t' X_0, \quad \forall t \in [0, T - 1], \end{aligned} \tag{39}$$

where

$$\begin{aligned} \mu_{o,t} &= c_{1,t} \nu^* + \frac{1}{2} c_{1,t}^2 \sigma^2 \\ \mu_t &= \left(\Phi^{*'} c_t - c_t - \alpha \right)' X_0 \end{aligned} \tag{40}$$

[**Proof :** see Appendix 7].

One may observe that, by choosing $\beta(t)$ as in (39), the extended model exactly fits the observed yield curve for each possible value of X_0 and of the parameters of the model.

6 S -FORWARD DYNAMICS

In many financial applications, a convenient numeraire is the zero-coupon bond whose maturity S is the same as the derivative to price. More precisely, the equivalent martingale measure is determined in this case, for every date $t \in [0, S]$, by the numeraire $N_t = \frac{B^*(t,S)}{B^*(0,S)}$, and it is referred to as S -forward probability and denoted by \mathbb{Q}_S . The one-period conditional (to I_t) density of the S -forward probability $\mathbb{Q}_{t,S}$ with respect to the historical probability

\mathbb{P}_t , and to the risk-neutral probability \mathbb{Q}_t , are respectively given by:

$$\begin{aligned} \frac{d\mathbb{Q}_{t,S}}{d\mathbb{P}_t} &= \frac{M_{t,t+1}B^*(t+1,S)}{B^*(t,S)} \\ \frac{d\mathbb{Q}_{t,S}}{d\mathbb{Q}_t} &= \frac{d\mathbb{Q}_{t,S}}{d\mathbb{P}_t} \frac{d\mathbb{P}_t}{d\mathbb{Q}_t} \\ &= E_t(M_{t,t+1}) \frac{B^*(t+1,S)}{B^*(t,S)} \\ &= \exp(-r_{t+1}) \frac{B^*(t+1,S)}{B^*(t,S)}; \end{aligned}$$

therefore, at the $(T-t)$ -periods horizon (where $T \leq S$), the S -forward probability $\mathbb{Q}_{t,S}^T$ has a (conditional to I_t) joint density with respect to the risk-neutral probability \mathbb{Q}_t^T given by:

$$\begin{aligned} \frac{d\mathbb{Q}_{t,S}^T}{d\mathbb{Q}_t^T} &= \prod_{\tau=t+1}^T \exp(-r_\tau) \frac{B^*(\tau,S)}{B^*(\tau-1,S)} \\ &= \frac{B^*(T,S)}{B^*(t,S)} \exp(-r_{t+1} - \dots - r_T), \end{aligned}$$

and the pricing formula (15) takes, for $S = T$, the following useful representation:

$$\begin{aligned} C^*(t,T) &= E_t^{\mathbb{Q}} [\exp(-r_{t+1} - \dots - r_T) g(x_T)] \\ &= B^*(t,T) E_t^{\mathbb{Q}_T} [g(x_T)], \end{aligned} \tag{41}$$

in which the problem of derivative pricing reduces to calculating an expectation of the payoff $g(x_T)$.

Proposition 13 : The S -forward dynamics of x_{t+1} has an $\text{AR}(p)$ representation of the following type:

$$x_{t+1} \stackrel{\mathbb{Q}_S}{=} \nu_S + \varphi_1^* x_t + \dots + \varphi_p^* x_{t+1-p} + \sigma^* \xi_{t+1}, \tag{42}$$

with

$$\nu_S = \nu^* - \sigma^* \omega(t+1, S),$$

where $\stackrel{\mathbb{Q}_S}{=}$ denotes the equality in distribution associated to the probability \mathbb{Q}_S , and where $\xi_{t+1} \sim \mathcal{IIN}(0, 1)$ under \mathbb{Q}_S . [**Proof :** see Appendix 8.]

This model can be represented in the following vectorial form :

$$X_{t+1} \stackrel{\mathbb{Q}_S}{=} \tilde{\nu}_S + \Phi^* X_t + \sigma^* \tilde{\xi}_{t+1}, \quad (43)$$

where $\tilde{\nu}_S = [\nu_S, 0, \dots, 0]'$ $= \tilde{\nu}^* - \sigma^* \omega(t+1, S)e_1$ and $\tilde{\xi}_{t+1} = [\xi_{t+1}, 0, \dots, 0]'$ are p -dimensional vectors; e_1 denotes the first element of the canonical basis of \mathbb{R}^p .

Proposition 14 : In the S -forward framework, the one-period geometric zero-coupon bond return process is described by the relation:

$$\begin{aligned} \rho(t+1, T) \stackrel{\mathbb{Q}_S}{=} & -\omega(t+1, T) \xi_{t+1} + r_{t+1} - \frac{1}{2} \omega(t+1, T)^2 \\ & + \omega(t+1, T) \omega(t+1, S), \end{aligned} \quad (44)$$

with a one-period risk premium given by :

$$\lambda_t^{\mathbb{Q}_S}(T) = \log E_t^{\mathbb{Q}_S} \exp [\rho(t+1, T)] - r_{t+1} = \omega(t+1, T) \omega(t+1, S).$$

[**Proof :** see Appendix 9].

Consequently, under the T -forward probability, the one-period risk premium per unit of $\omega(t+1, T)$ is given by the $\omega(t+1, T)$ itself.

Proposition 15 : For every fixed maturity T , the yield process $R^*(\cdot, T) = [R^*(t, T), 0 \leq t \leq T]$, under the S -forward probability \mathbb{Q}_S , has the following representation:

$$\begin{aligned} R^*(t, T) \stackrel{\mathbb{Q}_S}{=} & -\frac{1}{T-t} \log \left[\frac{B^*(0, T)}{B^*(0, t)} \right] + \frac{1}{T-t} \sum_{j=1}^t [\omega(j, T) - \omega(j, t)] \xi_j \\ & - \frac{1}{T-t} \sum_{j=1}^t \omega(j, S) [\omega(j, T) - \omega(j, t)] \\ & + \frac{1}{T-t} \frac{1}{2} \sum_{j=1}^t [\omega(j, T)^2 - \omega(j, t)^2]. \end{aligned} \quad (45)$$

For every maturity T , and conditionally to the information \underline{x}_0 , the processes $\{R^*(\cdot, T)\}$ are, also under the S -forward probability \mathbb{Q}_S , joint MA processes, driven by the same white noise ξ_t , and the past information is summarized in term structure at date $t = 0$. [**Proof :** see Appendix 10.]

Corollary 3 : The zero-coupon bond price process $B^*(\cdot, T) = [B^*(t, T), 0 \leq t \leq T]$ is, for every maturity T and under the T -forward probability \mathbb{Q}_T , given by :

$$\begin{aligned} B^*(t, T) \stackrel{\mathbb{Q}_T}{=} & \frac{B^*(0, T)}{B^*(0, t)} \exp \left(\sum_{j=1}^t [\omega(j, t) - \omega(j, T)] \xi_j \right. \\ & \left. + \frac{1}{2} \sum_{j=1}^t [\omega(j, t) - \omega(j, T)]^2 \right). \end{aligned} \quad (46)$$

Proposition 16 : For any fixed maturity T , the forward rate $f(t, T) = \log[B^*(t, T)] - \log[B^*(t, T + 1)]$ satisfies, under the S -forward probability, the following relation:

$$\begin{aligned}
f(t, T) \stackrel{\mathbb{Q}_S}{=} & f(0, T) + \sum_{j=1}^t [\omega(j, T + 1) - \omega(j, T)] \xi_j \\
& + \frac{1}{2} \sum_{j=1}^t [\omega(j, T + 1)^2 - \omega(j, T)^2] \\
& - \sum_{j=1}^t \omega(j, S) [\omega(j, T + 1) - \omega(j, T)];
\end{aligned} \tag{47}$$

[Proof : It follows immediately from the definition of forward rate and from Proposition 15].

Conditionally to the information \underline{x}_0 , and for every maturity T , the processes $\{f(\cdot, T)\}$ are, also under the S -forward probability \mathbb{Q}_S , joint MA processes and the past information appears through the forward rate at the date $t = 0$.

If we consider $S = T + 1$, relation (47) becomes :

$$\begin{aligned}
f(t, T) \stackrel{\mathbb{Q}_{T+1}}{=} & f(0, T) + \sum_{j=1}^t [\omega(j, T + 1) - \omega(j, T)] \xi_j \\
& - \frac{1}{2} \sum_{j=1}^t [\omega(j, T + 1) - \omega(j, T)]^2
\end{aligned} \tag{48}$$

$$\stackrel{\mathbb{Q}_{T+1}}{=} f(0, T) + \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, T) \xi_j - \frac{1}{2} \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, T)^2,$$

and, therefore, the process $\frac{B^*(t, T)}{B^*(t, T+1)} = \exp f(t, T)$ is, for every $t \leq T$, a \mathbb{Q}_{T+1} -martingale.

Corollary 4 : Under the S -forward probability \mathbb{Q}_S , the one-period forward rate increment $\Delta f(t, T) = f(t + 1, T) - f(t, T)$ is given by:

$$\Delta f(t, T) \stackrel{\mathbb{Q}_S}{=} \mu^{\mathbb{Q}_S}(t + 1, T) + \sigma^{\mathbb{Q}_S}(t + 1, T) \xi_{t+1}, \tag{49}$$

where

$$\begin{aligned}
\sigma^{\mathbb{Q}_S}(t+1, T) &= \sigma^{\mathbb{Q}}(t+1, T) = \omega(t+1, T+1) - \omega(t+1, T), \\
\mu^{\mathbb{Q}_S}(t+1, T) &= \frac{1}{2}[\omega(t+1, T+1)^2 - \omega(t+1, T)^2] \\
&\quad - \omega(t+1, S)[\omega(t+1, T+1) - \omega(t+1, T)] \\
&= \frac{1}{2}\sigma^{\mathbb{Q}}(t+1, T) \left[\sum_{\tau=t+1}^T \sigma^{\mathbb{Q}}(t+1, \tau) + \sum_{\tau=t+1}^{T-1} \sigma^{\mathbb{Q}}(t+1, \tau) \right] \\
&\quad - \sigma^{\mathbb{Q}}(t+1, T) \sum_{\tau=t+1}^{S-1} \sigma^{\mathbb{Q}}(t+1, \tau)
\end{aligned}$$

Under the S -forward probability, the process $f = (f(t, T), t \geq 0)$ has independent increments and a dynamics fully specified by its conditional risk-neutral volatilities. In the particular case $S = T + 1$, we have:

$$\begin{aligned}
\sigma^{\mathbb{Q}_{T+1}}(t+1, T) &= \sigma^{\mathbb{Q}}(t+1, T), \\
\mu^{\mathbb{Q}_{T+1}}(t+1, T) &= -\frac{1}{2}[\omega(t+1, T+1) - \omega(t+1, T)]^2 \\
&= -\frac{1}{2}\sigma^{\mathbb{Q}}(t+1, T)^2,
\end{aligned}$$

coherently with the martingale property of $\exp f(t, T)$ under \mathbb{Q}_{T+1} .

With the specification of the $(T + 1)$ -forward dynamics of the forward-rate process in terms of its volatility structure, we can represent the short-term rate process (r_t) in the following way:

Proposition 17 : Under the $(T + 1)$ -forward probability \mathbb{Q}_{T+1} , and for any maturity T , the short-term interest rate $r_{T+1} = f(T, T)$ is given by the expression:

$$r_{T+1} \stackrel{\mathbb{Q}_{T+1}}{=} f(0, T) + \sum_{j=1}^T \sigma^{\mathbb{Q}}(j, T)\xi_j - \frac{1}{2} \sum_{j=1}^T \sigma^{\mathbb{Q}}(j, T)^2. \quad (50)$$

We observe from (50) that $E_0^{\mathbb{Q}_{T+1}}(\exp r_{T+1}) = \exp f(0, T)$, that is, the linear short-term forward rate $F(0, T, T + 1)$ spanning the interval $(T, T + 1)$ is, under the \mathbb{Q}_{T+1} -forward probability, an unbiased predictor of the future linear short-term interest rate $L(T, T + 1)$ spanning the same period.

7 THE ENDOGENOUS CASE

7.1 The Term Structure

In what we have presented above, the factor x_t was exogenous. In the term structure literature several models are specified assuming $x_t = r_{t+1}$: the shape and the dynamics of the yield curve is driven by the short-rate process. This is a convenient framework, given that we can specify the historical dynamics of the factor starting from the observed stylized facts on the short-rate.

In this case, the results presented in the previous sections remain valid, except for the absence of arbitrage opportunity restriction for r_{t+1} , which requires $\alpha = e_1$ and $\beta = 0$, with e_1 denoting the first element of the canonical basis of \mathbb{R}^p . Consequently, the initial conditions in the recursive equations presented in Proposition 1 become $c_1 = -e_1$ and $d_1 = 0$.

In addition, the endogenous framework is useful to study how a shock on the short rate $r_{t+1} = R(t, 1)$ is propagated on the surface $R_{\mathcal{T}, \mathcal{H}} = [R(t + \tau, h), \tau \in \mathcal{T}, h \in \mathcal{H}]$, where $\mathcal{T} = \{0, \dots, T - t - 1\}$ and $\mathcal{H} = (1, \dots, H)$.

7.2 Propagation of Short Rate Shocks on the Yield Surface

The result presented in Proposition 1 describes, conditionally to X_t , the yields as a deterministic function of the time to maturity h , for a fixed date t . In many financial and economic contexts one needs to study which is the propagation of a shock, on the short-term interest rates, in the yield curve at different dates and for several maturities (e.g.: a Central Bank that needs to set a monetary policy). This means that we are interested in the dynamics of the process $R_{\mathcal{H}} = [R(t, h), 0 \leq t < T, h \in \mathcal{H}]$, for a given set of residual times to maturity $\mathcal{H} = (1, \dots, H)$.

If we consider a fixed time to maturity h , the process $R = [R(t, h), 0 \leq t < T]$ can be described by the following Proposition.

Proposition 18 : For a fixed time to maturity h , the process $R = [R(t, h), 0 \leq t < T]$ is an ARMA($p, p - 1$) process of the following type :

$$\Psi(L)R(t, h) = \sigma \mathbf{C}_h(L)\varepsilon_t + \mathbf{C}_h(1)\nu + \Psi(1)\delta_h, \quad (51)$$

where $\mathbf{C}_h(L) = -(c_{1,h} + c_{2,h}L + \dots + c_{p,h}L^{p-1})/h$ is a polynomial of degree $(p - 1)$ in the lag operator L , $\delta_h = -(d_h/h)$, and where the AR polynomial, applying to t , is given by $\Psi(L) = (1 - \varphi_1 L - \dots - \varphi_p L^p)$. [**Proof:** see Appendix 11.]

We observe that the AR polynomial is independent of h , while the MA polynomial is not.

Proposition 19 : For a given set of residual time to maturities $\mathcal{H} = (1, \dots, H)$, and for any status of the factor, the stochastic evolution of the yield curve process $R_{\mathcal{H}} = [R(t, h), 0 \leq t < T, h \in \mathcal{H}]$ takes the following particular H -variate VARMA($p, p - 1$) representation:

$$\Psi(L) \begin{pmatrix} R(t, 1) \\ R(t, 2) \\ \vdots \\ R(t, H) \end{pmatrix} = \sigma \begin{pmatrix} \mathbf{C}_1(L) \\ \mathbf{C}_2(L) \\ \vdots \\ \mathbf{C}_H(L) \end{pmatrix} \varepsilon_t + \begin{pmatrix} (\alpha'e)\nu + \Psi(1)\delta_1 \\ \mathbf{C}_2(1)\nu + \Psi(1)\delta_2 \\ \vdots \\ \mathbf{C}_H(1)\nu + \Psi(1)\delta_H \end{pmatrix}. \quad (52)$$

In our endogenous setting, with $x_t = r_{t+1}$, the yield curve process $R_{\mathcal{H}}$ is described by relation (52), with $R(t, 1) = r_{t+1}$, $\alpha'e = 1$, $\mathbf{C}_1(L) = 1$ and $\delta_1 = 0$. Consequently, in the endogenous setting the short rate process is Markovian (of order p) by definition, while, in the exogenous framework, the short rate dynamics is non-Markovian because of the ARMA($p, p - 1$) specification.

In order to study how a shock on the short rate is propagated on the surface $R_{\mathcal{T}, \mathcal{H}}$ we need to determine, for every maturity $h \in \mathcal{H}$, the infinite moving average [MA(∞)] representation of the process R .

Proposition 20 : Under the condition that the polynomials $\mathbf{C}_h(\cdot)$ and $\Psi(\cdot)$ have no common roots, and that $\Psi(z) \neq 0$ for each complex number z such that $|z| \leq 1$, the process $R = [R(t, h), 0 \leq t < T]$ has, for each maturity $h \in \mathcal{H}$, the following MA(∞) representation:

$$R(t, h) = \Psi(1)^{-1} \mathbf{C}_h(1)\nu + \delta_h + \sigma \Theta_h(L) \varepsilon_t, \quad (53)$$

with

$$\Theta_h(L) = \sum_{j=0}^{+\infty} \theta_{j,h} L^j = \Psi(L)^{-1} \mathbf{C}_h(L).$$

For each $\tau \in \mathcal{T}$ and $h \in \mathcal{H}$, the effect on $R(t + \tau, h)$ of a unit shock on ε_t is therefore measured by the MA coefficient $\sigma \theta_{\tau,h}$.

7.3 Exact Fitting of the Initial Term Structure in the Endogenous Framework

In section 5 we have presented the problem of matching the initial theoretical and the market term structure of interest rates in the case where an

exogenous factor (x_t) drives the term structure shapes and dynamics. In particular, the fact to consider the parameter β as the deterministic function of time derived in Proposition 12, guarantee the exact fitting of the currently-observed yield curve R_M^* .

In the endogenous setting, because of the absence of arbitrage opportunities ($\beta = 0$), the above mentioned approach must be applied to a different parameter. In particular, the exact fitting of R_M^* is possible if we replace the parameter ν in the historical dynamics of r_{t+1} by a time-dependent function $\tilde{\nu}(t)$.

Let us denote by $\{B_o^*(t, T), t \geq 0, T \geq t\}$ the term structure corresponding to any fixed values of $(\nu, \varphi, \sigma, \gamma_o, \gamma)$ and to the observed initial values of the short rate $(r_0, r_{-1}, \dots, r_{-p+1})$. This term structure is thus defined by the historical dynamics :

$$\Psi(L)r_{t+1} = \nu + \sigma\varepsilon_t, \quad t \geq 0, \quad \varepsilon_{t+1} \sim \mathcal{IIN}(0, 1), \quad (54)$$

where $\Psi(L) = (1 - \varphi_1 L - \dots - \varphi_p L^p)$, the initial values $(r_0, r_{-1}, \dots, r_{-p+1})$, and the SDF in (5) or, equivalently, by the risk-neutral dynamics :

$$\Psi^*(L)r_{t+1} \stackrel{\mathbb{Q}}{=} \nu + \gamma_o \sigma + \sigma \eta_t, \quad t \geq 0, \quad \eta_{t+1} \sim \mathcal{IIN}(0, 1) \quad (55)$$

where $\Psi^*(L) = (1 - \varphi_1^* L - \dots - \varphi_p^* L^p)$, with $\varphi_i^* = \varphi_i + \sigma \gamma_i$, and the initial values $(r_0, r_{-1}, \dots, r_{-p+1})$.

Let us now consider the "extended" term structure $\{B_e^*(t, T), t \geq 0, T \geq t\}$ corresponding to the time varying deterministic parameter $\nu + \tilde{\nu}(t)$, to the same parameters $(\varphi, \sigma, \gamma_o, \gamma)$ and the same initial values $(r_0, r_{-1}, \dots, r_{-p+1})$. This new term structure is thus defined by the historical dynamics :

$$\Psi(L)r_{t+1} \stackrel{\mathbb{P}_e}{=} \nu + \tilde{\nu}(t) + \sigma\varepsilon_t, \quad t \geq 0, \quad \varepsilon_{t+1} \sim \mathcal{IIN}(0, 1), \quad (56)$$

with the same initial values, and the same SDF, or, equivalently, by the risk-neutral dynamics :

$$\Psi^*(L)r_{t+1} \stackrel{\mathbb{Q}_e}{=} \nu + \tilde{\nu}(t) + \gamma_o \sigma + \sigma \eta_t, \quad t \geq 0, \quad \eta_{t+1} \sim \mathcal{IIN}(0, 1) \quad (57)$$

and the same initial values [\mathbb{Q}_e denotes the risk-neutral probability associated to the extended historical dynamics \mathbb{P}_e].

If we consider now the process

$$z_t = r_t - \zeta(t-1) \quad \text{with } t \geq -p+1, \quad (58)$$

where $\zeta(t)$ is the sequence of real numbers defined by :

$$\begin{aligned}\zeta(-1) &= \dots = \zeta(-p) = 0 \quad \text{and} \\ \Psi^*(L)\zeta(t) &= \tilde{\nu}(t), \quad t \geq 0,\end{aligned}\tag{59}$$

the risk-neutral dynamics of the process z_t is defined by :

$$\Psi^*(L)z_{t+1} \stackrel{\mathbb{Q}_e}{=} \nu + \gamma_o\sigma + \sigma\eta_t, \quad t \geq 0, \quad \eta_{t+1} \sim \mathcal{IIN}(0, 1)\tag{60}$$

with $z_0 = r_0, \dots, z_{-p+1} = r_{-p+1}$.

In other words, the process (z_t) has, under \mathbb{Q}_e , exactly the same dynamics as (r_t) under \mathbb{Q} , and therefore :

$$\begin{aligned}B_e^*(0, T) &= E_0^{\mathbb{Q}_e} \exp(-r_1 - \dots - r_T) \\ &= E_0^{\mathbb{Q}_e} \exp\left(-\sum_{t=0}^{T-1} \zeta(t) - \sum_{t=1}^T z_t\right) \\ &= \exp\left(-\sum_{t=0}^{T-1} \zeta(t)\right) E_0^{\mathbb{Q}} \exp(-\sum_{t=1}^T r_t) \\ &= \exp\left(-\sum_{t=0}^{T-1} \zeta(t)\right) B_o^*(0, T).\end{aligned}\tag{61}$$

So, using the same method as in Proposition 12, we can adjust $B_e^*(0, T)$ to the market initial term structure $B_M^*(0, T)$, by choosing

$$\zeta(t) = f_M(0, t) - f_o(0, t), \quad \text{with } t \geq 0\tag{62}$$

where $f_M(0, t)$ is the observed forward rate and $f_o(0, t)$ the forward rate associated with $B_o^*(0, t)$ [i.e., $f_o(0, t) = \log B_o^*(0, t) - \log B_o^*(0, t+1)$], or, equivalently :

$$\begin{aligned}\tilde{\nu}(t) &= \Psi^*(L)\zeta(t), \quad t \geq 0, \\ \text{with } \zeta(t) &= f_M(0, t) - f_o(0, t), \quad t \geq 0, \\ \text{and } \zeta(-1) &= \dots = \zeta(-p) = 0.\end{aligned}\tag{63}$$

Note that $f_o(0, t) = r_1 = f_M(0, t)$ and, therefore, $\zeta(0) = \tilde{\nu}(0) = 0$.

The result is summarized in the following proposition.

Proposition 21 : For any values of the parameters $(\nu, \varphi, \sigma, \gamma_o, \gamma)$, the extended term structure associated with the SDF (5) and the historical dynamics :

$$\Psi(L)r_{t+1} \stackrel{\mathbb{P}_e}{=} \nu + \tilde{\nu}(t) + \sigma\varepsilon_t, \quad t \geq 0, \quad \varepsilon_{t+1} \sim \mathcal{IIN}(0, 1),$$

where

$$\tilde{\nu}(t) = \Psi^*(L)\zeta(t), \quad t \geq 0, \quad (64)$$

$$\zeta(t) = f_M(0, t) - f_o(0, t), \quad t \geq 0,$$

$$\zeta(-1) = \dots = \zeta(-p) = 0,$$

fits exactly the observed term structure at $t = 0$, if $f_M(0, t)$ is the observed forward rate and $f_o(0, t)$ the theoretical forward rate obtained from the "non extended" dynamics in which $\tilde{\nu}(t)$ is replaced by 0, and we have $\tilde{\nu}(0) = 0$.

Example : If we consider the Extended Gaussian AR(1) Factor-Based Term Structure Model, we have the discrete time equivalent of the (continuous-time) Extended Vasicek model.

Under the risk-neutral measure \mathbb{Q} , the dynamics of the instantaneous rate is given, for the Vasicek model, by :

$$dr_t = a(b - r_t)dt - \sigma dW_t, \quad r_0 = r > 0, \quad (65)$$

where a , σ and b are positive scalar coefficients, and where W_t is a standard Brownian motion under \mathbb{Q} ; it is well known that the risk-neutral extended dynamics is :

$$dr_t = \left[\partial_t f_M(0, t) + a f_M(0, t) + \frac{\sigma^2}{2a}(1 - \exp(-2at)) \right] - ar_t dt - \sigma dW_t. \quad (66)$$

where $\partial_t f_M(0, t)$ denotes the derivative of the market forward rate observed at $t = 0$ [see Chapter 3 in Brigo and Mercurio (2001)].

With regard to the endogenous Extended Gaussian AR(1) Factor-Based Term Structure Model, the one-period increment is given, under \mathbb{Q} , by :

$$\Delta r_{t+1} = \nu^* + \tilde{\nu}(t) + (\varphi^* - 1)r_t + \sigma\eta_t, \quad \eta_{t+1} \sim \mathcal{IIN}(0, 1), \quad (67)$$

and we know, from Proposition 21, that :

$$\begin{aligned}
\tilde{\nu}(t) &= (1 - \varphi^* L)[f_M(0, t) - f_o(0, t)] \\
&= f_M(0, t) - f_M(0, t - 1) + (1 - \varphi^*)f_M(0, t - 1) \\
&\quad - [f_o(0, t) - \varphi^* f_o(0, t - 1)],
\end{aligned} \tag{68}$$

with :

$$\begin{aligned}
f_o(0, t) - \varphi^* f_o(0, t - 1) &= [(c_t - c_{t+1}) - \varphi^*(c_{t-1} - c_t)]r_1 \\
&\quad + (d_t - d_{t+1}) - \varphi^*(d_{t-1} - d_t) \\
&= (d_t - d_{t+1}) - \varphi^*(d_{t-1} - d_t) \\
&= \nu^*[\varphi^* c_{1,t} - c_{1,t-1}] - \frac{\sigma^2}{2}[c_t^2 - \varphi^* c_{t-1}^2] \\
&= \nu^* - \frac{\sigma^2}{2}[c_t^2 - \varphi^* c_{t-1}^2] \\
&= \nu^* - \frac{\sigma^2}{2} \left[\frac{(1 - \varphi^{*t})^2}{(1 - \varphi^*)^2} - \varphi^* \frac{(1 - \varphi^{*(t-1)})^2}{(1 - \varphi^*)^2} \right] \\
&= \nu^* - \frac{\sigma^2}{2(1 - \varphi^*)} [1 - \varphi^{*2t-1}].
\end{aligned} \tag{69}$$

If we substitute (69) in (67) we find the discrete-time equivalent of the extended Vasicek risk-neutral dynamics (66) :

$$\begin{aligned}
\Delta r_{t+1} &= \left[\Delta f_M(0, t) + (1 - \varphi^*)f_M(0, t - 1) + \frac{\sigma^2}{2(1 - \varphi^*)} (1 - \varphi^{*2t-1}) \right] \\
&\quad - (1 - \varphi^*)r_t + \sigma \eta_t,
\end{aligned} \tag{70}$$

which is similar to (66).

8 GAUSSIAN VAR(p) FACTOR-BASED TERM STRUCTURE MODELS

8.1 Historical Dynamics, SDF and Affine Term Structure

Let us assume that the exogenous factor $x_{t+1} = (x_{1,t+1}, \dots, x_{n,t+1})'$ driving the term structure is an n -dimensional VAR(p) process of the following type:

$$\begin{aligned} x_{t+1} &= \nu + \varphi_1 x_t + \dots + \varphi_p x_{t+1-p} + \sigma \varepsilon_{t+1} \\ &= \nu + \varphi X_t + \sigma \varepsilon_{t+1}, \end{aligned} \tag{71}$$

where ε_{t+1} is an n -dimensional gaussian white noise with $\mathcal{N}(0, I)$ distribution [I denotes the $(n \times n)$ identity matrix]; σ and φ_j , for each $j \in \{1, \dots, p\}$, are $(n \times n)$ matrices [σ can be chosen, for instance, lower triangular], and $\varphi = [\varphi_1, \dots, \varphi_p]$ is an $(n \times np)$ matrix; ν is an n -dimensional vector, while $X_t = (x'_t, \dots, x'_{t+1-p})'$ is an (np) -dimensional vector. This model can equivalently be represented in the following (np) -dimensional AR(1) form :

$$X_{t+1} = \Phi X_t + [\nu + \sigma \varepsilon_{t+1}] e_1, \tag{72}$$

where e_1 is a vector of size np , with all entries equal to zero except for the first n elements which are all equal to one, and where

$$\Phi = \begin{bmatrix} \varphi_1 & \varphi_2 & \dots & \varphi_p \\ 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 1 & 0 \end{bmatrix}$$

is a $(np \times np)$ matrix. Using the notation :

$$\Gamma_t = [\Gamma_{1,t}, \dots, \Gamma_{n,t}]',$$

where $\Gamma_{i,t} = \gamma_{o,i} + \tilde{\gamma}_i' X_t$, $i \in \{1, \dots, n\}$, and $\Gamma_t = \gamma_o + \gamma X_t$, with $\gamma_o = [\gamma_{o,1}, \dots, \gamma_{o,n}]'$ an n -dimensional vector and $\gamma = [\tilde{\gamma}_1, \dots, \tilde{\gamma}_n]'$ an $(n \times np)$ matrix, the SDF is defined as :

$$M_{t,t+1} = \exp \left[-\beta - \alpha' X_t + \Gamma_t' \varepsilon_{t+1} - \frac{1}{2} \Gamma_t' \Gamma_t \right]. \tag{73}$$

Moreover, assuming the absence of arbitrage opportunities for r_{t+1} we get $r_{t+1} = \beta + \alpha' X_t$. It is also easy to verify that the risk premium, for an asset providing the payoff $\exp(-b' x_{t+1})$ at $t + 1$, is $\lambda_t = b' \sigma \Gamma_t$.

The term structure of zero coupon-bond prices is given by the following proposition :

Proposition 22 : In the Gaussian VAR(p) Factor-Based Term Structure Model, the price at date t of the zero-coupon bond with time to maturity h is :

$$B(t, h) = \exp(c'_h X_t + d_h), \quad (74)$$

where c_h and d_h satisfies, for $h \geq 1$, the recursive equations :

$$\begin{cases} c_h &= -\alpha + \Phi' c_{h-1} + (\sigma\gamma)' c_{1,h-1} \\ &= -\alpha + \Phi^* c_{h-1}, \\ d_h &= -\beta + c'_{1,h-1}(\nu + \sigma\gamma_o) + \frac{1}{2}c'_{1,h-1}\sigma\sigma'c_{1,h-1} + d_{h-1}, \end{cases} \quad (75)$$

with, using the notation $\gamma = [\gamma_1, \dots, \gamma_p]$ and where the γ_i 's are $(n \times n)$ matrices :

$$\Phi^* = \begin{bmatrix} \varphi_1 + \sigma\gamma_1 & \varphi_2 + \sigma\gamma_2 & \dots & \varphi_p + \sigma\gamma_p \\ 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 1 & 0 \end{bmatrix} \quad \text{a } (np \times np) \text{ matrix}$$

with initial conditions $c_0 = 0, d_0 = 0$ (or $c_1 = -\alpha, d_1 = -\beta$); $c_{1,h}$ indicates the vector of the first n components of the (np) -dimensional vector c_h . [**Proof :** see Appendix 13.]

Corollary 5 : The yields to maturity associated to formula (74) are :

$$\begin{aligned} R(t, h) &= -\frac{1}{h} \log B(t, h) \\ &= -\frac{c'_h}{h} X_t - \frac{d_h}{h}, \quad h \text{ varying,} \end{aligned} \quad (76)$$

and they are affine functions of the factor X_t , that is of the p most recent lagged values of the n -dimensional factor x_{t+1} .

With regard to the one-period geometric zero-coupon bond return process $\rho = [\rho(t, T), 0 \leq t \leq T]$, with $\rho(t+1, T) = \log [B^*(t+1, T)] - \log [B^*(t, T)]$, it is easy to verify that :

$$\begin{aligned} \rho(t+1, T) &= r_{t+1} - \frac{1}{2} \omega(t+1, T)' \omega(t+1, T) \\ &\quad + \omega(t+1, T)' \Gamma_t - \omega(t+1, T)' \varepsilon_{t+1}, \end{aligned} \tag{77}$$

where $\omega(t+1, T) = -(\sigma' c_{1, T-t-1})$ is an n -dimensional row vector [the proof is a generalization, to the multivariate setting, of the proof in Appendix 2]. The associated risk premium, between t and $t+1$, is given by :

$$\begin{aligned} \lambda_t(T) &= \omega(t+1, T)' \Gamma_t \\ &= \sum_{i=1}^n \omega_i(t+1, T) \Gamma_{i,t}, \end{aligned} \tag{78}$$

where $\omega(t+1, T) = [\omega_1(t+1, T), \dots, \omega_n(t+1, T)]'$. One may notice that, in this multivariate setting, the magnitude of $\lambda_t(T)$ is given by a linear combination of the n scalar risk premia $\Gamma_{i,t} = \gamma_{o,i} + \tilde{\gamma}'_i X_t$; moreover, for a given matrix γ different from zero, $\lambda_t(T)$ is function of the p most recent lagged values of the n -dimensional factor x_{t+1} .

8.2 Risk-Neutral Dynamics

The Laplace transform of the one-period (conditional to \underline{x}_t) risk-neutral distribution of x_{t+1} is given by :

$$\begin{aligned} E_t^{\mathbb{Q}}[\exp(u' x_{t+1})] &= E_t \left[\frac{M_{t,t+1}}{E_t(M_{t,t+1})} \exp(u' x_{t+1}) \right] \\ &= E_t \left[\exp \left(\Gamma'_t \varepsilon_{t+1} - \frac{1}{2} \Gamma'_t \Gamma_t + u' x_{t+1} \right) \right] \\ &= \exp \left[u' (\nu + \varphi X_t) - \frac{1}{2} \Gamma'_t \Gamma_t \right] \\ &\quad E_t[\exp(\Gamma_t + \sigma' u)' \varepsilon_{t+1}] \\ &= \exp \left[u' [(\nu + \sigma \gamma_o) + (\varphi + \sigma \gamma) X_t] + \frac{1}{2} u' \sigma \sigma' u \right], \end{aligned}$$

Therefore we get :

Proposition 23 : Under the risk neutral probability \mathbb{Q} , x_{t+1} is an n -

dimensional VAR(p) process of the following type:

$$\begin{aligned} x_{t+1} &\stackrel{\mathbb{Q}}{=} \nu^* + \varphi_1^* x_t + \dots + \varphi_p^* x_{t+1-p} + \sigma^* \eta_{t+1} \\ &\stackrel{\mathbb{Q}}{=} \nu^* + \varphi^* X_t + \sigma^* \eta_{t+1}, \end{aligned} \quad (79)$$

with

$$\begin{aligned} \nu^* &= (\nu + \sigma \gamma_o) \\ \varphi_j^* &= (\varphi_j + \sigma \gamma_j), \quad \text{for } j \in \{1, \dots, p\} \\ \varphi^* &= [\varphi_1^*, \dots, \varphi_p^*] \\ \sigma^* &= \sigma, \end{aligned}$$

where $\stackrel{\mathbb{Q}}{=}$ denotes the equality in distribution (associated to the probability \mathbb{Q}); η_{t+1} is (under \mathbb{Q}) an n -dimensional gaussian white noise with $\mathcal{N}(0, I)$ distribution, and I denotes the $(n \times n)$ identity matrix.

This model can be represented in the following vectorial form :

$$X_{t+1} \stackrel{\mathbb{Q}}{=} \Phi^* X_t + [\nu^* + \sigma^* \eta_{t+1}] e_1, \quad (80)$$

where e_1 is the vector of size np , with all entries equal to zero except for the first n elements which are all equal to one, and where Φ^* is defined in Proposition 22.

With regard to the one-period zero-coupon bond return process, under the risk-neutral probability we have that :

$$\rho(t+1, T) \stackrel{\mathbb{Q}}{=} r_{t+1} - \frac{1}{2} \omega(t+1, T)' \omega(t+1, T) - \omega(t+1, T)' \eta_{t+1}, \quad (81)$$

with a risk premium $\lambda_t^{\mathbb{Q}}(T) = 0$ [the proof is a generalization, to the multivariate setting, of the proof in Appendix 3].

8.3 Moving Average or HJM Representations in the Multivariate Framework

The purpose of this section is to generalize the results presented in section 4 to the case where the factor (x_t) is the n -dimensional process defined in (71). Following the same steps as in the univariate setting, and using relation (81), we have :

Proposition 24 : In the Gaussian VAR(p) Factor-Based Term Structure Model, for every fixed maturity T and under the risk-neutral probability \mathbb{Q} :

- the yield process $R^*(\cdot, T) = [R^*(t, T), 0 \leq t \leq T]$ has the following representation:

$$\begin{aligned} R^*(t, T) \stackrel{\mathbb{Q}}{=} & -\frac{1}{T-t} \log \left[\frac{B^*(0, T)}{B^*(0, t)} \right] + \frac{1}{T-t} \sum_{j=1}^t [\omega(j, T) - \omega(j, t)]' \eta_j \\ & + \frac{1}{T-t} \frac{1}{2} \sum_{j=1}^t [\omega(j, T)' \omega(j, T) - \omega(j, t)' \omega(j, t)], \end{aligned} \quad (82)$$

where $\eta_j \stackrel{\mathbb{Q}}{\sim} \mathcal{IIN}(0, I)$. Conditionally to the information x_0 , and for every maturity T , the processes $\{R^*(\cdot, T)\}$ are, under the risk-neutral probability \mathbb{Q} , joint MA processes, driven by a n -dimensional gaussian white noise, with time-varying coefficients, and the past information appears through the term structure at date $t = 0$;

- the forward rate process $f(t, T) = \log[B^*(t, T)] - \log[B^*(t, T+1)]$ satisfies :

$$\begin{aligned} f(t, T) \stackrel{\mathbb{Q}}{=} & f(0, T) + \sum_{j=1}^t [\omega(j, T+1) - \omega(j, T)]' \eta_j \\ & + \frac{1}{2} \sum_{j=1}^t [\omega(j, T+1)' \omega(j, T+1) - \omega(j, T)' \omega(j, T)] \\ \stackrel{\mathbb{Q}}{=} & f(0, T) + \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, T)' \eta_j \\ & + \frac{1}{2} \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, T)' \left[\sum_{\tau=j}^T \sigma^{\mathbb{Q}}(j, \tau) + \sum_{\tau=j}^{T-1} \sigma^{\mathbb{Q}}(j, \tau) \right] \end{aligned} \quad (83)$$

where $f(0, T) = \log \left[\frac{B^*(0, T)}{B^*(0, T+1)} \right]$ and $\sigma^{\mathbb{Q}}(j, \tau) = \omega(j, \tau+1) - \omega(j, \tau)$ for each $j \leq \tau \leq T$.

As for the yield processes $\{R^*(\cdot, T)\}$, conditionally to the information x_0 , and for every maturity T , the processes $\{f(\cdot, T)\}$ are, under the risk-neutral probability \mathbb{Q} , joint MA processes and the past information appears through the forward rates at the date $t = 0$.

- the one-period forward rate increment $\Delta f(t, T) = f(t+1, T) - f(t, T)$ is give by:

$$\Delta f(t, T) \stackrel{\mathbb{Q}}{=} \mu^{\mathbb{Q}}(t+1, T) + \sigma^{\mathbb{Q}}(t+1, T)' \eta_{t+1}, \quad (84)$$

where

$$\begin{aligned}
\sigma^{\mathbb{Q}}(t+1, T) &= \omega(t+1, T+1) - \omega(t+1, T), \\
\mu^{\mathbb{Q}}(t+1, T) &= \frac{1}{2} [\omega(t+1, T+1)' \omega(t+1, T+1) \\
&\quad - \omega(t+1, T)' \omega(t+1, T)] \\
&= \frac{1}{2} \sigma^{\mathbb{Q}}(t+1, T)' \\
&\quad \times \left[\sum_{\tau=t+1}^T \sigma^{\mathbb{Q}}(t+1, \tau) + \sum_{\tau=t+1}^{T-1} \sigma^{\mathbb{Q}}(t+1, \tau) \right];
\end{aligned}$$

- the short-term interest rate $r_{t+1} = f(t, t)$ is given by the expression:

$$\begin{aligned}
r_{t+1} &\stackrel{\mathbb{Q}}{=} f(0, t) + \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, t)' \eta_j \\
&\quad + \frac{1}{2} \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, t)' \left[\sum_{\tau=j}^t \sigma^{\mathbb{Q}}(j, \tau) + \sum_{\tau=j}^{t-1} \sigma^{\mathbb{Q}}(j, \tau) \right].
\end{aligned} \tag{85}$$

8.4 S -Forward Dynamics

The S -forward dynamics of the n -dimensional factor x_{t+1} , has an VAR(p) representation of the following type:

$$x_{t+1} \stackrel{\mathbb{Q}_S}{=} \nu_S + \varphi_1^* x_t + \dots + \varphi_p^* x_{t+1-p} + \sigma^* \xi_{t+1}, \tag{86}$$

with

$$\nu_S = \nu^* - \sigma^* \omega(t+1, S),$$

where $\stackrel{\mathbb{Q}_S}{=}$ denotes the equality in distribution associated to the probability \mathbb{Q}_S , and where $\xi_{t+1} \sim \mathcal{IIN}(0, I)$ under \mathbb{Q}_S [the proof is a generalization, to the multivariate setting, of the proof in Appendix 8].

This model can be represented in the following vectorial form :

$$X_{t+1} \stackrel{\mathbb{Q}_S}{=} \Phi^* X_t + [\nu_S + \sigma^* \xi_{t+1}] e_1, \tag{87}$$

where e_1 denotes the vector of size np , with all entries equal to zero except for the first n elements which are all equal to one.

Moreover, the one-period geometric zero-coupon bond return process is given by:

$$\begin{aligned} \rho(t+1, T) \stackrel{\mathbb{Q}_S}{=} & r_{t+1} - \omega(t+1, T)' \xi_{t+1} - \frac{1}{2} \omega(t+1, T)' \omega(t+1, T) \\ & + \omega(t+1, T)' \omega(t+1, S), \end{aligned} \quad (88)$$

with one-period risk premium given by :

$$\lambda_t^{\mathbb{Q}_S}(T) = \log E_t^{\mathbb{Q}_S} \exp[\rho(t+1, T)] - r_{t+1} = \omega(t+1, T)' \omega(t+1, S).$$

Starting from relations (87) and (88), we can determine the generalizations, to the multivariate S -forward framework, of the results presented in section 6. More precisely, we have :

Proposition 25 : In the Gaussian VAR(p) Factor-Based Term Structure Model, for every fixed maturity T and under the S -forward probability \mathbb{Q}_S :

- the yield process $R^*(\cdot, T) = [R^*(t, T), 0 \leq t \leq T]$ has the following representation:

$$\begin{aligned} R^*(t, T) \stackrel{\mathbb{Q}_S}{=} & -\frac{1}{T-t} \log \left[\frac{B^*(0, T)}{B^*(0, t)} \right] + \frac{1}{T-t} \sum_{j=1}^t [\omega(j, T) - \omega(j, t)]' \xi_j \\ & - \frac{1}{T-t} \sum_{j=1}^t [\omega(j, T) - \omega(j, t)]' \omega(j, S) \\ & + \frac{1}{T-t} \frac{1}{2} \sum_{j=1}^t [\omega(j, T)' \omega(j, T) - \omega(j, t)' \omega(j, t)]; \end{aligned} \quad (89)$$

for every maturity T , and conditionally to the information \underline{x}_0 , the processes $\{R^*(\cdot, T)\}$ are, also under the S -forward probability \mathbb{Q}_S , joint MA processes, driven by a n -dimensional gaussian white noise, and the past information is summarized in term structure at date $t = 0$;

- the forward rate $f(t, T) = \log[B^*(t, T)] - \log[B^*(t, T+1)]$ satisfies the following relation:

$$\begin{aligned} f(t, T) \stackrel{\mathbb{Q}_S}{=} & f(0, T) + \sum_{j=1}^t [\omega(j, T+1) - \omega(j, T)]' \xi_j \\ & + \frac{1}{2} \sum_{j=1}^t [\omega(j, T+1)' \omega(j, T+1) - \omega(j, T)' \omega(j, T)] \\ & - \sum_{j=1}^t [\omega(j, T+1) - \omega(j, T)]' \omega(j, S); \end{aligned} \quad (90)$$

Conditionally to the information x_0 , and for every maturity T , the processes $\{f(\cdot, T)\}$ are, also under the S -forward probability \mathbb{Q}_S , joint MA processes and the past information appears through the forward rate at the date $t = 0$.

If we consider $S = T + 1$, relation (90) becomes:

$$\begin{aligned}
f(t, T) &\stackrel{\mathbb{Q}_{T+1}}{=} f(0, T) + \sum_{j=1}^t [\omega(j, T+1) - \omega(j, T)]' \xi_j \\
&\quad - \frac{1}{2} \sum_{j=1}^t [\omega(j, T+1) - \omega(j, T)]' [\omega(j, T+1) - \omega(j, T)] \\
&\stackrel{\mathbb{Q}_{T+1}}{=} f(0, T) + \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, T)' \xi_j - \frac{1}{2} \sum_{j=1}^t \sigma^{\mathbb{Q}}(j, T)' \sigma^{\mathbb{Q}}(j, T),
\end{aligned} \tag{91}$$

and, therefore, the process $\frac{B^*(t, T)}{B^*(t, T+1)} = \exp f(t, T)$ is, for every $t \leq T$, a \mathbb{Q}_{T+1} -martingale.

- the one-period forward rate increment $\Delta f(t, T) = f(t+1, T) - f(t, T)$ is given by:

$$\Delta f(t, T) \stackrel{\mathbb{Q}_S}{=} \mu^{\mathbb{Q}_S}(t+1, T) + \sigma^{\mathbb{Q}_S}(t+1, T)' \xi_{t+1}, \tag{92}$$

where

$$\begin{aligned}
\sigma^{\mathbb{Q}_S}(t+1, T) &= \sigma^{\mathbb{Q}}(t+1, T) = \omega(t+1, T+1) - \omega(t+1, T), \\
\mu^{\mathbb{Q}_S}(t+1, T) &= \frac{1}{2} [\omega(t+1, T+1)' \omega(t+1, T+1) \\
&\quad - \omega(t+1, T)' \omega(t+1, T)] \\
&\quad - [\omega(t+1, T+1) - \omega(t+1, T)]' \omega(t+1, S) \\
&= \frac{1}{2} \sigma^{\mathbb{Q}}(t+1, T)' \\
&\quad \times \left[\sum_{\tau=t+1}^T \sigma^{\mathbb{Q}}(t+1, \tau) + \sum_{\tau=t+1}^{T-1} \sigma^{\mathbb{Q}}(t+1, \tau) \right] \\
&\quad - \sigma^{\mathbb{Q}}(t+1, T)' \sum_{\tau=t+1}^{S-1} \sigma^{\mathbb{Q}}(t+1, \tau),
\end{aligned}$$

the process $f = (f(t, T), t \geq 0)$ has independent increments and a dynamics fully specified by its conditional risk-neutral volatilities. In

the particular case $S = T + 1$, we have:

$$\begin{aligned}\sigma^{\mathbb{Q}_{T+1}}(t+1, T) &= \sigma^{\mathbb{Q}}(t+1, T), \\ \mu^{\mathbb{Q}_{T+1}}(t+1, T) &= -\frac{1}{2} \sigma^{\mathbb{Q}}(t+1, T)' \sigma^{\mathbb{Q}}(t+1, T),\end{aligned}$$

coherently with the martingale property of $\exp f(t, T)$ under \mathbb{Q}_{T+1} .

- for $S = T + 1$, the short-term interest rate $r_{T+1} = f(T, T)$ is given by the expression:

$$r_{T+1} \stackrel{\mathbb{Q}_{T+1}}{=} f(0, T) + \sum_{j=1}^T \sigma^{\mathbb{Q}}(j, T)' \xi_j - \frac{1}{2} \sum_{j=1}^T \sigma^{\mathbb{Q}}(j, T)' \sigma^{\mathbb{Q}}(j, T). \quad (93)$$

8.5 The Endogenous Case

In the multivariate endogenous setting, the n -dimensional factor (x_t) can be considered as a vector of yields at different maturities in which the first component is assumed to be the short rate r_{t+1} ; more precisely, we assume:

$$x_t = \begin{bmatrix} R(t, h_1) \\ R(t, h_2) \\ \vdots \\ R(t, h_n) \end{bmatrix} \quad (94)$$

where $R(t, 1) = r_{t+1}$ and $h_1 < h_2 < \dots < h_n$. In this case, the absence of arbitrage conditions for the n yields in (94) imply :

$$\begin{aligned}(i) \quad c_1 &= -e_1, \quad d_1 = 0, \\ (ii) \quad c_{h_j} &= -h_j e_{h_j}, \quad d_{h_j} = 0, \quad \forall j \in \{2, \dots, n\},\end{aligned} \quad (95)$$

where e_{h_j} denotes the h_j^{th} element of the canonical basis in \mathbb{R}^{np} . The first set of conditions is used as initial value in the recursive formula of Proposition 22; the second set of $(n-1)$ conditions imply restrictions on the parameters $(\Phi^*, \nu^*, \sigma^*)$ which must be taken into account at the estimation stage.

9 EMPIRICAL ANALYSIS

9.1 Description of the Data

The CRSP data set on the U. S. term structure of interest rates (treasury zero-coupon bond yields), that we consider in the following application, covers the period from June 1964 to December 1995 and contains 379 monthly observations for each of the nine maturities : 1, 3, 6 and 9 months and 1, 2, 3, 4 and 5 years⁸. Summary statistics about the above mentioned (annualized) yields are presented in Table 1 : the term structure is, on average, upward sloping and the yields with larger standard deviation, positive skewness and kurtosis are those with shorter maturities. Moreover, yields are highly autocorrelated with a persistence which is increasing with the time to maturity.

Table 1 : Summary Statistics on U. S. Monthly Yields from June 1964 to December 1995. $ACF(k)$ indicates the empirical autocorrelation between yields $R(t, h)$ and $R(t, h - k)$, with h and k expressed on a monthly basis.

Maturity	1-m	3-m	6-m	9-m	1-yr	2-yr	3-yr	4-yr	5-yr
Mean	0.0645	0.0672	0.0694	0.0709	0.0713	0.0734	0.0750	0.0762	0.0769
Std. Dev.	0.0265	0.0271	0.0270	0.0269	0.0260	0.0252	0.0244	0.0240	0.0237
Skewness	1.2111	1.2118	1.1518	1.1013	1.0307	0.9778	0.9615	0.9263	0.8791
Kurtosis	4.5902	4.5237	4.3147	4.1605	3.9098	3.6612	3.5897	3.5063	3.3531
Minimum	0.0265	0.0277	0.0287	0.0299	0.0311	0.0366	0.0387	0.0397	0.0398
Maximum	0.1640	0.1612	0.1655	0.1644	0.1581	0.1564	0.1556	0.1582	0.1500
ACF(1)	0.9587	0.9731	0.9747	0.9745	0.9727	0.9780	0.9797	0.9802	0.9822
ACF(5)	0.8288	0.8531	0.8579	0.8588	0.8604	0.8783	0.8915	0.8986	0.9053
ACF(10)	0.7278	0.7590	0.7691	0.7699	0.7683	0.7885	0.8021	0.8075	0.8212
ACF(20)	0.4303	0.4631	0.4880	0.4996	0.5156	0.5742	0.6051	0.6193	0.6431
ACF(30)	0.2548	0.2682	0.3016	0.3213	0.3518	0.4358	0.4725	0.4994	0.5187
ACF(40)	0.1362	0.1415	0.1677	0.1853	0.2160	0.3056	0.3427	0.3780	0.3961

9.2 Estimated Models

9.2.1 Advantages of an Endogenous Multi-Lag Approach

We have seen in the previous sections that the scalar and multivariate Multi-Lag Term Structure Models can be characterized by an exogenous or endogenous factor (x_t). In the present empirical analysis we follow an endogenous approach, given that it gives several important advantages coming from the observations we have about the factor, that is, the short rate in the scalar case, or yields at different maturities in the multivariate framework. First,

⁸The same data set is used in the papers of Longstaff and Schwartz (1992) and Bansal and Zhou (2002). We are grateful to Ravi Bansal and Hao Zhou for providing us the data set.

thanks to data, we are able to detect stylized facts on interest rates which give us the possibility to justify the $AR(p)$ model we propose for the historical dynamics of (x_t) : indeed, a large empirical literature on bond yields show that interest rates have an historical multi-lag dynamics [see, among the others, Hamilton (1989), Christiansen and Lund (2003), Cochrane and Piazzesi (2005)]. Second, observations about the Gaussian-distributed factor lead to an exact maximum likelihood estimation of historical parameters: in this way, we are able to test hypotheses using likelihood ratio statistics, and rank the models in terms of various information criteria. Finally, the difference between directly observed and estimated factor values determine model residuals that can be used to derive various diagnostic criteria.

By a comparison with this Multi-Lag Endogenous discrete-time approach, the classical continuous-time affine term structure approach *à la* Duffie and Kan (1996) and Dai and Singleton (2000) has some different features. First, the factors are in general assumed not observable and therefore justifications for the (historical) factors dynamics, along with a precise econometric analysis of model residuals, are not possible. Second, in order to reconstruct a time series of the latent factors, for an exact maximum likelihood estimation, prices of some zero-coupon bonds are assumed to be perfectly observed in order to inverse the pricing equations [see Chen and Scott (1987) and Pearson and Sun (1994)]; this inversion technique depends on the zero-coupon bonds selected values of the parameters, which are not initially available, and therefore the reconstructed time series is model-sensitive [see Collin-Dufresne, Goldstein and Jones (2004) for a discussion]. Third, the class of discrete-time affine (Compound Autoregressive)⁹ processes is much larger than the discrete-time counterpart of the continuous-time affine class¹⁰ [see Gourieroux, Monfort and Polimenis (2005), and Darolles, Gourieroux and Jasiak (2006)].

9.2.2 Estimation Method

The methodology we follow to estimate the parameters of the endogenous Multi-Lag term structure models is based on a consistent two-step procedure.

⁹A Compound Autoregressive (Car) process is a Markovian process with an exponential-affine conditional Laplace transform [see Darolles, Gourieroux, Jasiak (2006) for details].

¹⁰For instance, the discrete-time Gaussian VAR(1) process has a continuous-time equivalent if and only if there exists a matrix ϑ such that $\varphi = \exp(-\vartheta)$, or, any Car(p) process [like the Gaussian VAR(p) process], with $p \geq 2$, cannot be the time discretized version of a continuous-time affine process.

In the first step, thanks to observations on the n -dimensional endogenous factor (x_t) , we estimate the $[n(1 + np) + (n(n + 1)/2)]$ -dimensional vector of parameters $\theta_{\mathbb{P}} = [\nu', vec(\varphi)', vech(\sigma\sigma')']'$, characterizing the historical dynamics (x_t) , by Maximum Likelihood (ML).

In the second step, using observations on yields with maturities different from those used in the first step and for a given estimates of $vech(\sigma\sigma')$, we estimate the $[n(1+np)]$ -dimensional vector of parameters $\theta_{\mathbb{Q}} = [(\nu^*)', vec(\varphi^*)']'$, characterizing the risk-neutral dynamics of (x_t) , by minimizing the sum of squared fitting errors between the observed and theoretical yields. We assume that the vector of measurement errors is a multivariate white noise, with contemporaneously homoskedastic and uncorrelated components. More precisely, in the scalar case, we estimate $\theta_{\mathbb{Q}}$ by nonlinear least squares (NLLS), while, in the multivariate case, these parameters are estimated by constrained NLLS. The constraints are imposed to satisfy restrictions (95)-(ii) implied by the absence of arbitrage opportunity principle [see Section 8.5].

Given the complete set of nine maturities of our data base, and given a number m of yields used to estimate the vector of historical parameters $\theta_{\mathbb{P}}$, we denote by H_m^* the set of remaining maturities used to estimate the vector of risk-neutral parameters $\theta_{\mathbb{Q}}$.

In the AR(p) Factor-Based case [see Section 9.3], x_t is the one-month yield to maturity $R(t, 1)$ expressed at a monthly frequency, while, in the bivariate VAR(p) Factor-Based case [see Section 9.4] the factor is given by :

$$x_t = [R(t, 1), R(t, 60) - R(t, 1)]',$$

where $[R(t, 60) - R(t, 1)]$ is the spread at date t between the five-year and one-month yield to maturity, expressed at a monthly frequency [see Ang and Bekaert (2002), and Ang, Piazzesi and Wei (2005) for similar specifications].

The NLLS estimator for the AR(p) case, is determined by :

$$\begin{cases} \hat{\theta}_{\mathbb{Q}} = Arg \min_{\theta_{\mathbb{Q}}} S^2(\theta_{\mathbb{Q}}), \\ S^2(\theta_{\mathbb{Q}}) = \sum_{t=p}^T \sum_{h \in H_1^*} [\tilde{R}(t, h) - R(t, h)]^2, \end{cases} \quad (96)$$

given the set H_1^* of maturities used to estimate the risk-neutral parameters; $\tilde{R}(t, h)$ is the observed yield, and $R(t, h)$ is the theoretical yield determined by formula (11).

The constrained NLLS estimator, in our bivariate model specification, is

given by :

$$\left\{ \begin{array}{l} \hat{\theta}_{\mathbb{Q}} = \text{Arg min}_{\theta_{\mathbb{Q}}} S^2(\theta_{\mathbb{Q}}) \\ S^2(\theta_{\mathbb{Q}}) = \sum_{t=p}^T \sum_{h \in H_2^*} [\tilde{R}(t, h) - R(t, h)]^2, \\ \text{s. t. } \sum_{t=p}^T [\tilde{R}(t, 60) - R(t, 60)]^2 = 0, \end{array} \right. \quad (97)$$

where $R(t, h)$ is the theoretical yield determined by formula (76). The constraint in the minimization program (97) guarantees the absence of arbitrage opportunity on the five-year yield to maturity.

9.3 Empirical Results for the AR(p) Factor-Based Term Structure Models

9.3.1 Historical Parameter Estimates

The maximum value of the mean Log-Likelihood and the values of the estimated vector of parameters $\theta_{\mathbb{P}} = (\nu, \varphi_1, \dots, \varphi_p, \sigma)'$ of the AR(p) Factor-Based Term Structure models, for $p \in \{1, \dots, 6\}$, are presented in Tables 2 and 3 [the t -values are given in parenthesis]. We also rank the models in terms of the Akaike Information Criterion (AIC).

Table 2 : AR(p) Factor-Based Term Structure models. Maximum value of the mean Log-Likelihood, AIC and parameter estimates of ν and σ . We denote with $mlogL$ the mean log-Likelihood of the AR(p) model : $mlogL = \log L(\theta_{\mathbb{P}} | x_1, \dots, x_{T-p}) / (T - p)$. (**) denotes a parameter significant at 0.05; (*) denotes a parameter significant at 0.1. The Akaike Information Criterion (AIC) is given by $2mlogL - (2k / (T - p))$, with k denoting the dimension of $\theta_{\mathbb{P}}$.

	AR(1)	AR(2)	AR(3)	AR(4)	AR(5)	AR(6)
$mlogL$	5.95657	5.95868	5.96082	5.96134	5.97224	5.97092
AIC	11.8973	11.8961	11.8950	11.8907	11.9071	11.8990
ν	0.00023 [2.6725] **	0.00021 [2.4822] **	0.00023 [2.6598] **	0.00021 [2.4761] **	0.00019 [2.1571] **	0.00019 [2.1262] **
σ^2	0.00000039 [13.7483] **	0.00000039 [13.7301] **	0.00000039 [13.7118] **	0.00000039 [13.6937] **	0.00000038 [13.6754] **	0.00000038 [13.6571] **

Table 3 : AR(p) Factor-Based Term Structure models Parameter estimates of $(\varphi_1, \dots, \varphi_p)$. (**) denotes a parameter significant at 0.05; (*) denotes a parameter significant at 0.1.

	AR(1)	AR(2)	AR(3)	AR(4)	AR(5)	AR(6)
φ_1	0.9580 ** [65.5620]	0.8798 ** [17.2393]	0.8861 ** [17.1525]	0.8912 ** [17.1688]	0.8814 ** [17.1628]	0.8806 ** [16.9714]
φ_2		0.0811 [1.5938]	0.1547 ** [2.2869]	0.1456 ** [2.0843]	0.1672 ** [2.4260]	0.1675 ** [2.3885]
φ_3			-0.0829 * [-1.6459]	-0.1372 * [-1.9204]	-0.1595 ** [-2.3048]	-0.1586 ** [-2.2623]
φ_4				0.0608 [1.1455]	-0.0790 [-1.1788]	-0.0798 [-1.1240]
φ_5					0.1557 ** [3.1048]	0.1510 ** [2.4443]
φ_6						0.0053 [0.1232]

An examination of the above displayed parameter estimates show, first of all, that the historical dynamics of the (one-month to maturity) short rate is not Markovian of order one, given that, in the AR(5) and AR(6) specifications, the parameters $(\varphi_1, \varphi_2, \varphi_3, \varphi_5)$ are always significant. Moreover, the AIC indicates these models as the preferred ones. Table 2 shows also that the constant term ν and of the volatility term σ are significant in each of the examined models.

Another indication of the key role played by the lagged values in determining, well specified models for the short rate historical dynamics, is given by the Ljung-Box test (for the Gaussian AR(p) short rate model residuals) presented in Table 4.

Table 4 : AR(p) Factor-Based Term Structure models. Ljung-Box test for model residuals. (**) denotes the null hypothesis accepted at 0.05; (*) denotes the null hypothesis accepted at 0.01.

Lags	AR(1)	AR(2)	AR(3)	AR(4)	AR(5)	AR(6)
5	13.2842 *	11.7456 *	9.2398 **	9.0748 **	0.2110 **	0.2933 **
10	20.7693 *	19.4151 *	17.4643 **	17.3474 **	8.8896 **	8.7842 **
15	33.2058	28.8190 *	25.5995 *	25.3238 *	18.3397 **	18.2022 **
20	43.9331	41.4980	39.3640	38.2511	28.2745 **	28.1044 **

We observe that, the models with small autoregressive orders are not able to pass the test for large lags, denoting a lack of these specifications in explaining the strong autocorrelation characterizing the short rate [see Table 1]. We are able to explain the short rate persistence also for large lags only

when the autoregressive order move to $p = 5$ and $p = 6$. Therefore, we find that putting several lagged values in the historical conditional mean of the factor, gives scalar models which are better specified, in explaining the short rate historical dynamics, than the discrete-time counterpart of the Ornstein-Uhlenbeck process proposed by Vasicek (1977).

9.3.2 Risk-Neutral Parameter Estimates

The minimum value of the mean nonlinear least square criterion $[S^2(\hat{\theta}_Q)/T^*]$ and the values of the estimated vector of risk-neutral parameters $\theta_Q = (\nu^*, \varphi_1^*, \dots, \varphi_p^*)$, with $p \in \{1, \dots, 6\}$, are presented in Tables 5 and 6 [the t -values are given in parenthesis]. We also rank the models in terms of the Root Mean Square Error (RMSE) and Mean Absolute Error (MAE).

Table 5 : AR(p) Factor-Based Term Structure models. Minimum value of the mean NLLS criterion, RMSE, MAE and parameter estimates of ν^* . (**) denotes a parameter significant at 0.05; (*) denotes a parameter significant at 0.1.

	AR(1)	AR(2)	AR(3)	AR(4)	AR(5)	AR(6)
$S^2(\hat{\theta}_Q)/T^*$	0.00000054	0.00000051	0.00000050	0.00000048	0.00000047	0.00000046
RMSE	0.000736	0.000716	0.000709	0.000696	0.000687	0.000679
MAE	0.000530	0.000526	0.000528	0.000524	0.000517	0.000509
ν^*	0.000110 [33.2526] **	0.000151 [22.6031] **	0.000152 [22.9266] **	0.000148 [22.9794] **	0.000148 [22.7051] **	0.000152 [22.4479] **

Table 6 : AR(p) Factor-Based Term Structure models. Parameter estimates of $(\varphi_1^*, \dots, \varphi_p^*)$. (**) denotes a parameter significant at 0.05; (*) denotes a parameter significant at 0.1.

	AR(1)	AR(2)	AR(3)	AR(4)	AR(5)	AR(6)
φ_1^*	0.9899 ** [1877]	0.5076 ** [9.6003]	0.7333 ** [14.2703]	0.7758 ** [15.4922]	0.7382 ** [14.2057]	0.7037 ** [13.3209]
φ_2^*		0.4788 ** [9.1313]	-0.0299 [-0.4132]	0.2291 ** [2.8931]	0.2947 ** [3.6124]	0.2998 ** [3.6802]
φ_3^*			0.2832 ** [7.5221]	-0.3860 ** [-5.3681]	-0.1600 ** [-2.0834]	-0.1069 [-1.3898]
φ_4^*				0.3685 ** [10.2233]	-0.1977 ** [-2.6864]	0.0123 [0.1609]
φ_5^*					0.3126 ** [8.4180]	-0.2173 ** [-2.9386]
φ_6^*						0.2961 ** [7.7697]

One may observe the significativity of all AR risk-neutral coefficients in the AR(4) and AR(5) model specifications, and the significativity of the coefficients φ_5^* and φ_6^* in the AR(6) case.

9.4 Empirical Results for the bivariate VAR(p) Factor-Based Term Structure Models

9.4.1 Historical Parameter Estimates

As in the scalar case, we present the maximum value of the mean Log-Likelihood and the values of the estimated vector of parameters $\theta_{\mathbb{P}} = [\nu', \text{vec}(\varphi)', \text{vech}(\sigma\sigma')']'$ of the bivariate VAR(p) Factor-Based Term Structure models, for an AR order $p = 1$ and $p = 2$. These results are presented in Tables 7 and 8 [the t -values are given in parenthesis]. We also rank the models in terms of the Akaike Information Criterion (AIC)¹¹.

Table 7 : VAR(p) Factor-Based Term Structure models. Maximum value of the mean Log-Likelihood, AIC and parameter estimates of (ν_1, ν_2) and $(\sigma_1^2, \sigma_{21}, \sigma_2^2)$. We denote with $mlogL$ the mean log-Likelihood of the VAR(p) model : $mlogL = \log L(\theta_{\mathbb{P}} | x_1, \dots, x_{T-p}) / (T - p)$. (**) denotes a parameter significant at 0.05; (*) denotes a parameter significant at 0.1. The Akaike Information Criterion (AIC) is given by $2mlogL - (2k / (T - p))$, with k denoting the dimension of $\theta_{\mathbb{P}}$.

	VAR(1)	VAR(2)
$mlogL$	12.6403	12.6837
AIC	25.2330	25.2984
ν_1	0.000065 [0.5856]	0.000132 [1.2262]
ν_2	0.000080 [0.8157]	0.000026 [0.2701]
σ_1^2	0.00000039 [5.94750] **	0.00000036 [6.02614] **
σ_{21}	-0.00000028 [-6.0995] **	-0.00000026 [-6.2100] **
σ_2^2	0.00000030 [7.6713] **	0.00000028 [8.0731] **

¹¹We have also estimated the historical parameters of the above mentioned bivariate VAR(p) model, for p larger than 2, but the AIC criterion has indicated the first two AR orders as the preferred ones.

Table 8 : VAR(1) and VAR(2) Factor-Based Term Structure models. Parameter estimates of (φ_1, φ_2) . (**) denotes a parameter significant at 0.05; (*) denotes a parameter significant at 0.1.

	VAR(1)		VAR(2)	
φ_1	0.9742	0.0719	1.3318	0.6207
	[59.8835] **	[2.2174] **	[15.0111] **	[7.0095] **
	0.0091	0.8769	-0.2744	0.4353
	[0.6388]	[30.7835] **	[-3.4988] **	[5.5601] **
φ_2			-0.3648	-0.5762
			[-3.6117] **	[-5.8201] **
			0.2893	0.4642
			[3.2397] **	[5.3020] **

If we consider the parameter estimates of Tables 7 and 8, we observe that the joint historical dynamics of short rate and spread is not Markovian of order one, given that, in the VAR(2) specification, the parameters in the second autoregressive matrix φ_2 are significantly different from zero. Moreover, the AIC indicates this model as the preferred one. Table 7 shows also that the constant term $(\nu_1, \nu_2)'$ is not significative for both AR orders.

9.4.2 Risk-Neutral Parameter Estimates

We present the minimum value of the mean nonlinear least square criterion $[S^2(\hat{\theta}_Q)/T^*]$ and the values of the estimated vector of risk-neutral parameters $\theta_Q = [(\nu^*)', \text{vec}(\varphi^*)']'$, for the bivariate VAR(1) and VAR(2) Factor-Based Term Structure models in Tables 9 and 10 [the t -values are given in parenthesis].

Table 9 : AR(p) Factor-Based Term Structure models. Minimum value of the mean NLLS criterion, RMSE, MAE and parameter estimates of (ν_1^*, ν_2^*) . (**) denotes a parameter significant at 0.05; (*) denotes a parameter significant at 0.1.

	VAR(1)	VAR(2)
$S^2(\hat{\theta}_Q)/T^*$	0.00000009	0.00000008
RMSE	0.000297	0.000283
MAE	0.000208	0.000198
ν_1^*	-0.000058	-0.000055
	[-6.6459] **	[-4.9423] **
ν_2^*	0.000072	0.000071
	[5.7860] **	[4.5783] **

Table 10 : VAR(1) and VAR(2) Factor-Based Term Structure models. Parameter estimates of $(\varphi_1^*, \varphi_2^*)$. (**) denotes a parameter significant at 0.05; (*) denotes a parameter significant at 0.1.

	VAR(1)		VAR(2)	
φ_1^*	1.0131	0.1105	1.3154	0.6020
	[805.8869] **	[34.5743] **	[28.4716] **	[9.5120] **
	-0.0156	0.9072	-0.2528	0.4142
	[-8.6611] **	[203.2978] **	[-3.5778] **	[4.2509] **
φ_2^*			-0.3004	-0.4890
			[-6.5177] **	[-7.8923] **
			0.2342	0.4839
			[3.3244] **	[5.0769] **

We find that, also in this bivariate risk-neutral (pricing) framework, the lagged values of the short rate and spread play an important role in the model specification. Moreover, one may observe the significativity of all risk-neutral AR coefficients in the VAR(2) specification.

Consequently, the goodness-of-fit of the VAR(2) Factor-Based Term Structure Model, outperform the results of the VAR(1) specification. In other words, a VAR(2) specification for the historical and risk-neutral dynamics of the factor driving term structure shapes, lead to propose a bivariate term structure model which is able to fit yields to maturity better than the VAR(1) (bivariate Vasicek) specification.

10 CONCLUSIONS

The purpose of this paper was to generalize the Gaussian family of discrete-time affine term structure models by the introduction of lags in the historical and risk-neutral factor dynamics, and in the specification of the stochastic risk correction coefficients.

We have studied the Gaussian AR(p) and VAR(p) Factor-Based Term Structure models, we have verified the important role played by the autoregressive order in the replication of term structure shapes coherent with observations. Moreover, several characterizations of the yield and short-term forward rate processes, under the risk-neutral (Moving Average representation) and S -forward probability, are proposed. We have also studied the problem of exact fitting of the initial term structure by means of the Extended AR(p) approach. These results are given for an exogenous factor, and for an endogenous setting where (x_t) is a vector of several yields.

In the last section of the paper, using monthly observations on the U.S. term structure of interest rates, from June 1964 to December 1995, we have estimated several endogenous factor models, with scalar and bivariate factors. We have verified that, the goodness-of-fit of the estimated $AR(p)$ and $VAR(p)$ Factor-Based Term Structure models improve when the autoregressive order increases, under both the historical and risk-neutral setting.

Appendix 1
Proof of Proposition 1

$$\begin{aligned}
B(t, h) &= \exp(c'_h X_t + d_h) \\
&= E_t[M_{t,t+1} \cdots M_{t+H-1,t+H}] \\
&= E_t[M_{t,t+1} B(t+1, h-1)] \\
&= \exp\left[-\beta - \alpha' X_t - \frac{1}{2} \Gamma_t^2 + d_{h-1}\right] \times E_t[\exp(\Gamma_t \varepsilon_{t+1} + c'_{h-1} X_{t+1})] \\
&= \exp\left[-\beta - \alpha' X_t - \frac{1}{2} \Gamma_t^2 + d_{h-1} + c'_{h-1} (\Phi X_t + \tilde{\nu})\right] \\
&\quad \times E_t[\exp(\Gamma_t + \sigma c_{1,h-1}) \varepsilon_{t+1}] \\
&= \exp\left[(-\alpha + \Phi' c_{h-1} + c_{1,h-1} \sigma \gamma)' X_t \right. \\
&\quad \left. + (-\beta + c_{1,h-1} \nu + \frac{1}{2} c_{1,h-1}^2 \sigma^2 + \gamma_o c_{1,h-1} \sigma + d_{h-1})\right], \\
\end{aligned} \tag{A.1}$$

and by identifying the coefficients we find the recursive relation presented in Proposition 1. \square

Appendix 2
Proof of Proposition 2

$$\begin{aligned}
\rho(t+1, T) &= \log[B^*(t+1, T)] - \log[B^*(t, T)] \\
&= c'_{T-t-1} X_{t+1} + d_{T-t-1} - c'_{T-t} X_t - d_{T-t} \\
&= c'_{T-t-1} [X_{t+1} - \Phi X_t - \tilde{\nu}] + (\beta + \alpha' X_t) \\
&\quad - \sigma c_{1,T-t-1} (\gamma_o + \gamma' X_t) - \frac{1}{2} c_{1,T-t-1}^2 \sigma^2 \\
&= (c_{1,T-t-1} \sigma) \varepsilon_{t+1} + (\beta + \alpha' X_t) - \sigma c_{1,T-t-1} (\gamma_o + \gamma' X_t) \\
&\quad - \frac{1}{2} c_{1,T-t-1}^2 \sigma^2. \\
\end{aligned} \tag{A.2}$$

Now, we have that, under the absence of arbitrage $r_{t+1} = (\beta + \alpha' X_t)$ and, consequently, the Proposition 2 is proved. \square

Appendix 3

Proof of Proposition 4

$$\begin{aligned}
\rho(t+1, T) &= \log [B^*(t+1, T)] - \log [B^*(t, T)] \\
&= c'_{T-t-1} X_{t+1} + d_{T-t-1} - c'_{T-t} X_t - d_{T-t} \\
&= c'_{T-t-1} X_{t+1} - (-\alpha' + c'_{T-t-1} \Phi^*) X_t + \beta - c_{1, T-t-1} \nu^* - \frac{1}{2} c_{1, T-t-1}^2 \sigma^2 \\
&= c'_{T-t-1} [X_{t+1} - \Phi^* X_t - \tilde{\nu}^*] + r_{t+1} - \frac{1}{2} c_{1, T-t-1}^2 \sigma^2 \\
&\stackrel{\mathbb{Q}}{=} r_{t+1} - \frac{1}{2} \omega(t+1, T)^2 - \omega(t+1, T) \eta_{t+1}. \quad \square
\end{aligned} \tag{A.3}$$

Appendix 4

Proof of Proposition 9

The conditional risk-neutral mean of the one-period forward rate increment

$$\Delta f(t, T) \stackrel{\mathbb{Q}}{=} \mu^{\mathbb{Q}}(t+1, T) + \sigma^{\mathbb{Q}}(t+1, T) \eta_{t+1}$$

can be written in the following way:

$$\begin{aligned}
\mu^{\mathbb{Q}}(t+1, T) &= \frac{1}{2} [\omega(t+1, T+1)^2 - \omega(t+1, T)^2] \\
&= \frac{1}{2} \sigma^{\mathbb{Q}}(t+1, T) [\omega(t+1, T+1) + \omega(t+1, T)],
\end{aligned} \tag{A.4}$$

and, given that $\omega(i, i) = 0$, we can write

$$\omega(t+1, T+1) = \sum_{\tau=t+1}^T \sigma^{\mathbb{Q}}(t+1, \tau) \quad \forall t \leq T;$$

consequently, relation (A.4) can be represented in the following way:

$$\mu^{\mathbb{Q}}(t+1, T) = \frac{1}{2} \sigma^{\mathbb{Q}}(t+1, T) \left[\sum_{\tau=t+1}^T \sigma^{\mathbb{Q}}(t+1, \tau) + \sum_{\tau=t+1}^{T-1} \sigma^{\mathbb{Q}}(t+1, \tau) \right],$$

and Proposition 9 is proved. \square

Appendix 5

Historical MA and HJM representations

A.5.1 MA REPRESENTATION OF THE HISTORICAL YIELD PROCESS

In this appendix we want to determine the conditions under which the joint historical dynamics of the yield processes $R^*(\cdot, T) = [R^*(t, T), 0 \leq t \leq T]$, with $R^*(t, T) = -\frac{1}{T-t} \log B^*(t, T)$ and $R^*(t, T) = R(t, T-t)$, can be represented under a Moving Average form. The derivation of the joint dynamics of the processes $R^*(\cdot, T)$ is based on the representation of the log-price stochastic process in terms of the conditional variances of $\rho = [\rho(t, T), 0 \leq t \leq T]$, where $\rho(t+1, T) = \log [B^*(t+1, T)] - \log [B^*(t, T)]$. Starting from relation (12) we can write the log-price as :

$$\begin{aligned}
 \log[B^*(t, T)] &= \sum_{j=1}^t \rho(j, T) + \log[B^*(0, T)] \\
 &= -\sum_{j=1}^t \omega(j, T)\varepsilon_j + \sum_{j=1}^t r_j \\
 &\quad + \sum_{j=1}^t \omega(j, T)(\gamma_o + \gamma' X_{j-1}) \\
 &\quad - \frac{1}{2} \sum_{j=1}^t \omega(j, T)^2 + \log[B^*(0, T)],
 \end{aligned} \tag{A.5}$$

and, consequently, we can give the following proposition.

Proposition A.1 : For every fixed maturity T , the zero-coupon bond price process $B^*(\cdot, T) = [B^*(t, T), 0 \leq t \leq T]$, under the historical probability \mathbb{P} , has the following specification:

$$\begin{aligned}
 B^*(t, T) &= B^*(0, T) \exp \left[\sum_{j=1}^t \left[r_j - \frac{1}{2} \omega(j, T)^2 + \omega(j, T) \Gamma_{j-1} \right. \right. \\
 &\quad \left. \left. - \sum_{j=1}^t \omega(j, T) \varepsilon_j \right] \right].
 \end{aligned} \tag{A.6}$$

If we put $T = t$ in (A.5), we find a relation for the sum of the short-rates:

$$\begin{aligned}
 \sum_{j=1}^t r_j &= \sum_{j=1}^t \omega(j, t) \varepsilon_j - \sum_{j=1}^t \omega(j, t) (\gamma_o + \gamma' X_{j-1}) \\
 &\quad + \frac{1}{2} \sum_{j=1}^t \omega(j, t)^2 - \log[B^*(0, t)]
 \end{aligned} \tag{A.7}$$

that we can substitute in (A.5) to find, for every date t and maturity $T > t$:

$$\begin{aligned} \log[B^*(t, T)] &= \log \left[\frac{B^*(0, T)}{B^*(0, t)} \right] - \sum_{j=1}^t [\omega(j, T) - \omega(j, t)] \varepsilon_j \\ &\quad + \sum_{j=1}^t [\omega(j, T) - \omega(j, t)] (\gamma_o + \gamma' X_{j-1}) \quad (\text{A.8}) \\ &\quad - \frac{1}{2} \sum_{j=1}^t [\omega(j, T)^2 - \omega(j, t)^2]. \end{aligned}$$

Conditionally to the information x_0 , and for every maturity T , the processes $\{R^*(\cdot, T)\}$ are, under the historical probability \mathbb{P} , joint MA processes with time-varying coefficients (driven by the same white noise ε_t) only if we assume $\gamma = 0$.

A.5.2 THE HJM REPRESENTATION OF HISTORICAL SHORT-TERM FORWARD RATES

Starting from the one-period forward rate $f(t, T) = \log[B^*(t, T)] - \log[B^*(t, T+1)]$, if we use relation (A.8) we have :

Proposition A.2 : Under the historical probability \mathbb{P} , and for any fixed maturity T , the forward rate $f(t, T)$ satisfies the following relation:

$$\begin{aligned} f(t, T) &= f(0, T) + \sum_{j=1}^t [\omega(j, T+1) - \omega(j, T)] \varepsilon_j \\ &\quad - \sum_{j=1}^t [\omega(j, T+1) - \omega(j, T)] (\gamma_o + \gamma' X_{j-1}) \quad (\text{A.9}) \\ &\quad + \frac{1}{2} \sum_{j=1}^t [\omega(j, T+1)^2 - \omega(j, T)^2], \end{aligned}$$

where $f(0, T) = \log \left[\frac{B^*(0, T)}{B^*(0, T+1)} \right]$.

Corollary A.1 : The one-period forward rate increment $\Delta f(t, T) = f(t+1, T) - f(t, T)$, is given by :

$$\Delta f(t, T) = \mu^{\mathbb{P}}(t+1, T) + \sigma^{\mathbb{P}}(t+1, T) \varepsilon_{t+1}, \quad (\text{A.10})$$

where

$$\begin{aligned}
\sigma^{\mathbb{P}}(t+1, T) &= [\omega(t+1, T+1) - \omega(t+1, T)] = \sigma^{\mathbb{Q}}(t+1, T) \\
\mu^{\mathbb{P}}(t+1, T) &= \frac{1}{2}[\omega(t+1, T+1)^2 - \omega(t+1, T)^2] - \sigma^{\mathbb{P}}(t+1, T)\Gamma_t \\
&= \mu^{\mathbb{Q}}(t+1, T) - \sigma^{\mathbb{P}}(t+1, T)\Gamma_t \\
&= \sigma^{\mathbb{Q}}(t+1, T) \left[\frac{\sum_{\tau=t+1}^T \sigma^{\mathbb{Q}}(t+1, \tau) + \sum_{\tau=t+1}^{T-1} \sigma^{\mathbb{Q}}(t+1, \tau)}{2} \right] \\
&\quad - \sigma^{\mathbb{Q}}(t+1, T)\Gamma_t
\end{aligned} \tag{A.11}$$

We observe that, as for the historical yield processes $\{R^*(\cdot, T)\}$, conditionally to the information \underline{x}_0 , and for every maturity T , the processes $\{f(\cdot, T)\}$ are, under the probability \mathbb{P} , joint MA processes (driven by the same white noise ε_t) only in the case $\Gamma_t = \gamma_o$.

Appendix 6

Proof of Proposition 11

Denoting $B_e^*(0, T)$ the zero coupon price in the extended model, we have :

$$\begin{aligned}
&B_e^*(0, T) \\
&= E_0^{\mathbb{Q}}[\exp(-r_1 - \dots - r_T)] \\
&= \exp\left[-\sum_{t=0}^{T-1} \beta(t)\right] E_0^{\mathbb{Q}}[\exp(-\alpha'X_0 - \dots - \alpha'X_{T-1})] \\
&= \exp\left[-\sum_{t=0}^{T-1} \beta(t)\right] E_0\left[\frac{M_{0,1} \cdot \dots \cdot M_{T-1,T}}{E_t(M_{0,1}) \cdot \dots \cdot E_{T-1}(M_{T-1,T})} \exp(-\alpha'X_0 - \dots - \alpha'X_{T-1})\right] \\
&= \exp\left[-\sum_{t=0}^{T-1} \beta(t)\right] E_0\left[M_{0,1}^o \cdot \dots \cdot M_{T-1,T}^o\right] \\
&= \exp\left[-\sum_{t=0}^{T-1} \beta(t)\right] B_o^*(0, T),
\end{aligned}$$

where $B_o^*(0, T) = E_0\left[M_{0,1}^o \cdot \dots \cdot M_{T-1,T}^o\right]$ is the price at date $t = 0$ of the zero-coupon bond maturing in T and obtained with the pricing kernel (4)

in which we assume $\beta = 0$. The result in Proposition 11 follows from $R_e^*(0, T) = -\log(B_e^*(0, T))/T$. \square

Appendix 7

Proof of Proposition 12

We are searching for the values of $\beta(t)$ such that $R_e^*(0, t) = R_M^*(0, t)$ for all t . From the result of Proposition 11, we can equivalently rewrite the above matching condition in the following way :

$$\begin{aligned} f_M(0, t) &= f_e(0, t) \\ &= \log B_e(0, t) - \log B_e(0, t + 1) \\ &= \beta(t) + f_o(0, t), \end{aligned}$$

and consequently, for each date t , we can always choose the matching function $\beta(t) = f_M(0, t) - f_o(0, t)$. Now, from formula (9), with $\beta = 0$, we obtain:

$$\begin{aligned} \beta(t) &= f_M(0, t) - [(c_t - c_{t+1})'X_0 + (d_t - d_{t+1})] \\ &= f_M(0, t) + (\Phi^{*'}c_t - c_t - \alpha)'X_0 + (c_{1,t}\nu^* + \frac{1}{2}c_{1,t}^2\sigma^2) \\ &= f_M(0, t) + \mu_{o,t} + \mu_t'X_0, \end{aligned}$$

with

$$\begin{aligned} \mu_{o,t} &= c_{1,t}\nu^* + \frac{1}{2}c_{1,t}^2\sigma^2 \\ \mu_t &= \left(\Phi^{*'}c_t - c_t - \alpha\right)'X_0 \end{aligned}$$

and the result in Proposition 12 is proved. \square

Appendix 8

Proof of Proposition 13

The S -forward Laplace transform of x_{t+1} , conditionally to $I_t = (\underline{x}_t)$, can be written in the following way:

$$\begin{aligned} &E_t^{\mathbb{Q}^S}[\exp(ux_{t+1})] \\ &= \frac{1}{B^*(t, S)} E_t^{\mathbb{Q}}[\exp(-r_{t+1} - \dots - r_S + ux_{t+1})]. \end{aligned} \tag{A.12}$$

From relation (28), we have that, under the risk-neutral measure \mathbb{Q} , the sum of short-term rates in the above formula can be written as:

$$\begin{aligned} \sum_{j=t+1}^S r_j &= \sum_{j=1}^S r_j - \sum_{j=1}^t r_j \\ &\stackrel{\mathbb{Q}}{=} \sum_{j=1}^S \omega(j, S) \eta_j - \sum_{j=1}^t \omega(j, t) \eta_j \\ &\quad + \frac{1}{2} \left[\sum_{j=1}^S \omega(j, S)^2 - \sum_{j=1}^t \omega(j, t)^2 \right] + \log \left[\frac{B^*(0, t)}{B^*(0, S)} \right] \end{aligned}$$

and, consequently, we get:

$$\begin{aligned} &E_t^{\mathbb{Q}^S} [\exp(ux_{t+1})] \\ &= \frac{\exp \left[-\frac{1}{2} \left[\sum_{j=1}^S \omega(j, S)^2 - \sum_{j=1}^t \omega(j, t)^2 \right] - \sum_{j=1}^t [\omega(j, S) - \omega(j, t)] \eta_j - \log \left[\frac{B^*(0, t)}{B^*(0, S)} \right] \right]}{B^*(t, S)} \times \\ &E_t^{\mathbb{Q}} \left[\exp \left(- \sum_{j=t+2}^S \omega(j, S) \eta_j - \omega(t+1, S) \eta_{t+1} + u[\nu^* + \varphi^* X_t + \sigma^* \eta_{t+1}] \right) \right] \\ &= k_{t, S} E_t^{\mathbb{Q}} \left[\exp \left(- \sum_{j=t+2}^S \omega(j, S) \eta_j \right) \right] \times \\ &E_t^{\mathbb{Q}} \left[\exp \left(u(\nu^* + \varphi^* X_t) + (u\sigma^* - \omega(t+1, S)) \eta_{t+1} \right) \right] \\ &= k'_{t, S} \exp \left[u \left[\nu^* + \varphi^* X_t - \sigma^* \omega(t+1, S) \right] + \frac{1}{2} u^2 \sigma^{*2} \right]; \end{aligned} \tag{A.13}$$

now, using (29) we have that

$$\begin{aligned} k_{t, S} &= \frac{\exp \left[-\frac{1}{2} \left[\sum_{j=1}^S \omega(j, S)^2 - \sum_{j=1}^t \omega(j, t)^2 \right] - \sum_{j=1}^t [\omega(j, S) - \omega(j, t)] \eta_j - \log \left[\frac{B^*(0, t)}{B^*(0, S)} \right] \right]}{B^*(t, S)} \\ &= \exp \left[-\frac{1}{2} \sum_{j=t+1}^S \omega(j, S)^2 \right] \end{aligned}$$

and that

$$k'_{t,S} = k_{t,S} \exp \left[\frac{1}{2} \sum_{j=t+1}^S \omega(j, S)^2 \right] = 1.$$

Consequently, Proposition 13 is proved. \square

Appendix 9

Proof of Proposition 14

$$\begin{aligned}
\rho(t+1, T) &= \log [B^*(t+1, T)] - \log [B^*(t, T)] \\
&= c'_{T-t-1} X_{t+1} + d_{T-t-1} - c'_{T-t} X_t - d_{T-t} \\
&= c'_{T-t-1} [X_{t+1} - \Phi^* X_t - \tilde{\nu}^*] + r_{t+1} - \frac{1}{2} c_{1, T-t-1}^2 \sigma^{*2} \\
&\stackrel{\mathbb{Q}_S}{=} c'_{T-t-1} \left[\sigma^* \left(\tilde{\xi}_{t+1} - \omega(t+1, S) e_1 \right) \right] + r_{t+1} - \frac{1}{2} c_{1, T-t-1}^2 \sigma^{*2} \\
&\stackrel{\mathbb{Q}_S}{=} r_{t+1} + \omega(t+1, T) \omega(t+1, S) \\
&\quad - \frac{1}{2} \omega(t+1, T)^2 - \omega(t+1, T) \xi_{t+1}. \quad \square
\end{aligned} \tag{A.14}$$

Appendix 10

Proof of Proposition 15

Let us consider the S -forward probability \mathbb{Q}_S determined by the numeraire $N_t = \frac{B^*(t, S)}{B^*(0, S)}$, with $t \leq T \leq S$. We have seen in Proposition 14 that :

$$\begin{aligned}
\rho(t+1, T) &\stackrel{\mathbb{Q}_S}{=} r_{t+1} - \frac{1}{2} \omega(t+1, T)^2 + \omega(t+1, T) \omega(t+1, S) \\
&\quad - \omega(t+1, T) \xi_{t+1},
\end{aligned} \tag{A.15}$$

and, consequently, the log-price zero-coupon bond process is given by :

$$\begin{aligned}
\log[B^*(t, T)] &\stackrel{\mathbb{Q}_S}{=} - \sum_{j=1}^t \omega(j, T) \xi_j + \sum_{j=1}^t \omega(j, T) \omega(j, S) + \sum_{j=1}^t r_j \\
&\quad - \frac{1}{2} \sum_{j=1}^t \omega(j, T)^2 + \log[B^*(0, T)].
\end{aligned} \tag{A.16}$$

If we put $T = t$ in (A.16), we find the following relation for the sum of the short-rates :

$$\begin{aligned} \sum_{j=1}^t r_j &\stackrel{\mathbb{Q}_S}{=} \sum_{j=1}^t \omega(j, t) \xi_j - \sum_{j=1}^t \omega(j, t) \omega(j, S) \\ &\quad + \frac{1}{2} \sum_{j=1}^t \omega(j, t)^2 - \log[B^*(0, t)], \end{aligned} \tag{A.17}$$

that we can substitute in (A.16) to find :

$$\begin{aligned} \log[B^*(t, T)] &\stackrel{\mathbb{Q}_S}{=} \log \left[\frac{B^*(0, T)}{B^*(0, t)} \right] + \sum_{j=1}^t [\omega(j, t) - \omega(j, T)] \xi_j \\ &\quad - \sum_{j=1}^t \omega(j, S) [\omega(j, t) - \omega(j, T)] \\ &\quad + \frac{1}{2} \sum_{j=1}^t [\omega(j, t)^2 - \omega(j, T)^2]. \end{aligned} \quad \square \tag{A.18}$$

Appendix 11

Proof of Proposition 18

We write relation (11) in the following way:

$$R(t, h) = \mathbf{C}_h(L)x_t + \delta_h,$$

where $\mathbf{C}_h(L) = -(c_{1,h} + c_{2,h}L + \dots + c_{p,h}L^{p-1})/h$ is the $(p-1)^{th}$ degree polynomial in the backward shift operator L , and where $\delta_h = -(d_h/h)$. Now, if we apply on the right-hand and left-hand side of this relation the operator $\Psi(L) = (1 - \varphi_1 L - \dots - \varphi_p L^p)$ operating on t , we can write :

$$\begin{aligned} \Psi(L)R(t, h) &= \mathbf{C}_h(L)\Psi(L)x_t + \Psi(1)\delta_h \\ &= \mathbf{C}_h(L)[\nu + \sigma\varepsilon_t] + \Psi(1)\delta_h \\ &= \sigma\mathbf{C}_h(L)\varepsilon_t + \mathbf{C}_h(1)\nu + \Psi(1)\delta_h. \end{aligned} \quad \square \tag{A.19}$$

Appendix 12
Proof of Proposition 22

$$\begin{aligned}
B(t, h) &= \exp(c'_h X_t + d_h) \\
&= E_t[M_{t,t+1} \cdots M_{t+H-1,t+H}] \\
&= E_t[M_{t,t+1} B(t+1, h-1)] \\
&= \exp\left[-\beta - \alpha' X_t - \frac{1}{2} \Gamma'_t \Gamma_t + d_{h-1}\right] \times E_t[\exp(\Gamma'_t \varepsilon_{t+1} + c'_{h-1} X_{t+1})] \\
&= \exp\left[-\beta - \alpha' X_t - \frac{1}{2} \Gamma'_t \Gamma_t + d_{h-1} + c'_{h-1} \Phi X_t + c'_{1,h-1} \nu\right] \\
&\quad \times E_t[\exp(\Gamma'_t + \sigma' c_{1,h-1})' \varepsilon_{t+1}] \\
&= \exp\left[\left(-\alpha + \Phi' c_{h-1} + (\sigma\gamma)' c_{1,h-1}\right)' X_t\right. \\
&\quad \left.+ \left(-\beta + c'_{1,h-1}(\nu + \sigma\gamma) + \frac{1}{2} c'_{1,h-1} \sigma \sigma' c_{1,h-1} + d_{h-1}\right)\right], \\
&\hspace{15em} \text{(A.20)}
\end{aligned}$$

and by identifying the coefficients we find the recursive relation presented in Proposition 22. \square

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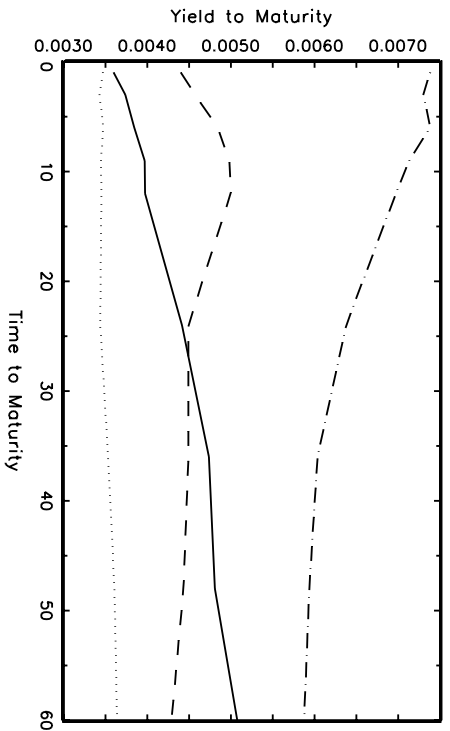


FIGURE A – Observed Yield Curves :
 Humpshaped (dashed), Flat (dotted):
 Monotone Decreasing (dots and dashes), Monotone Increasing (solid):

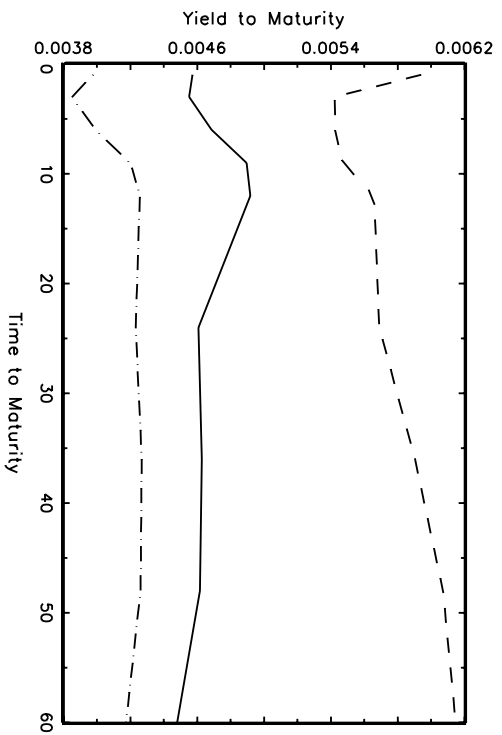


FIGURE B – Observed Yield Curves : J-SHAPES
 J-shaped (dashed; dots and dashes), J-Humpshaped (solid):

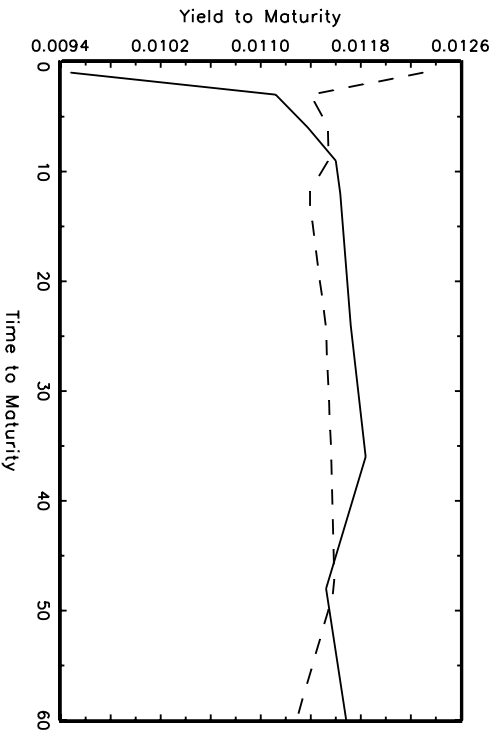


FIGURE C – Observed Yield Curves : L-SHAPES
 L-shaped (dashed); Inverted L-shaped (solid):

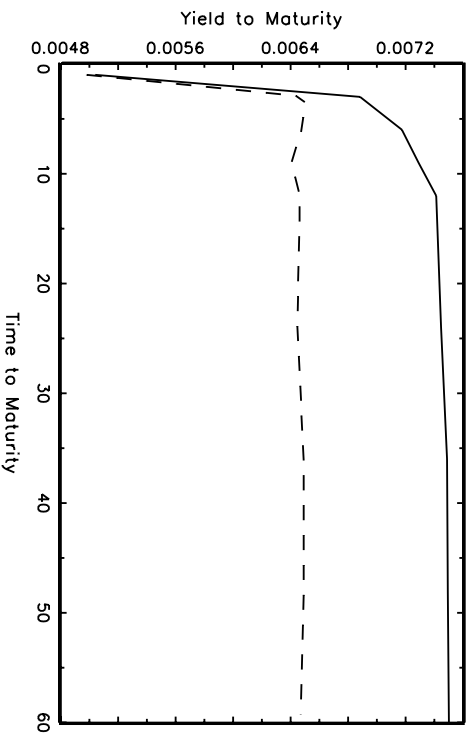


FIGURE D – Observed Yield Curves : L-SHAPES
 Inverted L-shaped (dashed and solid):

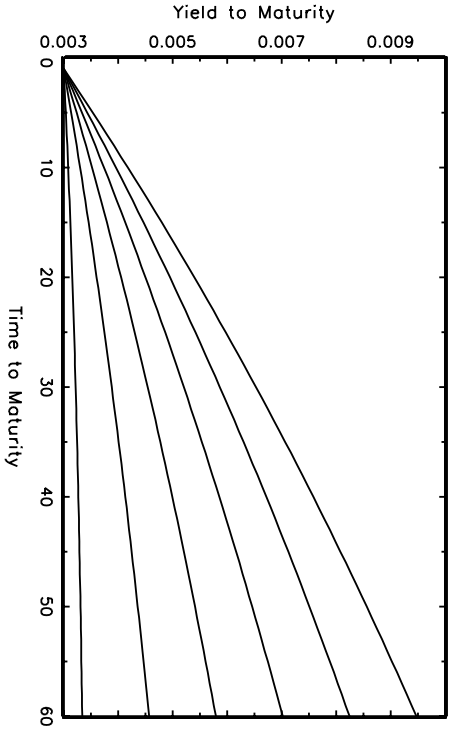


FIGURE 1 – Gaussian AR(1) model:
 $\phi_{\text{star}} = 0.99$, $r = 0.003$; $\sigma^2 = 0.00000039$;
 $\nu_{\text{star}} = 0.00005$ (bottom curve) to 0.00030 (top curve);

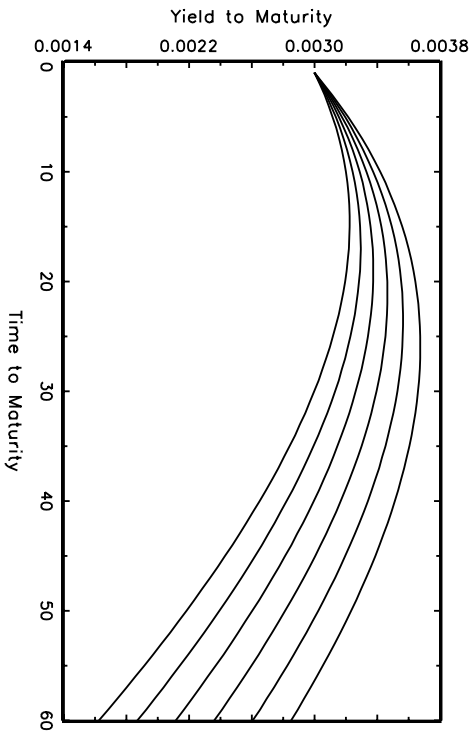


FIGURE 2 – Gaussian AR(1) model:
 $\phi_{\text{star}} = 0.99$, $r = 0.003$; $\sigma^2 = 0.000008$;
 $\nu_{\text{star}} = 0.00010$ (bottom curve) to 0.00015 (top curve);

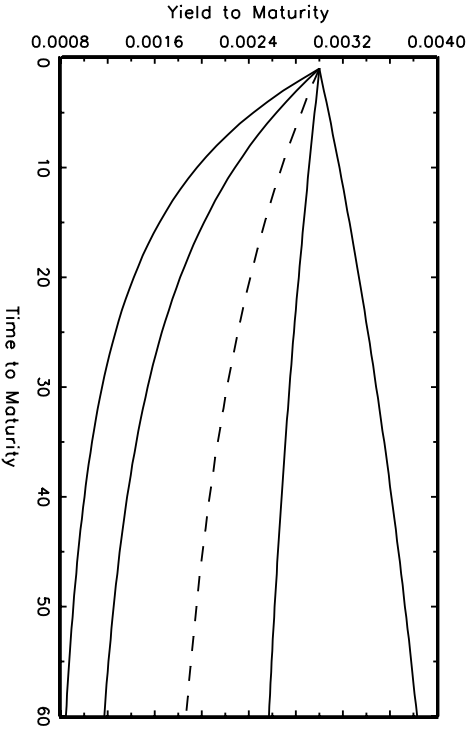


FIGURE 3 – Gaussian AR(1) model:
 $\phi = 0.95$; $\phi_{\text{star}} = 0.87$ (bottom curve) to 0.99 (top curve);
 $r = 0.003$; $\nu_{\text{star}} = 0.00007$; $\sigma^2 = 0.00000039$;

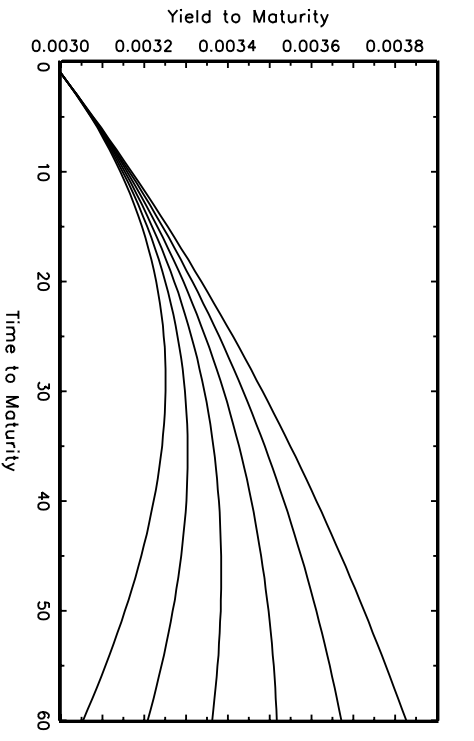


FIGURE 4 – Gaussian AR(1) model:
 $\phi_{\text{star}} = 0.99$, $r = 0.003$; $\nu_{\text{star}} = 0.00007$;
 $\sigma^2 = 0.0000004$ (top curve) to 0.0000024 (bottom curve);

FIGURE 5 – Gaussian AR(2):
 $\phi_1_star = 0.74$, $\phi_2_star = 0.25$;
 $r_{-}(t+1) = 0.0036$, $r_{-}(t) = 0.0030$; $\sigma^2 = 0.00000039$;
 $\nu_star = 0.00005$ (bottom curve) to 0.00030 (top curve)

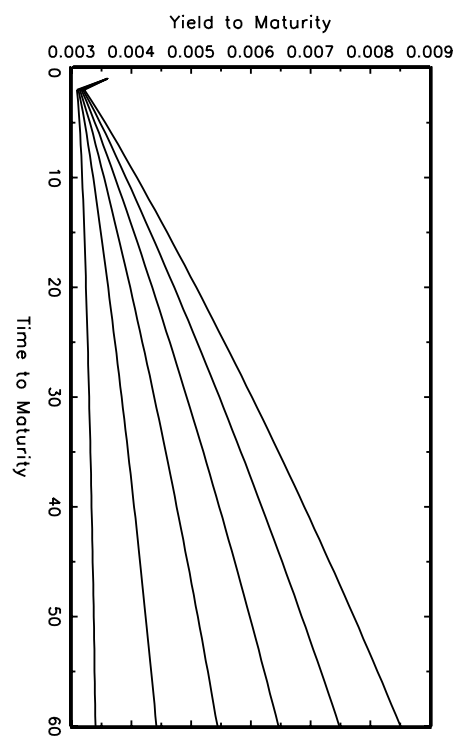


FIGURE 7 – Gaussian AR(2):
 $\phi_1_star = 0.74$, $\phi_2_star = 0.25$;
 $r_{-}(t+1) = 0.0032$, $r_{-}(t) = 0.0036$; $\sigma^2 = 0.000008$;
 $\nu_star = 0.00010$ (bottom curve) to 0.00015 (top curve)

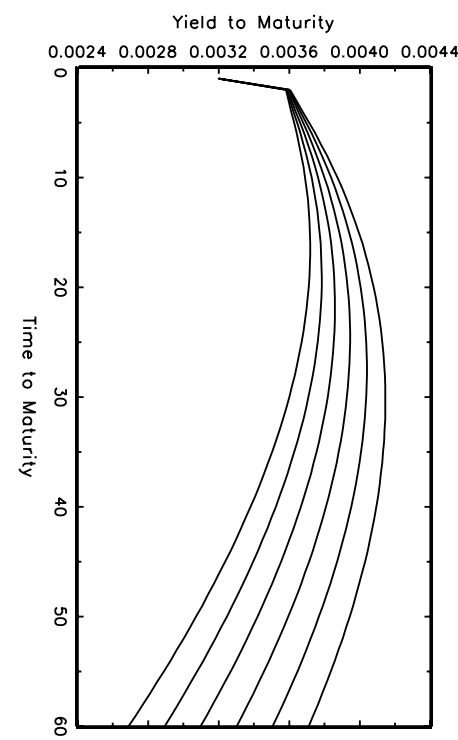


FIGURE 6 – Gaussian AR(2):
 $\phi_1_star = 0.74$, $\nu_star = 0.00008$;
 $\sigma^2 = 0.00000039$, $r_{-}(t+1) = 0.0030$, $r_{-}(t) = 0.0036$;
 $\phi_2_star = 0.04$ (bottom curve) to 0.24 (top curve);

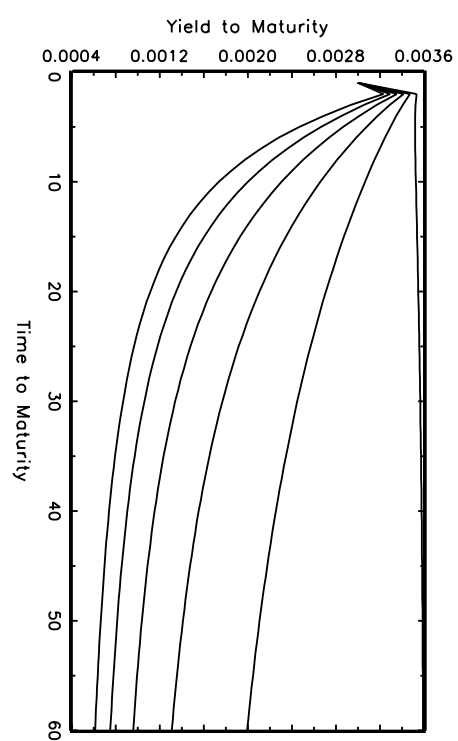
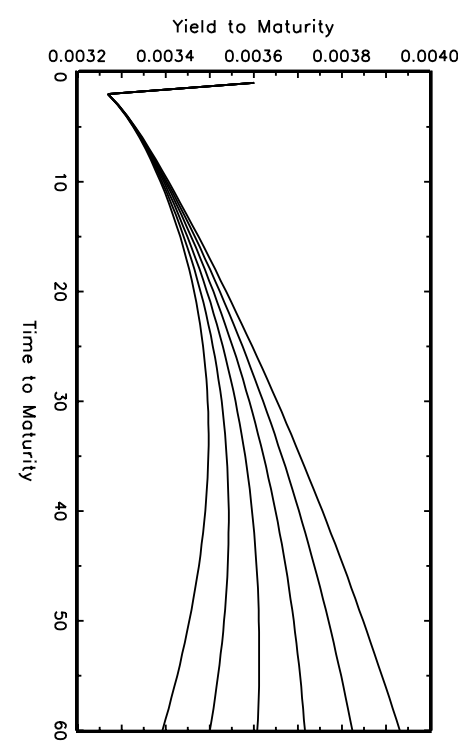


FIGURE 8 – Gaussian AR(2):
 $\phi_1_star = 0.74$, $\phi_2_star = 0.25$;
 $r_{-}(t+1) = 0.0036$, $r_{-}(t) = 0.0032$; $\nu_star = 0.00007$;
 $\sigma^2 = 0.0000004$ (top curve) to 0.0000024 (bottom curve);



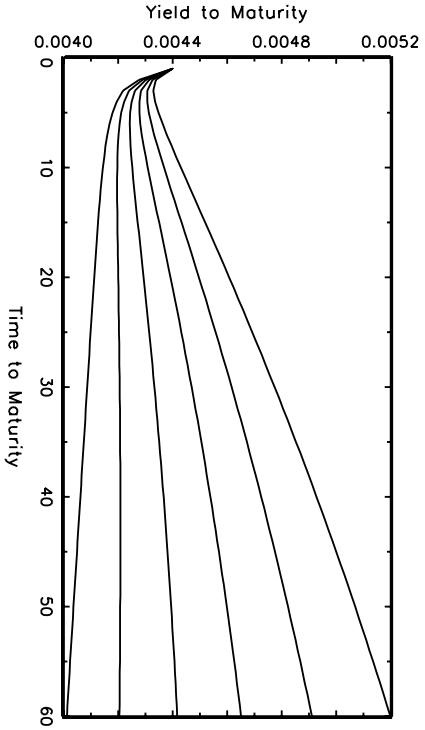


FIGURE 9 – AR(3):
 $\nu^* = 0.00008$, $\sigma^2 = 0.00000039$;
 $\phi_1^* = 0.79$ (bottom) to 0.865 (top);
 $\phi_2^* = 0.12$ (bottom) to 0.1075 (top);
 $\phi_3^* = 0.07$ (bottom) to 0.02 (top)

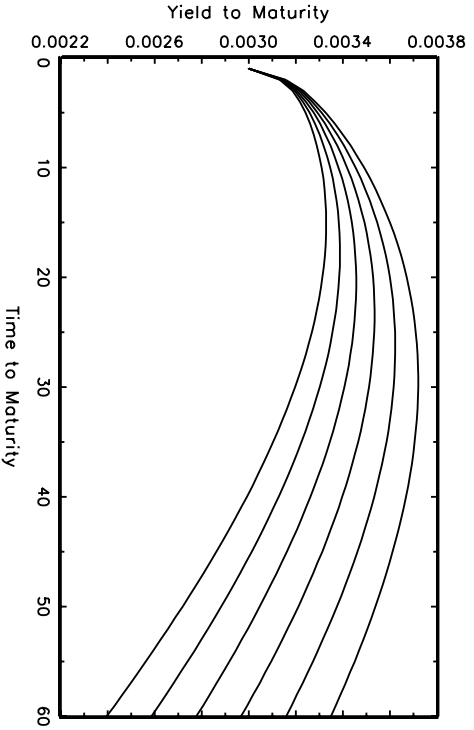


FIGURE 11 – AR(3):
 $\sigma^2 = 0.000008$; $\nu^* = 0.00010$ (bottom) to 0.00015 (top);
 $\phi_1^* = 0.766$, $\phi_2^* = 0.17$, $\phi_3^* = 0.05$;

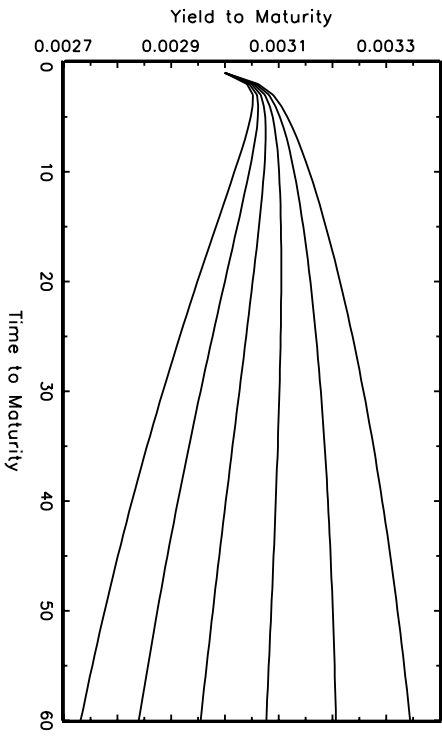


FIGURE 10 – AR(3):
 $\phi_1^* = 0.79$, $\phi_3^* = 0.07$, $\nu^* = 0.00008$;
 $\sigma^2 = 0.00000039$; $\phi_2^* = 0.12$ (top) to 0.1075 (bottom):

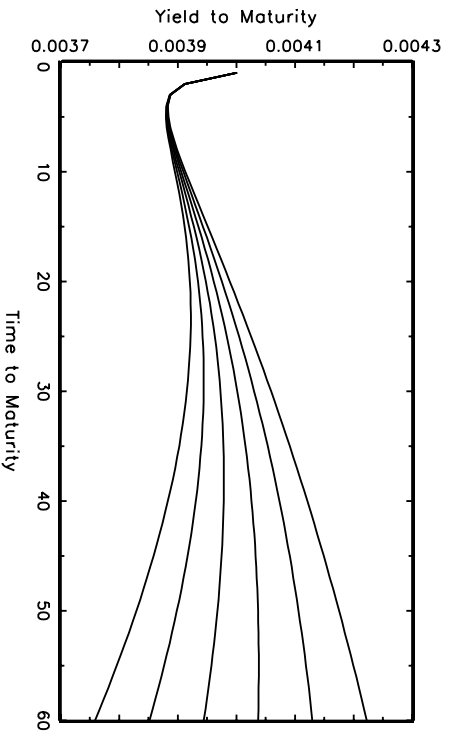


FIGURE 12 – Gaussian AR(3):
 $\phi_1^* = 0.766$, $\phi_2^* = 0.17$;
 $\phi_3^* = 0.05$; $\nu^* = 0.00008$;
 $\sigma^2 = 0.0000004$ (top) to 0.0000024 (bottom):

FIGURE 13 – AR(4):
 $\nu^* = 0.00009$, $\sigma^2 = 0.00000039$;
 $\phi_1^* = 0.73$, $\phi_2^* = 0.26$, $\phi_3^* = -0.07$;
 $\phi_4^* = 0.055$ (bottom) to 0.08 (top)

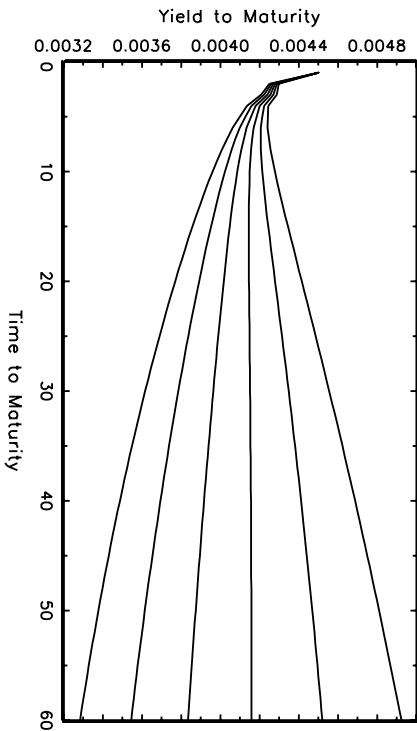


FIGURE 15 – AR(4):
 $\phi_1^* = 0.73$, $\phi_2^* = 0.25$, $\phi_3^* = -0.07$;
 $\phi_4^* = 0.08$, $\sigma^2 = 0.000008$;
 $\nu^* = 0.00010$ (bottom) to 0.00015 (top);

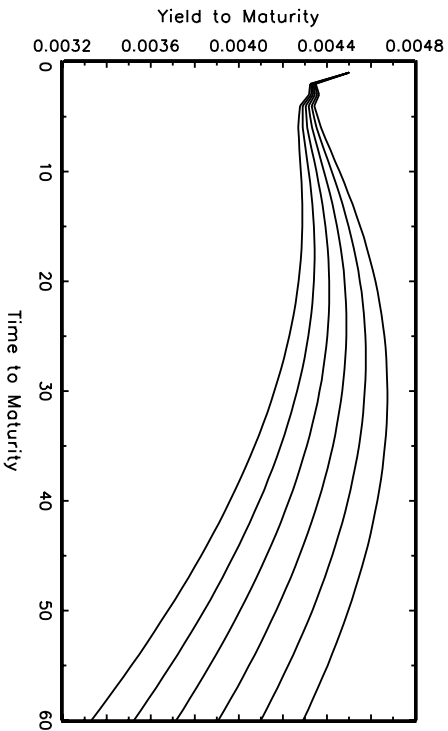


FIGURE 14 – AR(4):
 $\phi_1^* = 0.72$, $\phi_2^* = 0.25$, $\nu^* = 0.00009$;
 $\sigma^2 = 0.00000039$; $\phi_3^* = -0.10$ (bottom) to -0.0075 (top);

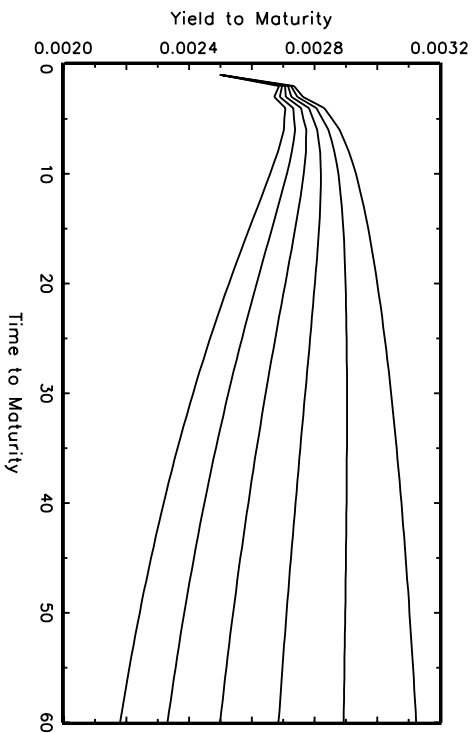


FIGURE 16 – Gaussian AR(4):
 $\phi_1^* = 0.73$, $\phi_2^* = 0.25$, $\phi_3^* = -0.07$;
 $\phi_4^* = 0.08$, $\nu^* = 0.00009$;
 $\sigma^2 = 0.00000004$ (top) to 0.0000024 (bottom);

