

# 1 VAR Dynamics

The vector of state variables is  $\mathbf{z}_t = (r_t^{(1)}, r_t^{(40)} - r_t^{(1)}, h_t, y_t, g_t, c_t, a_t, d_t)^\top$ . Its stationary counterpart is  $\tilde{\mathbf{z}}_t = (r_t^{(1)} - r_{t-1}^{(1)}, r_t^{(40)} - r_t^{(1)}, h_t, y_t - y_{t-1}, g_t, c_t, a_t, d_t)^\top$ , which follows a VAR(1) process (cf. VAR\_Framework.pdf):

$$\tilde{\mathbf{z}}_{t+1} = \boldsymbol{\alpha} + \tilde{\Phi} \tilde{\mathbf{z}}_t + \Sigma^{1/2} \boldsymbol{\epsilon}_{t+1},$$

where  $\tilde{\boldsymbol{\alpha}}$  is an  $8 \times 1$  vector,  $\tilde{\Phi}$  is an  $8 \times 8$  coefficient matrix, and  $\Sigma^{1/2}$  is the Cholesky decomposition of the residual covariance matrix  $\Sigma$ .

However,  $\mathbf{z}_t$  essentially follows a VAR(2) process

$$\mathbf{z}_{t+1} = \boldsymbol{\alpha} + \Phi_1 \mathbf{z}_t + \Phi_2 \mathbf{z}_{t-1} + \Sigma^{1/2} \boldsymbol{\epsilon}_{t+1},$$

with  $\Phi_1 = \tilde{\Phi} + \Psi$ , where  $\Psi$  is zero everywhere except  $\Psi_{1,1} = \Psi_{4,4} = 1$ , and  $\Phi_2$  is zero except the first and fourth columns being the corresponding columns of  $-\tilde{\Phi}$ .

## 2 Stochastic Discount Factors

### 2.1 The Model

Following [1], [2], [3], we develop stochastic discount factors, that is, pricing kernels, to price all nominal asset in the market.

Let  $\xi_{t+1}$  be the Radon-Nikodym derivative which converts the risk-neutral measure  $\mathbb{Q}$  to the data-generating measure  $\mathbb{P}$ ; let  $X_{t+1}$  be the payoff of an asset. The following holds

$$\mathbb{E}_t^{\mathbb{Q}}(X_{t+1}) = \mathbb{E}_t^{\mathbb{P}}(\xi_{t+1} X_{t+1}) / \xi_t.$$

Assume that  $\xi_{t+1}$  follows the log-normal process:

$$\xi_{t+1} = \xi_t \exp \left( -\frac{1}{2} \boldsymbol{\lambda}_t^\top \boldsymbol{\lambda}_t - \boldsymbol{\lambda}_t^\top \boldsymbol{\epsilon}_{t+1} \right),$$

where  $\boldsymbol{\lambda}_t$  are the market prices of risk process associated with the sources of uncertainty  $\boldsymbol{\epsilon}_t$ . We parametrise  $\boldsymbol{\lambda}_t$  as an affine process:

$$\boldsymbol{\lambda}_t = \boldsymbol{\lambda}_0 + \lambda_1 \mathbf{z}_t,$$

where  $\boldsymbol{\lambda}_0 \in \mathbb{R}^8$  affects the long-run mean yields and  $\lambda_1 \in \mathbb{R}^{8 \times 8}$  affects the time-varying risk premiums.

The pricing kernel is defined as:

$$s_{t+1} = \frac{\xi_{t+1}}{\xi_t} \exp(-r_t) = \exp \left( -\mathbf{e}_1^\top \mathbf{z}_t - \frac{1}{2} \boldsymbol{\lambda}_t^\top \boldsymbol{\lambda}_t - \boldsymbol{\lambda}_t^\top \boldsymbol{\epsilon}_{t+1} \right), \quad (1)$$

where  $r_t$  is the one-period short rate and  $\mathbf{e}_1 = (1, 0, 0, 0, 0, 0, 0, 0)^\top$ .

The price of an asset  $P_t$  paying  $X_{t+1}$  at time  $t + 1$  is:

$$P_t = E_t^P(s_{t+1}X_{t+1}).$$

In particular, the price of an  $n$ -period nominal bond  $p_t^{(n)}$  satisfies the recursive equation:

$$p_t^{(n)} = E_t^P(s_{t+1}p_{t+1}^{(n-1)}),$$

with the terminal condition  $p_t^{(0)} = 1$ .

The bond price is then an exponential linear function of the state vector:

$$p_t^{(n)} = \exp\left(A_n + \mathbf{B}_n^\top \mathbf{z}_t + \mathbf{C}_n^\top \mathbf{z}_{t-1}\right).$$

The absence of arbitrage can be achieved by setting

$$\begin{aligned} A_{n+1} &= A_n + \mathbf{B}_n^\top (\boldsymbol{\alpha} - \Sigma^{1/2} \boldsymbol{\lambda}_0) + \frac{1}{2} \mathbf{B}_n^\top \Sigma \mathbf{B}_n, \\ \mathbf{B}_{n+1} &= -\mathbf{e}_1 + (\Phi_1 - \Sigma^{1/2} \boldsymbol{\lambda}_1)^\top \mathbf{B}_n + \mathbf{C}_n, \\ \mathbf{C}_{n+1} &= \Phi_2^\top \mathbf{B}_n. \end{aligned}$$

To ensure the consistency between this equation and the observables  $r^{(1)}, r^{(40)}$  in the VAR model [3], we require initial values to be  $A_1 = 0, \mathbf{B}_1 = -\mathbf{e}_1, \mathbf{C}_1 = \mathbf{0}$ , and terminal values to be  $A_{40} = 0, \mathbf{B}_{40} = -40\mathbf{e}_1 - 40\mathbf{e}_2, \mathbf{C}_{40} = \mathbf{0}$ .

The continuously compounding yield  $r_t^{(n)}$  on an  $n$ -quarter zero-coupon bond is an affine function of the state vector:

$$r_t^{(n)} = -\frac{\log p_t^{(n)}}{n} = -\frac{A_n}{n} - \frac{\mathbf{B}_n^\top}{n} \mathbf{z}_t - \frac{\mathbf{C}_n^\top}{n} \mathbf{z}_{t-1}.$$

## 2.2 Estimation

Given the VAR dynamics, we can estimate the risk parameters  $\boldsymbol{\lambda}_0, \boldsymbol{\lambda}_1$  by minimising the sum squared error:

$$\min_{\boldsymbol{\lambda}_0, \boldsymbol{\lambda}_1} \sum_{t=1}^T \sum_{n=1}^N \left( \hat{r}_t^{(n)} - r_t^{(n)} \right)^2.$$

Different to [1], we calibrate the model to zero-coupon bonds which mature every quarter from 3-month to 10-year over the same time window as VAR (112 quarters):

$$\min_{\boldsymbol{\lambda}_0, \boldsymbol{\lambda}_1} \sum_{t=1}^{112} \sum_{n=1}^{40} \left( \hat{r}_t^{(n)} - r_t^{(n)} \right)^2.$$

## 2.3 Results

The estimated parameters are presented in table 1.

$\lambda_0$	$\lambda_1$							
-0.03009	-0.00677	0.00190	0.00376	0.00615	0.00116	0.00082	0.00363	0.00511
-0.01263	0.00874	0.00330	0.00514	0.00769	0.00494	0.01166	0.00778	0.01118
-0.01793	-0.00400	0.00189	0.00345	-0.00518	0.00723	-0.00883	0.00291	0.01206
0.02454	0.01599	0.01590	0.01790	0.01134	0.00546	0.01049	0.00909	0.01011
-0.01413	-0.00320	0.00489	0.00568	-0.00123	0.00773	0.00787	0.00216	-0.01536
-0.01316	-0.00164	0.01035	0.00173	0.00064	-0.00192	0.00954	0.00140	0.01201
-0.03393	-0.02448	-0.01474	-0.00440	0.04978	-0.00573	-0.00387	-0.00522	0.01668
-0.02980	-0.03279	-0.00689	0.00840	0.00221	-0.00160	-0.00581	0.00300	0.01806

Table 1: Estimated parameters in the market price of risk.

We simulated 10,000 trajectories of the pricing kernels over the next 100 years. Figure 1 shows the fitted historical and simulated stochastic discount factors. Observe that the range for both the historical and simulated SDFs are larger than the range in [1], probably because a larger time window was considered. The SDF around 2020 was greater than 1, which is consistent with the interest rate being negative. Similar to [1], the range of simulated trajectories almost doubled that of historical SDFs.

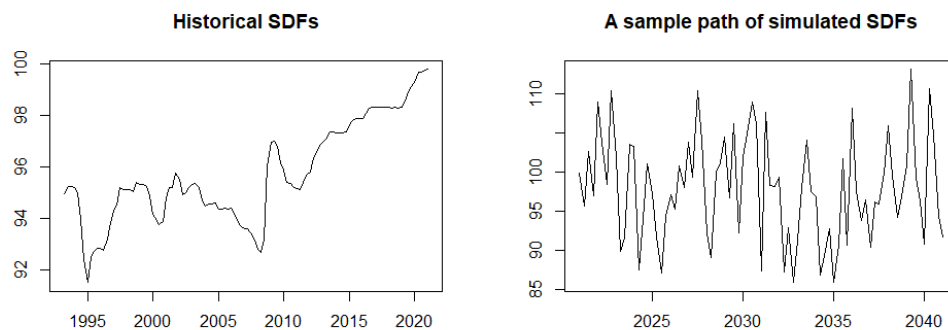


Figure 1: Stochastic Discount Factors.

Table 2 reports the correlations between the historical stochastic discount factors and the state variables.

Correlation	$r^{(1)}$	$r^{(40)} - r^{(1)}$	$h_t$	$y_t$	$g_t$	$c_t$	$a_t$	$d_t$
SDF	-0.99993	0.04220	0.20235	0.40931	-0.13179	-0.29297	0.00139	-0.03617

Table 2: Correlations between the stochastic discount factors and state variables.

## References

- [1] Daniel H Alai, Hua Chen, Daniel Cho, Katja Hanewald, and Michael Sherris. Developing equity release markets: Risk analysis for reverse mortgages and home reversions. *North American Actuarial Journal*, 18(1):217–241, 2014.
- [2] Andrew Ang and Monika Piazzesi. A no-arbitrage vector autoregression of term structure dynamics with macroeconomic and latent variables. *Journal of Monetary economics*, 50(4):745–787, 2003.
- [3] Andrew Ang, Monika Piazzesi, and Min Wei. What does the yield curve tell us about gdp growth? *Journal of econometrics*, 131(1-2):359–403, 2006.