

Exercise 10.3 (Calibration of two-factor Vasicek model)

The canonical form of two-factor Vasicek model is as follows.

$$\begin{aligned} dY_1(t) &= -\lambda_1 Y_1(t) dt + d\tilde{W}_1(t) \\ dY_2(t) &= -\lambda_{21} Y_1(t) dt - \lambda_2 Y_2(t) dt + d\tilde{W}_2(t) \\ R(t) &= \delta_0(t) + \delta_1 Y_1(t) + \delta_2 Y_2(t). \end{aligned}$$

Here δ_0 is a nonrandom function of t . Let

$$\begin{aligned} B(t, T) &= \tilde{\mathbb{E}} \left[e^{-\int_t^T R(u) du} | \mathcal{F}(t) \right] \\ &= f(t, T, Y_1(t), Y_2(t)). \end{aligned}$$

Here

$$f(t, T, y_1, y_2) = e^{-y_1 C_1(t, T) - y_2 C_2(t, T) - A(t, T)}$$

Assume the model parameters $\lambda_1 > 0, \lambda_2 > 0, \lambda_{21}, \delta_1, \delta_2$ are given. Find $\delta_0(T)$ in terms of $\frac{\partial}{\partial T} \log B(0, T)$ and the model parameters such that

$$f(0, T, Y_1(0), Y_2(0)) = B(0, T), \quad T \geq 0.$$

Proof

Let $\tau = T - t$. From the textbook, the following holds

$$\begin{aligned} \tilde{C}'_1(\tau) &= -\lambda_1 \tilde{C}_1(\tau) - \lambda_{21}(\tau) + \delta_1 \\ \tilde{C}'_2(\tau) &= -\lambda_2 \tilde{C}_2(\tau) + \delta_2 \\ A'(\tau) &= -\frac{1}{2} \tilde{C}_1^2(\tau) - \frac{1}{2} \tilde{C}_2^2(\tau) + \delta_0(\tau) \end{aligned}$$

Moreover, if $\lambda_1 \neq \lambda_2$, then

$$C_1(t, T) = \tilde{C}_1(\tau) = \frac{1}{\lambda_1} \left(\delta_1 - \frac{\lambda_{21}\delta_2}{\lambda_2} \right) \left(1 - e^{-\lambda_1\tau} \right) + \frac{\lambda_{21}\delta_2}{\lambda_2(\lambda_1 - \lambda_2)} \left(e^{-\lambda_2\tau} - e^{-\lambda_1\tau} \right)$$

And if $\lambda_1 = \lambda_2$, then

$$C_1(t, T) = \tilde{C}_1(\tau) = \frac{1}{\lambda_1} \left(\delta_1 - \frac{\lambda_{21}\delta_1}{\lambda_2} \right) \left(1 - e^{-\lambda_1\tau} \right) + \frac{\lambda_{21}\delta_2}{\lambda_1} \tau e^{-\lambda_1\tau}$$

Also

$$C_2(t, T) = \tilde{C}_2(\tau) = \frac{\delta_2}{\lambda_2} \left(1 - e^{-\lambda_2\tau} \right)$$

Finally, from (10.2.55),

$$A(t, T) = A(\tau) = \int_0^\tau \left[-\frac{1}{2} C_1^2(u) - \frac{1}{2} C_2^2(u) + \delta_0(t+u) \right] du$$

Therefore,

$$A(0, T) = \int_0^T \left[-\frac{1}{2} C_1^2(u, T) - \frac{1}{2} C_2^2(u, T) + \delta_0(u) \right] du$$

So

$$\begin{aligned}\frac{\partial}{\partial T} A(0, T) &= -\frac{1}{2} C_1^2(T, T) - \frac{1}{2} C_2^2(T, T) + \delta_0(T) \\ &= \delta_0(T).\end{aligned}$$

On the other hand,

$$\log B(0, T) = -Y_1(0)C_1(0, T) - Y_2(0)C_2(0, T) - A(0, T)$$

Thus,

$$\delta_0(T) = -\frac{\partial}{\partial T} \log B(0, T) - Y_1(0) \frac{\partial}{\partial T} C_1(0, T) - Y_2(0) \frac{\partial}{\partial T} C_2(0, T)$$

Computing $C_1(0, T)$ and $C_2(0, T)$ using their closed-form formulas as above completes the proof.