

CQF Module 4 Examination Solutions

January 2016 Cohort

dW is the usual increment of a Brownian motion.

1. We know the ZCB price $Z(r, t; T)$ satisfies the BPE

$$\frac{\partial Z}{\partial t} + \frac{1}{2}w^2 \frac{\partial^2 Z}{\partial r^2} + (u - \lambda w) \frac{\partial Z}{\partial r} - rZ = 0, \quad Z(r, T; T) = 1$$

Expand $Z(r, t; T)$ for small times t to maturity T , i.e. in powers of $(T - t)$

$$Z \sim a(r) + b(r)(T - t) + c(r)(T - t)^2 + \dots$$

for the unknown coefficients. Firstly, $Z(r, T; T) = 1 \longrightarrow a(r) = 1$

$$Z \sim 1 + b(r)(T - t) + c(r)(T - t)^2 + \dots$$

$$\frac{\partial Z}{\partial t} = -b(r) - 2c(r)(T - t)$$

$$\frac{\partial Z}{\partial r} = b'(r)(T - t) + c'(r)(T - t)^2; \quad \frac{\partial^2 Z}{\partial r^2} = b''(r)(T - t) + c''(r)(T - t)^2$$

Substituting these terms into the BPE gives

$$\begin{aligned} & -b(r) - 2c(r)(T - t) + \frac{1}{2}w^2 (b''(r)(T - t) + c''(r)(T - t)^2) + \\ & (u - \lambda w) (b'(r)(T - t) + c'(r)(T - t)^2) \\ & = r(1 + b(r)(T - t) + c(r)(T - t)^2). \end{aligned}$$

Now compare coefficients of powers of $(T - t)$

$O(1)$:

$$-b(r) = r \longrightarrow b(r) = -r \implies b'(r) = -1 \implies b''(r) = 0$$

$O((T - t))$:

$$\begin{aligned} -2c(r) + (u - \lambda w)(-1) &= -r^2 \\ c(r) &= \frac{1}{2}(r^2 - (u - \lambda w)) = \frac{1}{2}(r^2 - \eta + \gamma r) \end{aligned}$$

So for $t \longrightarrow T$ we have

$$Z \sim 1 - r(T - t) + \frac{1}{2}(r^2 - \eta + \gamma r)(T - t)^2 + \dots$$

Now write $\log Z = \log(1 + x)$, where $x \equiv -r(T - t) + \frac{1}{2}(r^2 - \eta + \gamma r)(T - t)^2$, so

$$\begin{aligned} \log Z &\sim -r(T - t) + \frac{1}{2}(r^2 - \eta + \gamma r)(T - t)^2 \\ &\quad - \frac{1}{2}(-r(T - t) + \frac{1}{2}(r^2 - \eta + \gamma r)(T - t)^2)^2 \\ &\sim -r(T - t) - \frac{1}{2}(T - t)^2(\eta - \gamma r) + \dots \end{aligned}$$

The bond price also satisfies $Z = \exp(-r_L(T - t)) \longrightarrow r_L = -\frac{1}{T-t} \log Z$, and using the expression for $\log Z$ above

$$\begin{aligned} r_L &= -\frac{1}{T-t} (-r(T - t) - \frac{1}{2}(T - t)^2(\eta - \gamma r)) \\ &= r + \frac{1}{2}(T - t)(\eta - \gamma r) \end{aligned}$$

So for Vasicek with one month LIBOR, $(T - t) = 1/12$, we find

$$r_L \sim r + \frac{1}{2}(\eta - \gamma r)(1/12).$$

2. For the SDE

$$dG = a(G, t) dt + b(G, t) dX$$

Itô's Lemma for $V = V(G)$ gives

$$dV = \left(a(G, t) \frac{dV}{dG} + \frac{1}{2} b^2(G, t) \frac{d^2V}{dG^2} \right) dt + b(G, t) \frac{dV}{dG} dX$$

where

$$\begin{aligned} a &\equiv \theta(t) + \frac{d(\log \sigma(t))}{dt} \log r \\ b &\equiv \sigma(t) \end{aligned}$$

Put $V = e^G \rightarrow \frac{dV}{dG} = e^G = \frac{d^2V}{dG^2}$ where $G = \log r$, hence

$$dr = r \left(\theta(t) + \frac{d(\log \sigma(t))}{dt} \log r + \frac{1}{2} \sigma^2(t) \right) dt + \sigma(t) r dX.$$

3. If $Z(r, t) = \exp(A(t; T) - rB(t; T))$ then $Z_r = -BV$ $Z_{rr} = B^2Z$
 Solution with $Z(r, T; T) = 1$ implies $B(T; T) = A(T; T) = 0$
 $Z_t = (A'(t) - rB'(t))V$ so subst. these in to the PDE above

$$\begin{aligned} A'(t; T) - rB'(t; T) + \frac{1}{2}c^2B^2 - (\eta(t) - \gamma r)B - r &= 0 \quad \forall r \\ (A' + \frac{1}{2}c^2B^2 - \eta(t)B) - r(B' - \gamma B + 1) &= 0 \\ \Rightarrow A' = -\frac{1}{2}c^2B^2 + \eta(t)B, & \quad B' = \gamma B - 1 \end{aligned}$$

From $Z(r, t; T) = \exp(A(t) - rB(t))$ we note that as $r \rightarrow \infty$, $Z \rightarrow 0$. Solving for $B(t; T)$

$$\frac{dB}{dt} = \gamma B - 1$$

You can solve this by the variable sep. method: $\frac{dB}{\gamma B - 1} = dt$. Now recall:

$$\int \frac{dx}{ax + 1} = \frac{1}{a} \ln|ax + 1| + K$$

therefore we have

$$\begin{aligned} \int_t^T \frac{dB}{\gamma B - 1} &= \int_t^T d\tau = \frac{1}{\gamma} \ln|\gamma B(\tau; T) - 1|_t^T = (T - t) \\ \ln \left| \frac{\gamma B(T; T) - 1}{\gamma B(t; T) - 1} \right| &= \gamma(T - t) \end{aligned}$$

we know $B(T; T) = 0$ so

$$\begin{aligned} \ln \left| \frac{\gamma B(t; T) - 1}{-1} \right|^{-1} &= \gamma(T - t) = -\ln|1 - \gamma B(t; T)| = \gamma(T - t) \\ 1 - \gamma B(t; T) &= \exp[-\gamma(T - t)] = \gamma B(t; T) = 1 - \exp[-\gamma(T - t)] \\ B(t; T) &= \frac{1}{\gamma} (1 - \exp[-\gamma(T - t)]) \end{aligned}$$

Then

$$A(t; T) = \frac{1}{2}c^2 \int_t^T B^2(\tau; T) d\tau - \int_t^T B(\tau; T) \eta(\tau) d\tau$$

and

$$\begin{aligned} \int_t^T B^2(\tau; T) d\tau &= \frac{1}{\gamma^2} \int_t^T (1 - 2e^{-\gamma(T-\tau)} + e^{-2\gamma(T-\tau)}) d\tau \\ &= \frac{1}{\gamma^2} \left((T - t) - \frac{2}{\gamma} e^{-\gamma(T-\tau)} \Big|_t^T + \frac{1}{2\gamma} e^{-2\gamma(T-\tau)} \Big|_t^T \right) \\ &= \frac{1}{\gamma^2} \left((T - t) - \frac{2}{\gamma} + \frac{2}{\gamma} e^{-\gamma(T-t)} + \frac{1}{2\gamma} - \frac{1}{2\gamma} e^{-2\gamma(T-t)} \right) \\ &= \frac{1}{\gamma^2} \left((T - t) - \frac{3}{2\gamma} + \frac{2}{\gamma} e^{-\gamma(T-t)} - \frac{1}{2\gamma} e^{-2\gamma(T-t)} \right) \end{aligned}$$

Hence

$$A(t; T) = - \int_t^T B(\tau; T) \eta(\tau) d\tau + \frac{c^2}{2\gamma^2} \left((T - t) - \frac{3}{2\gamma} + \frac{2}{\gamma} e^{-\gamma(T-t)} - \frac{1}{2\gamma} e^{-2\gamma(T-t)} \right)$$

4. This is an Ornstein-Uhlenbeck process and an analytic solution for this equation exists. The SDE can be written as $dU_t + \gamma U_t dt = \sigma dX_t$.

Multiply both sides by an integrating factor $e^{\gamma t}$

$$\begin{aligned} e^{\gamma t} (dU_t + \gamma U_t) dt &= \sigma e^{\gamma t} dX_t \\ d(e^{\gamma t} U_t) &= \sigma e^{\gamma t} dX_t \end{aligned}$$

Integrating over $[0, t]$ gives

$$U_t = ue^{-\gamma t} + \sigma \int_0^t e^{\gamma(s-t)} dX_s$$

where $U(0) = u = U_0$. Now taking expectations

$$\begin{aligned} \mathbb{E}[U_t] &= \mathbb{E}[ue^{-\gamma t}] + \sigma \int_0^t e^{\gamma(s-t)} \mathbb{E}[dX_s] \\ &= ue^{-\gamma t}. \end{aligned}$$

To calculate the variance we have $\mathbb{V}[U_t] = \mathbb{E}[U_t^2] - \mathbb{E}^2[U_t]$

$$\begin{aligned} &= \mathbb{E} \left[\left(ue^{-\gamma t} + \sigma \int_0^t e^{\gamma(s-t)} dW_s \right)^2 \right] - u^2 e^{-2\gamma t} \\ &= \mathbb{E} [u^2 e^{-2\gamma t}] + \sigma^2 \mathbb{E} \left[\left(\int_0^t e^{\gamma(s-t)} dW_s \right)^2 \right] + 2\sigma ue^{-\gamma t} \underbrace{\mathbb{E} \left[\int_0^t e^{\gamma(s-t)} dW_s \right]}_{\text{Itô integral}} - u^2 e^{-2\gamma t} \\ &= \sigma^2 \mathbb{E} \left[\left(\int_0^t e^{\gamma(s-t)} dW_s \right)^2 \right] \end{aligned}$$

Now use Itô's Isometry

$$\mathbb{E} \left[\left(\int_0^t Y_s dW_s \right)^2 \right] = \mathbb{E} \left[\int_0^t Y_s^2 ds \right],$$

So

$$\begin{aligned} \mathbb{V}[U_t] &= \sigma^2 \mathbb{E} \left[\int_0^t e^{2\gamma(s-t)} ds \right] \\ &= \frac{\sigma^2}{2\gamma} e^{2\gamma(s-t)} \Big|_0^t = \frac{\sigma^2}{2\gamma} (1 - e^{-2\gamma t}) \end{aligned}$$

5.

$$\begin{aligned}
\frac{dZ}{Z} &= r(t)dt \\
\int_t^T d(\log Z(s; T)) &= \int_t^T r(s)ds \\
-\log Z(t; T) &= \int_t^T r(s)ds \quad \text{because } \log Z(T; T) = \log(1) = 0 \\
Z(t; T) &= \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_t^T r(s)ds} \right]
\end{aligned}$$

This is consistent with Feynman-Kac for a bond as a derivative *wrt* the stochastic variable $r(t)$ with boundary condition (payoff) $Z(r, T; T)$ and pricing equation (since we are not given any coefficients of $dr(t)$)

$$\frac{\partial Z}{\partial t}(r, t; T) - r(t)Z(r, t; T) = 0$$

$$\begin{aligned}
Z(r, t; T) &= \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_t^T r(s)ds} Z(r, T; T) | \mathcal{F}_t \right] & Z(r, T; T) &= 1 \quad \forall r_T \\
&= \text{discounting with stochastic } r(t) \text{ remains inside the expectation} \\
Z(r, t; T) &= \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_t^T r(s)ds} | \mathcal{F}_t \right]
\end{aligned}$$

In order to satisfy the expectation, the pricing is carried out by Monte-Carlo. Within discrete implementation of HJM framework, use **the first column** of the simulated rates matrix – the shortest maturity instantaneous forward rate – as a proxy to short rate $r(t) = f(t, t)$:

$$\begin{aligned}
Z(t; T) &= \mathbb{E} \left[e^{-\int_t^T r(s)ds} \right] \\
&= \exp \left(- \sum r_t \Delta t \right)
\end{aligned}$$

The answer to the computational side of the question is a convergence diagram like one below. The longer maturity T selected, the smaller ZCB price will be compared to 1.

Excel Notes: Use *SUM()* on the first column to include rows up to the timestep that correspond to maturity, e.g., $\delta t = 0.01$ so for $T = 0.5$ there are 50 rows to add up.

Press *F9* to update the random numbers. Record the history of bond prices and plot the running average for $N = 1, 2, 3 \dots 1000$ simulations – this average should converge to a stable value with minimal variance. Writing a macro that saves the history of simulations would be useful.

