

1.

Let (Ω, \mathcal{F}, P) be a probability space. As $\mathcal{G}_u \subseteq \mathcal{G}_t$, we have that

$$\begin{aligned} E[E[X_t|\mathcal{G}_t]|\mathcal{G}_u] &= \int_{\mathcal{G}_u} E[X_t|\mathcal{G}_t]dP(\mathcal{G}_u) \\ &= \int_{\mathcal{G}_u} \int_{\mathcal{G}_t} X dP(\mathcal{G}_t)dP(\mathcal{G}_u) \\ &= \int_{\mathcal{G}_t} \int_{\mathcal{G}_u} X dP(\mathcal{G}_u)dP(\mathcal{G}_t) \\ &= E[X_t|\mathcal{G}_u] \end{aligned}$$

Similarly,

$$\begin{aligned} E \left[\int_0^t E[Y_s|\mathcal{G}_s] ds \middle| \mathcal{G}_u \right] &= E \left[\int_0^t \int_{\mathcal{G}_s} Y dP(\mathcal{G}_s) ds \middle| \mathcal{G}_u \right] \\ &= \int_{\mathcal{G}_u} \int_0^t \int_{\mathcal{G}_s} Y dP(\mathcal{G}_s) ds dP(\mathcal{G}_u) \\ &= \int_0^t \int_{\mathcal{G}_u} \int_{\mathcal{G}_s} Y dP(\mathcal{G}_s) dP(\mathcal{G}_u) ds \\ &= \int_0^t E[Y_s|\mathcal{G}_u] ds \end{aligned}$$

We also know that

$$\begin{aligned} \max \left(\int_{\mathcal{G}_t} X_+ dP(\mathcal{G}_t), \int_{\mathcal{G}_t} X_- dP(\mathcal{G}_t) \right) &\leq \max \left(\int_{\Omega} X_+ dP, \int_{\Omega} X_- dP \right) \\ &= \max(E[X_+], E[X_-]) < \infty \end{aligned}$$

Therefore, $E[|X_t||\mathcal{G}_t] < \infty$. Similarly, using Fubini and the steps above, we can also show that

$$\int_0^t E[|Y_s||\mathcal{G}_s] ds < \infty$$

Therefore,

$$\begin{aligned} &E[E[X_t|\mathcal{G}_t]|\mathcal{G}_u] - E \left[\int_0^t E[Y_s|\mathcal{G}_s] ds \middle| \mathcal{G}_u \right] \\ &= E[X_t|\mathcal{G}_u] - \int_0^t E[Y_s|\mathcal{G}_u] ds \end{aligned}$$

which confirms it is indeed a martingale.

2.

To match the state equations, we have that

$$a_{i,j}^1 = iK^1, \quad s_{i,j}^1 = i^2$$

and

$$a_{i,j}^2 = rj, \quad s_{i,j}^2 = \frac{r}{K^2}(j^2 + \alpha_{21}ij)$$

and

$$\begin{aligned} Lf(i, j) = & a_{i,j}^1[f(i+1, j) - f(i, j)] + a_{i,j}^2[f(i, j+1) - f(i, j)] \\ & + s_{i,j}^1[f(i-1, j) - f(i, j)] + s_{i,j}^2[f(i, j-1) - f(i, j)] \end{aligned}$$

Then the 2 state equations are consistent with the martingale problem:

$$f(X_t^1, X_t^2) - f(X_0^1, X_0^2) - \int_0^t Lf(X_u^1, X_u^2) du$$

which is a $\sigma(X_s^1, X_s^2, s \leq t)$ -martingale.

3.

We know that

$$F_{X_1 \vee X_2}(x) = (F_{X_i}(x))^2$$

Hence,

$$\begin{aligned} f_{X_1 \vee X_2}(x) &= 2F_{X_i}(x)f_{X_i}(x) \\ &= 2 \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{u^2}{2}} du \\ &= \frac{e^{-\frac{x^2}{2}}}{\pi} \int_{-\infty}^x e^{-\frac{u^2}{2}} du \end{aligned}$$

Therefore, we can calculate

$$f_{X_1 \vee X_2}(0) = \frac{1}{\pi} \underbrace{\int_{-\infty}^0 e^{-\frac{u^2}{2}} du}_{\sqrt{\pi/2}} = \frac{1}{\sqrt{2\pi}}$$

$$\begin{aligned} &\int_0^1 f_{X_1 \vee X_2}(x) dx \\ &= F_{X_1 \vee X_2}(1) - F_{X_1 \vee X_2}(0) \\ &= F_{X_i}(1)^2 - F_{X_i}(0)^2 \\ &= 0.84134475^2 - 0.5^2 \\ &= 0.45786098835 \end{aligned}$$

4.

Let $g(B_t, t) = f\left(\frac{\sigma^2}{2}t\right)$, hence we have that

$$\frac{\partial}{\partial x}g(B, x) = \frac{\sigma^2}{2}e^{\frac{\sigma^2 x}{2}}\left(f' \circ e^{\frac{\sigma^2 x}{2}}\right)$$

$$\frac{\Delta}{2}g(B, s) = 0$$

and that

$$g(B_t, t) - g(0, 0) - \int_0^t \frac{\Delta}{2}g(B_s, s) + \frac{\partial}{\partial s}g(B_s, s)ds$$

is a $\sigma(B_u, u \leq t)$ -martingale problem. However, we have that

$$\frac{\Delta}{2}g(B_s, s) + \frac{\partial}{\partial s}g(B_s, s) = \frac{\sigma^2}{2} \exp\left(\frac{\sigma^2 s}{2}\right) f'\left(\exp\left(\frac{\sigma^2 s}{2}\right)\right)$$

Hence, if we let $X_t = \exp\left(\frac{\sigma^2}{2}t\right)$, then we get

$$f(X_t) - f(X_0) - \int_0^t \frac{\sigma^2 X_u}{2} f'(X_u) du$$

is a martingale. If there is another solution Y_t that has $Y_0 = X_0$. Let $f(x) = \ln(x)$ so that $f'(x) = \frac{1}{x}$ and therefore, $X_u f'(X_u) = 1$. Thus,

$$\ln(X_t) - \ln(X_0) - \int_0^t \frac{\sigma^2}{2} du = \ln(Y_t) - \ln(Y_0) - \int_0^t \frac{\sigma^2}{2} du$$

and hence $\ln(X_t) = \ln(Y_t) \implies X_t = Y_t$ for all t which proves its uniqueness.