Solutions

CQF

- 1. Use Itô's formula to determine whether the following are martingales:
 - (i) $Y(t) = e^{1/2t} \cos X(t)$;
 - (ii) $Y(t) = e^{\alpha t} \sin X(t)$ for some constant α with $0 < \alpha < 1$. Does the answer depend on the value of α ?
 - (iii) $Y(t) = (X(t) + t) \exp \left\{-\frac{1}{2}t X(t)\right\}.$

A diffusion process Y(t) is a martingale if the drift coefficient of its SDE is identically 0.

(i) Consider the function $F(s,x) = e^{1/2s} \cos x$.

$$\frac{\partial F}{\partial s}(s,x) = \frac{1}{2}e^{1/2s}\cos x$$

$$\frac{\partial F}{\partial x}(s,x) = -e^{1/2s}\sin x$$

$$\frac{\partial^2 F}{\partial x^2}(s,x) = -e^{1/2s}\cos x$$

By Itô, the dynamics of the process Y(t) = F(t, X(t)) is given by

$$dY(t) = \left(\frac{\partial F}{\partial t}(t, X(t)) + \frac{1}{2}\frac{\partial^2 F}{\partial x^2}(s, x)\right)dt - e^{1/2s}\sin X(t)dX(t)$$
$$= -e^{1/2s}\sin X(t)dX(t)$$

Hence Y(t) is a martingale.

(ii) Consider the function $F(s,x) = e^{\alpha s} \sin x$ with α in (0,1).

$$\frac{\partial F}{\partial s}(s,x) = \alpha e^{\alpha s} \sin x$$

$$\frac{\partial F}{\partial x}(s,x) = e^{\alpha s} \cos x$$

$$\frac{\partial^2 F}{\partial x^2}(s,x) = -e^{\alpha s} \sin x$$

By Itô, the dynamics of the process Y(t) = F(t, X(t)) is given by

$$dY(t) = \left(\frac{\partial F}{\partial t}(t, X(t)) + \frac{1}{2}\frac{\partial^2 F}{\partial x^2}(s, x)\right)dt + e^{\alpha s}\cos X(t)dX(t)$$
$$= \left(\alpha - \frac{1}{2}\right)e^{\alpha t}\sin X(t)dt + e^{\alpha s}\cos X(t)dX(t)$$

For Y(t) to be a martingale we must have $\alpha = \frac{1}{2}$. With any other choice of value for α , Y(t) is not a martingale.

(iii) Consider the function $F(s,x) = (x+s) \exp\left\{-\frac{1}{2}s - x\right\}$.

$$\frac{\partial F}{\partial s}(s,x) = \left(1 - \frac{1}{2}(x+s)\right) \exp\left\{-\frac{1}{2}s - x\right\}$$

$$\frac{\partial F}{\partial x}(s,x) = (1 - x - s) \exp\left\{-\frac{1}{2}s - x\right\}$$

$$\frac{\partial^2 F}{\partial x^2}(s,x) = (-2 + x + s) \exp\left\{-\frac{1}{2}s - x\right\}$$

By Itô, the dynamics of the process Y(t) = F(t, X(t)) is given by

$$dY(t) = \left(\frac{\partial F}{\partial t}(t, X(t)) + \frac{1}{2}\frac{\partial^2 F}{\partial x^2}(s, x)\right)dt + (1 - x - s)\exp\left\{-\frac{1}{2}s - x\right\}dX(t)$$

$$= \left(1 - \frac{1}{2}(x + s) + \frac{1}{2}\left[-2 + x + s\right]\right)\exp\left\{-\frac{1}{2}s - x\right\}dt$$

$$+ (1 - x - s)\exp\left\{-\frac{1}{2}s - x\right\}dX(t)$$

$$= \left(1 - \frac{1}{2}(x + s) + \frac{1}{2}\left[-2 + x + s\right]\right)\exp\left\{-\frac{1}{2}s - x\right\}dt$$

$$+ (1 - x - s)\exp\left\{-\frac{1}{2}s - x\right\}dX(t)$$

$$= (1 - x - s)\exp\left\{-\frac{1}{2}s - x\right\}dX(t)$$

Hence Y(t) is a martingale.

2. Moments of the Brownian Motion X(t) - Consider the function $m_n(t)$ defined as

$$m_n(t) = \mathbf{E}[X^n(t)], \qquad n = 1, 2, \dots$$
 (1)

where X(t) is a standard Brownian motion.

Applying Itô's formula, show that:

$$m_n(t) = \frac{1}{2}n(n-1)\int_0^t m_{n-2}(s)ds \tag{2}$$

for n = 2, 3, ...

Deduce from (2) that

$$m_4(t) = 3t^2 \tag{3}$$

compute $m_6(t)$

Answer: Because of the expectation, we cannot tackle expression (1) upfront.

Consider instead the auxiliary function $g_n(t,x) = x^n$ for $n \ge 2$. Note the relation between $g_n(t,x)$ and $m_n(t)$:

$$m_n(t) = \mathbf{E}[g_n(t, X(t))]$$

Applying Itô's lemma to the function g_n and the standard Brownian motion, we get

$$g_n(t) = g_n(0) + \int_0^t \frac{\partial g_n}{\partial s} ds + \int_0^t \frac{\partial g_n}{\partial x} dX(s) + \frac{1}{2} \int_0^t \frac{\partial^2 g_n}{\partial x^2} ds$$
$$= n \int_0^t X^{n-1}(s) dX(s) + \frac{1}{2} n(n-1) \int_0^t X^{n-2}(s) ds$$

since $g_n(0) = 0$

Take expectation on both sides to get:

$$m_n(t) = \mathbf{E}[g_n(t, X(t))]$$

= $n\mathbf{E}\left[\int_0^t X^{n-1}(s)dX(s) + \frac{1}{2}n(n-1)\int_0^t X^{n-2}(s)ds\right]$

By linearity of expectation,

$$m_n(t) = n\mathbf{E}\left[\int_0^t X^{n-1}(s)dX(s)\right] + \frac{1}{2}n(n-1)\mathbf{E}\left[\int_0^t X^{n-2}(s)ds\right]$$

Recall that $\int_0^t X^{n-1}(s)dX(s)$ is an Itô integral and it is therefore a martingale, so $\mathbf{E}[\int_0^t X^{n-1}(s)dX(s)] = 0$.

Interchanging the order of integration to take the expectation inside the integral (that's Fubini's theorem), we finally get

$$m_n(t) = \frac{1}{2}n(n-1)\int_0^t \mathbf{E}\left[X^{n-2}(s)\right]ds$$

= $\frac{1}{2}n(n-1)\int_0^t m_{n-2}(s)ds$

Now let's apply this formula for n = 4:

$$m_4(t) = 6 \int_0^t \mathbf{E} \left[X^2(s) \right] ds$$
$$= 6 \int_0^t s ds$$
$$= 3t^2$$

What about n = 6?

$$m_6(t) = 15 \int_0^t \mathbf{E} \left[X^4(s) \right] ds$$
$$= 45 \int_0^t s^2 ds$$
$$= 15t^3$$

3. Let $Y_t = X_t^4$ where X_t is a Brownian motion. Using Itô's lemma, express the SDE for Y_t . Then, deduce the stochastic integral for Y_t over [0,T]. Finally, deduce from the stochastic integral an expression for $\mathbf{E}[Y_t]$.

First, note that $Y_t = f(X_t)$ where $f(x) = x^4$. Hence,

$$\frac{\partial f}{\partial t} = 0$$

$$\frac{\partial f}{\partial x} = 4x^3$$

$$\frac{\partial f}{\partial x} = 12x^2$$

By Itô's lemma,

$$dY_t = 6X_t^2 dt + 4X_t^3 dX_t$$

Since X_t is a Brownian motion then $X_0 = 0$ and therefore $Y_0 = 0$. Thus, integrating the SDE over [0, T], we get

$$Y_T = 6 \int_0^T X_t^2 dt + 4 \int_0^T X_t^3 dX_t$$

taking the expectation and by linearity of the expectation operator,

$$\mathbf{E}[Y_T] = 6\mathbf{E} \left[\int_0^T X_t^2 dt \right] + 4\mathbf{E} \left[\int_0^T X_t^3 dX_t \right]$$

Now, the Itô integral $\int_0^T X_t^3 dX_t$ is a martingale and hence $\mathbf{E}\left[\int_0^T X_t^3 dX_t\right] = 0$. Also, by Fubini's Theorem, we can change the order of integration and therefore slide the expectation inside $\int_0^T X_t^2 dt$. Hence,

$$\mathbf{E}[Y_T] = 6 \int_0^T \mathbf{E}\left[X_t^2\right] dt$$

Now, $\mathbf{E}[X_t^2] = \mathbf{E}[(X_t - X_0)^2] = t - 0 = t$. Therefore,

$$\mathbf{E}[Y_T] = 6 \int_0^T t dt = 3T^2$$

4. Let $X_n, n = 1, ...$ be i.i.d random variables where $P(X_n = 1) = p$ and $P(X_n = -1) = 1 - p$. You can think of X_n as being the nth coin toss in a sequence. Let $S_n, n = 1, ...$ be the associated random walk, defined as

$$S_n = X_1 + X_2 + \ldots + X_n \tag{4}$$

 S_n can be viewed as the P&L of the entire coin toss game. We also introduce the filtration \mathcal{F}_n generated by the X_n and such that X_n is \mathcal{F}_n -adapted.

Find conditions under which the random walk is (a) a martingale, (b) a submartingale (c) a supermartingale.

Answer:

Let's start from the conditional expectation

$$\mathbf{E}\left[S_n|\mathcal{F}_{n-1}\right] = \mathbf{E}\left[S_{n-1} + X_n|\mathcal{F}_{n-1}\right]$$

Now, S_{n-1} is \mathcal{F}_{n-1} adapted because X_{n-1} and therefore X_{n-2}, \ldots, X_1 are \mathcal{F}_{n-1} adapted. As a result,

$$\mathbf{E}[S_{n-1} + X_n | \mathcal{F}_{n-1}] = S_{n-1} + \mathbf{E}[X_n | \mathcal{F}_{n-1}]$$

Because the X_n are i.i.d. random variables,then X_n is independent from \mathcal{F}_{n-1} and as a result

$$\mathbf{E}\left[X_n|\mathcal{F}_{n-1}\right] = \mathbf{E}\left[X_n\right] \\ = 2p - 1$$

Thus,

$$\mathbf{E}[S_n|\mathcal{F}_{n-1}] = S_{n-1} + 2p - 1$$

We can now solve the question:

(a) S_n is martingale iff

$$\mathbf{E}\left[S_n|\mathcal{F}_{n-1}\right] = S_n$$

which only occurs if $p = \frac{1}{2}$;

(b) S_n is submartingale iff

$$\mathbf{E}\left[S_n|\mathcal{F}_{n-1}\right] > S_n$$

which only occurs if $p > \frac{1}{2}$;

(c) S_n is supermartingale iff

$$\mathbf{E}\left[S_n|\mathcal{F}_{n-1}\right] < S_n$$

which only occurs if $p < \frac{1}{2}$.

5. **Discrete Time Martingale**: Let Y_1, \ldots, Y_n be a sequence of independent random variables such that $\mathbf{E}[Y_i] = 0$ for $i = 1, \ldots, n$. Let \mathcal{F}_n be the filtration generated by the sequence Y_1, \ldots, Y_n . Consider the random variable $S_n = \sum_{i=1}^n Y_i$. Prove that S_n is a martingale for all n.

Reminder - proving that a process S_n is a martingale involves proving that $\mathbf{E}[|S_n|] < \infty$ and that $\mathbf{E}[S_{n+1}|\mathcal{F}_n] = S_n$

First,

$$\mathbf{E}[|S_n|] = \mathbf{E}[|Y_1 + Y_2 + \dots + Y_n|]$$

$$\leq \mathbf{E}[|Y_1| + |Y_2| + \dots + |Y_n|]$$

$$= \mathbf{E}[|Y_1|] + \mathbf{E}[|Y_2|] + \dots + \mathbf{E}[|Y_n|]$$

$$< \infty$$

since we have a finite sum of finite finite numbers.

Second,

$$\mathbf{E}[S_{n+1}|\mathcal{F}_n] = \mathbf{E}[S_n + Y_{n+1}|\mathcal{F}_n]$$
$$= \mathbf{E}[S_n|\mathcal{F}_n] + \mathbf{E}[Y_{n+1}|\mathcal{F}_n]$$

by linearity of the expectation operator.

Now, since S_n is \mathcal{F}_n -measurable (i.e. if we have the filtration \mathcal{F}_n we know what S_n is), then $\mathbf{E}[S_n|\mathcal{F}_n] = S_n$.

Also, since $Y_1, \ldots, Y_n, Y_{n+1}$ are independent, then Y_{n+1} is independent from \mathcal{F}_n and hence $\mathbf{E}[Y_{n+1}|\mathcal{F}_n] = \mathbf{E}[Y_{n+1}] = 0$.

Therefore,

$$\mathbf{E}[S_{n+1}|\mathcal{F}_n] = \mathbf{E}[S_n]$$

and we can conclude that S_n is a martingale for all n.