

Exer. 2.3

If M_n is a func. of ξ_1, \dots, ξ_n , then

$\varphi(M_n)$ is a func. of ξ_1, \dots, ξ_n ,

so $\{M_n\}_{n=0}^N$ is adapted. \checkmark

By Thm 2.3.2 (v) (conditional Jensen's inequ.),

$$E_n[\varphi(M_{n+1})] \geq \varphi(E_n[M_{n+1}])$$

Since $\{M_n\}_{n=0}^N$ is a mart.,

$$E_n[M_{n+1}] = M_n.$$

Therefore,

$$E_n[\varphi(M_{n+1})] \geq \varphi(M_n),$$

so $\{\varphi(M_n)\}_{n=0}^N$ is a submart.

$$(i) E_n[M_{n+1}] = E_n[M_n + X_{n+1}]$$

$$\stackrel{(i)}{=} E_n[M_n] + E_n[X_{n+1}]$$

$$\stackrel{(ii)}{=} M_n + E_n[X_{n+1}]$$

$$\stackrel{(iv)}{=} M_n + E[X_{n+1}]$$

$$= M_n + 0 = M_n.$$

$$(ii) E_n[S_{n+1}] = E_n \left[e^{\sigma(M_n + X_{n+1})} \left(\frac{2}{e^\sigma + e^{-\sigma}} \right)^{n+1} \right]$$

$$= \left(\frac{2}{e^\sigma + e^{-\sigma}} \right)^{n+1} E_n [e^{\sigma M_n} e^{\sigma X_{n+1}}]$$

$$\stackrel{(ii)+(iv)}{=} \left(\frac{2}{e^\sigma + e^{-\sigma}} \right)^{n+1} e^{\sigma M_n} E [e^{\sigma X_{n+1}}]$$

$$= \left(\frac{2}{e^\sigma + e^{-\sigma}} \right) S_n \left(\frac{1}{2} e^\sigma + \frac{1}{2} e^{-\sigma} \right) = S_n.$$

Exer. 2.5

$$\begin{aligned}
(i) M_n^2 &= \sum_{j=0}^{n-1} (M_{j+1}^2 - M_j^2) \\
&= \sum_{j=0}^{n-1} (M_{j+1} - M_j)(M_{j+1} + M_j) \\
&= \sum_{j=0}^{n-1} (M_{j+1} - M_j)((M_{j+1} - M_j) + 2M_j) \\
&= \sum_{j=0}^{n-1} (M_{j+1} - M_j)^2 + 2 \sum_{j=0}^{n-1} M_j (M_{j+1} - M_j) \\
&= \left(\sum_{j=0}^{n-1} X_{j+1}^2 \right) + 2I_n \\
&= \left(\sum_{j=0}^{n-1} 1 \right) + 2I_n = n + 2I_n \\
\Rightarrow I_n &= \frac{1}{2} M_n^2 - \frac{n}{2}
\end{aligned}$$

$$(ii) E_n[f(I_{n+1})]$$

$$= E_n\left[f\left(\frac{M_{n+1}^2 - n}{2}\right)\right]$$

$$= E_n\left[f\left(\frac{(M_n + X_{n+1})^2 - n}{2}\right)\right] = h(M_n),$$

where

$$h(m) = E\left[f\left(\frac{(m + X_{n+1})^2 - n}{2}\right)\right]$$

$$= \frac{1}{2} f\left(\frac{(m+1)^2 - n}{2}\right) + \frac{1}{2} f\left(\frac{(m-1)^2 - n}{2}\right).$$

Note that

$$h(-m) = \frac{1}{2} f\left(\frac{(-m+1)^2 - n}{2}\right) + \frac{1}{2} f\left(\frac{(-m-1)^2 - n}{2}\right)$$

$$= \frac{1}{2} f\left(\frac{(m-1)^2 - n}{2}\right) + \frac{1}{2} f\left(\frac{(m+1)^2 - n}{2}\right)$$

$$= h(m),$$

so $h(m) = h(|m|)$ for all m .

p.5

Define $g(i) = h(\sqrt{2i+n})$.

Since $|M_n| = \sqrt{2I_n+n}$, we get

$$g(I_n) = h(|M_n|) = h(M_n) = E_n[f(I_{n+1})].$$

Exer. 2.6

By the mart. prop., $E_n[M_{n+1}] = M_n$.

Also, $E_n[M_n] = M_n$.

So $E_n[M_{n+1} - M_n] = 0$. Now...

$$E_n[I_{n+1}] = E_n[I_n + \Delta_n(M_{n+1} - M_n)]$$

$$= E_n[I_n] + E_n[\Delta_n(M_{n+1} - M_n)]$$

$$= I_n + \Delta_n E_n[M_{n+1} - M_n]$$

$$= I_n + \Delta_n \cdot 0 = I_n$$

Exer. 2.7

$$S_0 = 4, u = 2, d = \frac{1}{2}, r = 0, N = 2$$

$$\tilde{p} = \frac{1+r-d}{u-d} = \frac{(\frac{1}{2})}{(\frac{3}{2})} = \frac{1}{3}, \tilde{q} = \frac{2}{3}$$

$$\text{Let } \Delta_0 = 0, \Delta_1 = 1_{\{\xi_1 = +\}}.$$

Since $r = 0$, $\{S_n\}$ is a mart. under \tilde{P} .

By Exer. 2.6, if $I_0 = 0$ and

$$I_n = \sum_{j=0}^{n-1} \Delta_j (S_{j+1} - S_j),$$

then $\{I_n\}$ is a mart. under \tilde{P} .

We will show $\{I_n\}$ is not Markov under \tilde{P} .

First,

$$I_0 = 0$$

$$I_1 = \Delta_0(S_1 - S_0) = 0$$

$$I_2 = \Delta_0(S_1 - S_0) + \Delta_1(S_2 - S_1)$$

$$= \Delta_1(S_2 - S_1)$$

$$= \begin{cases} S_2 - S_1 & \text{if } \xi_1 = H, \\ 0 & \text{if } \xi_1 = T. \end{cases}$$

$$I_2(H, H) = S_2(H, H) - S_1(H, H)$$

$$= 16 - 8 = 8$$

$$I_2(H, T) = S_2(H, T) - S_1(H, T)$$

$$= 4 - 8 = -4$$

$$I_2(T, H) = I_2(T, T) = 0$$

Suppose $\{I_n\}$ is Markov under \tilde{P} .

Then

$$\tilde{E}_1[I_2^2] = g(I_1) = \underbrace{g(0)}_{\substack{\text{constant,} \\ \text{not random}}} \text{ for some } g.$$

$$\therefore \tilde{E}_1[I_2^2](H) = \tilde{E}_1[I_2^2](\tau).$$

$$\begin{aligned} \tilde{E}_1[I_2^2](H) &= \tilde{E}[I_2^2 | \xi_1 = H] \\ &= \tilde{E}[I_2(H, \xi_2)^2] \end{aligned}$$

$$= \tilde{p} I_2(H, H)^2 + \tilde{q} I_2(H, \tau)^2$$

$$= \frac{1}{3} \cdot 8^2 + \frac{2}{3} (-4)^2$$

$$= \frac{64}{3} + \frac{32}{3} = \frac{96}{3} = 32$$

$$\begin{aligned}\tilde{E}_1 [I_2^2] (T) &= \tilde{p} I_2(T, H)^2 + \tilde{q} I_2(T, T)^2 \\ &= 0,\end{aligned}$$

contradiction.

Exer. 2.11

(i) For all $x \in \mathbb{R}$

$$(x - K) + (K - x)^+ = \begin{cases} x - K + K - x & \text{if } K > x, \\ x - K & \text{if } K \leq x \end{cases}$$

$$= \begin{cases} x - K & \text{if } x \geq K, \\ 0 & \text{if } x < K \end{cases}$$

$$= (x - K)^+.$$

Therefore,

$$F_N + P_N = (S_N - K) + (K - S_N)^+ \\ = (S_N - K)^+ = C_N$$

(ii)

$$F_n + P_n = \tilde{\mathbb{E}}_n \left[\frac{F_N}{(1+r)^{N-n}} \right] + \tilde{\mathbb{E}}_n \left[\frac{P_N}{(1+r)^{N-n}} \right] \\ = \tilde{\mathbb{E}}_n \left[\frac{F_N + P_N}{(1+r)^{N-n}} \right] \\ = \tilde{\mathbb{E}}_n \left[\frac{C_N}{(1+r)^{N-n}} \right] = C_n$$

(iii)

$$F_0 = \mathbb{E} \left[\frac{F_N}{(1+r)^N} \right] = \tilde{\mathbb{E}} \left[\frac{S_N - K}{(1+r)^N} \right] \\ = \tilde{\mathbb{E}} \left[\frac{S_N}{(1+r)^N} \right] - \frac{K}{(1+r)^N} = S_0 - \frac{K}{(1+r)^N}$$

↑
because $\left\{ \frac{S_n}{(1+r)^n} \right\}$ is
a mart. under $\tilde{\mathbb{P}}$

Exer. 2.13

(i) Since $S_{n+1} = Y_{n+1} S_n$ in class, let's use T_n instead for the sum, i.e.

$$T_n = \sum_{k=0}^n S_k \text{ and we want to show}$$

that $\{(S_n, T_n)\}$ is Markov (under \tilde{P}).

$$\begin{aligned} & \tilde{E}_n [f(S_{n+1}, T_{n+1})] \\ &= \tilde{E}_n [f(Y_{n+1} S_n, T_n + S_{n+1})] \\ &= \tilde{E}_n [f(Y_{n+1} S_n, T_n + Y_{n+1} S_n)] \\ &= g(S_n, T_n), \end{aligned}$$

where

$$g(s, t) = \tilde{E} [f(Y_{n+1} s, t + Y_{n+1} s)]$$

$$= \tilde{p} f(us, t + us) + \tilde{q} f(ds, t + ds)$$

$$(ii) \quad V_N = f\left(\frac{1}{N+1} \sum_{n=0}^N S_n\right) = f\left(\frac{T_N}{N+1}\right)$$

$$\boxed{v_n(s, t) = f\left(\frac{t}{N+1}\right)}$$

$$V_n = \frac{1}{1+r} \tilde{E}_n [V_{n+1}]$$

$$= \frac{1}{1+r} \tilde{E}_n [v_{n+1}(S_{n+1}, T_{n+1})]$$

$$= \frac{1}{1+r} \tilde{E}_n [v_{n+1}(Y_{n+1} S_n, T_n + Y_{n+1} S_n)]$$

$$v_n(s, t) = \frac{1}{1+r} \tilde{E} [v_{n+1}(Y_{n+1} s, t + Y_{n+1} s)]$$

$$V_n(s, t) = \frac{1}{1+r} \left(\tilde{p} v_{n+1}(us, t+us) + \tilde{q} v_{n+1}(ds, t+ds) \right)$$