

MATH 285, HW#7, SOLUTIONS, SPRING 07

1.  $\{X_i\}$  are iid  $P(X_i = 1) = p$ ,  $P(X_i = -1) = q$ ,  $p < 1/2$

$$S_n = x + \sum_{i=1}^n X_i$$

a)  $M_n = S_n + (q-p)n$  is a martingale wrt.

$$\mathcal{F}_n = \sigma(S_0, \dots, S_n)$$

i)  $M_n \in \mathcal{F}_n$ . Let  $\mathcal{G}_n = \sigma(M_0, \dots, M_n)$ .

Then  $M_n \in \mathcal{G}_n$  by definition of  $\mathcal{G}_n$ .  $\mathcal{G}_n = \mathcal{F}_n$

since at any time  $n$   $S_n$  and  $M_n$  differ by a deterministic constant. So  $M_n \in \mathcal{F}_n$ .

ii)  $E[|M_n|] < \infty$  for all  $n$ .

$$|M_n| \leq |x| + \left| \sum_{i=1}^n X_i \right| + |q-p|n \leq |x| + n + 2n$$

so  $E[|M_n|] < |x| + 3n < \infty$  for any  $n$ .

$M_n$  is integrable.

$$\text{iii) } E[M_{n+1} | \mathcal{F}_n] = E[S_n + X_{n+1} + (q-p)(n+1) | \mathcal{F}_n] =$$

$$S_n + E[X_{n+1}] + (q-p)(n+1) = S_n + (p-q) +$$

$$(q-p)(n+1) = S_n + (q-p)n = M_n.$$

The second equality is true because  $S_n \in \mathcal{F}_n$ ,

$X_{n+1} \perp \mathcal{F}_n$ ,  $(q-p)(n+1)$  is a constant and

conditional expectation is a linear operator.

b)  $M_n = \left(\frac{q}{p}\right)^{S_n}$  is a martingale wrt.

$$\mathcal{F}_n = \sigma(M_0, \dots, M_n).$$

i)  $M_n \in \mathcal{F}_n$  by definition of  $\mathcal{F}_n$

ii)  $E|M_n| < \infty$  for all  $n$ . Since  $q > p$   $\frac{q}{p} > 1$ .

$$|S_n| \leq |x| + \left| \sum_{i=1}^n \Delta_i \right| \leq x + \sum_{i=1}^n |\Delta_i| \leq x + n$$

(Note  $x > 0$  so  $|x| = x$ ).

$$\text{Since } \frac{q}{p} > 1 \quad \left| \left(\frac{q}{p}\right)^{S_n} \right| \leq \left(\frac{q}{p}\right)^{|S_n|} \leq \left(\frac{q}{p}\right)^{x+n}$$

so  $E|M_n| < \left(\frac{q}{p}\right)^{x+n} < \infty$  for all  $n$ .

$$\begin{aligned} \text{iii) } E[M_{n+1} | \mathcal{F}_n] &= E\left[\left(\frac{q}{p}\right)^{S_n} \left(\frac{q}{p}\right)^{\Delta_{n+1}} | \mathcal{F}_n\right] \\ &= \left(\frac{q}{p}\right)^{S_n} E\left[\left(\frac{q}{p}\right)^{\Delta_{n+1}}\right] \text{ since } S_n \in \mathcal{F}_n \text{ and} \end{aligned}$$

$$\begin{aligned} \Delta_{n+1} \perp \mathcal{F}_n &= \left(\frac{q}{p}\right)^{S_n} \left(\frac{q}{p} \cancel{p} \cancel{q} \cancel{p} \cancel{q}\right) = \\ &= \left(\frac{q}{p}\right)^{S_n} = M_n \end{aligned}$$

c)  $T = \inf\{n \geq 0 : S_n = 0 \text{ or } \pm\}$  is a stopping time for  $\mathcal{F}_n = \sigma(S_0, \dots, S_n)$

$$\{T = k\} = \{S_0 \notin \{0, \pm\}, \dots, S_{k-1} \notin \{0, \pm\},$$

$S_k \in \{0, \pm\}\} \in \mathcal{F}_k$  since  $k$  is arbitrary the result follows.

d) Let  $M_n$  be as in d). We are assuming that  $T < \infty$  a.s. Also whenever

$$n \leq T \quad |M_n| \leq \left(\frac{q}{p}\right)^{d+x} \quad \text{so } d$$

Thm 4.1.1 part ii)  $EM_T = EM_0$ .

$$EM_0 = \left(\frac{q}{p}\right)^x = EM_T$$

$$EM_T = \left(\frac{q}{p}\right)^d \mathbb{P}(S_T = d) + \mathbb{P}(S_T = 0)$$

We have that  $\mathbb{P}(S_T = d) + \mathbb{P}(S_T = 0) = 1$ .

We get that

$$\mathbb{P}(S_T = 0) = \frac{\left(\frac{q}{p}\right)^x - \left(\frac{q}{p}\right)^d}{1 - \left(\frac{q}{p}\right)^d}$$

$\{S_T = 0\} = \{S_n \text{ reaches } 0 \text{ before } d\}$ .

2.  $X = \{X_t, t \geq 0\}$  is a standard  
1 dim'l BM. Want to show that

$Y = \{Y_t, t \geq 0\} = \left\{ \frac{X_{at}}{\sqrt{a}}, t \geq 0 \right\}$  is also a  
standard 1 dim'l BM.

i)  $Y_0 = X_0 = 0$  as

ii) for any  $0 = t_0 < t_1 < \dots < t_n$

$$\{X_{at_1} - X_{at_0}, X_{at_2} - X_{at_1}, \dots, X_{at_n} - X_{at_{n-1}}\}$$

are independent. Scaling by  $\frac{1}{\sqrt{a}}$  does

not change independence so

$$\left\{ \frac{1}{\sqrt{a}} (X_{at_1} - X_{at_0}), \dots, \frac{1}{\sqrt{a}} (X_{at_n} - X_{at_{n-1}}) \right\}$$

$$= \{Y_{t_1} - Y_{t_0}, \dots, Y_{t_n} - Y_{t_{n-1}}\} \text{ are independent.}$$

iii)  $E(Y_t - Y_s) = \frac{1}{\sqrt{a}} E(X_{at} - X_{as}) = 0$

$$\begin{aligned} \text{Var}(Y_t - Y_s) &= \text{Var}\left(\frac{1}{\sqrt{a}} (X_{at} - X_{as})\right) \\ &= \frac{1}{a} \text{Var}(X_{at} - X_{as}) = \frac{1}{a} a(t-s) \end{aligned}$$

So  $Y_t - Y_s$  is normal rv with mean 0 and variance  $(t-s)$ .

ii)  $t \longrightarrow \frac{X_{at}}{\sqrt{a}}$  is continuous :

$$t \longrightarrow at \longrightarrow X_{at} \longrightarrow \frac{X_{at}}{\sqrt{a}}$$

since it is a composition of continuous mappings.

Note :  $Y_t - Y_s$  is normal because

$X_{at} - X_{as}$  is normal because  $X$  is a standard 1 dim'l BM.

$\{Y_t, t \geq 0\} = \left\{ \frac{X_{at}}{\sqrt{a}}, t \geq 0 \right\}$  is a standard 1 dim'l BM.

3.  $X = \{X_t, t \geq 0\}$  is a standard 1 dim BM. Compute  $\mathbb{P}(X_2 > 0 \mid X_1 > 0)$

$$\begin{aligned}\mathbb{P}(X_2 > 0 \mid X_1 > 0) &= \frac{\mathbb{P}(X_2 > 0, X_1 > 0)}{\mathbb{P}(X_1 > 0)} \\ &= 2 \mathbb{P}(X_2 > 0, X_1 > 0) \quad \text{Since } \mathbb{P}(X_1 > 0) \\ &= 1/2 \quad \text{by symmetry of BM}\end{aligned}$$

$$\mathbb{P}(X_2 > 0, X_1 > 0) = \mathbb{P}((X_2 - X_1) + X_1 > 0, X_1 > 0)$$

Let  $Y = (X_2 - X_1)$

Then  $Y \sim N(0, 1)$  and  $Y \perp\!\!\!\perp X_1$ .

$$\mathbb{P}(X_2 > 0, X_1 > 0) = \mathbb{P}(Y > -X_1, X_1 > 0)$$

The joint density of  $(X_1, Y)$  is

$$f(x, y) = \frac{1}{2\pi} \exp\left(-\frac{(x^2 + y^2)}{2}\right)$$

So

$$\mathbb{P}(Y > -X_1, X_1 > 0) = \int_0^{\infty} \int_{-x}^{\infty} f(x, y) dy dx$$

$$= \int_0^{\infty} \int_{-x}^{\infty} \frac{1}{2\pi} \exp\left(-\frac{(x^2+y^2)}{2}\right) dy dx$$

switching to polar coordinates

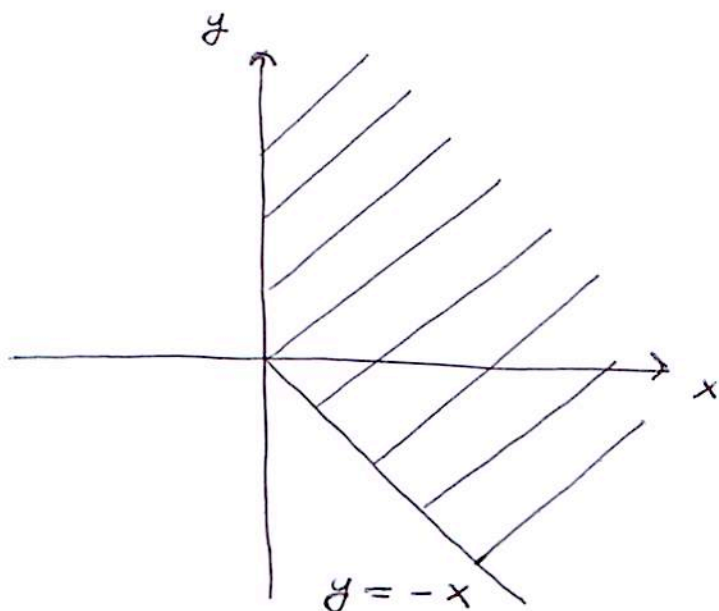
$$= \frac{1}{2\pi} \int_{-\pi/4}^{\pi/2} \int_0^{\infty} \exp\left(-\frac{r^2}{2}\right) r dr d\theta$$

$$u = r^2 \quad du = 2r dr \quad r dr = \frac{du}{2}$$

$$= \frac{1}{4\pi} \int_{-\pi/4}^{\pi/2} \int_0^{\infty} \exp\left(-\frac{u}{2}\right) du d\theta$$

$$= \frac{1}{4\pi} \int_{-\pi/4}^{\pi/2} -2 \exp\left(-\frac{u}{2}\right) \Big|_0^{\infty} d\theta$$

$$= \frac{1}{2\pi} \int_{-\pi/4}^{\pi/2} 1 d\theta = \frac{1}{2\pi} \left( \frac{\pi}{2} + \frac{\pi}{4} \right) = \frac{3}{8}$$



So

$$P(X_2 > 0 | X_1 > 0)$$

$$= 3/4$$