## Probability theory II Assignment 3

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4.6.6 Let  $X_n \in [0, 1]$  be adapted to  $\mathcal{F}_n$ . Let  $\alpha, \beta > 0$  with  $\alpha + \beta = 1$  and suppose  $P(X_{n+1} = \alpha + \beta X_n | \mathcal{F}_n) = X_n \qquad P(X_{n+1} = \beta X_n | \mathcal{F}_n) = 1 - X_n$ Show  $P(\lim_n X_n = 0 \text{ or } 1) = 1$  and if  $X_0 = \theta$ , then  $P(\lim_n X_n = 1) = \theta$ .

*Proof.*  $X_n \in \mathcal{F}_n$  and  $0 \le X_n \le 1 \quad \forall n \in \mathbb{N}$ . By assumption,

$$X_{n=1}|\mathcal{F}_n = \begin{cases} \alpha + \beta X_n & \text{with prob. } X_n \\ \beta X_n & \text{with prob. } 1 - X_n \end{cases}$$

Thus, we have

$$E[X_{n+1}|\mathcal{F}_n] = (\alpha + \beta X_n)X_n + \beta X_n(1 - X_n) = (\alpha + \beta)X_n = X_n$$

Therefore,  $X_n$  is a martingale. Since  $X_n$  is uniformly bounded by 0 and 1, we can say  $X_n$  is uniformly integrable martingale. Also,  $\sup_n E|X_n| \leq 1$  and by martingale convergence thm,  $X_n \to X$  a.s. for some integrable X. By Vitali lemma,  $X_n \to X$  in  $L^1$  and  $E(X_n) \to E(X)$ . Note that since  $X_n$  is a martingale. we have  $\theta = E(X_0) = E(X_1) = \cdots = E(X_n) \quad \forall n \in \mathbb{N}$ . Combining these, we have  $\theta = E(X_0) = E(X_n) = E(X) \quad \forall n \in \mathbb{N}$ 

Now consider the case that  $X_n = x$  where 0 < x < 1. Then

$$X_{n+1}|X_n = \begin{cases} \alpha + \beta x & \text{with prob } x\\ \beta x & \text{with prob } 1 - x \end{cases}$$

Thus, if 0 < x < 1 then the sequence  $X_n$  goes out of  $\varepsilon$ -ball containing x at the very next step  $X_{n+1}$ . Therefore, the limit X cannot have value  $x \in (0,1)$ . On the other hand, if  $X_n = 0$  then

$$X_{n+1}|X_n = \begin{cases} \alpha & \text{with prob } 0\\ 0 & \text{with prob } 1 \end{cases} = 0 \ a.s.$$

Else, if  $X_n = 1$  then

$$X_{n+1}|X_n = \begin{cases} \alpha + \beta = 1 & \text{with prob } 1\\ \beta & \text{with prob } 0 \end{cases} = 1 \ a.s.$$

Therefore X = 0 or 1 a.s. and  $\theta = E(X) = 0 \cdot P(X = 0) + 1 \cdot P(X = 1)$  so  $P(X = 1) = \theta$ 

**4.6.7** Show that if  $\mathcal{F}_n \uparrow \mathcal{F}_\infty$  and  $Y_n \to Y$  in  $L^1$ , then  $E(Y_n | \mathcal{F}_n) \to E(Y | \mathcal{F}_\infty)$  in  $L^1$ .

*Proof.* We want to show the  $L^1$  convergence i.e.

$$E|E(Y_n|\mathcal{F}_n) - E(Y|\mathcal{F}_\infty)| \to 0$$

Note that by triangle inequality, we have

$$E|E(Y_n|\mathcal{F}_n) - E(Y|\mathcal{F}_\infty)| \le E|E(Y_n|\mathcal{F}_n) - E(Y|\mathcal{F}_n)| + E|E(Y|\mathcal{F}_n) - E(Y|\mathcal{F}_\infty)|$$

By Levy's thm,  $E(Y|\mathcal{F}_n) \to E(Y|\mathcal{F}_\infty)$  in  $L^1$  so that  $E|E(Y|\mathcal{F}_n) - E(Y|\mathcal{F}_\infty)| \to 0$ Also,

$$E|E(Y_n|\mathcal{F}_n) - E(Y|\mathcal{F}_n)| \le E[E[|Y_n - Y||\mathcal{F}_n]] = E|Y_n - Y| \to 0$$

by the assumption that  $Y_n \to Y$  in  $L^1$ . Therefore

$$E|E(Y_n|\mathcal{F}_n) - E(Y|\mathcal{F}_\infty)| \to 0$$

In other words,  $E(Y|\mathcal{F}_n) \to E(Y|\mathcal{F}_\infty)$  in  $L^1$ 

4.8.1 Generalize Theorem 4.8.2 to show that if  $L \le M$  are stopping times and  $Y_{M \land n}$  is a uniformly integrable submartingale, then  $EY_L \le EY_M$  and

$$Y_L \leq E(Y_M | \mathcal{F}_L)$$

*Proof.* We can prove this thm on four steps.

i.  $E(Y_0) \leq E(Y_N) \leq E(Y_\infty)$  for any stopping time NFor uniformly integrable submartingale  $Y_n$ , we have  $Y_n \to Y_\infty$  a.s. for some integrable r.v.  $Y_\infty$ . Also, for a stopping time N,  $X_n = Y_{N \wedge n}$  is also a uniformly integrable submartingale so that  $X_n \to X_\infty$  a.s. for some integrable r.v.  $X_\infty$ . Note that  $X_\infty = Y_N$  a.s. (This is derived when we consider two cases where  $N = \infty$  and  $N < \infty$ ). Applying Vitali lemma to both  $Y_n \to Y_\infty$  and  $Y_{N \wedge n} \to Y_N$ , we have  $E(Y_n) \to E(Y_\infty)$  and  $E(Y_{N \wedge n}) \to E(Y_N)$ . Using bounded optional stopping thm, we have

$$E(Y_0) \le E(Y_{N \wedge n}) \le E(Y_n)$$

Taking  $n \to \infty$ , we have

$$E(Y_0) < E(Y_N) < E(Y_\infty)$$

ii.  $E(Y_L) \leq E(Y_M)$  given  $L \leq M$ 

For uniformly integrable submartingale  $Y_n$ , we have  $Y_n \to Y_\infty$  a.s. for some integrable r.v.  $Y_\infty$ .  $X_n = Y_{M \wedge n}$  is also a uniformly integrable submartingale so that  $X_n \to X_\infty$  a.s. for some integrable r.v.  $X_\infty$ . Note that  $X_\infty = Y_M$  a.s. By the result of the first step, we have

$$E(X_0) < E(X_L) < E(X_\infty)$$

Since  $X_L = Y_{M \wedge L} = Y_L$  and  $X_{\infty} = Y_M$  a.s. we have

$$E(Y_L) \leq E(Y_M)$$

- iii. If  $A \in \mathcal{F}_L$  then  $N = L \cdot I_A + M \cdot I_{A^C}$  is a stopping time. This is illustrated in exercise 4.4.3 of homework 2
- iv. Show  $E[Y_L I_A] \leq E[Y_M I_A] \quad \forall A \in \mathcal{F}_L$ This is illustrated in exercise 4.4.4 of homework 2

## 4.8.2 If $X_n \ge 0$ is a supermartingale, then $P(\sup X_n > \lambda) \le EX_0/\lambda$ .

*Proof.* We have learned in class that for nonnegative supermartingale  $X_n$ , if N is a stopping time then  $E(X_0) \ge E(X_N)$ . Note that  $N = \inf\{n : X_n > \lambda\}$  is a stopping time. On  $(N < \infty)$ , we have  $X_N > \lambda$  so

$$X_N I(N < \infty) \ge \lambda I(N < \infty)$$
  $E[X_N I(N < \infty)] \ge \lambda P(N < \infty)$ 

Since  $X_n \ge 0$ , we get  $E(X_0) \ge E(X_N) \ge E[X_N I(N < \infty)] \ge \lambda P(N < \infty)$ . Hence

$$P(N < \infty) \le \frac{E[X_0]}{\lambda}$$

Note that

$$(N < \infty) = (\{n : X_n > \lambda\} \neq \phi) = (X_n > \lambda \text{ for some } n \in \mathbb{N}) = (\sup_n X_n > \lambda)$$

Therefore, we can conclude that

$$P(\sup_{n} X_n > \lambda) \le \frac{E[X_0]}{\lambda}$$

4.8.4 Wald's second equation. Let  $S_n = \xi_1 + \cdots + \xi_n$ , where the  $\xi_i$  are independent with  $E\xi_i = 0$  and var  $(\xi_i) = \sigma^2$ . Use the martingale from the previous problem to show that if T is a stopping time with  $ET < \infty$ , then  $ES_T^2 = \sigma^2 ET$ .

*Proof.* As we can see in the example 4.2.2 in the textbook,  $M_n = S_n^2 - n\sigma^2$  is a martingale . Since T is a stopping time ,  $M_{T \wedge n}$  is also a martingale . Let  $P_n = M_{T \wedge n}$  . Then  $E(P_1) = E(P_n)$  by the property of martingale.

$$E(P_1) = E(M_1) = E(S_1^2 - \sigma^2) = E(\xi_1^2 - \sigma^2) = 0$$

$$E(P_n) = E(M_{T \wedge n}) = E[S_{n \wedge T}^2 - (n \wedge T)\sigma^2]$$

Thus we have  $E[S^2_{T \wedge n}] = \sigma^2 E[T \wedge n] \quad \forall n \in \mathbb{N}$ . Note that  $E(T) < \infty$  assumption implies that  $T < \infty$  a.s. Thus  $S_{T \wedge n} \to S_T$  a.s. Also, since  $T \wedge n \nearrow T$  and  $E[T \wedge n] \nearrow E[T]$  by MCT, we get

$$\sup_{n} E[S_{T \wedge n}^{2}] = \sigma^{2} \sup_{n} E[T \wedge n] = \sigma^{2} E[T] < \infty$$

Hence, we have  $\sup_n E[S^2_{T \wedge n}] < \infty$ . Note that  $S_n$  is also a martingale so that  $S_{T \wedge n}$  is a martingale too. By martingale  $L^p$  convergence thm,  $S_{T \wedge n} \to S_T$  a.s. and in  $L^2$ . This  $L^2$  convergence gives us the fact that  $E[S^2_{T \wedge n}] \to E[S^2_T]$ . Therefore, taking  $n \to \infty$  for  $E[S^2_{T \wedge n}] = \sigma^2 E[T \wedge n] \quad \forall \, n \in \mathbb{N}$ , we finally get

$$E[S_T^2] = \sigma^2 E[T]$$

4.8.3 Let  $S_n = \xi_1 + \dots + \xi_n$  where the  $\xi_i$  are independent with  $E\xi_i = 0$  and  $\text{var}(X_i) = \sigma^2$ .  $S_n^2 - n\sigma^2$  is a martingale. Let  $T = \min\{n : |S_n| > a\}$ . Use Theorem 4.8.2 to show that  $ET > a^2/\sigma^2$ .

*Proof.* Notice that T is a stopping time. If  $E(T) = \infty$  then the inequality trivially holds so assume that  $E(T) < \infty$ . Then, by the Wald's second equation we proved in the previous exercise, we have  $E[S_T^2] = \sigma^2 E[T]$ . Observe that on  $(T < \infty)$ , we get  $S_T^2 > a^2$ . Hence we have

$$E[S_T^2] \ge E[S_T^2 I(T < \infty)] \ge a^2 E[I(T < \infty)] = a^2 P(T < \infty)$$

The assumption  $E[T] < \infty$  implies that  $T < \infty$  a.s. thus  $P(T < \infty) = 1$ . Therefore,

$$E[T] = E[S_T^2]/\sigma^2 \ge a^2/\sigma^2$$

4.8.5 Variance of the time of gambler's ruin. Let  $\xi_1, \xi_2, \ldots$  be independent with  $P(\xi_i = 1) = p$  and  $P(\xi_i = -1) = q = 1 - p$ , where p < 1/2. Let  $S_n = S_0 + \xi_1 + \cdots + \xi_n$  and let  $V_0 = \min\{n \ge 0 : S_n = 0\}$ . Theorem 4.8.9 tells us that  $E_x V_0 = x/(1-2p)$ . The aim of this problem is to compute the variance of  $V_0$ . If we let  $Y_i = \xi_i - (p-q)$  and note that  $EY_i = 0$  and

$$var(Y_i) = var(X_i) = EX_i u^2 - (EX_i)^2$$

then it follows that  $(S_n - (p-q)n)^2 - n(1 - (p-q)^2)$  is a martingale. (a) Use this to conclude that when  $S_0 = x$  the variance of  $V_0$  is

$$x \cdot \frac{1 - (p - q)^2}{(q - p)^3}$$

(b) Why must the answer in (a) be of the form cx?

Proof. Here x>0 is assumed. (Otherwise, variance becomes negative which is nonsense) Note that  $E[Y_1]=0$  and  $Var[Y_1]=1-(p-q)^2$ .  $S_n-n(p-q)=S_0+Y_1+\cdots+Y_n$ . We have observed that  $S_n^2-n\sigma^2$  is martingale provided  $S_n=\xi_1+\cdots+\xi_n$  and  $\xi_i$  i.i.d. with  $E[\xi_1]=0$  and  $Var[\xi_1]\sigma^2$  Here,  $Y_i$  plays a role of  $\xi_i$  in the sense that it has zero mean. Thus,

$$(S_n - n(p-q))^2 - n(1 - (p-q))^2$$

is a martingale. In the lecture, we have learned that  $T_a < \infty$  a.s. for a < 0. Here, it is replaced by  $V_0 < \infty$  a.s. for 0 < x. (The former has starting point 0 and target point a, while the latter has starting point x and target point 0). Since  $E[V_0] = \frac{x}{1-2p} < \infty$ , we can apply Wald second identity. Let  $\tilde{S}_n = Y_1 + \cdots + Y_n = S_n - n(p-q) - S_0$ . Applying Wald second identity, we have

$$E[\tilde{S}_{V_0}^2] = (1 - (p - q)^2)E[V_0]$$

. Plugging in  $\tilde{S}_n = S_n - n(p-q) - S_0$  and  $E[V_0] = \frac{x}{q-p}$ , we get

$$E[(S_{V_0} - V_0(p-q) - x)^2] = (1 - (p-q)^2) \frac{x}{q-p}$$

Note that  $S_{V_0}=0$  a.s. since  $V_0<\infty$  a.s. Thus, we can derive that

$$E[(S_{V_0} - V_0(p - q) - x)^2] = E[(-V_0(p - q) - x)^2] = E[(V_0(p - q) + x)^2]$$

$$= E[\{(p - q)(V_0 + \frac{x}{p - q})\}^2] = (q - p)^2 E[(V_0 - \frac{x}{q - p})^2]$$

$$= (q - p)^2 E[\{V_0 - E(V_0)\}^2] = (q - p)^2 Var(V_0)$$

Therefore, we have

$$Var(V_0) = \frac{x(1 - (p - q)^2)}{(q - p)^3}$$

4.8.7 Let  $S_n$  be a symmetric simple random walk starting at 0, and let  $T = \inf\{n : S_n \notin (-a,a)\}$ , where a is an integer. Find constants b and c so that  $Y_n = S_n^4 - 6nS_n^2 + bn^2 + cn$  is a martingale, and use this to compute  $ET^2$ .

*Proof.*  $S_n = \xi_1 + \dots + \xi_n$  with  $\xi_i$  i.i.d. where  $P(\xi_1 = 1) = P(\xi_1 = -1) = 0.5$ .  $E(\xi_1) = 0$  and  $Var(\xi_1) = 1$ . Note that  $S_n^2 - n$  is a martingale. Since T is a stopping time,  $S_{T \wedge n}^2 - (T \wedge n)$  is a martingale.

$$E[S^2_{T\wedge n}-(T\wedge n)]=E[S^2_1-1]=E[\xi^2_1]-1=0$$

Hence  $E[S^2_{T \wedge n}] = E[T \wedge n] \quad \forall \, n \in \mathbb{N}$ . Here, we shall claim that  $T < \infty$  a.s.

Note that wherever  $S_n$  lies between (-a, a), if we have 2a consecutive steps of size +1 then we will exit the interval (-a, a). It can be written as

 $(T > m \cdot 2a) \subset (m \text{ times fail of "} 2a \text{ consecutive steps of size } +1")$ 

$$P(T > 2ma) \le \left(1 - \left(\frac{1}{2}\right)^{2a}\right)^m$$

Taking  $m\to\infty$ , we have  $P(T=\infty)=0$ . Hence,  $T<\infty$  a.s. Thus  $S_{T\wedge n}\to S_T$  a.s. and  $S_{T\wedge n}^2\to S_T^2$  a.s. Observe that  $S_T^2=a^2$  a.s. and  $|S_{T\wedge n}^2|\le a^2$  so that applying BCT, we get

$$E[S^2_{T\wedge n}] \to E[S^2_T] = E[a^2] = a^2$$

Also, by MCT,  $E[T \wedge n] \nearrow E[T]$ . From the equality  $E[S^2_{T \wedge n}] = E[T \wedge n] \quad \forall n \in \mathbb{N}$ , we can conclude that  $E[T] = a^2$  Now, we shall calculate  $E[Y_{n+1} | \mathcal{F}_n] = Y_n$  to make  $Y_n$  be a martingale.

$$E[Y_{n+1}|\mathcal{F}_n] = E[(S_n + \xi_{n+1})^4 - 6(n+1)(S_n + \xi_{n+1})^2 + b(n+1)^2 + c(n+1)|\mathcal{F}_n]$$
  
=  $S_n^4 + 6S_n^2 - 6(n+1)S_n^2 - 6(n+1) + bn^2 + b(2n+1) + cn + c$   
=  $S_n^5 - 6nS_n^2 + bn^2 + cn + (2b-6)n + (b+c-5) = Y_n$ 

To satisfy the last equation, b=3 and c=2 is the right choice. Since  $Y_{T\wedge n}$  is a martingale,  $E[Y_1]=E[Y_{T\wedge n}]$  and we have

$$E[S_{T \wedge n}^4] - 6E[(T \wedge n)S_{T \wedge n}^2] + 3E[(T \wedge n)^2] + 2E[T \wedge n] = 0$$

Taking  $n \to$ , by DCT and MCT, we get

$$a^4 - 6a^2E[T] + 3E[T^2] + 2E[T] = 0$$

Using  $E[T] = a^2$ , we get  $E[T^2] = (5a^4 - 2a^2)/3$