

# STA447/2006: Midterm Exam #2

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**This exam contains 11 pages.**

**Total marks: 100 pts**

**Time Allowed: 150 minutes**

**Question 1.** [30 points, 3 points for each question] Mark each of the following statements with T (true) or F (false). *No justification is required.* Your grade will be solely based on your true-or-false choices.

- (1) For a discrete-time stochastic process  $(X_n)_{n=0,1,2,\dots}$ , if  $T_1$  and  $T_2$  are both stopping times satisfying  $T_1 \geq T_2$ , then  $T_1 - T_2$  is a stopping time. **Answer: F**
- (2) Let  $T_1, T_2, T_3$  be three stopping times. Let  $T_{(2)}$  be the second smallest one among  $T_1, T_2, T_3$ . Then  $T_{(2)}$  is a stopping time. **Answer: T**
- (3) Let  $T := \min \{t \geq 10 : X_t = X_{t-10}\}$ . The random time  $T$  is a stopping time. **Answer: T**
- (4) If  $(Z_i)_{i \geq 0}$  are i.i.d. zero-mean random variables, the following process  $(X_n)_{n \geq 1}$  is a martingale

$$X_n := \left( \sum_{i=1}^n Z_i \right)^2 - n\mathbb{E}[Z_0^2].$$

**Answer: T**

- (5) Let  $(X_n)_{n \geq 0}$  be a discrete-time martingale such that  $X_n \geq 0$  for all  $n$ . Then there exists a random variable  $X_\infty$  such that  $X_n \rightarrow X_\infty$  a.s. **Answer: T**
- (6) Let  $(X_n)_{n=0,1,2,\dots}$  be a discrete-time martingale, if for any  $\varepsilon > 0$  and any  $n \geq 0$ , there exists a constant  $K > 0$  depending on  $\varepsilon$  and  $n$  such that

$$\mathbb{E} \left[ |X_n| \mathbf{1}_{|X_n| \geq K} \right] \leq \varepsilon,$$

then the process  $(X_n)_{n \geq 0}$  is uniformly integrable. **Answer: F**

- (7) Let  $(X_n)_{n=0,1,2,\dots}$  be a martingale. If there exists a stopping time such that  $\mathbb{E}[X_T] = \mathbb{E}[X_0]$ , then  $(X_n)_{n \geq 0}$  is uniformly integrable. **Answer: F**
- (8) Let  $X_n := \mathbb{E}[X | \mathcal{F}_n]$  for some random variable  $X$  and some filtration  $(\mathcal{F}_n)_{n \geq 0}$ , and  $T$  is a stopping time. Then  $X_T$  is well-defined and  $\mathbb{E}[X_T] = \mathbb{E}[X_0]$ . **Answer: T**
- (9) Let  $Z_1, Z_2, \dots$  be i.i.d. random variables with finite mean  $\mu$ . Let  $X_n := \sum_{i=1}^n Z_i$ . Let  $T := \arg \max_{1 \leq n \leq 10} X_n$ . Then  $\mathbb{E}[X_T] = \mu \mathbb{E}[T]$ . **Answer: F**

- (10) Let  $(X_n)_{n \geq 0}$  be a martingale satisfying  $\sup_{n \geq 0} \mathbb{E}[|X_n|^2] < +\infty$ . Then we have  $\mathbb{E}[\sup_{n \geq 0} |X_n|^2] < +\infty$ . **Answer: T**

**Rubrics:** Each item is worth 3 points. No justification is required, and no partial credit is given.

**Question 2.** [24 points] Let  $X_0 = 1$  and  $X_t = W_t X_{t-1}$  where  $W_t$  are i.i.d. random variables, following the distribution

$$W_t = \begin{cases} \frac{3}{2} & \text{with probability } 1/2, \\ \frac{1}{2} & \text{with probability } 1/2, \end{cases}$$

and the random time  $T_a := \inf\{t > 0 : X_t \geq a\}$  for some constant  $a > 1$ .

**Part (a) [9 pt].** Is  $(X_t)_{t \geq 0}$  uniformly integrable? How about the stopped process  $(X_{t \wedge T_a})_{t \geq 0}$ ? Justify your answer.

**Answer:** Since  $\mathbb{E}[W_t] = 1$ , we have

$$\mathbb{E}[X_t | \mathcal{F}_{t-1}] = X_{t-1} \mathbb{E}[W_t] = X_{t-1},$$

so  $(X_t)_{t \geq 0}$  is a nonnegative martingale with  $\mathbb{E}[X_t] = 1$  for all  $t$ .

However,

$$\log X_t = \sum_{i=1}^t \log W_i,$$

and by the strong law of large numbers,

$$\frac{1}{t} \log X_t \rightarrow \frac{1}{2} \log \frac{3}{2} + \frac{1}{2} \log \frac{1}{2} = \frac{1}{2} \log \frac{3}{4} < 0 \quad \text{a.s.}$$

Therefore  $X_t \rightarrow 0$  almost surely. If  $(X_t)$  were uniformly integrable, then by the martingale convergence theorem we would have

$$\mathbb{E}[X_\infty] = \lim_{t \rightarrow \infty} \mathbb{E}[X_t] = 1,$$

but  $X_\infty = 0$  almost surely, a contradiction. Hence  $(X_t)_{t \geq 0}$  is *not* uniformly integrable.

For the stopped process, note that if  $t < T_a$ , then  $X_t < a$ , while on the event  $T_a < +\infty$ ,

$$X_{T_a} = W_{T_a} X_{T_a-1} \leq \frac{3}{2} a$$

because  $X_{T_a-1} < a$  and  $W_{T_a} \leq 3/2$ . Hence

$$0 \leq X_{t \wedge T_a} \leq \frac{3}{2} a \quad \text{for all } t.$$

So  $(X_{t \wedge T_a})_{t \geq 0}$  is a bounded martingale, and therefore it is uniformly integrable.

**Rubrics:** 3 points for showing that  $X_t \rightarrow 0$  almost surely, or giving an equivalent argument based on  $\sum_{i=1}^t \log W_i$ .

3 points for concluding that  $(X_t)_{t \geq 0}$  is not uniformly integrable.

3 points for showing that  $(X_{t \wedge T_a})_{t \geq 0}$  is bounded, hence uniformly integrable.

**Part (b) [9 pt].** Show that

$$\frac{2}{3a} \leq \mathbb{P}(T_a < +\infty) \leq \frac{1}{a}$$

**Answer:** By part (a), the stopped process  $(X_{t \wedge T_a})_{t \geq 0}$  is uniformly integrable, so we can invoke the optional stopping theorem for possibly infinite stopping times (from lecture) to obtain that

$$\mathbb{E}[X_{T_a}] = \mathbb{E}[X_0] = 1,$$

which can be rewritten as

$$1 = \mathbb{E}[X_{T_a} \mathbf{1}_{T_a < +\infty}] + \mathbb{E}[X_{T_a} \mathbf{1}_{T_a = +\infty}] = \mathbb{E}[X_{T_a} | T_a < +\infty] \cdot \mathbb{P}(T_a < +\infty),$$

since  $X_{T_a} = \lim_{t \rightarrow \infty} X_{t \wedge T_a} = 0$  on the event  $\{T_a = +\infty\}$ .

On the event  $\{T_a < +\infty\}$ , we have

$$a \leq X_{T_a} \leq \frac{3}{2}a.$$

So we have

$$a \leq \mathbb{E}[X_{T_a} | T_a < +\infty] \leq \frac{3}{2}a.$$

Substituting this bound into the above equation gives

$$\frac{2}{3a} \leq \mathbb{P}(T_a < +\infty) \leq \frac{1}{a}.$$

**Rubrics:** 4 points for correctly applying optional stopping to get  $\mathbb{E}[X_{T_a}] = 1$ .

If you get the correct equation but without justifying the use of optional stopping, you can still get 2 points for this part.

5 points for using  $a \leq X_{T_a} \leq 3a/2$  on  $\{T_a < +\infty\}$  to derive both bounds.

If only one side of the bound is proved correctly, you get 3 points out of 5 for this part.

**Part (c) [6 pt].** Define the random variable

$$Z := \max_{t \geq 0} X_t.$$

Use the result in part (a) to show that

$$\mathbb{P}(Z < +\infty) = 1, \quad \text{but} \quad \mathbb{E}[Z] = +\infty.$$

You can use the result in part (b) even if you cannot solve part (b).

**Answer:** From part (a), we know that  $X_t \rightarrow 0$  almost surely. Hence on every sample path there exists a random time  $N$  such that  $X_t \leq 1$  for all  $t \geq N$ . Therefore

$$Z = \max_{t \geq 0} X_t = \max \{X_0, X_1, \dots, X_{N-1}, \sup_{t \geq N} X_t\}$$

is finite almost surely. Thus

$$\mathbb{P}(Z < +\infty) = 1.$$

On the other hand, for every  $a > 1$ ,

$$\{Z \geq a\} = \{T_a < +\infty\},$$

so part (b) gives

$$\mathbb{P}(Z \geq a) = \mathbb{P}(T_a < +\infty) \geq \frac{2}{3a}.$$

Using the tail-sum lower bound for nonnegative random variables,

$$\mathbb{E}[Z] \geq \sum_{k=2}^{\infty} \mathbb{P}(Z \geq k) \geq \sum_{k=2}^{\infty} \frac{2}{3k} = +\infty.$$

Therefore  $Z$  is finite almost surely, but  $\mathbb{E}[Z] = +\infty$ .

**Rubrics:** 3 points for proving  $Z < +\infty$  almost surely, for example from  $X_t \rightarrow 0$  a.s.

3 points for proving  $\mathbb{E}[Z] = +\infty$  using the lower tail bound from part (b) and a tail-sum argument.

**Question 3.** [21 points] Let  $(X_n)_{n \geq 0}$  be a one-dimensional symmetric simple random walk, i.e.,  $X_0 = 0$  and  $X_n = \sum_{i=1}^n Z_i$  where  $Z_i$  are i.i.d. random variables such that  $\mathbb{P}(Z_i = 1) = \mathbb{P}(Z_i = -1) = 1/2$ .

For  $\theta \in \mathbb{R}$ , define  $\lambda(\theta) := \log\left(\frac{1}{2}e^\theta + \frac{1}{2}e^{-\theta}\right)$ . We can construct a process  $(M_n(\theta))_{n \geq 0}$  as follows:

$$M_n(\theta) := e^{\theta X_n - n\lambda(\theta)}.$$

**Part (a) [4 points].** Prove that the process  $(M_n(\theta))_{n \geq 0}$  is a martingale, for any  $\theta \in \mathbb{R}$ .

**Answer:** Let  $\mathcal{F}_n = \sigma(Z_1, \dots, Z_n)$ . Since  $X_n = X_{n-1} + Z_n$ , we have

$$\begin{aligned} \mathbb{E}[M_n(\theta) \mid \mathcal{F}_{n-1}] &= \mathbb{E}[e^{\theta X_n - n\lambda(\theta)} \mid \mathcal{F}_{n-1}] \\ &= e^{\theta X_{n-1} - (n-1)\lambda(\theta)} \mathbb{E}[e^{\theta Z_n} \mid \mathcal{F}_{n-1}] \\ &= e^{\theta X_{n-1} - (n-1)\lambda(\theta)} \left(\frac{1}{2}e^\theta + \frac{1}{2}e^{-\theta}\right) \\ &= e^{\theta X_{n-1} - (n-1)\lambda(\theta)} e^{\lambda(\theta)} \\ &= e^{\theta X_{n-1} - (n-1)\lambda(\theta)} = M_{n-1}(\theta). \end{aligned}$$

Therefore  $(M_n(\theta))_{n \geq 0}$  is a martingale for every  $\theta \in \mathbb{R}$ .

**Rubrics:** 2 points for conditioning on  $\mathcal{F}_{n-1}$  and using  $X_n = X_{n-1} + Z_n$ .

2 points for computing  $\mathbb{E}[e^{\theta Z_n}] = e^{\lambda(\theta)}$  and concluding the martingale property.

**Part (b) [9 points].** For a positive integer  $a$ , define the stopping time

$$T := \inf\{n > 0 : |X_n| \geq a\}.$$

Use the optional stopping theorem to show that

$$\mathbb{E}[e^{-\lambda(\theta)T}] = \frac{2}{e^{\theta a} + e^{-\theta a}}.$$

[Hint: You may first show that  $X_T$  is independent of  $T$  and use this fact to compute  $\mathbb{E}[M_T(\theta)]$ . If you cannot show the independence, you can still get partial credit by proving the desired result by assuming the independence.]

**Answer:** Let  $\lambda = \lambda(\theta)$ . Since  $T$  is the first time the walk exits  $\{-a+1, \dots, a-1\}$ , we have  $|X_{n \wedge T}| \leq a$  for every  $n$ . Hence

$$0 \leq M_{n \wedge T}(\theta) = e^{\theta X_{n \wedge T} - (n \wedge T)\lambda} \leq e^{|\theta|a},$$

so  $(M_n(\theta))_{n \geq 0}$  is bounded up to time  $T$ . By optional stopping,

$$\mathbb{E}[M_T(\theta)] = \mathbb{E}[M_0(\theta)] = 1.$$

Next we show that  $X_T$  is independent of  $T$ . Since the walk moves by  $\pm 1$  and  $a$  is an integer, we must have  $X_T \in \{-a, a\}$ . For any fixed  $n$ , there is a sign-flip bijection between paths of length  $n$  with  $T = n$  and  $X_T = a$ , and paths of length  $n$  with  $T = n$  and  $X_T = -a$ . Therefore

$$\mathbb{P}(T = n, X_T = a) = \mathbb{P}(T = n, X_T = -a) = \frac{1}{2}\mathbb{P}(T = n).$$

Thus  $X_T$  is independent of  $T$ , and

$$\mathbb{P}(X_T = a) = \mathbb{P}(X_T = -a) = \frac{1}{2}.$$

Now compute

$$\begin{aligned} 1 &= \mathbb{E}[M_T(\theta)] = \mathbb{E}[e^{\theta X_T - \lambda T}] \\ &= \mathbb{E}[e^{-\lambda T}] \mathbb{E}[e^{\theta X_T}] \\ &= \mathbb{E}[e^{-\lambda T}] \cdot \frac{e^{\theta a} + e^{-\theta a}}{2}. \end{aligned}$$

Therefore,

$$\mathbb{E}[e^{-\lambda T}] = \frac{2}{e^{\theta a} + e^{-\theta a}}.$$

This is the desired identity.

**Rubrics:** 3 points for correctly applying optional stopping to the stopped martingale.

3 points for showing that  $X_T$  is independent of  $T$  so that  $\mathbb{P}(X_T = \pm a) = 1/2$ .

3 points for computing  $\mathbb{E}[M_T(\theta)]$  and deriving the stated formula.

**Part (c) [8 points].** Fix  $\delta > 0$  and  $n \in \mathbb{N}_+$ , show that

$$\mathbb{P}\left(\max_{0 \leq t \leq n} X_t \leq \sqrt{2n \log(1/\delta)}\right) \geq 1 - \delta.$$

Hint: you may use without proof the calculus fact that  $\frac{1}{2}e^\theta + \frac{1}{2}e^{-\theta} \leq e^{\theta^2/2}$  for any  $\theta \in \mathbb{R}$ . The martingale constructed in part (a) may be useful for this question.

**Answer:** Fix  $\theta > 0$  and  $b > 0$ . If  $\max_{0 \leq t \leq n} X_t \geq b$ , then for some  $t \leq n$  we have

$$M_t(\theta) = e^{\theta X_t - t\lambda(\theta)} \geq e^{\theta b - n\lambda(\theta)},$$

because  $t \leq n$  and  $\lambda(\theta) \geq 0$ . Therefore

$$\left\{ \max_{0 \leq t \leq n} X_t \geq b \right\} \subseteq \left\{ \max_{0 \leq t \leq n} M_t(\theta) \geq e^{\theta b - n\lambda(\theta)} \right\}.$$

Since  $(M_t(\theta))_{0 \leq t \leq n}$  is a nonnegative martingale, Doob's maximal inequality gives

$$\begin{aligned} \mathbb{P}\left(\max_{0 \leq t \leq n} X_t \geq b\right) &\leq \mathbb{P}\left(\max_{0 \leq t \leq n} M_t(\theta) \geq e^{\theta b - n\lambda(\theta)}\right) \\ &\leq e^{-\theta b + n\lambda(\theta)} \mathbb{E}[M_n(\theta)] \\ &= e^{-\theta b + n\lambda(\theta)}. \end{aligned}$$

Using the given inequality,

$$\lambda(\theta) = \log\left(\frac{1}{2}e^\theta + \frac{1}{2}e^{-\theta}\right) \leq \frac{\theta^2}{2},$$

so

$$\mathbb{P}\left(\max_{0 \leq t \leq n} X_t \geq b\right) \leq e^{-\theta b + n\theta^2/2}.$$

Choose  $\theta = b/n$ . Then

$$\mathbb{P}\left(\max_{0 \leq t \leq n} X_t \geq b\right) \leq e^{-b^2/(2n)}.$$

Now set

$$b = \sqrt{2n \log(1/\delta)}.$$

Then

$$\mathbb{P}\left(\max_{0 \leq t \leq n} X_t \geq \sqrt{2n \log(1/\delta)}\right) \leq \delta,$$

which is equivalent to

$$\mathbb{P}\left(\max_{0 \leq t \leq n} X_t \leq \sqrt{2n \log(1/\delta)}\right) \geq 1 - \delta.$$

**Rubrics:** 4 points for reducing the event  $\{\max_{t \leq n} X_t \geq b\}$  to a maximal event for the martingale  $M_t(\theta)$  and applying Doob's inequality.

2 points for using  $\lambda(\theta) \leq \theta^2/2$  to obtain an exponential bound.

2 points for optimizing in  $\theta$  and substituting  $b = \sqrt{2n \log(1/\delta)}$ .

**Question 4.** [9 points] Let  $p$  be a probability distribution on  $\{0, 1, 2, \dots\}$  satisfying  $\mathbb{E}_{Z \sim p}[Z] = 1$ . Consider a stochastic process  $(X_n)_{n \geq 0}$  defined as follows:  $X_0 = 1$  and for any  $n \geq 1$ , we have

$$X_n = \sum_{i=1}^{X_{n-1}} Z_{n,i},$$

where  $Z_{n,i}$  are i.i.d. random variables following the distribution  $p$  and independent of  $X_{n-1}$ . If  $X_n = 0$  for some  $n$ , we let  $X_m = 0$  for all  $m > n$ , i.e., the process goes extinct once it hits zero. Assuming that  $p$  is not a degenerate distribution that puts all its mass on 1, show that

$$\mathbb{P}(\text{the process goes extinct}) = 1.$$

**Answer:** Let  $\mathcal{F}_n = \sigma(Z_{k,i} : k \leq n, i \geq 1)$ . Since the  $Z_{n,i}$  are i.i.d. with mean  $\mathbb{E}[Z] = 1$  and are independent of  $X_{n-1}$ ,

$$\mathbb{E}[X_n | \mathcal{F}_{n-1}] = \mathbb{E}\left[\sum_{i=1}^{X_{n-1}} Z_{n,i} \mid \mathcal{F}_{n-1}\right] = X_{n-1} \cdot \mathbb{E}[Z] = X_{n-1}.$$

Hence  $(X_n)_{n \geq 0}$  is a martingale. Since  $X_n \geq 0$  for all  $n$  and  $\sup_n \mathbb{E}[X_n] = \mathbb{E}[X_0] = 1 < \infty$ , it is a non-negative  $L^1$ -bounded martingale.

By the martingale convergence theorem,  $X_n \rightarrow X_\infty$  a.s. for some non-negative random variable  $X_\infty$ . Note that  $X_n \in \{0, 1, 2, \dots\}$  for every  $n$ . A sequence of integers can converge only if it is eventually constant, so  $X_n = X_\infty$  for all sufficiently large  $n$ , almost surely.

For  $k \geq 1$ , if  $X_n = k$ , then the next value is  $X_{n+1} = \sum_{i=1}^k Z_{n+1,i}$ , which has mean  $k$  and variance  $k \text{Var}(Z) > 0$ . So  $X_{n+1}$  is not almost surely equal to  $k$ . Therefore,  $k$  is not an absorbing state, and the process can escape from  $k$  with positive probability, and  $k$  cannot be the eventual constant value of  $X_n$ . Hence  $\mathbb{P}(X_\infty = k) = 0$  for every  $k \geq 1$ . We can conclude that  $X_\infty = 0$  almost surely, so the process goes extinct with probability one.

**Rubrics:** 4 points for identifying  $(X_n)_{n \geq 0}$  as a non-negative martingale and applying the martingale convergence theorem to obtain an a.s. limit  $X_\infty$ .

3 points for arguing that  $X_\infty$  is integer-valued and therefore eventually constant.

2 points for showing that the only possible eventual constant value is zero, hence the process goes extinct with probability one.

**Question 5.** [16 points] Sum of independent random variables

**Part (a) [8 points].** Let  $(X_n)_{n \geq 0}$  be i.i.d. normal random variables with mean zero and variance one. Define the process

$$S_n = \sum_{i=1}^n X_i^2.$$

Fix  $a > 0$ , define the stopping time

$$T := \inf\{n > 0 : |X_n| \geq a\}.$$

Compute  $\mathbb{E}[S_T]$ . You can express the answer using the cumulative distribution function of the standard normal distribution.

**Answer:** Let

$$p := \mathbb{P}(|X_1| \geq a) = 2(1 - \Phi(a)),$$

where  $\Phi$  is the standard normal cdf. Since the  $X_n$  are i.i.d., the stopping time  $T$  is geometric with success probability  $p$ , so

$$\mathbb{E}[T] = \frac{1}{p} = \frac{1}{2(1 - \Phi(a))}.$$

Now

$$S_T = \sum_{i=1}^T X_i^2,$$

and the random variables  $X_i^2$  are i.i.d., nonnegative, with mean

$$\mathbb{E}[X_1^2] = 1.$$

Therefore Wald's identity applies and gives

$$\mathbb{E}[S_T] = \mathbb{E}[T]\mathbb{E}[X_1^2] = \frac{1}{2(1 - \Phi(a))}.$$

**Rubrics:** 3 points for identifying  $T$  as a geometric random variable with parameter  $2(1 - \Phi(a))$ .

5 points for correctly applying Wald's identity, or an equivalent argument, to obtain  $\mathbb{E}[S_T] = 1/\{2(1 - \Phi(a))\}$ .

**Part (b) [8 points].** Let  $(Y_n)_{n \geq 0}$  be i.i.d. random variables with  $\mathbb{P}(Y_n = 1) = \mathbb{P}(Y_n = -1) = 1/2$ . Define the process

$$Z_n := \sum_{i=1}^n 2^{i-1} Y_i.$$

Define the stopping time

$$T := \inf\{n > 0 : Y_n > 0\}.$$

Compute  $\mathbb{E}[T]$ , and show that  $\mathbb{E}[Z_T] \neq \mathbb{E}[Z_0]$ .

**Answer:** Since  $T$  is the first time that  $Y_n = 1$ , it is geometric with parameter  $1/2$ . Hence

$$\mathbb{E}[T] = \frac{1}{1/2} = 2.$$

On the event  $\{T = t\}$ , we must have

$$Y_1 = Y_2 = \dots = Y_{t-1} = -1, \quad Y_t = 1.$$

Therefore

$$\begin{aligned} Z_T &= \sum_{i=1}^{t-1} 2^{i-1}(-1) + 2^{t-1} \\ &= -\sum_{i=1}^{t-1} 2^{i-1} + 2^{t-1} \\ &= -(2^{t-1} - 1) + 2^{t-1} = 1. \end{aligned}$$

So  $Z_T = 1$  almost surely, and hence

$$\mathbb{E}[Z_T] = 1.$$

But  $Z_0 = 0$ , so

$$\mathbb{E}[Z_T] = 1 \neq 0 = \mathbb{E}[Z_0].$$

**Rubrics:** 3 points for correctly computing  $\mathbb{E}[T] = 2$ .

5 points for showing that  $Z_T = 1$  almost surely, and therefore  $\mathbb{E}[Z_T] \neq \mathbb{E}[Z_0]$ .