Math 171, Fall 2016, UCLA		Instructor: Steven Heilman
Name:	UCLA ID:	Date:
Signature:		
(By signing here, I certify that I h	nave taken this test wh	nile refraining from cheating.)

Final Exam

This exam contains 16 pages (including this cover page) and 10 problems. Check to see if any pages are missing. Enter all requested information on the top of this page.

You may *not* use your books, notes, or any calculator on this exam. You are required to show your work on each problem on this exam. The following rules apply:

• You have 180 minutes to complete the exam.

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- If you use a theorem or proposition from class or the notes or the book you must indicate this and explain why the theorem may be applied. It is okay to just say, "by some theorem/proposition from class."
- Organize your work, in a reasonably neat and coherent way, in the space provided. Work scattered all over the page without a clear ordering will receive very little credit.
- Mysterious or unsupported answers will not receive full credit. A correct answer, unsupported by calculations, explanation, or algebraic work will receive no credit; an incorrect answer supported by substantially correct calculations and explanations might still receive partial credit.
- If you need more space, use the back of the pages; clearly indicate when you have done this. Scratch paper is at the end of the document.

Do not write in the table to the right. Good luck!^a

Problem	Points	Score
1	15	
2	10	
3	10	
4	10	
5	10	
6	5	
7	10	
8	10	
9	10	
10	10	
Total:	100	

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Reference sheet

Below are some definitions that may be relevant.

A (finite or countable) Markov Chain is a stochastic process $(X_0, X_1, X_2, ...)$ together with a finite or countable set Ω , which is called the **state space** of the Markov Chain, and function $P: \Omega \times \Omega \to [0, 1]$. The random variables $X_0, X_1, ...$ take values in the finite set Ω . P is **stochastic**, that is all of its entries are nonnegative and

$$\sum_{y \in \Omega} P(x, y) = 1, \quad \forall y \in \Omega.$$

And the stochastic process satisfies the following **Markov property**: for all $x, y \in \Omega$, for any $n \ge 1$, and for all events H_{n-1} of the form $H_{n-1} = \bigcap_{k=0}^{n-1} \{X_k = x_k\}$, where $x_k \in \Omega$ for all $0 \le k \le n-1$, such that $\mathbf{P}(H_{n-1} \cap \{X_n = x\}) > 0$, we have

$$\mathbf{P}(X_{n+1} = y \mid H_{n-1} \cap \{X_n = x\}) = \mathbf{P}(X_{n+1} = y \mid X_n = x) = P(x, y).$$

Suppose we have a Markov Chain X_0, X_1, \ldots with state space Ω . Let $y \in \Omega$. Define the **first return time** of y to be the following random variable: $T_y := \min\{n \geq 1 : X_n = y\}$. Also, define $\rho_{yy} := \mathbf{P}_y(T_y < \infty)$.

If $\rho_{yy} = 1$, we say the state $y \in \Omega$ is **recurrent**. If $\rho_{yy} < 1$, we say the state $y \in \Omega$ is **transient**. A Markov chain is **irreducible** if any state can reach any other state, with some positive probability, if the chain runs long enough.

We say that π is a **stationary distribution** if $\pi(x) \geq 0$ for every $x \in \Omega$, $\sum_{x \in \Omega} \pi(x) = 1$, and if π satisfies $\pi = \pi P$ (that is, $\pi(x) = \sum_{y \in \Omega} \pi(y) P(y, x)$ for every $x \in \Omega$.)

Let P be the transition matrix of a finite Markov chain with state space Ω . We say that the Markov chain is **reversible** if there exists a probability distribution π on Ω satisfying the following **detailed balance condition**: $\pi(x)P(x,y) = \pi(y)P(y,x)$, $\forall x,y \in \Omega$.

Let μ, ν be probability distributions on a finite state space Ω . We define the **total variation distance** between μ and ν to be $||\mu - \nu||_{TV} := \max_{A \subset \Omega} |\mu(A) - \nu(A)|$.

Let $(X_0, X_1, ...)$ be a real-valued stochastic process. A **real-valued martingale with respect to** $(X_0, X_1, ...)$ is a stochastic process $(M_0, M_1, ...)$ such that $\mathbf{E}|M_n| < \infty$ for all $n \geq 0$, and for any $m_0, x_0, ..., x_n \in \mathbf{R}$,

$$\mathbf{E}(M_{n+1} - M_n | X_n = x_n, \dots, X_0 = x_0, M_0 = m_0) = 0.$$

A **stopping time** for a martingale M_0, M_1, \ldots is a random variable T taking values in $0, 1, 2, \ldots, \cup \{\infty\}$ such that, for any integer $n \geq 0$, the event $\{T = n\}$ is determined by M_0, X_0, \ldots, X_n . More formally, for any integer $n \geq 1$, there is a set $B_n \subseteq \mathbf{R}^{n+2}$ such that

 $\{T=n\}=\{(M_0,X_0,\ldots,X_n)\in B_n\}$. Put another way, the indicator function $1_{\{T=n\}}$ is a function of the random variables M_0,X_0,\ldots,X_n .

Let X be a random variables on a sample space Ω . Let $A \subseteq \Omega$ with $\mathbf{P}(A) > 0$. Then the **conditional expectation of** X **given** A, denoted $\mathbf{E}(X|A)$ is

$$\mathbf{E}(X|A) := \frac{\mathbf{E}(X \cdot 1_A)}{\mathbf{P}(A)}.$$

Suppose we have a partition of a sample space Ω . That is, we have sets $A_1, \ldots, A_k \subseteq \Omega$ such that $A_i \cap A_j = \emptyset$ for all $i, j \in \{1, \ldots, k\}$ with $i \neq j$, and $\bigcup_{i=1}^k A_i = \Omega$. Denote $\mathcal{A} = \{A_1, \ldots, A_k\}$. Define $\mathbf{E}(X|\mathcal{A})$ to be a random variable that takes the value $\mathbf{E}(X|A_i)$ on the set A_i .

Let $\lambda > 0$. Recall that a random variable T is **exponential with parameter** λ if T has the density function given by $f_T(x) = \lambda e^{-\lambda x}$ for all $x \geq 0$, and $f_T(x) = 0$ otherwise.

Let $\lambda > 0$. Let τ_1, τ_2, \ldots be independent exponential random variables with parameter λ . Let $T_0 = 0$, and for any $n \ge 1$, let $T_n := \tau_1 + \cdots + \tau_n$. A **Poisson Process** with parameter $\lambda > 0$ is a set of integer-valued random variables $\{N(s)\}_{s>0}$ defined by $N(s) := \max\{n \ge 0 : T_n \le s\}$.

Let τ_1, τ_2, \ldots be nonnegative independent identically distributed variables. Let $T_0 = 0$, and for any $n \ge 1$, let $T_n := \tau_1 + \cdots + \tau_n$. A **Renewal process** is a set of integer-valued random variables $\{N(s)\}_{s>0}$ defined by $N(s) := \max\{n \ge 0 : T_n \le s\}$.

Standard Brownian motion with B(0) = 0 is uniquely characterized by the following properties:

- (i) (Independent increments) For any $0 < t_1 < \cdots < t_n$, the random variables $B(t_2) B(t_1), \ldots, B(t_n) B(t_{n-1})$ are all independent.
- (ii) (Stationary Gaussian increments) for any 0 < s < t, B(t) B(s) is a Gaussian random variable with mean zero and variance t s.
- (iii) (Continuous Sample Paths) With probability 1, the function $t \mapsto B(t)$ is continuous

- 1. Label the following statements as TRUE or FALSE. If the statement is true, explain your reasoning. If the statement is false, provide a counterexample and explain your reasoning. (In this question, you can freely cite results from the homeworks.)
 - (a) (3 points) Let $\{N(s)\}_{s\geq 0}$ be a Poisson Process with parameter $\lambda=1$. Then

$$N(4) - N(3), N(3) - N(2), N(2) - N(1), N(1)$$

are all independent random variables.

(b) (3 points) Let $\{N(s)\}_{s\geq 0}$ be a renewal process. Then N(1) and N(0) are independent random variables.

(c) (3 points) Suppose we have a renewal process $\{N(s)\}_{s\geq 0}$ with arrival increments τ_1, τ_2, \ldots Let $\mu := \mathbf{E}\tau_1$. Assume that $0 < \mu < \infty$. Then

$$\mathbf{P}\left(\lim_{s\to\infty}\frac{N(s)}{s}=\frac{1}{\mu}\right)=1.$$

TRUE FALSE (circle one)

(d) (3 points) Let $\{B(t)\}_{t \geq 0}$ be a standard Brownian motion. Let 0 < s < t. Then

$$\mathbf{E}B(s)B(t) = t.$$

TRUE FALSE (circle one)

(e) (3 points) Let $\{B(t)\}_{t\geq 0}$ be a standard Brownian motion. Let a,b>0. Let $T_a:=\inf\{t\geq 0\colon B(t)=a\}$. Then

$$\mathbf{P}(T_a < T_{-b}) = \frac{b}{a+b}$$

TRUE FALSE (circle one)

2. (10 points) Let $\Omega = [0,1]$. Let **P** be the uniform probability law on Ω . Let $X \colon [0,1] \to \mathbf{R}$ be a random variable such that $X(t) = t^3$ for all $t \in [0,1]$. Let

$$\mathcal{A} = \{[0, 1/4), [1/4, 1/2), [1/2, 3/4), [3/4, 1]\}$$

Compute explicitly the function $\mathbf{E}(X|\mathcal{A})$.

3. (10 points) Give an example of a martingale that is a Markov chain. (Your example should be a discrete time stochastic process Y_0, Y_1, Y_2, \ldots)

4. (10 points) Give an example of a martingale that is **not** a Markov chain. (Your example should be a discrete time stochastic process Y_0, Y_1, Y_2, \ldots)

5. (10 points) Let $X \ge 0$ be a random variable such that $\mathbf{P}(X > 0) > 0$. Show that

$$\mathbf{E}(X \mid X > 0) \le \frac{\mathbf{E}X^2}{\mathbf{E}X}.$$

(Hint: you can freely use the Cauchy-Schwarz inequality: $(\mathbf{E}XY)^2 \leq \mathbf{E}X^2\mathbf{E}Y^2$.)

6. (5 points) Suppose you run a (busy) car wash, and the number of cars that come to the car wash between time 0 and time s > 0 is a Poisson poisson with rate $\lambda = 1$. Suppose every car has either one, two, three, or four people in it. The probability that a car has one, two, three or four people in it is 1/4, 1/2, 1/12 and 1/6, respectively.

What is the average number of cars with four people that have arrived by time s = 60?

7. (10 points) Let $X_1, X_2, \ldots, Y_1, Y_2, \ldots, Z_1, Z_2, \ldots$ be random variables. Let $a, b \in \mathbf{R}$. Assume that $X_n \leq Y_n \leq Z_n$ for any $n \geq 1$. Assume that $\mathbf{P}(\lim_{n \to \infty} X_n = a) = 1$ and $\mathbf{P}(\lim_{n \to \infty} Z_n = a) = 1$. Prove that $\mathbf{P}(\lim_{n \to \infty} Y_n = a) = 1$.

8. (10 points) Let
$$A := \{1+1, 1+1/2, 1+1/3, 1+1/4, \ldots\}$$
.
Let $B := \{1-1, 1-1/2, 1-1/3, 1-1/4, 1-1/5, \ldots\}$.

- Find $\inf(A)$, the greatest lower bound of A.
- Find $\inf(B)$.

- 9. (10 points) Let $\{B(t)\}_{t\geq 0}$ be a standard Brownian motion (so that B(0)=0). For any x>0, let $T_x:=\inf\{t\geq 0\colon B(t)=x\}$.
 - Show the bound $\mathbf{P}(-x < B(t) < x) \ge \frac{x}{20\sqrt{t}}$ holds for all $t > x^2$.
 - Show that $\mathbf{E}T_x = \infty$. (Hint: use a reflection principle.)

10. (10 points) Let $\{B(t)\}_{t\geq 0}$ be a standard Brownian motion. Let $x_1,\ldots,x_n\in \mathbf{R}$, and let $t_n>\cdots>t_1>0$. Show that the event

$$\{B(t_1) = x_1, \dots, B(t_n) = x_n\}$$

has a multivariate normal distribution. That is, the joint density of $(B(t_1), \ldots, B(t_n))$ is

$$f(x_1, \dots, x_n) = f_{t_1}(x_1) f_{t_2-t_1}(x_2 - x_1) \cdots f_{t_n-t_{n-1}}(x_n - x_{n-1})$$

where

$$f_t(x) = \frac{1}{\sqrt{2\pi t}} e^{-x^2/(2t)}, \quad \forall x \in \mathbf{R}, t > 0.$$

(Scratch paper)

(More scratch paper)