

## 3.7 Exercises

1. For  $\mathcal{F} = \{\phi, \Omega\}$ , show that  $E[X|\mathcal{F}] = E[X]$ .
2. Give the proof of Proposition 3.2 when  $X$  and  $Y$  are jointly continuous.
3. If  $E[|X_i|] < \infty$ ,  $i = 1, \dots, n$ , show that

$$E\left[\sum_{i=1}^n X_i | \mathcal{F}\right] = \sum_{i=1}^n E[X_i | \mathcal{F}]$$

4. Prove that if  $f$  is a convex function, then

$$E[f(X)|\mathcal{F}] \geq f(E[X|\mathcal{F}])$$

provided the expectations exist.

5. Let  $X_1, X_2, \dots$ , be independent random variables with mean 1. Show that  $Z_n = \prod_{i=1}^n X_i$ ,  $n \geq 1$ , is a martingale.
6. If  $E[X_{n+1}|X_1, \dots, X_n] = a_n X_n + b_n$  for constants  $a_n, b_n$ ,  $n \geq 0$ , find constants  $A_n, B_n$  so that  $Z_n = A_n X_n + B_n$ ,  $n \geq 0$ , is a martingale with respect to the filtration  $\sigma(X_0, \dots, X_n)$ .
7. Consider a population of individuals as it evolves over time, and suppose that, independent of what occurred in prior generations, each individual in generation  $n$  independently has  $j$  offspring with probability  $p_j$ ,  $j \geq 0$ . The offspring of individuals of generation  $n$  then make up generation  $n+1$ . Assume that  $m = \sum_j j p_j < \infty$ . Let  $X_n$  denote the number of individuals in generation  $n$ , and define a martingale related to  $X_n$ ,  $n \geq 0$ . The process  $X_n$ ,  $n \geq 0$  is called a branching process.
8. Suppose  $X_1, X_2, \dots$ , are independent and identically distributed random variables with mean zero and finite variance  $\sigma^2$ . If  $T$  is a stopping time with finite mean, show that

$$\text{Var}\left(\sum_{i=1}^T X_i\right) = \sigma^2 E(T).$$

problem

bound on the traveling salesman problem 0.5.1, we take  $Z$  to be Consider the integrand  $X_{i-1}$ . If  $S$  denotes the as in Example 5.7.2

$$\min_{j>i} \|v_j - y_i\|$$

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$v_i\}$ ). It follows that

$$\min_{j>i} \|v_j - v_i\|$$

$$\min_{j>i} \|Y_j - Y_i\|.$$

$\min_{j>i} \|Y_j - y\| \geq r$  for at a distance of  $r$  or less me case occurs when  $y$  ult,

$$\leq e^{-\frac{(n-i)\pi r^2}{4}}.$$

$$\frac{\pi r^2}{4} dr$$

$$\frac{(n-i)\pi r^2}{4} dr$$

use the crude inequality

$$2\sqrt{2},$$

then the sum  $\sum_{i=1}^n c_i^2$  figuring in Proposition 10.5.1 can be bounded by

$$\sum_{i=1}^n c_i^2 \leq (2\sqrt{2})^2 + 4^2 \sum_{i=1}^{n-1} \frac{1}{n-i} \leq 8 + 16(\ln n + 1).$$

This in turn translates into the Azuma-Hoeffding bound

$$\Pr[|D_n - E(D_n)| \geq \lambda] \leq 2e^{-\frac{\lambda^2}{48 + 32 \ln n}}.$$

## 10.6 Problems

1. Define the random variables  $Y_n$  inductively taking by  $Y_0 = 1$  and  $Y_{n+1}$  to be uniformly distributed on the interval  $(0, Y_n)$ . Show that the sequence  $X_n = 2^n Y_n$  is a martingale.
2. An urn contains  $b$  black balls and  $w$  white balls. Each time we randomly withdraw a ball, we replace it by  $c + 1$  balls of the same color. Let  $X_n$  be the fraction of white balls after  $n$  draws. Demonstrate that  $X_n$  is a martingale.
3. Let  $Y_1, Y_2, \dots$  be a sequence of independent random variables with zero means and common variance  $\sigma^2$ . If  $X_n = Y_1 + \dots + Y_n$ , then show that  $X_n^2 - n\sigma^2$  is a martingale.
4. Let  $Y_1, Y_2, \dots$  be a sequence of i.i.d. random variables with common moment generating function  $M(t) = E(e^{tY_1})$ . Prove that

$$X_n = M(t)^{-n} e^{t(Y_1 + \dots + Y_n)}$$

is a martingale whenever  $M(t) < \infty$ .

5. Let  $Y_n$  be a finite-state, discrete-time Markov chain with transition matrix  $P = (p_{ij})$ . If  $v$  is a column eigenvector for  $P$  with nonzero eigenvalue  $\lambda$ , then verify that  $X_n = \lambda^{-n} v_{Y_n}$  is a martingale, where  $v_{Y_n}$  is coordinate  $Y_n$  of  $v$ .
6. Suppose  $Y_n$  is the number of particles at the  $n$ th generation of a branching process. If  $s_\infty$  is the extinction probability, prove that  $X_n = s_\infty^{Y_n}$  is a martingale. (Hint: If  $Q(s)$  is the progeny generating function, then  $Q(s_\infty) = s_\infty$ .)
7. In Example 10.3.2, show that  $\text{Var}(X_\infty) = \frac{\sigma^2}{\mu(\mu-1)}$  by differentiating equation (10.12) twice. This result is consistent with the mean square convergence displayed in equation (10.9).