

MASSACHUSETTS INSTITUTE OF TECHNOLOGY

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 Problem Set 10

Fall 2008
 due 11/26/2008

Exercise 1. Let $\{X_n\}$ be a sequence of identically distributed random variables, with finite variance. Suppose that $\text{cov}(X_i, X_j) \leq \alpha^{|i-j|}$, for every i and j , where $|\alpha| < 1$. Show that the sample mean $(X_1 + \cdots + X_n)/n$ converges to $\mathbb{E}[X_1]$, in probability.

Exercise 2. Suppose that a random variable does not admit a finite upper bound, that is, $F_X(x) < 1$ for all $x \in \mathbb{R}$. Show that

$$\lim_{s \rightarrow \infty} \frac{\log M(s)}{s} = \infty.$$

Exercise 3. Let X_1, X_2, \dots be i.i.d. exponential random variables with parameter $\lambda = 1$. Let $S_n = X_1 + \cdots + X_n$. What is the Chernoff upper bound for $\mathbb{P}(S_n \geq na)$?

Exercise 4. (“Change of measure” for fast simulation.)

Consider a nonnegative random variable X whose PDF is close to being exponential, of the form

$$f_X(x) = g(x)e^{-x},$$

where $g(x)$ is a nonnegative function that satisfies $1/2 \leq g(x) \leq 2$ for all x , and $\int g(x)e^{-x} dx = 1$. Let a be a large constant. We wish to estimate $p = \mathbb{P}(X \geq a)$ using Monte Carlo simulation. We assume that we are able to generate random variables drawn from the distribution of X , as well as from an exponential distribution.

The straightforward simulation method is to generate n random samples, drawn from the distribution of X , let N be the number of samples that satisfy $X_i \geq a$, and form the estimate $\hat{P} = N/n$. Clearly, $\mathbb{E}[\hat{P}] = p$.

(a) Show that for large enough a , we have $\text{var}(\hat{P}) \geq e^{-a}/(3n)$.

Part (a) shows that the standard deviation of the estimation error $\hat{P} - p$ is of order $O(e^{-a/2})$, which is larger than the quantity p to be estimated by a $O(e^{a/2})$ factor. This is an instance of a general phenomenon: probabilities of rare events are difficult to estimate by simulation.

Consider now a random variable Y whose PDF is exponential, with parameter $\lambda = 1/a$.

$$f_Y(x) = \frac{e^{-x/a}}{a} = \exp\left\{\left(1 - \frac{1}{a}\right)x\right\} \cdot \frac{1}{a \cdot g(x)} \cdot f_X(x).$$

We generate n random samples Y_i , drawn from the distribution of Y , and estimate p by

$$Q = \frac{1}{n} \sum_{i=1}^n I_{Y_i \geq a} \frac{f_X(Y_i)}{f_Y(Y_i)} = \frac{1}{n} \sum_{i=1}^n I_{Y_i \geq a} \cdot ag(Y_i) \cdot \exp\left\{-\left(1 - \frac{1}{a}\right)Y_i\right\}.$$

- (b) Show that $\mathbb{E}[Q] = p$.
- (c) Show that the standard deviation σ_Q of Q is “comparable” to p , in the sense that σ_Q/p does not grow exponentially with a .

Exercise 5. A coin is tossed independently n times. The probability of heads at each toss is p . At each time k (with $k = 2, \dots, n$), we obtain a unit reward at time $k + 1$ if the k th toss is heads and the previous toss was tails. Let R be the total reward obtained.

- (a) Each time k (with $k < n$) that a tail is obtained, there is a probability p that the next toss is heads, in which case a unit reward is obtained at time $k + 1$. Let T be the number of tails in tosses $1, \dots, n - 1$. Is it true that conditional on $T = t$, the reward R has a binomial (conditional) distribution with parameters t and p ? Justify your answer.
- (b) Let A_k be the event that a reward is obtained at time k .
- (i) Are the events A_k and A_{k+1} independent?
 - (ii) Are the events A_k and A_{k+2} independent?
- (c) Find the expected value of R .
- (d) Find the variance of R .
- (e) If the number n of coin tosses is infinite, what is the expected value of the number of tosses until the reward becomes equal to some given number k ?
- (f) Suppose that $n = 1000$ and $p = 1/2$. Find an approximation to the probability that $R \geq 260$. You may leave your answer in the form $\Phi(c)$, where $\Phi(\cdot)$ is the cumulative distribution function of a standard normal random variable, and c is some number.

Exercise 6. For any positive integer k , let

$$h_k = 1 + \frac{1}{2} + \cdots + \frac{1}{k}.$$

Consider a Poisson process and let $X_k = 1$ if and only if there has been at least one arrival during the interval $[h_k, h_{k+1})$. Show that X_k converges to zero in probability, but not almost surely.

Exercise 7. Let $N(\cdot)$ be a Poisson process with rate λ . Find the covariance of $N(s)$ and $N(t)$.

Exercise 8. Based on your understanding of the Poisson process, determine the numerical values of a and b in the following expression and explain your reasoning.

$$\int_t^\infty \frac{\lambda^5 \tau^4 e^{-\lambda\tau}}{4!} d\tau = \sum_{k=a}^b \frac{(\lambda t)^k e^{-\lambda t}}{k!}.$$

Exercise 9. [Drill problem, does not have to be turned in.]

Fred is giving out samples of dog food. He makes calls door to door, but he leaves a sample (one can) only on those calls for which the door is answered *and* a dog is in residence. On any call the probability of the door being answered is $3/4$, and the probability that any household has a dog is $2/3$. Assume that the events “Door answered” and “A dog lives here” are independent and also that the outcomes of all calls are independent.

- (a) Determine the probability that Fred gives away his first sample on his third call.
- (b) Given that he has given away exactly four samples on his first eight calls, determine the conditional probability that Fred will give away his fifth sample on his eleventh call.
- (c) Determine the probability that he gives away his second sample on his fifth call.
- (d) Given that he did not give away his second sample on his second call, determine the conditional probability that he will leave his second sample on his fifth call.
- (e) We will say that Fred “needs a new supply” immediately *after* the call on which he gives away his last can. If he starts out with two cans, determine the probability that he completes at least five calls before he needs a new supply.

- (f) If he starts out with exactly m cans, determine the expected value and variance of D_m , the number of homes with dogs which he passes up (because of no answer) before he needs a new supply.

Hint: The formula $\text{var}(X) = \text{var}(\mathbb{E}[X | Y]) + \mathbb{E}[\text{var}(X | Y)]$ may be useful. Also, if T_i are i.i.d. geometric random variables, with parameter p , and $Y_k = T_1 + \cdots + T_n$, then the PMF of Y_k (known as a Pascal PMF) is of the form

$$p_{Y_k}(t) = \binom{t-1}{k-1} p^k (1-p)^{t-k}, \quad t = k, k+1, \dots$$

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