

MASSACHUSETTS INSTITUTE OF TECHNOLOGY
Department of Electrical Engineering & Computer Science
6.041SC Probabilistic Systems Analysis and Applied Probability
Lecture 12 Bonus Video Solution

Problem 27.* We toss n times a biased coin whose probability of heads, denoted by q , is the value of a random variable Q with given mean μ and positive variance σ^2 . Let X_i be a Bernoulli random variable that models the outcome of the i th toss (i.e., $X_i = 1$ if the i th toss is a head). We assume that X_1, \dots, X_n are conditionally independent, given $Q = q$. Let X be the number of heads obtained in the n tosses.

- (a) Use the law of iterated expectations to find $\mathbf{E}[X_i]$ and $\mathbf{E}[X]$.
- (b) Find $\text{cov}(X_i, X_j)$. Are X_1, \dots, X_n independent?
- (c) Use the law of total variance to find $\text{var}(X)$. Verify your answer using the covariance result of part (b).

Solution. (a) We have, from the law of iterated expectations and the fact $\mathbf{E}[X_i | Q] = Q$,

$$\mathbf{E}[X_i] = \mathbf{E}[\mathbf{E}[X_i | Q]] = \mathbf{E}[Q] = \mu.$$

Since $X = X_1 + \dots + X_n$, it follows that

$$\mathbf{E}[X] = \mathbf{E}[X_1] + \dots + \mathbf{E}[X_n] = n\mu.$$

(b) We have, for $i = j$, using the conditional independence assumption,

$$/ \quad \mathbf{E}[X_i X_j | Q] = \mathbf{E}[X_i | Q] \mathbf{E}[X_j | Q] = Q^2,$$

and

$$\mathbf{E}[X_i X_j] = \mathbf{E}[\mathbf{E}[X_i X_j | Q]] = \mathbf{E}[Q^2].$$

Thus,

$$\text{cov}(X_i, X_j) = \mathbf{E}[X_i X_j] - \mathbf{E}[X_i] \mathbf{E}[X_j] = \mathbf{E}[Q^2] - \mu^2 = \sigma^2.$$

Since $\text{cov}(X_i, X_j) > 0$, X_1, \dots, X_n are not independent.

Also, for $i = j$, using the observation that $X_i^2 = X_i$,

$$\begin{aligned} \text{var}(X_i) &= \mathbf{E}[X_i^2] - (\mathbf{E}[X_i])^2 \\ &= \mathbf{E}[X_i] - (\mathbf{E}[X_i])^2 \\ &= \mu - \mu^2. \end{aligned}$$

(c) Using the law of total variance, and the conditional independence of X_1, \dots, X_n , we have

$$\begin{aligned}
 \text{var}(X) &= \mathbf{E} \text{ var}(X | Q) + \text{var} \mathbf{E}[X | Q] \\
 &= \mathbf{E} \text{ var}(X_1 + \dots + X_n | Q) + \text{var} \mathbf{E}[X_1 + \dots + X_n | Q] \\
 &= \mathbf{E} nQ(1 - Q) + \text{var}(nQ) \\
 &= n\mathbf{E}[Q - Q^2] + n^2 \text{var}(Q) \\
 &= n(\mu - \mu^2 - \sigma^2) + n^2 \sigma^2 \\
 &= n(\mu - \mu^2) + n(n - 1)\sigma^2.
 \end{aligned}$$

To verify the result using the covariance formulas of part (b), we write

$$\begin{aligned}
 \text{var}(X) &= \text{var}(X_1 + \dots + X_n) \\
 &= \sum_{i=1}^n \text{var}(X_i) + \sum_{\{(i,j) | i \neq j\}} \text{cov}(X_i, X_j) \\
 &= n\text{var}(X_1) + n(n - 1)\text{cov}(X_1, X_2) \\
 &= n(\mu - \mu^2) + n(n - 1)\sigma^2.
 \end{aligned}$$

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