Assignment 5

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https://github.com/jatin-tarachandani/AI1103/blob/ main/Assignment_5/main.tex

Problem Statement

Let X, Y, have the joint discrete distribution such that $X|Y = y \sim \text{Binomial}(y, 0.5) \text{ and } Y \sim \text{Poisson}(\lambda), \lambda > 1$ 0, where λ is an unknown parameter. Let T = T(X, Y)be any unbiased estimator of λ . Then

- 1. $Var(T) \leq Var(Y)$ for all λ
- 2. $Var(T) \ge Var(Y)$ for all λ
- 3. $Var(T) \ge \lambda$ for all λ
- 4. Var(T) = Var(Y) for all λ

Solution

Definition 2.1. The pmf for a binomial distribution *Proof.* with parameters n and p is

$$\Pr(X = x; n, p) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x} \quad (2.0.1)$$

Definition 2.2. The pmf for a Poisson distribution with parameter λ is:

$$\Pr(Y = y; \lambda) = \frac{\lambda^{y}}{v!} e^{-\lambda}$$
 (2.0.2)

Lemma 2.1. Since Y has a Poisson distribution, we know that:

$$Var(Y) = \lambda \tag{2.0.3}$$

Definition 2.3. Suppose we have an estimator T(X,Y), acting on a set of 2 RVs X, Y; which estimates a parameter θ , then if

$$E(T(X,Y)) = \theta \tag{2.0.4}$$

the estimator is unbiased.

We define $p_{XY}(x, y)$ as the joint discrete distribution of X and Y. From Bayes Theorem,

$$p_{XY}(x, y) = \Pr(Y = y) \Pr(X = x | Y = y)$$
 (2.0.5)

$$= {}^{y}C_{x}\frac{1}{2^{y}}\frac{\lambda^{y}}{y!}e^{-\lambda}$$
 (2.0.6)

Definition 2.4. If T(X, Y) is an unbiased estimator of a parameter λ , the Cramer-Rao bound states that:

$$Var\left(T\left(X,Y\right)\right) \ge -\frac{1}{E\left(\frac{\partial^{2}\ln\left(p_{XY}\left(x,y\right)\right)\right)}{\partial\lambda^{2}}\right)}$$
 (2.0.7)

Lemma 2.2.

$$\frac{\partial^2 \ln(p_{XY}(x,y))}{\partial \lambda^2} = -\frac{y}{\lambda^2}$$
 (2.0.8)

$$p_{XY}(x, y) = {}^{y}C_{x}\frac{1}{2^{y}}\frac{\lambda^{y}}{y!}e^{-\lambda}$$
 (2.0.9)

$$\frac{\partial^{2} \ln(p_{XY}(x,y))}{\partial \lambda^{2}} = \frac{\partial^{2} \left(\frac{\ln^{y} C_{x} - y \ln 2^{y}}{+ y \ln \lambda - \ln y! - \lambda}\right)}{\partial \lambda^{2}}$$
(2.0.10)

$$=\frac{\partial\left(\frac{y}{\lambda}-1\right)}{\partial\lambda}\tag{2.0.11}$$

$$= -\frac{y}{\lambda^2}$$
 (2.0.12)

which is the required result.

From using (2.0.7) on T(X, Y),

$$Var(T(X,Y)) \ge -\frac{1}{E\left(\frac{\partial^2 \ln(p_{XY}(x,y))}{\partial \lambda^2}\right)}$$
 (2.0.13)

(2.0.14)

From Lemma 2.2,

$$Var(T(X,Y)) \ge -\frac{1}{E\left(-\frac{y}{\lambda^2}\right)}$$
 (2.0.15)

$$\geq \frac{\lambda^2}{E(y)} \tag{2.0.16}$$

$$\geq \lambda$$
 (2.0.17)

because the expectation value of a Poisson distribution with parameter λ is λ .

The correct options are options (2) and (3), since by (2.0.17), we see that

$$Var(T) \ge \lambda = Var(Y)$$

(from (2.0.3)). (1) and (4) do not hold for all λ .