

Assignment 2

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17 March 2022

Q.108 UGC 2017

Let X_1, X_2, \dots, X_n be a random sample from $f_\theta(x)$, a probability density function or a probability mass function. Defined

$$s_n^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1}, \quad \bar{X}_n = \frac{\sum_{i=1}^n X_i}{n}$$

Then s_n^2 is unbiased for θ if

- 1) $f_\theta(x) = \frac{e^{-\theta} \theta^x}{x!}; x > 0$
- 2) $f_\theta(x) = \frac{e^{-\frac{x^2}{2\theta}}}{\sqrt{2\pi\theta}} - \infty < x < \infty, \theta > 0$
- 3) $f_\theta(x) = \frac{e^{-\frac{x}{\theta}}}{\theta} x > 0, \theta > 0$
- 4) $f_\theta(x) = \theta e^{-\theta x} x > 0, \theta > 0$

Answer: Option 1 and Option 2 are correct.

1 Solution

Regardless of the distribution, the expected value of the sample variance is equal to the population variance. Hence the sample variance is said to be an unbiased estimator of population variance.

$$E[s^2] = \sigma^2 \tag{1}$$

Proof:

Let X_1, X_2, \dots, X_n be n independent observations from a population with mean μ and variance σ^2

$$E[\sum X_i] = \sum E[X_i] \tag{2}$$

$$E[cX] = cE[X] \quad (3)$$

We also know that,

$$\sigma^2 = E[X^2] - E[X]^2; \quad E[X^2] = \sigma^2 + \mu^2 \quad (4)$$

If we take n samples from this population, then the sample mean would be μ and sample variance would be $\frac{\sigma^2}{n}$. Then, according to 4,

$$E[\bar{X}^2] = \frac{\sigma^2}{n} + \mu^2 \quad (5)$$

Now,

$$\begin{aligned} E[s_n^2] &= E\left[\frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1}\right] \\ E\left[\sum_{i=1}^n (X_i - \bar{X})^2\right] &= E\left[\sum_{i=1}^n (X_i^2 - 2X_i\bar{X} + \bar{X}^2)\right] \\ &= E\left[\sum_{i=1}^n (X_i^2) - 2\bar{X} \sum_{i=1}^n (X_i) + \sum_{i=1}^n (\bar{X}^2)\right] \\ &= E\left[\sum_{i=1}^n (X_i^2) - 2\bar{X}n\bar{X} + n\bar{X}^2\right] \\ &= E\left[\sum_{i=1}^n (X_i^2) - n\bar{X}^2\right] \\ &= \left[\sum_{i=1}^n (E[X_i^2]) - E[n\bar{X}^2]\right] \end{aligned}$$

Now, substituting 4 and 5,

$$\begin{aligned} &= \left[\sum_{i=1}^n (\sigma^2 + \mu^2) - n\left(\frac{\sigma^2}{n} + \mu^2\right)\right] \\ &= [n\sigma^2 + n\mu^2 - \sigma^2 - n\mu^2] = (n-1)\sigma^2 \end{aligned}$$

$$E[s_n^2] = \frac{(n-1)\sigma^2}{(n-1)} = \sigma^2 \quad (6)$$

Hence the sample variance is said to be an unbiased estimator of the population variance.

2 Checking Options

2.1 Option 1

Option 1 is a probability distribution function following the poisson distribution. A poisson distribution has the same mean and variance. Since the variance for this distribution is θ , s_n^2 is unbiased for this θ . Option 1 is correct

2.2 Option 2

Option 2 is a probability distribution function following the normal distribution with mean 0 and standard deviation of $\sqrt{\theta}$. So the variance of this distribution is θ , therefore, s_n^2 is unbiased for this θ . Option 2 is correct

2.3 Option 3

Option 3 is a probability distribution function following the exponential distribution. The mean for this distribution is θ and variance is θ^2 . Here s_n^2 is unbiased for this θ^2 . Option 3 is incorrect.

2.4 Option 4

Option 4 is a probability distribution function following exponential distribution. The mean for this distribution is $\frac{1}{\theta}$ and variance $\frac{1}{\theta^2}$. s_n^2 is unbiased for $\frac{1}{\theta^2}$. Option 4 is incorrect