

Exercise 2: Determine whether S_n is an unbiased estimator of σ . In case it is not an unbiased estimator, which one is larger $E[S_n]$ or σ ?

Solution:

Consider the sample variance:

$$\overline{S}_n^2 = 1/n * \sum_{i=1}^n (X_i - M_n)^2 \text{ [here, } M_n = \text{Sample Mean]}$$

Now,

$$\begin{aligned} E[\overline{S}_n^2] &= 1/n * E[\sum_{i=1}^n X_i^2 - 2 * M_n * \sum_{i=1}^n X_i + n * M_n^2] \\ &= E[(1/n) * \sum_{i=1}^n X_i^2 - 2 * M_n + M_n^2] \\ &= E[(1/n) * \sum_{i=1}^n X_i^2 - M_n^2] \\ &= \mu^2 + \sigma^2 - (\mu^2 + \frac{\sigma^2}{n}) \\ &\quad [E(X)^2 = \mu^2 + \sigma^2 \text{ and } E[M_n^2] = \mu^2 + \frac{\sigma^2}{n}] \\ &= \frac{n-1}{n} * \sigma^2 \end{aligned}$$

Conclude that \overline{S}_n^2 is a biased estimator of the variance. Nevertheless, note that if n is relatively large, the bias is very small [1]. Since $E[\overline{S}_n^2] = \frac{n-1}{n} * \sigma^2$, so obtain an unbiased estimator of σ^2 by multiplying \overline{S}_n^2 by $\frac{n}{n-1}$. Thus, define

$$\begin{aligned} S_n^2 &= \frac{1}{n-1} * \sum_{i=1}^n (X_i - M_n)^2 \\ &= \frac{1}{n-1} * (\sum_{i=1}^n X_i^2 - n * M_n^2) \end{aligned}$$

By the above discussion, S_n^2 is an unbiased estimator of the variance. Note that if n is large, the difference between S_n^2 and \overline{S}_n^2 is very small. Also define the sample standard deviation as

$$S_n = \sqrt{S_n^2}$$

Although the sample standard deviation is usually used as an estimator for the standard deviation, it is a biased estimator. To see this, note that S is random, so $Var(S) > 0$. Thus,

$$0 < Var(S) = E[\overline{S}_n^2] - (E[S_n])^2 = \sigma^2 - (E[S_n])^2.$$

Therefore, $E[S_n] < \sigma$, which means that S is a biased estimator of σ .

Reference:

1. https://www.probabilitycourse.com/chapter8/8_2_2_point_estimators_for_mean_and_var.php

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It is great that you cited this reference,
but you need to paraphrase them
in your own words.

