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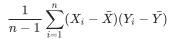
## unbiased estimate of the covariance

Asked 2 years, 10 months ago Active 1 year, 5 months ago Viewed 8k times



How can I prove that







is an unbiased estimate of the covariance Cov(X,Y) where  $\bar{X} = \frac{1}{n} \sum_{i=1}^{n} X_i$  and  $\bar{Y} = \frac{1}{n} \sum_{i=1}^{n} Y_i$  and  $(X_1,Y_1),\ldots,(X_n,Y_n)$  an independent sample from random vector (X,Y)?





Michael Hardy 214k 23 210 491

asked Nov 17 '16 at 23:00

Mama Bullki 86 1 6

- Probably the reason why someone down-voted this question and someone voted to close it is that questions posted here should not be phrased in language suitable for assigning homework. Michael Hardy Nov 18 '16 at 0:09
  - Do the multiplication, and deal with expectations of the resulting terms. BruceET Nov 18 '16 at 0:58
- 2 One cannot show that it is an "unbiased estimate of the covariance". Perhaps you intend: unbiased estimator of the covariance wolfies Nov 18 '16 at 3:26
  - @BruceET: Would you do something substantially different from what is in my answer posted below? Michael Hardy Nov 18 '16 at 23:04
  - That seems to work nicely. Proof that  $E(S^2) = \sigma^2$  is similar, but easier. Perhaps my clue was too simplistic (omitting the  $-\mu + \mu = 0$  trick). BruceET Nov 18 '16 at 23:20  $\nearrow$

## 2 Answers



Let  $\mu = \mathrm{E}(X)$  and  $\nu = \mathrm{E}(Y)$ . Then

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$$\begin{split} &\sum_{i=1}^{n} (X_i - \bar{X})(Y_i - \bar{Y}) \\ &= \sum_{i=1}^{n} \left( (X_i - \mu) + (\mu - \bar{X}) \right) \left( (Y_i - \nu) + (\nu - \bar{Y}) \right) \\ &= \left( \sum_{i} (X_i - \mu)(Y_i - \nu) \right) + \left( \sum_{i} (X_i - \mu)(\nu - \bar{Y}) \right) \\ &+ \left( \sum_{i} (\mu - \bar{X})(Y_i - \nu) \right) + \left( \sum_{i} (\mu - \bar{X})(\nu - \bar{Y}) \right). \end{split}$$

The expected value of the first of the four terms above is

$$\sum_i \mathrm{E}\left((X_i-\mu)(Y_i-\mu)
ight) = \sum_i \mathrm{cov}(X_i,Y_i) = n\,\mathrm{cov}(X,Y).$$

The expected value of the second term is

$$egin{aligned} \sum_i - \operatorname{cov}(X_i, ar{Y}) &= \sum_i - \operatorname{cov}igg(X_i, rac{Y_1 + \cdots + Y_n}{n}igg) \ &= -n\operatorname{cov}igg(X_1, rac{Y_1 + \cdots + Y_n}{n}igg) = -\operatorname{cov}(X_1, Y_1 + \cdots + Y_n) \ &= -\operatorname{cov}(X_1, Y_1) + 0 + \cdots + 0 = -\operatorname{cov}(X, Y). \end{aligned}$$

The third term is similarly that same number.

The fourth term is

$$\sum_{i}^{\text{No "i" appears here.}} \underbrace{\cos(\bar{X}, \bar{Y})}_{\text{cov}(\bar{X}, \bar{Y})} = n \cos(\bar{X}, \bar{Y}) = n \cos\left(\frac{1}{n} \sum_{i} X_{i}, \frac{1}{n} \sum_{i} Y_{i}\right)$$

$$= n \cdot \frac{1}{n^{2}} \left(\underbrace{\cdots + \cot(X_{i}, Y_{j}) + \cdots}_{n^{2} \text{ terms}}\right).$$

This last sum is over all pairs of indices i and j. But the covariances are 0 except the ones in which i = j. Hence there are just n nonzero terms, and we have

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I leave the rest as an exercise.

edited Apr 13 '18 at 20:39



answered Nov 18 '16 at 1:09





Additional Comment, after some thought, following an exchange of Comments with @MichaelHardy:

6

His answer closely parallels the usual demonstration that  $E(S^2) = \sigma^2$  and is easy to follow. However, the proof below, in abbreviated notation I hope is not too cryptic, may be more direct.



$$(n-1)S_{xy}=\sum (X_i-ar{X})(Y_i-ar{Y})=\sum X_iY_i-nar{X}ar{Y}=\sum X_iY_i-rac{1}{n}\sum X_i\sum Y_i.$$

Hence,

$$egin{align} (n-1)E(S_{xy}) &= E\left(\sum X_i Y_i
ight) - rac{1}{n} E\left(\sum X_i \sum Y_i
ight) \ &= n \mu_{xy} - rac{1}{n} [n \mu_{xy} + n(n-1) \mu_x \mu_y] \ &= (n-1)[\mu_{xy} - \mu_x \mu_y] = (n-1) \sigma_{xy}, \end{split}$$

So the expectation of the sample covariance  $S_{xy}$  is the population covariance  $\sigma_{xy} = \text{Cov}(X, Y)$ , as claimed.

Note that  $\mathrm{E}(\sum X_i \sum Y_i)$  has  $n^2$  terms, among which  $\mathrm{E}(X_i Y_i) = \mu_{xy}$  and  $\mathrm{E}(X_i Y_j) = \mu_x \mu_y$ .

edited Nov 19 '16 at 16:23

Michael Hardy
214k 23 210

answered Nov 19 '16 at 6:39

BruceET 38k 7 17 43