Exercices 1

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Consider a random variable X which takes values 1, 2, 3 and 4 with corresponding probabilities 0.1, 0.2, 0.4 and 0.3.

- a What is E(X)?
- b What is $E(X^2)$?
- c What is Var(X)?

Hint: Recall that $Var(X) = E[(X - E(X))^2]$.

X and Y are two random variables such that E(X) = 10, $E(X^2) = 150$, E(Y) = 5, $E(Y^2) = 75$ and E(XY) = 20.

- a What is Cor(X, Y)?
- b What is Cov(5X, 10Y)?
- c What is Var(5X + 10Y)?

The following frequency table was observed from a two-item test where each item was scored 0/1.

	Item $1 = 0$	$Item\ 1 = 1$
Item 2 = 0	42	20
Item 2 = 1	22	16

- a Estimate the difficulty of each item.
- **b** Estimate the variance of the total score.

The following covariance matrix was observed from a three-item test.

$$\Sigma_{X_1,X_2,X_3} = \begin{bmatrix} 1.19 & 0.28 & 0.22 \\ 0.28 & 1.26 & 0.40 \\ 0.22 & 0.40 & 1.47 \end{bmatrix}.$$

- a Calculate the sample variance of $X_1 + X_2 + X_3$.
- b Calculate the estimated correlation between X_1 and X_3 .

Show that the sample mean is an unbiased estimator of the expected value of a $R.V.\ X.$ I.e. derive the expected value of

$$\bar{x} = \frac{\sum_{i=1}^{n} X_i}{n}.$$

Assume that observations X_i are independent. Derive the variance of

$$\bar{x} = \frac{\sum_{i=1}^{n} X_i}{n}.$$

Show that

$$\mathsf{MSE}(\hat{\theta}) = \mathsf{Var}(\hat{\theta}) + [\mathsf{Bias}(\hat{\theta})]^2.$$

Derive the expected value of

$$s_{xy}^* = \frac{\sum_{i=1}^n X_i Y_i - n\bar{x}\bar{y}}{n-1},$$

where observations $k, l, k \neq l$ are independent.

L3 Task 1: solution

$$E(X) = 1 \times 0.1 + 2 \times 0.2 + 3 \times 0.4 + 4 \times 0.3 = 2.9$$

$$E(X^{2}) = 1^{2} \times 0.1 + 2^{2} \times 0.2 + 3^{2} \times 0.4 + 4^{2} \times 0.3 = 9.3$$

$$Var(X) = E(X^{2}) - [E(X)]^{2} = 9.3 - 2.9^{2} = 0.89$$
To see that
$$Var(X) = E(X^{2}) - [E(X)]^{2},$$

$$E[(X - E(X))^{2}] = E[X^{2} + E(X) \times E(X) - 2 \times X \times E(X)]$$

$$= E(X^{2}) + E(X) \times E(X) - 2E(X) \times E(X)$$

$$= E(X^{2}) - E(X) \times E(X)$$

$$= E(X^{2}) - [E(X)]^{2}.$$

L3 Task 2: solution

We can first note that

$$Cov(X, Y) = E(XY) - E(X) \times E(Y) = 20 - 50 = -30$$

$$Var(X) = E(X^{2}) - (E(X))^{2} = 150 - 100 = 50$$

$$Var(Y) = E(Y^{2}) - (E(Y))^{2} = 75 - 25 = 50$$

We thus obtain

$$Cor(X, Y) = \frac{Cov(X, Y)}{\sigma_X \sigma_Y} = \frac{-30}{\sqrt{50}\sqrt{50}} = -30/50 = -0.6$$

$$Cov(5X, 10Y) = 5 \times 10 \times Cov(X, Y) = -1500$$

$$Var(5X + 10Y) = 5^2 Var(X) + 10^2 Var(Y) + 2 \times 5 \times 10 \times Cov(X, Y)$$

$$= 25 \times 50 + 100 \times 50 + 100 \times (-30) = 3250$$

L3 Task 3: solution

The difficulty is, per the textbook, defined as the mean score of each item. Hence we obtain $\hat{p}_1=(20+16)/100=0.36$ and $\hat{p}_2=(22+16)/100=0.38$.

From the formula given, we have that the variance of the sum score is equal to the sum of the variances and covariances for all items/item pairs. We have

$$s_1^2 = \hat{p}_1 \times (1 - \hat{p}_1) = 0.36 \times 0.64 = 0.2304$$

 $s_2^2 = \hat{p}_2 \times (1 - \hat{p}_2) = 0.38 \times 0.62 = 0.2356$

and

$$s_{12} = \hat{p}_{12} - \hat{p}_1 \times \hat{p}_2 = 0.16 - 0.36 \times 0.38 = 0.0232.$$

Hence we obtain

$$s_{1+2} = 0.2304 + 0.2356 + 2 \times 0.0232 = 0.5124$$



L3 Task 4: solution

$$Var(X_1 + X_2 + X_3) = \sum_{i=1}^{3} \sum_{j=1}^{3} s_{ij} = 5.72.$$

$$\hat{\mathsf{Cor}}(X_1, X_3) = \frac{\hat{\mathsf{Cov}}(X_1, X_3)}{\hat{\sigma}_{X_1} \hat{\sigma}_{X_3}} = \frac{0.22}{\sqrt{1.19}\sqrt{1.47}} \approx 0.166.$$

L3 Task 5: solution

$$E(\bar{x}) = E\left(\frac{\sum_{i=1}^{n} X_i}{n}\right)$$

$$= \frac{\sum_{i=1}^{n} E(X_i)}{n}$$

$$= \frac{nE(X)}{n}$$

$$= E(X).$$

We thus have that

$$E(\bar{x} - E(X)) = E(X) - E(X) = 0,$$

and \bar{x} is an unbiased estimator of E(X).

L3 Task 6: solution

$$Var(\bar{x}) = Var\left(\frac{\sum_{i=1}^{n} X_i}{n}\right)$$

$$= \frac{\sum_{i=1}^{n} Var(X_i)}{n^2}$$

$$= \frac{nVar(X)}{n^2}$$

$$= \frac{Var(X)}{n}$$

L3 Task 7: solution

$$\begin{aligned} \mathsf{MSE}(\hat{\theta}) &= E((\hat{\theta} - \theta)^2) \\ &= E(\hat{\theta}^2 + \theta^2 - 2\hat{\theta}\theta) \\ &= E(\hat{\theta}^2) + \theta^2 - 2\theta E(\hat{\theta}) \\ &= E(\hat{\theta}^2) + (\theta - E(\hat{\theta}))^2 - (E(\hat{\theta}))^2 \\ &= E(\hat{\theta}^2) - (E(\hat{\theta}))^2 + (\theta - E(\hat{\theta}))^2 \\ &= \mathsf{Var}(\hat{\theta}) + (\mathsf{Bias}(\hat{\theta}))^2. \end{aligned}$$

L3 Task 8: solution I

We have

$$E\left(\sum_{i=1}^{n} X_i X_i\right) = \sum_{i=1}^{n} E(X_i X_i) = nE(XY)$$

L3 Task 8: solution II

and

$$E\left[\sum_{i=1}^{n} \bar{x}\bar{y}\right] = nE\left(\sum_{i=1}^{n} X_i/n \sum_{j=1}^{n} X_j/n\right)$$

$$= (1/n)E\left(\sum_{i=1}^{n} X_i \sum_{j=1}^{n} X_j\right)$$

$$= (1/n) \sum_{i=1}^{n} E\left(X_i \sum_{j=1}^{n} X_j\right)$$

$$= (1/n)(n-1)nE(X)E(Y) + (1/n)nE(XY)$$

$$= (n-1)E(X)E(Y) + E(XY).$$

L3 Task 8: solution III

We thus obtain

$$E\left[\sum_{i=1}^{n} X_{i}X_{i} - n\bar{x}\bar{y}\right] = nE(XY) - (n-1)E(X)E(Y) - E(XY)$$
$$= (n-1)E(XY) - (n-1)E(X)E(Y).$$

Hence, we have that

$$E(s_{xy}) = \frac{n-1}{n} [E(XY) - E(X)E(Y)]$$
$$= \frac{n-1}{n} \sigma_{xy}.$$

Consequently, an unbiased estimator of σ_{xy} is

$$s_{xy}^* = \frac{\sum_{i=1}^n (X_i - \bar{x})(X_i - \bar{y})}{n-1}.$$

