Probability and Statistics I
21. Expected Values, Covariance, Correlation

5.2 Expected Values, Covariance, and Correlation

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5.2.1 Expected Values

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Expected Values for Jointly Distributed RVs

Suppose that X and Y are jointly distributed discrete random variables with joint pmf p(x, y). The expected value of any function of X and Y, h(X, y), is

$$E[h(X,Y)] = \sum_{x} \sum_{y} h(x,y)p(x,y)$$

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Expected Values for Jointly Distributed RVs

Suppose that X and Y are jointly distributed continuous random variables with pdf f(x,y). The expected value of any function of X and Y, h(X,y), is

$$E[h(X,Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(x,y)f(x,y) \ dxdy$$

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Example 21.1 ctd

The simplest possible joint distribution is that for two Bernoulli random variables. Suppose that X and Y take the values 0 and 1 according to the following joint pmf:

$$\begin{array}{c|ccccc} x & 0 & 1 \\ y & 0 & 1 & 0 & 1 \\ \hline \rho(x,y) & \rho_{00} & \rho_{01} & \rho_{10} & \rho_{11}. \end{array}$$

a) What is the expected value of XY?

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5.2.2 Covariance and Correlation

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Covariance

The covariance between two random variables X and Y is

$$Cov(X, Y) = E[(X - \mu_X)(Y - \mu_Y)]$$

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Covariance

The covariance between two discrete random variables X and Y is

$$Cov(X, Y) = E[(X - \mu_X)(Y - \mu_Y)]$$
$$= \sum_{x} \sum_{y} h(x, y) p(x, y).$$

The covariance between two continuous random variables X and Y is

$$Cov(X, Y) = E[(X - \mu_X)(Y - \mu_Y)]$$
$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(x, y) f(x, y) \ dxdy.$$

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Shortcut Formula for Covariances

The covariance of any two random variables X and Y is

$$Cov(X, Y) = E[(X - E(X))(Y - E(Y))].$$

It is often more efficient to use the shortcut formula

$$Cov(X, Y) = E(XY) - E(X)E(Y).$$

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Correlation

The correlation (coefficient) of X and Y is

$$\operatorname{Corr}(X,Y) = \rho(X,Y) = \frac{\operatorname{Cov}(X,Y)}{\sigma_X \sigma_Y}.$$

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Correlation

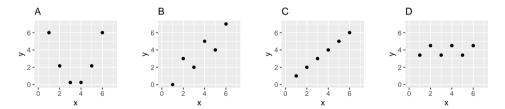
Correlation measures the strength of the linear relationship between two variables.

- Corr(X, Y) = -1: Y = aX + b for some a < 0.
- $-1 < \operatorname{Corr}(X, Y) < 0$: $Y \approx aX + b$ for some a < 0.
- Corr(X, Y) = 0: the best fitting line has slope 0.
- 0 < Corr(X, Y) < 1: $Y \approx aX + b$ for some a <> 0.
- Corr(X, Y) = 1: Y = aX + b for some a > 0.

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Example 21.1

Each of the following plots represents the joint pmf of two random variables, X and Y. The points, (x, y) represent the possible values of (X, Y). The distribution places equal probability, 1/6, at each point.

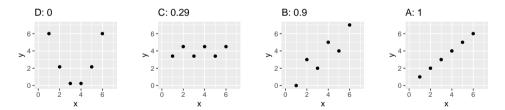


Order the plots according to their correlation.

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Example 21.1 ctd

Each of the following plots represents the joint pmf of two random variables, X and Y. The points, (x, y) represent the possible values of (X, Y). The distribution places equal probability, 1/6, at each point.



Order the plots according to their correlation.

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Example 21.1 ctd

The simplest possible joint distribution is that for two Bernoulli random variables. Suppose that X and Y take the values 0 and 1 according to the following joint pmf:

$$egin{array}{c|cccc} x & 0 & 1 \\ y & 0 & 1 & 0 & 1 \\ \hline p(x,y) & p_{00} & p_{01} & p_{10} & p_{11}. \end{array}$$

- What is the expected value of XY?
- ② What are the covariance and correlation of X and Y?

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5.2.3 Sums of Random Variables

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Sums of Random Variables

Let X_1, X_2, \ldots be any sequence of random variables and c_1, c_2, \ldots be any sequence of constants. Then

$$E\left[\sum c_k X_k\right] = \sum c_k E(X_k).$$

The sum may either be finite or infinite.

In the case of two random variables

$$E(aX + bY) = aE(X) + bE(Y).$$

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Variance of a Sum of Random Variables

Let X and Y be any two random variables and $a, b \in \Re$. Then

$$Var(aX + bY) = a^{2}Var(X) + 2abCov(X, Y) + b^{2}Var(Y).$$

If X and Y are independent then $Cov(X, Y) = 0^1$ In this case

$$Var(aX + bY) = a^{2}Var(X) + b^{2}Var(Y).$$

¹The reverse is **not** true in general.

Example 21.1 ctd

The simplest possible joint distribution is that for two Bernoulli random variables. Suppose that X and Y take the values 0 and 1 according to the following joint pmf:

$$\begin{array}{c|ccccc} x & 0 & 1 \\ y & 0 & 1 & 0 & 1 \\ \hline p(x,y) & p_{00} & p_{01} & p_{10} & p_{11}. \end{array}$$

- What is the expected value of XY?
- ② What are the covariance and correlation of X and Y?
- **3** What are the mean and variance of Z = 2X + 4Y?

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5.2.4 Products of Independent Random Variables

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Products of Independent Random Variables

Let X_1, X_2, \ldots be any set of independent random variables. Then

$$E\left[\prod X_k\right]=\prod E(X_k).$$

In the case of two independent random variables

$$E(XY)=E(X)E(Y).$$

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Example 21.1 ctd

The simplest possible joint distribution is that for two Bernoulli random variables. Suppose that X and Y take the values 0 and 1 according to the following joint pmf:

X	0		1	
y	0	1	0	1
p(x,y)	p_{00}	p_{01}	p_{10}	p_{11} .

- What is the expected value of XY?
- What are the covariance and correlation of X and Y?
- **3** What are the mean and variance of Z = 2X + 4Y?
- Under what conditions are X and Y independent? What is are the mean XY in this case?

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Questions?

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Exercise 21.1

Consider rolling two fair, three-sided die. Let S denote the sum of the values showing on the two die and D the absolute value of the difference. E.g., if one die shows the value 1 and the second shows the value 2 then S=3 and D=1, regardless of which was thrown first.

- a) Construct a table showing the joint pmf of S and D.
- b) Compute the marginal pmf of both S and D.
- c) Compute the expected value and variance of S and D.
- d) Compute the covariance and correlation of S and D.
- e) Are S and D independent? Justify your answer.

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