
5.3.1 Covariance and Correlation

Consider two random variables X and Y . Here, we define the **covariance** between X and Y , written $\text{Cov}(X, Y)$. The covariance gives some information about how X and Y are statistically related. Let us provide the definition, then discuss the properties and applications of covariance.

The **covariance** between X and Y is defined as

$$\text{Cov}(X, Y) = E[(X - EX)(Y - EY)] = E[XY] - (EX)(EY).$$

Note that

$$\begin{aligned} E[(X - EX)(Y - EY)] &= E[XY - X(EY) - (EX)Y + (EX)(EY)] \\ &= E[XY] - (EX)(EY) - (EX)(EY) + (EX)(EY) \\ &= E[XY] - (EX)(EY). \end{aligned}$$

Intuitively, the covariance between X and Y indicates how the values of X and Y move relative to each other. If large values of X tend to happen with large values of Y , then $(X - EX)(Y - EY)$ is positive on average. In this case, the covariance is positive and we say X and Y are positively correlated. On the other hand, if X tends to be small when Y is large, then $(X - EX)(Y - EY)$ is negative on average. In this case, the covariance is negative and we say X and Y are negatively correlated.

Example 5.32

Suppose $X \sim \text{Uniform}(1, 2)$, and given $X = x$, Y is exponential with parameter $\lambda = x$. Find $\text{Cov}(X, Y)$.

Solution

We can use $\text{Cov}(X, Y) = EXY - EXEY$. We have $EX = \frac{3}{2}$ and

$$\begin{aligned}
EY &= E[E[Y|X]] && \text{(law of iterated expectations (Equation 5.17))} \\
&= E\left[\frac{1}{X}\right] && \text{(since } Y|X \sim \text{Exponential}(X)\text{)} \\
&= \int_1^2 \frac{1}{x} dx \\
&= \ln 2.
\end{aligned}$$

We also have

$$\begin{aligned}
EXY &= E[E[XY|X]] && \text{(law of iterated expectations)} \\
EXY &= E[XE[Y|X]] && \text{(since } E[X|X = x] = x\text{)} \\
&= E\left[X\frac{1}{X}\right] && \text{(since } Y|X \sim \text{Exponential}(X)\text{)} \\
&= 1.
\end{aligned}$$

Thus,

$$\text{Cov}(X, Y) = E[XY] - (EX)(EY) = 1 - \frac{3}{2} \ln 2.$$

Now we discuss the properties of covariance.

Lemma 5.3

The covariance has the following properties:

1. $\text{Cov}(X, X) = \text{Var}(X)$;
2. if X and Y are independent then $\text{Cov}(X, Y) = 0$;
3. $\text{Cov}(X, Y) = \text{Cov}(Y, X)$;
4. $\text{Cov}(aX, Y) = a\text{Cov}(X, Y)$;
5. $\text{Cov}(X + c, Y) = \text{Cov}(X, Y)$;
6. $\text{Cov}(X + Y, Z) = \text{Cov}(X, Z) + \text{Cov}(Y, Z)$;
7. more generally,

$$\text{Cov}\left(\sum_{i=1}^m a_i X_i, \sum_{j=1}^n b_j Y_j\right) = \sum_{i=1}^m \sum_{j=1}^n a_i b_j \text{Cov}(X_i, Y_j).$$