



Statistics Examples

Mean, Variance, and Standard Deviation

Let X_1, X_2, \dots, X_n be n observations of a random variable X . We wish to measure the average of X_1, X_2, \dots, X_n in some sense. One of the most commonly used statistics is the mean, μ_X , defined by the formula

$$\mu_X = \bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

Next, we wish to obtain some measure of the variability of the data. The statistics most often used are the variance σ_X^2 and the standard deviation $\sigma_X = \sqrt{\sigma_X^2}$. We have

$$\sigma_X = \sqrt{\frac{1}{n} \left\{ \sum_{i=1}^n X_i^2 - \frac{1}{n} \left(\sum_{i=1}^n X_i \right)^2 \right\}}$$

It is easy to show that the variance is simply the mean squared deviation from the mean.

Covariance and Correlation

Next, let $(X_1, Y_1), (X_2, Y_2), \dots, (X_n, Y_n)$ be n pairs of values of two random variables X and Y . We wish to measure the degree to which X and Y vary together, as opposed to being independent. The first statistic we will calculate is the covariance σ_{XY} given by

$$\sigma_{XY} = \frac{1}{n} \left\{ \sum_{i=1}^n X_i Y_i - \frac{1}{n} \left(\sum_{i=1}^n X_i \right) \left(\sum_{i=1}^n Y_i \right) \right\}$$

Actually, a much better measure of correlation can be obtained from the formula

$$\rho_{XY} = \frac{\sum_{i=1}^n X_i Y_i - \frac{1}{n} \left(\sum_{i=1}^n X_i \right) \left(\sum_{i=1}^n Y_i \right)}{\sqrt{\left\{ \sum_{i=1}^n X_i^2 - \frac{1}{n} \left(\sum_{i=1}^n X_i \right)^2 \right\} \left\{ \sum_{i=1}^n Y_i^2 - \frac{1}{n} \left(\sum_{i=1}^n Y_i \right)^2 \right\}}}$$

The quantity ρ_{XY} is known as the coefficient of correlation of X and Y .

The Covariance Matrix

Covariances and variances are sometimes arranged in a matrix known as a covariance matrix. In our case, the covariance matrix will be a 2×2 matrix:

$$\mathbf{C} = \begin{bmatrix} \sigma_X^2 & \sigma_{XY} \\ \sigma_{XY} & \sigma_Y^2 \end{bmatrix}$$

The eigenvalues of the covariance matrix are sometimes of interest. These are obtained in the usual way by solving the characteristic equation:

$$p(\lambda) = \det(\lambda \mathbf{I} - \mathbf{C}) = \begin{vmatrix} \lambda - \sigma_X^2 & -\sigma_{XY} \\ -\sigma_{XY} & \lambda - \sigma_Y^2 \end{vmatrix}$$