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Linear combinations of normal random variables

One property that makes the normal distribution extremely tractable from an analytical viewpoint is its closure under linear combinations: the linear combination of two [independent random variables](#) having a [normal distribution](#) also has a normal distribution. The following sections present a multivariate generalization of this elementary property and then discuss some special cases.

Linear transformation of a multivariate normal random vector

A linear transformation of a [multivariate normal random vector](#) also has a multivariate normal distribution, as illustrated by the following proposition.

Proposition Let \mathbf{X} be a p -dimensional multivariate normal random vector with mean $\boldsymbol{\mu}$ and [covariance matrix](#) $\boldsymbol{\Sigma}$. Let \mathbf{a} be an $1 \times p$ real vector and \mathbf{B} an $n \times p$ full-rank real matrix. Then the n -dimensional random vector \mathbf{Y} defined by

$\mathbf{Y} = \mathbf{B}\mathbf{X} + \mathbf{a}$ has a multivariate normal distribution with mean

and covariance matrix

Proof

The following examples present some important special cases of the above property.

Example 1 - Sum of two independent normal random variables

The sum of two independent normal random variables has a normal distribution, as stated in the following:

Example Let X be a random variable having a normal distribution with mean μ_X and variance σ_X^2 . Let Y be a random variable, independent of X , having a normal distribution with mean μ_Y and variance σ_Y^2 . Then, the random variable Z defined as:

has a normal distribution with mean

and variance

Proof

Example 2 - Sum of more than two mutually independent normal random variables

The sum of more than two independent normal random variables also has a normal distribution, as shown in the following example.

Example Let X_1, X_2, \dots, X_n be mutually independent normal random variables, having means $\mu_1, \mu_2, \dots, \mu_n$ and variances $\sigma_1^2, \sigma_2^2, \dots, \sigma_n^2$. Then, the random variable Y defined as

$Y = X_1 + X_2 + \dots + X_n$ has a normal distribution with mean

and variance

Proof

Example 3 - Linear combinations of mutually independent normal random variables

The properties illustrated in the previous two examples can be further generalized to linear combinations of mutually independent normal random variables.

Example Let X_1, \dots, X_n be mutually independent normal random variables, having means μ_1, \dots, μ_n and variances $\sigma_1^2, \dots, \sigma_n^2$. Let a_1, \dots, a_n be constants. Then, the random variable Y defined as

$Y = a_1 X_1 + \dots + a_n X_n$ has a normal distribution with mean

and variance

Proof

Example 4 - Linear transformation of a normal random variable

A special case of the above proposition obtains when X has dimension 1 (i.e., it is a random variable).

Example Let X be a normal random variable with mean μ and variance σ^2 . Let a and b be two constants (with $b \neq 0$). Then the random variable Y defined by

$Y = a + bX$ has a normal distribution with mean

and variance

Proof

Example 5 - Linear combinations of mutually independent normal random vectors

The property illustrated in Example 3 can be generalized to linear combinations of mutually independent normal random vectors.

Example Let $\mathbf{X}_1, \mathbf{X}_2, \dots, \mathbf{X}_k$ be mutually independent normal random vectors, having means $\boldsymbol{\mu}_1, \boldsymbol{\mu}_2, \dots, \boldsymbol{\mu}_k$ and covariance matrices $\boldsymbol{\Sigma}_1, \boldsymbol{\Sigma}_2, \dots, \boldsymbol{\Sigma}_k$. Let $\mathbf{A}_1, \mathbf{A}_2, \dots, \mathbf{A}_k$ be real full-rank matrices. Then, the random vector defined as

$\mathbf{Y} = \mathbf{A}_1 \mathbf{X}_1 + \mathbf{A}_2 \mathbf{X}_2 + \dots + \mathbf{A}_k \mathbf{X}_k$ has a normal distribution with mean

and covariance matrix

■ Proof

Solved exercises

Below you can find some exercises with explained solutions.

Exercise 1

Let

be a multivariate normal random vector with mean

and covariance matrix

Find the distribution of the random variable defined as

■ Solution

Exercise 2

Let Z_1, \dots, Z_n be mutually independent standard normal random variables. Let c be a constant. Find the distribution of the random variable Y defined as

Solution

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