

Theorem If $X_i \sim N(\mu_i, \sigma_i^2)$ for $i = 1, 2, \dots, n$ and X_1, X_2, \dots, X_n are mutually independent, then

$$\sum_{i=1}^n a_i X_i \sim N\left(\sum_{i=1}^n a_i \mu_i, \sum_{i=1}^n a_i^2 \sigma_i^2\right)$$

for nonzero real constants a_1, a_2, \dots, a_n .

Proof The moment generating function of X_i is

$$M_{X_i}(t) = \exp\left(\mu_i t + \frac{1}{2} \sigma_i^2 t^2\right) \quad -\infty < t < \infty; i = 1, \dots, n.$$

The moment generating function of $a_i X_i$ is

$$M_{a_i X_i}(t) = \exp\left(a_i \mu_i t + \frac{1}{2} a_i^2 \sigma_i^2 t^2\right) \quad -\infty < t < \infty; i = 1, \dots, n.$$

Since X_1, X_2, \dots, X_n are mutually independent, the moment generating function of the linear combination is

$$M_{a_1 X_1 + a_2 X_2 + \dots + a_n X_n}(t) = \prod_{i=1}^n M_{a_i X_i}(t) = \exp\left(\left(\sum_{i=1}^n a_i \mu_i\right) t + \frac{1}{2} \left(\sum_{i=1}^n a_i^2 \sigma_i^2\right) t^2\right)$$

for $-\infty < t < \infty$. Therefore

$$\sum_{i=1}^n a_i X_i \sim N\left(\sum_{i=1}^n a_i \mu_i, \sum_{i=1}^n a_i^2 \sigma_i^2\right).$$

APPL illustration: For $n = 3$ and $a_1 = a_2 = a_3 = 1$, the APPL statements

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X1 := NormalRV(mu1, sigma1);
X2 := NormalRV(mu2, sigma2);
X3 := NormalRV(mu3, sigma3);
Convolution(X1, X2, X3);
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yield the probability density function of a $N(\mu_1 + \mu_2 + \mu_3, \sigma_1^2 + \sigma_2^2 + \sigma_3^2)$ random variable.