

Note Sample variance. (s^2)

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

Note Here in sample variance, we have divided it by $(n-1)$ in place of n because if we divide it by n then sample variance will not be unbiased for population variance and we will lose very important characteristic in statistics. Since, it is little bit deep concept so just skip these things and remember the formula of sample variance.

population

Total (set) of observations or collection of some measurements is known as population.

Sample: A subset of population is known as sample, and the process of selecting sample from population is known as sampling.

002 2.518 2.511 0 s_2

0.0 0.0 0.0 0.0 (s_2)

$$0.0 \times 2.518 + 0.0 \times 2.511 + 0.0 \times 0 + 0.0 \times 0 = 0$$

$$2.518$$

0.0 0.0 0.0 0.0

$$0.0 \times (2.518 - 0.0) + 0.0 \times (2.511 - 0.0) + 0.0 \times (0 - 0.0) + 0.0 \times (0 - 0.0) = 0$$

$$2.518$$

Q. Let x_1, x_2, \dots, x_n are iid (and) follow the $N(\mu, \sigma^2)$

where μ and σ^2 are characteristic of population.
then find the sampling distribution of \bar{x} .

Soln

$$\bar{x} = \frac{1}{n} (x_1 + x_2 + \dots + x_n)$$

$$= \frac{1}{n} \sum_{i=1}^n x_i$$

Note: We know that if $x_i; i=1, 2, \dots, n$ follows the normal distribution then any linear combination of these x_i also follows the normal dist. (we prove it) later.

Now, we have to find the distribution of \bar{x}

$$\bar{x} = \frac{1}{n} (x_1 + x_2 + \dots + x_n)$$

$$E(\bar{x}) = E\left[\frac{1}{n} (x_1 + x_2 + \dots + x_n)\right]$$

$$= \frac{1}{n} E(x_1 + x_2 + \dots + x_n)$$

$$= \frac{1}{n} [E(x_1) + E(x_2) + \dots + E(x_n)]$$

$$= \frac{1}{n} (\mu + \mu + \dots + \mu) \quad \text{(n-times)} = \frac{n\mu}{n} = \mu.$$

Now, variance of \bar{x} ,

$$V(\bar{x}) = V\left(\frac{1}{n} (x_1 + x_2 + \dots + x_n)\right)$$
$$= \frac{1}{n^2} [V(x_1) + V(x_2) + \dots + V(x_n)]$$

$$= \frac{1}{n^2} (\sigma^2 + \sigma^2 + \dots + \sigma^2) \quad \text{(n-times)}$$

$$V(\bar{x}) = \frac{n\sigma^2}{n^2} = \frac{\sigma^2}{n}$$

$$\therefore \text{Thus, } \bar{x} \sim N\left(\mu, \frac{\sigma^2}{n}\right).$$

Result: If $X_i \sim N(\mu, \sigma^2)$ and X_i are i.i.d then any linear combination also follow the same distribution.

Proof Consider $Z = a_1 X_1 + a_2 X_2 + \dots + a_n X_n$

$$E(Z) = E(a_1 X_1 + a_2 X_2 + \dots + a_n X_n)$$

$$= a_1 E(X_1) + a_2 E(X_2) + \dots + a_n E(X_n)$$

$$= a_1 \mu + a_2 \mu + \dots + a_n \mu$$

$$= (a_1 + a_2 + \dots + a_n) \mu$$

Note: We know that the linear combination of independent normal variables is also normal.

$$V(Z) = V(a_1 X_1 + a_2 X_2 + \dots + a_n X_n)$$

$$= (a_1^2 + a_2^2 + \dots + a_n^2) \sigma^2$$

Thus; $Z \sim N((a_1 + a_2 + \dots + a_n) \mu, (a_1^2 + a_2^2 + \dots + a_n^2) \sigma^2)$.

Note Each X_i are independent and have the same distribution (i.i.d) then $Cov(X_i, X_j) = 0 \quad \forall i \neq j$.

$$\begin{pmatrix} (X_1) V \\ (X_2) V \\ \vdots \\ (X_n) V \end{pmatrix} \left(\frac{(X_1 + \dots + X_n)}{n} \right) V = (\bar{X}) V$$

$$(a_1^2 + a_2^2 + \dots + a_n^2) \sigma^2 \frac{1}{n} =$$

$$\frac{S_0}{n} = \frac{S_0 n}{S_N} = (\bar{X}) V$$

$$\left(\frac{S_0}{n}, \mu \right) \sim \bar{X} \quad \text{Thus } \therefore$$