

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

$$S^2 = \frac{1}{(n-1)} \sum (x_i - \bar{x})^2$$

Theorem

1) \bar{X} and S^2 are independent

$$2) \frac{(n-1)S^2}{\sigma^2} = \frac{\sum (x_i - \bar{x})^2}{\sigma^2} \sim \chi^2(n-1)$$

If X_1, X_2, \dots, X_n are a random sample of size n from $N(\mu, \sigma^2)$

Proof

$$1) \underbrace{\sum \frac{(x_i - \mu)^2}{\sigma^2}}_{V \sim \chi^2(n)} = \underbrace{\sum \frac{(x_i - \bar{x})^2}{\sigma^2}}_{V_1 \sim ?} + \underbrace{\frac{n(x_i - \mu)^2}{\sigma^2}}_{V_2 \sim \chi^2(1)}$$

$$\underbrace{\left(\frac{(x_i - \mu)}{\sigma} \right)^2}_{Z} \sim \chi^2(1)$$

This shows that the second part of the theorem is true if the first part is true

If \bar{X} and S^2 are independent, then we have:

$$M_V(t) = M_{V_1}(t) M_{V_2}(t)$$

$\chi^2(n) \qquad \qquad \chi^2(1)$

$$\frac{1}{(1-2t)^{n/2}} = M_{V_1}(t) \left(\frac{1}{(1-2t)^{1/2}} \right)$$

So

$$M_{V_1}(t) = \frac{(1-2t)^{1/2}}{(1-2t)^{n/2}} = \frac{1}{(1-2t)^{\frac{n-1}{2}}}$$

$$V_1 \sim \chi^2(n-1)$$

Note this is the $\chi^2(n-1)$ mgf.