

**TOOL:** For  $n$  independent samples,  $X_i$ , from a normal (or gaussian) distribution with mean  $\mu$  and variance  $\sigma^2$ , the probability density function of the normalized sample variance,

$$\chi^2 = \frac{(n-1)S^2}{\sigma^2} = v \frac{S^2}{\sigma^2}$$

$$\text{(where } v = n - 1, S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2, \text{ and } \bar{X} = \frac{1}{n} \sum_{i=1}^n X_i \text{)}$$

has a chi-squared distribution with  $v$  degrees of freedom:

$$\chi^2 \sim \chi_v^2$$

or

$$f_{\chi^2}(x) = \begin{cases} \frac{1}{2^{v/2} \Gamma(v/2)} x^{(v/2)-1} e^{-x/2} & x > 0 \\ 0 & x \leq 0 \end{cases}$$

**PROOF:** See [derivation of  \$\chi^2\$  distribution](#) | [\(pdf\)](#).

**REF:** Ronald E. Walpole, Raymond H. Myers, Sharon L. Myers, and Keying Ye, *Probability and Statistics for Engineers and Scientists*, 8th Ed., Upper Saddle River, NJ: Prentice Hall, 2007.