Case 1: One Variable (Population)

Assumptions	σ is known	σ is unknown	$n(p_0) > 5$ and $n(1 - p_0) > 5$
Parameter of interest:	Mean, μ	Mean, μ	Proportion, p
Confidence Interval Formula:	$\bar{x} \pm z(\alpha/2) \frac{\sigma}{\sqrt{n}}$	$ar{x} \pm t(df, \alpha/2) rac{s}{\sqrt{n}}$ with $df = n - 1$	$p' \pm z(\alpha/2) \sqrt{\frac{p'q'}{n}}$
Name of Hypothesis	One sample z -test,	One sample <i>t</i> -test,	One sample test of
Test, H_0	H_0 : $\mu = \mu_0$	H_0 : $\mu = \mu_0$	$proportion H_0: p = p_0$
Test Statistic Formula:	$z^* = \frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}}$	$t^* = rac{ar{x} - \mu_0}{s/\sqrt{n}}$ with $df = n-1$	$z^* = \frac{p' - p_0}{\sqrt{\frac{p_0 q_0}{n}}}$
p-value:	H_a : $\mu > \mu_0$, p-value= $P(z \ge z^*)$ H_a : $\mu < \mu_0$, p-value= $P(z \le z^*)$ H_a : $\mu \ne \mu_0$, p-value= $P(z \ge z^*)$	$\begin{aligned} &H_a \colon \mu > \mu_0, \text{ p-value} = P(t(df) \geq t^*) \\ &H_a \colon \mu < \mu_0, \text{ p-value} = P(t(df) \leq t^*) \\ &H_a \colon \mu \neq \mu_0, \text{ p-value} = 2 \times P(t(df) \geq t^*) \end{aligned}$	$\begin{split} &H_a : p > p_0 \text{, p-value} = P(z \geq z^*) \\ &H_a : p < p_0 \text{, p-value} = P(z \leq z^*) \\ &H_a : p \neq p_0 \text{, p-value} = 2 \times P(z \geq z^*) \end{split}$

Case 3: More than Two Populations (Multinomial Experiment, Contingency Table, Test of Homogeneity - ANOVA)

	Multinomial Experiment, Contingency Table, Homogeneity		Analysis of Variance (ANOVA)
Parameter of interest:	Probability: p_1, p_2, \cdots, p_k	Probability	Mean: $\mu_1, \mu_2, \cdots, \mu_c$
H_{0}	$H_0: p_1 = p_{10}, \cdots, p_k = p_{k0}$	H_0 : Independency (Homogeneity)	$H_0: \mu_1 = \mu_2 = \dots = \mu_c$
Test Statistic Formula:	$\chi^{2*} = \sum_{\text{all cells}} \frac{(O-E)^2}{E}$		$F^* = \frac{\text{MS(factor)}}{\text{MS(error)}} = \frac{\text{SS(factor)/df(factor)}}{\text{SS(error)/df(error)}}$
df and other items	$df = k - 1$ $E_i = n \times p_{i0}$	$df = (r - 1)(c - 1)$ $E_{ij} = \frac{R_i \times C_j}{n}$	$df(factor) = c - 1, df(error) = n - c$ $SS(factor) = \left(\frac{C_1^2}{k_1} + \frac{C_2^2}{k_2} + \dots + \frac{C_c^2}{k_c}\right) - \frac{(\sum x)^2}{n}$ $SS(error) = \sum (x^2) - \left(\frac{C_1^2}{k_1} + \frac{C_2^2}{k_2} + \dots + \frac{C_c^2}{k_c}\right)$
p-value:	p-value= $P(\chi^2(df) \ge \chi^{2*})$		p-value= $P(F(df_n, df_d) \ge F^*)$