

Population Parameter	Sample Statistic	SE	Critical Value	Test Statistic for a HT
$p$	$\hat{p}$	$\sqrt{\frac{\hat{p}\hat{q}}{n}}$ for a CI $\sqrt{\frac{p_0q_0}{n}}$ for a HT	$z_{\alpha/2}$ for a CI $z_{\alpha/2}$ or $z_\alpha$ for a HT	$z = \frac{\hat{p} - p_0}{SE_{\hat{p}}}$
$p_1 - p_2$	$\hat{p}_1 - \hat{p}_2$	$\sqrt{\frac{\hat{p}_1\hat{q}_1}{n_1} + \frac{\hat{p}_2\hat{q}_2}{n_2}}$ for a CI $\sqrt{\bar{p}\bar{q}\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}$ for a HT where $\bar{p} = \frac{X_1 + X_2}{n_1 + n_2}$ , based on the assumption that $H_0: p_1 = p_2$ is true	$z_{\alpha/2}$ for a CI $z_{\alpha/2}$ or $z_\alpha$ for a HT	$z = \frac{(\hat{p}_1 - \hat{p}_2) - 0}{SE_{\hat{p}_1 - \hat{p}_2}}$
$\mu$	$\bar{x}$	$\frac{\sigma}{\sqrt{n}}$ for a CI or HT if $\sigma$ is known $\frac{s}{\sqrt{n}}$ for a CI or a HT if $\sigma$ is unknown	$z_{\alpha/2}$ for a CI if $\sigma$ is known; $t_{\alpha/2, df}$ if $\sigma$ is unknown $z_{\alpha/2}$ or $z_\alpha$ for a HT if $\sigma$ is known; $t_{\alpha, df}$ or $t_{\alpha/2, df}$ if $\sigma$ is unknown $df = n - 1$ for $t$	$z = \frac{\bar{x} - \mu_0}{SE_{\bar{x}}}$ if $\sigma$ is known $t = \frac{\bar{x} - \mu_0}{SE_{\bar{x}}}$ if $\sigma$ is unknown
$\mu_d$	$\bar{d}$	$\frac{s_d}{\sqrt{n}}$ for a CI or a HT for matched pairs	$t_{\alpha/2, df}$ for a CI $t_{\alpha, df}$ or $t_{\alpha/2, df}$ for a HT $df = n - 1$ where $n$ is the number of pairs	$t = \frac{\bar{d} - \mu_{d_0}}{SE_{\bar{d}}}$

Population Parameter	Sample Statistic	SE	Critical Value	Test Statistic for a HT
$\mu_1 - \mu_2$	$\bar{x}_1 - \bar{x}_2$	$\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$ for a CI or HT, for independent samples, $\sigma_1^2 \neq \sigma_2^2$ $\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$ for a CI or HT, for independent samples, $\sigma_1^2, \sigma_2^2$ unknown and assume $\sigma_1^2 \neq \sigma_2^2$ $\sqrt{s_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}$ for a CI or HT for independent samples, $\sigma_1^2, \sigma_2^2$ unknown but assume $\sigma_1^2 = \sigma_2^2$ $s_p^2 = \frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1 + n_2 - 2}$	$z_{\alpha/2}$ for a CI if $\sigma$ is known; $t_{\alpha/2, df}$ if $\sigma$ is unknown	$z = \frac{(\bar{x}_1 - \bar{x}_2) - 0}{SE_{\bar{x}_1 - \bar{x}_2}}$ if $\sigma$ is known;  $t = \frac{(\bar{x}_1 - \bar{x}_2) - 0}{SE_{\bar{x}_1 - \bar{x}_2}}$ if $\sigma$ is unknown
$\beta_1$	$b_1$	$s_{b_1} = \frac{s_e}{\sqrt{SS_{xx}}}$ where $s_e = \sqrt{s_e^2} = \sqrt{\frac{SSE}{n-2}} = \sqrt{\frac{\sum_{i=1}^n (y_i - \hat{y})^2}{n-2}}$ and $SS_{xx} = \sum_{i=1}^n (x_i - \bar{x})^2$	$t_{\alpha/2, df}$ for a CI $t_{\alpha, df}$ or $t_{\alpha/2, df}$ for a HT $df = n - 2$ where $n$ is the number of paired observations in the regression	$t = \frac{b_1 - 0}{s_{b_1}}$