

We are always trying to make inference about a population parameter from a sample statistic. We have three methods for inference:

1. Simulation/randomization methods
2. Exact/probability methods
3. Distributional approximations

We have focused (perhaps too much) on the distributional approximation methods. However, all of the problems we have studied could be solved with any of the three methods.

Parameter	Statistic	Hypothesis test	Confidence interval	Conditions
$p$	$\hat{p}$	$H_0 : p = p_0, z = \frac{\hat{p}-p_0}{SE}, SE = \sqrt{\frac{p_0(1-p_0)}{n}}$ $H_A : p \neq p_0$	$\hat{p} \pm z^* SE, SE = \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$	<b>I</b> , $np > 10, n(1-p) > 10$
$p_1 - p_2$	$\hat{p}_1 - \hat{p}_2$	$H_0 : p_1 - p_2 = 0, z = \frac{\hat{p}_1 - \hat{p}_2 - 0}{SE}, SE = \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$ $H_A : p_1 - p_2 \neq 0$	$\hat{p}_1 - \hat{p}_2 \pm z^* SE, SE = \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$	<b>I</b> , <b>N</b>
$X^2$	$\hat{X}^2$	$H_0 : \text{the counts are the same}, X^2 = \sum \frac{(O_i - E_i)^2}{E_i}$ $H_A : \text{the counts are different}$		<b>I</b> , 5 successes, $df > 2$
$\mu$	$\bar{x}$	$H_0 : \mu = \mu_0, t = \frac{\bar{x} - \mu_0}{SE}, SE = \frac{s}{\sqrt{n}}$ $H_A : \mu \neq \mu_0$	$\bar{x} \pm t^* SE, SE = \frac{s}{\sqrt{n}}$	<b>I</b> , <b>N</b>
$\mu_{diff}$	$\bar{x}_{diff}$	$H_0 : \mu_{diff} = \mu_0, t = \frac{\bar{x}_{diff} - \mu_0}{SE_{diff}}, SE_{diff} = \frac{s_{diff}}{\sqrt{n}}$ $H_A : \mu_{diff} \neq \mu_0$	$\bar{x} \pm t^* SE, SE = \frac{s_{diff}}{\sqrt{n}}$	<b>I</b> , <b>N</b>
$\mu_1 - \mu_2$	$\bar{x}_1 - \bar{x}_2$	$H_0 : \mu_1 - \mu_2 = \mu_0, t = \frac{\bar{x}_1 - \bar{x}_2 - \mu_0}{SE}, SE = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$ $H_A : \mu_1 - \mu_2 \neq \mu_0$	$\bar{x}_1 - \bar{x}_2 \pm t^* SE, SE = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$	<b>I</b> (between and within), <b>N</b>
$\mu_1, \mu_2, \dots, \mu_n$	$\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n$	$H_0 : \mu_1 = \mu_2 = \dots = \mu_n, F = \frac{MSG}{MSE}$ $H_A : \text{at least one of the } \mu_i \text{ is different}$		<b>I</b> , <b>N</b> , <b>E</b>
$\beta_i$	$b_i$	$H_0 : \beta_i = 0, t = \frac{b_i - 0}{SE}$ $H_A : \beta_i \neq 0$	$b_1 \pm t^* SE$	<b>L</b> , <b>I</b> , <b>N</b> , <b>E</b>

Where

- **Linearity**
- **Independence**
- **Normality**
- **Equality of variance**