

## Formulas

### Confidence Intervals

$$\sigma \text{ Known: } \mu_1 - \mu_2 = \bar{x}_1 - \bar{x}_2 \pm z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

$$\sigma \text{ Unknown: } \mu_1 - \mu_2 = \bar{x}_1 - \bar{x}_2 \pm t_{n_1+n_2-2, \alpha/2} \sqrt{S_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}$$

$$\text{where: } S_p^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2}$$

$$\text{Proportions: } p_1 - p_2 = \hat{p}_1 - \hat{p}_2 \pm z_{\alpha/2} \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$$

### Hypothesis Testing

1. Select the correct formula ( $\sigma$  known,  $\sigma$  Unknown, or Proportions).
2. Calculate the Statistic.
3. Find the Critical Value (use  $\alpha$  for one tail and  $\alpha/2$  for two tails).
4. If  $Stat \geq C.V.$  reject the null hypothesis (yes).

If  $Stat < C.V.$  cannot reject the null hypothesis (no).

$$\sigma \text{ Known: } z = \frac{\bar{x}_1 - \bar{x}_2 - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

$$\sigma \text{ Unknown: } t_{n_1+n_2-2} = \frac{\bar{x}_1 - \bar{x}_2 - (\mu_1 - \mu_2)}{\sqrt{S_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

$$\text{Proportions: } z = \frac{\hat{p}_1 - \hat{p}_2 - (p_1 - p_2)}{\sqrt{\hat{p}(1-\hat{p}) \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

p-value for  $z$ : for one tail  $0.5 - F(z)$ . For two tails  $2[0.5 - F(z)]$ .

## One-way ANOVA

$k$  columns,  $n_1, \dots, n_k$  rows in each column,  $x_{ij}$  value in column  $i$  row  $j$ ,  $\bar{x}_i$  mean of column  $i$ ,  $\bar{\bar{x}}$  grand mean,  $n = n_1 + \dots + n_k$ .

$$SST = \sum_{i=1}^k n_i (\bar{x}_i - \bar{\bar{x}})^2$$

$$SSE = \sum_{i=1}^k \sum_{j=1}^{n_i} (x_{ij} - \bar{x}_i)^2$$

$$MST = \frac{SST}{k-1}$$

$$MSE = \frac{SSE}{n-k}$$

$$F_{k-1, n-k} = \frac{MST}{MSE}.$$

## Two-way ANOVA

$k$  columns,  $b$  rows,  $x_{ij}$  value in column  $i$  row  $j$ ,  $\bar{x}_i^c$  mean of column  $i$ ,  $\bar{x}_j^r$  mean of row  $j$ ,  $\bar{\bar{x}}$  grand mean .

$$SST = b \sum_{i=1}^k (\bar{x}_i^c - \bar{\bar{x}})^2$$

$$SSE = \sum_{i=1}^k \sum_{j=1}^b (x_{ij} - \bar{x}_i^c - \bar{x}_j^r + \bar{\bar{x}})^2$$

$$MST = \frac{SST}{k-1}$$

$$MSE = \frac{SSE}{kb-k-b+1}$$

$$F_{k-1, kb-k-b+1} = \frac{MST}{MSE}.$$