

# Formula Sheet

# Math 222

## Probability rules

$$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$
$$P(A \text{ and } B) = P(A)P(B|A)$$

## Discrete expected value & variance

$$E(X) = \mu = \sum p_k x_k$$
$$\text{Var}(X) = \sigma^2 = \sum p_k (x_k - \mu)^2$$

## Expected value & variance rules

$$E(X + Y) = E(X) + E(Y)$$
$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)^*$$
$$E(cX) = cE(X)$$
$$\text{Var}(cX) = c^2 \text{Var}(X)$$

\* Only if  $X$  and  $Y$  are independent.

## General form of a confidence interval

estimate  $\pm$  margin of error, where  
margin of error = (critical value)  $\times$   $SE_{\text{estimate}}$

## Standard errors\*

$$SE_{\bar{x}} = \frac{s}{\sqrt{n}}$$
$$SE_{\bar{x}_1 - \bar{x}_2} = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$
$$SE_{\hat{p}} = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$$
$$SE_{\hat{p}_1 - \hat{p}_2} = \sqrt{\frac{\hat{p}_1(1 - \hat{p}_1)}{n_1} + \frac{\hat{p}_2(1 - \hat{p}_2)}{n_2}}$$

\* When testing a null hypothesis about proportions, it is better to use the hypothesized population proportion  $p_0$  rather than  $\hat{p}$  in the formula for standard error for one-sample tests, and it is better to use the pooled proportion  $\hat{p}$  rather than either individual sample proportion  $\hat{p}_1$  or  $\hat{p}_2$  in the formula for standard error in two-sample tests.

## Inference about regression

$$SE_{b_1} = \frac{s}{s_x \sqrt{n-1}} = \frac{r\sqrt{n-2}}{1-r^2}$$
$$SE_{\mu_y} = s \sqrt{\frac{1}{n} + \frac{(x^* - \bar{x})^2}{s_x^2(n-1)}} = \sqrt{\frac{s^2}{n} + (x^* - \bar{x})^2 SE_{b_1}^2}$$
$$SE_{\hat{y}} = s \sqrt{1 + \frac{1}{n} + \frac{(x^* - \bar{x})^2}{s_x^2(n-1)}} = \sqrt{s^2 + \frac{s^2}{n} + (x^* - \bar{x})^2 SE_{b_1}^2}$$

where  $s = \sqrt{MSE}$  is the standard error of the residuals. Use  $df = n - 2$  for simple linear regression. Recall that  $r^2 = \frac{SSM}{SST}$  and  $MST = s_y^2$ . With that you can fill in the whole ANOVA table.

**Inference in ANOVA** Use the pooled standard deviation  $s_p = \sqrt{MSE}$  as the best guess for  $\sigma$ . Recall also that  $SSG = \sum_{i=1}^I N_i (\bar{x}_i - \bar{x})^2$  and  $SSE = \sum_{i=1}^I (N_i - 1) s_i^2$ .

## General form of a test statistic

$$\frac{\text{statistic} - \text{hypothesized valued}}{SE_{\text{statistic}}}$$

## Least squares regression line

$$\hat{y} = b_0 + b_1 x, \text{ where } b_1 = r \frac{s_y}{s_x} \text{ and } b_0 = \bar{y} - b_1 \bar{x}$$

## Chi-squared formulas

$$\chi^2 = \sum \frac{(\text{observed count} - \text{expected count})^2}{\text{expected count}}$$

## Expected counts for 2-way tables

$$E_{ij} = \frac{\text{row total} \times \text{column total}}{\text{table total}}$$