

# Formula Sheet – Exam 4

## Chapter 9: Hypothesis Testing

### *Regions of rejection*

This is the same for means and proportions, just replace  $\mu$  with  $p$

Hypothesis for comparing means	$H_0 : \mu_1 - \mu_2 = 0$ $H_A : \mu_1 - \mu_2 \neq 0$	$H_0 : \mu_1 - \mu_2 \geq 0$ $H_A : \mu_1 - \mu_2 < 0$	$H_0 : \mu_1 - \mu_2 \leq 0$ $H_A : \mu_1 - \mu_2 > 0$
Region for Rejection (large samples)	$z < -z_{\alpha/2}$ or $z > z_{\alpha/2}$	$z < -z_{\alpha}$	$z > z_{\alpha}$
Region for Rejection (small samples)	$t < -t_{\alpha/2}$ or $t > t_{\alpha/2}$	$t < -t_{\alpha}$	$t > t_{\alpha}$

### ■ Difference in Population Means

$$z = \frac{(\bar{x}_1 - \bar{x}_2)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

### ■ Difference in Population Proportions

$$z = \frac{(\bar{p}_1 - \bar{p}_2)}{\sqrt{\bar{p}(1-\bar{p})\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} \quad \text{where } \bar{p} = \frac{n_1\bar{p}_1 + n_2\bar{p}_2}{n_1 + n_2}$$

### Differences in Small Samples using the t-statistic

### ■ Difference in population mean

$$t = \frac{(\bar{x}_1 - \bar{x}_2)}{\sqrt{s^2\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} \quad \text{where } s^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}$$

## Chapter 12: Regression Formula

### Regression Model

$$y_i = \beta_0 + \beta_1 x_i + \varepsilon_i \quad \text{where } \varepsilon_i \sim N(0, \sigma) \text{ and independent}$$

### Sum of Squares

$$SS_{xx} = \sum_{i=1}^n (x_i - \bar{x})^2 \quad SS_{xy} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) \quad SS_{yy} = \sum_{i=1}^n (y_i - \bar{y})^2$$

### Ordinary Least Squares Estimators

$$\hat{\beta}_1 = \frac{SS_{xy}}{SS_{xx}} \quad \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x$$

$$y_i = \hat{\beta}_0 + \hat{\beta}_1 x_i + e_i$$

### Sum of Square Errors, Variance and Standard Error

$$SSE = \sum (e_i)^2 = \sum (y_i - \bar{y})^2 = SS_{YY} - \hat{\beta}_1 SS_{XY}$$

$$s^2 = \frac{SSE}{n-2}$$

$$s = \sqrt{\frac{SSE}{n-2}}$$

### Sampling Distribution of Estimated Slope

$$\sigma_{\hat{\beta}_1} = \frac{\sigma}{\sqrt{SS_{XX}}} \quad (\text{population})$$

$$s_{\hat{\beta}_1} = \frac{s}{\sqrt{SS_{XX}}} \quad (\text{estimated value})$$

$$t = \frac{\hat{\beta}_1}{s_{\hat{\beta}_1}} \quad (\text{has a } t\text{-distribution with } n-2 \text{ degrees of freedom})$$

$$\hat{\beta}_1 \pm t_{\alpha/2} s_{\hat{\beta}_1} \quad (1-\alpha) \text{ confidence interval for } \beta_1$$

### Correlation and Fit

$$R = \frac{SS_{XY}}{\sqrt{SS_{XX} SS_{YY}}} \quad (\text{Correlation Coefficient})$$

$$R^2 = 1 - \frac{SSE}{SS_{YY}} \quad (\text{Coefficient of Determination})$$

### Estimation and prediction confidence intervals

A  $(1-\alpha)$  confidence interval for the mean value of  $y$  at  $x = x_t$  is

$$\hat{\beta}_0 + \hat{\beta}_1 x_t \pm t_{\alpha/2} s \sqrt{\frac{1}{n} + \frac{(x_t - \bar{x})^2}{SS_{XX}}}$$

A  $(1-\alpha)$  confidence interval for the value of  $y$  at  $x = x_t$  is

$$\hat{\beta}_0 + \hat{\beta}_1 x_t \pm t_{\alpha/2} s \sqrt{1 + \frac{1}{n} + \frac{(x_t - \bar{x})^2}{SS_{XX}}}$$