# STAT 105 Exam II Reference Sheet

#### **Numeric Summaries**

mean 
$$\bar{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$$

population variance 
$$\sigma^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2$$

population standard deviation 
$$\sigma = \sqrt{\frac{1}{n} \sum_{i=1}^{n} (x_i - \bar{x})^2}$$

sample variance 
$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

sample standard deviation 
$$s = \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x})^2}$$

## Linear Relationships

Form  $y \approx \beta_0 + \beta_1 x$ 

Fitted linear relationship  $\hat{y} = b_0 + b_1 x$ 

Least squares estimates  $b_1 = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^{n} (x_i - \bar{x})^2}$ 

$$b_1 = \frac{\sum_{i=1}^{n} x_i y_i - n\bar{x}\bar{y}}{\sum_{i=1}^{n} x_i^2 - n\bar{x}^2}$$

$$b_0 = \bar{y} - b_1 \bar{x}$$

Residuals  $e_i = y_i - \hat{y}_i$ 

sample correlation coeffecient  $r = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^{n} (x_i - \bar{x})^2 \sum_{i=1}^{n} (x_i - \bar{x})^2}}$ 

$$r = \frac{\sum_{i=1}^{n} x_i y_i - n\bar{x}\bar{y}}{\sqrt{\left(\sum_{i=1}^{n} x_i^2 - n\bar{x}^2\right)\left(\sum_{i=1}^{n} y_i^2 - n\bar{y}^2\right)}}$$

coeffecient of determination  $R^2 = (r)^2$ 

$$\frac{\sum_{i=1}^{n} (y_i - \bar{y})^2 - \sum_{i=1}^{n} (y_i - \hat{y}_i)^2}{\sum_{i=1}^{n} (y_i - \bar{y})^2}$$

### Multivariate Relationships

Form  $y \approx \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \ldots + \beta_k x_k$ 

Fitted relationship  $\hat{y} \approx b_0 + b_1 x_1 + b_2 x_2 + \ldots + b_k x_k$ 

Residuals  $e_i = y_i - \hat{y}_i$ 

Sums of Squares  $SSTO = \sum_{i=1}^{n} (y_i - \bar{y})^2$ 

$$SSE = \sum_{i=1}^{n} (y_i - \hat{y}_i)^2$$

$$SSR = SSTO - SSE = \sum_{i=1}^{n} (\hat{y}_i - \bar{y})^2$$

coeffecient of determination  $R^2 = \frac{\sum_{i=1}^n (y_i - \bar{y})^2 - \sum_{i=1}^n (y_i - \hat{y}_i)^2}{\sum_{i=1}^n (y_i - \bar{y}_i)^2}$ 

$$R^2 = \frac{\text{SSTO} - \text{SSE}}{\text{SSTO}}$$

$$R^2 = \frac{\text{SSR}}{\text{SSTO}}$$

$$\frac{\sum_{i=1}^{n} (y_i - \bar{y})^2 - \sum_{i=1}^{n} (y_i - \hat{y}_i)^2}{\sum_{i=1}^{n} (y_i - \bar{y})^2}$$

#### **Functions**

**Quantile Function** Q(p) For a dataset consisting of n values that are ordered so that  $x_1 \le x_2 \le \ldots \le x_n$  and value p where  $0 \le p \le 1$ , let  $i = \lfloor n \cdot p + 0.5 \rfloor$ . Then the quantile function at p is:

$$Q(p) = \begin{cases} x_i & \lfloor n \cdot p + 0.5 \rfloor = n \cdot p + 0.5 \\ x_i + (n \cdot p - i + 0.5)(x_{i+1} - x_i) & \lfloor n \cdot p + 0.5 \rfloor \neq n \cdot p + 0.5 \end{cases}$$

### Discrete Random Variables

Probability function  $P[X = x] = f_X(x)$ 

Cumulative probability function  $P[X \le x] = F_X(x)$ 

Expected Value  $\mu = E(X) = \sum_{x} x f_X(x)$ 

Variance  $\sigma^2 = Var(X) = \sum_{x} (x - \mu)^2 f_X(x)$ 

Standard Deviation  $\sigma = \sqrt{Var(X)}$ 

# Continuous Random Variables

Probability density function  $P[a \le X \le b] = \int_a^b f_X(x) dx$ 

Cumulative probability function  $P[X \leq x] = F_X(x) = \int_{-\infty}^x f_X(t) dt$ 

Expected Value  $\mu = E(X) = \int_{-\infty}^{\infty} x f_X(x) dx$ 

Variance  $\sigma^2 = Var(X) = \int_{-\infty}^{\infty} (x-\mu)^2 f_X(x) dx$ 

Standard Deviation  $\sigma = \sqrt{Var(X)}$