Sample Mean

$$\bar{x} = \frac{\sum x_i}{n} \tag{3.1}$$

Population Mean

$$\mu = \frac{\sum x_i}{N} \tag{3.2}$$

Weighted Mean

$$\bar{x} = \frac{\sum w_i x_i}{\sum w_i} \tag{3.3}$$

Geometric Mean

$$\bar{x}_g = \sqrt[n]{(x_1)(x_2)\cdots(x_n)} = [(x_1)(x_2)\cdots(x_n)]^{1/n}$$
 (3.4)

Location of the pth Percentile

$$L_p = \frac{p}{100}(n+1) \tag{3.5}$$

Interquartile Range

$$IQR = Q_3 - Q_1 (3.6)$$

Population Variance

$$\sigma^2 = \frac{\Sigma (x_i - \mu)^2}{N} \tag{3.7}$$

Sample Variance

$$s^2 = \frac{\sum (x_i - \bar{x})^2}{n - 1}$$
 (3.8)

Standard Deviation

Sample standard deviation =
$$s = \sqrt{s^2}$$
 (3.9)

Population standard deviation =
$$\sigma = \sqrt{\sigma^2}$$
 (3.10)

Coefficient of Variation

$$\left(\frac{\text{Standard deviation}}{\text{Mean}} \times 100\right)\%$$
 (3.11)

z-Score

$$z_i = \frac{x_i - \bar{x}}{s} \tag{3.12}$$

Sample Covariance

$$s_{xy} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{n - 1}$$
 (3.13)

Population Covariance

$$\sigma_{xy} = \frac{\sum (x_i - \mu_x)(y_i - \mu_y)}{N}$$
 (3.14)

Pearson Product Moment Correlation Coefficient: Sample Data

$$r_{xy} = \frac{s_{xy}}{s_x s_y} \tag{3.15}$$

Pearson Product Moment Correlation Coefficient: Population Data

$$\rho_{xy} = \frac{\sigma_{xy}}{\sigma_x \sigma_y} \tag{3.16}$$

Counting Rule for Combinations

$$C_n^N = \binom{N}{n} = \frac{N!}{n!(N-n)!} \tag{4.1}$$

Counting Rule for Permutations

$$P_n^N = n! \binom{N}{n} = \frac{N!}{(N-n)!}$$
 (4.2)

Computing Probability Using the Complement

$$P(A) = 1 - P(A^c) {(4.5)}$$

Addition Law

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$
 (4.6)

Conditional Probability

$$P(A \mid B) = \frac{P(A \cap B)}{P(B)} \tag{4.7}$$

$$P(B \mid A) = \frac{P(A \cap B)}{P(A)}$$
(4.8)

Multiplication Law

$$P(A \cap B) = P(B)P(A \mid B)$$
 (4.11)

$$P(A \cap B) = P(A)P(B \mid A)$$
 (4.12)

Multiplication Law for Independent Events

$$P(A \cap B) = P(A)P(B) \tag{4.13}$$

Bayes' Theorem

$$P(A_i \mid B) = \frac{P(A_i)P(B \mid A_i)}{P(A_1)P(B \mid A_1) + P(A_2)P(B \mid A_2) + \dots + P(A_n)P(B \mid A_n)}$$
 (4.19)

Discrete Uniform Probability Function

$$f(x) = 1/n \tag{5.3}$$

Expected Value of a Discrete Random Variable

$$E(x) = \mu = \sum x f(x) \tag{5.4}$$

Variance of a Discrete Random Variable

$$Var(x) = \sigma^2 = \sum (x - \mu)^2 f(x)$$
 (5.5)

Covariance of Random Variables x and y

$$\sigma_{xy} = [Var(x+y) - Var(x) - Var(y)]/2$$
 (5.6)

Correlation between Random Variables x and y

$$\rho_{xy} = \frac{\sigma_{xy}}{\sigma_x \sigma_y} \tag{5.7}$$

Expected Value of a Linear Combination of Random Variables x and y

$$E(ax + by) = aE(x) + bE(y)$$
(5.8)

Variance of a Linear Combination of Two Random Variables

$$Var(ax + by) = a^{2}Var(x) + b^{2}Var(y) + 2ab\sigma_{xy}$$
(5.9)

where σ_{xy} is the covariance of x and y

Number of Experimental Outcomes Providing Exactly x Successes in n Trials

$$\binom{n}{x} = \frac{n!}{x!(n-x)!} \tag{5.10}$$

Binomial Probability Function

$$f(x) = \binom{n}{x} p^{x} (1-p)^{(n-x)}$$
 (5.12)

Expected Value for the Binomial Distribution

$$E(x) = \mu = np \tag{5.13}$$

Variance for the Binomial Distribution

$$Var(x) = \sigma^2 = np(1-p)$$
 (5.14)

Poisson Probability Function

$$f(x) = \frac{\mu^x e^{-\mu}}{x!}$$
 (5.15)

Hypergeometric Probability Function

$$f(x) = \frac{\binom{r}{x} \binom{N-r}{n-x}}{\binom{N}{n}}$$
 (5.16)

Expected Value for the Hypergeometric Distribution

$$E(x) = \mu = n \left(\frac{r}{N}\right) \tag{5.17}$$

Variance for the Hypergeometric Distribution

$$Var(x) = \sigma^2 = n \left(\frac{r}{N}\right) \left(1 - \frac{r}{N}\right) \left(\frac{N - n}{N - 1}\right)$$
 (5.18)

Uniform Probability Density Function

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{for } a \le x \le b \\ 0 & \text{elsewhere} \end{cases}$$
 (6.1)

Normal Probability Density Function

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2}(\frac{x-\mu}{\sigma})^2}$$
 (6.2)

Converting to the Standard Normal Random Variable

$$z = \frac{x - \mu}{\sigma} \tag{6.3}$$

Exponential Probability Density Function

$$f(x) = \frac{1}{\mu} e^{-x/\mu}$$
 for $x \ge 0$ (6.4)

Exponential Distribution: Cumulative Probabilities

$$P(x \le x_0) = 1 - e^{-x_0/\mu} \tag{6.5}$$

Expected Value of \bar{x}

$$E(\bar{x}) = \mu \tag{7.1}$$

Standard Deviation of \bar{x} (Standard Error)

Finite Population

Infinite Population

$$\sigma_{\bar{x}} = \sqrt{\frac{N-n}{N-1}} \left(\frac{\sigma}{\sqrt{n}} \right)$$

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$
(7.2)

Expected Value of \bar{p}

$$E(\overline{p}) = p \tag{7.4}$$

Standard Deviation of \bar{p} (Standard Error)

Finite Population

Infinite Population

$$\sigma_{\overline{p}} = \sqrt{\frac{N-n}{N-1}} \sqrt{\frac{p(1-p)}{n}} \qquad \sigma_{\overline{p}} = \sqrt{\frac{p(1-p)}{n}}$$
 (7.5)

Interval Estimate of a Population Mean: σ Known

$$\bar{x} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \tag{8.1}$$

Interval Estimate of a Population Mean: σ Unknown

$$\bar{x} \pm t_{\alpha/2} \frac{s}{\sqrt{n}} \tag{8.2}$$

Sample Size for an Interval Estimate of a Population Mean

$$n = \frac{(z_{\alpha/2})^2 \sigma^2}{E^2}$$
 (8.3)

Interval Estimate of a Population Proportion

$$\overline{p} \pm z_{\alpha/2} \sqrt{\frac{\overline{p}(1-\overline{p})}{n}}$$
 (8.6)

Sample Size for an Interval Estimate of a Population Proportion

$$n = \frac{(z_{\alpha/2})^2 p^* (1 - p^*)}{E^2}$$
 (8.7)