

## FREQUENTLY USED FORMULAS

$n$  = sample size     $N$  = population size     $f$  = frequency

### Chapter 2

Class Width =  $\frac{\text{high} - \text{low}}{\text{number classes}}$  (increase to next integer)

Class Midpoint =  $\frac{\text{upper limit} + \text{lower limit}}{2}$

Lower boundary = lower boundary of previous class  
+ class width

### Chapter 3

Sample mean  $\bar{x} = \frac{\sum x}{n}$

Population mean  $\mu = \frac{\sum x}{N}$

Weighted average =  $\frac{\sum xw}{\sum w}$

Range = largest data value – smallest data value

Sample standard deviation  $s = \sqrt{\frac{\sum(x - \bar{x})^2}{n - 1}}$

Computation formula  $s = \sqrt{\frac{\sum x^2 - (\sum x)^2/n}{n - 1}}$

Population standard deviation  $\sigma = \sqrt{\frac{\sum(x - \mu)^2}{N}}$

Sample variance  $s^2$

Population variance  $\sigma^2$

Sample Coefficient of Variation  $CV = \frac{s}{\bar{x}} \cdot 100$

Sample mean for grouped data  $\bar{x} = \frac{\sum xf}{n}$

Sample standard deviation for grouped data

$$s = \sqrt{\frac{\sum(x - \bar{x})^2 f}{n - 1}} = \sqrt{\frac{\sum x^2 f - (\sum xf)^2/n}{n - 1}}$$

### Chapter 4

Probability of the complement of event  $A$   
 $P(\text{not } A) = 1 - P(A)$

Multiplication rule for independent events  
 $P(A \text{ and } B) = P(A) \cdot P(B)$

General multiplication rules  
 $P(A \text{ and } B) = P(A) \cdot P(B, \text{ given } A)$   
 $P(A \text{ and } B) = P(B) \cdot P(A, \text{ given } B)$

Addition rule for mutually exclusive events  
 $P(A \text{ or } B) = P(A) + P(B)$

General addition rule  
 $P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$

Permutation rule  $P_{n,r} = \frac{n!}{(n - r)!}$

Combination rule  $C_{n,r} = \frac{n!}{r!(n - r)!}$

### Chapter 5

Mean of a discrete probability distribution  $\mu = \sum xP(x)$

Standard deviation of a discrete probability distribution

$$\sigma = \sqrt{\sum(x - \mu)^2 P(x)}$$

Given  $L = a + bx$

$$\mu_L = a + b\mu$$

$$\sigma_L = |b|\sigma$$

Given  $W = ax_1 + bx_2$  ( $x_1$  and  $x_2$  independent)

$$\mu_W = a\mu_1 + b\mu_2$$

$$\sigma_W = \sqrt{a^2\sigma_1^2 + b^2\sigma_2^2}$$

For Binomial Distributions

$r$  = number of successes;  $p$  = probability of success;  
 $q = 1 - p$

Binomial probability distribution  $P(r) = C_{n,r} p^r q^{n-r}$

Mean  $\mu = np$

Standard deviation  $\sigma = \sqrt{npq}$

Geometric Probability Distribution

$n$  = number of trial on which first success occurs

$$P(n) = p(1 - p)^{n-1}$$

Poisson Probability Distribution

$r$  = number of successes

$\lambda$  = mean number of successes over given interval

$$P(r) = \frac{e^{-\lambda} \lambda^r}{r!}$$

### Chapter 6

Raw score  $x = z\sigma + \mu$

Standard score  $z = \frac{x - \mu}{\sigma}$

### Chapter 7

Mean of  $\bar{x}$  distribution  $\mu_{\bar{x}} = \mu$

Standard deviation of  $\bar{x}$  distribution  $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$

Standard score for  $\bar{x}$   $z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$

Mean of  $\hat{p}$  distribution  $\mu_{\hat{p}} = p$

Standard deviation of  $\hat{p}$  distribution  $\sigma_{\hat{p}} = \sqrt{\frac{pq}{n}}$ ;  $q = 1 - p$

## Chapter 8

### Confidence Interval

for  $\mu$

$$\bar{x} - E < \mu < \bar{x} + E$$

where  $E = z_c \frac{\sigma}{\sqrt{n}}$  when  $\sigma$  is known

$$E = t_c \frac{s}{\sqrt{n}} \text{ when } \sigma \text{ is unknown}$$

with  $d.f. = n - 1$

for  $p$  ( $np > 5$  and  $n(1 - p) > 5$ )

$$\hat{p} - E < p < \hat{p} + E$$

$$\text{where } E = z_c \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$$

$$\hat{p} = \frac{r}{n}$$

for  $\mu_1 - \mu_2$  (independent samples)

$$(\bar{x}_1 - \bar{x}_2) - E < \mu_1 - \mu_2 < (\bar{x}_1 - \bar{x}_2) + E$$

where  $E = z_c \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$  when  $\sigma_1$  and  $\sigma_2$  are known

$$E = t_c \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}} \text{ when } \sigma_1 \text{ or } \sigma_2 \text{ is unknown}$$

with  $d.f. = \text{smaller of } n_1 - 1 \text{ and } n_2 - 1$

(Note: Software uses Satterthwaite's approximation for degrees of freedom  $d.f.$ )

for difference of proportions  $p_1 - p_2$

$$(\hat{p}_1 - \hat{p}_2) - E < p_1 - p_2 < (\hat{p}_1 - \hat{p}_2) + E$$

$$\text{where } E = z_c \sqrt{\frac{\hat{p}_1 \hat{q}_1}{n_1} + \frac{\hat{p}_2 \hat{q}_2}{n_2}}$$

$$\hat{p}_1 = r_1/n_1; \hat{p}_2 = r_2/n_2$$

$$\hat{q}_1 = 1 - \hat{p}_1; \hat{q}_2 = 1 - \hat{p}_2$$

### Sample Size for Estimating

$$\text{means } n = \left( \frac{z_c \sigma}{E} \right)^2$$

proportions

$$n = p(1 - p) \left( \frac{z_c}{E} \right)^2 \text{ with preliminary estimate for } p$$

$$n = \frac{1}{4} \left( \frac{z_c}{E} \right)^2 \text{ without preliminary estimate for } p$$

## Chapter 9

### Sample Test Statistics for Tests of Hypotheses

$$\text{for } \mu \text{ (}\sigma \text{ known)} \quad z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$$

$$\text{for } \mu \text{ (}\sigma \text{ unknown)} \quad t = \frac{\bar{x} - \mu}{s/\sqrt{n}}; d.f. = n - 1$$

$$\text{for } p \text{ (} np > 5 \text{ and } nq > 5) \quad z = \frac{\hat{p} - p}{\sqrt{pq/n}}$$

$$\text{where } q = 1 - p; \hat{p} = r/n$$

$$\text{for paired differences } d \quad t = \frac{\bar{d} - \mu_d}{s_d/\sqrt{n}}; d.f. = n - 1$$

for difference of means,  $\sigma_1$  and  $\sigma_2$  known

$$z = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

for difference of means,  $\sigma_1$  or  $\sigma_2$  unknown

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

$$d.f. = \text{smaller of } n_1 - 1 \text{ and } n_2 - 1$$

(Note: Software uses Satterthwaite's approximation for degrees of freedom  $d.f.$ )

for difference of proportions

$$z = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\frac{\bar{p}\bar{q}}{n_1} + \frac{\bar{p}\bar{q}}{n_2}}}$$

$$\text{where } \bar{p} = \frac{r_1 + r_2}{n_1 + n_2} \text{ and } \bar{q} = 1 - \bar{p}$$

$$\hat{p}_1 = r_1/n_1; \hat{p}_2 = r_2/n_2$$

## Chapter 10

### Regression and Correlation

Pearson product moment correlation coefficient

$$r = \frac{n\sum xy - (\sum x)(\sum y)}{\sqrt{n\sum x^2 - (\sum x)^2} \sqrt{n\sum y^2 - (\sum y)^2}}$$

Least-squares line  $\hat{y} = a + bx$

$$\text{where } b = \frac{n\sum xy - (\sum x)(\sum y)}{n\sum x^2 - (\sum x)^2}$$

$$a = \bar{y} - b\bar{x}$$

Coefficient of determination =  $r^2$

Sample test statistic for  $r$

$$t = \frac{r\sqrt{n-2}}{\sqrt{1-r^2}} \text{ with } d.f. = n - 2$$

Standard error of estimate  $S_e = \sqrt{\frac{\Sigma y^2 - a\Sigma y - b\Sigma xy}{n-2}}$

Confidence interval for  $y$

$$\hat{y} - E < y < \hat{y} + E$$

$$\text{where } E = t_c S_e \sqrt{1 + \frac{1}{n} + \frac{n(x - \bar{x})^2}{n\Sigma x^2 - (\Sigma x)^2}}$$

with  $d.f. = n - 2$

Sample test statistic for slope  $b$

$$t = \frac{b}{S_e} \sqrt{\Sigma x^2 - \frac{1}{n}(\Sigma x)^2} \text{ with } d.f. = n - 2$$

Confidence interval for  $\beta$

$$b - E < \beta < b + E$$

$$\text{where } E = \frac{t_c S_e}{\sqrt{\Sigma x^2 - \frac{1}{n}(\Sigma x)^2}} \text{ with } d.f. = n - 2$$

## Chapter 11

$$\chi^2 = \sum \frac{(O - E)^2}{E} \text{ where } E = \frac{(\text{row total})(\text{column total})}{\text{sample size}}$$

Tests of Independence  $d.f. = (R - 1)(C - 1)$

Goodness of fit  $d.f. = (\text{number of categories}) - 1$

Confidence Interval for  $\sigma^2$ ;  $d.f. = n - 1$

$$\frac{(n-1)s^2}{\chi^2_U} < \sigma^2 < \frac{(n-1)s^2}{\chi^2_L}$$

Sample test statistic for  $\sigma^2$

$$\chi^2 = \frac{(n-1)s^2}{\sigma^2} \text{ with } d.f. = n - 1$$

Testing Two Variances

$$\text{Sample test statistic } F = \frac{s_1^2}{s_2^2}$$

$$\text{where } s_1^2 \geq s_2^2$$

$$d.f._N = n_1 - 1; d.f._D = n_2 - 1$$

ANOVA

$k$  = number of groups;  $N$  = total sample size

$$SS_{TOT} = \Sigma x_{TOT}^2 - \frac{(\Sigma x_{TOT})^2}{N}$$

$$SS_{BET} = \sum_{\text{all groups}} \left( \frac{(\Sigma x_i)^2}{n_i} \right) - \frac{(\Sigma x_{TOT})^2}{N}$$

$$SS_W = \sum_{\text{all groups}} \left( \Sigma x_i^2 - \frac{(\Sigma x_i)^2}{n_i} \right)$$

$$SS_{TOT} = SS_{BET} + SS_W$$

$$MS_{BET} = \frac{SS_{BET}}{d.f._{BET}} \text{ where } d.f._{BET} = k - 1$$

$$MS_W = \frac{SS_W}{d.f._W} \text{ where } d.f._W = N - k$$

$$F = \frac{MS_{BET}}{MS_W} \text{ where } d.f. \text{ numerator} = d.f._{BET} = k - 1;$$

$$d.f. \text{ denominator} = d.f._W = N - k$$

Two-Way ANOVA

$r$  = number of rows;  $c$  = number of columns

$$\text{Row factor } F: \frac{MS \text{ row factor}}{MS \text{ error}}$$

$$\text{Column factor } F: \frac{MS \text{ column factor}}{MS \text{ error}}$$

$$\text{Interaction } F: \frac{MS \text{ interaction}}{MS \text{ error}}$$

with degrees of freedom for

$$\text{row factor} = r - 1 \quad \text{interaction} = (r - 1)(c - 1)$$

$$\text{column factor} = c - 1 \quad \text{error} = rc(n - 1)$$

## Chapter 12

Sample test statistic for  $x$  = proportion of plus signs to all signs ( $n \geq 12$ )

$$z = \frac{x - 0.5}{\sqrt{0.25/n}}$$

Sample test statistic for  $R$  = sum of ranks

$$z = \frac{R - \mu_R}{\sigma_R} \text{ where } \mu_R = \frac{n_1(n_1 + n_2 + 1)}{2} \text{ and}$$

$$\sigma_R = \sqrt{\frac{n_1 n_2 (n_1 + n_2 + 1)}{12}}$$

Spearman rank correlation coefficient

$$r_s = 1 - \frac{6\Sigma d^2}{n(n^2 - 1)} \text{ where } d = x - y$$

Sample test statistic for runs test

$R$  = number of runs in sequence