

## Formula Sheet (Page 1)

1. Addition Rule (events not mutually exclusive):  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$

2. Addition Rule (events not mutually exclusive):

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

3. Addition Rule (mutually exclusive events):  $P(A \cup B) = P(A) + P(B)$

4. Multiplication Rule (dependent events):  $P(A \cap B) = P(A)P(B|A)$

5. Multiplication Rule (independent events):  $P(A \cap B) = P(A)P(B)$

6. Conditional Probability:  $P(B|A) = \frac{P(A \cap B)}{P(A)}$

7. Binomial:  $f(x) = \binom{n}{x} p^x (1-p)^{n-x}$ ,  $x = 0, 1, \dots, n$ ;  $\binom{n}{x} = \frac{n!}{x!(n-x)!}$ ;  $\mu = np$ ,  $\sigma^2 = np(1-p)$

8. Negative Binomial:  $f(x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$ ,  $x = r, r+1, r+2, \dots$ ;  $\mu = \frac{r}{p}$ ,  $\sigma^2 = \frac{r(1-p)}{p^2}$

9. Hypergeometric Distribution:  $f(x) = \frac{\binom{K}{x} \binom{N-K}{n-x}}{\binom{N}{n}}$ ,  $\mu = np$ ,  $\sigma^2 = np(1-p) \left( \frac{N-n}{N-1} \right)$

10. Poisson Distribution:  $f(x) = \frac{e^{-\lambda} \lambda^x}{x!}$ ,  $x = 0, 1, 2, \dots$ ;  $\mu = \sigma^2 = \lambda$

11. Normal Distribution:  $f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2}$ ,  $-\infty < x < \infty$

12. Exponential Distribution:  $f(x) = \lambda e^{-\lambda x}$ ,  $0 < x < \infty$ ;  $\mu = \frac{1}{\lambda}$ ,  $\sigma^2 = \frac{1}{\lambda^2}$

13. Marginal Distributions:  $f_X(x) = \int f_{XY}(x, y) dy$ ,  $f_Y(y) = \int f_{XY}(x, y) dx$

14. Correlation:  $\rho_{XY} = \frac{\text{cov}(X, Y)}{\sqrt{V(X)V(Y)}} = \frac{\sigma_{XY}}{\sigma_X \sigma_Y}$

15. Transformation to standard normal:  $z = \frac{X - \mu}{\sigma}$

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

17. Outliers:  $Q_1 - 1.5 \text{ IQR}$ ,  $Q_3 + 1.5 \text{ IQR}$

18. Normal probability plot:  $\Phi(z_j) = \frac{j-0.5}{n}$ ,  $j = 1, 2, \dots, n$

19. Central Limit Theorem formula:  $z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$

20.  $z$  confidence interval for the mean:  $\bar{x} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$

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**21.**  $t$  confidence interval for the mean:  $\bar{x} \pm t_{\alpha/2, n-1} \frac{s}{\sqrt{n}}$

**22.** Confidence interval for a proportion:  $\hat{p} \pm z_{\alpha/2} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$

**23.**  $z$  test for a mean:  $Z_0 = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}}$ ;  $\beta = \Phi\left(z_{\alpha/2} - \frac{\delta\sqrt{n}}{\sigma}\right) - \Phi\left(-z_{\alpha/2} - \frac{\delta\sqrt{n}}{\sigma}\right)$

**24.**  $t$  test for a mean:  $T_0 = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$  **25.**  $z$  test for proportions:  $Z_0 = \frac{X - np_0}{\sqrt{np_0(1-p_0)}} = \frac{\hat{p} - p_0}{\sqrt{p_0(1-p_0)/n}}$ ;  
 $\beta = \Phi\left(\frac{p_0 - p + z_{\alpha/2} \sqrt{p_0(1-p_0)/n}}{\sqrt{p(1-p)/n}}\right) - \Phi\left(\frac{p_0 - p - z_{\alpha/2} \sqrt{p_0(1-p_0)/n}}{\sqrt{p(1-p)/n}}\right)$

Confidence interval for a difference in means:

**26.** Variances equal:  $\bar{x}_1 - \bar{x}_2 \pm t_{\alpha/2, n_1+n_2-2} s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$ ,  $s_p^2 = \frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1+n_2-2}$

**27.** Variances unequal:  $\bar{x}_1 - \bar{x}_2 \pm t_{\alpha/2, \nu} \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$ ,  $\nu = \left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)^2 \left/ \left(\frac{(s_1^2/n_1)^2}{n_1-1} + \frac{(s_2^2/n_2)^2}{n_2-1}\right)\right.$

**28.**  $t$  test for comparing two means (variances equal):  $t_{n_1+n_2-2} = \frac{\bar{x}_1 - \bar{x}_2}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$

**29.**  $t$  test for comparing two means (variances unequal):  $t_\nu = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{(s_1^2/n_1) + (s_2^2/n_2)}}$ ,

**30.** Single variable Least Squares Regression line:  $\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$ ,  $\hat{\beta}_1 = \frac{s_{xy}}{s_{xx}}$ , where

$$s_{xx} = \sum_{i=1}^n x_i^2 - n\bar{x}^2 \text{ and } s_{xy} = \sum_{i=1}^n x_i y_i - n\bar{x}\bar{y}$$

**31.**  $t$ -test for single variable regression:  $t_{n-2} = \frac{\hat{\beta}_1}{\sqrt{\hat{\sigma}^2/s_{xx}}}$ , where  $\hat{\sigma}^2 = \frac{SS_E}{n-2}$

**32.** Residual sum of squares:  $SS_E = \sum_{i=1}^n (y_i - \hat{y}_i)^2$  **33.** Regression sum of squares:  $SS_R = \hat{\beta}_1 s_{xy}$

**34.** Total sum of squares:  $SS_T = \sum_{i=1}^n y_i^2 - n\bar{y}^2$

**35.** Prediction Interval:  $\hat{y}_0 \pm t_{\alpha/2, n-2} \sqrt{\hat{\sigma}^2 \left(1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{s_{xx}}\right)}$

**36.** Sample correlation coefficient:  $r = \frac{s_{xy}}{\sqrt{s_{xx}s_{yy}}}$  **37.** Coefficient of determination:  $R^2 = \frac{SS_R}{SS_T}$

**38.** Total sum of squares:  $SS_T = \sum_{i=1}^a \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{..})^2 = \sum_{i=1}^a \sum_{j=1}^{n_i} y_{ij}^2 - \frac{y_{..}^2}{N}$  (d.f. =  $N - 1$ )

**39.** Error sum of squares:  $SS_E = \sum_{i=1}^a \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.})^2 = \sum_{i=1}^a (n_i - 1)s_i^2$  (d.f. =  $N - a$ )

**40.** Treatment sum of squares:  $SS_{\text{Treatments}} = \sum_{i=1}^a n_i (\bar{y}_{i.} - \bar{y}_{..})^2 = \sum_{i=1}^a \frac{y_{i.}^2}{n_i} - \frac{y_{..}^2}{N}$  (d.f. =  $a - 1$ )

**41.** Fisher's LSD Test:  $LSD = t_{\alpha/2, N-a} \sqrt{MSE \left(\frac{1}{n_i} + \frac{1}{n_j}\right)}$ , **42.** Fisher's CI:  $\bar{y}_{i.} - \bar{y}_{j.} \pm LSD$