

Formulas and Tables by Mario F. Triola

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Ch. 3: Descriptive Statistics	Ch. 7: Confidence Intervals (one population)
$\bar{x} = \frac{\sum x}{n} \quad \text{Mean}$ $\bar{x} = \frac{\sum (f \cdot x)}{\sum f} \quad \text{Mean (frequency table)}$ $s = \sqrt{\frac{\sum (x - \bar{x})^2}{n - 1}} \quad \text{Standard deviation}$ $s = \sqrt{\frac{n(\sum x^2) - (\sum x)^2}{n(n - 1)}} \quad \text{Standard deviation (shortcut)}$ $s = \sqrt{\frac{n[\sum (f \cdot x^2)] - [\sum (f \cdot x)]^2}{n(n - 1)}} \quad \text{Standard deviation (frequency table)}$ <p>variance = s^2</p>	$\hat{p} - E < p < \hat{p} + E \quad \text{Proportion}$ <p style="text-align: center;">where $E = z_{\alpha/2} \sqrt{\frac{\hat{p}\hat{q}}{n}}$</p> <hr/> $\bar{x} - E < \mu < \bar{x} + E \quad \text{Mean}$ <p style="text-align: center;">where $E = t_{\alpha/2} \frac{s}{\sqrt{n}}$ (σ unknown)</p> <p style="text-align: center;">or $E = z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$ (σ known)</p> <hr/> $\frac{(n - 1)s^2}{\chi^2_R} < \sigma^2 < \frac{(n - 1)s^2}{\chi^2_L} \quad \text{Variance}$
Ch. 4: Probability	Ch. 7: Sample Size Determination
$P(A \text{ or } B) = P(A) + P(B)$ if A, B are mutually exclusive $P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$ if A, B are not mutually exclusive $P(A \text{ and } B) = P(A) \cdot P(B)$ if A, B are independent $P(A \text{ and } B) = P(A) \cdot P(B A)$ if A, B are dependent $P(\bar{A}) = 1 - P(A)$ Rule of complements ${}_nP_r = \frac{n!}{(n - r)!}$ Permutations (no elements alike) $\frac{n!}{n_1! n_2! \dots n_k!}$ Permutations (n_1 alike, . . .) ${}_nC_r = \frac{n!}{(n - r)! r!}$ Combinations	$n = \frac{[z_{\alpha/2}]^2 0.25}{E^2} \quad \text{Proportion}$ $n = \frac{[z_{\alpha/2}]^2 \hat{p}\hat{q}}{E^2} \quad \text{Proportion } (\hat{p} \text{ and } \hat{q} \text{ are known})$ $n = \left[\frac{z_{\alpha/2} \sigma}{E} \right]^2 \quad \text{Mean}$
Ch. 5: Probability Distributions	Ch. 8: Test Statistics (one population)
$\mu = \sum [x \cdot P(x)]$ Mean (prob. dist.) $\sigma = \sqrt{\sum [x^2 \cdot P(x)] - \mu^2}$ Standard deviation (prob. dist.) $P(x) = \frac{n!}{(n - x)! x!} \cdot p^x \cdot q^{n-x}$ Binomial probability $\mu = n \cdot p$ Mean (binomial) $\sigma^2 = n \cdot p \cdot q$ Variance (binomial) $\sigma = \sqrt{n \cdot p \cdot q}$ Standard deviation (binomial) $P(x) = \frac{\mu^x \cdot e^{-\mu}}{x!}$ Poisson distribution where $e = 2.71828$	$z = \frac{\hat{p} - p}{\sqrt{\frac{pq}{n}}} \quad \text{Proportion—one population}$ $t = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} \quad \text{Mean—one population } (\sigma \text{ unknown})$ $z = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \quad \text{Mean—one population } (\sigma \text{ known})$ $\chi^2 = \frac{(n - 1)s^2}{\sigma^2} \quad \text{Standard deviation or variance—one population}$
Ch. 6: Normal Distribution	
$z = \frac{x - \mu}{\sigma} \text{ or } \frac{x - \bar{x}}{s} \quad \text{Standard score}$ $\mu_{\bar{x}} = \mu$ Central limit theorem $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$ Central limit theorem (Standard error)	