

Formula Sheet

Sample Statistics:

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

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

Linear Transformations/Combinations:

Mean: $E(aX + b) = aE(X) + b, E(\sum_{i=1}^n a_i X_i) = \sum_{i=1}^n a_i E(X_i)$

Variance: $V(aX + b) = a^2 V(X), V(\sum_{i=1}^n a_i X_i) = \sum_{i=1}^n \sum_{j=1}^n a_i a_j \text{cov}(X_i, X_j)$

Discrete:

Binomial:

$$p(x; n, p) = \binom{n}{x} p^x (1-p)^{n-x}, x \in \{0, 1, \dots, n\}$$

$$E(X) = np, V(X) = np(1-p)$$

Poisson:

$$p(x; \lambda) = \frac{e^{-\lambda} \lambda^x}{x!}, x \in \{0, 1, 2, 3, \dots\}$$

$$E(X) = V(X) = \lambda$$

Negative Binomial:

$$p(x; r, p) = \binom{x-1}{r-1} p^r (1-p)^{x-r}, x \in \{r, r+1, r+2, \dots\}$$

$$E(X) = \frac{r}{p}, V(X) = \frac{r(1-p)}{p^2}$$

Continuous:

Uniform:

$$f(x; a, b) = \frac{1}{b-a}, x \in [a, b]$$

$$E(X) = \frac{a+b}{2}, V(X) = \frac{(b-a)^2}{12}$$

Normal:

$$f(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(x-\mu)^2}, x \in (-\infty, \infty)$$

$$E(X) = \mu, V(X) = \sigma^2$$

Central Limit Theorem: $n > 30$

$$\bar{X} \overset{\text{approx}}{\sim} \mathcal{N}\left(\mu, \frac{\sigma^2}{n}\right)$$

One Sample Inference: Use $\mathcal{N}(0, 1)$ or t quantiles where necessary, as well as σ^2 or s^2 where necessary, depending on whether the variance is known or unknown.

For a random variable Y , let Y_c denote the c^{th} percentile, i.e. $P(Y \leq Y_c) = c$.

Two sided $(1 - \alpha)100\%$ C.I. for μ :

$$\begin{aligned}\bar{x} \pm z_{1-\alpha/2} \frac{\sigma}{\sqrt{n}} & \quad \sigma^2 \text{ known} \\ \bar{x} \pm t_{(1-\alpha/2, n-1)} \frac{s}{\sqrt{n}} & \quad \sigma^2 \text{ unknown}\end{aligned}$$

Sample size for C.I. width $\leq k$:

$$n \geq \left[\left(2z_{1-\alpha/2} \frac{\sigma}{k} \right)^2 \right]$$

Test Statistic for μ :

$$\begin{aligned}T.S. &= \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \stackrel{H_0}{\sim} \mathcal{N}(0, 1) & \sigma^2 \text{ known} \\ T.S. &= \frac{\bar{x} - \mu_0}{s/\sqrt{n}} \stackrel{H_0}{\sim} t_{n-1} & \sigma^2 \text{ unknown}\end{aligned}$$

Agresti-Coull CI for a proportion: if $\tilde{n} = n + 4$, $\tilde{p} = (x + 2)/\tilde{n}$, then

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

Two sided $(1 - \alpha)100\%$ C.I. (variance):

$$\left(\frac{(n-1)s^2}{\chi^2_{(1-\alpha/2, n-1)}}, \frac{(n-1)s^2}{\chi^2_{(\alpha/2, n-1)}} \right)$$

Test statistic (variance):

$$T.S. = \frac{(n-1)s^2}{\sigma_0^2} \stackrel{H_0}{\sim} \chi^2_{n-1}$$

Two sample inference

Two sided C.I. (means):

$$v = \frac{\left(\frac{s_x^2}{n_x} + \frac{s_y^2}{n_y}\right)^2}{\frac{(s_x^2/n_x)^2}{n_x-1} + \frac{(s_y^2/n_y)^2}{n_y-1}}$$
$$s_p^2 = \frac{(n_x - 1)s_x^2 + (n_y - 1)s_y^2}{n_x + n_y - 2}$$

$$\bar{x} - \bar{y} \pm z_{1-\alpha/2} \sqrt{\frac{\sigma_x^2}{n_x} + \frac{\sigma_y^2}{n_y}} \quad \sigma_x^2, \sigma_y^2 \text{ both known}$$

$$\bar{x} - \bar{y} \pm t_{1-\alpha/2, n_x+n_y-2} \sqrt{\frac{s_p^2}{n_x} + \frac{s_p^2}{n_y}} \quad \sigma_x^2, \sigma_y^2 \text{ unknown, assume } \sigma_x^2 = \sigma_y^2$$

$$\bar{x} - \bar{y} \pm t_{1-\alpha/2, v} \sqrt{\frac{s_x^2}{n_x} + \frac{s_y^2}{n_y}} \quad \sigma_x^2, \sigma_y^2 \text{ unknown, do not assume } \sigma_x^2 = \sigma_y^2$$

Two sided C.I. (proportions):

$$\tilde{p}_x - \tilde{p}_y \pm z_{1-\alpha/2} \sqrt{\frac{\tilde{p}_x(1-\tilde{p}_x)}{\tilde{n}_x} + \frac{\tilde{p}_y(1-\tilde{p}_y)}{\tilde{n}_y}} \text{ where } \tilde{n}_x = n_x + 2 \text{ and } \tilde{p}_x = \frac{x+1}{\tilde{n}_x}$$

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