

## Handout

### Confidence Intervals:

Confidence interval,  
Normal distribution

If parameter  $\theta$  has an unbiased, Normally distributed estimator  $\hat{\theta}$ , then

$$\hat{\theta} \pm z_{\alpha/2} \cdot \sigma(\hat{\theta}) = \left[ \hat{\theta} - z_{\alpha/2} \cdot \sigma(\hat{\theta}), \hat{\theta} + z_{\alpha/2} \cdot \sigma(\hat{\theta}) \right]$$

is a  $(1 - \alpha)100\%$  confidence interval for  $\theta$ .

If the distribution of  $\hat{\theta}$  is *approximately* Normal, we get an *approximately*  $(1 - \alpha)100\%$  confidence interval.

If we do not know the population std. dev. but we know the  $n$  is large, then  $\sigma(\theta)$  can be replaced by  $s(\theta)$

Confidence interval  
for the difference of means;  
known standard deviations

$$\bar{X} - \bar{Y} \pm z_{\alpha/2} \sqrt{\frac{\sigma_X^2}{n} + \frac{\sigma_Y^2}{m}}$$

Confidence interval  
for a population proportion

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

Confidence interval  
for the difference  
of proportions

$$\hat{p}_1 - \hat{p}_2 \pm z_{\alpha/2} \sqrt{\frac{\hat{p}_1(1 - \hat{p}_1)}{n_1} + \frac{\hat{p}_2(1 - \hat{p}_2)}{n_2}}$$

Common Z values:

$$\begin{aligned} z_{0.10} &= 1.282, & z_{0.05} &= 1.645, & z_{0.025} &= 1.960 \\ z_{0.01} &= 2.326, & z_{0.005} &= 2.576. \end{aligned}$$

Can also obtained from Z-table or from T-table with  $v = \infty$

If  $n$  is small,

Confidence interval  
for the mean;  
 $\sigma$  is unknown

$$\bar{X} \pm t_{\alpha/2} \frac{s}{\sqrt{n}}$$

where  $t_{\alpha/2}$  is a critical value from T-distribution with  $n - 1$  degrees of freedom

Confidence interval for the difference of means; equal, unknown standard deviations

$$\bar{X} - \bar{Y} \pm t_{\alpha/2} s_p \sqrt{\frac{1}{n} + \frac{1}{m}}$$

where  $s_p$  is the *pooled standard deviation*, a root of the pooled variance in (9.11) and  $t_{\alpha/2}$  is a critical value from T-distribution with  $(n + m - 2)$  degrees of freedom

Pooled Std. Deviation:

$$s_p^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2 + \sum_{i=1}^m (Y_i - \bar{Y})^2}{n + m - 2} = \frac{(n-1)s_X^2 + (m-1)s_Y^2}{n + m - 2}. \quad (9.11)$$

Confidence interval for the difference of means; unequal, unknown standard deviations

$$\bar{X} - \bar{Y} \pm t_{\alpha/2} \sqrt{\frac{s_X^2}{n} + \frac{s_Y^2}{m}}$$

where  $t_{\alpha/2}$  is a critical value from T-distribution with  $\nu$  degrees of freedom given by formula (9.12)

Statterthwaite approximation of degrees of freedom (9.12):

$$\nu = \frac{\left( \frac{s_X^2}{n} + \frac{s_Y^2}{m} \right)^2}{\frac{s_X^4}{n^2(n-1)} + \frac{s_Y^4}{m^2(m-1)}}$$

**Hypothesis Testing (Z-tests):**

$$\text{Right Tailed Test (H}_A: \theta > \theta_0\text{): } \begin{cases} \text{reject } H_0 & \text{if } Z \geq z_\alpha \\ \text{accept } H_0 & \text{if } Z < z_\alpha \end{cases}$$

$$\text{Left Tailed Test (H}_A: \theta < \theta_0\text{): } \begin{cases} \text{reject } H_0 & \text{if } Z \leq -z_\alpha \\ \text{accept } H_0 & \text{if } Z > -z_\alpha \end{cases}$$

$$\text{Two Tailed Test (H}_A: \theta \neq \theta_0\text{): } \begin{cases} \text{reject } H_0 & \text{if } |Z| \geq z_{\alpha/2} \\ \text{accept } H_0 & \text{if } |Z| < z_{\alpha/2} \end{cases}$$

**Hypothesis Testing (t-tests):**

For a **right-tail alternative**,

$$\begin{cases} \text{reject } H_0 & \text{if } t \geq t_\alpha \\ \text{accept } H_0 & \text{if } t < t_\alpha \end{cases}$$

For a **left-tail alternative**,

$$\begin{cases} \text{reject } H_0 & \text{if } t \leq -t_\alpha \\ \text{accept } H_0 & \text{if } t > -t_\alpha \end{cases}$$

For a **two-sided alternative**,

$$\begin{cases} \text{reject } H_0 & \text{if } |t| \geq t_{\alpha/2} \\ \text{accept } H_0 & \text{if } |t| < t_{\alpha/2} \end{cases}$$

### Summary of Z - Tests

Null hypothesis	Parameter, estimator	If $H_0$ is true:		Test statistic
		$\mathbf{E}(\hat{\theta})$	$\text{Var}(\hat{\theta})$	
$H_0$	$\theta, \hat{\theta}$			$Z = \frac{\hat{\theta} - \theta_0}{\sqrt{\text{Var}(\hat{\theta})}}$
One-sample Z-tests for means and proportions, based on a sample of size $n$				
$\mu = \mu_0$	$\mu, \bar{X}$	$\mu_0$	$\frac{\sigma^2}{n}$	$\frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$
$p = p_0$	$p, \hat{p}$	$p_0$	$\frac{p_0(1-p_0)}{n}$	$\frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$
Two-sample Z-tests comparing means and proportions of two populations, based on independent samples of size $n$ and $m$				
$\mu_X - \mu_Y = D$	$\mu_X - \mu_Y, \bar{X} - \bar{Y}$	$D$	$\frac{\sigma_X^2}{n} + \frac{\sigma_Y^2}{m}$	$\frac{\bar{X} - \bar{Y} - D}{\sqrt{\frac{\sigma_X^2}{n} + \frac{\sigma_Y^2}{m}}}$
$p_1 - p_2 = D$	$p_1 - p_2, \hat{p}_1 - \hat{p}_2$	$D$	$\frac{p_1(1-p_1)}{n} + \frac{p_2(1-p_2)}{m}$	$\frac{\hat{p}_1 - \hat{p}_2 - D}{\sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n} + \frac{\hat{p}_2(1-\hat{p}_2)}{m}}}$
$p_1 = p_2$	$p_1 - p_2, \hat{p}_1 - \hat{p}_2$	0	$p(1-p) \left( \frac{1}{n} + \frac{1}{m} \right),$ where $p = p_1 = p_2$	$\frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\hat{p}(1-\hat{p}) \left( \frac{1}{n} + \frac{1}{m} \right)}}$ where $\hat{p} = \frac{n\hat{p}_1 + m\hat{p}_2}{n+m}$

### Summary of t-tests

Hypothesis $H_0$	Conditions	Test statistic $t$	Degrees of freedom
$\mu = \mu_0$	Sample size $n$ ; unknown $\sigma$	$t = \frac{\bar{X} - \mu_0}{s/\sqrt{n}}$	$n - 1$
$\mu_X - \mu_Y = D$	Sample sizes $n, m$ ; unknown but equal standard deviations, $\sigma_X = \sigma_Y$	$t = \frac{\bar{X} - \bar{Y} - D}{s_p \sqrt{\frac{1}{n} + \frac{1}{m}}}$	$n + m - 2$
$\mu_X - \mu_Y = D$	Sample sizes $n, m$ ; unknown, unequal standard deviations, $\sigma_X \neq \sigma_Y$	$t = \frac{\bar{X} - \bar{Y} - D}{\sqrt{\frac{s_X^2}{n} + \frac{s_Y^2}{m}}}$	Satterthwaite approximation, formula (9.12)

**P Values (Reject  $H_0$  if  $P < 0.01$ , Accept  $H_0$  if  $P > 0.1$ , Not enough evidence otherwise):**

P values for Z tests:

Hypothesis $H_0$	Alternative $H_A$	P-value	Computation
$\theta = \theta_0$	right-tail $\theta > \theta_0$	$P\{Z \geq Z_{\text{obs}}\}$	$1 - \Phi(Z_{\text{obs}})$
	left-tail $\theta < \theta_0$	$P\{Z \leq Z_{\text{obs}}\}$	$\Phi(Z_{\text{obs}})$
	two-sided $\theta \neq \theta_0$	$P\{ Z  \geq  Z_{\text{obs}} \}$	$2(1 - \Phi( Z_{\text{obs}} ))$

P values for t tests:

Hypothesis $H_0$	Alternative $H_A$	P-value	Computation
$\theta = \theta_0$	right-tail $\theta > \theta_0$	$P\{t \geq t_{\text{obs}}\}$	$1 - F_{\nu}(t_{\text{obs}})$
	left-tail $\theta < \theta_0$	$P\{t \leq t_{\text{obs}}\}$	$F_{\nu}(t_{\text{obs}})$
	two-sided $\theta \neq \theta_0$	$P\{ t  \geq  t_{\text{obs}} \}$	$2(1 - F_{\nu}( t_{\text{obs}} ))$

**Confidence Intervals (variance):**

**Confidence interval  
for the variance**

$$\left[ \frac{(n-1)s^2}{\chi_{\alpha/2}^2}, \frac{(n-1)s^2}{\chi_{1-\alpha/2}^2} \right]$$

**Confidence interval  
for the standard  
deviation**

$$\left[ \sqrt{\frac{(n-1)s^2}{\chi_{\alpha/2}^2}}, \sqrt{\frac{(n-1)s^2}{\chi_{1-\alpha/2}^2}} \right]$$

**Hypothesis tests for variance (can also be used for Std Dev by conv question to variance):**

Null Hypothesis	Alternative Hypothesis	Test statistic	Rejection region	P-value
$\sigma^2 = \sigma_0^2$	$\sigma^2 > \sigma_0^2$	$\frac{(n-1)s^2}{\sigma_0^2}$	$\chi_{\text{obs}}^2 > \chi_{\alpha}^2$	$P\{\chi^2 \geq \chi_{\text{obs}}^2\}$
	$\sigma^2 < \sigma_0^2$		$\chi_{\text{obs}}^2 < \chi_{\alpha}^2$	$P\{\chi^2 \leq \chi_{\text{obs}}^2\}$
	$\sigma^2 \neq \sigma_0^2$		$\chi_{\text{obs}}^2 \geq \chi_{\alpha/2}^2$ or $\chi_{\text{obs}}^2 \leq \chi_{1-\alpha/2}^2$	$2 \min \left( P\{\chi^2 \geq \chi_{\text{obs}}^2\}, P\{\chi^2 \leq \chi_{\text{obs}}^2\} \right)$

**Testing ratio of Variances (can also be used for Std Dev by conv question to variance):**

Null Hypothesis $H_0 : \frac{\sigma_X^2}{\sigma_Y^2} = \theta_0$		Test statistic $F_{\text{obs}} = \frac{s_X^2}{s_Y^2} / \theta_0$
Alternative Hypothesis	Rejection region	P-value Use $F(n-1, m-1)$ distribution
$\frac{\sigma_X^2}{\sigma_Y^2} > \theta_0$	$F_{\text{obs}} \geq F_{\alpha}(n-1, m-1)$	$\mathbf{P}\{F \geq F_{\text{obs}}\}$
$\frac{\sigma_X^2}{\sigma_Y^2} < \theta_0$	$F_{\text{obs}} \leq F_{\alpha}(n-1, m-1)$	$\mathbf{P}\{F \leq F_{\text{obs}}\}$
$\frac{\sigma_X^2}{\sigma_Y^2} \neq \theta_0$	$F_{\text{obs}} \geq F_{\alpha/2}(n-1, m-1)$ or $F_{\text{obs}} < 1/F_{\alpha/2}(m-1, n-1)$	$2 \min(\mathbf{P}\{F \geq F_{\text{obs}}\}, \mathbf{P}\{F \leq F_{\text{obs}}\})$