

## INFERENCE GUIDE

### CONFIDENCE INTERVALS

Use **confidence intervals** to **estimate** a parameter with a particular **confidence level, C**.

**IDENTIFY:** Identify the parameter and the confidence level.

**CHOOSE:** Choose and name the appropriate interval.

**CHECK:** Check that conditions for the procedure are met.

**CALCULATE:**

**CI: point estimate  $\pm$  critical value  $\times$  SE of estimate**

$df$  = (if applicable)  
( \_\_\_\_\_, \_\_\_\_\_ )

**CONCLUDE:**

We are C% confident that the true [parameter] is between \_\_\_\_\_ and \_\_\_\_\_. (Put the parameter in *context*.)

We have evidence that [...], because [...]. OR  
We do not have evidence that [...], because [...].

When the parameter is: **a single proportion  $p$**

**CHOOSE:** **1-Proportion Z-Interval** to estimate  $p$ , or  
**1-Proportion Z-Test** to test  $H_0: p = p_0$ .

**CHECK:**

- Data come from a random sample or process.
- for CI:  $n\hat{p} \geq 10$  and  $n(1 - \hat{p}) \geq 10$ .
- for Test:  $np_0 \geq 10$  and  $n(1 - p_0) \geq 10$ .

**CALCULATE:** (1-PropZInt or 1-PropZTest)

**point estimate:** sample proportion  $\hat{p}$

**SE of estimate:** for CI, use  $\sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$ ; for Test, use  $\sqrt{\frac{p_0(1-p_0)}{n}}$

When the parameter is: **a difference of proportions  $p_1 - p_2$**

**CHOOSE:** **2-Proportion Z-Interval** to estimate  $p_1 - p_2$ , or  
**2-Proportion Z-Test** to test  $H_0: p_1 = p_2$ .

**CHECK:**

- Data come from 2 independent random samples or 2 randomly assigned treatments.
  - $n_1\hat{p}_1 \geq 10$ ,  $n_1(1 - \hat{p}_1) \geq 10$ ,  
 $n_2\hat{p}_2 \geq 10$ ,  $n_2(1 - \hat{p}_2) \geq 10$ .
- Note: use  $\hat{p}_c$ , the pooled proportion, in place of  $\hat{p}_1$  and  $\hat{p}_2$  when checking condition for the 2-Proportion Z-Test

**CALCULATE:** (2-PropZInt or 2-PropZTest)

**point estimate:** difference of sample proportions  $\hat{p}_1 - \hat{p}_2$

**SE of estimate:**

CI, use  $\sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$ ; Test, use  $\sqrt{\hat{p}_c(1-\hat{p}_c) \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}$

### HYPOTHESIS TESTS

Use **hypothesis tests** to **test**  $H_0$  versus  $H_A$  at a particular **significance level,  $\alpha$** .

**IDENTIFY:** Identify the hypotheses and the significance level.

**CHOOSE:** Choose and name the appropriate test.

**CHECK:** Check that conditions for the procedure are met.

**CALCULATE:**

**standardized test statistic** =  $\frac{\text{point estimate} - \text{null value}}{\text{SE of estimate}}$

$df$  = (if applicable)  
p-value =

**CONCLUDE:**

p-value  $< \alpha$ , so we reject  $H_0$ .

We have evidence that  $[H_A]$ . (Put  $H_A$  in *context*.)

OR

p-value  $> \alpha$ , so we do NOT reject  $H_0$ .

We do NOT have evidence that  $[H_A]$ . (Put  $H_A$  in *context*.)

When the parameter is: **a single mean  $\mu$**

**CHOOSE:** **1-Sample T-Interval** to estimate  $\mu$ , or  
**1-Sample T-Test** to test  $H_0: \mu = \mu_0$ .

**CHECK:**

- Data come from a random sample or process.
- $n \geq 30$ , OR population known to be nearly normal, OR population could to be nearly normal because data has no excessive skew or outliers (draw graph).

**CALCULATE:** (TInterval or T-Test)

**point estimate:** sample mean  $\bar{x}$

**SE of estimate:**  $\frac{s}{\sqrt{n}}$

$df = n - 1$

When the parameter is: **a difference of means  $\mu_1 - \mu_2$**

**CHOOSE:** **2-Sample T-Interval** to estimate  $\mu_1 - \mu_2$ , or  
**2-Sample T-Test** to test  $H_0: \mu_1 = \mu_2$ .

**CHECK:**

- Data come from 2 independent random samples or 2 randomly assigned treatments.
- $n_1 \geq 30$  and  $n_2 \geq 30$ , OR *both* populations known to be nearly normal, OR *both* populations could be nearly normal because both data sets have no excessive skew or outliers (draw 2 graphs).

**CALCULATE:** (2-SampTInt or 2-SampTTest)

**point estimate:** difference of sample means  $\bar{x}_1 - \bar{x}_2$

**SE of estimate:**  $\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$

$df$ : use technology

When the parameter is: **a mean of differences  $\mu_{diff}$**

CHOOSE: **1-Sample T-Interval** to estimate  $\mu_{diff}$ , or  
**1-Sample T-Test** to test  $H_0: \mu_{diff} = 0$ .

CHECK:

- There is paired data from a random sample or matched pairs experiment.
- $n_{diff} \geq 30$ , OR population of differences known to be nearly normal, OR population of differences could be nearly normal because observed differences have no excessive skew or outliers (draw graph of *differences*).

CALCULATE: (TInterval or T-Test)

point estimate: mean of sample difference  $\bar{x}_{diff}$

SE of estimate:  $\frac{s_{diff}}{\sqrt{n_{diff}}}$

$df = n_{diff} - 1$

When the parameter is: **the slope  $\beta$  of a regression line**

CHOOSE: **T-Interval for the slope** to estimate  $\beta$ , or  
**T-Test for the slope** to test  $H_0: \beta = 0$ .

CHECK:

- There is  $(x, y)$  data from a random sample or experiment.
- The residual plot shows no pattern making a linear model reasonable. (More specifically, the residuals should be independent, nearly normal, and have constant standard deviation.)

CALCULATE: (LinRegTInt or LinRegTTest)

point estimate: sample slope  $b$

SE of estimate: SE of slope (from computer output)

$df = n - 2$

The  $\chi^2$  tests for categorical variables: **chi-square statistic** =  $\sum \frac{(\text{observed} - \text{expected})^2}{\text{expected}}$

When comparing the distribution of **one categorical variable to a fixed/specified population distribution**

CHOOSE:  **$\chi^2$  Goodness of Fit Test**

CHECK:

- Data come from a random sample or process.
- All expected counts  $\geq 5$ . (To calculate expected counts for each category, multiply the sample size by the expected proportion under  $H_0$ .)

CALCULATE: ( $\chi^2$ GOF-Test)

$\chi^2 =$

$df = \# \text{ of categories} - 1$

When comparing the distribution of **a categorical variable across 2 or more populations/treatments**

CHOOSE:  **$\chi^2$  Test for Homogeneity**

CHECK:

- Data come from 2 or more independent random samples or 2 or more randomly assigned treatments.
- All expected counts  $\geq 5$ . (Calculate expected counts and verify this to be true.)

CALCULATE: ( $\chi^2$ -Test, then 2ND MATRIX, EDIT, 2: [B] to find expected counts)

$\chi^2 =$

$df = (\# \text{ of rows} - 1)(\# \text{ of cols} - 1)$

When looking for **association or dependence between two categorical variables**

CHOOSE:  **$\chi^2$  Test for Independence**

CHECK:

- Data come from a random sample or process.
- All expected counts  $\geq 5$ . (Calculate expected counts and verify this to be true.)

CALCULATE: ( $\chi^2$ -Test, then 2ND MATRIX, EDIT, 2: [B] to find expected counts)

$\chi^2 =$

$df = (\# \text{ of rows} - 1)(\# \text{ of cols} - 1)$