

## Finding Confidence interval for the population proportion:

What we need:

Random probability sample

Conditions for Binomial:

- Fix number of trials
- Trials are independent
- Two outcomes: Success / Failure
- At least 10 'Yes' and 10 'No'

$n$  - sample size;  $p^{\wedge}$  - probability of success (or sample proportion of success);  $q^{\wedge}$  - probability of failure (or sample proportion of failure)

$n * p^{\wedge} \geq 10$ ;  $n * q^{\wedge} \geq 10$

**Margin of Error (E)** - the maximum possible difference between  $p$  (population proportion) and  $p^{\wedge}$  (sample proportion for success; is a point estimate for  $p$ ).

$$E = Z_{\alpha/2} * \sqrt{p^{\wedge} * q^{\wedge} / n} \quad (\text{the square root included } p^{\wedge}, q^{\wedge}, \text{ and } n)$$

If  $p^{\wedge}$  and  $q^{\wedge}$  are not accurate,  $p^{\wedge} * q^{\wedge} = 0.5 * 0.5$  and after that we use the above formula

$Z_{\alpha/2}$  - Critical value and we use Confidence level to find it:

for 90% Confidence level  $\rightarrow Z_{\alpha/2} = 1,645$ ;

for 95% Confidence level  $\rightarrow Z_{\alpha/2} = 1,96$ ;

for 98% Confidence level  $\rightarrow Z_{\alpha/2} = 2,326$ ;

for 99% Confidence level  $\rightarrow Z_{\alpha/2} = 2,576$ ;

## Construct Confidence interval for the population proportion:

$$p^{\wedge} - E < p < p^{\wedge} + E$$

## How to find required sample size for the Survey with given E:

a) If we know  $p^{\wedge}$  and  $q^{\wedge}$ :

$$n = (Z_{\alpha/2} * Z_{\alpha/2}) * p^{\wedge} * q^{\wedge} / (E * E)$$

b) If we don't know  $p^{\wedge}$  and  $q^{\wedge}$ :

$$n = (Z_{\alpha/2} * Z_{\alpha/2}) * 0.25 / (E * E)$$

## From the given Confidence interval find $p^{\wedge}$ and E:

$$p^{\wedge} = (\text{upper boundary} + \text{lower boundary}) / 2$$

$$E = (\text{upper boundary} - \text{lower boundary}) / 2$$

**How to estimate the difference between two populations proportion:**

$$p_1 - p_2 \pm Z_{\alpha/2} * \sqrt{(p^{\wedge}_1 * q^{\wedge}_1 / n_1) + (p^{\wedge}_2 * q^{\wedge}_2 / n_2)} \text{ (the square root end after } n_2)$$

**Finding Confidence interval for the population mean, when population standard deviation is known ( $\sigma$ ):**

What we need:

Random probability sample

Population standard deviation is known ( $\sigma$ )

$n > 30$  **or** Population is normally distributed

$\bar{X}$  is a sample mean (point estimate) for population mean ( $\mu$ )

**Margin of Error (E)** - the maximum possible difference between  $p$  (population proportion) and  $p^{\wedge}$  (sample proportion for success; is a point estimate for  $p$ ).

$$E = Z_{\alpha/2} * \sigma / \sqrt{n}$$

$\sigma / \sqrt{n}$  - standard error

$Z_{\alpha/2}$  - Critical value and we use Confidence level to find it:

for 90% Confidence level  $\rightarrow Z_{\alpha/2} = 1,645$ ;

for 95% Confidence level  $\rightarrow Z_{\alpha/2} = 1,96$ ;

for 98% Confidence level  $\rightarrow Z_{\alpha/2} = 2,326$ ;

for 99% Confidence level  $\rightarrow Z_{\alpha/2} = 2,576$ ;

**Construct Confidence interval for the population mean:**

$$\bar{X} - E < \mu < \bar{X} + E$$

**From the given Confidence interval find  $\bar{X}$  and E:**

$$\bar{X} = (\text{upper boundary} + \text{lower boundary}) / 2$$

$$E = (\text{upper boundary} - \text{lower boundary}) / 2$$

**How to find required sample size for the Survey with given E and  $\sigma$  :**

$$n = (Z_{\alpha/2} * Z_{\alpha/2}) * (\sigma * \sigma) / (E * E)$$

**How to estimate the difference between two population means, if we have two independent groups, and  $\sigma$  is known for each population (example- BMI between men and women Mexican-American ):**

$$\mu_1 - \mu_2 \pm Z_{\alpha/2} * \sqrt{((\sigma_1 * \sigma_1) / n_1) + ((\sigma_2 * \sigma_2) / n_2)} \text{ (the square root end after } n_2)$$

**Finding Confidence interval for the population mean, when population standard deviation is unknown (more realistic case):**

What we need:

Random probability sample

$n > 30$  **or** Population is normally distributed

$\bar{X}$  is a sample mean (point estimate) for population mean ( $\mu$ )

**If we don't know  $\sigma$ , we can't use  $Z_{\alpha/2}$ . Instead we use  $T_{\alpha/2}$  (T - score).**

**Critical values are given by  $T_{\alpha/2}$  .**

**If the sample size is big enough  $T_{\alpha/2} = Z_{\alpha/2}$**

**Steps to find T-score:**

Calculate Degrees of Freedom ( **D.F. = n - 1**), n is a sample size;

Calculate  $\alpha$  (Alpha);  $\alpha$  is a Significance Level and is a complement of **Critical level**; For example if **Critical Level** is 95%,  $\alpha = 5\%$  or 0.05)

Use statistics table to find T - score;

**Margin of Error (E)** - the maximum possible difference between p (population proportion) and  $p^{\wedge}$  (sample proportion for success; is a point estimate for p).

$$E = T_{\alpha/2} * s / \sqrt{n}$$

s - sample standard deviation

$s / \sqrt{n}$  - estimate standard error (sample standard error)

**Construct Confidence interval for the population mean:**

$$\bar{X} - E < \mu < \bar{X} + E$$

**From the given Confidence interval find  $\bar{X}$  and E:**

$$\bar{X} = (\text{upper boundary} + \text{lower boundary}) / 2$$

$$E = (\text{upper boundary} - \text{lower boundary}) / 2$$

**How to estimate the difference between two population means, if we have two independent groups (for example- BMI between men and women Mexican-American) and standard deviation for these two populations is unknown:**

a) First approach - assumption that  $(\sigma_1 * \sigma_1)$  is not equal to  $(\sigma_2 * \sigma_2)$

Degree of Freedom =  $\min(n_1 - 1; n_2 - 1)$ ; take the minimum of these two

After that find  $T_{\alpha/2}$  from the statistics table.

$$\mu_1 - \mu_2 \pm T_{\alpha/2} * \sqrt{((s_1 * s_1) / n_1) + ((s_2 * s_2) / n_2)} \text{ (the square root end after } n_2)$$

b) Second approach - assumption that  $(\sigma_1 * \sigma_1)$  is equal to  $(\sigma_2 * \sigma_2)$

Degree of Freedom =  $n_1 + n_2 - 2$

After that find  $T_{\alpha/2}$  from the statistics table

$$\mu_1 - \mu_2 \pm T_{\alpha/2} * \sqrt{(((n_1 - 1)(s_1^2 * s_1)) + ((n_2 - 1)(s_2^2 * s_2))) / (n_1 + n_2 - 2)} \text{ (the square root end after 2)}$$

$$* \sqrt{(1 / n_1) + (1 / n_2)} \text{ (the square root end after } n_2 \text{)}$$

**How to estimate the difference between two population means, if we have paired data (for example older twin education vs younger twin education)**

n - is the same for these two groups

Degree of Freedom = n - 1

After that find  $T_{\alpha/2}$  from the statistics table.

$$\mu_1 - \mu_2 \pm T_{\alpha/2} * (S_d / \sqrt{n})$$

$$S_d = S_1 - S_2$$

**Finding Confidence interval for Variance ( $\sigma^2$ ) and Standard Deviation ( $\sigma$ ). Chi-Squared Distribution. One population**

**Chi-Squared Distribution is not symmetrical, it's right-skewed.**

**Values are only positive, because the distribution has only one tail to the right.**

**If Degrees of Freedom goes up the distribution becomes more symmetrical.**

**Chi-Squared Distribution gives Critical value to the Right ( $X_R^2$ ) for current Confidence Level ( 90%, 95%, 98%, 99%).**

**Find Left and Right critical values**

n = 12, 95% Confidence Level, Degrees of Freedom = 11

right\_critical = scipy.stats.chi2.ppf(1-.025, df = 11) is  $\chi^2_R = 21.92$

left\_critical = scipy.stats.chi2.ppf(.025, df=11) is  $\chi^2_L = 3.816$

**Construct Confidence interval for the population variance:**

$$(n - 1) (s^2) / \chi^2_R < \sigma^2 < (n - 1) (s^2) / \chi^2_L$$

**Construct Confidence interval for the population standard deviation:**

$$\sqrt{(n - 1) (s^2) / \chi^2_R} < \sigma < \sqrt{(n - 1) (s^2) / \chi^2_L}$$

(the square roots end after the two Chi-squared which we found from Statistics Table or with Python)

## How to estimate the ratio between two population variances

**Find F left critical value and F right critical value**

F\_right = scipy.stats.f.ppf (q = 1 - .025, df1, df2)

F\_left = scipy.stats.f.ppf(q = .025, df1, df2)

first is the population with bigger size

**Construct confidence interval for the difference**

$$1 / F_{\text{right}} * (\text{variance}_1 / \text{variance}_2) < \text{variance}_1 / \text{variance}_2 < 1 / F_{\text{left}} *$$

$$(\text{variance}_1 / \text{variance}_2)$$