

Confidence Interval 2

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Example

Example

A study is conducted to estimate the mean BP (blood pressure) for a single target popⁿ. A sample is collected and mean, s.d. and sample size are tabulated.

Find 95% CI for mean BP for

the ~~the~~ target population. ~~Assume~~
Given: The popⁿ BP follows Normal dist".

Example

Case 1.

$$\underline{n = 3534, \bar{x} = 127.3, \text{s.d.} = 19.}$$

$n > 30$, Normal popⁿ, $1 - \alpha = .95$, $\alpha = 0.05$,

$$\alpha/2 = 0.025$$

$$95\% \text{ CI} \rightarrow \left(\bar{x} - z_{0.025} \times \frac{s}{\sqrt{n}}, \bar{x} + z_{0.025} \times \frac{s}{\sqrt{n}} \right)$$

$$\left(127.3 - 1.96 \times \frac{19}{\sqrt{3524}}, 127.3 + 1.96 \times \frac{19}{\sqrt{3524}} \right)$$
$$(126.7, 127.9)$$

Example

Case 2. $n = 10$, $\bar{x} = 121.2$, $s = 11.1$.

$n < 30$, Normal popn, $\frac{\alpha}{2} = 0.025$

$$95\% \text{ CI} : \left(\bar{x} - t_{0.025, 9} \times \frac{s}{\sqrt{n}}, \bar{x} + t_{0.025, 9} \times \frac{s}{\sqrt{n}} \right)$$

$$t_{0.025, 9} = 2.262 \quad (\text{from t-table}).$$

$$95\% \text{ CI} : \left(121.2 - 2.262 \times \frac{11.1}{\sqrt{10}}, 121.2 + 2.262 \times \frac{11.1}{\sqrt{10}} \right) \\ (113.3, 129.1)$$

Interpretation:

CI for Variance

CI for Variance

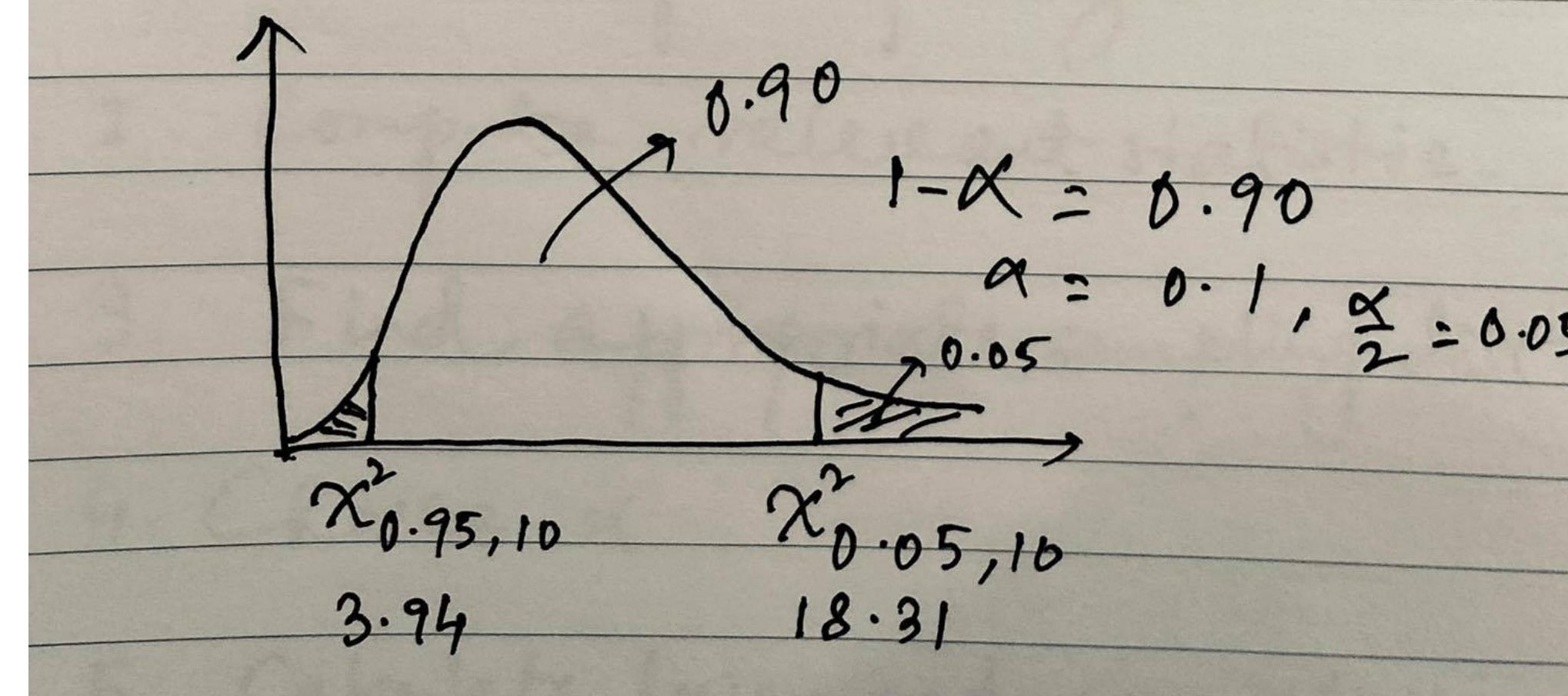
Assumption: Population is normal.

X_1, X_2, \dots, X_n are iid random sample

from $N(\mu, \sigma^2)$. Let $s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$

Then $\chi^2 = \frac{(n-1)s^2}{\sigma^2} \sim \chi^2_{(n-1)}$
Sampling distⁿ.

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



CI for Variance

Unknown parameter : σ^2

Population : Normal

Sample variance : s^2

random sample of n obs. from Normal
popn

$$100(1-\alpha)\% \text{ CI} : \frac{(n-1)s^2}{\chi^2_{\alpha/2, n-1}} \leq \sigma^2 \leq \frac{(n-1)s^2}{\chi^2_{1-\alpha/2, n-1}}$$

Algorithm for developing CI

General Algorithm for obtaining C.I.

1. Collect sample of size n
2. Compute relevant statistic.
3. Find appropriate sampling distⁿ
4. Choose α
5. Calculate lower and upper limits
from the statistical table.

CI for Difference in Means

CI for Difference in Means

Case 1: Variances Known (σ_1^2, σ_2^2).

Popⁿ 1, popⁿ 2

$x_{11}, x_{12}, \dots, x_{1n_1}$ i.i.d random sample popⁿ 1.

$x_{21}, x_{22}, \dots, x_{2n_2}$ - - - - - from popⁿ 2.

popⁿ 1 (x_1), popⁿ 2 (x_2) are independent

Both populations are normal.

Estimator of $\mu_1 - \mu_2 \rightarrow \bar{x}_1 - \bar{x}_2$

$$E(\bar{x}_1 - \bar{x}_2) = ? \quad \text{Var}(\bar{x}_1 - \bar{x}_2) = ?$$

Sampling Dist

Both populations are normal.

Estimator of $\mu_1 - \mu_2 \rightarrow \bar{x}_1 - \bar{x}_2$

$$E(\bar{x}_1 - \bar{x}_2) = ? \quad \text{Var}(\bar{x}_1 - \bar{x}_2) = ?$$

$$E(\bar{x}_1 - \bar{x}_2) = \mu_1 - \mu_2$$

$$\text{Var}(\bar{x}_1 - \bar{x}_2) = \text{Var}(\bar{x}_1) + \text{Var}(\bar{x}_2) = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$

$$\text{Then, } Z = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0, 1)$$

Sampling Distribution

$$P(-z_{\alpha/2} \leq Z \leq z_{\alpha/2}) = 1 - \alpha$$

$$\text{P } P\left(-z_{\alpha/2} \leq \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \leq z_{\alpha/2}\right) = 1 - \alpha.$$

$$\Rightarrow P\left(\bar{X}_1 - \bar{X}_2 - z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}} \leq \mu_1 - \mu_2 \leq \bar{X}_1 - \bar{X}_2 + z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}\right) = 1 - \alpha.$$

Note: When σ_1^2, σ_2^2 are unknown,

The sample variances s_1^2, s_2^2 can be

Substitute s_1^2, s_2^2 for large samples.

If $n_1, n_2 \geq 30$, this holds even if

The populations are not normally distributed.

CI for Case 1

Sample sizes: n_1 & n_2 , sample means: \bar{x}_1 , \bar{x}_2

Independent random samples from independent normal populations with known variances

σ_1^2 & σ_2^2 resp.

100(1 - α)% CI for $(\mu_1 - \mu_2)$:

$$\bar{x}_1 - \bar{x}_2 - 2\alpha_{1/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}} \leq \mu_1 - \mu_2 \leq \bar{x}_1 - \bar{x}_2 + 2\alpha_{1/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}.$$

S-pooled

Case 2: Variances unknown

Assume $\sigma_1^2 = \sigma_2^2 = \sigma^2$, Normal popⁿ, small samples

$$\hat{\sigma}_1^2 = s_1^2, \hat{\sigma}_2^2 = s_2^2$$

But $s_1^2 \neq s_2^2$

Pooled estimator of σ^2 : s_p^2

$$s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}$$

$$Z = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sigma \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim N(0, 1)$$

$$T = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim t_{n_1 + n_2 - 2}$$

CI for Case 2A

$\bar{x}_1, \bar{x}_2, s_1^2, s_2^2$: sample means and variances

i.i.d random samples of sizes n_1 and n_2

two independent Normal populations

$$\sigma_1^2 = \sigma_2^2 = \sigma^2 \text{ (unknown)}$$

100(1- α)% CI on the difference in

means $(\mu_1 - \mu_2)$:

$$[(\bar{x}_1 - \bar{x}_2) - t_{\alpha/2, n_1+n_2-2} s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}] \leq \mu_1 - \mu_2$$

$$\leq (\bar{x}_1 - \bar{x}_2) + t_{\alpha/2, n_1+n_2-2} s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

where $s_p = \sqrt{\frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1+n_2-2}}$

CI for Case 2B

2B: $\sigma_1^2 \neq \sigma_2^2$, small samples, Normal pop's

$\bar{x}_1, \bar{x}_2, s_1^2, s_2^2$: means and vars of two i.i.d random samples

sample size: n_1, n_2

two independent Normal pop's with

$\sigma_1^2 \neq \sigma_2^2$ (unknown)

100(1- α)% CI for $(\mu_1 - \mu_2)$ (Approx):

$$(\bar{x}_1 - \bar{x}_2) - t_{\alpha/2} \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}} \leq \mu_1 - \mu_2 \leq (\bar{x}_1 - \bar{x}_2) + t_{\alpha/2} \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

$$\gamma = \frac{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)^2}{\frac{(s_1^2/n_1)^2}{n_1-1} + \frac{(s_2^2/n_2)^2}{n_2-1}}, \text{ take closest integer}$$