

Statistics

L4: Estimation

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2025/2026

Random sample

Aim of statistics: to draw conclusions about population from a set of observed data.

Sample – a known and measurable entity representing a population having unknown features

Observations: x_1, x_2, \dots, x_n
(values of random variables X_1, X_2, \dots, X_n)

Estimation

θ - unknown parameter in a population distribution, $f(x)$

Estimator of θ – statistic (the function) that describes the method of computing of the estimate of θ :

$$\hat{\Theta} = \hat{\Theta}(X_1, X_2, \dots, X_n)$$

Estimator = statistic = **random variable** having its distribution!

Observations - the results of experiment: x_1, x_2, \dots, x_n

Estimate of θ :

$$\hat{\theta} = \hat{\Theta}(x_1, x_2, \dots, x_n)$$

Point estimation – population mean

Observed feature in a population – random variable X with distribution with unknown population mean μ

Parameter (unknown): population mean μ

Sample: X_1, X_2, \dots, X_n

Estimator of μ : sample mean, $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$

Observations: x_1, x_2, \dots, x_n

Estimate of μ : \bar{x}

$$\hat{\mu} = \bar{x}$$

Example 1

The mechanical engineer who designed the physical-therapy device collected the following data for the amount of time (in hours) spent by the test patients using his machine:

8; 12; 26; 10; 23; 21; 16; 22; 18; 17; 36; 9.

Estimate the mean time spent until recovery by all patients who use the same therapy.

Interval estimation

θ – unknown parameter

$\hat{\Theta}$ – parameter estimator

Definition

If

$$P \left(L(\hat{\Theta}) < \theta < U(\hat{\Theta}) \right) = 1 - \alpha$$

then

$$\left(L(\hat{\Theta}); U(\hat{\Theta}) \right)$$

is called a $(1 - \alpha) \cdot 100\%$ **confidence interval**, and the probability $(1 - \alpha)$ is a **confidence level**.

Sample mean distribution – revisited

(1) $X_i \sim N(\mu, \sigma)$, μ, σ – known:

$$\overline{X} \sim N\left(\mu, \frac{\sigma}{\sqrt{n}}\right) \Rightarrow \frac{\overline{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0, 1)$$

(2) X_i with μ, σ – known, distribution of X_i is arbitrary, sample is large:

$$\overline{X}_{\text{app}} \sim N\left(\mu, \frac{\sigma}{\sqrt{n}}\right) \Rightarrow \frac{\overline{X} - \mu}{\frac{\sigma}{\sqrt{n}}}_{\text{app}} \sim N(0, 1)$$

(3) $X_i \sim N(\mu, \sigma)$, μ – known, σ – unknown: $\frac{\overline{X} - \mu}{\frac{S}{\sqrt{n}}} \sim t_{n-1}$

Confidence interval for μ

(1)

Assumption: $X_i \sim N(\mu, \sigma)$, μ - unknown, σ - known

We are $100(1 - \alpha)\%$ confident, that the interval

$$\left(\bar{X} - z_{1-\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}, \bar{X} + z_{1-\alpha/2} \cdot \frac{\sigma}{\sqrt{n}} \right)$$

covers the true unknown **population** mean μ .

$z_{1-\alpha/2}$ – quantile of $N(0, 1)$: `qnorm(1 - $\alpha/2$)`

Confidence interval for μ

(2)

Assumptions: arbitrary population distribution with unknown μ ,
large sample ($n \geq 30$)

We are $100(1 - \alpha)\%$ confident, that the interval

$$\left(\bar{X} - z_{1-\alpha/2} \cdot \frac{S}{\sqrt{n}}, \bar{X} + z_{1-\alpha/2} \cdot \frac{S}{\sqrt{n}} \right)$$

covers the true unknown **population** mean μ .

$z_{1-\alpha/2}$ – quantile of $N(0, 1)$: `qnorm(1 - $\alpha/2$)`

Confidence interval for μ

(3)

Assumption: $X_i \sim N(\mu, \sigma)$, μ - unknown, σ - unknown

We are $100(1 - \alpha)\%$ confident, that the interval

$$\left(\bar{X} - t_{n-1, 1-\alpha/2} \cdot \frac{S}{\sqrt{n}}, \bar{X} + t_{n-1, 1-\alpha/2} \cdot \frac{S}{\sqrt{n}} \right)$$

covers the true unknown **population** mean μ .

$t_{n-1, 1-\alpha/2}$ – quantile of t_{n-1} : $\text{qt}(1 - \alpha/2, n - 1)$

Example 1 - cont.

The mechanical engineer who designed the physical-therapy device collected the following data for the amount of time (in hours) spent by the test patients using his machine:

8; 12; 26; 10; 23; 21; 16; 22; 18; 17; 36; 9.

Assuming normality of the distribution of time, estimate with 95% of confidence the mean time spent until recovery by all patients who use the same therapy.

Example 1 - cont.

Confidence intervals for μ in R

(1) $X_i \sim N(\mu, \sigma)$, σ – known:

`z.test(data, sigma.x = σ , conf.level = $1 - \alpha$)`

(2) distribution of X_i is arbitrary, sample is large:

σ – known

`zsum.test(mean(data), σ , n , conf.level = $1 - \alpha$)`

σ – unknown

`zsum.test(mean(data), sd(data), n , conf.level = $1 - \alpha$)`

(3) $X_i \sim N(\mu, \sigma)$, σ – unknown:

`t.test(data, conf.level = $1 - \alpha$)`

CAUTION! For (1) and (2) - **BSDA** package required

Example 1 - cont.

The mechanical engineer who designed the physical-therapy device collected the following data for the amount of time (in hours) spent by the test patients using his machine:

8; 12; 26; 10; 23; 21; 16; 22; 18; 17; 36; 9.

Assuming normality of the distribution of time, estimate with 95% of confidence the mean time spent until recovery by all patients who use the same therapy.

Example 1 - cont.

Point estimation – population variance

Observed feature in a population – random variable X with distribution with unknown population variance σ^2

Parameter (unknown): population variance σ^2

Sample: X_1, X_2, \dots, X_n

Estimator of σ^2 : sample variance, $S^2 = \frac{1}{n-1} \left(\sum_{i=1}^n X_i^2 - n\bar{X}^2 \right)$

Observations: x_1, x_2, \dots, x_n

Estimate of σ^2 : s^2

$$\hat{\sigma}^2 = s^2$$

Example 1 - cont.

The mechanical engineer who designed the physical-therapy device collected the following data for the amount of time (in hours) spent by the test patients using his machine:

8; 12; 26; 10; 23; 21; 16; 22; 18; 17; 36; 9.

Estimate the standard deviation of the time spent until recovery by all patients who use the same therapy.

Confidence interval for σ^2

Assumption: $X_i \sim N(\mu, \sigma)$, σ - unknown

We are $100(1 - \alpha)\%$ confident, that the interval

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

covers the true unknown **population** variance σ^2 .

$\chi_{n-1,\beta}^2$ – quantiles of χ_{n-1}^2 : `qchisq(β , $n - 1$)`

Confidence interval for σ^2 in R:

```
sigma.test(data, conf.level = 1 -  $\alpha$ )
```

CAUTION! `TeachingDemos` package required

Example 1 - cont.

The mechanical engineer who designed the physical-therapy device collected the following data for the amount of time (in hours) spent by the test patients using his machine:

8; 12; 26; 10; 23; 21; 16; 22; 18; 17; 36; 9.

Assuming normality of the distribution of time, estimate with 95% of confidence the variance and standard deviation of time spent until recovery by all patients who use the same therapy.

Example 1 - cont.

Point estimation – population proportion

Observed feature in a population – random variable X with distribution $\text{bin}(1, p)$ with unknown probability of success p

Parameter (unknown): population proportion p

Sample: X_1, X_2, \dots, X_n

Estimator of p : sample proportion, $\hat{p} = \frac{T}{n}$

$T = \sum_{i=1}^n X_i$ - number of “successes” in a sample

Observations: x_1, x_2, \dots, x_n

Estimate of p : \hat{p}

Example 2

A school district is trying to determine its students' reaction to a proposed dress code. To do so, the school selected a random sample of 150 students and questioned them. If 70 were in favor of the proposal, then estimate the proportion of all students who are in favor.

Confidence interval for p

Assumption: **large** sample ($n \geq 100$)

We are $100(1 - \alpha)\%$ confident, that the interval

$$\left(\hat{p} - z_{1-\alpha/2} \cdot \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}, \hat{p} + z_{1-\alpha/2} \cdot \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}} \right)$$

covers the true unknown **population** proportion p .

$z_{1-\alpha/2}$ – quantile of $N(0, 1)$: `qnorm(1 - $\alpha/2$)`

Confidence interval for p in R:

<code>binom.test(T, n, conf.level = $1 - \alpha$)</code>	(exact)
<code>prop.test(T, n, conf.level = $1 - \alpha$)</code>	(approximate)

Example 2 - cont.

A school district is trying to determine its students' reaction to a proposed dress code. To do so, the school selected a random sample of 150 students and questioned them. If 70 were in favor of the proposal, then estimate with 99% of confidence the true proportion of all students who are in favor of the proposal.

Example 2 - cont.

Example 3

On December 24, 1991, *The New York Times* reported that a poll indicated that 46% of the population was in favor of the way that President Bush was handling the economy, with a margin of error of $\pm 3\%$. What does this mean? Can we conclude how many people were questioned if it is known, that the standard confidence level in media is 95%?

