

# 1 Key Concepts and Basic

## Statistics

- ① Key concepts
- ② Descriptive statistics
- ③ Frequency and probability-distributions
- ④ Hypothesis testing
- ⑤ P-values
- ⑥ Interval estimation
- ⑦ Suggested exercises

# ① Key concepts



Def. Population: All the units to be studied



Def. Sample: A subset of the population

## Some common sampling methods:

\* Simple random sampling: All subsets of the same size have the same probability of being drawn

↑ Usually means: All units have an equal probability of being selected

\* Systematic sampling: Selecting  
 . . . . .  
 the units at regular intervals  
 from a "list" of all the  
 population members

↑ E.g. an alphabetically  
 ordered list

\* Stratified sampling: Classify  
 . . . . .  
 the population into strata  
 (i.e. categories) and then  
 sample from each stratum  
 (i.e. category)

\* Convenience sampling: Select  
 . . . . .  
 those that are close at hand/  
 easy to investigate

## ② Descriptive statistics

### Measures of central tendency

→ Intended to indicate where the majority of the values are

→ The sample mean:

$$\bar{x} = \frac{\sum x}{n}$$

Example: 8, 10, 9, 7, 8

$$\bar{x} = \frac{8+10+9+7+8}{5} = \underline{\underline{8.4}}$$

→ The median: The middle value that separates the highest values from the lowest

→ The mode: The most frequent value 1.5

→ The weighted mean

## Measures of variability:

→ Intended to indicate the degree of variability or dispersion

Example: Sample 1 = 8, 10, 9, 7, 8  $\bar{x} = 8.4$

— " — Sample 2 = 6, 7, 11, 8, 10  $\bar{x} = 8.4$

Sample 1:     x     xx     x     x

— (1-2):

       x     x     x                     x             x



→ The sample variance:

$$s^2 = \frac{\sum (x - \bar{x})^2}{n-1}$$

Example:

sample 1			sample 2		
x	(x - $\bar{x}$ )	(x - $\bar{x}$ ) <sup>2</sup>	x	(x - $\bar{x}$ )	(x - $\bar{x}$ ) <sup>2</sup>
8	-0.4	0.16	6	-2.4	5.76
10	1.6	2.56	7	-1.4	1.96
9	0.6	0.36	11	2.6	6.76
7	-1.4	1.96	8	-0.4	0.16
8	-0.4	0.16	10	1.6	2.56
SUM		5.2			17.2

$$s^2 = \frac{\sum (x - \bar{x})^2}{n-1} = \frac{5.2}{4} = \underline{\underline{1.3}}$$

$$s^2 = \frac{17.2}{4} = \underline{\underline{4.3}}$$

→ The sample standard deviation:

$$s = \sqrt{s^2} = \sqrt{\frac{\sum (x - \bar{x})^2}{n-1}}$$

Examples: Sample 1 =  $s = \sqrt{1.3} = 1.14$

2 =  $s = \sqrt{4.3} = 2.07$

→ The sample range: Highest value minus the lowest

Examples: Sample 1:  $10 - 7 = \underline{\underline{3}}$

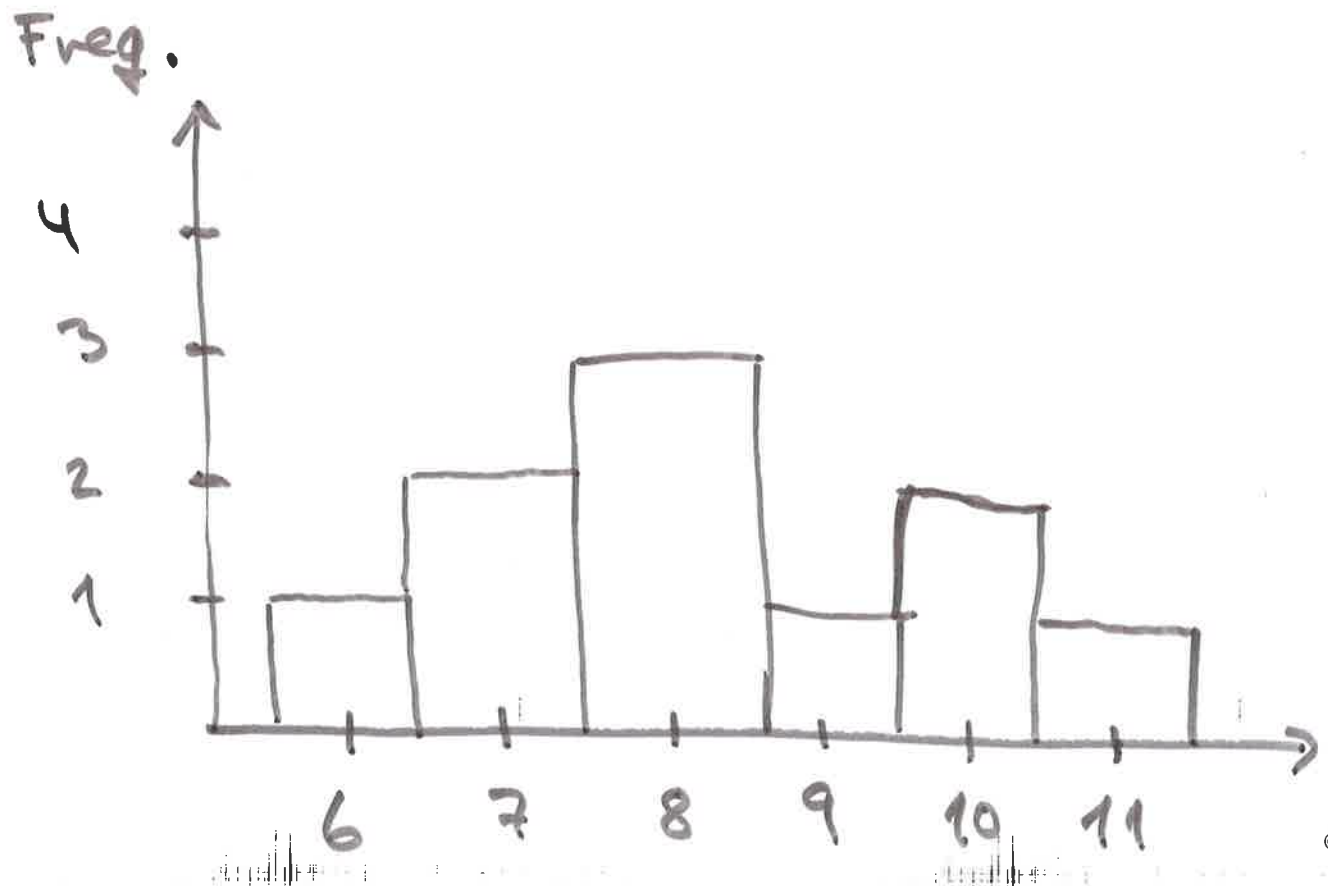
— 1 — 2:  $11 - 6 = \underline{\underline{5}}$

### ③ Frequency and probability distributions

Def. Frequency distribution: A description of the number of times each value of a variable appears in the sample

Example: A histogram, i.e. a bar graph with no space between the bars

Sample = 8, 10, 9, 7, 8, 6, 7, 11, 8, 10



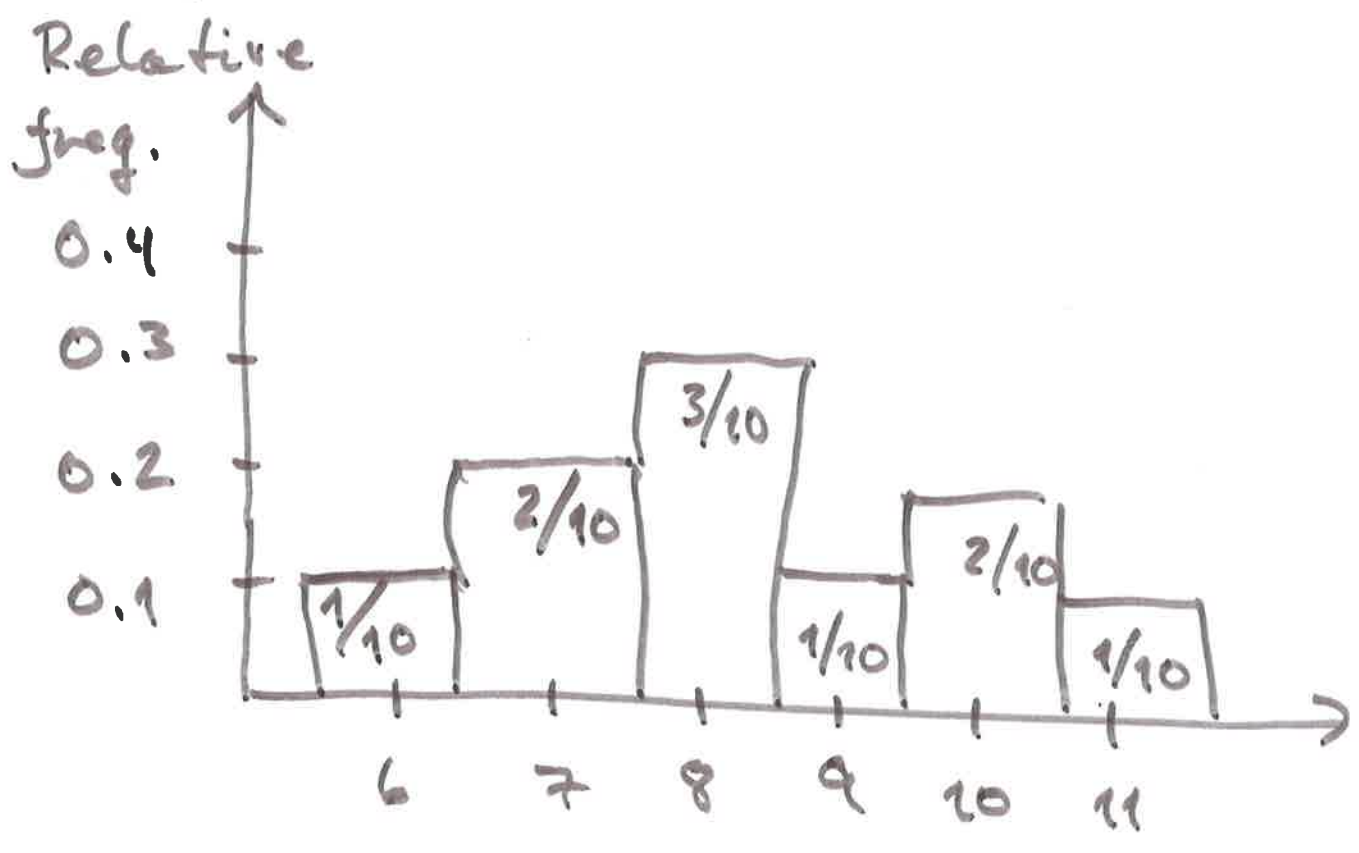


Def. Relative frequency distribution:

A description of the proportion of times each value appears in the sample

Example: A histogram

Sample = Same as previous ( $n = 10$ )



Def. Probability distribution:

A histogram of relative frequencies

Example: Previous!

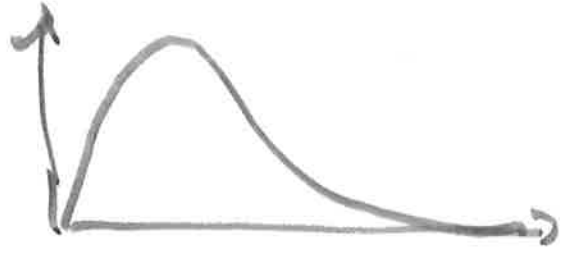
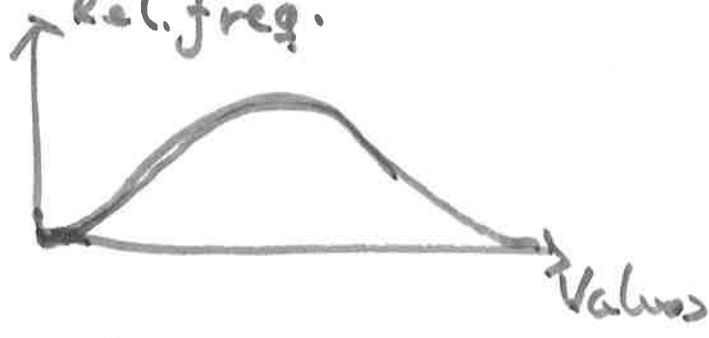
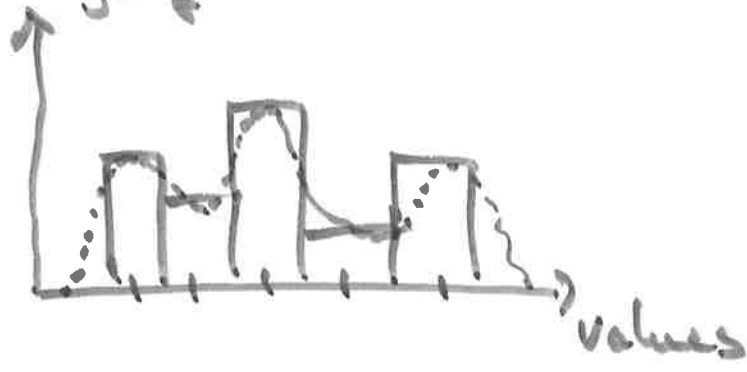
Discrete  
(categorical)  
probability  
distributions

vs.

Continuous  
probability dist-  
ributions

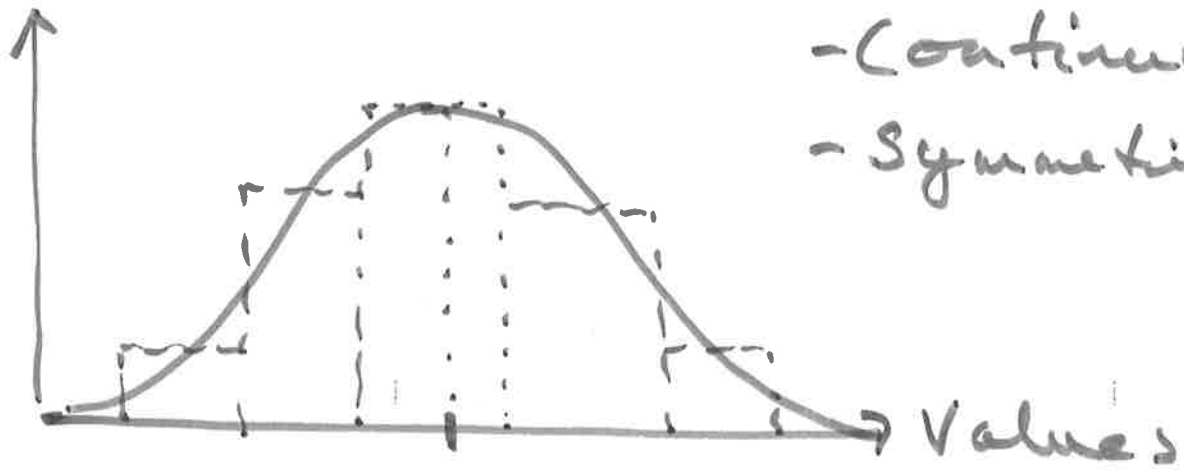
Rel. freq.

Rel. freq.



The normal distribution:

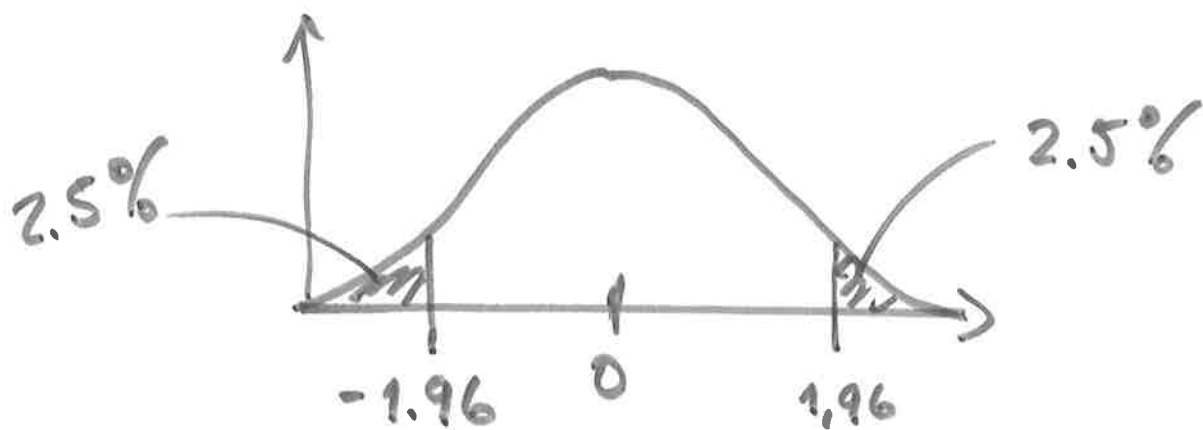
Rel. freq.



Mean

- Continuous
- Symmetric

Some properties of the standard normal distribution: 1.11



→ mean is 0

→ Area under the curved line is 1 or 100%

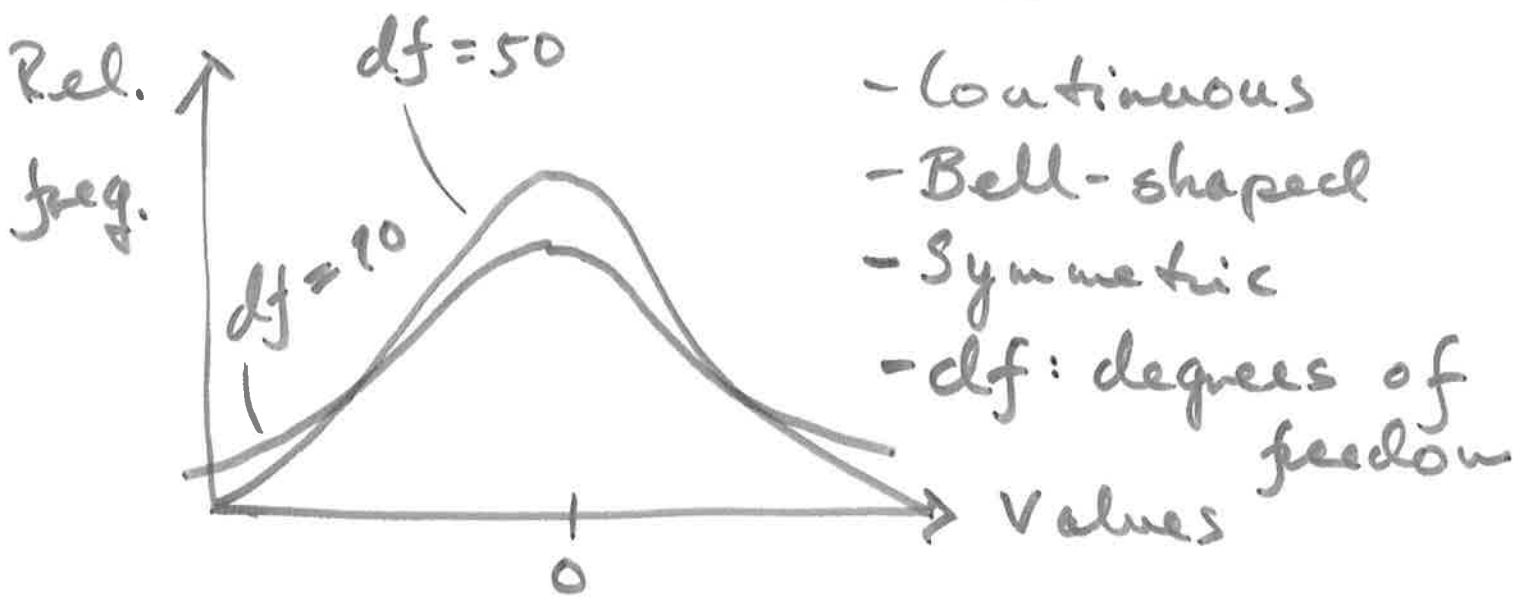
→ Symmetry:

- Area to <sup>the</sup> left of 0 is 0.5 or 50%

- Area to the right of 0 is 0.5

- Area to the left of -1.96 is 2.5%

# The t-distribution:



## Some properties:

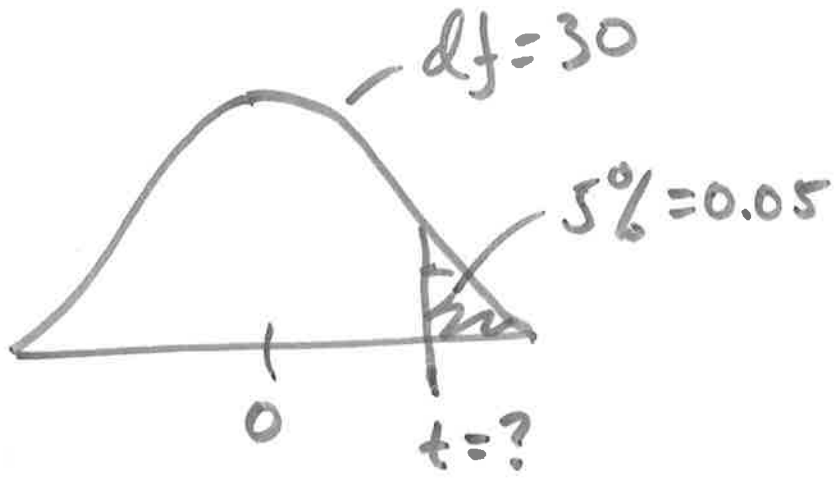
- There is in fact an infinite number of t-distributions, one for each df
- df is usually approximately equal to the number of observations
- The "flatness" of the t-distribution depends on df

→ When  $df \rightarrow \infty$ , then the  $t$ -distribution becomes a standard normal distribution

Examples:

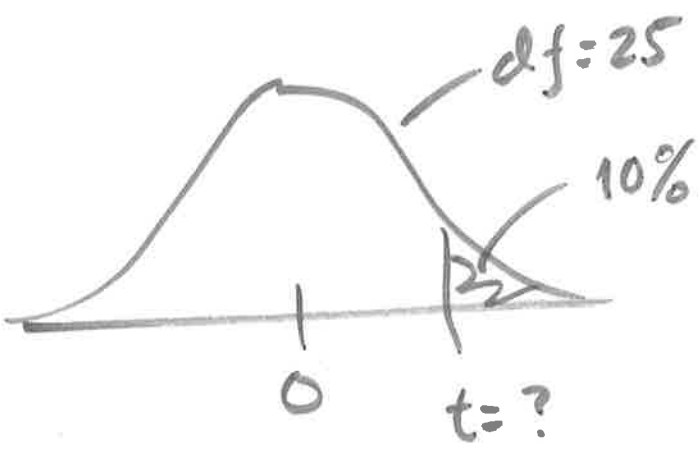
What is  $t$ ?

$t_{0.05}(30) = 1.697$



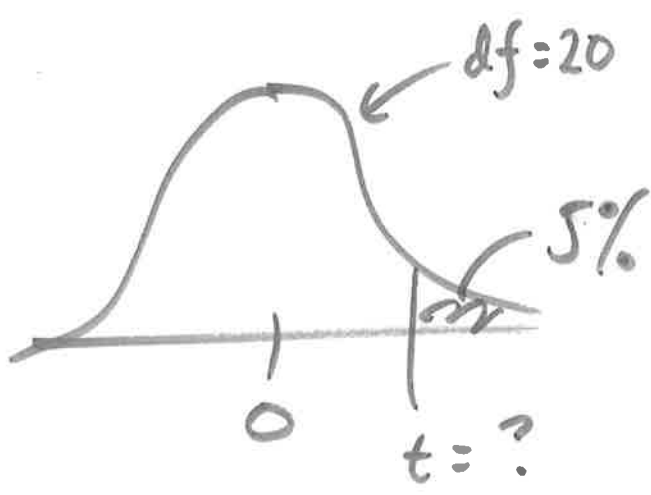
What is  $t$ ?

$t_{0.10}(25) = 1.316$



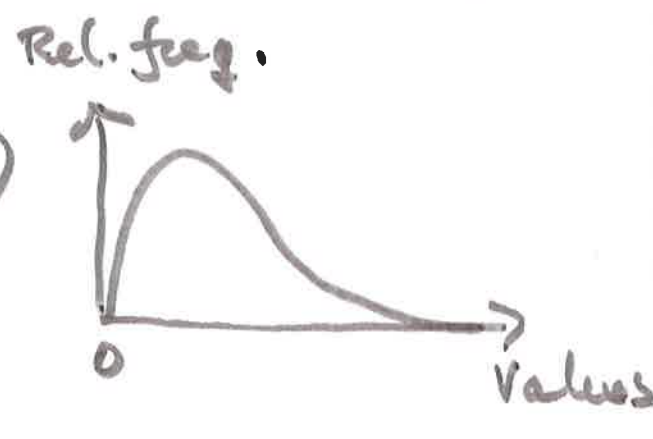
What is  $t$ ?

$t_{0.05}(20) = 1.725$

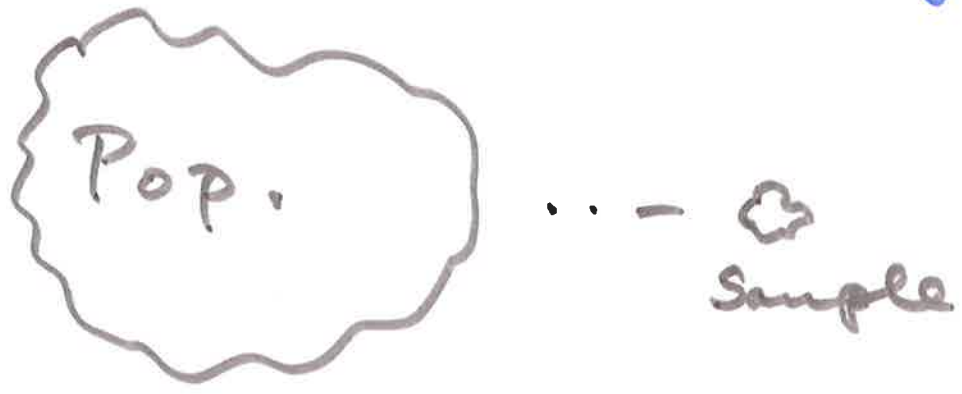


# Other important probability distributions:

- Chi-squared dist.
- F-distribution



## ④ Hypothesis testing



Def. Hypothesis: A claim about a population

Def. Null hypothesis ( $H_0$ ): A claim about the population in terms of an equality

Examples:

→  $\mu = 20 : H_0$  (population mean equal to 20)  
 $\uparrow$

→  $H_0 : \mu = 9$

→  $H_0 : \mu = 2$

Def. Alternative hypothesis ( $H_A$ ):

A claim about the population of the "not  $H_0$ " type

Examples:

→  $H_A : \mu \neq 20$      $H_A : \mu > 20$      $H_A : \mu < 20$

→  $H_A : \mu \neq 9$      $H_A : \mu > 9$      $H_A : \mu < 9$

→

NOTE: The investigator chooses  $H_0$  and  $H_A$

## Some conventions:

- $H_A$  is usually the claim you would like to test or investigate
- $H_0$  is what previous knowledge or findings suggest
- Moral and/or health-reasons that come into play

Def. Significance level ( $\alpha$ ): The probability of rejecting  $H_0$  when it is true

NOTE: The investigator chooses  $\alpha$

Examples:  $\alpha = 0.01 = 1\%$   
 $\alpha = 0.05 = 5\%$   
 $\alpha = 0.10 = 10\%$  } most common ones



1.17

Def. Test expression (statistic):

A formula whose value indicates whether to keep or reject  $H_0$

Test expressions that are t-distributed typically have the following form:

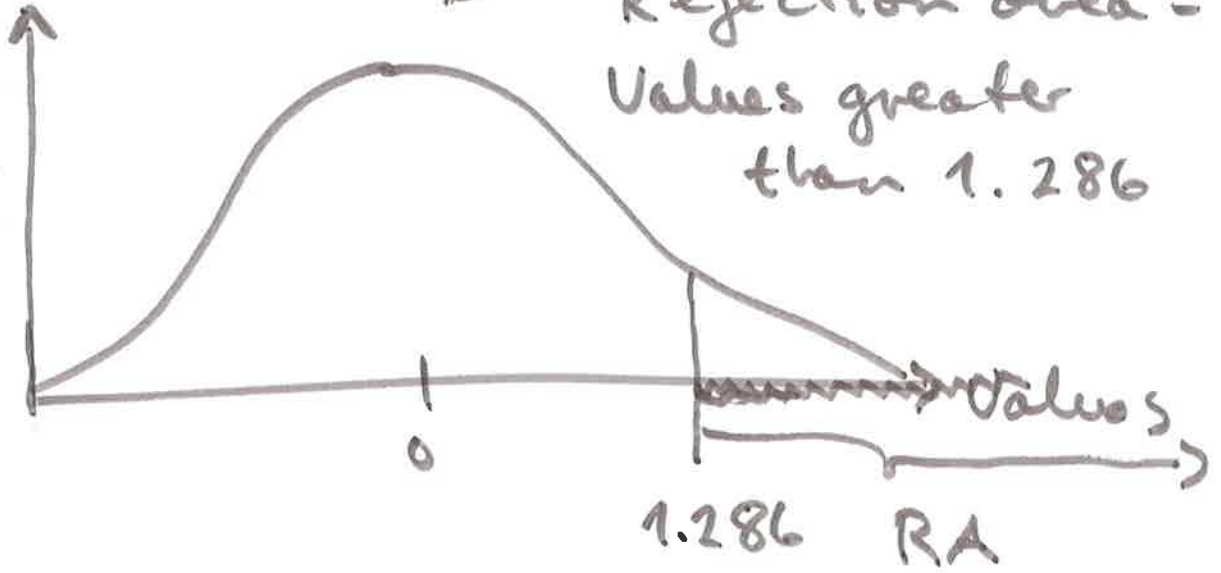
$$\frac{\bar{x} - H_0 \text{ value}}{s}$$

sample estimate                      sample standard dev.

Def. Rejection area: The values of the test-expression that makes us reject  $H_0$

Example:

Relative  
freq.

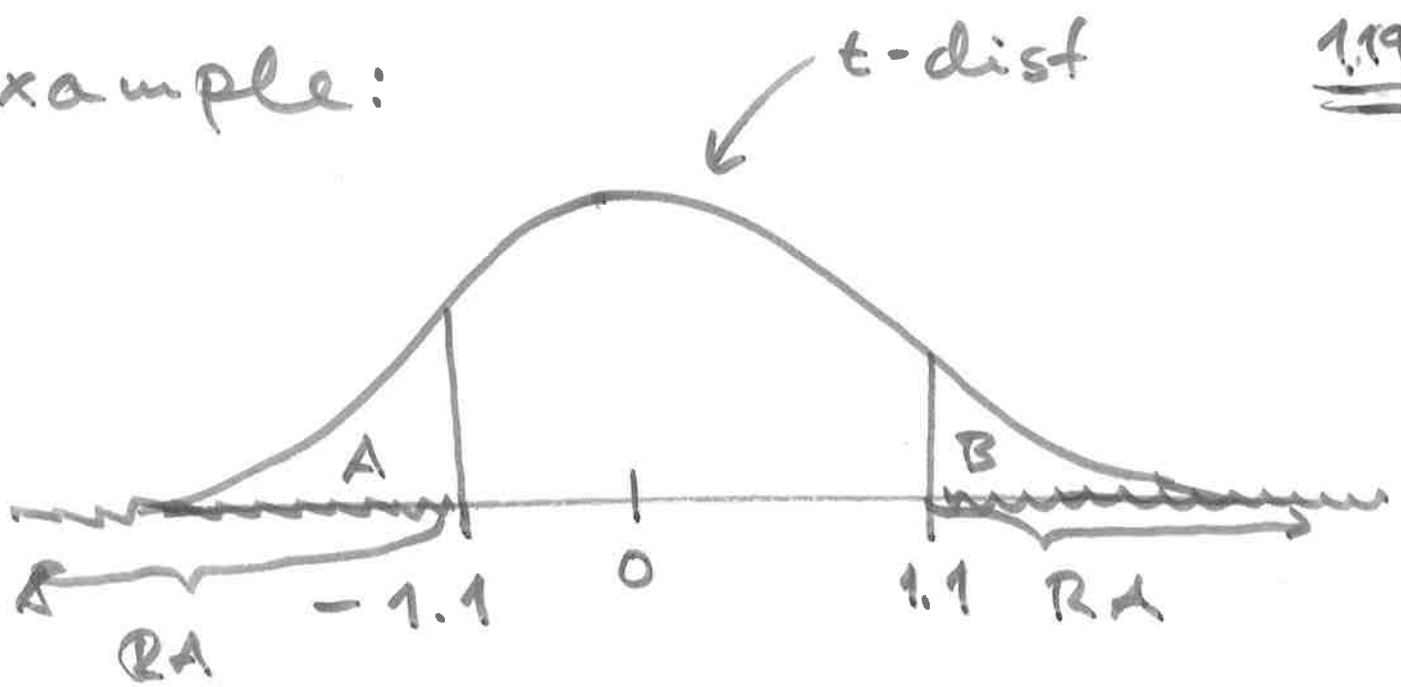


Def. Critical value(s): The bound-  
ary (or boundaries) of the re-  
jection area

Example: Above (i.e. previous  
example) we have 1 critical  
value, and it is equal to 1.286

Example:

1.19



Rejection area = Values greater than 1.1 and values smaller than -1.1. Critical values = 1.1 and -1.1.

Testing in 4 steps:

Step 1: Choose  $\alpha$ ,  $H_0$  and  $H_A$

2: Find the critical value(s) and identify the rejection area

3: Compute the value of the test-expression

4: Conclude: Reject  $H_0$  if the test-value lies in the rejection area, otherwise keep  $H_0$

Type 1 and 2 errors:

	$H_0$ True	$H_A$ true
Reject $H_0$	Type 1 error	Correct!
Keep $H_0$	Correct!	Type 2 error

Testing the value of a population mean

Step 1: Choose  $\alpha$ ,  $H_0$  and  $H_A$

1.21

2: Identify the rejection area using a t-distribution with  $n-1$  df

3: Compute the test-expression

$$\frac{\bar{x} - H_0 \text{ value}}{s/\sqrt{n}}$$

4: Conclude: Reject  $H_0$  if the test-value lies in the rejection area, otherwise keep  $H_0$

Example: Travel time (in minutes)

Claim: Average travel time is greater than 20 minutes

Sample: 20, 15, 17, 55, 20, 10, 20, 90, 15  
18, 40

$$n = 11$$

$$\bar{x} = 29.09$$

$$s = 24.0394$$

1.22

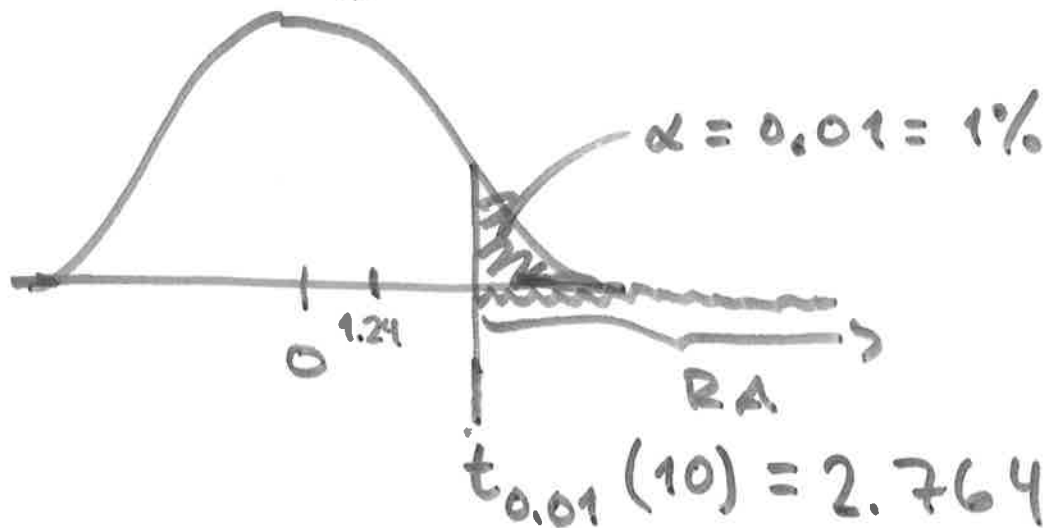
Step 1:  $\alpha = 1\%$

$$H_0: \mu = 20$$

$$H_1: \mu > 20$$

2: Identify the rejection area:

$$df = n - 1 = 11 - 1 = 10$$



3: Compute the test-value:

$$\frac{29.09 - 20}{\frac{24.0394}{\sqrt{11}}} = 1.2417$$

4: Conclusion: We keep  $H_0$ . That is,

We have found support in 1.23  
favour of the claim that the  
average travel time is  
20 minutes

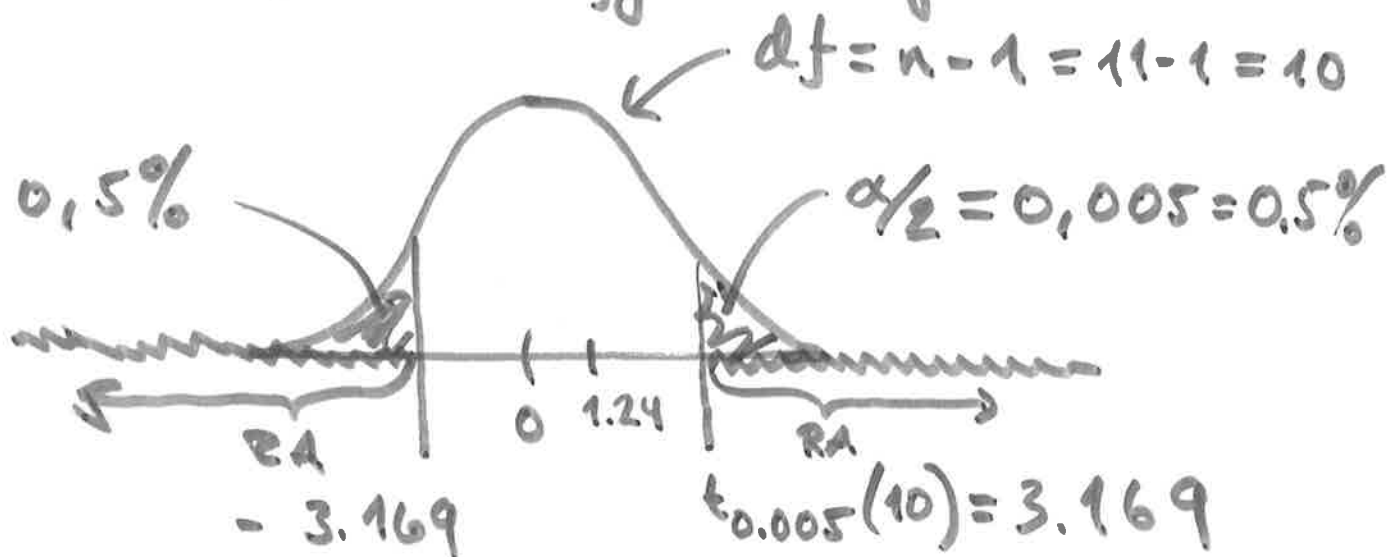
Example:

- Claim: Average travel time is different from 20 minutes
- Sample: Same as previous

Step 1:  $\alpha = 0.01 = 1\%$        $H_0: \mu = 20$

$H_1: \mu \neq 20$

2: Identify the rejection area:



RA: Values higher than 3.169 and values lower than -3.169

3: Value of test-expression:

1.24

$$29.09 \rightarrow \frac{\bar{X} - H_0 \text{ value}}{s/\sqrt{n}} = 1.2417$$

$\leftarrow 20$

$\leftarrow 11$

4: Conclusion: We keep  $H_0$



## ⑤ P-values

Intuition: A number between 0 and 1 that summarises the information from a hypothesis test

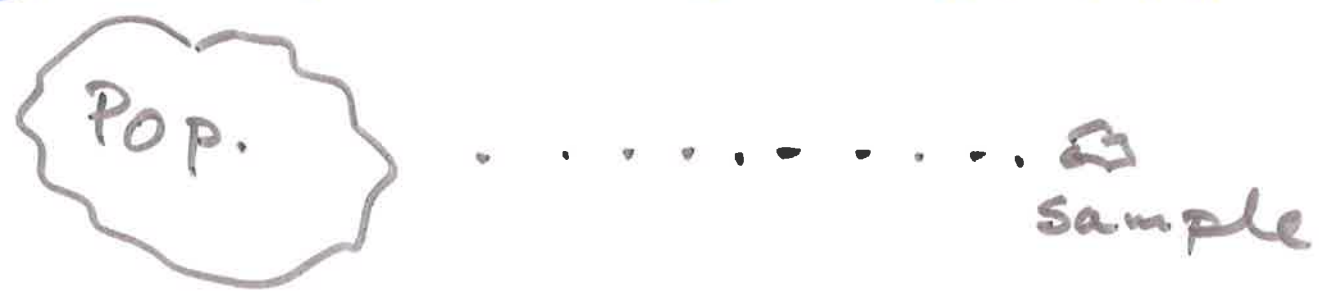
Def. P-value: The smallest significance level at which the null hypothesis can be rejected

In practice: Reject  $H_0$  if the p-value is smaller than  $\alpha$

↑ The p-value game & dance!

NOTE: In order to compute a p-value "by hand",  $H_0$  and  $H_1$  must already have been defined, and the test-value must already have been computed

### (b) Interval estimation



$\mu$	} sample counterparts:	$\bar{x}$
$\sigma^2$		$s^2$
$\sigma$		$s$

Def.  $(1-\alpha) \cdot 100\%$  confidence interval: An interval that contains the population value of interest (e.g.  $\mu, \sigma^2, \sigma$ , etc.) with  $(1-\alpha) \cdot 100\%$  degree of certainty

Example: 90% confidence interval for the mean time it takes from you wake up until you leave the house (Q2)

Sample = 110, 45, 75, 90, 70, 90, 40, 30, 45, 60

n = 10     $\bar{x} = 65.5$      $s_x^2 = 674.72$      $s_x = 25.98$

$$L = \bar{x} - t_{\alpha/2}(df) \cdot s_x / \sqrt{n} = 50.43$$

Annotations:  $\bar{x} = 65.5$ ,  $\alpha/2 = 0.05$ ,  $df = 10 - 1 = 9$ ,  $t_{0.05}(9) = 1.833$ ,  $n = 10$ .

$$U = \bar{x} + t_{\alpha/2}(df) \cdot s_x / \sqrt{n} = 80.57$$

Annotations:  $\bar{x} = 65.5$ ,  $t_{\alpha/2}(df) = 1.833$ ,  $s_x / \sqrt{n}$  is shared with the L calculation.

Interpretation: The interval [50.43, 80.57] contains the population mean with 90% degree of certainty

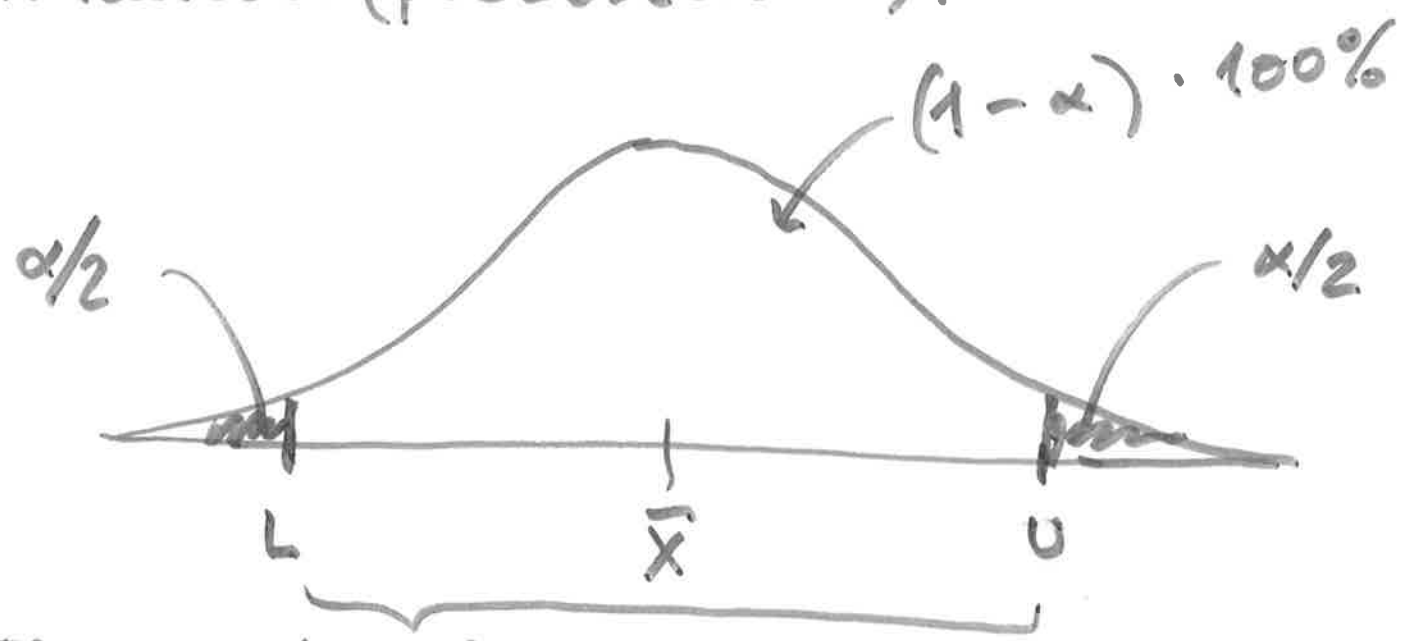
Example: Suppose  $[6, 11]$  is a 95% confidence interval for  $\mu$ , then  $\mu$  lies between 6 and 11 with 95% degree of certainty

Computation of confidence interval for  $\mu$

Lower bound =  $\bar{x} - t_{\alpha/2}(df) \cdot s_x / \sqrt{n}$   
 $\uparrow$   
 $n-1$

Upper bound =  $\bar{x} + t_{\alpha/2}(df) \cdot s_x / \sqrt{n}$

Intuition (probabilistic):



This is the  $(1 - \alpha) \cdot 100\%$  confidence interval

## ⑦ Suggested exercises

Exercise set 1: 1(a)-g), 4(a)iv),

4(b)iv), 4(c)i), 5a)i), 5a)iii),

5b)i), 5b)iii), 5c)i), 7