

## 浙江大学爱丁堡大学联合学院 ZJU-UoE Institute

## The t-test: practical applications and variants

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## **Learning objectives**

At the end of this lecture, you should be able to:

- Describe variants of the t-test
  - 1-tailed vs 2-tailed
  - 1- vs 2-sample
  - · Paired vs unpaired
- Choose an appropriate type of t-test for a given problem
- Describe and use non-parametric alternatives to the *t*-test



## Our earlier example

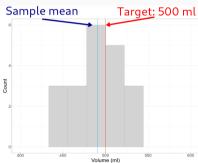
Is the factory filling each bottle with enough Guinness?



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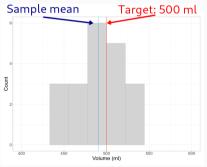


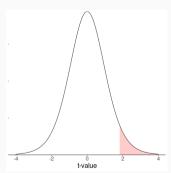


## Our earlier example

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## Variants of the *t*-test

1-tailed vs 2-tailed

#### One-tailed vs two-tailed t-test

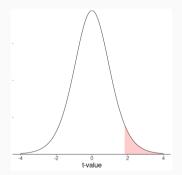
Compare the following two questions:

- Is the factory filling each bottle with enough Guinness?
- Is the factory filling each bottle with a volume **different** from 500ml?

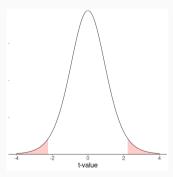
#### One-tailed vs two-tailed t-test

## Compare the following two questions:

- Is the factory filling each bottle with **enough** Guinness?
- Is the factory filling each bottle with a volume **different** from 500ml?

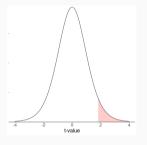


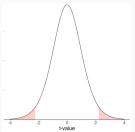
1-tailed - Area of interest is only on one side of the distribution (ex. 5%)



2-tailed - Area of interest is on both sides of the distribution (ex. 2.5% on each side)

## Which one should I use?





#### One-tailed

- If we look for outcomes in one direction only, e.g. A > B (or A < B)</li>
- $H_0$ : A is not greater (or smaller) than B
- $H_1$ : A is greater (or smaller) than B

#### **Two-tailed**

- If we look for outcomes in both directions, e.g.  $A \neq B$
- Ho: A is equal to B
- $H_1$ : A is not equal to B

#### Critical values for 1-tailed and 2-tailed tests

For the same significance level, the critical values for 1-tailed and 2-tailed tests are different!

	One tailed			Two tailed			
C	i.f.	t <sub>.100</sub>	t.050*	t <sub>.025</sub> **	t <sub>.010</sub>	t <sub>.005</sub>	d.f.
	1	3.078	6.314	12.706	31.821	63.657	1
	2	1.886	2.920	4.303	6.965	9.925	2
	3	1.638	2.353	3.182	4.541	5.841	3
	4	1.533	2.132	2.776	3.747	4.604	4
	5	1.476	2.015	2.571	3.365	4.032	5
	6	1.440	1.943	2.447	3.143	3.707	6
	7	1.415	1.895	2.365	2.998	3.499	7
	8	1.397	1.860	2.306	2.896	3.355	8
	9	1.383	1.833	2.262	2.821	3.250	9
	10	1.372	1.812	2.228	2.764	3.169	10

## Variants of the *t*-test

1-sample vs 2-sample

## 1-sample vs 2-sample t-test

#### Compare the following two questions



Is the factory filling each bottle with a volume different from 500ml?



Is the factory in Dublin filling each bottle with a volume different from the factory in Glasgow?

## 1-sample vs 2-sample t-test

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difference from the ref. value  $t = \frac{\bar{x} - \mu}{s / \sqrt{n}}$ 

sample standard deviation (estim, of the population  $\sigma$ )

standard error of the mean



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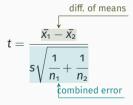
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$$\dot{s} = \frac{\bar{X_1} - \bar{X_2}}{\sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}}$$

# Equal variances Student's t-test

**Unequal variances** 

## 2-sample t-test - an example

Example:

**Sample 1**: 4, 6, 8, 10 **Sample 2**: 1, 5, 3, 4

$$H_0$$
:  $\mu_1 = \mu_2$   
 $H_A$ :  $\mu_1 \neq \mu_2$ 

What should we do to test this hypothesis?

## Variants of the *t*-test

Paired vs unpaired

## Paired vs unpaired t-test

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**Unpaired t-test**: Different subjects are measured at different times or under different conditions.

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Unpaired t-test: Different subjects are measured at different times or under different conditions.

**Paired t-test example** You want to study the effect of a drug on the blood pressure of 10 patients. You measure the blood pressure of each patient before and after the administration of the drug.

## Paired t-test - calculating t statistics

The paired t-test is effectively a 1-sample t-test on the differences between the two measurements (with reference value = 0).

avg. difference between paired measurements

$$t = \frac{\frac{1}{d}}{\left(\frac{s_d}{\sqrt{n}}\right)}$$

standard deviation of the differences

#### t-test variants in R

Use the t.test() function in R to perform a t-test.

```
t.test(x, y = NULL,
alternative = c("two.sided", "less", "greater"),
mu = 0, paired = FALSE, var.equal = FALSE,
conf.level = 0.95, ...)
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#### 1 sample

mu = reference value

#### 1-tailed

alternative = "greater" Or "less"

#### **Paired**

paired = TRUE

#### 2 sample

Must provide two vectors of values as x and y For Welch's t-testvar.equal = FALSE ()

#### 2-tailed

alternative = "two.sided" (default)

## Unpaired

paired = FALSE (default)

State the hypothesis. What is  $H_0$  and what is  $H_A$ ?

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- 2. What direction are you looking at? 1-tailed or 2-tailed?
- 3. Are you comparing 1-sample or 2-samples?
- 4. Are the observations paired or unpaired?

## What to do when assumptions are not met?

#### Normality of the sample(s)

- If the deviation from normality is small, the t-test is robust to it, and we can still use it.
- If the deviation from normality is large, we can try transforming the data and see if the data become more normal. Examples of transformations include log, square root, inverse, etc.
- If the data cannot be transformed, we can use a **non-parametric** test instead (see next slides).

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## Homogeneity of variances

• We can use Welch's t-test instead of Student's t-test.

#### **Non-parametric tests**

A **non-parametric** (or distribution-free) test is a statistical test that does not assume that the data follow a particular distribution.

As alternatives to the *t*-test, we can use the following non-parametric tests:

- Wilcoxon signed-rank test for paired data or one-sample data
- · Mann-Whitney U test for unpaired data

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The idea of these two tests is very similar, and in R you can use the same function wilcox.test() to perform both tests.

```
Wilcoxon: wilcox.test(x, y, paired = TRUE, ...)
Wilcoxon, one-sample: wilcox.test(x, mu = 0, ...)
Mann-Whitney: wilcox.test(x, y, paired = FALSE, ...)
```

## The idea behind non-parametric tests

Wilcoxon and Mann-Whitney tests are based on the **rank** of the observations. They do not assume a particular distribution of the data, but they do assume that the observations are **independent** and that the **distribution of the observations is the same** (or similar) in the two groups.

They compare medians rather than means.

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Note: they have **less statistical power** than the *t*-test, so they are less likely to detect a difference when there is one. However, they are more robust to deviations from normality and to outliers.

## How non-parametric tests work

## **Mann-Whitney test:**

- Combine all observations and rank them from smallest to largest
- · Sum the ranks in each group
- Calculate the test statistic U and compare it to the critical value (we're not going to do this by hand, but R will do it for you!)

## How non-parametric tests work

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#### Wilcoxon test:

- Calculate the difference between the two observations in each pair
- Rank the differences from smallest to largest
- Sum the ranks of the positive differences and the ranks of the negative differences
- Calculate the test statistic W and compare it to the critical value

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