

Introduction to Statistics

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Parametric and non parametric tests

T-tests: summary

T-test in general Used to compare means

One-sample t-test Compare the mean of a sample to a given number

> Two-sample t-test Compare the means of two samples

Paired t-test Compare the difference between pairs of related data points

Assumptions for the t-test

- This test is one of the most widely used tests. However, it can be used only if the background assumptions are satisfied.
 - Data values must be **independent**.
 - Data in each group must be obtained via a random sample from the population.
 - The variances for the two independent groups are **equal**.
 - Data in each group are **normally** distributed.
 - Data values are continuous.

T-test in practice

- Analysts generally don't use the standard (Student) t-test.
- The Welch t-test takes into account different sample variances.
- If the variances are actually the same, it provides the same results as the standard Student t-test.

```
> t.test(KO_WT$weight ~ KO_WT$genotype)
```

Welch Two Sample t-test

Graphical tools to assess normality



DATASET 1

Normality tests

- Examples: Shapiro-Wilk, Kolmogorov-Smirnov.
- 1 weight\$weight 0.902 0.166
- Normality tests are statistical tests
 - They test *against* normality (and can not show/prove normality)
 - They may not have enough power
 - Conversely, if n is large, even a very small departure from normality may be significant.

What happens if the data is not normally distributed ?

- Transform the data using the log function
 - Examples of data that should be logged:
 - Highly asymetrical data
 - Data spanning several orders of magnitude
 - Data originating from ratios
- Another possible solution: non-parametric tests
 - Wilcoxon rank-sum test (WRS)
 - Mann–Whitney U test
 - Mann–Whitney–Wilcoxon (MWW)



Mann-Whitney U test

- General formulation:
 - All the observations from both groups are independent of each other
 - The variable is ordinal or numerical
 - H₀: the distributions of both populations are equal
 - H₁: the distributions are not equal

Mann–Whitney U test: an example

Consider a Phase II clinical trial designed to investigate the effectiveness of a new drug to reduce symptoms of asthma in children. Participants are asked to record the number of episodes of shortness of breath over a 1 week period following receipt of either the new drug or a placebo. Is the difference significant ?



Mann–Whitney U test: an example

		Total sample	e (ordered)	Ranks		
Placebo	New drug	Placebo	New drug	Placebo	New drug	
7	3		1		1	
5	6		2		2	
6	4		3		3	
4	2	4	4	4.5	4.5	
12	1	5		6		
		6	6	7.5	7.5	
		7		9		
		12		10		

R1 = 4.5+6+7.5+9+10 = 37

R2 = 1+2+3+4.5+7.5= 18

$$U_{1} = n_{1}n_{2} + \frac{n_{1}(n_{1} + 1)}{2} - R_{1} = 5(5) + \frac{5(6)}{2} - 37 = 3$$
$$U_{2} = n_{1}n_{2} + \frac{n_{2}(n_{2} + 1)}{2} - R_{2} = 5(5) + \frac{5(6)}{2} - 18 = 22$$

Test statistic for the Mann Whitney U Test is denoted $U = min(U_1, U_2)$

In R: wilcox.test(data\$placebo,data\$drug)

Alpha = .05 (two-tailed)

n1\ ⁿ²	2	3	4	5	6	7	8	9	10
2							0	0	0
3				0	1	1	2	2	3
4			0	1	2	3	4	4	5
5		0	1	2	3	5	6	7	8
6		1	2	3	5	6	8	10	11
7		1	3	5	6	8	10	12	14
8	0	2	4	6	8	10	13	15	17
9	0	2	4	7	10	12	15	17	20
10	0	3	5	8	11	14	17	20	23

Mann–Whitney U test: an example

Decision for the test: if U < *U*_{crit} we reject H0

We do not reject H0 because 3 > 2.

We do not have statistically significant evidence at α =0.05, to show that the two populations of numbers of episodes of shortness of breath are not equal.

Low power because of small sample sizes ?

```
> t.test(placebo,newdrug)
```

```
Welch Two Sample t-test
```

data: placebo and newdrug

t = 2.199, df = 6.6639, p-value = 0.06574

alternative hypothesis: true difference in means is not equal to 0

Take home message

- How to handle (non-)normality
 - Start by making sure that your data is of the right type
 - If the data spans several orders of magnitude, is made of ratios or is very asymmetrical, consider taking the log.
 - If n is large (>20 or 30), then you should not have to worry.
 - As long as your data is not "too" far from normality (and you are careful with borderline p-values), the t-test will return a meaning full value.
- How to handle (non-)normality: alternatively
 - Perform both a t-test and a Wilcoxon rank-sum test
 - If they return a coherent result, it is stronger that if you performed only one test.
 - If the results are incoherent, look at the data to find out what produced this.