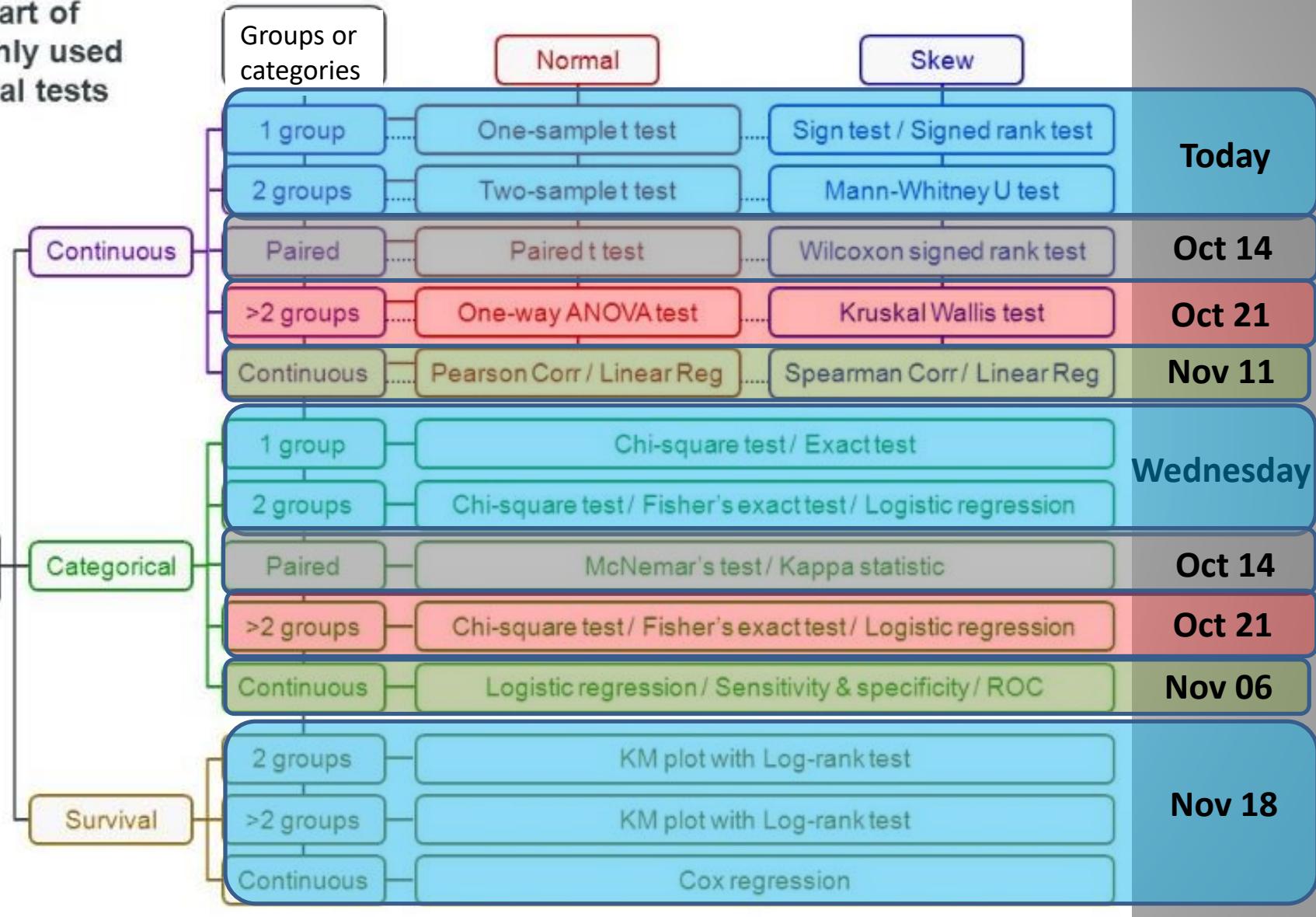


Comparing one or two independent groups of continuous data:

One sample and two sample (AKA independent or unpaired) t-tests,  
Wilcoxon signed rank test, Mann-Whitney U (Wilcoxon rank sum)

Kathleen Torkko  
October 07, 2019

# Flow chart of commonly used statistical tests



# Objectives

Learn to use and interpret parametric and non-parametric tests for testing differences between:

one mean compared to a known mean

one distribution to a known distribution

two independent (unpaired) means

two independent distributions

Learn about the assumptions for each test

## Example

- *Maximum relaxation of bladder muscles in rats:* Frazier, Schneider and Michel (2006) measured how well the neurotransmitter norepinephrine relaxes bladder muscles in rats. Here we will compare the maximal relaxation between old and young rats.
- *Data:*
  - Old: 20.8, 2.8, 50.0, 33.3, 29.4, 38.9, 29.4, 52.6, 14.3.
  - Young: 45.5, 55.0, 60.7, 61.5, 61.1, 65.5, 42.9, 37.5.

# History of the t-test

## How beer created a statistical test



At Guinness Brewery in Ireland until the early 1900s, hops quality were determined by “looks and fragrance” but they wanted to use a more scientific method: proportion of soft resins to hard.

Problem: small number of hops batches to test. The statistical test available at the time, the z-statistic, needed a sample size >30

Also, the z-statistic needs a known population mean and SD for comparison

How could they test the consistency of the resin proportion across small batches of hops and when the population mean and SD are not known?

They turned to William Gosset (mathematician, scientist, and experimental brewer; Employed by Guinness Brewery, Dublin, Ireland, from 1899-1935).



# History of the t-test

Guiness asked Gosset to work on the problem because he had studied a bit of Math at Oxford and was less scared of this kind of work than the other brewers.

He developed the t-test (which used a t-distribution) around 1905 (while on sabbatical working in Karl Pearson's lab) for dealing with small samples in brewing quality control.

Published in 1908 under pseudonym "Student" ("Student's t-test")

## THE PROBABLE ERROR OF A MEAN

BY STUDENT

### Introduction

Any experiment may be regarded as forming an individual of a "population" of experiments which might be performed under the same conditions. A series of experiments is a sample drawn from this population.

Now any series of experiments is only of value in so far as it enables us to form a judgment as to the statistical constants of the population to which the experiments belong. In a greater number of cases the question finally turns on the value of a mean, either directly, or as the mean difference between the two quantities.

If the number of experiments be very large, we may have precise information

The t-test was designed to work in small sample sizes and when the population mean and SD are unknown (otherwise you would use a Z score)



For continuous data that meet assumptions of normality and homoscedasticity:  
t-tests are used to compare ***two*** means

### **Independent t-test**

data collected from two independent groups

Prism calls these “unpaired” t-tests

### **One-sample t-test**

used to compare one mean to a known or standard mean

**For paired data** used the paired t-test – next week

**For >2 groups**, use the ANOVA - which will be covered later

*t-tests are called t-tests because they use the t-statistic and the t-distribution*

# The t Statistic

## Independent t-test

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{N_1} + \frac{s_2^2}{N_2}}}$$

The t statistic allows researchers to use sample data to test hypotheses about an unknown population mean.

The t statistic does not require any knowledge of the population standard deviation and variance.

## One sample t-test

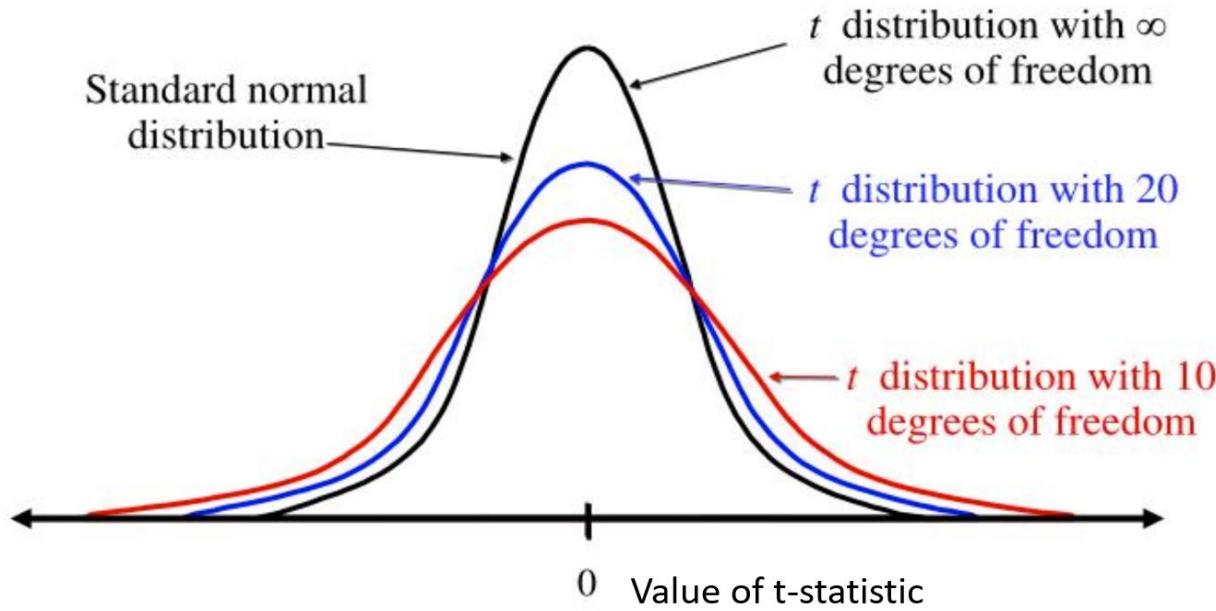
$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$$

Uses *degrees of freedom* (df; calculated from sample sizes) to determine *critical values* for the *t-distribution* for a specific value of df.

$$z = \frac{x - \mu_0}{\sigma}$$

## *t* Distribution

The probability distribution of the t-statistic values



The t-distribution has

- slightly less area near the expected central value
- more area in the "tails" (fatter; helps account for uncertainty with small n)
- a larger spread
- a different shape for different sample sizes

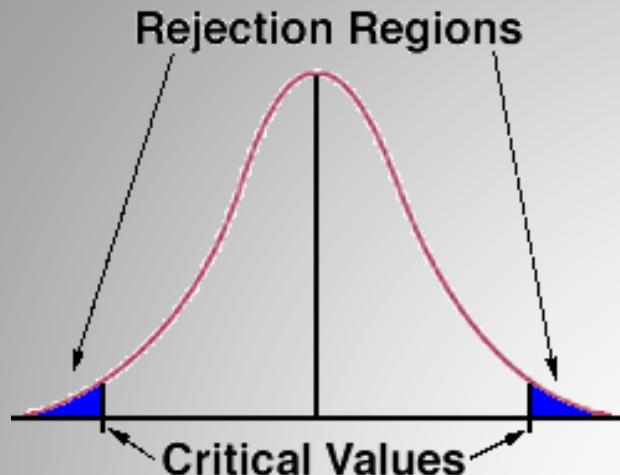
# The t-statistic and the t-distribution

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{N_1} + \frac{s_2^2}{N_2}}}$$

Example: A t-test is performed on two data samples and it calculates a t-statistic value of 2.5. What does that mean?

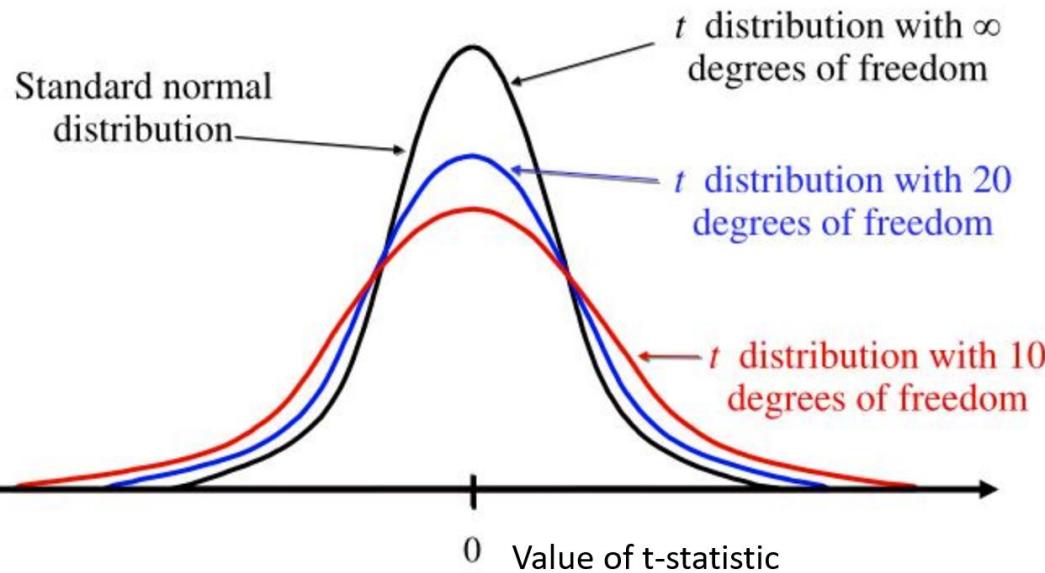
By itself, a t-statistic = 2.5 doesn't mean anything on its own

It needs to be compared to a t-distribution with specific degrees of freedom to determine critical values and *rejection regions*



Rejection regions = reject the null hypothesis

Upper blue area = 2.5% of t-statistic values  
 Lower blue area = 2.5% of t-statistic values



Critical values at  $\alpha=0.05$  for a two-tailed test:

*t* with

2 df	+/- 4.303
10 df	+/- 2.228
20 df	+/- 2.086
120 df	+/- 1.980
$\infty$ df	+/- 1.960

## Critical values of Student's $t$ distribution with $v$ degrees of freedom

Probability less than the critical value ( $t_{1-\alpha, v}$ )

$v$	0.90	0.95	0.975	0.99	0.995	0.999
1.	3.078	6.314	12.706	31.821	63.657	318.313
2.	1.886	2.920	4.303	6.965	9.925	22.327
3.	1.638	2.353	3.182	4.541	5.841	10.215
4.	1.533	2.132	2.776	3.747	4.604	7.173
5.	1.476	2.015	2.571	3.365	4.032	5.893
6.	1.440	1.943	2.447	3.143	3.707	5.208
7.	1.415	1.895	2.365	2.998	3.499	4.782
8.	1.397	1.860	2.306	2.896	3.355	4.499
9.	1.383	1.833	2.262	2.821	3.250	4.296
10.	1.372	1.812	2.228	2.764	3.169	4.143
11.	1.363	1.796	2.201	2.718	3.106	4.024
12.	1.356	1.782	2.179	2.681	3.055	3.929
13.	1.350	1.771	2.160	2.650	3.012	3.852
14.	1.345	1.761	2.145	2.624	2.977	3.787
15.	1.341	1.753	2.131	2.602	2.947	3.733
16.	1.337	1.746	2.120	2.583	2.921	3.686
17.	1.333	1.740	2.110	2.567	2.898	3.646
18.	1.330	1.734	2.101	2.552	2.878	3.610
19.	1.328	1.729	2.093	2.539	2.861	3.579
20.	1.325	1.725	2.086	2.528	2.845	3.552
21.	1.323	1.721	2.080	2.518	2.831	3.527
22.	1.321	1.717	2.074	2.509	2.816	3.505

$$v = df = (n_1 - 1) + (n_2 - 1)$$

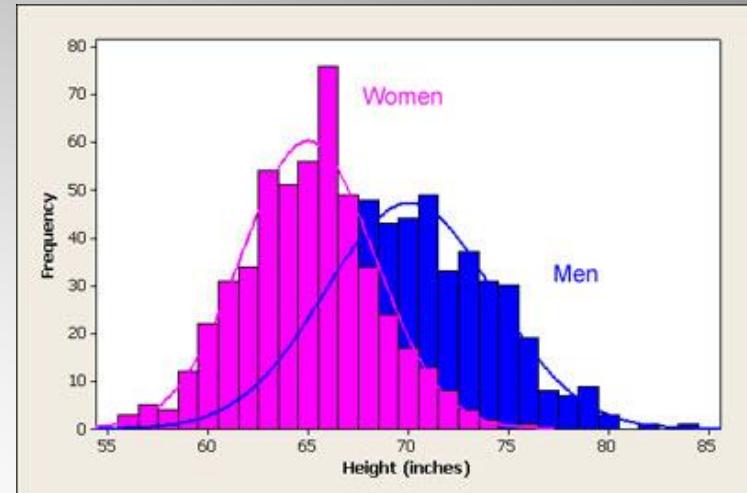
Alpha level for a two-tailed test:  $0.05/2 = 0.025$  or 0.975

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{N_1} + \frac{s_2^2}{N_2}}}$$

# Let's start with the Independent t-test

Continuous variable (not nominal or ordinal)

Can be done only in 2 independent groups



Both samples are random samples from their respective populations

iid: Independent and Identically Distributed random variables

$H_0$ : mean height of women is the same as the mean height of men

$H_A$ : mean height of women is different from the mean height of men

$$H_0: \mu_{\text{women}} = \mu_{\text{men}}$$

$$H_A: \mu_{\text{women}} \neq \mu_{\text{men}}$$

$$H_0: \mu_{\text{women}} - \mu_{\text{men}} = 0$$

$$H_A: \mu_{\text{women}} - \mu_{\text{men}} \neq 0$$

## iid: Independent and Identically Distributed

Example: a coin toss to count the number of heads in ten tosses

Each toss is *independent* since every time you flip a coin, the previous result doesn't influence your current result

They are *identically distributed*, since every time you flip a coin, the chances of getting head (or tail) are identical, no matter if its the 1st or the 100th toss (probability distribution is identical over time). If the coin is "fair" the chances are 0.5 for each event (getting head or tail).

# Assumptions of independent t-tests

## Every test has them

Textbooks usually list the following assumptions:

1. The data must be randomly sampled from normally distributed populations  
Or, the sample data for each group must be normally (symmetrically) distributed
2. The two samples must have equal variances (homoscedasticity).
3. Each observation must be independent from the others.

The third of these is the most important assumption

Determines use of an independent or a paired t-test

Samples from real data rarely meets the first two assumptions

Real populations are rarely truly normally distributed

But the t-test is a *robust* test concerning normality and homoscedasticity violations

## There is no formal definition of a "robust statistical test"

Robust tests are any statistical test that yield good performance when data are mostly unaffected by outliers or have only small departures from model assumptions in a given dataset.

*t*-tests are *robust* to violations of the assumptions of normality and equal variances, particularly in certain circumstances

larger samples sizes (*i.e.*, >30)

numbers are equal in each group (known as a balanced design)  
the data distribution of the groups are similar



## *Testing t-test Assumptions*

**Normality:** Plot frequency histograms and box plots  
for each group of independent samples

Statistical tests (e.g., Shapiro-Wilk) for normality in terms of the null hypothesis (but remember the issues with normality tests)

$H_0$ : Sample is drawn from a **normal** population

$H_A$ : Sample is drawn from a **non-normal** population

**Homoscedasticity:** Scatter plots, compare sample standard deviations between the 2 groups

As a rough estimate, one SD should be no more than twice the other

Do an F-test to formally test for differences (remember problems)

**Independence:** Determine from study design

# The F test for Equal Variances Used by Prism

Divide the larger sample SD by the smaller

$$F = s_1^2 / s_2^2$$

s= the standard deviation of the sample

Degrees of freedom = n-1, calculated for both the numerator and denominator

$H_0$ : variance of sample1 = variance of sample2

$H_A$ : variance of sample1  $\neq$  variance of sample2

EXAMPLE two-tailed test

SD sample 1 = 9.6, n=25

SD sample 2 = 10.9, n=21

$$F = 10.9^2 / 9.6^2 = 118.81 / 92.16 = 1.289$$

To find critical value

Calculate degrees of freedom for the numerator = 21-1 = 20

Calculate degrees of freedom for the denominator = 25-1 = 24

Set Alpha level for a two-tailed test: 0.05/2 = 0.025

IV. Percentage Points of the F Distribution (continued)

		$F_{0.025, v_1, v_2}$														$\alpha = 0.025$				
		Degrees of Freedom for the Numerator ( $v_1$ )																		
$v_1 \backslash v_2$	1	2	3	4	5	6	7	8	9	10	12	15	20	24	30	40	60	120	$\infty$	
1	647.8	799.5	864.2	899.6	921.8	937.1	948.2	956.7	963.3	968.6	976.7	984.9	993.1	997.2	1001	1006	1010	1014	1018	
2	38.51	39.00	39.17	39.25	39.30	39.33	39.36	39.37	39.39	39.40	39.41	39.43	39.45	39.46	39.46	39.47	39.48	39.49	39.50	
3	17.44	16.04	15.44	15.10	14.88	14.73	14.62	14.54	14.47	14.42	14.34	14.25	14.17	14.12	14.08	14.04	13.99	13.95	13.90	
4	12.22	10.65	9.98	9.60	9.36	9.20	9.07	8.98	8.90	8.84	8.75	8.66	8.56	8.51	8.46	8.41	8.36	8.31	8.26	
5	10.01	8.43	7.76	7.39	7.15	6.98	6.85	6.76	6.68	6.62	6.52	6.43	6.33	6.28	6.23	6.18	6.12	6.07	6.02	
6	8.81	7.26	6.60	6.23	5.99	5.82	5.70	5.60	5.52	5.46	5.37	5.27	5.17	5.12	5.07	5.01	4.96	4.90	4.85	
7	8.07	6.54	5.89	5.52	5.29	5.12	4.99	4.90	4.82	4.76	4.67	4.57	4.47	4.42	4.36	4.31	4.25	4.20	4.14	
8	7.57	6.06	5.42	5.05	4.82	4.65	4.53	4.43	4.36	4.30	4.20	4.10	4.00	3.95	3.89	3.84	3.78	3.73	3.67	
9	7.21	5.71	5.08	4.72	4.48	4.32	4.20	4.10	4.03	3.96	3.87	3.77	3.67	3.61	3.56	3.51	3.45	3.39	3.33	
10	6.94	5.46	4.83	4.47	4.24	4.07	3.95	3.85	3.78	3.72	3.62	3.52	3.42	3.37	3.31	3.26	3.20	3.14	3.08	
11	6.72	5.26	4.63	4.28	4.04	3.88	3.76	3.66	3.59	3.53	3.43	3.33	3.23	3.17	3.12	3.06	3.00	2.94	2.88	
12	6.55	5.10	4.47	4.12	3.89	3.73	3.61	3.51	3.44	3.37	3.28	3.18	3.07	3.02	2.96	2.91	2.85	2.79	2.72	
13	6.41	4.97	4.35	4.00	3.77	3.60	3.48	3.39	3.31	3.25	3.15	3.05	2.95	2.89	2.84	2.78	2.72	2.66	2.60	
14	6.30	4.86	4.24	3.89	3.66	3.50	3.38	3.29	3.21	3.15	3.05	2.95	2.84	2.79	2.73	2.67	2.61	2.55	2.49	
15	6.20	4.77	4.15	3.80	3.58	3.41	3.29	3.20	3.12	3.06	2.96	2.86	2.76	2.70	2.64	2.59	2.52	2.46	2.40	
16	6.12	4.69	4.08	3.73	3.50	3.34	3.22	3.12	3.05	2.99	2.89	2.79	2.68	2.63	2.57	2.51	2.45	2.38	2.32	
17	6.04	4.62	4.01	3.66	3.44	3.28	3.16	3.06	2.98	2.92	2.82	2.72	2.62	2.56	2.50	2.44	2.38	2.32	2.25	
18	5.98	4.56	3.95	3.61	3.38	3.22	3.10	3.01	2.93	2.87	2.77	2.67	2.56	2.50	2.44	2.38	2.32	2.26	2.19	
19	5.92	4.51	3.90	3.56	3.33	3.17	3.05	2.96	2.88	2.82	2.72	2.62	2.51	2.45	2.39	2.33	2.27	2.20	2.13	
20	5.87	4.46	3.86	3.51	3.29	3.13	3.01	2.91	2.84	2.77	2.68	2.57	2.46	2.41	2.35	2.29	2.22	2.16	2.09	
21	5.83	4.42	3.82	3.48	3.25	3.09	2.97	2.87	2.80	2.73	2.64	2.53	2.42	2.37	2.31	2.25	2.18	2.11	2.04	
22	5.79	4.38	3.78	3.44	3.22	3.05	2.93	2.84	2.76	2.70	2.60	2.50	2.39	2.33	2.27	2.21	2.14	2.08	2.00	
23	5.75	4.35	3.75	3.41	3.18	3.02	2.90	2.81	2.73	2.67	2.57	2.47	2.36	2.30	2.24	2.18	2.11	2.04	1.97	
24	5.72	4.32	3.72	3.38	3.15	2.99	2.87	2.78	2.70	2.64	2.54	2.44	2.33	2.27	2.21	2.15	2.08	2.01	1.94	
25	5.69	4.29	3.69	3.35	3.13	2.97	2.85	2.75	2.68	2.61	2.51	2.41	2.30	2.24	2.18	2.12	2.05	1.98	1.91	
26	5.66	4.27	3.67	3.33	3.10	2.94	2.82	2.73	2.65	2.59	2.49	2.39	2.28	2.22	2.16	2.09	2.03	1.95	1.88	
27	5.63	4.24	3.65	3.31	3.08	2.92	2.80	2.71	2.63	2.57	2.47	2.36	2.25	2.19	2.13	2.07	2.00	1.93	1.85	
28	5.61	4.22	3.63	3.29	3.06	2.90	2.78	2.69	2.61	2.55	2.45	2.34	2.25	2.17	2.11	2.05	1.98	1.91	1.83	
29	5.59	4.20	3.61	3.27	3.04	2.88	2.76	2.67	2.59	2.53	2.43	2.32	2.21	2.15	2.09	2.03	1.96	1.89	1.81	
30	5.57	4.18	3.59	3.25	3.03	2.87	2.75	2.65	2.57	2.51	2.41	2.31	2.20	2.14	2.07	2.01	1.94	1.87	1.79	
40	5.42	4.05	3.46	3.13	2.90	2.74	2.62	2.53	2.45	2.39	2.29	2.18	2.07	2.01	1.94	1.88	1.80	1.72	1.64	
60	5.29	3.93	3.34	3.01	2.79	2.63	2.51	2.41	2.33	2.27	2.17	2.06	1.94	1.88	1.82	1.74	1.67	1.58	1.48	
120	5.15	3.80	3.23	2.89	2.67	2.52	2.39	2.30	2.22	2.16	2.05	1.94	1.82	1.76	1.69	1.61	1.53	1.43	1.31	
$\infty$	5.02	3.69	3.12	2.79	2.57	2.41	2.29	2.19	2.11	2.05	1.94	1.83	1.71	1.64	1.57	1.48	1.39	1.27	1.00	

11

## Results of Our Example

$$F = 10.9^2 / 9.6^2 = 118.81 / 92.16 = 1.29$$

Critical value from the table = 2.41

Because  $1.29 < 2.41$ , then  $p > 0.05$  (if greater than critical value,  $p \leq 0.05$ )

Fail to reject the null and accept that the variances are the same

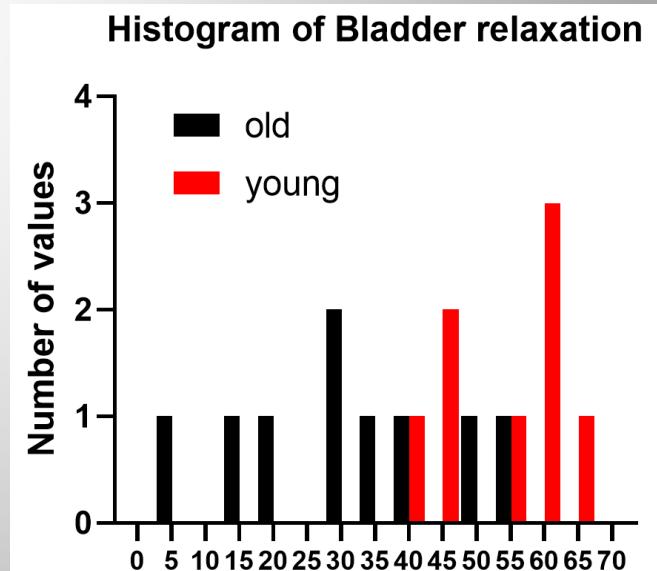
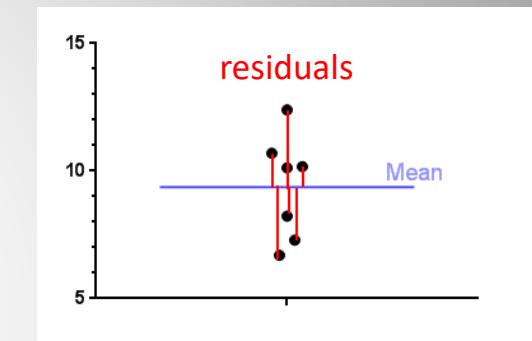
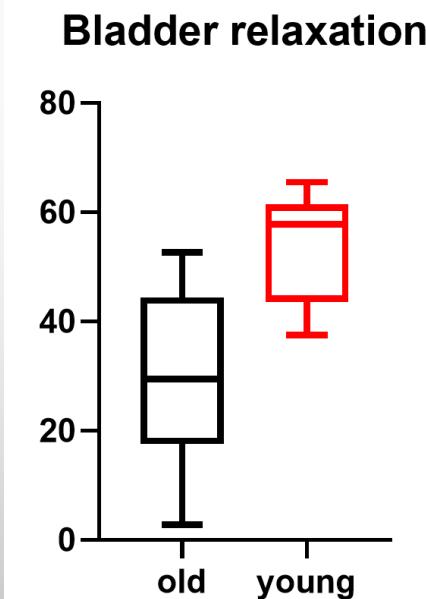
# Example using bladder relaxation data in old and young rats

$H_0$ : the means are equal

$H_A$ : the means are not equal

Normality?  
Homoscedasticity?  
Independent?

	A	B
	old	young
1 Number of values	9	8
2		
3 Minimum	2.800	37.50
4 25% Percentile	17.55	43.55
5 Median	29.40	57.85
6 75% Percentile	44.45	61.40
7 Maximum	52.60	65.50
8 Range	49.80	28.00
9		
10 Mean	30.17	53.71
11 Std. Deviation	16.09	10.36
12 Std. Error of Mean	5.365	3.664
13		
14 Skewness	-0.2130	-0.5515
15 Kurtosis	-0.3964	-1.463
16		



# Example using bladder relaxation data in old and young rats

Welcome to GraphPad Prism

GraphPad Prism Version 8.2.1 (441)

New table & graph

XY  
Column  
Grouped  
Contingency  
Survival  
Parts of whole  
Multiple variables  
Nested

Existing file  
Open a file  
LabArchives  
Clone a graph  
Graph portfolio

Prism Tips

Column tables have one grouping variable, with each group defined by a column

Data table:

Control Treated

Y Y

1 2

Control Treated

Learn more

Data table:  
 Enter or import data into a new table  
 Start with sample data to follow a tutorial

Options:  
 Enter replicate values, stacked into columns  
 Enter paired or repeated measures data - each subject on a separate row  
 Enter and plot error values already calculated elsewhere

Enter: Mean, SD, N

Project1:Data 1 - GraphPad Prism 8.2.1 (441)

File Edit View Insert Change Arrange Family Window Help

Prism File Sheet Undo Clipboard Analysis Change Import Draw

Search...

Group A Group B Group C

old young Title

	old	young	Title
1	20.8	45.5	
2	2.8	55.0	
3	50.0	60.7	
4	33.3	61.5	
5	29.4	61.1	
6	38.9	65.5	
7	29.4	42.9	
8	52.6	37.5	
9		14.3	
10			
11			
12			
13			
14			

Bladder relaxation

New Data Table...

Project info 1

New Info...

New Analysis...

Bladder relaxation

New Graph...

Layouts

Analyze

# T-test for differences in independent groups of continuous data

## With F Test for Equal Variances (Homoscedasticity)

### Unpaired (independent) t-test

Analyze Data

Built-in analysis

Which analysis?

- Transform, Normalize...**
  - Transform
  - Transform concentrations (X)
  - Normalize
  - Prune rows
  - Remove baseline and column math
  - Transpose X and Y
  - Fraction of total
- XY analyses**
- Column analyses**
  - t tests (and nonparametric tests)**
    - One-way ANOVA (and nonparametric or mixed)
    - One sample t and Wilcoxon test
    - Descriptive statistics
    - Normality and Lognormality Tests
    - Frequency distribution
    - ROC Curve
    - Bland-Altman method comparison
    - Identify outliers
    - Analyze a stack of P values
- Grouped analyses**
- Contingency table analyses**
- Survival analyses**
- Parts of whole analyses**

Analyze which data sets?

- A:old
- B:young

Select All   Deselect All

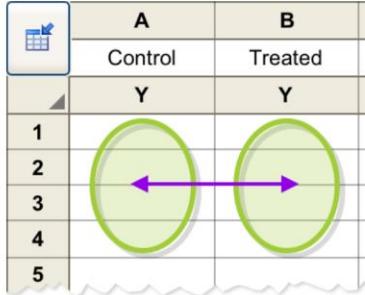
Help   Cancel   OK

Parameters: t tests (and Nonparametric Tests)

Experimental Design   Residuals   Options

**Experimental design**

Unpaired  
 Paired



**Assume Gaussian distribution?**

Yes. Use parametric test.  
 No. Use nonparametric test.

**Choose test**

Unpaired t test. Assume both populations have the same SD  
 Unpaired t test with Welch's correction. Do not assume equal SDs

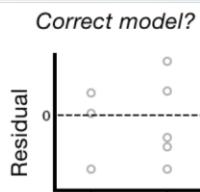
Learn   Cancel   OK

## Parameters: t tests (and Nonparametric Tests)

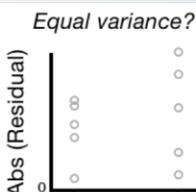
Experimental Design Residuals Options

### What graphs to create?

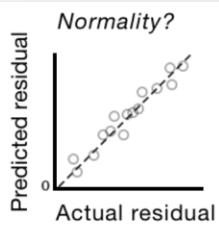
Residual plot



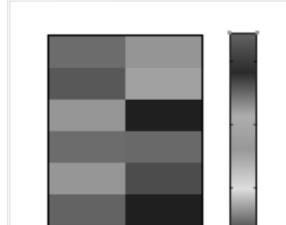
Homoscedasticity plot



QQ plot



Heatmap plot



### Diagnostics for residuals

Are the residuals Gaussian?

Normality tests of Anderson-Darling, D'Agostino, Shapiro-Wilk and Kolmogorov-Smirnov.

Make options on this tab be the default for future tests.

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One of the assumptions of t-tests is that the *residuals* from a Gaussian distribution.

## Which residual graph to use?

**Residual plot.** The X axis is the actual value of the value (unpaired tests) or difference (paired test). The Y axis is the residual. This lets you spot residuals that are much larger or smaller than the rest.

**Homoscedasticity plot.** The X axis is the actual value of the value (unpaired tests) or difference (paired test). The Y axis is the absolute value of the residual.

**QQ plot.** The X axis is the actual residual. The Y axis is the predicted residual, computed from the percentile of the residual (among all residuals) and assuming sampling from a Gaussian distribution. Most useful for t-tests

## Parameters: t tests (and Nonparametric Tests)

X

Experimental Design Residuals Options

**Calculations**P value:  One-tailed  Two-tailed (recommended)

Report differences as: young - old

Confidence level: 95% ▾

Definition of statistical significance: P &lt; 0.05

**Graphing options** Graph differences (paired) Graph ranks (nonparametric) Graph correlation (paired) Graph CI of difference between means**Additional results** Descriptive statistics for each data set t test: Also compare models using AICc Mann-Whitney: Also compute the CI of difference between medians

Assumes both distributions have the same shape.

 Wilcoxon: When both values on a row are identical, use method of Pratt

If this option is unchecked, those rows are ignored and the results will match prior version of Prism

**Output**

Show this many significant digits (for everything except P values): 4 ▾

P value style: GP: 0.1234 (ns), 0.0332 (\*), 0.0021 (\*\*), 0.00 ▾ N = 6 ▾

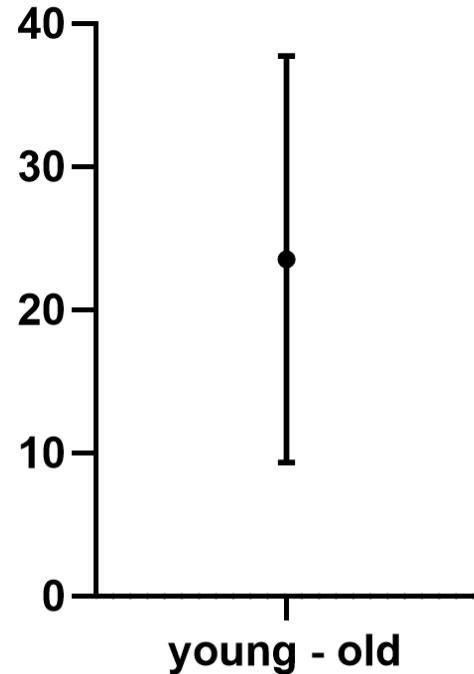
 Make options on this tab be the default for future tests.

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# Difference between means



## Unpaired t test Tabular results

2	
3	Column B
4	vs.
5	Column A
6	

### Unpaired t test

8	P value	0.0030
9	P value summary	**
10	Significantly different ( $P < 0.05$ )?	Yes
11	One- or two-tailed P value?	Two-tailed
12	t, df	t=3.531, df=15

### How big is the difference?

15	Mean of column A	30.17
16	Mean of column B	53.71
17	Difference between means (B - A) $\pm$ SEM	23.55 $\pm$ 6.667
18	95% confidence interval	9.335 to 37.76
19	R squared (eta squared)	0.4540

### F test to compare variances

22	F, DFn, Dfd	2.412, 8, 7
23	P value	0.2631
24	P value summary	ns
25	Significantly different ( $P < 0.05$ )?	No

Prism, unlike most statistics programs, reports a  $R^2$  value. It quantifies the fraction of all the variation in the samples that is accounted for by a difference between the group means. If  $R^2=0.36$ , that means that 36% of all the variation among values is attributed to differences between the two group means, leaving 64% of the variation that comes from scatter among values within the groups.

## Critical values of Student's $t$ distribution with $v$ degrees of freedom

Probability less than the critical value ( $t_{1-\alpha, v}$ )

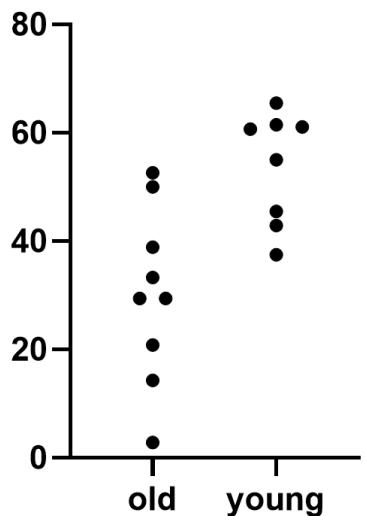
$v$	0.90	0.95	0.975	0.99	0.995	0.999
1.	3.078	6.314	12.706	31.821	63.657	318.313
2.	1.886	2.920	4.303	6.965	9.925	22.327
3.	1.638	2.353	3.182	4.541	5.841	10.215
4.	1.533	2.132	2.776	3.747	4.604	7.173
5.	1.476	2.015	2.571	3.365	4.032	5.893
6.	1.440	1.943	2.447	3.143	3.707	5.208
7.	1.415	1.895	2.365	2.998	3.499	4.782
8.	1.397	1.860	2.306	2.896	3.355	4.499
9.	1.383	1.833	2.262	2.821	3.250	4.296
10.	1.372	1.812	2.228	2.764	3.169	4.143
11.	1.363	1.796	2.201	2.718	3.106	4.024
12.	1.356	1.782	2.179	2.681	3.055	3.929
13.	1.350	1.771	2.160	2.650	3.012	3.852
14.	1.345	1.761	2.145	2.624	2.977	3.787
15.	1.341	1.753	2.131	2.602	2.947	3.733
16.	1.337	1.746	2.120	2.583	2.921	3.686
17.	1.333	1.740	2.110	2.567	2.898	3.646
18.	1.330	1.734	2.101	2.552	2.878	3.610
19.	1.328	1.729	2.093	2.539	2.861	3.579
20.	1.325	1.725	2.086	2.528	2.845	3.552
21.	1.323	1.721	2.080	2.518	2.831	3.527
22.	1.321	1.717	2.074	2.509	2.816	3.505

$$v = df = (n_1 - 1) + (n_2 - 1)$$

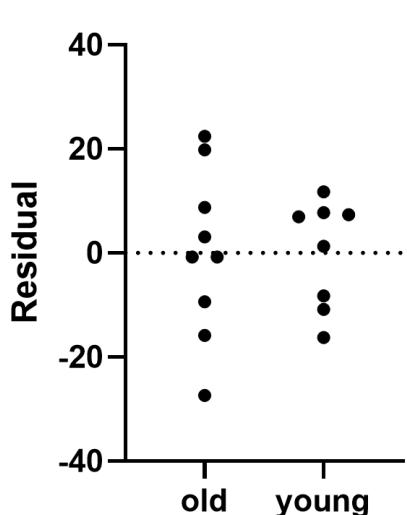
Alpha level for a two-tailed test:  $0.05/2 = 0.025$  or 0.975

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{N_1} + \frac{s_2^2}{N_2}}}$$

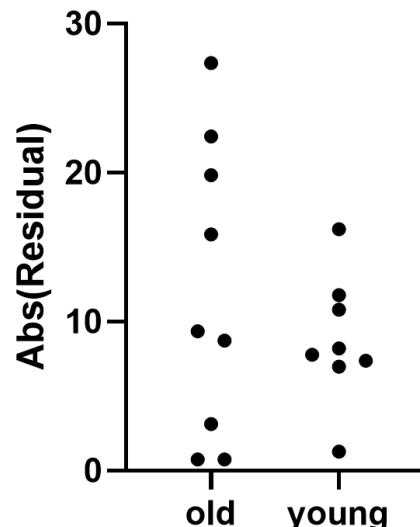
**Scatter Plot**  
**Bladder relaxation**



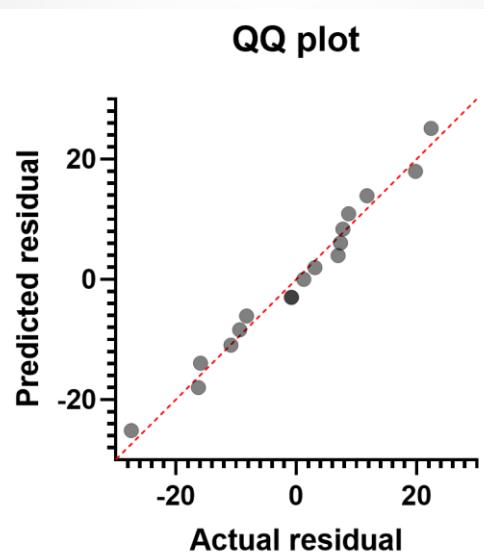
**Residual plot**



**Homoscedasticity plot**



**QQ plot**



# For unequal variances (heteroscedascity)

If the variances are too different, i.e.  
there is heteroscedasticity

Parameters: t tests (and Nonparametric Tests)

Experimental Design   Residuals   Options

**Experimental design**

Unpaired  
 Paired

	A	B
	Control	Treated
Y	Y	Y
1		
2		
3		
4		
5		

**Assume Gaussian distribution?**

Yes. Use parametric test.  
 No. Use nonparametric test.

**Choose test**

Unpaired t test. Assume both populations have the same SD  
 Unpaired t test with Welch's correction. Do not assume equal SDs

Learn   Cancel   OK

Unpaired t test	
Tabular results	
	1 Table Analyzed
2	
3 Column B	young
4 vs.	vs.
5 Column A	old
6	
7 Unpaired t test	
8 P value	0.0030
9 P value summary	**
10 Significantly different ( $P < 0.05$ )?	Yes
11 One- or two-tailed P value?	Two-tailed
12 t, df	t=3.531, df=15
13	
14 How big is the difference?	
15 Mean of column A	30.17
16 Mean of column B	53.71
17 Difference between means (B - A) $\pm$ SEM	23.55 $\pm$ 6.667
18 95% confidence interval	9.335 to 37.76
19 R squared (eta squared)	0.4540
20	
21 F test to compare variances	
22 F, DFn, Dfd	2.412, 8, 7
23 P value	0.2631
24 P value summary	ns
25 Significantly different ( $P < 0.05$ )?	No
26	
Bladder relaxation	
3 Column B	young
4 vs.	vs.
5 Column A	old
6	
7 Unpaired t test with Welch's correction	
8 P value	0.0028
9 P value summary	**
10 Significantly different ( $P < 0.05$ )?	Yes
11 One- or two-tailed P value?	Two-tailed
12 Welch-corrected t, df	t=3.624, df=13.78
13	
14 How big is the difference?	
15 Mean of column A	30.17
16 Mean of column B	53.71
17 Difference between means (B - A) $\pm$ SEM	23.55 $\pm$ 6.497
18 95% confidence interval	9.591 to 37.50
19 R squared (eta squared)	0.4881
20	
21 F test to compare variances	
22 F, DFn, Dfd	2.412, 8, 7
23 P value	0.2631
24 P value summary	ns
25 Significantly different ( $P < 0.05$ )?	No

What happens if your data violate the independent t test assumptions? (hint: biased results)

Assumption of independence for the sample values is violated  
The independent (unpaired) t-test is not appropriate.

Assumption of normality is violated, or outliers are present  
The t-test may not be the most powerful test available  
A non-parametric test or employing a transformation may result in a less biased answer

Assumption of homoscedasticity is violated  
The Welch-Satterthwaite t-test provides means of performing a t-test adjusted for the inequality of the variances  
  
Or transform the data or use a non-parametric test

## Example: The Independent t-test by hand

Is there a difference in Test X by Gender?

Males	Females
$\bar{X}_1 = 22.5$	$\bar{X}_2 = 25.6$
$SD_1 = 2.5$	$SD_2 = 3.0$
$N_1 = 9$	$N_2 = 9$

For the purposes of this example, we have already determined that all test assumptions have been met.

$H_0$ : the means are equal

$H_A$ : the means are not equal

# Determining Critical Value

$$df = (n_1 - 1) + (n_2 - 1)$$

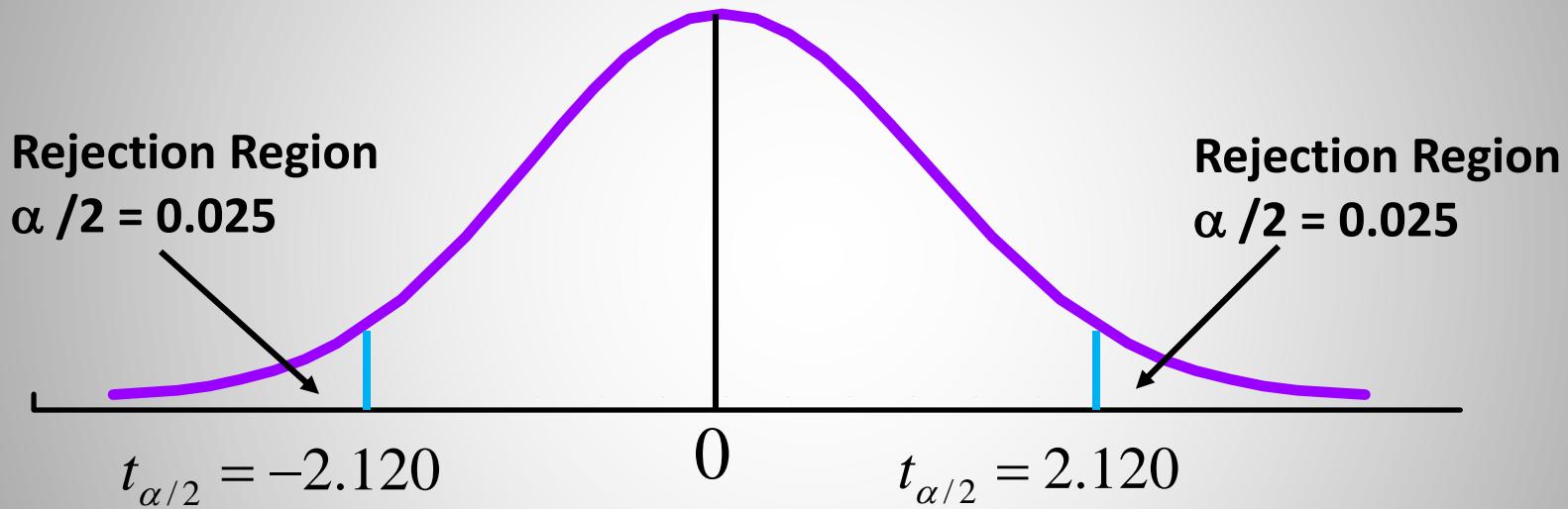
$$df = (9 - 1) + (9 - 1) = 16$$

	0.1	0.05	0.025	0.01	0.005
1 tail $\alpha$ =	0.2	0.1	0.05	0.02	0.01
2 tails $\alpha$ =					
df = 1	3.078	6.314	12.706	31.821	63.656
2	1.886	2.920	4.303	6.965	9.925
3	1.638	2.353	3.182	4.541	5.841
4	1.533	2.132	2.776	3.747	4.604
5	1.476	2.015	2.571	3.365	4.032
6	1.440	1.943	2.447	3.143	3.707
7	1.415	1.895	2.365	2.998	3.499
8	1.397	1.860	2.306	2.896	3.355
9	1.383	1.833	2.262	2.821	3.250
10	1.372	1.812	2.228	2.764	3.169
11	1.363	1.796	2.201	2.718	3.106
12	1.356	1.782	2.179	2.681	3.055
13	1.350	1.771	2.160	2.650	3.012
14	1.345	1.761	2.145	2.624	2.977
15	1.341	1.753	2.131	2.602	2.947
16	1.337	1.746	2.120	2.583	2.921
17	1.333	1.740	2.110	2.567	2.898
18	1.330	1.734	2.101	2.552	2.878
19	1.328	1.729	2.093	2.539	2.861
20	1.325	1.725	2.086	2.528	2.845

## Hypothesis Tests for Two Means (two-tailed test, df=16)

$H_0$ : the means are equal

$H_A$ : the means are not equal



If  $-2.120 \leq t \leq 2.120$ , fail to reject  $H_0$

## Example: The Independent t-test by hand

Is there a difference in Test X by Gender?

Males	Females
$\bar{X}_1 = 22.5$	$\bar{X}_2 = 25.6$
$SD_1 = 2.5$	$SD_2 = 3.0$
$N_1 = 9$	$N_2 = 9$

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{N_1} + \frac{s_2^2}{N_2}}}$$

$$\bar{x}_1 - \bar{x}_2 = 22.5 - 25.6 = -3.1$$

$$s_1^2 = 2.5^2 = 6.25 \quad s_2^2 = 3.0^2 = 9$$

$$6.25/9 = 0.69 \quad 9/9 = 1$$

$$\sqrt{0.69 + 1} = \sqrt{1.69} = 1.3$$

$$-3.1/1.3 = -2.38 = \text{t statistic}$$

$$df = (9-1) + (9-1) = 16$$

Critical value of  $t(16) = (+/-) 2.120$  (two-tailed test)

$-2.38 < -2.120$ , then  $p < 0.05$  and we reject the null hypothesis of no differences between groups.

## What about confidence intervals?

95% confidence intervals for the independent (unpaired) t-test  
(Confidence intervals for the difference in means)

$$\text{Lower bound: } (\bar{X}_1 - \bar{X}_2) - t_{\alpha/2} * \sqrt{\frac{s_1^2}{n_1} + \frac{s_{12}^2}{n_2}}$$

$$\text{Upper bound: } (\bar{X}_1 - \bar{X}_2) + t_{\alpha/2} * \sqrt{\frac{s_1^2}{n_1} + \frac{s_{12}^2}{n_2}}$$

# Independent t-test Example with CI: Effect of Treatment on Test Y

Placebo      Treatment

18	22
21	25
16	17
22	24
19	16
24	29
17	20
21	23
23	19
18	20
14	15
16	15
16	18
19	26
18	18
20	24
12	18
22	25
15	19
17	16

$H_0$ : the means are equal

$H_A$ : the means are not equal

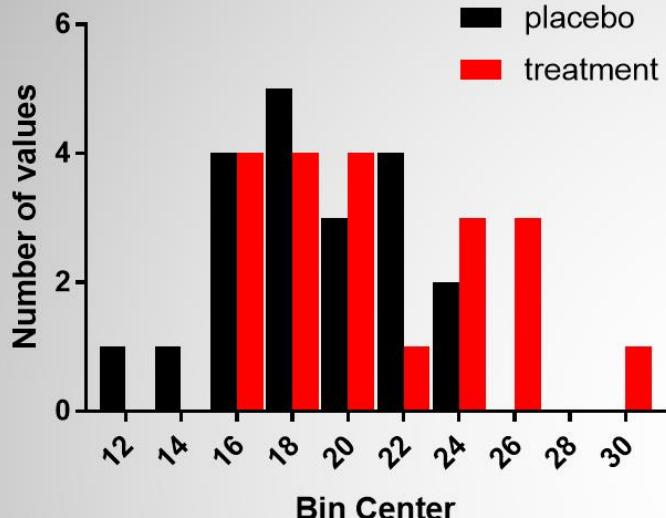
Assumption of Independence

The observations are independent by treatment. The two groups are different people.

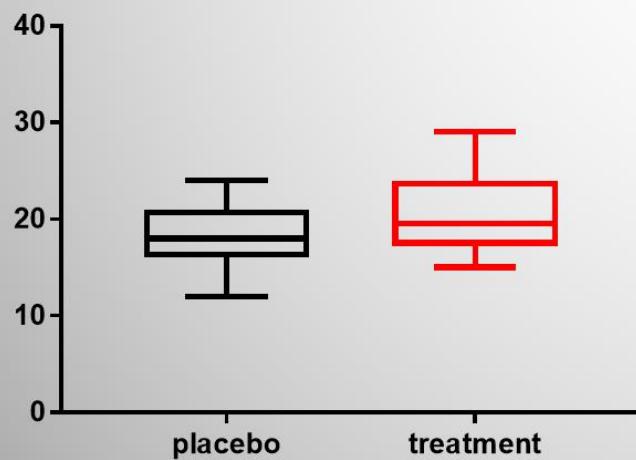
Mean	18.40	20.45
N	20	20

# Testing normality (symmetry)

Histogram of Placebo vs Treatment



Placebo vs Treatment



	placebo	treatment	
1	Number of values	20	20
2			
3	Minimum	12	15
4	25% Percentile	16	17.25
5	Median	18	19.5
6	75% Percentile	21	24
7	Maximum	24	29
8			
9	Mean	18.4	20.45
10	Std. Deviation	3.152	4.058
11	Std. Error of Mean	0.7049	0.9075
12			
13	Skewness	-0.05198	0.4398
14	Kurtosis	-0.4966	-0.7904
15			
16	D'Agostino & Pearson normality test		
17	K2	0.1646	1.545
18	P value	0.9210	0.4619
19	Passed normality test (alpha=0.05)?	Yes	Yes
20	P value summary	ns	ns
21			

	Col. stats	A	B
		placebo	treatment
1	Number of values	20	20
2			
3	Minimum	12.00	15.00
4	25% Percentile	16.00	17.25
5	Median	18.00	19.50
6	75% Percentile	21.00	24.00
7	Maximum	24.00	29.00
8			
9	Mean	18.40	20.45
10	Std. Deviation	3.152	4.058
11	Std. Error of Mean	0.7049	0.9075
12			
13	Skewness	-0.05198	0.4398
14	Kurtosis	-0.4966	-0.7904
15			
16	D'Agostino & Pearson normality test		
17	K2	0.1646	1.545
18	P value	0.9210	0.4619
19	Passed normality test (alpha=0.05)?	Yes	Yes
20	P value summary	ns	ns
21			
22	Shapiro-Wilk normality test		
23	W	0.9820	0.9423
24	P value	0.9569	0.2654
25	Passed normality test (alpha=0.05)?	Yes	Yes
26	P value summary	ns	ns

## Testing Homoscedasticity

SD  
 Treatment SD 4.058  
 Placebo 3.152  
*Ratio* 1.28  
*Rule of thumb* <2

### F test

--		
21	F test to compare variances	
22	F,DFn, Dfd	1.658, 19, 19
23	P value	0.2795
24	P value summary	ns
25	Significantly different? (P < 0.05)	No
26		
27		

F test p = 0.28, fail to reject the null of no differences in variances and conclude the variances are equal

Normal/Symmetrical	Yes
Equal Variances	Yes
Independent	Yes
2 groups of continuous data	Yes



Independent t-test

Unpaired t test	
Tabular results	
1 Table Analyzed	Placebo vs Treatment
2	
3 Column B	treatment
4 vs.	vs.
5 Column A	placebo
6	
7 Unpaired t test	
8 P value	0.0824
9 P value summary	ns
10 Significantly different ( $P < 0.05$ )?	No
11 One- or two-tailed P value?	Two-tailed
12 t, df	$t=1.784$ , df=38
13	
14 How big is the difference?	
15 Mean of column A	18.40
16 Mean of column B	20.45
17 Difference between means (B - A) $\pm$ SEM	$2.050 \pm 1.149$
18 95% confidence interval	-0.2762 to 4.376
19 R squared (eta squared)	0.07728
20	
21 F test to compare variances	
22 F, DFn, Dfd	1.658, 19, 19
23 P value	0.2795
24 P value summary	ns
25 Significantly different ( $P < 0.05$ )?	No
26	

## WRITE UP RESULTS

For this class:

The data for both groups passed the assumptions for an independent t-test (independence, normality/symmetry, homoscedasticity). With  $p=0.08$  (two tailed independent t-test,  $t=1.78$ ,  $df=38$ ,  $\alpha=0.05$ ), we fail to reject the null hypothesis of no difference in mean values of Test Y in placebo (18.4) and treatment groups (20.5). Further, the 95%CI for the difference in means contained 0 (-0.28 to 4.38). Therefore, we conclude treatment has no effect on the value of Test Y.

What people may usually write:

We conclude treatment has no effect on the value of Test Y ( $p=0.08$ ).

what is the type of the test and what is the result

## **Box 2. Checklist for clear statistical reporting for t-tests and ANOVAs**

Basic biomedical science studies often include several small experiments. Providing information about statistical tests in the legend of each table or figure makes it easy for readers to determine what test was performed for each set of data and confirm that the test is appropriate.

### **t-tests**

- # State whether the test was unpaired (for comparing independent groups) or paired (for non-independent data, including repeated measurements on the same individual or matched participants, specimens or samples).
- # State whether the test assumed equal or unequal variance between groups.
- # Report the t-statistic, degrees of freedom and exact p-value.
- # To focus on the magnitude of the difference, it is strongly recommended to report effect sizes with confidence intervals.

### **ANOVAs**

- # Specify the number of factors included in the ANOVA (i.e., one- vs. two-way ANOVA).
- # For each factor, specify the name and level of the factor and state whether the factor was entered as a within-subjects (i.e., independent) factor or as a between-subjects (i.e., non-independent) factor.
- # If the ANOVA has two or more factors, specify whether the interaction term was included. If the ANOVA has three or more factors and includes interaction terms, specify which interaction terms were included.
- # Report the F-statistic, degrees of freedom and exact p-value for each factor or interaction.
- # Specify if a post-hoc test was performed. If post-hoc tests were performed, specify the type of post-hoc test and, if applicable, the test statistic and p-value.
- # To focus on the magnitude of the difference, it is strongly recommended to report effect sizes with confidence intervals.

DOI: <https://doi.org/10.7554/eLife.36163.005>

## How do you interpret the confidence intervals?

Confidence intervals for the independent t-test (difference in mean)

the confidence interval is a range of likely values for the difference in population means

the interval either contains the population value or it does not

How big is the difference?	
Mean of column A	18.40
Mean of column B	20.45
Difference between means (B - A) $\pm$ SEM	2.050 $\pm$ 1.149
95% confidence interval	-0.2762 to 4.376

Although our estimate of the difference in means is 2.05, because the 95%CI includes the null value (0; no difference), we conclude there is no difference in groups (based on our sample)

## What about the *non-parametric* tests?

Nonparametric tests make no assumptions about the distribution of the populations

Use **ranks** of the data values from low to high and analyzes the value of the ranks, ignoring the actual values of the data

They can be useful when:

Sample sizes are small and data distribution cannot be determined

Sample sizes are larger, but the data severely violate the normality assumption of standard statistical tests, e.g., data are heavily skewed or have “outliers”

Nonparametric tests can still depend on other assumptions  
e.g., Data are independent

# Properties of parametric and non-parametric datasets

	<b>Parametric</b>	<b>Non-parametric</b>
Distribution	Assumed to be Gaussian (normal)	Any
Variance	Assumed to be homoscedastic	Any
Data type	Continuous	Any also for ordinal data
Central measure	Mean	Distribution (or median in specific circumstances)

**But, use parametric tests where possible!**

Parametric tests usually have a higher statistical power than non-parametric tests and are thus likelier to avoid a false negative result (type 2 error)

Choices of non-parametric tests are limited: Compared to parametric methods, there are only a few non-parametric tests

The information in the data is not fully utilized in a non-parametric test

Confidence intervals from parametric tests are easier to interpret

## What Happens If the Wrong Type of Test Is Used?

When parametric test assumptions are met

parametric methods will have more power to detect differences than non-parametric methods

When parametric test assumptions are violated

parametric methods will lack power

analyzing data not normally distributed with parametric tests will increase Type II error (concluding no difference when there is one)

## What if I choose the ‘wrong’ test? Relationship to sample size

Distribution	Test	Small Sample	Large Sample
Gaussian	Nonparametric	Nonparametric tests may have much lower power than parametric tests to detect differences with small sample sizes. P values will be too large.	Little problem. With large samples, nonparametric tests are nearly as powerful as parametric tests. P values will be very slightly larger.
Non-Gaussian	Parametric	Misleading. With small sample sizes, parametric tests are not very robust to violations of the normality assumption. P values may be inaccurate.	Little problem. With large sample sizes, parametric tests are robust to violations of the normality assumption.

## Some notes...

Don't do both parametric and non-parametric tests and pick up the result with the smaller p-value

Don't refer to the data as "non-parametric". Data are not "parametric" or "non-parametric", the method or the test is

# Some commonly used parametric tests and their non-parametric equivalents

They do not compare the median, but the distribution

	<b>Parametric test</b>	<b>Non-parametric test</b>
Correlation	Pearson	Spearman
One sample, Compare to another	One-sample t-test	Wilcoxon signed rank test
Two groups, independent data	Independent t –test	Mann-Whitney <i>U</i> test (the same as the Wilcoxon rank-sum test)
Paired data	Paired t-test	Sign test, Wilcoxon signed rank test
>two groups, independent data	One-way analysis of variance (ANOVA)	Kruskal-Wallis test

## Independent t-test non-parametric equivalent: Mann-Whitney U (Wilcoxon rank sum)

the Mann-Whitney test does not compare medians of the two groups  
but the distribution and location of ranks of the data in each group

Data are grouped together and ranked from smallest to largest  
Ranks for each group are summed and adjusted for sample size

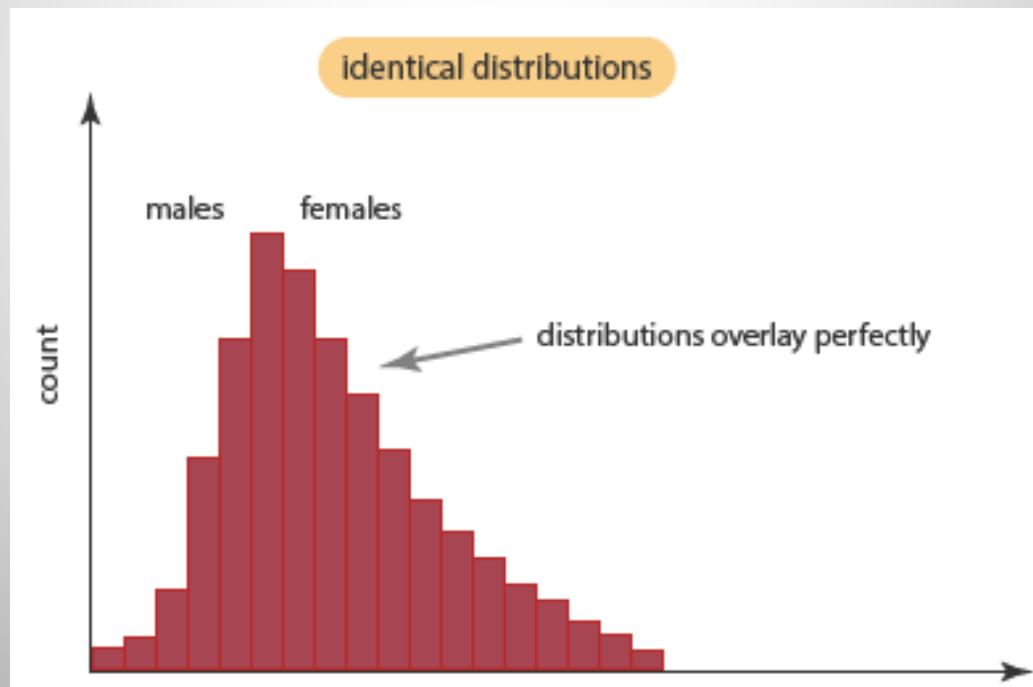
Assumptions: Observations are randomly sampled from a larger population and each value is obtained independently (iid)

## Independent t-test non-parametric equivalent: Mann-Whitney U (Wilcoxon rank sum)

Question: Are the rank scores of the two populations the same or different in terms of distribution and/or location (median)

$H_0$ : the distribution of rank scores for the two groups are the same

$H_A$ : the distribution of rank scores for the two groups are not the same



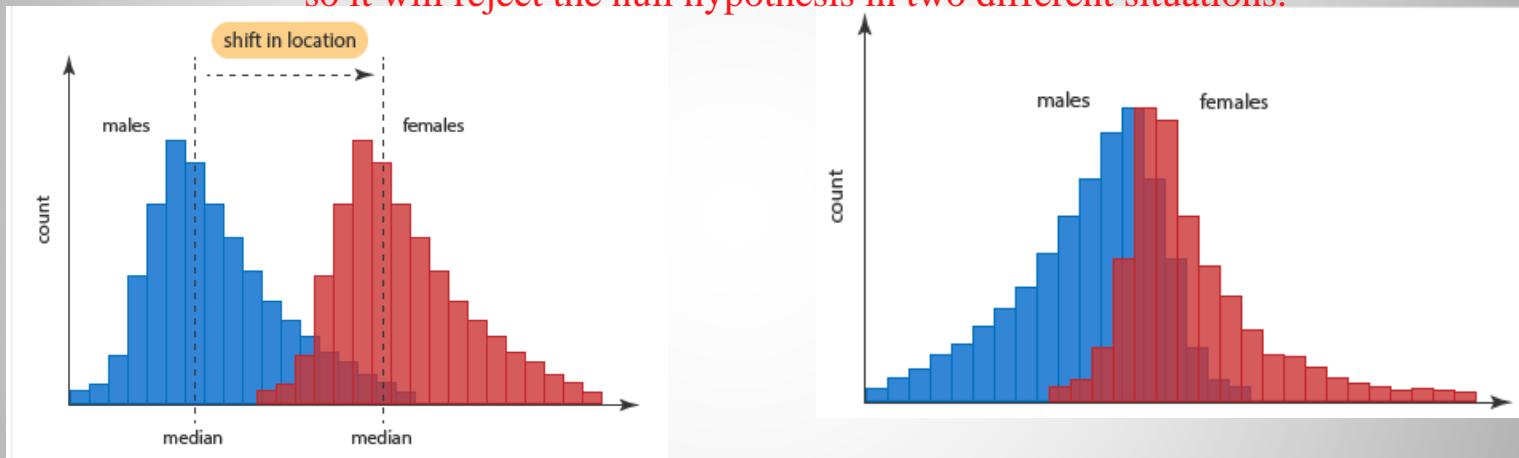
# Independent t-test non-parametric equivalent: Mann-Whitney U (Wilcoxon rank sum)

those two are the same

## Different distributions

Same shape, different median

so it will reject the null hypothesis in two different situations.



If  $p \leq 0.05$  and the medians of the actual data (not ranks) are different, conclude that the medians are different.

If  $p \leq 0.05$  and the medians of the actual data (not ranks) are the same, conclude that the distributions are different.

## Procedure for the Mann-Whitney U test

- Rank all the values in the dataset without paying attention to the group from which the value is drawn. If there are ties, assign the tied values the average of the ranks.
- Sum the ranks in the group 1 and name it  $R_1$  and do the same for group 2 to get  $R_2$ .
- Calculate  $U_1 = R_1 - \frac{n_1(n_1+1)}{2}$  and  $U_2 = R_2 - \frac{n_2(n_2+1)}{2}$ .
- The smaller of  $U_1$  and  $U_2$  is Mann-Whitney U statistic.
- Use Mann-Whitney test table to get the p-value or use Prism!

## Old Rat Young Rat Data

# Compute Mann-Whitney statistic by hand

- Old: 20.8, 2.8, 50.0, 33.3, 29.4, 38.9, 29.4, 52.6, 14.3.
- Young: 45.5, 55.0, 60.7, 61.5, 61.1, 65.5, 42.9, 37.5.
- Ranks:
  - Old: 3, 1, 11, 6, 4.5, 8, 4.5, 12, 2.
  - Young: 10, 13, 14, 16, 15, 17, 9, 7.
- $R_1 = 3+1+11+6+4.5+8+4.5+12+2=52$
- $R_2 = 0+13+14+16+15+17+9+7=101$

Rank	Value
1	2.8
2	14.3
3	20.8
4	29.4
5	29.4
6	33.3
7	37.5
8	38.9
9	42.9
10	45.5
11	50.0
12	52.6
13	55.0
14	60.7
15	61.1
16	61.5
17	65.5

Tie

## Compute Mann-Whitney statistic by hand

- $R_1 = 52, R_2 = 101.$
- $n_1 = 9, n_2 = 8.$
- Therefore,  $U_1 = 52 - \frac{9(9+1)}{2} = 7$  and  $U_2 = 101 - \frac{8(8+1)}{2} = 65.$
- Mann-Whitney U-statistic is the smaller one, i.e. 7.



Table 3 Critical values of  $U$  (5% significance).

$n_1 \backslash n_2$	1	2	3	4	5	6	7	8	9	10	11
1											
2							0	0	0	0	
3			0	1	1	2	2	3	3	3	
4		0	1	2	3	4	4	5	6		
5	0	1	2	3	5	6	7	8	9		
6	1	2	3	5	6	8	10	11	13		
7	1	3	5	6	8	10	12	14	16		
8	0	2	4	6	8	10	13	15	17	19	
9	0	2	4	7	10	12	15	17	20	23	
10	0	3	5	8	11	14	17	20	23	26	
11	0	3	6	9	13	16	19	23	26	30	

Reject  $H_0$  if  $U_{\text{smallest}} \leq \text{critical value}$

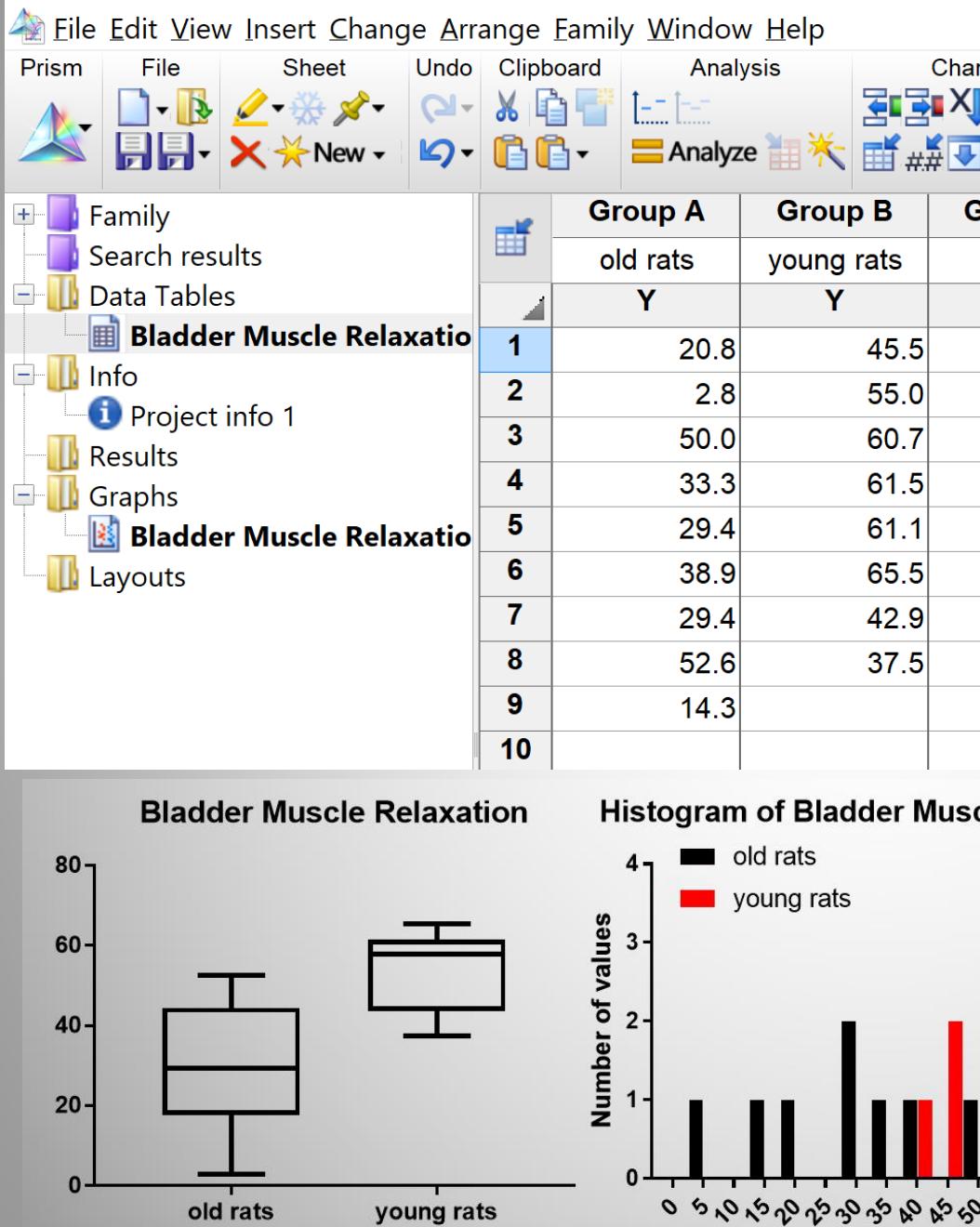
Reject  $H_0$  if  $U \leq 15$ .

Fail to reject  $H_0$  if  $U > 15$ .

What is the value of the  $U$  statistic?

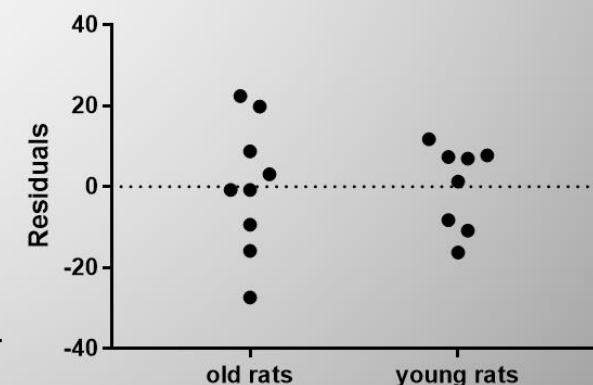
Do we reject or fail to reject the null hypothesis?

Can you conclude that the medians are different?



We saw this earlier...

	Col. stats	A	B
		old	young
1	Number of values	9	8
2			
3	Minimum	2.8	37.5
4	25% Percentile	17.55	43.55
5	Median	29.4	57.85
6	75% Percentile	44.45	61.4
7	Maximum	52.6	65.5
8			
9	Mean	30.17	53.71
10	Std. Deviation	16.09	10.36
11	Std. Error of Mean	5.365	3.664
12			
13	Skewness	-0.213	-0.5515
14	Kurtosis	-0.3964	-1.463
15			



To do the Mann-Whitney U test, start with the same Column table as for the t-test

Analyze Data

Built-in analysis

Which analysis?

A:old  
 B:young

**Transform, Normalize...**  
Transform  
Transform concentrations (X)  
Normalize  
Prune rows  
Remove baseline and column math  
Transpose X and Y  
Fraction of total

**XY analyses**  
**Column analyses**  
**t tests (and nonparametric tests)**  
One-way ANOVA (and nonparametric or mixed)  
One sample t and Wilcoxon test  
Descriptive statistics  
Normality and Lognormality Tests  
Frequency distribution  
ROC Curve  
Bland-Altman method comparison  
Identify outliers  
Analyze a stack of P values

**Grouped analyses**  
**Contingency table analyses**  
**Survival analyses**  
**Parts of whole analyses**

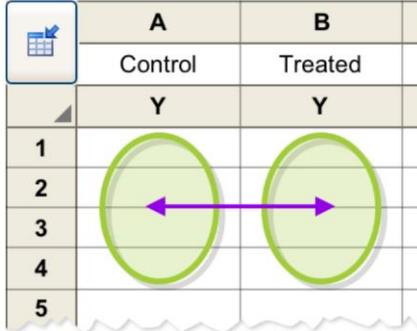
Select All   Deselect All

Help   Cancel   OK

Parameters: t tests (and Nonparametric Tests)

Experimental Design   Residuals   Options

**Experimental design**  
 Unpaired  
 Paired



**Assume Gaussian distribution?**  
 Yes. Use parametric test.  
 No. Use nonparametric test. ←

**Choose test**  
 Mann-Whitney test. Compare ranks  
 Kolmogorov-Smirnov test. Compare cumulative distributions

Learn   Cancel   OK

Parameters: t tests (and Nonparametric Tests)

Experimental Design Residuals Options

**Calculations**

P value:  One-tailed  Two-tailed (recommended)

Report differences as: young - old

Confidence level: 95%

Definition of statistical significance: P < 0.05

**Graphing options**

Graph differences (paired)  
 Graph ranks (nonparametric)  
 Graph correlation (paired)  
 Graph CI of difference between medians

**Additional results**

Descriptive statistics for each data set  
 t test: Also compare models using AICc  
 Mann-Whitney: Also compute the CI of difference between medians  
 Assumes both distributions have the same shape.  
 Wilcoxon: When both values on a row are identical, use method of Pratt  
 If this option is unchecked, those rows are ignored and the results will match prior version of Prism

**Output**

Show this many significant digits (for everything except P values): 4

P value style: GP: 0.1234 (ns), 0.0332 (\*), 0.0021 (\*\*), 0.00 N = 6

Make options on this tab be the default for future tests.

Learn Cancel OK

Mann-Whitney test	
Tabular results	
2	
3	Column B less power for the test
4	vs. young
5	Column A vs.
6	old
7	<b>Mann Whitney test</b>
8	P value 0.0035
9	Exact or approximate P value? Exact
10	P value summary **
11	Significantly different (P < 0.05)? Yes
12	One- or two-tailed P value? Two-tailed
13	Sum of ranks in column A,B 52 , 101
14	Mann-Whitney U 7
15	
16	<b>Difference between medians</b>
17	Median of column A 29.40, n=9
18	Median of column B 57.85, n=8
19	Difference: Actual 28.45
20	Difference: Hodges-Lehmann 23.95
21	95.36% CI of difference 8.500 to 39.90
22	Exact or approximate CI? Exact
23	

Prism computed an exact P value (0.0035), which takes into account ties among values. Note that most other programs do not compute exact P values when there are tied values, but would instead report an approximate P value (0.0061).

Prism reports the **difference in medians** in two ways.

Subtract the median of one group from the median of the other

Compute the Hodges-Lehmann estimate.

Many think it is the best estimate for the difference between population medians

Since the nonparametric test works with ranks, it is usually not possible to get a confidence interval with exact 95% confidence.

Prism finds a close confidence level, and reports what it is. For example, you might get a 96.2% confidence interval when you asked for a 95% interval.

You can either report the precise confidence level ("96.2%") or just report the confidence level you requested ("95%"). The latter approach is used more commonly.

Unpaired t test		
Tabular results		
2		
3	Column B	
4	vs.	
5	Column A	
6		
7	<b>Unpaired t test</b>	
8	P value	0.0030
9	P value summary	**
10	Significantly different ( $P < 0.05$ )?	Yes
11	One- or two-tailed P value?	Two-tailed
12	t, df	t=3.531, df=15
13		
14	<b>How big is the difference?</b>	
15	Mean of column A	30.17
16	Mean of column B	53.71
17	Difference between means (B - A) $\pm$ SEM	23.55 $\pm$ 6.667
18	95% confidence interval	9.335 to 37.76
19	R squared (eta squared)	0.4540
20		
21	<b>F test to compare variances</b>	
22	F, DFn, Dfd	2.412, 8, 7
23	P value	0.2631
24	P value summary	ns
25	Significantly different ( $P < 0.05$ )?	No

1	Table Analyzed	Bladder relaxation
2		
3	Column B	young
4	vs.	vs.
5	Column A	old
6		
7	<b>Unpaired t test with Welch's correction</b>	
8	P value	0.0028
9	P value summary	**
10	Significantly different ( $P < 0.05$ )?	Yes
11	One- or two-tailed P value?	Two-tailed
12	Welch-corrected t, df	t=3.624, df=13.78
13		
14	<b>How big is the difference?</b>	
15	Mean of column A	30.17
16	Mean of column B	53.71
17	Difference between means (B - A) $\pm$ SEM	23.55 $\pm$ 6.497
18	95% confidence interval	9.591 to 37.50
19	R squared (eta squared)	0.4881
20		
21	<b>F test to compare variances</b>	
22	F, DFn, Dfd	2.412, 8, 7
23	P value	0.2631
24	P value summary	ns
25	Significantly different ( $P < 0.05$ )?	No
26		

Mann-Whitney test		
Tabular results		
2		
3	Column B	young
4	vs.	vs.
5	Column A	old
6		
7	<b>Mann Whitney test</b>	
8	P value	0.0035
9	Exact or approximate P value?	Exact
10	P value summary	**
11	Significantly different ( $P < 0.05$ )?	Yes
12	One- or two-tailed P value?	Two-tailed
13	Sum of ranks in column A,B	52 , 101
14	Mann-Whitney U	7
15		
16	<b>Difference between medians</b>	
17	Median of column A	29.40, n=9
18	Median of column B	57.85, n=8
19	Difference: Actual	28.45
20	Difference: Hodges-Lehmann	23.95
21	95.36% CI of difference	8.500 to 39.90
22	Exact or approximate CI?	Exact
23		

Prism computed an exact P value (0.0035), which takes into account ties among values. Note that most other programs do not compute exact P values when there are tied values, but would instead report an approximate P value (0.0061).

# Write up results for Mann-Whitney

## For this class:

The data for both groups failed the assumptions for an independent t-test (independence, normality/symmetry, homoscedasticity) so we performed the non-parametric Mann Whitney U test. With  $p=0.004$  (two-sided Mann-Whitney test,  $\alpha=0.05$ ), we reject the null hypothesis of no difference in rank score distribution of muscle relaxation between old and young rats. We conclude that older rats have lower levels of muscle relaxation (median = 4.5) compared to younger rats (13.5).

## What people may usually write:

We conclude that older rats have lower levels of muscle relaxation (median = 4.5) compared to younger rats (13.5;  $p=0.004$ ).

# The One Sample t-Test

The one sample t test is used to compare a single sample to a population with a known mean.

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$$

$\bar{x}$  = sample means,  
 $\mu_0$  = population mean,  
 $s$  = sample SD,  
 $n$  = sample size

df = n-1

$$\begin{aligned} H_0: \bar{x} &= 0 \\ H_A: \bar{x} &\neq 0 \end{aligned}$$

$$\begin{aligned} H_0: \bar{x} &= 100 \\ H_A: \bar{x} &\neq 100 \end{aligned}$$

It doesn't have to be zero, it could be any number you want to compare to your data

- Assumptions -

The data should follow the normal distribution.

The sample is a random sample from its population (iid).

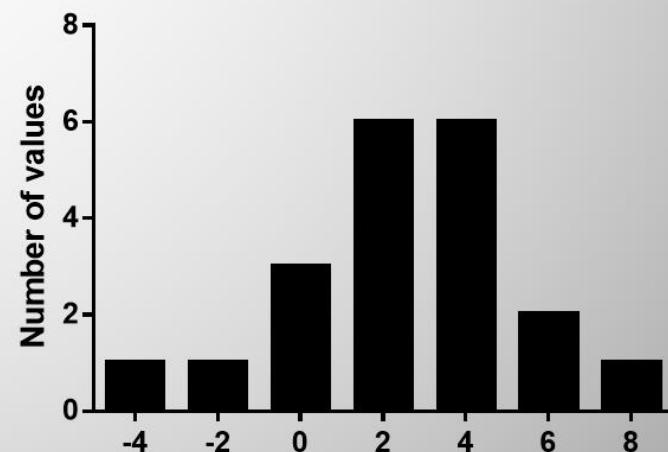
## Data in Excel

	A	B	C	D
1	Patient	Before Treatment	After Treatment	Difference
2	1	18	22	4
3	2	21	25	4
4	3	16	17	1
5	4	22	24	2
6	5	19	16	-3
7	6	24	29	5
8	7	17	20	3
9	8	21	23	2
10	9	23	19	-4
11	10	18	20	2
12	11	14	15	1
13	12	16	15	-1
14	13	16	18	2
15	14	19	26	7
16	15	18	18	0

## Data in Prism

	Group A	Group B	
	Before Treatment	After Treatment	
	18	22	
	21	25	
	16	17	
	22	24	
	19	16	
	24	29	
	17	20	
	21	23	
	23	19	
	18	20	
	14	15	
	16	15	
	16	18	
	19	26	
	18	18	
	20	24	
	12	18	
	22	25	
	15	19	
	17	16	

Histogram of Difference



$$\begin{aligned} H_0: \bar{x} &= 0 \\ H_A: \bar{x} &\neq 0 \end{aligned}$$

## Analyze Data

X

### Built-in analysis

#### Which analysis?

##### **Transform, Normalize...**

- Transform
- Transform concentrations (X)
- Normalize
- Prune rows
- Remove baseline and column math
- Transpose X and Y
- Fraction of total

##### **XY analyses**

##### **Column analyses**

- t tests (and nonparametric tests)
- One-way ANOVA (and nonparametric or mixed)
- One sample t and Wilcoxon test**
- Descriptive statistics
- Normality and Lognormality Tests
- Frequency distribution
- ROC Curve
- Bland-Altman method comparison
- Identify outliers
- Analyze a stack of P values

##### **Grouped analyses**

##### **Contingency table analyses**

##### **Survival analyses**

##### **Parts of whole analyses**



#### Analyze which data sets?

- A:Before Treatment
- B:After Treatment
- C:Difference

Select All

Deselect All

Help

Cancel

OK

Parameters: One sample t and Wilcoxon test

Experimental Design Options

**Choose test**

One sample t test  
Compare the mean of your sample with a hypothetical mean.  
Assumes sampling from a Gaussian distribution.

Wilcoxon signed-rank test  
Compare the median of your sample with a hypothetical median.  
Nonparametric.

Calculate CI of the discrepancy

**Hypothetical value**

Hypothetical value. Often 0.0, 1.0 or 100:

For the Wilcoxon test, if a value in data set matches the hypothetical value:

Ignore that value entirely

Include that value using method of Pratt (not commonly used)

Learn Cancel OK

Parameters: One sample t and Wilcoxon test

Experimental Design Options

**Subcolumns**

Average the replicates in each row, and then perform the calculation for each column

Perform the calculation for each subcolumn separately

Treat all the values in all the subcolumns as one set of data

**Calculations**

Significance level (alpha)

**Output**

Show this many significant digits (for everything except P values):

P value style: GP: 0.1234 (ns), 0.0332 (\*), 0.0021 (\*\*), 0.000 N =

Make these choices the default for future analyses.

Learn Cancel OK



	One sample t test	A Difference
1	Theoretical mean	0.000
2	Actual mean	2.050
3	Number of values	20
4		
5	<b>One sample t test</b>	
6	t, df	t=3.231, df=19
7	P value (two tailed)	0.0044
8	P value summary	**
9	Significant (alpha=0.05)?	Yes
10		
11	<b>How big is the discrepancy?</b>	
12	Discrepancy	2.050
13	SD of discrepancy	2.837
14	SEM of discrepancy	0.6344
15	95% confidence interval	0.7221 to 3.378
16	R squared (partial eta squared)	0.3546
17		

## WRITE UP RESULTS

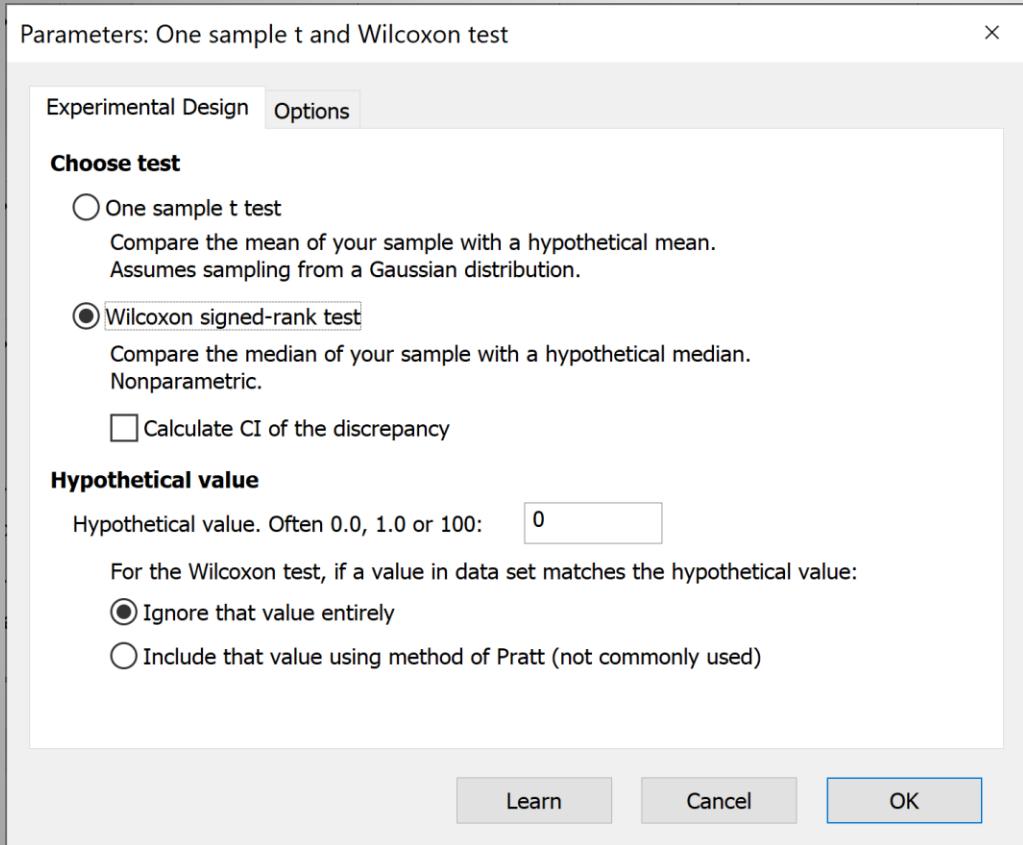
For this class:

The data passed the assumptions for a one sample t-test (data are normal/symmetrical). With  $p=0.004$  (two-tailed one sample t-test,  $t=3.231$ ,  $df=19$ ,  $\alpha=0.05$ ), we reject the null hypothesis that the mean of our data is zero. Further, the 95%CI do not include 0 (0.7 to 3.4). We conclude that with a mean increase of 2.05 over placebo, treatment increases the value of Test Y.

What people may usually write:

We found that treatment increases the value of Test Y by 2.05 ( $p=0.004$ ).

# The non-parametric alternative to the one-sample t-test: the Wilcoxon signed rank test



$$H_0: \text{the median} = 0$$

$$H_A: \text{the median} \neq 0$$

It doesn't have to be zero, it could be any number you want to compare to your data

	A	Difference
1	Theoretical median	0.000
2	Actual median	2.000
3	Number of values	20
4		
5	<b>Wilcoxon Signed Rank Test</b>	
6	Sum of signed ranks (W)	132.0
7	Sum of positive ranks	161.0
8	Sum of negative ranks	-29.00
9	P value (two tailed)	0.0058
10	Exact or estimate?	Exact
11	P value summary	**
12	Significant (alpha=0.05)?	Yes
13		
14	<b>How big is the discrepancy?</b>	
15	Discrepancy	2.000
16		

	A	Difference
1	Theoretical mean	0.000
2	Actual mean	2.050
3	Number of values	20
4		
5	<b>One sample t test</b>	
6	t, df	t=3.231, df=19
7	P value (two tailed)	0.0044
8	P value summary	**
9	Significant (alpha=0.05)?	Yes
10		
11	<b>How big is the discrepancy?</b>	
12	Discrepancy	2.050
13	SD of discrepancy	2.837
14	SEM of discrepancy	0.6344
15	95% confidence interval	0.7221 to 3.378
16	R squared (partial eta squared)	0.3546

## Write up results

### For this class:

Because the data did not pass the assumptions for a one-sample t-test (data are normal/symmetrical), a Wilcoxon signed rank test was used. With  $p=0.006$  (two-tailed Wilcoxon signed rank test,  $\alpha=0.05$ ), we reject the null hypothesis that the median of our data is zero. We conclude that with a median increase of 2 points over placebo, treatment increases the value of Test Y.

### What people may usually write:

We found that treatment increases the value of Test Y by 2 points ( $p=0.006$ )