

P-values and statistical tests

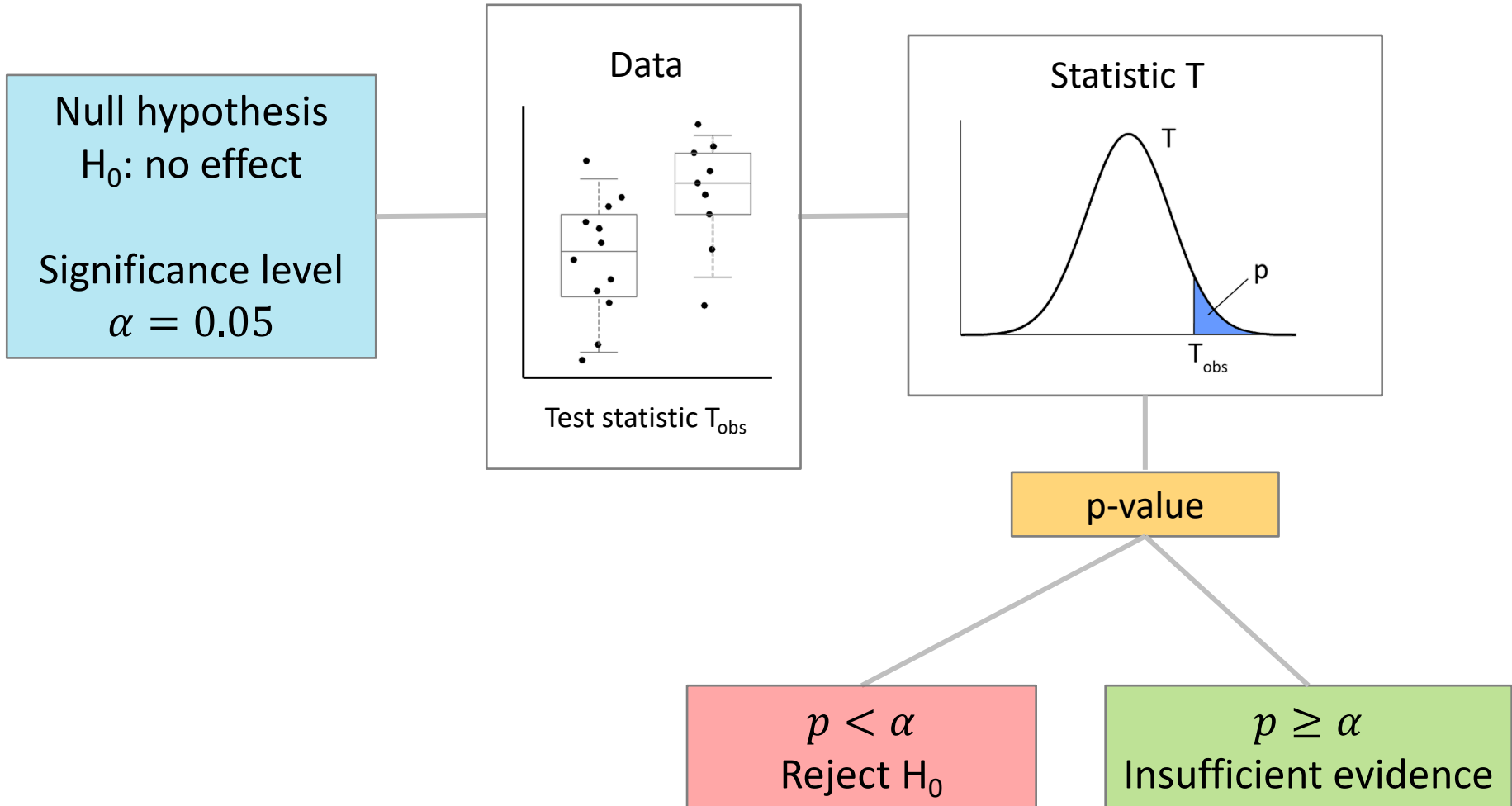
5. Non-parametric methods

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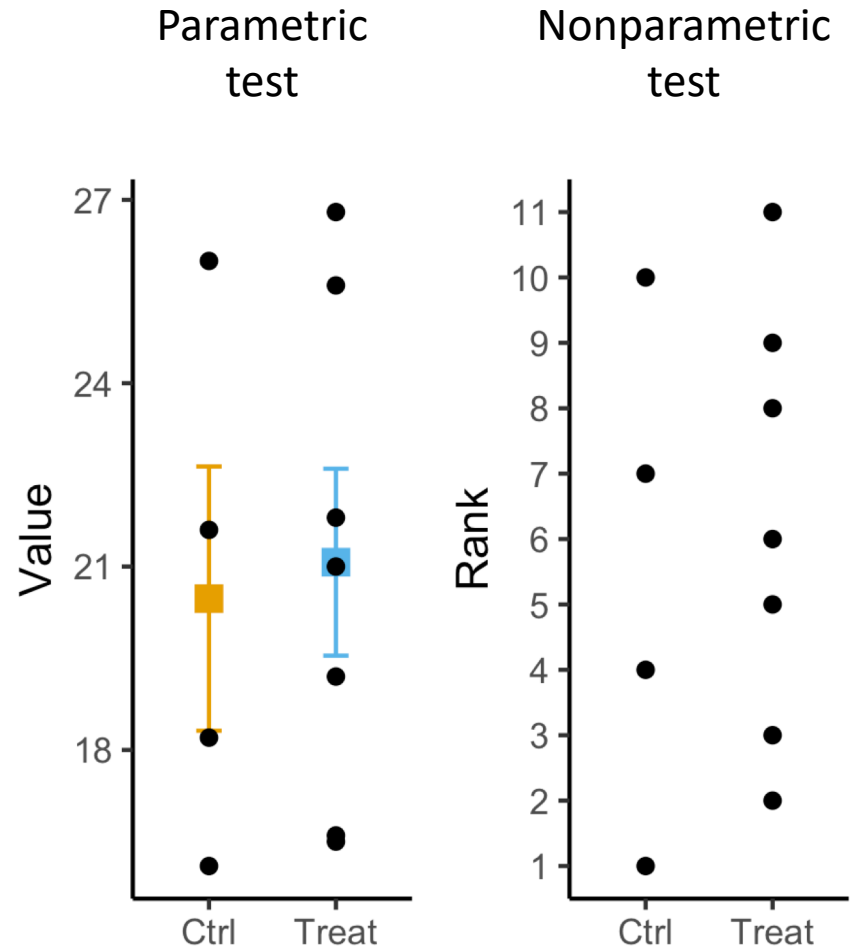
Hand-outs available at <http://is.gd/statlec>

Statistical test



Nonparametric methods

- Parametric methods:
 - require finding parameters (e.g. mean)
 - sensitive to distributions
 - don't work in some cases
 - more powerful
- Nonparametric methods:
 - based on ranks
 - distribution-free
 - wider application
 - less powerful



Mann-Whitney test

(Wilcoxon rank-sum test)

a nonparametric alternative to t-test

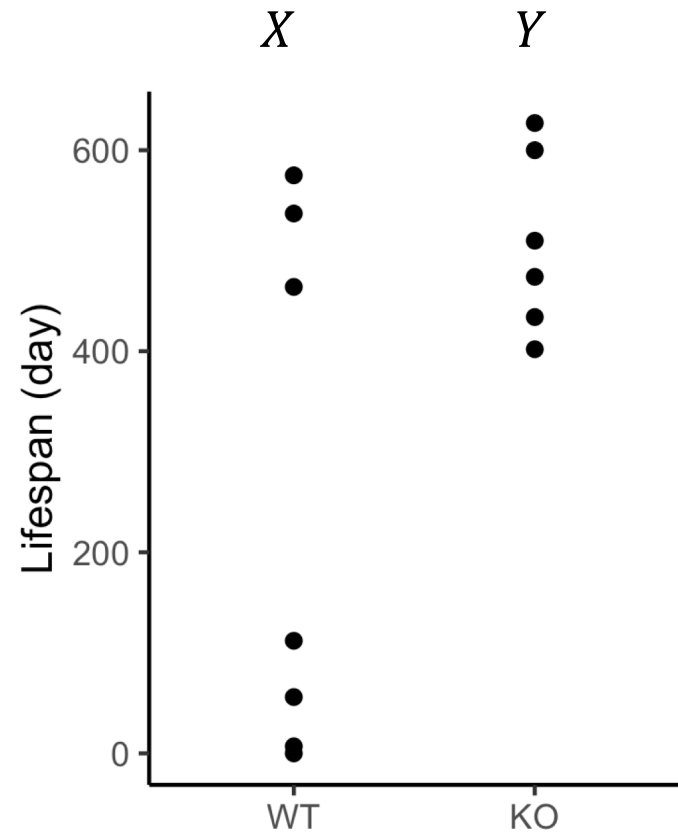
Mann-Whitney test

- Two samples representing random variables X and Y

- Null hypothesis: there is no shift in location (and/or change in shape)

$$H_0: P(X > Y) = P(Y > X)$$

- Only ranks matter, not actual values



Mann-Whitney test

- Two samples:

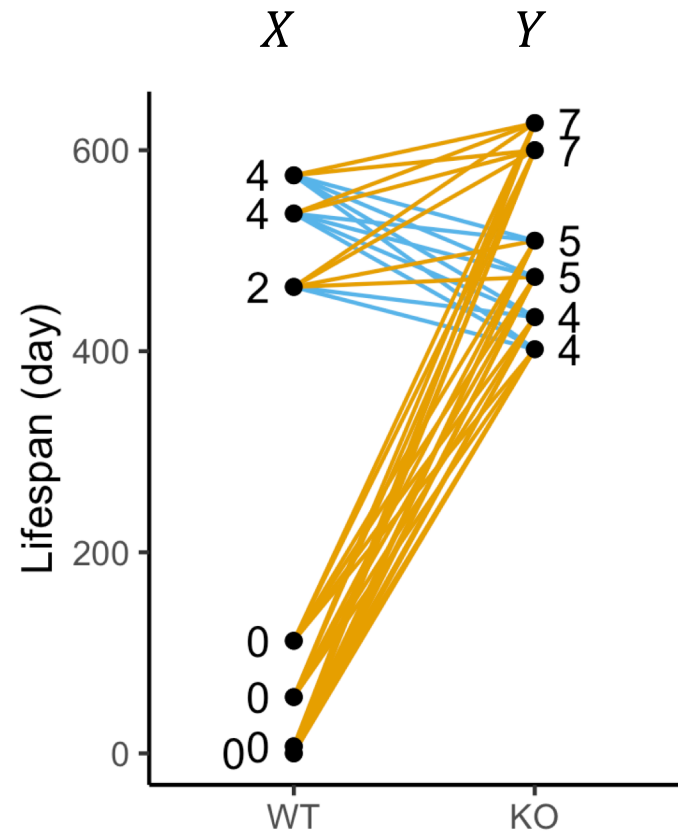
$$x_1, x_2, \dots, x_{n_x}$$

$$y_1, y_2, \dots, y_{n_y}$$

- For each x_i count the number of y_j , such that $x_i > y_j$
- The sum of these counts over all x_i is U_x
- Do the same for y_j and find U_y

- Test statistic

$$U = \min(U_x, U_y)$$



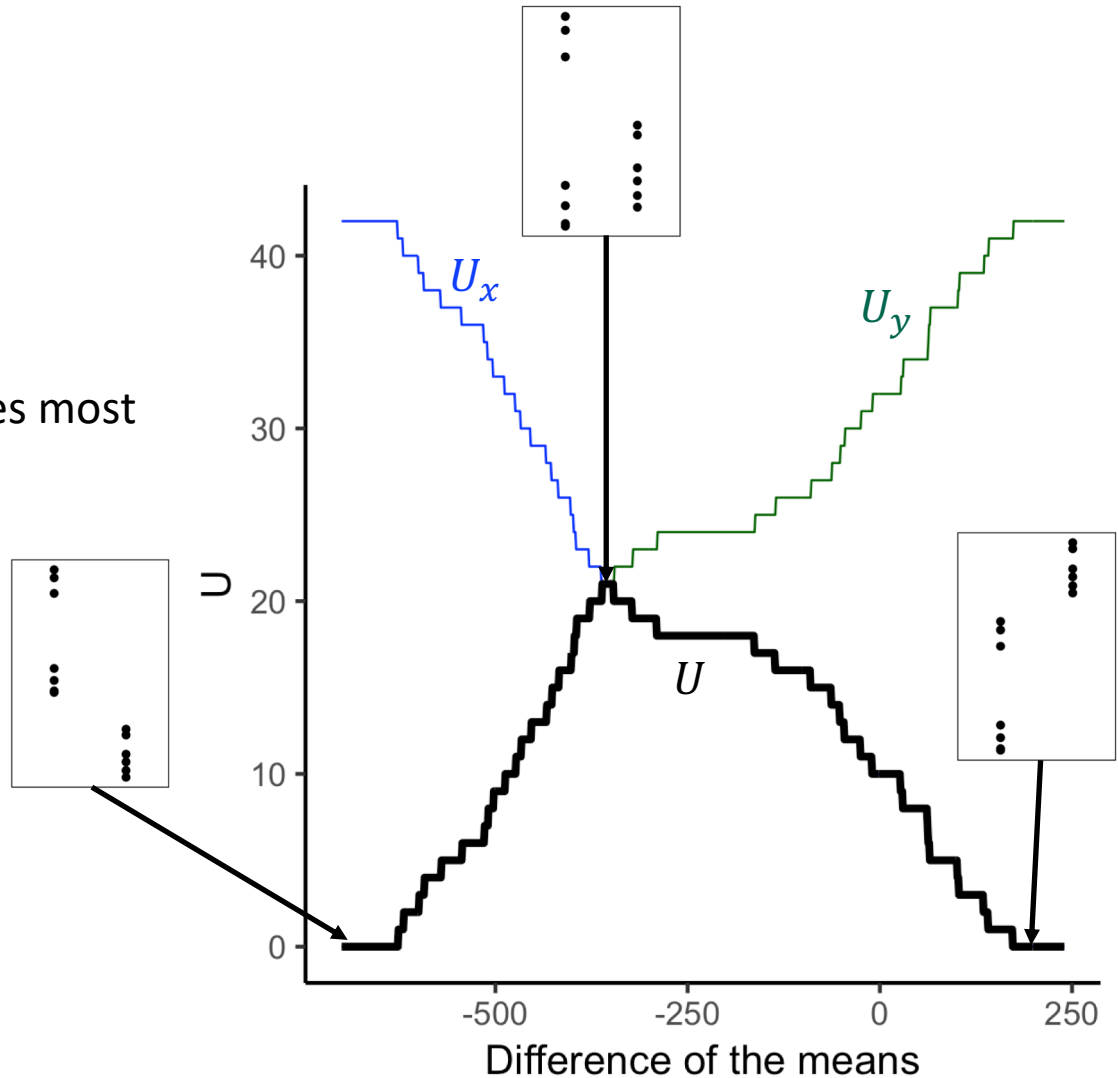
$$U_x = 10$$

$$U_y = 32$$

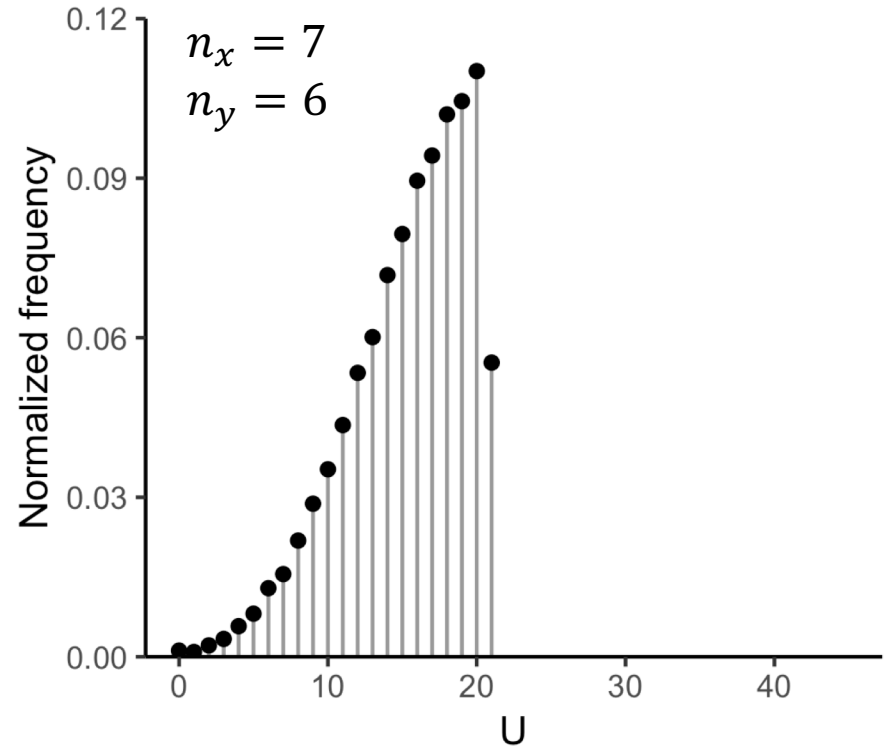
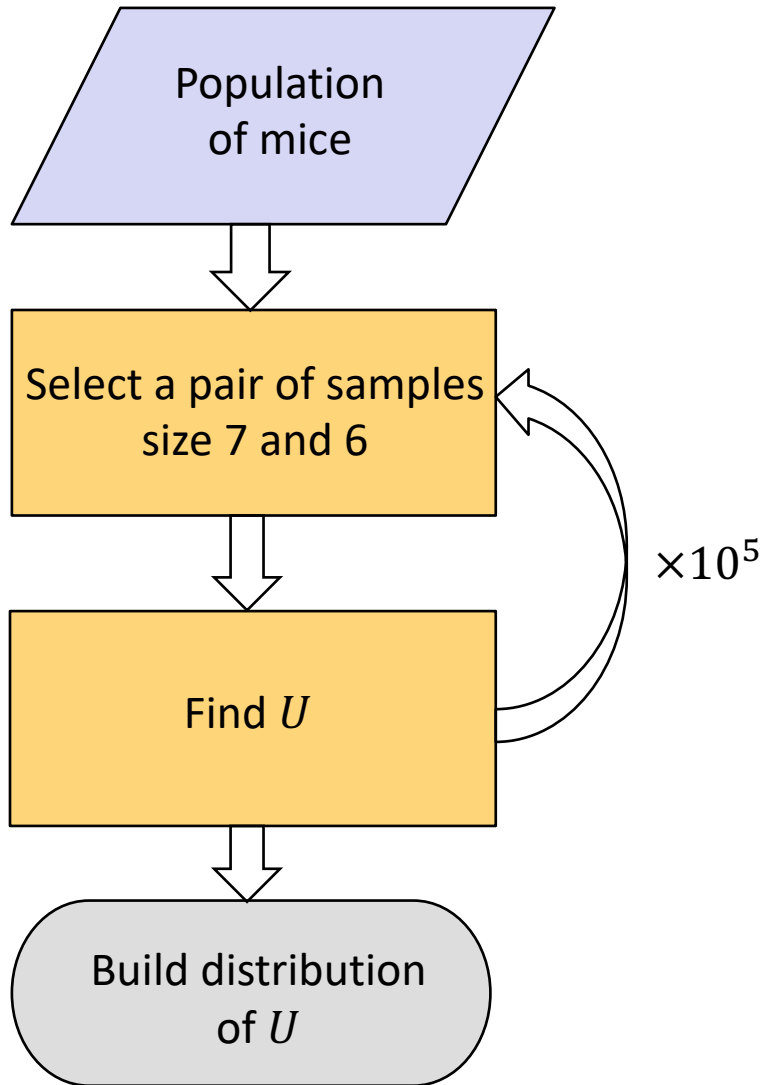
$$U = 10$$

Mann-Whitney test

- U measures difference in location between the samples
- With no overlap $U = 0$
- Direction not important
- $U = \max = \left\lfloor \frac{n_x n_y}{2} \right\rfloor$ when samples most similar



Null distribution



Null distribution represents all random samples when the null hypothesis is true

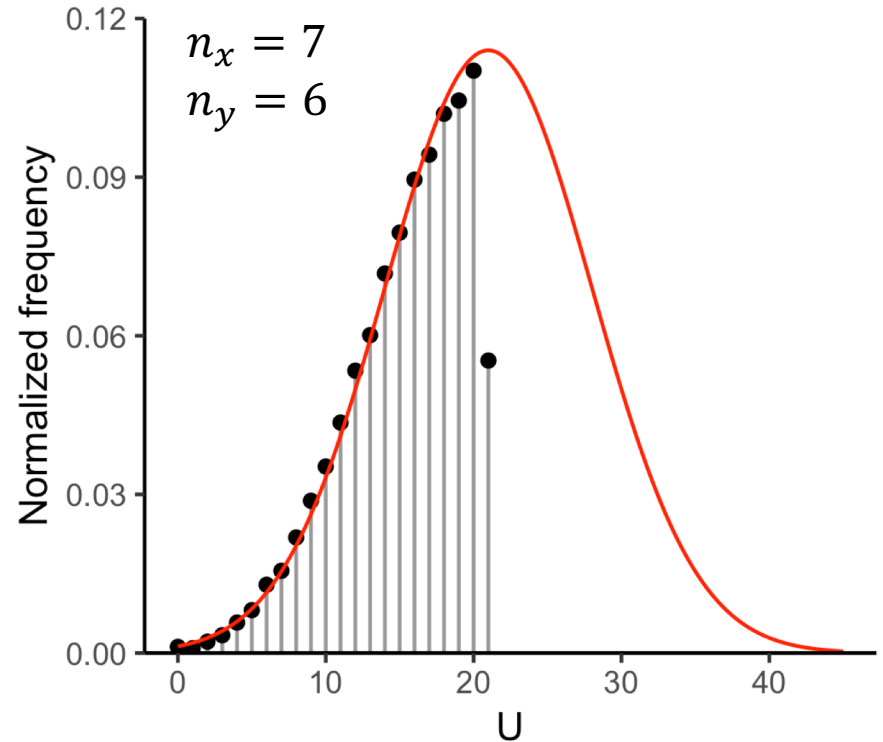
Null distribution

- For large samples U is approximately normally distributed (half of it) with

$$\mu_U = \frac{n_x n_y}{2}$$

$$\sigma_U = \sqrt{\frac{n_x n_y (n_x + n_y + 1)}{12}}$$

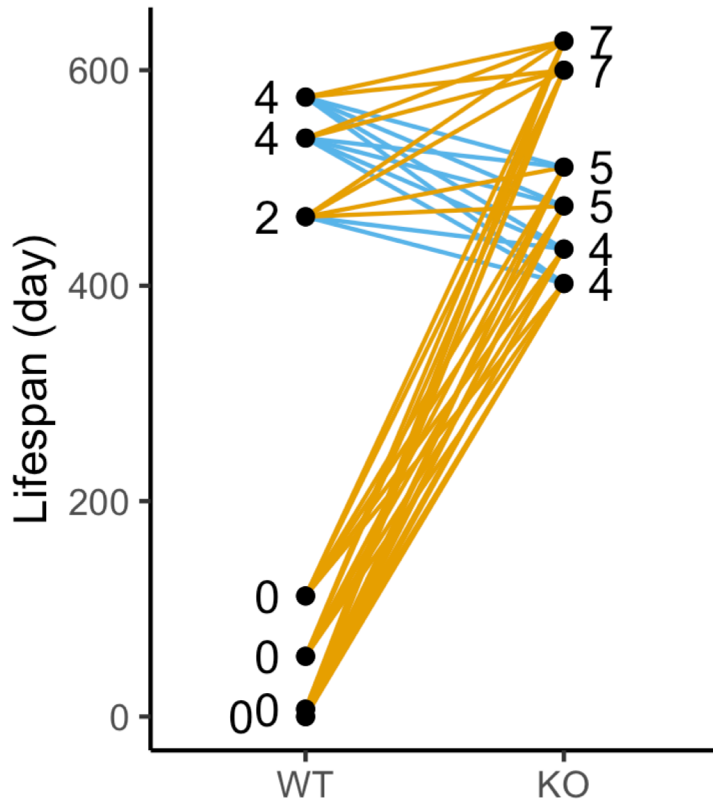
- For smaller samples exact solutions are available (tables or software)



$$\mu_U = \frac{7 \times 6}{2} = 21$$

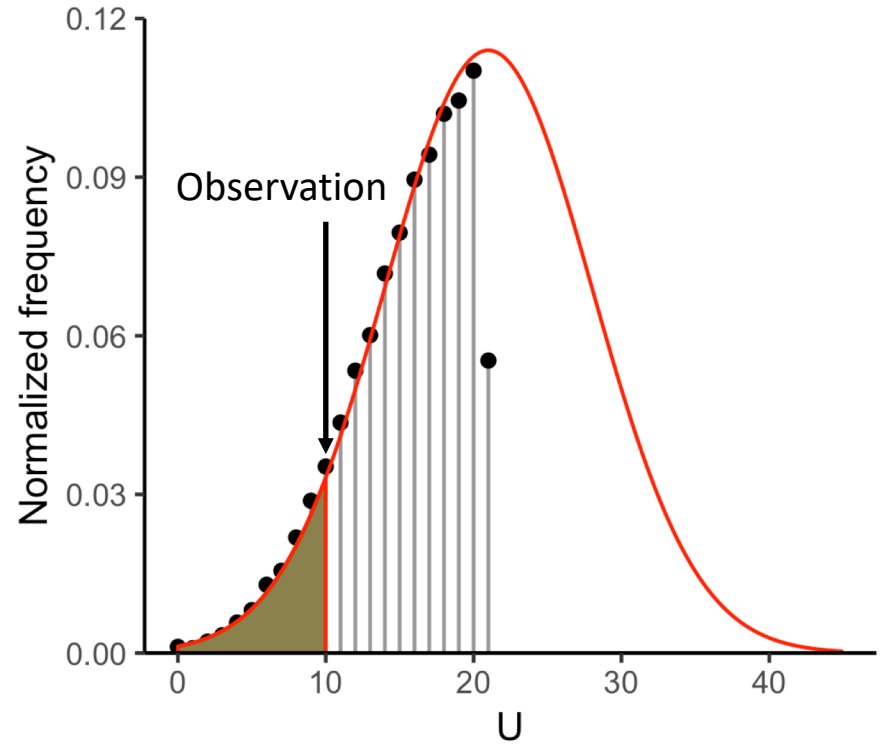
$$\sigma_U = \sqrt{\frac{7 \times 6 \times (7 + 6 + 1)}{12}} = 7$$

P-value



$$U_x = 10 \quad U_y = 32$$

$$U = 10$$



$$\mu_U = 21$$

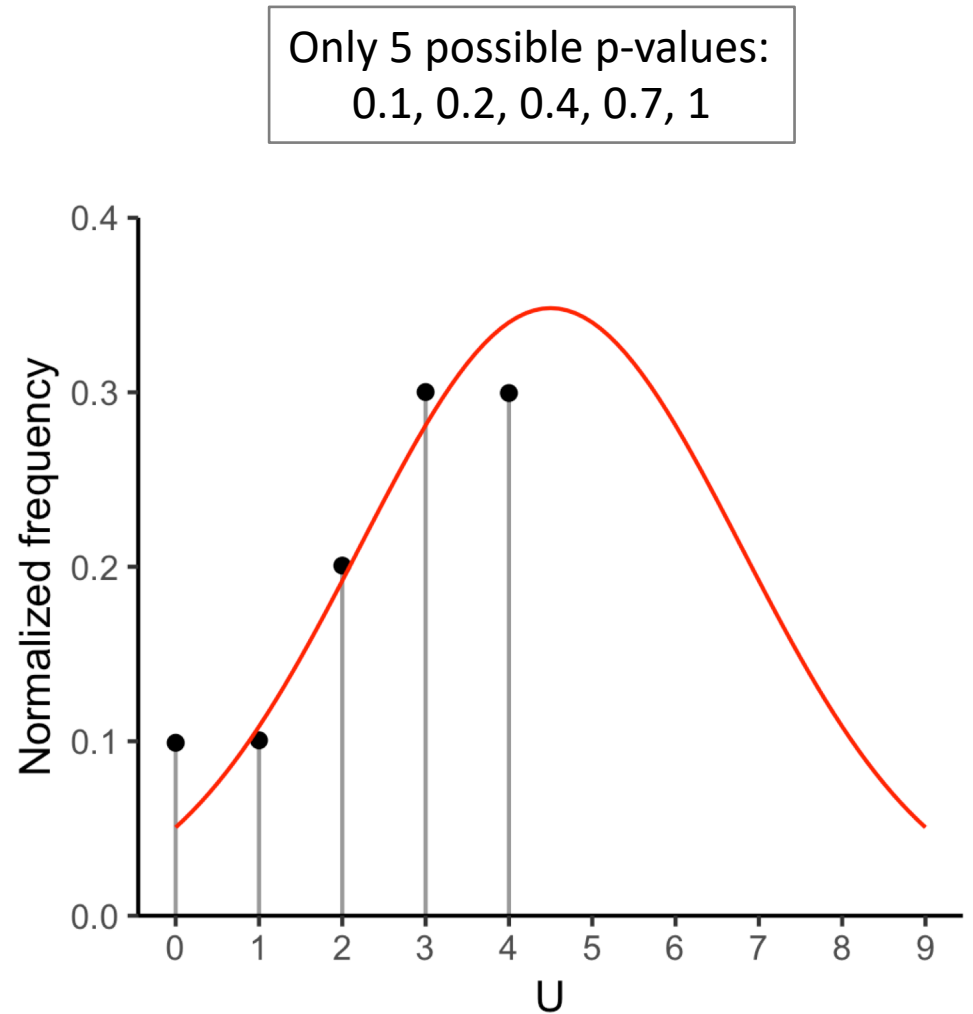
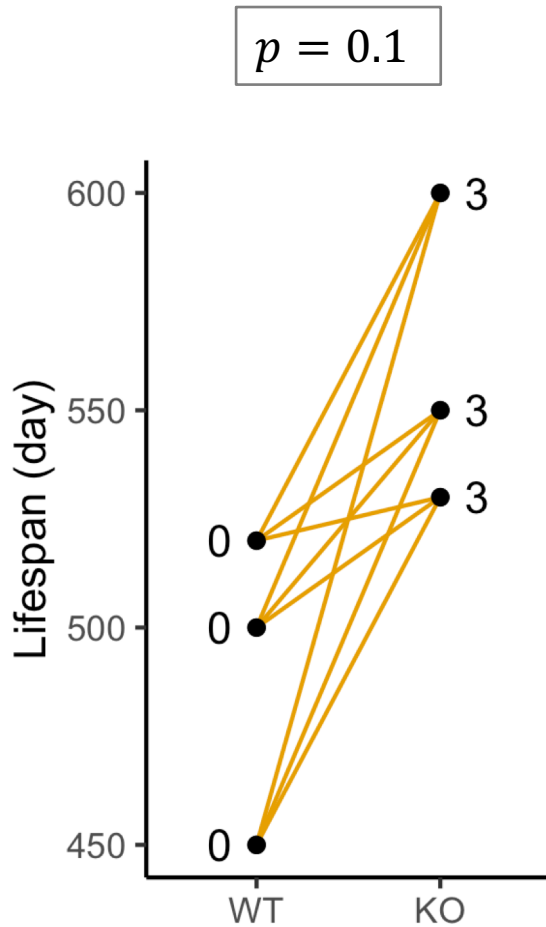
$$\sigma_U = 7$$

$$Z = \frac{U - \mu_U}{\sigma_U} = -1.57$$

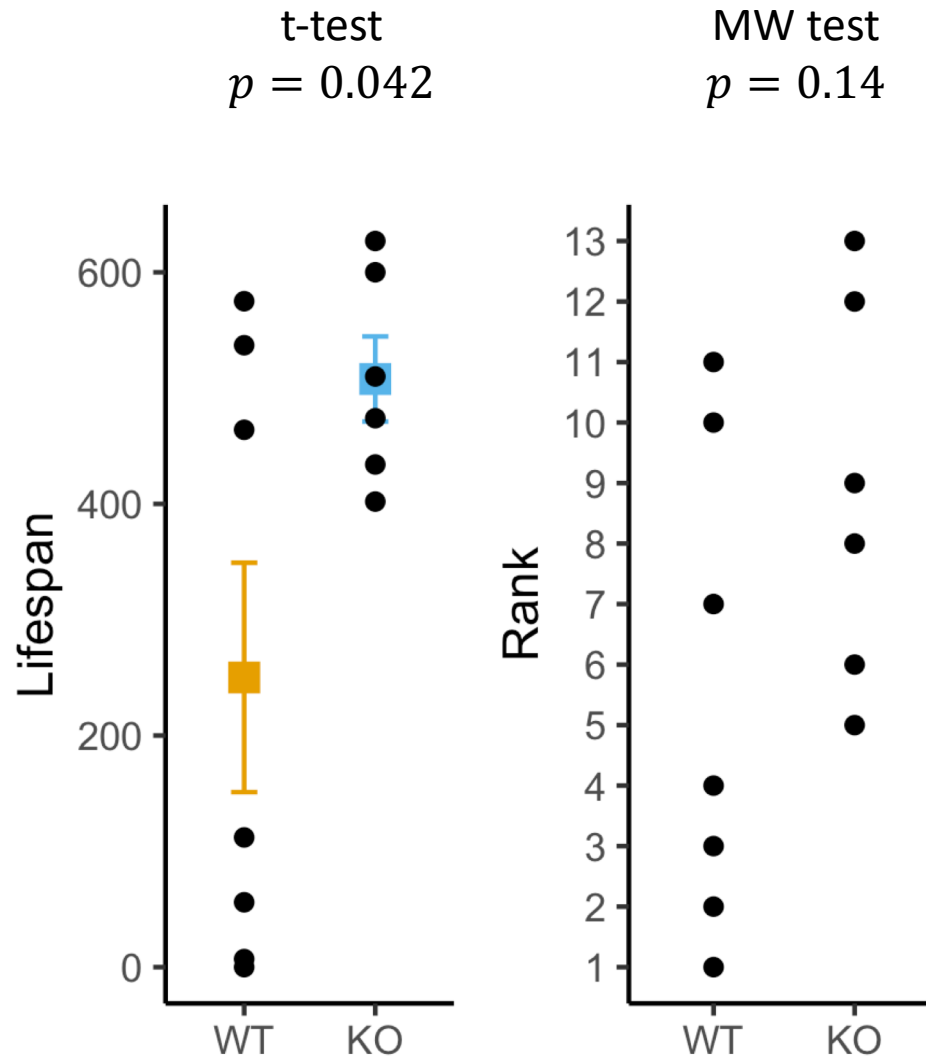
$$p = 0.12$$

$$\text{Exact solution: } p = 0.14$$

Limited usage for small samples

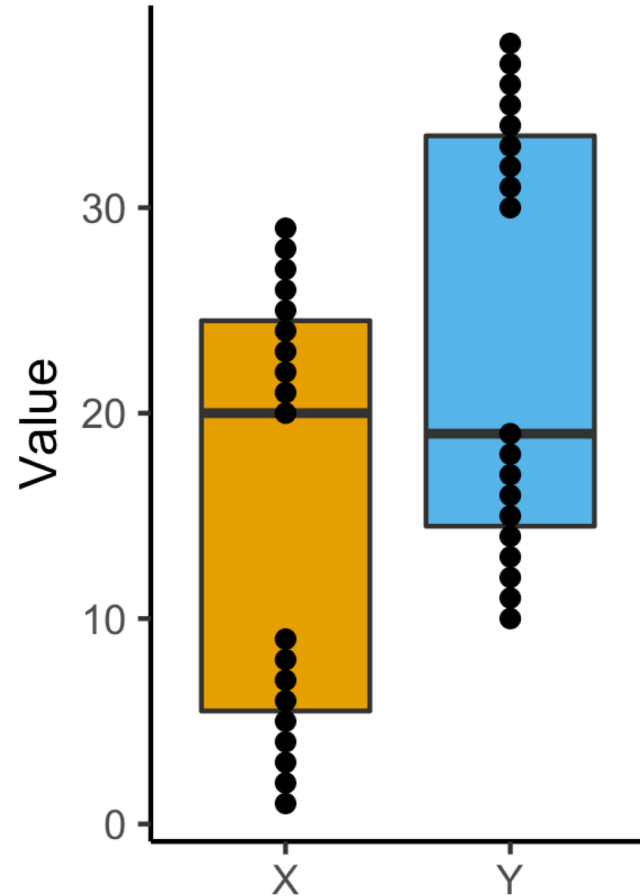


Comparison to t-test



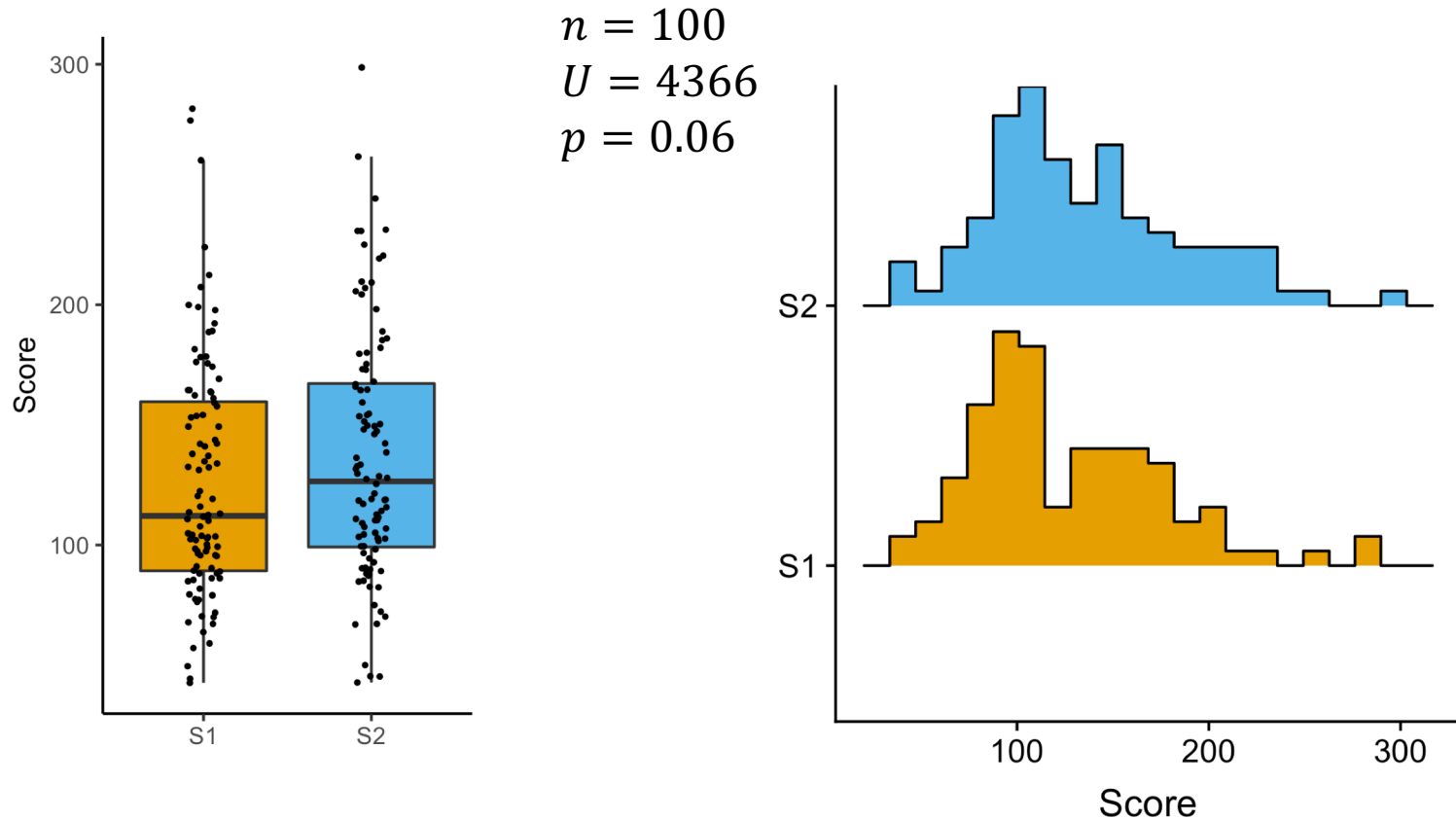
Mann-Whitney can compare medians, but...

- Consider two samples in the figure
- Yes, I know they are contrived
- Medians are similar, but $\text{med } X > \text{med } Y$
- Mann-Whitney test gives $U = 100$ and one-sided $p = 0.02$
- Y exceeds X !
- Mann-Whitney test is sensitive to change in location (median) and/or shape
- If shapes are similar, then MW test can be a test of medians
- Otherwise, use Mood's test for medians



What is Mann-Whitney test good for?

- If data are distributed (roughly) normally, use t-test
- MW test is good for weird distributions, e.g. 'scores'



What is Mann-Whitney test good for?

- Ordinal variables, e.g., APGAR score
- New pre-natal care program in a rural community

Usual care	8	7	6	2	5	8	7	3
New program	9	8	7	8	10	9	6	

- $U = 9.5$
- $p = 0.03$

How to do it in R?

```
> x <- c(0, 7, 56, 112, 464, 537, 575)
> y <- c(402, 434, 472, 510, 600, 627)
# Mann-Whitney test
> wilcox.test(x, y)
```

Wilcoxon rank sum test

data: x and y

W = 10, p-value = 0.1375

alternative hypothesis: true location shift is not equal to 0

```
# Mood's test for medians
```

```
> mood.test(x,y)
```

Mood two-sample test of scale

data: x and y

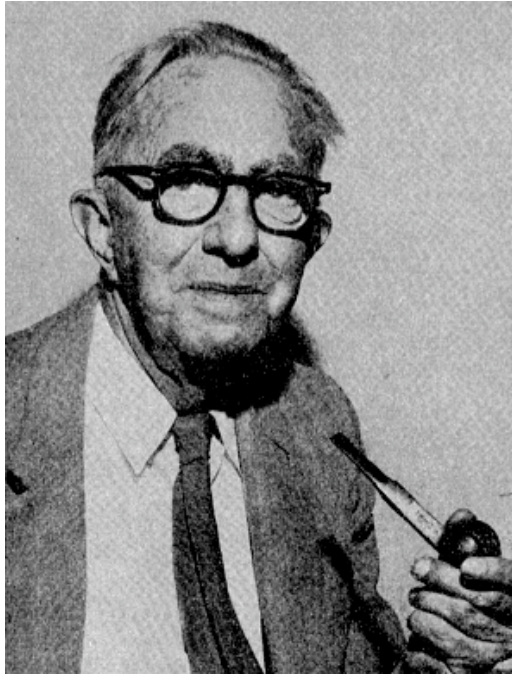
Z = 0.55995, p-value = 0.5755

alternative hypothesis: two.sided

Mann-Whitney test: summary

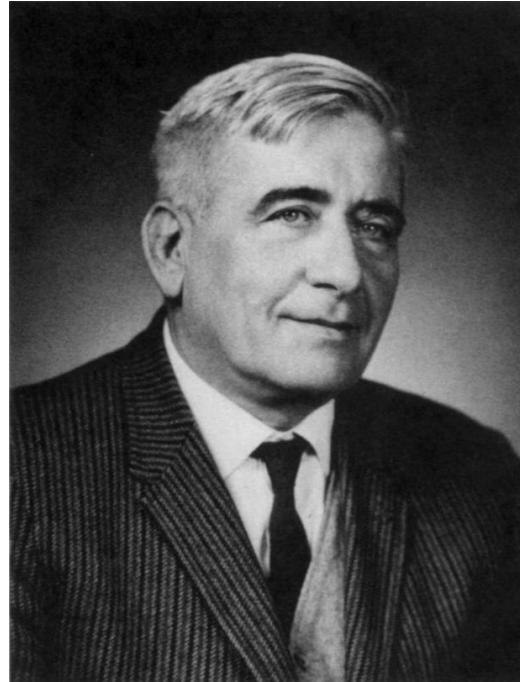
Input	two samples of n_1 and n_2 values values can be ordinal
Assumptions	Samples are random and independent (no before/after tests) If used for medians, both distributions must be the same
Usage	Compare location and shape of two samples
Null hypothesis	There is no shift in location and/or change in shape Stronger version: both samples are from the same distribution
Comments	Also known as Wilcoxon rank-sum test Non-parametric counterpart of t-test Less powerful than t-test (use t-test if distributions symmetric) Not very useful for small samples Doesn't really give the effect size

Mann-Whitney-Wilcoxon



Frank Wilcoxon
(1892-1965)

Wilcoxon, F. (1945) "Individual Comparisons by Ranking Methods" *Biometrics Bulletin* **1**, 80-83



Henry Berthold Mann
(1905-2000)

Mann, H. B.; Whitney, D. R. (1947). "On a Test of Whether one of Two Random Variables is Stochastically Larger than the Other" *Annals of Mathematical Statistics* **18**, 50-60



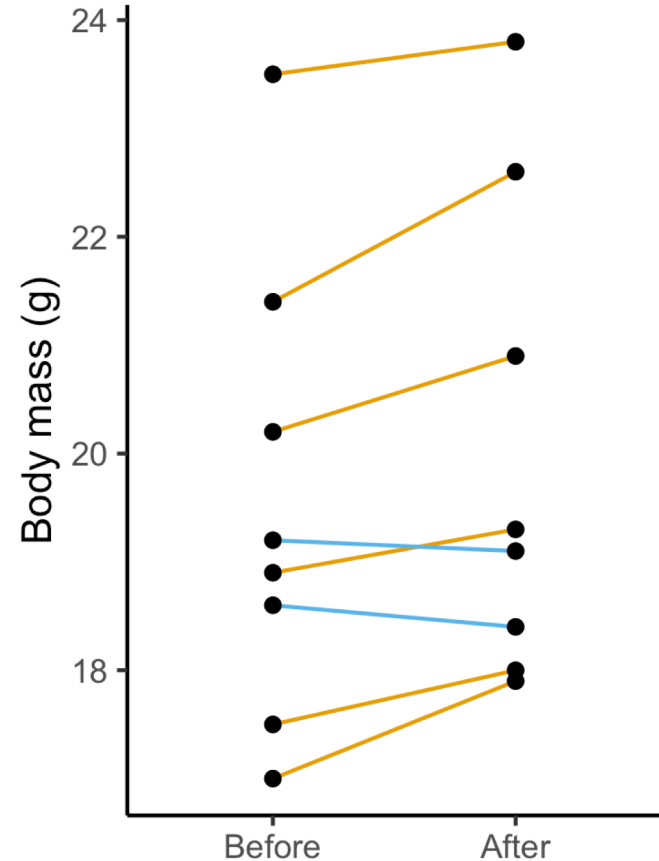
Donald Ransom Whitney
(1915-2007)

Wilcoxon signed-rank test

a nonparametric alternative to paired t-test

Paired data

- Samples are paired
- For example: mouse weight before and after obesity treatment
- Null hypothesis: difference between pairs follows a symmetric distribution around zero
- Example: mouse body mass (g)



Before:	21.4	20.2	23.5	17.5	18.6	17.0	18.9	19.2
After:	22.6	20.9	23.8	18.0	18.4	17.9	19.3	19.1

Wilcoxon signed-rank test

- Find the differences:

$$\Delta_i = |y_i - x_i|$$

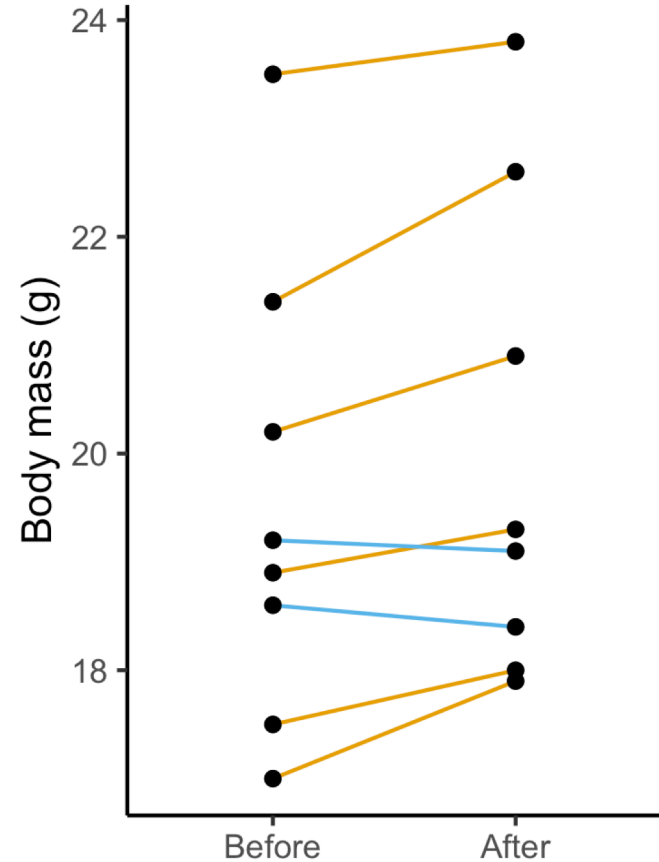
$$s_i = \text{sgn}(y_i - x_i)$$

- Order and rank the pairs according to Δ_i

R_i - rank of the i -th pair

- Test statistic:

$$W = \sum_{i=1}^n s_i R_i$$



Wilcoxon signed-rank test

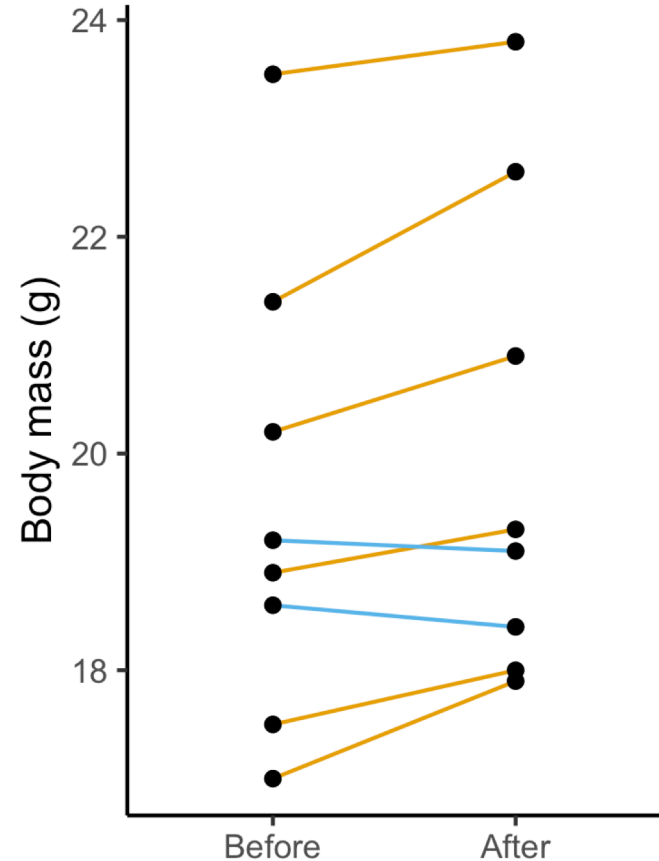
$$\Delta_i = |y_i - x_i|$$

$$s_i = \text{sgn}(y_i - x_i)$$

R_i - rank of the i -th pair

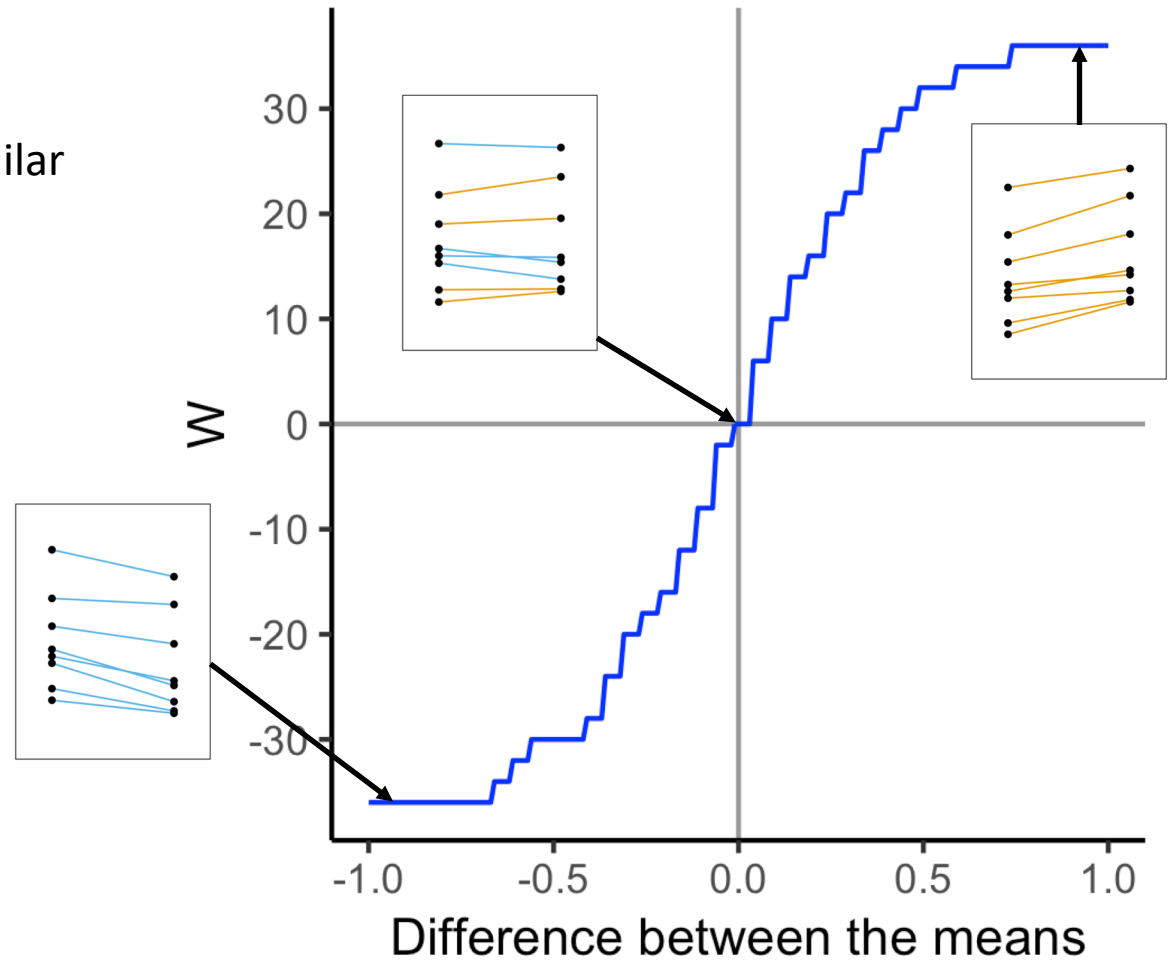
$$W = \sum_{i=1}^n s_i R_i$$

x_i	y_i	Δ_i	R_i	s_i	$s_i R_i$
19.2	19.1	0.1	1	-1	-1
18.6	18.4	0.2	2	-1	-2
23.5	23.8	0.3	3	1	3
18.9	19.3	0.4	4	1	4
17.5	18.0	0.5	5	1	5
20.2	20.9	0.7	6	1	6
17.0	17.9	0.9	7	1	7
21.4	22.6	1.2	8	1	8
					<hr/>
					30

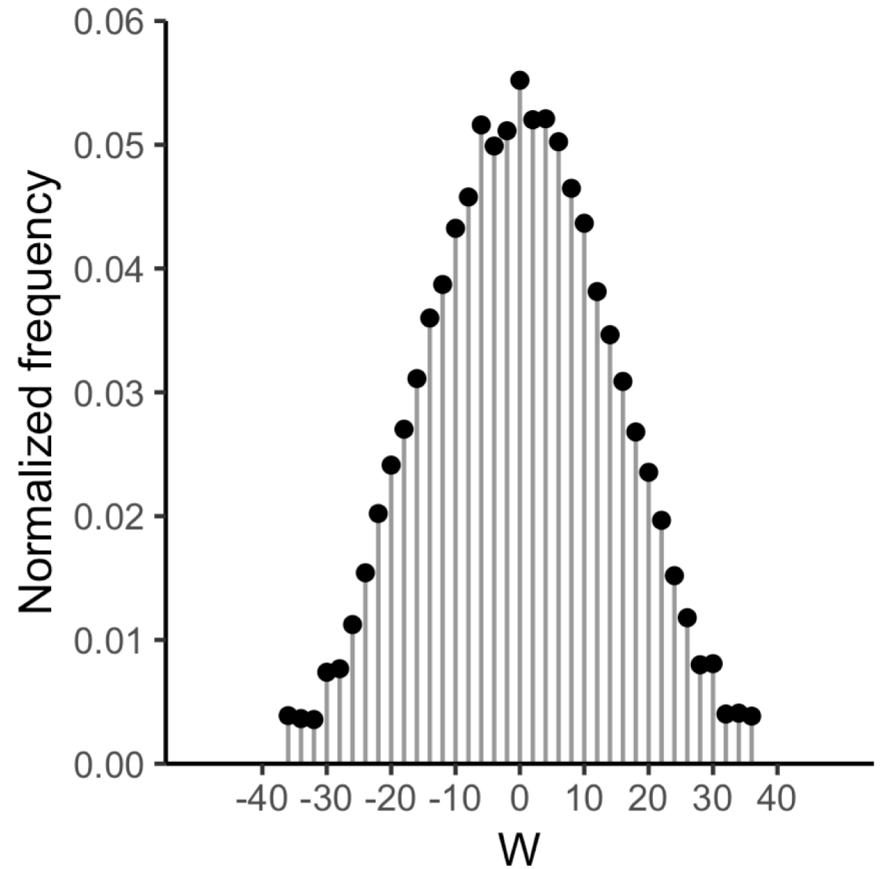
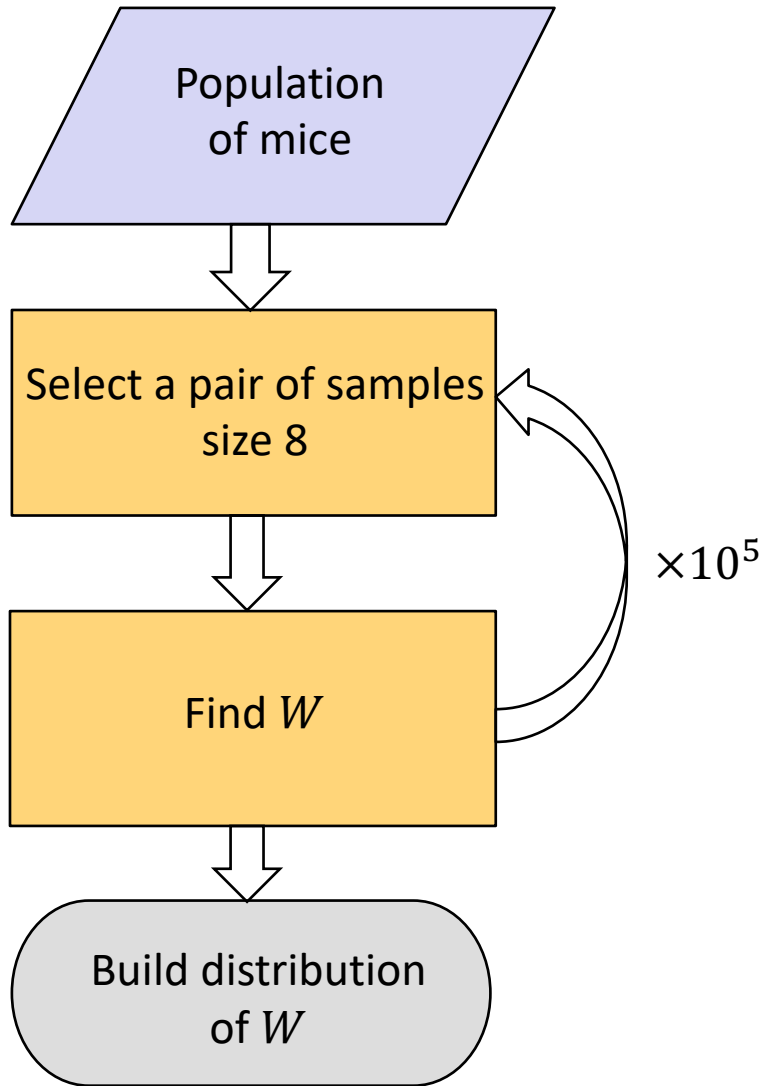


Wilcoxon signed-rank test

- W measures difference in location between pairs of points
- Direction is important
- $W = 0$ when samples most similar



Null distribution



Null distribution represents all random samples when the null hypothesis is true

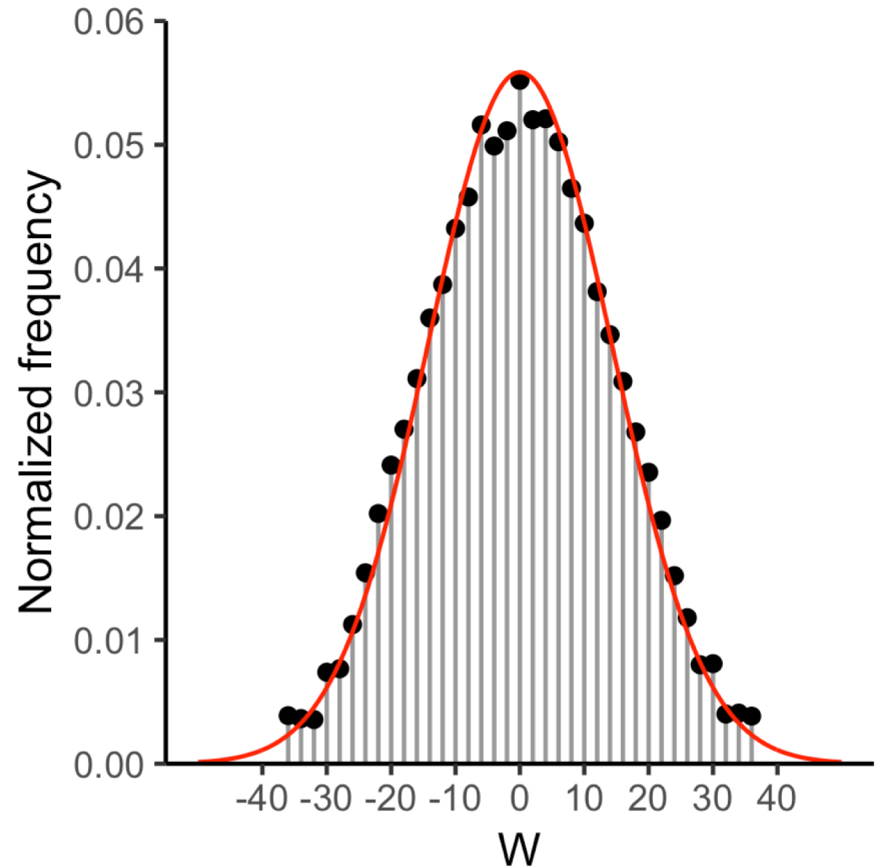
Null distribution

- For large samples W is approximately normally distributed with

$$\mu_W = 0$$

$$\sigma_W = \sqrt{\frac{n(n+1)(2n+1)}{6}}$$

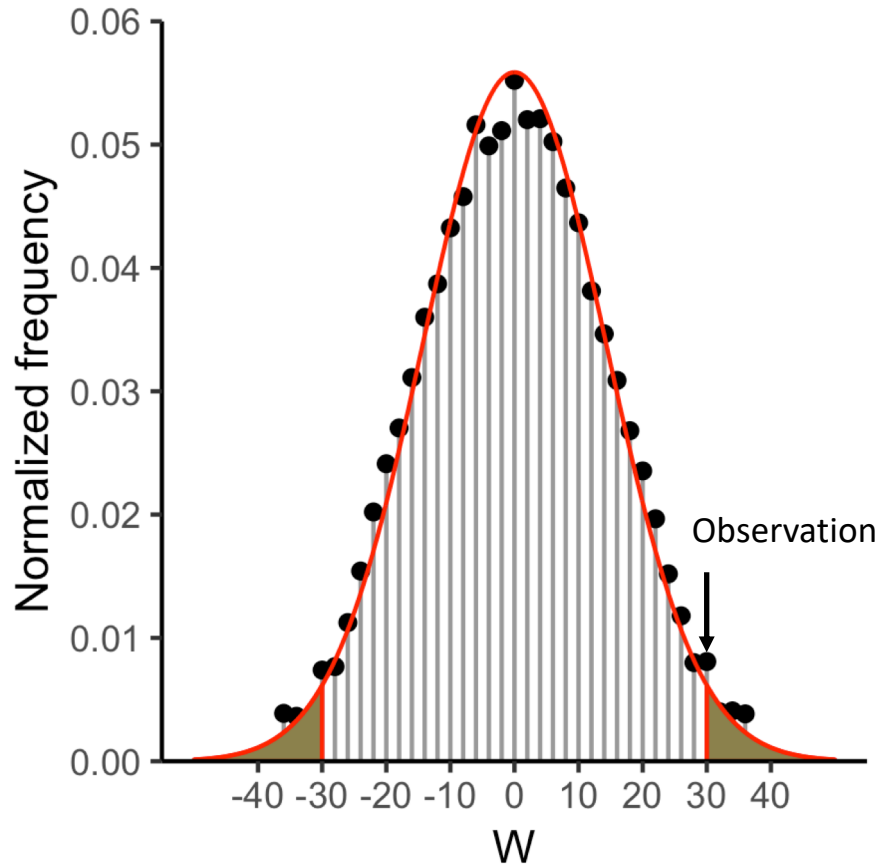
- For smaller samples exact solutions are available (tables or software)



$$n = 8$$

$$\sigma_W = \sqrt{\frac{8 \times 9 \times 17}{6}} = \sqrt{204} \approx 14.3$$

P-value



x_i	y_i	Δ_i	R_i	s_i	$s_i R_i$
19.2	19.1	0.1	1	-1	-1
18.6	18.4	0.2	2	-1	-2
23.5	23.8	0.3	3	1	3
18.9	19.3	0.4	4	1	4
17.5	18.0	0.5	5	1	5
20.2	20.9	0.7	6	1	6
17.0	17.9	0.9	7	1	7
21.4	22.6	1.2	8	1	8
					30

$$n = 8$$

$$W = 30$$

$$\sigma_W = 14.3$$

$$Z = W/\sigma_W = 2.10$$

$$p = 0.036$$

$$p_{\text{exact}} = 0.039$$

How to do it in R?

```
# Paired t-test  
> before <- c(21.4, 20.2, 23.5, 17.5, 18.6, 17.0, 18.9, 19.2)  
> after <- c(22.6, 20.9, 23.8, 18.0, 18.4, 17.9, 19.3, 19.1)  
> wilcox.test(before, after, paired=TRUE)
```

wilcoxon signed rank test

data: before and after

V = 3, p-value = 0.03906

alternative hypothesis: true location shift is not equal to 0

Wilcoxon signed-rank test: summary

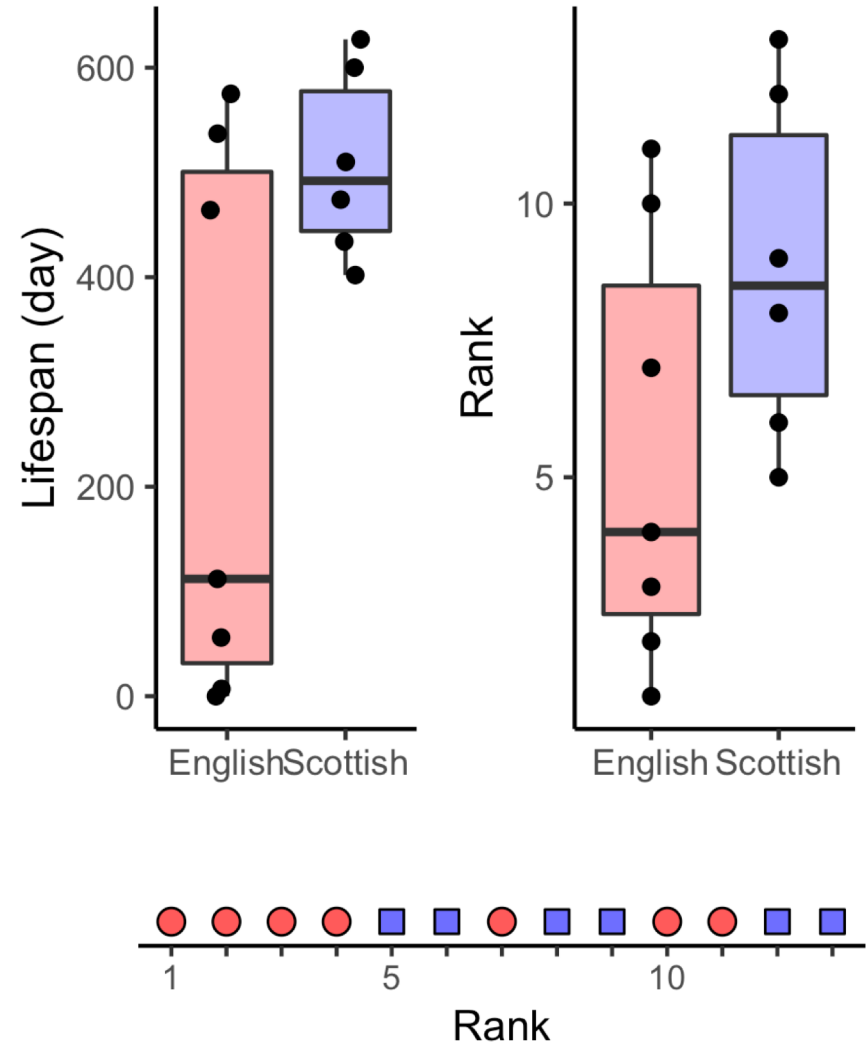
Input	Sample of n pairs of data (<i>before</i> and <i>after</i>) Values can be ordinal
Assumptions	Pairs should be random and independent
Usage	Discover change in individual points between <i>before</i> and <i>after</i>
Null hypothesis	There is no change between <i>before</i> and <i>after</i> is zero The difference between <i>before</i> and <i>after</i> follows a symmetric distribution around zero
Comments	Non-parametric counterpart of paired t-test Paired data only Doesn't care about distributions Not very useful for small samples

Kruskal-Wallis test

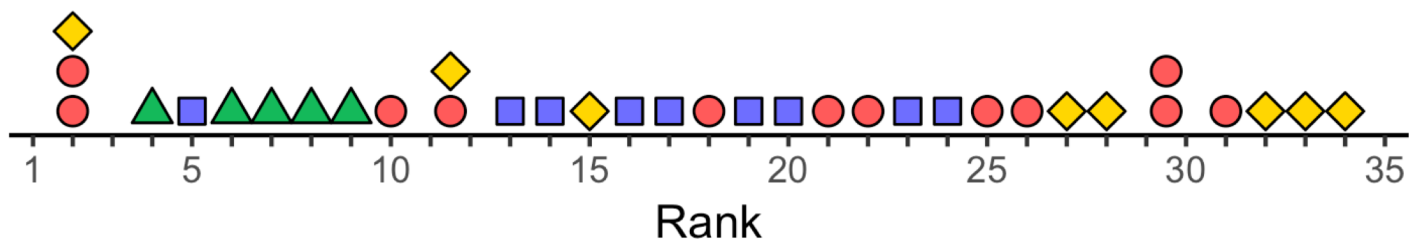
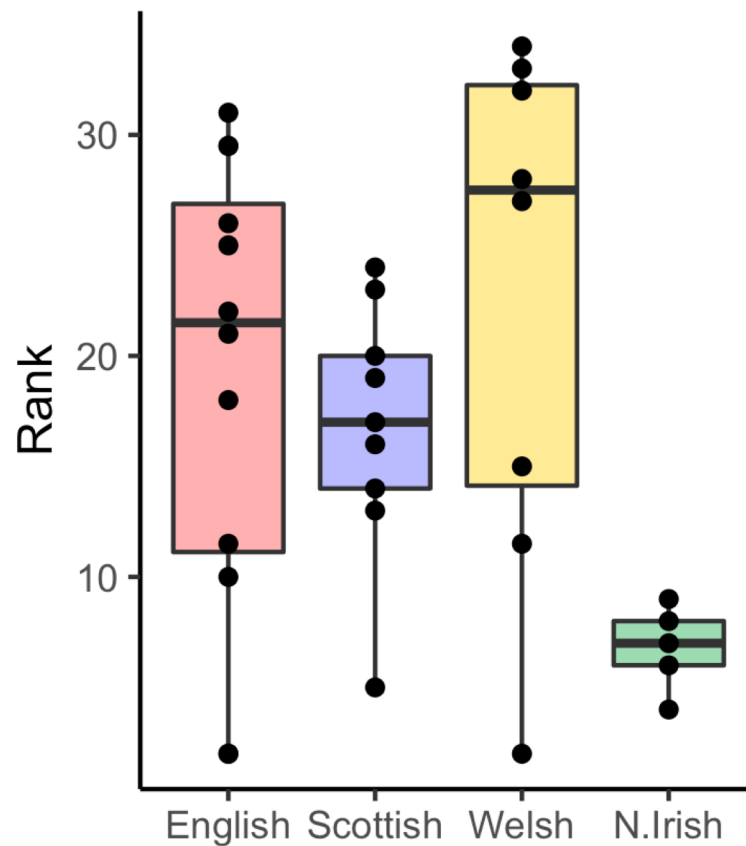
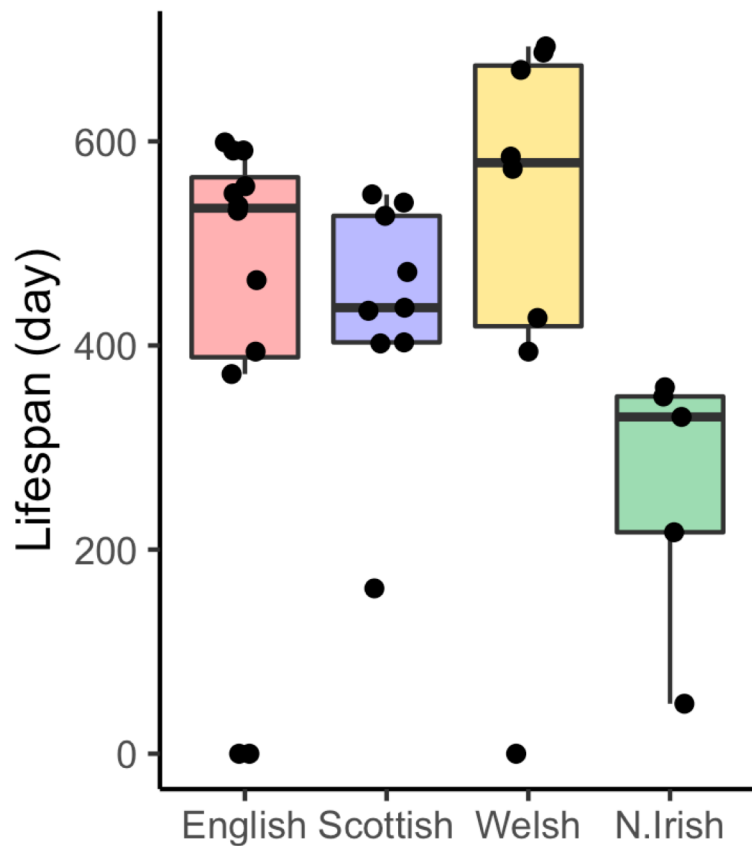
a nonparametric alternative to one-way ANOVA

Alternative formulation of the Mann-Whitney test

- Rank pooled data from the smallest to the largest
- Null hypothesis: both samples are randomly distributed between available rank slots
- Can be extended to more than 2 samples



Ranked ANOVA



Test statistic: use variance between groups

- Sum of square residuals

$$SS_B = \sum_{g=1}^n n_g (\bar{r}_g - \bar{r})^2$$

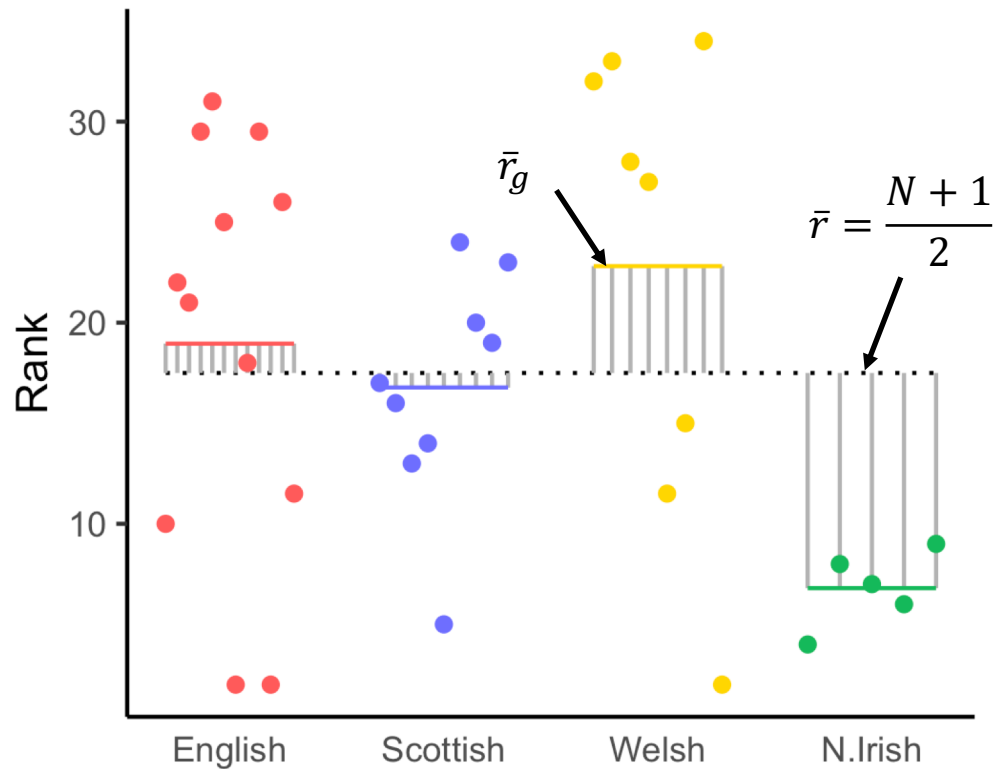
- Rank variance (ranks are 1, ..., N)

$$\sigma^2 = \frac{1}{12} N(N + 1)$$

- Test statistic

$$H = \frac{SS_B}{\sigma^2}$$

$$H = \frac{12}{N(N + 1)} \sum_{g=1}^n n_g \left(\bar{r}_g - \frac{N + 1}{2} \right)^2$$

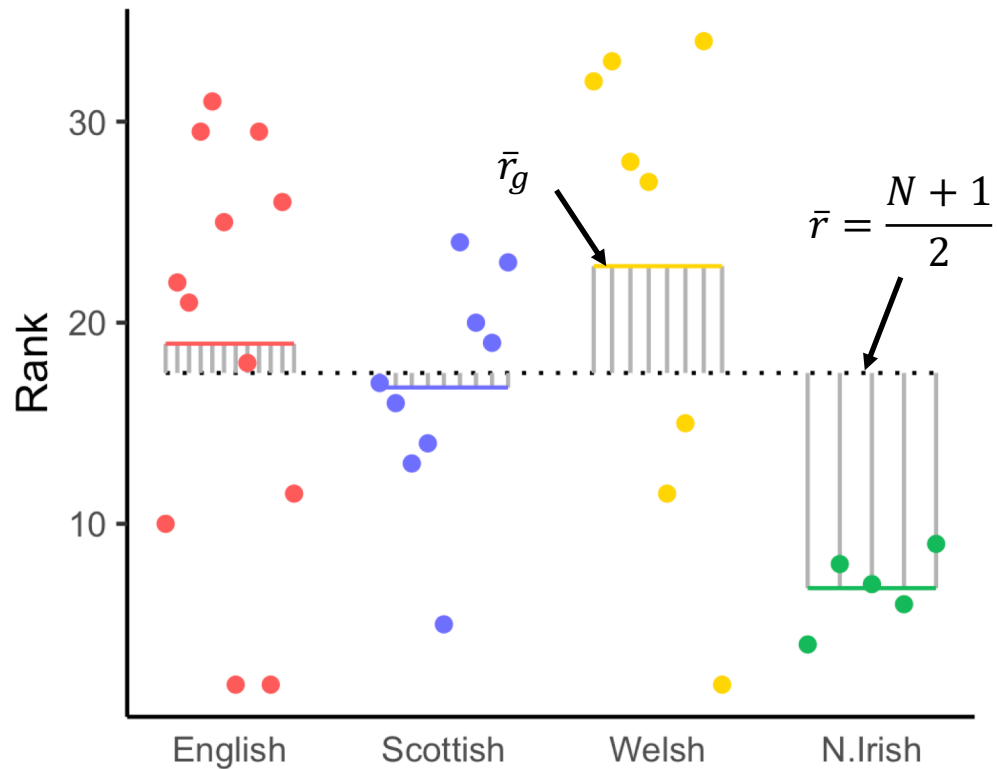


Test statistic

$$H = \frac{12}{N(N+1)} \sum_{g=1}^n n_g \left(\bar{r}_g - \frac{N+1}{2} \right)^2$$

- where
 - n_g – number of points in group g
 - \bar{r}_g – mean rank in group g
 - $\bar{r} = (N+1)/2$ – mean rank
 - N – number of all points
 - n – number of groups
- H is distributed with χ^2 distribution with $n - 1$ degrees of freedom
- Null hypothesis: mean group rank is the same as total mean rank

$$H_0: \bar{r}_g = \frac{N+1}{2}$$



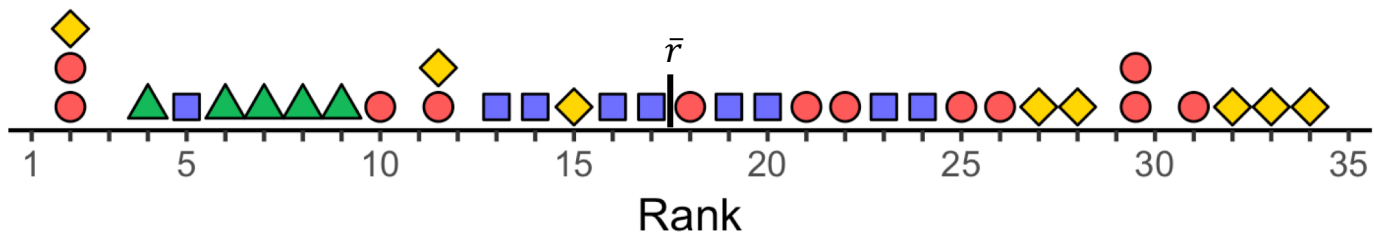
$$H = \frac{1}{\sigma^2} \sum_{g=1}^n n_g (\bar{r}_g - \bar{r})^2$$

$$\bar{r} = \frac{N + 1}{2} = 17.5$$

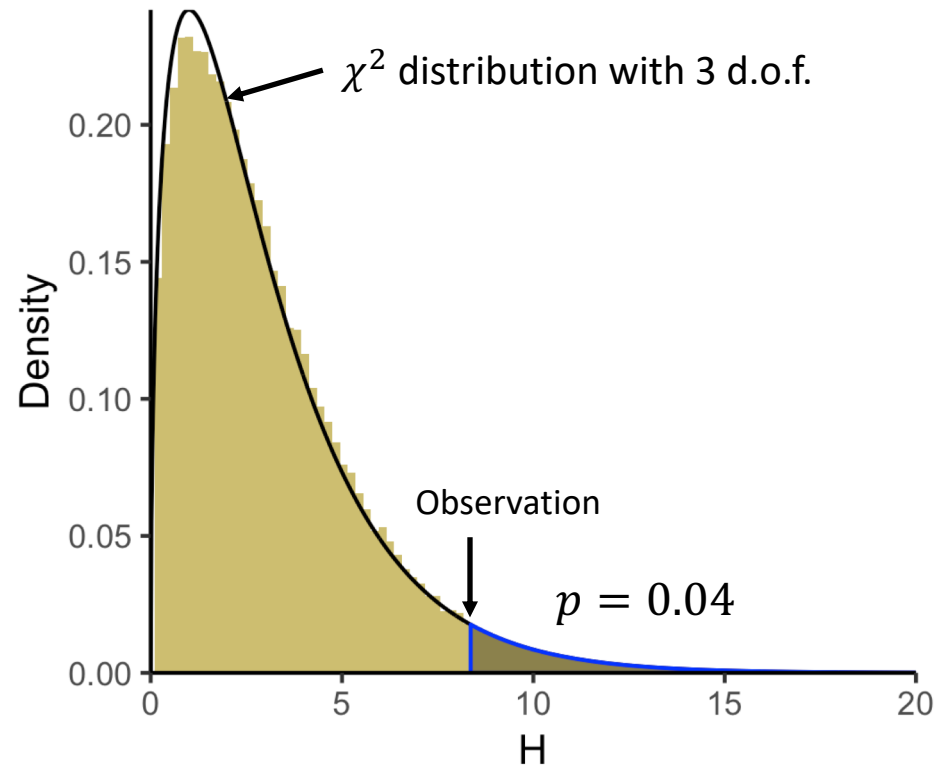
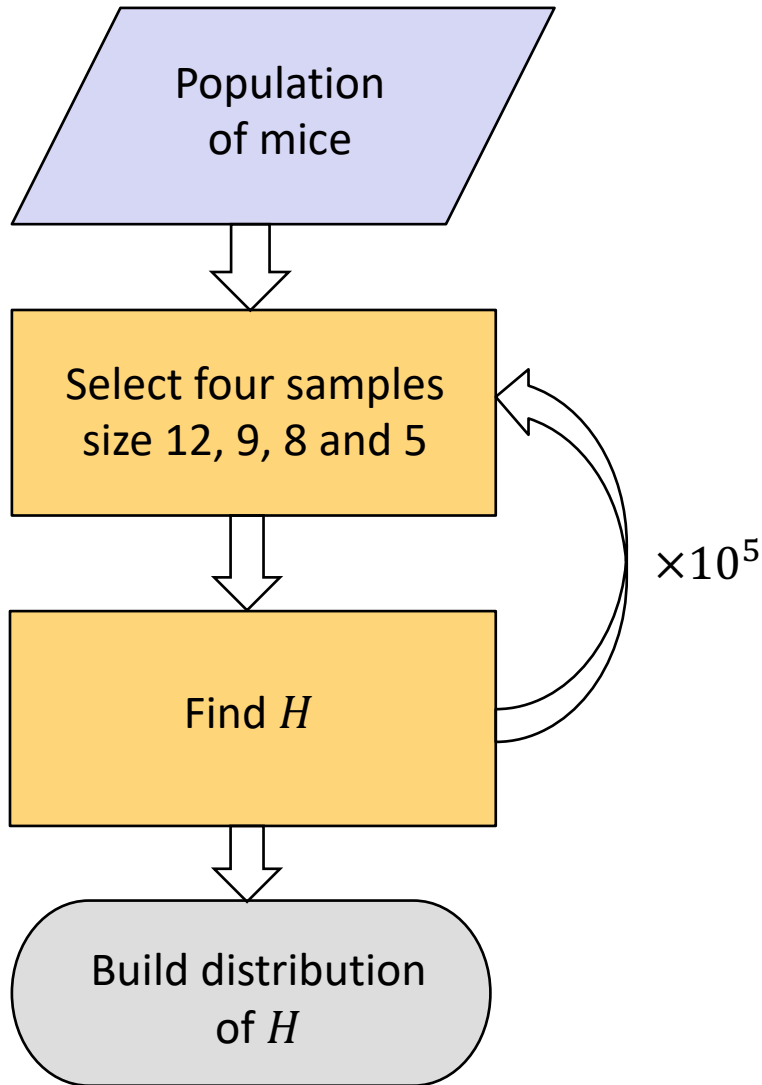
$$\sigma^2 = \frac{N(N + 1)}{12} = 99.2$$

		English	Scottish	Welsh	N. Irish
Number	n_g	12	9	8	5
Mean rank	\bar{r}_g	18.96	16.78	22.81	6.80
Contribution to H	$\frac{n_g(\bar{r}_g - \bar{r})^2}{\sigma^2}$	0.258	0.047	2.27	5.77

$$H = 8.36$$

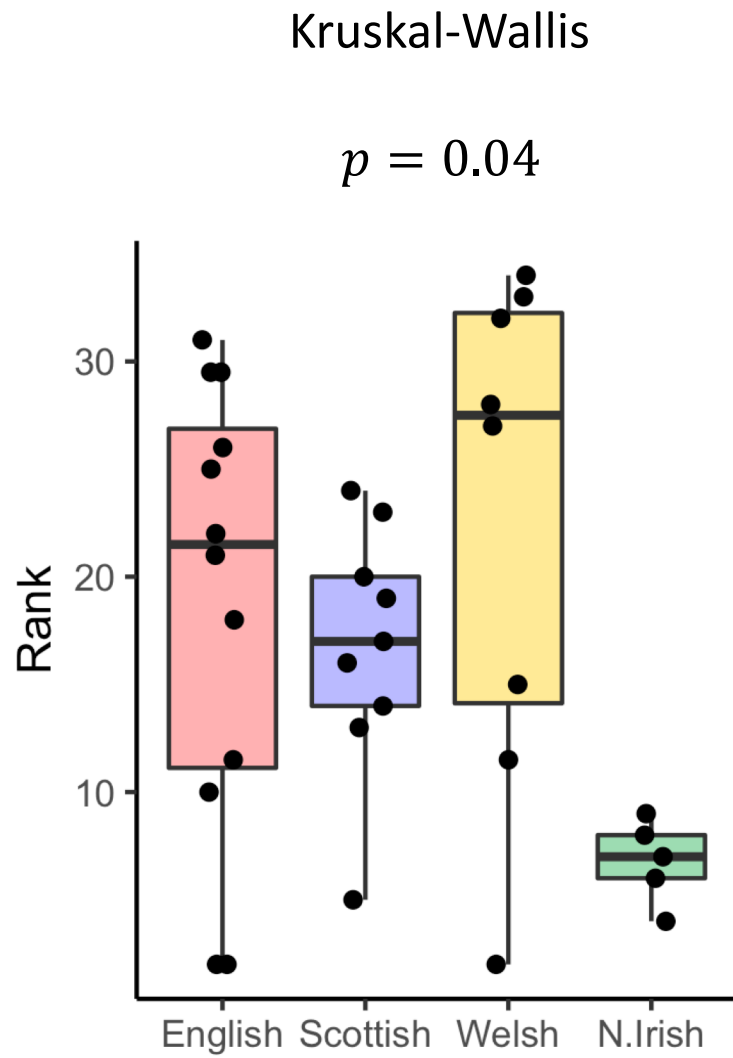
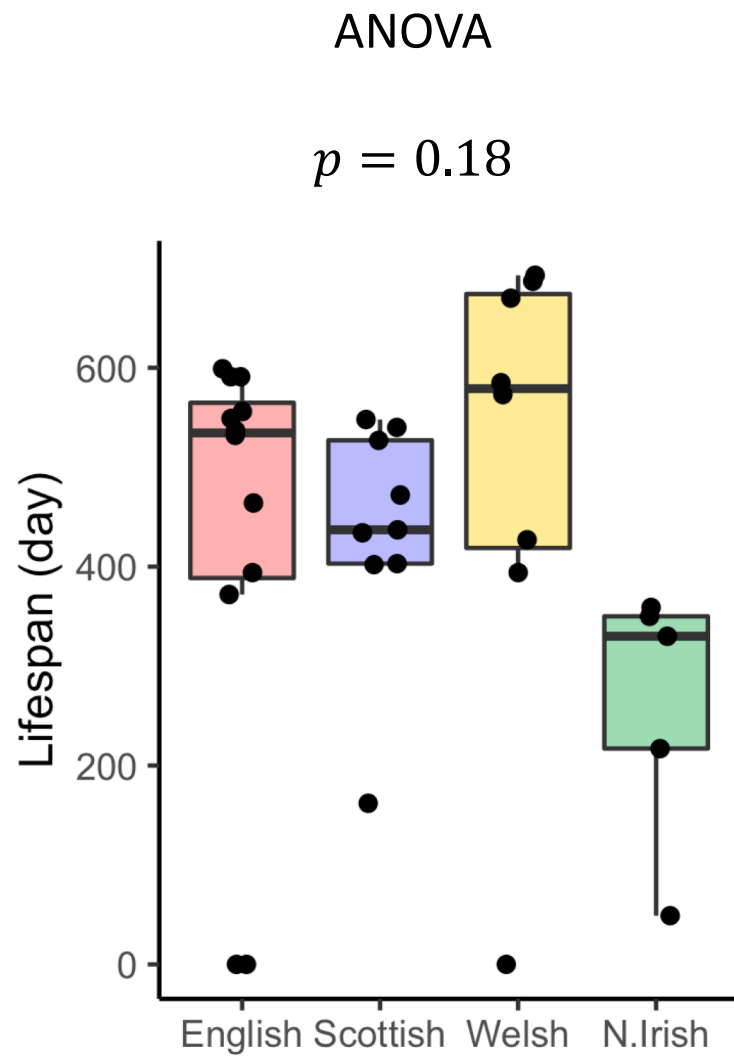


Null distribution



Null distribution represents all random samples when the null hypothesis is true

Comparison to ANOVA



How to do it in R?

```
> mice <- read.table('http://tiny.cc/mice_kruskal', header=TRUE)
> kruskal.test(Lifespan ~ Country, data=mice)
```

Kruskal-wallis rank sum test

data: Lifespan by Country

Kruskal-wallis chi-squared = 8.3617, df = 3, p-value = 0.0391

What about two-way test?

- Scheirer-Ray-Hare extension to Kruskal-Wallis test
- Briefly: replace values with ranks and carry out two-way ANOVA

Scheirer C.J., Ray W.S. and Hare N (1976), The Analysis of Ranked Data Derived from Completely Randomized Factorial Designs, *Biometrics*, **32**, 429-434

Kruskal-Wallis test: summary

Input	n samples of values N values divided into n groups
Assumptions	Samples are random and independent
Usage	Compare location and shape of n samples
Null hypothesis	Mean rank in each group is the same as total mean rank There is no change between groups
Comments	Doesn't care about distributions

Hand-outs available at <http://tiny.cc/statlec>