# Lecture 8: Non-parametric Comparison of Location

**GENOME 560** 

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#### Review

What do we mean by nonparametric?

What is a desirable location statistic for ordinal data?

What are NP equivalents of a one-sample t-test?

## Goals

 Comparing the medians of two samples using the Wilcoxon Rank Sum test

 Comparing the medians of many mutually independent samples using the Kruskal-Wallis test

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Pool N =  $n_x + n_y$  observations  $H_1: M_x \neq M_y$ 

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- The test statistic  $T_x$  is the sum of the ranks of X

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 Reject H<sub>0</sub> if T<sub>x</sub> is very large or very small compared to possible values of T<sub>x</sub> for n = N

Let's say we have measured a transcript level in 6 preoperative patients (X) and 3 post-operative patients (Y). Does the surgery change transcript levels?

 $H_0: M_x = M_y$ 

 $H_1: M_x \neq M_y$ 

$$X: (2.5, 3.0, 6.2, 9.1, 14.3, 14.7)$$
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Observation	Sample	Rank
2.5	X	
3	X	
6.2	X	
9.1	X	
14.3	Χ	
14.7	Χ	
14.1	Υ	
15.6	Υ	
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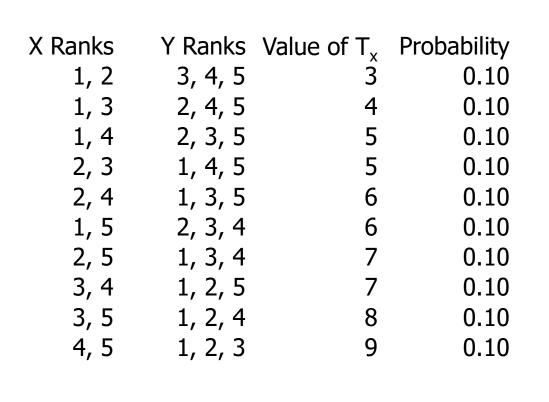
	<u> </u>	
Observation	Sample	Rank
2.5	Χ	1
3	X	2
6.2	X	3
9.1	X	4
14.1	Υ	5
14.3	X	6
14.7	X	7
15.6	Υ	8
16.7	Υ	9

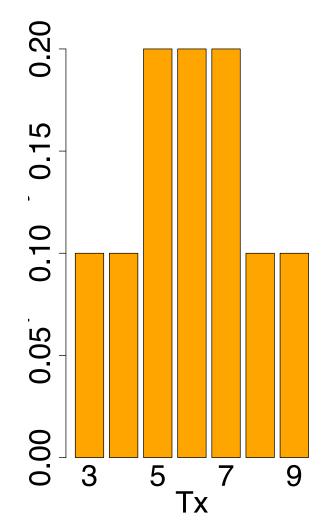
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		•					
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14	ł.1	Υ	5				
14	ł.3	X	6				
14	1.7	X	7				
15	5.6	Υ	8				
16	5.7	Υ	9				
$T_{:}$	x' = 1	(1+2)	2 + 3 +	-4 +	6+	7) =	23

- Consider a case where n<sub>x</sub> = 2 and n<sub>y</sub> = 3
- We know ranks must be 1, 2, 3, 4, 5
- Again, the issue is how to assign these ranks amongst the samples X and Y

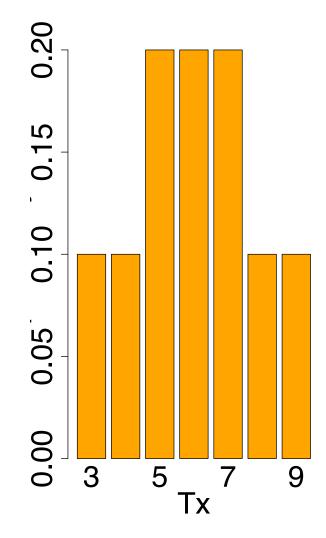
- Consider a case where  $n_x = 2$  and  $n_y = 3$
- We know ranks must be 1, 2, 3, 4, 5
- Again, the issue is how to assign these ranks amongst the samples X and Y
- There are  $\binom{5}{2} = 10$  ways of assigning five ranks to two samples
- Each way is equally likely under the null hypothesis so each has a probability of 10%





X Ranks	Y Ranks	Value of T <sub>x</sub>	Probability
1, 2	3, 4, 5	3	0.10
1, 3	2, 4, 5	4	0.10
1, 4	2, 3, 5	5	0.10
2, 3	1, 4, 5	5	0.10
2, 4	1, 3, 5	6	0.10
1, 5	2, 3, 4	6	0.10
2, 5	1, 3, 4	7	0.10
3, 4	1, 2, 5	7	0.10
3, 5	1, 2, 4	8	0.10
4, 5	1, 2, 3	9	0.10

$$P(T_x \le 4) = 0.2$$



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$T_x$	= (1 + 2)	2 + 3	+4+6+7) = 23
p =	= 0.095		

Accept H<sub>0</sub>

## Frank Wilcoxon



Wilcoxon lived from 1892 to 1965. He was a polymath, working as an oilman and a tree surgeon before training as a physical chemist, working in plant research and then in process control in industry. In a single paper in 1945 he published both tests that bear his name.

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- Rank the pooled samples
- Sum the ranks for each sample to get individual sample rank sums  $R_1, R_2, ..., R_k$

Under the null hypothesis, what should be true about the relationship between any two rank sums R<sub>i</sub>, R<sub>i</sub>?

$$R_1, R_2, ..., R_k$$

The sum of all the sample rank sums is

$$R_1 + R_2 + \dots + R_k = \frac{N(N+1)}{2}$$

## Kruskal-Wallis Test Outcome

• Given the way the test statistic/hypotheses are constructed, what does a rejection of  $H_0$  mean?

## Nonparametric Location Tests

- Can be used to perform one or two sample tests with fewer assumptions about the distribution from which the sample(s) are drawn
- Usage of sign and rank (rather than interval, as with parametric tests) enable this and confer other benefits
  - More robust (immune to outliers)
  - Can be used on ordinal data
- NP tests still have assumptions, and still must be used with care (e.g. zeroes for sign test, ties, similarity of distributions for rank-sum test)

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If we assume normality and identicality of variance, then a two sample t-test gives:

$$p = 0.02$$

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- Different nonparametric tests perform better or worse in this regard (efficiency)
- All will do better than their parametric counterparts when assumptions are violated
- The Mann-Whitney-Wilcoxon test is particularly good, giving up little power even for normally distributed data

## R Goals

Executing nonparametric tests in R

 Playing around with different distribution shapes and test assumptions

Examining effect size vs. test outcome

## Reading/Resources

- http://www.statsoft.com/Textbook/Nonparametric-Statistics/button/2
- http://sci2s.ugr.es/keel/pdf/algorithm/articulo/wilcoxon1 945.pdf
- http://www.mayo.edu/mayo-edu-docs/center-fortranslational-science-activities-documents/berd-5-6.pdf
- Nonparametric statistics: an introduction, Jean Gibbons (available online through UW libraries at http://goo.gl/NERixX)