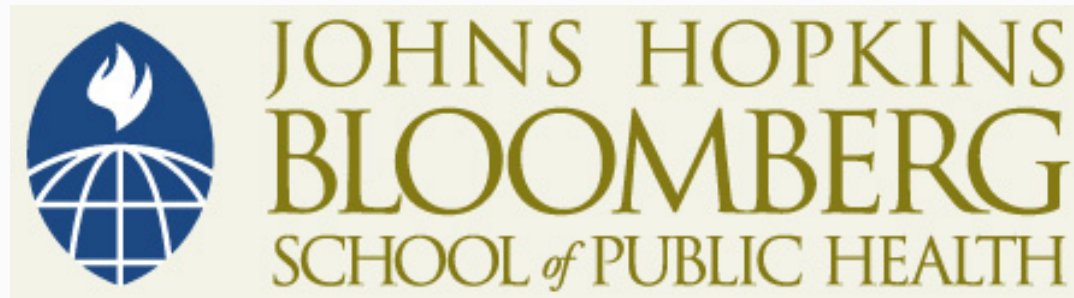


This work is licensed under a [Creative Commons Attribution-NonCommercial-ShareAlike License](https://creativecommons.org/licenses/by-nc-sa/4.0/). Your use of this material constitutes acceptance of that license and the conditions of use of materials on this site.



Copyright 2009, The Johns Hopkins University and John McGready. All rights reserved. Use of these materials permitted only in accordance with license rights granted. Materials provided "AS IS"; no representations or warranties provided. User assumes all responsibility for use, and all liability related thereto, and must independently review all materials for accuracy and efficacy. May contain materials owned by others. User is responsible for obtaining permissions for use from third parties as needed.



JOHNS HOPKINS
BLOOMBERG
SCHOOL *of* PUBLIC HEALTH

Section F (Optional)

Non-Parametric Analogue to the Two Sample t-test

Alternative to the Two Sample T-Test

- Nonparametric test for comparing two groups
- “Non-parametric” refers to a class of tests that do not assume anything about distribution of the data
- Nonparametric test for comparing two groups
 - Mann-Whitney Rank Sum Test (Wilcoxon Rank Sum Test)
 - Also called Mann-Whitney-Wilcoxon (a mouthful)
- Tries to answer the following question:
 - Are the two population distributions different?

Advantages

- Does not assume populations being compared are normally distributed
 - The two-sample t-test requires that assumption with very small samples sizes
- Uses only ranks
- Not sensitive to outliers

Disadvantage of the Nonparametric Test

- Nonparametric methods are often less sensitive (powerful) for finding true differences because they throw away information (they use only ranks)
- Need full data set, not just summary statistics
- Results do not include any confidence intervals quantifying range of possibility for true difference between populations

Example: Health Education Study

- Evaluate an intervention to educate high school students about health and lifestyle over a two-month period
- 10 students randomized to “intervention” or “control” group
- $x = \text{post test score} - \text{pre-test score}$ is outcome to compare between the intervention and control groups

Example: Health Education Study

- x = post- pretest score for both groups
- Intervention (I) 5 0 7 2 19
- Control (C) 6 -5 -6 1 4
 - Only five individuals in each sample!!!
 - We want to compare the control and intervention groups to assess whether the “improvement” (post-pre) in scores are different, taking random sampling error into account

Example: Health Education Study

- With such a small sample size, we need to be sure score improvements are normally distributed if we want to use t-test (BIG assumption)
- Possible approach:
 - Mann-Whitney-Wilcoxon non-parametric test!

Example: Health Education Study

- First step—rank the pooled data (ignore groupings)

	-6	-5	0	1	2	4	5	5	7	19
— Rank	1	2	3	4	5	6	7	8	9	10

Example: Health Education Study

- Second step—“reattach” group status

	-6	-5	0	1	2	4	5	5	7	19
— Rank	1	2	3	4	5	6	7	8	9	10
— Group	C	C	I	C	I	C	I	C	I	I

Example: Health Education Study

- Find the average rank in each of the two groups
- Intervention group average rank

$$\frac{3 + 5 + 6 + 9 + 10}{5} = 6.8$$

- Control group average rank

$$\frac{1 + 2 + 4 + 6 + 8}{5} = 4.2$$

Example: Health Education Study

- Statisticians have developed formulas and tables to determine the probability of observing such an extreme discrepancy in ranks (6.8 vs. 4.2) by chance alone
 - This is the p-value
- In the health education study, the p-value was .17
 - The interpretation is that the Mann-Whitney test did not show any significant difference in test score “improvement” between the intervention and control group ($p = .17$)

Notes

- The two-sample t-test would give a different answer ($p = .14$)
- Different statistical procedures can give different p-values
- If the largest observation, 19, was changed, the p-value based on the Mann-Whitney test would not change but the two-sample t-test would change

Notes

- The t-test or the nonparametric test?
 - Statisticians will not always agree, but there are some guidelines
 - Use non-parametric test if sample size is small and you have no reason to believe data is “well behaved” (normally distributed)
 - Only “ranks” available

Using Stata to Perform Mann-Whitney-Wilcoxon

- Data, as entered

```
. list diff int_cntrl
```

```
      +-----+
      | diff   int_cntrl |
      +-----+
  1. |      4           0 |
  2. |      1           0 |
  3. |     -6           0 |
  4. |     -5           0 |
  5. |      6           0 |
      +-----+
  6. |     19           1 |
  7. |      2           1 |
  8. |      7           1 |
  9. |      0           1 |
 10. |      5           1 |
      +-----+
```

Using Stata to Perform Mann-Whitney-Wilcoxon

- “ranksum” command
 - Syntax:
 - ▶ `ranksum varname, by(group_var)`

```
. ranksum diff, by( int_cntrl)

Two-sample Wilcoxon rank-sum (Mann-Whitney) test

      int_cntrl |      obs   rank sum   expected
-----+-----
           0 |         5        21       27.5
           1 |         5        34       27.5
-----+-----
    combined |        10        55        55

unadjusted variance      22.92
adjustment for ties      0.00
-----
adjusted variance      22.92

Ho: diff(int_cntrl==0) = diff(int_cntrl==1)
      z =  -1.358
      Prob > |z| =  0.1745
```


Using Stata to Perform Mann-Whitney-Wilcoxon

- “ranksum” command

- Syntax:

- ▶ `ranksum varname, by(group_var)`

```
. ranksum diff, by( int_cntrl)
```

```
Two-sample Wilcoxon rank-sum (Mann-Whitney) test
```

int_cntrl	obs	rank sum	expected
0	5	21	27.5
1	5	34	27.5
combined	10	55	55

```
unadjusted variance      22.92
```

```
adjustment for ties      0.00
```

```
adjusted variance      22.92
```

```
Ho: diff(int_cn~l==0) = diff(int_cn~l==1)
```

```
z = -1.358
```

```
Prob > |z| = 0.1745
```

Using Stata to Perform t-test

- “*ttest*” command without “*i*” on end when data already in Stata
 - Syntax:

▶ *ttest varname, by(group_var)*

```
. ttest diff, by( int_cntrl)
```

Two-sample t test with equal variances

Group	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
0	5	0	2.387467	5.338539	-6.628672	6.628672
1	5	6.6	3.325658	7.436397	-2.633506	15.83351
combined	10	3.3	2.221361	7.02456	-1.725068	8.325068
diff		-6.6	4.093898		-16.04055	2.840545

```
diff = mean(0) - mean(1)                                t = -1.6122
Ho: diff = 0                                           degrees of freedom = 8
```

```
Ha: diff < 0
Pr(T < t) = 0.0728
```

```
Ha: diff != 0
Pr(|T| > |t|) = 0.1456
```

```
Ha: diff > 0
Pr(T > t) = 0.9272
```

Summary: Educational Intervention Example

■ Statistical methods

- 10 high school students were randomized to either receive a two-month health and lifestyle education program (or no program)
- Each student was administered a test regarding health and lifestyle issues prior to randomization (and after the two-month period)

Summary: Educational Intervention Example

■ Statistical methods

- Differences in the two test scores (after-before) were computed for each student
- Mean and median test score changes were computed for each of the two study groups
- A Mann-Whitney rank sum test was used to determine if there was a statistically significant difference in test score change between the intervention and control groups at the end of the two-month study period

Summary: Educational Intervention Example

■ Result

- Participants randomized to the educational intervention scored a median five points higher on the test given at the end of the two-month study period, as compared to the test administered prior to the intervention
- Participants randomized to receive no educational intervention scored a median one point higher on the test given at the end of the two-month study period
- The difference in test score improvements between the intervention and control groups was not statistically significant ($p = .17$)