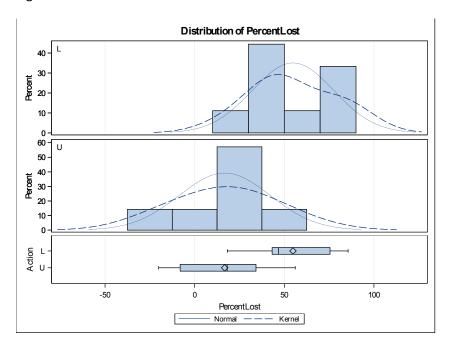
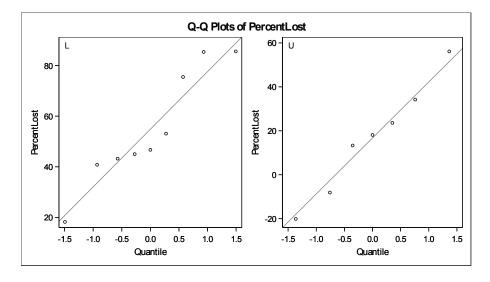
## Question 2)

## **Assumptions**

Assuming one knows little enough about the **normalcy** and **variance** of the population distribution to warrant using a non-parametric test like the Rank-Sum, a potential outlier in the logged data (value = 18.2) may also make Rank-Sum appropriate in this case. An initial t-test—boxplots and qqplots for which are produced below (Figure 1)—was run to preliminarily determine normalcy and variance. Furthermore, the assumption of observation **independence** should be adequately satisfied by the "randomly located transect patterns".

Figure 1:





#### **Analysis**

To test for the impact of logging on fire-affected plots, a Wilcoxon Rank-Sum test was run to test the *null-hypothesis* (*H*<sub>0</sub>) that: there is **no difference** between logged and unlogged plots **in the mean percentage of seedlings lost**:

$$\mu_L - \mu_U = 0$$

The *alternative-hypothesis* ( $H_A$ ) would be that: the mean percentage of seedlings lost is **greater** in logged plots than in unlogged plots:

$$\mu_L - \mu_U > 0$$

Assuming the null-hypothesis, statistical software determined that, for this test,  $S_{\text{stat}} = 36$ . The corresponding Z-scores (**including continuity correction**) and p-values were as follows:

$$Z = -2.4346 \rightarrow p$$
-value (one-sided) = .0075

On this basis, one would **reject the null-hypothesis** ( $H_0$ ), determining that: there is statistically significant evidence to suggest that the **mean percentage of seedlings lost is greater in logged plots** than in unlogged plots affected by fire.

The associated **99% confidence interval,** according to **Monte Carlo estimates** for the Exact Test, is as follows: **[.0043, .0083]**.

Given the evidence for this observational study—where random samples are selected from existing distinct populations of logged and unlogged fire-affected plots—inferences regarding the difference in percentage of seedlings lost **can be drawn to the populations** being examined. However, as assignment to groups (logged and unlogged) was not randomized, inferences regarding **causal relations cannot be drawn.** 

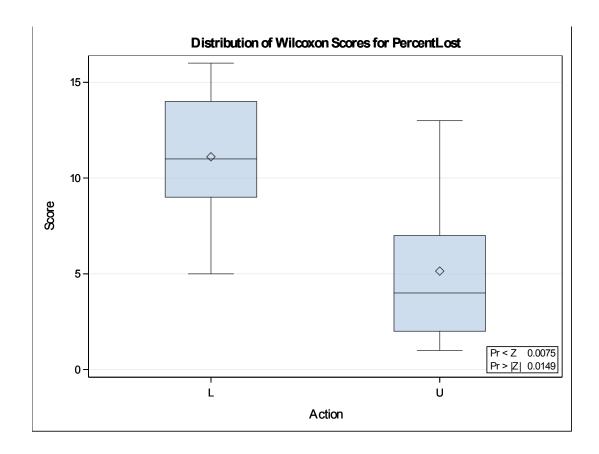
Tables and boxplots for the results are produced below (Figure 2):

Figure 2:

Wilcoxon Scores (Rank Sums) for Variable PercentLost Classified by Variable Action								
Action	ion Sum of Expected Std Dev Mea Under H0 Under H0 Scores							
L	9	100.0	76.50	9.447222	11.111111			
U	7	36.0	59.50	9.447222	5.142857			

Wilcoxon Two-	Wilcoxon Two-Sample Test					
Statistic (S)	36.0000					
Normal Approximation						
Z	-2.4346					
One-Sided Pr < Z	0.0075					
Two-Sided Pr >  Z	0.0149					
t Approximation						
One-Sided Pr < Z	0.0139					
Two-Sided Pr >  Z	0.0279					
Z includes a continuity correction of 0.5.						

Monte Carlo Estimates for the Test	Exact
One-Sided Pr <= S	
Estimate	0.0063
99% Lower Conf Limit	0.0043
99% Upper Conf Limit	0.0083
Two-Sided Pr >=  S - Mean	
Estimate	0.0120
99% Lower Conf Limit	0.0092
99% Upper Conf Limit	0.0148
Number of Samples	10000
Initial Seed	1234



## Question 3)

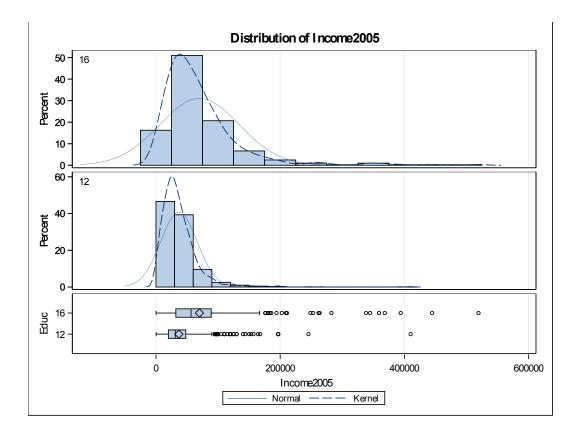
## Parts A + B)

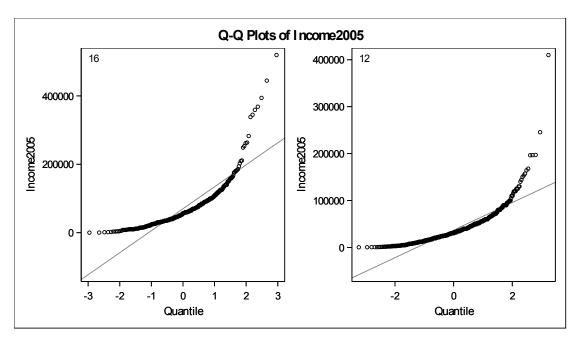
#### **Assumptions**

An initial t-test—boxplots and qqplots for which are produced below (Figure 3)—was run to preliminarily determine normalcy and variance. The data are clearly **right-skewed** and have **unequal variances**; of the two groups, the one with the larger mean also has the larger variance. Moreover, judging from the description of the sampling procedure for the study's design, the assumption of observation **independence** should be adequately satisfied. Although these conditions are suited for a logarithmic transformation, knowledge of the CLT tells us that the study's adequate sample size (n = 1426) should ensure reasonable results, compensating somewhat for skewedness and variance. Figures for assumption testing (Figure 3) are produced below:

Figure 3:

Educ	N	Mean	Std Dev	Std Err	Minimum	Maximum
16	406	69997.0	64256.8	3189.0	200.0	519340
12	1020	36864.9	29369.7	919.6	300.0	410008
Diff (1-2)		33132.1	42326.9	2483.8		





#### **Analysis**

To test for the impact of education on income level, a one-sided Welch's two-sample t-test was run to test the *null-hypothesis* ( $H_0$ ) that: there is **no difference** in mean income level between those with 16 and those with 12 years of education:

$$\mu_{16} - \mu_{12} = 0$$

The *alternative-hypothesis* ( $H_A$ ) would be that: mean income level is **greater** for those with 16 years of education than it is for those with 12 years:

$$\mu_{16} - \mu_{U12} > 0$$

Assuming the null-hypothesis and unequal variance (**Satterthwaite**), with a sample size of n = 1426, with df = 1424, and  $\alpha = .05$ , the critical value would be:  $\mathbf{t}_{crit} = \mathbf{1}.961$ .

Statistical software determined that, for this test  $t_{\text{stat}} \approx 9.98$ .

Correspondingly, the **p-value (one-sided)** = **.0001**.

On this basis, one would **reject the null-hypothesis** ( $H_0$ ), determining that: there is statistically significant evidence to suggest that the mean income for those with 16 years of education is **greater** than it is for those with 12 years.

The associated **95% confidence interval** for the difference in means, providing **two-sides** for intuitive clarity, is as follows: [**26610.40**, **39653.80**].

#### Part C)

Given the design description and the evidence for this observational study—in which neither sampling nor allocation is randomized, relying on a subset of those that voluntarily responded to a survey—inferences can only be drawn to the population of 41-49 year old survey respondents, though this limited scope of inference may still be informative. However, inferences regarding causal relations cannot be drawn.

Further tables for the results (Figure 4) are produced below:

Figure 4:

Educ	Method	Mean	95% CL Mean		Std Dev	95% (	CL Std ev
16		69997.0	63727.9	76266.1	64256.8	60120.1	69009.5
12		36864.9	35060.4	38669.4	29369.7	28148.2	30702.9
Diff (1-2)	Pooled	33132.1	28259.8	38004.3	42326.9	40828.0	43940.9
Diff (1-2)	Satterthwaite	33132.1	26610.4	39653.8			

Method	Variances	DF	t Value	Pr >  t
Pooled	Equal	1424	13.34	<.0001
Satterthwaite	Unequal	473.85	9.98	<.0001

Equality of Variances						
Method Num DF Den DF F Value Pr > F						
Folded F	405	1019	4.79	<.0001		

### Part D)

Comparing these results with the numbers, tables, and figures produced by the log-transformed analysis prepared for HW3, it is arguable that the **log-transformed analysis is more appropriate**. Although the assumption of independence holds for both kinds of tests, one should note that the data are very right-skewed, and the variance is higher for the group with the higher mean. As the **median** is more resistant to such skewedness than the **mean**, a log transformation—which provides inference on the former parameter—is especially suitable for this kind of data.

# Question 4)

Part A)

## Rank Sum Figures

All	Group	Order	Rank			
18.8	N	1	1			
20	N	2	2			
20.1	N	3	3			
20.9	N	4	4.5			
20.9	N	5	4.5			
21.4	N	6	6			
22	Т	7	7			
22.7	N	8	8			
22.9	N	9	9			
23	Т	10	10			
24.5	Т	11	11			
25.8	Т	12	12			
30	Т	13	13			
37.6	Т	14	14			
38.5	Т	15	15			
				T-Stat	82	
				n1	7	
				R-bar	8	
				Mean(T)	56	
				n2	8	
				S(R)	4.468141	
				SD(T)	8.633269	
				Cont. Correction	-0.5	
				Z-Stat	2.95369	
				P-val (1-sided)	(199843)	= .00157
				Crit Val	1.771	
				df	13	
				alpha	0.05	

## Conclusion

On the basis of the above evidence, one would **reject the null-hypothesis** that there is no difference in the mean metabolic expenditures (kcal/kg/day) of patients with and without trauma:

[**P-val** (one-sided) = .00157]

## Part B)

Statistical software was used to verify the rank-sum and the Z-statistic above, producing the following tables (Figure 5):

Figure 5:

Wilcoxon Scores (Rank Sums) for Variable Measure Classified by Variable Trauma								
Trauma	Sum of Expected Std Dev Mear N Scores Under H0 Under H0 Score							
N	8	38.0	64.0	8.633269	4.750000			
T 7 82.0 56.0 8.633269 11.714286								
	Average scores were used for ties.							

Wilcoxon Two-Sample Test				
Statistic	82.0000			
Normal Approximation				
Z	2.9537			
One-Sided Pr > Z	0.0016			
Two-Sided Pr >  Z	0.0031			
t Approximation				
One-Sided Pr > Z	0.0052			
Two-Sided Pr >  Z	0.0105			
Z includes a continuity	y correction of 0.5.			

Kruskal-Wallis Test			
Chi-Square	9.0698		
DF	1		
Pr > Chi-Square	0.0026		

Rounding variations aside, the statistical software, which uses a continuity correction of (.5) by default, found a **one-sided p-value (.0016)** roughly equal to that which was identified by hand **(.00157)**. Furthermore, the software verified the value of the **Z-statistic (82)**.

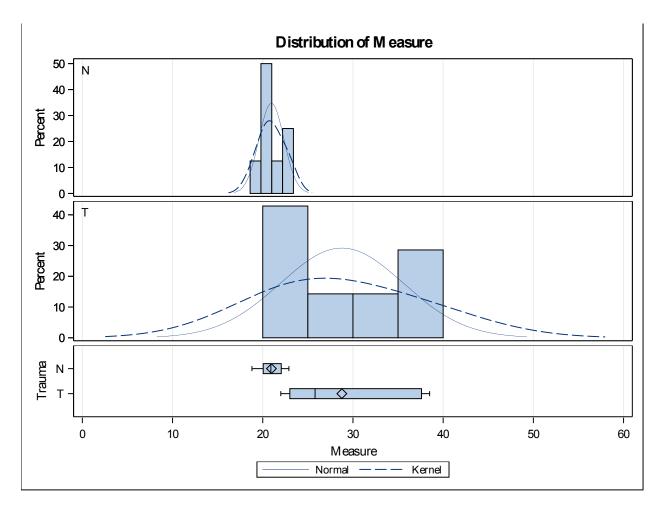
## Parts C.i and C.ii)

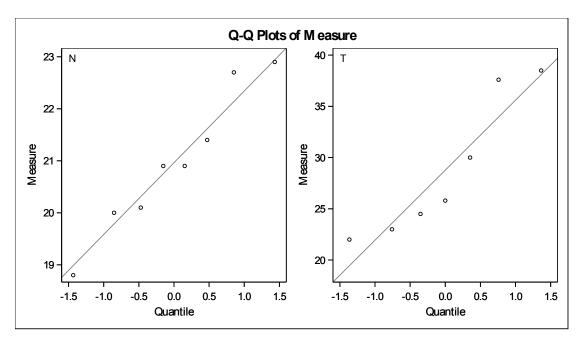
## **Assumptions**

An initial t-test—histograms, box-plots and qq-plots for which are produced below (Figure 6)—was run to preliminarily determine normalcy and variance. Although the data appear **roughly normal**, they have substantially **unequal variances**; of the two groups, the one with the larger mean also has the larger variance. Moreover, judging from the sparse description of study's design, one can assume that observation **independence** is adequately satisfied. However, since the sample size is not especially large (n = 15), it is difficult to make assumptions regarding the normalcy of the distribution for the population being examined. For such reasons, a rank-sum test would be appropriate in this case.

Figure 6:

Trauma	N	Mean	Std Dev	Std Err	Minimum	Maximum
N	8	20.9625	1.3794	0.4877	18.8000	22.9000
T	7	28.7714	6.8354	2.5835	22.0000	38.5000
Diff (1-2)		-7.8089	4.7528	2.4598		





Parts C.iii. and C.iv)

#### **Analysis**

To test for the impact of trauma on metabolic expenditure, a Wilcoxon Rank-Sum test was run to test the *null-hypothesis* ( $H_0$ ) that: there is **no difference** in mean metabolic expenditure between those with and without trauma:

$$\mu_T - \mu_N = 0$$

The *alternative-hypothesis* ( $H_A$ ) would be that: mean metabolic expenditure is **greater** for those with trauma than for those without:

$$\mu_T - \mu_N > 0$$

Assuming the null-hypothesis, with a sample size of n = 15, with df = 13, and  $\alpha = .05$ , the critical value would be  $\mathbf{S}_{crit} = \mathbf{8}$ .

Statistical software determined that, for this test,  $S_{\text{stat}} = 82$ .

The corresponding Z-scores (including continuity correction) and p-values were as follows:

## Z = 2.9537→ p-value (one-sided) = .0016

On this basis, one would **reject the null-hypothesis** ( $H_0$ ), determining that: there is statistically significant evidence to suggest that **mean metabolic expenditure** is **greater for those with trauma** than it is for those without.

The associated **95% confidence interval for Location Shift (Trauma - Nontrauma),** according to the **Hodges-Lehmann Estimation** for the Exact Test, is as follows: **[1.900, 16.7000]**, with a midpoint of **(9.3000)**.

Given the evidence for this sparsely described observational study—in which neither sampling nor allocation appear to be randomized, and where collections of available units are selected from existing distinct populations—inferences regarding the difference in mean metabolic expenditure can only be drawn to the particular population of trauma patients examined by the study, though the results are still informative within that limited scope of inference. However, inferences regarding causal relations cannot be drawn.

A table for the Hodges-Lehman Estimation (Figure 7) is included below:

Figure 7:

Hodges-Lehmann Estimation							
L	Location Shift (T - N) 5.3000						
Туре	95% Confidence Limits		Interval Midpoint	Asymptotic Standard Error			
Asymptotic (Moses)	1.9000	16.7000	9.3000	3.7756			
Exact	1.9000	16.7000	9.3000				

## Question 5) Part A) Signed Rank Figures

Child	Before	After	Difference	Ordered Mag	Order	Rank	Ranks (+)	Ranks (-)		
1	. 85	75	10	5	1	1.5	1.5			
2	70	50	20	5	2	1.5		1.5		
3	40	50	-10	10	3	4		4		
4	65	40	15	10	4	4	4			
5	80	20	60	10	5	4	4			
$\epsilon$	75	65	10	15	6	6	6			
7	55	40	5	20	7	7	7			
8	20	25	-5	40	8	8	8			
g	70	30	40	60	9	9	9			
						SR-Stat	39.5		(SR-Stat) - H0 Expected	17
						n1	9			
						Mean(S) (norm approx	22.5			
						SD(S) (norm approx.)	8.440972			
						Cont. Corr	-0.5			
						Z-Stat	1.954751			
						P-val (norm approx.)	(19747) =	= .0253		
						Crit-Val	1.86			
						df	8			
						alpha	0.05			
						Poss. Assignments	512			
						# as extreme	13			

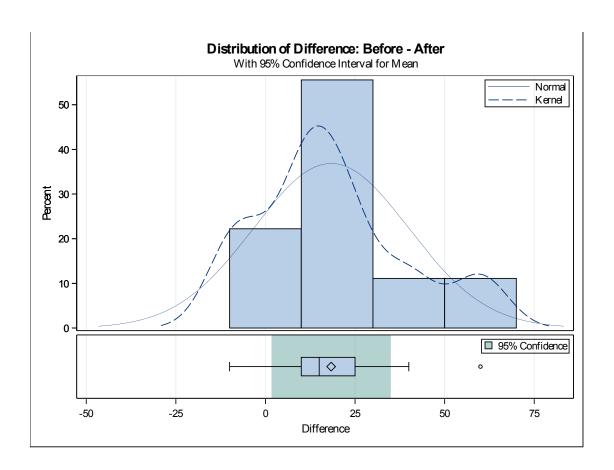
## Part B)

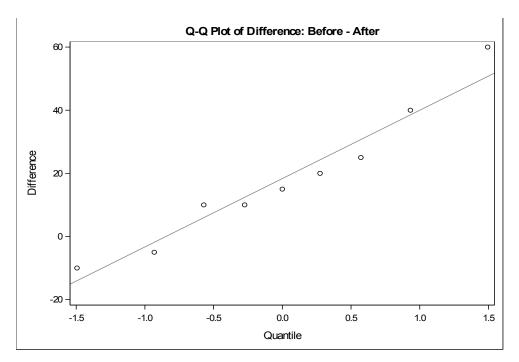
## **Assumptions**

A paired t-test—histograms, box-plots, and qq-plots for which are produced below (Figure 8)—was run to preliminarily determine normalcy and variance. With the exception of a single outlier (differential value = 60), the data appears **roughly normal**, with a **variance** of (sd = 21.6506). Moreover, judging from the sparse description of study's design, one can assume that observation **independence** is adequately satisfied. However, since the sample size is not especially large (n = 9), it is difficult to make assumptions regarding the normalcy of the distribution for the population being examined. For such reasons, a signed-rank test would be appropriate in this case, especially for the sake of mitigating the influence of the outlying observation.

Figure 8:

Ν	Mean	Std Dev	Std Err	Minimum	Maximum
9	18.3333	21.6506	7.2169	-10.0000	60.0000





**Analysis** 

To test for the impact of yoga on puzzle completion time, a **Signed-Rank test** was run to test the *null-hypothesis* (*H*<sub>0</sub>) that: there is **no difference** in mean completion time before and after yoga treatment:

$$\mu_B - \mu_A = 0$$

The *alternative-hypothesis* ( $H_A$ ) would be that: mean puzzle completion time is greater for patients before treatment:

$$\mu_B - \mu_A > 0$$

Hand-written analysis determined that, for this test  $SR_{stat} = 39.5$ .

The SR-sum expected under  $H_0$  would be: **Mean(S) = 22.5**, producing the following difference from a **test** for location shift: (39.5-22.5) = **17** 

The corresponding Z-scores (including continuity correction) and p-values were as follows:

#### $Z = 1.954751 \rightarrow p$ -value (normal approx.) = .0253

On this basis, one would **reject the null-hypothesis** ( $H_0$ ), determining that: there is moderate statistical evidence to suggest that **mean puzzle completion time is greater before treatment**.

Given the evidence for this sparsely described observational study—in which neither sampling nor allocation appear to be randomized, since participation in the study was voluntary and all subjects underwent the treatment—inferences regarding the difference in mean puzzle completion time can only be drawn to the particular population of autistic children examined by the study, though the

results are still informative within that limited scope of inference. However, as sampling was determined by voluntary participation, inferences regarding **causal relations cannot be drawn.** 

#### Part C)

Statistical software was used to identify the **test statistic** (**t**<sub>stat</sub>) and **p-value** for a **Paired T-test** of the data. Figures and **confidence intervals** for the following statistics are produced below (Figure 9):

$$t_{stat} = 2.54 \rightarrow p$$
-value = .0347

Figure 9:

Mean	95% CL Mean		Std Dev	95% CL Std Dev	
18.3333	1.6912	34.9755	21.6506	14.6241	41.4777

DF	t Value	<b>Pr</b> >  t
8	2.54	0.0347

#### Part D)

**Analysis** 

To test for the impact of yoga on puzzle completion time, a paired t-test was run to test the *null-hypothesis* (*H*<sub>0</sub>) that: there is **no difference** in mean completion time before and after yoga treatment:

$$\mu_B - \mu_A = 0$$

The *alternative-hypothesis*  $(H_A)$  would be that: mean puzzle completion time is **greater** for patients before treatment:

$$\mu_B - \mu_A > 0$$

Assuming the null-hypothesis, with a sample size of n = 9, with df = 8, and  $\alpha = .05$ , the critical value would be  $\mathbf{t}_{crit} = \mathbf{1.860}$ .

Statistical software determined that, for this test,  $\mathbf{t}_{\text{stat}} = \mathbf{2.54}$ .

Correspondingly, the p-value = .0347

On this basis, one would **reject the null-hypothesis** ( $H_0$ ), determining that: there is moderate statistical evidence to suggest that **mean puzzle completion time is greater before treatment**.

The associated **95% confidence interval** for the difference in mean completion time, providing **two-sides** for intuitive clarity, is as follows: [**1.6912**, **34.9755**].

As before, given the evidence for this sparsely described observational study—in which neither sampling nor allocation appear to be randomized, since participation in the study was voluntary and all subjects underwent the treatment—inferences regarding the difference in mean puzzle completion time **can only be drawn to the particular population of autistic children examined by the study**, though the results are still informative within that limited scope of inference. However, as sampling was determined by voluntary participation, inferences regarding **causal relations cannot be drawn.** 

Further tables for the results (Figure 10) are produced below:

Figure 10:

Moments						
N	9	Sum Weights	9			
Mean	18.3333333	Sum Observations	165			
<b>Std Deviation</b>	21.6506351	Variance	468.75			
Skewness	0.7310904	Kurtosis	0.5328254			
<b>Uncorrected SS</b>	6775	Corrected SS	3750			
Coeff Variation	118.094373	Std Error Mean	7.21687836			

	Basic Statistical Measures					
Location Variability						
Mean	18.33333	<b>Std Deviation</b>	21.65064			
Median	15.00000	Variance	468.75000			
Mode	10.00000	Range	70.00000			
		Interquartile Range	15.00000			

Tests for Location: Mu0=0						
Test	Statistic p Value					
Student's t	t	2.540341	<b>Pr</b> >  t	0.0347		
Sign	M	2.5	Pr >=  M	0.1797		
Signed Rank	S	18.5	Pr >=  S	0.0313		

#### Part E)

#### **Concluding Analysis**

On the basis of the resulting evidence above, although a comparison of the Signed-Rank test and Paired T-test for this data produced roughly similar **p-values** (Hand Written Signed Rank = .0253; SAS Signed Rank = .0313; SAS Paired T-test = .0347), closer examination of the data and study design suggested that Signed-Rank would be an appropriate test, on account of its resistance to outliers, its lack of assumptions regarding normalcy and variance for this rather small sample size, and its sensitivity to the magnitude of observed differences, all of which would produce more reliable test results in the event of a non-normal and highly variable population distribution, presumably fraught with outliers.

#### **Bonus Question**

### Part A)

A permutation distribution was built for the Rank-Sum statistic for the trauma data above, using 5000 permutations. Tables (Figure 11) comparing the results for the **mean**, **standard deviation**, and **p-values** with those identified in Question #3 are produced below.

Figure 11:

Rank Sum Permutation – Results

Data Scores for Variable Measure Classified by Variable Trauma								
Trauma	Trauma N Scores Under H0 Under H0 Scores							
N	8	167.70	196.853333	11.790497	20.962500			
T	7	201.40	172.246667	11.790497	28.771429			

Data Scores Two-Sample Test			
Statistic (S)	201.400		
Z	2.4726		
One-Sided Pr > Z	0.0067		
Two-Sided Pr >  Z	0.0134		

Monte Carlo Estimates for the Exact Test			
One-Sided Pr >= S			
Estimate	0.0002		
99% Lower Conf Limit	<.0001		
99% Upper Conf Limit	0.0007		
Two-Sided Pr >=  S - Mean			
Estimate	0.0002		
99% Lower Conf Limit	<.0001		
99% Upper Conf Limit	0.0007		
Number of Samples	5000		
Initial Seed	2345		

# Rank Sum Test (from Question #3 above) – Results

Wilcoxon Scores (Rank Sums) for Variable Measure Classified by Variable Trauma								
Trauma Sum of Expected Std Dev Mean Scores Under H0 Under H0 Score								
N	8	38.0	64.0	8.633269	4.750000			
Т	7	82.0	56.0	8.633269	11.71428 6			
	Average scores were used for ties.							

Wilcoxon Two-Sample Test	
Statistic	82.0000
Normal Approximation	
Z	2.9537
One-Sided Pr > Z	0.0016
Two-Sided Pr >  Z	0.0031
t Approximation	
One-Sided Pr > Z	0.0052
Two-Sided Pr >  Z	0.0105
Z includes a continuity correction of 0.5.	

## Conclusion

Comparative analysis revealed a higher, but still significant p-value for the **permuted version** of the Rank-Sum test, where the **(one-sided) p-val = .0067**.