

1. If all of the observations in treatment 1 is larger than treatment 2, it must have the largest difference between means.

$$\Rightarrow \text{number of } D's \geq D_{\text{obs}} = 1$$

$$m=4, n_1=4$$

$$\Rightarrow \binom{m+n}{m} = \frac{8!}{4!4!} = \frac{8 \cdot 7 \cdot 6 \cdot 5 \cdots 1}{4 \cdot 3 \cdot 2} = 7 \cdot 2 \cdot 5 = 70$$

$$\Rightarrow P_{\text{upper}} = \frac{1}{70} = 0,0143$$

2. See appendix for table

a) p-value =  $\frac{P's \geq D_{\text{obs}}}{\binom{m+n}{m}} = \frac{9}{20} = 0,45$  (Assuming upper)

b)  $T_1 = \frac{1}{m} + \frac{1}{n}$ ,  $T = 12 \Rightarrow T_1 = \frac{1}{3} + \frac{1}{3} = \frac{2}{3}$

$$T_1 = 75, T = 12 \Rightarrow 75 \left( \frac{1}{3} + \frac{1}{3} \right) - \frac{1}{3} = 9$$

$$\Rightarrow \frac{9}{20} = 0,45$$

3. See appendix for table

a) Given that  $m+n$  is a larger sample, we will use the normal approximation

$$H_0: \Theta_1 = \Theta_2 \quad H_a: \Theta_1 > \Theta_2, \quad \alpha = 0,05$$

$$\text{P-value} = P\left(Z \geq \frac{W - \frac{1}{2}m(N+1)}{\sqrt{\frac{1}{12}mn(N+1)}}\right)$$

Ranked: 0 0 0 0 1  
 2 2 2 2 2 2 2 2 2 3 3 4 5 6 8

Ranks: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25  
 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41

$$\text{Avg ranks: } 0 = 2,5 \quad 1 = 1,5 \quad 2 = 30,5 \quad 3 = 36,5$$

$$\Rightarrow W = 10 \cdot 1,5 + 9 \cdot 30,5 + 2 \cdot 36,5 \\ + 38 + 39 + 40 \\ = 614,5$$

$$\frac{1}{2}m(N+1) = 504$$

$$\sqrt{\frac{1}{12}mn(N+1)} = 37,7\dots$$

$$\Rightarrow \text{P-value} = P(Z \geq 2,924) = 0,0018 < \alpha$$

Thus we reject  $H_0$  and conclude that there is sufficient evidence that  $\Theta_1 > \Theta_2$

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$$b) H_0: \theta_1 \leq \theta_2 \quad H_a: \theta_1 > \theta_2$$

$$t = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{s_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}} = 1,554 \quad (\text{from R})$$

$$\Rightarrow p\text{-value} = 0,066 \quad (\text{from R})$$

Hence we fail to reject  $H_0$  and conclude that there is insufficient evidence that  $\theta_1$  is larger than  $\theta_2$ .

8 See appendix for code

As b the number of simulated permutations increases, the p-value gets closer to that of the t-test. This suggest that as  $n+m$  increases, the underlying distributions approaches normal distributions as by the CLT. That implies that the t-test is a good approximation for larger permutations.

Given that t-tests check for differences in means and wilcoxon tests for differences in location parameter, the two tests differ. Now, if the data was approximately normal distribution, there would be no difference. Thus, we can conclude that there is skewness to the data.

OBS. belongs to question 7

$$10 \quad H_0: G_1 = G_2 \quad H_a: G_1 < G_2 \quad \alpha = 0,05$$

(see app-  
endix for  
code)

From the Permutation test we saw:

$$V_1 = -3,02 \leftarrow$$

$$\#\text{V's} \leq V_1 = 6$$

$$\binom{m+n}{m} = 462$$

$$\Rightarrow p\text{-val} = \frac{6}{462} = 0,0130 \leftarrow$$

Hence, at the 8% significance level, we can conclude that there is sufficient evidence that  $G_1$  is less than  $G_2$

$$12 \quad k_a = 3+1 \quad \text{See appendix for code}$$

$$k_b = 27$$

$$\Rightarrow -12 < \Delta < -1 \quad \text{at } \alpha = 0,05$$

$$\text{Hedges-Lehman } \Delta = -6$$

(18)  $H_0: G_{\text{Exp}} = G_{\text{con}}$   $H_1: G_{\text{Exp}} < G_{\text{con}}$  See appendix

Wilcoxon:  $\alpha = 0,05$

$W = 41$ , p-value = 0,036

Van der Waerden

$V_1 = -2,355$ , p-value = 0,0485

Savage

Observed Savage Score = -2,355

$\Rightarrow$  p-value = 0,0485

Permutation

p-value = 0,107

In our case we would reject  $H_0$  with the Wilcoxon, van der Waerden, and Savage Scores but not the permutation test.

The difference arise because the first three explicitly test for a difference in location parameter through ordering of the data. The permutation test checks for difference in means.

The power will be low for the permutation test if the distributions are different but have the same mean. But, looking at the data, it seems plausible that the control group has a higher mean, thus should the first three be used.

# HW2\_STA104

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2022-10-18

## Appendix

### Code: Question 2 & 3.

```
# Get packages -----
library(dplyr)
library(tidyr)

# Create data frame of 6C3 -----
t1 <- c(10, 15, 50)
t2 <- c(12, 17, 19)
obs <- append(t1, t2)

df <- data.frame(combn(obs, 3))
df <- as.data.frame(t(df))

# Find mean difference and treatment sum -----
perm <- df %>%
  mutate(sumTreatment1=V1+V2+V3) %>%
  mutate(sumOther=123-sumTreatment1) %>%
  mutate(mean1=sumTreatment1/3, mean2=sumOther/3) %>%
  mutate(diff=mean1-mean2) %>%
  select(V1, V2, V3, diff, sumTreatment1)

name <- rownames(perm)

# Add the rest of the observations as second treatment group -----
df2 <- data.frame()

for (row in 1:nrow(perm)) {
  vec <- obs

  v1 <- perm[row, "V1"]
  v2 <- perm[row, "V2"]
  v3 <- perm[row, "V3"]

  df2 <- df2 %>% rbind(vec[! vec %in% c(v1, v2, v3)])
}

df2 <- df2 %>% select(V4=X12, V5=X17, V6=X19)
```

```

perm <- perm %>%
  cbind(df2)

# Create final table -----
perm <- perm %>%
  rowwise() %>%
  mutate(
    median1 = median(c(V1, V2, V3)),
    median2 = median(c(V4, V5, V6))
  ) %>%
  mutate(medianDiff = median1 - median2) %>%
  select(V1, V2, V3, meanDiff=diff, sumTreatment1, medianDiff)

perm <- data.frame(perm)
rownames(perm) <- name
rownames(perm)[rownames(perm) == "X1"] <- "X1*****" # Observed sample

perm <- perm %>% arrange(desc(meanDiff))

```

**Table: Question 2 & 3.**

	V1	V2	V3	meanDiff	sumTreatment1	medianDiff
X19	50	17	19	16.33333	86	7
X13	15	50	19	15.00000	84	7
X12	15	50	17	13.66667	82	5
X18	50	12	19	13.00000	81	4
X7	10	50	19	11.66667	79	4
X17	50	12	17	11.66667	79	2
X6	10	50	17	10.33333	77	2
X11	15	50	12	10.33333	77	-2
X1*****	10	15	50	9.00000	75	-2
X5	10	50	12	7.00000	72	-5
X16	15	17	19	-7.00000	51	5
X20	12	17	19	-9.00000	48	2
X10	10	17	19	-10.33333	46	2
X15	15	12	19	-10.33333	46	-2
X4	10	15	19	-11.66667	44	-2
X14	15	12	17	-11.66667	44	-4
X3	10	15	17	-13.00000	42	-4
X9	10	12	19	-13.66667	41	-5
X8	10	12	17	-15.00000	39	-7
X2	10	15	12	-16.33333	37	-7

**Code: Question 7 b).**

```

# Two sample t-test -----
rural <- c(3,2,1,1,2,1,3,2,2,2,2,5,1,4,1,1,1,1,6,2,2,2,1,1)

```

```

urban <- c(1,0,1,1,0,0,1,1,1,8,1,1,1,0,1,1,2)

t.test(rural, urban, alternative = c("greater"))

##
##  Welch Two Sample t-test
##
## data: rural and urban
## t = 1.554, df = 27.699, p-value = 0.06577
## alternative hypothesis: true difference in means is greater than 0
## 95 percent confidence interval:
## -0.07669184      Inf
## sample estimates:
## mean of x mean of y
## 2.041667 1.235294

```

### Code: Question 8.

```

#simulation with b permutations randomly selected.
outcome=c(rural, urban)
b=100000
treat=c(rep(1,length(rural)),rep(2,length(urban)))
diffobs=mean(rural)-mean(urban)
d=c()
p=c()
for(i in 1:b){
  permut=sample(outcome)
  d[i]=mean(permut[treat==1])-mean(permut[treat==2])
  p[i]=(d[i]>=diffobs)+0
}
pvalue=sum(p)/b

```

### Code: Question 10.

```

# Create data frame of VW scores -----
s1 <- c(5, 11, 16, 8, 12)
s2 <- c(17, 14, 15, 21, 19, 13)
data <- c(s1, s2)

df <- data.frame(data)

section <- c(rep(1, length(s1)), rep(2, length(s2)))

VW <- c(-1.383, -0.967, -0.674, -0.431, -0.210, 0, 0.210, 0.431, 0.674, 0.967, 1.383)

V <- sum(
  df %>%
  cbind(section) %>%

```

```

mutate(rank=rank(data)) %>%
arrange(rank) %>%
cbind(VW) %>%
filter(section==1) %>%
select(VW)
)

# Create dictionary of value:VW pairs -----
dictdata <- df %>%
  cbind(section) %>%
  mutate(rank=rank(data)) %>%
  arrange(rank) %>%
  cbind(VW) %>%
  select(data, VW)

dict <- list()
for(i in 1:nrow(dictdata)){
  dict[[as.character(dictdata[i,1])]] <- dictdata[i,2]
}

# Create data frame of 11C5 -----
df <- data.frame(combn(data, 5))

df <- as.data.frame(t(df))

df[["VW"]] <- 0 # Empty vector to fill

# Add VW score for each permutation group -----
for(row in 1:nrow(df)){
  for(col in 1:(ncol(df)-1)){
    df[row, "VW"] <- df[row, "VW"] + as.numeric(unlist(dict[as.character(df[row,col])]))
  }
}

rownames(df)[rownames(df)=="X1"] <- "X1*****" # Mark our observed value

# Observed less than or equal to V -----
D <- nrow(
  df %>%
  filter(VW <= V)
)

tot <- nrow(df)

pval <- D/tot

# Order df for table -----
ordDf <- df %>% arrange(VW)

```

**Table: Question 10.**

\*Only the head of data frame given amount of permutations

	V1	V2	V3	V4	V5	VW
X34	5	11	8	12	13	-3.665
X30	5	11	8	12	14	-3.455
X31	5	11	8	12	15	-3.245
X43	5	11	8	14	13	-3.234
X1*****	5	11	16	8	12	-3.024
X46	5	11	8	15	13	-3.024

**Code: Question 12.**

```
# Create data frame of pairwise differences -----
s1 <- c(5, 11, 16, 8, 12)
s2 <- c(17, 14, 15, 21, 19, 13)

diffs <- c()
for(row in 1:length(s2)){
  for(col in 1:length(s1)){
    diff <- s1[col] - s2[row]
    diffs <- append(diffs, diff)
  }
}

diffs <- sort(diffs)

# Table for visual aid -----
pairwise <- data.frame("5"=rep(0, length(s2)),
                       "11"=rep(0, length(s2)),
                       "16"=rep(0, length(s2)),
                       "8"=rep(0, length(s2)),
                       "12"=rep(0, length(s2)))

rownames(pairwise) <- as.character(s2)

for(row in 1:length(s2)){
  for(col in 1:length(s1)){
    pairwise[row, col] <- s1[col] - s2[row]
  }
}

# Find ka and kb -----
df <- data.frame(diffs)

rank <- 1:nrow(df)

limits <- df %>%
  cbind(rank) %>%
  filter(rank==(3+1) | rank==27) # From A4 table
```

```

rownames(limits) <- c("Lower", "Upper")

hl <- median(diffs)

```

### Tables: Question 12.

Pairwise Comparison

	X5	X11	X16	X8	X12
17	-12	-6	-1	-9	-5
14	-9	-3	2	-6	-2
15	-10	-4	1	-7	-3
21	-16	-10	-5	-13	-9
19	-14	-8	-3	-11	-7
13	-8	-2	3	-5	-1

Confidence Interval Limits

	diffs	rank
Lower	-12	4
Upper	-1	27

### Code: Question 18.

```

exp <- c(11,33,48,34,112,369,64,44)
con <- c(177,80,141,332)

# ----- Wilcoxon -----
p <- wilcox.test(exp, con, alternative = ("less"))$p.value
W <- wilcox.test(exp, con, alternative = ("less"))$statistic + 8*9/2

# ----- Van der Waerden -----
# Get V1 -----
data <- c(exp, con)

df <- data.frame(data)

group <- c(rep("Experiment", length(exp)), rep("Control", length(con)))

VW <- c(-1.426, -1.020, -0.736, -0.502, -0.293, -0.097, 0.097, 0.293, 0.502, 0.736, 1.020, 1.426)

V <- sum(

```

```

df %>%
  cbind(group) %>%
  mutate(rank=rank(data)) %>%
  arrange(rank) %>%
  cbind(VW) %>%
  filter(group=="Experiment") %>%
  select(VW)
)

# Create dictionary of value:VW pairs -----
dictdata <- df %>%
  cbind(group) %>%
  mutate(rank=rank(data)) %>%
  arrange(rank) %>%
  cbind(VW) %>%
  select(data, VW)

dict <- list()
for(i in 1:nrow(dictdata)){
  dict[[as.character(dictdata[i,1])]] <- dictdata[i,2]
}

# Savage scores dict -----
N <- nrow(df)

savage <- c(1/N)
for(i in 2:N){
  savage[i] <- savage[i-1] + 1/(N-(i-1))
}

savDictData <- df %>%
  cbind(group) %>%
  mutate(rank=rank(data)) %>%
  arrange(rank) %>%
  cbind(savage) %>%
  select(data, savage)

dictSavage <- list()
for(i in 1:nrow(savDictData)){
  dictSavage[[as.character(savDictData[i,1])]] <- savDictData[i,2]
}

# Create data frame of 12C8 -----
df <- data.frame(combn(data, 8))

df <- as.data.frame(t(df))

df[["VW"]] <- 0 # Empty vector to fill

# Add VW score for each permutation group -----
for(row in 1:nrow(df)){
  for(col in 1:(ncol(df)-1)){
    
```

```

        df[row, "VW"] <- df[row, "VW"] + as.numeric(unlist(dict[as.character(df[row,col]))))
    }
}

rownames(df)[rownames(df)=="X1"] <- "X1*****" # Mark our observed value

# Observed less than or equal to V -----
D <- nrow(
  df %>%
  filter(VW <= V)
)

tot <- nrow(df)

pval <- D/tot


# ----- Exponential -----
# Add savage score to existing df -----
df["sav"] <- 0

for(row in 1:nrow(df)){
  for(col in 1:(ncol(df)-2)){
    df[row, "sav"] <- df[row, "sav"] + as.numeric(unlist(dict[as.character(df[row,col]))))
  }
}

savObs <- as.numeric(df %>%
  filter(row.names(df) == "X1*****") %>%
  select(sav))

dSav <- nrow(
  df %>%
  filter(sav <= savObs )
)

total <- nrow(df)

p <- dSav/total


# ----- Permutation -----
Dperm <- nrow(df %>%
  select(1:8) %>%
  mutate(sumOfExperiment=V1+V2+V3+V4+V5+V6+V7+V8) %>%
  arrange(sumOfExperiment) %>%
  filter(sumOfExperiment <= 715))

pvaluePerm <- Dperm/total

```