

BACS HW – Week 7

Question 1) Let's **develop some intuition** about the data and results:

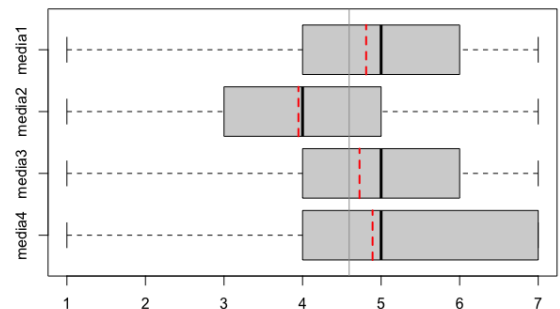
- a. What are the *means* of viewers' intentions to share (INTEND.0) on each of the four media types?

```
media_list <- list()
media_list$media1 <- (read.csv("pls-media1.csv",header = TRUE))$INTEND.0
media_list$media2 <- (read.csv("pls-media2.csv",header = TRUE))$INTEND.0
media_list$media3 <- (read.csv("pls-media3.csv",header = TRUE))$INTEND.0
media_list$media4 <- (read.csv("pls-media4.csv",header = TRUE))$INTEND.0
media_means <- sapply(media_list, mean)
cat("The mean of media1 is",media_means[1],"\nThe mean of media2 is",media_means[2],"\nThe mean of media3 is",media_means[3],"\nThe mean of media4 is",media_means[4])
```

The mean of media1 is 4.809524
 The mean of media2 is 3.947368
 The mean of media3 is 4.725
 The mean of media4 is 4.891304

- b. Visualize the *distribution and mean* of intention to share, across all four media.
 (Your choice of data visualization; Try to put them all on the same plot and make it look sensible)

```
boxplot(rev(media_list), horizontal=TRUE)
segments(x0=unlist(rev(media_means)),
y0=c(1,2,3,4)-0.4, x1=unlist(rev(media_means)),
y1=c(1,2,3,4)+0.4, lty="dashed", col="red", lwd=2)
abline(v=mean(media_means), col=8)
```



- c. From the visualization alone, do you feel that media type makes a difference on intention to share?

I observe that media2 is the most different from the other three media types. Therefore, I think media type could make a difference on intention to share.

Question 2) Let's try **traditional one-way ANOVA**:

- a. State the null and alternative hypotheses when comparing INTEND.0 across four groups in ANOVA

H_{null} : The mean of intentions in four groups would be the same.

$H_{alternative}$: The mean of intentions in four groups would not be all the same.

b. Let's compute the F-statistic ourselves:

i. Show the code and results of computing MSTR, MSE, and F

```
# SSTR: I use sapply to compute the length.
SSTR <- sum((sapply(media_list,length)*(sapply(media_list,mean)-mean(media_means))^2))
df_mstr <- length(media_list)-1
MSTR <- SSTR/df_mstr
SSE <- sum((sapply(media_list,length)-1)*sapply(media_list, var))
df_mse <- sum(sapply(media_list,length)) - length(media_list)
MSE <- SSE/df_mse
F <- MSTR/MSE
cat("The value (MSTR):",MSTR,"The value (MSE):", MSE,"The value (F):", F)
The value (MSTR): 7.53239
The value (MSE): 2.869151
The value (F): 2.625303
```

ii. Compute the p-value of F, from the null F-distribution; is the F-value significant?
If so, state your conclusion for the hypotheses.

```
F_cut <- qf(p=0.95, df1=df_mstr, df2=df_mse)
p_value <- pf(F, df_mstr, df_mse, lower.tail=FALSE)
cat("The value (F):", F,"The value (F-cut):", F_cut,"The p-value is",p_value)
The value (F): 2.625303
The value (F-cut): 2.660406
The p-value is 0.05230686
```

Because the p-value(=0.05230686) is larger than 0.05, I couldn't reject the null hypothesis.

c. Conduct the same one-way ANOVA using the aov() function in R – confirm that you got similar results.

```
# Create a new dataframe to combine media1,2,3,4.
m1 <- data.frame(media_num=rep("m1",42), intention=media_list$media1)
m2 <- data.frame(media_num=rep("m2",38), intention=media_list$media2)
m3 <- data.frame(media_num=rep("m3",40), intention=media_list$media3)
m4 <- data.frame(media_num=rep("m4",46), intention=media_list$media4)
medias_intention <- rbind(m1, m2, m3, m4)
summary(aov(medias_intention$intention ~factor(medias_intention$media_num)))
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
factor(medias_intention\$media_num)	3	22.5	7.508	2.617	0.0529
Residuals	162	464.8	2.869		

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Yes, I got similar results.

- d. Regardless of your conclusions, conduct a post-hoc Tukey test (feel free to use the TukeyHSD() function in R) to see if any *pairs of media have significantly different means* – what do you find?

```
anova_model <- aov(medias_intention$intention ~ factor(medias_intention$media_num))
TukeyHSD(anova_model, conf.level = 0.05)
```

	diff	lwr	upr	p adj
m2-m1	-0.86215539	-1.06562977	-0.6586810	0.1085727
m3-m1	-0.08452381	-0.28530983	0.1162622	0.9959223
m4-m1	0.08178054	-0.11218249	0.2757436	0.9959032
m3-m2	0.77763158	0.57175512	0.9835080	0.1825044
m4-m2	0.94393593	0.74470805	1.1431638	0.0573229
m4-m3	0.16630435	-0.03017708	0.3627858	0.9687417

From the result, I don't think there are any pairs of media that have significantly different means.

- e. Do you feel the classic requirements of one-way ANOVA were met?
(Feel free to use any combination of methods we saw in class or any analysis we haven't covered)

1. Each treatment/population's **response variable is normally distributed**
2. The **variance (s^2) of the response variables is the same** for all treatments/populations
3. The **observations are independent**: the response variables are not related between groups



1. Check normally distributed

I use the Shapiro-Wilk test to check if it is normally distributed.

In the Shapiro-Wilk test, when the p-value > 0.05, it is normally distributed.

But I found that the data in these media types are **not normally distributed** because the p-value is 6.886e-08 which is smaller than 0.05.

My code:

```
shapiro.test(medias_intention$intention
data: medias_intention$intention
W = 0.92051, p-value = 6.886e-08
```

2. Check variance is the same for multi groups

I try to use Levene's test which is an alternative to Bartlett's test when the data is not normally distributed. In Levene's test for my data, p-value is larger than 0.05 which means that the variance (s^2) of the response variables is the same.

The reference link: <http://www.sthda.com/english/wiki/compare-multiple-sample-variances-in-r>

My code:

```
install.packages("car")
library("car")
leveneTest(medias_intention$intention ~ medias_intention$media_num, data = medias_intention)
```

Levene's Test for Homogeneity of Variance (center = median)

	Df	F value	Pr(>F)
group	3	1.5403	0.2061
	162		

Warning message:

In leveneTest.default(y = y, group = group, ...) : group coerced to factor.

3. From the article, it mentions that our researcher runs an experiment where each of these four alternative media is shown to a different panel of randomly assigned people. Therefore, I think the observations are independent.

From the previous check, only the first assumption is not met.

Question 3) Let's use the **non-parametric Kruskal Wallis** test:

- a. State the null and alternative hypotheses (in terms of distribution or difference of mean ranks)

H_{null} : The distribution of mean ranks are the same.

$H_{alternative}$: The distribution of mean ranks are not all the same.

- b. Let's compute (an approximate) Kruskal Wallis H ourselves:

- i. Show the code and results of computing H

```
# Rank all the combined values across groups
rank_medias <- rank(medias_intention$intention)
# combine rank into medias_intention
medias_rank <- cbind(medias_intention, rank_medias)
# split the same rank into same group
group_rank <- split(medias_rank$rank_medias, medias_rank$media_num)
group_ranksum <- sapply(group_rank, sum)
group_rank_length <- sapply(group_rank, length)
N <- sum(group_rank_length)
H <- (12 / (N * (N + 1))) * sum((group_ranksum^2) / group_rank_length) - 3 * (N + 1)
paste("H value is", H)
[1] "H value is 8.45465979544389"
```

- ii. Compute the p-value of H, from the null chi-square distribution; is the H value significant?
If so, state your conclusion of the hypotheses.

```
kw_p <- 1 - pchisq(H, df = 4 - 1)
paste("The p-value of H is", kw_p)
[1] "The p-value of H is 0.037492918119218"
```

The p-value of H is smaller than 0.05, which means that the H value isn't significant.
I could reject the null hypothesis.

- c. Conduct the same test using the `kruskal.wallis()` function in R – confirm that you got similar results.

```
kruskal.test(medias_rank$rank_medias ~ medias_rank$media_num, data = medias_rank)
```

Kruskal-Wallis rank sum test

data: medias_rank\$rank_medias by medias_rank\$media_num

Kruskal-Wallis chi-squared = 8.8283, df = 3, p-value = 0.03166

- d. Regardless of your conclusions, conduct a post-hoc Dunn test (feel free to use the `dunnTest()` function from the FSA package) to see if any *pairs of media are significantly different* – what do you find?

Code:

```
dunnTest(medias_rank$rank_medias ~ medias_rank$media_num, data = medias_rank, method = "bonferroni")
```

Thought:

I found that the p-value of the comparison (m2-m4) is 0.04542535 (<0.05) which means the pair (m2&m4) is significantly different..

**Dunn (1964) Kruskal-Wallis multiple comparison
p-values adjusted with the Bonferroni method.**

	Comparison	Z	P.unadj	P.adj
1	m1 - m2	2.30087819	0.021398517	0.12839110
2	m1 - m3	-0.09233644	0.926430736	1.00000000
3	m2 - m3	-2.36408588	0.018074622	0.10844773
4	m1 - m4	-0.31452459	0.753122646	1.00000000
5	m2 - m4	-2.65613380	0.007904225	0.04742535
6	m3 - m4	-0.21613379	0.828883460	1.00000000