BMEG 802 – Advanced Biomedical Experimental Design and Analysis

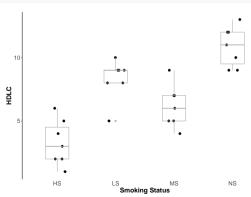
Assignment 4

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Question 1

```
install.packages("ggplot2")
hdlc <- c(12, 10, 11, 13, 9, 9, 12, 9, 8, 5, 9, 9, 10, 8, 5, 4, 7, 9, 5, 7, 6, 3,
 smoker <- c("NS", "NS", "NS", "NS", "NS", "NS", "LS", 
library(ggplot2)
  # create data frame
 data1 <- data.frame(</pre>
               SMOKER=smoker.
               HDLC=hdlc,
               sd=c(1) # dummy input
```

Graphing with GG Plot - Box and Whisker



```
HDLC <- c(12, 10, 11, 13, 9, 9, 12, 9, 8, 5, 9, 9, 10, 8, 5, 4, 7, 9, 5, 7
m1 <- lm(HDLC ~ factor(Smoker))
anova(m1)
## Analysis of Variance Table
##
## Response: HDLC
##
              Df Sum Sq Mean Sq F value Pr(>F)
## factor(Smoker) 3 216.857 72.286 26.06 1.005e-07 ***
## Residuals 24 66.571 2.774
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
Group1 = hdlc[1:7]
Group2 = hdlc[8:14]
Group3 = hdlc[15:21]
Group4 = hdlc[22:28]
shapiro.test(Group1)$p.value
## [1] 0.4196491
shapiro.test(Group2)$p.value
## [1] 0.06607612
shapiro.test(Group3)$p.value
## [1] 0.7385166
shapiro.test(Group4)$p.value
## [1] 0.7724094
```

Since p > 0.05 for all, no violations of normality

```
bartlett.test(HDLC ~ factor(Smoker))

##

## Bartlett test of homogeneity of variances

##

## data: HDLC by factor(Smoker)

## Bartlett's K-squared = 0.12176, df = 3, p-value = 0.9891

Since p > 0.05, no violation of homogeneity of variances
```

```
install.packages("effectsize")
library(effectsize)
omega squared(m1)
## For one-way between subjects designs, partial omega squared is equvilan
## Returning omega squared.
## Parameter | Omega2 | 90% CI
## factor(Smoker) | 0.73 | [0.54, 0.82]
```

```
options(scipen = 999) # shows all values in decimal form
pval_1v2 = t.test(Group1, Group2, alternative = "two.sided")$p.value # G1 vs G2
pval 1v3 = t.test(Group1, Group3, alternative = "two.sided") p.value # G1 vs G3
pval 1v4 = t.test(Group1, Group4, alternative = "two.sided") p.value # G1 vs G4
pval 2v3 = t.test(Group2, Group3, alternative = "two.sided") p.value # G2 vs G3
pval_2v4 = t.test(Group2, Group4, alternative = "two.sided") p.value # G2 vs G4
pval 3v4 = t.test(Group3, Group4, alternative = "two.sided") p.value # G3 vs G4
pvals = c(pval 1v2, pval 1v3, pval 1v4, pval 2v3, pval 2v4, pval 3v4)
pvals
## [1] 0.010506033434 0.000155856817 0.000002627869 0.030962641135 0.000145470449
## [6] 0.009697653922
p.adjust(pvals, method = "holm", n = length(pvals))
## [1] 0.02909296177 0.00072735225 0.00001576721 0.03096264113 0.00072735225
## [6] 0.02909296177
```

```
install.packages("effsize")
library(effsize)
options(scipen = 999) # shows all values in decimal form
d 1v2 = cohen.d(Group1, Group2, var.equal = False) $estimate # G1 vs G2
d 1v3 = cohen.d(Group1, Group3, var.equal = False) $estimate # G1 vs G3
d 1v4 = cohen.d(Group1, Group4, var.equal = False)$estimate # G1 vs G4
d 2v3 = cohen.d(Group2, Group3, var.equal = False) $estimate # G2 vs G3
d_2v4 = cohen.d(Group2, Group4, var.equal = False)$estimate # G2 vs G4
d 3v4 = cohen.d(Group3, Group4, var.equal = False)$estimate # G3 vs G4
d = c(d 1v2, d 1v3, d 1v4, d 2v3, d 2v4, d 3v4)
d effsize
```

[1] 1.618625 2.899874 4.479318 1.306407 2.933693 1.643065

There was a significant main effect of smoking status on HDLC [F(3,24) = 26.06, p < 0.001, ω^2 = 0.73]. All groups were significantly different from one another (p < 0.05 and d > 1.31 for all mean comparisons). It is worth noting that the light smokers had significantly higher HDLC compared to moderate smokers.

```
kruskal.test(HDLC ~ factor(Smoker))

##

## Kruskal-Wallis rank sum test

##

## data: HDLC by factor(Smoker)

## Kruskal-Wallis chi-squared = 20.744, df = 3, p-value = 0.000119
```

Question 2

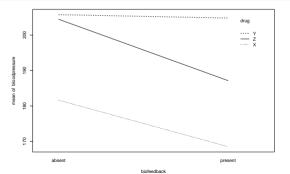
```
##
        Balanced one-way analysis of variance power calculation
##
##
            groups = 4
##
                 n = 25.22645
##
       between.var = 15
##
        within.var = 100
##
         sig.level = 0.05
##
             power = 0.8
##
## NOTE: n is number in each group
```

We should collect 26 participants per group to be sufficiently powered.

Question 3

```
dx_bf <- rnorm(20, mean = 170, sd=12.91)
dy_bf <- rnorm(20, mean = 203, sd=13.91)
dz_bf <- rnorm(20, mean = 188, sd=13.84)
dx_nbf <- rnorm(20, mean = 186, sd=11.84)
dy_nbf <- rnorm(20, mean = 201, sd=11.93)
dz_nbf <- rnorm(20, mean = 210, sd=12.81)</pre>
```

```
bloodpressure <- c(dx_bf, dy_bf, dz_bf, dx_nbf, dy_nbf, dz_nbf)
biofeedback <- factor(c(rep("present",60),rep("absent",60)))
drug <- factor(rep(c(rep("X",20),rep("Y",20),rep("Z",20)),2))
bpdata <- data.frame(bloodpressure, biofeedback, drug)
interaction.plot(biofeedback, drug, bloodpressure)</pre>
```



```
myanova <- aov(bloodpressure ~ biofeedback*drug)
summary(myanova)</pre>
```

```
##
                  Df Sum Sq Mean Sq F value
                                                      Pr(>F)
## biofeedback
                       3282
                             3282 21.830
                                                  0.00000821 ***
                      19087 9544 63.476 < 0.0000000000000000 ***
## drug
                   2 1441 720 4.792
## biofeedback:drug
                                                       0.01 *
## Residuals
            114 17140
                           150
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
shapiro.test(bloodpressure[1:20])$p.value # group 1
## [1] 0.2561596
shapiro.test(bloodpressure[21:40])$p.value # group 2
## [1] 0.0679432
shapiro.test(bloodpressure[41:60])$p.value # group 3
## [1] 0.2260702
shapiro.test(bloodpressure[61:80])$p.value # group 4
## [1] 0.9518
shapiro.test(bloodpressure[81:100])$p.value # group 5
## [1] 0.5419854
shapiro.test(bloodpressure[101:120])$p.value # group 6
## [1] 0.8039848
No violations of normality
```

```
Test for sphericity (homogeneous of variance) in each main effect and interaction
bartlett.test(bloodpressure ~ interaction(drug, biofeedback))$p.value
## [1] 0.2097777
bartlett.test(bloodpressure ~ drug)$p.value
## [1] 0.8007082
bartlett.test(bloodpressure ~ biofeedback)$p.value
## [1] 0.05050119
```

```
install.packages("effectsize")
```

```
library(effectsize)
omega_squared(myanova)
```

```
## Parameter | Omega2 (partial) | 90% CI

## ------

## biofeedback | 0.15 | [0.06, 0.25]

## drug | 0.51 | [0.40, 0.59]

## biofeedback:drug | 0.06 | [0.00, 0.13]
```

[5] 0.000000391381721 0.686476275943622

```
# biofeedback
pval_dxbf_v_dybf = t.test(dx_bf, dy_bf, alternative = "two.sided")$p.value
pval_dxbf_v_dzbf = t.test(dx_bf, dz_bf, alternative = "two.sided")$p.value
pval_dxbf_v_dzbf = t.test(dy_bf, dz_bf, alternative = "two.sided")$p.value
# no biofeedback

pval_dxnbf_v_dynbf = t.test(dx_nbf, dy_nbf, alternative = "two.sided")$p.value
pval_dxnbf_v_dznbf = t.test(dx_nbf, dz_nbf, alternative = "two.sided")$p.value
pval_dxnbf_v_dznbf = t.test(dx_nbf, dz_nbf, alternative = "two.sided")$p.value
pval_dynbf_v_dznbf = t.test(dy_nbf, dz_nbf, alternative = "two.sided")$p.value
pvals = c(pval_dxbf_v_dybf, pval_dxbf_v_dzbf, pval_dybf_v_dzbf, pval_dxnbf_v_dynbf, pval_dxnbf_v_dznbf, pval_dynbf_v_dznbf)
pvals

## [1] 0.0000000005284255 0.0000309517829200 0.0005426306515590 0.0000001150953406
## [5] 0.0000000782763442 0.6864762759436223
p.adjust(pvals, method = "holm", n = length(pvals))
```

[1] 0.00000003170553 0.000092855348760 0.001085261303118 0.000000460381362

install.packages("effsize")

```
library(effsize)
options(scipen = 999) # shows all values in decimal form
d1v2 = cohen.d(dx bf, dy bf, var.equal = False) $estimate #
d1v3 = cohen.d(dx bf, dz bf, var.equal = False) sestimate #
d2v3 = cohen.d(dy_bf, dz_bf, var.equal = False) $estimate #
d4v5 = cohen.d(dx nbf, dy nbf, var.equal = False) $estimate #
d4v6 = cohen.d(dx nbf, dz nbf, var.equal = False) sestimate #
d5v6 = cohen.d(dy nbf, dz nbf, var.equal = False) sestimate #
d = c(d1v2, d1v3, d2v3, d4v5, d4v6, d5v6)
abs(d effsize1)
```

[1] 2.7230155 1.5089173 1.1961034 2.0602965 2.1383831 0.1286744

There was a significant interaction between drug and biofeedback on blood pressure $(p < 0.001, \omega_p^2 = 0.11)$. All mean comparisons showed that Drugs X, Y, and Z were significantly different from each other both in the presence of biofeedback (p < 0.05, d > 0.9 for all comparisons) and the absence of biofeedback (p < 0.05, d > 0.72 for all comparisons)

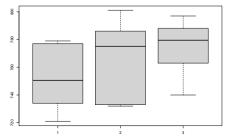
install.packages("pwr")

```
library(pwr)
# u = levels, v = dof residuals (df residuals = <math>a * b * (n-1))
pwr.f2.test(u=2, v=NULL, f2=0.1^2, sig.level=0.05, power=0.90)
##
        Multiple regression power calculation
##
##
##
                11 = 2
##
                v = 1265.395
                f2 = 0.01
##
         sig.level = 0.05
##
             power = 0.9
1265.395 / (3 * 2) + 1
```

[1] 211.8992

We would need to collect 212 participants.

Question 4.



```
m1 <- aov(dv ~ treatment + Error(subject/treatment), data=mydata)</pre>
summary(m1)
##
## Error: subject
            Df Sum Sq Mean Sq F value Pr(>F)
##
## Residuals 5 8548 1710
##
## Error: subject:treatment
##
            Df Sum Sq Mean Sq F value Pr(>F)
## treatment 2 1575.0 787.5 14.43 0.00113 **
## Residuals 10 545.7 54.6
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
```

```
install.packages("car")
response <- with(mydata,cbind(dv[treatment==1], dv[treatment==2], dv[treatment==3]
mlm1 <- lm(response ~ 1)
rfactor <- factor(c("r1", "r2", "r3"))
design = data.frame(rfactor)
library(car)

## Loading required package: carData
mlm1.aov <- Anova(mlm1, idata=design, idesign = ~rfactor, type="III")</pre>
```

summary(mlm1.aov, multivariate=FALSE)

```
## Warning in summary. Anova. mlm(mlm1.aov. multivariate = FALSE): HF eps > 1 treated
## as 1
##
## Univariate Type III Repeated-Measures ANOVA Assuming Sphericity
##
                Sum Sq num Df Error SS den Df F value
                                                               Pr(>F)
## (Intercept) 10520285
                                8547.8
                                           5 6153.773 0.000000006378 ***
## rfactor
                  1575
                                 545.7 10 14.432
                                                            0.001128 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
##
## Mauchly Tests for Sphericity
##
          Test statistic p-value
## rfactor
                 0 75802 0 57459
## Greenhouse-Geisser and Huvnh-Feldt Corrections
   for Departure from Sphericity
           GG eps Pr(>F[GG])
## rfactor 0.80516 0.00286 **
## ---
## Signif codes: 0 | *** 0 001 | ** 0 01 | * 1 0 05 | 1 0 1 | 1 1
```

```
pval = 0.00286
GGeps = 0.80516
df1_adj = 2 * GGeps # adjusted df
df2_adj = 10 * GGeps
Fscore = qf(1 - pval, df1=df1_adj, df2=df2_adj) # chi-square function
df1_adj
## [1] 1.61032
df2_adj
## [1] 8.0516
Fscore
## [1] 14.43117
```

Pov 1 1230 75/5 6153 773 1

summary(mlm1.aov, multivariate=TRUE) ## Warning in summary. Anova.mlm(mlm1.aov, multivariate = TRUE): HF eps > 1 treated ## as 1 ## ## Type III Repeated Measures MANOVA Tests: ## Term: (Intercept) ## Response transformation matrix: (Intercept) ## [1.] ## [2.] ## [3.] ## Sum of squares and products for the hypothesis: (Intercept) ## (Intercept) 31560854 ## ## Multivariate Tests: (Intercept) Df test stat approx F num Df den Df Pr(>F) ## Pillai 0.9992 6153.773 5 0.000000063781 *** ## Wilke 0.0008 6153.773 1 5 0.0000000063781 *** ## Hotelling-Lawlev 1 1230.7545 6153.773 1 5 0.000000063781 ***

5 0 0000000063781 ***

```
install.packages("rstatix")
library(rstatix)
## Attaching package: 'rstatix'
## The following objects are masked from 'package:effectsize':
##
      cohens_d, eta_squared
## The following object is masked from 'package:stats':
      filter
friedman test(dv ~ treatment | subject, data=mvdata)
## # A tibble: 1 v 6
    .y. n statistic df
                                   p method
## * <chr> <int> <dbl> <dbl> <dbl> <chr>
## 1 dv 6 7 2.0.0302 Friedman test
```

Sphericity not violated (Mauchly Test: p=0.575). Main Effect of Time is significant without (p=0.001) and with the Greenhouse-Geisser Corrections (p=0.003), with the Multivariate Approach (p=0.007), and Friedman Test (p=0.030).

Let's quickly split our data into each treatment

```
treat1 = response[1:6]
treat2 = response[7:12]
treat3 = response[13:18]
```

Performed on the PAIRED differences

```
shapiro.test(treat2-treat1)$p.value #
## [1] 0.9786552
shapiro.test(treat3-treat1)$p.value #
## [1] 0.1051113
shapiro.test(treat3-treat2)$p.value #
## [1] 0.2199981
Normality is not Violated, since all p > 0.05
```

```
library(effectsize)
omega_squared(mlm1.aov)

## Parameter | Omega2 (partial) | 90% CI
## ------
## rfactor | 0.86 | [0.29, 0.94]
```

```
p1v2 = t.test(treat1, treat2, paired = TRUE, alternative = "two.sided") $p. p1v3 = t.test(treat1, treat3, paired = TRUE, alternative = "two.sided") $p. p2v3 = t.test(treat2, treat3, paired = TRUE, alternative = "two.sided") $p. pvals = c(p1v2, p1v3, p2v3)

pvals_holm = p.adjust(pvals, method = "holm", n = length(pvals))

sprintf("%.5f", pvals_holm) #outputs in decimal (not scientific notation)

## [1] "0.04023" "0.00239" "0.19328"
```

```
library(effsize)
cohen.d(treat1, treat2, paired = TRUE)$estimate #
## [1] -0.4824522
cohen.d(treat1, treat3, paired = TRUE)$estimate #
## [1] -0.9637995
cohen.d(treat2, treat3, paired = TRUE)$estimate #
## [1] -0.2168753
```

There was a significant main of effect of Time (F(1.6,8.1) = 14.4, p = 0.003, ω^2 = 0.86). Time 2 (p = 0.040, d = 0.48) and Time 3 (p = 0.002, d = 0.96) were significantly greater than Time 1. Thus, this medication actually led to more coughs and can be considered ineffective.

Note: interpretation from any omnibus test is acceptable.

```
library(WebPower)
## Loading required package: MASS
##
## Attaching package: 'MASS'
## The following object is masked from 'package:rstatix':
##
##
       select
## Loading required package: lme4
## Loading required package: Matrix
## Registered S3 methods overwritten by 'lme4':
     method
                                     from
     cooks.distance.influence.merMod car
     influence.merMod
                                     car
     dfheta influence merMod
                                     car
     dfbetas.influence.merMod
                                     car
## Loading required package: lavaan
## This is lavaan 0.6-7
## lavaan is BETA software! Please report any bugs.
## Loading required package: parallel
```

```
## Repeated-measures ANOVA analysis
##
## n f ng nm nscor alpha power
## 72.05928 0.4 1 3 0.8 0.05 0.8
##
## NOTE: Power analysis for within-effect test
## URL: http://psychstat.org/rmanova
```

73 Participants would be needed.