PSYC 7720 Lab

Lab 10 Activity

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Directions:

- A. Download the angle_noise_data.csv. This is a two-factor within subjects design with two levels of the noise factor and three levels of the angle factor. In the data, abs represents the condition where noise is absent and pres represents where noise is present. The 0, 4, and 8 denote the three different angle conditions. The measurements in each cell are aggregated reaction times.
- B. Answer the following questions and save the code you used in an R script.
- C. You have until the end of lab to complete.

```
library(tidyverse)
library(ez)
library(emmeans)
options(contrasts = c("contr.sum","contr.poly"))
theme_set(theme_classic())
```

Questions:

1. Convert the angle_noise_data from wide format to long format. Name the first column *id*, the second column *noise*, the third column *angle*, the fourth column *rt*, and the dataframe *long_data*. Print the head and the dimensions of the long data.

```
long_data <- read.csv("angle_noise_data.csv") %>%
  rownames to column(var = "id") %>%
  pivot_longer(-id,
               names_to = "ws", values_to = "rt") %>%
  separate(ws, into = c("noise", "angle"), sep = -1,
           remove = TRUE) %>%
  mutate_if(is_character, factor) %>%
  mutate(angle = factor(angle, ordered = TRUE))
head(long_data)
## # A tibble: 6 x 4
##
     id
           noise angle
##
     <fct> <fct> <ord> <int>
## 1 1
           abs
                 0
                         420
## 2 1
                 4
                         420
           abs
## 3 1
           abs
                 8
                         480
## 4 1
           pres 0
                         480
## 5 1
                         600
           pres 4
                         780
## 6 1
           pres
dim(long_data)
```

```
## [1] 60 4
```

2. Ensure that the within-subjects conditions are appropriately implemented as qualitative and/or quantitative factors. Print the appropriate structure of *long_data* using str().

```
str(long_data)
```

```
## Classes 'tbl_df', 'tbl' and 'data.frame': 60 obs. of 4 variables:
## $ id : Factor w/ 10 levels "1","10","2","3",..: 1 1 1 1 1 1 3 3 3 3 ...
## $ noise: Factor w/ 2 levels "abs","pres": 1 1 1 2 2 2 1 1 1 2 ...
## $ angle: Ord.factor w/ 3 levels "0"<"4"<"8": 1 2 3 1 2 3 1 2 3 1 ...
## $ rt : int 420 420 480 480 600 780 420 480 480 360 ...</pre>
```

3. Run and interpret the results of the two-way factorial RM ANOVA using aov.

```
aov_mod <- aov(rt ~ noise * angle + Error(id/(noise * angle)), data = long_data)
summary(aov_mod)</pre>
```

```
##
## Error: id
##
            Df Sum Sq Mean Sq F value Pr(>F)
## Residuals 9 292140
                        32460
##
## Error: id:noise
            Df Sum Sq Mean Sq F value
##
                                       Pr(>F)
## noise
             1 285660 285660
                               33.77 0.000256 ***
## Residuals 9 76140
                         8460
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Error: id:angle
            Df Sum Sq Mean Sq F value
                                       Pr(>F)
             2 289920 144960
                               40.72 2.09e-07 ***
## Residuals 18 64080
                         3560
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Error: id:noise:angle
              Df Sum Sq Mean Sq F value
## noise:angle 2 105120
                          52560
                                 45.31 9.42e-08 ***
              18 20880
## Residuals
                           1160
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

4. Run and interpret the results of the two-way factorial RM ANOVA using ezANOVA.

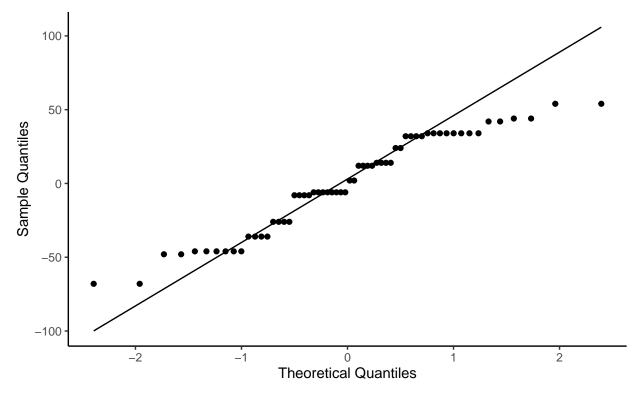
```
## $ANOVA

## Effect DFn DFd SSn SSd F p p<.05 ges

## 1 (Intercept) 1 9 19425660 292140 598.44917 1.526600e-09 * 0.9771999
```

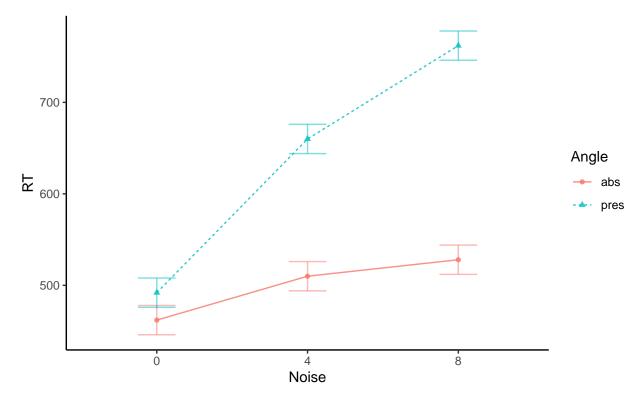
```
angle 2 18 289920 64080 40.71910 2.086763e-07 * 0.3901179
         noise 1 9 285660 76140 33.76596 2.559737e-04 * 0.3866017
## 4 angle:noise 2 18 105120 20880 45.31034 9.424093e-08 * 0.1882656
## $`Mauchly's Test for Sphericity`
## Effect W p p<.05
## 2 angle 0.9601060 0.8497219
## 4 angle:noise 0.8937772 0.6381418
## $`Sphericity Corrections`
## Effect GGe p[GG] p[GG] <.05 HFe p[HF] p[HF] <.05
        angle 0.9616365 3.401705e-07 * 1.217564 2.086763e-07
## 4 angle:noise 0.9039771 3.453931e-07
                                       * 1.117870 9.424093e-08
##
## $aov
##
## Call:
## aov(formula = formula(aov_formula), data = data)
## Grand Mean: 569
##
## Stratum 1: id
##
## Terms:
##
                 Residuals
## Sum of Squares
                   292140
## Deg. of Freedom
## Residual standard error: 180.1666
## Stratum 2: id:angle
##
## Terms:
                angle Residuals
## Sum of Squares 289920 64080
## Deg. of Freedom 2
## Residual standard error: 59.66574
## Estimated effects are balanced
##
## Stratum 3: id:noise
##
## Terms:
                 noise Residuals
## Sum of Squares 285660 76140
## Deg. of Freedom 1
## Residual standard error: 91.97826
## Estimated effects are balanced
## Stratum 4: id:angle:noise
##
## Terms:
##
                angle:noise Residuals
```

```
## Sum of Squares
                         105120
                                    20880
## Deg. of Freedom
                                       18
## Residual standard error: 34.05877
## Estimated effects are balanced
# Check the residuals for normality
resid <- proj(ez_mod$aov)[[3]][, "Residuals"]</pre>
resid %>%
  tibble() %>%
  ggplot(aes(sample = `.`)) +
  geom_qq() +
  stat_qq_line() +
  labs(x = "Theoretical Quantiles",
       y = "Sample Quantiles")
```



angle = 0:

```
contrast
                estimate
                           SE
                                df t.ratio p.value
##
    pres - abs
                      30 26.8 14.1 1.119
                                            0.2818
##
## angle = 4:
##
    contrast
                estimate
                           SE
                                df t.ratio p.value
    pres - abs
                     150 26.8 14.1 5.595
                                            0.0001
##
##
## angle = 8:
##
    contrast
                estimate
                           SE
                                 df t.ratio p.value
                     234 26.8 14.1 8.729
                                            <.0001
    pres - abs
ez_plot <- ezPlot(</pre>
  data = long_data,
  dv = rt,
  wid = id,
  within = .(angle, noise),
  x = angle,
  split = noise,
  x_lab = "Noise",
  split_lab = "Angle",
  y_lab = "RT"
ez_plot
```



• There was a significant two-way interaction between *noise* and *angle* on *rt*. Following up this interaction, it seems that the contrasts between the two noise conditions is not significant when *angle* is 0, but is significant when *angle* is 4 and 8. Another way to interpet this interaction is that there is a stronger positive linear trend when *noise* is present relative to when it is absent.