

HW9

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Q1

Answer 4 question for each scenario

- i. Would this scenario create systematic or random error (or both or neither)?
- ii. Which part of the t-statistic or significance (diff, sd, n, alpha) would be affected?
- iii. Will it increase or decrease our power to reject the null hypothesis?
- iv. Which kind of error (Type I or Type II) becomes more likely because of this scenario?

a. only collected data from a pool of young consumers, and missed many older customers who you suspect might use the product much less every day.

- i. Systematic error
- ii. diff
- iii. Yes, it will increase the power to reject the null hypothesis.
- iv. Type II error

b. 20 of the respondents should be removed from the data.

- i. Random error
- ii. n
- iii. No
- iv. Type II error.

c. Relaxing the C.I. from 95% to 90%

- i. A systematic error
- ii. Alpha
- iii. Yes , it will increase the power to reject the null hypothesis.
- iv. Type I error.

d. over-reports usage of older users.

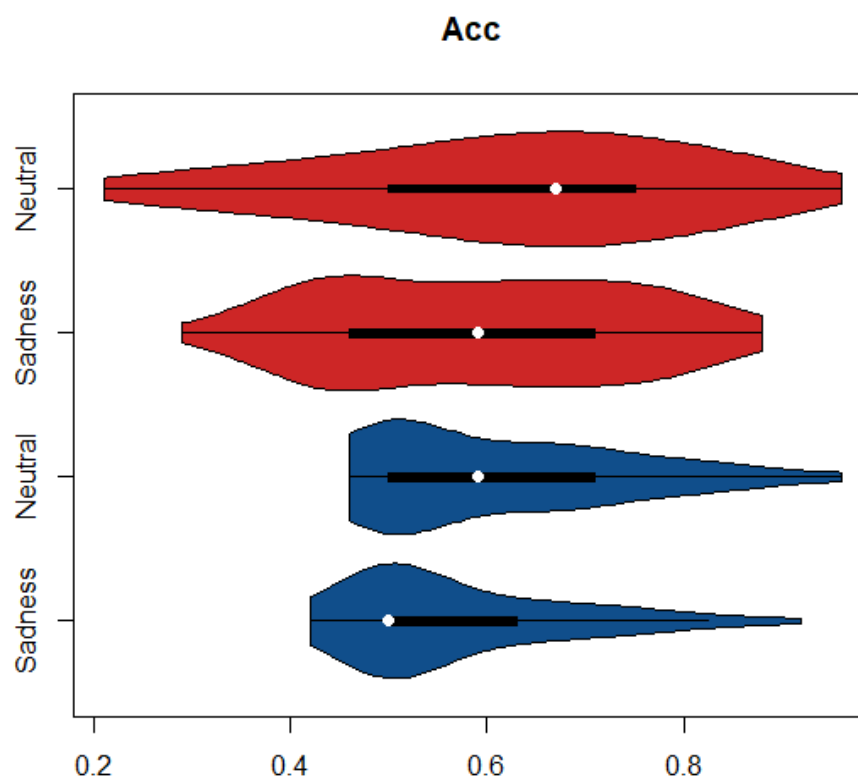
- i. Both systematic error and random error
- ii. Diff and sd
- iii. Yes , it will decrease the power to reject the null hypothesis.
- iv. Type I error

Q2

- a. Visualize the differences between blue-yellow accuracy (BY_ACC) and red-green accuracy (RG_ACC) for both the sad and neutral viewers (Emotion_Condition).

```
library(vioplot)
```

```
vioplot(BY_data[BY_data$Emotion_Condition == 'Sadness'], $ACC,
        BY_data[BY_data$Emotion_Condition == 'Neutral'], $ACC,
        RG_data[RG_data$Emotion_Condition == 'Sadness'], $ACC,
        RG_data[RG_data$Emotion_Condition == 'Neutral'], $ACC,
        main = 'Acc', names = c('Sadness', 'Neutral', 'Sadness', 'N
neutral'),
        col = c('dodgerblue4', 'dodgerblue4', 'firebrick3', 'fireb
rick3'),
        horizontal = TRUE)
```



The red violin plots mean the data from red-green color-axis, the blue ones are from blue-yellow color-axis.

- b. Run a t-test (traditional) to check if there is a significant difference in blue-yellow accuracy between sad and neutral participants at 95% confidence.

```
t.test(BY_data[BY_data$Emotion_Condition == 'Sadness'], $ACC,
        BY_data[BY_data$Emotion_Condition == 'Neutral'], $ACC,
        conf.level = 0.95)
```

```
##
## Welch Two Sample t-test
##
## data:  BY_data[BY_data$Emotion_Condition == "Sadness", ]$ACC
C and BY_data[BY_data$Emotion_Condition == "Neutral", ]$ACC
## t = -2.0435, df = 125.61, p-value = 0.04309
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.086308149 -0.001384159
## sample estimates:
## mean of x mean of y
## 0.5690769 0.6129231
```

From the p-value, $0.04309 < 0.05$ so that there is no significant difference in blue-yellow accuracy between sad and neutral participants.

- c. Run a t-test (traditional) to check if there is a significant difference in red-green accuracy between sad and neutral participants at 95% confidence.

```
t.test(RG_data[RG_data$Emotion_Condition == 'Sadness'],]$ACC,
      RG_data[RG_data$Emotion_Condition == 'Neutral'],]$ACC,
      conf.level = 0.95)
```

```
##
## Welch Two Sample t-test
##
## data:  RG_data[RG_data$Emotion_Condition == "Sadness", ]$ACC
C and RG_data[RG_data$Emotion_Condition == "Neutral", ]$ACC
## t = -0.87491, df = 121.98, p-value = 0.3833
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.08432635 0.03263405
## sample estimates:
## mean of x mean of y
## 0.5944615 0.6203077
```

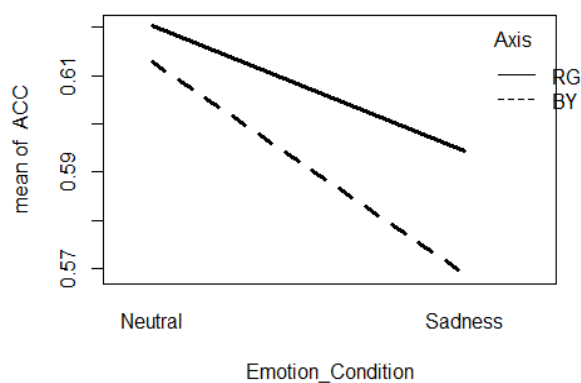
From the p-value, $0.3833 > 0.05$ so that there is significant difference in red-green accuracy between sad and neutral participants.

- d. Do the above t-tests support a claim that there is an interaction between emotion and color axis?
- No. It is hard to find out that there has any interaction between.
- e. Running a factorial design ANOVA, are any of these three factors (emotion/color-axis/interaction) possibly influencing color perception accuracy at any meaningful level of confidence?

```
all_data <- rbind(BY_data, RG_data)
summary(aov(formula = ACC ~ Axis + Emotion_Condition + Axis:Emotion_Condition, data=all_data))
```

```
##              Df Sum Sq Mean Sq F value Pr(>F)
## Axis          1  0.017  0.01745    0.806 0.3703
## Emotion_Condition 1  0.079  0.07893    3.644 0.0574 .
## Axis:Emotion_Condition 1  0.005  0.00526    0.243 0.6224
## Residuals      256  5.545  0.02166
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
with(all_data,
      interaction.plot(
        x.factor = Emotion_Condition,
        trace.factor = Axis,
        response = ACC,
        lwd = 3
      ))
```



From the probability on the Axis, emotion condition and the interaction between there are all above 0.05, which are significant different. Also, the interaction seems influencing a lot in this case. By the interaction plot, the distance between two lines change that means there are interaction between.