# Visualization Literacy Analysis

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#### Read in data files

```
#get the first 6? characters of each data file
#qet unique values of these
# this is list of subject ids
raw_file_names <- list.files("Raw Data")</pre>
first_six <- substr(raw_file_names, 1, 6)</pre>
sub_ids <- unique(first_six)</pre>
length(sub ids)
## [1] 122
fast_RTs <- data.frame(ParticipantId = character(),</pre>
                        TrialName = character(),
                        type = character(),
                        time = numeric()
)
rt_data <- NULL
for (i in 1:length(sub_ids)){
  temp_main_file <- read_csv(paste0("Raw Data/", sub_ids[i], "_maindata.csv")) %>%
    mutate(AnswerRT = TimeToBeginInput - TimeToReadQuestion)
  #three potential RTs to exclude by:
  ## total RT (reading + answering)
  ## reading RT
  ## answering RT (i'm thinking this one)
  if (any(temp_main_file$TimeToReadQuestion < 2000, na.rm = T)){</pre>
    which_index <- which(temp_main_file$TimeToReadQuestion < 2000)</pre>
    for (j in which_index){
      fast_RTs <- add_row(fast_RTs, ParticipantId = sub_ids[i],</pre>
                           TrialName = temp_main_file$TrialName[j],
                           type = "ReadingRT",
                           time = temp_main_file$TimeToReadQuestion[j])
    }
  if (any(temp_main_file$AnswerRT < 2000, na.rm = T)){</pre>
```

```
which_index <- which(temp_main_file$AnswerRT < 2000)</pre>
    for (j in which_index){
      fast_RTs <- add_row(fast_RTs, ParticipantId = sub_ids[i],</pre>
                           TrialName = temp_main_file$TrialName[j],
                           type = "AnswerRT",
                           time = temp_main_file$AnswerRT[j])
    }
  }
  if (any(temp_main_file$TimeToBeginInput < 2000, na.rm = T)){</pre>
    which_index <- which(temp_main_file$TimeToBeginInput < 2000)</pre>
    for (j in which_index) {
      fast_RTs <- add_row(fast_RTs, ParticipantId = sub_ids[i],</pre>
                           TrialName = temp_main_file$TrialName[j],
                           type = "TotalRT",
                           time = temp_main_file$TimeToBeginInput[j])
    }
  rt_data <- rt_data %>%
    bind_rows(temp_main_file)
}
rt_data <- rt_data %>%
  rename(readRT = TimeToReadQuestion, totalRT = TimeToBeginInput)
#read in trialtype key (I created this from an early version of the previous paper)
trial_type_key <- read.csv("trial_type_key.csv", stringsAsFactors = F)</pre>
rt_data <- rt_data %>%
  mutate(TrialType = trial_type_key$TrialType[match(TrialName, trial_type_key$TrialName)]) %>%
  mutate(TrialType = as.factor(TrialType))
```

### Remove outliers

```
dim(rt_data)[1]

## [1] 2074

#122 participants, 17 trials
122*17

## [1] 2074

#remove answerRTs below 2000ms first
rt_data_remove <- rt_data %>%
    filter(AnswerRT >= 2000)
dim(rt_data_remove)[1]

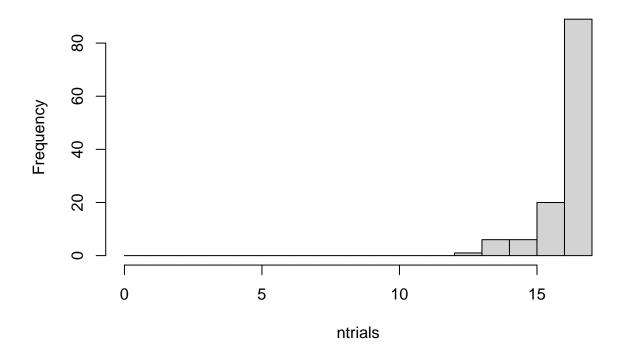
## [1] 2067

#drops 7 trials

rt_data_summary <- rt_data %>%
```

```
group_by(TrialName) %>%
  summarize(meanAnswerRT = mean(AnswerRT, na.rm = T),
            sdAnswerRT = sd(AnswerRT, na.rm = T),
            UB = meanAnswerRT + 3*sdAnswerRT,
            LB = meanAnswerRT - 3*sdAnswerRT)
rt_data_summary
## # A tibble: 17 x 5
##
      TrialName
                  meanAnswerRT sdAnswerRT
                                                 UB
                                                          LB
##
      <chr>
                           <dbl>
                                      <dbl>
                                              <dbl>
                                                       <dbl>
## 1 BarChartQ1
                          27611.
                                     14762. 71897. -16675.
## 2 BarChartQ2
                          16921.
                                     13338. 56935. -23093.
## 3 BarChartQ3
                                      9948. 45452. -14235.
                          15609.
                                      8978. 37222. -16646.
## 4 BarChartQ4
                          10288.
## 5 LineChartQ1
                                     29505. 137699. -39329.
                          49185.
## 6 LineChartQ2
                          40127.
                                     31660. 135107. -54854.
## 7 LineChartQ3
                          27190.
                                     17504. 79701. -25322.
## 8 LineChartQ4
                                     16568. 64483. -34925.
                          14779.
## 9 LineChartQ5
                                     17250. 79628. -23874.
                          27877.
## 10 ScatterplotQ1
                                     24294. 105682. -40080.
                          32801.
                                     23980. 101576. -42304.
## 11 ScatterplotQ2
                          29636.
## 12 ScatterplotQ3
                          45580.
                                     33014. 144623. -53463.
## 13 ScatterplotQ4
                                     25402. 112194. -40219.
                          35987.
## 14 ScatterplotQ5
                          59805.
                                     42684. 187859. -68248.
## 15 SurfacePlotQ1
                                     36042. 164733. -51516.
                          56608.
## 16 SurfacePlotQ2
                          65110.
                                     50062. 215297. -85076.
## 17 SurfacePlotQ3
                          48373.
                                     37424. 160646. -63900.
rt data no outliers <- rt data remove %>%
  group_by(TrialName) %>%
  filter((!(abs(AnswerRT - mean(AnswerRT)) > 3*sd(AnswerRT))))
dim(rt_data_no_outliers)[1]
## [1] 2020
#drops 47 more trials
rt_data_no_outliers %>%
  group by(ParticipantId) %>%
  summarize(ntrials = n()) %>%
  with(hist(ntrials, breaks = 0:17))
```

# **Histogram of ntrials**



##maybe consider replacing outliers with means instead of removing them?

### Compare conditions using individual questions

```
#compare read times (should be no differences of condition)
#compare answer times (potentially a difference)
# first, make .csv files in wide format to double check in statuiew
read_rt_wider <- rt_data_no_outliers %>%
  select(ParticipantId, Condition, TrialName, readRT) %>%
  pivot_wider(names_from = TrialName, values_from = readRT) %>%
  mutate(noutliers = rowSums(is.na(across(3:19))))
answer_rt_wider <- rt_data_no_outliers %>%
  select(ParticipantId, Condition, TrialName, AnswerRT) %>%
  pivot_wider(names_from = TrialName, values_from = AnswerRT) %>%
  mutate(noutliers = rowSums(is.na(across(3:19))),
         mean_all = rowMeans(across(3:19), na.rm = T))
write.csv(read_rt_wider, file = "readRTs.csv", row.names = F)
write.csv(answer_rt_wider, file = "answerRTs.csv", row.names = F)
#within ANOVA can't run with missing cells, so remove participants that had outliers removed
#this loses a lot of data. would probably be better to replace outliers with means, but
#i figure we don't really want to use individual question as the within variable anyway
```

```
#see below for analysis with question TYPE as the within variable
rt_data_no_outliers2 <- rt_data_no_outliers %>%
  group_by(ParticipantId) %>%
  filter(n() > 16)
#READING RTs
#make a little plot
readplot1 <- rt_data_no_outliers2 %>%
  group_by(ParticipantId, Condition) %>%
  summarize(overallmean = mean(readRT)) %>%
  group_by(Condition) %>%
  summarize(overall_condition_mean = mean(overallmean),
            se = std.error(overallmean),
            n = n(),
            CI = qt(0.975, df=n-1)*se)
## `summarise()` has grouped output by 'ParticipantId'. You can override using the `.groups` argument.
ggplot(readplot1, aes(Condition,
                      overall_condition_mean,
                      group = 1,
                      ymin = overall_condition_mean - CI,
                      ymax = overall_condition_mean + CI)) +
  theme_classic() +
  geom_point(size = 4) +
  geom_errorbar(width = .15, size = 0.85) +
  geom_line(size = 0.85) +
  labs(y = "Reading RTs (ms)", title = "Read RTs")
```

```
Read RTs

10000

9000

VR VR Monitor Condition

VR Monitor Stereo
```

```
## Warning: Converting "ParticipantId" to factor for ANOVA.
```

## Warning: Converting "TrialName" to factor for ANOVA.

## Warning: Converting "Condition" to factor for ANOVA.

## Warning: Data is unbalanced (unequal N per group). Make sure you specified a well-## considered value for the type argument to ezANOVA().

```
#get the partial eta-squared too
aovEffectSize(read_rt_anova, effectSize = "pes")$ANOVA
```

```
p p<.05
##
                  Effect DFn DFd
                                           SSn
                                                       SSd
## 1
             (Intercept)
                               86 110050120600 8403185411 1126.276514 3.448191e-51
## 2
               Condition
                           2
                               86
                                     327980221 8403185411
                                                              1.678310 1.927454e-01
               TrialName 16 1376
                                  13109978306 38691894000
                                                              29.139389 6.716381e-76
## 4 Condition:TrialName 32 1376
                                     944710922 38691894000
                                                              1.049899 3.919419e-01
##
            pes
## 1 0.92905909
```

```
## 2 0.03756431
## 3 0.25307924
## 4 0.02383430
#no significant effect of condition on read times
#ANSWERING RTs
#make a little plot
answerplot1 <- rt_data_no_outliers2 %>%
 group_by(ParticipantId, Condition) %>%
  summarize(overallmean = mean(AnswerRT)) %>%
  group_by(Condition) %>%
  summarize(overall_condition_mean = mean(overallmean),
            se = std.error(overallmean),
           n = n(),
           CI = qt(0.975, df=n-1)*se)
## `summarise()` has grouped output by 'ParticipantId'. You can override using the `.groups` argument.
ggplot(answerplot1, aes(Condition,
                      overall_condition_mean,
                      group = 1,
                      ymin = overall_condition_mean - CI,
                      ymax = overall_condition_mean + CI)) +
 theme_classic() +
 geom_point(size = 4) +
 geom_errorbar(width = .15, size = 0.85) +
  geom_line(size = 0.85) +
 labs(y = "Answering RTs (ms)", title = "Answer RTs")
```

```
Answer RTs

35000

(©) 32500

27500

VR Wonitor Condition

VR Monitor Stereo
```

```
## Warning: Converting "ParticipantId" to factor for ANOVA.
```

## Warning: Converting "TrialName" to factor for ANOVA.

## Warning: Converting "Condition" to factor for ANOVA.

## Warning: Data is unbalanced (unequal N per group). Make sure you specified a well-## considered value for the type argument to ezANOVA().

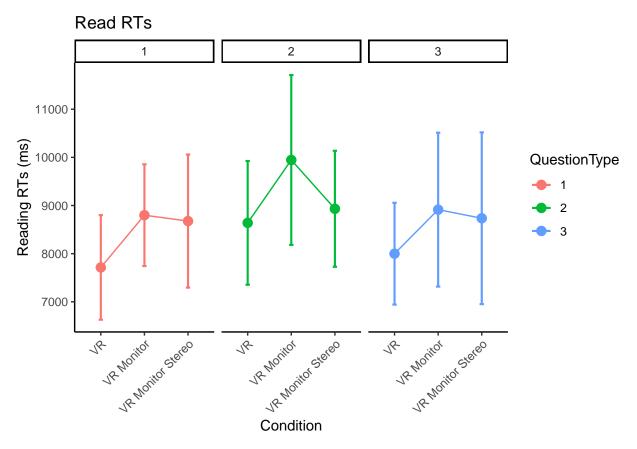
aovEffectSize(answer\_rt\_anova, effectSize = "pes")\$ANOVA

```
p p<.05
##
                  Effect DFn DFd
                                           SSn
                                                        SSd
## 1
             (Intercept)
                           1
                               86 1.418165e+12 9.84632e+10 1238.658028
                                                                         7.596851e-53
## 2
                           2
                               86 8.509034e+09 9.84632e+10
                                                               3.715992 2.832199e-02
               Condition
## 3
               TrialName 16 1376 3.160056e+11 5.13627e+11
                                                             52.910937 9.625859e-131
## 4 Condition:TrialName 32 1376 1.329243e+10 5.13627e+11
                                                               1.112821 3.051122e-01
##
            pes
## 1 0.93507758
## 2 0.07954433
```

```
## 3 0.38089828
## 4 0.02522669
```

# Compare conditions for question type (instead of by individual question)

```
trial_type_means <- rt_data_no_outliers %>%
  group_by(ParticipantId, Condition, TrialType) %>%
  summarize(mean_readRT = mean(readRT),
           mean_answerRT = mean(AnswerRT),
           n = n()
## `summarise()` has grouped output by 'ParticipantId', 'Condition'. You can override using the `.group
read_rt_type_wider <- trial_type_means %>%
  select(ParticipantId, Condition, TrialType, mean_readRT) %>%
 pivot wider(names from = TrialType, values from = mean readRT, names prefix="Type")
answer_rt_type_wider <- trial_type_means %>%
  select(ParticipantId, Condition, TrialType, mean_answerRT) %>%
  pivot_wider(names_from = TrialType, values_from = mean_answerRT, names_prefix="Type")
write.csv(read_rt_type_wider, file = "readtypeRTs.csv", row.names = F)
write.csv(answer_rt_type_wider, file = "answertypeRTs.csv", row.names = F)
#READ RTs
#make the little plot
superbPlot(read_rt_type_wider,
   BSFactors = "Condition",
   WSFactors = "QuestionType(3)",
   variables = c("Type1", "Type2", "Type3"),
   statistic = "mean",
   errorbar = "CI".
              = 0.95,
   adjustments = list(
                   = "difference"
       purpose
   ),
   plotStyle = "line",
   factorOrder = c("Condition", "QuestionType")
) +
theme_classic() +
  theme(axis.text.x = element_text(angle = 45, vjust = 1, hjust=1))+
facet_wrap(vars(QuestionType))+
labs(y = "Reading RTs (ms)", title = "Read RTs")
```



```
## Warning: Converting "ParticipantId" to factor for ANOVA.
```

 $\mbox{\tt \#\#}$  Warning: Converting "Condition" to factor for ANOVA.

## Warning: Data is unbalanced (unequal N per group). Make sure you specified a well-## considered value for the type argument to ezANOVA().

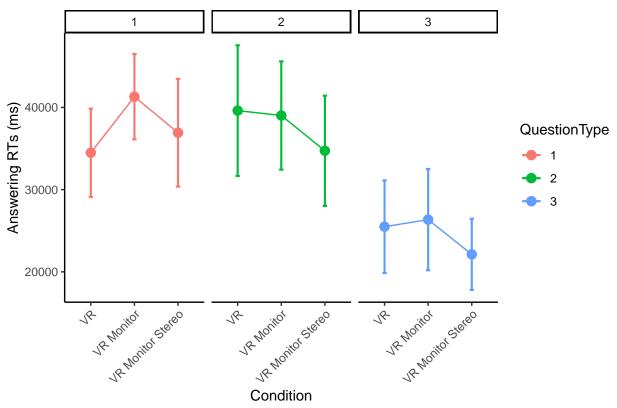
```
#get the partial eta-squared too
aovEffectSize(read_rt_type_anova, effectSize = "pes")$ANOVA
```

```
p p<.05
##
                  Effect DFn DFd
                                          SSn
                                                     SSd
                                                                     F
## 1
                           1 119 26950178706 2148179088 1492.9254659 3.448588e-69
             (Intercept)
                                                            2.2762861 1.071180e-01
## 2
               Condition
                           2 119
                                     82182691 2148179088
## 3
               TrialType
                           2 238
                                     40014704 1107710827
                                                            4.2987300 1.465597e-02
                                     9637742 1107710827
                                                            0.5176853 7.228079e-01
## 4 Condition:TrialType
                           4 238
             pes
## 1 0.926175247
## 2 0.036847247
```

```
## 3 0.034864349
## 4 0.008625546
```

```
#ANSWER RTx
#make the little plot
superbPlot(answer_rt_type_wider,
    BSFactors = "Condition",
    WSFactors = "QuestionType(3)",
    variables = c("Type1", "Type2", "Type3"),
    statistic = "mean",
               = "CI",
    errorbar
               = 0.95,
    adjustments = list(
                     = "difference"
        purpose
    ),
    plotStyle = "line",
    factorOrder = c("Condition", "QuestionType")
) +
theme_classic() +
theme(axis.text.x = element_text(angle = 45, vjust = 1, hjust=1))+
facet_wrap(vars(QuestionType))+
labs(y = "Answering RTs (ms)", title = "Answer RTs")
```

## **Answer RTs**



```
within = TrialType,
                              between = Condition,
                              type = 3,
                              detailed = TRUE,
                              return_aov = T
)
## Warning: Converting "ParticipantId" to factor for ANOVA.
## Warning: Converting "Condition" to factor for ANOVA.
## Warning: Data is unbalanced (unequal N per group). Make sure you specified a well-
## considered value for the type argument to ezANOVA().
#qet the partial eta-squared too
aovEffectSize(answer_rt_type_anova, effectSize = "pes")$ANOVA
                  Effect DFn DFd
##
                                         SSn
                                                     SSd
                                                                  F
                                                                               p p<.05
## 1
             (Intercept) 1 119 395128217295 49497187991 949.958165 1.446491e-58
## 2
              Condition 2 119 1004601416 49497187991 1.207620 3.025441e-01
## 3
              TrialType 2 238 13395765201 21142570734 75.397457 4.324238e-26
## 4 Condition:TrialType 4 238
                                   896611493 21142570734 2.523269 4.168886e-02
##
## 1 0.88867665
## 2 0.01989239
## 3 0.38785207
## 4 0.04068261
```