

Feedback MTurk Study

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Introduction

Load Data

```
# ?register_google
# register_google(key = "AIzaSyCTk2a5vIEqcvgz9KmQmItoNF7J8_hiMMk")
#
# #uses Google API to obtain location data based on longitude and latitude....dont use unless necessary
# d_respondents_only[, c("houzenumber", "street", "city", "county", "state", "zip", "country") := revg
# #
# head(d_respondents_only)
# #
# #
# fwrite(d_respondents_only, file='datatable_clean_survey_responses_v2.dta')
```

```
d_respondents <- fread('datatable_clean_survey_responses_v2.dta')

setnames(d_respondents,
  old = c('Duration (in seconds)'),
  new = c('Survey_Duration'))
head(d_respondents)
```

##	StartDate	EndDate	Status	IPAddress	Progress
## 1:	2020-11-09 20:46:55	2020-11-09 20:50:39	IP Address	174.88.123.135	100
## 2:	2020-11-09 20:47:33	2020-11-09 20:51:24	IP Address	172.93.166.91	100
## 3:	2020-11-09 20:47:23	2020-11-09 20:51:35	IP Address	68.36.215.223	100
## 4:	2020-11-09 20:46:32	2020-11-09 20:51:43	IP Address	99.75.53.174	100
## 5:	2020-11-09 20:47:44	2020-11-09 20:52:08	IP Address	24.35.119.43	100
## 6:	2020-11-09 20:46:47	2020-11-09 20:52:39	IP Address	98.212.214.93	100
##	Survey_Duration	Finished	RecordedDate	ResponseId	
## 1:	223	TRUE	2020-11-09 20:50:39	R_VLuUQ4C82PP9HEd	
## 2:	231	TRUE	2020-11-09 20:51:25	R_29cCZD1XK1dpmdY	
## 3:	251	TRUE	2020-11-09 20:51:35	R_31VN8EncJofnqnV	
## 4:	310	TRUE	2020-11-09 20:51:43	R_50vJlmoFTK1IeB	
## 5:	264	TRUE	2020-11-09 20:52:08	R_1dFaKMSjyE3FJHg	
## 6:	351	TRUE	2020-11-09 20:52:39	R_25vjj4Ik4Dkm2UN	
##	RecipientLastName	RecipientFirstName	RecipientEmail	ExternalReference	
## 1:	NA	NA	NA	NA	
## 2:	NA	NA	NA	NA	
## 3:	NA	NA	NA	NA	
## 4:	NA	NA	NA	NA	
## 5:	NA	NA	NA	NA	
## 6:	NA	NA	NA	NA	

```

##      LocationLatitude LocationLongitude DistributionChannel UserLanguage
## 1:      43.68      -79.29      anonymous      EN
## 2:      33.75      -84.39      anonymous      EN
## 3:      42.66      -83.12      anonymous      EN
## 4:      42.00      -88.14      anonymous      EN
## 5:      40.08      -82.97      anonymous      EN
## 6:      42.01      -88.10      anonymous      EN
##      Amazon_Turk_ID Gender Q82_3_TEXT Age_Range      Education_Level
## 1:  A4D99Y82KOLC8   Male      NA      35-44      Trade school
## 2:  A1AC47WJLW4G7   Male      NA      25-34 Master's degree and above
## 3:  A77K8W55MJEKX   Female     NA      45-54      Bachelor's degree
## 4:  A17TKHT8FEVHOR   Male      NA      25-34      Associate's degree
## 5:  A1AOWM00JMOF7Z   Female     NA      25-34      Trade school
## 6:  A2V08C41JJIQY9   Male      NA      25-34 Master's degree and above
##      Q1      Q2      Q3      Q4      Q5      Q6      Q7
## 1: Pneumonia      Normal      Normal Pneumonia      Normal Pneumonia Pneumonia
## 2: Pneumonia      Normal      Normal Pneumonia      Normal      Normal      Normal
## 3: Pneumonia Pneumonia Pneumonia Pneumonia Pneumonia Pneumonia      Normal
## 4:      Normal Pneumonia Pneumonia      Normal Pneumonia Pneumonia Pneumonia
## 5:      Normal Pneumonia Pneumonia      Normal Pneumonia Pneumonia Pneumonia
## 6: Pneumonia      Normal      Normal Pneumonia Pneumonia Pneumonia      Normal
##      Q8      Q9      Q10
## 1:      Normal Pneumonia Pneumonia
## 2:      Normal      Normal Pneumonia
## 3:      Normal Pneumonia Pneumonia
## 4:      Normal Pneumonia      Normal
## 5:      Normal Pneumonia      Normal
## 6: Pneumonia Pneumonia Pneumonia
##
## 1:
## 2:
## 3:
## 4:
## 5: The sentiment that this place brings, and how much hospitality means to them. How open, diverse and
## 6:
##      Q70_First Click Q70_Last Click Q70_Page Submit Q70_Click Count      Q11
## 1:      NA      NA      NA      NA      NA      Normal
## 2:      NA      NA      NA      NA      NA      Normal
## 3:      NA      NA      NA      NA      NA Pneumonia
## 4:      NA      NA      NA      NA      NA Pneumonia
## 5:      31.08      31.08      77.39      1      Normal
## 6:      NA      NA      NA      NA      NA      Normal
##      Q12      Q13      Q14      Q15      Q16      Q17      Q18
## 1:      Normal Pnuemonia Pneumonia Pneumonia      Normal      Normal Pneumonia
## 2:      Normal Pnuemonia      Normal      Normal Pneumonia      Normal Pneumonia
## 3:      Normal Pnuemonia Pneumonia Pneumonia Pneumonia Pneumonia Pneumonia
## 4:      Normal Pnuemonia      Normal      Normal      Normal Pneumonia      Normal
## 5: Pneumonia      Normal Pneumonia Pneumonia      Normal Pneumonia Pneumonia
## 6:      Normal Pnuemonia Pneumonia      Normal      Normal Pneumonia Pneumonia
##      Q19      Q20
## 1: Pneumonia      Normal
## 2:      Normal Pneumonia
## 3: Pneumonia Pneumonia
## 4: Pneumonia Pneumonia

```

```

## 5: Normal Pneumonia
## 6: Pneumonia Pneumonia
##
## 1:
## 2:
## 3:
## 4:
## 5: It brings awareness to a serious issue that can harm people. It's an advertisement to bring people
## 6:
## Q90_First Click Q90_Last Click Q90_Page Submit Q90_Click Count Q21
## 1: NA NA NA NA Pneumonia
## 2: NA NA NA NA Normal
## 3: NA NA NA NA Pneumonia
## 4: NA NA NA NA Normal
## 5: 10.13 68.74 70.97 2 Pneumonia
## 6: NA NA NA NA Normal
## Q22 Q23 Q24 Q25 Q26 Q27 Q28
## 1: Pneumonia Normal Pneumonia Normal Pneumonia Pneumonia Normal
## 2: Pneumonia Pneumonia Normal Normal Normal Pneumonia Normal
## 3: Pneumonia Normal Pneumonia Pneumonia Pneumonia Pneumonia Pneumonia
## 4: Pneumonia Pneumonia Pneumonia Pneumonia Pneumonia Normal Pneumonia
## 5: Pneumonia Pneumonia Normal Pneumonia Normal Normal Pneumonia
## 6: Pneumonia Pneumonia Normal Pneumonia Normal Normal Pneumonia
## Q29 Q30 Q36
## 1: Normal Pneumonia
## 2: Pneumonia Normal
## 3: Pneumonia Pneumonia
## 4: Pneumonia Normal
## 5: Pneumonia Pneumonia
## 6: Normal Normal
## Self_Reflect_Q1
## 1:
## 2:
## 3:
## 4: I think I did pretty good. I was not expecting to do as well as I did.
## 5:
## 6:
## Q61_First Click Q61_Last Click Q61_Page Submit Q61_Click Count Q41
## 1: NA NA NA NA
## 2: NA NA NA NA
## 3: NA NA NA NA
## 4: 32.36 43.79 111.6 3
## 5: NA NA NA NA
## 6: NA NA NA NA
## Self_Reflect_Q2
## 1:
## 2:
## 3:
## 4: I think I did incredible. I only got 2 wrong. This was harder than the previous page.
## 5:
## 6:
## Q62_First Click Q62_Last Click Q62_Page Submit Q62_Click Count
## 1: NA NA NA NA
## 2: NA NA NA NA

```

3: NA NA NA NA

4: 10.06 10.06 100.3 1

5: NA NA NA NA

6: NA NA NA NA

##

1:

2:

3:

4:

5:

6: Image 2Correct diagnosis: Normal\nYou chose: \${q://QID5/ChoiceGroup/SelectedChoices}\n

Q63_First Click Q63_Last Click Q63_Page Submit Q63_Click Count Q43

1: NA NA NA NA

2: NA NA NA NA

3: NA NA NA NA

4: NA NA NA NA

5: NA NA NA NA

6: 1.205 100.7 108 16

Q64_First Click Q64_Last Click Q64_Page Submit Q64_Click Count Q45

1: NA NA NA NA NA

2: NA NA NA NA NA

3: NA NA NA NA NA

4: NA NA NA NA NA

5: NA NA NA NA NA

6: 76.03 101.2 102.4 3 NA

Q65_First Click Q65_Last Click Q65_Page Submit Q65_Click Count Q47

1: NA NA NA NA NA

2: 14.69 16.23 47.68 2 NA

3: NA NA NA NA NA

4: NA NA NA NA NA

5: NA NA NA NA NA

6: NA NA NA NA NA

Q66_First Click Q66_Last Click Q66_Page Submit Q66_Click Count Q46

1: NA NA NA NA NA

2: 5.75 18.28 46.59 2 NA

3: NA NA NA NA NA

4: NA NA NA NA NA

5: NA NA NA NA NA

6: NA NA NA NA NA

Q67_First Click Q67_Last Click Q67_Page Submit Q67_Click Count Q48

1: 0.855 57.41 58.20 29 NA

2: NA NA NA NA NA

3: 16.263 16.26 50.06 1 NA

4: NA NA NA NA NA

5: NA NA NA NA NA

6: NA NA NA NA NA

Q68_First Click Q68_Last Click Q68_Page Submit Q68_Click Count Total_Score

1: 0.530 60.605 61.22 15 16

2: NA NA NA NA 12

3: 9.427 9.427 49.63 1 15

4: NA NA NA NA 21

5: NA NA NA NA 14

6: NA NA NA NA 18

Random ID Assignment Q1_Score Q2_Score Q3_Score Q4_Score Q5_Score Q6_Score

Q38

```

## 1:      14409      FL_41      0      1      0      1      1      1
## 2:      58508      FL_16      0      1      0      1      1      0
## 3:      96075      FL_41      0      0      1      1      0      1
## 4:      74553      FL_14      1      0      1      0      0      1
## 5:      35543      FL_17      1      0      1      0      0      1
## 6:      84565      FL_15      0      1      0      1      0      1
##      Q7_Score Q8_Score Q9_Score Q10_Score Q11_Score Q12_Score Q13_Score Q14_Score
## 1:      1      1      1      0      0      1      0      1
## 2:      0      1      0      0      0      1      0      0
## 3:      0      1      1      0      1      1      0      1
## 4:      1      1      1      1      1      1      0      0
## 5:      1      1      1      1      0      0      0      1
## 6:      0      0      1      0      0      1      0      1
##      Q15_Score Q16_Score Q17_Score Q18_Score Q19_Score Q20_Score Q21_Score
## 1:      0      1      0      0      0      1      1
## 2:      1      0      0      0      0      0      1
## 3:      0      0      1      0      1      0      0
## 4:      1      1      1      1      1      0      1
## 5:      0      1      1      0      0      0      0
## 6:      1      1      1      0      1      0      1
##      Q22_Score Q23_Score Q24_Score Q25_Score Q26_Score Q27_Score Q28_Score
## 1:      0      0      0      0      1      1      0
## 2:      0      1      1      0      0      1      0
## 3:      0      0      0      1      1      1      1
## 4:      0      1      0      1      1      0      1
## 5:      0      1      1      1      0      0      1
## 6:      0      1      1      1      0      0      1
##      Q29_Score Q30_Score Assignment_Group TaskPhase1_Score TaskPhase2_Score
## 1:      1      0 Negative Images      0.7      0.5
## 2:      0      1 Positive Images      0.4      0.2
## 3:      0      0 Negative Images      0.5      0.5
## 4:      0      1 Self-Reflect      0.7      0.7
## 5:      0      0 Control      0.7      0.3
## 6:      1      1 Medical Feedback      0.4      0.6
##      TaskPhase3_Score housenumber      street      city
## 1:      0.3      351      Glen Manor Drive      Toronto
## 2:      0.5      262 Capitol Avenue Southeast      Atlanta
## 3:      0.4      440      Bedlington Drive Rochester Hills
## 4:      0.6      1200      Sycamore Avenue      Hanover Park
## 5:      0.4      1913      Brookfield Road      Columbus
## 6:      0.7      617      Boxwood Drive      Schaumburg
##      county      state      zip      country
## 1:      Canada      Ontario M4E 2X8      Canada
## 2: United States      Georgia      30312 United States
## 3: United States      Michigan      48307 United States
## 4: United States      Illinois      60133 United States
## 5: United States      Ohio      43229 United States
## 6: United States      Illinois      60193 United States

```

```
nrow(d_respondents)
```

```
## [1] 350
```

```
#remove duplicate Amazon Turk IDs
```

```
nrow(d_respondents) #350 rows
```

```
## [1] 350
```

```
d_respondents <- d_respondents[ !duplicated(d_respondents$Amazon_Turk_ID) , ] #350 rows
```

EDA

Helper Functions

```
create_heatmap <- function(var1, var2) {  
  ### Create a heatmap for a table of frequencies between two variables ###  
  df <- data.frame(table(var1,var2))  
  
  ggplot(df,aes(x=var1,y=var2)) +  
    geom_tile(aes(fill=Freq,color=Freq),show.legend=FALSE,alpha=.8) +  
    geom_text(aes(label=Freq)) +  
    scale_fill_continuous(high = "darkslategray4", low = "powderblue")  
}  
  
g_legend<-function(a.gplot){  
  #extract legend from a ggplot object  
  #https://stackoverflow.com/questions/13649473/add-a-common-legend-for-combined-ggplots  
  #https://github.com/hadley/ggplot2/wiki/Share-a-legend-between-two-ggplot2-graphs  
  tmp <- ggplot_gtable(ggplot_build(a.gplot))  
  leg <- which(sapply(tmp$grobs, function(x) x$name) == "guide-box")  
  legend <- tmp$grobs[[leg]]  
  return(legend)}  
  
#some EDA  
  
#d_respondents[ , table(state, country)]  
  
table(d_respondents$state, d_respondents$country) %>%  
  as.data.frame() %>%  
  arrange(desc(Freq)) %>%  
  filter(Freq>0)
```

##	Var1	Var2	Freq
## 1	Tamil Nadu	India	107
## 2	California	United States	72
## 3	New York	United States	22
## 4	Kansas	United States	21
## 5	Texas	United States	15
## 6	Florida	United States	9
## 7	Massachusetts	United States	7
## 8	Missouri	United States	6
## 9	Connecticut	United States	5
## 10	Georgia	United States	5
## 11	Indiana	United States	5
## 12	Michigan	United States	5
## 13	New Jersey	United States	5
## 14	Illinois	United States	4
## 15	Virginia	United States	4
## 16	Kerala	India	3

```
## 17      Maharashtra      India 3
## 18      Colorado United States 3
## 19      Kentucky United States 3
## 20      Maryland United States 3
## 21      North Carolina United States 3
## 22      Oregon United States 3
## 23      Ontario Canada 2
## 24      Alabama United States 2
## 25      Idaho United States 2
## 26      Minnesota United States 2
## 27      Mississippi United States 2
## 28      Nevada United States 2
## 29      Ohio United States 2
## 30      Pennsylvania United States 2
## 31      Washington United States 2
## 32      Qarku i Tiranës Albania 1
## 33      Khulna Division Bangladesh 1
## 34      Bahia Brazil 1
## 35      Atacama Chile 1
## 36 Provence-Alpes-Côte d'Azur France 1
## 37      Departamento de Olancho Honduras 1
## 38      Andhra Pradesh India 1
## 39      Karnataka India 1
## 40      Sardegna Italy 1
## 41      England United Kingdom 1
## 42      Arizona United States 1
## 43      Iowa United States 1
## 44      Louisiana United States 1
## 45      Maine United States 1
## 46      Nebraska United States 1
## 47      Oklahoma United States 1
## 48      South Carolina United States 1
## 49      South Dakota United States 1
## 50      Tennessee United States 1
```

```
table(d_respondents$country) %>%
  as.data.frame() %>%
  arrange(desc(Freq))
```

```
##      Var1 Freq
## 1 United States 225
## 2      India 115
## 3      Canada 2
## 4      Albania 1
## 5      Bangladesh 1
## 6      Brazil 1
## 7      Chile 1
## 8      France 1
## 9      Honduras 1
## 10      Italy 1
## 11 United Kingdom 1
```

```
table(d_respondents$Total_Score) %>%
  as.data.frame() %>%
  arrange(desc(Var1))
```

```
##      Var1 Freq
## 1      27    1
## 2      26    1
## 3      25    4
## 4      24   12
## 5      23   15
## 6      22   16
## 7      21   22
## 8      20   27
## 9      19   21
## 10     18   31
## 11     17   40
## 12     16   40
## 13     15   30
## 14     14   30
## 15     13   19
## 16     12   18
## 17     11   13
## 18     10    6
## 19      9    3
## 20      8    1
```

```
d_respondents %>%
  group_by(Assignment_Group) %>%
  summarise(mean = mean(Total_Score),
            count = n(),
            time_duration = mean(Survey_Duration))
```

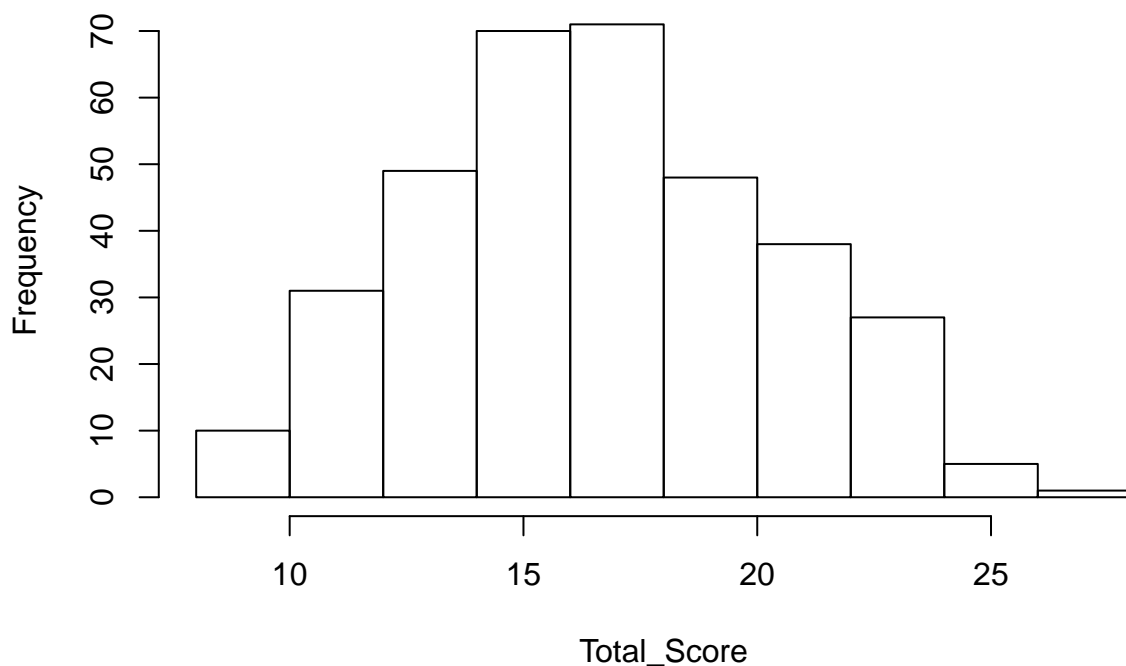
```
## `summarise()` ungrouping output (override with `.groups` argument)
```

```
## # A tibble: 5 x 4
##   Assignment_Group mean count time_duration
##   <chr>          <dbl> <int>      <dbl>
## 1 Control        16.7    69      638.
## 2 Medical Feedback 17.8    70      656.
## 3 Negative Images  16.5    72      783
## 4 Positive Images  17.3    70      505.
## 5 Self-Reflect    17.2    69      612.
```

```
#d_respondents[, .(count = .N, avg = mean(Total_Score)), by=Assignment_Group] #same thing
```

```
d_respondents[, hist(Total_Score)]
```


Histogram of Total_Score



```
## $breaks
## [1] 8 10 12 14 16 18 20 22 24 26 28
##
## $counts
## [1] 10 31 49 70 71 48 38 27 5 1
##
## $density
## [1] 0.014286 0.044286 0.070000 0.100000 0.101429 0.068571 0.054286 0.038571
## [9] 0.007143 0.001429
##
## $mids
## [1] 9 11 13 15 17 19 21 23 25 27
##
## $xname
## [1] "Total_Score"
##
## $equidist
## [1] TRUE
##
## attr("class")
## [1] "histogram"
```

```
tapply(d_respondents$Total_Score, d_respondents$Assignment_Group, summary)
```

```
## $Control
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      8.0   14.0   16.0   16.7   19.0   24.0
##
## $`Medical Feedback`
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
```

```
##      10.0      16.0      17.5      17.8      20.0      24.0
##
## $`Negative Images`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      9.0      13.0      16.0      16.5      19.2      25.0
##
## $`Positive Images`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      9.0      15.0      17.0      17.3      20.0      27.0
##
## $`Self-Reflect`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      9.0      14.0      17.0      17.2      20.0      25.0
```

```
tapply(d_respondents$Total_Score, d_respondents$Assignment_Group, sd)
```

```
##           Control Medical Feedback  Negative Images  Positive Images
##           3.659                3.279                3.996                3.817
##      Self-Reflect
##           3.882
```

```
d_respondents[, sd(Total_Score)]
```

```
## [1] 3.743
```

```
library(ggmap)
?register_google
register_google(key = "AIzaSyCTk2a5vIEqcvgz9KmQmItoNF7J8_hiMMk")
# ggmap_show_api_key()

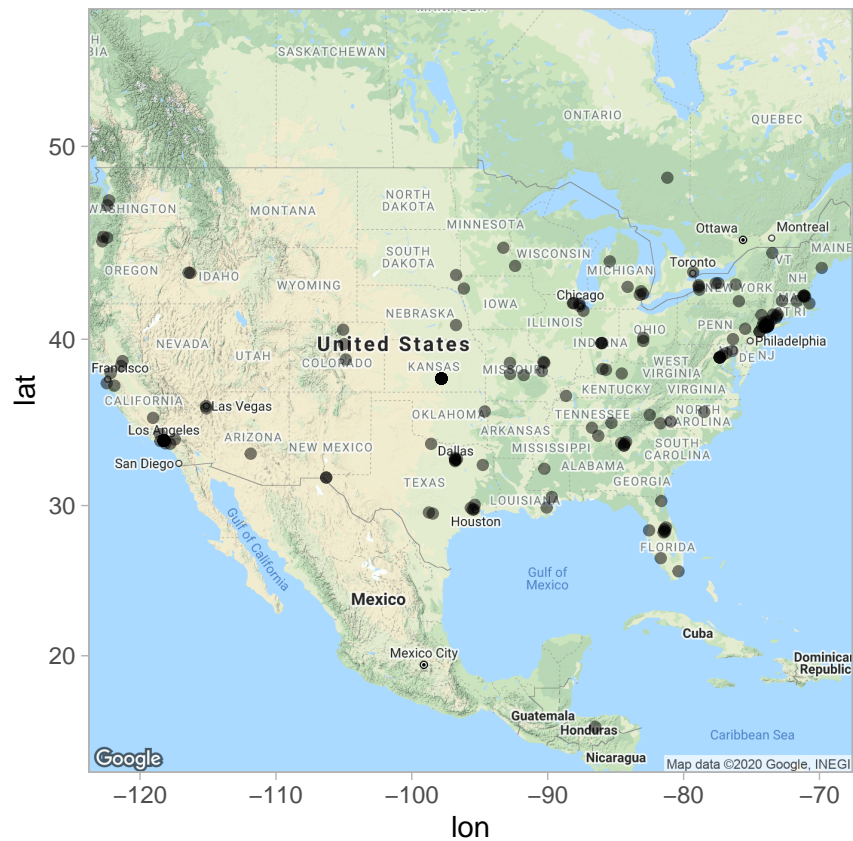
us_map<-get_map(location='united states', zoom=4, maptype = "terrain",
                source='google',color='color')
```

```
## Source : https://maps.googleapis.com/maps/api/staticmap?center=united%20states&zoom=4&size=640x640&s
```

```
## Source : https://maps.googleapis.com/maps/api/geocode/json?address=united+states&key=xxx
```

```
ggmap(us_map) + geom_point(x=d_respondents$LocationLongitude, y = d_respondents$LocationLatitude, show_
```

```
## Warning: `show_guide` has been deprecated. Please use `show.legend` instead.
```



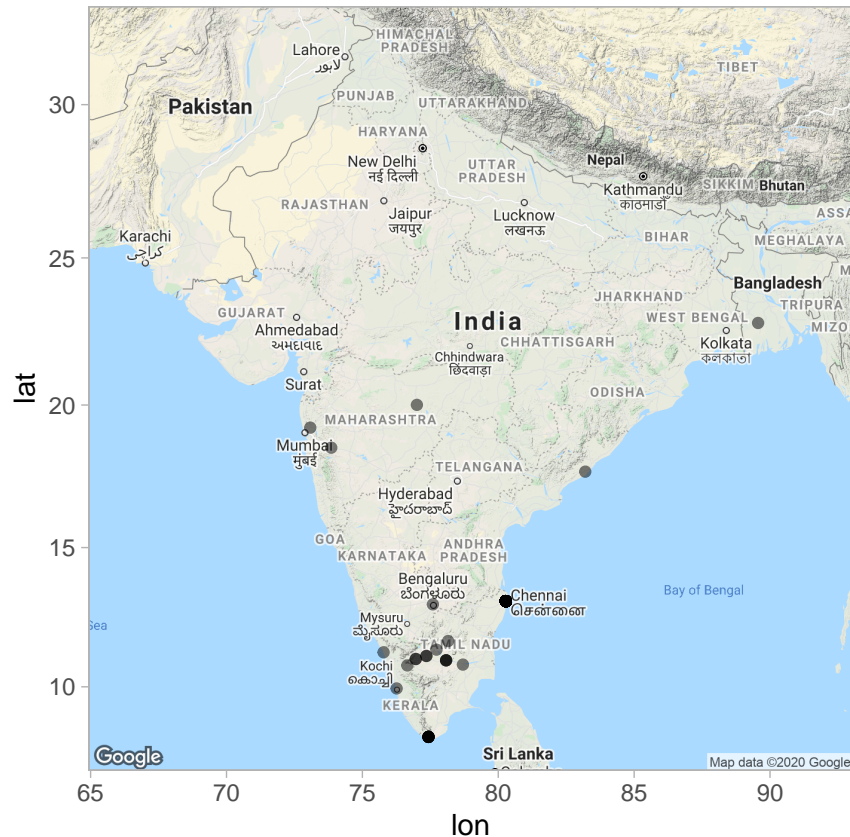
```
india_map<-get_map(location='india', zoom=5, maptype = "terrain",
                  source='google',color='color')
```

```
## Source : https://maps.googleapis.com/maps/api/staticmap?center=india&zoom=5&size=640x640&scale=2&map
```

```
## Source : https://maps.googleapis.com/maps/api/geocode/json?address=india&key=xxx
```

```
ggmap(india_map) + geom_point(x=d_respondents$LocationLongitude, y = d_respondents$LocationLatitude, sh
```

```
## Warning: `show_guide` has been deprecated. Please use `show.legend` instead.
```



Randomization Check

<http://www.sthda.com/english/wiki/chi-square-goodness-of-fit-test-in-r>

```
respondent_counts <- d_respondents[ , .(N), keyby=Assignment_Group]

respondent_counts_chisq_test <- chisq.test(respondent_counts[,2], p=c(1/5, 1/5, 1/5, 1/5, 1/5))
pander(respondent_counts_chisq_test, style = 'rmarkdown')
```

Table 1: Chi-squared test for given probabilities:
respondent_counts[, 2]

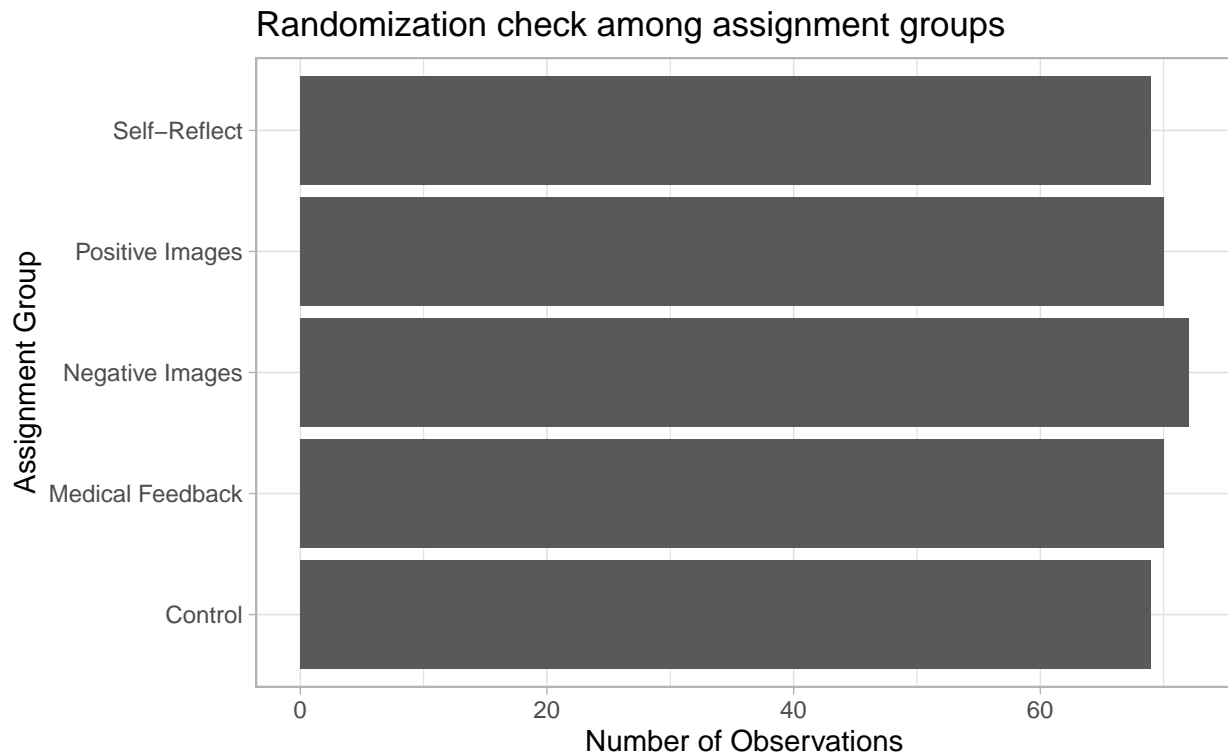
Test statistic	df	P value
0.08571	4	0.9991

```
respondent_counts %>%
  ggplot(aes(x = Assignment_Group, y=N)) +
  geom_bar(stat = 'identity') +
  coord_flip() +
  xlab('Assignment Group') +
  ylab('Number of Observations') +
  labs(title='Randomization check among assignment groups',
       caption = paste0('Assuming equal distribution among assignment groups, a chi-squared goodness of
round(respondent_counts_chisq_test$parameter,4),' degrees of \nfreedom ', 'yield
round(respondent_counts_chisq_test$p.value,4),
```

```

', suggesting that the observed proportions are not significantly different from
theme(plot.caption = element_text(hjust = 0))

```



Assuming equal distribution among assignment groups, a chi-squared goodness of fit test with 4 degrees of freedom yields $p=0.9991$, suggesting that the observed proportions are not significantly different from the expected proportions at a significance level of 0.05.

*#p-value = 0.9991, which is greater than significance level of 0.05.
 #We can conclude that the observed proportions are not significantly different from the expected proportions*

Covariate Balance Check

#let's consider adding age bins and education bins

```

d_respondents[ Age_Range == "18-24", age_bin := 1]
d_respondents[ Age_Range == "25-34", age_bin := 2]
d_respondents[ Age_Range == "35-44", age_bin := 3]
d_respondents[ Age_Range == "45-54", age_bin := 4]
d_respondents[ Age_Range == "55-64", age_bin := 5]
d_respondents[ Age_Range == "Above 65", age_bin := 6]

d_respondents[ Education_Level == "Associate's degree", edu_bin := 1]
d_respondents[ Education_Level == "Bachelor's degree", edu_bin := 2]
d_respondents[ Education_Level == "High school", edu_bin := 3]
d_respondents[ Education_Level == "Master's degree and above", edu_bin := 4]
d_respondents[ Education_Level == "Some high school", edu_bin := 5]
d_respondents[ Education_Level == "Trade school", edu_bin := 6]

d_respondents[ Assignment_Group == "Control", assign_bin := 1]
d_respondents[ Assignment_Group == "Medical Feedback", assign_bin := 2]

```

```

d_respondents[ Assignment_Group == "Negative Images", assign_bin := 3]
d_respondents[ Assignment_Group == "Positive Images", assign_bin := 4]
d_respondents[ Assignment_Group == "Self-Reflect", assign_bin := 5]

d_respondents[ , US_Dummy := ifelse(country == "United States", 1, 0)]

d_respondents[ , Male_Dummy := ifelse(Gender == "Male", 1, 0)]

#add treatment dummy

d_respondents[ , Treatment_Dummy := ifelse(Assignment_Group != "Control", 1, 0)]

#head(d_respondents)

d_respondents %>%
  group_by(Assignment_Group) %>%
  summarise(num_respondents = n(),
            pre_treatment_avg = mean(TaskPhase1_Score),
            taskphase2_avg = mean(TaskPhase2_Score),
            taskphase3_avg = mean(TaskPhase3_Score))

## `summarise()` ungrouping output (override with `.groups` argument)

## # A tibble: 5 x 5
##   Assignment_Group num_respondents pre_treatment_a~ taskphase2_avg
##   <chr>              <int>          <dbl>          <dbl>
## 1 Control              69            0.607            0.461
## 2 Medical Feedback     70            0.634            0.523
## 3 Negative Images      72            0.578            0.494
## 4 Positive Images      70            0.614            0.514
## 5 Self-Reflect         69            0.599            0.526
## # ... with 1 more variable: taskphase3_avg <dbl>

d_respondents %>%
  group_by(Assignment_Group) %>%
  summarise(num_respondents = n(),
            avg_age_bin = mean(age_bin),
            avg_edu_bin = mean(edu_bin),
            male = mean(Male_Dummy),
            US = mean(US_Dummy))

## `summarise()` ungrouping output (override with `.groups` argument)

## # A tibble: 5 x 6
##   Assignment_Group num_respondents avg_age_bin avg_edu_bin male    US
##   <chr>              <int>          <dbl>          <dbl> <dbl> <dbl>
## 1 Control              69            2.68            2.61 0.609 0.652
## 2 Medical Feedback     70            2.63            2.47 0.586 0.529
## 3 Negative Images      72            2.62            2.58 0.583 0.625
## 4 Positive Images      70            2.86            2.6  0.586 0.714
## 5 Self-Reflect         69            2.83            2.42 0.594 0.696

d_respondents %>%
  group_by(Assignment_Group) %>%
  summarise(num_respondents = n(),
            )

```

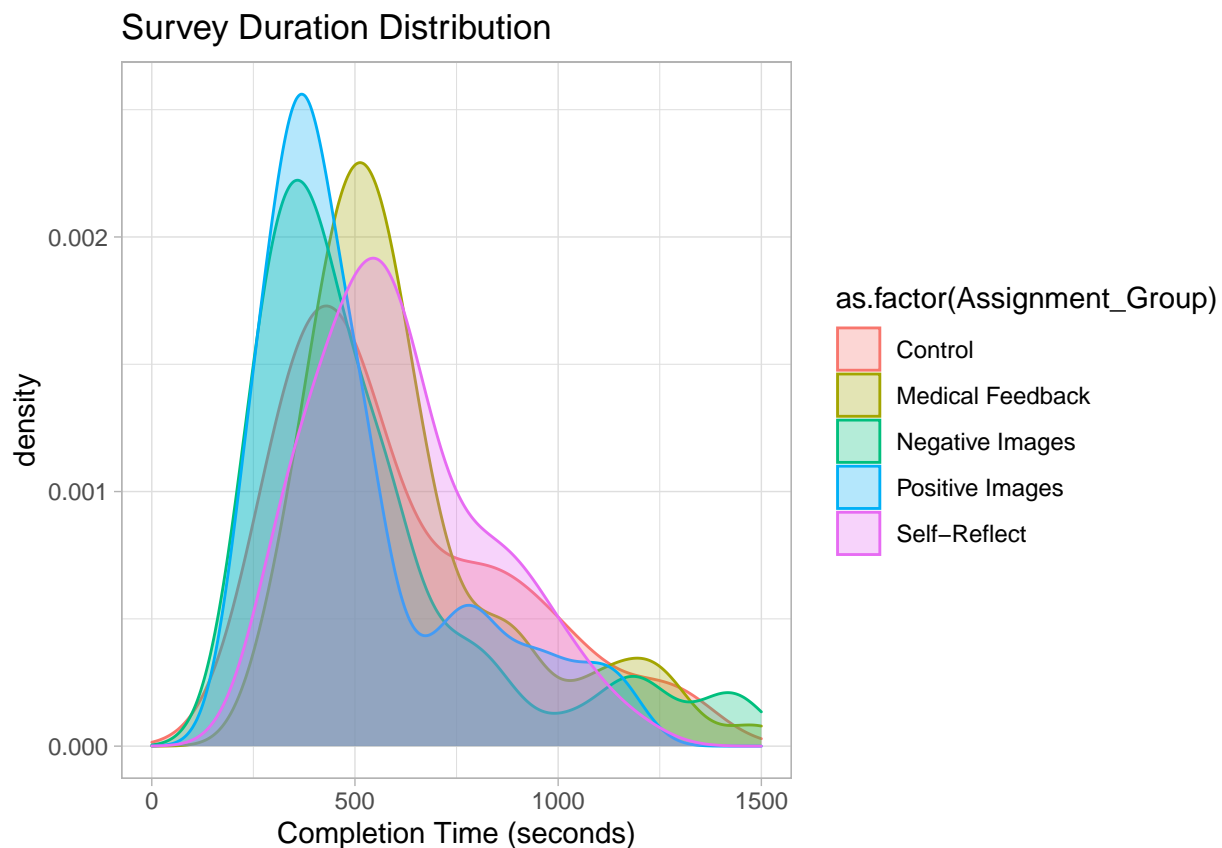
```
## `summarise()` ungrouping output (override with `.groups` argument)

## # A tibble: 5 x 2
##   Assignment_Group num_respondents
##   <chr>             <int>
## 1 Control             69
## 2 Medical Feedback    70
## 3 Negative Images     72
## 4 Positive Images     70
## 5 Self-Reflect        69
```

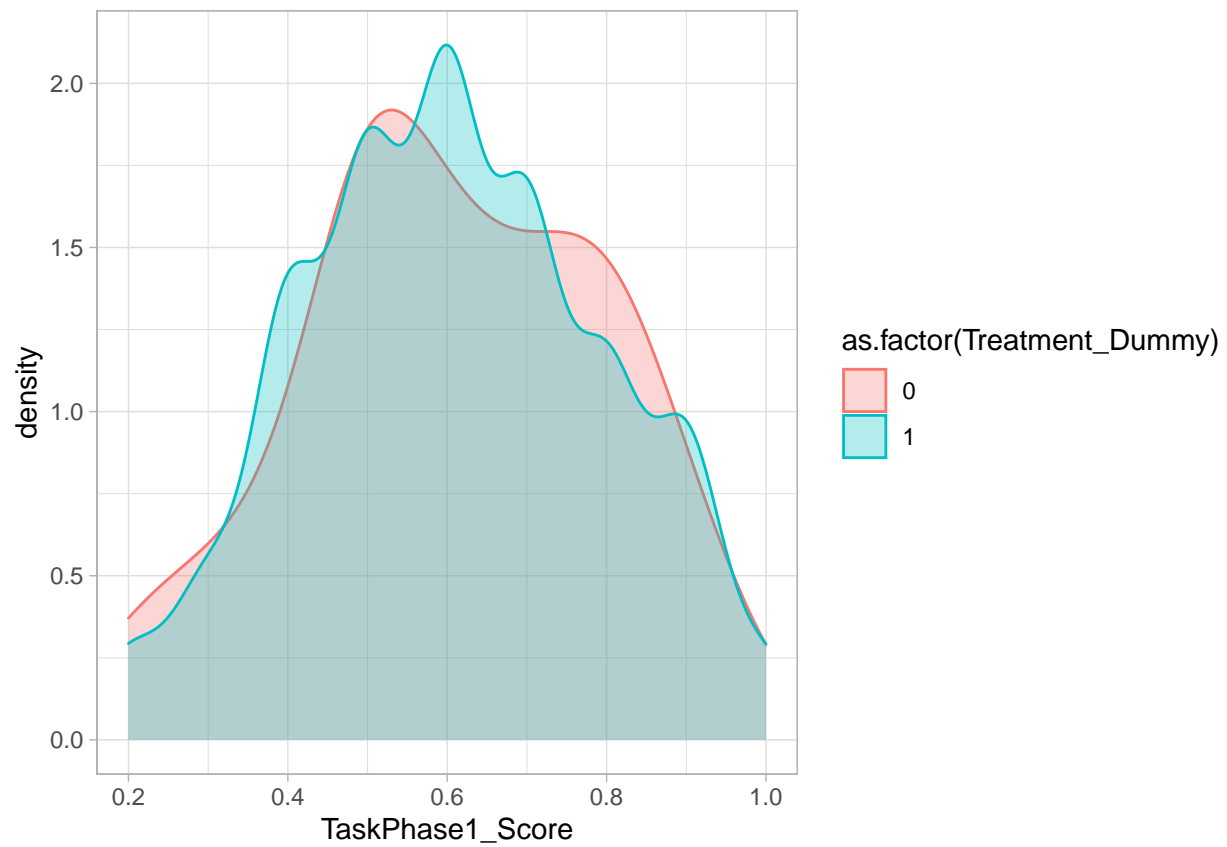
Visuals

```
#Density distribution of Survey Duration
ggplot(d_respondents, aes(x=Survey_Duration, colour=as.factor(Assignment_Group), fill = as.factor(Assignment_Group))) +
  geom_density(alpha = 0.3) +
  xlim(0, 1500) +
  xlab("Completion Time (seconds)") +
  ggtitle("Survey Duration Distribution")
```

```
## Warning: Removed 6 rows containing non-finite values (stat_density).
```

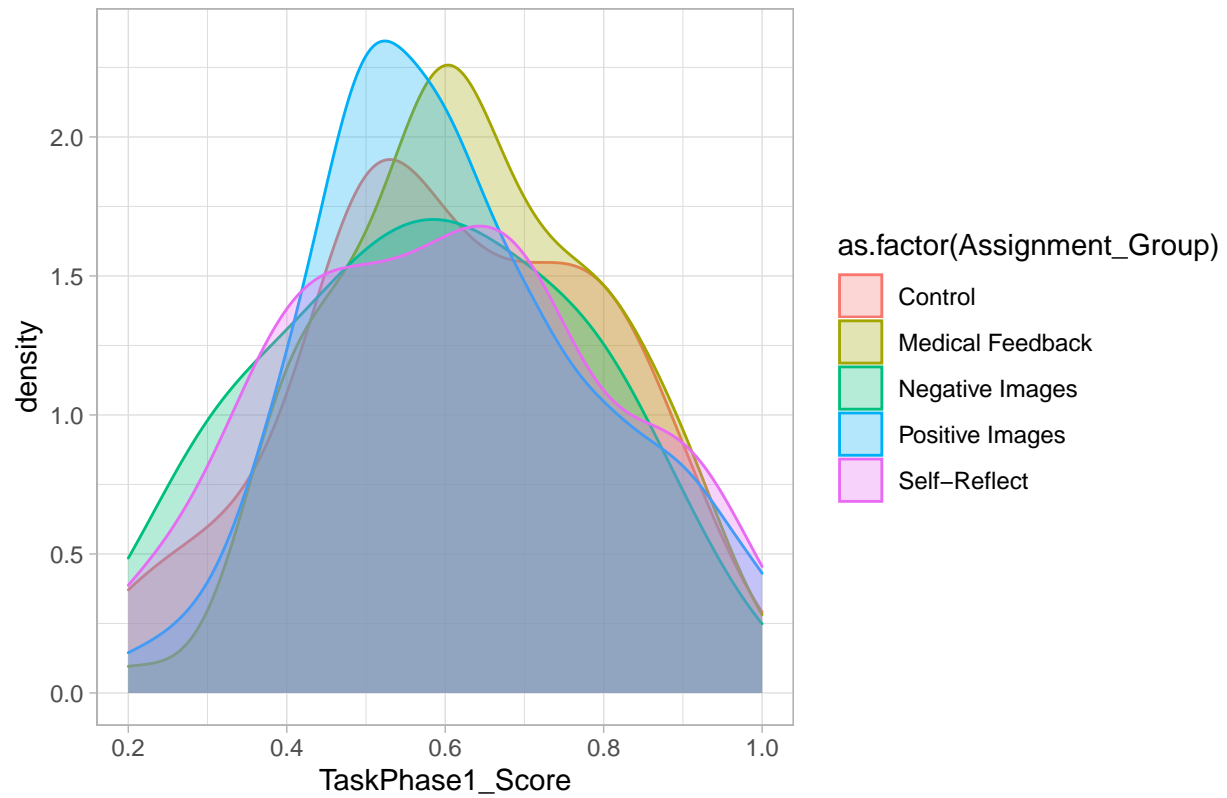


```
#Comparing pretreatment values
ggplot(d_respondents, aes(x=TaskPhase1_Score, fill = as.factor(Treatment_Dummy), colour=as.factor(Treatment_Dummy))) +
```



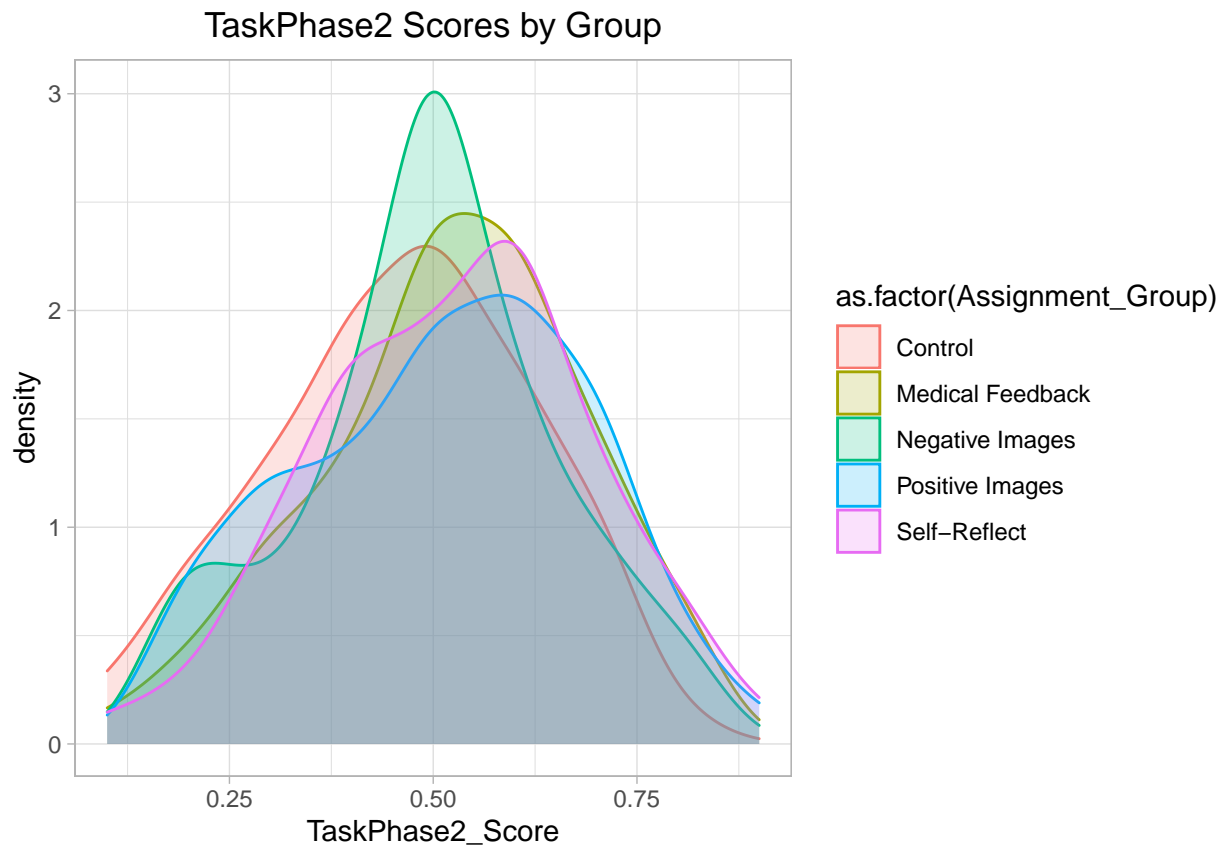
```
ggplot(d_respondents, aes(x=TaskPhase1_Score, fill = as.factor(Assignment_Group), colour=as.factor(Assignment_Group)))
```


PreTreatment Scores by Group



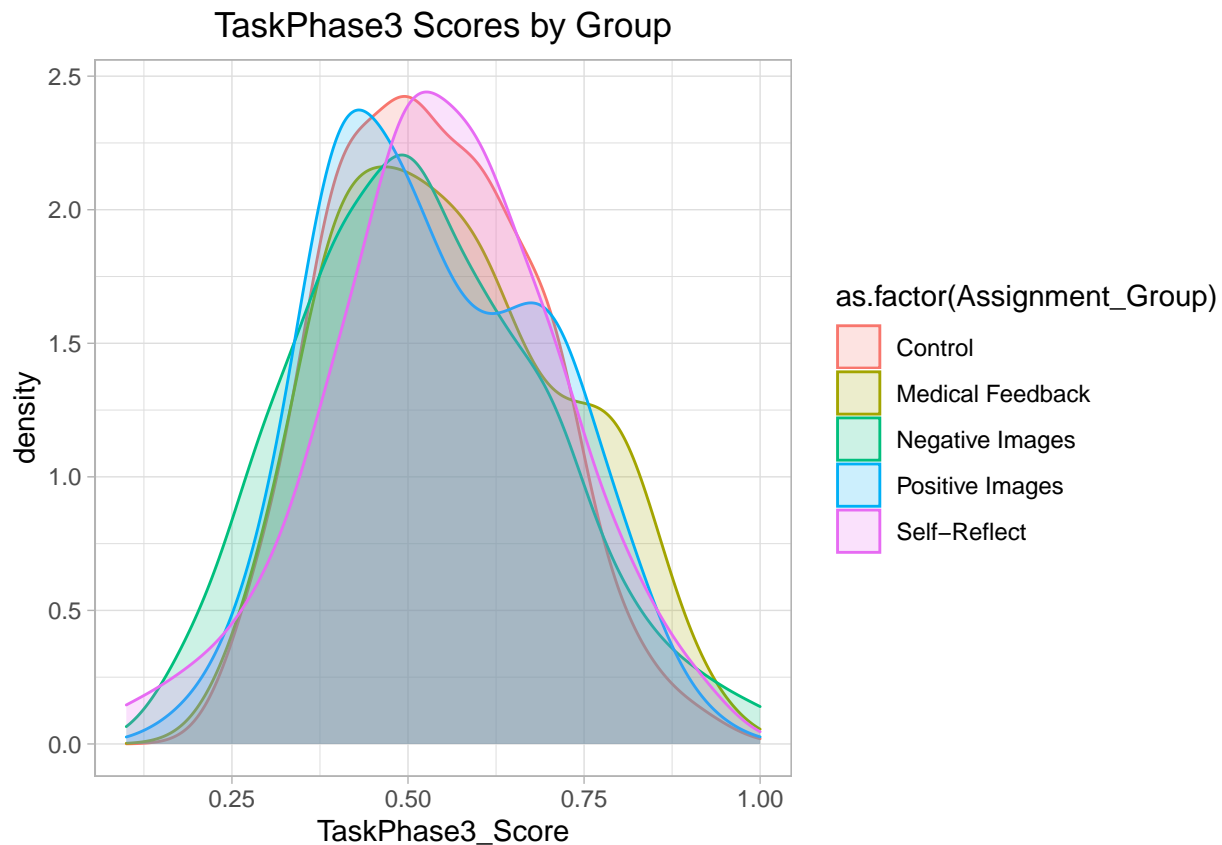
#Comparing taskphase2 values

```
ggplot(d_respondents, aes(x=TaskPhase2_Score, fill = as.factor(Assignment_Group), colour=as.factor(Assi
```



```
#Comparing taskphase3 values
```

```
ggplot(d_respondents, aes(x=TaskPhase3_Score, fill = as.factor(Assignment_Group), colour=as.factor(Assi
```



TODO add these boxplots for binary treatment/control groups

```
task2a_bp <- ggplot(d_respondents, aes(x = Treatment_Dummy, y=TaskPhase1_Score, colour=as.factor(Treatment_Dummy))) +
  geom_boxplot() +
  stat_summary(fun.y = mean, geom = "errorbar", aes(ymax = ..y.., ymin = ..y..), width = .75, linetype = "solid") +
  xlab('') +
  ylab('Task Score (%)') +
  ggtitle("Pre Treatment Scores") +
  scale_y_continuous(labels = scales::percent, limits = c(0,1)) +
  theme(axis.text.x = element_blank(),
        axis.ticks = element_blank(),
        plot.title = element_text(hjust = 0.5, size=10),
        legend.position = "bottom",
        legend.title = element_blank())
```

Warning: `fun.y` is deprecated. Use `fun` instead.

```
task2b_bp <- ggplot(d_respondents, aes(x = Treatment_Dummy, y=TaskPhase2_Score, colour=as.factor(Treatment_Dummy))) +
  geom_boxplot() +
  stat_summary(fun.y = mean, geom = "errorbar", aes(ymax = ..y.., ymin = ..y..), width = .75, linetype = "solid") +
  xlab('') +
  ylab('') +
  ggtitle("Task Phase 2 Scores") +
  scale_y_continuous(labels = scales::percent, limits = c(0,1)) +
  theme(axis.text.x = element_blank(),
        axis.ticks = element_blank(),
        plot.title = element_text(hjust = 0.5, size=10),
        legend.position = "none")
```

```
## Warning: `fun.y` is deprecated. Use `fun` instead.
```

```
task2c_bp <- ggplot(d_respondents, aes(x = Treatment_Dummy, y=TaskPhase3_Score, colour=as.factor(Treatment_Dummy))) +
  geom_boxplot() +
  stat_summary(fun.y = mean, geom = "errorbar", aes(ymax = ..y.., ymin = ..y..), width = .75, linetype="dashed") +
  xlab('') +
  ylab('') +
  ggtitle("Task Phase 3 Scores") +
  scale_y_continuous(labels = scales::percent, limits = c(0,1)) +
  theme(axis.text.x = element_blank(),
        axis.ticks = element_blank(),
        plot.title = element_text(hjust = 0.5, size=10),
        legend.position = "none")
```

```
## Warning: `fun.y` is deprecated. Use `fun` instead.
```

```
mylegend_2<-g_legend(task2a_bp)
```

```
grid.arrange(arrangeGrob(task2a_bp + theme(legend.position="none"), task2b_bp, task2c_bp, ncol=3),
  mylegend_2,
  nrow=2,
  heights=c(10,1),
  top = textGrob("Compare task scores in different phases\n", just='right', gp=gpar(fontsize=14)))
```

Compare task scores in different phases



```
# boxplots for multiple treatment groups
```

```
task1a_bp <- ggplot(d_respondents, aes(x = Assignment_Group, y=TaskPhase1_Score, colour=as.factor(Assignment_Group))) +
  geom_boxplot() +
```

```

stat_summary(fun.y = mean, geom = "errorbar", aes(ymax = ..y.., ymin = ..y..), width = .75, linetype = "solid",
xlab('') +
ylab('Task Score (%)') +
ggtitle("Pre Treatment Scores") +
scale_y_continuous(labels = scales::percent,limits = c(0,1)) +
theme(axis.text.x = element_blank(),
axis.ticks = element_blank(),
plot.title = element_text(hjust = 0.5,size=10),
legend.position = "bottom",
legend.title = element_blank())

## Warning: `fun.y` is deprecated. Use `fun` instead.

task1b_bp <- ggplot(d_respondents, aes(x = Assignment_Group, y=TaskPhase2_Score, colour=as.factor(Assignment_Group)) +
geom_boxplot() +
stat_summary(fun.y = mean, geom = "errorbar", aes(ymax = ..y.., ymin = ..y..), width = .75, linetype = "solid",
xlab('') +
ylab('') +
ggtitle("Task Phase 2 Scores") +
scale_y_continuous(labels = scales::percent,limits = c(0,1)) +
theme(axis.text.x = element_blank(),
axis.ticks = element_blank(),
plot.title = element_text(hjust = 0.5,size=10),
legend.position = "none")

## Warning: `fun.y` is deprecated. Use `fun` instead.

task1c_bp <- ggplot(d_respondents, aes(x = Assignment_Group, y=TaskPhase3_Score, colour=as.factor(Assignment_Group)) +
geom_boxplot() +
stat_summary(fun.y = mean, geom = "errorbar", aes(ymax = ..y.., ymin = ..y..), width = .75, linetype = "solid",
xlab('') +
ylab('') +
ggtitle("Task Phase 3 Scores") +
scale_y_continuous(labels = scales::percent,limits = c(0,1)) +
theme(axis.text.x = element_blank(),
axis.ticks = element_blank(),
plot.title = element_text(hjust = 0.5,size=10),
legend.position = "none")

## Warning: `fun.y` is deprecated. Use `fun` instead.

mylegend_1<-g_legend(task1a_bp)

grid.arrange(arrangeGrob(task1a_bp + theme(legend.position="none"),task1b_bp,task1c_bp,ncol=3),
mylegend_1,
nrow=2,
heights=c(10,1),
top = textGrob("Compare task scores in different phases\n",just='right',gp=gpar(fontsize=14)))

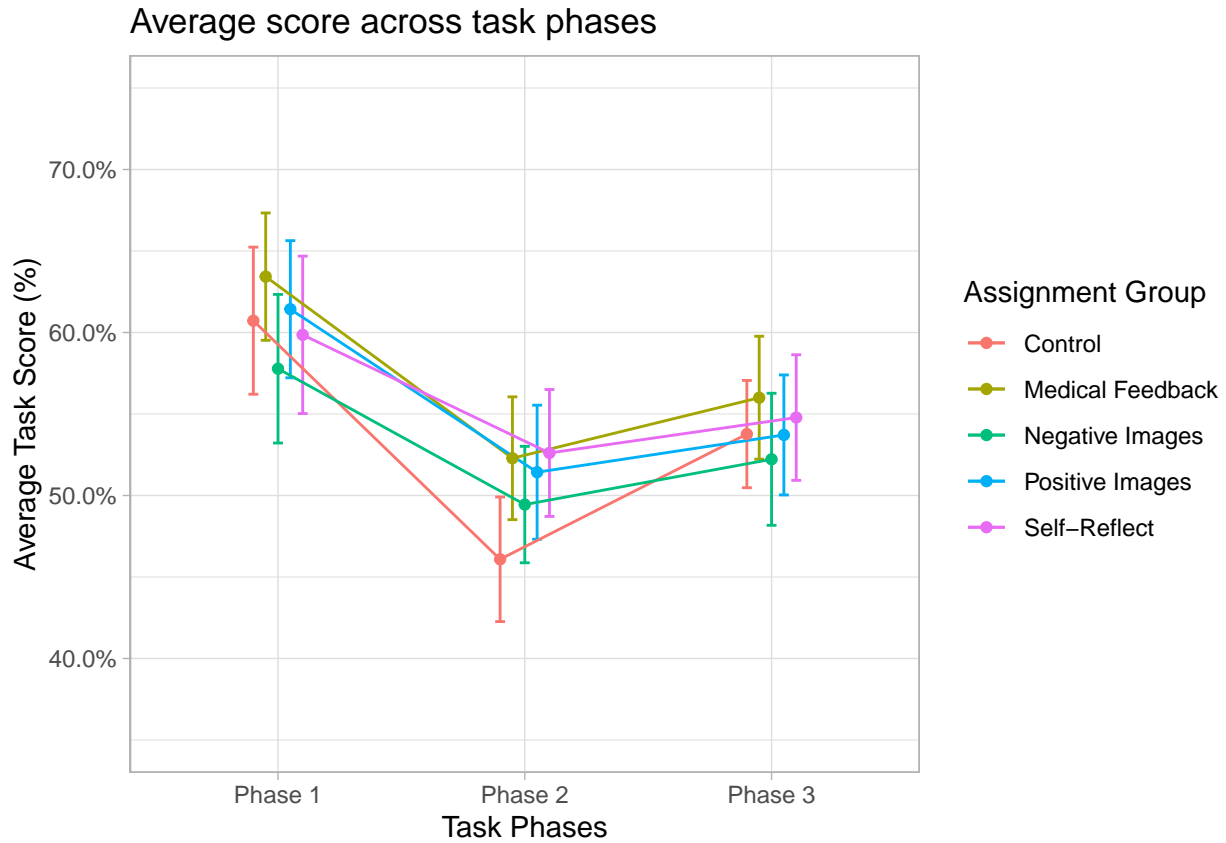
```

Compare task scores in different phases



```
# Compare score across time for all groups
# https://sphweb.bumc.bu.edu/otlt/MPH-Modules/BS/BS704_Confidence_Intervals/BS704_Confidence_Intervals_
# TODO finish formatting
# TODO duplicate for treatment dummy as well
summary_task_score <- (melt(d_respondents,id.vars=c('Assignment_Group'),
                           measure.vars = c('TaskPhase1_Score','TaskPhase2_Score','TaskPhase3_Score'))
  ,.('avg_score'=mean(value),'sd_score'=sd(value),'obs'=.N),keyby=.(Assignment_Group,variable))[,
    ,se:=1.96*sd_score/sqrt(obs)]

summary_task_score %>%
  ggplot( aes(x=variable, y=avg_score, group=Assignment_Group, color=Assignment_Group)) +
  geom_errorbar(aes(ymin=avg_score-1.96*sd_score/sqrt(obs), ymax=avg_score+1.96*sd_score/sqrt(obs)),
    width=.2,
    position=position_dodge(0.25)) +
  geom_line(position=position_dodge(0.25)) +
  geom_point(position=position_dodge(0.25)) +
  scale_y_continuous(labels = scales::percent,limits = c(.35,.75)) +
  scale_x_discrete(breaks=c("TaskPhase1_Score", "TaskPhase2_Score","TaskPhase3_Score"),
    labels=c("Phase 1", "Phase 2", "Phase 3")) +
  xlab('Task Phases') +
  ylab('Average Task Score (%)') +
  labs(title='Average score across task phases', color = "Assignment Group")
```



```
# TODO add this to the appendix
kable(summary_task_score)
```

Assignment_Group	variable	avg_score	sd_score	obs	se
Control	TaskPhase1_Score	0.6072	0.1912	69	0.0451
Control	TaskPhase2_Score	0.4609	0.1620	69	0.0382
Control	TaskPhase3_Score	0.5377	0.1394	69	0.0329
Medical Feedback	TaskPhase1_Score	0.6343	0.1667	70	0.0391
Medical Feedback	TaskPhase2_Score	0.5229	0.1608	70	0.0377
Medical Feedback	TaskPhase3_Score	0.5600	0.1610	70	0.0377
Negative Images	TaskPhase1_Score	0.5778	0.1973	72	0.0456
Negative Images	TaskPhase2_Score	0.4944	0.1546	72	0.0357
Negative Images	TaskPhase3_Score	0.5222	0.1754	72	0.0405
Positive Images	TaskPhase1_Score	0.6143	0.1796	70	0.0421
Positive Images	TaskPhase2_Score	0.5143	0.1755	70	0.0411
Positive Images	TaskPhase3_Score	0.5371	0.1571	70	0.0368
Self-Reflect	TaskPhase1_Score	0.5986	0.2047	69	0.0483
Self-Reflect	TaskPhase2_Score	0.5261	0.1651	69	0.0390
Self-Reflect	TaskPhase3_Score	0.5478	0.1632	69	0.0385

Gender

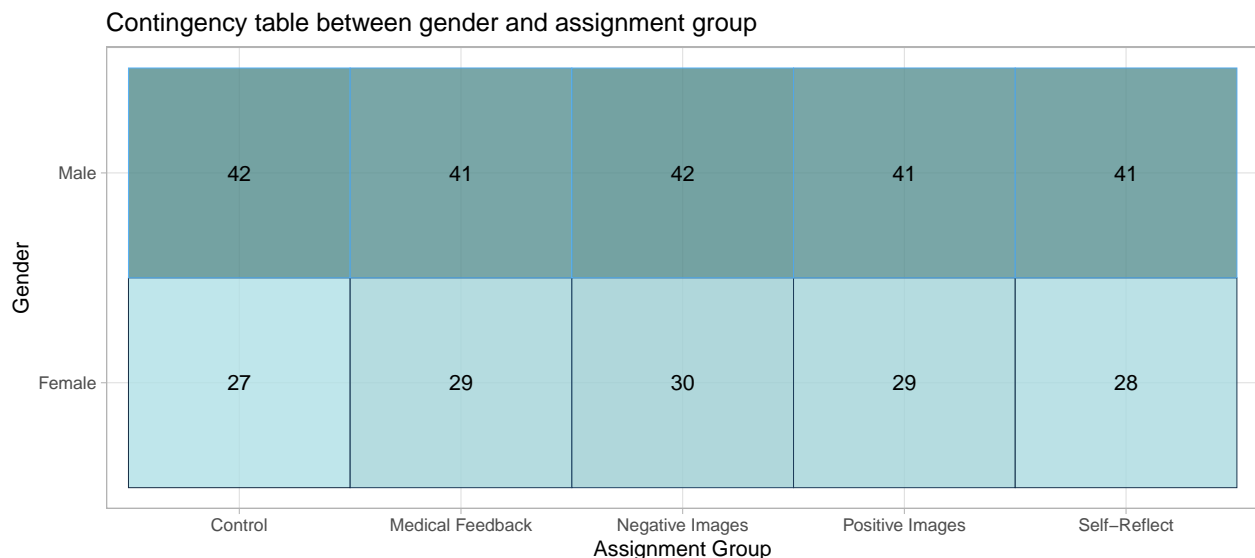
```
# TODO format figures and captions
#check balance between gender
```

```
gender_chiqq <- chisq.test(d_respondents[, table(Assignment_Group, Gender)])
```

```
pander(gender_chiq, style='rmarkdown')
```

Quitting from lines 570-586 (EDA.Rmd) Error in pander(gender_chiq, style = "rmarkdown") : object 'gender_chiq' not found Calls: ... withCallingHandlers -> withVisible -> eval -> eval -> pander

```
create_heatmap(var1 = d_respondents$Assignment_Group, var2 = d_respondents$Gender) +
  xlab('Assignment Group') +
  ylab('Gender') +
  labs(title = 'Contingency table between gender and assignment group',
       caption = paste0('Assuming gender distributions are the same among assignment groups, a chi-squared test for independence with 4
                           degrees of freedom yields p=', round(gender_chi$parameter, 4), '\ndegrees of freedom ', 'yields p=',
                           round(gender_chi$p.value, 4),
                           ', suggesting that there is no relationship between gender and assignment groups at a
                           significance level of 0.05.'))
  theme(plot.caption = element_text(hjust = 0))
```



Assuming gender distributions are the same among assignment groups, a chi-squared test for independence with 4 degrees of freedom yields $p=0.9981$, suggesting that there is no relationship between gender and assignment groups at a significance level of 0.05.

Age Range

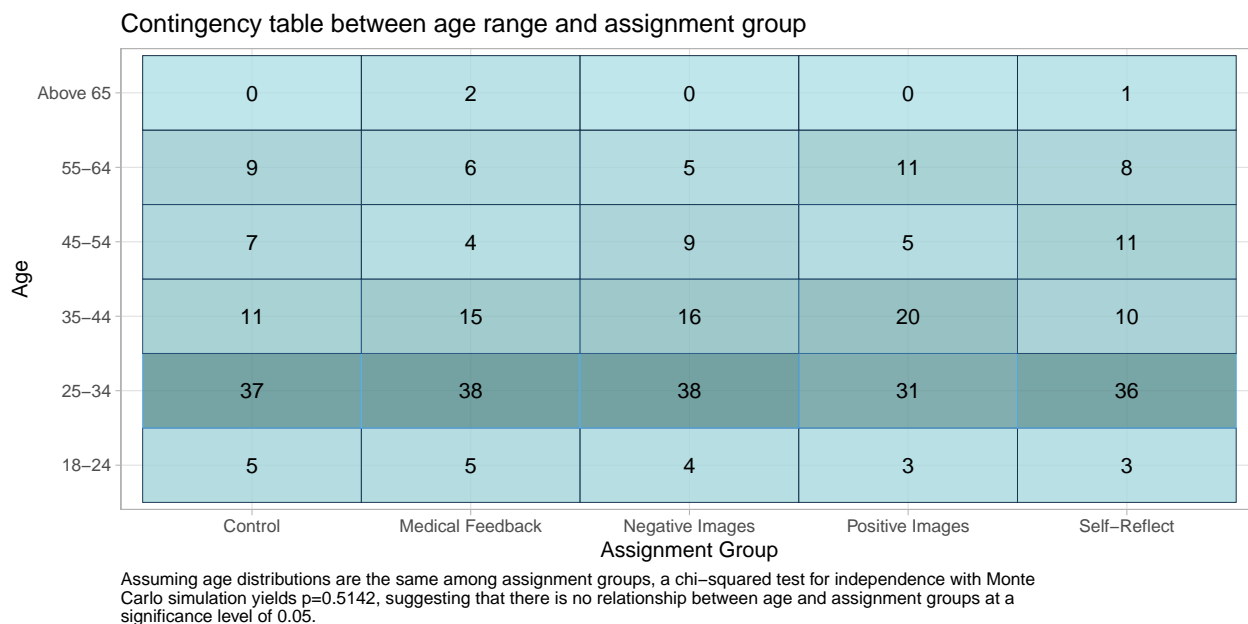
```
# TODO format figures and captions
#check balance between age-range

# expected frequency count for each cell of the contingency table should be at least 5. Since this is not the case, we use a simulation-based p-value.
# https://stats.stackexchange.com/questions/81483/warning-in-r-chi-squared-approximation-may-be-incorrect
age_chisq <- chisq.test(d_respondents[, table(Assignment_Group, Age_Range)], simulate.p.value = TRUE)
pander(age_chisq, style='rmarkdown')
```

Table 3: Pearson's Chi-squared test with simulated p-value (based on 2000 replicates): `d_respondents[, table(Assignment_Group, Age_Range)]`

Test statistic	df	P value
19.22	NA	0.5142


```
create_heatmap(var1 = d_respondents$Assignment_Group,var2 = d_respondents$Age_Range) +
  xlab('Assignment Group') +
  ylab('Age') +
  labs(title = 'Contingency table between age range and assignment group',
        caption = paste0('Assuming age distributions are the same among assignment groups, a chi-squared
                           round(age_chisq$p.value,4),
                           ', suggesting that there is no relationship between age and assignment groups a
  theme(plot.caption = element_text(hjust = 0))
```



Education Level

```
# TODO format figures and captions
#check balance between education levels
edu_chisq <- chisq.test(d_respondents[, table(Assignment_Group, Education_Level)],simulate.p.value = T)
pander(edu_chisq,style='rmarkdown')
```

Table 4: Pearson's Chi-squared test with simulated p-value (based on 2000 replicates): `d_respondents[, table(Assignment_Group, Education_Level)]`

Test statistic	df	P value
28.7	NA	0.07146

```
create_heatmap(var1 = d_respondents$Assignment_Group,var2 = d_respondents$Education_Level) +
  xlab('Assignment Group') +
  ylab('Education Level') +
  labs(title = 'Contingency table between education and assignment group',
        caption = paste0('Assuming education distributions are the same among assignment groups, a chi-s
                           round(edu_chisq$p.value,4),
                           ', suggesting that there is no relationship \nbetween education and assignment
  theme(plot.caption = element_text(hjust = 0))
```

Contingency table between education and assignment group

Education Level	Trade school	1	1	3	2	1
	Some high school	0	0	1	0	0
	Master's degree and above	20	14	13	19	11
	High school	1	1	3	0	7
	Bachelor's degree	44	54	50	45	46
	Associate's degree	3	0	2	4	4
		Control	Medical Feedback	Negative Images	Positive Images	Self-Reflect
		Assignment Group				

Assuming education distributions are the same among assignment groups, a chi-squared test for independence with Monte Carlo simulation yields $p=0.0715$, suggesting that there is no relationship between education and assignment groups at a significance level of 0.05.

Country: US, non-US

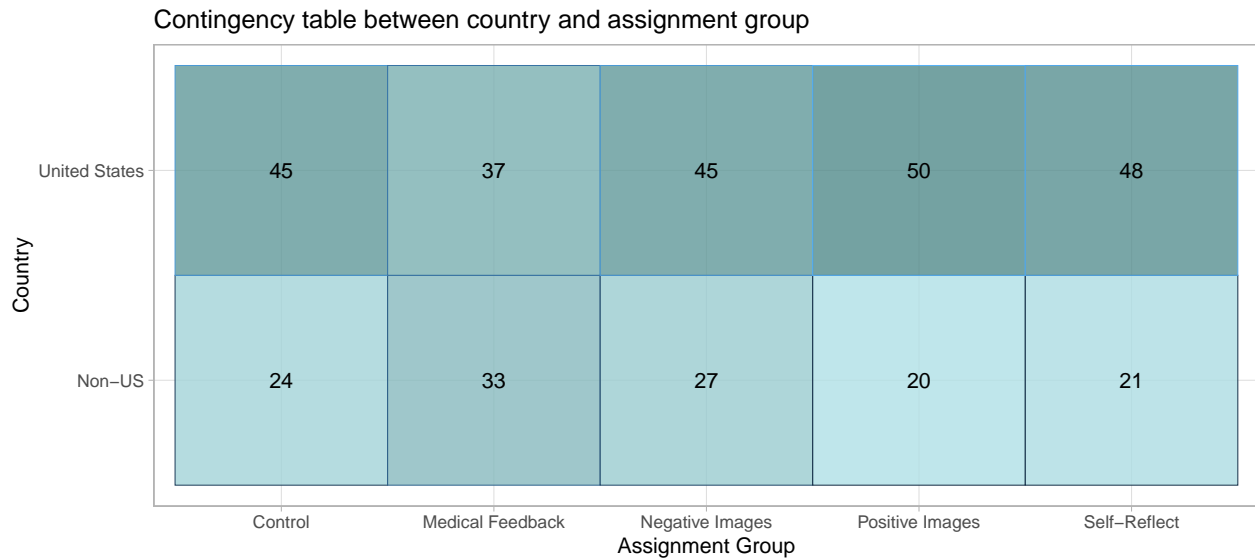
```
# TODO format figures and captions
# out.width = "80%"
# check balance between US and non-US respondents

us_chisq <- chisq.test(d_respondents[, table(Assignment_Group, US_Dummy)])
pander(us_chisq, style='rmarkdown')
```

Table 5: Pearson's Chi-squared test: `d_respondents[, table(Assignment_Group, US_Dummy)]`

Test statistic	df	P value
6.502	4	0.1647

```
create_heatmap(var1 = d_respondents$Assignment_Group, var2 = d_respondents$US_Dummy) +
  xlab('Assignment Group') +
  ylab('Country') +
  scale_y_discrete(breaks=c("0", "1"),
    labels=c("Non-US", "United States")) +
  labs(title = 'Contingency table between country and assignment group',
    caption = paste0('Assuming country distributions are the same among assignment groups, a chi-squared test with Monte Carlo simulation yields p=',
      round(us_chisq$parameter,4), ' degrees of freedom ', 'yields p=',
      round(us_chisq$p.value,4),
      ', suggesting that there is no relationship between country and assignment \n groups at a significance level of 0.05.'))
  theme(plot.caption = element_text(hjust = 0))
```



Assuming country distributions are the same among assignment groups, a chi-squared test for independence with 4 degrees of freedom yields $p=0.1647$, suggesting that there is no relationship between country and assignment groups at a significance level of 0.05.

```
# ATE of treatment on Total Score
```

```
d_respondents[ Treatment_Dummy == 1, mean(Total_Score)] - d_respondents[ Treatment_Dummy == 0, mean(Tot
```

```
## [1] 0.5143
```

```
sd(d_respondents$Total_Score)
```

```
## [1] 3.743
```

```
# ATE of treatment on TaskPhase2 Score
```

```
d_respondents[ Treatment_Dummy == 1, mean(TaskPhase2_Score)] - d_respondents[ Treatment_Dummy == 0, mean
```

```
## [1] 0.05337
```

```
sd(d_respondents$TaskPhase2_Score)
```

```
## [1] 0.1645
```

```
#trying 2SLS...but dont think it applies here
```

```
# d_respondents[, lm(Total_Score ~ Education_Level)]
```

```
# d_respondents[, ivreg(Total_Score ~ Education_Level | Assignment_Group)]
```

```
power.t.test( delta = .05, sd=.16, sig.level = 0.05, power=0.8)
```

```
##
```

```
## Two-sample t test power calculation
```

```
##
```

```
## n = 161.7
```

```
## delta = 0.05
```

```
## sd = 0.16
```

```
## sig.level = 0.05
```

```
## power = 0.8
```

```
## alternative = two.sided
```

```
##
```

```
## NOTE: n is number in *each* group
```

Analysis

Helper Functions

```
get_robust_se <- function(model){  
  # Get robust SE for use in stargazer  
  vcov <- vcovHC(model,type = "HC1")  
  return(sqrt(diag(vcov)))  
}
```

Task Phase 2 Analysis

```
# does any treatment have an effect on task phase 2 score?  
mod_task2_a <- d_respondents[, lm(TaskPhase2_Score ~ Treatment_Dummy)]  
  
mod_task2_b <- d_respondents[, lm(TaskPhase2_Score ~ Treatment_Dummy +  
  TaskPhase1_Score +  
  as.factor(Gender) +  
  as.factor(Education_Level) +  
  as.factor(Age_Range))]  
  
stargazer(mod_task2_a,  
  mod_task2_b,  
  se = list(get_robust_se(mod_task2_a),get_robust_se(mod_task2_b)),  
  omit = c("Education_Level","Age_Range"),  
  add.lines = list(c('Education Fixed Effects', 'No','Yes'),  
    c('Age Fixed Effects','No','Yes')),  
  header=FALSE,  
  type='latex')  
  
#add an F test to compare  
pander(anova(mod_task2_a, mod_task2_b, test='F'),style='rmarkdown')
```

Table 7: Analysis of Variance Table

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
348	9.287	NA	NA	NA	NA
336	8.338	12	0.9498	3.19	0.0002426

```
#does the specific treatment group have an effect on task phase 2 score?  
mod_task2_c <- d_respondents[, lm(TaskPhase2_Score ~ as.factor(Assignment_Group))]  
  
mod_task2_d <- d_respondents[, lm(TaskPhase2_Score ~ as.factor(Assignment_Group) +  
  TaskPhase1_Score +  
  as.factor(Gender) +  
  as.factor(Education_Level) +  
  as.factor(Age_Range))]  
  
# Do you think that there are features of the data that might systematically predict that people will r  
# TODO update this heterogeneity issue. I'm not quite sure this applies because they're both considered  
# mod5 <- d_respondents[, lm(TaskPhase2_Score ~ Treatment_Dummy + as.factor(assign_bin) +  
  # Treatment_Dummy * as.factor(assign_bin))]
```

Table 6:

	<i>Dependent variable:</i>	
	TaskPhase2_Score	
	(1)	(2)
Treatment_Dummy	0.053** (0.022)	0.051** (0.022)
TaskPhase1_Score		0.240*** (0.047)
as.factor(Gender)Male		-0.010 (0.017)
Constant	0.461*** (0.019)	0.281*** (0.072)
Education Fixed Effects	No	Yes
Age Fixed Effects	No	Yes
Observations	350	350
R ²	0.017	0.117
Adjusted R ²	0.014	0.083
Residual Std. Error	0.163 (df = 348)	0.158 (df = 336)
F Statistic	5.911** (df = 1; 348)	3.433*** (df = 13; 336)

Note:

*p<0.1; **p<0.05; ***p<0.01

```
stargazer(mod_task2_c,
          mod_task2_d,
          se = list(get_robust_se(mod_task2_c), get_robust_se(mod_task2_d)),
          omit = c("Education_Level", "Age_Range"),
          add.lines = list(c('Education Fixed Effects', 'No', 'Yes'),
                           c('Age Fixed Effects', 'No', 'Yes')),
          header=FALSE,
          type='latex')
```

Table 8:

	<i>Dependent variable:</i>	
	TaskPhase2_Score	
	(1)	(2)
as.factor(Assignment_Group)Medical Feedback	0.062** (0.027)	0.055* (0.029)
as.factor(Assignment_Group)Negative Images	0.034 (0.027)	0.039 (0.027)
as.factor(Assignment_Group)Positive Images	0.053* (0.029)	0.050* (0.027)
as.factor(Assignment_Group)Self-Reflect	0.065** (0.028)	0.058** (0.029)
TaskPhase1_Score		0.238*** (0.048)
as.factor(Gender)Male		-0.010 (0.017)
Constant	0.461*** (0.019)	0.282*** (0.073)
Education Fixed Effects	No	Yes
Age Fixed Effects	No	Yes
Observations	350	350
R ²	0.021	0.119
Adjusted R ²	0.010	0.076
Residual Std. Error	0.164 (df = 345)	0.158 (df = 333)
F Statistic	1.874 (df = 4; 345)	2.805*** (df = 16; 333)

Note:

*p<0.1; **p<0.05; ***p<0.01

```
pander(anova(mod_task2_c, mod_task2_d, test='F'), style='rmarkdown')
```

Table 9: Analysis of Variance Table

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
345	9.244	NA	NA	NA	NA

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
333	8.323	12	0.921	3.071	0.0003943

Task Phase 3 Analysis

```
# test final task and any treatment
mod_task3_a <- d_respondents[, lm(TaskPhase3_Score ~ Treatment_Dummy)]
mod_task3_b <- d_respondents[, lm(TaskPhase3_Score ~ Treatment_Dummy +
                                TaskPhase1_Score +
                                as.factor(Gender) +
                                as.factor(Education_Level) +
                                as.factor(Age_Range))]

stargazer(mod_task3_a,
           mod_task3_b,
           se = list(get_robust_se(mod_task3_a), get_robust_se(mod_task3_b)),
           omit = c("Education_Level", "Age_Range"),
           add.lines = list(c('Education Fixed Effects', 'No', 'Yes'),
                           c('Age Fixed Effects', 'No', 'Yes')),
           header=FALSE,
           type='latex')

pander(anova(mod_task3_a, mod_task3_b, test='F'), style='rmarkdown')
```

Table 11: Analysis of Variance Table

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
348	8.865	NA	NA	NA	NA
336	8.117	12	0.7479	2.58	0.002743

```
# test final task and specific treatment
mod_task3_c <- d_respondents[, lm(TaskPhase3_Score ~ as.factor(Assignment_Group))]
mod_task3_d <- d_respondents[, lm(TaskPhase3_Score ~ as.factor(Assignment_Group) +
                                TaskPhase1_Score +
                                as.factor(Gender) +
                                as.factor(Education_Level) +
                                as.factor(Age_Range))]

stargazer(mod_task3_c,
           mod_task3_d,
           se = list(get_robust_se(mod_task3_c), get_robust_se(mod_task3_d)),
           omit = c("Education_Level", "Age_Range"),
           add.lines = list(c('Education Fixed Effects', 'No', 'Yes'),
                           c('Age Fixed Effects', 'No', 'Yes')),
           header=FALSE,
           type='latex')

pander(anova(mod_task3_c, mod_task3_d, test='F'), style='rmarkdown')
```

Table 10:

	<i>Dependent variable:</i>	
	TaskPhase3_Score	
	(1)	(2)
Treatment_Dummy	0.004 (0.019)	0.002 (0.019)
TaskPhase1_Score		0.161*** (0.047)
as.factor(Gender)Male		-0.004 (0.017)
Constant	0.538*** (0.017)	0.515*** (0.064)
Education Fixed Effects	No	Yes
Age Fixed Effects	No	Yes
Observations	350	350
R ²	0.0001	0.084
Adjusted R ²	-0.003	0.049
Residual Std. Error	0.160 (df = 348)	0.155 (df = 336)
F Statistic	0.034 (df = 1; 348)	2.384*** (df = 13; 336)
<i>Note:</i> *p<0.1; **p<0.05; ***p<0.01		

Table 12:

	<i>Dependent variable:</i>	
	TaskPhase3_Score	
	(1)	(2)
as.factor(Assignment_Group)Medical Feedback	0.022 (0.026)	0.011 (0.026)
as.factor(Assignment_Group)Negative Images	-0.015 (0.027)	-0.011 (0.026)
as.factor(Assignment_Group)Positive Images	-0.001 (0.025)	0.004 (0.025)
as.factor(Assignment_Group)Self-Reflect	0.010 (0.026)	0.005 (0.026)
TaskPhase1_Score		0.157*** (0.047)
as.factor(Gender)Male		-0.004 (0.017)
Constant	0.538*** (0.017)	0.518*** (0.064)
Education Fixed Effects	No	Yes
Age Fixed Effects	No	Yes
Observations	350	350
R ²	0.006	0.087
Adjusted R ²	-0.005	0.043
Residual Std. Error	0.160 (df = 345)	0.156 (df = 333)
F Statistic	0.545 (df = 4; 345)	1.971** (df = 16; 333)
<i>Note:</i> *p<0.1; **p<0.05; ***p<0.01		

Table 13: Analysis of Variance Table

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
345	8.81	NA	NA	NA	NA
333	8.099	12	0.7113	2.437	0.004751

Wearing Off Effects

```

mod_task3_e <- d_respondents[ , lm(TaskPhase3_Score ~ TaskPhase2_Score)]
mod_task3_f <- d_respondents[ , lm(TaskPhase3_Score ~ TaskPhase2_Score + Treatment_Dummy)]
mod_task3_g <- d_respondents[ , lm(TaskPhase3_Score ~ TaskPhase2_Score + as.factor(Assignment_Group))]
mod_task3_h <- d_respondents[ , lm(TaskPhase3_Score ~ TaskPhase2_Score +
                                   as.factor(Assignment_Group) +
                                   as.factor(Gender) +
                                   as.factor(Education_Level) +
                                   as.factor(Age_Range))]

stargazer(mod_task3_e,
           mod_task3_f,
           mod_task3_g,
           mod_task3_h,
           se = list(get_robust_se(mod_task3_e),
                     get_robust_se(mod_task3_f),
                     get_robust_se(mod_task3_h)),
           omit = c("Education_Level", "Age_Range"),
           add.lines = list(c('Education Fixed Effects', 'No', 'No', 'No', 'Yes'),
                             c('Age Fixed Effects', 'No', 'No', 'No', 'Yes')),
           header=FALSE,
           type='latex')

pander(anova(mod_task3_e, mod_task3_f, test='F'), style='rmarkdown')

```

Table 16: Analysis of Variance Table

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
348	8.326	NA	NA	NA	NA
347	8.322	1	0.004358	0.1817	0.6702

```

pander(anova(mod_task3_g, mod_task3_h, test='F'), style='rmarkdown')

```

Table 17: Analysis of Variance Table

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
344	8.286	NA	NA	NA	NA
333	7.863	11	0.4227	1.627	0.08955

```

##
mod_task3_i <- d_respondents[ , lm(TaskPhase3_Score ~ TaskPhase1_Score + TaskPhase2_Score)]
mod_task3_j <- d_respondents[ , lm(TaskPhase3_Score ~ TaskPhase1_Score + TaskPhase2_Score + Treatment_D

```

Table 14:

	<i>Dependent variable:</i>		
	TaskPhase3_Score		
	(1)	(2)	(3)
TaskPhase2_Score	0.239*** (0.050)	0.242*** (0.051)	0.238*** (0.052)
Treatment_Dummy		-0.009 (0.019)	
as.factor(Assignment_Group)Medical Feedback			0.008 (0.027)
as.factor(Assignment_Group)Negative Images			-0.023 (0.026)
as.factor(Assignment_Group)Positive Images			-0.013 (0.025)
as.factor(Assignment_Group)Self-Reflect			-0.005 (0.025)
as.factor(Gender)Male			
Constant	0.420*** (0.026)	0.426*** (0.028)	0.428*** (0.062)
Education Fixed Effects	No	No	No
Age Fixed Effects	No	No	No
Observations	350	350	350
R ²	0.061	0.061	0.065
Adjusted R ²	0.058	0.056	0.052
Residual Std. Error	0.155 (df = 348)	0.155 (df = 347)	0.155 (df = 344)
F Statistic	22.540*** (df = 1; 348)	11.330*** (df = 2; 347)	4.815*** (df = 5; 344)

Note:

Table 15:

(Intercept)	TaskPhase2_Score	as.factor(Assignment_Group)Medical Feedback	as.factor(Assignment_Group)Negative Images
0.028	0.051	0.026	0.026

```

mod_task3_k <- d_respondents[, lm(TaskPhase3_Score ~ TaskPhase1_Score + TaskPhase2_Score + as.factor(A
mod_task3_l <- d_respondents[, lm(TaskPhase3_Score ~ TaskPhase1_Score + TaskPhase2_Score +
                                as.factor(Assignment_Group) +
                                as.factor(Gender) +
                                as.factor(Education_Level) +
                                as.factor(Age_Range))]]

stargazer(mod_task3_i,
           mod_task3_j,
           mod_task3_k,
           mod_task3_l,
           se = list(get_robust_se(mod_task3_i),
                     get_robust_se(mod_task3_j),
                     get_robust_se(mod_task3_k),
                     get_robust_se(mod_task3_l)),
           header=FALSE,
           type='latex')

pander(anova(mod_task3_i, mod_task3_j, test = 'F'),style='rmarkdown')

```

Table 19: Analysis of Variance Table

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
347	8.181	NA	NA	NA	NA
346	8.178	1	0.002531	0.1071	0.7437

```

pander(anova(mod_task3_k, mod_task3_l, test = 'F'),style='rmarkdown')

```

Table 20: Analysis of Variance Table

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
343	8.151	NA	NA	NA	NA
332	7.739	11	0.4128	1.61	0.09435

```

# lm(TaskPhase3_Score ~ TaskPhase2_Score) vs lm(TaskPhase3_Score ~ TaskPhase1_Score + TaskPhase2_Score)
pander(anova(mod_task3_e, mod_task3_i, test = 'F'),style='rmarkdown')

```

Table 21: Analysis of Variance Table

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
348	8.326	NA	NA	NA	NA
347	8.181	1	0.1455	6.172	0.01345

Playground

```

# compare self-reflect against medical feedback groups?
#make dummies
d_respondents[, Self_Reflect_Dummy := ifelse(Assignment_Group == "Self-Reflect", 1, 0)]
d_respondents[, Med_Feedback_Dummy := ifelse(Assignment_Group == "Medical Feedback", 1, 0)]

```

Table 18:

	<i>Dependent variable:</i>		
	TaskPhase3_Score		
	(1)	(2)	(3)
TaskPhase1_Score	0.113** (0.046)	0.113** (0.046)	0.109* (0.046)
TaskPhase2_Score	0.202*** (0.053)	0.204*** (0.054)	0.202* (0.053)
Treatment_Dummy		-0.007 (0.019)	
as.factor(Assignment_Group)Medical Feedback			0.007 (0.026)
as.factor(Assignment_Group)Negative Images			-0.01 (0.026)
as.factor(Assignment_Group)Positive Images			-0.01 (0.024)
as.factor(Assignment_Group)Self-Reflect			-0.00 (0.024)
as.factor(Gender)Male			
as.factor(Education_Level)Bachelor's degree			
as.factor(Education_Level)High school			
as.factor(Education_Level)Master's degree and above			
as.factor(Education_Level)Some high school			
as.factor(Education_Level)Trade school			
as.factor(Age_Range)25-34			
as.factor(Age_Range)35-44			
as.factor(Age_Range)45-54			
as.factor(Age_Range)55-64			
as.factor(Age_Range)Above 65			

```

mod_test_dummies1 <- d_respondents[ , lm(TaskPhase2_Score ~ Treatment_Dummy + Self_Reflect_Dummy)]
mod_test_dummies2 <- d_respondents[ , lm(TaskPhase2_Score ~ Treatment_Dummy + Med_Feedback_Dummy)]

stargazer(mod_test_dummies1,
  mod_test_dummies2,
  se = list(get_robust_se(mod_test_dummies1),
    get_robust_se(mod_test_dummies2)),
  header=FALSE,
  type = 'latex')

##
## \begin{table}[!htbp] \centering
##   \caption{}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lcc}
## \ll[-1.8ex]\hline
## \hline \ll[-1.8ex]
## & \multicolumn{2}{c}{\textit{Dependent variable:}} \ll
## \cline{2-3}
## \ll[-1.8ex] & \multicolumn{2}{c}{TaskPhase2\_Score} \ll
## \ll[-1.8ex] & (1) & (2)\ll
## \hline \ll[-1.8ex]
## Treatment\_Dummy & 0.050$^{**}$ & 0.051$^{**}$ \ll
## & (0.022) & (0.023) \ll
## & & \ll
## Self\_Reflect\_Dummy & 0.016 & \ll
## & (0.023) & \ll
## & & \ll
## Med\_Feedback\_Dummy & & 0.011 \ll
## & & (0.022) \ll
## & & \ll
## Constant & 0.461$^{***}$ & 0.461$^{***}$ \ll
## & (0.019) & (0.019) \ll
## & & \ll
## \hline \ll[-1.8ex]
## Observations & 350 & 350 \ll
## R$^{2}$ & 0.018 & 0.017 \ll
## Adjusted R$^{2}$ & 0.012 & 0.012 \ll
## Residual Std. Error (df = 347) & 0.163 & 0.164 \ll
## F Statistic (df = 2; 347) & 3.192$^{**}$ & 3.079$^{**}$ \ll
## \hline
## \hline \ll[-1.8ex]
## \textit{Note:} & \multicolumn{2}{r}{\textit{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01}} \ll
## \end{tabular}
## \end{table}

```

Playground 2

```

### linear model playground
d_test <- d_respondents[,c("Assignment_Group", "TaskPhase1_Score", "TaskPhase2_Score", "TaskPhase3_Score",

#does treatment have an effect on total score?
mod_test1 <- d_test[ , lm(TaskPhase2_Score ~ TaskPhase1_Score + Treatment_Dummy)]

```

```

mod_test2 <- d_test[, lm(TaskPhase2_Score ~ TaskPhase1_Score + Treatment_Dummy + (TaskPhase1_Score * Tr

#does treatment and pretreatment score have an effect on total score?

###
# seems that if i ad in TaskPhase1 to the linear model, the RSEs disappear...
mod_test3 <- d_test[, lm(TaskPhase2_Score ~ Treatment_Dummy +
                        TaskPhase1_Score +
                        as.factor(Education_Level) +
                        as.factor(Gender) +
                        as.factor(Age_Range)
                        )]

coeftest(mod_test3, vcov = vcovHC(mod_test3,"HC1"))

```

```

##
## t test of coefficients:
##
##
## Estimate Std. Error t value
## (Intercept) 0.28108 0.07213 3.90
## Treatment_Dummy 0.05071 0.02217 2.29
## TaskPhase1_Score 0.24027 0.04682 5.13
## as.factor(Education_Level)Bachelor's degree -0.00683 0.04856 -0.14
## as.factor(Education_Level)High school 0.04068 0.05619 0.72
## as.factor(Education_Level)Master's degree and above -0.01698 0.05128 -0.33
## as.factor(Education_Level)Some high school -0.12065 0.05108 -2.36
## as.factor(Education_Level)Trade school 0.02867 0.06926 0.41
## as.factor(Gender)Male -0.00995 0.01735 -0.57
## as.factor(Age_Range)25-34 0.04469 0.03768 1.19
## as.factor(Age_Range)35-44 0.04198 0.03952 1.06
## as.factor(Age_Range)45-54 0.06975 0.04178 1.67
## as.factor(Age_Range)55-64 0.08035 0.04252 1.89
## as.factor(Age_Range)Above 65 0.12575 0.05172 2.43
## Pr(>|t|)
## (Intercept) 0.00012 ***
## Treatment_Dummy 0.02281 *
## TaskPhase1_Score 4.9e-07 ***
## as.factor(Education_Level)Bachelor's degree 0.88827
## as.factor(Education_Level)High school 0.46957
## as.factor(Education_Level)Master's degree and above 0.74080
## as.factor(Education_Level)Some high school 0.01874 *
## as.factor(Education_Level)Trade school 0.67922
## as.factor(Gender)Male 0.56658
## as.factor(Age_Range)25-34 0.23642
## as.factor(Age_Range)35-44 0.28890
## as.factor(Age_Range)45-54 0.09593 .
## as.factor(Age_Range)55-64 0.05968 .
## as.factor(Age_Range)Above 65 0.01556 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	0.200	0.500	0.600	0.606	0.700	1.000

```
## % Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu
## % Date and time: Sun, Nov 29, 2020 - 19:05:13
## \begin{table}[\!htbp] \centering
##   \caption{}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \hline
## \hline \hline
## & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \hline & \multicolumn{3}{c}{TaskPhase2\_Score} \\
## \hline & (1) & (2) & (3) \\
## \hline
## TaskPhase1\_Score & 0.249$^{***}$ & 0.153 & 0.240$^{***}$ \\
## & (0.044) & (0.095) & (0.047) \\
## & & & \\
## as.factor(Education\_Level)Bachelor's degree & & & $-0.007$ \\
## & & & (0.049) \\
## & & & \\
## as.factor(Education\_Level)High school & & & 0.041 \\
## & & & (0.056) \\
## & & & \\
## as.factor(Education\_Level)Master's degree and above & & & $-0.017$ \\
## & & & (0.051) \\
## & & & \\
## as.factor(Education\_Level)Some high school & & & $-0.121$^{**}$ \\
## & & & (0.051) \\
## & & & \\
## as.factor(Education\_Level)Trade school & & & 0.029 \\
## & & & (0.069) \\
## & & & \\
## as.factor(Gender)Male & & & $-0.010$ \\
## & & & (0.017) \\
## & & & \\
## as.factor(Age\_Range)25-34 & & & 0.045 \\
## & & & (0.038) \\
## & & & \\
## as.factor(Age\_Range)35-44 & & & 0.042 \\
## & & & (0.040) \\
## & & & \\
## as.factor(Age\_Range)45-54 & & & 0.070$^{*}$ \\
## & & & (0.042) \\
## & & & \end{table}
```



```

## & & & \\
## as.factor(Age\_Range)55-64 & & & 0.080$^{*}$ \\
## & & & (0.043) \\
## & & & \\
## as.factor(Age\_Range)Above 65 & & & 0.126$^{**}$ \\
## & & & (0.052) \\
## & & & \\
## Treatment\_Dummy & 0.054$^{**}$ & $-0.019 & 0.051$^{**}$ \\
## & (0.021) & (0.065) & (0.022) \\
## & & & \\
## TaskPhase1\_Score:Treatment\_Dummy & & 0.120 & \\
## & & (0.107) & \\
## & & & \\
## Constant & 0.310$^{***}$ & 0.368$^{***}$ & 0.281$^{***}$ \\
## & (0.032) & (0.057) & (0.072) \\
## & & & \\
## \hline \\[-1.8ex]
## Observations & 350 & 350 & 350 \\
## R$^{2}$ & 0.098 & 0.101 & 0.117 \\
## Adjusted R$^{2}$ & 0.092 & 0.093 & 0.083 \\
## Residual Std. Error & 0.157 (df = 347) & 0.157 (df = 346) & 0.158 (df = 336) \\
## F Statistic & 18.780$^{***}$ (df = 2; 347) & 12.920$^{***}$ (df = 3; 346) & 3.433$^{***}$ (df = 13; 346) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:} & \multicolumn{3}{r}{ $^{*}$p<$0.1; $^{**}$p<$0.05; $^{***}$p<$0.01} \\
## \end{tabular}
## \end{table}

mod_test4 <- d_test[ , lm(TaskPhase3_Score ~ TaskPhase2_Score)]
coeftest(mod_test4)

##
## t test of coefficients:
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.4205     0.0267   15.77 <2e-16 ***
## TaskPhase2_Score  0.2389     0.0503    4.75  3e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# use Robust SE
mod_test2 <- d_respondents[, lm(TaskPhase2_Score ~ Treatment_Dummy + as.factor(Education_Level) + (Treatment_Dummy:as.factor(Education_Level)))]
mod_test2$vcovHC_ <- vcovHC(mod_test2)
coeftest(mod_test2, vcov = mod_test2$vcovHC_)

##
## t test of coefficients:
##
##               Estimate
## (Intercept)      0.53333
## Treatment_Dummy    0.00667
## as.factor(Education_Level)Bachelor's degree -0.07424
## as.factor(Education_Level)High school -0.03333
## as.factor(Education_Level)Master's degree and above -0.07333
## as.factor(Education_Level)Some high school -0.14000
## as.factor(Education_Level)Trade school -0.23333

```

```

## Treatment_Dummy:as.factor(Education_Level)Bachelor's degree      0.04142
## Treatment_Dummy:as.factor(Education_Level)High school            0.07515
## Treatment_Dummy:as.factor(Education_Level)Master's degree and above 0.04386
## Treatment_Dummy:as.factor(Education_Level)Trade school           0.30762
##                                                                    Std. Error
## (Intercept)                                                        NA
## Treatment_Dummy                                                    NA
## as.factor(Education_Level)Bachelor's degree                      NA
## as.factor(Education_Level)High school                            NA
## as.factor(Education_Level)Master's degree and above              NA
## as.factor(Education_Level)Some high school                      NA
## as.factor(Education_Level)Trade school                          NA
## Treatment_Dummy:as.factor(Education_Level)Bachelor's degree      NA
## Treatment_Dummy:as.factor(Education_Level)High school            NA
## Treatment_Dummy:as.factor(Education_Level)Master's degree and above NA
## Treatment_Dummy:as.factor(Education_Level)Trade school           NA
##                                                                    t value
## (Intercept)                                                        NA
## Treatment_Dummy                                                    NA
## as.factor(Education_Level)Bachelor's degree                      NA
## as.factor(Education_Level)High school                            NA
## as.factor(Education_Level)Master's degree and above              NA
## as.factor(Education_Level)Some high school                      NA
## as.factor(Education_Level)Trade school                          NA
## Treatment_Dummy:as.factor(Education_Level)Bachelor's degree      NA
## Treatment_Dummy:as.factor(Education_Level)High school            NA
## Treatment_Dummy:as.factor(Education_Level)Master's degree and above NA
## Treatment_Dummy:as.factor(Education_Level)Trade school           NA
##                                                                    Pr(>|t|)
## (Intercept)                                                        NA
## Treatment_Dummy                                                    NA
## as.factor(Education_Level)Bachelor's degree                      NA
## as.factor(Education_Level)High school                            NA
## as.factor(Education_Level)Master's degree and above              NA
## as.factor(Education_Level)Some high school                      NA
## as.factor(Education_Level)Trade school                          NA
## Treatment_Dummy:as.factor(Education_Level)Bachelor's degree      NA
## Treatment_Dummy:as.factor(Education_Level)High school            NA
## Treatment_Dummy:as.factor(Education_Level)Master's degree and above NA
## Treatment_Dummy:as.factor(Education_Level)Trade school           NA

```