

# ImpactDataViz 241 Final Project

```
# load packages
library(data.table)
library(foreign)
library(sandwich)
library(stargazer)

##
## Please cite as:

## Hlavac, Marek (2018). stargazer: Well-Formatted Regression and Summary Statistics Tables.

## R package version 5.2.2. https://CRAN.R-project.org/package=stargazer

library(lmtest)

## Loading required package: zoo

## Warning: package 'zoo' was built under R version 3.6.2

##
## Attaching package: 'zoo'

## The following objects are masked from 'package:base':
##
##      as.Date, as.Date.numeric
```

## 1. Parse the survey data into a data.table

```
d <- fread("data/final.csv")
#d <- fread("data/pilot.csv")

# covariates
setnames(d, 'Q3', 'state')
setnames(d, 'Q4', 'gender')
setnames(d, 'Q5', 'age')
setnames(d, 'Q6', 'ethnicity_multi')
setnames(d, 'Q7', 'political_party')
setnames(d, 'Q8', 'education')
setnames(d, 'Q9', 'covid_sick')
setnames(d, 'Q10', 'covid_hospitalized')
setnames(d, 'Q11', 'covid_died')
```

```

# duration of survey time
setnames(d, 'Duration (in seconds)', 'duration_of_survey')

# which block was active? (did the user see the treatment or
# control data viz)
setnames(d, 'Q15', 'treatment_viz_is_accurate')
setnames(d, 'Q17', 'control_viz_is_accurate')

# outcome questions about COVID attitudes
setnames(d, 'Q18', 'outcome_spread')
setnames(d, 'Q19', 'outcome_death')

# which block was active determines if
# subject received treatment data viz or control
# data viz
d[, treatment := ifelse(is.na(treatment_viz_is_accurate), 0, 1)]
d = d[!is.na(outcome_spread) & !is.na(outcome_death),]

# ethnicity allows for multiple choice
# for covariates, just grab the first one
ethnicity_single = rep(0,nrow(d))
i = 1
for (eth_entry in d[,ethnicity_multi]) {
  eth_tokens = unlist(strsplit(eth_entry, ","))
  ethnicity_single[i] = as.numeric(eth_tokens[1])
  i = i + 1
}
d[, ethnicity := ethnicity_single ]

# counts in control vs treatment
n_control = nrow(d[treatment == 0, ])
n_treatment = nrow(d[treatment == 1, ])

# Multiple plot function
#
# ggplot objects can be passed in ..., or to plotlist (as a list of ggplot objects)
# - cols:   Number of columns in layout
# - layout: A matrix specifying the layout. If present, 'cols' is ignored.
#
# If the layout is something like matrix(c(1,2,3,3), nrow=2, byrow=TRUE),
# then plot 1 will go in the upper left, 2 will go in the upper right, and
# 3 will go all the way across the bottom.
#
multiplot <- function(..., plotlist=NULL, file, cols=1, layout=NULL) {
  library(grid)

  # Make a list from the ... arguments and plotlist
  plots <- c(list(...), plotlist)

  numPlots = length(plots)

  # If layout is NULL, then use 'cols' to determine layout
  if (is.null(layout)) {

```

```

# Make the panel
# ncol: Number of columns of plots
# nrow: Number of rows needed, calculated from # of cols
layout <- matrix(seq(1, cols * ceiling(numPlots/cols)),
                 ncol = cols, nrow = ceiling(numPlots/cols))
}

if (numPlots==1) {
  print(plots[[1]])
} else {
  # Set up the page
  grid.newpage()
  pushViewport(viewport(layout = grid.layout(nrow(layout), ncol(layout))))

  # Make each plot, in the correct location
  for (i in 1:numPlots) {
    # Get the i,j matrix positions of the regions that contain this subplot
    matchidx <- as.data.frame(which(layout == i, arr.ind = TRUE))

    print(plots[[i]], vp = viewport(layout.pos.row = matchidx$row,
                                     layout.pos.col = matchidx$col))
  }
}
}

```

## 2. EDA

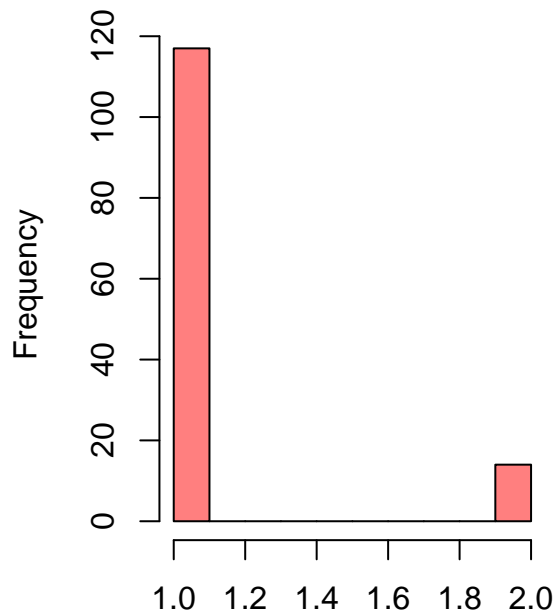
2.1 Are the data visualizations in control and treatment considered accurate by subjects?

```

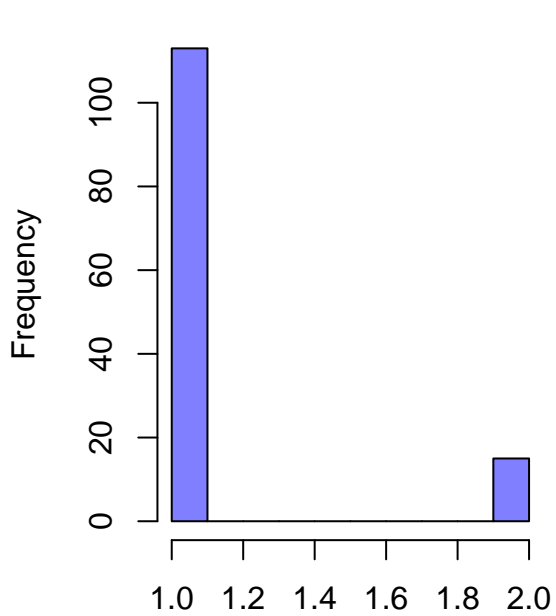
par(mfrow=c(1,2))
hist(d[, treatment_viz_is_accurate], col=rgb(1,0,0,0.5), xlim=c(1,2))
hist(d[, control_viz_is_accurate], col=rgb(0,0,1,0.5), xlim=c(1,2))
mtext("I trust that the information presented in the data visualization is accurate", side=1, outer=TRUE)

```

histogram of d[, treatment\_viz\_is\_accurate]      histogram of d[, control\_viz\_is\_accurate]



d[, treatment\_viz\_is\_accurate]



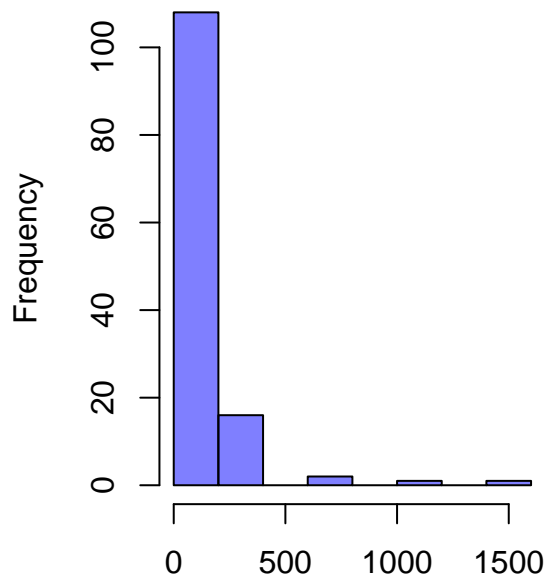
d[, control\_viz\_is\_accurate]

I trust that the information presented in the data visualization is accurate

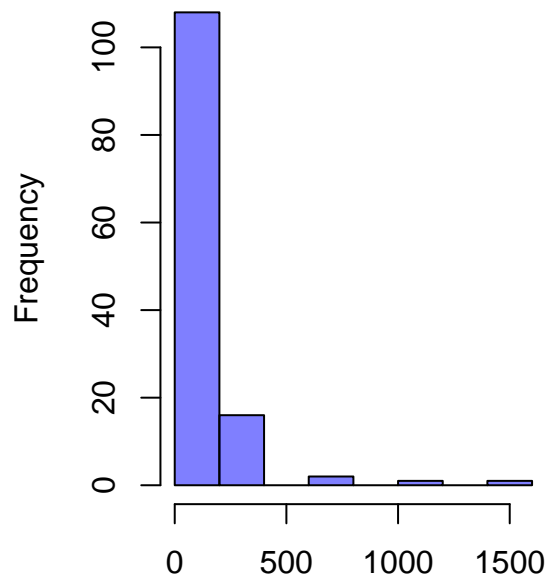
## 2.2 Check Time

```
par(mfrow=c(1,2))
hist(d[treatment == 0,duration_of_survey], col=rgb(0,0,1,0.5))
hist(d[treatment == 1,duration_of_survey], col=rgb(0,0,1,0.5))
```

gram of d[treatment == 0, duration\_of\_survey]



d[treatment == 0, duration\_of\_survey]



d[treatment == 0, duration\_of\_survey]

```
mean(d[treatment == 0, duration_of_survey])
```

```
## [1] 153.1875
```

```
mean(d[treatment == 0, duration_of_survey])
```

```
## [1] 153.1875
```

### 3. Covariate balance

#### 3.1 Compare distributions using violin plote

```
library('ggplot2')

options(repr.plot.width = 14, repr.plot.height = 8)

p1 = ggplot(d, aes(x=as.factor(treatment), y=gender, fill=as.factor(treatment))) +
  geom_violin()

p2 = ggplot(d, aes(x=as.factor(treatment), y=age, fill=as.factor(treatment))) +
  geom_violin()
```

```

p3 = ggplot(d, aes(x=as.factor(treatment), y=as.numeric(ethnicity), fill=as.factor(treatment))) +
  geom_violin()

p4 = ggplot(d, aes(x=as.factor(treatment), y=political_party, fill=as.factor(treatment))) +
  geom_violin()

p5 = ggplot(d, aes(x=as.factor(treatment), y=education, fill=as.factor(treatment))) +
  geom_violin()

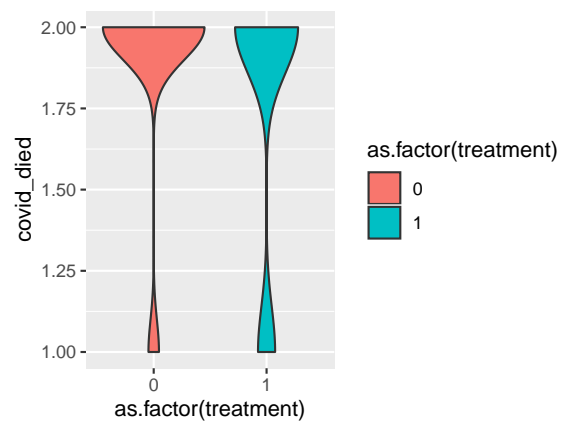
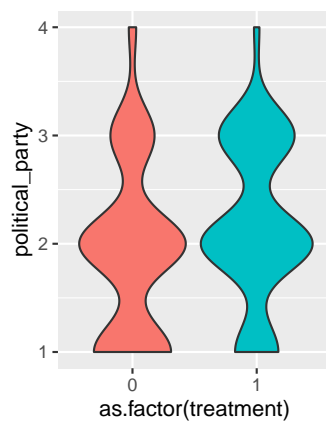
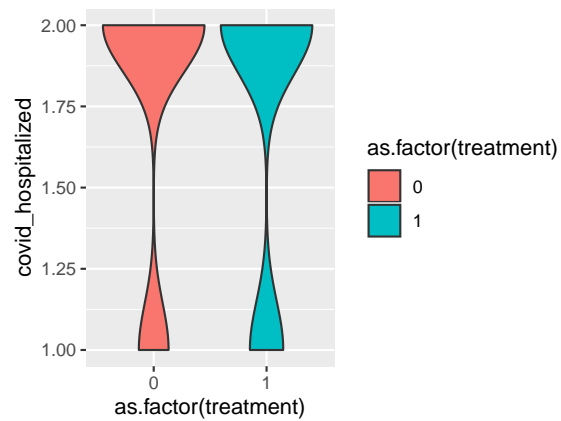
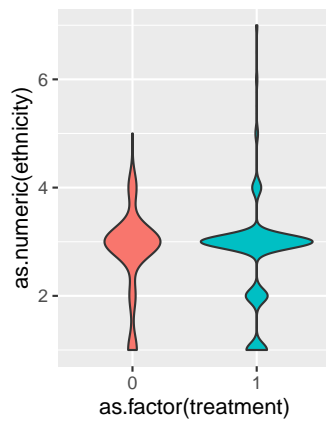
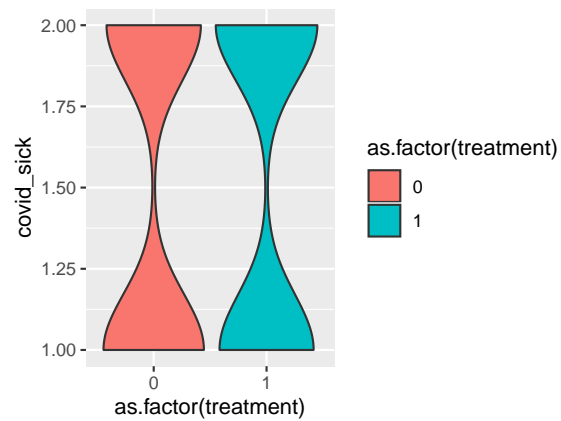
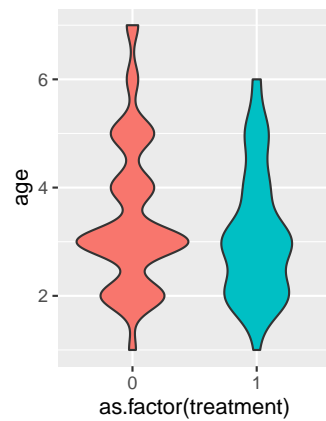
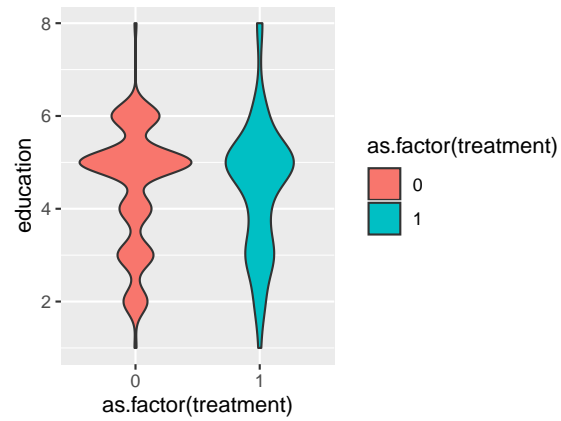
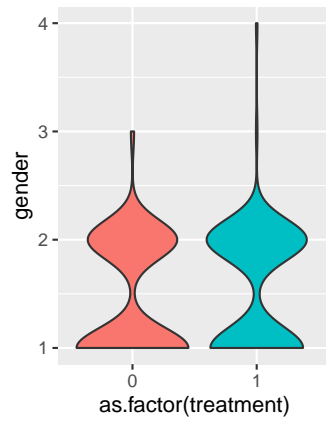
p6 = ggplot(d, aes(x=as.factor(treatment), y=covid_sick, fill=as.factor(treatment))) +
  geom_violin()

p7 = ggplot(d, aes(x=as.factor(treatment), y=covid_hospitalized, fill=as.factor(treatment))) +
  geom_violin()

p8 = ggplot(d, aes(x=as.factor(treatment), y=covid_died, fill=as.factor(treatment))) +
  geom_violin()

multiplot(p1, p2, p3, p4, p5, p6, p7, p8, cols=2)

```



## 4. Estimate ATE

```
estimate_ate <- function(dt, outcome, treatment, treat_val) {  
  ## This takes a data.table, the name of the outcome variable, and the name  
  ## of the treatment indicator.  
  
  g <- dt[, .(group_mean = mean(get(outcome))), keyby = .(get(treatment))]  
  ate <- g[, diff(group_mean)]  
  
  return(ate)  
}  
ri <-function(num=10000){  
  
  res <- NA  
  for (i in 1:num) {  
    res[i] <- d[, .(group_mean = mean(views)), keyby = .(sample(success))][, diff(group_mean)]  
  }  
  
  return (res)  
}
```

### 4.1 Estimate the ATE for the outcome on how concerned subject is about COVID-19 continued spread

```
g <- d[, .(group_mean = mean(outcome_spread)), keyby = .(treatment)]  
g
```

```
##      treatment group_mean  
## 1:           0   3.132812  
## 2:           1   3.473282
```

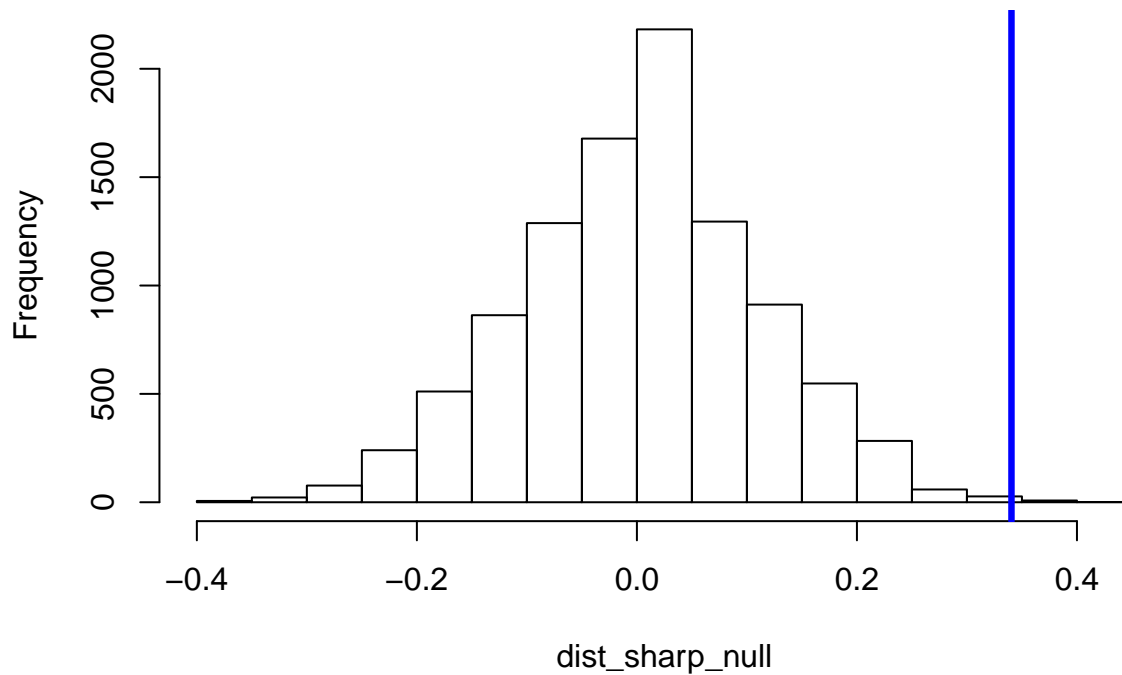
```
ate_spread <- g[, diff(group_mean)]  
ate_spread
```

```
## [1] 0.3404699
```

```
res <- NA  
for (i in 1:10000) {  
  res[i] <- d[, .(group_mean = mean(outcome_spread)), keyby = .(sample(treatment))][, diff(group_mean)]  
}  
dist_sharp_null <- res  
hist(dist_sharp_null)  
abline(v=ate_spread, lwd=3, col='blue')  
abline(v=abs(ate_spread), lwd=3, col='blue')
```



**Histogram of dist\_sharp\_null**



```
p_value_one_tailed <- mean(dist_sharp_null >= ate_spread)
p_value_one_tailed
```

```
## [1] 0.0015
```

```
p_value_two_tailed <- mean(abs(dist_sharp_null) >= abs(ate_spread))
p_value_two_tailed
```

```
## [1] 0.0021
```

#### 4.2 Estimate the ATE for the outcome on emotional reaction to COVID-19 deaths in US

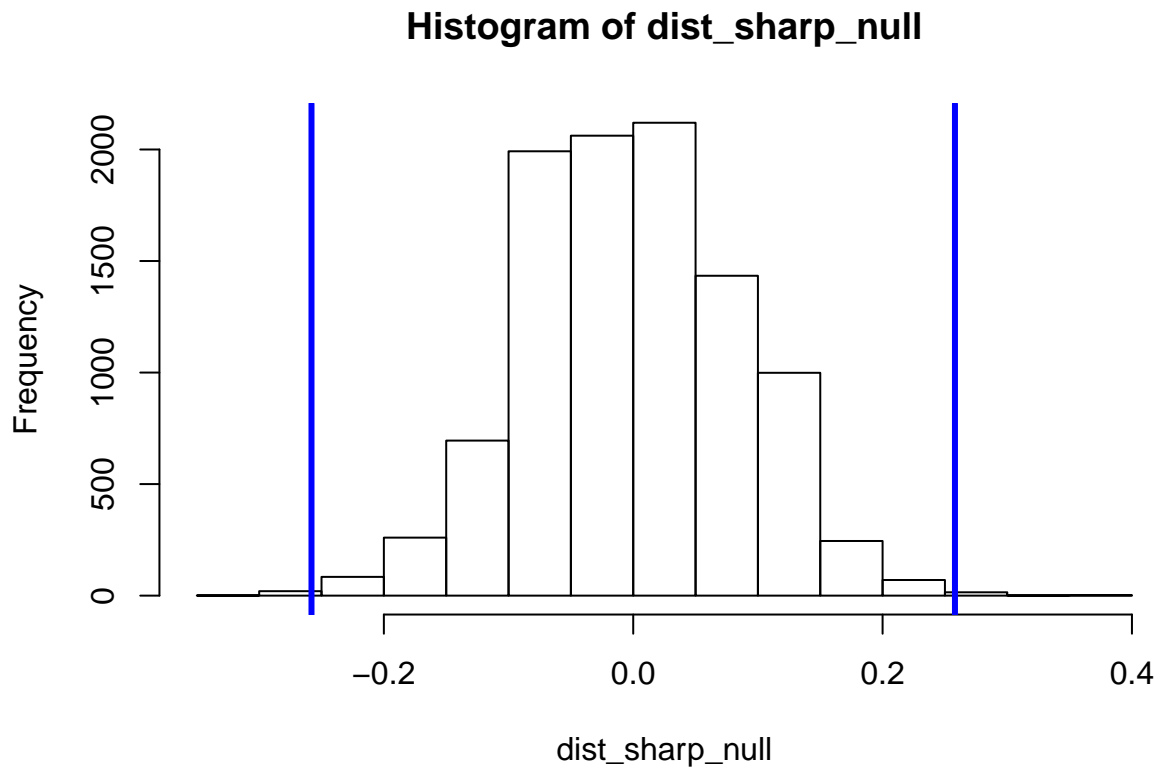
```
g <- d[, .(group_mean = mean(outcome_death)), keyby = .(treatment)]
g
```

```
##      treatment group_mean
## 1:           0  2.398438
## 2:           1  2.656489
```

```
ate_death <- g[, diff(group_mean)]
ate_death
```

```
## [1] 0.258051
```

```
res <- NA
for (i in 1:10000) {
  res[i] <- d[, .(group_mean = mean(outcome_death)), keyby = .(sample(treatment))][, diff(group_mean)]
}
dist_sharp_null <- res
hist(dist_sharp_null)
abline(v=ate_death, lwd=3, col='blue')
abline(v=ate_death*-1, lwd=3, col='blue')
```



```
p_value_one_tailed <- mean(dist_sharp_null >= ate_death)
p_value_one_tailed
```

```
## [1] 0.0018
```

```
p_value_two_tailed <- mean(abs(dist_sharp_null) >= abs(ate_death))
p_value_two_tailed
```

```
## [1] 0.003
```

## 5. Linear Regression

### 5.1 Perform linear regression on concern over COVID-19 spread outcome

```
model_spread = lm(outcome_spread ~ treatment, d)
model_spread1 = lm(outcome_spread ~ treatment
+ as.factor(gender)
+ as.factor(age)
+ as.factor(ethnicity)
+ as.factor(political_party)
+ as.factor(education)
+ as.factor(covid_sick)
+ as.factor(covid_hospitalized)
+ as.factor(covid_died), d)
stargazer(model_spread, model_spread1, type="text")
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               outcome_spread
##                               (1)                (2)
## -----
## treatment                    0.340***          0.339***
##                               (0.107)          (0.104)
##
## as.factor(gender)2           -0.063
##                               (0.102)
##
## as.factor(gender)3           0.886*
##                               (0.508)
##
## as.factor(gender)4           0.646
##                               (0.795)
##
## as.factor(age)2              0.323
##                               (0.347)
##
## as.factor(age)3              0.295
##                               (0.344)
##
## as.factor(age)4              0.181
##                               (0.362)
##
## as.factor(age)5              0.370
##                               (0.364)
##
## as.factor(age)6              0.393
##                               (0.425)
##
## as.factor(age)7              1.156**
##                               (0.494)
##
```

```

## as.factor(ethnicity)2          0.195
##                               (0.219)
##
## as.factor(ethnicity)3         -0.087
##                               (0.170)
##
## as.factor(ethnicity)4          0.230
##                               (0.233)
##
## as.factor(ethnicity)5         -0.471
##                               (0.508)
##
## as.factor(ethnicity)6        -2.519***
##                               (0.799)
##
## as.factor(ethnicity)7          0.715
##                               (0.824)
##
## as.factor(political_party)2    0.675***
##                               (0.126)
##
## as.factor(political_party)3    0.357**
##                               (0.149)
##
## as.factor(political_party)4   -0.222
##                               (0.324)
##
## as.factor(education)2         -0.050
##                               (0.494)
##
## as.factor(education)3          0.194
##                               (0.483)
##
## as.factor(education)4          0.017
##                               (0.492)
##
## as.factor(education)5          0.188
##                               (0.477)
##
## as.factor(education)6          0.337
##                               (0.492)
##
## as.factor(education)7         -0.166
##                               (0.906)
##
## as.factor(education)8          0.502
##                               (0.569)
##
## as.factor(covid_sick)2        -0.200
##                               (0.121)
##
## as.factor(covid_hospitalized)2 -0.166
##                               (0.165)
##

```

```
## as.factor(covid_died)2                                0.092
##                                                         (0.181)
##
## Constant                                              3.133***
##                                                         (0.076)
##                                                         2.450***
##                                                         (0.634)
## -----
## Observations                                         259
## R2                                                    0.038
## Adjusted R2                                          0.034
## Residual Std. Error      0.858 (df = 257)          0.768 (df = 229)
## F Statistic      10.183*** (df = 1; 257)  3.614*** (df = 29; 229)
## =====
## Note:                                               *p<0.1; **p<0.05; ***p<0.01
```

## 5.2 Perform linear regression on reaction to COVID-19 deaths outcome

```
model_death = lm(outcome_death ~ treatment, d)
model_death1 = lm(outcome_death ~ treatment
  + as.factor(gender)
  + as.factor(age)
  + as.factor(ethnicity)
  + as.factor(political_party)
  + as.factor(education)
  + as.factor(covid_sick)
  + as.factor(covid_hospitalized)
  + as.factor(covid_died),
  d)

stargazer(model_death, model_death1, type="text")
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               outcome_death
##                               (1)                (2)
## -----
## treatment                    0.258***          0.250***
##                               (0.084)          (0.085)
##
## as.factor(gender)2                                0.050
##                                                     (0.084)
##
## as.factor(gender)3                                0.573
##                                                     (0.415)
##
## as.factor(gender)4                                0.437
##                                                     (0.649)
##
## as.factor(age)2                                   -0.371
##                                                     (0.283)
```

```

##
## as.factor(age)3                -0.323
##                                (0.281)
##
## as.factor(age)4                -0.345
##                                (0.296)
##
## as.factor(age)5                -0.258
##                                (0.297)
##
## as.factor(age)6                -0.279
##                                (0.348)
##
## as.factor(age)7                0.170
##                                (0.404)
##
## as.factor(ethnicity)2          0.041
##                                (0.179)
##
## as.factor(ethnicity)3          -0.103
##                                (0.139)
##
## as.factor(ethnicity)4          -0.019
##                                (0.190)
##
## as.factor(ethnicity)5          -0.261
##                                (0.415)
##
## as.factor(ethnicity)6          -1.667**
##                                (0.653)
##
## as.factor(ethnicity)7          0.374
##                                (0.674)
##
## as.factor(political_party)2    0.543***
##                                (0.103)
##
## as.factor(political_party)3    0.340***
##                                (0.122)
##
## as.factor(political_party)4    -0.107
##                                (0.265)
##
## as.factor(education)2          0.233
##                                (0.404)
##
## as.factor(education)3          0.216
##                                (0.395)
##
## as.factor(education)4          0.135
##                                (0.402)
##
## as.factor(education)5          0.088
##                                (0.390)

```

```

##
## as.factor(education)6          0.263
##                               (0.402)
##
## as.factor(education)7         -0.475
##                               (0.741)
##
## as.factor(education)8          0.290
##                               (0.465)
##
## as.factor(covid_sick)2         -0.102
##                               (0.099)
##
## as.factor(covid_hospitalized)2 -0.084
##                               (0.135)
##
## as.factor(covid_died)2         0.067
##                               (0.148)
##
## Constant                      2.398***
##                               (0.059)
##                               2.327***
##                               (0.518)
## -----
## Observations                   259
## R2                             0.036
## Adjusted R2                    0.032
## Residual Std. Error            0.672 (df = 257)
## F Statistic                    9.533*** (df = 1; 257)
##                               0.628 (df = 229)
##                               2.657*** (df = 29; 229)
## =====
## Note:                          *p<0.1; **p<0.05; ***p<0.01

```