Cheap Tweets

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Intro

```
#Read in needed packages
library(stargazer)
```

```
##
## Please cite as:
##  Hlavac, Marek (2022). stargazer: Well-Formatted Regression and Summary Statistics Tables.
##  R package version 5.2.3. https://CRAN.R-project.org/package=stargazer
```

```
library(lmtest)
```

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

```
#set WD

setwd("/Users/harri/Dropbox (MIT)/Audience Costs and Twitter/SUBMISSION 2023/ISQ Final Accepted/Replica
```

Public Sample

```
DF_PublicSample <-read.csv("Data, Public_Sample.csv")
```

Demographic Table Stats

Gender

```
table(DF_PublicSample$Q.Female)
```

```
##
##   0   1
## 476 442
```

```
length(na.omit(DF_PublicSample$Q.Female[DF_PublicSample$Q.Female==0]))/length(DF_PublicSample$Q.Female)
```

```
## [1] 0.4872057
```

```
sum(na.omit(DF_PublicSample$Q.Female))/length(DF_PublicSample$Q.Female) #females
```

```
## [1] 0.4524053
```

```
sum(is.na(DF_PublicSample$Q.Female))/length(DF_PublicSample$Q.Female) #NAs/other
```

```
## [1] 0.06038895
```

Race

```
sum(na.omit(DF_PublicSample$Q.White))/length((DF_PublicSample$Q.White))
```

```
## [1] 0.6806551
```

```
sum(na.omit(DF_PublicSample$Q.Black))/length((DF_PublicSample$Q.Black))
```

```
## [1] 0.09825998
```

```
sum(na.omit(DF_PublicSample$Q.AIorAN))/length((DF_PublicSample$Q.AIorAN))
```

```
## [1] 0.01330604
```

```
sum(na.omit(DF_PublicSample$Q.Asian))/length((DF_PublicSample$Q.Asian))
```

```
## [1] 0.0419652
```

```
sum(na.omit(DF_PublicSample$Q.NHorPI))/length((DF_PublicSample$Q.NHorPI))
```

```
## [1] 0.008188332
```

```
sum(na.omit(DF_PublicSample$Q.Hispanic))/length((DF_PublicSample$Q.Hispanic))
```

```
## [1] 0.07369498
```

```
sum(na.omit(DF_PublicSample$Q.Mixed))/length((DF_PublicSample$Q.Mixed))
```

```
## [1] 0.0174002
```

```
sum(na.omit(DF_PublicSample$Q.Other))/length((DF_PublicSample$Q.Other))
```

```
## [1] 0.01023541
```

```
sum(na.omit(DF_PublicSample$Q.Other_Mixed))/length((DF_PublicSample$Q.Other_Mixed))
```

```
## [1] 0.02763562
```

```
sum(is.na(DF_PublicSample$Q.Race))/length(DF_PublicSample$Q.Race)
```

```
## [1] 0.05629478
```

Age

```
summary(DF_PublicSample$Q.Age)
```

```
##    Min. 1st Qu.  Median    Mean 3rd Qu.    Max.    NA's
##   18.00   32.00   46.00   46.33   60.00   88.00      52
```

Education

```
table(DF_PublicSample$Q.Education)
```

```
##
##   1   2   3   4   5   6   7   8
##   8  35 196 195 122 229  30 110
```

```
sum((na.omit(DF_PublicSample$Q.HighSchool)))/length((DF_PublicSample$Q.HighSchool))
```

```
## [1] 0.9027636
```

```
sum((na.omit(DF_PublicSample$Q.Bach)))/length((DF_PublicSample$Q.Bach))
```

```
## [1] 0.3776868
```

```
sum(is.na(DF_PublicSample$Q.Education))/length(DF_PublicSample$Q.Education)
```

```
## [1] 0.05322416
```

Income

```
table(DF_PublicSample$Q.Income)
```

```
##
##   1   2   3   4   5   6   7   8
## 166 225 165 113  67  71  31  81
```

```
length(na.omit(DF_PublicSample$Q.Income[DF_PublicSample$Q.Income==1]))/length(DF_PublicSample$Q.Income)
```

```
## [1] 0.1699079
```

```
length(na.omit(DF_PublicSample$Q.Income[DF_PublicSample$Q.Income==2]))/length(DF_PublicSample$Q.Income)
```

```
## [1] 0.2302968
```

```
length(na.omit(DF_PublicSample$Q.Income[DF_PublicSample$Q.Income==3]))/length(DF_PublicSample$Q.Income)
```

```
## [1] 0.1688843
```

```
length(na.omit(DF_PublicSample$Q.Income[DF_PublicSample$Q.Income==4]))/length(DF_PublicSample$Q.Income)
```

```
## [1] 0.1156602
```

```
length(na.omit(DF_PublicSample$Q.Income[DF_PublicSample$Q.Income==5]))/length(DF_PublicSample$Q.Income)
```

```
## [1] 0.06857728
```

```
length(na.omit(DF_PublicSample$Q.Income[DF_PublicSample$Q.Income==6]))/length(DF_PublicSample$Q.Income)
```

```
## [1] 0.07267144
```

```
length(na.omit(DF_PublicSample$Q.Income[DF_PublicSample$Q.Income==7]))/length(DF_PublicSample$Q.Income)
```

```
## [1] 0.03172979
```

```
length(na.omit(DF_PublicSample$Q.Income[DF_PublicSample$Q.Income==8]))/length(DF_PublicSample$Q.Income)
```

```
## [1] 0.08290686
```

```
sum(is.na(DF_PublicSample$Q.Income))/length(DF_PublicSample$Q.Income)
```

```
## [1] 0.0593654
```

Political ID

```
table(DF_PublicSample$Q.Political_ID)
```

```
##
##   1   2   3   4   5
## 118 155 361 183 103
```

```
sum(na.omit(DF_PublicSample$Q.Liberal))/length((DF_PublicSample$Q.Liberal))
```

```
## [1] 0.2794268
```

```
sum(na.omit(DF_PublicSample$Q.Moderate))/length((DF_PublicSample$Q.Moderate))
```

```
## [1] 0.3694985
```

```
sum(na.omit(DF_PublicSample$Q.Conservative))/length((DF_PublicSample$Q.Conservative))
```

```
## [1] 0.2927329
```

```
sum(is.na(DF_PublicSample$Q.Political_ID))/length(DF_PublicSample$Q.Political_ID)
```

```
## [1] 0.05834186
```

Veteran

```
table(DF_PublicSample$Q.Veteran)
```

```
##
##   0   1
## 803 124
```

```
length(na.omit(DF_PublicSample$Q.Veteran[DF_PublicSample$Q.Veteran==0]))/length(DF_PublicSample$Q.Vetera
```

```
## [1] 0.8219038
```

```
length(na.omit(DF_PublicSample$Q.Veteran[DF_PublicSample$Q.Veteran==1]))/length(DF_PublicSample$Q.Vetera
```

```
## [1] 0.1269191
```

```
sum(is.na(DF_PublicSample$Q.Veteran))/length(DF_PublicSample$Q.Veteran)
```

```
## [1] 0.05117707
```

Analysis

Credibility

```
## a simple regression for mean and se estimates


######## Credibility


Cred_Full_1 <- Credibility ~ 0 + as.factor(Tweet)
lm_Cred_Full_1 <- lm(Cred_Full_1,
                     data = DF_PublicSample)

summary(lm_Cred_Full_1)
```

```
##
## Call:
## lm(formula = Cred_Full_1, data = DF_PublicSample)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -2.2963 -1.2179  0.7037  0.7821  1.7821
##
## Coefficients:
##                    Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  3.29630    0.05885   56.01   <2e-16 ***
## as.factor(Tweet)1  3.21792    0.05855   54.96   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.297 on 975 degrees of freedom
## Multiple R-squared:  0.8633, Adjusted R-squared:  0.863
## F-statistic:  3079 on 2 and 975 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_PublicSample$Credibility[DF_PublicSample$Tweet==1]))
```

```
## [1] 491
```

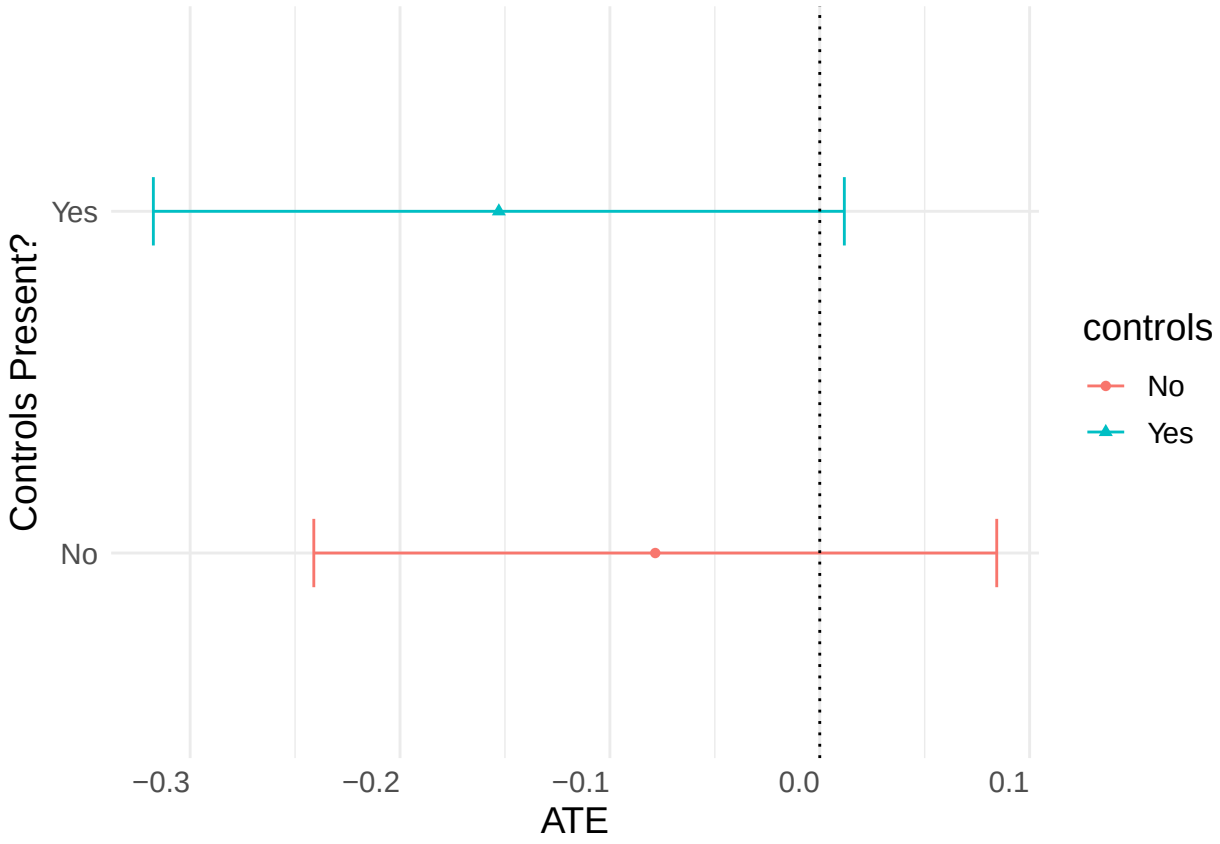
```
length(na.omit(DF_PublicSample$Credibility[DF_PublicSample$Tweet==0]))
```

```
## [1] 486
```

```
#time for some regressions
#let's do a couple different models

##DV: credibility, IV: Tweet binary

## Model 1: No Demographics

Cred_Form_1 <- Credibility ~ Tweet

## Model 5: Factor Demographics

Cred_Form_5 <-
  Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran

## Model 7: Slightly simplified for the Appendix

Cred_Form_7 <- Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran

#regression
```

```r
Cred_1 <- lm(Cred_Form_1,
             data = DF_PublicSample,
             na.action=na.omit)

Cred_5 <- lm(Cred_Form_5,
             data = DF_PublicSample,
             na.action=na.omit)

Cred_7 <- lm(Cred_Form_7,
             data = DF_PublicSample,
             na.action=na.omit)

library(stargazer)


stargazer(Cred_1, Cred_5, Cred_7, title = "Perceived Credibility with Covariates", no.space = TRUE)
```

```r
# Credibility ATE Graph


Cred_Alone <- as.data.frame(matrix(data = c("credibility", coeftest(Cred_5)[2, 1:2],"Yes",
                                    "credibility", coeftest(Cred_1)[2, 1:2], "No"),
                                    ncol = 4, byrow = TRUE))

colnames(Cred_Alone) <- c("dv", "estimate", "se", "controls")

#making into correct operators
Cred_Alone$estimate <- as.numeric(Cred_Alone$estimate)
Cred_Alone$se <- as.numeric(Cred_Alone$se)

#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Cred_Alone$ci <- Cred_Alone$se*q

library(ggplot2)
## graph time
pd <- position_dodge(0.5)

ggplot(Cred_Alone, aes(x = controls, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Controls Present?") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```
#ggtitle("Credibility, ATE of Tweet Medium, Public")
```

Manipulation Check

```
##### stargazer tables for MC

# Threat/Treatment Check

## Model 1: No Demographics

Threat_Form_1 <- Threat_Check_Binary ~ Tweet


## Model 5: Factor Demographics

Threat_Form_5 <- Threat_Check_Binary ~ Tweet  + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran



## Model 7: Slightly simplified for the Appendix

Threat_Form_7 <- Threat_Check_Binary ~ Tweet + Q.Female +
```

```
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran

#regression

Threat_1 <- lm(Threat_Form_1,
               data = DF_PublicSample,
               na.action=na.omit)

Threat_5 <- lm(Threat_Form_5,
               data = DF_PublicSample,
               na.action=na.omit)

Threat_7 <- lm(Threat_Form_7,
               data = DF_PublicSample,
               na.action=na.omit)


stargazer(Threat_1, Threat_5, Threat_7, title = "Treatment Check Question with Covariates", no.space = T
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:35
## \begin{table}[!htbp] \centering
##   \caption{Treatment Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Threat\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.215$^{***}$ & $-$0.225$^{***}$ & $-$0.226$^{***}$ \\
##   & (0.031) & (0.031) & (0.031) \\
##   Q.Female &  & 0.080$^{**}$ & 0.083$^{**}$ \\
##   &  & (0.033) & (0.033) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.062 & $-$0.053 \\
##   &  & (0.053) & (0.054) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & $-$0.044 & $-$0.061 \\
##   &  & (0.133) & (0.132) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & $-$0.002 & 0.025 \\
##   &  & (0.077) & (0.077) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & 0.007 & 0.008 \\
##   &  & (0.179) & (0.180) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & $-$0.028 & $-$0.034 \\
##   &  & (0.061) & (0.061) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & 0.237$^{**}$ & 0.243$^{**}$ \\
##   &  & (0.116) & (0.116) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.262$^{*}$ & 0.242 \\
##   &  & (0.157) & (0.159) \\
```

```
##   Q.Age &  & 0.001 & 0.001 \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.108 &  \\
##   &  & (0.079) &  \\
##   Q.Bach &  & 0.053 & 0.057 \\
##   &  & (0.037) & (0.037) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & $-$0.041 &  \\
##   &  & (0.058) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & $-$0.080 &  \\
##   &  & (0.049) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.065 &  \\
##   &  & (0.051) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & $-$0.0
##   &  & (0.058) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & $-$0.0
##   &  & (0.057) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & 0.079 &
##   &  & (0.052) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & 0.066 &
##   &  & (0.051) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 0.031 &  \\
##   &  & (0.045) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.095$^{**}$ &  \\
##   &  & (0.040) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.058 &  \\
##   &  & (0.039) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.039
##   &  & (0.032) &  \\
##   Q.Income &  &  & $-$0.003 \\
##   &  &  & (0.008) \\
##   Q.Conservative &  &  & 0.020 \\
##   &  &  & (0.039) \\
##   Q.Liberal &  &  & $-$0.033 \\
##   &  &  & (0.039) \\
##   Q.Veteran &  & $-$0.147$^{***}$ & $-$0.142$^{***}$ \\
##   &  & (0.048) & (0.048) \\
##   Constant & 0.695$^{***}$ & 0.501$^{***}$ & 0.619$^{***}$ \\
##   & (0.022) & (0.092) & (0.065) \\
##  \hline \\[-1.8ex]
## Observations & 977 & 906 & 906 \\
## R$^{2}$ & 0.048 & 0.120 & 0.094 \\
## Adjusted R$^{2}$ & 0.047 & 0.096 & 0.078 \\
## Residual Std. Error & 0.481 (df = 975) & 0.466 (df = 881) & 0.470 (df = 890) \\
## F Statistic & 48.735$^{***}$ (df = 1; 975) & 4.994$^{***}$ (df = 24; 881) & 6.139$^{***}$ (df = 15; 8
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```
# Target Check

## Model 1: No Demographics
```

```r
Target_Form_1 <- Target_Check_Binary ~ Tweet


## Model 5: Factor Demographics

Target_Form_5 <- Target_Check_Binary ~ Tweet  +  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran



## Model 7: Slightly simplified for the Appendix

Target_Form_7 <- Target_Check_Binary ~ Tweet  + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran


#regression

Target_1 <- lm(Target_Form_1,
               data = DF_PublicSample,
               na.action=na.omit)

Target_5 <- lm(Target_Form_5,
               data = DF_PublicSample,
               na.action=na.omit)

Target_7 <- lm(Target_Form_7,
               data = DF_PublicSample,
               na.action=na.omit)


stargazer(Target_1, Target_5, Target_7, title = "Target Check Question with Covariates", no.space = TRU
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:35
## \begin{table}[!htbp] \centering
##   \caption{Target Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Target\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
```

```
##  Tweet & 0.033 & 0.037 & 0.036 \\
##   & (0.031) & (0.032) & (0.032) \\
##   Q.Female &  & 0.074$^{**}$ & 0.071$^{**}$ \\
##   &  & (0.034) & (0.034) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.077 & $-$0.073 \\
##   &  & (0.055) & (0.055) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & 0.194 & 0.190 \\
##   &  & (0.137) & (0.135) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & 0.040 & 0.059 \\
##   &  & (0.079) & (0.078) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & 0.019 & 0.020 \\
##   &  & (0.184) & (0.184) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & 0.026 & 0.016 \\
##   &  & (0.063) & (0.063) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & 0.100 & 0.097 \\
##   &  & (0.119) & (0.119) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.174 & 0.156 \\
##   &  & (0.162) & (0.162) \\
##   Q.Age &  & 0.003$^{***}$ & 0.003$^{***}$ \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.143$^{*}$ &  \\
##   &  & (0.081) &  \\
##   Q.Bach &  & 0.008 & 0.013 \\
##   &  & (0.038) & (0.037) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & $-$0.067 &  \\
##   &  & (0.060) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & 0.017 &  \\
##   &  & (0.050) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & $-$0.004 &  \\
##   &  & (0.053) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & $-$0.0
##   &  & (0.059) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & $-$0.08
##   &  & (0.058) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & 0.006 &
##   &  & (0.054) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & 0.040 &
##   &  & (0.052) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 0.068 &  \\
##   &  & (0.047) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.071$^{*}$ &  \\
##   &  & (0.041) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.008 &  \\
##   &  & (0.040) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.029
##   &  & (0.033) &  \\
##   Q.Income &  &  & $-$0.013 \\
##   &  &  & (0.009) \\
##   Q.Conservative &  &  & 0.018 \\
##   &  &  & (0.040) \\
##   Q.Liberal &  &  & $-$0.042 \\
##   &  &  & (0.040) \\
##   Q.Veteran &  & $-$0.140$^{***}$ & $-$0.137$^{***}$ \\
##   &  & (0.049) & (0.049) \\
```

```
##   Constant & 0.580$^{***}$ & 0.280$^{***}$ & 0.474$^{***}$ \\
##   & (0.022) & (0.095) & (0.066) \\
##  \hline \\[-1.8ex]
## Observations & 977 & 906 & 906 \\
## R$^{2}$ & 0.001 & 0.057 & 0.045 \\
## Adjusted R$^{2}$ & 0.0001 & 0.031 & 0.029 \\
## Residual Std. Error & 0.491 (df = 975) & 0.480 (df = 881) & 0.480 (df = 890) \\
## F Statistic & 1.090 (df = 1; 975) & 2.209$^{***}$ (df = 24; 881) & 2.809$^{***}$ (df = 15; 890) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```
# Support Check

## Model 1: No Demographics

Support_Form_1 <- Support_Check_Binary ~ Tweet


## Model 5: Factor Demographics

Support_Form_5 <- Support_Check_Binary ~ Tweet  + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran



## Model 7: Slightly simplified for the Appendix

Support_Form_7 <- Support_Check_Binary ~ Tweet  + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran


#regression

Support_1 <- lm(Support_Form_1,
             data = DF_PublicSample,
             na.action=na.omit)

Support_5 <- lm(Support_Form_5,
             data = DF_PublicSample,
             na.action=na.omit)

Support_7 <- lm(Support_Form_7,
             data = DF_PublicSample,
             na.action=na.omit)
```

```
stargazer(Support_1, Support_5, Support_7, title = "Support Check Question with Covariates", no.space =
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:36
## \begin{table}[!htbp] \centering
##   \caption{Support Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Support\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & 0.032 & 0.026 & 0.019 \\
##   & (0.030) & (0.031) & (0.031) \\
##   Q.Female &  & $-$0.034 & $-$0.026 \\
##   &  & (0.033) & (0.033) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.038 & $-$0.037 \\
##   &  & (0.053) & (0.053) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & $-$0.018 & $-$0.068 \\
##   &  & (0.131) & (0.130) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & 0.122 & 0.138$^{*}$ \\
##   &  & (0.075) & (0.076) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & 0.098 & 0.126 \\
##   &  & (0.176) & (0.177) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & 0.013 & 0.003 \\
##   &  & (0.060) & (0.060) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & $-$0.011 & $-$0.011 \\
##   &  & (0.114) & (0.114) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.153 & 0.125 \\
##   &  & (0.155) & (0.156) \\
##   Q.Age &  & 0.002$^{**}$ & 0.003$^{***}$ \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.330$^{***}$ &  \\
##   &  & (0.077) &  \\
##   Q.Bach &  & 0.086$^{**}$ & 0.098$^{***}$ \\
##   &  & (0.036) & (0.036) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & $-$0.008 &  \\
##   &  & (0.057) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & $-$0.0003 &  \\
##   &  & (0.048) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.033 &  \\
##   &  & (0.050) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & $-$0.04
##   &  & (0.057) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & $-$0.0
##   &  & (0.056) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & $-$0.0
##   &  & (0.051) &  \\
```

```
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & 0.034 &
##   &  & (0.050) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 0.091$^{**}$ &  \\
##   &  & (0.045) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.009 &  \\
##   &  & (0.039) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.018 &  \\
##   &  & (0.038) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.082
##   &  & (0.031) &  \\
##   Q.Income &  &  & 0.001 \\
##   &  &  & (0.008) \\
##   Q.Conservative &  &  & 0.113$^{***}$ \\
##   &  &  & (0.039) \\
##   Q.Liberal &  &  & 0.038 \\
##   &  &  & (0.039) \\
##   Q.Veteran &  & $-$0.043 & $-$0.042 \\
##   &  & (0.047) & (0.047) \\
##   Constant & 0.636$^{***}$ & 0.217$^{**}$ & 0.440$^{***}$ \\
##   & (0.022) & (0.091) & (0.064) \\
##  \hline \\[-1.8ex]
## Observations & 977 & 906 & 906 \\
## R$^{2}$ & 0.001 & 0.076 & 0.049 \\
## Adjusted R$^{2}$ & 0.0001 & 0.051 & 0.033 \\
## Residual Std. Error & 0.477 (df = 975) & 0.459 (df = 881) & 0.463 (df = 890) \\
## F Statistic & 1.117 (df = 1; 975) & 3.010$^{***}$ (df = 24; 881) & 3.075$^{***}$ (df = 15; 890) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```r
# Sea Check

## Model 1: No Demographics

Sea_Form_1 <- Sea_Check_Binary ~ Tweet


## Model 5: Factor Demographics

Sea_Form_5 <- Sea_Check_Binary ~ Tweet  + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran



## Model 7: Slightly simplified for the Appendix

Sea_Form_7 <- Sea_Check_Binary ~ Tweet  + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
```

```
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran


#regression

Sea_1 <- lm(Sea_Form_1,
             data = DF_PublicSample,
             na.action=na.omit)

Sea_5 <- lm(Sea_Form_5,
             data = DF_PublicSample,
             na.action=na.omit)

Sea_7 <- lm(Sea_Form_7,
             data = DF_PublicSample,
             na.action=na.omit)


stargazer(Sea_1, Sea_5, Sea_7, title = "Sea Check Question with Covariates", no.space = TRUE)
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:36
## \begin{table}[!htbp] \centering
##   \caption{Sea Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Sea\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.0002 & 0.002 & $-$0.002 \\
##   & (0.031) & (0.032) & (0.032) \\
##   Q.Female &  & 0.121$^{***}$ & 0.128$^{***}$ \\
##   &  & (0.034) & (0.034) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.012 & $-$0.013 \\
##   &  & (0.054) & (0.054) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & $-$0.019 & $-$0.030 \\
##   &  & (0.135) & (0.134) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & 0.038 & 0.057 \\
##   &  & (0.078) & (0.078) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & 0.070 & 0.072 \\
##   &  & (0.182) & (0.182) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & $-$0.034 & $-$0.036 \\
##   &  & (0.062) & (0.062) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & 0.184 & 0.176 \\
##   &  & (0.118) & (0.117) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.182 & 0.168 \\
##   &  & (0.160) & (0.160) \\
```

```
##   Q.Age &  & $-$0.001 & $-$0.001 \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.095 &  \\
##   &  & (0.080) &  \\
##   Q.Bach &  & $-$0.090$^{**}$ & $-$0.086$^{**}$ \\
##   &  & (0.037) & (0.037) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & 0.022 &  \\
##   &  & (0.059) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & $-$0.040 &  \\
##   &  & (0.050) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.048 &  \\
##   &  & (0.052) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & $-$0.0
##   &  & (0.059) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & $-$0.08
##   &  & (0.058) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & $-$0.0
##   &  & (0.053) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & $-$0.0
##   &  & (0.052) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 0.072 &  \\
##   &  & (0.046) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.021 &  \\
##   &  & (0.040) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.064 &  \\
##   &  & (0.039) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.018
##   &  & (0.032) &  \\
##   Q.Income &  &  & 0.005 \\
##   &  &  & (0.008) \\
##   Q.Conservative &  &  & 0.017 \\
##   &  &  & (0.040) \\
##   Q.Liberal &  &  & $-$0.009 \\
##   &  &  & (0.040) \\
##   Q.Veteran &  & $-$0.117$^{**}$ & $-$0.110$^{**}$ \\
##   &  & (0.049) & (0.048) \\
##   Constant & 0.621$^{***}$ & 0.602$^{***}$ & 0.641$^{***}$ \\
##   & (0.022) & (0.094) & (0.065) \\
##  \hline \\[-1.8ex]
## Observations & 977 & 906 & 906 \\
## R$^{2}$ & 0.00000 & 0.055 & 0.042 \\
## Adjusted R$^{2}$ & $-$0.001 & 0.029 & 0.025 \\
## Residual Std. Error & 0.486 (df = 975) & 0.474 (df = 881) & 0.475 (df = 890) \\
## F Statistic & 0.00005 (df = 1; 975) & 2.137$^{***}$ (df = 24; 881) & 2.571$^{***}$ (df = 15; 890) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```
#simple manipulation check percentages
#using DF_PublicSample

## overall
```

```r
sum(DF_PublicSample$Threat_Check_Binary)/length(DF_PublicSample$Threat_Check_Binary)
```

```
## [1] 0.5875128
```

```r
sum(DF_PublicSample$Target_Check_Binary)/length(DF_PublicSample$Target_Check_Binary)
```

```
## [1] 0.5967247
```

```r
sum(DF_PublicSample$Support_Check_Binary)/length(DF_PublicSample$Support_Check_Binary)
```

```
## [1] 0.6519959
```

```r
sum(DF_PublicSample$Sea_Check_Binary)/length(DF_PublicSample$Sea_Check_Binary)
```

```
## [1] 0.6212897
```

```r
## Tweet Treatments
sum(DF_PublicSample$Threat_Check_Binary[DF_PublicSample$Treatment==1 | DF_PublicSample$Treatment==2])/le
```

```
## [1] NaN
```

```r
sum(DF_PublicSample$Target_Check_Binary[DF_PublicSample$Treatment==1 | DF_PublicSample$Treatment==2])/le
```

```
## [1] NaN
```

```r
sum(DF_PublicSample$Support_Check_Binary[DF_PublicSample$Treatment==1 | DF_PublicSample$Treatment==2])/l
```

```
## [1] NaN
```

```r
sum(DF_PublicSample$Sea_Check_Binary[DF_PublicSample$Treatment==1 | DF_PublicSample$Treatment==2])/lengt
```

```
## [1] NaN
```

```r
## Official Statement Treatments
sum(DF_PublicSample$Threat_Check_Binary[DF_PublicSample$Treatment==3 | DF_PublicSample$Treatment==4])/le
```

```
## [1] NaN
```

```r
sum(DF_PublicSample$Target_Check_Binary[DF_PublicSample$Treatment==3 | DF_PublicSample$Treatment==4])/le
```

```
## [1] NaN
```

```r
sum(DF_PublicSample$Support_Check_Binary[DF_PublicSample$Treatment==3 | DF_PublicSample$Treatment==4])/l
```

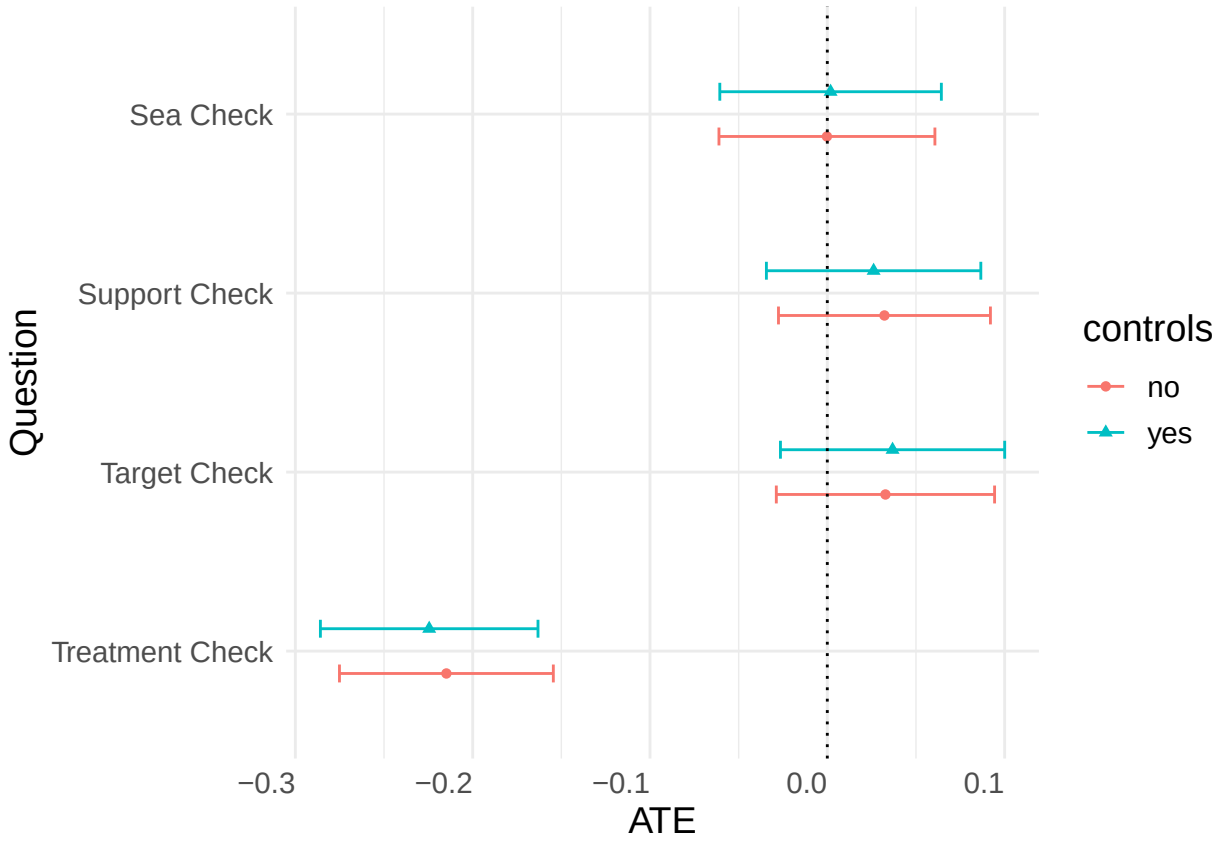
```
## [1] NaN
```

```r
sum(DF_PublicSample$Sea_Check_Binary[DF_PublicSample$Treatment==3 | DF_PublicSample$Treatment==4])/lengt
```

```
## [1] NaN
```

```r
## graphs for manipulation check

#put all the names, estimates, SEs, and behavior into one df for graphing
Results_DF <- as.data.frame(matrix(data = c("Treatment Check", coeftest(Threat_1)[2, 1:2],"no",
                                    "Treatment Check", coeftest(Threat_5)[2, 1:2], "yes",
                                    "Target Check", coeftest(Target_1)[2, 1:2],"no",
                                    "Target Check", coeftest(Target_5)[2, 1:2], "yes",
                                    "Support Check", coeftest(Support_1)[2, 1:2], "no",
                                    "Support Check", coeftest(Support_5)[2, 1:2], "yes",
                                    "Sea Check", coeftest(Sea_1)[2, 1:2], "no",
                                    "Sea Check", coeftest(Sea_5)[2, 1:2], "yes"),
                                    ncol = 4, byrow = TRUE))
```

```r
colnames(Results_DF) <- c("dv", "estimate", "se", "controls")

#making into correct operators
Results_DF$dv <- factor(Results_DF$dv, levels = c("Treatment Check", "Target Check", "Support Check", "S
Results_DF$estimate <- as.numeric(Results_DF$estimate)
Results_DF$se <- as.numeric(Results_DF$se)

#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Results_DF$ci <- Results_DF$se*q


## graph time
library(ggplot2)
pd <- position_dodge(0.5)

ggplot(Results_DF, aes(x = dv, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Question") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```r
#ggtitle("Tweet Imagery ATE by MC or AC Question") +

## mean and se tables for threat check
```

```
####Threat/Treatment Check
Threat_Full_1 <- Threat_Check_Binary ~ 0 + as.factor(Tweet)


lm_Threat_Full_1 <- lm(Threat_Full_1,
                       data = DF_PublicSample)

summary(lm_Threat_Full_1)
```

```
##
## Call:
## lm(formula = Threat_Full_1, data = DF_PublicSample)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6955 -0.4807  0.3045  0.3045  0.5193
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.69547    0.02181   31.88   <2e-16 ***
## as.factor(Tweet)1  0.48065    0.02170   22.15   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4809 on 975 degrees of freedom
## Multiple R-squared:  0.6071, Adjusted R-squared:  0.6063
## F-statistic: 753.4 on 2 and 975 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_PublicSample$Threat_Check_Binary[DF_PublicSample$Tweet==1]))
```

```
## [1] 491
```

```
length(na.omit(DF_PublicSample$Threat_Check_Binary[DF_PublicSample$Tweet==0]))
```

```
## [1] 486
```

```
##### Target Check
Target_Full_1 <- Target_Check_Binary ~ 0 + as.factor(Tweet)

lm_Target_Full_1 <- lm(Target_Full_1,
                       data = DF_PublicSample)

summary(lm_Target_Full_1)
```

```
##
## Call:
## lm(formula = Target_Full_1, data = DF_PublicSample)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6130 -0.5803  0.3870  0.4198  0.4198
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.58025    0.02226   26.06   <2e-16 ***
## as.factor(Tweet)1  0.61303    0.02215   27.68   <2e-16 ***
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4908 on 975 degrees of freedom
## Multiple R-squared:  0.5972, Adjusted R-squared:  0.5963
## F-statistic: 722.7 on 2 and 975 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_PublicSample$Target_Check_Binary[DF_PublicSample$Tweet==1]))
```

```
## [1] 491
```

```
length(na.omit(DF_PublicSample$Target_Check_Binary[DF_PublicSample$Tweet==0]))
```

```
## [1] 486
```

```
##### Support Check

Support_Full_1 <- Support_Check_Binary ~ 0 + as.factor(Tweet)

lm_Support_Full_1 <- lm(Support_Full_1,
                     data = DF_PublicSample)

summary(lm_Support_Full_1)
```

```
##
## Call:
## lm(formula = Support_Full_1, data = DF_PublicSample)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6680 -0.6358  0.3320  0.3642  0.3642
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.63580    0.02162   29.41   <2e-16 ***
## as.factor(Tweet)1  0.66802    0.02151   31.06   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4766 on 975 degrees of freedom
## Multiple R-squared:  0.6524, Adjusted R-squared:  0.6517
## F-statistic: 914.9 on 2 and 975 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_PublicSample$Support_Check_Binary[DF_PublicSample$Tweet==1]))
```

```
## [1] 491
```

```
length(na.omit(DF_PublicSample$Support_Check_Binary[DF_PublicSample$Tweet==0]))
```

```
## [1] 486
```

```
##### Sea Check

Sea_Full_1 <- Sea_Check_Binary ~ 0 + as.factor(Tweet)

lm_Sea_Full_1 <- lm(Sea_Full_1,
                     data = DF_PublicSample)
```

```
summary(lm_Sea_Full_1)
```

```
##
## Call:
## lm(formula = Sea_Full_1, data = DF_PublicSample)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6214 -0.6212  0.3786  0.3788  0.3788
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.62140    0.02203   28.21   <2e-16 ***
## as.factor(Tweet)1  0.62118    0.02191   28.35   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4856 on 975 degrees of freedom
## Multiple R-squared:  0.6213, Adjusted R-squared:  0.6205
## F-statistic: 799.8 on 2 and 975 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_PublicSample$Sea_Check_Binary[DF_PublicSample$Tweet==1]))
```

```
## [1] 491
```

```
length(na.omit(DF_PublicSample$Sea_Check_Binary[DF_PublicSample$Tweet==0]))
```

```
## [1] 486
```

Interaction Effects

```
# looking at interactions between partisanship and tweet
# and between age and Tweet
# and between veteran and Tweet

### two new formulas for interaction. 8 is interaction with partisanship. 9 is for age

Cred_Form_8 <- Credibility ~ Tweet*Q.Liberal + Tweet*Q.Conservative + Q.Female + relevel(as.factor(Q.Ra
    Q.Age + Q.Bach + Q.Income  +  Q.Veteran

Cred_Form_9 <- Credibility ~ Tweet*Q.Age + Q.Female + relevel(as.factor(Q.Race), ref = 1) +
    Q.Bach + Q.Income + Q.Conservative + Q.Liberal +
    Q.Veteran

Cred_Form_V <- Credibility ~ Tweet*Q.Veteran + Q.Female + relevel(as.factor(Q.Race), ref = 1) +
    Q.Bach + Q.Income + Q.Conservative + Q.Liberal + Q.Age




Cred_8 <- lm(Cred_Form_8, data = DF_PublicSample)
Cred_9 <- lm(Cred_Form_9, data = DF_PublicSample)
Cred_V <- lm(Cred_Form_V, data = DF_PublicSample)
```

```
stargazer(Cred_8, Cred_9, Cred_V, title = "Crediblity with Interaction Terms with Covariates", no.space
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:36
## \begin{table}[!htbp] \centering
##   \caption{Crediblity with Interaction Terms with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Credibility} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.096 & $-$0.068 & $-$0.114 \\
##   & (0.134) & (0.247) & (0.090) \\
##   Q.Liberal & 0.426$^{***}$ & 0.371$^{***}$ & 0.375$^{***}$ \\
##   & (0.148) & (0.104) & (0.104) \\
##   Q.Conservative & $-$0.522$^{***}$ & $-$0.553$^{***}$ & $-$0.553$^{***}$ \\
##   & (0.143) & (0.104) & (0.104) \\
##   Q.Female & $-$0.058 & $-$0.059 & $-$0.065 \\
##   & (0.089) & (0.088) & (0.088) \\
##   relevel(as.factor(Q.Race), ref = 1)2 & 0.158 & 0.155 & 0.153 \\
##   & (0.143) & (0.143) & (0.143) \\
##   relevel(as.factor(Q.Race), ref = 1)3 & $-$0.373 & $-$0.386 & $-$0.381 \\
##   & (0.353) & (0.352) & (0.352) \\
##   relevel(as.factor(Q.Race), ref = 1)4 & 0.086 & 0.093 & 0.096 \\
##   & (0.204) & (0.204) & (0.204) \\
##   relevel(as.factor(Q.Race), ref = 1)5 & 0.375 & 0.365 & 0.384 \\
##   & (0.479) & (0.479) & (0.479) \\
##   relevel(as.factor(Q.Race), ref = 1)6 & $-$0.144 & $-$0.147 & $-$0.138 \\
##   & (0.163) & (0.163) & (0.163) \\
##   relevel(as.factor(Q.Race), ref = 1)7 & $-$0.195 & $-$0.194 & $-$0.193 \\
##   & (0.310) & (0.309) & (0.309) \\
##   relevel(as.factor(Q.Race), ref = 1)8 & $-$0.091 & $-$0.099 & $-$0.074 \\
##   & (0.422) & (0.422) & (0.422) \\
##   Q.Age & $-$0.005$^{**}$ & $-$0.005 & $-$0.005$^{**}$ \\
##   & (0.003) & (0.004) & (0.003) \\
##   Tweet:Q.Veteran &  &  & $-$0.250 \\
##   &  &  & (0.248) \\
##   Q.Bach & 0.040 & 0.038 & 0.036 \\
##   & (0.097) & (0.097) & (0.097) \\
##   Q.Income & 0.029 & 0.029 & 0.029 \\
##   & (0.022) & (0.022) & (0.022) \\
##   Q.Veteran & 0.308$^{**}$ & 0.309$^{**}$ & 0.441$^{**}$ \\
##   & (0.128) & (0.127) & (0.184) \\
##   Tweet:Q.Liberal & $-$0.104 &  &  \\
##   & (0.205) &  &  \\
##   Tweet:Q.Conservative & $-$0.066 &  &  \\
##   & (0.201) &  &  \\
##   Tweet:Q.Age &  & $-$0.002 &  \\
```

```
##   &  & (0.005) &  \\
##   Constant & 3.501$^{***}$ & 3.486$^{***}$ & 3.505$^{***}$ \\
##   & (0.179) & (0.207) & (0.173) \\
##  \hline \\[-1.8ex]
## Observations & 906 & 906 & 906 \\
## R$^{2}$ & 0.108 & 0.108 & 0.109 \\
## Adjusted R$^{2}$ & 0.091 & 0.092 & 0.093 \\
## Residual Std. Error & 1.250 (df = 888) & 1.250 (df = 889) & 1.249 (df = 889) \\
## F Statistic & 6.322$^{***}$ (df = 17; 888) & 6.713$^{***}$ (df = 16; 889) & 6.776$^{***}$ (df = 16; 8
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

AC Sensitivity Analysis

Just three ACs

```
DF4 <- subset(DF_PublicSample, DF_PublicSample$Target_Check_Binary==1)
DF5 <- subset(DF4, DF4$Support_Check_Binary==1)
DF6 <- subset(DF5, DF5$Sea_Check_Binary==1)
```

Subset

```
## a simple regression for mean and se estimates


######## Credibility


AC.Cred_Full_1 <- Credibility ~ 0 + as.factor(Tweet)
AC.lm_Cred_Full_1 <- lm(AC.Cred_Full_1,
                     data = DF6)

summary(AC.lm_Cred_Full_1)
```

Analysis

```
##
## Call:
## lm(formula = AC.Cred_Full_1, data = DF6)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -2.2458 -1.1258  0.7542  0.8742  1.8742
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  3.12579    0.10281    30.4   <2e-16 ***
## as.factor(Tweet)1  3.24581    0.09689    33.5   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```
## Residual standard error: 1.296 on 336 degrees of freedom
## Multiple R-squared:  0.859,  Adjusted R-squared:  0.8581
## F-statistic:  1023 on 2 and 336 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF6$Credibility[DF6$Tweet==1]))
```

```
## [1] 179
```

```
length(na.omit(DF6$Credibility[DF6$Tweet==0]))
```

```
## [1] 159
```

```
#time for some regressions

#let's do a couple different models


##DV: credibility, IV: Tweet binary

## Model 1: No Demographics

AC.Cred_Form_1 <- Credibility ~ Tweet

## Model 2: Demographics in Erik's Paper


## Model 5: Factor Demographics

AC.Cred_Form_5 <-
  Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran


## Model 7: Slightly simplified for the Appendix

AC.Cred_Form_7 <- Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran

#regression

AC.Cred_1 <- lm(AC.Cred_Form_1,
             data = DF6,
             na.action=na.omit)

AC.Cred_5 <- lm(AC.Cred_Form_5,
             data = DF6,
             na.action=na.omit)

AC.Cred_7 <- lm(AC.Cred_Form_7,
```

```
              data = DF6,
              na.action=na.omit)

library(stargazer)


stargazer(AC.Cred_1, AC.Cred_5, AC.Cred_7, title = "Perceived Credibility, AC Questions Correct", no.spa
```

Just with MC

```
DF7 <- subset(DF_PublicSample, DF_PublicSample$Threat_Check_Binary==1)
```

Subset

```
## a simple regression for mean and se estimates


######## Credibility


MC.Cred_Full_1 <- Credibility ~ 0 + as.factor(Tweet)
MC.lm_Cred_Full_1 <- lm(MC.Cred_Full_1,
                     data = DF7)

summary(MC.lm_Cred_Full_1)
```

Analysis

```
##
## Call:
## lm(formula = MC.Cred_Full_1, data = DF7)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -2.1923 -1.1923 -0.1525  0.8475  1.8475
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  3.19231    0.07065   45.19   <2e-16 ***
## as.factor(Tweet)1  3.15254    0.08455   37.29   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.299 on 572 degrees of freedom
## Multiple R-squared:  0.8571, Adjusted R-squared:  0.8566
## F-statistic:  1716 on 2 and 572 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF7$Credibility[DF7$Tweet==1]))
```

```
## [1] 236
```

```
length(na.omit(DF7$Credibility[DF7$Tweet==0]))
```

```
## [1] 338
```

```r
#time for some regressions

#let's do a couple different models


##DV: credibility, IV: Tweet binary

## Model 1: No Demographics

MC.Cred_Form_1 <- Credibility ~ Tweet

## Model 2: Demographics in Erik's Paper


## Model 5: Factor Demographics

MC.Cred_Form_5 <-
  Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran


## Model 7: Slightly simplified for the Appendix

MC.Cred_Form_7 <- Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran

#regression

MC.Cred_1 <- lm(MC.Cred_Form_1,
             data = DF7,
             na.action=na.omit)

MC.Cred_5 <- lm(MC.Cred_Form_5,
             data = DF7,
             na.action=na.omit)

MC.Cred_7 <- lm(MC.Cred_Form_7,
             data = DF7,
             na.action=na.omit)

library(stargazer)


stargazer(MC.Cred_1, MC.Cred_5, MC.Cred_7, title = "Perceived Credibility, MC Question Correct", no.spa
```

Expert Experiment

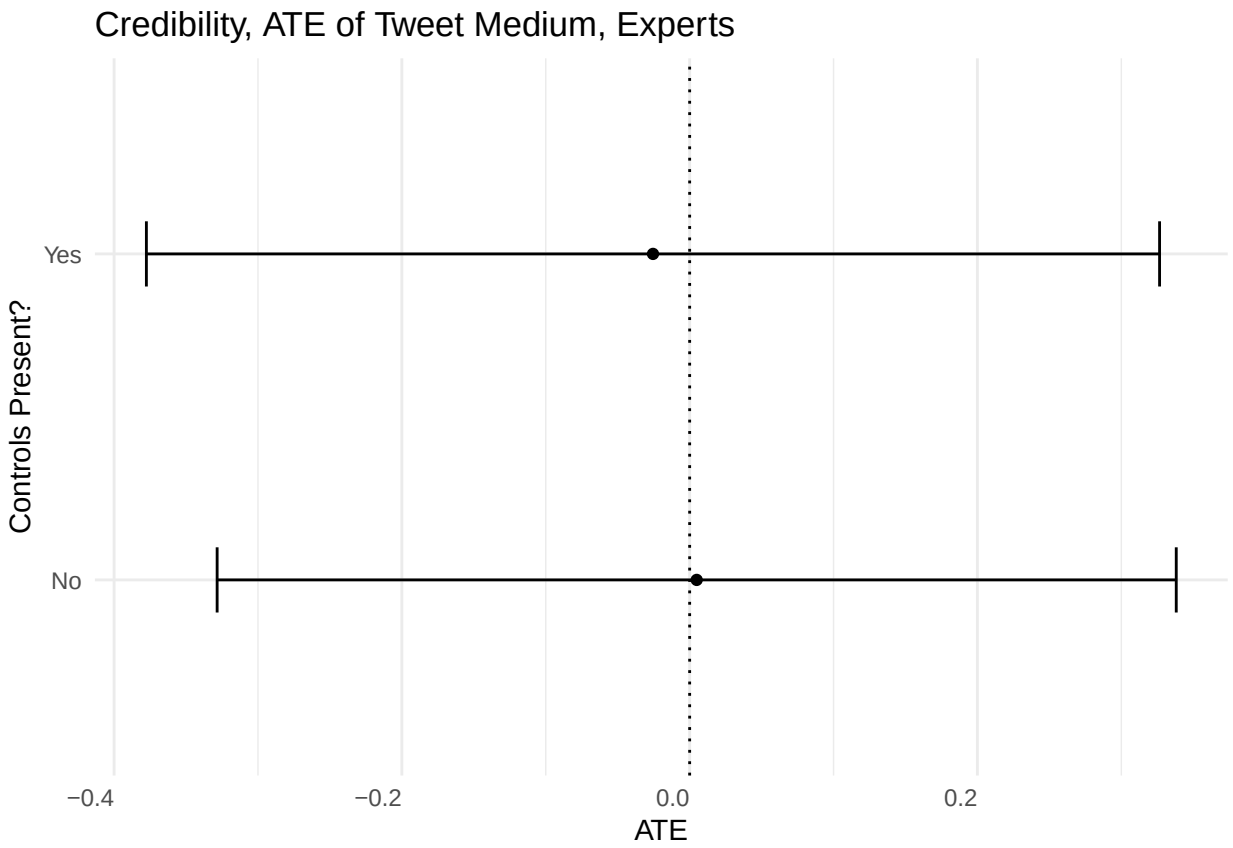
```
DF_ExpertSample <-read.csv("Data, Expert_Sample.csv")
```

Analysis

```
# regression formulas

## credibility

### Model 1: No Demographics

Cred_form_1 <- Credibility ~ Tweet

### Model 5: All demographics, and ordered factors

Cred_form_5 <- Credibility ~ Tweet + AfterKabul + female + Veteran_Binary + Gov_Binary  + bach +
  as.factor(Country) + Age + as.factor(Current_Affiliation) + factor(Political_ID, ordered = TRUE, level
```

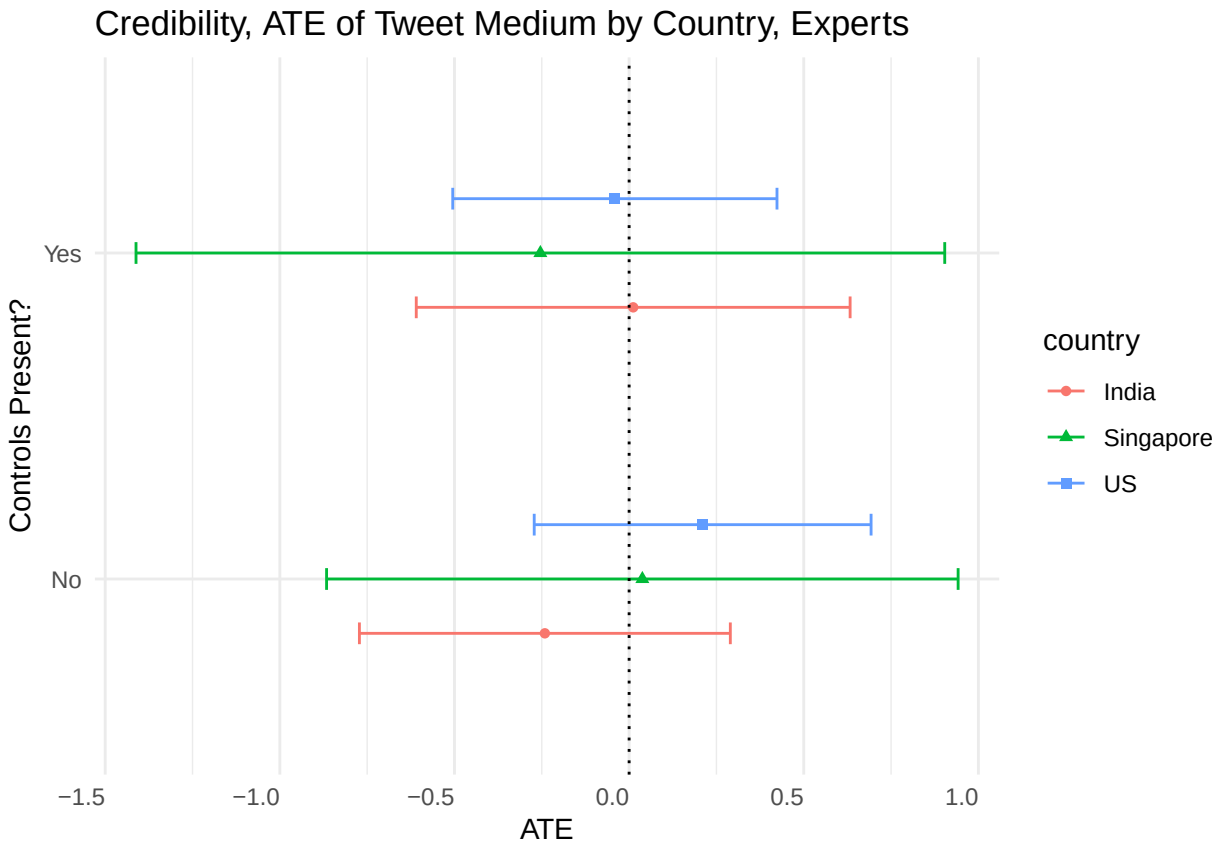
```
# Regressions

library(stats)
library(lmtest)

##credibility
lm_cred_5 <- lm(formula = Cred_form_5, data = DF_ExpertSample)
lm_cred_1 <- lm(formula = Cred_form_1, data = DF_ExpertSample)



Cred_DF <- as.data.frame(matrix(data = c("credibility", coeftest(lm_cred_5)[2, 1:2],"Yes",
                                         "credibility", coeftest(lm_cred_1)[2, 1:2], "No"),
                                         ncol = 4, byrow = TRUE))

colnames(Cred_DF) <- c("dv", "estimate", "se", "controls")


#making into correct operators
Cred_DF$estimate <- as.numeric(Cred_DF$estimate)
Cred_DF$se <- as.numeric(Cred_DF$se)


#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))

Cred_DF$ci <- Cred_DF$se*q


## graph time
library(ggplot2)
pd <- position_dodge(0.5)

## cred graph
```

```
ggplot(Cred_DF, aes(x = controls, y = estimate))  +
  geom_point(aes(), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Controls Present?") + ylab("ATE") +
  ggtitle("Credibility, ATE of Tweet Medium, Experts") + geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1)) + coord_flip()
```




```
# let's repeat the graphs, but break them out by country

DF_US <- subset(DF_ExpertSample, DF_ExpertSample$Country==1)
DF_Singapore <- subset(DF_ExpertSample, DF_ExpertSample$Country==2)
DF_India <- subset(DF_ExpertSample, DF_ExpertSample$Country==3)

# Credibility

##have to make a new model 5 formula without country
Cred_form_5_NC <- Credibility ~ Tweet + AfterKabul + female + Veteran_Binary +
    Gov_Binary + highschool + bach + Age +
    as.factor(Current_Affiliation) + factor(Political_ID, ordered = TRUE,
    levels = c(1, 2, 3, 4, 5))

# Regressions

lm_cred_5_US <- lm(formula = Cred_form_5_NC, data = DF_US)
lm_cred_1_US <- lm(formula = Cred_form_1, data = DF_US)
lm_cred_5_Singapore <- lm(formula = Cred_form_5_NC, data = DF_Singapore)
lm_cred_1_Singapore <- lm(formula = Cred_form_1, data = DF_Singapore)
```

```r
lm_cred_5_India <- lm(formula = Cred_form_5_NC, data = DF_India)
lm_cred_1_India <- lm(formula = Cred_form_1, data = DF_India)


Cred_DF_Country <- as.data.frame(matrix(data = c("credibility", coeftest(lm_cred_5_US)[2, 1:2],"Yes", "U
                                    "credibility", coeftest(lm_cred_1_US)[2, 1:2], "No", "US",
                                    "credibility", coeftest(lm_cred_5_Singapore)[2, 1:2],"Yes", "Singapo
                                    "credibility", coeftest(lm_cred_1_Singapore)[2, 1:2], "No", "Singapo
                                    "credibility", coeftest(lm_cred_5_India)[2, 1:2],"Yes", "India",
                                    "credibility", coeftest(lm_cred_1_India)[2, 1:2], "No", "India"),
                                    ncol = 5, byrow = TRUE))


colnames(Cred_DF_Country) <- c("dv", "estimate", "se", "controls", "country")


#making into correct operators
Cred_DF_Country$estimate <- as.numeric(Cred_DF_Country$estimate)
Cred_DF_Country$se <- as.numeric(Cred_DF_Country$se)
Cred_DF_Country$country <- factor(Cred_DF_Country$country)


#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))

Cred_DF_Country$ci <- Cred_DF_Country$se*q



## cred graph
ggplot(Cred_DF_Country, aes(x = controls, y = estimate, color = country))  +
  geom_point(aes(color=country, shape=country), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Controls Present?") + ylab("ATE") +
  ggtitle("Credibility, ATE of Tweet Medium by Country, Experts") + geom_hline(yintercept = 0, linetype=
  theme(axis.text.x = element_text(hjust = 1)) + coord_flip()
```




```
#stargazer of the above
library(stargazer)

stargazer(lm_cred_1, lm_cred_5, title = "Credibility, Full Elite Sample", no.space = TRUE)
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:37
## \begin{table}[!htbp] \centering
##   \caption{Credibility, Full Elite Sample}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lcc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{2}{c}{\textit{Dependent variable:}} \\
## \cline{2-3}
## \\[-1.8ex] & \multicolumn{2}{c}{Credibility} \\
## \\[-1.8ex] & (1) & (2)\\
## \hline \\[-1.8ex]
##  Tweet & 0.005 & $-$0.025 \\
##   & (0.170) & (0.180) \\
##   AfterKabul &  & 0.032 \\
##   &  & (0.266) \\
##   female &  & $-$0.406$^{*}$ \\
##   &  & (0.209) \\
##   Veteran\_Binary &  & $-$0.424$^{**}$ \\
##   &  & (0.211) \\
```

```
##   Gov\_Binary &  & 0.019 \\
##   &  & (0.245) \\
##   bach &  & 0.517 \\
##   &  & (0.382) \\
##   as.factor(Country)2 &  & 0.384 \\
##   &  & (0.333) \\
##   as.factor(Country)3 &  & $-$0.244 \\
##   &  & (0.223) \\
##   Age &  & 0.013$^{*}$ \\
##   &  & (0.007) \\
##   as.factor(Current\_Affiliation)2 &  & 0.350 \\
##   &  & (0.297) \\
##   as.factor(Current\_Affiliation)3 &  & 0.308 \\
##   &  & (0.358) \\
##   as.factor(Current\_Affiliation)4 &  & 0.176 \\
##   &  & (0.252) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.489 \\
##   &  & (0.417) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.346 \\
##   &  & (0.364) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.358 \\
##   &  & (0.288) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & 0.190 \\
##   &  & (0.202) \\
##   Constant & 3.104$^{***}$ & 2.097$^{***}$ \\
##   & (0.121) & (0.612) \\
##  \hline \\[-1.8ex]
## Observations & 234 & 212 \\
## R$^{2}$ & 0.00000 & 0.112 \\
## Adjusted R$^{2}$ & $-$0.004 & 0.039 \\
## Residual Std. Error & 1.300 (df = 232) & 1.266 (df = 195) \\
## F Statistic & 0.001 (df = 1; 232) & 1.533$^{*}$ (df = 16; 195) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{2}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```
stargazer(lm_cred_1_US, lm_cred_5_US, title = "Credibility, US Elite Sample", no.space = TRUE)
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:37
## \begin{table}[!htbp] \centering
##   \caption{Credibility, US Elite Sample}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lcc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{2}{c}{\textit{Dependent variable:}} \\
## \cline{2-3}
## \\[-1.8ex] & \multicolumn{2}{c}{Credibility} \\
## \\[-1.8ex] & (1) & (2)\\
## \hline \\[-1.8ex]
##  Tweet & 0.211 & $-$0.041 \\
```

```
##   & (0.246) & (0.237) \\
##   AfterKabul &  & 0.032 \\
##   &  & (0.283) \\
##   female &  & $-$0.672$^{**}$ \\
##   &  & (0.261) \\
##   Veteran\_Binary &  & $-$0.223 \\
##   &  & (0.290) \\
##   Gov\_Binary &  & 0.233 \\
##   &  & (0.473) \\
##   highschool &  &  \\
##   &  &  \\
##   bach &  & 0.670 \\
##   &  & (0.533) \\
##   Age &  & 0.018$^{*}$ \\
##   &  & (0.010) \\
##   as.factor(Current\_Affiliation)2 &  & 0.480 \\
##   &  & (0.474) \\
##   as.factor(Current\_Affiliation)3 &  & 0.849 \\
##   &  & (0.774) \\
##   as.factor(Current\_Affiliation)4 &  & $-$0.175 \\
##   &  & (0.342) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$1.984$^{***}$ \\
##   &  & (0.630) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.950$^{*}$ \\
##   &  & (0.546) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.014 \\
##   &  & (0.378) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & 0.174 \\
##   &  & (0.258) \\
##   Constant & 3.000$^{***}$ & 1.330$^{*}$ \\
##   & (0.172) & (0.782) \\
##  \hline \\[-1.8ex]
## Observations & 117 & 111 \\
## R$^{2}$ & 0.006 & 0.257 \\
## Adjusted R$^{2}$ & $-$0.002 & 0.149 \\
## Residual Std. Error & 1.330 (df = 115) & 1.207 (df = 96) \\
## F Statistic & 0.732 (df = 1; 115) & 2.375$^{***}$ (df = 14; 96) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{2}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```
stargazer(lm_cred_1_Singapore, lm_cred_5_Singapore, title = "Credibility, India Elite Sample", no.space
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:37
## \begin{table}[!htbp] \centering
##   \caption{Credibility, India Elite Sample}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lcc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{2}{c}{\textit{Dependent variable:}} \\
```

```
## \cline{2-3}
## \\[-1.8ex] & \multicolumn{2}{c}{Credibility} \\
## \\[-1.8ex] & (1) & (2)\\
## \hline \\[-1.8ex]
##  Tweet & 0.038 & $-$0.254 \\
##   & (0.461) & (0.591) \\
##   AfterKabul &  & 0.780 \\
##   &  & (0.975) \\
##   female &  & 1.925$^{*}$ \\
##   &  & (0.951) \\
##   Veteran\_Binary &  & 1.227 \\
##   &  & (0.750) \\
##   Gov\_Binary &  & 0.272 \\
##   &  & (0.721) \\
##   highschool &  &  \\
##   &  &  \\
##   bach &  & 3.042 \\
##   &  & (1.644) \\
##   Age &  & 0.103$^{**}$ \\
##   &  & (0.041) \\
##   as.factor(Current\_Affiliation)2 &  & $-$0.249 \\
##   &  & (0.986) \\
##   as.factor(Current\_Affiliation)3 &  & 5.190$^{*}$ \\
##   &  & (2.394) \\
##   as.factor(Current\_Affiliation)4 &  & 1.320 \\
##   &  & (1.027) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 2.296 \\
##   &  & (1.685) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$1.321 \\
##   &  & (1.660) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 1.677$^{*}$ \\
##   &  & (0.848) \\
##   Constant & 3.417$^{***}$ & $-$5.368 \\
##   & (0.319) & (4.336) \\
##  \hline \\[-1.8ex]
## Observations & 23 & 21 \\
## R$^{2}$ & 0.0003 & 0.710 \\
## Adjusted R$^{2}$ & $-$0.047 & 0.170 \\
## Residual Std. Error & 1.105 (df = 21) & 0.982 (df = 7) \\
## F Statistic & 0.007 (df = 1; 21) & 1.316 (df = 13; 7) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{2}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```
stargazer(lm_cred_1_India, lm_cred_5_India, title = "Credibility, Singapore Elite Sample", no.space = TR
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:37
## \begin{table}[!htbp] \centering
##   \caption{Credibility, Singapore Elite Sample}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lcc}
```

```
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{2}{c}{\textit{Dependent variable:}} \\
## \cline{2-3}
## \\[-1.8ex] & \multicolumn{2}{c}{Credibility} \\
## \\[-1.8ex] & (1) & (2)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.241 & 0.012 \\
##   & (0.271) & (0.317) \\
##   AfterKabul &  &  \\
##   &  &  \\
##   female &  & $-$0.210 \\
##   &  & (0.438) \\
##   Veteran\_Binary &  & $-$0.857$^{**}$ \\
##   &  & (0.405) \\
##   Gov\_Binary &  & $-$0.123 \\
##   &  & (0.373) \\
##   highschool &  &  \\
##   &  &  \\
##   bach &  & 1.251 \\
##   &  & (0.896) \\
##   Age &  & 0.013 \\
##   &  & (0.015) \\
##   as.factor(Current\_Affiliation)2 &  & 0.190 \\
##   &  & (0.528) \\
##   as.factor(Current\_Affiliation)3 &  & $-$0.308 \\
##   &  & (0.526) \\
##   as.factor(Current\_Affiliation)4 &  & $-$0.064 \\
##   &  & (0.465) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 0.750 \\
##   &  & (0.633) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.271 \\
##   &  & (0.600) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.592 \\
##   &  & (0.573) \\
##   factor(Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & 0.297 \\
##   &  & (0.409) \\
##   Constant & 3.163$^{***}$ & 1.668 \\
##   & (0.200) & (1.053) \\
##  \hline \\[-1.8ex]
## Observations & 94 & 80 \\
## R$^{2}$ & 0.009 & 0.161 \\
## Adjusted R$^{2}$ & $-$0.002 & $-$0.004 \\
## Residual Std. Error & 1.309 (df = 92) & 1.322 (df = 66) \\
## F Statistic & 0.793 (df = 1; 92) & 0.974 (df = 13; 66) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{2}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```
#regressions for means and SEs

##because of small size, I have to get the basic values without covariates
```

```r
#full sample
Cred_Full_1 <- Credibility ~ 0 + as.factor(Tweet)



lm_Cred_Full_1 <- lm(Cred_Full_1, data = DF_ExpertSample)


summary(lm_Cred_Full_1)
```

```
##
## Call:
## lm(formula = Cred_Full_1, data = DF_ExpertSample)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -2.1092 -1.1092  0.8908  0.8956  1.8957
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0   3.1043     0.1213   25.60   <2e-16 ***
## as.factor(Tweet)1   3.1092     0.1192   26.08   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.3 on 232 degrees of freedom
## Multiple R-squared:  0.852,  Adjusted R-squared:  0.8507
## F-statistic: 667.8 on 2 and 232 DF,  p-value: < 2.2e-16
```

```r
length(na.omit(DF_ExpertSample$Credibility[DF_ExpertSample$Tweet==1]))
```

```
## [1] 119
```

```r
length(na.omit(DF_ExpertSample$Credibility[DF_ExpertSample$Tweet==0]))
```

```
## [1] 115
```

```r
# by country

Cred_Full_1_NC <- Credibility ~ 0 + as.factor(Tweet)


## US

lm_Cred_Full_1_NC_US <- lm(Cred_Full_1_NC, data = DF_US)


summary(lm_Cred_Full_1_NC_US)
```

```
##
## Call:
## lm(formula = Cred_Full_1_NC, data = DF_US)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
```

```
## -2.2105 -1.2105  0.7895  1.0000  2.0000
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0   3.0000     0.1717   17.47   <2e-16 ***
## as.factor(Tweet)1   3.2105     0.1762   18.22   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.33 on 115 degrees of freedom
## Multiple R-squared:  0.8471, Adjusted R-squared:  0.8445
## F-statistic: 318.6 on 2 and 115 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_US$Credibility[DF_US$Tweet==1]))
```

```
## [1] 57
```

```
length(na.omit(DF_US$Credibility[DF_US$Tweet==0]))
```

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

```
## Singapore
lm_Cred_Full_1_NC_Singapore <- lm(Cred_Full_1_NC, data = DF_Singapore)

mean(DF_Singapore$Credibility[DF_Singapore$Tweet==1])
```

```
## [1] 3.454545
```

```
summary(lm_Cred_Full_1_NC_Singapore)
```

```
##
## Call:
## lm(formula = Cred_Full_1_NC, data = DF_Singapore)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -1.4546 -1.4167  0.5454  0.5833  1.5833
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0   3.4167     0.3190   10.71 5.75e-10 ***
## as.factor(Tweet)1   3.4545     0.3332   10.37 1.02e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.105 on 21 degrees of freedom
## Multiple R-squared:  0.9137, Adjusted R-squared:  0.9054
## F-statistic: 111.1 on 2 and 21 DF,  p-value: 6.767e-12
```

```
length(na.omit(DF_Singapore$Credibility[DF_Singapore$Tweet==1]))
```

```
## [1] 11
```

```
length(na.omit(DF_Singapore$Credibility[DF_Singapore$Tweet==0]))
```

```
## [1] 12
```

```
## India
lm_Cred_Full_1_NC_India <- lm(Cred_Full_1_NC, data = DF_India)
```

```
summary(lm_Cred_Full_1_NC_India)
```

```
##
## Call:
## lm(formula = Cred_Full_1_NC, data = DF_India)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -2.1628 -1.1628  0.8372  1.0784  2.0784
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0   3.1628     0.1996   15.85   <2e-16 ***
## as.factor(Tweet)1   2.9216     0.1832   15.94   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.309 on 92 degrees of freedom
## Multiple R-squared:  0.846,  Adjusted R-squared:  0.8426
## F-statistic: 252.7 on 2 and 92 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_India$Credibility[DF_India$Tweet==1]))
```

```
## [1] 51
```

```
length(na.omit(DF_India$Credibility[DF_India$Tweet==0]))
```

```
## [1] 43
```

Follow On 1, President and Language

```
DF_Follow1 <-read.csv("Data, Follow_1_President.csv")
```

Demographic Table Stats

```
### demographics

table(DF_Follow1$Q.Female)
```

```
##
##   0   1
## 666 706
```

```
length(na.omit(DF_Follow1$Q.Female[DF_Follow1$Q.Female==0]))/length(DF_Follow1$Q.Female)
```

```
## [1] 0.4815618
```

```
sum(na.omit(DF_Follow1$Q.Female))/length(DF_Follow1$Q.Female)
```

```
## [1] 0.5104845
```

```
sum(is.na(DF_Follow1$Q.Female))/length(DF_Follow1$Q.Female)
```

```
## [1] 0.007953724
```

```
sum(na.omit(DF_Follow1$Q.White))/length((DF_Follow1$Q.White))
```

```
## [1] 0.7151121
```

```
sum(na.omit(DF_Follow1$Q.Black))/length((DF_Follow1$Q.Black))
```

```
## [1] 0.1070137
```

```
sum(na.omit(DF_Follow1$Q.AIorAN))/length((DF_Follow1$Q.AIorAN))
```

```
## [1] 0.007953724
```

```
sum(na.omit(DF_Follow1$Q.Asian))/length((DF_Follow1$Q.Asian))
```

```
## [1] 0.05061461
```

```
sum(na.omit(DF_Follow1$Q.NHorPI))/length((DF_Follow1$Q.NHorPI))
```

```
## [1] 0.002169197
```

```
sum(na.omit(DF_Follow1$Q.Hispanic))/length((DF_Follow1$Q.Hispanic))
```

```
## [1] 0.08098337
```

```
sum(na.omit(DF_Follow1$Q.Mixed))/length((DF_Follow1$Q.Mixed))
```

```
## [1] 0.02313811
```

```
sum(na.omit(DF_Follow1$Q.Other))/length((DF_Follow1$Q.Other))
```

```
## [1] 0.01301518
```

```
sum(na.omit(DF_Follow1$Q.Other_Mixed))/length((DF_Follow1$Q.Other_Mixed))
```

```
## [1] 0.03615329
```

```
sum(is.na(DF_Follow1$Q.Race))/length(DF_Follow1$Q.Race)
```

```
## [1] 0
```

```
summary(DF_Follow1$Q.Age)
```

```
##    Min. 1st Qu.  Median    Mean 3rd Qu.    Max.    NA's
##   18.00   31.00   45.00   45.83   60.00   92.00       5
```

```
table(DF_Follow1$Q.Education)
```

```
##
##   1   2   3   4   5   6   7   8
##  11  59 339 335 156 297  28 158
```

```
sum((na.omit(DF_Follow1$Q.HighSchool)))/length((DF_Follow1$Q.HighSchool))
```

```
## [1] 0.9493854
```

```
sum((na.omit(DF_Follow1$Q.Bach)))/length((DF_Follow1$Q.Bach))
```

```
## [1] 0.3492408
```

```
sum(is.na(DF_Follow1$Q.Education))/length(DF_Follow1$Q.Education)
```

```
## [1] 0
```

```
table(DF_Follow1$Q.Income)
```

```
##
##   1   2   3   4   5   6   7   8
## 249 327 257 186 126  84  50 104
```

```
length(na.omit(DF_Follow1$Q.Income[DF_Follow1$Q.Income==1]))/length(DF_Follow1$Q.Income)
```

```
## [1] 0.1800434
```

```
length(na.omit(DF_Follow1$Q.Income[DF_Follow1$Q.Income==2]))/length(DF_Follow1$Q.Income)
```

```
## [1] 0.2364425
```

```
length(na.omit(DF_Follow1$Q.Income[DF_Follow1$Q.Income==3]))/length(DF_Follow1$Q.Income)
```

```
## [1] 0.1858279
```

```
length(na.omit(DF_Follow1$Q.Income[DF_Follow1$Q.Income==4]))/length(DF_Follow1$Q.Income)
```

```
## [1] 0.1344902
```

```
length(na.omit(DF_Follow1$Q.Income[DF_Follow1$Q.Income==5]))/length(DF_Follow1$Q.Income)
```

```
## [1] 0.09110629
```

```
length(na.omit(DF_Follow1$Q.Income[DF_Follow1$Q.Income==6]))/length(DF_Follow1$Q.Income)
```

```
## [1] 0.06073753
```

```
length(na.omit(DF_Follow1$Q.Income[DF_Follow1$Q.Income==7]))/length(DF_Follow1$Q.Income)
```

```
## [1] 0.03615329
```

```
length(na.omit(DF_Follow1$Q.Income[DF_Follow1$Q.Income==8]))/length(DF_Follow1$Q.Income)
```

```
## [1] 0.07519884
```

```
sum(is.na(DF_Follow1$Q.Income))/length(DF_Follow1$Q.Income)
```

```
## [1] 0
```

```
table(DF_Follow1$Q.Political_ID)
```

```
##
##   1   2   3   4   5
## 142 260 598 252 131
```

```
sum(na.omit(DF_Follow1$Q.Liberal))/length((DF_Follow1$Q.Liberal))
```

```
## [1] 0.2906725
```

```
sum(na.omit(DF_Follow1$Q.Moderate))/length((DF_Follow1$Q.Moderate))
```

```
## [1] 0.4323933
```

```
sum(na.omit(DF_Follow1$Q.Conservative))/length((DF_Follow1$Q.Conservative))
```

```
## [1] 0.2769342
```

```
sum(is.na(DF_Follow1$Q.Political_ID))/length(DF_Follow1$Q.Political_ID)
```

```
## [1] 0
```

```
table(DF_Follow1$Q.Veteran)
```

```
##
##    0    1
## 1220  163
```

```
length(na.omit(DF_Follow1$Q.Veteran[DF_Follow1$Q.Veteran==0]))/length(DF_Follow1$Q.Veteran)
```

```
## [1] 0.8821403
```

```
length(na.omit(DF_Follow1$Q.Veteran[DF_Follow1$Q.Veteran==1]))/length(DF_Follow1$Q.Veteran)
```

```
## [1] 0.1178597
```

```
sum(is.na(DF_Follow1$Q.Veteran))/length(DF_Follow1$Q.Veteran)
```

```
## [1] 0
```

```
table(DF_Follow1$Q.Twitter_Use )
```

```
##
##   1   2   3   4   5
## 765 279 139 130  70
```

```
length(na.omit(DF_Follow1$Q.Twitter_Use[DF_Follow1$Q.Twitter_Use==1]))/length(DF_Follow1$Q.Twitter_Use)
```

```
## [1] 0.5531453
```

```
length(na.omit(DF_Follow1$Q.Twitter_Use[DF_Follow1$Q.Twitter_Use==2]))/length(DF_Follow1$Q.Twitter_Use)
```

```
## [1] 0.2017354
```

```
length(na.omit(DF_Follow1$Q.Twitter_Use[DF_Follow1$Q.Twitter_Use==3]))/length(DF_Follow1$Q.Twitter_Use)
```

```
## [1] 0.1005061
```

```
length(na.omit(DF_Follow1$Q.Twitter_Use[DF_Follow1$Q.Twitter_Use==4]))/length(DF_Follow1$Q.Twitter_Use)
```

```
## [1] 0.09399855
```

```
length(na.omit(DF_Follow1$Q.Twitter_Use[DF_Follow1$Q.Twitter_Use==5]))/length(DF_Follow1$Q.Twitter_Use)
```

```
## [1] 0.05061461
```

```
sum(is.na(DF_Follow1$Q.Twitter_Use))/length(DF_Follow1$Q.Twitter_Use)
```

```
## [1] 0
```

Analysis

Mean Estimates with Standard Error

```
Cred_Full_1 <- Credibility ~ 0 + as.factor(TweetMedium)
lm_Cred_Full_1 <- lm(Cred_Full_1,
                     data = DF_Follow1)

summary(lm_Cred_Full_1)
```

```
##
## Call:
## lm(formula = Cred_Full_1, data = DF_Follow1)
```

```
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.38559 -1.03556 -0.01322  0.98678  1.99565
##
## Coefficients:
##                          Estimate Std. Error t value Pr(>|t|)
## as.factor(TweetMedium)1  3.00435    0.08766   34.27   <2e-16 ***
## as.factor(TweetMedium)2  3.01322    0.08824   34.15   <2e-16 ***
## as.factor(TweetMedium)3  3.03556    0.08863   34.25   <2e-16 ***
## as.factor(TweetMedium)4  3.38559    0.08654   39.12   <2e-16 ***
## as.factor(TweetMedium)5  3.30603    0.08728   37.88   <2e-16 ***
## as.factor(TweetMedium)6  3.09871    0.08709   35.58   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.329 on 1377 degrees of freedom
## Multiple R-squared:  0.8491, Adjusted R-squared:  0.8484
## F-statistic:  1291 on 6 and 1377 DF,  p-value: < 2.2e-16
```

```
length(DF_Follow1$Credibility[DF_Follow1$TweetMedium==1])
```

```
## [1] 230
```

```
length(DF_Follow1$Credibility[DF_Follow1$TweetMedium==2])
```

```
## [1] 227
```

```
length(DF_Follow1$Credibility[DF_Follow1$TweetMedium==3])
```

```
## [1] 225
```

```
length(DF_Follow1$Credibility[DF_Follow1$TweetMedium==4])
```

```
## [1] 236
```

```
length(DF_Follow1$Credibility[DF_Follow1$TweetMedium==5])
```

```
## [1] 232
```

```
length(DF_Follow1$Credibility[DF_Follow1$TweetMedium==6])
```

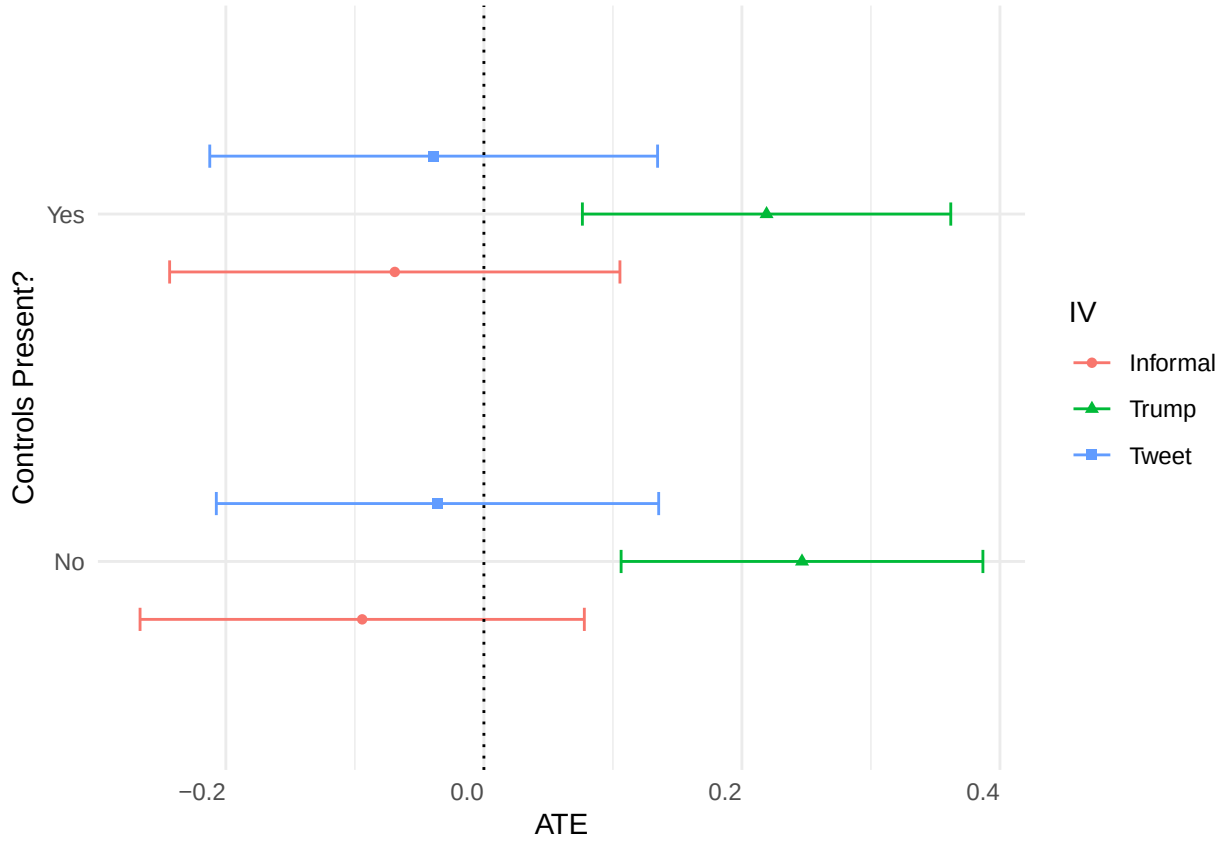
```
## [1] 233
```

```
#time for some regressions


## Model 1: No Demographics

Cred_Form_1 <- Credibility ~ Tweet + Informal + Trump

## Model 5: Factor Demographics

Cred_Form_5 <- Credibility ~ Tweet + Informal + Trump   +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
```

```
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))

## Model 7: Slightly simplified for the Appendix

Cred_Form_7 <- Credibility ~ Tweet + Informal + Trump  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some


#regression



Cred_1 <- lm(Cred_Form_1,
             data = DF_Follow1,
             na.action=na.omit)
Cred_5 <- lm(Cred_Form_5,
             data = DF_Follow1,
             na.action=na.omit)
Cred_7 <- lm(Cred_Form_7,
             data = DF_Follow1,
             na.action=na.omit)


stargazer(Cred_1, Cred_5, Cred_7, title = "Credibility, Follow-On", no.space = TRUE)
```

```
#graph time


Cred_DF <- as.data.frame(matrix(data = c("Tweet", coeftest(Cred_5)[2, 1:2],"Yes",
                                         "Tweet", coeftest(Cred_1)[2, 1:2], "No",
                                         "Informal", coeftest(Cred_5)[3, 1:2],"Yes",
                                         "Informal", coeftest(Cred_1)[3, 1:2], "No",
                                         "Trump", coeftest(Cred_5)[4, 1:2],"Yes",
                                         "Trump", coeftest(Cred_1)[4, 1:2], "No"),
                                        ncol = 4, byrow = TRUE))


colnames(Cred_DF) <- c("IV", "estimate", "se", "controls")


#making into correct operators
Cred_DF$estimate <- as.numeric(Cred_DF$estimate)
Cred_DF$se <- as.numeric(Cred_DF$se)


#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))

Cred_DF$ci <- Cred_DF$se*q
```

```
library(ggplot2)
## graph time
pd <- position_dodge(0.5)

## cred graph
ggplot(Cred_DF, aes(x = controls, y = estimate, color = IV))  +
  geom_point(aes(color=IV, shape=IV), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Controls Present?") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1)) + coord_flip()
```




```
#ggtitle("ATE on Credibility, all IVs")
```

Manipulation Check

```
##### stargazer tables for MC
#### note: this version has removed the "backsdown" binary variable

# Threat/Treatment Check

## Model 1: No Demographics

Threat_Form_1 <- Threat_Check_Binary ~ Tweet + Informal + Trump


## Model 5: Factor Demographics
```

```
Threat_Form_5 <- Threat_Check_Binary ~ Tweet + Informal + Trump   +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))




## Model 7: Slightly simplified for the Appendix

Threat_Form_7 <- Threat_Check_Binary ~ Tweet + Informal + Trump  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Threat_1 <- lm(Threat_Form_1,
               data = DF_Follow1,
               na.action=na.omit)

Threat_5 <- lm(Threat_Form_5,
               data = DF_Follow1,
               na.action=na.omit)

Threat_7 <- lm(Threat_Form_7,
               data = DF_Follow1,
               na.action=na.omit)


stargazer(Threat_1, Threat_5, Threat_7, title = "Treatment Check Question with Covariates", no.space =
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac
## % Date and time: Sun, Dec 31, 2023 - 14:16:38
## \begin{table}[!htbp] \centering
##   \caption{Treatment Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Threat\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.109$^{***}$ & $-$0.106$^{***}$ & $-$0.106$^{***}$ \\
##   & (0.032) & (0.032) & (0.032) \\
```

```
##   Informal & $-$0.023 & $-$0.037 & $-$0.035 \\
##   & (0.032) & (0.032) & (0.032) \\
##   Trump & 0.025 & 0.022 & 0.022 \\
##   & (0.026) & (0.026) & (0.026) \\
##   Q.Female &  & $-$0.044$^{*}$ & $-$0.040 \\
##   &  & (0.027) & (0.027) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.135$^{***}$ & $-$0.134$^{***}$ \\
##   &  & (0.044) & (0.044) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & $-$0.005 & $-$0.019 \\
##   &  & (0.146) & (0.146) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & $-$0.184$^{***}$ & $-$0.190$^{***}$ \\
##   &  & (0.061) & (0.061) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & $-$0.362 & $-$0.393 \\
##   &  & (0.278) & (0.278) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & $-$0.026 & $-$0.028 \\
##   &  & (0.050) & (0.050) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & $-$0.052 & $-$0.071 \\
##   &  & (0.089) & (0.088) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.014 & 0.006 \\
##   &  & (0.122) & (0.121) \\
##   Q.Age &  & $-$0.001 & $-$0.001 \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.038 &  \\
##   &  & (0.062) &  \\
##   Q.Bach &  & 0.079$^{**}$ & 0.076$^{**}$ \\
##   &  & (0.031) & (0.031) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & $-$0.030 &  \\
##   &  & (0.048) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & $-$0.047 &  \\
##   &  & (0.040) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.001 &  \\
##   &  & (0.044) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & 0.069 &
##   &  & (0.047) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & 0.077$
##   &  & (0.047) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & 0.080$
##   &  & (0.044) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & 0.024 &
##   &  & (0.041) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.010 &  \\
##   &  & (0.041) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.013 &  \\
##   &  & (0.035) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.029 &  \\
##   &  & (0.033) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.036
##   &  & (0.026) &  \\
##   Q.Income &  &  & 0.0004 \\
##   &  &  & (0.007) \\
##   Q.Conservative &  &  & 0.013 \\
##   &  &  & (0.033) \\
##   Q.Liberal &  &  & 0.043 \\
##   &  &  & (0.032) \\
```

```
##   Q.Veteran &  & $-$0.097$^{**}$ & $-$0.096$^{**}$ \\
##   &  & (0.041) & (0.041) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.009 &  \\
##   &  & (0.043) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & 0.061 &  \\
##   &  & (0.042) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.110$^{***}$ &  \\
##   &  & (0.038) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.024 &
##   &  & (0.040) &  \\
##   Q.Twitter\_Some &  &  & 0.011 \\
##   &  &  & (0.027) \\
##   Constant & 0.687$^{***}$ & 0.737$^{***}$ & 0.751$^{***}$ \\
##   & (0.026) & (0.076) & (0.057) \\
##  \hline \\[-1.8ex]
## Observations & 1,383 & 1,368 & 1,368 \\
## R$^{2}$ & 0.015 & 0.054 & 0.041 \\
## Adjusted R$^{2}$ & 0.013 & 0.033 & 0.028 \\
## Residual Std. Error & 0.483 (df = 1379) & 0.477 (df = 1337) & 0.478 (df = 1349) \\
## F Statistic & 6.903$^{***}$ (df = 3; 1379) & 2.544$^{***}$ (df = 30; 1337) & 3.221$^{***}$ (df = 18;
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```r
# President Check

## Model 1: No Demographics

President_Form_1 <- President_Check_Binary ~ Tweet + Informal + Trump


## Model 5: Factor Demographics

President_Form_5 <- President_Check_Binary ~ Tweet + Informal + Trump   +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))




## Model 7: Slightly simplified for the Appendix

President_Form_7 <- President_Check_Binary ~ Tweet + Informal + Trump  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some
```

```r
#regression

President_1 <- lm(President_Form_1,
             data = DF_Follow1,
             na.action=na.omit)

President_5 <- lm(President_Form_5,
             data = DF_Follow1,
             na.action=na.omit)

President_7 <- lm(President_Form_7,
             data = DF_Follow1,
             na.action=na.omit)


stargazer(President_1, President_5, President_7, title = "President Check Question with Covariates", no
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:38
## \begin{table}[!htbp] \centering
##   \caption{President Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{President\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & 0.044$^{**}$ & 0.041$^{**}$ & 0.045$^{**}$ \\
##   & (0.018) & (0.018) & (0.018) \\
##   Informal & $-$0.013 & $-$0.017 & $-$0.020 \\
##   & (0.018) & (0.018) & (0.018) \\
##   Trump & 0.027$^{*}$ & 0.023 & 0.024 \\
##   & (0.015) & (0.015) & (0.015) \\
##   Q.Female &  & $-$0.026$^{*}$ & $-$0.024 \\
##   &  & (0.015) & (0.015) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  &  & $-$0.087$^{***}$ & $-$0.085$^{***}$ \\
##   &  & (0.025) & (0.025) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  &  & $-$0.006 & $-$0.017 \\
##   &  & (0.083) & (0.083) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  &  & $-$0.038 & $-$0.036 \\
##   &  & (0.035) & (0.035) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  &  & $-$0.616$^{***}$ & $-$0.621$^{***}$ \\
##   &  & (0.159) & (0.159) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  &  & 0.007 & 0.003 \\
##   &  & (0.029) & (0.029) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  &  & $-$0.046 & $-$0.045 \\
##   &  & (0.050) & (0.050) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  &  & $-$0.020 & $-$0.005 \\
##   &  & (0.069) & (0.069) \\
```

```
##   Q.Age &  & $-$0.0002 & 0.00000 \\
##   &  & (0.0005) & (0.0005) \\
##   Q.HighSchool &  & 0.077$^{**}$ &  \\
##   &  & (0.035) &  \\
##   Q.Bach &  & 0.039$^{**}$ & 0.041$^{**}$ \\
##   &  & (0.018) & (0.018) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & 0.036 &  \\
##   &  & (0.027) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & 0.003 &  \\
##   &  & (0.023) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.010 &  \\
##   &  & (0.025) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & $-$0.0
##   &  & (0.027) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & $-$0.0
##   &  & (0.027) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & $-$0.0
##   &  & (0.025) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & 0.043$
##   &  & (0.023) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 0.012 &  \\
##   &  & (0.023) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.017 &  \\
##   &  & (0.020) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.012 &  \\
##   &  & (0.019) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.009
##   &  & (0.015) &  \\
##   Q.Income &  &  & 0.005 \\
##   &  &  & (0.004) \\
##   Q.Conservative &  &  & 0.002 \\
##   &  &  & (0.019) \\
##   Q.Liberal &  &  & $-$0.002 \\
##   &  &  & (0.018) \\
##   Q.Veteran &  & $-$0.020 & $-$0.020 \\
##   &  & (0.023) & (0.023) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.052$^{**}$ &  \\
##   &  & (0.024) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.026 &  \\
##   &  & (0.024) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.029 &  \\
##   &  & (0.022) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & 0.025 &
##   &  & (0.023) &  \\
##   Q.Twitter\_Some &  &  & $-$0.005 \\
##   &  &  & (0.016) \\
##   Constant & 0.877$^{***}$ & 0.825$^{***}$ & 0.881$^{***}$ \\
##   & (0.015) & (0.043) & (0.033) \\
##  \hline \\[-1.8ex]
## Observations & 1,383 & 1,368 & 1,368 \\
## R$^{2}$ & 0.007 & 0.055 & 0.038 \\
## Adjusted R$^{2}$ & 0.005 & 0.034 & 0.025 \\
## Residual Std. Error & 0.278 (df = 1379) & 0.272 (df = 1337) & 0.273 (df = 1349) \\
## F Statistic & 3.131$^{**}$ (df = 3; 1379) & 2.589$^{***}$ (df = 30; 1337) & 2.973$^{***}$ (df = 18;
```

```
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```
# Target Check

## Model 1: No Demographics

Target_Form_1 <- Target_Check_Binary ~ Tweet + Informal + Trump


## Model 5: Factor Demographics

Target_Form_5 <- Target_Check_Binary ~ Tweet + Informal + Trump   +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))




## Model 7: Slightly simplified for the Appendix

Target_Form_7 <- Target_Check_Binary ~ Tweet + Informal + Trump  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Target_1 <- lm(Target_Form_1,
               data = DF_Follow1,
               na.action=na.omit)

Target_5 <- lm(Target_Form_5,
               data = DF_Follow1,
               na.action=na.omit)

Target_7 <- lm(Target_Form_7,
               data = DF_Follow1,
               na.action=na.omit)


stargazer(Target_1, Target_5, Target_7, title = "Target Check Question with Covariates", no.space = TRU
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:38
```

```
## \begin{table}[!htbp] \centering
##   \caption{Target Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Target\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & 0.033 & 0.028 & 0.036 \\
##   & (0.033) & (0.033) & (0.032) \\
##   Informal & $-$0.123$^{***}$ & $-$0.130$^{***}$ & $-$0.133$^{***}$ \\
##   & (0.033) & (0.033) & (0.033) \\
##   Trump & $-$0.035 & $-$0.049$^{*}$ & $-$0.049$^{*}$ \\
##   & (0.027) & (0.027) & (0.027) \\
##   Q.Female &  & $-$0.026 & $-$0.024 \\
##   &  & (0.027) & (0.027) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.158$^{***}$ & $-$0.152$^{***}$ \\
##   &  & (0.045) & (0.045) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & 0.120 & 0.103 \\
##   &  & (0.149) & (0.149) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & 0.011 & 0.010 \\
##   &  & (0.062) & (0.062) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & $-$0.528$^{*}$ & $-$0.517$^{*}$ \\
##   &  & (0.284) & (0.285) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & 0.084 & 0.077 \\
##   &  & (0.051) & (0.051) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & $-$0.022 & $-$0.025 \\
##   &  & (0.091) & (0.090) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.111 & 0.118 \\
##   &  & (0.124) & (0.124) \\
##   Q.Age &  & 0.002$^{**}$ & 0.002$^{***}$ \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.135$^{**}$ &  \\
##   &  & (0.063) &  \\
##   Q.Bach &  & 0.019 & 0.025 \\
##   &  & (0.032) & (0.031) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & 0.028 &  \\
##   &  & (0.049) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & 0.058 &  \\
##   &  & (0.041) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.040 &  \\
##   &  & (0.045) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & $-$0.0
##   &  & (0.048) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & 0.037 &
##   &  & (0.048) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & $-$0.0
##   &  & (0.045) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & 0.047 &
##   &  & (0.042) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.056 &  \\
```

```
##   &  & (0.042) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.019 &  \\
##   &  & (0.036) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.004 &  \\
##   &  & (0.033) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.020
##   &  & (0.027) &  \\
##   Q.Income &  &  & 0.002 \\
##   &  &  & (0.007) \\
##   Q.Conservative &  &  & $-$0.014 \\
##   &  &  & (0.033) \\
##   Q.Liberal &  &  & 0.030 \\
##   &  &  & (0.032) \\
##   Q.Veteran &  & 0.054 & 0.048 \\
##   &  & (0.042) & (0.042) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.104$^{**}$ &  \\
##   &  & (0.044) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.010 &  \\
##   &  & (0.043) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.048 &  \\
##   &  & (0.039) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.010 &
##   &  & (0.040) &  \\
##   Q.Twitter\_Some &  &  & $-$0.033 \\
##   &  &  & (0.028) \\
##   Constant & 0.595$^{***}$ & 0.368$^{***}$ & 0.508$^{***}$ \\
##   & (0.027) & (0.078) & (0.059) \\
##  \hline \\[-1.8ex]
## Observations & 1,383 & 1,368 & 1,368 \\
## R$^{2}$ & 0.012 & 0.056 & 0.042 \\
## Adjusted R$^{2}$ & 0.010 & 0.035 & 0.029 \\
## Residual Std. Error & 0.494 (df = 1379) & 0.488 (df = 1337) & 0.489 (df = 1349) \\
## F Statistic & 5.677$^{***}$ (df = 3; 1379) & 2.666$^{***}$ (df = 30; 1337) & 3.245$^{***}$ (df = 18;
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```r
# Support Check

## Model 1: No Demographics

Support_Form_1 <- Support_Check_Binary ~ Tweet + Informal + Trump


## Model 5: Factor Demographics

Support_Form_5 <- Support_Check_Binary ~ Tweet + Informal + Trump   +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
```

```
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))




## Model 7: Slightly simplified for the Appendix

Support_Form_7 <- Support_Check_Binary ~ Tweet + Informal + Trump  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Support_1 <- lm(Support_Form_1,
             data = DF_Follow1,
             na.action=na.omit)

Support_5 <- lm(Support_Form_5,
             data = DF_Follow1,
             na.action=na.omit)

Support_7 <- lm(Support_Form_7,
             data = DF_Follow1,
             na.action=na.omit)


stargazer(Support_1, Support_5, Support_7, title = "Support Check Question with Covariates", no.space =
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:38
## \begin{table}[!htbp] \centering
##   \caption{Support Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Support\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.016 & $-$0.015 & $-$0.012 \\
##   & (0.032) & (0.031) & (0.031) \\
##   Informal & 0.025 & 0.013 & 0.011 \\
##   & (0.032) & (0.031) & (0.031) \\
##   Trump & $-$0.006 & $-$0.022 & $-$0.019 \\
##   & (0.026) & (0.025) & (0.025) \\
##   Q.Female &  & $-$0.147$^{***}$ & $-$0.146$^{***}$ \\
##   &  & (0.026) & (0.026) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.039 & $-$0.036 \\
```

```
##   &  & (0.043) & (0.043) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & 0.066 & 0.067 \\
##   &  & (0.142) & (0.141) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & $-$0.050 & $-$0.041 \\
##   &  & (0.059) & (0.059) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & 0.308 & 0.289 \\
##   &  & (0.270) & (0.269) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & $-$0.031 & $-$0.034 \\
##   &  & (0.049) & (0.048) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & $-$0.056 & $-$0.056 \\
##   &  & (0.086) & (0.085) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.003 & $-$0.009 \\
##   &  & (0.118) & (0.117) \\
##   Q.Age &  & 0.003$^{***}$ & 0.003$^{***}$ \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.047 &  \\
##   &  & (0.060) &  \\
##   Q.Bach &  & 0.159$^{***}$ & 0.165$^{***}$ \\
##   &  & (0.030) & (0.030) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & $-$0.013 &  \\
##   &  & (0.047) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & $-$0.017 &  \\
##   &  & (0.039) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & $-$0.041 &  \\
##   &  & (0.043) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & $-$0.0
##   &  & (0.045) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & 0.044 &
##   &  & (0.045) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & 0.005 &
##   &  & (0.043) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & 0.066$
##   &  & (0.040) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.067$^{*}$ &  \\
##   &  & (0.040) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & 0.026 &  \\
##   &  & (0.034) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.028 &  \\
##   &  & (0.032) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.052
##   &  & (0.025) &  \\
##   Q.Income &  &  & $-$0.002 \\
##   &  &  & (0.007) \\
##   Q.Conservative &  &  & 0.048 \\
##   &  &  & (0.032) \\
##   Q.Liberal &  &  & 0.088$^{***}$ \\
##   &  &  & (0.031) \\
##   Q.Veteran &  & 0.032 & 0.031 \\
##   &  & (0.040) & (0.040) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.002 &  \\
##   &  & (0.041) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & 0.019 &  \\
##   &  & (0.041) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.018 &  \\
```

```
##   &  & (0.037) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & 0.012 &
##   &  & (0.038) &  \\
##   Q.Twitter\_Some &  &  & $-$0.008 \\
##   &  &  & (0.027) \\
##   Constant & 0.649$^{***}$ & 0.533$^{***}$ & 0.533$^{***}$ \\
##   & (0.026) & (0.074) & (0.055) \\
##  \hline \\[-1.8ex]
## Observations & 1,383 & 1,368 & 1,368 \\
## R$^{2}$ & 0.001 & 0.084 & 0.078 \\
## Adjusted R$^{2}$ & $-$0.002 & 0.063 & 0.066 \\
## Residual Std. Error & 0.480 (df = 1379) & 0.463 (df = 1337) & 0.463 (df = 1349) \\
## F Statistic & 0.239 (df = 3; 1379) & 4.081$^{***}$ (df = 30; 1337) & 6.324$^{***}$ (df = 18; 1349) \
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

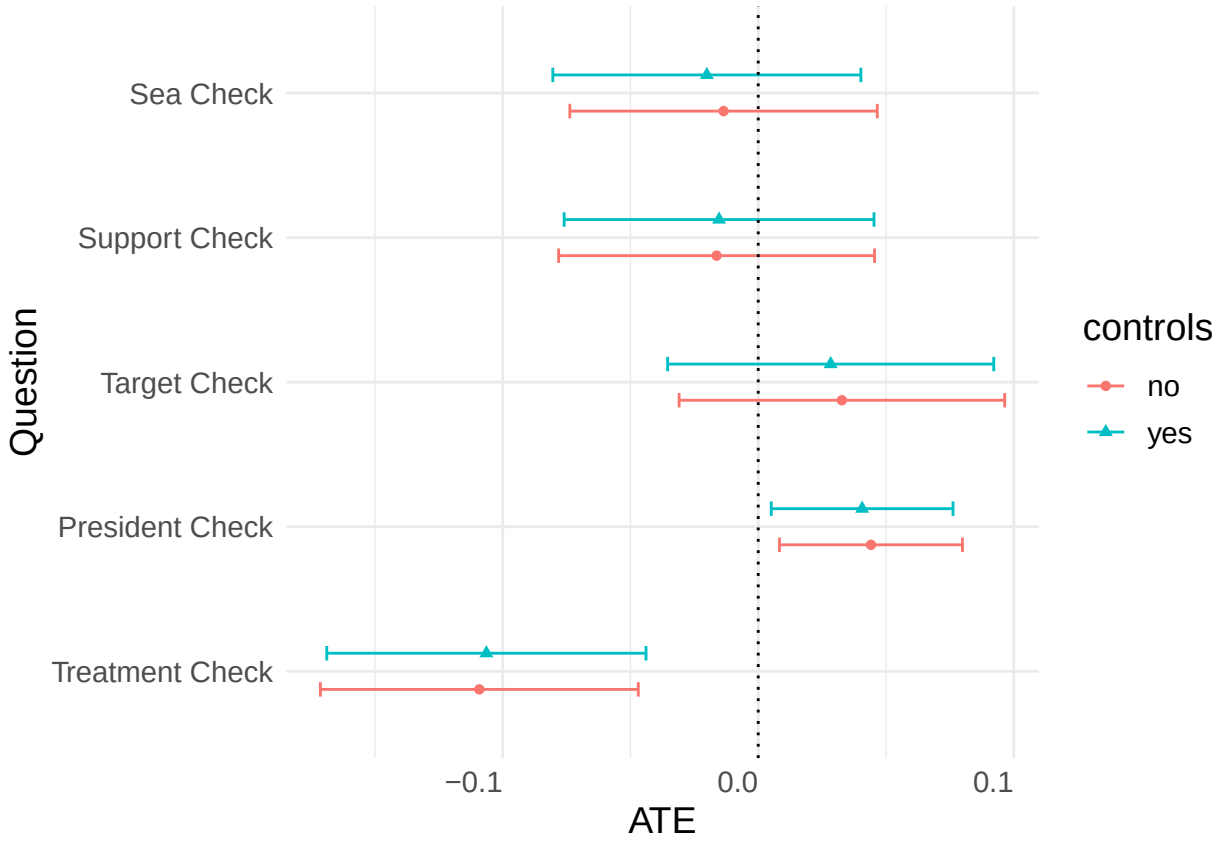
```r
# Sea Check

## Model 1: No Demographics

Sea_Form_1 <- Sea_Check_Binary ~ Tweet + Informal + Trump


## Model 5: Factor Demographics

Sea_Form_5 <- Sea_Check_Binary ~ Tweet + Informal + Trump   +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))




## Model 7: Slightly simplified for the Appendix

Sea_Form_7 <- Sea_Check_Binary ~ Tweet + Informal + Trump  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Sea_1 <- lm(Sea_Form_1,
            data = DF_Follow1,
            na.action=na.omit)
```

```r
Sea_5 <- lm(Sea_Form_5,
            data = DF_Follow1,
            na.action=na.omit)

Sea_7 <- lm(Sea_Form_7,
            data = DF_Follow1,
            na.action=na.omit)


stargazer(Sea_1, Sea_5, Sea_7, title = "Sea Check Question with Covariates", no.space = TRUE)
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:39
## \begin{table}[!htbp] \centering
##   \caption{Sea Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Sea\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.014 & $-$0.020 & $-$0.017 \\
##   & (0.031) & (0.031) & (0.031) \\
##   Informal & 0.028 & 0.029 & 0.028 \\
##   & (0.031) & (0.031) & (0.031) \\
##   Trump & 0.026 & 0.028 & 0.023 \\
##   & (0.025) & (0.025) & (0.025) \\
##   Q.Female &  & $-$0.007 & $-$0.006 \\
##   &  & (0.026) & (0.026) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.101$^{**}$ & $-$0.097$^{**}$ \\
##   &  & (0.042) & (0.042) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & 0.173 & 0.163 \\
##   &  & (0.141) & (0.141) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & $-$0.082 & $-$0.092 \\
##   &  & (0.059) & (0.059) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & $-$0.427 & $-$0.407 \\
##   &  & (0.269) & (0.269) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & 0.031 & 0.036 \\
##   &  & (0.049) & (0.048) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & 0.054 & 0.050 \\
##   &  & (0.086) & (0.085) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & $-$0.131 & $-$0.133 \\
##   &  & (0.118) & (0.117) \\
##   Q.Age &  & $-$0.004$^{***}$ & $-$0.004$^{***}$ \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.105$^{*}$ &  \\
##   &  & (0.060) &  \\
##   Q.Bach &  & $-$0.020 & $-$0.014 \\
##   &  & (0.030) & (0.030) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & $-$0.044 &  \\
```

```
##   &  & (0.046) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & 0.010 &  \\
##   &  & (0.039) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.079$^{*}$ &  \\
##   &  & (0.042) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & 0.055 &
##   &  & (0.045) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & 0.053 &
##   &  & (0.045) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & 0.055 &
##   &  & (0.043) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & 0.006 &
##   &  & (0.039) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.010 &  \\
##   &  & (0.040) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.057$^{*}$ &  \\
##   &  & (0.034) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.045 &  \\
##   &  & (0.032) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.013
##   &  & (0.025) &  \\
##   Q.Income &  &  & $-$0.003 \\
##   &  &  & (0.007) \\
##   Q.Conservative &  &  & $-$0.008 \\
##   &  &  & (0.031) \\
##   Q.Liberal &  &  & $-$0.026 \\
##   &  &  & (0.030) \\
##   Q.Veteran &  & $-$0.081$^{**}$ & $-$0.083$^{**}$ \\
##   &  & (0.040) & (0.040) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.028 &  \\
##   &  & (0.041) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & 0.048 &  \\
##   &  & (0.040) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.066$^{*}$ &  \\
##   &  & (0.037) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.027 &
##   &  & (0.038) &  \\
##   Q.Twitter\_Some &  &  & $-$0.015 \\
##   &  &  & (0.026) \\
##   Constant & 0.667$^{***}$ & 0.736$^{***}$ & 0.887$^{***}$ \\
##   & (0.025) & (0.074) & (0.055) \\
##  \hline \\[-1.8ex]
## Observations & 1,383 & 1,368 & 1,368 \\
## R$^{2}$ & 0.001 & 0.046 & 0.032 \\
## Adjusted R$^{2}$ & $-$0.001 & 0.025 & 0.019 \\
## Residual Std. Error & 0.467 (df = 1379) & 0.461 (df = 1337) & 0.462 (df = 1349) \\
## F Statistic & 0.628 (df = 3; 1379) & 2.160$^{***}$ (df = 30; 1337) & 2.497$^{***}$ (df = 18; 1349) \
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

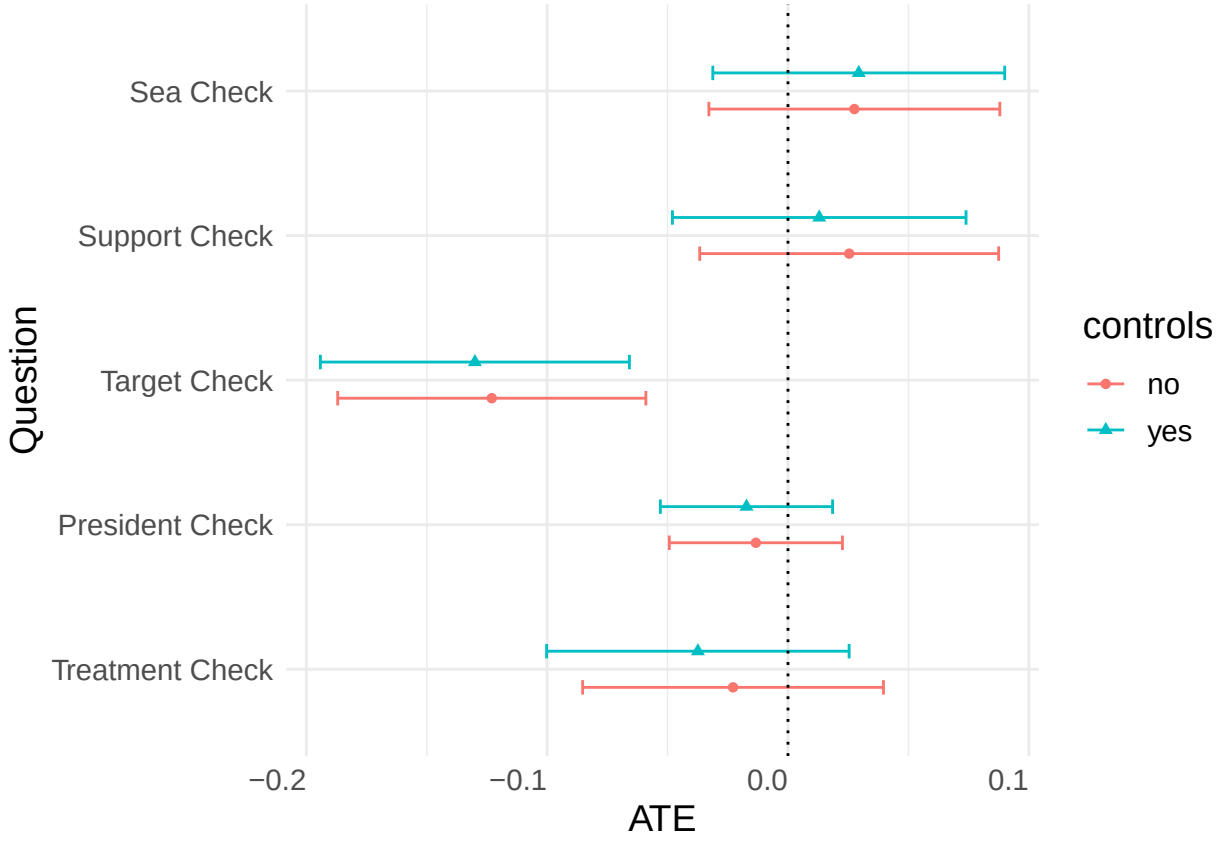
```r
## more detail on manipulation check

## ATE of Tweet Treatment

#put all the names, estimates, SEs, and behavior into one df for graphing
Results_DF <- as.data.frame(matrix(data = c("Treatment Check", coeftest(Threat_1)[2, 1:2],"no",
                                    "Treatment Check", coeftest(Threat_5)[2, 1:2], "yes",
                                    "President Check", coeftest(President_1)[2, 1:2],"no",
                                    "President Check", coeftest(President_5)[2, 1:2], "yes",
                                    "Target Check", coeftest(Target_1)[2, 1:2],"no",
                                    "Target Check", coeftest(Target_5)[2, 1:2], "yes",
                                    "Support Check", coeftest(Support_1)[2, 1:2], "no",
                                    "Support Check", coeftest(Support_5)[2, 1:2], "yes",
                                    "Sea Check", coeftest(Sea_1)[2, 1:2], "no",
                                    "Sea Check", coeftest(Sea_5)[2, 1:2], "yes"),
                                    ncol = 4, byrow = TRUE))

colnames(Results_DF) <- c("dv", "estimate", "se", "controls")

#making into correct operators
Results_DF$dv <- factor(Results_DF$dv, levels = c("Treatment Check", "President Check", "Target Check",
Results_DF$estimate <- as.numeric(Results_DF$estimate)
Results_DF$se <- as.numeric(Results_DF$se)

#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Results_DF$ci <- Results_DF$se*q


## graph time
library(ggplot2)
pd <- position_dodge(0.5)

ggplot(Results_DF, aes(x = dv, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Question") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```
#ggtitle("Tweet Imagery ATE by MC or AC Question") +
```

```
## more detail on manipulation check

## ATE of Informal Language Treatment

#put all the names, estimates, SEs, and behavior into one df for graphing
Results_DF <- as.data.frame(matrix(data = c("Treatment Check", coeftest(Threat_1)[3, 1:2],"no",
                                    "Treatment Check", coeftest(Threat_5)[3, 1:2], "yes",
                                    "President Check", coeftest(President_1)[3, 1:2],"no",
                                    "President Check", coeftest(President_5)[3, 1:2], "yes",
                                    "Target Check", coeftest(Target_1)[3, 1:2],"no",
                                    "Target Check", coeftest(Target_5)[3, 1:2], "yes",
                                    "Support Check", coeftest(Support_1)[3, 1:2], "no",
                                    "Support Check", coeftest(Support_5)[3, 1:2], "yes",
                                    "Sea Check", coeftest(Sea_1)[3, 1:2], "no",
                                    "Sea Check", coeftest(Sea_5)[3, 1:2], "yes"),
                                    ncol = 4, byrow = TRUE))

colnames(Results_DF) <- c("dv", "estimate", "se", "controls")

#making into correct operators
Results_DF$dv <- factor(Results_DF$dv, levels = c("Treatment Check", "President Check", "Target Check",
Results_DF$estimate <- as.numeric(Results_DF$estimate)
Results_DF$se <- as.numeric(Results_DF$se)

#adding in CIs
```

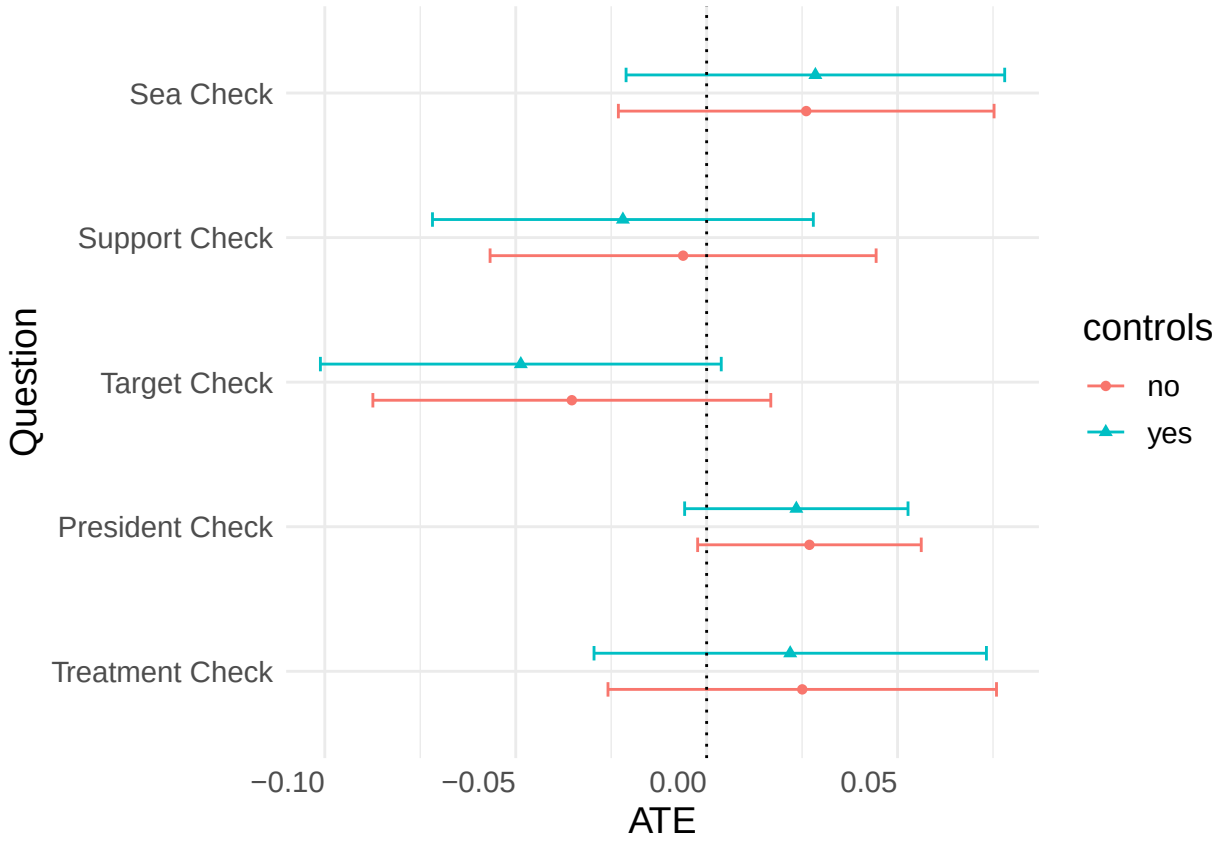
```
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Results_DF$ci <- Results_DF$se*q


## graph time
library(ggplot2)
pd <- position_dodge(0.5)

ggplot(Results_DF, aes(x = dv, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Question") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```
#ggtitle("Informal Language ATE by MC or AC Question") +
```

```
## more detail on manipulation check

## ATE of Trump as President Treatment

#put all the names, estimates, SEs, and behavior into one df for graphing
Results_DF <- as.data.frame(matrix(data = c("Treatment Check", coeftest(Threat_1)[4, 1:2],"no",
                                    "Treatment Check", coeftest(Threat_5)[4, 1:2], "yes",
                                    "President Check", coeftest(President_1)[4, 1:2],"no",
                                    "President Check", coeftest(President_5)[4, 1:2], "yes",
                                    "Target Check", coeftest(Target_1)[4, 1:2],"no",
```

```
                                    "Target Check", coeftest(Target_5)[4, 1:2], "yes",
                                    "Support Check", coeftest(Support_1)[4, 1:2], "no",
                                    "Support Check", coeftest(Support_5)[4, 1:2], "yes",
                                    "Sea Check", coeftest(Sea_1)[4, 1:2], "no",
                                    "Sea Check", coeftest(Sea_5)[4, 1:2], "yes"),
                                    ncol = 4, byrow = TRUE))

colnames(Results_DF) <- c("dv", "estimate", "se", "controls")

#making into correct operators
Results_DF$dv <- factor(Results_DF$dv, levels = c("Treatment Check", "President Check", "Target Check",
Results_DF$estimate <- as.numeric(Results_DF$estimate)
Results_DF$se <- as.numeric(Results_DF$se)

#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Results_DF$ci <- Results_DF$se*q


## graph time
library(ggplot2)
pd <- position_dodge(0.5)

ggplot(Results_DF, aes(x = dv, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Question") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```
#ggtitle("Trump as President ATE by MC or AC Question") +
```

```
## mean and se tables for threat check

####Threat/Treatment Check
Threat_Full_1 <- Threat_Check_Binary ~ 0 + as.factor(TweetMedium)


lm_Threat_Full_1 <- lm(Threat_Full_1,
                       data = DF_Follow1)

summary(lm_Threat_Full_1)
```

Mean and SEs

```
##
## Call:
## lm(formula = Threat_Full_1, data = DF_Follow1)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.7609 -0.5771  0.2391  0.3966  0.5200
##
## Coefficients:
##                           Estimate Std. Error t value Pr(>|t|)
## as.factor(TweetMedium)1  0.76087    0.03159   24.08   <2e-16 ***
## as.factor(TweetMedium)2  0.57709    0.03180   18.15   <2e-16 ***
```

```
## as.factor(TweetMedium)3  0.48000    0.03194   15.03   <2e-16 ***
## as.factor(TweetMedium)4  0.63983    0.03119   20.51   <2e-16 ***
## as.factor(TweetMedium)5  0.60345    0.03146   19.18   <2e-16 ***
## as.factor(TweetMedium)6  0.65236    0.03139   20.78   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4792 on 1377 degrees of freedom
## Multiple R-squared:  0.6311, Adjusted R-squared:  0.6295
## F-statistic: 392.6 on 6 and 1377 DF,  p-value: < 2.2e-16
```

```
length(DF_Follow1$Threat_Check_Binary[DF_Follow1$TweetMedium==1]) #statement
```

```
## [1] 230
```

```
length(DF_Follow1$Threat_Check_Binary[DF_Follow1$TweetMedium==2])
```

```
## [1] 227
```

```
length(DF_Follow1$Threat_Check_Binary[DF_Follow1$TweetMedium==3])
```

```
## [1] 225
```

```
length(DF_Follow1$Threat_Check_Binary[DF_Follow1$TweetMedium==4]) #statement
```

```
## [1] 236
```

```
length(DF_Follow1$Threat_Check_Binary[DF_Follow1$TweetMedium==5])
```

```
## [1] 232
```

```
length(DF_Follow1$Threat_Check_Binary[DF_Follow1$TweetMedium==6])
```

```
## [1] 233
```

```
####Threat/Treatment Check
Threat_Full_1 <- Threat_Check_Binary ~ 0 + as.factor(Tweet)


lm_Threat_Full_1 <- lm(Threat_Full_1,
                       data = DF_Follow1)

summary(lm_Threat_Full_1)
```

```
##
## Call:
## lm(formula = Threat_Full_1, data = DF_Follow1)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6996 -0.5791  0.3004  0.4209  0.4209
##
## Coefficients:
##                    Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.69957    0.02235   31.30   <2e-16 ***
## as.factor(Tweet)1  0.57906    0.01593   36.34   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```
## Residual standard error: 0.4825 on 1381 degrees of freedom
## Multiple R-squared:  0.6249, Adjusted R-squared:  0.6244
## F-statistic:  1150 on 2 and 1381 DF,  p-value: < 2.2e-16
```

```
length(DF_Follow1$Threat_Check_Binary[DF_Follow1$Tweet==1])
```

```
## [1] 917
```

```
length(DF_Follow1$Threat_Check_Binary[DF_Follow1$Tweet==0])
```

```
## [1] 466
```

```
##### President Check
President_Full_1 <- President_Check_Binary ~ 0 + as.factor(TweetMedium)

lm_President_Full_1 <- lm(President_Full_1,
                     data = DF_Follow1)

summary(lm_President_Full_1)
```

```
##
## Call:
## lm(formula = President_Full_1, data = DF_Follow1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.94850  0.05286  0.07759  0.10667  0.13478
##
## Coefficients:
##                          Estimate Std. Error t value Pr(>|t|)
## as.factor(TweetMedium)1  0.86522    0.01829   47.31   <2e-16 ***
## as.factor(TweetMedium)2  0.94714    0.01841   51.45   <2e-16 ***
## as.factor(TweetMedium)3  0.89333    0.01849   48.32   <2e-16 ***
## as.factor(TweetMedium)4  0.91525    0.01805   50.70   <2e-16 ***
## as.factor(TweetMedium)5  0.92241    0.01821   50.66   <2e-16 ***
## as.factor(TweetMedium)6  0.94850    0.01817   52.20   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2773 on 1377 degrees of freedom
## Multiple R-squared:  0.9163, Adjusted R-squared:  0.916
## F-statistic:  2514 on 6 and 1377 DF,  p-value: < 2.2e-16
```

```
length(DF_Follow1$President_Check_Binary[DF_Follow1$Tweet==0])
```

```
## [1] 466
```

```
length(DF_Follow1$President_Check_Binary[DF_Follow1$Tweet==1])
```

```
## [1] 917
```

```
##### Target Check
Target_Full_1 <- Target_Check_Binary ~ 0 + as.factor(TweetMedium)

lm_Target_Full_1 <- lm(Target_Full_1,
                     data = DF_Follow1)

summary(lm_Target_Full_1)
```

```
##
## Call:
## lm(formula = Target_Full_1, data = DF_Follow1)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6256 -0.5466  0.3745  0.4534  0.5193
##
## Coefficients:
##                          Estimate Std. Error t value Pr(>|t|)
## as.factor(TweetMedium)1  0.60870    0.03261   18.67   <2e-16 ***
## as.factor(TweetMedium)2  0.62555    0.03282   19.06   <2e-16 ***
## as.factor(TweetMedium)3  0.49333    0.03297   14.96   <2e-16 ***
## as.factor(TweetMedium)4  0.54661    0.03219   16.98   <2e-16 ***
## as.factor(TweetMedium)5  0.59483    0.03247   18.32   <2e-16 ***
## as.factor(TweetMedium)6  0.48069    0.03240   14.84   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4945 on 1377 degrees of freedom
## Multiple R-squared:  0.5638, Adjusted R-squared:  0.5619
## F-statistic: 296.6 on 6 and 1377 DF,  p-value: < 2.2e-16
```

```
length(DF_Follow1$Target_Check_Binary[DF_Follow1$TweetMedium==1])
```

```
## [1] 230
```

```
length(DF_Follow1$Target_Check_Binary[DF_Follow1$TweetMedium==2])
```

```
## [1] 227
```

```
length(DF_Follow1$Target_Check_Binary[DF_Follow1$TweetMedium==3])
```

```
## [1] 225
```

```
length(DF_Follow1$Target_Check_Binary[DF_Follow1$TweetMedium==4])
```

```
## [1] 236
```

```
length(DF_Follow1$Target_Check_Binary[DF_Follow1$TweetMedium==5])
```

```
## [1] 232
```

```
length(DF_Follow1$Target_Check_Binary[DF_Follow1$TweetMedium==6])
```

```
## [1] 233
```

```
##### Support Check

Support_Full_1 <- Support_Check_Binary ~ 0 + as.factor(TweetMedium)

lm_Support_Full_1 <- lm(Support_Full_1,
                        data = DF_Follow1)

summary(lm_Support_Full_1)
```

```
##
## Call:
## lm(formula = Support_Full_1, data = DF_Follow1)
```

```
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6756 -0.6304  0.3390  0.3656  0.3750
##
## Coefficients:
##                          Estimate Std. Error t value Pr(>|t|)
## as.factor(TweetMedium)1  0.63043    0.03163   19.93   <2e-16 ***
## as.factor(TweetMedium)2  0.63436    0.03184   19.93   <2e-16 ***
## as.factor(TweetMedium)3  0.67556    0.03198   21.13   <2e-16 ***
## as.factor(TweetMedium)4  0.66102    0.03122   21.17   <2e-16 ***
## as.factor(TweetMedium)5  0.62500    0.03149   19.85   <2e-16 ***
## as.factor(TweetMedium)6  0.63519    0.03142   20.21   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4797 on 1377 degrees of freedom
## Multiple R-squared:  0.644,  Adjusted R-squared:  0.6425
## F-statistic: 415.2 on 6 and 1377 DF,  p-value: < 2.2e-16
```

```
length(DF_Follow1$Support_Check_Binary[DF_Follow1$TweetMedium==1])
```

```
## [1] 230
```

```
length(DF_Follow1$Support_Check_Binary[DF_Follow1$TweetMedium==2])
```

```
## [1] 227
```

```
length(DF_Follow1$Support_Check_Binary[DF_Follow1$TweetMedium==3])
```

```
## [1] 225
```

```
length(DF_Follow1$Support_Check_Binary[DF_Follow1$TweetMedium==4])
```

```
## [1] 236
```

```
length(DF_Follow1$Support_Check_Binary[DF_Follow1$TweetMedium==5])
```

```
## [1] 232
```

```
length(DF_Follow1$Support_Check_Binary[DF_Follow1$TweetMedium==6])
```

```
## [1] 233
```

```
##### Sea Check

Sea_Full_1 <- Sea_Check_Binary ~ 0 + as.factor(TweetMedium)

lm_Sea_Full_1 <- lm(Sea_Full_1,
                    data = DF_Follow1)

summary(lm_Sea_Full_1)
```

```
##
## Call:
## lm(formula = Sea_Full_1, data = DF_Follow1)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
```

```
## -0.7210 -0.6667  0.3130  0.3263  0.3524
##
## Coefficients:
##                          Estimate Std. Error t value Pr(>|t|)
## as.factor(TweetMedium)1  0.68696    0.03078   22.32   <2e-16 ***
## as.factor(TweetMedium)2  0.64758    0.03098   20.90   <2e-16 ***
## as.factor(TweetMedium)3  0.66667    0.03112   21.42   <2e-16 ***
## as.factor(TweetMedium)4  0.67373    0.03039   22.17   <2e-16 ***
## as.factor(TweetMedium)5  0.68534    0.03065   22.36   <2e-16 ***
## as.factor(TweetMedium)6  0.72103    0.03058   23.58   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4668 on 1377 degrees of freedom
## Multiple R-squared:  0.6811, Adjusted R-squared:  0.6798
## F-statistic: 490.3 on 6 and 1377 DF,  p-value: < 2.2e-16
```

```
length(DF_Follow1$Sea_Check_Binary[DF_Follow1$TweetMedium==1])
```

```
## [1] 230
```

```
length(DF_Follow1$Sea_Check_Binary[DF_Follow1$TweetMedium==2])
```

```
## [1] 227
```

```
length(DF_Follow1$Sea_Check_Binary[DF_Follow1$TweetMedium==3])
```

```
## [1] 225
```

```
length(DF_Follow1$Sea_Check_Binary[DF_Follow1$TweetMedium==4])
```

```
## [1] 236
```

```
length(DF_Follow1$Sea_Check_Binary[DF_Follow1$TweetMedium==5])
```

```
## [1] 232
```

```
length(DF_Follow1$Sea_Check_Binary[DF_Follow1$TweetMedium==6])
```

```
## [1] 233
```

Interaction Effects

```
# looking at interactions between Trump and both informal language and Tweet
# looking at interactions between informal language and Tweet and between Tweet and Twitter Usage
# looking at interactions between Trump and political ideology


### two new formulas for interaction. 8 is interaction with partisanship. 9 is for age

Cred_Form_8 <- Credibility ~ Tweet*Trump + Informal*Trump  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

Cred_Form_9 <- Credibility ~ Tweet*Q.Twitter_Some + Trump  +
```

```
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran


Cred_Form_10 <- Credibility ~ Tweet + Informal + Trump*Q.Conservative + Trump*Q.Liberal  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Veteran + Q.Twitter_Some




Cred_8 <- lm(Cred_Form_8, data = DF_Follow1)
Cred_9 <- lm(Cred_Form_9, data = DF_Follow1)
Cred_10 <- lm(Cred_Form_10, data = DF_Follow1)


stargazer(Cred_8, Cred_9, Cred_10, title = "Crediblity with Interaction Terms and Covariates", no.space
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:40
## \begin{table}[!htbp] \centering
##   \caption{Crediblity with Interaction Terms and Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Credibility} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & 0.019 & $-$0.098 & $-$0.027 \\
##   & (0.125) & (0.103) & (0.084) \\
##   Trump & 0.368$^{***}$ & 0.236$^{***}$ & 0.186$^{*}$ \\
##   & (0.124) & (0.072) & (0.105) \\
##   Informal & 0.018 &  & $-$0.119 \\
##   & (0.126) &  & (0.084) \\
##   Q.Female & 0.001 & $-$0.001 & 0.038 \\
##   & (0.074) & (0.074) & (0.071) \\
##   relevel(as.factor(Q.Race), ref = 1)2 & 0.302$^{**}$ & 0.308$^{**}$ & 0.293$^{**}$ \\
##   & (0.122) & (0.122) & (0.116) \\
##   relevel(as.factor(Q.Race), ref = 1)3 & 0.458 & 0.500 & 0.422 \\
##   & (0.404) & (0.404) & (0.385) \\
##   relevel(as.factor(Q.Race), ref = 1)4 & 0.447$^{***}$ & 0.448$^{***}$ & 0.443$^{***}$ \\
##   & (0.168) & (0.168) & (0.160) \\
##   relevel(as.factor(Q.Race), ref = 1)5 & $-$1.122 & $-$1.116 & $-$0.960 \\
##   & (0.771) & (0.771) & (0.735) \\
##   relevel(as.factor(Q.Race), ref = 1)6 & $-$0.167 & $-$0.163 & $-$0.152 \\
```

```
##   & (0.139) & (0.139) & (0.132) \\
##   relevel(as.factor(Q.Race), ref = 1)7 & 0.124 & 0.117 & 0.156 \\
##   & (0.245) & (0.245) & (0.233) \\
##   relevel(as.factor(Q.Race), ref = 1)8 & 0.057 & 0.035 & 0.135 \\
##   & (0.336) & (0.336) & (0.321) \\
##   Q.Age & 0.004$^{*}$ & 0.004$^{*}$ & 0.004$^{*}$ \\
##   & (0.002) & (0.002) & (0.002) \\
##   Q.Bach & $-$0.134 & $-$0.141$^{*}$ & $-$0.110 \\
##   & (0.085) & (0.085) & (0.081) \\
##   Q.Income & $-$0.009 & $-$0.008 & $-$0.011 \\
##   & (0.019) & (0.019) & (0.018) \\
##   Q.Conservative & $-$0.046 & $-$0.060 & $-$0.647$^{***}$ \\
##   & (0.091) & (0.090) & (0.121) \\
##   Q.Liberal & $-$0.159$^{*}$ & $-$0.166$^{*}$ & 0.328$^{***}$ \\
##   & (0.087) & (0.087) & (0.118) \\
##   Q.Veteran & 0.116 & 0.120 & 0.096 \\
##   & (0.114) & (0.114) & (0.109) \\
##   Tweet:Q.Twitter\_Some &  & 0.053 &  \\
##   &  & (0.153) &  \\
##   Q.Twitter\_Some & 0.060 & 0.023 & 0.058 \\
##   & (0.076) & (0.127) & (0.072) \\
##   Tweet:Trump & $-$0.107 &  &  \\
##   & (0.176) &  &  \\
##   Trump:Informal & $-$0.182 &  &  \\
##   & (0.177) &  &  \\
##   Trump:Q.Conservative &  &  & 1.176$^{***}$ \\
##   &  &  & (0.168) \\
##   Trump:Q.Liberal &  &  & $-$0.956$^{***}$ \\
##   &  &  & (0.165) \\
##   Constant & 2.873$^{***}$ & 2.955$^{***}$ & 2.959$^{***}$ \\
##   & (0.168) & (0.165) & (0.159) \\
##  \hline \\[-1.8ex]
## Observations & 1,368 & 1,368 & 1,368 \\
## R$^{2}$ & 0.033 & 0.031 & 0.121 \\
## Adjusted R$^{2}$ & 0.019 & 0.018 & 0.108 \\
## Residual Std. Error & 1.325 (df = 1347) & 1.326 (df = 1349) & 1.263 (df = 1347) \\
## F Statistic & 2.294$^{***}$ (df = 20; 1347) & 2.362$^{***}$ (df = 18; 1349) & 9.271$^{***}$ (df = 20
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

AC Sensitivity Analysis

Just ACs

```
DF4 <- subset(DF_Follow1, DF_Follow1$Target_Check_Binary==1)
DF5 <- subset(DF4, DF4$Support_Check_Binary==1)
DF6 <- subset(DF5, DF5$Sea_Check_Binary==1)
```

Subset

Analysis

```
Cred_Full_1 <- Credibility ~ 0 + as.factor(TweetMedium)
lm_Cred_Full_1 <- lm(Cred_Full_1,
                     data = DF6)

summary(lm_Cred_Full_1)
```

```
##
## Call:
## lm(formula = Cred_Full_1, data = DF6)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -2.6250 -1.0952  0.3750  0.9474  2.1395
##
## Coefficients:
##                           Estimate Std. Error t value Pr(>|t|)
## as.factor(TweetMedium)1   2.8605     0.1438   19.89   <2e-16 ***
## as.factor(TweetMedium)2   3.0526     0.1530   19.95   <2e-16 ***
## as.factor(TweetMedium)3   3.0952     0.1680   18.42   <2e-16 ***
## as.factor(TweetMedium)4   3.4382     0.1414   24.32   <2e-16 ***
## as.factor(TweetMedium)5   3.6250     0.1491   24.31   <2e-16 ***
## as.factor(TweetMedium)6   3.1875     0.1667   19.12   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.334 on 452 degrees of freedom
## Multiple R-squared:  0.8557, Adjusted R-squared:  0.8538
## F-statistic: 446.9 on 6 and 452 DF,  p-value: < 2.2e-16
```

```
length(DF6$Credibility[DF6$TweetMedium==1])
```

```
## [1] 86
```

```
length(DF6$Credibility[DF6$TweetMedium==2])
```

```
## [1] 76
```

```
length(DF6$Credibility[DF6$TweetMedium==3])
```

```
## [1] 63
```

```
length(DF6$Credibility[DF6$TweetMedium==4])
```

```
## [1] 89
```

```
length(DF6$Credibility[DF6$TweetMedium==5])
```

```
## [1] 80
```

```
length(DF6$Credibility[DF6$TweetMedium==6])
```

```
## [1] 64
```

```
#time for some regressions


## Model 1: No Demographics
```

```
Cred_Form_1 <- Credibility ~ Tweet + Informal + Trump

## Model 5: Factor Demographics

Cred_Form_5 <- Credibility ~ Tweet + Informal + Trump   +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))

## Model 7: Slightly simplified for the Appendix

Cred_Form_7 <- Credibility ~ Tweet + Informal + Trump  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some


#regression



Cred_1 <- lm(Cred_Form_1,
             data = DF6,
             na.action=na.omit)
Cred_5 <- lm(Cred_Form_5,
             data = DF6,
             na.action=na.omit)
Cred_7 <- lm(Cred_Form_7,
             data = DF6,
             na.action=na.omit)


stargazer(Cred_1, Cred_5, Cred_7, title = "Credibility, AC Questions Correct", no.space = TRUE)
```

Just MCs

```
DF7 <- subset(DF_Follow1, DF_Follow1$President_Check_Binary==1)
DF8 <- subset(DF7, DF7$Threat_Check_Binary==1)
```

Subset

```
Cred_Full_1 <- Credibility ~ 0 + as.factor(TweetMedium)
lm_Cred_Full_1 <- lm(Cred_Full_1,
                     data = DF8)

summary(lm_Cred_Full_1)
```

Analysis

```
##
## Call:
## lm(formula = Cred_Full_1, data = DF8)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.51389 -1.03165  0.04724  0.99010  2.04724
##
## Coefficients:
##                          Estimate Std. Error t value Pr(>|t|)
## as.factor(TweetMedium)1   3.0316     0.1072   28.27   <2e-16 ***
## as.factor(TweetMedium)2   2.9528     0.1196   24.68   <2e-16 ***
## as.factor(TweetMedium)3   3.0099     0.1341   22.44   <2e-16 ***
## as.factor(TweetMedium)4   3.5139     0.1123   31.28   <2e-16 ***
## as.factor(TweetMedium)5   3.4104     0.1165   29.29   <2e-16 ***
## as.factor(TweetMedium)6   3.0470     0.1104   27.59   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.348 on 807 degrees of freedom
## Multiple R-squared:  0.8482, Adjusted R-squared:  0.8471
## F-statistic: 751.5 on 6 and 807 DF,  p-value: < 2.2e-16
```

```
length(DF8$Credibility[DF8$TweetMedium==1])
```

```
## [1] 158
```

```
length(DF8$Credibility[DF8$TweetMedium==2])
```

```
## [1] 127
```

```
length(DF8$Credibility[DF8$TweetMedium==3])
```

```
## [1] 101
```

```
length(DF8$Credibility[DF8$TweetMedium==4])
```

```
## [1] 144
```

```
length(DF8$Credibility[DF8$TweetMedium==5])
```

```
## [1] 134
```

```
length(DF8$Credibility[DF8$TweetMedium==6])
```

```
## [1] 149
```

```
#time for some regressions


## Model 1: No Demographics

Cred_Form_1 <- Credibility ~ Tweet + Informal + Trump

## Model 5: Factor Demographics

Cred_Form_5 <- Credibility ~ Tweet + Informal + Trump   +
```

```
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))

## Model 7: Slightly simplified for the Appendix

Cred_Form_7 <- Credibility ~ Tweet + Informal + Trump  +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some


#regression



Cred_1 <- lm(Cred_Form_1,
             data = DF8,
             na.action=na.omit)
Cred_5 <- lm(Cred_Form_5,
             data = DF8,
             na.action=na.omit)
Cred_7 <- lm(Cred_Form_7,
             data = DF8,
             na.action=na.omit)


stargazer(Cred_1, Cred_5, Cred_7, title = "Credibility, MC Questions Correct", no.space = TRUE)
```

Follow On 2, Iran

```
DF_Follow2 <-read.csv("Data, Follow_2_Iran.csv")
```

Demographic Table Stats

```
### demographics

table(DF_Follow2$Q.Female)
```

```
##
##   0   1
## 709 774
```

```
length(na.omit(DF_Follow2$Q.Female[DF_Follow2$Q.Female==0]))/length(DF_Follow2$Q.Female)
```

```
## [1] 0.4758389
```

```
sum(na.omit(DF_Follow2$Q.Female))/length(DF_Follow2$Q.Female)
```

```
## [1] 0.5194631
```

```
sum(is.na(DF_Follow2$Q.Female))/length(DF_Follow2$Q.Female)
```

```
## [1] 0.004697987
```

```
sum(na.omit(DF_Follow2$Q.White))/length((DF_Follow2$Q.White))
```

```
## [1] 0.7194631
```

```
sum(na.omit(DF_Follow2$Q.Black))/length((DF_Follow2$Q.Black))
```

```
## [1] 0.1228188
```

```
sum(na.omit(DF_Follow2$Q.AIorAN))/length((DF_Follow2$Q.AIorAN))
```

```
## [1] 0.005369128
```

```
sum(na.omit(DF_Follow2$Q.Asian))/length((DF_Follow2$Q.Asian))
```

```
## [1] 0.03892617
```

```
sum(na.omit(DF_Follow2$Q.NHorPI))/length((DF_Follow2$Q.NHorPI))
```

```
## [1] 0.001342282
```

```
sum(na.omit(DF_Follow2$Q.Hispanic))/length((DF_Follow2$Q.Hispanic))
```

```
## [1] 0.08255034
```

```
sum(na.omit(DF_Follow2$Q.Mixed))/length((DF_Follow2$Q.Mixed))
```

```
## [1] 0.01879195
```

```
sum(na.omit(DF_Follow2$Q.Other))/length((DF_Follow2$Q.Other))
```

```
## [1] 0.01073826
```

```
sum(na.omit(DF_Follow2$Q.Other_Mixed))/length((DF_Follow2$Q.Other_Mixed))
```

```
## [1] 0.0295302
```

```
sum(is.na(DF_Follow2$Q.Race))/length(DF_Follow2$Q.Race)
```

```
## [1] 0
```

```
summary(DF_Follow2$Q.Age)
```

```
##    Min. 1st Qu.  Median    Mean 3rd Qu.    Max.    NA's
##   18.00   31.00   44.00   45.24   58.00   97.00       3
```

```
table(DF_Follow2$Q.Education)
```

```
##
##   1   2   3   4   5   6   7   8
##   5  51 364 422 190 334  28  96
```

```
sum((na.omit(DF_Follow2$Q.HighSchool)))/length((DF_Follow2$Q.HighSchool))
```

```
## [1] 0.9624161
```

```
sum((na.omit(DF_Follow2$Q.Bach)))/length((DF_Follow2$Q.Bach))
```

```
## [1] 0.3073826
```

```
sum(is.na(DF_Follow2$Q.Education))/length(DF_Follow2$Q.Education)
```

```
## [1] 0
```

```
table(DF_Follow2$Q.Income)
```

```
##
##   1   2   3   4   5   6   7   8
## 267 338 295 213 152 100  47  78
```

```
length(na.omit(DF_Follow2$Q.Income[DF_Follow2$Q.Income==1]))/length(DF_Follow2$Q.Income)
```

```
## [1] 0.1791946
```

```
length(na.omit(DF_Follow2$Q.Income[DF_Follow2$Q.Income==2]))/length(DF_Follow2$Q.Income)
```

```
## [1] 0.2268456
```

```
length(na.omit(DF_Follow2$Q.Income[DF_Follow2$Q.Income==3]))/length(DF_Follow2$Q.Income)
```

```
## [1] 0.1979866
```

```
length(na.omit(DF_Follow2$Q.Income[DF_Follow2$Q.Income==4]))/length(DF_Follow2$Q.Income)
```

```
## [1] 0.142953
```

```
length(na.omit(DF_Follow2$Q.Income[DF_Follow2$Q.Income==5]))/length(DF_Follow2$Q.Income)
```

```
## [1] 0.1020134
```

```
length(na.omit(DF_Follow2$Q.Income[DF_Follow2$Q.Income==6]))/length(DF_Follow2$Q.Income)
```

```
## [1] 0.06711409
```

```
length(na.omit(DF_Follow2$Q.Income[DF_Follow2$Q.Income==7]))/length(DF_Follow2$Q.Income)
```

```
## [1] 0.03154362
```

```
length(na.omit(DF_Follow2$Q.Income[DF_Follow2$Q.Income==8]))/length(DF_Follow2$Q.Income)
```

```
## [1] 0.05234899
```

```
sum(is.na(DF_Follow2$Q.Income))/length(DF_Follow2$Q.Income)
```

```
## [1] 0
```

```
table(DF_Follow2$Q.Political_ID)
```

```
##
##   1   2   3   4   5
## 150 257 646 292 145
```

```
sum(na.omit(DF_Follow2$Q.Liberal))/length((DF_Follow2$Q.Liberal))
```

```
## [1] 0.2731544
```

```
sum(na.omit(DF_Follow2$Q.Moderate))/length((DF_Follow2$Q.Moderate))
```

```
## [1] 0.433557
```

```
sum(na.omit(DF_Follow2$Q.Conservative))/length((DF_Follow2$Q.Conservative))
```

```
## [1] 0.2932886
```

```
sum(is.na(DF_Follow2$Q.Political_ID))/length(DF_Follow2$Q.Political_ID)
```

```
## [1] 0
```

```
table(DF_Follow2$Q.Veteran)
```

```
##
##    0    1
## 1342  148
```

```
length(na.omit(DF_Follow2$Q.Veteran[DF_Follow2$Q.Veteran==0]))/length(DF_Follow2$Q.Veteran)
```

```
## [1] 0.9006711
```

```
length(na.omit(DF_Follow2$Q.Veteran[DF_Follow2$Q.Veteran==1]))/length(DF_Follow2$Q.Veteran)
```

```
## [1] 0.09932886
```

```
sum(is.na(DF_Follow2$Q.Veteran))/length(DF_Follow2$Q.Veteran)
```

```
## [1] 0
```

```
#Twitter_Use

table(DF_Follow2$Q.Twitter_Use)
```

```
##
##   1   2   3   4   5
## 775 264 183 174  94
```

```
length(na.omit(DF_Follow2$Q.Twitter_Use[DF_Follow2$Q.Twitter_Use==1]))/length(DF_Follow2$Q.Twitter_Use)
```

```
## [1] 0.5201342
```

```
length(na.omit(DF_Follow2$Q.Twitter_Use[DF_Follow2$Q.Twitter_Use==2]))/length(DF_Follow2$Q.Twitter_Use)
```

```
## [1] 0.1771812
```

```
length(na.omit(DF_Follow2$Q.Twitter_Use[DF_Follow2$Q.Twitter_Use==3]))/length(DF_Follow2$Q.Twitter_Use)
```

```
## [1] 0.1228188
```

```
length(na.omit(DF_Follow2$Q.Twitter_Use[DF_Follow2$Q.Twitter_Use==4]))/length(DF_Follow2$Q.Twitter_Use)
```

```
## [1] 0.1167785
```

```
length(na.omit(DF_Follow2$Q.Twitter_Use[DF_Follow2$Q.Twitter_Use==5]))/length(DF_Follow2$Q.Twitter_Use)
```

```
## [1] 0.06308725
```

```
sum(is.na(DF_Follow2$Q.Twitter_Use))/length(DF_Follow2$Q.Twitter_Use)
```

```
## [1] 0
```

```
colnames(DF_Follow2)
```

```
##  [1] "ResponseId"          "Tweet"               "Credibility_Iran"
##  [4] "Threat_Check_Binary" "Target_Check_Binary" "Q.Female"
##  [7] "Q.Race"              "Q.White"             "Q.Black"
```

```
## [10] "Q.AIorAN"            "Q.Asian"             "Q.NHorPI"
## [13] "Q.Hispanic"          "Q.Mixed"             "Q.Other"
## [16] "Q.Other_Mixed"       "Q.Age"               "Q.Education"
## [19] "Q.HighSchool"        "Q.Bach"              "Q.Income"
## [22] "Q.Political_ID"      "Q.Liberal"           "Q.Moderate"
## [25] "Q.Conservative"      "Q.Veteran"           "Q.Twitter_Use"
## [28] "Q.Twitter_None"      "Q.Twitter_Some"
```

Analysis

Credibility

```
## a simple regression for mean and se estimates


######## Credibility


Cred_Full_1 <- Credibility_Iran ~ 0 + as.factor(Tweet)
lm_Cred_Full_1 <- lm(Cred_Full_1,
                     data = DF_Follow2)

summary(lm_Cred_Full_1)
```

```
##
## Call:
## lm(formula = Cred_Full_1, data = DF_Follow2)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -2.5778 -0.5778  0.4222  0.4434  1.4434
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  3.57781    0.04436   80.66   <2e-16 ***
## as.factor(Tweet)1  3.55659    0.04400   80.83   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.206 on 1488 degrees of freedom
## Multiple R-squared:  0.8976, Adjusted R-squared:  0.8974
## F-statistic:  6520 on 2 and 1488 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_Follow2$Credibility_Iran[DF_Follow2$Tweet==1]))
```

```
## [1] 751
```

```
length(na.omit(DF_Follow2$Credibility_Iran[DF_Follow2$Tweet==0]))
```

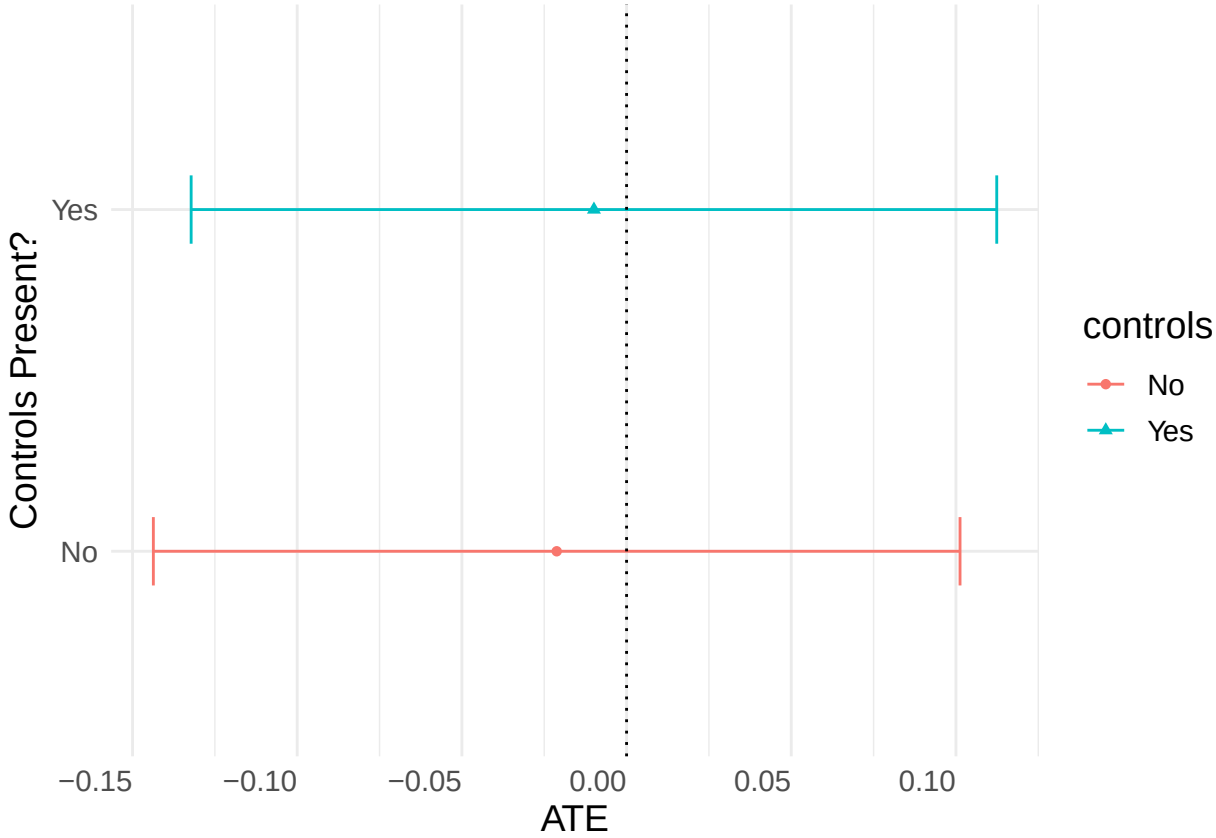
```
## [1] 739
```

```
#time for some regressions

#let's do a couple different models

##DV: credibility, IV: Tweet binary
```

```r
## Model 1: No Demographics

Cred_Form_1 <- Credibility_Iran ~ Tweet

## Model 2: Demographics in Erik's Paper


## Model 5: Factor Demographics

Cred_Form_5 <-
  Credibility_Iran ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))


## Model 7: Slightly simplified for the Appendix

Cred_Form_7 <- Credibility_Iran ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Cred_1 <- lm(Cred_Form_1,
             data = DF_Follow2,
             na.action=na.omit)

Cred_5 <- lm(Cred_Form_5,
             data = DF_Follow2,
             na.action=na.omit)

Cred_7 <- lm(Cred_Form_7,
             data = DF_Follow2,
             na.action=na.omit)

library(stargazer)


stargazer(Cred_1, Cred_5, Cred_7, title = "Perceived Credibility with Covariates", no.space = TRUE)
```

```r
#Credibility not separated



Cred_Alone <- as.data.frame(matrix(data = c("credibility", coeftest(Cred_5)[2, 1:2],"Yes",
                                    "credibility", coeftest(Cred_1)[2, 1:2], "No"),
                                    ncol = 4, byrow = TRUE))
```

```r
colnames(Cred_Alone) <- c("dv", "estimate", "se", "controls")

#making into correct operators
Cred_Alone$estimate <- as.numeric(Cred_Alone$estimate)
Cred_Alone$se <- as.numeric(Cred_Alone$se)

#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Cred_Alone$ci <- Cred_Alone$se*q

library(ggplot2)
## graph time
pd <- position_dodge(0.5)

ggplot(Cred_Alone, aes(x = controls, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Controls Present?") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```r
#ggtitle("Credibility, ATE of Tweet Medium, Iran")
```

Manipulation Check

```r
##### stargazer tables for MC
#### note: this version has removed the "backsdown" binary variable
```

```r
# Threat/Treatment Check

## Model 1: No Demographics

Threat_Form_1 <- Threat_Check_Binary ~ Tweet


## Model 5: Factor Demographics

Threat_Form_5 <- Threat_Check_Binary ~ Tweet  + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))



## Model 7: Slightly simplified for the Appendix

Threat_Form_7 <- Threat_Check_Binary ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Threat_1 <- lm(Threat_Form_1,
               data = DF_Follow2,
               na.action=na.omit)

Threat_5 <- lm(Threat_Form_5,
               data = DF_Follow2,
               na.action=na.omit)

Threat_7 <- lm(Threat_Form_7,
               data = DF_Follow2,
               na.action=na.omit)


stargazer(Threat_1, Threat_5, Threat_7, title = "Treatment Check Question with Covariates", no.space =
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac
## % Date and time: Sun, Dec 31, 2023 - 14:16:41
## \begin{table}[!htbp] \centering
##   \caption{Treatment Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
```

```
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Threat\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.126$^{***}$ & $-$0.126$^{***}$ & $-$0.127$^{***}$ \\
##   & (0.021) & (0.021) & (0.021) \\
##   Q.Female &  & 0.057$^{***}$ & 0.061$^{***}$ \\
##   &  & (0.022) & (0.021) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.101$^{***}$ & $-$0.107$^{***}$ \\
##   &  & (0.034) & (0.033) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & 0.069 & 0.047 \\
##   &  & (0.151) & (0.150) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & $-$0.077 & $-$0.072 \\
##   &  & (0.055) & (0.055) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & $-$0.272 & $-$0.226 \\
##   &  & (0.283) & (0.281) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & $-$0.029 & $-$0.025 \\
##   &  & (0.039) & (0.038) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & $-$0.008 & $-$0.003 \\
##   &  & (0.077) & (0.076) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & $-$0.087 & $-$0.096 \\
##   &  & (0.100) & (0.100) \\
##   Q.Age &  & 0.0003 & 0.0004 \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.037 &  \\
##   &  & (0.056) &  \\
##   Q.Bach &  & 0.084$^{***}$ & 0.084$^{***}$ \\
##   &  & (0.024) & (0.024) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & $-$0.013 &  \\
##   &  & (0.041) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & 0.007 &  \\
##   &  & (0.034) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.011 &  \\
##   &  & (0.036) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & 0.011 &
##   &  & (0.039) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & 0.012 &
##   &  & (0.038) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & $-$0.0
##   &  & (0.035) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & $-$0.0
##   &  & (0.031) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.032 &  \\
##   &  & (0.032) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.046$^{*}$ &  \\
##   &  & (0.028) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.021 &  \\
##   &  & (0.026) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.022
##   &  & (0.021) &  \\
##   Q.Income &  &  & $-$0.002 \\
##   &  &  & (0.006) \\
##   Q.Conservative &  &  & $-$0.024 \\
##   &  &  & (0.025) \\
```

```
##   Q.Liberal &  &  & 0.013 \\
##   &  &  & (0.025) \\
##   Q.Veteran &  & 0.002 & $-$0.001 \\
##   &  & (0.035) & (0.035) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.016 &  \\
##   &  & (0.032) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.030 &  \\
##   &  & (0.030) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.028 &  \\
##   &  & (0.029) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.014 &
##   &  & (0.029) &  \\
##   Q.Twitter\_Some &  &  & 0.027 \\
##   &  &  & (0.022) \\
##   Constant & 0.858$^{***}$ & 0.769$^{***}$ & 0.801$^{***}$ \\
##   & (0.015) & (0.065) & (0.045) \\
##  \hline \\[-1.8ex]
## Observations & 1,490 & 1,480 & 1,480 \\
## R$^{2}$ & 0.024 & 0.054 & 0.050 \\
## Adjusted R$^{2}$ & 0.023 & 0.036 & 0.040 \\
## Residual Std. Error & 0.399 (df = 1488) & 0.396 (df = 1451) & 0.395 (df = 1463) \\
## F Statistic & 36.824$^{***}$ (df = 1; 1488) & 2.977$^{***}$ (df = 28; 1451) & 4.815$^{***}$ (df = 16
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

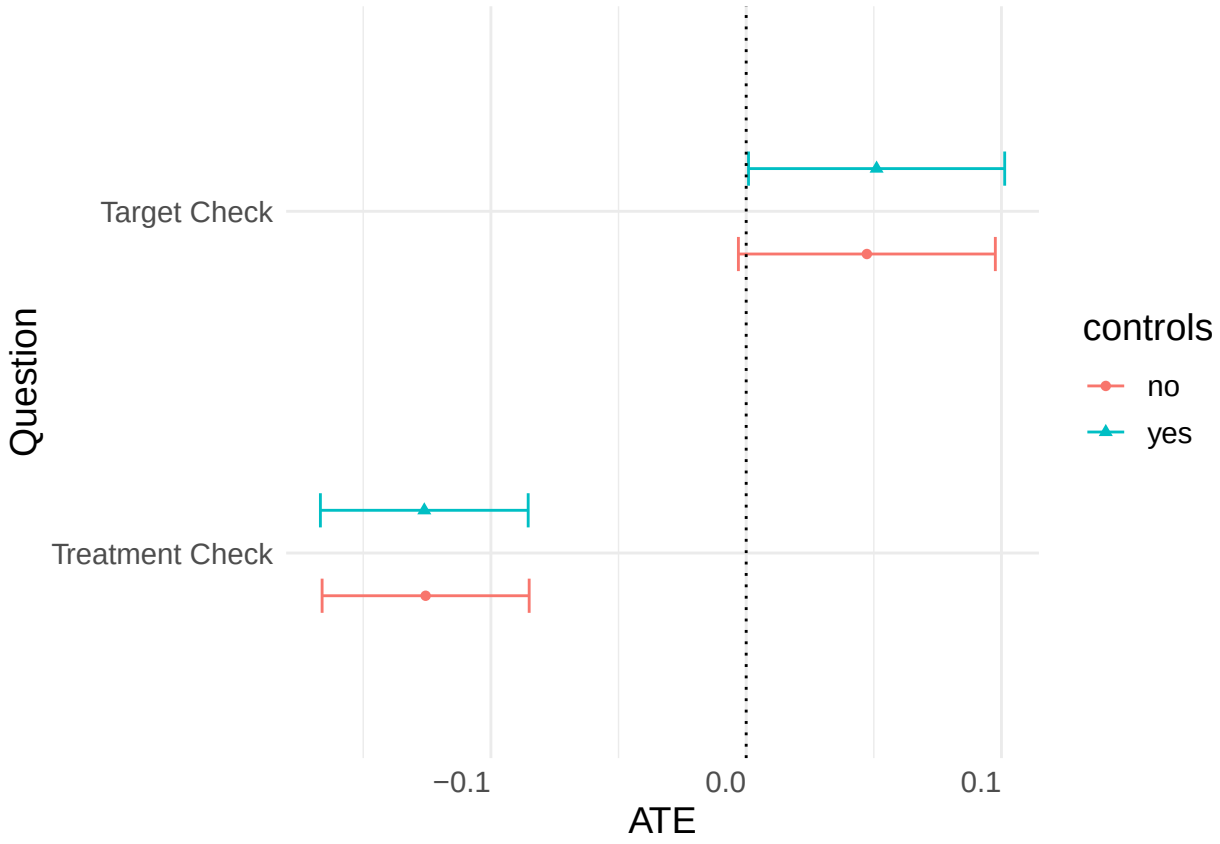
```r
# Target Check

## Model 1: No Demographics

Target_Form_1 <- Target_Check_Binary ~ Tweet


## Model 5: Factor Demographics

Target_Form_5 <- Target_Check_Binary ~ Tweet  +  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))



## Model 7: Slightly simplified for the Appendix

Target_Form_7 <- Target_Check_Binary ~ Tweet  + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some
```

```r
#regression

Target_1 <- lm(Target_Form_1,
               data = DF_Follow2,
               na.action=na.omit)

Target_5 <- lm(Target_Form_5,
               data = DF_Follow2,
               na.action=na.omit)

Target_7 <- lm(Target_Form_7,
               data = DF_Follow2,
               na.action=na.omit)


stargazer(Target_1, Target_5, Target_7, title = "Target Check Question with Covariates", no.space = TRU
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:41
## \begin{table}[!htbp] \centering
##   \caption{Target Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Target\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & 0.047$^{*}$ & 0.051$^{**}$ & 0.047$^{*}$ \\
##   & (0.026) & (0.026) & (0.026) \\
##   Q.Female &  & $-$0.012 & $-$0.005 \\
##   &  & (0.027) & (0.026) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.088$^{**}$ & $-$0.102$^{**}$ \\
##   &  & (0.041) & (0.041) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & 0.177 & 0.158 \\
##   &  & (0.187) & (0.186) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & $-$0.040 & $-$0.020 \\
##   &  & (0.068) & (0.068) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & $-$0.541 & $-$0.531 \\
##   &  & (0.349) & (0.347) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & $-$0.037 & $-$0.036 \\
##   &  & (0.048) & (0.048) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & 0.084 & 0.086 \\
##   &  & (0.094) & (0.094) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & $-$0.169 & $-$0.182 \\
##   &  & (0.124) & (0.123) \\
##   Q.Age &  & 0.004$^{***}$ & 0.004$^{***}$ \\
##   &  & (0.001) & (0.001) \\
##   Q.HighSchool &  & 0.033 &  \\
##   &  & (0.070) &  \\
##   Q.Bach &  & 0.030 & 0.027 \\
```

```
##   &  & (0.030) & (0.030) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & 0.080 &  \\
##   &  & (0.050) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & 0.042 &  \\
##   &  & (0.042) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.051 &  \\
##   &  & (0.045) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & $-$0.08
##   &  & (0.048) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & $-$0.1
##   &  & (0.047) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & $-$0.07
##   &  & (0.043) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & $-$0.05
##   &  & (0.038) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 0.055 &  \\
##   &  & (0.039) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.029 &  \\
##   &  & (0.035) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.015 &  \\
##   &  & (0.032) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.004
##   &  & (0.026) &  \\
##   Q.Income &  &  & 0.006 \\
##   &  &  & (0.007) \\
##   Q.Conservative &  &  & 0.006 \\
##   &  &  & (0.031) \\
##   Q.Liberal &  &  & $-$0.035 \\
##   &  &  & (0.031) \\
##   Q.Veteran &  & 0.045 & 0.043 \\
##   &  & (0.044) & (0.043) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.056 &  \\
##   &  & (0.040) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.026 &  \\
##   &  & (0.037) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.051 &  \\
##   &  & (0.035) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.009 &
##   &  & (0.036) &  \\
##   Q.Twitter\_Some &  &  & 0.007 \\
##   &  &  & (0.027) \\
##   Constant & 0.541$^{***}$ & 0.355$^{***}$ & 0.362$^{***}$ \\
##   & (0.018) & (0.080) & (0.055) \\
##  \hline \\[-1.8ex]
## Observations & 1,490 & 1,480 & 1,480 \\
## R$^{2}$ & 0.002 & 0.049 & 0.037 \\
## Adjusted R$^{2}$ & 0.002 & 0.030 & 0.026 \\
## Residual Std. Error & 0.496 (df = 1488) & 0.488 (df = 1451) & 0.489 (df = 1463) \\
## F Statistic & 3.391$^{*}$ (df = 1; 1488) & 2.656$^{***}$ (df = 28; 1451) & 3.504$^{***}$ (df = 16; 14
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```r
## more detail on manipulation check

#put all the names, estimates, SEs, and behavior into one df for graphing
Results_DF <- as.data.frame(matrix(data = c("Treatment Check", coeftest(Threat_1)[2, 1:2],"no",
                                    "Treatment Check", coeftest(Threat_5)[2, 1:2], "yes",
                                    "Target Check", coeftest(Target_1)[2, 1:2],"no",
                                    "Target Check", coeftest(Target_5)[2, 1:2], "yes"),
                                    ncol = 4, byrow = TRUE))

colnames(Results_DF) <- c("dv", "estimate", "se", "controls")

#making into correct operators
Results_DF$dv <- factor(Results_DF$dv, levels = c("Treatment Check", "Target Check"))
Results_DF$estimate <- as.numeric(Results_DF$estimate)
Results_DF$se <- as.numeric(Results_DF$se)

#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Results_DF$ci <- Results_DF$se*q


## graph time
library(ggplot2)
pd <- position_dodge(0.5)

ggplot(Results_DF, aes(x = dv, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Question") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```
#ggtitle("Tweet Imagery ATE by MC or AC Question") +
```

```
## mean and se tables for threat check

####Threat/Treatment Check
Threat_Full_1 <- Threat_Check_Binary ~ 0 + as.factor(Tweet)


lm_Threat_Full_1 <- lm(Threat_Full_1,
                       data = DF_Follow2)

summary(lm_Threat_Full_1)
```

```
##
## Call:
## lm(formula = Threat_Full_1, data = DF_Follow2)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.8579  0.1421  0.1421  0.2676  0.2676
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.85792    0.01469   58.40   <2e-16 ***
## as.factor(Tweet)1  0.73236    0.01457   50.26   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```
## Residual standard error: 0.3993 on 1488 degrees of freedom
## Multiple R-squared:  0.7996, Adjusted R-squared:  0.7993
## F-statistic:  2968 on 2 and 1488 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_Follow2$Threat_Check_Binary[DF_Follow2$Tweet==1]))
```

```
## [1] 751
```

```
length(na.omit(DF_Follow2$Threat_Check_Binary[DF_Follow2$Tweet==0]))
```

```
## [1] 739
```

```
##### Target Check
Target_Full_1 <- Target_Check_Binary ~ 0 + as.factor(Tweet)

lm_Target_Full_1 <- lm(Target_Full_1,
                       data = DF_Follow2)

summary(lm_Target_Full_1)
```

```
##
## Call:
## lm(formula = Target_Full_1, data = DF_Follow2)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.5886 -0.5413  0.4114  0.4587  0.4587
##
## Coefficients:
##                    Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.54127    0.01823   29.70   <2e-16 ***
## as.factor(Tweet)1  0.58855    0.01808   32.55   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4955 on 1488 degrees of freedom
## Multiple R-squared:  0.5661, Adjusted R-squared:  0.5655
## F-statistic: 970.6 on 2 and 1488 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_Follow2$Target_Check_Binary[DF_Follow2$Tweet==1]))
```

```
## [1] 751
```

```
length(na.omit(DF_Follow2$Target_Check_Binary[DF_Follow2$Tweet==0]))
```

```
## [1] 739
```

Interaction Effects

```
# looking at interactions between partisanship and tweet
# and between age and Tweet

### two new formulas for interaction. 8 is interaction with partisanship. 9 is for age

Cred_Form_8 <- Credibility_Iran ~ Tweet*Q.Liberal + Tweet*Q.Conservative +
  Q.Female + relevel(as.factor(Q.Race), ref = 1) +
    Q.Age + Q.Bach + Q.Income  +  Q.Veteran + Q.Twitter_Some
```

```r
Cred_Form_9 <- Credibility_Iran ~ Tweet*Q.Twitter_Some +
  Q.Female + relevel(as.factor(Q.Race), ref = 1) +
    Q.Bach + Q.Income + Q.Conservative + Q.Liberal +
    Q.Veteran + Q.Twitter_Some



Cred_8 <- lm(Cred_Form_8, data = DF_Follow2)
Cred_9 <- lm(Cred_Form_9, data = DF_Follow2)

stargazer(Cred_8, Cred_9, title = "Crediblity with Interaction Terms with Covariates", no.space = TRUE)
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:41
## \begin{table}[!htbp] \centering
##   \caption{Crediblity with Interaction Terms with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lcc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{2}{c}{\textit{Dependent variable:}} \\
## \cline{2-3}
## \\[-1.8ex] & \multicolumn{2}{c}{Credibility\_Iran} \\
## \\[-1.8ex] & (1) & (2)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.041 & $-$0.0004 \\
##   & (0.095) & (0.086) \\
##   Q.Liberal & $-$0.178$^{*}$ & $-$0.091 \\
##   & (0.108) & (0.076) \\
##   Q.Conservative & 0.014 & 0.017 \\
##   & (0.108) & (0.076) \\
##   Q.Female & 0.348$^{***}$ & 0.354$^{***}$ \\
##   & (0.064) & (0.064) \\
##   relevel(as.factor(Q.Race), ref = 1)2 & 0.156 & 0.121 \\
##   & (0.099) & (0.097) \\
##   relevel(as.factor(Q.Race), ref = 1)3 & $-$0.446 & $-$0.453 \\
##   & (0.453) & (0.453) \\
##   relevel(as.factor(Q.Race), ref = 1)4 & $-$0.338$^{**}$ & $-$0.375$^{**}$ \\
##   & (0.165) & (0.162) \\
##   relevel(as.factor(Q.Race), ref = 1)5 & $-$1.063 & $-$1.079 \\
##   & (0.845) & (0.845) \\
##   relevel(as.factor(Q.Race), ref = 1)6 & $-$0.149 & $-$0.183 \\
##   & (0.116) & (0.114) \\
##   relevel(as.factor(Q.Race), ref = 1)7 & $-$0.234 & $-$0.231 \\
##   & (0.229) & (0.229) \\
##   relevel(as.factor(Q.Race), ref = 1)8 & $-$0.459 & $-$0.443 \\
##   & (0.300) & (0.301) \\
##   Q.Age & 0.002 &  \\
##   & (0.002) &  \\
##   Q.Bach & $-$0.085 & $-$0.069 \\
##   & (0.073) & (0.072) \\
##   Q.Income & 0.008 & 0.008 \\
```

```
##   & (0.017) & (0.017) \\
##   Q.Veteran & 0.167 & 0.186$^{*}$ \\
##   & (0.106) & (0.105) \\
##   Tweet:Q.Twitter\_Some &  & $-$0.007 \\
##   &  & (0.124) \\
##   Q.Twitter\_Some & 0.067 & 0.051 \\
##   & (0.066) & (0.089) \\
##   Tweet:Q.Liberal & 0.153 &  \\
##   & (0.152) &  \\
##   Tweet:Q.Conservative & $-$0.031 &  \\
##   & (0.148) &  \\
##   Constant & 3.299$^{***}$ & 3.384$^{***}$ \\
##   & (0.139) & (0.096) \\
##  \hline \\[-1.8ex]
## Observations & 1,480 & 1,483 \\
## R$^{2}$ & 0.035 & 0.033 \\
## Adjusted R$^{2}$ & 0.023 & 0.022 \\
## Residual Std. Error & 1.189 (df = 1461) & 1.191 (df = 1466) \\
## F Statistic & 2.924$^{***}$ (df = 18; 1461) & 3.109$^{***}$ (df = 16; 1466) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{2}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

AC Sensitivity Analysis

Just ACs

```
DF4 <- subset(DF_Follow2, DF_Follow2$Target_Check_Binary==1)
```

Subset

```
## a simple regression for mean and se estimates


######## Credibility


Cred_Full_1 <- Credibility_Iran ~ 0 + as.factor(Tweet)
lm_Cred_Full_1 <- lm(Cred_Full_1,
                     data = DF4)

summary(lm_Cred_Full_1)
```

Analysis

```
##
## Call:
## lm(formula = Cred_Full_1, data = DF4)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -2.6606 -0.6606  0.3394  1.3394  1.3400
```

```
##
## Coefficients:
##                    Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  3.66000    0.05913   61.90   <2e-16 ***
## as.factor(Tweet)1  3.66063    0.05625   65.08   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.183 on 840 degrees of freedom
## Multiple R-squared:  0.9057, Adjusted R-squared:  0.9055
## F-statistic:  4033 on 2 and 840 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF4$Credibility_Iran[DF4$Tweet==1]))
```

```
## [1] 442
```

```
length(na.omit(DF4$Credibility_Iran[DF4$Tweet==0]))
```

```
## [1] 400
```

```
#time for some regressions

#let's do a couple different models

##DV: credibility, IV: Tweet binary

## Model 1: No Demographics

Cred_Form_1 <- Credibility_Iran ~ Tweet

## Model 2: Demographics in Erik's Paper


## Model 5: Factor Demographics

Cred_Form_5 <-
  Credibility_Iran ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))


## Model 7: Slightly simplified for the Appendix

Cred_Form_7 <- Credibility_Iran ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Cred_1 <- lm(Cred_Form_1,
             data = DF4,
```

```
            na.action=na.omit)

Cred_5 <- lm(Cred_Form_5,
             data = DF4,
             na.action=na.omit)

Cred_7 <- lm(Cred_Form_7,
             data = DF4,
             na.action=na.omit)

library(stargazer)


stargazer(Cred_1, Cred_5, Cred_7, title = "Perceived Credibility, AC Question Correct", no.space = TRUE)
```

Just MCs

```
DF5 <- subset(DF_Follow2, DF_Follow2$Threat_Check_Binary==1)
```

Subset

```
## a simple regression for mean and se estimates


######## Credibility


Cred_Full_1 <- Credibility_Iran ~ 0 + as.factor(Tweet)
lm_Cred_Full_1 <- lm(Cred_Full_1,
                     data = DF5)

summary(lm_Cred_Full_1)
```

Analysis

```
##
## Call:
## lm(formula = Cred_Full_1, data = DF5)
##
## Residuals:
##    Min     1Q Median     3Q    Max
## -2.640 -0.640  0.360  1.360  1.385
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  3.61514    0.04784   75.57   <2e-16 ***
## as.factor(Tweet)1  3.64000    0.05136   70.87   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.204 on 1182 degrees of freedom
## Multiple R-squared:  0.9008, Adjusted R-squared:  0.9006
## F-statistic:  5367 on 2 and 1182 DF,  p-value: < 2.2e-16
```

```r
length(na.omit(DF5$Credibility_Iran[DF5$Tweet==1]))
```

```
## [1] 550
```

```r
length(na.omit(DF5$Credibility_Iran[DF5$Tweet==0]))
```

```
## [1] 634
```

```r
#time for some regressions

#let's do a couple different models

##DV: credibility, IV: Tweet binary

## Model 1: No Demographics

Cred_Form_1 <- Credibility_Iran ~ Tweet

## Model 2: Demographics in Erik's Paper


## Model 5: Factor Demographics

Cred_Form_5 <-
  Credibility_Iran ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))


## Model 7: Slightly simplified for the Appendix

Cred_Form_7 <- Credibility_Iran ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Cred_1 <- lm(Cred_Form_1,
             data = DF5,
             na.action=na.omit)

Cred_5 <- lm(Cred_Form_5,
             data = DF5,
             na.action=na.omit)

Cred_7 <- lm(Cred_Form_7,
             data = DF5,
             na.action=na.omit)

library(stargazer)
```

```
stargazer(Cred_1, Cred_5, Cred_7, title = "Perceived Credibility, MC Question Correct", no.space = TRUE)
```

Follow On 3, Vetting

```
DF_Follow3 <-read.csv("Data, Follow_3_Vetting.csv")
```

Demographic Table Stats

```
### demographics

table(DF_Follow3$Q.Female)
```

```
##
##   0   1
## 523 566
```

```
length(na.omit(DF_Follow3$Q.Female[DF_Follow3$Q.Female==0]))/length(DF_Follow3$Q.Female)
```

```
## [1] 0.4758872
```

```
sum(na.omit(DF_Follow3$Q.Female))/length(DF_Follow3$Q.Female)
```

```
## [1] 0.5150136
```

```
sum(is.na(DF_Follow3$Q.Female))/length(DF_Follow3$Q.Female)
```

```
## [1] 0.009099181
```

```
sum(na.omit(DF_Follow3$Q.White))/length((DF_Follow3$Q.White))
```

```
## [1] 0.7088262
```

```
sum(na.omit(DF_Follow3$Q.Black))/length((DF_Follow3$Q.Black))
```

```
## [1] 0.1155596
```

```
sum(na.omit(DF_Follow3$Q.AIorAN))/length((DF_Follow3$Q.AIorAN))
```

```
## [1] 0.006369427
```

```
sum(na.omit(DF_Follow3$Q.Asian))/length((DF_Follow3$Q.Asian))
```

```
## [1] 0.04913558
```

```
sum(na.omit(DF_Follow3$Q.NHorPI))/length((DF_Follow3$Q.NHorPI))
```

```
## [1] 0.002729754
```

```
sum(na.omit(DF_Follow3$Q.Hispanic))/length((DF_Follow3$Q.Hispanic))
```

```
## [1] 0.08644222
```

```
sum(na.omit(DF_Follow3$Q.Mixed))/length((DF_Follow3$Q.Mixed))
```

```
## [1] 0.02092812
```

```
sum(na.omit(DF_Follow3$Q.Other))/length((DF_Follow3$Q.Other))
```

```
## [1] 0.0100091
```

```
sum(na.omit(DF_Follow3$Q.Other_Mixed))/length((DF_Follow3$Q.Other_Mixed))
```

```
## [1] 0.03093722
```

```
sum(is.na(DF_Follow3$Q.Race))/length(DF_Follow3$Q.Race)
```

```
## [1] 0
```

```
summary(DF_Follow3$Q.Age)
```

```
##    Min. 1st Qu.  Median    Mean 3rd Qu.    Max.    NA's
##    0.38   30.00   42.00   49.95   57.00 6969.00       1
```

```
table(DF_Follow3$Q.Education)
```

```
##
##   1   2   3   4   5   6   7   8
##   5  40 278 250 158 244  28  96
```

```
sum((na.omit(DF_Follow3$Q.HighSchool)))/length((DF_Follow3$Q.HighSchool))
```

```
## [1] 0.9590537
```

```
sum((na.omit(DF_Follow3$Q.Bach)))/length((DF_Follow3$Q.Bach))
```

```
## [1] 0.3348499
```

```
sum(is.na(DF_Follow3$Q.Education))/length(DF_Follow3$Q.Education)
```

```
## [1] 0
```

```
table(DF_Follow3$Q.Income)
```

```
##
##   1   2   3   4   5   6   7   8
## 163 254 229 180  90  75  32  76
```

```
length(na.omit(DF_Follow3$Q.Income[DF_Follow3$Q.Income==1]))/length(DF_Follow3$Q.Income)
```

```
## [1] 0.1483167
```

```
length(na.omit(DF_Follow3$Q.Income[DF_Follow3$Q.Income==2]))/length(DF_Follow3$Q.Income)
```

```
## [1] 0.2311192
```

```
length(na.omit(DF_Follow3$Q.Income[DF_Follow3$Q.Income==3]))/length(DF_Follow3$Q.Income)
```

```
## [1] 0.2083712
```

```
length(na.omit(DF_Follow3$Q.Income[DF_Follow3$Q.Income==4]))/length(DF_Follow3$Q.Income)
```

```
## [1] 0.1637853
```

```
length(na.omit(DF_Follow3$Q.Income[DF_Follow3$Q.Income==5]))/length(DF_Follow3$Q.Income)
```

```
## [1] 0.08189263
```

```
length(na.omit(DF_Follow3$Q.Income[DF_Follow3$Q.Income==6]))/length(DF_Follow3$Q.Income)
```

```
## [1] 0.06824386
```

```
length(na.omit(DF_Follow3$Q.Income[DF_Follow3$Q.Income==7]))/length(DF_Follow3$Q.Income)
```

```
## [1] 0.02911738
```

```
length(na.omit(DF_Follow3$Q.Income[DF_Follow3$Q.Income==8]))/length(DF_Follow3$Q.Income)
```

```
## [1] 0.06915378
```

```
sum(is.na(DF_Follow3$Q.Income))/length(DF_Follow3$Q.Income)
```

```
## [1] 0
```

```
table(DF_Follow3$Q.Political_ID)
```

```
##
##   1   2   3   4   5
## 108 199 470 206 116
```

```
sum(na.omit(DF_Follow3$Q.Liberal))/length((DF_Follow3$Q.Liberal))
```

```
## [1] 0.2793449
```

```
sum(na.omit(DF_Follow3$Q.Moderate))/length((DF_Follow3$Q.Moderate))
```

```
## [1] 0.4276615
```

```
sum(na.omit(DF_Follow3$Q.Conservative))/length((DF_Follow3$Q.Conservative))
```

```
## [1] 0.2929936
```

```
sum(is.na(DF_Follow3$Q.Political_ID))/length(DF_Follow3$Q.Political_ID)
```

```
## [1] 0
```

```
table(DF_Follow3$Q.Veteran)
```

```
##
##    0    1
## 1015   84
```

```
length(na.omit(DF_Follow3$Q.Veteran[DF_Follow3$Q.Veteran==0]))/length(DF_Follow3$Q.Veteran)
```

```
## [1] 0.9235669
```

```
length(na.omit(DF_Follow3$Q.Veteran[DF_Follow3$Q.Veteran==1]))/length(DF_Follow3$Q.Veteran)
```

```
## [1] 0.07643312
```

```
sum(is.na(DF_Follow3$Q.Veteran))/length(DF_Follow3$Q.Veteran)
```

```
## [1] 0
```

```
#Twitter_Use

table(DF_Follow3$Q.Twitter_Use)
```

```
##
##   1   2   3   4   5
## 521 218 143 140  77
```

```
length(na.omit(DF_Follow3$Q.Twitter_Use[DF_Follow3$Q.Twitter_Use==1]))/length(DF_Follow3$Q.Twitter_Use)
```

```
## [1] 0.4740673
```

```
length(na.omit(DF_Follow3$Q.Twitter_Use[DF_Follow3$Q.Twitter_Use==2]))/length(DF_Follow3$Q.Twitter_Use)
```

```
## [1] 0.1983621
```

```
length(na.omit(DF_Follow3$Q.Twitter_Use[DF_Follow3$Q.Twitter_Use==3]))/length(DF_Follow3$Q.Twitter_Use)
```

```
## [1] 0.1301183
```

```
length(na.omit(DF_Follow3$Q.Twitter_Use[DF_Follow3$Q.Twitter_Use==4]))/length(DF_Follow3$Q.Twitter_Use)
```

```
## [1] 0.1273885
```

```
length(na.omit(DF_Follow3$Q.Twitter_Use[DF_Follow3$Q.Twitter_Use==5]))/length(DF_Follow3$Q.Twitter_Use)
```

```
## [1] 0.07006369
```

```
sum(is.na(DF_Follow3$Q.Twitter_Use))/length(DF_Follow3$Q.Twitter_Use)
```

```
## [1] 0
```

Analysis

Credibility

```
## a simple regression for mean and se estimates


######## Credibility


Cred_Full_1 <- Credibility ~ 0 + as.factor(Tweet)
lm_Cred_Full_1 <- lm(Cred_Full_1,
                     data = DF_Follow3)

summary(lm_Cred_Full_1)
```

```
##
## Call:
## lm(formula = Cred_Full_1, data = DF_Follow3)
##
## Residuals:
##    Min     1Q Median     3Q    Max
## -1.940 -0.940  0.060  1.066  2.066
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  2.93443    0.05395   54.40   <2e-16 ***
## as.factor(Tweet)1  2.94000    0.05390   54.55   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.264 on 1097 degrees of freedom
## Multiple R-squared:  0.844,  Adjusted R-squared:  0.8437
## F-statistic:  2967 on 2 and 1097 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_Follow3$Credibility[DF_Follow3$Tweet==1]))
```

```
## [1] 550
```

```
length(na.omit(DF_Follow3$Credibility[DF_Follow3$Tweet==0]))
```

```
## [1] 549
```

```r
#time for some regressions

#let's do a couple different models

##DV: credibility, IV: Tweet binary

## Model 1: No Demographics

Cred_Form_1 <- Credibility ~ Tweet

## Model 2: Demographics in Erik's Paper


## Model 5: Factor Demographics

Cred_Form_5 <-
  Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))


## Model 7: Slightly simplified for the Appendix

Cred_Form_7 <- Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Cred_1 <- lm(Cred_Form_1,
             data = DF_Follow3,
             na.action=na.omit)

Cred_5 <- lm(Cred_Form_5,
             data = DF_Follow3,
             na.action=na.omit)

Cred_7 <- lm(Cred_Form_7,
             data = DF_Follow3,
             na.action=na.omit)

library(stargazer)


stargazer(Cred_1, Cred_5, Cred_7, title = "Perceived Credibility with Covariates", no.space = TRUE)
```

Vetting

```
## a simple regression for mean and se estimates


######## Vetting


Vet_Full_1 <- Vetting ~ 0 + as.factor(Tweet)
lm_Vet_Full_1 <- lm(Vet_Full_1,
                    data = DF_Follow3)

summary(lm_Vet_Full_1)
```

```
##
## Call:
## lm(formula = Vet_Full_1, data = DF_Follow3)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.7418 -0.7377  0.2582  0.2623  0.2623
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.73770    0.01874   39.36   <2e-16 ***
## as.factor(Tweet)1  0.74182    0.01873   39.62   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4392 on 1097 degrees of freedom
## Multiple R-squared:  0.7398, Adjusted R-squared:  0.7393
## F-statistic:  1559 on 2 and 1097 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_Follow3$Vetting[DF_Follow3$Tweet==1]))
```

```
## [1] 550
```

```
length(na.omit(DF_Follow3$Vetting[DF_Follow3$Tweet==0]))
```

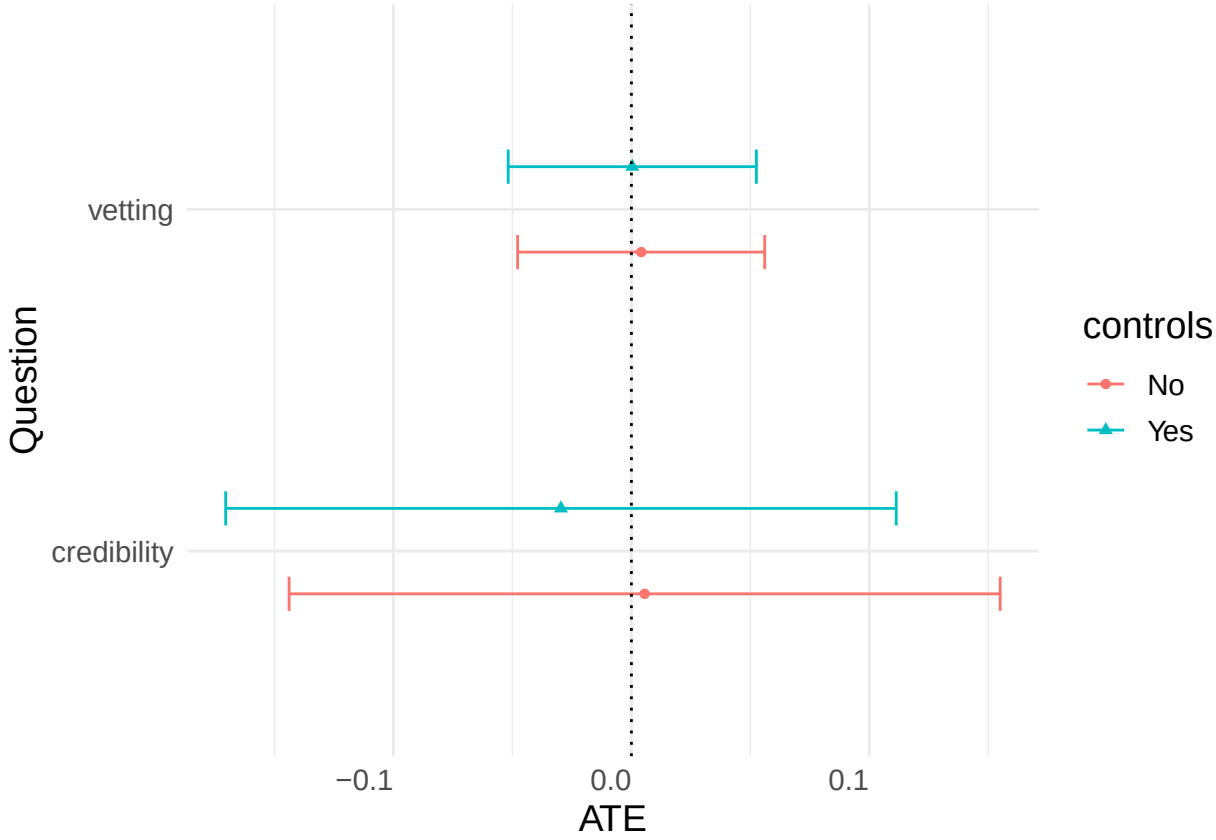
```
## [1] 549
```

```
#time for some regressions

#let's do a couple different models

##DV: Vetting, IV: Tweet binary

## Model 1: No Demographics

Vet_Form_1 <- Vetting ~ Tweet

## Model 2: Demographics in Erik's Paper


## Model 5: Factor Demographics
```

```
Vet_Form_5 <-
  Vetting ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))


## Model 7: Slightly simplified for the Appendix

Vet_Form_7 <- Vetting ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Vet_1 <- lm(Vet_Form_1,
             data = DF_Follow3,
             na.action=na.omit)

Vet_5 <- lm(Vet_Form_5,
             data = DF_Follow3,
             na.action=na.omit)

Vet_7 <- lm(Vet_Form_7,
             data = DF_Follow3,
             na.action=na.omit)

library(stargazer)


stargazer(Vet_1, Vet_5, Vet_7, title = "Perceived Vetting with Covariates", no.space = TRUE)
```

```
#Credibility and vetting



Cred_Alone <- as.data.frame(matrix(data = c("credibility", coeftest(Cred_5)[2, 1:2],"Yes",
                                    "credibility", coeftest(Cred_1)[2, 1:2], "No",
                                    "vetting", coeftest(Vet_5)[2, 1:2],"Yes",
                                    "vetting", coeftest(Vet_1)[2, 1:2], "No"),
                                    ncol = 4, byrow = TRUE))

colnames(Cred_Alone) <- c("dv", "estimate", "se", "controls")

#making into correct operators
Cred_Alone$dv <- factor(Cred_Alone$dv, levels = c("credibility", "vetting"))
Cred_Alone$estimate <- as.numeric(Cred_Alone$estimate)
Cred_Alone$se <- as.numeric(Cred_Alone$se)

#adding in CIs
```

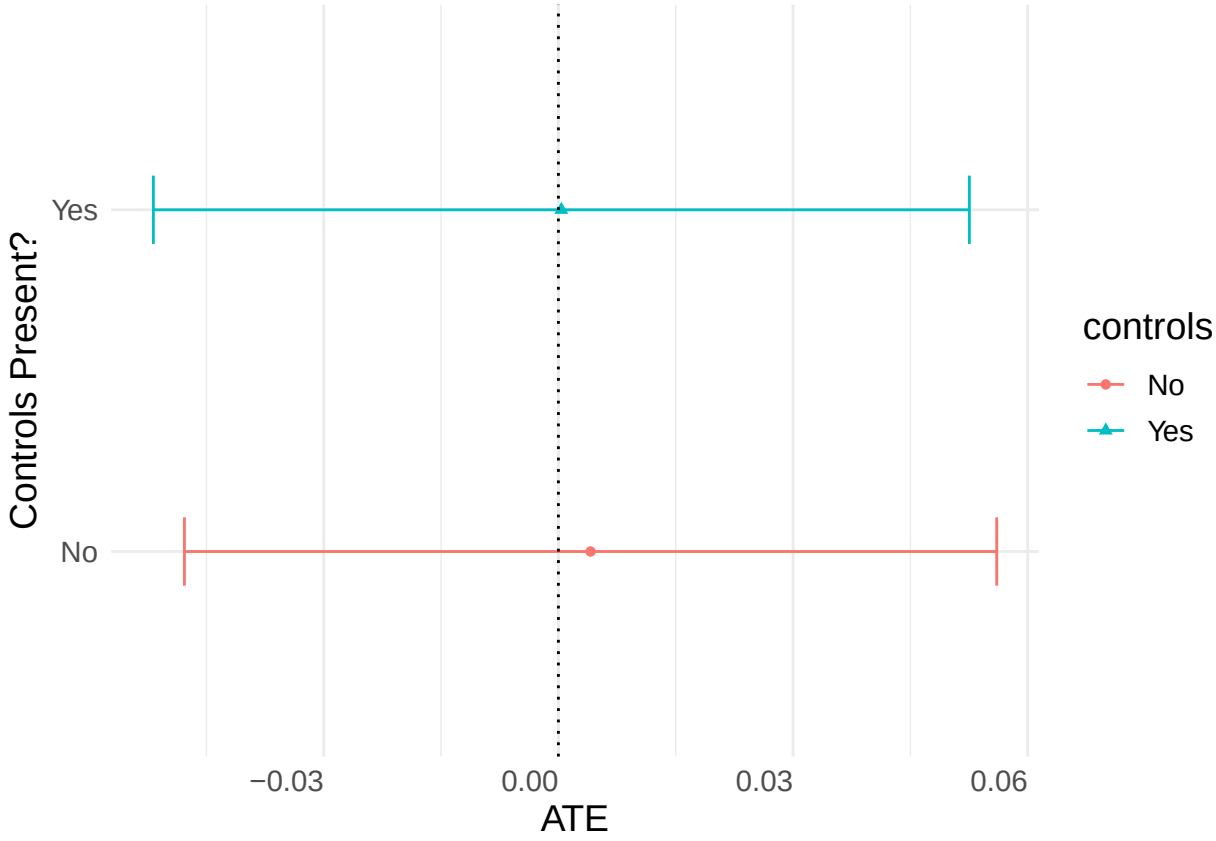
```
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Cred_Alone$ci <- Cred_Alone$se*q

library(ggplot2)
## graph time
pd <- position_dodge(0.5)

ggplot(Cred_Alone, aes(x = dv, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Question") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```
#ggtitle("Credibility, ATE of Tweet Medium, Iran")
```

```
#Credibility and vetting



Cred_Alone <- as.data.frame(matrix(data = c("vetting", coeftest(Vet_5)[2, 1:2],"Yes",
                                    "vetting", coeftest(Vet_1)[2, 1:2], "No"),
                                    ncol = 4, byrow = TRUE))

colnames(Cred_Alone) <- c("dv", "estimate", "se", "controls")

#making into correct operators
```

```
Cred_Alone$dv <- factor(Cred_Alone$dv, levels = c("vetting"))
Cred_Alone$estimate <- as.numeric(Cred_Alone$estimate)
Cred_Alone$se <- as.numeric(Cred_Alone$se)

#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Cred_Alone$ci <- Cred_Alone$se*q

library(ggplot2)
## graph time
pd <- position_dodge(0.5)

ggplot(Cred_Alone, aes(x = controls, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Controls Present?") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```
#ggtitle("Credibility, ATE of Tweet Medium, Iran")
```

Manipulation Check

```
# Target Check

## Model 1: No Demographics
```

```r
Target_Form_1 <- Target_Check_Binary ~ Tweet


## Model 5: Factor Demographics

Target_Form_5 <- Target_Check_Binary ~ Tweet +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))




## Model 7: Slightly simplified for the Appendix

Target_Form_7 <- Target_Check_Binary ~ Tweet +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Target_1 <- lm(Target_Form_1,
               data = DF_Follow3,
               na.action=na.omit)

Target_5 <- lm(Target_Form_5,
               data = DF_Follow3,
               na.action=na.omit)

Target_7 <- lm(Target_Form_7,
               data = DF_Follow3,
               na.action=na.omit)


stargazer(Target_1, Target_5, Target_7, title = "Target Check Question with Covariates", no.space = TRU
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:42
## \begin{table}[!htbp] \centering
##   \caption{Target Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Target\_Check\_Binary} \\
```

```
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & 0.042 & 0.032 & 0.038 \\
##   & (0.029) & (0.029) & (0.029) \\
##   Q.Female &  & $-$0.003 & $-$0.0003 \\
##   &  & (0.030) & (0.030) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.118$^{**}$ & $-$0.135$^{***}$ \\
##   &  & (0.046) & (0.046) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & 0.087 & 0.043 \\
##   &  & (0.179) & (0.180) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & $-$0.147$^{**}$ & $-$0.149$^{**}$ \\
##   &  & (0.067) & (0.067) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & $-$0.014 & $-$0.009 \\
##   &  & (0.272) & (0.274) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & $-$0.069 & $-$0.065 \\
##   &  & (0.052) & (0.052) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & $-$0.134 & $-$0.121 \\
##   &  & (0.111) & (0.111) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.275$^{*}$ & 0.311$^{**}$ \\
##   &  & (0.150) & (0.150) \\
##   Q.Age &  & 0.0001 & 0.0001 \\
##   &  & (0.0001) & (0.0001) \\
##   Q.HighSchool &  & 0.043 &  \\
##   &  & (0.077) &  \\
##   Q.Bach &  & $-$0.002 & $-$0.001 \\
##   &  & (0.034) & (0.034) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & $-$0.022 &  \\
##   &  & (0.056) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & $-$0.112$^{**}$ &  \\
##   &  & (0.046) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.058 &  \\
##   &  & (0.049) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & 0.087 &
##   &  & (0.056) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & 0.050 &
##   &  & (0.054) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & 0.010 &
##   &  & (0.048) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & $-$0.0
##   &  & (0.044) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.103$^{**}$ &  \\
##   &  & (0.043) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & 0.074$^{*}$ &  \\
##   &  & (0.039) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.069$^{*}$ &  \\
##   &  & (0.036) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & 0.012 &
##   &  & (0.029) &  \\
##   Q.Income &  &  & 0.004 \\
##   &  &  & (0.008) \\
##   Q.Conservative &  &  & 0.001 \\
##   &  &  & (0.035) \\
##   Q.Liberal &  &  & 0.049 \\
##   &  &  & (0.036) \\
```

```
##   Q.Veteran &  & $-$0.014 & 0.001 \\
##   &  & (0.056) & (0.056) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.121$^{***}$ &  \\
##   &  & (0.042) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.060 &  \\
##   &  & (0.040) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.039 &  \\
##   &  & (0.037) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.084$
##   &  & (0.038) &  \\
##   Q.Twitter\_Some &  &  & $-$0.030 \\
##   &  &  & (0.030) \\
##   Constant & 0.641$^{***}$ & 0.605$^{***}$ & 0.655$^{***}$ \\
##   & (0.020) & (0.081) & (0.043) \\
##  \hline \\[-1.8ex]
## Observations & 1,099 & 1,089 & 1,089 \\
## R$^{2}$ & 0.002 & 0.050 & 0.024 \\
## Adjusted R$^{2}$ & 0.001 & 0.025 & 0.010 \\
## Residual Std. Error & 0.473 (df = 1097) & 0.468 (df = 1060) & 0.471 (df = 1072) \\
## F Statistic & 2.217 (df = 1; 1097) & 2.007$^{***}$ (df = 28; 1060) & 1.678$^{**}$ (df = 16; 1072) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```
# Support Check

## Model 1: No Demographics

Support_Form_1 <- Support_Check_Binary ~ Tweet


## Model 5: Factor Demographics

Support_Form_5 <- Support_Check_Binary ~ Tweet +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))




## Model 7: Slightly simplified for the Appendix

Support_Form_7 <- Support_Check_Binary ~ Tweet +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some
```

```r
#regression

Support_1 <- lm(Support_Form_1,
             data = DF_Follow3,
             na.action=na.omit)

Support_5 <- lm(Support_Form_5,
             data = DF_Follow3,
             na.action=na.omit)

Support_7 <- lm(Support_Form_7,
             data = DF_Follow3,
             na.action=na.omit)


stargazer(Support_1, Support_5, Support_7, title = "Support Check Question with Covariates", no.space =
```

```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:43
## \begin{table}[!htbp] \centering
##   \caption{Support Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Support\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & $-$0.030 & $-$0.035 & $-$0.026 \\
##   & (0.029) & (0.029) & (0.029) \\
##   Q.Female &  & $-$0.093$^{***}$ & $-$0.092$^{***}$ \\
##   &  & (0.030) & (0.030) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.120$^{***}$ & $-$0.130$^{***}$ \\
##   &  & (0.046) & (0.046) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & $-$0.016 & $-$0.040 \\
##   &  & (0.179) & (0.179) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & 0.070 & 0.076 \\
##   &  & (0.067) & (0.067) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & 0.258 & 0.284 \\
##   &  & (0.272) & (0.273) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & 0.011 & 0.012 \\
##   &  & (0.052) & (0.052) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & $-$0.187$^{*}$ & $-$0.187$^{*}$ \\
##   &  & (0.111) & (0.111) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.029 & 0.044 \\
##   &  & (0.150) & (0.150) \\
##   Q.Age &  & $-$0.0001 & $-$0.0001 \\
##   &  & (0.0001) & (0.0001) \\
##   Q.HighSchool &  & 0.238$^{***}$ &  \\
##   &  & (0.077) &  \\
```

```
##   Q.Bach &  & 0.066$^{*}$ & 0.072$^{**}$ \\
##   &  & (0.034) & (0.034) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & 0.040 &  \\
##   &  & (0.056) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & 0.007 &  \\
##   &  & (0.046) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & 0.066 &  \\
##   &  & (0.049) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & 0.048 &
##   &  & (0.056) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & $-$0.0
##   &  & (0.054) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & 0.017 &
##   &  & (0.048) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & $-$0.0
##   &  & (0.044) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.089$^{**}$ &  \\
##   &  & (0.044) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & 0.008 &  \\
##   &  & (0.039) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.055 &  \\
##   &  & (0.036) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.025
##   &  & (0.029) &  \\
##   Q.Income &  &  & 0.012 \\
##   &  &  & (0.008) \\
##   Q.Conservative &  &  & 0.007 \\
##   &  &  & (0.035) \\
##   Q.Liberal &  &  & 0.054 \\
##   &  &  & (0.035) \\
##   Q.Veteran &  & 0.013 & 0.019 \\
##   &  & (0.056) & (0.056) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 0.004 &  \\
##   &  & (0.042) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & 0.034 &  \\
##   &  & (0.040) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.033 &  \\
##   &  & (0.037) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.020 &
##   &  & (0.038) &  \\
##   Q.Twitter\_Some &  &  & $-$0.003 \\
##   &  &  & (0.030) \\
##   Constant & 0.668$^{***}$ & 0.489$^{***}$ & 0.648$^{***}$ \\
##   & (0.020) & (0.081) & (0.043) \\
##  \hline \\[-1.8ex]
## Observations & 1,099 & 1,089 & 1,089 \\
## R$^{2}$ & 0.001 & 0.060 & 0.042 \\
## Adjusted R$^{2}$ & 0.0001 & 0.035 & 0.028 \\
## Residual Std. Error & 0.476 (df = 1097) & 0.468 (df = 1060) & 0.470 (df = 1072) \\
## F Statistic & 1.113 (df = 1; 1097) & 2.424$^{***}$ (df = 28; 1060) & 2.964$^{***}$ (df = 16; 1072) \
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
```

```
## \end{table}
```

```
# Sea Check

## Model 1: No Demographics

Sea_Form_1 <- Sea_Check_Binary ~ Tweet


## Model 5: Factor Demographics

Sea_Form_5 <- Sea_Check_Binary ~ Tweet +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))




## Model 7: Slightly simplified for the Appendix

Sea_Form_7 <- Sea_Check_Binary ~ Tweet +
  Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Sea_1 <- lm(Sea_Form_1,
            data = DF_Follow3,
            na.action=na.omit)

Sea_5 <- lm(Sea_Form_5,
            data = DF_Follow3,
            na.action=na.omit)

Sea_7 <- lm(Sea_Form_7,
            data = DF_Follow3,
            na.action=na.omit)


stargazer(Sea_1, Sea_5, Sea_7, title = "Sea Check Question with Covariates", no.space = TRUE)
```

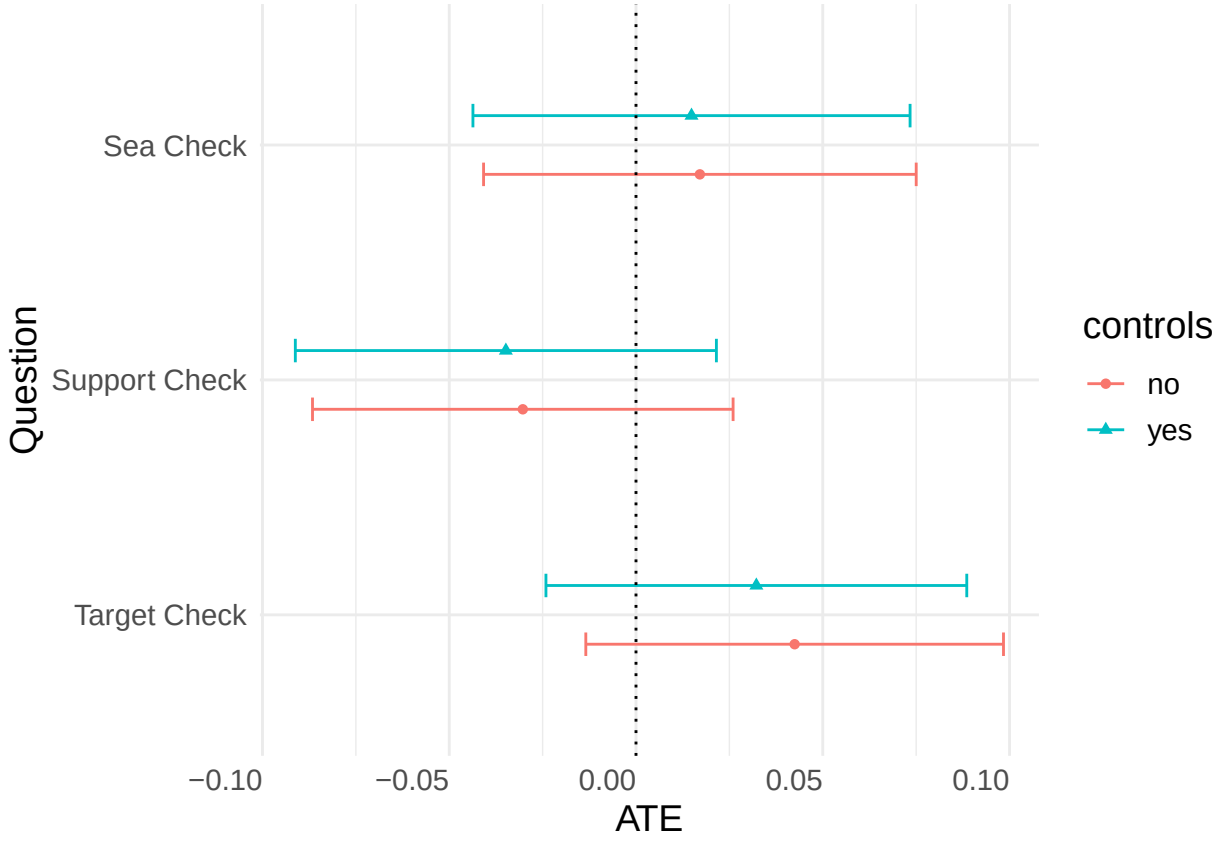
```
##
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac a
## % Date and time: Sun, Dec 31, 2023 - 14:16:43
## \begin{table}[!htbp] \centering
##   \caption{Sea Check Question with Covariates}
##   \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccc}
```

```
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
##  & \multicolumn{3}{c}{\textit{Dependent variable:}} \\
## \cline{2-4}
## \\[-1.8ex] & \multicolumn{3}{c}{Sea\_Check\_Binary} \\
## \\[-1.8ex] & (1) & (2) & (3)\\
## \hline \\[-1.8ex]
##  Tweet & 0.017 & 0.015 & 0.016 \\
##   & (0.030) & (0.030) & (0.030) \\
##   Q.Female &  & 0.060$^{*}$ & 0.060$^{*}$ \\
##   &  & (0.031) & (0.031) \\
##   relevel(as.factor(Q.Race), ref = 1)2 &  & $-$0.125$^{***}$ & $-$0.127$^{***}$ \\
##   &  & (0.048) & (0.048) \\
##   relevel(as.factor(Q.Race), ref = 1)3 &  & $-$0.055 & $-$0.067 \\
##   &  & (0.186) & (0.185) \\
##   relevel(as.factor(Q.Race), ref = 1)4 &  & 0.065 & 0.068 \\
##   &  & (0.070) & (0.069) \\
##   relevel(as.factor(Q.Race), ref = 1)5 &  & 0.473$^{*}$ & 0.482$^{*}$ \\
##   &  & (0.283) & (0.282) \\
##   relevel(as.factor(Q.Race), ref = 1)6 &  & $-$0.038 & $-$0.033 \\
##   &  & (0.054) & (0.053) \\
##   relevel(as.factor(Q.Race), ref = 1)7 &  & 0.070 & 0.055 \\
##   &  & (0.115) & (0.114) \\
##   relevel(as.factor(Q.Race), ref = 1)8 &  & 0.196 & 0.192 \\
##   &  & (0.155) & (0.155) \\
##   Q.Age &  & $-$0.0001 & $-$0.0001 \\
##   &  & (0.0001) & (0.0001) \\
##   Q.HighSchool &  & $-$0.013 &  \\
##   &  & (0.079) &  \\
##   Q.Bach &  & $-$0.060$^{*}$ & $-$0.056 \\
##   &  & (0.035) & (0.035) \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).L &  & 0.018 &  \\
##   &  & (0.058) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).Q &  & $-$0.081$^{*}$ &  \\
##   &  & (0.048) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)).C &  & $-$0.054 &  \\
##   &  & (0.051) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$4 &  & 0.048 &
##   &  & (0.058) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$5 &  & 0.039 &
##   &  & (0.056) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$6 &  & $-$0.0
##   &  & (0.050) &  \\
##   factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8))$\hat{\mkern6mu}$7 &  & $-$0.0
##   &  & (0.046) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & $-$0.014 &  \\
##   &  & (0.045) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.063 &  \\
##   &  & (0.040) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & $-$0.006 &  \\
##   &  & (0.037) &  \\
##   factor(Q.Political\_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.052
##   &  & (0.030) &  \\
##   Q.Income &  &  & 0.007 \\
```

```
##   &  &  & (0.008) \\
##   Q.Conservative &  &  & 0.016 \\
##   &  &  & (0.036) \\
##   Q.Liberal &  &  & 0.025 \\
##   &  &  & (0.037) \\
##   Q.Veteran &  & $-$0.160$^{***}$ & $-$0.154$^{***}$ \\
##   &  & (0.058) & (0.057) \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).L &  & 0.034 &  \\
##   &  & (0.043) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).Q &  & $-$0.025 &  \\
##   &  & (0.042) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5)).C &  & 0.017 &  \\
##   &  & (0.039) &  \\
##   factor(Q.Twitter\_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))$\hat{\mkern6mu}$4 &  & $-$0.020 &
##   &  & (0.040) &  \\
##   Q.Twitter\_Some &  &  & 0.055$^{*}$ \\
##   &  &  & (0.031) \\
##   Constant & 0.594$^{***}$ & 0.642$^{***}$ & 0.548$^{***}$ \\
##   & (0.021) & (0.084) & (0.044) \\
##  \hline \\[-1.8ex]
## Observations & 1,099 & 1,089 & 1,089 \\
## R$^{2}$ & 0.0003 & 0.040 & 0.031 \\
## Adjusted R$^{2}$ & $-$0.001 & 0.015 & 0.016 \\
## Residual Std. Error & 0.490 (df = 1097) & 0.486 (df = 1060) & 0.485 (df = 1072) \\
## F Statistic & 0.335 (df = 1; 1097) & 1.576$^{**}$ (df = 28; 1060) & 2.138$^{***}$ (df = 16; 1072) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:}  & \multicolumn{3}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
```

```r
## more detail on manipulation check

## ATE of Tweet Treatment

#put all the names, estimates, SEs, and behavior into one df for graphing
Results_DF <- as.data.frame(matrix(data = c("Target Check", coeftest(Target_1)[2, 1:2],"no",
                                    "Target Check", coeftest(Target_5)[2, 1:2], "yes",
                                    "Support Check", coeftest(Support_1)[2, 1:2], "no",
                                    "Support Check", coeftest(Support_5)[2, 1:2], "yes",
                                    "Sea Check", coeftest(Sea_1)[2, 1:2], "no",
                                    "Sea Check", coeftest(Sea_5)[2, 1:2], "yes"),
                                    ncol = 4, byrow = TRUE))

colnames(Results_DF) <- c("dv", "estimate", "se", "controls")

#making into correct operators
Results_DF$dv <- factor(Results_DF$dv, levels = c("Target Check", "Support Check", "Sea Check"))
Results_DF$estimate <- as.numeric(Results_DF$estimate)
Results_DF$se <- as.numeric(Results_DF$se)

#adding in CIs
q <- as.numeric(qnorm(p=.05/2, lower.tail=FALSE))
Results_DF$ci <- Results_DF$se*q
```

```
## graph time
library(ggplot2)
pd <- position_dodge(0.5)

ggplot(Results_DF, aes(x = dv, y = estimate, color = controls))  +
  geom_point(aes(color=controls, shape=controls), position = pd) +
  geom_errorbar(aes(ymin = estimate - ci, ymax = estimate + ci), width = .2, position = pd) +
  theme_minimal() + xlab("Question") + ylab("ATE") +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(hjust = 1), text = element_text(size = 14)) + coord_flip()
```




```
#ggtitle("Tweet Imagery ATE by MC or AC Question") +
```

```
## mean and se tables

##### Target Check
Target_Full_1 <- Target_Check_Binary ~ 0 + as.factor(Tweet)

lm_Target_Full_1 <- lm(Target_Full_1,
                       data = DF_Follow3)

summary(lm_Target_Full_1)
```

```
##
## Call:
## lm(formula = Target_Full_1, data = DF_Follow3)
```

```
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6836 -0.6412  0.3164  0.3588  0.3588
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.64117    0.02018   31.77   <2e-16 ***
## as.factor(Tweet)1  0.68364    0.02016   33.91   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4728 on 1097 degrees of freedom
## Multiple R-squared:  0.6631, Adjusted R-squared:  0.6625
## F-statistic:  1080 on 2 and 1097 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_Follow3$Target_Check_Binary[DF_Follow3$Tweet==1]))
```

```
## [1] 550
```

```
length(na.omit(DF_Follow3$Target_Check_Binary[DF_Follow3$Tweet==0]))
```

```
## [1] 549
```

```
##### Support Check
Support_Full_1 <- Support_Check_Binary ~ 0 + as.factor(Tweet)

lm_Support_Full_1 <- lm(Support_Full_1,
                    data = DF_Follow3)

summary(lm_Support_Full_1)
```

```
##
## Call:
## lm(formula = Support_Full_1, data = DF_Follow3)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6685 -0.6382  0.3315  0.3618  0.3618
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.66849    0.02032   32.90   <2e-16 ***
## as.factor(Tweet)1  0.63818    0.02030   31.44   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4761 on 1097 degrees of freedom
## Multiple R-squared:  0.6537, Adjusted R-squared:  0.653
## F-statistic:  1035 on 2 and 1097 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_Follow3$Support_Check_Binary[DF_Follow3$Tweet==1]))
```

```
## [1] 550
```

```
length(na.omit(DF_Follow3$Support_Check_Binary[DF_Follow3$Tweet==0]))
```

```
## [1] 549
```

```
##### Sea Check
Sea_Full_1 <- Sea_Check_Binary ~ 0 + as.factor(Tweet)

lm_Sea_Full_1 <- lm(Sea_Full_1,
                    data = DF_Follow3)

summary(lm_Sea_Full_1)
```

```
##
## Call:
## lm(formula = Sea_Full_1, data = DF_Follow3)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -0.6109 -0.5938  0.3891  0.4062  0.4062
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.59381    0.02090   28.41   <2e-16 ***
## as.factor(Tweet)1  0.61091    0.02088   29.25   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4898 on 1097 degrees of freedom
## Multiple R-squared:  0.6025, Adjusted R-squared:  0.6018
## F-statistic: 831.3 on 2 and 1097 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_Follow3$Sea_Check_Binary[DF_Follow3$Tweet==1]))
```

```
## [1] 550
```

```
length(na.omit(DF_Follow3$Sea_Check_Binary[DF_Follow3$Tweet==0]))
```

```
## [1] 549
```

AC Sensitivity Analysis

Just three ACs

```
DF4 <- subset(DF_Follow3, DF_Follow3$Target_Check_Binary==1)
DF5 <- subset(DF4, DF4$Support_Check_Binary==1)
DF6 <- subset(DF5, DF5$Sea_Check_Binary==1)
```

Subset

Analysis

```
## a simple regression for mean and se estimates


######## Credibility


AC.Cred_Full_1 <- Credibility ~ 0 + as.factor(Tweet)
```

```
AC.lm_Cred_Full_1 <- lm(AC.Cred_Full_1,
                       data = DF6)

summary(AC.lm_Cred_Full_1)
```

Credibility

```
##
## Call:
## lm(formula = AC.Cred_Full_1, data = DF6)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -2.0603 -1.0603 -0.0098  0.9902  1.9902
##
## Coefficients:
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  3.06030    0.09075   33.72   <2e-16 ***
## as.factor(Tweet)1  3.00980    0.08964   33.58   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.28 on 401 degrees of freedom
## Multiple R-squared:  0.8496, Adjusted R-squared:  0.8488
## F-statistic:  1132 on 2 and 401 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF6$Credibility[DF6$Tweet==1]))
```

```
## [1] 204
```

```
length(na.omit(DF6$Credibility[DF6$Tweet==0]))
```

```
## [1] 199
```

```
#time for some regressions

#let's do a couple different models


##DV: credibility, IV: Tweet binary

## Model 1: No Demographics

AC.Cred_Form_1 <- Credibility ~ Tweet

## Model 2: Demographics in Erik's Paper


## Model 5: Factor Demographics

AC.Cred_Form_5 <-
  Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
```

```
  Q.Veteran


## Model 7: Slightly simplified for the Appendix

AC.Cred_Form_7 <- Credibility ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran

#regression

AC.Cred_1 <- lm(AC.Cred_Form_1,
              data = DF6,
              na.action=na.omit)

AC.Cred_5 <- lm(AC.Cred_Form_5,
              data = DF6,
              na.action=na.omit)

AC.Cred_7 <- lm(AC.Cred_Form_7,
              data = DF6,
              na.action=na.omit)

library(stargazer)


stargazer(AC.Cred_1, AC.Cred_5, AC.Cred_7, title = "Perceived Credibility, AC Questions Correct", no.spa
```

```
## a simple regression for mean and se estimates


######## Vetting


Vet_Full_1 <- Vetting ~ 0 + as.factor(Tweet)
lm_Vet_Full_1 <- lm(Vet_Full_1,
                    data = DF6)

summary(lm_Vet_Full_1)
```

Vetting

```
##
## Call:
## lm(formula = Vet_Full_1, data = DF6)
##
## Residuals:
##    Min     1Q Median     3Q    Max
## -0.848  0.152  0.152  0.196  0.196
##
## Coefficients:
```

```
##                   Estimate Std. Error t value Pr(>|t|)
## as.factor(Tweet)0  0.80402    0.02688   29.91   <2e-16 ***
## as.factor(Tweet)1  0.84804    0.02655   31.95   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3792 on 401 degrees of freedom
## Multiple R-squared:  0.8269, Adjusted R-squared:  0.826
## F-statistic: 957.7 on 2 and 401 DF,  p-value: < 2.2e-16
```

```
length(na.omit(DF_Follow3$Vetting[DF_Follow3$Tweet==1]))
```

```
## [1] 550
```

```
length(na.omit(DF_Follow3$Vetting[DF_Follow3$Tweet==0]))
```

```
## [1] 549
```

```
#time for some regressions

#let's do a couple different models

##DV: Vetting, IV: Tweet binary

## Model 1: No Demographics

Vet_Form_1 <- Vetting ~ Tweet

## Model 2: Demographics in Erik's Paper


## Model 5: Factor Demographics

Vet_Form_5 <-
  Vetting ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.HighSchool + Q.Bach +
  factor(Q.Income, ordered = TRUE, levels = c(1, 2, 3, 4, 5, 6, 7, 8)) +  #treats income as an ordered
  factor(Q.Political_ID, ordered = TRUE, levels = c(1, 2, 3, 4, 5)) + #treats party as an ordered facto
  Q.Veteran + factor(Q.Twitter_Use, ordered = TRUE, levels = c(1, 2, 3, 4, 5))


## Model 7: Slightly simplified for the Appendix

Vet_Form_7 <- Vetting ~ Tweet + Q.Female +
  relevel(as.factor(Q.Race), ref = 1) + #makes white the reference cat
  Q.Age + Q.Bach + Q.Income +  #treats income as a continuous variable, removes high school dummy
  Q.Conservative + Q.Liberal + #uses party binaries
  Q.Veteran + Q.Twitter_Some

#regression

Vet_1 <- lm(Vet_Form_1,
            data = DF6,
            na.action=na.omit)
```

```
Vet_5 <- lm(Vet_Form_5,
             data = DF6,
             na.action=na.omit)

Vet_7 <- lm(Vet_Form_7,
             data = DF6,
             na.action=na.omit)

library(stargazer)


stargazer(Vet_1, Vet_5, Vet_7, title = "Perceived Vetting, All AC Questions Correct", no.space = TRUE)
```

Convert File