

Main Analysis

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TODO

1. Find other measures of polarization
 - Cook, redistrict scores of how far learning
2. introduce more controls
 - twitter followers [done]
 - state under 45 population [done]

Loading Data

```
dataset <- read.csv(file = "../python_env/data/final_dataset.csv", header = TRUE)
dataset$is_attack<- as.integer(as.logical(dataset$is_attack))
dataset$is_policy<- as.integer(as.logical(dataset$is_policy))
dataset$is_contrast<- as.integer(as.logical(dataset$is_contrast))
dataset$incumbency<- as.integer(as.logical(dataset$incumbency))
dataset$competitiveness<- as.integer(as.logical(dataset$competitiveness))

# convert party label
convert_party <- function(p) {
  if(p == "R") {
    return("R");
  } else if (p == "D") {
    return("D");
  } else {
    return("T");
  }
}

dataset$party <- as.factor(sapply(dataset$party, convert_party))
dataset <- na.omit(dataset)
#View(dataset)
```

Group Data

```
getmode <- function(v) {
  uniqv <- unique(v)
  uniqv[which.max(tabulate(match(v, uniqv)))]
}
```

```

get_viciousness <- function(att, pol, con) {
  print(att);
  if (att == 0) {
    return(0);
  }

  if (pol == 1) {
    if (con == 1) {
      return(1);
    } else {
      return(2);
    }
  }
  return(3);
}

dataset <- dataset %>% mutate(viciousness = is_attack * 3 - is_policy - is_contrast)

relevent_set <- dataset %>%
  filter(party != "T") %>%
  group_by(user) %>%
  summarise(
    # controls
    "gender" = factor(mean(gender)),
    "incumbency" = factor(mean(incumbency)),
    "competitiveness" = factor(mean(competitiveness)),
    "party" = factor(getmode(party)),
    "follower_count" = mean(follower_count),
    "young_percent" = mean(percentAge20To44),
    "total_tweet_count" = log(n() + 0.00001),
    # IVs: elite
    "s_dist" = mean(s_dist),
    "h_dist" = mean(h_dist),
    "total_dist" = mean(s_dist) + mean(h_dist),
    # IVs: mass
    "ideoConsistVariance" = mean(ideoConsistVariance),
    "ideoFirstDiff" = mean(ideoFirstDiff),
    # DVs: PTVI
    "attack_per" = mean(is_attack),
    "viciousness" = mean(viciousness),
    "total_attck_count" = log(sum(is_attack) + 0.00001),
    # DVs: affects
    "anticipation" = mean(anticipation),
    "positive" = mean(positive),
    "joy" = mean(joy),
    "trust" = mean(trust),
    "surprise" = mean(surprise),
    "anger" = mean(anger),
    "sadness" = mean(sadness),
    "negative" = mean(negative),
    "fear" = mean(fear),
    "disgust" = mean(disgust))

```

Making Sure There is No Confounding Variables

```
correlations <- relevent_set %>% select(-c(attack_per, viciousness, total_attck_count, anticipation, pos
```

```
correlations
```

```
## # A tibble: 53 x 12
##   gender incumbency competitiveness party follower_count young_percent
##   <fct>  <fct>      <fct>          <fct>          <dbl>      <dbl>
## 1 1      0          1          R              8159        0.761
## 2 0      1          0          R             1608455.    0.788
## 3 1      0          0          R              62445.        0.799
## 4 1      1          0          R             140846.        0.807
## 5 1      0          1          R              3245         0.934
## 6 1      1          0          R             22574.         0.850
## 7 1      0          1          D              7271         0.946
## 8 1      0          1          D             32197.         0.701
## 9 1      1          1          D             41350.         0.722
## 10 1     0          0          R             68740.         0.844
## # ... with 43 more rows, and 6 more variables: total_tweet_count <dbl>,
## #   s_dist <dbl>, h_dist <dbl>, total_dist <dbl>, ideoConsistVariance <dbl>,
## #   ideoFirstDiff <dbl>
```

```
correlate(as.data.frame(correlations))
```

```
##
## CORRELATIONS
## =====
## - correlation type: pearson
## - correlations shown only when both variables are numeric
##
##           gender incumbency competitiveness party follower_count
## gender                .                .                .                .
## incumbency            .                .                .                .
## competitiveness      .                .                .                .
## party                 .                .                .                .
## follower_count        .                .                .                .
## young_percent         .                .                .                0.125
## total_tweet_count     .                .                .                -0.122
## s_dist                .                .                .                0.231
## h_dist                .                .                .                0.118
## total_dist            .                .                .                0.174
## ideoConsistVariance    .                .                .                -0.054
## ideoFirstDiff         .                .                .                -0.057
##
##           young_percent total_tweet_count s_dist h_dist total_dist
## gender                .                .                .                .
## incumbency            .                .                .                .
## competitiveness      .                .                .                .
## party                 .                .                .                .
## follower_count        0.125            -0.122 0.231 0.118 0.174
## young_percent         .                0.057 0.330 0.340 0.342
## total_tweet_count     0.057            . 0.158 0.121 0.141
## s_dist                0.330            0.158 . 0.926 0.979
## h_dist                0.340            0.121 0.926 . 0.984
## total_dist            0.342            0.141 0.979 0.984 .
```

```
## ideoConsistVariance      -0.115          0.006  0.437  0.361      0.405
## ideoFirstDiff            -0.097          0.104  0.534  0.431      0.489
##                ideoConsistVariance ideoFirstDiff
## gender                        .          .
## incumbency                    .          .
## competitiveness              .          .
## party                          .          .
## follower_count               -0.054      -0.057
## young_percent                -0.115      -0.097
## total_tweet_count            0.006       0.104
## s_dist                       0.437       0.534
## h_dist                       0.361       0.431
## total_dist                   0.405       0.489
## ideoConsistVariance          .          0.871
## ideoFirstDiff                0.871      .
```

Multivariable Regression

Political Tweet Viciousness Index (PTVI)

Model 1: Mass

```
model_at_idvar <- lm(attack_per ~ ideoConsistVariance + gender + incumbency + competitiveness + party +
model_v_idvar <- lm(viciousness ~ ideoConsistVariance + gender + incumbency + competitiveness + party +
model_at_iddist <- lm(attack_per ~ ideoFirstDiff + gender + incumbency + competitiveness + party + follow
model_v_iddist <- lm(viciousness ~ ideoFirstDiff + gender + incumbency + competitiveness + party + follow
```

```
print("Attack Percentage & Ideology Consistency Variance")
```

```
## [1] "Attack Percentage & Ideology Consistency Variance"
```

```
summary(model_at_idvar)
```

```
##
## Call:
## lm(formula = attack_per ~ ideoConsistVariance + gender + incumbency +
##      competitiveness + party + follower_count + young_percent +
##      total_tweet_count, data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.115410 -0.048144 -0.001426  0.043340  0.141021
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -2.854e-03  1.168e-01  -0.024  0.980622
## ideoConsistVariance -2.573e-03  2.095e-03  -1.228  0.225963
## gender0        -3.040e-02  2.852e-02  -1.066  0.292310
## incumbency1    -3.394e-02  2.143e-02  -1.584  0.120308
## competitiveness0 -2.269e-02  2.142e-02  -1.059  0.295236
## partyD         2.200e-03  1.924e-02   0.114  0.909466
## follower_count  -9.632e-09  3.355e-08  -0.287  0.775401
## young_percent    7.625e-03  9.731e-02   0.078  0.937897
## total_tweet_count  3.176e-02  8.671e-03   3.663  0.000666 ***
## ---
```

```

## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.06727 on 44 degrees of freedom
## Multiple R-squared:  0.4239, Adjusted R-squared:  0.3192
## F-statistic: 4.047 on 8 and 44 DF,  p-value: 0.00113
print("PTVI & Ideology Consistency Variance")

## [1] "PTVI & Ideology Consistency Variance"
summary(model_v_idvar)

##
## Call:
## lm(formula = viciousness ~ ideoConsistVariance + gender + incumbency +
##     competitiveness + party + follower_count + young_percent +
##     total_tweet_count, data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.28587 -0.12734 -0.00053  0.08887  0.36109
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -1.387e-02  2.952e-01  -0.047  0.962736
## ideoConsistVariance -6.054e-03  5.294e-03  -1.144  0.258958
## gender0        -8.163e-02  7.206e-02  -1.133  0.263423
## incumbency1    -9.292e-02  5.413e-02  -1.716  0.093111 .
## competitiveness -4.752e-02  5.413e-02  -0.878  0.384721
## partyD         6.793e-03  4.861e-02   0.140  0.889491
## follower_count -2.705e-08  8.477e-08  -0.319  0.751192
## young_percent   6.136e-03  2.459e-01   0.025  0.980202
## total_tweet_count 8.082e-02  2.191e-02   3.689  0.000616 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.17 on 44 degrees of freedom
## Multiple R-squared:  0.4292, Adjusted R-squared:  0.3254
## F-statistic: 4.136 on 8 and 44 DF,  p-value: 0.0009531
print("Attack Percentage & Ideology Consistency First Difference")

## [1] "Attack Percentage & Ideology Consistency First Difference"
summary(model_at_iddist)

##
## Call:
## lm(formula = attack_per ~ ideoFirstDiff + gender + incumbency +
##     competitiveness + party + follower_count + young_percent +
##     total_tweet_count, data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.114153 -0.050330 -0.007159  0.043934  0.157643
##
## Coefficients:

```

```
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -6.733e-02  1.095e-01  -0.615  0.54179
## ideoFirstDiff  -4.021e-03  9.799e-03  -0.410  0.68357
## gender0        -3.368e-02  2.894e-02  -1.164  0.25088
## incumbency1    -3.251e-02  2.178e-02  -1.493  0.14260
## competitiveness0 -1.776e-02  2.132e-02  -0.833  0.40923
## partyD         2.161e-03  1.953e-02   0.111  0.91241
## follower_count -9.671e-09  3.406e-08  -0.284  0.77781
## young_percent   1.482e-02  9.874e-02   0.150  0.88133
## total_tweet_count 3.268e-02  8.825e-03   3.703  0.00059 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.06829 on 44 degrees of freedom
## Multiple R-squared:  0.4064, Adjusted R-squared:  0.2985
## F-statistic: 3.766 on 8 and 44 DF,  p-value: 0.001948
```

```
print("PTVI & Ideology Consistency First Difference")
```

```
## [1] "PTVI & Ideology Consistency First Difference"
```

```
summary(model_v_iddist)
```

```
##
## Call:
## lm(formula = viciousness ~ ideoFirstDiff + gender + incumbency +
##     competitiveness + party + follower_count + young_percent +
##     total_tweet_count, data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.28318 -0.13485 -0.02111  0.10256  0.39134
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -1.713e-01  2.762e-01  -0.620  0.538372
## ideoFirstDiff  -8.468e-03  2.472e-02  -0.343  0.733558
## gender0        -8.915e-02  7.301e-02  -1.221  0.228571
## incumbency1    -8.968e-02  5.493e-02  -1.633  0.109684
## competitiveness0 -3.570e-02  5.377e-02  -0.664  0.510220
## partyD         6.708e-03  4.926e-02   0.136  0.892305
## follower_count -2.707e-08  8.592e-08  -0.315  0.754237
## young_percent   2.392e-02  2.491e-01   0.096  0.923934
## total_tweet_count 8.291e-02  2.226e-02   3.724  0.000554 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1722 on 44 degrees of freedom
## Multiple R-squared:  0.4138, Adjusted R-squared:  0.3072
## F-statistic: 3.883 on 8 and 44 DF,  p-value: 0.001553
```

Model 2: Elite

```
model_at_hdist <- lm(attack_per ~ h_dist + gender + incumbency + competitiveness + party + follower_count)
model_v_hdist <- lm(viciousness ~ h_dist + gender + incumbency + competitiveness + party + follower_count)
```

```

model_at_sdist <- lm(attack_per ~ s_dist + gender + incumbency + competitiveness + party + follower_count)
model_v_sdist <- lm(viciousness ~ s_dist + gender + incumbency + competitiveness + party + follower_count)
model_at_tdist <- lm(attack_per ~ total_dist + gender + incumbency + competitiveness + party + follower_count)
model_v_tdist <- lm(viciousness ~ total_dist + gender + incumbency + competitiveness + party + follower_count)

```

```
print("Attack Percentage & House Ideology Distance")
```

```
## [1] "Attack Percentage & House Ideology Distance"
```

```
summary(model_at_hdist)
```

```

##
## Call:
## lm(formula = attack_per ~ h_dist + gender + incumbency + competitiveness +
##     party + follower_count + young_percent + total_tweet_count,
##     data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.115590 -0.057152 -0.003737  0.045405  0.166838
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -8.692e-02  9.366e-02  -0.928  0.358424
## h_dist         2.000e-02  2.732e-02   0.732  0.468034
## gender0       -2.967e-02  2.909e-02  -1.020  0.313302
## incumbency1   -3.314e-02  2.165e-02  -1.531  0.132927
## competitiveness0 -1.578e-02  2.117e-02  -0.745  0.460091
## partyD         1.017e-03  1.951e-02   0.052  0.958686
## follower_count -1.262e-08  3.420e-08  -0.369  0.714000
## young_percent  -7.083e-03  1.039e-01  -0.068  0.945958
## total_tweet_count 3.179e-02  8.786e-03   3.618  0.000761 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.068 on 44 degrees of freedom
## Multiple R-squared:  0.4113, Adjusted R-squared:  0.3043
## F-statistic: 3.843 on 8 and 44 DF,  p-value: 0.001676

```

```
print("PTVI & House Ideology Distance")
```

```
## [1] "PTVI & House Ideology Distance"
```

```
summary(model_v_hdist)
```

```

##
## Call:
## lm(formula = viciousness ~ h_dist + gender + incumbency + competitiveness +
##     party + follower_count + young_percent + total_tweet_count,
##     data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.28645 -0.14127 -0.01258  0.09816  0.41493
##
## Coefficients:

```

```
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -2.100e-01  2.357e-01  -0.891  0.377847
## h_dist         5.655e-02  6.875e-02   0.823  0.415225
## gender0        -7.840e-02  7.322e-02  -1.071  0.290138
## incumbency1    -9.108e-02  5.448e-02  -1.672  0.101706
## competitiveness0 -3.073e-02  5.329e-02  -0.577  0.567126
## partyD         3.449e-03  4.912e-02   0.070  0.944336
## follower_count -3.560e-08  8.609e-08  -0.414  0.681191
## young_percent  -4.050e-02  2.615e-01  -0.155  0.877641
## total_tweet_count 8.062e-02  2.211e-02   3.646  0.000701 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1712 on 44 degrees of freedom
## Multiple R-squared:  0.4211, Adjusted R-squared:  0.3159
## F-statistic: 4.001 on 8 and 44 DF,  p-value: 0.001234
print("Attack Percentage & Senate Ideology Distance")

## [1] "Attack Percentage & Senate Ideology Distance"
summary(model_at_sdist)

##
## Call:
## lm(formula = attack_per ~ s_dist + gender + incumbency + competitiveness +
##     party + follower_count + young_percent + total_tweet_count,
##     data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.115882 -0.051002 -0.001499  0.046396  0.167301
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -8.747e-02  9.309e-02  -0.940  0.35252
## s_dist         3.174e-02  3.159e-02   1.005  0.32045
## gender0        -2.731e-02  2.914e-02  -0.937  0.35376
## incumbency1    -3.308e-02  2.153e-02  -1.536  0.13164
## competitiveness0 -1.688e-02  2.101e-02  -0.804  0.42593
## partyD         9.855e-04  1.938e-02   0.051  0.95968
## follower_count -1.806e-08  3.482e-08  -0.519  0.60660
## young_percent  -1.320e-02  1.024e-01  -0.129  0.89796
## total_tweet_count 3.092e-02  8.821e-03   3.505  0.00106 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.06764 on 44 degrees of freedom
## Multiple R-squared:  0.4175, Adjusted R-squared:  0.3116
## F-statistic: 3.943 on 8 and 44 DF,  p-value: 0.001382
print("PTVI & Senate Ideology Distance")

## [1] "PTVI & Senate Ideology Distance"
```



```
summary(model_v_sdist)
```

```
##
## Call:
## lm(formula = vicousness ~ s_dist + gender + incumbency + competitiveness +
##      party + follower_count + young_percent + total_tweet_count,
##      data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.28724 -0.13714 -0.01346  0.10373  0.41583
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -2.118e-01  2.341e-01  -0.904  0.370701
## s_dist         8.773e-02  7.945e-02   1.104  0.275485
## gender0       -7.207e-02  7.329e-02  -0.983  0.330770
## incumbency1   -9.090e-02  5.416e-02  -1.678  0.100354
## competitiveness0 -3.385e-02  5.284e-02  -0.641  0.525053
## partyD        3.439e-03  4.875e-02   0.071  0.944086
## follower_count -5.045e-08  8.759e-08  -0.576  0.567566
## young_percent  -5.579e-02  2.574e-01  -0.217  0.829453
## total_tweet_count 7.826e-02  2.219e-02   3.527  0.000996 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1701 on 44 degrees of freedom
## Multiple R-squared:  0.4281, Adjusted R-squared:  0.3241
## F-statistic: 4.117 on 8 and 44 DF,  p-value: 0.0009881
```

```
print("PTVI & Total Ideology Distance")
```

```
## [1] "PTVI & Total Ideology Distance"
```

```
summary(model_at_tdist)
```

```
##
## Call:
## lm(formula = attack_per ~ total_dist + gender + incumbency +
##      competitiveness + party + follower_count + young_percent +
##      total_tweet_count, data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.115732 -0.055537 -0.002525  0.045785  0.167409
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -8.692e-02  9.338e-02  -0.931  0.357032
## total_dist     1.307e-02  1.494e-02   0.875  0.386388
## gender0       -2.849e-02  2.912e-02  -0.978  0.333231
## incumbency1   -3.312e-02  2.159e-02  -1.534  0.132177
## competitiveness0 -1.616e-02  2.108e-02  -0.767  0.447387
## partyD        9.271e-04  1.945e-02   0.048  0.962199
## follower_count -1.507e-08  3.445e-08  -0.437  0.663932
```

```
## young_percent      -1.125e-02  1.034e-01  -0.109 0.913823
## total_tweet_count  3.140e-02  8.798e-03   3.569 0.000882 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.06783 on 44 degrees of freedom
## Multiple R-squared:  0.4143, Adjusted R-squared:  0.3079
## F-statistic: 3.891 on 8 and 44 DF,  p-value: 0.001527
print("PTVI & Total Ideology Distance")

## [1] "PTVI & Total Ideology Distance"
summary(model_v_tdist)

##
## Call:
## lm(formula = viciousness ~ total_dist + gender + incumbency +
##      competitiveness + party + follower_count + young_percent +
##      total_tweet_count, data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.28684 -0.14003 -0.01372  0.10095  0.41632
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -2.101e-01  2.350e-01  -0.894  0.376000
## total_dist     3.650e-02  3.758e-02   0.971  0.336680
## gender0       -7.521e-02  7.326e-02  -1.027  0.310255
## incumbency1   -9.102e-02  5.432e-02  -1.676  0.100928
## competitiveness0 -3.184e-02  5.304e-02  -0.600  0.551408
## partyD         3.240e-03  4.894e-02   0.066  0.947517
## follower_count -4.234e-08  8.667e-08  -0.489  0.627576
## young_percent  -5.127e-02  2.601e-01  -0.197  0.844653
## total_tweet_count  7.954e-02  2.214e-02   3.593  0.000819 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1707 on 44 degrees of freedom
## Multiple R-squared:  0.4246, Adjusted R-squared:  0.32
## F-statistic: 4.058 on 8 and 44 DF,  p-value: 0.001106
```

Affects

Disgust: - strong positive association with h_dist & s_dist - slight positive association with ideoConsistVariance - slight negative association with competitiveness0

Anger: - strong positive association with h_dist & s_dist

Negative: - slight positive association with h_dist & s_dist

Fear: - slight positive association with s_dist

Anticipation: - slight negative association with h_dist & s_dist

Positive: - strong positive association with incumbency1

Surprise: - slight positive association with incumbency1

Joy: - none

Trust: - none

Sadness: - none

```
summary(lm(anger ~ h_dist + gender + incumbency + competitiveness + party + follower_count + total_tweet_count, data = relevent_set))

##
## Call:
## lm(formula = anger ~ h_dist + gender + incumbency + competitiveness +
##     party + follower_count + total_tweet_count, data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.16040 -0.04804 -0.01002  0.04739  0.25104
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    4.450e-02  7.964e-02   0.559   0.5791
## h_dist         7.982e-02  3.438e-02   2.322   0.0248 *
## gender0        4.122e-03  3.881e-02   0.106   0.9159
## incumbency1     9.123e-03  2.882e-02   0.317   0.7530
## competitiveness0 -8.498e-04  2.808e-02  -0.030   0.9760
## partyD         2.102e-02  2.599e-02   0.809   0.4230
## follower_count  7.475e-08  4.549e-08   1.643   0.1073
## total_tweet_count 1.559e-02  1.172e-02   1.331   0.1900
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.09076 on 45 degrees of freedom
## Multiple R-squared:  0.2348, Adjusted R-squared:  0.1157
## F-statistic: 1.972 on 7 and 45 DF,  p-value: 0.0802
```

```
summary(lm(anger ~ s_dist + gender + incumbency + competitiveness + party + follower_count + total_tweet_count, data = relevent_set))

##
## Call:
## lm(formula = anger ~ s_dist + gender + incumbency + competitiveness +
##     party + follower_count + total_tweet_count, data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.154586 -0.050210 -0.009989  0.041249  0.262747
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    3.975e-02  7.855e-02   0.506   0.6153
## s_dist         1.024e-01  3.988e-02   2.568   0.0136 *
## gender0        9.296e-03  3.862e-02   0.241   0.8109
## incumbency1     9.402e-03  2.848e-02   0.330   0.7429
## competitiveness0 -5.260e-03  2.775e-02  -0.190   0.8505
## partyD         2.182e-02  2.566e-02   0.850   0.3996
## follower_count  5.964e-08  4.612e-08   1.293   0.2026
## total_tweet_count 1.322e-02  1.170e-02   1.131   0.2642
## ---
```

```

## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.0897 on 45 degrees of freedom
## Multiple R-squared:  0.2526, Adjusted R-squared:  0.1364
## F-statistic: 2.173 on 7 and 45 DF,  p-value: 0.0548
summary(lm(anger ~ ideoConsistVariance + gender + incumbency + competitiveness + party + follower_count

##
## Call:
## lm(formula = anger ~ ideoConsistVariance + gender + incumbency +
##      competitiveness + party + follower_count + total_tweet_count,
##      data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.162049 -0.064164 -0.006017  0.052976  0.233144
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    4.464e-02  1.208e-01   0.370   0.7134
## ideoConsistVariance  2.042e-03  2.964e-03   0.689   0.4944
## gender0        -1.151e-02  4.050e-02  -0.284   0.7776
## incumbency1     8.463e-03  3.034e-02   0.279   0.7816
## competitiveness0  1.598e-03  3.035e-02   0.053   0.9582
## partyD          2.492e-02  2.730e-02   0.913   0.3663
## follower_count   9.298e-08  4.722e-08   1.969   0.0551 .
## total_tweet_count  1.902e-02  1.229e-02   1.548   0.1285
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.09554 on 45 degrees of freedom
## Multiple R-squared:  0.152, Adjusted R-squared:  0.02014
## F-statistic: 1.153 on 7 and 45 DF,  p-value: 0.3486
summary(lm(anger ~ ideoFirstDiff + gender + incumbency + competitiveness + party + follower_count + tot

##
## Call:
## lm(formula = anger ~ ideoFirstDiff + gender + incumbency + competitiveness +
##      party + follower_count + total_tweet_count, data = relevent_set)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.163401 -0.059608 -0.007662  0.052258  0.235541
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    8.566e-02  1.056e-01   0.811   0.4215
## ideoFirstDiff   4.326e-03  1.372e-02   0.315   0.7540
## gender0        -8.644e-03  4.064e-02  -0.213   0.8325
## incumbency1     7.248e-03  3.053e-02   0.237   0.8134
## competitiveness0 -2.163e-03  2.984e-02  -0.072   0.9425
## partyD          2.499e-02  2.742e-02   0.912   0.3668
## follower_count   9.287e-08  4.743e-08   1.958   0.0565 .

```

```
## total_tweet_count 1.817e-02 1.236e-02 1.470 0.1485
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.09594 on 45 degrees of freedom
## Multiple R-squared:  0.145, Adjusted R-squared:  0.01199
## F-statistic: 1.09 on 7 and 45 DF, p-value: 0.3856
```