

Speaking Truth to Twitter

Team 3

Hertie School of Governance

May 12, 2016

Outline

- 1 Implementation
- 2 Descriptive Statistics
- 3 Results
- 4 Limitations
- 5 Conclusion

Outline

- 1 Implementation
- 2 Descriptive Statistics
- 3 Results
- 4 Limitations
- 5 Conclusion

Main Changes

- We only focussed on Trump, not Clinton
- Our sample was drawn from unconnected accounts which had recently liked a Trump tweet (4420)
- We randomly assigned 1000 accounts to our treatment group and 3420 to our control group

Implementation

- We created 5 similar Twitter accounts (**@twi_truth**, **@truth_to_twitt**, **@truthToTwitt**, **@SpeakingTw**, **@facts_for_twitt**) - see figure 1
- We regularly created Twitter Apps for each account. Robots used these to automatically tweet the treatment groups
- We sent nearly 7000 tweets over 19 days (see table 1)
- Our server automatically monitored our observation group, recording 1,475,347 tweets and 170,516 likes

Implementation

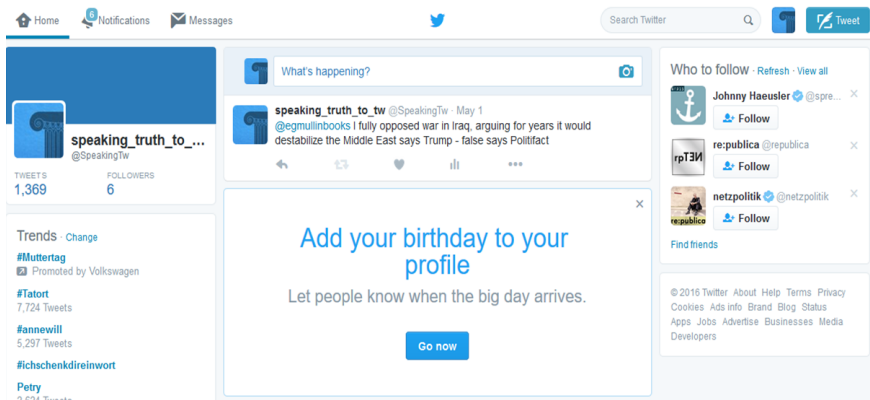


Figure 1 : Example Twitter profile

Implementation

| Tweet number | Text | Truth | Start date |
|--------------|--|-------|------------|
| 1 | @LostinMemphis Trump says most wire transfers to Mexico from undocumented immigrants-half true says award-winning website Politifact | 0 | 2016-04-14 |
| 2 | @LostinMemphis Trump says his deficit to Clinton much smaller than Reagan's against Carter-false says award-winning website Politifact | -2 | 2016-04-20 |
| 3 | @LostinMemphis Trump says Ted Cruz is mathematically out of winning the race - mostly true says politifact | 1 | 2016-04-22 |
| 4 | @LostinMemphis Trump says PA lost 35%, and Harrisburg 40%, of manufacturing jobs since 2001 - Mostly true says politifact | 1 | 2016-04-25 |
| 5 | @LostinMemphis Trump says football coach Rex Ryan won championships in NY twice - false says Politifact. He never did | -2 | 2016-04-27 |
| 6 | @LostinMemphis Trump says ISIS makes millions of dollars a week by selling Libyan oil - false says Politifact | -2 | 2016-04-29 |
| 7 | @LostinMemphis Trump says he fully opposed war in Iraq arguing for years it would destabilize the Middle East - false says Politifact | -2 | 2016-04-30 |

Table 1 : Example tweets

Responses

We received

- 164 retweets
- 369 likes
- 417 responses

Responses

truthAboutTrump @truthToTwitt · Apr 15
@BigJoeyStud Trump says most wire transfers to Mexico from undocumented immigrants- half true says award-winning website Politifact

JoSepth @BigJoeyStud
@truthToTwitt fuck off
8:31 AM - 15 Apr 2016

speaking_truth_to_tw @truth_to_twitt · Apr 14
@roch2779 Trump says most wire transfers to Mexico from undocumented immigrants- half true says award-winning website Politifact

Rochelle @roch2779
@truth_to_twitt what half is true?
10:17 AM - 14 Apr 2016

speaking_truth_to_tw @SpeakingTw · Apr 30
@pbeaman5169 I fully opposed war in Iraq, arguing for years it would destabilize the Middle East says Trump - false says Politifact

Paula McKim @pbeaman5169
@SpeakingTw I don't need you to "preach" to me about Trump! I support #TRUMP2016 end of discussion!
5:11 PM - 30 Apr 2016

Charles D Hedrick @chashedrick
That proves nothing! My Omnifacts says you're full of what makes the grass grow green! They say he has to said it!!

trumpfact @facts_for_twitt
@chashedrick I fully opposed war in Iraq, arguing for years it would destabilize the Middle East says Trump - false says Politifact
6:48 PM - 1 May 2016
Alabaster, AL

Figure 2 : Some interesting comments

Outline

- 1 Implementation
- 2 Descriptive Statistics
- 3 Results
- 4 Limitations
- 5 Conclusion

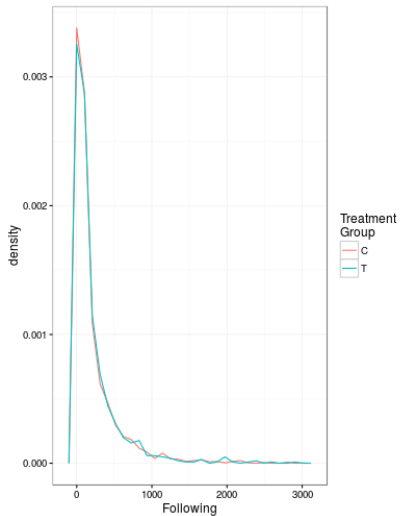
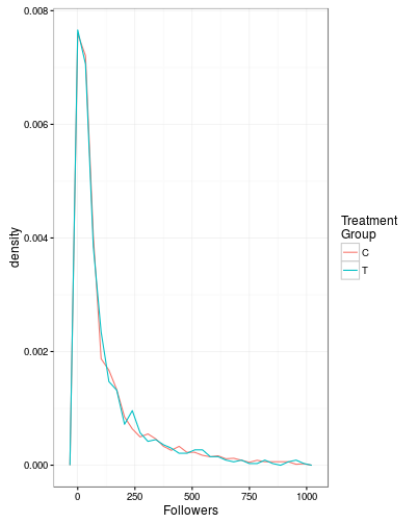
Descriptive Statistics

Descriptive statistics before the treatment period are similar in treatment and control groups

| Variable | Mean | | Median | | SEM | |
|-------------------|---------|-----------|---------|-----------|---------|-----------|
| | Control | Tweetment | Control | Tweetment | Control | Tweetment |
| Avg tweet "Trump" | 2.46 | 2.41 | 1.00 | 0.86 | 0.08 | 0.13 |
| Avg likes | 0.57 | 0.56 | 0.14 | 0.14 | 0.02 | 0.03 |
| Avg #MAGA | 0.26 | 0.19 | 0.00 | 0.00 | 0.01 | 0.02 |
| Avg mentions | 1.43 | 1.41 | 0.57 | 0.57 | 0.05 | 0.09 |
| Avg retweets | 0.59 | 0.59 | 0.00 | 0.00 | 0.03 | 0.06 |
| Followers | 120.81 | 113.38 | 55.00 | 51.0 | 3.06 | 5.50 |
| Following | 210.94 | 201.82 | 93.50 | 91.00 | 5.90 | 11.06 |

Table 2 : Descriptive statistics before the treatment period

Descriptive Statistics



Outline

- 1 Implementation
- 2 Descriptive Statistics
- 3 Results**
- 4 Limitations
- 5 Conclusion

Results: Data and Dependent Variables

From the like and tweet data we collected, we used the following as dependent variables (all per user per day) - tweet data excludes replies to our tweets

- number of likes of tweets by Donald Trump
- number of retweets of tweets by Donald Trump
- number of tweets using the hashtag “#MakeAmericaGreatAgain”
- number of tweets including the key word “Trump”

Results: Difference in Means

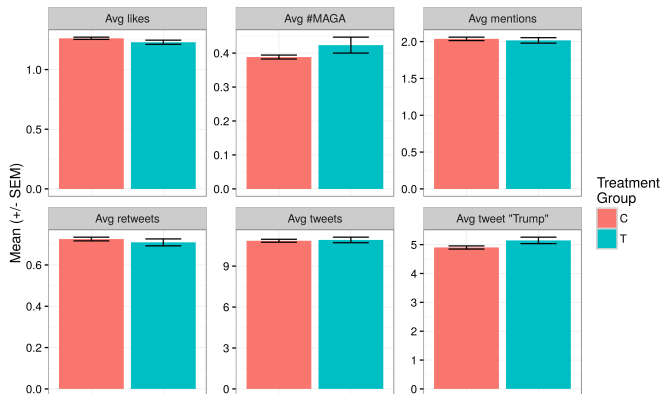


Figure 3 : Per user per day means of each dependent variable in treatment and control groups during the treatment period

Results: Difference in Means

| variable | control mean | treatment mean | p-value |
|-------------------|--------------|----------------|---------|
| Avg likes | 1.26 | 1.23 | 0.09 * |
| Avg tweets | 10.86 | 10.92 | 0.79 |
| Avg retweets | 0.73 | 0.71 | 0.39 |
| Avg mentions | 2.04 | 2.02 | 0.63 |
| Avg #MAGA | 0.39 | 0.42 | 0.15 |
| Avg tweet "Trump" | 4.90 | 5.15 | 0.05 ** |

Table 3 : A t-test on the difference in means between treatment and control groups

Results: Differences over time

Likes

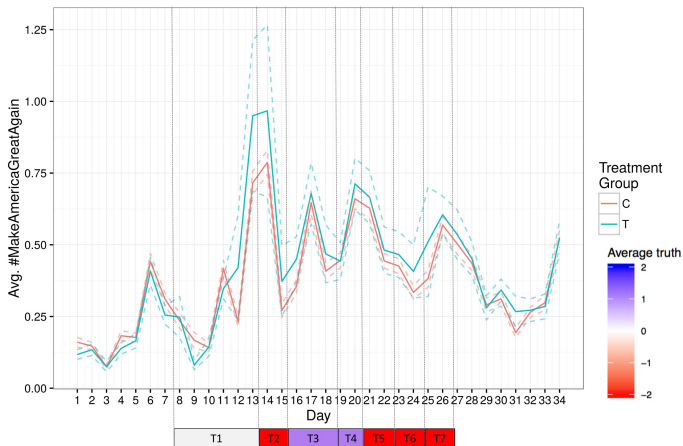


Figure 4 : Likes over time

Results: Differences over time

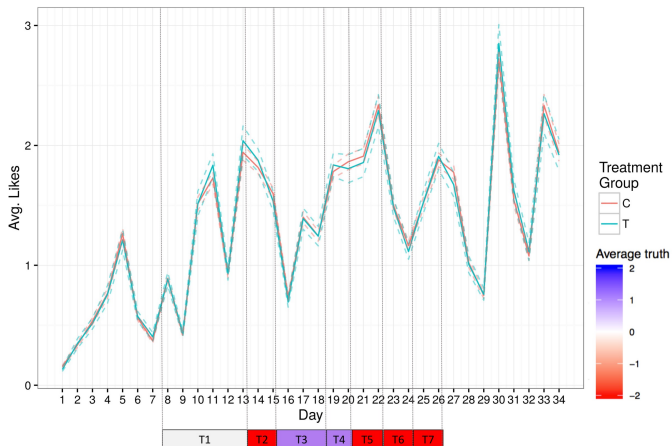


Figure 5 : #MakeAmericaGreatAgain over time

Results: Fixed Effects Model

Table 4 :

| | <i>Dependent variable:</i> | | | | | |
|------------------------------|----------------------------|-------------------|-------------------|-------------------|-------------------|-------------------|
| | y | | | | | |
| | likes | tweets | retweets | mentions | MAGA | keywords |
| | (1) | (2) | (3) | (4) | (5) | (6) |
| temptweet1 | 0.001 (0.081) | 0.098 (0.512) | -0.010 (0.078) | -0.069 (0.127) | 0.055 (0.094) | 0.091 (0.284) |
| temptweet2 | 0.133 (0.130) | -0.048 (0.971) | 0.020 (0.083) | -0.015 (0.172) | 0.144 (0.190) | 0.241 (0.530) |
| temptweet3 | -0.065 (0.076) | 0.797 (1.081) | -0.024 (0.058) | 0.044 (0.181) | 0.034 (0.082) | 0.735 (0.571) |
| temptweet4 | -0.085 (0.104) | 0.126 (1.239) | -0.072 (0.083) | -0.102 (0.254) | -0.012 (0.074) | 0.182 (0.716) |
| temptweet5 | -0.128 (0.106) | -0.315 (1.192) | -0.042 (0.084) | -0.096 (0.204) | 0.028 (0.067) | 0.139 (0.634) |
| temptweet6 | -0.040 (0.105) | -1.940 (1.638) | -0.023 (0.081) | -0.120 (0.252) | 0.009 (0.087) | -0.325 (0.921) |
| temptweet7 | -0.020 (0.081) | -0.705 (1.038) | 0.005 (0.068) | -0.077 (0.164) | 0.064 (0.118) | 0.046 (0.585) |
| F-Test (-tive Tweets) | 2.378 | 1.238 | 0.172 | 0.249 | 3.406 | 0.293 |
| Pr(>F) (-tive Tweets) | 0.05 | 0.292 | 0.953 | 0.91 | 0.009 | 0.883 |
| Observations | 150,246 | 150,246 | 150,246 | 150,246 | 150,246 | 150,246 |
| R ² | 0.0001 | 0.00005 | 0.00002 | 0.00001 | 0.0001 | 0.0001 |
| Adjusted R ² | 0.0001 | 0.00005 | 0.00002 | 0.00001 | 0.0001 | 0.0001 |
| F Statistic (df = 7; 150205) | 1.851* | 1.025 | 0.343 | 0.292 | 2.412** | 1.204 |

Results: Fixed Effects Model

Table 5 :

| | <i>Dependent variable:</i> | | | | | |
|------------------------------|----------------------------|-------------------|-------------------|-------------------|-------------------|-------------------|
| | y | | | | | |
| | likes (1) | tweets (2) | retweets (3) | mentions (4) | MAGA (5) | keywords (6) |
| posdummy | -0.027 (0.057) | 0.192 (0.945) | -0.038 (0.053) | -0.050 (0.157) | -0.005 (0.034) | 0.268 (0.501) |
| negdummy | -0.009 (0.031) | -0.552 (0.409) | 0.021 (0.029) | -0.015 (0.080) | 0.073 (0.086) | -0.103 (0.231) |
| neutdummy | 0.052 (0.055) | 0.331 (0.519) | 0.014 (0.058) | 0.003 (0.109) | 0.016 (0.035) | 0.111 (0.274) |
| F-Test (-tive Tweets) | 2.378 | 1.238 | 0.172 | 0.249 | 3.406 | 0.293 |
| Pr(>F) (-tive Tweets) | 0.05 | 0.292 | 0.953 | 0.91 | 0.009 | 0.883 |
| Observations | 150,246 | 150,246 | 150,246 | 150,246 | 150,246 | 150,246 |
| R ² | 0.00004 | 0.00003 | 0.00002 | 0.00001 | 0.0001 | 0.00002 |
| Adjusted R ² | 0.00004 | 0.00002 | 0.00002 | 0.00001 | 0.0001 | 0.00002 |
| F Statistic (df = 3; 145791) | 1.909 | 1.234 | 1.115 | 0.277 | 3.269** | 1.193 |

Note:

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

Outline

- 1 Implementation
- 2 Descriptive Statistics
- 3 Results
- 4 Limitations**
- 5 Conclusion

Limitations

- Self selection in the sample: *Only active trump followers were selected in our study* and *People had the option to opt-out*
- 10 individuals asked to be withdrawn during treatment. Effect measured thus ITT effect
- Being recognized as a robot
- Bias from manipulation of the twitter feed
- Outcome is **likes** or **tweets** per day while the tweeting has been done at a certain time during the day...
- Collinearity of variables in case of the lasting turn-on model!

Outline

- 1 Implementation
- 2 Descriptive Statistics
- 3 Results
- 4 Limitations
- 5 Conclusion

Interpretation of results

Two different hypotheses:

- 1 rational updaters
- 2 motivated reasoners

Results are unclear:

- Some changes in engagement with Trump are observable
- Different variables react in different directions
- Hard to attribute effects to individual tweets or to truth levels of tweets

Conclusion

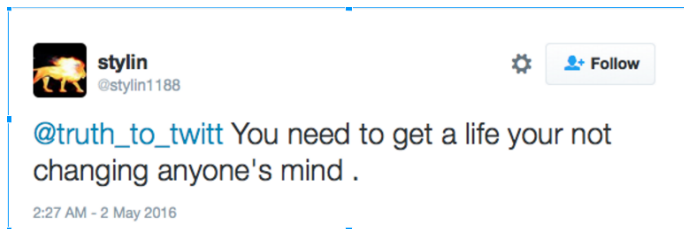


Figure 6 : Your not changing any minds