

Preliminary Results

Group 3

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We calculate two types of time fixed effect models that use differently specified tweet dummies. In table 1, the dummies are switched on on the day the user receives a tweet and stay on. In table 2, they are switched on as before, but switched off when the user receives a subsequent tweet.

We test the joint significance of all negative tweets (tweets 2,5,6 and 7).

We model the effects of the tweet dummies for a list of dependent variables such that

$$y_{it} = \beta_1 T1_{it} + \beta_2 T2_{it} + \beta_3 T3_{it} + \beta_4 T4_{it} + \beta_5 T5_{it} + \beta_6 T6_{it} + \beta_7 T7_{it} + \alpha_t + u_{it}$$

Table 1:

	<i>Dependent variable:</i>					
	likes	tweets	retweets	y mentions	MAGA	keywords
	(1)	(2)	(3)	(4)	(5)	(6)
tweet1	0.051 (0.049)	0.107 (0.540)	0.020 (0.049)	-0.072 (0.116)	0.082** (0.038)	0.148 (0.278)
tweet2	0.118 (0.080)	-0.214 (0.884)	0.040 (0.081)	0.107 (0.191)	0.113* (0.063)	0.288 (0.456)
tweet3	-0.227*** (0.083)	1.196 (0.920)	-0.068 (0.084)	0.051 (0.198)	-0.133** (0.065)	0.580 (0.474)
tweet4	0.019 (0.085)	-0.710 (0.938)	-0.045 (0.086)	-0.128 (0.202)	-0.045 (0.066)	-0.463 (0.483)
tweet5	-0.056 (0.093)	-0.345 (1.029)	0.041 (0.094)	0.038 (0.222)	0.032 (0.073)	-0.067 (0.530)
tweet6	0.131 (0.117)	-2.030 (1.294)	0.015 (0.118)	0.009 (0.279)	-0.006 (0.092)	-0.451 (0.667)
tweet7	-0.046 (0.104)	2.304** (1.151)	-0.003 (0.105)	0.057 (0.248)	0.012 (0.082)	0.471 (0.594)
F-Test (Negative Tweets)	0.988	1.047	0.186	0.184	0.921	0.27
Pr(>F) (Negative Tweets)	0.412	0.381	0.946	0.947	0.451	0.898
Observations	119,968	119,968	119,968	119,968	119,968	119,968
R ²	0.0001	0.0001	0.00001	0.00001	0.0002	0.0002
Adjusted R ²	0.0001	0.0001	0.00001	0.00001	0.0002	0.0002
F Statistic (df = 7; 119929)	1.653	1.103	0.233	0.230	3.702***	3.263***

Note:

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

Table 2:

	<i>Dependent variable:</i>					
	y					
	likes	tweets	retweets	mentions	MAGA	keywords
	(1)	(2)	(3)	(4)	(5)	(6)
temptweet1	0.042 (0.049)	0.257 (0.542)	0.015 (0.049)	-0.020 (0.117)	0.080** (0.038)	0.226 (0.279)
temptweet2	0.169*** (0.065)	-0.100 (0.720)	0.059 (0.066)	0.038 (0.155)	0.195*** (0.051)	0.440 (0.371)
temptweet3	-0.058 (0.054)	1.089* (0.596)	-0.008 (0.054)	0.086 (0.129)	0.062 (0.042)	1.016*** (0.307)
temptweet4	-0.039 (0.067)	0.379 (0.743)	-0.054 (0.068)	-0.041 (0.160)	0.017 (0.053)	0.553 (0.383)
temptweet5	-0.095 (0.067)	0.034 (0.737)	-0.013 (0.067)	-0.004 (0.159)	0.048 (0.052)	0.486 (0.380)
temptweet6	0.036 (0.100)	-1.996* (1.101)	0.002 (0.101)	0.006 (0.237)	0.042 (0.078)	0.035 (0.568)
temptweet7	-0.010 (0.033)	0.308 (0.363)	-0.001 (0.033)	0.063 (0.078)	0.054** (0.026)	0.506*** (0.187)
F-Test (Negative Tweets)	2.259	1.029	0.212	0.176	5.011	2.578
Pr(>F) (Negative Tweets)	0.06	0.39	0.932	0.951	0	0.035
Observations	119,968	119,968	119,968	119,968	119,968	119,968
R ²	0.0001	0.0001	0.00001	0.00001	0.0002	0.0002
Adjusted R ²	0.0001	0.0001	0.00001	0.00001	0.0002	0.0002
F Statistic (df = 7; 119929)	1.603	1.129	0.224	0.179	3.675***	3.317***

Note:

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

For simplicity, we also run t-tests comparing the means of the treatment and control groups during the time when tweets were being sent out

```
## [1] "like_n"
##
## Welch Two Sample t-test
##
## data: treatment_period[treatment_period$t_group == "T", y] and treatment_period[treatment_period$t_group == "C", y]
## t = -0.083947, df = 27203, p-value = 0.9331
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.04863419 0.04463938
## sample estimates:
## mean of x mean of y
## 1.500419 1.502417
##
## [1] "tweet_n"
##
## Welch Two Sample t-test
##
## data: treatment_period[treatment_period$t_group == "T", y] and treatment_period[treatment_period$t_group == "C", y]
## t = 1.0303, df = 26145, p-value = 0.3029
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.2477424 0.7967770
## sample estimates:
## mean of x mean of y
## 11.91413 11.63961
##
## [1] "trump_rt_n"
##
## Welch Two Sample t-test
##
## data: treatment_period[treatment_period$t_group == "T", y] and treatment_period[treatment_period$t_group == "C", y]
## t = 0.48883, df = 26195, p-value = 0.625
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.03418874 0.05690758
## sample estimates:
## mean of x mean of y
## 0.8425216 0.8311622
##
## [1] "trump_mention_n"
##
## Welch Two Sample t-test
##
## data: treatment_period[treatment_period$t_group == "T", y] and treatment_period[treatment_period$t_group == "C", y]
## t = 1.0549, df = 30253, p-value = 0.2915
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.04738074 0.15781026
## sample estimates:
## mean of x mean of y
## 2.290748 2.235533
```

```

##
## [1] "MAGA_n"
##
## Welch Two Sample t-test
##
## data: treatment_period[treatment_period$t_group == "T", y] and treatment_period[treatment_period$t_group == "C", y]
## t = 2.0278, df = 18488, p-value = 0.0426
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 0.002042893 0.120324163
## sample estimates:
## mean of x mean of y
## 0.5007790 0.4395955
##
## [1] "trump_keyword_n"
##
## Welch Two Sample t-test
##
## data: treatment_period[treatment_period$t_group == "T", y] and treatment_period[treatment_period$t_group == "C", y]
## t = 3.3522, df = 24227, p-value = 0.000803
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 0.2065765 0.7882761
## sample estimates:
## mean of x mean of y
## 5.869847 5.372420

```