

March 28, 2021

This paper uses Google search terms containing offensive language as a proxy measure for racial animus. Following the practice of similar works (Stephens-Davidowitz 2014; Chae, Clouston, Hatzenbuehler, et al. 2015; Chae, Clouston, Martz, et al. 2018; Isoya and Yamada 2021), I use coded language to refer to these terms. The words themselves can be found in Table ??

1 Introduction

2 Methods

2.1 Tools Used

Surveys of major social science journals routinely fail to reproduce the findings of a plurality or majority of published articles, even when using data provided with the paper (Nuijten, Hartgerink, et al. 2015; Nuijten and Polanin 2020; Eubank 2016). In many cases, failures to replicate are due to coding errors or mistakes in transcribing the results of a calculation into the published manuscript (Eubank 2016).

To ensure that the methods of this paper have been properly implemented and the finding are reproducible, I tested the analysis routines using the *testthat* package in R (Wickham 2011) and the *unittest* module in Python (Van Rossum and Drake 2009). I use Kintr to integrate statistical calculations into the paper, eliminating the possibility of transcription errors (Xie 2014).

Finally, I provide a Docker image (Merkel 2014) with the reproducibility materials to ensure others can replicate the figures in this paper on their own systems (Boettiger 2015). The net result is “one-click reproducibility” (Nüst et al. 2020); readers can reproduce this exact paper with the push of a button on their own systems from the reproducibility materials.

Figure 1: Sinclair News Anchors Reading a “Must-Run” Script (May 2018)



(a) Images from 30 of 210 Stations Reading the Script (Burke 2018)

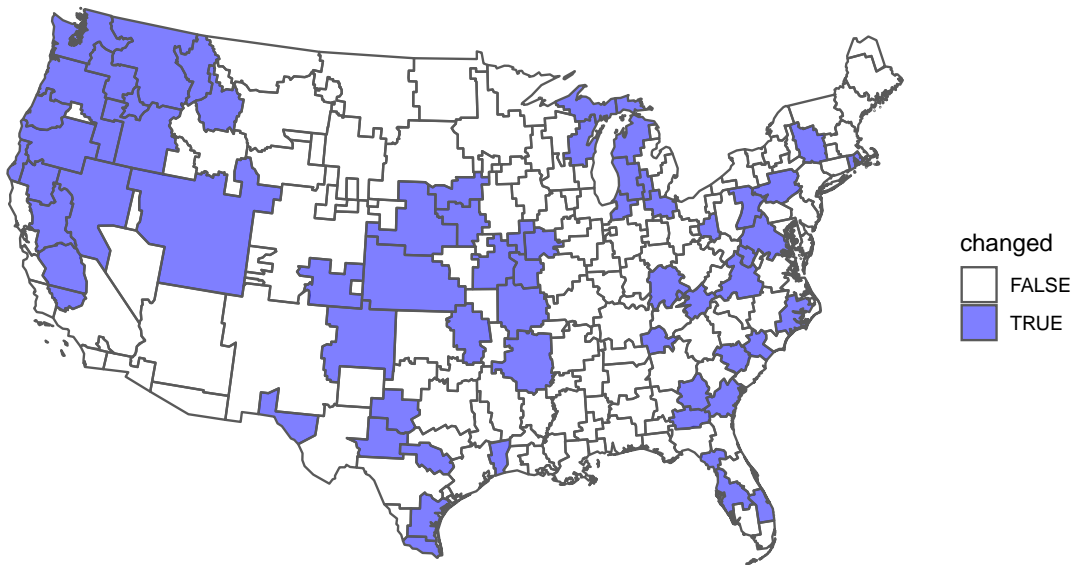
(A): But we’re concerned about the troubling trend of irresponsible, one sided news stories plaguing our country. The sharing of biased and false news has become all too common on social media.

(B): More alarming, some media outlets publish these same fake stories... stories that just aren’t true, without checking facts first.

(A): Unfortunately, some members of the media use their platforms to push their own personal bias and agenda to control ‘exactly what people think’...This is extremely dangerous to a democracy.

(b) Transcript of Segment (Cohen 2018)

Map of Sinclair Stations That Bought or Sold in 2004–2021



References

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A Codings for Offensive Words

Code	Word
Word 1	nigger

B Code In This Document

```
1 ## ----echo=FALSE, warn=FALSE, include=FALSE-----
2 # Set options to knit this document, load libraries
3 knitr::opts_chunk$set(out.width="100%", message=F, cache=T, echo=F, warn=F, include=F)
4 library(tidyverse)
5 library(sf)
6
7
8 ## -----
9 # Load and create data to show markets Sinclair has expanded to / moved out of
10 sinclair_expansions <- read_csv("../data/clean_sinclair_data") %>%
11   nest(-code) %>%
12   mutate(
13     any_true = data %>% map_lgl(~any(.$sinclair_present)),
14     any_false = data %>% map_lgl(~any(!.$sinclair_present)),
15     changed = any_true & any_false
16   )
17
18 dma_boundaries <- st_read("../data/dma_boundaries/dma_boundary.shp")
19 dma_boundaries <- merge(dma_boundaries, sinclair_expansions, by.x="dma0", by.y="code")
20
21
22 ## ----include = T-----
23 # Plot markets Sinclair has expanded to / moved out of
24 ggplot() +
25   geom_sf(data=dma_boundaries, aes(fill=changed), alpha=.5) +
26   scale_fill_manual(values = c(NA, "blue")) +
27   ggtitle("Map of Sinclair Stations That Bought or Sold in 2004-2021") +
28   theme_void()
29
30
31 ## ----echo=FALSE, include=T-----
32 # Pull R code out of this document to put into the appendix
33 code <- knitr::purl("diss.Rnw", quiet=T)
```

C Web Scraping Code

C.1 Master Web Scraping Script

```
1 #!/usr/bin/python
2 import glob
3 import os
4 import pickle as pkl
5 from between_regions import between_reigion_many
6 from in_region import get_region_trend
7 from utils import itr_split_overlap
8 import tqdm
9
10 def extract_keywords(filename):
11     with open(filename, "r") as f:
12         keywords = f.read().splitlines()
13     return keywords
14
15
16 def get_basename(filename):
17     name_w_ext = os.path.basename(filename)
18     basename = os.path.splitext(name_w_ext)[0]
19     return (basename)
20
21
22 def run_keywords(name, keywords):
23     regions = pkl.load(open("dma_abbreviations.pkl", "rb"))
24     filename = f"../data/google_trends_data/{name}_time_serires.csv"
```

```

25 with open(filename, "w") as f:
26     f.write("row,date,score,ispartial,code,term\n")
27     for keyword in keywords:
28         for region in tqdm.tqdm(regions):
29             df = get_region_trend(keyword, region, True)
30
31             if df is not None:
32                 df.to_csv(f, header=False)
33
34 filename = f"../data/google_trends_data/{name}_between_regions.csv"
35 with open(filename, "w") as f:
36     df = between_reigion_many(itr_split_overlap(keywords, 5, 1),
37                               censor=True,
38                               timeframe="all",
39                               geo="US",
40                               gprop="")
41     df.to_csv(f, header=True)
42
43
44 if __name__ == "__main__":
45     keyword_files = sorted(glob.glob("keywords/*.csv"))
46     base_names = [get_basename(filename) for filename in keyword_files]
47     keywords = [extract_keywords(filename) for filename in keyword_files]
48     first = (list(zip(base_names, keywords)))[0]
49     run_keywords(first[0], first[1])

```

C.2 Collecting Data for Between-Region Comparisons

```

1  #!/usr/bin/env python
2  from pytrends.request import TrendReq
3  from utils import censor_string
4
5
6  def between_region(query, censor, **kwargs):
7      pytrends = TrendReq(hl='en-US', tz=360)
8      pytrends.build_payload(query, cat=0, **kwargs)
9      df = pytrends.interest_by_region(resolution='DMA',
10                                     inc_low_vol=True,
11                                     inc_geo_code=True)
12
13     df = df.set_index("geoCode")
14     df.index.name = 'code'
15     if censor:
16         df = df.rename(censor_string, axis="columns")
17     return (df)
18
19 def between_reigion_many(iterable, censor, **kwargs):
20     iterable = iter(iterable)
21     chunk1 = next(iterable)
22     df1 = between_region(chunk1, censor, **kwargs)
23
24     for chunk in iterable:
25         chunk2 = chunk
26         shared = list(set(chunk1) & set(chunk2))[0]
27         if censor:
28             shared = censor_string(shared)
29         df2 = between_region(chunk2, censor, **kwargs)
30
31         mean1 = df1[shared].mean()
32         mean2 = df2[shared].mean()
33         normaliation_factor = mean1 / mean2
34
35         df2 = df2 * normaliation_factor
36         df2 = df2.drop(columns=[shared])
37
38         df1 = df1.join(df2)
39
40         chunk1 = chunk2
41
42     return (df1)

```

C.3 Collecting Within-Region Data

```

1  #!/usr/bin/python
2  from pytrends.request import TrendReq

```

```

3 from utils import censor_string
4 import time
5
6 def get_region_trend(query, region, censor):
7     pytrends = TrendReq(hl='en-US', tz=360)
8     try:
9         pytrends.build_payload(kw_list=[query],
10                                geo=f"{region}",
11                                timeframe="all")
12         df = pytrends.interest_over_time()
13         if not df.empty:
14             df.columns = ["n", "ispartial"]
15             df.index.name = 'date'
16             df.reset_index(inplace=True)
17             df["region"] = region
18
19             df["query"] = query
20
21             if censor:
22                 df["query"] = df["query"].apply(censor_string)
23
24             return df
25     except:
26         if censor:
27             print(f"Rate error: {censor_string(query)} in {region}")
28         else:
29             print(f"Rate error: {query} in {region}")
30
31     time.sleep(60)
32     get_region_trend(query, region, censor)

```

C.4 Utility Functions

```

1 import itertools as it
2 import requests
3 import random
4
5 # Returns list of N random words from MIT dictionary
6 def random_words(n):
7     word_site = "https://www.mit.edu/~ecprice/wordlist.10000"
8
9     response = requests.get(word_site)
10    words = response.text.splitlines()
11
12    keywords = random.sample(words, n)
13
14    return (keywords)
15
16
17 # Tests if computer is connected to internet (used in tests)
18 def connected():
19     url = "http://google.com"
20     timeout = 5
21     try:
22         requests.get(url, timeout=timeout)
23         return (True)
24     except (requests.ConnectionError, requests.Timeout):
25         return (False)
26
27
28 # Censors strings so that sensitive words aren't uploaded to github / used in scripts
29 def censor_string(string):
30     return (string[0] + "_" * (len(string) - 2) + string[-1])
31
32
33 # credit to Ilja Everila for this implementation
34 # https://stackoverflow.com/questions/48381870/a-better-way-to-split-a-sequence-in-chunks-
35 # with-overlaps
36 def itr_split_overlap(iterable, size, overlap):
37     if overlap >= size:
38         raise ValueError("overlap must be smaller than size")
39
40     itr = iter(iterable)
41
42     next_ = tuple(it.islice(itr, size))

```

```

43
44     yield next_
45
46     prev = next_[-overlap:] if overlap else ()
47
48     while True:
49         chunk = tuple(it.islice(itr, size - overlap))
50
51         if not chunk:
52             break
53
54         next_ = (*prev, *chunk)
55         yield next_
56
57         if overlap:
58             prev = next_[-overlap:]

```

D Analysis Code

```

1  #!/usr/bin/Rscript
2  # Load Libraries for Analysis
3  library(tidyverse)
4  library(lubridate)
5  # Load utility functions
6  source("utils.R")
7
8  between_data <- read_csv("../data/google_trends_data/set_1_between_regions.csv") %>%
9    pivot_longer(~code, names_to = "term", values_to = "overall") %>%
10    mutate(term = censor_string(term))
11
12  search_data <- read_csv("../data/google_trends_data/set_1_time_serires.csv") %>%
13    mutate(year = year(date)) %>%
14    mutate(term = censor_string(term)) %>%
15    group_by(year, code, term) %>%
16    nest() %>%
17    mutate(
18      score = data %>% map_dbl(~ mean(.x$score, na.rm = T)),
19      code = str_extract(code, "[0-9]+") %>% as.numeric()
20    )
21
22  search_data <- search_data %>%
23    dplyr::select(-data) %>%
24    distinct(term, code, year, .keep_all = TRUE) %>%
25    pivot_wider(names_from = term, values_from = score) %>%
26    mutate(overall_score = rowSums(across(everything()), na.rm = T))
27
28
29  full_data <- search_data %>%
30    right_join(sinclair_data) %>%
31    filter(year != 2021)
32
33  lm(overall_score ~ as.factor(year) + as.factor(code) + sinclair_present, data = full_data)
34    summary()
35
36  lm(overall_score ~ as.factor(year) + as.factor(code) + sinclair_present + year:as.factor(code)
37    , data = full_data) %>%
38    summary()

```

D.1 Utility Functions

```

1  censor_string <- function(string) {
2    substr(string, 2, nchar(string) - 1) <- paste0(rep("_", nchar(string) - 2), collapse = "")
3    names(string) <- NULL
4    return(string)
5  }
6  censor_string <- Vectorize(censor_string, USE.NAMES=F)
7
8  years_before <- function(bool) {
9    if (any(bool)) {
10      out <- numeric(length = length(bool))
11      start <- min(which(bool))
12
13      before <- (seq(start, 1) - 1) * -1

```

```

14   if (start != length(bool)) {
15     after <- seq(1, length(bool) - start)
16   } else {
17     after <- c()
18   }
19
20   out <- c(before, after)
21 } else {
22   out <- rep(NA_real_, length(bool))
23 }
24
25
26 return(out)
27 }

```

E Unit tests

E.1 For Python Code

```

1 import unittest
2 import pandas as pd
3 from between_regions import between_region
4 from utils import connected
5
6
7 @unittest.skipIf(not connected(), "not connected to the internet")
8 class TestBetweenRegion(unittest.TestCase):
9     def test_between_region(self):
10
11         result_1 = between_region(["socks"],
12                                   timeframe="2016-12-14 2017-01-25",
13                                   geo="US",
14                                   gprop="")
15         result_2 = between_region(["socks", "shoes", "fish"],
16                                   timeframe="2016-12-14 2017-01-25",
17                                   geo="US",
18                                   gprop="")
19
20         expected_1 = pd.read_parquet("tests/test_data/between_region_1.parquet")
21         expected_2 = pd.read_parquet("tests/test_data/between_region_2.parquet")
22
23         self.assertTrue(expected_1.equals(result_1))
24         self.assertTrue(expected_2.equals(result_2))

```

```

1 import unittest
2 from utils import itr_split_overlap, censor_string
3
4 class TestItrSplitOverlap(unittest.TestCase):
5     def test_itr(self):
6         test_list = ["one", "two", "three", "four", "five"]
7
8         expected_result = [('one', 'two'), ('two', 'three'), ('three', 'four'),
9                             ('four', 'five')]
10        self.assertEqual(expected_result,
11                          list(itr_split_overlap(test_list, 2, 1)))
12
13        expected_result = [('one', 'two', 'three', 'four'),
14                            ('two', 'three', 'four', 'five')]
15        self.assertEqual(expected_result,
16                          list(itr_split_overlap(test_list, 4, 3)))
17
18        def test_exceptions(self):
19            with self.assertRaises(ValueError):
20                test_list = ["one", "two", "three", "four", "five"]
21
22                list(itr_split_overlap(test_list, 2, 3))
23
24
25 class CensorString(unittest.TestCase):
26     def test_censor(self):
27         self.assertEqual("h_y", censor_string("hey"))
28         self.assertEqual("f____r", censor_string("fender"))
29         self.assertEqual("fr", censor_string("fr"))

```


E.2 For R Code

```
1 library(testthat)
2 source("../utils.R")
3
4 test_that("censor string works", {
5   expect_equal(censor_string("hey"), "h_y")
6   expect_equal(censor_string("hy"), "hy")
7   expect_equal(censor_string("watermelon"), "w_____n")
8   expect_equal(censor_string("fantastic"), "f_____c")
9 })
10
11 test_that("censor string vectorized correctly", {
12   expect_equal(censor_string(c("hello", "there")), c("h___o", "t___e"))
13   expect_equal(censor_string(c("watermelon", "paper")), c("w_____n", "p___r"))
14 })
15
16 test_that("years before works", {
17   expect_equal(years_before(c(F, F, F, T, T)), c(-3, -2, -1, 0, 1))
18   expect_equal(years_before(c(F, F, F, F, T)), c(-4, -3, -2, -1, 0))
19   expect_equal(years_before(c(T, T, T, T, T)), c(0, 1, 2, 3, 4))
20 })
```