

```
In [1]: # Initialize OK
from client.api.notebook import Notebook
ok = Notebook('hw3.ok')
```

```
=====
Assignment: hw3
OK, version v1.13.11
=====
```

Homework 3: Trump, Twitter, and Text

Due Date: Monday 2/25, 11:59 pm PST

Welcome to the third homework assignment of Data 100/200! In this assignment, we will work with Twitter data in order to analyze Donald Trump's tweets.

Collaboration Policy

Data science is a collaborative activity. While you may talk with others about the homework, we ask that you **write your solutions individually**. If you do discuss the assignments with others please **include their names** below.

Collaborators: *list collaborators here*

```
In [2]: # Run this cell to set up your notebook
import csv
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import json
import zipfile

# Ensure that Pandas shows at least 280 characters in columns, so we can see
pd.set_option('max_colwidth', 280)

%matplotlib inline
plt.style.use('fivethirtyeight')
import seaborn as sns
sns.set()
sns.set_context("talk")
import re
```

Score Breakdown

| Question | Points |
|------------|--------|
| Question 1 | 2 |
| Question 2 | 1 |

| Question | Points |
|-------------|--------|
| Question 3 | 2 |
| Question 4a | 1 |
| Question 4b | 2 |
| Question 4c | 2 |
| Question 5a | 1 |
| Question 5b | 1 |
| Question 5c | 1 |
| Question 5d | 2 |
| Question 5e | 2 |
| Question 6a | 1 |
| Question 6b | 1 |
| Total | 19 |

Part 1: Importing the Data

We will again use the `fetch_and_cache` utility to download the dataset.

```
In [3]: # Download the dataset
from ds100_utils import fetch_and_cache
data_url = 'http://www.ds100.org/sp19/assets/datasets/hw3-realdonaldtrump_t
file_name = 'hw3-realdonaldtrump_tweets.json.zip'

dest_path = fetch_and_cache(data_url=data_url, file=file_name)
print(f'Located at {dest_path}')
```

[#####]


Downloaded hw3-realdonaldtrump_tweets.json.zip!
MD5 hash of file: f26e90f51b3d7b304d8db1ad5bee2f57
Located at data/hw3-realdonaldtrump_tweets.json.zip

Now that we've downloaded the tweets, let's unzip them and load them into our notebook. Run the cell below to unzip and read tweets from the json file into a list named `all_tweets`.

```
In [5]: # Unzip the dataset
my_zip = zipfile.ZipFile(dest_path, 'r')
with my_zip.open('hw3-realdonaldtrump_tweets.json', 'r') as f:
    all_tweets = json.load(f)
```

Here is what a typical tweet from `all_tweets` looks like:

```
In [6]: from pprint import pprint # to get a more easily-readable view.
        pprint(all_tweets[-1])
```

```
{'contributors': None,
 'coordinates': None,
 'created_at': 'Tue Oct 16 18:40:18 +0000 2018',
 'display_text_range': [0, 174],
 'entities': {'hashtags': [], 'symbols': [], 'urls': [], 'user_mentions':
 []},
 'favorite_count': 52115,
 'favorited': False,
 'full_text': 'Just spoke with the Crown Prince of Saudi Arabia who total
ly '
                'denied any knowledge of what took place in their Turkish '
                'Consulate. He was with Secretary of State Mike Pompeo...',
 'geo': None,
 'id': 1052268011900555265,
 'id_str': '1052268011900555265',
 'in_reply_to_screen_name': None,
 'in_reply_to_status_id': None,
 'in_reply_to_status_id_str': None,
 'in_reply_to_user_id': None,
 'in_reply_to_user_id_str': None,
 'is_quote_status': False,
 'lang': 'en',
 'place': None,
 'retweet_count': 13493,
 'retweeted': False,
 'source': '<a href="http://twitter.com/download/iphone" '
            'rel="nofollow">Twitter for iPhone</a>',
 'truncated': False,
 'user': {'contributors_enabled': False,
          'created_at': 'Wed Mar 18 13:46:38 +0000 2009',
          'default_profile': False,
          'default_profile_image': False,
          'description': '45th President of the United States of America
 ',
          'entities': {'description': {'urls': []},
                        'url': {'urls': [{'display_url': 'Instagram.com/re
alDonaldTrump',
                                         'expanded_url': 'http://www.Inst
agram.com/realDonaldTrump',
                                         'indices': [0, 23],
                                         'url': 'https://t.co/OMxB0x7xC
5'}}]}},
          'favourites_count': 7,
          'follow_request_sent': False,
          'followers_count': 58311576,
          'following': True,
          'friends_count': 45,
          'geo_enabled': True,
          'has_extended_profile': False,
          'id': 25073877,
          'id_str': '25073877',
          'is_translation_enabled': True,
          'is_translator': False,
```

```

        'lang': 'en',
        'listed_count': 100264,
        'location': 'Washington, DC',
        'name': 'Donald J. Trump',
        'notifications': False,
        'profile_background_color': '6D5C18',
        'profile_background_image_url': 'http://abs.twimg.com/images/themes/themes1/bg.png',
        'profile_background_image_url_https': 'https://abs.twimg.com/images/themes/themes1/bg.png',
        'profile_background_tile': True,
        'profile_banner_url': 'https://pbs.twimg.com/profile_banners/25073877/1550087458',
        'profile_image_url': 'http://pbs.twimg.com/profile_images/874276197357596672/kUuht00m_normal.jpg',
        'profile_image_url_https': 'https://pbs.twimg.com/profile_images/874276197357596672/kUuht00m_normal.jpg',
        'profile_link_color': '1B95E0',
        'profile_sidebar_border_color': 'BDDCAD',
        'profile_sidebar_fill_color': 'C5CEC0',
        'profile_text_color': '333333',
        'profile_use_background_image': True,
        'protected': False,
        'screen_name': 'realDonaldTrump',
        'statuses_count': 40563,
        'time_zone': None,
        'translator_type': 'regular',
        'url': 'https://t.co/OMxB0x7xC5',
        'utc_offset': None,
        'verified': True}}

```

Question 1

Construct a DataFrame called `trump` containing data from all the tweets stored in `all_tweets`. The index of the DataFrame should be the ID of each tweet (looks something like 907698529606541312). It should have these columns:

- `time`: The time the tweet was created encoded as a datetime object. (Use `pd.to_datetime` to encode the timestamp.)
- `source`: The source device of the tweet.
- `text`: The text of the tweet.
- `retweet_count`: The retweet count of the tweet.

Finally, **the resulting DataFrame should be sorted by the index.**

Warning: Some tweets will store the text in the `text` field and other will use the `full_text` field.

```
In [20]: def inTweet(tweet):
    if "text" in tweet:
        return tweet['text']
    else:
        return tweet['full_text']

trump = pd.DataFrame({'time': pd.to_datetime([tw['created_at'] for twt in
    'source': [tw['source'] for twt in all_tweets],
    'text': [inTweet(twt) for twt in all_tweets],
    'retweet_count': [tw["retweet_count"] for twt in all
    },
    index=[tw['id'] for twt in all_tweets],
    columns=['time', 'source', 'text', 'retweet_count'],
).sort_index()

trump.head()
```

Out[20]:

| | time | source | text | retw |
|--------------------|---------------------|--|--|------|
| 690171032150237184 | 2016-01-21 13:56:11 | Twitter for Android | "@bigop1: @realDonaldTrump @SarahPalinUSA https://t.co/3kYQGqeVyD" | |
| 690171403388104704 | 2016-01-21 13:57:39 | Twitter for Android | "@AmericanAsPie: @glennbeck @SarahPalinUSA Remember when Glenn gave out gifts to ILLEGAL ALIENS at crossing the border? Me too!" | |
| 690173226341691392 | 2016-01-21 14:04:54 | Twitter for Android | So sad that @CNN and many others refused to show the massive crowd at the arena yesterday in Oklahoma. Dishonest reporting! | |
| 690176882055114758 | 2016-01-21 14:19:26 | Twitter for Android | Sad sack @JebBush has just done another ad on me, with special interest money, saying I won't beat Hillary - I WILL. But he can't beat me. | |
| 690180284189310976 | 2016-01-21 14:32:57 | Twitter for Android | Low energy candidate @JebBush has wasted \$80 million on his failed presidential campaign. Millions spent on me. He should go home and relax! | |

```
In [21]: ok.grade("q1");
```

```
~~~~~
Running tests
```

```
-----
Test summary
  Passed: 9
  Failed: 0
[ooooooooook] 100.0% passed
```

Part 2: Tweet Source Analysis

In the following questions, we are going to find out the characteristics of Trump tweets and the devices used for the tweets.

First let's examine the source field:

```
In [22]: trump['source'].unique()
```

```
Out[22]: array(['<a href="http://twitter.com/download/android" rel="nofollow">Twitter for Android</a>',
               '<a href="http://twitter.com/download/iphone" rel="nofollow">Twitter for iPhone</a>',
               '<a href="http://twitter.com" rel="nofollow">Twitter Web Client</a>',
               '<a href="https://mobile.twitter.com" rel="nofollow">Mobile Web (M5)</a>',
               '<a href="http://instagram.com" rel="nofollow">Instagram</a>',
               '<a href="http://twitter.com/#!/download/ipad" rel="nofollow">Twitter for iPad</a>',
               '<a href="https://studio.twitter.com" rel="nofollow">Media Studio</a>',
               '<a href="https://periscope.tv" rel="nofollow">Periscope</a>',
               '<a href="https://ads.twitter.com" rel="nofollow">Twitter Ads</a>',
               '<a href="https://studio.twitter.com" rel="nofollow">Twitter Media Studio</a>'],
              dtype=object)
```

Question 2

Notice how sources like "Twitter for Android" or "Instagram" are surrounded by HTML tags. In the cell below, clean up the `source` field by removing the HTML tags from each `source` entry.

Hints:

- Use `trump['source'].str.replace` along with a regular expression.

- You may find it helpful to experiment with regular expressions at regex101.com (<https://regex101.com/>).

```
In [23]: ## Uncomment and complete  
trump['source'] = trump['source'].str.replace(r"<[^>]*>", "")
```

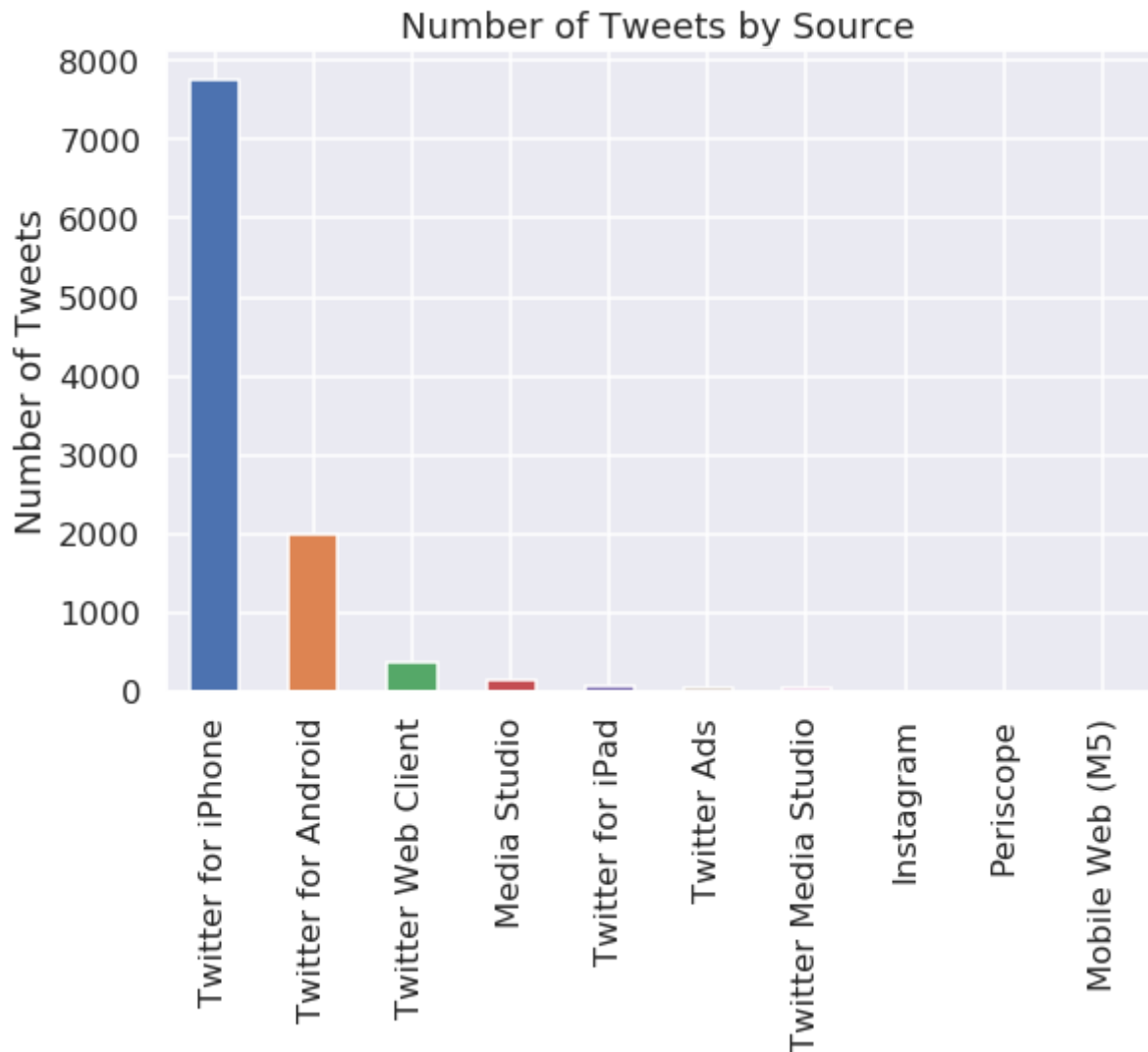
```
In [24]: ok.grade("q2");
```

~~~~~  
Running tests

-----  
Test summary  
    Passed: 1  
    Failed: 0  
[ooooooooook] 100.0% passed

In the following plot, we see that there are two device types that are more commonly used than others.

```
In [25]: plt.figure(figsize=(8, 6))
trump['source'].value_counts().plot(kind="bar")
plt.ylabel("Number of Tweets")
plt.title("Number of Tweets by Source");
```



### Question 3

Now that we have cleaned up the `source` field, let's now look at which device Trump has used over the entire time period of this dataset.

To examine the distribution of dates we will convert the date to a fractional year that can be plotted as a distribution.

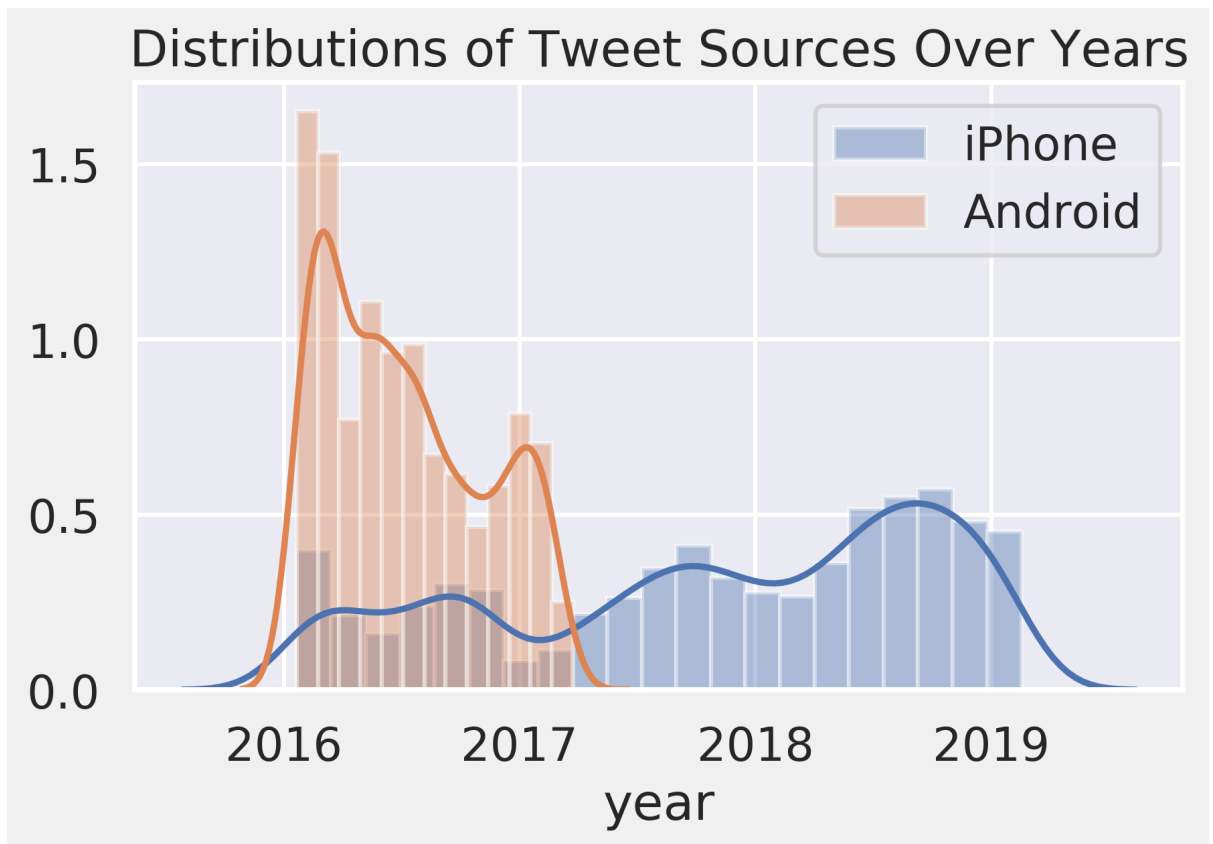
(Code borrowed from <https://stackoverflow.com/questions/6451655/python-how-to-convert-datetime-dates-to-decimal-years> (<https://stackoverflow.com/questions/6451655/python-how-to-convert-datetime-dates-to-decimal-years>))



```
In [26]: import datetime
def year_fraction(date):
    start = datetime.date(date.year, 1, 1).toordinal()
    year_length = datetime.date(date.year+1, 1, 1).toordinal() - start
    return date.year + float(date.toordinal() - start) / year_length

trump['year'] = trump['time'].apply(year_fraction)
```

Now, use `sns.distplot` to overlay the distributions of Trump's 2 most frequently used web technologies over the years. Your final plot should look like:

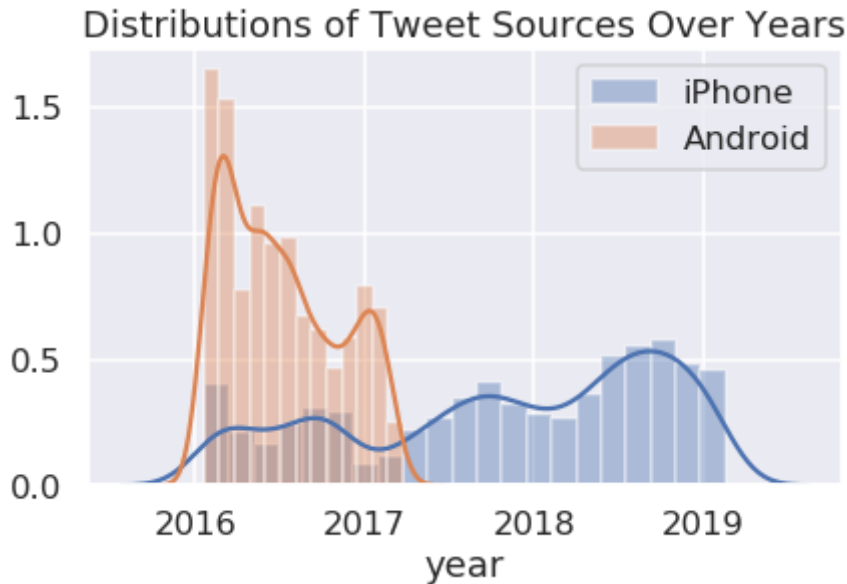


```
In [60]: popular_dev = trump['source'].value_counts().head(2).index

for item in popular_dev:
    filterItem = trump['source'] == item
    sns.distplot(trump[filterItem]['year'], label=item[12:])

plt.title('Distributions of Tweet Sources Over Years')
plt.legend()
```

Out[60]: <matplotlib.legend.Legend at 0x7f56045ec438>



## Question 4

Is there a difference between Trump's tweet behavior across these devices? We will attempt to answer this question in our subsequent analysis.

First, we'll take a look at whether Trump's tweets from an Android device come at different times than his tweets from an iPhone. Note that Twitter gives us his tweets in the [UTC timezone](https://www.wikiwand.com/en/List_of_UTC_time_offsets) ([https://www.wikiwand.com/en/List\\_of\\_UTC\\_time\\_offsets](https://www.wikiwand.com/en/List_of_UTC_time_offsets)) (notice the +0000 in the first few tweets).

```
In [42]: for tweet in all_tweets[:3]:
          print(tweet['created_at'])
```

```
Wed Oct 12 14:00:48 +0000 2016
Wed Oct 12 13:46:43 +0000 2016
Wed Oct 12 12:59:05 +0000 2016
```

We'll convert the tweet times to US Eastern Time, the timezone of New York and Washington D.C., since those are the places we would expect the most tweet activity from Trump.

```
In [43]: trump['est_time'] = (
    trump['time'].dt.tz_localize("UTC") # Set initial timezone to UTC
    .dt.tz_convert("EST") # Convert to Eastern Time
)
trump.head()
```

Out[43]:

|                    | time                | source              | text                                                                                                                                          | retweet_count | year        | est_time            |
|--------------------|---------------------|---------------------|-----------------------------------------------------------------------------------------------------------------------------------------------|---------------|-------------|---------------------|
| 690171032150237184 | 2016-01-21 13:56:11 | Twitter for Android | "@bigop1: @realDonaldTrump @SarahPalinUSA https://t.co/3kYQGqeVyD"                                                                            | 1059          | 2016.054645 | 2016-01-21 08:56:05 |
| 690171403388104704 | 2016-01-21 13:57:39 | Twitter for Android | "@AmericanAsPie: @glennbeck @SarahPalinUSA Remember when Glenn gave out gifts to ILLEGAL ALIENS at crossing the border? Me too!"              | 1339          | 2016.054645 | 2016-01-21 08:57:05 |
| 690173226341691392 | 2016-01-21 14:04:54 | Twitter for Android | So sad that @CNN and many others refused to show the massive crowd at the arena yesterday in Oklahoma. Dishonest reporting!                   | 2006          | 2016.054645 | 2016-01-21 09:04:05 |
| 690176882055114758 | 2016-01-21 14:19:26 | Twitter for Android | Sad sack @JebBush has just done another ad on me, with special interest money, saying I won't beat Hillary - I WILL. But he can't beat me.    | 2266          | 2016.054645 | 2016-01-21 09:19:05 |
| 690180284189310976 | 2016-01-21 14:32:57 | Twitter for Android | Low energy candidate @JebBush has wasted \$80 million on his failed presidential campaign. Millions spent on me. He should go home and relax! | 2886          | 2016.054645 | 2016-01-21 09:32:05 |

## Question 4a

Add a column called `hour` to the `trump` table which contains the hour of the day as floating point number computed by:

$$\text{hour} + \frac{\text{minute}}{60} + \frac{\text{second}}{60^2}$$

- **Hint:** See the cell above for an example of working with [dt accessors](https://pandas.pydata.org/pandas-docs/stable/getting_started/basics.html#basics-dt-accessors) ([https://pandas.pydata.org/pandas-docs/stable/getting\\_started/basics.html#basics-dt-accessors](https://pandas.pydata.org/pandas-docs/stable/getting_started/basics.html#basics-dt-accessors)).

```
In [49]: trump['hour'] = trump['est_time'].dt.hour + trump['est_time'].dt.minute/60
```

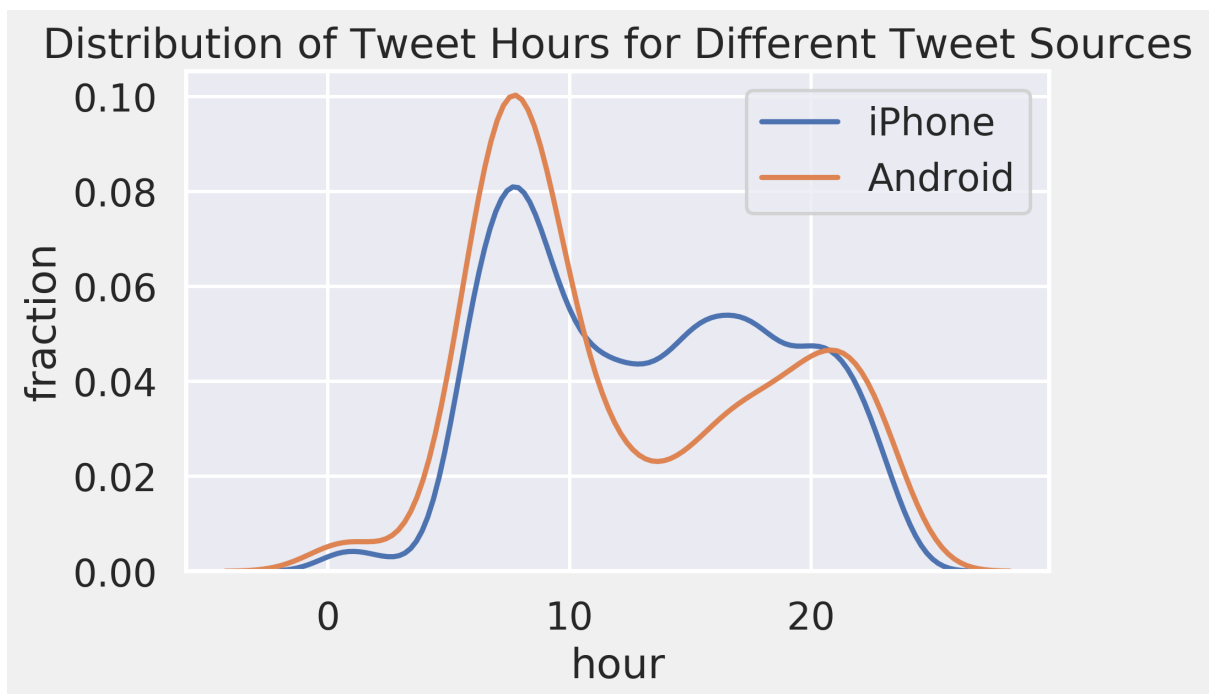
```
In [50]: ok.grade("q4a");
```

~~~~~  
Running tests

Test summary
Passed: 1
Failed: 0
[ooooooooook] 100.0% passed

Question 4b

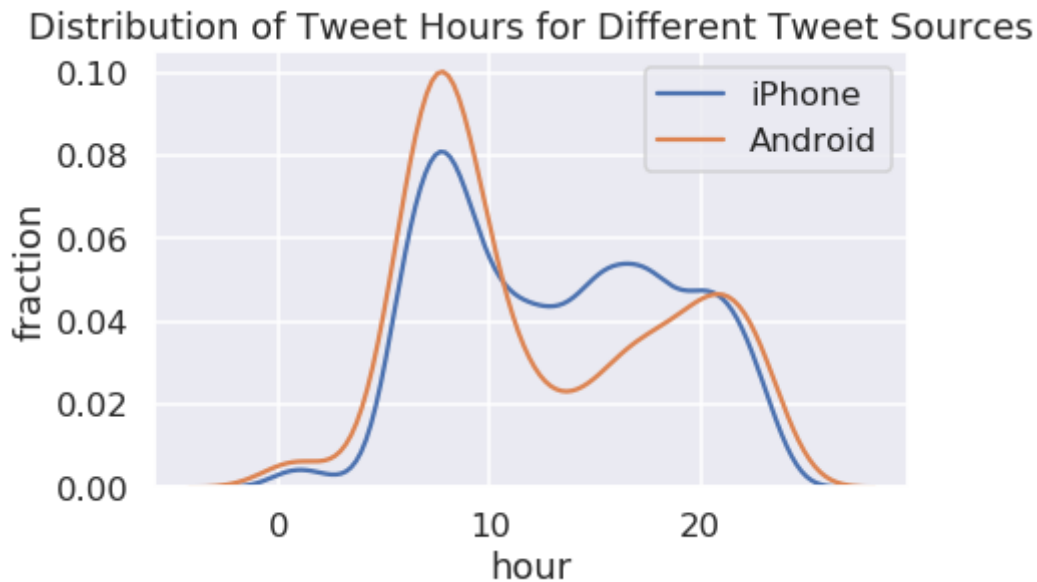
Use this data along with the seaborn `distplot` function to examine the distribution over hours of the day in eastern time that trump tweets on each device for the 2 most commonly used devices. Your plot should look similar to the following:



```
In [62]: ### make your plot here
popular_dev = trump['source'].value_counts().head(2).index
for device in popular_dev:
    filterItem = trump['source'] == device
    sns.distplot(trump[filterItem]['hour'], label = device[12:], hist=False)

plt.title('Distribution of Tweet Hours for Different Tweet Sources')
plt.xlabel('hour')
plt.ylabel('fraction')
plt.legend()
```

Out[62]: <matplotlib.legend.Legend at 0x7f5604e060b8>

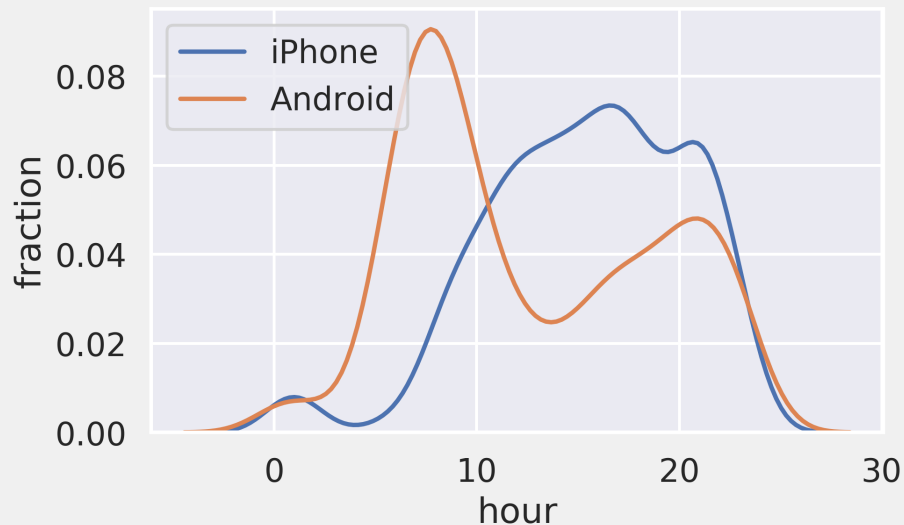


Question 4c

According to [this Verge article \(https://www.theverge.com/2017/3/29/15103504/donald-trump-iphone-using-switched-android\)](https://www.theverge.com/2017/3/29/15103504/donald-trump-iphone-using-switched-android), Donald Trump switched from an Android to an iPhone sometime in March 2017.

Let's see if this information significantly changes our plot. Create a figure similar to your figure from question 4b, but this time, only use tweets that were tweeted before 2017. Your plot should look similar to the following:

Distribution of Tweet Hours for Different Tweet Sources (pre-2017)

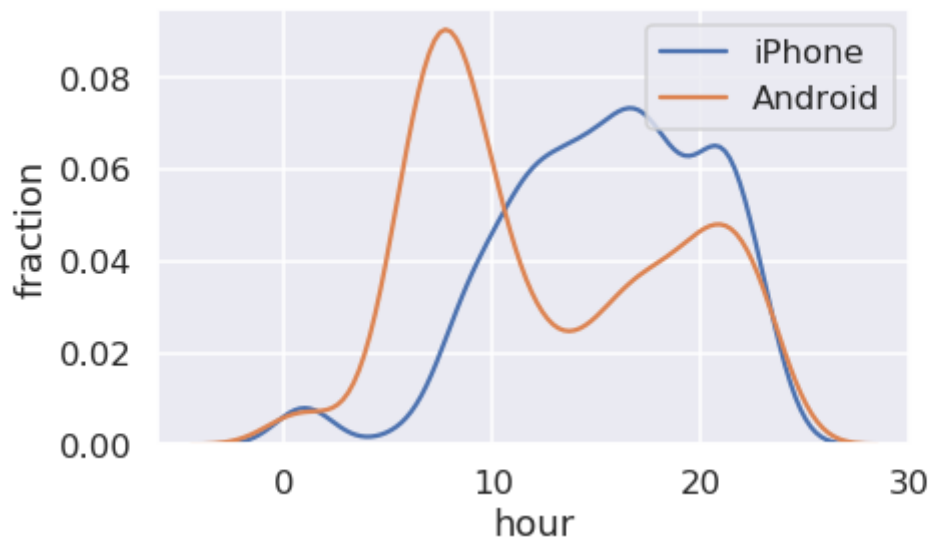


```
In [76]: ### make your plot here
for device in popular_dev:
    filterItem = (trump['source'] == device) & (trump['year'] < 2017)
    sns.distplot(trump[filterItem]['hour'], label = device[12:], hist=False)

plt.title('Distribution of Tweet Hours for Different Tweet Sources (pre-2017)')
plt.xlabel('hour')
plt.ylabel('fraction')
plt.legend()
```

Out[76]: <matplotlib.legend.Legend at 0x7f5604ebdf60>

Distribution of Tweet Hours for Different Tweet Sources (pre-2017)



Question 4d

During the campaign, it was theorized that Donald Trump's tweets from Android devices were written by him personally, and the tweets from iPhones were from his staff. Does your figure give support to this theory? What kinds of additional analysis could help support or reject this claim?

The figure does support the theory that during the campaign, Trump used an Android phone to tweet messages while most of his staff used iPhones, based on the increased activity at around hour 8.

Part 3: Sentiment Analysis

It turns out that we can use the words in Trump's tweets to calculate a measure of the sentiment of the tweet. For example, the sentence "I love America!" has positive sentiment, whereas the sentence "I hate taxes!" has a negative sentiment. In addition, some words have stronger positive / negative sentiment than others: "I love America." is more positive than "I like America."

We will use the [VADER \(Valence Aware Dictionary and sEntiment Reasoner\)](https://github.com/cjhutto/vaderSentiment) (<https://github.com/cjhutto/vaderSentiment>) lexicon to analyze the sentiment of Trump's tweets. VADER is a lexicon and rule-based sentiment analysis tool that is specifically attuned to sentiments expressed in social media which is great for our usage.

The VADER lexicon gives the sentiment of individual words. Run the following cell to show the first few rows of the lexicon:

```
In [77]: print(''.join(open("vader_lexicon.txt").readlines()[:10]))
```

| | | | |
|-------------|------|---------|--|
| \$: | -1.5 | 0.80623 | [-1, -1, -1, -1, -3, -1, -3, -1, -2, -1] |
| %) | -0.4 | 1.0198 | [-1, 0, -1, 0, 0, -2, -1, 2, -1, 0] |
| %-) | -1.5 | 1.43178 | [-2, 0, -2, -2, -1, 2, -2, -3, -2, -3] |
| &-: | -0.4 | 1.42829 | [-3, -1, 0, 0, -1, -1, -1, 2, -1, 2] |
| &: | -0.7 | 0.64031 | [0, -1, -1, -1, 1, -1, -1, -1, -1, -1] |
| (' } { ') | 1.6 | 0.66332 | [1, 2, 2, 1, 1, 2, 2, 1, 3, 1] |
| (% | -0.9 | 0.9434 | [0, 0, 1, -1, -1, -1, -2, -2, -1, -2] |
| ('-: | 2.2 | 1.16619 | [4, 1, 4, 3, 1, 2, 3, 1, 2, 1] |
| (': | 2.3 | 0.9 | [1, 3, 3, 2, 2, 4, 2, 3, 1, 2] |
| ((-: | 2.1 | 0.53852 | [2, 2, 2, 1, 2, 3, 2, 2, 3, 2] |

Question 5

As you can see, the lexicon contains emojis too! Each row contains a word and the *polarity* of that word, measuring how positive or negative the word is.

(How did they decide the polarities of these words? What are the other two columns in the lexicon? See the link above.)

Question 5a

Read in the lexicon into a DataFrame called `sent`. The index of the DataFrame should be the words in the lexicon. `sent` should have one column named `polarity`, storing the polarity of each word.

- **Hint:** The `pd.read_csv` function may help here.

```
In [82]: sent = pd.read_csv('vader_lexicon.txt', sep='\t', usecols=[0, 1], names=['token', 'sentiment'])
sent.head()
```

Out[82]:

| | polarity |
|-------|----------|
| token | |
| \$: | -1.5 |
| %) | -0.4 |
| %-) | -1.5 |
| &-: | -0.4 |
| &: | -0.7 |

```
In [83]: ok.grade("q5a");
```

```
~~~~~
Running tests

-----
Test summary
  Passed: 4
  Failed: 0
[ooooooooook] 100.0% passed
```

Question 5b

Now, let's use this lexicon to calculate the overall sentiment for each of Trump's tweets. Here's the basic idea:

1. For each tweet, find the sentiment of each word.
2. Calculate the sentiment of each tweet by taking the sum of the sentiments of its words.

First, let's lowercase the text in the tweets since the lexicon is also lowercase. Set the `text` column of the `trump` DataFrame to be the lowercased text of each tweet.


```
In [84]: trump['text'] = trump['text'].str.lower()
trump.head()
```

Out[84]:

| | time | source | text | retweet_count | year | est_tim |
|---------------------------|------------------------|---------------------------|---|---------------|-------------|-------------------------------|
| 690171032150237184 | 2016-01-21 13:56:11 | Twitter for Android | "@bigop1: @realdonaldtrump @sarahpalinusa https://t.co/3kyqgqevyd" | 1059 | 2016.054645 | 2016-01-2 08:56:11 05:0 |
| 690171403388104704 | 2016-01-21 13:57:39 | Twitter for Android | "@americanaspie: @glennbeck @sarahpalinusa remember when glenn gave out gifts to illegal aliens at crossing the border? me too!" | 1339 | 2016.054645 | 2016-01-2 08:57:39 05:0 |
| 690173226341691392 | 2016-01-21 14:04:54 | Twitter for Android | so sad that @cnn and many others refused to show the massive crowd at the arena yesterday in oklahoma. dishonest reporting! | 2006 | 2016.054645 | 2016-01-2 09:04:54 05:0 |
| 690176882055114758 | 2016-01-21 14:19:26 | Twitter for Android | sad sack @jebbush has just done another ad on me, with special interest money, saying i won't beat hillary - i will. but he can't beat me. | 2266 | 2016.054645 | 2016-01-2 09:19:26 05:0 |
| 690180284189310976 | 2016-01-21 14:32:57 | Twitter for Android | low energy candidate @jebbush has wasted \$80 million on his failed presidential campaign. millions spent on me. he should go home and relax! | 2886 | 2016.054645 | 2016-01-2 09:32:57 05:0 |

```
In [85]: ok.grade("q5b");
```

~~~~~  
Running tests

-----  
Test summary

Passed: 1

Failed: 0

[ooooooooook] 100.0% passed

## Question 5c

Now, let's get rid of punctuation since it will cause us to fail to match words. Create a new column called `no_punc` in the `trump` DataFrame to be the lowercased text of each tweet with all punctuation replaced by a single space. We consider punctuation characters to be *any character that isn't a Unicode word character or a whitespace character*. You may want to consult the Python documentation on regexes for this problem.

(Why don't we simply remove punctuation instead of replacing with a space? See if you can figure this out by looking at the tweet data.)

```
In [90]: # Save your regex in punct_re
punct_re = r'^\w\s]'
trump['no_punc'] = trump['text'].str.replace(punct_re, ' ')
```

```
In [91]: ok.grade("q5c");
```

~~~~~  
Running tests

Test summary

Passed: 10

Failed: 0

[ooooooooook] 100.0% passed

Question 5d

Now, let's convert the tweets into what's called a [tidy format](https://cran.r-project.org/web/packages/tidyr/vignettes/tidy-data.html) (<https://cran.r-project.org/web/packages/tidyr/vignettes/tidy-data.html>) to make the sentiments easier to calculate. Use the `no_punc` column of `trump` to create a table called `tidy_format`. The index of the table should be the IDs of the tweets, repeated once for every word in the tweet. It has two columns:

1. `num` : The location of the word in the tweet. For example, if the tweet was "i love america", then the location of the word "i" is 0, "love" is 1, and "america" is 2.
2. `word` : The individual words of each tweet.

The first few rows of our `tidy_format` table look like:

| | num | word |
|--------------------|-----|------------|
| 894661651760377856 | 0 | i |
| 894661651760377856 | 1 | think |
| 894661651760377856 | 2 | senator |
| 894661651760377856 | 3 | blumenthal |
| 894661651760377856 | 4 | should |

Note that your DataFrame may look different from the one above. However, you can double check that your tweet with ID 894661651760377856 has the same rows as ours. Our tests don't check whether your table looks exactly like ours.

As usual, try to avoid using any for loops. Our solution uses a chain of 5 methods on the `trump` DataFrame, albeit using some rather advanced Pandas hacking.

- **Hint 1:** Try looking at the `expand` argument to pandas' `str.split`.
- **Hint 2:** Try looking at the `stack()` method.
- **Hint 3:** Try looking at the `level` parameter of the `reset_index` method.

```
In [94]: tidy_format = trump['no_punc'].str.split(expand=True).stack().reset_index(1)
tidy_format.head(5)
```

Out[94]:

| | num | word |
|--------------------|-----|-----------------|
| 690171032150237184 | 0 | bigop1 |
| 690171032150237184 | 1 | realdonaldtrump |
| 690171032150237184 | 2 | sarahpalinusa |
| 690171032150237184 | 3 | https |
| 690171032150237184 | 4 | t |

```
In [95]: ok.grade("q5d");
```

~~~~~  
Running tests

-----  
Test summary

Passed: 2

Failed: 0

[ooooooooook] 100.0% passed

## Question 5e

Now that we have this table in the tidy format, it becomes much easier to find the sentiment of each tweet: we can join the table with the lexicon table.

Add a `polarity` column to the `trump` table. The `polarity` column should contain the sum of the sentiment polarity of each word in the text of the tweet.

### Hints:

- You will need to merge the `tidy_format` and `sent` tables and group the final answer.
- If certain words are not found in the `sent` table, set their polarities to 0.

```
In [96]: trump['polarity'] = (tidy_format.merge(sent, how='left', left_on='word', ri
trump[['text', 'polarity']].head())
```

Out[96]:

|                    |                                                                                                                                               | text | polarity |
|--------------------|-----------------------------------------------------------------------------------------------------------------------------------------------|------|----------|
| 690171032150237184 | "@bigop1: @realdonaldtrump @sarahpalinusa https://t.co/3kyqgqevyd"                                                                            |      | 0.0      |
| 690171403388104704 | "@americanaspie: @glennbeck @sarahpalinusa remember when glenn gave out gifts to illegal aliens at crossing the border? me too!"              |      | -2.6     |
| 690173226341691392 | so sad that @cnn and many others refused to show the massive crowd at the arena yesterday in oklahoma. dishonest reporting!                   |      | -6.0     |
| 690176882055114758 | sad sack @jebbush has just done another ad on me, with special interest money, saying i won't beat hillary - i will. but he can't beat me.    |      | 4.3      |
| 690180284189310976 | low energy candidate @jebbush has wasted \$80 million on his failed presidential campaign. millions spent on me. he should go home and relax! |      | -2.6     |

```
In [97]: ok.grade("q5e");
```

```
~~~~~
Running tests

Test summary
 Passed: 6
 Failed: 0
[ooooooooook] 100.0% passed
```

Now we have a measure of the sentiment of each of his tweets! Note that this calculation is rather basic; you can read over the VADER readme to understand a more robust sentiment analysis.

Now, run the cells below to see the most positive and most negative tweets from Trump in your dataset:

```
In [98]: print('Most negative tweets:')
for t in trump.sort_values('polarity').head()['text']:
 print('\n ', t)
```

Most negative tweets:

the trump portrait of an unsustainable border crisis is dead on. "in the last two years, ice officers made 266,000 arrests of aliens with criminal records, including those charged or convicted of 100,000 assaults, 30,000 sex crimes & 4000 violent killings." america's southern....

it is outrageous that poisonous synthetic heroin fentanyl comes pouring into the u.s. postal system from china. we can, and must, end this now! the senate should pass the stop act – and firmly stop this poison from killing our children and destroying our country. no more delay!

the rigged russian witch hunt goes on and on as the "originators and founders" of this scam continue to be fired and demoted for their corrupt and illegal activity. all credibility is gone from this terrible hoax, and much more will be lost as it proceeds. no collusion!

...this evil anti-semitic attack is an assault on humanity. it will take all of us working together to extract the poison of anti-semitism from our world. we must unite to conquer hate.

james comey is a proven leaker & liar. virtually everyone in washington thought he should be fired for the terrible job he did-until he was, in fact, fired. he leaked classified information, for which he should be prosecuted. he lied to congress under oath. he is a weak and.....

```
In [99]: print('Most positive tweets:')
for t in trump.sort_values('polarity', ascending=False).head()['text']:
 print('\n ', t)
```

Most positive tweets:

congratulations to patrick reed on his great and courageous masters win! when patrick had his amazing win at doral 5 years ago, people saw his great talent, and a bright future ahead. now he is the masters champion!

congratulations to a truly great football team, the clemson tigers, on an incredible win last night against a powerful alabama team. a big win also for the great state of south carolina. look forward to seeing the team, and their brilliant coach, for the second time at the w.h.

my supporters are the smartest, strongest, most hard working and most loyal that we have seen in our countries history. it is a beautiful thing to watch as we win elections and gather support from all over the country. as we get stronger, so does our country. best numbers ever!

thank you to all of my great supporters, really big progress being made. other countries wanting to fix crazy trade deals. economy is roaring. supreme court pick getting great reviews. new poll says trump, at over 90%, is the most popular republican in history of the party. wow!

thank you, @wvgovernor jim justice, for that warm introduction. tonight, it was my great honor to attend the "greenbrier classic – salute to service dinner" in west virginia! god bless our veterans. god bless america – and happy independence day to all! <https://t.co/v35qvcn8m6> (<https://t.co/v35qvcn8m6>)

## Question 6

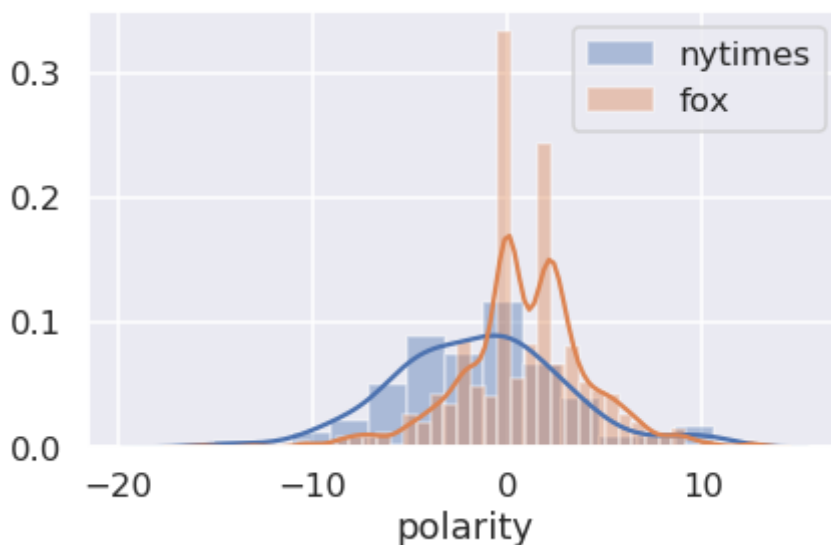
Now, let's try looking at the distributions of sentiments for tweets containing certain keywords.

### Question 6a

In the cell below, create a single plot showing both the distribution of tweet sentiments for tweets containing `nytimes`, as well as the distribution of tweet sentiments for tweets containing `fox`.

```
In [104]: sns.distplot(trump[trump['text'].str.lower().str.contains("nyt")]['polarity'])
sns.distplot(trump[trump['text'].str.lower().str.contains("fox")]['polarity'])
plt.legend()
```

Out[104]: <matplotlib.legend.Legend at 0x7f5604029240>



## Question 6b

Comment on what you observe in the plot above. Can you find other pairs of keywords that lead to interesting plots? (If you modify your code in 6a, remember to change the words back to `nytimes` and `fox` before submitting for grading).

Clearly, the president appears to say more positive things about Fox than the New York Times. It also appears that the president appears to say more positive things about Fox than CNN.

```
In [105]: # Save your notebook first, then run this cell to submit.
import jassign.to_pdf
jassign.to_pdf.generate_pdf('hw3.ipynb', 'hw3.pdf')
ok.submit()
```

Generating PDF...

Saved hw3.pdf

Saving notebook... Saved 'hw3.ipynb'.

Submit... 100% complete

Submission successful for user: kushal.singh@berkeley.edu

URL: <https://okpy.org/cal/data100/sp19/hw3/submissions/jZZ1xl> (<https://okpy.org/cal/data100/sp19/hw3/submissions/jZZ1xl>)

In [ ]: