Text as Data: Homework 1

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In this homework assignment we're going to analyze the first presidential debate from the 2012 election.

Problem 1

To analyze the debate, we first need to load the debate and parse the content. On the coursewebsite, you'll find the file debate1.html. Download the file and open it in a browser. We will use BeautifulSoup to parse HTML file containing the debate transcript.

• Load the webpage into Python and use BeautifulSoup to create a searchable version of the debate. What tags can you use to identify statements?

• Note that not all of the statements contain information about the speaker. Devise a rule to assign the unlabeled statements to speakers. For substantive reasons, we would like to define a single statement as any uninterrupted speech from a candidate. We'll say a candidate is interrupted when the transcript says that a new speaker has begun. In other words, cross talk doesn't count as an interruption. Create a list with just the text (not the tags) of each statement as an element. Some statements are split among several tags; these will need to be concatenated according to the rule you devised above. Remember to filter out notes about audience behavior.

```
1 # we know that there are three speakers
2 # Speakers: FORMER GOV. MITT ROMNEY, R-MASS; PRESIDENT BARACK OBAMA;
3 # JIM LEHRER, MODERATOR
```

```
4 # but even if we didn't know which names to search for, it appears
5 # they are labeled by all caps
6 # which is how we'll identify who is speaking and speaker changes
8 # create empty vector to be filled with statements
9 statements = []
10 # prior speaker is set to NULL, but will be filled with the most
11 # recent speaker
priorSpeaker = ''
14 # iterate over each text block (excluding the introduction and ending)
15 for i in pageText[6:477]:
    # first, convert all  from bs4 object to strings to be searched
    # and get rid of HTML in strings
    # '\' will appear, but it's just to escape the apostrophes
18
      cleanedStatements = re.sub(re.compile('<.*?>'), '', str(i))
      # then check if there is a fully capitalized word at the beginning
2.0
      # of each statement
21
      speakerLabelled = re.search(', [A-Z]+:', cleanedStatements)
22
      # and if there is...
23
      if speakerLabelled:
24
        # record who the current speaker is (by checking which portion
25
        # of the string matched the regex))
26
        currentSpeaker = speakerLabelled.group()
27
        # if the current speaker matches the prior speaker, add cleaned
28
        # statement to the last full statement that was added
29
        # and remove current speaker from every other statement
30
        # except the first
31
          if currentSpeaker == priorSpeaker:
            # since no index is specified, .pop() removes and returns
33
            # the last item in the list
34
            # Note: there is an extra space added because otherwise
35
            # append will crunch words together
36
              statements.append(statements.pop() + " " +
37
              cleanedStatements.replace(currentSpeaker, ''))
38
          # if the current speaker is different than prior speaker,
39
          # add cleaned statement on its own
40
41
              statements.append(cleanedStatements)
42
              # and reset prior speaker to the most recently recorded
43
     speaker
              priorSpeaker = speakerLabelled.group(0)
44
      # if there is no speaker listed (does not match regex search),
45
      # add cleaned statement to the last full statement that was added
46
47
        statements.append(statements.pop() + " " + cleanedStatements)
```

Problem 2

Now we're going to do some more preprocessing to create a dataset that includes useful information about our texts. We will use a curated dictionary list from Neal Caren. The positive

words are at http://www.unc.edu/ \sim ncaren/haphazard/positive.txt and the negative words are at http://www.unc.edu/ \sim ncaren/haphazard/negative.txt.

• Load the positive and negative words into python. Use the porter, snowball and lancaster stemmers from the nltk package to create stemmed versions of the dictionaries.

```
# create function to load sentimental dictionaries
def loadWords(type, stemmer):
    # open url specifying positive or negative dictionary
    url = urlopen('http://www.unc.edu/~ncaren/haphazard/' + type + '.txt').
    # since they are in .txt files, we need to split each word
    # create the unstemmed dictionary
6
    unstemmedDict = url.split(' \ ')
    # determine which stemmer should be used
    # (1) Porter
    if stemmer="Porter":
      # for each word in dictionary, stem
      stemmedDict = [nltk.stem.PorterStemmer().stem(word) for word in
     unstemmedDict]
    # (2) Snowball
    elif stemmer="Snowball":
14
      # for each word in dictionary, stem
      stemmedDict = [nltk.stem.SnowballStemmer('english').stem(word) for
     word in unstemmedDict]
    # (3) Lancaster
17
    elif stemmer="Lancaster":
18
      stemmedDict = [nltk.stem.LancasterStemmer().stem(word) for word in
19
     unstemmedDict]
    else:
20
      stemmedDict = unstemmedDict
21
    # return both stemmed and unstemmed dictionaries
22
    # by using set() instead of keeping them as lists
    # removes duplicates
24
    return [unstemmedDict, set(stemmedDict)]
25
27 # get basic positive and negative, unstemmed dictionaries
positiveWords = loadWords('positive', stemmer="None").pop(0)
negativeWords = loadWords('negative', stemmer="None").pop(0)
31 # run dictionary acquisition and stemming function for all stemmers
32 # (1) Porter
33 stemmedPositivePorter = loadWords('positive', stemmer="Porter").pop(1)
34 stemmedNegativePorter = loadWords('negative', stemmer="Porter").pop(1)
35
36 \# (2) Snowball
37 stemmedPositiveSnowball = loadWords('positive', stemmer="Snowball").pop
38 stemmedNegativeSnowball = loadWords('negative', stemmer="Snowball").pop
      (1)
39
```

- Using the original and stemmed dictionaries, we're going to create a statement by statement data set of the speech. The data set should have the following columns:
 - 1) Statement number (place in debate)
 - 2) Speaker
 - 3) Number of non-stop words spoken
 - 4) Number of positive words
 - 5) Number of negative words
 - 6) Number of lancaster stemmed positive words
 - 7) Number of lancaster stemmed negative words
 - 8) Number of porter stemmed positive words
 - 9) Number of porter stemmed negative words
 - 10) Number of snowball stemmed positive words
 - 11) Number of snowball stemmed negative words

To create the data set, create a set of nested dictionaries that map each statement in the list created in Problem 1 to the each of the attributes described above. To calculate the values for items 3 - 11 above, you'll need to do the following to each statement:

- Discard punctuation
- Remove capitalization
- Remove stop words with the list of words provided here:
 'http://jmlr.org/papers/volume5/lewis04a/a11-smart-stop-list/english.stop'
- Tokenize the words
- Apply each of the stemmers, determining which of the words appear in the corresponding stemmed dictionaries

Write your dataset as a .csv file and save it to a working directory. Turn it in with your homework.

```
1 # create function that will easily check how many words are in
2 # corresponding dictionary list
3 def wordCount(inputStatement, dictionaries):
    return len([x for x in inputStatement if x in dictionaries])
5 # create function to pull necessary info from each statement
6 def statementInfo(statement, documentContent, count):
    # first, need to discard punctuation
    removedPunctuation = re.sub("\W", " ", i)
    # capitalization
9
    removedCaps = removedPunctuation.lower()
10
    # and tokenization
    reducedStatements = nltk.word_tokenize(removedCaps)
13
      # append documentContent with relevant info
14
    documentContent.append({
    # add to statementIter
16
17
    "statementNumber": count,
    "speaker": re.search(``^[A\!-\!Z]+`, statement).group()\;,
18
    # record the number of ___ in statements w/ no punctuation, caps,
19
    # and reduced tokens:
20
    # non-stop words
    #"NstopWords": len([x for x in reducedStatements if x not in stop_words]),
22
    # number of positive words
23
    "NposWords": wordCount(reducedStatements, positiveWords),
24
      # number of negative words
25
      "NnegWords": wordCount(reducedStatements, negativeWords),
26
      # number of words in each positive and negative using:
      # (1) Porter stem
28
      "NposPorter": wordCount([nltk.stem.PorterStemmer().stem(y) for y in
     reducedStatements], stemmedPositivePorter),
      "NnegPorter": wordCount([nltk.stem.PorterStemmer().stem(y) for y in
30
     reducedStatements], stemmedNegativePorter),
      # (2) Snowball stem
      "NposSnowball": wordCount([nltk.stem.SnowballStemmer('english').stem(y)
     for y in reducedStatements, stemmedPositiveSnowball,
      "NnegSnowball": wordCount([nltk.stem.SnowballStemmer('english').stem(y)
33
     for y in reducedStatements], stemmedNegativeSnowball),
      # (3) Lancaster stem
34
      "NposLancaster": wordCount([nltk.stem.LancasterStemmer().stem(y) for y in
35
     reducedStatements], stemmedPositiveLancaster),
      "NnegLancaster": wordCount([nltk.stem.LancasterStemmer().stem(y) for y in
36
     reducedStatements], stemmedNegativeLancaster)})
37
38 # create empty list to fill with statement info
39 statementCharacteristics = []
40 # begin document iterations at 0
statementIter = 0
42 for i in statements:
   # execute statementInfo function for each statement
   # begin document iterations at 1
    statementIter +=1
```

```
statementInfo(i, statementCharacteristics, count=statementIter)

# with data now assigned to dictionary

# write content to .csv

with open('Documents/Git/WUSTL_textAnalysis/statmentInfo.csv', 'wb') as f:

w = csv.DictWriter(f, fieldnames=("statementNumber", "speaker",
"NposWords", "NnegWords", "NposPorter", "NnegPorter",
"NposSnowball", "NnegSnowball", "NposLancaster", "NnegLancaster"))

w. writeheader()

for item in statementCharacteristics:

w. writerow(item)
```

Problem 3

Using our new data set, let's make some observations about the debate

- Load the data into R
- Create a visualization that compares the overall positive and negative word rate for Obama, Romney, and Lehrer. What patterns do you notice? There is no one right answer, be creative!
- Using your data set, examine trends in each candidate's statements and Lehrer's speeches. Do you notice any
 - i) Trends in the measured tone?
 - ii) Response to the other candidate's tone (examining who spoke previously)?
 - iii) Overall interesting patterns? (this is an intentionally vague question)