```
In [56]:
         from textstat.textstat import textstat
         %matplotlib inline
         import numpy as np
         import pandas as pd
         import scipy
         import sklearn
         #import spacy
         import matplotlib.pvplot as plt
         import seaborn as sns
         import re
         from collections import Counter
         from sklearn.feature extraction.text import TfidfTransformer
         from sklearn.feature extraction.text import CountVectorizer
         from sklearn.feature extraction.text import TfidfVectorizer
         from sklearn import linear model
         from sklearn.metrics import accuracy score
         from sklearn.naive bayes import MultinomialNB
         from sklearn.naive bayes import GaussianNB
         from gensim.models import doc2vec
         from collections import namedtuple
         #nlp = spacy.load('en')
         import en core web sm
         nlp = en_core_web_sm.load()
         from nltk.corpus import stopwords
         from sklearn.model selection import train test split
         import pyphen
```

For this challenge, you will need to choose a corpus of data from nltk or another source that includes categories you can predict and create an analysis pipeline that includes the following steps:

- 1. Data cleaning / processing / language parsing
- 2. Create features using two different NLP methods: For example, BoW vs tf-idf.
- 3. Use the features to fit supervised learning models for each feature set to predict the category outcomes.
- 4. Assess your models using cross-validation and determine whether one model performed better.
- 5. Pick one of the models and try to increase accuracy by at least 5 percentage points.

In [58]: # Import the data we just downloaded and installed. from nltk.corpus import inaugural

> # Grab and process the raw data. inaugural.fileids()

```
Out[58]: ['1789-Washington.txt',
           '1793-Washington.txt',
           '1797-Adams.txt',
           '1801-Jefferson.txt',
           '1805-Jefferson.txt',
           '1809-Madison.txt',
           '1813-Madison.txt',
           '1817-Monroe.txt',
           '1821-Monroe.txt',
           '1825-Adams.txt',
           '1829-Jackson.txt',
           '1833-Jackson.txt',
           '1837-VanBuren.txt',
           '1841-Harrison.txt',
           '1845-Polk.txt',
           '1849-Taylor.txt',
           '1853-Pierce.txt',
           '1857-Buchanan.txt',
           '1861-Lincoln.txt',
           '1865-Lincoln.txt',
           '1869-Grant.txt',
           '1873-Grant.txt',
           '1877-Hayes.txt',
           '1881-Garfield.txt',
           '1885-Cleveland.txt',
           '1889-Harrison.txt',
           '1893-Cleveland.txt',
           '1897-McKinley.txt',
           '1901-McKinley.txt',
           '1905-Roosevelt.txt',
           '1909-Taft.txt',
           '1913-Wilson.txt',
           '1917-Wilson.txt',
           '1921-Harding.txt',
           '1925-Coolidge.txt',
           '1929-Hoover.txt',
           '1933-Roosevelt.txt',
           '1937-Roosevelt.txt',
           '1941-Roosevelt.txt',
           '1945-Roosevelt.txt',
           '1949-Truman.txt',
           '1953-Eisenhower.txt',
           '1957-Eisenhower.txt',
           '1961-Kennedy.txt',
           '1965-Johnson.txt',
           '1969-Nixon.txt',
           '1973-Nixon.txt',
           '1977-Carter.txt',
           '1981-Reagan.txt',
           '1985-Reagan.txt',
           '1989-Bush.txt',
           '1993-Clinton.txt',
           '1997-Clinton.txt',
           '2001-Bush.txt',
           '2005-Bush.txt',
           '2009-Obama.txt']
```

1. Data cleaning / processing / language parsing

Raw format of the data

Resources: https://towardsdatascience.com/intro-to-nlp-using-inaugural-speeches-of-presidents-8c7ca32cbdfe)

In [60]: words=list(inaugural.words(fileids = names[55]))
 words

```
Out[60]: ['My',
            'fellow',
            'citizens',
            ':',
'I',
            'stand',
            'here',
            'today',
            'humbled',
            'by',
            'the',
            'task',
            'before',
            'us',
            ٠,',
            'grateful',
            'for',
            'the',
            'trust',
            'you',
            'have',
            'bestowed',
            ٠,٠,
            'mindful',
            'of',
            'the',
            'sacrifices',
            'borne',
            'by',
            'our',
            'ancestors',
            ۱.',
            'I',
            'thank',
            'President',
            'Bush',
            'for',
            'his',
            'service',
            'to',
            'our',
            'nation',
            ',',
            'as',
            'well',
            'as',
            'the',
            'generosity',
            'and',
            'cooperation',
            'he',
            'has',
            'shown',
            'throughout',
            'this',
            'transition',
```

```
'Forty',
'-',
'four',
'Americans',
'have',
'now',
'taken',
'the',
'presidential',
'oath',
'.',
'The',
'words',
'have',
'been',
'spoken',
'during',
'rising',
'tides',
'of',
'prosperity',
'and',
'the',
'still',
'waters',
'of',
'peace',
'.',
'Yet',
٠,',
'every',
'so',
'often',
'the',
'oath',
'is',
'taken',
'amidst',
'gathering',
'clouds',
'and',
'raging',
'storms',
٠٠',
'At',
'these',
'moments',
٠,',
'America',
'has',
'carried',
'on',
'not',
'simply',
'because',
'of',
'the',
```

```
'skill',
'or',
'vision',
'of',
'those',
'in',
'high',
'office',
٠,',
'but',
'because',
'We',
'the',
'People',
'have',
'remained',
'faithful',
'to',
'the',
'ideals',
'of',
'our',
'forbearers',
ر'ر'
'and',
'true',
'to',
'our',
'founding',
'documents',
٠.',
'So',
'it',
'has',
'been',
'.',
'So',
'it',
'must',
'be',
'with',
'this',
'generation',
'of',
'Americans',
٠٠',
'That',
'we',
'are',
'in',
'the',
'midst',
'of',
'crisis',
'is',
'now',
'well',
```

```
'understood',
١.',
'Our',
'nation',
'is',
'at',
'war',
٠,٠,
'against',
'a',
'far',
'-',
'reaching',
'network',
'of',
'violence',
'and',
'hatred',
٠.',
'Our',
'economy',
'is',
'badly',
'weakened',
',',
'a',
'consequence',
'of',
'greed',
'and',
'irresponsibility',
'on',
'the',
'part',
'of',
'some',
٠,',
'but',
'also',
'our',
'collective',
'failure',
'to',
'make',
'hard',
'choices',
'and',
'prepare',
'the',
'nation',
'for',
'a',
'new',
'age',
١.',
'Homes',
'have',
```

```
'been',
'lost',
';',
'jobs',
'shed',
';',
'businesses',
'shuttered',
٠.',
'Our',
'health',
'care',
'is',
'too',
'costly',
';',
'our',
'schools',
'fail',
'too',
'many',
';',
'and',
'each',
'day',
'brings',
'further',
'evidence',
'that',
'the',
'ways',
'we',
'use',
'energy',
'strengthen',
'our',
'adversaries',
'and',
'threaten',
'our',
'planet',
١.',
'These',
'are',
'the',
'indicators',
'of',
'crisis',
'subject',
'to',
'data',
'and',
'statistics',
٠.',
'Less',
'measurable',
```

```
'but',
'no',
'less',
'profound',
'is',
'a',
'sapping',
'of',
'confidence',
'across',
'our',
'land',
'--',
'a',
'nagging',
'fear',
'that',
'America',
""",
's',
'decline',
'is',
'inevitable',
٠,٠,
'that',
'the',
'next',
'generation',
'must',
'lower',
'its',
'sights',
'.',
'Today',
'I',
'say',
'to',
'you',
'that',
'the',
'challenges',
'we',
'face',
'are',
'real',
٠٠',
'They',
'are',
'serious',
'and',
'they',
'are',
'many',
'.',
'They',
'will',
'not',
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'be',
'met',
'easily',
'or',
'in',
'a',
'short',
'span',
'of',
'time',
١.',
'But',
'know',
'this',
',',
'America',
'--',
'they',
'will',
'be',
'met',
٠.',
'On',
'this',
'day',
ر ٔ ر ٔ
'we',
'gather',
'because',
'we',
'have',
'chosen',
'hope',
'over',
'fear',
',',
'unity',
'of',
'purpose',
'over',
'conflict',
'and',
'discord',
'.',
'On',
'this',
'day',
ر ٔ ر ٔ
'we',
'come',
'to',
'proclaim',
'an',
'end',
'to',
'the',
'petty',
```

```
'grievances',
'and',
'false',
'promises',
٠,٠,
'the',
'recriminations',
'and',
'worn',
'-',
'out',
'dogmas',
'that',
'for',
'far',
'too',
'long',
'have',
'strangled',
'our',
'politics',
'.',
'We',
'remain',
'a',
'young',
'nation',
٠,٠,
'but',
'in',
'the',
'words',
'of',
'Scripture',
',',
'the',
'time',
'has',
'come',
'to',
'set',
'aside',
'childish',
'things',
٠.',
'The',
'time',
'has',
'come',
'to',
'reaffirm',
'our',
'enduring',
'spirit',
';<sup>'</sup>,
'to',
'choose',
```

```
'our',
'better',
'history',
';',
'to',
'carry',
'forward',
'that',
'precious',
'gift',
',',
'that',
'noble',
'idea',
٠,',
'passed',
'on',
'from',
'generation',
'to',
'generation',
':',
'the',
'God',
'-',
'given',
'promise',
'that',
'all',
'are',
'equal',
'all',
'are',
'free',
ر','
'and',
'all',
'deserve',
'a',
'chance',
'to',
'pursue',
'their',
'full',
'measure',
'of',
'happiness',
٠٠',
'In',
'reaffirming',
'the',
'greatness',
'of',
'our',
'nation',
٠,',
```

```
'we',
'understand',
'that',
'greatness',
'is',
'never',
'a',
'given',
'.',
'It',
'must',
'be',
'earned',
١.',
'Our',
'journey',
'has',
'never',
'been',
'one',
'of',
'shortcuts',
'or',
'settling',
'for',
'less',
٠.',
'It',
'has',
'not',
'been',
'the',
'path',
'for',
'the',
'faint',
'-',
'hearted',
'--',
'for',
'those',
'who',
'prefer',
'leisure',
'over',
'work',
٠,',
'or',
'seek',
'only',
'the',
'pleasures',
'of',
'riches',
'and',
'fame',
٠٠',
```

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'Rather',
۰,۰,
'it',
'has',
'been',
'the',
'risk',
'-',
'takers',
٠,٠,
'the',
'doers',
٠,٠,
'the',
'makers',
'of',
'things',
"'",
'some',
'celebrated',
'but',
'more',
'often',
'men',
'and',
'women',
'obscure',
'in',
'their',
'labor',
٠,٠,
'who',
'have',
'carried',
'us',
'up',
'the',
'long',
',',
'rugged',
'path',
'towards',
'prosperity',
'and',
'freedom',
٠.',
'For',
'us',
'they',
'packed',
'up',
'their',
'few',
'worldly',
'possessions',
'and',
```

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'traveled',
'across',
'oceans',
'in',
'search',
'of',
'a',
'new',
'life',
٠.',
'For',
'us',
٠,',
'they',
'toiled',
'in',
'sweatshops',
'and',
'settled',
'the',
'West',
';',
'endured',
'the',
'lash',
'of',
'the',
'whip',
'and',
'plowed',
'the',
'hard',
'earth',
۱.',
'For',
'us',
',',
'they',
'fought',
'and',
'died',
',',
'in',
'places',
'like',
'Concord',
'and',
'Gettysburg',
';',
'Normandy',
'and',
'Khe',
'Sahn',
٠٠',
'Time',
'and',
'again',
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```
'these',
'men',
'and',
'women',
'struggled',
'and',
'sacrificed',
'and',
'worked',
'till',
'their',
'hands',
'were',
'raw',
'so',
'that',
'we',
'might',
'live',
'a',
'better',
'life',
٠.',
'They',
'saw',
'America',
'as',
'bigger',
'than',
'the',
'sum',
'of',
'our',
'individual',
'ambitions',
';',
'greater',
'than',
'all',
'the',
'differences',
'of',
'birth',
'or',
'wealth',
'or',
'faction',
'.',
'This',
'is',
'the',
'journey',
'we',
'continue',
'today',
٠٠',
'We',
```

```
'remain',
'the',
'most',
'prosperous',
٠,٠,
'powerful',
'nation',
'on',
'Earth',
٠.',
'Our',
'workers',
'are',
'no',
'less',
'productive',
'than',
'when',
'this',
'crisis',
'began',
'.',
'Our',
'minds',
'are',
'no',
'less',
'inventive',
',',
'our',
'goods',
'and',
'services',
'no',
'less',
'needed',
'than',
'they',
'were',
'last',
'week',
'or',
'last',
'month',
'or',
'last',
'year',
'.',
'Our',
'capacity',
'remains',
'undiminished',
٠٠',
'But',
'our',
'time',
'of',
```

```
'standing',
'pat',
',',
of',
'protecting',
'narrow',
'interests',
'and',
'putting',
'off',
'unpleasant',
'decisions',
'--',
'that',
'time',
'has',
'surely',
'passed',
'.',
'Starting',
'today',
٠,',
'we',
'must',
'pick',
'ourselves',
'up',
٠, ٔ,
'dust',
'ourselves',
'off',
',',
'and',
'begin',
'again',
'the',
'work',
'of',
'remaking',
'America',
٠٠',
'For',
'everywhere',
'we',
'look',
',',
'there',
'is',
'work',
'to',
'be',
'done',
٠٠',
'The',
'state',
'of',
'our',
```

```
'economy',
'calls',
'for',
'action',
',',
'bold',
'and',
'swift',
٠,',
'and',
'we',
'will',
'act',
'--',
'not',
'only',
'to',
'create',
'new',
'jobs',
٠, ٔ ,
'but',
'to',
'lay',
'a',
'new',
'foundation',
'for',
'growth',
٠٠',
'We',
'will',
'build',
'the',
'roads',
'and',
'bridges',
٠,',
'the',
'electric',
'grids',
'and',
'digital',
'lines',
'that',
'feed',
'our',
'commerce',
'and',
'bind',
'us',
'together',
'.',
'We',
'will',
'restore',
'science',
```

```
'to',
'its',
'rightful',
'place',
٠,٠,
'and',
'wield',
'technology',
's',
'wonders',
'to',
'raise',
'health',
'care',
"'",
's',
'quality',
'and',
'lower',
'its',
'cost',
١.',
'We',
'will',
'harness',
'the',
'sun',
'and',
'the',
'winds',
'and',
'the',
'soil',
'to',
'fuel',
'our',
'cars',
'and',
'run',
'our',
'factories',
٠٠',
'And',
'we',
'will',
'transform',
'our',
'schools',
'and',
'colleges',
'and',
'universities',
'to',
'meet',
'the',
'demands',
```

In [62]:

```
'of',
           'a',
           'new',
           'age',
           ٠.',
           'All',
           'this',
           'we',
           'can',
           'do',
           ١.',
           'All',
           'this',
           'we',
           'will',
           'do',
           '.',
           'Now',
           ٠,٠,
           'there',
           'are',
           'some',
           'who',
           'question',
           'the',
           'scale',
           'of',
           'our',
           'ambitions',
           '--',
           'who',
           ...]
In [61]: nltk.FreqDist(words)
Out[61]: FreqDist({',': 130, 'the': 126, '.': 108, 'and': 105, 'of': 82, 'to': 66, 'ou
         r': 58, 'we': 50, 'that': 48, 'a': 47, ...})
          stop_words = stopwords.words('english')
          add_to_stop_words=[',', '.', '-', ';', ':', '--', "'", '(',')']
          stop_words.extend(add_to_stop_words)
          stop_words=set(stop_words)
```

In [63]: stop_words

```
Out[63]: {"'",
            'a',
            'about',
            'above',
            'after',
            'again',
            'against',
            'ain',
            'all',
            'am',
            'an',
            'and',
            'any',
            'are',
            'aren',
            "aren't",
            'as',
            'at',
            'be',
            'because',
            'been',
            'before',
            'being',
            'below',
            'between',
            'both',
            'but',
            'by',
            'can',
            'couldn',
            "couldn't",
            'd',
            'did',
            'didn',
            "didn't",
            'do',
            'does',
            'doesn',
            "doesn't",
            'doing',
            'don',
            "don't",
            'down',
            'during',
            'each',
            'few',
            'for',
            'from',
            'further',
```

```
'had',
'hadn',
"hadn't",
'has',
'hasn',
"hasn't",
'have',
'haven',
"haven't",
'having',
'he',
'her',
'here',
'hers',
'herself',
'him',
'himself',
'his',
'how',
'i',
'if',
'in',
'into',
'is',
'isn',
"isn't",
'it',
"it's",
'its',
'itself',
'just',
'11',
'm',
'ma',
'me',
'mightn',
"mightn't",
'more',
'most',
'mustn',
"mustn't",
'my',
'myself',
'needn',
"needn't",
'no',
'nor',
'not',
'now',
'o',
'of',
'off',
'on',
'once',
'only',
'or',
'other',
```

```
'our',
'ours',
'ourselves',
'out',
'over',
'own',
're',
's',
'same',
'shan',
"shan't",
'she',
"she's",
'should',
"should've",
'shouldn',
"shouldn't",
'so',
'some',
'such',
't',
'than',
'that',
"that'll",
'the',
'their',
'theirs',
'them',
'themselves',
'then',
'there',
'these',
'they',
'this',
'those',
'through',
'to',
'too',
'under',
'until',
'up',
've',
'very',
'was',
'wasn',
"wasn't",
'we',
'were',
'weren',
"weren't",
'what',
'when',
'where',
'which',
'while',
'who',
'whom',
```

```
'why',
           'will',
           'with',
           'won',
           "won't",
           'wouldn',
           "wouldn't",
           'y',
           'you',
           "you'd",
           "you'll",
           "you're",
           "you've",
           'your',
           'yours',
           'yourself',
           'yourselves'}
In [64]: for i in stop words:
              if i in words:
                  while i in words:
                      words.remove(i)
In [65]: nltk.FreqDist(words)
Out[65]: FreqDist({'us': 23, 'nation': 12, 'We': 12, 'new': 11, 'America': 10, 'The':
         9, 'Our': 9, 'every': 8, 'must': 8, 'For': 8, ...})
In [66]: def text cleaner(text):
              # Visual inspection identifies a form of punctuation spaCy does not
              # recognize: the double dash '--'. Better get rid of it now!
              text = re.sub(r'--',' ',text)
              text = re.sub("[\[].*?[\]]", "", text)
              text = ' '.join(text.split())
              return text
```

```
In [68]:
         #Presidential parties from Washington to Obama
         #link for all presidential parties: https://www.presidentsusa.net/partyofpresi
         dents.html
         #dem_list - democrat
         #rep list - repbulican
         #fed list - federalist
         #demrep_list - democratic-republican
         #whig list - whig
         dem list = ['2009-Obama.txt','1997-Clinton.txt','1993-Clinton.txt','1977-Carte
         r.txt', '1965-Johnson.txt', '1885-Cleveland.txt', '1893-Cleveland.txt',
                      '1853-Pierce.txt', '1857-Buchanan.txt', '1829-Jackson.txt', '1833-
         Jackson.txt', '1837-VanBuren.txt', '1845-Polk.txt',
                      '1961-Kennedy.txt','1949-Truman.txt','1945-Roosevelt.txt','1941-Ro
         osevelt.txt','1937-Roosevelt.txt','1933-Roosevelt.txt']
         rep_list = ['1953-Eisenhower.txt', '1957-Eisenhower.txt', '1969-Nixon.txt', '1
         973-Nixon.txt', '1981-Reagan.txt', '1985-Reagan.txt', '1989-Bush.txt','2001-Bu
         sh.txt', '2005-Bush.txt'
                      '1861-Lincoln.txt', '1865-Lincoln.txt', '1869-Grant.txt', '1873-Gr
         ant.txt', '1877-Hayes.txt', '1881-Garfield.txt', '1889-Harrison.txt',
                      '1897-McKinley.txt', '1901-McKinley.txt', '1905-Roosevelt.txt', '1
         909-Taft.txt', '1921-Harding.txt', '1925-Coolidge.txt', '1929-Hoover.txt']
         fed list = ['1789-Washington.txt','1793-Washington.txt','1797-Adams.txt']
         demrep list = ['1801-Jefferson.txt', '1805-Jefferson.txt', '1809-Madison.txt',
         '1813-Madison.txt',
                        '1817-Monroe.txt','1821-Monroe.txt','1825-Adams.txt']
         whig_list = [ '1841-Harrison.txt', '1849-Taylor.txt']
         #John Tyler - no address following Harrison's death
         #Millard Fillmore - no address following Taylor's death
         #Andrew Johnson - no address following Lincoln's death
```

Using textstat

https://pypi.org/project/textstat/ (https://pypi.org/project/textstat/)

```
In [69]: import textstat
```

```
In [70]:
         # make lists with the text and features
         speeches = []
         raw text = []
         clean text = []
         reading_ease = []
         smog_index = []
         flesch_kincaid_grade = []
         coleman liau index = []
         readability = []
         chall_readability = []
         diffwords = []
         linsear_write_formula = []
         gunning_fog = []
         text standard = []
         party1 = []
         twords = []
         for p in inaugural.fileids():
             speeches.append(p)
             x = inaugural.raw(p)
             raw text.append(x)
             clean = text cleaner(x)
             clean_text.append(clean)
             ease = textstat.flesch_reading_ease(x)
             reading_ease.append(ease)
             smog = textstat.smog index(x)
             smog index.append(smog)
             fk_grade = textstat.flesch_kincaid_grade(x)
             flesch kincaid grade.append(fk grade)
             liau = textstat.coleman_liau_index(x)
             coleman_liau_index.append(liau)
             read = textstat.automated_readability_index(x)
             readability.append(read)
             read2 = textstat.dale_chall_readability_score(x)
             chall_readability.append(read2)
             words = textstat.difficult words(x)
             diffwords.append(words)
             write = textstat.linsear_write_formula(x)
             linsear write formula.append(write)
             fog = textstat.gunning_fog(x)
             gunning_fog.append(fog)
             standard = textstat.text_standard(x)
             text_standard.append(standard)
             token words = nltk.word tokenize(clean)
             twords.append(token_words)
             if p in dem list:
                  party1.append(1)
                                    #democrat
             elif p in rep_list:
                 party1.append(2) #republican
             elif p in fed list:
                 party1.append(3) #federalist
             elif p in demrep_list:
                  party1.append(4) #democrat-republican
             elif p in whig list:
```

```
party1.append(5) #whig
else:
   party1.append(-1) #unknown
```

```
In [71]: # make dataframe of the lists of features
         sp = pd.DataFrame()
         sp['speeches'] = speeches
         sp['raw_text'] = raw_text
         sp['clean text'] = clean text
         sp['reading_ease'] = reading_ease
         sp['smog_index'] = smog_index
         sp['flesch kincaid grade'] = flesch kincaid grade
         sp['coleman_liau_index'] = coleman_liau_index
         sp['readability'] = readability
         sp['chall readability'] = chall readability
         sp['diffwords'] = diffwords
         sp['linsear_write_formula'] = linsear_write_formula
         sp['gunning_fog'] = gunning_fog
         sp['text standard'] = text standard
         sp['party'] = party1
         sp['tokens'] = twords
```

In [72]: pd.set_option('display.max_columns', None)

In [73]: sp.head(5)

Out[73]:

	speeches	raw_text	clean_text	reading_ease	smog_index	flesch_kinca
0	1789- Washington.txt	Fellow-Citizens of the Senate and of the House	Fellow- Citizens of the Senate and of the House	-9.13	24.5	34.3
1	1793- Washington.txt	Fellow citizens, I am again called upon by the	Fellow citizens, I am again called upon by the	25.80	19.9	20.8
2	1797- Adams.txt	When it was first perceived, in early times, t	When it was first perceived, in early times, t	-26.58	26.2	41.0
3	1801- Jefferson.txt	Friends and Fellow Citizens:\n\nCalled upon to	Friends and Fellow Citizens: Called upon to un	31.32	17.5	20.8
4	1805- Jefferson.txt	Proceeding, fellow citizens, to that qualifica	Proceeding, fellow citizens, to that qualifica	8.99	21.0	29.4

WordNetLemmatizer

https://pythonprogramming.net/lemmatizing-nltk-tutorial/ (https://pythonprogramming.net/lemmatizing-nltk-tutorial/)

```
In [74]: from nltk.stem import WordNetLemmatizer
wordnet_lemmatizer = WordNetLemmatizer()
```

```
In [75]: # this picks the most common words in each speech:
         allwords = []
         def bag_of_words(text):
             # Filter out punctuation and stop words.
             for word in text:
                 word = word.lower()
                 word = wordnet lemmatizer.lemmatize(word)
                 if word.isalnum() == True:
                      if word not in stopwords.words('english'):
                          allwords.append(word)
                      else:
                          continue
                 else:
                      continue
             # Return the most common words.
             return [item[0] for item in Counter(allwords).most_common(4000)]
In [76]: # this goes through the whole speech dataframe to find the BOW
         all_common_words = []
         i = 0
         for i in range(0,sp.shape[0]):
             z = bag of words(sp.tokens[i])
```

```
all_common_words = []
i = 0
for i in range(0,sp.shape[0]):
    z = bag_of_words(sp.tokens[i])
    all_common_words.append(z)
    z = []
    i =+ 1

# Can flatten list of lists [[word, word],[word,word]] ~ sum(list_of_list_s, [])
```

```
In [77]: # this finds the common words among all speeches top words:

cw=[]

for i in range(0,sp.shape[0]):
    for word in all_common_words[i]:
        if word not in cw:
              cw.append(word)
        else:
        continue
```

```
In [78]: print(len(cw))
5263
```

```
In [79]:
         wordcount = pd.DataFrame(columns=cw)
         wordcount['text_sentence'] = sp.clean_text
         #wordcount['text_source'] = sp.party
         wordcount.loc[:, cw] = 0
         list_of_words = []
         for i in range(0,sp.shape[0]):
             for word in sp.tokens[i]:
                 word = word.lower()
                 word = wordnet_lemmatizer.lemmatize(word)
                 if word.isalnum() == True:
                      if word not in stopwords.words('english'):
                         if word in cw:
                              list_of_words.append(word)
             # Populate the row with word counts.
             for w in list_of_words:
                 wordcount.loc[i, w] += 1
             # reset list again
             list_of_words = []
```

In [80]: wordcount.head(10)

Out[80]:

	every	government	public	may	present	country	duty	ha	wa	one	hand	citizen	οι
0	9	9	6	6	5	5	5	5	4	4	4	4	4
1	0	1	0	1	1	1	0	1	0	0	0	1	0
2	5	18	6	13	2	10	3	7	8	1	0	3	0
3	2	13	4	8	0	4	2	4	1	6	0	5	1
4	4	3	14	10	3	5	8	5	3	1	1	10	0
5	2	0	6	1	1	5	3	4	0	2	0	0	0
6	5	3	2	2	0	6	1	6	7	2	1	3	0
7	14	21	8	10	4	11	9	19	6	2	3	9	6
8	13	14	5	15	3	8	6	35	16	6	0	14	3
9	5	21	9	3	1	10	9	27	9	4	3	3	0

```
In [81]: sp.rename(columns={'party': 'political_party'}, inplace=True)
    result = pd.concat([wordcount, sp], axis=1)
```

In [82]: result.head(10)

Out[82]:

	every	government	public	may	present	country	duty	ha	wa	one	hand	citizen	οι
0	9	9	6	6	5	5	5	5	4	4	4	4	4
1	0	1	0	1	1	1	0	1	0	0	0	1	0
2	5	18	6	13	2	10	3	7	8	1	0	3	0
3	2	13	4	8	0	4	2	4	1	6	0	5	1
4	4	3	14	10	3	5	8	5	3	1	1	10	0
5	2	0	6	1	1	5	3	4	0	2	0	0	0
6	5	3	2	2	0	6	1	6	7	2	1	3	0
7	14	21	8	10	4	11	9	19	6	2	3	9	6

	every	government	public	may	present	country	duty	ha	wa	one	hand	citizen	οι
8	13	14	5	15	3	8	6	35	16	6	0	14	3
9	5	21	9	3	1	10	9	27	9	4	3	3	0

In [83]: result.corr()

Out[83]: