# **AIWIR ASSIGNMENT-2**

## SEC: 6E

# Submitted by:

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# Bi word phrase query retrieval using inverted index and positional index

### CODE:

```
In [1]: import pandas
            from nltk.tokenize import sent_tokenize
            from nltk.tokenize import word_tokenize
            from nltk.stem import PorterStemmer
            from nltk.corpus import stopwords
            import string
            import time
            stemmer = PorterStemmer()
   In [2]: data=pandas.read_csv("C:/Users/Rahul Bhat/Desktop/file_name.csv")
            data.drop(['Score','Id','Subreddit','URL','Num of Comments','Text','Date Created'], axis=1, inplace=True)
            data.head()
  Out[2]:
                                              Title Political Lean
             0 No matter who someone is, how they look like, ...
             1 Biden speech draws 38.2 million U.S. TV viewers
             2
                                     State of the union
                                                         Liberal
                     We Should Just Give Poor People Money
                                     Do it for the Dew
                                                         Liberal
   In [3]: sentences = []
            words = []
            filtered = []
In [4]: for i in data. Title:
                  i = i.translate(str.maketrans('', '', string.punctuation))
                  a = sent_tokenize(i)
                  sentences.append(a)
                  b = word_tokenize(i)
                  words.append(b)
                  c = [w.lower() for w in b]
                  filtered.append(c)
         data.head()
Out[4]:
                                            Title Political Lean
          0 No matter who someone is, how they look like, ...
          1 Biden speech draws 38.2 million U.S. TV viewers
                                                       Liberal
                  We Should Just Give Poor People Money
          3
                                                       Liberal
                                   Do it for the Dew
In [5]: data['sentence_tokenized'] = sentences
         data['word_tokenized'] = words
         data['filtered'] = filtered
         data.head()
Out[5]:
                                              Political
                                      Title
                                                                     sentence_tokenized
                                                                                                       word tokenized
                                                                                                                                           filtered
              No matter who someone is, how they
                                                       [No matter who someone is how they look
                                                                                          [No, matter, who, someone, is, how,
                                                                                                                       [no, matter, who, someone, is, how,
                                               Liberal
                                                         Biden speech draws 38.2 million U.S.
                                                                                                                        [biden, speech, draws, 382, million,
                                               Liberal
```

```
In [6]: data['stemmed_words']=data['filtered'].apply(lambda f: [stemmer.stem(word) for word in f])
Out [6]:
                                       Political
                                Title
                                                          sentence_tokenized
                                                                                        word_tokenized
                                                                                                                      filtered
                                                                                                                                         stemmed_words
                                         Lean
               No matter who someone is,
                                                                              [No, matter, who, someone, is,
                                                                                                        [no, matter, who, someone,
                                                                                                                                [no, matter, who, someon, is,
                                                  [No matter who someone is how
                    how they look like, ...
                                                              they look like w ...
                                                                                        how, they, look ...
                                                                                                              is, how, they, look ...
                                                                                                                                          how, they, look,...
                                               [Biden speech draws 382 million US TV viewers]
                Biden speech draws 38.2 million U.S. TV viewers
                                                                                [Biden, speech, draws, 382, million, US, TV, v...
                                                                                                       [biden, speech, draws, 382,
                                                                                                                                  [biden, speech, draw, 382,
                                         Liberal
                                                                                                                                         million, us, tv, vi...
                                                                                                               million, us, tv, v...
          2
                      State of the union
                                        Liberal
                                                            [State of the union]
                                                                                     [State, of, the, union]
                                                                                                             [state, of, the, union]
                                                                                                                                       [state, of, the, union]
                We Should Just Give Poor
                                                 [We Should Just Give Poor People
                                                                              [We, Should, Just, Give, Poor,
                                                                                                       [we, should, just, give, poor,
                                                                                                                                 [we, should, just, give, poor,
                                        Liberal
                         People Money
                                                                                         People, Money
                                                                                                                 people, money)
                       Do it for the Dew
                                        Liberal
                                                             [Do it for the Dew]
                                                                                      [Do, it, for, the, Dew]
                                                                                                              [do, it, for, the, dew]
                                                                                                                                       [do, it, for, the, dew]
In [8]: pos index = {}
          for docid,tList in enumerate(data['stemmed_words']):
              for position, w in enumerate (tList):
                   if w in pos_index.keys():
                        pos index[w][0]+=1
                        if docid in pos index[w][1]:
                            pos index[w][1][docid].append(position)
                        else:
                           pos index[w][1][docid]=[position]
                        temp={docid : [position]}
                        pos_index[w]=[1,temp]
 In [9]: for i in pos index:
                print(i, ":", pos index[i], end='\n\n')
           no: [250, {0: [0], 31: [16], 46: [0], 93: [0], 101: [8], 128: [21, 29], 177: [2], 249: [0], 362: [0], 392: [5], 41
           3: [1], 414: [0, 4], 455: [15], 494: [9], 549: [29], 587: [0, 3], 606: [13], 626: [3], 627: [2], 676: [19], 684: [1
           9], 728: [4], 826: [0], 837: [0], 907: [3], 917: [2], 1101: [0], 1106: [10], 1250: [4], 1361: [4], 1369: [0], 1450:
           [2], 1490: [4], 1556: [4], 1588: [1], 1746: [2], 1756: [38], 1820: [1, 4], 1997: [33], 1998: [11],
           [36], 2262: [16], 2288: [2], 2330: [0], 2331: [0], 2426: [19], 2551: [38], 2558: [0], 2563: [9], 2652: [3], 2661:
            [4], 2669: [0], 2698: [5], 2718: [2], 2753: [17], 2777: [11], 2801: [0], 2931: [1], 2976: [14], 3067: [0], 3188:
           [1], 3276: [12], 3286: [21], 3388: [3], 3371: [7], 3380: [2], 3450: [5], 3597: [0], 3637: [30], 3754: [1], 3994: [1], 4119: [24], 4220: [1], 4381: [6], 4386: [21], 4501: [9], 4505: [10], 4531: [11], 4815: [17], 4835: [1], 5019: [4, 7], 5074: [0], 5083: [6], 5184: [0], 5248: [8], 5259: [2], 5264: [7], 5291: [1], 5437: [2], 5533: [9], 5770:
           [0, 2], 5930: [4, 11], 5931: [4, 11], 5948: [2], 5983: [3], 6151: [13], 6191: [4], 6277: [8], 6309: [0, 9], 6312:
            [4], 6329: [5], 6363: [15], 6430: [3], 6447: [3], 6458: [2], 6565: [3], 6584: [4], 6627: [16], 6641:
            [9], 6719: [9], 6745: [3], 6763: [21], 6828: [11], 6872: [13], 6873: [3], 6897: [43], 6899: [16], 6906: [6], 6908:
                 6985: [4], 7007: [3], 7041: [9], 7157: [11], 7166: [9], 7184: [19], 7220: [14], 7238: [0], 7289: [1], 7477: [1
           9], 7496: [12], 7519: [18], 7524: [4], 7554: [8], 7560: [1], 7650: [1], 7721: [0], 7782: [20], 7828: [7], 7838: [1], 7961: [0], 8017: [0], 8036: [8], 8255: [14], 8434: [16], 8519: [5], 8615: [41], 8634: [22], 8853: [0], 9027:
           [4], 9068: [4], 9322: [22, 29], 9335: [3], 9355: [3], 9395: [3], 9399: [3], 9456: [8], 9457: [8], 9556: [2], 9650: [5], 9672: [0], 9688: [4], 9728: [3], 9763: [15], 9774: [15, 47], 9883: [13], 9981: [7], 10040: [11], 10064: [1, 4,
           7], 10094: [3], 10214: [3], 10224: [8], 10230: [21, 24], 10381: [10], 10382: [3], 10383: [3], 10423: [23], 10433:
            [13], 10454: [12], 10467: [4], 10471: [1], 10485: [2], 10603: [12], 10617: [2], 10699: [6], 10703: [2], 10819: [2
 In [10]: inv index = {}
            for pos, term in enumerate(data['stemmed_words']):
                  word in term:
                 if word in inv_index:
                   if pos not in inv_index[word]:
                      inv index[word].append(pos)
                 else:
                    inv_index[word]=[]
                    inv index[word].append(pos)
            for i in inv index:
              print(i,":",inv_index[i], end='\n\n')
                8028, 8034, 8047, 8056, 8085, 8094, 8097, 8121, 8127,
             4, 8295, 8299, 8329, 8330, 8335, 8346, 8361, 8382, 8386, 8398, 8401, 8402, 8418, 8422, 8423, 8427, 8432, 8445, 846
            2, 8467, 8469, 8473, 8477, 8482, 8485, 8496, 8498, 8502, 8521, 8526, 8528, 8532, 8536, 8537, 8548, 8552,
                                                                                                                                         8572. 858
            3. 8586. 8587. 8591. 8611. 8624. 8626. 8628. 8633. 8648. 8649. 8652. 8655. 8660. 8666. 8675. 8678. 8683.
                                                                                                                                         8690. 869
            1, 8696, 8729, 8730, 8732, 8733, 8735, 8739, 8740, 8759, 8767, 8775, 8779, 8804, 8810, 8814, 8819, 8820, 8830, 883
            1, 8845, 8849, 8858, 8860, 8866, 8868, 8872, 8873, 8884, 8886, 8914, 8947, 8963, 8969, 8974, 8980, 8984, 8986, 902
             6, 9028, 9030, 9034, 9036, 9038, 9049, 9063, 9070, 9093, 9103, 9114, 9116, 9117, 9134, 9142, 9143, 9154, 9158,
                                                                                                                                                917
                      9179,
             4, 9177,
                              9185, 9194, 9221, 9223,
                                                           9233,
                                                                  9235, 9236, 9260,
                                                                                       9263, 9264, 9274, 9294, 9301, 9302, 9307, 9337,
                                                                                                                                                935
            8, 9366, 9390, 9394, 9423, 9430, 9441, 9447, 9449, 9450, 9468, 9478, 9479, 9493, 9494, 9503, 9506, 9518, 9531, 954
            0, 9547, 9551, 9552, 9555, 9556, 9565, 9566, 9569, 9570, 9579, 9581, 9588, 9595, 9596, 9609, 9619, 9631, 9638, 964
            3, 9646, 9660, 9674, 9694, 9696, 9697, 9698, 9701, 9704, 9705, 9706, 9707, 9709, 9713, 9722, 9738, 9757, 9774, 979
            0, 9811, 9813, 9814, 9834, 9840, 9849, 9856, 9859, 9865, 9866, 9868, 9907, 9911, 9912, 9920, 9923, 9926, 9928, 994
            1, 9943, 9955, 9957, 9977, 9983, 10006, 10012, 10013, 10033, 10035, 10038, 10065, 10066, 10080, 10081, 10092, 1009
             6, 10098, 10105, 10131, 10132, 10168, 10170, 10177, 10178, 10213, 10214, 10215, 10226, 10236, 10238, 10239, 10256,
            10261, 10269, 10288, 10298, 10301, 10305, 10314, 10316, 10323, 10324, 10326, 10337, 10343, 10358, 10373, 10374, 103
            76, 10378, 10387, 10388, 10391, 10392, 10399, 10404, 10408, 10423, 10424, 10428, 10433, 10438, 10449, 10459, 10476,
```

10480, 10491, 10497, 10500, 10502, 10505, 10516, 10519, 10523, 10526, 10531, 10535, 10537, 10543, 10550, 10551, 10560, 10561, 10570, 10572, 10585, 10601, 10626, 10633, 10634, 10648, 10652, 10678, 10680, 10689, 10698, 10714, 10719, 10720, 10728, 10765, 10767, 10778, 10782, 10793, 10822, 10848, 10850, 10857, 10859, 10862, 10866, 10873, 10876, 108

```
In [12]: query=input().lower().split(" ")
    start time=time.time()
    inv_results=[]
           if(len(query)>2):
               print("Invalid query")
           else:
                try:
    inv_results=[value for value in inv_index[query[0]] if value in inv_index[query[1]]]
                except:
                     print ("No matches found")
                     inv_results=set(inv_results)
                     results=[]
for res in inv_results:
    word1_poslist=pos_index[query[0]][1][res]
    word2_poslist=pos_index[query[1]][1][res]
                          for i in word1_poslist:
                               if (i+1 in word2_poslist):
                                    results.append(res)
break
                     for i in results:
                          print(data['Title'][i]+'\n')
                 finally:
                      print("Query Execution time =", time.time()-start_time)
```

#### Outputs:

#### 1. Mixture of upper case and lower case letters in query

print("Query Execution time =", time.time()-start time)

No matter

No matter who someone is, how they look like, what language they speak, what they wear, remember the human. For the sake of humanity, the working class can and must unite across all arbitrary boundaries.

Dems and corporate media keep saying "Vote blue no matter who" but that only applies until a socialist wins the democ ratic primary, in which case they introduce an "independent" establishment candidate to make sure the blue doesn't win.

This is the result of "Vote Blue no Matter Who".

Reminder: No matter what leftists worldwide disagree on, these are the 10 fundamental demands ALL socialists should a lways agree on

"no matter which way elections go the pro-labour sentiment will not be reflected among the Democrats and Republicans. This gives a glimpse, however, of the potential for a militant labour movement to build independent working-class political representation"

blue no matter who, amirite?!

Query Execution time = 0.0020034313201904297

### 2. All small case letters in query

look like

No matter who someone is, how they look like, what language they speak, what they wear, remember the huma n. For the sake of humanity, the working class can and must unite across all arbitrary boundaries.

What Commitment to Diversifying the Federal Bench Looks Like — The president's success in diversifying our federal courts has gone underreported this year. It is our collective responsibility to champion this progress and to call for its continuation

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Looks like a comrade may be running the Kotaku Twitter account.

Bloomberg News Co-Founder Twists Numbers to Make MI Gov. Whitmer Look Like Economic Genius

How national elections will look like if you excluded votes from confederate states

What U.S. abortion access looks like, in graphics

Julie Bindel: The red light district of Amsterdam could soon be a distant memory | "What once looked like a revolutionary approach to prostitution in Holland is now clearly seen as a disaster, by all except those who seek to make a profit"

What Winning Looks Like

Rebekah Jones, fired Florida employee receives whistleblower status. Governor was lying about the COVID c ases in the entire state of FL to make himself look like he was doing an amazing job. Ms Jones was trashed in the media by her employer and had a police raid on her home with guns drawn.

What does American illiberal democracy look like?

Looks like I will be Boycotting Kelloggs products

This is what accountability looks like: Ex-Cop Kim Potter sentenced to 24 months in Daunte Wright killin g.

Looks like Former Trumpist Tom Brady got Fed Up with Trumps "Big Lie"

CNN Airs HUGE Explosion In the Sky Over Kyiv: 'It Looks Like a Large Aircraft'

What will an upcoming peak of the Red Scare look like and how concerned should we be?

YouGov - What do Americans think socialism looks like?

Trump and Mr Pillow Don't care about the 2020 election anymore. Maybe what they're trying to do is convin ce enough people that it will be okay to try and steal the 2024 election. Lindell's whole argument looks like nonsense to us. Maybe it's a diversion.

What Does Trans-Inclusive Abortion Advocacy Look Like?

Was reminded of the Movie "War Machine" and Will Poulter plays an Army Infantry soldier in Afghanistan. I thought shit, he looks like a kid and that 18 year olds are not that far from being a kid. So I want to k now, how many 18 year old Americans died in that bullshit war?

Query Execution time = 0.008003950119018555

# 3. Query length greater than 2

Invalid query

### Link to dataset:

https://www.kaggle.com/datasets/neelgajare/liberals-vs-conservatives-on-reddit-13000-posts