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#Import packages
         import json
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import fitz
         import re
         import string
         import datetime
         import math
         from glob import glob
         from zipfile import ZipFile
         import nltk
         from nltk.corpus import stopwords
         from nltk.tokenize import word_tokenize
         from nltk.stem import WordNetLemmatizer
         from wordcloud import WordCloud
         from gensim.models.word2vec import Word2Vec
         from sklearn.metrics.pairwise import cosine similarity
         from operator import itemgetter
In [ ]: def retrieve(n):
             zip file = ZipFile('pdf json.zip','r')
             random files = np.random.choice(zip file.namelist(), n)
             for names in zip file.namelist():
                 if names in random files:
                     zip file.extract(names)
             #Transform json files to dataframe
             json path = 'pdf json/*.json'
             files = glob(json path)
             data = []
             for file in files:
                 with open(file) as pdf:
                     json data = json.loads(pdf.read())
                     data.append(json data)
             df = pd.DataFrame(data)
             #Drop columns that will not be used
             df = df.drop(columns = ['bib entries', 'ref entries', 'back matter'])
             with open('metadata.csv', 'rb') as file:
                 metadata = pd.read_csv(file)
             #Drop useless columns
             metadata = metadata.drop(columns = ['cord uid', 'source x','pmcid','pubmed id', 'license','mag id', 'who covidence id', 'pmc json files', 'arxiv id','s2 id', 'pdf json files']
             #Joining ison dataframe and metatdata.csv
             df_merge = df.merge(metadata, how = 'left', left_on = 'paper_id', right_on = 'sha')
             #Remane abstract column
             df_merge = df_merge.rename(columns = {"abstract_y": "abstract"})
             #Merge all body text into one string and store in main_content column
             #Remove body_text afterwards
             for i, row in df_merge.iterrows():
                 try:
                     body_text = pd.DataFrame(df_merge.loc[i]['body_text'])
                     for text in body_text['text']:
                         content += text
                     df_merge.at[i, 'main_content'] = content
                 except IndexError:
                     df_merge.at[i, 'main_content'] = np.nan
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df_merge = df_merge.drop(columns = ['body_text'])
             #Set the country of first author with non-empty country as the country of the article is from
             #Remove metatdata afterwards
             for i, row in df_merge.iterrows():
                 try:
                     country name = df merge.loc[i]['metadata']['authors'][c]['affiliation']['location']['country']
                     while(country name == np.nan):
                         c += 1
                         country name = df merge.loc[i]['metadata']['authors'][c]['affiliation']['location']['country']
                     df merge.at[i, 'country'] = country name
                 except:
                     continue
             df merge = df merge.drop(columns = ['metadata'])
             #Replace null under abstract with empty string
             df_merge['abstract'].fillna("", inplace = True)
             #Remove duplicated columns to reduce dimension
             df merge = df merge.drop(columns = ['abstract x', 'sha'])
             df merge = df merge[~df merge['title'].isna()]
             df merge = df merge.reset index(drop = True)
             #Save cleaned dataframe
             df merge.to pickle('merged df')
In [ ]: def text process1(text):
             #lowercase all characters
             text = text.lower()
             #remove punctuation
             text = re.sub(r'[%s]' % re.escape(punc), ' ', text)
             #remove unicode text
             text = re.sub(r'[^\x00-\x7F]+', ' ', text)
             #remove the numbers
             text = re.sub(r'[0-9]', '', text)
             #remove double space
             text = re.sub(r'\setminus s\{2,\}', '', text)
             return text
         def text_process2(text):
             #Lowercase all characters
             text = text.lower()
             #remove punctuation
             text = re.sub(r'[%s]' % re.escape(string.punctuation), ' ', text)
             #remove unicode text
             text = re.sub(r'[^\x00-\x7F]+', ' ', text)
             #remove the numbers
             text = re.sub(r'[0-9]', '', text)
             #remove double space
             text = re.sub(r'\setminus s\{2,\}', '', text)
             return text
         #Edit the two lines below according to the comment to retrieve a new set of articles
         retrieve new = False #Set to True to retrieve a new set of articles
         n = 6000 #Set this value to number of articles to be retrieved
         if(retrieve_new):
             retrieve(n)
         #Load clean dataframe
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df_merge = pd.read_pickle('merged_df')
         #Check
         df_merge
        ###EDA before training model
In [ ]:
         #Plot top 10 count of articles by country
         df_merge['country'].value_counts().head(10).plot(kind = 'bar', figsize = (12, 7), title = 'Number of articles by country')
         plt.xticks(rotation = 45)
         plt.savefig('Number of articles by country.png')
In [ ]:
         #Plot number of articles published after 2003 in the sample
         df_plot = df_merge[["publish_time"]].copy()
         df_plot['publish_time'] = pd.to_datetime(df_plot['publish_time'])
         minimum = pd.to_datetime(datetime.date(2003, 1, 1))
         df = df_plot[df_plot['publish_time'] >= minimum]
         p = df['publish_time'].groupby(df['publish_time'].dt.to_period("M")).agg('count')
         #Plot top 10 count of articles by country
         p.plot(kind = 'line', figsize = (12, 7), title = 'Number of articles published after 2003')
         plt.xlabel('Year of publication')
         plt.savefig('Number of articles published after 2003.png')
In [ ]:
         ###Text preprocessing and normalization
         abstract_tokenized = []
         main_content_tokenized = []
         sentence_list = []
         #Load stopwords
         stop words = set(stopwords.words('english'))
         stop_words.add("et")
         stop words.add("al")
         stop_words.add("also")
         stop_words.add("may")
         punc = string.punctuation
         punc = punc.replace(".", "")
         #Read abstract
         for i, row in df_merge.iterrows():
             abstract_sent = []
             try:
                 text = df_merge.at[i, 'abstract']
                 splitted_text = text.split(sep = '.')
                 for sentence in splitted_text:
                     if(sentence != ''):
                         sentence = text process1(sentence)
                         #tokenize and Lemmatize the sentence
                         word tokens = word tokenize(sentence)
                         token_text = [w for w in word_tokens if not w.lower() in stop_words]
                         lemmatizer = WordNetLemmatizer()
                         lemmatized_text = []
                         for word in token text:
                             lemmatized_text.append(lemmatizer.lemmatize(word))
                         abstract_sent.append(lemmatized_text)
                         sentence_list.append(lemmatized_text)
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abstract_tokenized.append(abstract_sent)
             except:
                  continue
          #Read main content
         for i, row in df merge.iterrows():
             body sent = []
             try:
                 text = df_merge.at[i, 'main_content']
                  splitted text = text.split(sep = '.')
                  for sentence in splitted_text:
                     if(sentence != ''):
                          sentence = text_process1(sentence)
                          #tokenize and Lemmatize the file
                          word tokens = word tokenize(sentence)
                         token text = [w for w in word tokens if not w.lower() in stop words]
                          lemmatizer = WordNetLemmatizer()
                          lemmatized text = []
                         for word in token text:
                              lemmatized text.append(lemmatizer.lemmatize(word))
                         body_sent.append(lemmatized_text)
                          sentence list.append(lemmatized text)
                  main_content_tokenized.append(body_sent)
             except:
                  continue
         sentence list = [s \text{ for } s \text{ in sentence list if } len(s) > 0]
         #WordCLoud
In [ ]:
          freq = pd.Series(np.concatenate(sentence_list)).value_counts()
         wordcloud = WordCloud(
                      width = 300, height = 200,
                     background_color = "white",
                      max_words = 200,
                      max_font_size = 40,
                      stopwords = stop_words,
                      scale = 5,
                      random_state = 0).generate_from_frequencies(freq)
         fig = plt.figure(1, figsize = (9,6))
         plt.axis('off')
         plt.imshow(wordcloud)
         plt.savefig('WordCloud.png')
In [ ]:
         #Word2Vec
         model = Word2Vec(sentence_list, vector_size = 100, window = 8, min_count = 1, sg = 0, workers = 4)
          #Save model
         model.save("Word2Vec.model")
         #Load model
         model = Word2Vec.load("Word2Vec.model")
In [ ]: | ##Take and clean query
         #Read input
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text = input("Please enter your query: ")
        text = text_process2(text)
        #tokenize and Lemmatize the file
        word_tokens = word_tokenize(text)
        token_text = [w for w in word_tokens if not w.lower() in stop_words]
        lemmatizer = WordNetLemmatizer()
        query = []
        for word in token text:
            query.append(lemmatizer.lemmatize(word))
        #Vectorize query
In [ ]:
        query_vec = [0.0 for i in range(100)]
        num = 0
        for i in query:
            try:
                vec = model.wv[i]
            except:
                continue
            else:
                query_vec += vec
                num +=1
        if(num != 0):
            query_vec /= num
        sims = [] #contains top 20 of most similary vector and corresponding sentence index
In [ ]:
         count = 0
        for i in range(len(sentence_list)):
            sent vec = [0.0 for i in range(100)]
            num = 0
            for j in sentence list[i]:
               try:
                   vec = model.wv[j]
                except:
                   continue
                else:
                   sent vec += vec
                   num +=1
            if(num != 0):
                sent vec /= num
            similarity = cosine_similarity(np.expand_dims(query_vec, 0), np.expand_dims(sent_vec, 0))
                sims.append(tuple((i, similarity)))
                count += 1
            else:
                if(similarity > min(sims, key = itemgetter(1))[1]):
                   sims[sims.index(min(sims, key = itemgetter(1)))] = tuple((i, similarity))
        sims.sort(key = lambda x: x[1], reverse = True)
In [ ]:
        #Generate most relevant sentence with source
        article = []
        print("Showing top 10 relevant articles")
        print("-----")
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for n in range(len(sims)):
   if(len(article) == 10):
      break
   for i in range(len(abstract_tokenized)):
      if(i not in article):
          for j in range(len(abstract tokenized[i])):
             if(abstract_tokenized[i][j] == sentence_list[sims[n][0]]):
                 text = df merge.loc[i]['abstract']
                 splitted text = text.split(sep = '.')
                 location = text.find(splitted text[j])
                 e = len(text) - 1
                 if(location - 150 > 0):
                    s = location - 150
                 if(location + 450 < len(text)):</pre>
                    e = location + 450
                 while(text[e].isalpha()):
                    e += 1
                 print(df_merge.loc[i]['title'] + str(i))
                 print("\n")
                 while(text[s].isalpha()):
                    s += 1
                 print("... " + text[s:e] + " ...")
                 if(df merge.loc[i]['url'] != np.nan):
                     print("Original document at: " + df_merge.loc[i]['url'])
                 else:
                    print("Original document at: Not available")
                 article.append(i)
                 print("-----")
   for i in range(len(main content tokenized)):
      if(i not in article):
          for j in range(len(main_content_tokenized[i])):
             if(main_content_tokenized[i][j] == sentence_list[sims[n][0]]):
                 text = df merge.loc[i]['main content']
                 splitted text = text.split(sep = '.')
                 location = text.find(splitted text[j])
                 e = len(text) - 1
                 if(location - 150 > 0):
                    s = location - 150
                 if(location + 450 < len(text)):</pre>
                    e = location + 450
                 while(text[e].isalpha()):
                    e += 1
                 print(df merge.loc[i]['title'] + str(i))
                 print("\n")
                 while(text[s].isalpha()):
                    s += 1
                 print("... " + text[s:e] + " ...")
                 if(df merge.loc[i]['url'] != np.nan):
                    print("Original document at: " + df_merge.loc[i]['url'])
                 else:
                    print("Original document at: Not available")
                 article.append(i)
                 print("-----")
```