@author: HN import codecs import math import re import os import time #import string import pandas as pd import numpy as np from collections import Counter, defaultdict from nltk import FreqDist from nltk.corpus import stopwords #from spacy.lemmatizer import Lemmatizer as lemma #from spacy.lang.en import LEMMA_INDEX, LEMMA_EXC, LEMMA_RULES #from nltk.tokenize import word_tokenize start = time.time(); script_dir = os.path.dirname(os.path.realpath('__file__')) #<-- absolute dir the script is in</pre> rel_path = "02_data/"; abs_file_path = os.path.join(script_dir, rel_path) Files = ["wallstreet.csv", "winter.csv", "christmas.csv", "britishcricket.csv", "scubadive.csv", "brexit.csv"]; out = ["allwallstreet.txt", "allwinter.txt", "allchristmas.txt", "allbritishcricket.txt", "allscubadive.txt", "allbrexit.txt"]; ind = -1; cleaned_tweets = defaultdict(list); all_hashtags = defaultdict(list); stop_words = set(stopwords.words('english')); for file in Files: ind += 1 outfile = os.path.join(abs_file_path, out[ind]) with codecs.open(outfile, "w", "utf-8") as out_data: readfile = os.path.join(abs_file_path, file) #print(readfile) with open(readfile, "r", encoding = "utf8") as my_input_file: for line in my_input_file: # clean data $|\{|\}|[|\}| \in |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| = |\{|\}| =$ line = re.sub("http[a-zA-Z0-9]+", "", line) line = re.sub("pictwitter"," pictwitter", line); line = re.sub(" # "," #", line); line = re.sub("#"," #", line); line = re.sub("# ","", line); line = line.lower(); # remove stop words word_list = line.split(' '); filt_words = [] for word in word_list: if ((word not in stop_words) and (word != '' and len(word)>2) and (not word.__contains__('pictwitter'))): filt_words.append(re.sub(" "," ", word)); if (word[0] == '#') and (len(word)> 1) : all_hashtags[ind].append(word); filt_tweet = ' '.join(filt_words); cleaned_tweets[ind].append(filt_tweet) out_data.write(filt_tweet+'\n') print("Checkpoint 1 completed in ", time.time() - start); Checkpoint 1 completed in 6.970265626907349 In [117]: #Data Cleansing part 2 start = time.time(); top_hashtags = [] for tag in all_hashtags.keys(): hashtag = FreqDist(all_hashtags[tag]).most_common(1)[0][0] top_hashtags.append(hashtag) selectedtweets = ["wallstreet.txt", "winter.txt", "christmas.txt", "cricket.txt", "scubadive.txt", "brexit.txt"]; ind = -1; hashtag_file = os.path.join(abs_file_path, "classes.txt") #Top hashtags in each catagory with codecs.open(hashtag_file, "w", "utf-8") as hashfile: for tag in top_hashtags: hashfile.write(tag + '\n') for file in out: ind += 1; readfile = os.path.join(abs_file_path, out[ind]) with codecs.open(readfile, "r", "utf-8") as read file: writefile = os.path.join(abs_file_path, selectedtweets[ind]) with codecs.open(writefile, "w", "utf-8") as write_file: for line in read file: if any(s in line for s in top_hashtags): write_file.write(line) print("Checkpoint 2 completed in ", time.time() - start); Checkpoint 2 completed in 0.8426642417907715 In [118]: #Data segregation start = time.time(); tweets = selectedtweets; hashtags = top_hashtags; # Dict = {Hashtag: Tweets} and bag of words HT = dict.fromkeys(hashtags,''); trainHT = defaultdict(list); testHT = defaultdict(list); ind = -1; bow = set(); #hashtweets = defaultdict(list); for file in tweets: infile = os.path.join(abs_file_path, file) with codecs.open(infile, "r", "utf-8") as in_data: for line in in_data: for tag in hashtags: if tag in line: flag = np.random.choice([0, 1], 1, p=[0.75, 0.25]) **if** flag == 1: #Test data: Dict = {Hashtag: Tweets} line = " ".join(filter(lambda x:x[0]!='#', line.split())) testHT[tag].append(line) else: # Train set: Dict = {Hashtag: Tweets} trainHT[tag].append(re.sub('#' , '', line)) # {Hashtag: [T1, T2, T3]} HT[tag] += re.sub('#' , '', line) # {Hashtag: ['T1 T2 T3']} print("Checkpoint 3 completed in ", time.time() - start); Checkpoint 3 completed in 86.5642876625061 In [154]: #TF-IDF Section start = time.time(); #Bag of words #1 creation for txt in HT.values(): bow = set(bow).union(set(txt.split(' '))) #2 Bag of words initialization bowDict = {} for h in HT.keys(): bowDict[h] = dict.fromkeys(bow, 0) #3 Bow updation for each class(hashtag) for w in HT[h].split(' ') : bowDict[h][w] += 1#Compute Term-Frequency def computeTF(bowDict, tweet): tfDict = {} tweetlen = len(tweet.split(' ')) for word, count in bowDict.items(): tfDict[word] = count/float(tweetlen) return tfDict tf = {} for h, t in HT.items(): tf[h] = computeTF(bowDict[h], t) #Compute Inverse document frequency def computeIDF(docList): idfDict = {} N = len(docList)idfDict = HT.fromkeys(docList[0].keys(), 0) for doc in docList: #Each class(hashtag) for word, val in doc.items(): **if** val > 0: if word not in idfDict.keys(): idfDict[word] = 1 else : idfDict[word] += 1 for word, val in idfDict.items(): idfDict[word] = math.log10(N / float(val)) return idfDict idfs = computeIDF([bowDict[h] for h in bowDict.keys()]) #Compute TF-IDF def computeTFIDF(tfhashtxt, idfs): tfidf = {} for word, val in tfhashtxt.items(): tfidf[word] = val*idfs[word] return tfidf tfidfBow = {}; topfivewords = {} for hashtag in HT.keys(): tfidfBow[hashtag] = computeTFIDF(tf[hashtag], idfs) topfivewords[hashtag] = dict(Counter(tfidfBow[hashtag]).most_common(100)) tfidf = {} for hashtag, rankwords in topfivewords.items(): tfidf[hashtag] = rankwords.keys() print("Checkpoint 4 completed in ", time.time() - start); Checkpoint 4 completed in 1.6940264701843262 In [155]: #Vectors set start = time.time(); """To transform train and test dataset into input and output vectors for the multiclass-logistic regression.""" rank = tfidfpd; trainXY = pd.DataFrame.from_dict(trainHT, orient='index') testXY = pd.DataFrame.from dict(testHT, orient='index') bow = set()for i in range(0, len(rank)): rankwords = rank.iloc[i] bow = set(bow).union(set(rankwords)) #Train data X Y start_time = time.time() X = list(bow); Y = list(rank.index.get values()); r, c = trainXY.shape; train = {}; train[tuple([0]*len(X))] = 0; start_time = time.time() for i in range(r): outY = ifor j in range(c): if trainXY.iloc[i, j] != None : inX = [0]*len(X)for feature in X: if feature in trainXY.iloc[i, j]: inX[X.index(feature)] = 1 if tuple(inX) not in train: train[tuple(inX)] = outY else : break trainpd = pd.DataFrame.from_dict(train, orient='index') elapsed_time = time.time() - start_time print('Train data created',elapsed_time) #Test data X Y X = list(bow); Y = list(rank.index.get values()); r, c = testXY.shape; test = {}; test[tuple([0]*len(X))] = 0; start_time = time.time() for i in range(r): outY = ifor j in range(c): if testXY.iloc[i, j] != None : inX = [0]*len(X)for feature in X: if feature in testXY.iloc[i, j]: inX[X.index(feature)] = 1 if tuple(inX) not in test: test[tuple(inX)] = outY else : break testpd = pd.DataFrame.from_dict(test, orient='index') elapsed_time = time.time() - start_time print('Test data created',elapsed_time) print("Checkpoint 5 completed in ", time.time() - start); print("Vector set created --> Success!!!"); Train data created 112.39547657966614 Test data created 37.8052933216095 Checkpoint 5 completed in 151.3086199760437 Vector set created --> Success!!! In [156]: import numpy as np import pickle import matplotlib.pyplot as plt from sklearn.linear_model import LogisticRegression from sklearn.metrics import confusion_matrix from sklearn.metrics import accuracy score from sklearn.metrics import precision recall fscore support as score #Logistic regression start = time.time(); #Train data train = trainpd; trainX = np.array(list(train.index.get_values())) trainY = [] for j in range(len(train)): trainY.append(list(train.iloc[j])[0]) trainY = np.array(trainY) #Test data test = testpd; testX = np.array(list(test.index.get values())) testY = [] for j in range(len(test)): testY.append(list(test.iloc[j])[0]) testY = np.array(testY) #Logistic regression. logreg = LogisticRegression(C=1e1, solver='lbfgs', multi_class='multinomial', max_iter = 3000) # Create an instance of Logistic Regression Classifier and fit the data. logreg.fit(trainX, trainY) PredY = logreg.predict(testX) accuracy = accuracy_score(testY, PredY) print('Accuracy: ', accuracy) #print(confusion matrix(testY, PredY)) #precision, recall, fscore, support = score(testY, PredY) #print('precision: {}'.format(precision)) #print('recall: {}'.format(recall)) #print('fscore: {}'.format(fscore)) #print('support: {}'.format(support)) print("Checkpoint 6 completed in ", time.time() - start); print("Task complete -- > Success!!!"); [[401 150 41 25 2 16] [7 113 13 6 1 0] [0 18 62 2 2 0] 8 114 12 554 14 4] [1 15 8 5 105 0] [27 143 60 3 3 392]] Accuracy: 0.6991834980661796 precision: [0.90315315 0.20433996 0.31632653 0.93109244 0.82677165 0.95145631] recall: [0.63149606 0.80714286 0.73809524 0.78470255 0.78358209 0.62420382] fscore: [0.74328082 0.32611833 0.44285714 0.85165257 0.8045977 0.75384615] support: [635 140 84 706 134 628] Checkpoint 6 completed in 10.545912265777588 Task complete -- > Success!!! In [161]: import matplotlib.pyplot as plt Featuresize = [5, 25, 50, 100] Accuracy = [44.11, 69.32, 69.45, 69.91]plt.plot(Featuresize, Accuracy, 'ro') plt.axis([0, 120, 40, 100]) plt.grid(True) plt.title('Num of features vs Accuracy for 3000 iterations with 0.1 regularization.') plt.xlabel('Feature size') plt.ylabel('Accuracy in %') plt.savefig('FvsA.png', bbox_inches='tight') Num of features vs Accuracy for 3000 iterations with 0.1 regularization. 90 Ξ. 70 50 40 100 120 Feature size In [166]: | # Confusion matrix mat = confusion_matrix(testY, PredY) pd.DataFrame(mat, columns=["wallstreet","winter", "christmas", "cricket", "scubadive", "brexit"], index = ["wallstreet","winter r", "christmas", "cricket", "scubadive", "brexit"]) Out[166]: wallstreet winter christmas cricket scubadive brexit 150 25 2 16 wallstreet 401 41 113 winter 13 18 christmas 114 554 cricket 12 14 scubadive 15 105 27 143 60 3 3 392 brexit In [192]: # Performance metric score = [list(precision), list(recall), list(fscore), list(support)]; rows = ['precision', 'recall', 'fscore', 'support']; pd.DataFrame(score,columns=["wallstreet","winter", "christmas", "cricket", "scubadive", "brexit"], index = ['precision', 'recal 1', 'fscore', 'support']) Out[192]: wallstreet winter christmas cricket scubadive brexit 0.903153 0.204340 0.316327 0.931092 0.826772 0.951456 precision

0.631496

0.743281

#print(tfidfpd) # <--uncomment</pre>

recall

fscore

In []: #TF-IDF top rank words

0.807143

0.326118

0.738095

0.442857

support 635.000000 140.000000 84.000000 706.000000 134.000000 628.000000

0.784703

0.851653

0.783582

0.804598

0.624204

0.753846

In [116]: #Data Cleansing part 1

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