```
import warnings
warnings.filterwarnings("ignore")
import pandas as pd
import sqlite3
import csv
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
from wordcloud import WordCloud
import re
import os
from sqlalchemy import create engine # database connection
import datetime as dt
from nltk.corpus import stopwords
from nltk.tokenize import word tokenize
from nltk.stem.snowball import SnowballStemmer
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.multiclass import OneVsRestClassifier
from sklearn.linear_model import SGDClassifier
from sklearn import metrics
from sklearn.metrics import f1_score,precision_score,recall_score
from sklearn import svm
from sklearn.linear model import LogisticRegression
from skmultilearn.adapt import mlknn
from skmultilearn.problem_transform import ClassifierChain
from skmultilearn.problem_transform import BinaryRelevance
from skmultilearn.problem_transform import LabelPowerset
from sklearn.naive_bayes import GaussianNB
from datetime import datetime
```

Stack Overflow: Tag Prediction

1.1 Description

Description

Stack Overflow is the largest, most trusted online community for developers to learn, share their programming knowledge, and build their careers.

Stack Overflow is something which every programmer use one way or another. Each month, over 50 million developers come to Stack Overflow to learn, share their knowledge, and build their careers. It features questions and answers on a wide range of topics in computer programming. The website serves as a platform for users to ask and answer questions, and, through membership and active participation, to vote questions and answers up or down and edit questions and answers in a fashion similar to a wiki or Digg. As of April 2014 Stack Overflow has over 4,000,000 registered users, and it exceeded 10,000,000 questions in late August 2015. Based on the type of tags assigned to questions, the top eight most discussed topics on the site are: Java, JavaScript, C#, PHP, Android, jQuery, Python and HTML.

Problem Statemtent

Suggest the tags based on the content that was there in the question posted on Stackoverflow.

Source: https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/

1.2 Source / useful links

Data Source: https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/data (https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/data)

Youtube: https://youtu.be/nNDqbUhtlRg (https://youtu.be/nNDqbUhtlRg)

Research paper: https://www.microsoft.com/en-us/research/wp-content/uploads/2016/02/tagging-1.pdf (<a href="https://www.microsoft.com/en-us/research/wp-content/

Research paper: https://dl.acm.org/citation.cfm?id=2660970&dl=ACM&coll=DL (https://dl.acm.org/citation.cfm?id=2660970&dl=ACM&coll=DL)

2. Machine Learning problem

2.1 Data

2.1.1 Data Overview

Refer: https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/data (https://www.kaggle.com/c

```
Train.csv contains 4 columns: Id,Title,Body,Tags.

Test.csv contains the same columns but without the Tags, which you are to predict.
Size of Train.csv - 6.75GB
```

Number of rows in Train.csv = 6034195

Size of Test.csv - 2GB

The questions are randomized and contains a mix of verbose text sites as well as sites related to math and programming. The number of questions from each site may vary, and no filtering has been performed on the questions (such as closed questions).

Data Field Explaination

Dataset contains 6,034,195 rows. The columns in the table are:

 ${\bf Id}\,$ - Unique identifier for each question

Title - The question's title

Body - The body of the question

Tags - The tags associated with the question in a space-seperated format (all lowercase, should not contain tabs '\t' or ampersands '&')

2.1.2 Example Data point

Title: Implementing Boundary Value Analysis of Software Testing in a C++ program?

Body :

```
#include<
iostream>\n
#include<
stdlib.h>\n\n
using namespace std;\n\n
int main()\n
{\n
         int n,a[n],x,c,u[n],m[n],e[n][4];\n
         cout<<"Enter the number of variables";\n</pre>
                                                             cin>>n;\n\n
         cout<<"Enter the Lower, and Upper Limits of the variables";\n</pre>
         for(int y=1; y<n+1; y++)\n
         \{\n
             cin>>m[y];\n
            cin>>u[y];\n
         }\n
         for(x=1; x<n+1; x++)\n
         {\n
            a[x] = (m[x] + u[x])/2; \n
         }\n
         c=(n*4)-4;\n
         for(int a1=1; a1<n+1; a1++)\n</pre>
         {n n}
            e[a1][0] = m[a1]; \n
             e[a1][1] = m[a1]+1; \n
            e[a1][2] = u[a1]-1; \n
            e[a1][3] = u[a1]; \n
         }\n
         for(int i=1; i<n+1; i++)\n</pre>
         {\n
            for(int l=1; l<=i; l++)\n
                 if(l!=1)\n
                 {\n
                     cout<<a[l]<<"\\t";\n
                 }\n
            }\n
             for(int j=0; j<4; j++)\n
             \{\n
                 cout<<e[i][j];\n</pre>
                 for(int k=0; k< n-(i+1); k++) n
                     cout<<a[k]<<"\\t";\n
                 }\n
                 cout<<"\\n";\n
            }\n
         }
              n\n
         system("PAUSE");\n
         return 0;
}\n
```

\n\n

The answer should come in the form of a table like $\n\$

```
1
              50
                                50\n
2
                                50\n
              50
99
              50
                                50\n
100
              50
                                50\n
50
              1
                                50\n
50
              2
                                50\n
50
              99
                                50\n
50
              100
                                50\n
50
              50
                                1\n
50
              50
                                2\n
50
              50
                                99\n
50
              50
                                100\n
```

 $n\n$

2.2 Mapping the real-world problem to a Machine Learning Problem

2.2.1 Type of Machine Learning Problem

It is a multi-label classification problem

Tags : 'c++ c'

Multi-label Classification: Multilabel classification assigns to each sample a set of target labels. This can be thought as predicting properties of a datapoint that are not mutually exclusive, such as topics that are relevant for a document. A question on Stackoverflow might be about any of C, Pointers, FileIO and/or memory-management at the same time or none of these.

__Credit__: http://scikit-learn.org/stable/modules/multiclass.html

2.2.2 Performance metric

Micro-Averaged F1-Score (Mean F Score): The F1 score can be interpreted as a weighted average of the precision and recall, where an F1 score reaches its best value at 1 and worst score at 0. The relative contribution of precision and recall to the F1 score are equal. The formula for the F1 score is:

```
F1 = 2 * (precision * recall) / (precision + recall)
```

In the multi-class and multi-label case, this is the weighted average of the F1 score of each class.

'Micro f1 score':

Calculate metrics globally by counting the total true positives, false negatives and false positives. This is a better metric when we have class imbalance.

'Macro f1 score':

Calculate metrics for each label, and find their unweighted mean. This does not take label imbalance into account.

https://www.kaggle.com/wiki/MeanFScore (https://www.kaggle.com/wiki/MeanFScore)
http://scikit-learn.org/stable/modules/generated/sklearn.metrics.f1_score.html (http://scikit-learn.org/stable/modules/generated/sklearn.metrics.f1 score.html)

Hamming loss: The Hamming loss is the fraction of labels that are incorrectly predicted. https://www.kaggle.com/wiki/HammingLoss (https://www.kaggle.com/wiki/HammingLoss)

3. Exploratory Data Analysis

3.1 Data Loading and Cleaning

3.1.1 Using Pandas with SQLite to Load the data

```
In [ ]:
```

```
#Creating db file from csv
if not os.path.isfile('train.db'):
    start = datetime.now()
   disk engine = create engine('sqlite:///train.db')
   start = dt.datetime.now()
   chunksize = 180000
    j = 0
    index start = 1
    for df in pd.read csv('Train.csv', names=['Id', 'Title', 'Body', 'Tags'], chunksize=chunksize, iterator=True,
encoding='utf-8', ):
        df.index += index_start
        i += 1
        print('{} rows'.format(j*chunksize))
        df.to sql('data', disk_engine, if_exists='append')
        index start = df.index[-1] + 1
   print("Time taken to run this cell :", datetime.now() - start)
```

3.1.2 Counting the number of rows

In []:

```
if os.path.isfile('train.db'):
    start = datetime.now()
    con = sqlite3.connect('train.db')
    num_rows = pd.read_sql_query("""SELECT count(*) FROM data""", con)
    #Always remember to close the database
    print("Number of rows in the database :","\n",num_rows['count(*)'].values[0])
    con.close()
    print("Time taken to count the number of rows :", datetime.now() - start)
else:
    print("Please download the train.db file from drive or run the above cell to genarate train.db file")
```

```
Number of rows in the database : 6034196 Time taken to count the number of rows : 0:01:15.750352
```

3.1.3 Checking for duplicates

In []:

```
if os.path.isfile('train.db'):
    start = datetime.now()
    con = sqlite3.connect('train.db')
    df_no_dup = pd.read_sql_query('SELECT Title, Body, Tags, COUNT(*) as cnt_dup FROM data GROUP BY Title, Body,
Tags', con)
    con.close()
    print("Time taken to run this cell :", datetime.now() - start)
else:
    print("Please download the train.db file from drive or run the first to genarate train.db file")
```

Time taken to run this cell : 0:04:33.560122

In []:

```
df_no_dup.head()
# we can observe that there are duplicates
```

Out[]:

cnt_dup	Tags	Body	Title	
1	C++ C	<pre><code>#include<iostream>\n#include&</code></pre>	Implementing Boundary Value Analysis of S	0
1	c# silverlight data-binding	I should do binding for datagrid dynamicall	Dynamic Datagrid Binding in Silverlight?	1
1	c# silverlight data-binding columns	I should do binding for datagrid dynamicall	Dynamic Datagrid Binding in Silverlight?	2
1	jsp jstl	I followed the guide in		

```
In [ ]:
print("number of duplicate questions :", num_rows['count(*)'].values[0]- df_no_dup.shape[0], "(",(1-((df_no_dup.s
hape[0])/(num_rows['count(*)'].values[0])))*100,"%)")
number of duplicate questions : 1827881 ( 30.2920389063 \% )
In [ ]:
# number of times each question appeared in our database
df no dup.cnt dup.value counts()
Out[]:
      2656284
1
2
      1272336
3
       277575
4
           90
5
           25
Name: cnt_dup, dtype: int64
In [ ]:
start = datetime.now()
df_no_dup["tag_count"] = df_no_dup["Tags"].apply(lambda text: len(text.split(" ")))
# adding a new feature number of tags per question
print("Time taken to run this cell :", datetime.now() - start)
df no dup.head()
Time taken to run this cell: 0:00:03.169523
Out[]:
                                Title
                                                                     Body
                                                                                              Tags cnt_dup tag_count
0
     Implementing Boundary Value Analysis of
                                                                     <
                                                                                                                   2
                                                                                              C++ C
                                        <code>#include&lt;iostream&gt;\n#include&...
      Dynamic Datagrid Binding in Silverlight?
                                                                                                                   3
 1
                                      I should do binding for datagrid dynamicall...
                                                                               c# silverlight data-binding
                                                                                                         1
2
                                                                               c# silverlight data-binding
      Dynamic Datagrid Binding in Silverlight?
                                     I should do binding for datagrid dynamicall...
                                                                                            columns
 3
          java.lang.NoClassDefFoundError:
                                       I followed the guide in <a href="http://sta...</p>
                                                                                             jsp jstl
                                                                                                         1
                                                                                                                   2
     java.sql.SQLException:[Microsoft][ODBC
                                          I use the following code\n\n
                                                                                                         2
                                                                                                                   2
                                                                                           java jdbc
                                                                  <code>...
In [ ]:
# distribution of number of tags per question
df no dup.tag count.value counts()
Out[]:
3
      1206157
2
      1111706
4
       814996
1
       568298
       505158
Name: tag_count, dtype: int64
In [ ]:
#Creating a new database with no duplicates
if not os.path.isfile('train_no_dup.db'):
```

disk dup = create engine("sqlite:///train no dup.db")

no_dup.to_sql('no_dup_train',disk_dup)

no_dup = pd.DataFrame(df_no_dup, columns=['Title', 'Body', 'Tags'])

```
In [ ]:
```

```
#creating the connection with database file.
if os.path.isfile('train_no_dup.db'):
    start = datetime.now()
    con = sqlite3.connect('train_no_dup.db')
    tag_data = pd.read_sql_query("""SELECT Tags FROM no_dup_train""", con)
    con.close()

# Let's now drop unwanted column.
    tag_data.drop(tag_data.index[0], inplace=True)
    #Printing first 5 columns from our data frame
    tag_data.head()
    print("Time taken to run this cell :", datetime.now() - start)
else:
    print("Please download the train.db file from drive or run the above cells to genarate train.db file")
```

Time taken to run this cell: 0:00:52.992676

3.2 Analysis of Tags

3.2.1 Total number of unique tags

```
In [ ]:
```

```
# Importing & Initializing the "CountVectorizer" object, which
#is scikit-learn's bag of words tool.

#by default 'split()' will tokenize each tag using space.
vectorizer = CountVectorizer(tokenizer = lambda x: x.split())
# fit_transform() does two functions: First, it fits the model
# and learns the vocabulary; second, it transforms our training data
# into feature vectors. The input to fit_transform should be a list of strings.
tag_dtm = vectorizer.fit_transform(tag_data['Tags'])
```

In []:

```
print("Number of unique tags :", tag_dtm.shape[1])

Number of data points : 4206314
Number of unique tags : 42048

In []:

#'get_feature_name()' gives us the vocabulary.
tags = vectorizer.get_feature_names()
#Lets look at the tags we have.
```

```
Some of the tages we have : ['.a', '.app', '.asp.net-mvc', '.aspxauth', '.bash-profile', '.class-file', '.cs-file', '.doc', '.drv', '.ds-store']
```

3.2.3 Number of times a tag appeared

print("Some of the tags we have :", tags[:10])

print("Number of data points :", tag_dtm.shape[0])

```
# https://stackoverflow.com/questions/15115765/how-to-access-sparse-matrix-elements
#Lets now store the document term matrix in a dictionary.
freqs = tag_dtm.sum(axis=0).A1
result = dict(zip(tags, freqs))
```

```
#Saving this dictionary to csv files.
if not os.path.isfile('tag_counts_dict_dtm.csv'):
    with open('tag_counts_dict_dtm.csv', 'w') as csv_file:
        writer = csv.writer(csv_file)
        for key, value in result.items():
            writer.writerow([key, value])
tag_df = pd.read_csv("tag_counts_dict_dtm.csv", names=['Tags', 'Counts'])
tag_df.head()
```

Out[]:

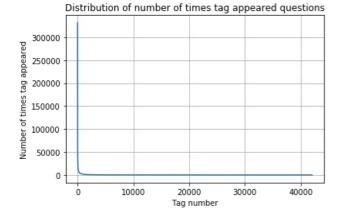
	Tags	Counts
0	.a	18
1	.арр	37
2	.asp.net-mvc	1
3	.aspxauth	21
4	.bash-profile	138

In []:

```
tag_df_sorted = tag_df.sort_values(['Counts'], ascending=False)
tag_counts = tag_df_sorted['Counts'].values
```

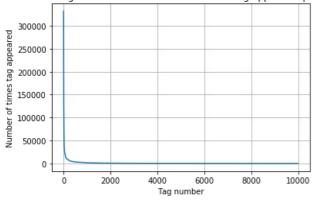
In []:

```
plt.plot(tag_counts)
plt.title("Distribution of number of times tag appeared questions")
plt.grid()
plt.xlabel("Tag number")
plt.ylabel("Number of times tag appeared")
plt.show()
```



```
plt.plot(tag_counts[0:10000])
plt.title('first 10k tags: Distribution of number of times tag appeared questions')
plt.grid()
plt.xlabel("Tag number")
plt.ylabel("Number of times tag appeared")
plt.show()
print(len(tag_counts[0:10000:25]), tag_counts[0:10000:25])
```

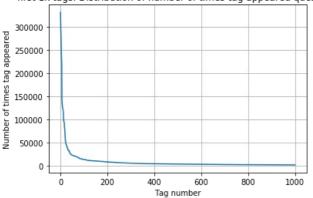
first 10k tags: Distribution of number of times tag appeared questions



400 [3315	EOE 440	329 224	120 17	728 133	364 11	162 100	20 0	148 8	3054 7151	
6466	5865	5370	4983	4526	4281	4144	3929	3750	3593	L
3453	3299	3123	2989	2891	2738	2647	2527	2431	2331	
2259	2186	2097	2020	1959	1900	1828	1770	1723	1673	
1631	1574	1532	1479	1448	1406	1365	1328	1300	1266	
1245	1222	1197	1181	1158	1139	1121	1101	1076	1056	
1038	1023	1006	983	966	952	938	926	911	891	
882	869	856	841	830	816	804	789	779	770	
752	743	733	725	712	702	688	678	671	658	
650	643	634	627	616	607	598	589	583	577	
568	559	552	545	540	533	526	518	512	506	
500	495	490	485	480	477	469	465	457	450	
447	442	437	432	426	422	418	413	408	403	
398	393	388	385	381	378	374	370	367	365	
361	357	354	350	347	344	342	339	336	332	
330	326	323	319	315	312	309	307	304	301	
299	296	293	291	289	286	284	281	278	276	
275	272	270	268	265	262	260	258	256	254	
252	250	249	247	245	243	241	239	238	236	
234	233	232	230	228	226	224	222	220	219	
217	215	214	212	210	209	207	205	204	203	
201	200	199	198	196	194	193	192	191	189	
188	186	185	183	182	181	180	179	178	177	
175	174	172	171	170	169	168	167	166	165	
164	162	161	160	159	158	157	156	156	155	
154	153	152	151	150	149	149	148	147	146	
145	144	143	142	142	141	140	139	138	137	
137	136	135	134	134	133	132	131	130	130	
129	128	128	127	126	126	125	124	124	123	
123	122	122	121	120	120	119	118	118	117	
117	116	116	115	115	114	113	113	112	111	
111	110	109	109	108	108	107	106	106	106	
105	105	104	104	103	103	102	102	101	101	
100	100	99	99	98	98	97	97	96	96	
95	95	94	94	93	93	93	92	92	91	
91	90	90	89	89	88	88	87	87	86	
86	86	85	85	84	84	83	83	83	82	
82	82	81	81	80	80	80	79	79	78	
78	78	78	77	77	76	76	76	75 73	75 731	
75	74	74	74	73	73	73	73	72	72]	

```
plt.plot(tag_counts[0:1000])
plt.title('first 1k tags: Distribution of number of times tag appeared questions')
plt.grid()
plt.xlabel("Tag number")
plt.ylabel("Number of times tag appeared")
plt.show()
print(len(tag_counts[0:1000:5]), tag_counts[0:1000:5])
```

first 1k tags: Distribution of number of times tag appeared questions



200 [331505 221533 122769 95160 62023 44829 37170 31897 26925 24537 1639]

```
plt.plot(tag_counts[0:500])
plt.title('first 500 tags: Distribution of number of times tag appeared questions')
plt.grid()
plt.xlabel("Tag number")
plt.ylabel("Number of times tag appeared")
plt.show()
print(len(tag_counts[0:500:5]), tag_counts[0:500:5])
```

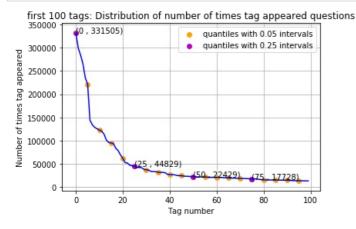
```
first 500 tags: Distribution of number of times tag appeared questions
  300000
ed
ed
  250000
tag
  200000
times
  150000
to
  100000
Number
   50000
         0
              Ò
                        100
                                    200
                                               300
                                                           400
                                                                       500
                                      Tag number
```

```
100 [331505 221533 122769 95160 62023 44829 37170 31897
                                                                   26925 24537
  22429
         21820
                 20957
                        19758
                                18905
                                        17728
                                               15533
                                                       15097
                                                               14884
                                                                      13703
  13364
                 12407
                                11228
                                                                      10224
         13157
                        11658
                                        11162
                                                10863
                                                       10600
                                                               10350
  10029
          9884
                  9719
                          9411
                                 9252
                                         9148
                                                 9040
                                                        8617
                                                                8361
                                                                       8163
   8054
          7867
                  7702
                          7564
                                 7274
                                         7151
                                                 7052
                                                        6847
                                                                6656
                                                                       6553
   6466
          6291
                  6183
                          6093
                                 5971
                                         5865
                                                 5760
                                                        5577
                                                                5490
                                                                       5411
   5370
          5283
                  5207
                          5107
                                 5066
                                         4983
                                                 4891
                                                        4785
                                                                4658
                                                                        4549
   4526
          4487
                  4429
                          4335
                                 4310
                                         4281
                                                 4239
                                                        4228
                                                                4195
                                                                        4159
   4144
          4088
                  4050
                          4002
                                  3957
                                         3929
                                                 3874
                                                        3849
                                                                3818
                                                                        3797
   3750
          3703
                  3685
                          3658
                                 3615
                                         3593
                                                 3564
                                                        3521
                                                                3505
                                                                        3483]
```

```
plt.plot(tag_counts[0:100], c='b')
plt.scatter(x=list(range(0,100,5)), y=tag_counts[0:100:5], c='orange', label="quantiles with 0.05 intervals")
# quantiles with 0.25 difference
plt.scatter(x=list(range(0,100,25)), y=tag_counts[0:100:25], c='m', label = "quantiles with 0.25 intervals")

for x,y in zip(list(range(0,100,25)), tag_counts[0:100:25]):
    plt.annotate(s="({} , {})".format(x,y), xy=(x,y), xytext=(x-0.05, y+500))

plt.title('first 100 tags: Distribution of number of times tag appeared questions')
plt.grid()
plt.xlabel("Tag number")
plt.ylabel("Number of times tag appeared")
plt.legend()
plt.show()
print(len(tag_counts[0:100:5]), tag_counts[0:100:5])
```



20 [331505 221533 122769 95160 62023 44829 37170 31897 26925 24537 22429 21820 20957 19758 18905 17728 15533 15097 14884 13703]

```
# Store tags greater than 10K in one list
lst_tags_gt_10k = tag_df[tag_df.Counts>10000].Tags
#Print the length of the list
print ('{} Tags are used more than 10000 times'.format(len(lst_tags_gt_10k)))
# Store tags greater than 100K in one list
lst_tags_gt_100k = tag_df[tag_df.Counts>100000].Tags
#Print the length of the list.
print ('{} Tags are used more than 100000 times'.format(len(lst_tags_gt_100k)))
```

153 Tags are used more than 10000 times 14 Tags are used more than 100000 times

Observations:

- 1. There are total 153 tags which are used more than 10000 times.
- 2. 14 tags are used more than 100000 times.
- 3. Most frequent tag (i.e. c#) is used 331505 times.
- 4. Since some tags occur much more frequenctly than others, Micro-averaged F1-score is the appropriate metric for this probelm.

3.2.4 Tags Per Question

In []:

```
#Storing the count of tag in each question in list 'tag_count'
tag_quest_count = tag_dtm.sum(axis=1).tolist()
#Converting list of lists into single list, we will get [[3], [4], [2], [2], [3]] and we are converting this to [
3, 4, 2, 2, 3]
tag_quest_count=[int(j) for i in tag_quest_count for j in i]
print ('We have total {} datapoints.'.format(len(tag_quest_count)))
print(tag_quest_count[:5])
```

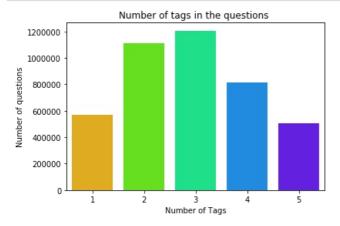
We have total 4206314 datapoints. [3, 4, 2, 2, 3]

In []:

```
print( "Maximum number of tags per question: %d"%max(tag_quest_count))
print( "Minimum number of tags per question: %d"%min(tag_quest_count))
print( "Avg. number of tags per question: %f"% ((sum(tag_quest_count)*1.0)/len(tag_quest_count)))
```

Maximum number of tags per question: 5 Minimum number of tags per question: 1 Avg. number of tags per question: 2.899440

```
sns.countplot(tag_quest_count, palette='gist_rainbow')
plt.title("Number of tags in the questions ")
plt.xlabel("Number of Tags")
plt.ylabel("Number of questions")
plt.show()
```



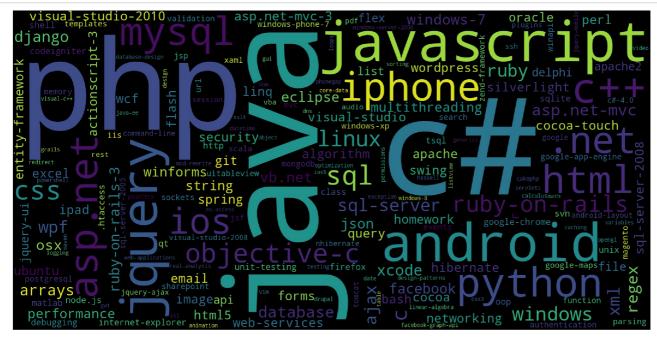
Observations:

- 1. Maximum number of tags per question: 5
- 2. Minimum number of tags per question: 1
- 3. Avg. number of tags per question: 2.899
- 4. Most of the questions are having 2 or 3 tags

3.2.5 Most Frequent Tags

In []:

```
# Ploting word cloud
start = datetime.now()
# Lets first convert the 'result' dictionary to 'list of tuples'
tup = dict(result.items())
#Initializing WordCloud using frequencies of tags.
                          background color='black',
wordcloud = WordCloud(
                          width=1600,
                          height=800,
                    ).generate_from_frequencies(tup)
fig = plt.figure(figsize=(30,20))
plt.imshow(wordcloud)
plt.axis('off')
plt.tight_layout(pad=0)
fig.savefig("tag.png")
plt.show()
print("Time taken to run this cell :", datetime.now() - start)
```



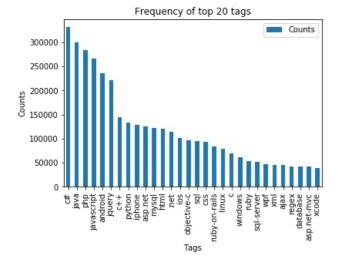
Time taken to run this cell: 0:00:05.470788

Observations:

A look at the word cloud shows that "c#", "java", "php", "asp.net", "javascript", "c++" are some of the most frequent tags.

3.2.6 The top 20 tags

```
i=np.arange(30)
tag_df_sorted.head(30).plot(kind='bar')
plt.title('Frequency of top 20 tags')
plt.xticks(i, tag_df_sorted['Tags'])
plt.xlabel('Tags')
plt.ylabel('Counts')
plt.show()
```



Observations:

- 1. Majority of the most frequent tags are programming language.
- 2. C# is the top most frequent programming language.
- 3. Android, IOS, Linux and windows are among the top most frequent operating systems.

3.3 Cleaning and preprocessing of Questions

3.3.1 Preprocessing

- 1. Sample 1M data points
- 2. Separate out code-snippets from Body
- 3. Remove Spcial characters from Question title and description (not in code)
- 4. Remove stop words (Except 'C')
- 5. Remove HTML Tags
- 6. Convert all the characters into small letters
- 7. Use SnowballStemmer to stem the words

```
def striphtml(data):
    cleanr = re.compile('<.*?>')
    cleantext = re.sub(cleanr, ' ', str(data))
    return cleantext
stop_words = set(stopwords.words('english'))
stemmer = SnowballStemmer("english")
```

```
In [ ]:
#http://www.sqlitetutorial.net/sqlite-python/create-tables/
def create connection(db file):
    """ create a database connection to the SQLite database
        specified by db_file
    :param db_file: database file
    :return: Connection object or None
        conn = sqlite3.connect(db file)
        return conn
    except Error as e:
        print(e)
    return None
def create_table(conn, create_table_sql):
     '"" create a table from the create_table_sql statement
    :param conn: Connection object
    :param create table sql: a CREATE TABLE statement
    :return:
    try:
        c = conn.cursor()
        c.execute(create table sql)
    except Error as e:
        print(e)
def checkTableExists(dbcon):
    cursr = dbcon.cursor()
    str = "select name from sqlite_master where type='table'"
    table_names = cursr.execute(str)
    print("Tables in the databse:")
    tables =table_names.fetchall()
    print(tables[0][0])
    return(len(tables))
def create_database_table(database, query):
    conn = create connection(database)
    if conn is not None:
        create table(conn, query)
        checkTableExists(conn)
        print("Error! cannot create the database connection.")
    conn.close()
sql create table = """CREATE TABLE IF NOT EXISTS QuestionsProcessed (question text NOT NULL, code text, tags te
xt, words_pre integer, words_post integer, is_code integer);""
create database table("Processed.db", sql create table)
Tables in the databse:
QuestionsProcessed
In [ ]:
# https://stackoverflow.com/questions/2279706/select-random-row-from-a-sqlite-table
start = datetime.now()
read db = 'train no dup.db
write_db = 'Processed.db'
if os.path.isfile(read db):
    conn_r = create_connection(read_db)
    if conn r is not None:
```

```
# https://stackoverflow.com/questions/2279706/select-random-row-from-a-sqlite-table
start = datetime.now()
read_db = 'train_no_dup.db'
write_db = 'Processed.db'
if os.path.isfile(read_db):
    conn_r = create_connection(read_db)
    if conn_r is not None:
        reader =conn_r.cursor()
        reader.execute("SELECT Title, Body, Tags From no_dup_train ORDER BY RANDOM() LIMIT 10000000;")

if os.path.isfile(write_db):
    conn_w = create_connection(write_db)
    if conn_w is not None:
        tables = checkTableExists(conn_w)
        writer =conn_w.cursor()
        if tables != 0:
              writer.execute("DELETE FROM QuestionsProcessed WHERE 1")
              print("Cleared All the rows")
print("Time taken to run this cell :", datetime.now() - start)
```

Time taken to run this cell: 0:06:32.806567

Tables in the databse: QuestionsProcessed Cleared All the rows

```
In [ ]:
```

conn w.close()

```
#http://www.bernzilla.com/2008/05/13/selecting-a-random-row-from-an-sqlite-table/
start = datetime.now()
preprocessed_data_list=[]
reader.fetchone()
questions_with_code=0
len pre=0
len post=0
questions proccesed = 0
for row in reader:
    is code = 0
    title, question, tags = row[0], row[1], row[2]
    if '<code>' in question:
        {\tt questions\_with\_code+=1}
        is code = 1
    x = len(question) + len(title)
    len pre+=x
    code = str(re.findall(r'<code>(.*?)</code>', question, flags=re.DOTALL))
    question=re.sub('<code>(.*?)</code>', '', question, flags=re.MULTILINE|re.DOTALL)
    question=striphtml(question.encode('utf-8'))
    title=title.encode('utf-8')
    question=str(title)+" "+str(question)
    question=re.sub(r'[^A-Za-z]+',' ',question)
    words=word tokenize(str(question.lower()))
    #Removing all single letter and and stopwords from question exceptt for the letter 'c'
    question=' '.join(str(stemmer.stem(j)) for j in words if j not in stop_words and (len(j)!=1 or j=='c'))
    len post+=len(question)
    tup = (question,code,tags,x,len(question),is code)
    questions proccesed += 1
    writer.execute("insert into QuestionsProcessed(question,code,tags,words pre,words post,is code) values (?,?,?
,?,?,?)",tup)
    if (questions proccesed%100000==0):
        print("number of questions completed=",questions proccesed)
no_dup_avg_len_pre=(len_pre*1.0)/questions_proccesed
no dup avg len post=(len post*1.0)/questions proccesed
print( "Avg. length of questions(Title+Body) before processing: %d"%no dup avg len pre)
print( "Avg. length of questions(Title+Body) after processing: %d"%no_dup_avg_len_post)
print ("Percent of questions containing code: %d"%((questions with code*100.0)/questions proccesed))
print("Time taken to run this cell :", datetime.now() - start)
number of questions completed= 100000
number of questions completed= 200000
number of questions completed= 300000
number of questions completed= 400000
number of questions completed= 500000
number of questions completed= 600000
number of questions completed= 700000
number of questions completed= 800000
number of questions completed= 900000
Avg. length of questions(Title+Body) before processing: 1169
Avg. length of questions(Title+Body) after processing: 327
Percent of questions containing code: 57
Time taken to run this cell: 0:47:05.946582
In [ ]:
conn_r.commit()
conn w.commit()
conn_r.close()
```

```
In [ ]:
```

Questions after preprocessed

('ef code first defin one mani relationship differ key troubl defin one zero mani relationship entit i ef object model look like use fluent api object composit pk defin batch id batch detail id use flu ent api object composit pk defin batch detail id compani id map exist databas tpt basic idea submitt edtransact zero mani submittedsplittransact associ navig realli need one way submittedtransact submittedsplittransact need dbcontext class onmodelcr overrid map class lazi load occur submittedtransact submittedsplittransact help would much appreci edit taken advic made follow chang dbcontext class ad follow onmodelcr overrid must miss someth get follow except thrown submittedtransact key batch id batch detail id zero one mani submittedsplittransact key batch detail id compani id rather assum convent creat relationship two object configur requir sinc obvious wrong',)

('explan new statement review section c code came accross statement block come accross new oper use way someon explain new call way',)

('error function notat function solv logic riddl iloczyni list structur list possibl candid solut li

st possibl coordin matrix wan na choos one candid compar possibl candid element equal wan na delet c oordin call function skasuj look like ni knowledg haskel cant see what wrong',)

('step plan move one isp anoth one work busi plan switch isp realli soon need chang lot inform dns w an wan wifi question guy help mayb peopl plan correct chang current isp new one first dns know recei v new ip isp major chang need take consider exchang server owa vpn two site link wireless connect km away citrix server vmware exchang domain control link place import server crucial step inform need k now avoid downtim busi regard ndavid',)

('use ef migrat creat databas googl migrat tutori af first run applic creat databas ef enabl migrat way creat databas migrat rune applic tri',)

('magento unit test problem magento site recent look way check integr magento site given point unit test jump one method would assum would big job write whole lot test check everyth site work anyon in volv unit test magento advis follow possibl test whole site custom modul nis exampl test would amaz given site heavili link databas would nbe possibl fulli test site without disturb databas better way automaticlli check integr magento site say integr realli mean fault site ship payment etc work corre ct',)

('find network devic without bonjour write mac applic need discov mac pcs iphon ipad connect wifi ne twork bonjour seem reason choic turn problem mani type router mine exampl work block bonjour servic need find ip devic tri connect applic specif port determin process run best approach accomplish task without violat app store sandbox',)

('send multipl row mysql databas want send user mysql databas column user skill time nnow want abl a

dd one row user differ time etc would code send databas nthen use help schema',)

('insert data mysql php powerpoint event powerpoint present run continu way updat slide present auto mat data mysql databas websit',)

```
#Taking 1 Million entries to a dataframe.
write_db = 'Processed.db'
if os.path.isfile(write_db):
    conn_r = create_connection(write_db)
    if conn_r is not None:
        preprocessed_data = pd.read_sql_query("""SELECT question, Tags FROM QuestionsProcessed""", conn_r)
conn_r.commit()
conn_r.close()
```

```
In []:
```

```
Out[ ]:
```

preprocessed_data.head()

	question	tags
0	resiz root window tkinter resiz root window re	python tkinter
1	ef code first defin one mani relationship diff	entity-framework-4.1
2	explan new statement review section c code cam	C++
3	error function notat function solv logic riddl	haskell logic
4	step plan move one isp anoth one work busi pla	dns isp

In []:

```
print("number of data points in sample :", preprocessed_data.shape[0])
print("number of dimensions :", preprocessed_data.shape[1])
```

```
number of data points in sample : 999999 number of dimensions : 2
```

4. Machine Learning Models

4.1 Converting tags for multilabel problems

```
        X
        y1
        y2
        y3
        y4

        x1
        0
        1
        1
        0

        x1
        1
        0
        0
        0

        x1
        0
        1
        0
        0
```

In []:

```
# binary='true' will give a binary vectorizer
vectorizer = CountVectorizer(tokenizer = lambda x: x.split(), binary='true')
multilabel_y = vectorizer.fit_transform(preprocessed_data['tags'])
```

We will sample the number of tags instead considering all of them (due to limitation of computing power)

In []:

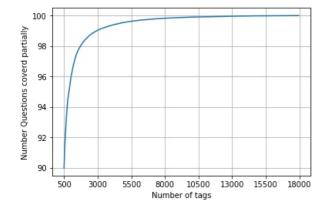
```
def tags_to_choose(n):
    t = multilabel_y.sum(axis=0).tolist()[0]
    sorted_tags_i = sorted(range(len(t)), key=lambda i: t[i], reverse=True)
    multilabel_yn=multilabel_y[:,sorted_tags_i[:n]]
    return multilabel_yn

def questions_explained_fn(n):
    multilabel_yn = tags_to_choose(n)
    x= multilabel_yn.sum(axis=1)
    return (np.count_nonzero(x==0))
```

```
questions_explained = []
total_tags=multilabel_y.shape[1]
total_qs=preprocessed_data.shape[0]
for i in range(500, total_tags, 100):
    questions_explained.append(np.round(((total_qs-questions_explained_fn(i))/total_qs)*100,3))
```

```
In [ ]:
```

```
fig, ax = plt.subplots()
ax.plot(questions_explained)
xlabel = list(500+np.array(range(-50,450,50))*50)
ax.set_xticklabels(xlabel)
plt.xlabel("Number of tags")
plt.ylabel("Number Questions coverd partially")
plt.grid()
plt.show()
# you can choose any number of tags based on your computing power, minimum is 50(it covers 90% of the tags)
print("with ",5500,"tags we are covering ",questions explained[50],"% of questions")
```



with 5500 tags we are covering 99.04 % of questions

In []:

```
multilabel_yx = tags_to_choose(5500)
print("number of questions that are not covered :", questions_explained_fn(5500),"out of ", total_qs)
```

number of questions that are not covered : 9599 out of 999999

In []:

```
print("Number of tags in sample :", multilabel_y.shape[1])
print("number of tags taken :", multilabel_yx.shape[1],"(",(multilabel_yx.shape[1]/multilabel_y.shape[1])*100,"%)
")
```

Number of tags in sample : 35422 number of tags taken : 5500 (15.527073570097679 %)

We consider top 15% tags which covers 99% of the questions

4.2 Split the data into test and train (80:20)

In []:

```
total_size=preprocessed_data.shape[0]
train_size=int(0.80*total_size)

x_train=preprocessed_data.head(train_size)
x_test=preprocessed_data.tail(total_size - train_size)

y_train = multilabel_yx[0:train_size,:]
y_test = multilabel_yx[train_size:total_size,:]
```

In []:

```
print("Number of data points in train data :", y_train.shape)
print("Number of data points in test data :", y_test.shape)
```

Number of data points in train data : (799999, 5500) Number of data points in test data : (200000, 5500)

4.3 Featurizing data

```
In [ ]:
start = datetime.now()
vectorizer = TfidfVectorizer(min_df=0.00009, max_features=200000, smooth_idf=True, norm="l2", \
                             tokenizer = lambda x: x.split(), sublinear_tf=False, ngram_range=(1,3))
x_train_multilabel = vectorizer.fit_transform(x_train['question'])
x_test_multilabel = vectorizer.transform(x_test['question'])
print("Time taken to run this cell :", datetime.now() - start)
Time taken to run this cell: 0:09:50.460431
In [ ]:
print("Dimensions of train data X:",x_train_multilabel.shape, "Y :",y_train.shape)
print("Dimensions of test data X:",x_test_multilabel.shape,"Y:",y_test.shape)
Diamensions of train data X: (799999, 88244) Y: (799999, 5500)
Diamensions of test data X: (200000, 88244) Y: (200000, 5500)
In [ ]:
# https://www.analyticsvidhya.com/blog/2017/08/introduction-to-multi-label-classification/
#https://stats.stackexchange.com/questions/117796/scikit-multi-label-classification
# classifier = LabelPowerset(GaussianNB())
from skmultilearn.adapt import MLkNN
classifier = MLkNN(k=21)
# train
classifier.fit(x train multilabel, y train)
# predict
predictions = classifier.predict(x_test_multilabel)
print(accuracy_score(y_test,predictions))
print(metrics.f1_score(y_test, predictions, average = 'macro'))
print(metrics.f1_score(y_test, predictions, average = 'micro'))
print(metrics.hamming_loss(y_test,predictions))
0.00
# we are getting memory error because the multilearn package
# is trying to convert the data into dense matrix
#MemoryError
                                           Traceback (most recent call last)
#<ipython-input-170-f0e7c7f3e0be> in <module>()
#----> classifier.fit(x train multilabel, y train)
Out[ ]:
```

"\nfrom skmultilearn.adapt import MLkNN\nclassifier = MLkNN(k=21)\n\n# train\nclassifier.fit(x_train _multilabel, y_train)\n\n# predict\npredictions = classifier.predict(x_test_multilabel)\nprint(accur acy_score(y_test,predictions))\nprint(metrics.fl_score(y_test, predictions, average = 'macro'))\nprint(metrics.fl_score(y_test, predictions, average = 'micro'))\nprint(metrics.hamming_loss(y_test,pred ictions))\n\n"

4.4 Applying Logistic Regression with OneVsRest Classifier

In []:

2

0.82

0.76

0.55

0.42

0.66

0.54

13446

12730

```
# this will be taking so much time try not to run it, download the lr with equal weight.pkl file and use to predi
ct
# This takes about 6-7 hours to run.
classifier = OneVsRestClassifier(SGDClassifier(loss='log', alpha=0.00001, penalty='l1'), n_jobs=-1)
classifier.fit(x_train_multilabel, y_train)
predictions = classifier.predict(x_test_multilabel)
print("accuracy :",metrics.accuracy_score(y_test,predictions))
print("macro f1 score :",metrics.f1_score(y_test, predictions, average = 'macro'))
print("micro f1 scoore :",metrics.f1 score(y test, predictions, average = 'micro'))
print("hamming loss:", metrics.hamming loss(y test, predictions))
print("Precision recall report :\n", metrics.classification report(y test, predictions))
accuracy : 0.081965
macro f1 score : 0.0963020140154
micro f1 scoore : 0.374270748817
hamming loss: 0.00041225090909090907
Precision recall report :
              precision
                           recall f1-score
                                              support
          0
                  0.62
                            0.23
                                      0.33
                                               15760
          1
                  0.79
                            0.43
                                      0.56
                                               14039
```

4 5 6 7 8 9 10 11 12 3 14 15 16 17 18 19 20 1 22 23 24 25 6 27 28 29 30 1 22 23 24 25 6 27 28 29 20 20 20 20 20 20 20 20 20 20 20 20 20
0.94 0.85 0.70 0.87 0.78 0.86 0.52 0.55 0.61 0.78 0.80 0.61 0.57 0.33 0.59 0.64 0.67 0.38 0.64 0.69 0.41 0.69 0.41 0.69 0.41 0.69 0.76 0.76 0.77 0.88 0.67 0.77 0.88 0.77 0.76 0.77
0.76 0.64 0.64 0.62 0.61 0.62 0.10 0.25 0.25 0.27 0.26 0.27 0.27 0.28 0.38 0.39 0.39 0.39 0.39 0.39 0.39 0.39 0.39 0.39 0.31 0.32 0.31 0.32 0.32 0.33 0.34 0.35 0.36 0.37 0.38 0.39 0.39 0.39 0.39 0.31 0.31 0.32 0.31 0.32 0.33 0.34 0.35 0.36 0.37 0.38 0.38 0.39 0.39 0.39 0.39 0.39 0.30 0.30 0.31 0.31 0.32 0.34 0.34 0.34 0.35 0.36 0.37 0.38
0.84 0.73 0.42 0.55 0.72 0.55 0.16 0.35 0.36 0.37 0.16 0.37 0.38 0.44 0.39 0.30 0.31 0.32 0.38 0.47 0.39 0.30 0.31 0.31 0.32 0.31 0.32 0.33 0.34 0.35 0.31 0.31 0.32 0.33 0.34 0.35 0.37 0.38 0.39 0.39 0.39 0.39 0.31 0.31 0.32 0.34 0.35 0.37 0.38 0.39
11229 10561 6958 6309 6032 6020 5707 5723 14722 4468 4545 4069 3638 3000 2585 2439 2157 2123 1801 1728 1725 1531 1568 1524 1531 1568 1524 1264 1265 1171 1173 1137 1105 1048 1058 1058 1058 1058 1058 1058 1058 105

87 88 89 90 91 92 93 94 95 96 97 98 99 100 101 102 103 104 105 107 108 109 110 111 112 113 114 115 116 117 118 119 120 121 123 124 125 126 127 128 131 142 143 144 145 147 148 149 151 152 153 154 155 156 167 168 169 169 169 169 169 169 169 169 169 169
0.93 0.91 0.58 0.71 0.44 0.71 0.66 0.86 0.57 0.64 0.89 0.22 0.64 0.92 0.68 0.72 0.68 0.73 0.69 0.68 0.70 0.68 0.70 0.68 0.70 0.68 0.70 0.68 0.70 0.68 0.70 0.68 0.70 0.70 0.70 0.70 0.70 0.70 0.70 0.7
0.58 0.520 0.42 0.047 0.10 0.669 0.37 0.10 0.609 0.31 0.609 0.318 0.052 0.0318 0.053 0.0318 0.053 0.0318 0.035 0.035 0.035 0.035 0.035 0.036 0.038
0.72 0.67 0.30 0.53 0.06 0.57 0.16 0.57 0.17
714 683 711 699 725 676 672 645 691 6645 6656 6656 6567 6566 657 578 564 5594 606 567 571 578 536 567 578 536 567 579 536 549 501 523 501 523 501 524 485 536 493 501 523 508 490 482 464 482 507 503 456 497 470 468 482 497 470 468 482 497 470 468 482 497 470 468 483 470 468 484 477 478 468 487 478 468 487 478 468 487 478 468 487 478 468 487 478 468 487 478 468 487 478 468 487 478 468 487 478 478 478 478 478 478 478 478 47

170 171 172 173 174 175 176 177 178 179 180 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 207 208 209 210 211 212 223 224 225 226 227 228 229 231 221 222 223 224 225 226 227 228 229 231 232 244 245 247 248 249 250 251 252
0.38 0.59 0.69 0.91 0.45 0.64 0.67 0.74 0.52 0.91 0.46 0.28 0.69 0.68 0.22 0.90 0.64 0.16 0.36 0.36 0.37 0.49 0.77 0.18 0.49 0.77 0.18 0.42 0.77 0.18 0.42 0.91 0.77 0.18 0.42 0.77 0.18 0.42 0.77 0.18 0.42 0.77 0.18 0.42 0.77 0.18 0.42 0.77 0.18 0.42 0.77 0.18 0.67 0.77 0.18 0.67 0.77 0.67 0.76 0.76 0.76 0.77 0.77 0.78 0.77
0.09 0.32 0.67 0.16 0.17 0.43 0.49 0.16 0.17 0.42 0.65 0.61 0.65 0.61 0.65 0.61 0.65 0.61 0.65 0.65 0.65 0.65 0.65 0.65 0.65 0.65
0.15 0.41 0.50 0.77 0.24 0.52 0.40 0.52 0.40 0.52 0.60
410 450 435 427 424 410 426 459 433 452 427 410 404 406 411 394 414 430 389 411 383 423 378 389 411 383 423 378 389 311 383 389 378 389 381 382 344 383 390 365 346 378 389 389 389 381 381 382 378 389 389 381 381 382 378 389 389 381 381 382 378 389 389 381 389 389 389 389 380 370 380 380 370 380 380 380 380 380 380 380 38

253 254 255 256 257 258 259 260 261 262 263 264 265 267 273 274 275 277 278 279 281 272 273 274 275 277 278 283 284 285 287 288 290 291 292 293 301 302 303 304 305 307 307 308 309 311 312 313 314 315 317 318 319 319 311 311 311 311 311 311 311 311
0.76 0.43 0.54 0.49 0.16 0.85 0.06 0.55 0.05 0.55 0.07 0.34 0.56 0.59 0.36 0.36 0.37 0.36 0.37 0.30 0.37 0.30 0.37 0.30 0.38 0.48 0.49 0.51 0.78 0.19 0.26 0.37 0.20 0.49 0.53 0.37 0.20 0.49 0.53 0.50 0.53 0.61 0.55 0.62 0.65 0.65 0.65 0.65 0.65 0.65 0.65 0.65
0.51 0.09 0.11 0.02 0.03 0.05 0.00 0.05 0.02 0.05
0.61 0.15 0.28 0.18 0.04 0.09 0.01 0.09 0.01 0.09
316 306 289 304 268 299 305 281 295 281 269 312 294 285 279 269 277 272 285 295 281 270 272 278 264 281 261 283 275 274 284 260 245 263 268 270 261 240 250 245 283 236 267 243 276 280 249 258 262 248 244 254 263 263 268 270 261 261 270 272 278 285 279 261 283 275 274 284 260 245 263 268 270 261 261 261 270 270 270 270 270 270 270 270 270 270

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0.57 0.20 0.00 0.22 0.66 0.57 0.45 0.17 0.28 0.37 0.48 0.51 0.57 0.44 0.58 0.77 0.96 0.47 0.90 0.06 0.50 0.43 0.27 0.48 0.76 0.90 0.41 0.00 0.76 0.93 0.10 0.20 0.41 0.43 0.41 0.00 0.87 0.95 0.17 0.28 0.12 0.84 0.06 0.31 0.24 0.36 0.31 0.24 0.36 0.31 0.24 0.36 0.31 0.28 0.36 0.31 0.29 0.40 0.22 0.62 0.96 0.31 0.72 0.40 0.22 0.62 0.96 0.31 0.72 0.40 0.22 0.62 0.96 0.31 0.72 0.40 0.22 0.62 0.96 0.31 0.72 0.40 0.22 0.62 0.96 0.31 0.72 0.40 0.22 0.62 0.96 0.71
0.30 0.01 0.00 0.02 0.30 0.22 0.03 0.12 0.03 0.12 0.01 0.09 0.12 0.39 0.12 0.39 0.12 0.39 0.12 0.39 0.12 0.39 0.14 0.00 0.16 0.00 0.17 0.37 0.37 0.26 0.37
0.40 0.03 0.00 0.04 0.14 0.29 0.04 0.17 0.13 0.20 0.18 0.20 0.18 0.27 0.18 0.29 0.57 0.10 0.12 0.14 0.25 0.14 0.25 0.14 0.15 0.16 0.16 0.17 0.18 0.19 0.19 0.10
211 212 222 227 216 231 233 232 209 216 222 243 222 248 205 177 234 230 195 209 205 211 230 211 220 219 222 213 199 200 199 212 214 197 212 210 211 213 216 195 187 191 178 193 187 193 189 194 183 189 191 206 221 195 187 198 199 200 188 199 200 188 189 199 200 188 189 199 200 188 189 199 200 188 189 191 206 221 195 187 203 205 218 196 207 217 219 208 209 209 209 209 209 209 209 209 209 209

419 420 421 4223 4244 4254 4274 4284 431 4334 4345 4374 4384 4384 4384 4384 4384 4384 4384
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177 168 187 209 177 182 187 185 185 185 185 175 190 185 189 184 200 167 209 200 169 170 182 156 170 170 176 194 175 187 170 176 194 175 187 170 182 172 190 183 182 173 171 173 184 175 162 176 177 167 192 168 188 163 160 180 180 182 177 176 192 168 188 163 160 180 181 177 175 185 167 197 167 192 168 188 163 160 180 180 181 177 175 185 167 197 167 197 168 188 163 160 180 180 181 171 174 162 169 157 167 175 185 167 197 1669 157 175 185 167 192 168 188 163 160 180 180 181 174 162 169 157 175 185 167 195 167 175 185 167 197 197 197 197 197 197 197 197 197 19

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143 177 177 152 179 171 151 162 158 164 149 174 164 152 175 168 145 165 151 171 160 139 165 148 178 152 143 174 135 163 174 135 163 127 130 155 163 148 152 143 147 153 165 149 140 143 147 153 165 148 149 159 140 143 147 151 157 163 165 148 149 159 140 143 147 159 165 148 149 159 165 149 149 149 149 149 159 169 169 169 169 169 169 169 169 169 16

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0.61 0.64 0.74 0.48 0.20 0.79 0.52 0.85 0.29 0.64 0.95 0.63 0.94 0.36 0.36 0.36 0.38 0.38 0.38 0.38 0.38 0.40 0.36 0.36 0.36 0.36 0.36 0.36 0.36 0.3
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134 151 150 141 137 154 126 144 130 148 115 142 123 150 134 165 150 137 133 146 129 151 138 124 144 150 130 127 141 133 148 117 125 123 148 117 129 113 120 121 123 148 149 121 125 123 148 149 121 125 123 148 149 121 121 125 126 127 128 129 121 121 121 122 123 124 127 128 129 129 120 121 121 122 123 124 125 126 127 128 129 129 120 120 121 121 122 123 124 125 126 127 128 129 129 120 120 120 121 121 122 123 124 126 127 128 129 129 120 120 120 121 121 122 123 124 125 126 127 128 129 129 120 120 120 120 120 120 120 120 120 120

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0.29 0.26 0.47 0.33 0.55 0.72 0.19 0.60 0.15 0.53 0.57 0.26 0.43 0.53 0.57 0.29 0.00 0.50 0.36 0.36 0.42 0.72 0.80 0.42 0.72 0.62 0.72 0.45 0.25 0.25 0.25 0.25 0.25 0.25 0.25 0.2
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131 127 125 111 127 130 130 126 104 127 130 131 140 114 112 115 128 122 109 108 125 117 127 129 118 151 112 119 109 122 102 107 105 113 98 100 131 112 119 105 113 110 130 101 112 197 115 129 101 112 199 105 117 115 129 101 112 199 105 117 115 129 101 110 110 110 110 110 110 110 110 11

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0.50 0.87 0.28 0.63 0.22 0.00 0.41 0.34 0.20 0.39 0.45 0.22 0.97 1.00 0.39 0.45 0.29 0.45 0.29 0.45 0.29 0.45 0.29 0.45 0.29 0.45 0.29 0.47 0.00 0.49 0.35 0.67 0.00 0.49 0.35 0.00 0.49 0.35 0.00 0.49 0.35 0.00 0.40 0.53 0.00 0.40 0.51 0.67 0.67 0.67 0.67 0.67 0.67 0.67 0.69 0.79 0.88 0.91 0.67 0.91 0.67 0.91 0.67 0.91 0.91 0.91 0.91 0.91 0.92 0.93 0.94 0.94 0.95 0.97 0.99 0.91 0.90 0.91 0.92 0.93 0.94 0.95 0.90 0.91 0.90 0.91 0.90 0.91 0.90 0.91 0.90 0.91 0.92 0.93 0.94 0.94 0.95
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91 92 104 102 111 96 86 105 92 88 101 94 101 88 81 109 101 97 91 88 101 80 97 91 88 101 80 97 91 88 102 99 98 88 102 99 98 88 102 99 88 104 90 104 91 91 91 91 91 91 91 91 91 91 91 91 91

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0.00 0.81 0.44 0.00 0.85 0.33 0.00 0.41 0.43 0.38 0.33 0.55 0.29 0.00 0.52 0.50 0.64 0.52 0.70 0.47 0.23 0.00 0.11 0.00 0.44 0.00 0.94 0.09 0.12 0.29 1.00 0.83 0.81 0.87 0.43 0.81 0.87 0.43 0.81 0.87 0.43 0.81 0.97 0.93 0.95 0.90 0.95 0.90 0.95 0.90 0.95 0.90 0.95 0.90 0.95 0.90 0.95 0.90 0.95 0.90 0.95 0.93 0.90 0.95 0.93 0.90 0.95 0.93 0.90 0.95 0.93 0.90 0.95 0.93 0.90 0.95 0.93 0.90 0.95 0.93 0.90 0.95 0.98 0.99 0.99 0.99 0.99 0.99 0.99 0.99
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0.00 0.51 0.16 0.00 0.41 0.04 0.00 0.14 0.06 0.15 0.05 0.12 0.11 0.00 0.02 0.12 0.40 0.23 0.34 0.16 0.13 0.00 0.02 0.02 0.04 0.05 0.06 0.00 0.05 0.05 0.06 0.00 0.05 0.05
92 78 81 87 95 89 73 85 76 85 76 85 76 87 88 88 76 87 88 88 77 88 88 78 88 78 88 78 88 78 88 78 88 78 88 78 88 78 88 78 88 78 88 78 7

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0.00 0.14 0.67 0.00 0.59 0.50 0.17 0.62 0.00 0.47 0.59 0.62 0.00 0.80 0.65 0.00 0.46 0.93 0.86 0.12 0.00 0.33 0.88 0.17 0.44 0.00 0.33 0.88 0.17 0.00 0.10
0.00 0.02 0.25 0.00 0.08 0.55 0.00 0.15 0.00 0.16 0.12 0.00 0.16 0.25 0.00 0.16 0.25 0.00 0.11 0.22 0.32 0.00 0.32 0.00 0.32 0.00 0.32 0.00 0.04 0.05 0.07 0.04 0.05 0.00 0.05 0.06 0.07 0.06 0.07 0.08 0.09 0.09 0.09 0.09 0.09 0.09 0.09
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74 62 71 72 75 72 81 74 75 90 88 87 74 87 87 87 87 87 87 87 87 87 87 87 87 87

1083 1084 1085 1086 1087 1088 1089 1090 1091 1092 1093 1094 1095 1096 1097 1098 1100 1101 1102 1103 1104 1105 1106 1107 1108 1109 1110 1111 1112 1113 1114 1115 1116 1117 1118 1119 1120 1121 1122 1123 1124 1125 1126 1127 1128 1129 1130 1131 1132 1133 1134 1135 1136 1137 1138 1139 1140 1141 1142 1143 1144 1145 1146 1147 1148 1149 1150 1151 1151 1151 1151 1151 1151 115
0.09 0.51 0.69 0.00 0.40 0.00 0.40 0.35 0.38 0.65 0.00 0.36 0.36 0.44 0.58 0.80 0.57 0.00 0.90 0.14 0.40 0.21 0.25 0.00 0.00 0.05 0.00 0.65 0.20 0.25 0.00 0.25 0.00 0.25 0.00 0.29 0.50 0.20 0.25 0.00 0.29 0.50 0.20 0.25 0.00 0.29 0.50 0.20 0.25 0.00 0.27 0.00 0.38 0.15 0.00 0.39 0.38 0.00 0.39 0.38 0.00 0.39 0.00 0.44 0.00 0.39 0.00 0.44 0.00 0.39 0.00 0.44 0.00 0.39 0.00 0.44 0.00 0.39 0.00 0.44 0.00 0.39 0.00 0.04 0.39 0.00 0.04 0.39 0.00 0.04 0.35 0.71 0.37 0.41 0.57 0.00 0.00 0.94 0.62 0.90 0.94 0.62 0.90 0.94 0.62 0.90 0.94
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0.03 0.34 0.38 0.09 0.05 0.05 0.05 0.15 0.28 0.00 0.15 0.28 0.00 0.19 0.48 0.32 0.01 0.00 0.03 0.00 0.03 0.00 0.03 0.00 0.03 0.00 0.03 0.00 0.03 0.00 0.03 0.00 0.00 0.03 0.00 0.03 0.00
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1166 1167 1168 1169 1170 1171 1172 1173 1174 1175 1176 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199 1200 1201 1202 1203 1204 1205 1206 1207 1208 1210 1211 1212 1213 1214 1215 1216 1217 1218 1219 1220 1221 1222 1223 1224 1225 1226 1227 1228 1230 1231 1232 1233 1234 1245 1246 1247 1248
0.00 0.46 0.33 0.35 0.80 0.60 0.29 0.23 0.45 0.98 0.87 0.00 0.97 0.70 0.88 0.12 0.00 0.97 0.70 0.88 0.12 0.00 0.33 0.82 0.17 0.45 0.50 0.50 0.60 0.60 0.60 0.60 0.60 0.6
0.00 0.21 0.03 0.11 0.05 0.31 0.03 0.04 0.14 0.60 0.42 0.00 0.07 0.12 0.30 0.02 0.00 0.01
0.00 0.29 0.06 0.17 0.10 0.41 0.06 0.07 0.22 0.74 0.57 0.00 0.00 0.03 0.21 0.44 0.03 0.00 0.06 0.30 0.14 0.03 0.03 0.14 0.03 0.04 0.05 0.06 0.03 0.18 0.10 0.06 0.03 0.18 0.10 0.06 0.03 0.18 0.10 0.06 0.07 0.00 0.06 0.03 0.18 0.10 0.06 0.17 0.09 0.09 0.09 0.09 0.09 0.09 0.09 0.0
73 56 63 73 58 59 63 76 62 45 79 58 71 66 67 67 66 67 67 67 66 67 67 67 67 67

1249 1250 1251 1252 1253 1254 1255 1256 1257 1258 1259 1260 1261 1262 1263 1264 1265 1266 1267 1271 1272 1273 1274 1275 1276 1271 1272 1273 1274 1275 1276 1277 1278 1279 1280 1281 1282 1283 1284 1285 1286 1287 1288 1289 1290 1291 1292 1293 1294 1295 1296 1301 1302 1303 1304 1305 1306 1307 1308 1309 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1311 1312 1313 1314 1315 1316 1317 1318 1319 1319 1321 1322 1323 1324 1325 1326 1327 1328 1329 1329 1329 1329 1329 1329 1329 1329
0.33 0.97 0.38 0.37 0.38 0.59 0.00 0.00 0.00 0.39 0.62 0.00 0.93 0.00 0.93 0.00 0.94 0.25 0.00 0.35 0.00 0.25
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0.09 0.64 0.21 0.15 0.31 0.73 0.00 0.00 0.00 0.20 0.00 0.36 0.00 0.36 0.00 0.60 0.00 0.60 0.12 0.00 0.60 0.12 0.00 0.12 0.00 0.14 0.00 0.12 0.01 0.01 0.01 0.01 0.02 0.01 0.01 0.02 0.01 0.02 0.03 0.04 0.05 0.05 0.06 0.07 0.08 0.09
74 56 56 56 56 57 56 56 56 57 56 56 56 57 56 56 57 56 56 57 56 57 57 57 57 57 57 57 57 57 57 57 57 57

1332	0.00	0.00	0.00	48
1333	0.00	0.00	0.00	51
1334	0.00	0.00	0.00	38
1335	0.91	0.42	0.58	50
1336	0.00	0.00	0.00	48
1337	0.38	0.10	0.15	52
1338	0.58	0.21	0.31	52
1339	0.25	0.04	0.06	56
1340	0.50	0.04	0.07	52
1341	1.00		0.03	58
1342	0.00	0.00	0.00	56
1343	0.33		0.06	62
1344	0.93	0.32	0.47	44
1345	0.38		0.10	53
1346	0.20	0.02	0.03	53
1347	0.00		0.00	52
1348	0.50	0.10	0.17	58
1349	0.64	0.36	0.46	50
1350 1351 1352	0.00 0.96	0.00	0.00 0.55	62 59
1352 1353 1354	0.00 0.63 0.67	0.00 0.24	0.00 0.35	57 50 55
1355 1356	0.00 0.17	0.11 0.00 0.02	0.19 0.00 0.03	55 56
1357 1358	0.17 0.16 0.20	0.02 0.08 0.04	0.03 0.11 0.06	38 53
1359	1.00	0.23	0.37	44
1360	1.00	0.23	0.38	56
1361	0.25	0.04	0.06	56
1362	1.00	0.33	0.49	46
1363	0.73	0.22	0.34	49
1364	0.00		0.00	66
1365	0.33	0.05	0.09	60
1366	0.86	0.11	0.19	56
1367	0.00	0.00	0.00	63
1368	0.53	0.15	0.23	67
1369	1.00	0.44	0.61	59
1370	0.94		0.48	49
1371	0.76	0.25	0.38	51
1372	0.20	0.02	0.04	50
1373	0.93	0.40	0.56	63
1374	0.20	0.02	0.03	55
1375	0.00	0.00	0.00	60
1376	0.52	0.18	0.27	60
1377	0.00	0.00	0.00	42
1378	0.94	0.30	0.45	54
1379	0.00	0.00	0.00	50
1380	0.00	0.00	0.00	45
1381	0.60	0.06	0.12	47
1382	0.11	0.02	0.03	54
1383 1384	0.33	0.04	0.08	45 52
1385	0.73	0.23	0.35	48
1386	0.60		0.11	50
1387	0.17	0.02	0.04	47
1388	0.75	0.16	0.26	57
1389	0.00	0.00	0.00	49
1390	0.55	0.27	0.36	44
1391	0.00	0.00	0.00	58
1392	0.77	0.19	0.30	54
1393	0.38	0.12	0.18	51
1394	0.50	0.02	0.04	51
1395	0.83	0.21	0.33	48
1396	0.67	0.13	0.22	61
1397	1.00	0.02	0.03	61
1398 1399	0.62 0.74	0.15 0.25	0.03 0.24 0.37	55 57
1400	0.50	0.06	0.11	49
1401	0.50	0.04	0.07	56
1402	0.54	0.13	0.22	52
1403	0.75	0.12	0.21	49
1404	0.92	0.80	0.86	41
1405	0.75	0.32	0.44	57
1406	0.33	0.02	0.04	54
1407	0.70	0.55	0.62	47
1408	0.38	0.07	0.12	41
1409	1.00	0.39	0.56	49
1410	1.00	0.44	0.61	48
1411	0.17	0.02	0.03	55
1412	0.73	0.13	0.23	60
1413	1.00	0.01	0.03	67
1414	0.00	0.00	0.00	50

1415 1416 1417 1418 1420 1421 1422 1423 1424 1425 1426 1427 1428 1431 1432 1433 1434 1435 1437 1438 1438 1439 1441 1442 1443 1444 1445 1446 1451 1451 1451 1451 1451
0.00 0.40 0.53 0.67 0.80 0.30 0.90 0.38 0.82 0.50 0.00 0.67 0.30 0.97 0.86 0.00 0.97 0.86 0.00 0.10 0.10 0.15 0.00 0.11 0.00 0.15 0.00 0.65 0.20 0.65 0.60 0.60 0.60 0.60 0.60 0.60 0.6
0.00 0.10 0.14 0.04 0.11 0.06 0.00 0.10 0.18 0.07 0.00 0.04 0.06 0.00 0.10 0.12 0.11 0.16 0.00 0.04 0.00 0.04 0.00 0.04 0.00 0.02 0.04 0.00 0.02 0.07 0.07 0.00 0.02 0.07 0.07
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1830 1831 1832 1833 1834 1835 1836 1837 1838 1840 1841 1842 1843 1844 1845 1846 1847 1848 1849 1850 1851 1852 1853 1854 1855 1856 1857 1858 1859 1860 1861 1862 1863 1864 1865 1866 1867 1878 1879 1870 1871 1872 1873 1874 1875 1876 1877 1878 1879 1880 1871 1872 1873 1874 1875 1876 1877 1878 1879 1880 1871 1872 1873 1874 1875 1876 1877 1878 1879 1880 1891 1892 1893 1894 1895 1896 1897 1898 1899 1900 1901 1902 1903 1904 1905 1906 1907 1908 1909 1910 1911 1912
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354 332 345 345 345 347 347 347 347 347 347 347 347 347 347

1913	0.00	0.00	0.00	38
1914	0.73	0.19	0.30	43
1915	0.84	0.52	0.64	31
1916	0.33	0.08	0.12	39
1917	0.00	0.00	0.00	38
1918	0.75	0.20	0.32	45
1919	0.58	0.19	0.29	37
1920	0.00	0.00	0.00	29
1921	0.00	0.00	0.00	31
1922	0.61	0.34	0.44	41
1923	0.17	0.02	0.03	54
1924	0.80	0.12	0.22	32
1925	0.00	0.00	0.00	32
1926	0.00	0.00	0.00	38
1927	0.94	0.38	0.54	42
1928	0.00	0.00	0.00	41
1929	0.00	0.00	0.00	47
1930	1.00	0.40	0.57	30
1931	1.00	0.05	0.09	41
1932	0.00	0.00	0.00	40
1933	0.62	0.19	0.29	43
1934	0.00	0.00	0.00	42
1935	0.33	0.06	0.10	36
1936	0.57	0.29	0.38	42
1937	1.00	0.03	0.05	36
1938	0.94	0.50	0.65	32
1939	1.00	0.12	0.21	50
1940	0.33	0.03	0.05	35
1941	0.00	0.00	0.00	41
1942	0.80	0.20	0.32	40
1943	0.00	0.00	0.00	38
1944	0.84	0.47	0.60	34
1945	0.00	0.00	0.00	42
1946	0.90	0.32	0.47	28
1947	0.00	0.00	0.00	37
1948	0.00	0.00	0.00	32
1949	0.00	0.00	0.00	32
1950	0.69	0.35	0.46	26
1951	0.00	0.00	0.00	49
1952	0.00	0.00	0.00	32
1953	0.50	0.03	0.06	31
1954	0.71	0.12	0.21	40
1955	0.00	0.00	0.00	47
1956	1.00	0.07	0.13	43
1957	0.00	0.00	0.00	38
1958	0.77	0.26	0.39	38
1959	0.00	0.00	0.00	34
1960	0.32	0.21	0.25	39
1961	1.00	0.03	0.06	34
1962	0.20	0.02	0.04	42
1963	0.60	0.09	0.16	32
1964	0.00	0.00	0.00	41
1965 1966 1967 1968 1969	0.33 0.00 0.00 0.86	0.02 0.00 0.00 0.60 0.24	0.04 0.00 0.00 0.71	42 37 41 30 25
1970 1971 1972 1973	0.50 0.50 0.00 0.00 0.00	0.24 0.15 0.00 0.00 0.00	0.32 0.23 0.00 0.00 0.00	40 43 42 32
1974	0.00	0.00	0.00	33
1975	1.00	0.21	0.35	28
1976	0.00	0.00	0.00	35
1977	0.92	0.22	0.36	49
1978	1.00	0.33	0.49	49
1979	0.00	0.00	0.00	34
1980	0.00	0.00	0.00	28
1981	1.00	0.24	0.38	34
1982	0.00	0.00	0.00	30
1983	0.50	0.03	0.05	40
1984	0.00	0.00	0.00	38
1985	0.00	0.00	0.00	42
1986	0.00	0.00	0.00	32
1987	0.00	0.00	0.00	37
1988	0.25	0.03	0.05	34
1989	0.75	0.15	0.24	41
1990	0.00	0.00	0.00	34
1991	0.00	0.00	0.00	34
1992	0.00	0.00	0.00	30
1993	0.67	0.17	0.27	36
1994	0.83	0.16	0.26	32
1995	0.00	0.00	0.00	38

1996 1997 1998 1999 2000 2001 2002 2003 2004 2005 2007 2008 2007 2012 2013 2014 2015 2017 2018 2019 2020 2021 2022 2023 2024 2025 2027 2028 2020 2031 2032 2031 2032 2033 2034 2035 2037 2038 2040 2041 2042 2043 2044 2045 2046 2047 2048 2049 2055 2056 2057 2058 2060 2061 2062 2066 2066 2067 2077 2078
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2162 2163 2164 2165 2166 2167 2168 2169 2170 2171 2172 2173 2174 2175 2176 2177 2178 2179 2180 2181 2182 2183 2184 2185 2186 2187 2188 2190 2191 2192 2193 2194 2195 2196 2207 2208 2201 2202 2203 2204 2212 2213 2214 2215 2216 2217 2218 2219 2221 2222 2223 2244 2255 2266 227 2228 2229 2230 2211 2212 2213 2214 2215 2216 2217 2218 2219 2220 2221 2222 2223 2224 2225 2226 2227 2228 2229 2230 2231 2212 2223 2224 2225 2226 2227 2228 2229 2230 2231 2212 2223 2224 2225 2226 2227 2228 2229 2230 2231 2222 2223 2224 2225 2226 2227 2228 2229 2230 2231 2222 2224 2222 2224 2222 2224 2222 2224 2224 2224 2242 2243 2244
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27 37 30 35 24 37 26 27 39 25 33 39 36 28 31 32 33 33 31 32 33 33 33 34 32 31 32 32 33 33 34 35 36 37 37 38 38 39 39 39 39 39 39 39 39 39 39 39 39 39

2245 2246 2247 2248 2249 2250 2251 2252 2253 2254 2255 2256 2257 2258 2259 2260 2261 2262 2263 2264 2265 2266 2267 2268 2269 2271 2272 2273 2274 2275 2278 2279 2280 2281 2282 2283 2284 2285 2286 2287 2288 2290 2291 2292 2293 2294 2295 2296 2297 2298 2290 2301 2302 2303 2304 2305 2307 2308 2309 2310 2311 2312 2331 2314 2315 2316 2317 2318 2319 2321 2322 2323 2324 2325 2326 2327
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2328 2329 2330 2331 2332 2333 2334 2335 2337 2338 2339 2340 2341 2342 2343 2344 2345 2346 2347 2348 2350 2351 2353 2356 2357 2358 2359 2360 2371 2372 2373 2373 2374 2375 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2377 2378 2379 2379 2379 2379 2379 2379 2379 2379
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25 28 30 29 32 34 30 26 33 36 37 18 30 32 32 32 33 31 32 32 33 31 32 32 33 31 32 32 33 31 32 32 33 31 32 32 33 33 34 35 36 37 37 38 38 38 38 38 38 38 38 38 38 38 38 38

2411 2412 2413 2414 2415 2416 2417 2418 2419 2420 2421 2422 2423 2424 2425 2427 2431 2432 2431 2432 2433 2434 2435 2437 2438 2439 2441 2442 2443 2444 2445 2446 2457 2458 2458 2458 2458 2458 2458 2458 2458
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2577 2578 2579 2580 2581 2582 2583 2584 2585 2588 2589 2591 2589 2591 2593 2594 2599 2601 2602 2603 2604 2605 2606 2607 2608 2609 2611 2612 2613 2614 2615 2616 2617 2622 2623 2624 2625 2626 2637 2638 2639 2640 2641 2642 2643 2644 2645 2646 2647 2648 2659 2650 2651 2652 2653 2654 2655 2656 2657 2658 2659
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2660 2661 2662 2663 2664 2665 2666 2667 2668 2669 2671 2672 2673 2674 2675 2676 2677 2678 2679 2680 2681 2682 2683 2684 2685 2688 2689 2691 2692 2693 2691 2692 2693 2694 2695 2696	0.00 0.00 0.83 0.00 0.00 0.00 0.00 0.00	0.00 0.00 0.19 0.00	0.00 0.00 0.31 0.00 0.00 0.00 0.00 0.00	24 18 26 28 22 28 31 18 32 24 25 26 23 24 26 19 21 30 28 21 32 23 24 26 23 24 26 27 23 24 26 27 28 29 20 20 20 20 20 20 20 20 20 20
2668 2669 2670 2671 2672 2673 2674 2675 2676 2677 2678 2680 2681 2682 2683 2684 2685 2686 2687 2688 2689 2690 2691 2692 2693 2694 2695 2696	0.00 0.00 0.00 0.00 0.00 0.93 0.50 1.00 0.00 0.00 0.00 0.00 0.00 0.00	0.00 0.00 0.00 0.00 0.00 0.56 0.04 0.13 0.00 0.00 0.00 0.00 0.00 0.00 0.11 0.00 0.00 0.11 0.00 0.00 0.00 0.00 0.00 0.00	0.00 0.00 0.00 0.00 0.00 0.70 0.23 0.00 0.00 0.00 0.00 0.00 0.00 0.0	18 32 24 22 23 25 26 23 24 26 19 21 30 28 26 23 28 21 32 27 23 24 26 23 24 26 29 20 20 20 20 20 20 20 20 20 20
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2909 2910 2911 2912 2913 2914 2915 2916 2917 2918 2919 2920 2921 2922 2923 2924 2925 2926 2927 2928 2930 2931 2932 2933 2934 2935 2936 2937 2938 2939 2940 2941 2942 2943 2944 2945 2946 2947 2948 2949 2950 2951 2952 2953 2954 2955 2956 2967 2968 2969 2971 2972 2973 2974 2975 2976 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2977 2978 2979 2980 2961 2962 2963 2964 2965 2967 2977 2978 2978 2979 2970 2971 2972 2973 2974 2975 2976 2977 2978 2988 29990 2991
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5227	0.00	0.00	0.00	13
5228	0.00	0.00	0.00	7
5229	0.00	0.00	0.00	6
5230	0.00	0.00	0.00	7
5231	0.00	0.00	0.00	10
5232	0.00	0.00	0.00	7
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5311 0.00 0.00 0.00 10 5312 0.00 0.00 0.00 11 5313 0.00 0.00 0.00 11 5314 0.00 0.00 0.00 11					
5313 0.00 0.00 0.00 11 5314 0.00 0.00 0.00 11	5311	0.00	0.00	0.00	10
5314 0.00 0.00 0.00 11					
5315 0.00 0.00 0.00 11	5314	0.00	0.00	0.00	11
	5315	0.00	0.00	0.00	11

5316	0.00	0.00	0.00	2
5317	0.00	0.00	0.00	5
5318	0.00	0.00	0.00	11
5319	0.00	0.00	0.00	12
5320	0.00	0.00	0.00	7
5321	0.00	0.00	0.00	7
5322	0.00	0.00	0.00	9
5323	0.00	0.00	0.00	9
5324	0.00	0.00	0.00	8
5325	0.00	0.00	0.00	10
5326	0.00	0.00	0.00	3
5327	0.00	0.00	0.00	13
	0.00	0.00	0.00	13
5328 5329	0.00	0.00	0.00	7
5330	0.00	0.00	0.00	8
5331	0.00	0.00	0.00	9
5332	0.00	0.00	0.00	8
5333	0.00	0.00	0.00	11
5334	0.00	0.00	0.00	11
5335	0.00	0.00	0.00	6
5336	0.00	0.00	0.00	6
5337	0.00	0.00	0.00	6
5338	0.00	0.00	0.00	11
5339	0.00	0.00	0.00	12
5340	0.00	0.00	0.00	9
5341		0.00	0.00	8
5342	0.00 0.00	0.00	0.00	8
5343	0.00	0.00	0.00	7
5344	0.00	0.00	0.00	5
5345	0.00	0.00	0.00	11
5346	0.00	0.00	0.00	13
5347	0.00	0.00	0.00	10
5348	0.00	0.00	0.00	11
5349	0.00	0.00	0.00	7
5350	0.00	0.00	0.00	10
5351	0.00	0.00	0.00	7
5352	0.00	0.00	0.00	7
5353	0.00	0.00	0.00	11
5354	0.00	0.00	0.00	12
5355	0.00	0.00	0.00	12
5356	0.00	0.00	0.00	10
5357	0.00	0.00	0.00	9
5358	0.00	0.00	0.00	8
5359	0.00	0.00	0.00	7
5360	0.00	0.00	0.00	10
5361	0.00	0.00	0.00	6
5362	0.00	0.00	0.00	6
5363	0.00	0.00	0.00	9
5364	0.00	0.00	0.00	
5365	0.00	0.00	0.00	17
5366	0.00	0.00	0.00	8
5367	0.00	0.00	0.00	9
5368	0.00	0.00	0.00	8
5369	0.00	0.00	0.00	8
5370	0.00	0.00	0.00	18
5371	0.00	0.00	0.00	14
5372	0.00	0.00	0.00	10
5373	0.00	0.00	0.00	7
5374	0.00	0.00	0.00	6
5375	0.00	0.00	0.00	12
5376	0.00	0.00	0.00	13
5377	0.00	0.00	0.00	9
5378	0.00	0.00	0.00	10
5379	0.00	0.00	0.00	10
5380	0.00	0.00	0.00	9
5381	0.00	0.00	0.00	7
5382	0.00	0.00	0.00	10
5383	0.00	0.00	0.00	9
5384	0.00	0.00	0.00	12
5385	0.00	0.00	0.00	15
5386	0.00	0.00	0.00	7
5387	0.00	0.00	0.00	8
5388	0.00	0.00	0.00	4
5389	0.00	0.00	0.00	7
5390	0.00	0.00	0.00	8
5391	0.00	0.00	0.00	4
5392	0.00	0.00	0.00	10
5393	0.00	0.00	0.00	7
5394	0.00	0.00	0.00	8
5395	0.00	0.00	0.00	16
5396	0.00	0.00	0.00	13
5397	0.00	0.00	0.00	11
5398	0.00	0.00	0.00	5

5399	0.00	0.00	0.00	5
5400	0.00	0.00	0.00	12
5401	0.00	0.00	0.00	7
5402	0.00	0.00	0.00	5
5403	0.00	0.00	0.00	12
5404	0.00	0.00	0.00	5
5405	0.00	0.00		10
5406	0.00	0.00	0.00	7
5407	0.00	0.00	0.00	12
5408	0.00	0.00	0.00	9
5409 5410	0.00 0.00	0.00	0.00 0.00	9
5411 5412	0.00 0.00	0.00	0.00 0.00	6 8 6
5413 5414 5415	0.00 0.00 0.00	0.00 0.00 0.00	0.00 0.00 0.00	8 16
5416	0.00	0.00	0.00	9
5417	0.00	0.00	0.00	11
5418	0.00	0.00	0.00	9
5419	0.00	0.00	0.00	14
5420	0.00	0.00	0.00	6
5421 5422	0.00 0.00	0.00	0.00	11 12
5423	0.00	0.00	0.00	8
5424	0.00	0.00	0.00	13
5425	0.00	0.00	0.00	4
5426	0.00	0.00	0.00	10
5427	0.00		0.00	9
5428	0.00	0.00	0.00	12
5429	0.00		0.00	11
5430	0.00	0.00	0.00	9
5431	0.00	0.00	0.00	15
5432	0.00	0.00	0.00	12
5433	0.00	0.00	0.00	8
5434	0.00	0.00	0.00	6
5435	0.00	0.00	0.00	12
5436	0.00	0.00	0.00	11
5437	0.00	0.00	0.00	10
5438	0.00	0.00	0.00	7
5439	0.00		0.00	9
5440	0.00	0.00	0.00	12
5441	0.00	0.00	0.00	10
5442	0.00	0.00	0.00	7
5443	0.00	0.00	0.00	12
5444	0.00		0.00	7
5445 5446	0.00 0.00	0.00	0.00	9 7
5447	0.00	0.00	0.00	6
5448	0.00	0.00	0.00	12
5449	0.00	0.00	0.00	9
5450	0.00	0.00	0.00	10
5451	0.00		0.00	6
5452	0.00	0.00	0.00	11
5453	0.00	0.00	0.00	7
5454	0.00	0.00	0.00	9
5455	0.00	0.00	0.00	11
5456	0.00	0.00	0.00	7
5457	0.00	0.00	0.00	9
5458	0.00	0.00	0.00	8
5459	0.00	0.00	0.00	11
5460	0.00	0.00	0.00	7
5461	0.00		0.00	11
5462	0.00	0.00	0.00	10
5463	0.00	0.00	0.00	9
5464	0.00	0.00	0.00	9
5465 5466	0.00	0.00	0.00 0.00	7 9
5467	0.00	0.00	0.00	14
5468		0.00	0.00	9
5469	0.00	0.00	0.00	12
5470	0.00	0.00	0.00	11
5471	0.00	0.00	0.00	8
5472 5473	0.00 0.00	0.00	0.00	15 4
5474	0.00	0.00	0.00	8
5475	0.00	0.00	0.00	9
5476	0.00	0.00	0.00	11
5477	0.00	0.00	0.00	8
5478	0.00	0.00	0.00	6
5479	0.00	0.00	0.00	7
5480	0.00	0.00	0.00	7
5481	0.00	0.00	0.00	10

```
5482
                    0.00
                               0.00
                                          0.00
                                                       12
       5483
                    0.00
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       5486
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       5487
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                                          0.00
                                                        7
       5488
                    0.00
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                                          0.00
                                                        10
       5489
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       5490
                    0.00
                               0.00
                                          0.00
                                                        12
       5491
                    0.00
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                                          0.00
                                                        6
                    0.00
                                                        8
       5492
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                                          0.00
       5493
                    0.00
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                                          0.00
                                                       13
       5494
                    0.00
                               0.00
                                          0.00
                                                        6
       5495
                    0.00
                               0.00
                                                       10
                                          0.00
       5496
                    0.00
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                    0.00
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                                                        9
       5497
       5498
                    0.00
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                                          0.00
                                                        6
       5499
                    0.00
                               0.00
                                          0.00
                                                       13
                    0.53
                               0.26
                                          0.33
                                                   530065
avg / total
```

In []:

```
from sklearn.externals import joblib
joblib.dump(classifier, 'lr_with_equal_weight.pkl')
```

4.5 Modeling with less data points (0.5M data points) and more weight to title and 500 tags only.

```
In [ ]:
```

```
sql_create_table = """CREATE TABLE IF NOT EXISTS QuestionsProcessed (question text NOT NULL, code text, tags text
, words_pre integer, words_post integer, is_code integer);"""
create_database_table("Titlemoreweight.db", sql_create_table)
```

Tables in the databse: QuestionsProcessed

```
In [ ]:
```

```
# http://www.salitetutorial.net/salite-delete/
# https://stackoverflow.com/questions/2279706/select-random-row-from-a-sqlite-table
read db = 'train no dup.db'
write db = 'Titlemoreweight.db'
train datasize = 400000
if os.path.isfile(read db):
    conn_r = create_connection(read_db)
    if conn r is not None:
        reader =conn_r.cursor()
        # for selecting first 0.5M rows
        reader.execute("SELECT Title, Body, Tags From no_dup_train LIMIT 500001;")
        # for selecting random points
        #reader.execute("SELECT Title, Body, Tags From no dup train ORDER BY RANDOM() LIMIT 500001;")
if os.path.isfile(write_db):
    conn_w = create_connection(write_db)
    if conn_w is not None:
        tables = checkTableExists(conn w)
        writer =conn w.cursor()
        if tables != 0:
            writer.execute("DELETE FROM QuestionsProcessed WHERE 1")
            print("Cleared All the rows")
```

Tables in the databse: QuestionsProcessed Cleared All the rows

4.5.1 Preprocessing of questions

- 1. Separate Code from Body
- 2. Remove Spcial characters from Question title and description (not in code)
- 3. Give more weightage to title: Add title three times to the question
- 4. Remove stop words (Except 'C')
- 5. Remove HTML Tags
- 6. Convert all the characters into small letters
- 7. Use SnowballStemmer to stem the words

Time taken to run this cell: 0:23:12.329039

In []:

```
#http://www.bernzilla.com/2008/05/13/selecting-a-random-row-from-an-sqlite-table/
start = datetime.now()
preprocessed data list=[]
reader.fetchone()
questions_with_code=0
len pre=0
len post=0
questions_proccesed = 0
for row in reader:
    is code = 0
    title, question, tags = row[0], row[1], str(row[2])
    \quad \textbf{if} \ ' \texttt{<} \texttt{code} \texttt{>} ' \ \ \textbf{in} \ \ \texttt{question} \texttt{:}
        {\tt questions\_with\_code+=1}
        is code = 1
    x = len(question)+len(title)
    len pre+=x
    code = str(re.findall(r'<code>(.*?)</code>', question, flags=re.DOTALL))
    question=re.sub('<code>(.*?)</code>', '', question, flags=re.MULTILINE|re.DOTALL)
    question=striphtml(question.encode('utf-8'))
    title=title.encode('utf-8')
    # adding title three time to the data to increase its weight
    # add tags string to the training data
    question=str(title)+" "+str(title)+" "+str(title)+" "+question
#
      if questions_proccesed<=train_datasize:</pre>
         question=str(title)+" "+str(title)+" "+str(title)+" "+question+" "+str(tags)
#
#
      else:
#
          question=str(title)+" "+str(title)+" "+str(title)+" "+question
    question=re.sub(r'[^A-Za-z0-9#+.\-]+','',question)
    words=word_tokenize(str(question.lower()))
    #Removing all single letter and and stopwords from question exceptt for the letter 'c'
    question=' '.join(str(stemmer.stem(j)) for j in words if j not in stop_words and (len(j)!=1 or j=='c'))
    len_post+=len(question)
    tup = (question,code,tags,x,len(question),is_code)
    questions\_proccesed += 1
    writer.execute("insert into QuestionsProcessed(question,code,tags,words pre,words post,is code) values (?,?,?
,?,?,?)",tup)
    if (questions_proccesed%100000==0):
        print("number of questions completed=",questions proccesed)
no dup avg len pre=(len pre*1.0)/questions proccesed
no_dup_avg_len_post=(len_post*1.0)/questions_proccesed
print( "Avg. length of questions(Title+Body) before processing: %d"%no_dup_avg_len_pre)
print( "Avg. length of questions(Title+Body) after processing: %d"%no_dup_avg_len_post)
print ("Percent of questions containing code: %d"%((questions with code*100.0)/questions proccesed))
print("Time taken to run this cell :", datetime.now() - start)
number of questions completed= 100000
number of questions completed= 200000
number of questions completed= 300000
number of questions completed= 400000
number of questions completed= 500000
Avg. length of questions(Title+Body) before processing: 1239
Avg. length of questions(Title+Body) after processing: 424
Percent of questions containing code: 57
```

```
In [ ]:
```

```
conn_r.commit()
conn_w.commit()
conn_r.close()
conn_w.close()
```

Sample quesitons after preprocessing of data

In []:

```
if os.path.isfile(write_db):
    conn_r = create_connection(write_db)
    if conn_r is not None:
        reader =conn_r.cursor()
        reader.execute("SELECT question From QuestionsProcessed LIMIT 10")
        print("Questions after preprocessed")
        print('='*100)
        reader.fetchone()
        for row in reader:
            print(row)
            print('-'*100)
        conn_r.commit()
        conn_r.close()
```

Questions after preprocessed

('dynam datagrid bind silverlight dynam datagrid bind silverlight dynam datagrid bind silverlight bind datagrid dynam code wrote code debug code block seem bind correct grid come column form come grid column although necessari bind nthank repli advance..',)

('java.lang.noclassdeffounderror javax servlet jsp tagext taglibraryvalid java.lang.noclassdeffounde rror javax servlet jsp tagext taglibraryvalid java.lang.noclassdeffounderror javax servlet jsp tagext taglibraryvalid follow guid link instal jstl got follow error tri launch jsp page java.lang.noclas sdeffounderror javax servlet jsp tagext taglibraryvalid taglib declar instal jstl 1.1 tomcat webapp tri project work also tri version 1.2 jstl still messag caus solv',)

('java.sql.sqlexcept microsoft odbc driver manag invalid descriptor index java.sql.sqlexcept microsoft odbc driver manag invalid descriptor index java.sql.sqlexcept microsoft odbc driver manag invalid descriptor index use follow code display caus solv',)

('better way updat feed fb php sdk better way updat feed fb php sdk better way updat feed fb php sdk novic facebook api read mani tutori still confused.i find post feed api method like correct second w ay use curl someth like way better',)

('btnadd click event open two window record ad btnadd click event open two window record ad btnadd c lick event open two window record ad open window search.aspx use code hav add button search.aspx nwh

en insert record btnadd click event open anoth window nafter insert record close window',)

('sql inject issu prevent correct form submiss php sql inject issu prevent correct form submiss php sql inject issu prevent correct form submiss php check everyth think make sure input field safe type sql inject good news safe bad news one tag mess form submiss place even touch life figur exact html use templat file forgiv okay entir php script get execut see data post none forum field post problem use someth titl field none data get post current use print post see submit noth work flawless statem ent though also mention script work flawless local machin use host come across problem state list in put test mess',)

('countabl subaddit lebesgu measur countabl subaddit lebesgu measur countabl subaddit lebesgu measur let lbrace rbrace sequenc set sigma -algebra mathcal want show left bigcup right leq sum left right countabl addit measur defin set sigma algebra mathcal think use monoton properti somewher proof star t appreci littl help nthank ad han answer make follow addit construct given han answer clear bigcup bigcup cap emptyset neq left bigcup right left bigcup right sum left right also construct subset mon oton left right leq left right final would sum leq sum result follow',)

('hql equival sql queri hql equival sql queri hql equival sql queri hql queri replac name class prop erti name error occur hql error',)

('undefin symbol architectur i386 objc class skpsmtpmessag referenc error undefin symbol architectur i386 objc class skpsmtpmessag referenc error undefin symbol architectur i386 objc class skpsmtpmessag g referenc error import framework send email applic background import framework i.e skpsmtpmessag so mebodi suggest get error collect2 ld return exit status import framework correct sorc taken framework follow mfmailcomposeviewcontrol question lock field updat answer drag drop folder project click co pi nthat',)

```
In [ ]:
```

```
#Taking 0.5 Million entries to a dataframe.
write_db = 'Titlemoreweight.db'
if os.path.isfile(write_db):
    conn_r = create_connection(write_db)
    if conn_r is not None:
        preprocessed_data = pd.read_sql_query("""SELECT question, Tags FROM QuestionsProcessed""", conn_r)
conn_r.commit()
conn_r.close()
```

In []:

```
preprocessed_data.head()
```

Out[]:

	question	tags
0	dynam datagrid bind silverlight dynam datagrid	c# silverlight data-binding
1	dynam datagrid bind silverlight dynam datagrid	c# silverlight data-binding columns
2	java.lang.noclassdeffounderror javax servlet j	jsp jstl
3	java.sql.sqlexcept microsoft odbc driver manag	java jdbc
4	better way updat feed fb php sdk better way up	facebook ani facebook-php-sdk

In []:

```
print("number of data points in sample :", preprocessed_data.shape[0])
print("number of dimensions :", preprocessed_data.shape[1])
```

```
number of data points in sample : 500000
number of dimensions : 2
```

Converting string Tags to multilable output variables

In []:

```
vectorizer = CountVectorizer(tokenizer = lambda x: x.split(), binary='true')
multilabel_y = vectorizer.fit_transform(preprocessed_data['tags'])
```

Selecting 500 Tags

In []:

```
questions_explained = []
total_tags=multilabel_y.shape[1]
total_qs=preprocessed_data.shape[0]
for i in range(500, total_tags, 100):
    questions_explained.append(np.round(((total_qs-questions_explained_fn(i))/total_qs)*100,3))
```

```
In [ ]:
```

```
fig, ax = plt.subplots()
ax.plot(questions_explained)
xlabel = list(500+np.array(range(-50,450,50))*50)
ax.set_xticklabels(xlabel)
plt.xlabel("Number of tags")
plt.ylabel("Number Questions coverd partially")
plt.grid()
plt.show()
# you can choose any number of tags based on your computing power, minimun is 500(it covers 90% of the tags)
print("with ",5500,"tags we are covering ",questions_explained[50],"% of questions")
print("with ",500,"tags we are covering ",questions_explained[0],"% of questions")
```

with 5500 tags we are covering 99.157 % of questions with 500 tags we are covering 90.956 % of questions

In []:

```
# we will be taking 500 tags
multilabel_yx = tags_to_choose(500)
print("number of questions that are not covered :", questions_explained_fn(500),"out of ", total_qs)
```

number of questions that are not covered : 45221 out of 500000

In []:

```
x_train=preprocessed_data.head(train_datasize)
x_test=preprocessed_data.tail(preprocessed_data.shape[0] - 400000)

y_train = multilabel_yx[0:train_datasize,:]
y_test = multilabel_yx[train_datasize:preprocessed_data.shape[0],:]
```

In []:

```
print("Number of data points in train data :", y_train.shape)
print("Number of data points in test data :", y_test.shape)
```

Number of data points in train data : (400000, 500) Number of data points in test data : (100000, 500)

4.5.2 Featurizing data with Tfldf vectorizer

In []:

Time taken to run this cell: 0:03:52.522389

In []:

```
print("Dimensions of train data X:",x_train_multilabel.shape, "Y :",y_train.shape)
print("Dimensions of test data X:",x_test_multilabel.shape,"Y:",y_test.shape)
```

Diamensions of train data X: $(400000, 94927) \ Y : (400000, 500)$ Diamensions of test data X: $(100000, 94927) \ Y : (100000, 500)$

4.5.3 Applying Logistic Regression with OneVsRest Classifier

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39 40

41

42

43

44

45

0.66

0.65 0.89

0.62

0.71

0.77

0.27

0.49

0.91

0.56

0.68

0.65

0.60 0.75

0.42

0.75

0.29

0.59

0.56

0.71

0.83

0.69

0.96

0.64

0.85

0.62

0.46

0.81

0.80

0.66

0.75

0.18

0.23

0.61

0.23

0.40

0.41

0.07

0.23

0.49

0.29

0.30

0.40

0.32

0.36

0.09

0.18

0.10

0.24

0.18

0.25

0.54

0.21

0.68

0.37

0.29

0.28

0.19

0.51

0.41

0.50

0.32

```
In [ ]:
start = datetime.now()
classifier = OneVsRestClassifier(SGDClassifier(loss='log', alpha=0.00001, penalty='l1'), n_jobs=-1)
classifier.fit(x_train_multilabel, y_train)
predictions = classifier.predict (x_test_multilabel)
print("Accuracy :",metrics.accuracy_score(y_test, predictions))
print("Hamming loss ", metrics.hamming_loss(y_test, predictions))
precision = precision_score(y_test, predictions, average='micro')
recall = recall_score(y_test, predictions, average='micro')
f1 = f1_score(y_test, predictions, average='micro')
print("Micro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1))
precision = precision_score(y_test, predictions, average='macro')
recall = recall_score(y_test, predictions, average='macro')
f1 = f1_score(y_test, predictions, average='macro')
print("Macro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1))
print (metrics.classification_report(y_test, predictions))
print("Time taken to run this cell :", datetime.now() - start)
Accuracy : 0.23623
Hamming loss 0.00278088
Micro-average quality numbers
Precision: 0.7216, Recall: 0.3256, F1-measure: 0.4488
Macro-average quality numbers
Precision: 0.5473, Recall: 0.2572, F1-measure: 0.3339
                          recall f1-score support
             precision
          0
                  0.94
                             0.64
                                       0.76
                                                 5519
          1
                  0.69
                             0.26
                                       0.38
                                                 8190
          2
                  0.81
                             0.37
                                       0.51
                                                 6529
          3
                  0.81
                             0.43
                                       0.56
                                                 3231
                  0.81
                             0.40
                                       0.54
                                                 6430
          5
                  0.82
                             0.33
                                       0.47
                                                 2879
          6
                  0.87
                             0.50
                                       0.63
                                                 5086
          7
                  0.87
                             0.54
                                       0.67
                                                 4533
                  0.60
                             0.13
                                       0.22
                                                 3000
          9
                  0.81
                             0.53
                                       0.64
                                                 2765
         10
                  0.59
                             0.17
                                       0.26
                                                 3051
         11
                  0.70
                             0.33
                                       0.45
                                                 3009
         12
                  0.64
                             0.24
                                       0.35
                                                 2630
         13
                  0.71
                             0.23
                                       0.35
                                                 1426
         14
                  0.90
                             0.53
                                       0.67
                                                 2548
```

0.28

0.34

0.72

0.33

0.51

0.53

0.11

0.31

0.64

0.38

0.42

0.49

0.42

0.48

0.15

0.29

0.14

0.35

0.27

0.37

0.66

0.32

0.79

0.47

0.43

0.38

0.27

0.63

0.54

0.57

0.45

2371

873

2151

2204

831

1860

2023

1513

1207

506

425

793

1291

1208

406

504

732

441

1645

1058

946

644

136

570

766

1132

174

210

433

626

852

46 47 48 49 50 51 52 53 54 55 56 67 68 69 70 71 72 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90 91 102 103 104 105 106 107 108 109 100 101 101 102 103 104 105 106 107 108 109 100 101 101 102 103 104 105 106 107 108 109 100 101 101 102 103 104 105 106 107 108 109 100 101 102 103 104 105 106 107 108 109 100 101 102 103 104 105 106 107 108 109 109 100 101 102 103 104 105 106 107 108 109 100 100 100 100 100 100 100
0.75 0.34 0.74 0.79 0.16 0.33 0.50 0.68 0.45 0.31 0.46 0.47 0.78 0.94 0.34 0.83 0.91 0.83 0.55 0.36 0.29 0.76 0.82 0.67 0.389 0.79 0.72 0.53 0.78 0.16 0.77 0.55 0.63 0.79 0.72 0.53 0.78 0.16 0.77 0.55 0.67 0.48 0.48 0.43 0.28 0.55 0.63 0.76 0.69 0.31 0.34 0.26 0.90 0.76 0.54 0.55 0.93 0.90 0.77 0.57 0.58 0.54 0.77 0.57 0.58 0.54 0.77 0.75 0.58 0.54 0.77 0.75 0.58 0.54 0.77 0.75 0.58 0.54 0.77 0.75 0.78 0.50 0.77 0.77 0.77 0.77 0.77 0.77 0.77
0.42 0.14 0.51 0.62 0.04 0.10 0.04 0.10 0.02 0.07 0.09 0.27 0.63 0.11 0.63 0.11 0.63 0.11 0.63 0.11 0.63 0.15 0.45 0.45 0.25 0.45 0.25 0.45 0.25 0.45 0.30 0.25 0.45 0.30 0.25 0.45 0.30 0.25 0.45 0.30 0.25 0.45 0.30 0.25 0.45 0.30 0.27 0.30 0.31 0.32 0.33 0.31 0.33 0.45 0.33 0.45 0.35 0.45 0.36 0.37 0.37 0.38 0.39 0.49 0.49 0.49 0.49 0.49 0.69 0.71 0.68 0.69 0.71 0.68 0.69 0.71 0.68 0.71 0.71 0.72 0.73 0.73 0.74 0.75
0.54 0.60 0.70 0.60 0.70 0.15 0.07 0.15 0.17
534 350 496 785 251 914 728 258 821 541 748 724 660 235 718 468 191 429 415 274 401 86 129 473 143 347 479 279 461 298 396 184 573 325 273 135 232 409 408 241 277 410 501 136 239 324 277 613 325 325 325 325 327 327 327 328 329 329 329 329 329 329 329 329

129 130 131 132 133 134 135 136 137 138 139 140 141 142 143 144 145 146 147 148 149 150 151 152 153 154 165 167 168 169 171 173 174 175 176 177 178 189 181 182 183 184 185 1867 189 191 192 193 194 195 196 191 192 193 194 195 196 191 192 193 194 195 196 197 198 199 191 192 193 194 195 196 197 198 199 191 192 193 194 195 196 197 198 199 191 192 193 194 195 196 197 198 199 191 192 193 194 195 196 197 198 199 190 191 192 193 194 195 196 197 198 199 190 191 192 193 194 195 196 197 198 199 190 191 192 193 194 195 196 197 198 199 190 191 191 192 193 194 195 196 197 198 199 190 191 191 192 193 194 195 196 197 198 199 190 190 190 190 190 190 190 190 190
0.45 0.44 0.66 0.94 0.89 0.31 0.68 0.57 0.77 0.39 0.66 0.59 0.65 0.65 0.66 0.98 0.62 0.98 0.62 0.98 0.62 0.98 0.60 0.76 0.29 0.60 0.70 0.60 0.70
0.10 0.08 0.01 0.26 0.54 0.72 0.047 0.15 0.18 0.17 0.18 0.17 0.18 0.17 0.18 0.19 0.19 0.10 0.19 0.10
0.16 0.14 0.02 0.37 0.69 0.79 0.08 0.25 0.16 0.25 0.16 0.47 0.26 0.47 0.19 0.62 0.14 0.70 0.14 0.70 0.14 0.70 0.34 0.17 0.34 0.26 0.37 0.34 0.19 0.68 0.17 0.34 0.26 0.37 0.34 0.31 0.31 0.32 0.31 0.32 0.31 0.32 0.31 0.32 0.32 0.33 0.34 0.35 0.36 0.37 0.36 0.37 0.37 0.38 0.39 0.39 0.39 0.39 0.39 0.39 0.39 0.39
144 150 210 361 453 124 91 128 218 243 149 318 159 274 362 118 164 461 159 166 346 350 55 387 150 281 202 130 245 177 130 336 220 229 316 283 197 101 231 370 258 101 89 193 309 172 95 346 322 232 125 145 77 182 257 216 242 165 263 174 136 202 134 230 90 185 156 160 266 284 136 297 101 291 202 125 145 77 182 257 216 242 232 125 145 77 182 257 216 242 165 263 174 136 202 134 230 90 185 156 160 266 284 136 297 193 220 140

295 297 298 299 300 301 302 303 304 305 307 308 309 311 313 313 314 315 317 318 319 321 322 323 324 327 328 329 331 331 331 331 331 331 331 331 331 33
0.24 0.22 0.77 0.93 0.18 0.00 0.57 0.96 0.95 0.95 0.29 0.00 0.78 0.48 0.75 0.50 0.48 0.50 0.45 0.45 0.45 0.45 0.45 0.45 0.45
0.08 0.08 0.08 0.09 0.09 0.00 0.35 0.00 0.35 0.00 0.35 0.00 0.35 0.00 0.35 0.00 0.35 0.00 0.35 0.36 0.37 0.35 0.36 0.37 0.37 0.37 0.38 0.37 0.37 0.38 0.39 0.30 0.31 0.31 0.32 0.34 0.35 0.36 0.37 0.37 0.38 0.37 0.38 0.37 0.38 0.39 0.31 0.31 0.31 0.32 0.34 0.35 0.36 0.37 0.38 0.37 0.38 0.38 0.38 0.39 0.39 0.30 0.31 0.31 0.32 0.34 0.35 0.36 0.37 0.38
0.12 0.12 0.06 0.51 0.09 0.00 0.44 0.74 0.13 0.00 0.14 0.34 0.34 0.34 0.34 0.35 0.30 0.15 0.40 0.58 0.17 0.62 0.30 0.15 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.17 0.05 0.05 0.05 0.05 0.05 0.05 0.05 0.0
241 72 107 61 77 111 126 73 176 230 156 146 98 78 94 162 116 57 65 138 195 69 134 148 161 104 156 134 232 92 197 126 115 198 125 81 94 56 260 60 110 71 66 150 54 195 79 38 43 68 73 116 111 63 104 44 40 136 54 131 63 104 44 120 228 269 80 140 136 54 136 56 154 171 57 186 186 78 78 186 78 78 186 78 78 78 78 78 78 78 78 78 78 78 78 78

378 379 381 382 383 384 385 387 389 391 393 394 401 402 403 404 405 407 408 409 401 407 408 409 409 409 409 409 409 409 409 409 409
0.27 0.33 1.00 0.19 0.28 0.50 0.36 0.50 0.36 0.59 0.62 0.62 0.78 0.62 0.78 0.64 0.73 0.64 0.73 0.64 0.73 0.64 0.73 0.64 0.90 0.46 0.37 0.64 0.90 0.47 0.46 0.57 0.47 0.47 0.47 0.48 0.90 0.61 0.90 0.62 0.95 0.95 0.95 0.63 0.95 0.95 0.95 0.95 0.95 0.95 0.95 0.95
0.03 0.07 0.40 0.03 0.08 0.08 0.08 0.09 0.15 0.09 0.15 0.09 0.11 0.09 0.11 0.09 0.11 0.09 0.11 0.09 0.11 0.09 0.11 0.09 0.11 0.09 0.09
0.06 0.12 0.57 0.05 0.12 0.36 0.09 0.38 0.02 0.78 0.27 0.00 0.17 0.16 0.33 0.09 0.37 0.17 0.10 0.17 0.10 0.17 0.10 0.17 0.10
86 14 122 104 66 110 155 50 64 93 102 108 178 115 42 134 112 176 125 224 63 59 63 98 162 83 19 92 41 43 160 50 19 175 72 95 97 48 83 40 91 90 37 66 73 56 33 76 81 150 29 389 167 123 39 82 66 93 87 86 104 100 123 71 109 48 76 38 81 132 66 93 87 86 104 100 123 71 109 48 76 38 81 132 66 93 87 86 104 100 123 71 109 48 76 38 81 132 66 93 87 86 104 100 123 71 109 48 76 38 81 132 66 93 87 86 104 100 123 71 109 48 76 38 81 132 66 93 87 86 104 100 123 71 109 48 76 38 81 132 66 93 87 86 104 100 123 71 109 48 76 38 119

```
461
                    0.79
                               0.14
                                           0.24
                                                        79
        462
                    0.69
                               0.23
                                           0.35
                                                        47
        463
                    0.20
                               0.04
                                           0.06
                                                       104
        464
                    0.66
                               0.33
                                           0.44
                                                       106
        465
                    0.50
                               0.11
                                           0.18
                                                        64
        466
                    0.56
                               0.28
                                           0.37
                                                       173
        467
                    0.81
                               0.36
                                          0.50
                                                       107
        468
                    0.82
                               0.11
                                          0.20
                                                       126
        469
                    0.00
                               0.00
                                           0.00
                                                       114
        470
                    0.94
                               0.79
                                           0.86
                                                       140
        471
                    0.92
                               0.28
                                           0.43
                                                        79
        472
                    0.41
                               0.30
                                           0.35
                                                       143
        473
                    0.69
                               0.30
                                           0.42
                                                       158
        474
                    0.36
                               0.07
                                          0.11
                                                       138
        475
                    0.00
                               0.00
                                          0.00
                                                        59
        476
                    0.57
                               0.30
                                          0.39
                                                        88
        477
                                                       176
                    0.86
                               0.56
                                           0.68
                    0.94
        478
                                          0.81
                                                        24
                               0.71
        479
                    0.09
                               0.01
                                           0.02
                                                        92
                               0.50
                                                       100
        480
                    0.82
                                          0.62
        481
                    0.47
                               0.17
                                           0.26
                                                       103
        482
                    0.47
                                                        74
                               0.23
                                          0.31
                                                       105
        483
                    0.85
                               0.57
                                           0.68
        484
                               0.02
                                          0.04
                                                        83
                    0.25
        485
                    0.17
                               0.01
                                           0.02
                                                        82
        486
                                                        71
                    0.36
                                          0.17
                               0.11
        487
                    0.43
                               0.18
                                           0.26
                                                       120
        488
                    0.33
                               0.02
                                          0.04
                                                       105
        489
                    0.72
                               0.30
                                           0.42
                                                        87
                    1.00
        490
                               0.81
                                          0.90
                                                        32
        491
                    0.00
                                          0.00
                                                        69
                               0.00
        492
                                          0.00
                                                        49
                    0.00
                               0.00
        493
                    0.00
                               0.00
                                           0.00
                                                       117
        494
                                          0.27
                    0.52
                               0.18
                                                        61
        495
                    0.98
                               0.65
                                           0.78
                                                       344
                                          0.25
        496
                    0.36
                               0.19
                                                        52
        497
                    0.60
                               0.18
                                           0.28
                                                       137
        498
                    0.33
                               0.04
                                          0.07
                                                        98
        499
                    0.65
                               0.16
                                           0.26
                                                        79
avg / total
                    0.67
                               0.33
                                           0.43
                                                    173812
```

Time taken to run this cell: 0:10:14.264591

In []:

```
joblib.dump(classifier, 'lr_with_more_title_weight.pkl')
```

Out[]:

['lr_with_more_title_weight.pkl']

In []:

```
start = datetime.now()
classifier_2 = OneVsRestClassifier(LogisticRegression(penalty='l1'), n_jobs=-1)
classifier 2.fit(x_train_multilabel, y_train)
predictions_2 = classifier_2.predict(x_test_multilabel)
print("Accuracy :",metrics.accuracy score(y test, predictions 2))
print("Hamming loss ",metrics.hamming_loss(y_test,predictions_2))
precision = precision_score(y_test, predictions_2, average='micro')
recall = recall score(y test, predictions 2, average='micro')
f1 = f1_score(y_test, predictions_2, average='micro')
print("Micro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1))
precision = precision_score(y_test, predictions_2, average='macro')
recall = recall score(y test, predictions 2, average='macro')
f1 = f1_score(y_test, predictions_2, average='macro')
print("Macro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1))
print (metrics.classification_report(y_test, predictions_2))
print("Time taken to run this cell :", datetime.now() - start)
```

Accuracy: 0.25108
Hamming loss 0.00270302
Micro-average quality numbers
Precision: 0.7172, Recall: 0.3672, F1-measure: 0.4858
Macro-average quality numbers

Precision: 0.	5570, Recall precision			: 0.3710 support
0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 40 40 41 42 43 44 44 45 46 47 48 48 49 40 40 40 40 40 40 40 40 40 40 40 40 40	0.94 0.70 0.80 0.82 0.80 0.82 0.86 0.87 0.60 0.62 0.65 0.65 0.65 0.65 0.65 0.65 0.65 0.65 0.65 0.69 0.60 0.71 0.76 0.29 0.52 0.89 0.60 0.71 0.76 0.29 0.52 0.89 0.65 0.69 0.69 0.69 0.69 0.69 0.70	necall 0.72 0.34 0.42 0.49 0.44 0.38 0.53 0.58 0.13 0.57 0.20 0.38 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.25 0.63 0.31 0.35 0.10 0.21 0.08 0.29 0.27 0.26 0.58 0.31 0.35 0.10 0.21 0.08 0.29 0.27 0.26 0.58 0.31 0.35 0.18 0.49 0.42 0.52 0.64 0.65 0.38 0.31 0.35 0.18 0.49 0.42 0.52 0.64 0.06 0.13 0.03	f1-score 0.82 0.45 0.55 0.61 0.57 0.52 0.66 0.70 0.22 0.67 0.30 0.49 0.40 0.43 0.70 0.34 0.37 0.74 0.35 0.52 0.58 0.14 0.33 0.68 0.38 0.45 0.52 0.47 0.51 0.17 0.33 0.12 0.39 0.38 0.45 0.52 0.47 0.51 0.17 0.33 0.12 0.39 0.38 0.45 0.52 0.47 0.51 0.17 0.33 0.12 0.39 0.38 0.45 0.52 0.47 0.51 0.17 0.33 0.12 0.39 0.38 0.45 0.52 0.47 0.51 0.17 0.33 0.12 0.39 0.38 0.38 0.68 0.35 0.78 0.47 0.55 0.78 0.47 0.55 0.78 0.47 0.57 0.22 0.62 0.71 0.09 0.19 0.06	5519 8190 6529 3231 6430 2879 5086 4533 3000 2765 3051 3009 2630 1426 2548 2371 873 2151 2204 831 1860 2023 1513 1207 506 425 793 1291 1208 406 504 732 441 1645 1058 946 644 136 570 766 1132 174 210 433 626 852 534 350 496 785 475 305 251
48 49 50 51	0.75 0.78 0.21 0.37	0.52 0.64 0.06 0.13	0.62 0.71 0.09 0.19	496 785 475 305

80 81 82 83 84 85 86 87 88 99 192 93 94 95 96 97 98 99 100 101 102 103 104 105 107 118 119 110 111 112 113 114 115 116 117 118 119 121 121 121 121 121 131 131 131 131 131
0.56 0.70 0.34 0.78 0.55 0.61 0.50 0.51 0.40 0.55 0.66 0.30 0.37 0.28 0.30 0.37 0.28 0.75 0.91 0.47 0.91 0.47 0.58 0.48 0.59 0.69 0.77 0.58 0.69 0.77 0.69 0.77 0.69 0.77 0.69 0.77 0.69 0.77 0.69 0.77
0.34 0.27 0.29 0.29 0.21 0.27 0.29 0.21 0.27 0.37 0.44 0.10 0.37 0.43 0.43 0.43 0.43 0.43 0.43 0.43 0.43 0.43 0.44 0.43 0.45 0.45 0.45 0.45 0.47 0.47 0.48 0.48 0.49
0.43 0.34 0.61 0.38 0.37 0.38 0.37
279 461 298 396 184 573 325 273 135 232 409 420 408 241 211 277 410 501 136 239 324 277 613 157 295 334 370 313 874 293 344 370 313 874 293 349 251 187 140 154 332 323 344 370 313 874 293 361 453 129 252 144 150 210 361 453 124 91 128 243 149 318 159 274 362 118 164 461 159 166 346 350 557 377 150 210 361 453 124 91 128 218 243 149 318 159 274 362 118 164 461 159 166 346 350 557 150 201 301 202 202 202

412 413 414 415 416 417 418 420 421 422 423 424 425 427 428 430 431 432 433 434 435 437 438 439 441 442 453 454 457 458 459 461 462 463 464 465 467 477 478 479 471 477 478 479 471 477 478 477 478 479 471 477 478 479 471 477 478 479 471 477 478 479 471 477 478 479 471 477 478 479 471 477 478 479 479 471 477 478 479 479 479 479 479 479 479 479 479 479
0.28 0.38 0.12 0.33 0.43 0.48 0.53 0.48 0.69 0.48 0.94 0.00 0.27 0.98 0.99 0.63 0.57 0.33 0.57 0.32 0.47 0.37 0.32 0.47 0.39 0.67 0.42 0.89 0.00 0.71 0.63 0.70 0.63 0.70 0.63 0.70
0.07 0.05 0.02 0.10 0.35 0.07 0.15 0.07 0.16 0.37 0.45 0.05 0.05 0.05 0.05 0.05 0.05 0.05
0.11 0.09 0.04 0.16 0.42 0.13 0.25 0.32 0.55 0.31 0.00 0.84 0.96 0.49 0.18 0.25 0.49 0.25 0.49 0.25 0.49 0.25 0.49 0.25 0.37 0.28 0.37 0.28 0.37 0.00 0.37 0.28 0.37 0.09 0.37 0.09 0.37 0.37 0.37 0.37 0.37 0.37 0.37 0.37
72 95 97 48 83 40 91 90 37 66 73 56 33 76 81 150 29 389 167 123 39 82 66 93 87 86 104 100 141 110 123 71 109 48 76 38 81 132 81 76 44 44 70 155 43 71 104 104 119 105 119 106 119 107 107 108 108 108 109 109 109 109 109 109 109 109 109 109

0.87	0.79	0.95	495
0.19	0.13	0.32	496
0.38	0.28	0.59	497
0.15	0.10	0.31	498
0.29	0.20	0.48	499
0.46	0.37	0.67	avg / total
	0.19 0.38 0.15 0.29	0.13 0.19 0.28 0.38 0.10 0.15 0.20 0.29	0.32 0.13 0.19 0.59 0.28 0.38 0.31 0.10 0.15 0.48 0.20 0.29

Time taken to run this cell: 1:09:41.236859