Stackoverflow tag(s) predictor

In [1]:

```
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. Business Problem

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 Diqq. 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

(https://www.microsoft.com/en-us/research/wp-content/uploads/2016/02/tagging-1.pdf)

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)

1.3 Real World / Business Objectives and Constraints

- 1. Predict as many tags as possible with high precision and recall.
- 2. Incorrect tags could impact customer experience on StackOverflow.
- 3. No strict latency constraints.

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/facebook-recruiting-iii-keyword-extraction/data)

All of the data is in 2 files: Train and Test.

```
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
```

Size of Test.csv - 2GB

Number of rows in Train.csv = 6034195

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:

```
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 lowe rcase, 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 variable</pre>
s";\n
                  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
                  \{ 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
                     {\n
                         if(1!=1)\n
                         {\n
                             cout<<a[1]<<"\\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
                         {\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\n
       <code>
       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
                                    100\n
       50
                    50
       </code>\n\n
       if the no of inputs is 3 and their ranges are\n
       1,100\n
       1,100\n
       1,100\n
       (could be varied too)\n\n
       The output is not coming, can anyone correct the code or tell me what
   \'s wrong?\n'
Tags : 'c++ c'
```

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

Multi-label Classification: Multilabel classification assigns to each sample a set of target labels. This can be thought as predicting properties of a data-point 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

3. Exploratory Data Analysis

3.1 Data Loading and Cleaning

3.1.1 Using Pandas with SQLite to Load the data

In [2]:

```
#Creating db file from csv
#Learn SQL: https://www.w3schools.com/sql/default.asp
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=chu
        df.index += index_start
        j+=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 [3]:
```

```
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 to)
```

```
Number of rows in the database : 6034196
Time taken to count the number of rows : 0:00:04.153227
```

In [4]:

```
num_rows
```

Out[4]:

count(*)

0 6034196

3.1.3 Checking for duplicates

In [5]:

```
#Learn SQL: https://www.w3schools.com/sql/default.asp
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
    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.
```

Time taken to run this cell: 0:05:25.691910

In [6]:

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

Out[6]:

	Title	Body	Tags	cnt_dup
0	Implementing Boundary Value Analysis of S	<pre><pre><code>#include&Itiostream>\n#include&</code></pre></pre>	c++ c	1
1	Dynamic Datagrid Binding in Silverlight?	I should do binding for datagrid dynamicall	c# silverlight data- binding	1
2	Dynamic Datagrid Binding in Silverlight?	I should do binding for datagrid dynamicall	c# silverlight data- binding columns	1
3	java.lang.NoClassDefFoundError: javax/serv	I followed the guide in <a href="http://sta</a 	jsp jstl	1
4	java.sql.SQLException:[Microsoft] [ODBC Dri	I use the following code\n\n <pre><code></code></pre>	java jdbc	2

In [7]:

```
print("number of duplicate questions :", num_rows['count(*)'].values[0]- df_no_dup.shape[0]
```

number of duplicate questions : 1827881 (30.292038906260256 %)

In [8]:

```
# number of times each question appeared in our database
df_no_dup.cnt_dup.value_counts()
```

Out[8]:

```
1 2656284
2 1272336
3 277575
4 90
5 25
6 5
```

Name: cnt_dup, dtype: int64

3.1.4 Removing the rows where no tags are present

In [13]:

```
sd=[]
start = datetime.now()
for i in range(df_no_dup.shape[0]):
    f=df_no_dup["Tags"][i]# no of characters==0
    if f==None:# when no tag given just remove that datapoint
        df_no_dup=df_no_dup.drop(i,axis=0) # remove this datapoint
    else:
        d=len(df_no_dup["Tags"][i].split(" "))
        sd.append(d)

print(datetime.now()-start)
```

0:03:36.320016

In [14]:

```
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: 0:00:02.329769

Out[14]:

	Title	Body	Tags	cnt_dup	t
0	Implementing Boundary Value Analysis of S	<pre><pre><code>#include&Itiostream>\n#include&</code></pre></pre>	c++ c	1	_
1	Dynamic Datagrid Binding in Silverlight?	I should do binding for datagrid dynamicall	c# silverlight data- binding	1	
2	Dynamic Datagrid Binding in Silverlight?	I should do binding for datagrid dynamicall	c# silverlight data- binding columns	1	
3	java.lang.NoClassDefFoundError: javax/serv	I followed the guide in			

```
In [15]:
```

```
# distribution of number of tags per question
df_no_dup.tag_count.value_counts()

Out[15]:
3    1206157
2    1111706
4    814996
1    568291
```

Creating no Duplicate database

Name: tag_count, dtype: int64

In [16]:

505158

5

```
#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 = pd.DataFrame(df_no_dup, columns=['Title', 'Body', 'Tags'])
    no_dup.to_sql('no_dup_train',disk_dup)
```

In [17]:

```
#This method seems more appropriate to work with this much data.
#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)
    #Always remember to close the database
    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
```

Time taken to run this cell: 0:00:16.116694

3.2 Analysis of Tags

3.2.1 Total number of unique tags

```
In [18]:
```

```
# 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'])
#Returns: X: array, [n_samples, n_features]
```

```
In [19]:
tag_data['Tags']
Out[19]:
                   c# silverlight data-binding
1
2
           c# silverlight data-binding columns
3
                                       isp istl
4
                                      java jdbc
5
                 facebook api facebook-php-sdk
4206310
                    wordpress wordpress-plugin
4206311
                                 php mysql text
4206312
            php codeigniter character-encoding
4206313
                        php email outlook mime
4206314
Name: Tags, Length: 4206314, dtype: object
In [20]:
print("Number of data points :", tag_dtm.shape[0])
print("Number of unique tags :", tag_dtm.shape[1])
Number of data points : 4206314
Number of unique tags: 42048
In [21]:
#'get_feature_name()' gives us the vocabulary.
tags = vectorizer.get feature names()
#Lets look at the tags we have.
print("Some of the tags we have :", tags[:10])
```

Some of the tags we have : ['.a', '.app', '.asp.net-mvc', '.aspxauth', '.bas

h-profile', '.class-file', '.cs-file', '.doc', '.drv', '.ds-store']

```
3.2.3 Number of times a tag appeared
```

In [22]:

```
# 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))
```

In [23]:

```
result
Out[23]:
{'.a': 18,
 '.app': 37,
 '.asp.net-mvc': 1,
 '.aspxauth': 21,
 '.bash-profile': 138,
 '.class-file': 53,
 '.cs-file': 14,
 '.doc': 47,
 '.drv': 1,
 '.ds-store': 8,
 '.each': 184,
 '.emf': 33,
 '.exe': 27,
 '.exe.config': 1,
 '.hgtags': 6,
 '.htaccess': 14884,
 '.htpasswd': 61,
 '.ico': 10.
In [24]:
```

#Savina this dictionary to csy files

```
#Saving this dictionary to csv files.
if not os.path.isfile('data/tag_counts_dict_dtm.csv'):
    with open('data/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("data/tag_counts_dict_dtm.csv", names=['Tags', 'Counts'])
tag_df.head()
```

Out[24]:

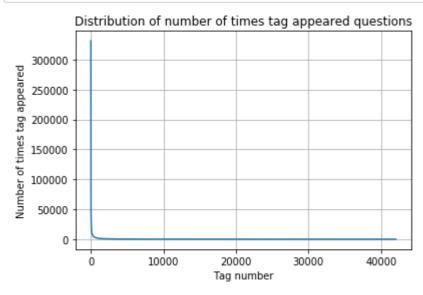
	Tags	Counts
0	.a	18
1	.арр	37
2	.asp.net-mvc	1
3	.aspxauth	21
4	.bash-profile	138

In [25]:

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

In [26]:

```
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()
```

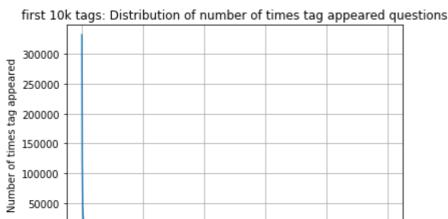


In [27]:

0

2000

```
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])
```



4000

Tag number

6000

8000

10000

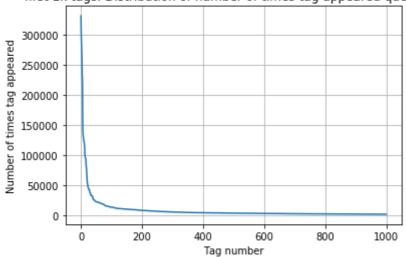
400	[3315	05 4 4	1829 22	2429 1		.3364 1	1162 1	0029	9148	8054	7151
6	5466	5865	5370	4983	4526	4281	4144	3929	3750	3593	
3	3453	3299	3123	2989	2891	. 2738	2647	2527	2431	2331	
2	2259	2186	2097	2020	1959			1770	1723	1673	
1	L631	1574	1532	1479	1448	1406	1365	1328	1300	1266	
1	L245	1222	1197		1158						
1	L038	1023		983							
		869		841							
	752	743	733	725	712					658	
	650	643	634								
	568	559		545							
	500	495		485							
	447	442									
	398	393		385							
	361	357		350							
	330	326	323	319							
	299	296	293	291							
	275	272		268							
	252	250	249								
	234	233	232								
	217	215	214	212							
	201	200	199	198							
	188	186	185	183							
	175	174		171							
	164	162		160							
	154	153		151							
	145	144		142							
	137	136	135	134							
	129	128		127							
	123	122									
	117	116		115							
	111	110					107				
	105	105	104	104							
	100	100	99	99	98	98	97	97	96	96	
alhoet	-8888/not	ehooke/F	Jownloade/	5 Stackov	orflow/SO	Tan Predict	or invnh				

95	95	94	94	93	93	93	92	92	91
91	90		89						
86	86	85							
		81		-	_				_
78		78							
75	74	74	74	73	73	73	73	72	72]

In [28]:

```
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])
```

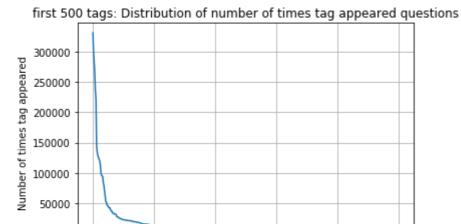




200 [331	505 221	533 122	769 95	160 62	023 44	829 37	170 31	897 26	925 24537
22429	21820		19758	18905	17728	15533	15097	14884	13703
13364	13157	12407	11658	11228	11162	10863	10600	10350	10224
		_		_	_				
10029	9884	9719	9411	9252	9148	9040	8617	8361	8163
8054	7867	7702	7564	7274	7151	7052	6847	6656	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
3453	3427	3396	3363	3326	3299	3272	3232	3196	3168
3123	3094	3073	3050	3012	2989	2984	2953	2934	2903
2891	2844	2819	2784	2754	2738	2726	2708	2681	2669
2647	2621	2604	2594	2556	2527	2510	2482	2460	2444
2431	2409	2395	2380	2363	2331	2312	2297	2290	2281
2259	2246	2222	2211	2198	2186	2162	2142	2132	2107
2097	2078	2057	2045	2036	2020	2011	1994	1971	1965
1959	1952	1940	1932	1912	1900	1879	1865	1855	1841
1828	1821	1813	1801	1782	1770	1760	1747	1741	1734
1723	1707	1697	1688	1683	1673	1665	1656	1646	1639]

In [29]:

```
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])
```



Tag number

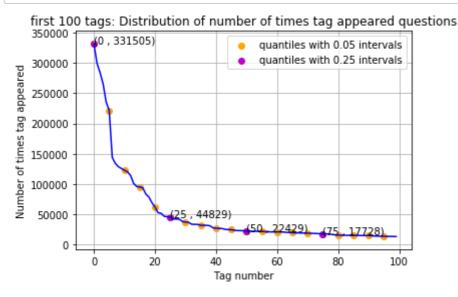
```
62023 44829
100 [331505 221533 122769 95160
                                                    37170
                                                            31897
                                                                    26925
                                                      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

In [30]:

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

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]

In [31]:

```
# 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 [32]:

```
#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 ar
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]

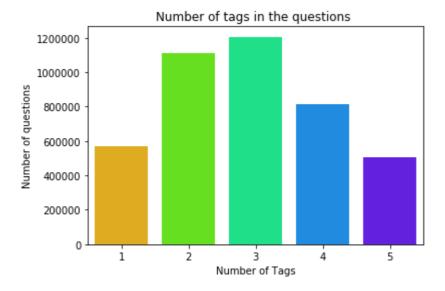
In [33]:

```
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

In [34]:

```
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()
```



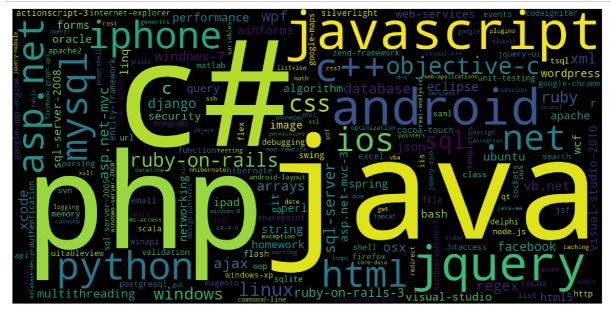
Observations:

- 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 [35]:

```
# 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.
wordcloud = WordCloud(
                          background_color='black',
                          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: 0:00:03.085306

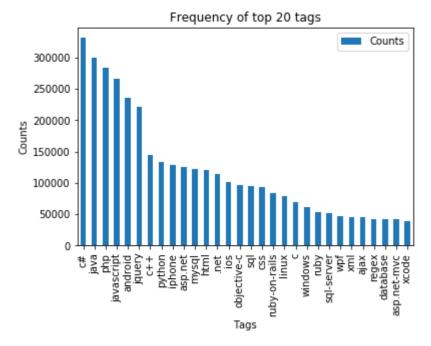
Observations:

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

In [36]:

```
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 0.5M 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

In [37]:

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

In [38]:

```
#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
    try:
        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)
    else:
        print("Error! cannot create the database connection.")
    conn.close()
sql_create_table = """CREATE TABLE IF NOT EXISTS QuestionsProcessed (question text NOT NULL
create database table("Processed.db", sql create table)
                                                                                           Þ
```

Tables in the databse: OuestionsProcessed

```
In [39]:
```

```
# http://www.sqlitetutorial.net/sqlite-delete/
# 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
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)
```

Tables in the databse: QuestionsProcessed Cleared All the rows Time taken to run this cell: 0:03:32.619944

__ we create a new data base to store the sampled and preprocessed questions __

In [40]:

```
import nltk
nltk.download('punkt')
[nltk_data] Downloading package punkt to
```

Out[40]:

True

In [41]:

```
#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:
        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
    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,
    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 pr
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
Avg. length of questions(Title+Body) before processing: 1171
Avg. length of questions(Title+Body) after processing: 326
```

Percent of questions containing code: 57 Time taken to run this cell: 0:09:45.271985



In [42]:

```
# dont forget to close the connections, or else you will end up with locks
conn_r.commit()
conn_w.commit()
conn_r.close()
conn_w.close()
```

In [43]:

```
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

('sql convers script delet bad data permiss bridg tabl goal write loop convers go contain acucor securitycontain tabl clean permiss acucor securitypermiss bridg tabl base script care end two script complet fine notic differ two clean queri vs saliva vs hair queri hair applic context secur queri saliva a pplic context secur',)

('combin result multipl subqueri one set tri exclud row tabl base id tabl ni tabl select result set like tri combin result subqueri one like filter big t abl look group concat union kind seem find someth actual work anyon idea',)

('version percona db packag instal ubuntu natti narwhal new percona amp ubuntu accord page percona avail use natti someon recommend version choos',)

('hide div view mvc develop mvc applic use razor syntax applic give comment facil ad partial view load comment record db imag see comment box call run t ime employe index view want remov div user click delet button work right ple as check imag code applic updat script',)

('sudo anoth user without specifi usernam current tri creat sudoer file ran someth figur end result look want user abl someth like instead run root like script run user look sudoer file tri ad line like work specifi user easi way script run specif user without specifi command line',)

('maximum size ellips net attempt use object draw ellips realli circl larg m illion pixel though sinc pictur box draw reason size ought get either imag s ometh approach line depend ellips locat use system draw graphic drawellips p en rectanglef draw ellips success construct rectanglef describ dimens want e xampl describ ellips much bigger actual visibl view space pass drawelips act ual noth sinc part edg elips intersect visibl region instead tri draw ellips bound produc method specifi ever throw microsoft document give indic inappro pri argument big ellips draw far away center break',)

('move new line haskel updat follow function haskel must print sale week sal e new line work way expect problem newlin charact code tri mani way new line charact work expect everyth print line whichi want need help thank updat fol low work compil error error come second line formatlin type decalar caus err

```
('verita netbackup nbazd daemon wont start im new work verita netbackup inst
al server solari im plan use nbac problem nbazd daemon dont start go activ m
onitor daemon right click nbazd daemon mark start daemon dont anyth select d
aemon detail nbazd status stop start servic',)

('revers dns lookup command line like get list domain point certain ip addre
ss way get inform command line noth like host nslookup dig return hostnam ip
address help part want return edit inform nan exampl websit return inform ht
tp www domaintool com revers ip hostnam',)
```

In [44]:

```
#Taking 0.5 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 QuestionsProces
conn_r.commit()
conn_r.close()
```

In [45]:

```
preprocessed_data.head()
```

Out[45]:

tags	question	
java synchronization	determin object lock synchron block java proce	0
sql-server-2008 tsql delete script	sql convers script delet bad data permiss brid	1
mysql set subquery	combin result multipl subqueri one set tri exc	2
mysql ruby ubuntu natty percona	version percona db packag instal ubuntu natti	3
javascript asp.net-mvc-3	hide div view mvc develop mvc applic use razor	4

In [46]:

```
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 : 499999 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 [47]:

```
# 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 [48]:

```
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))
```

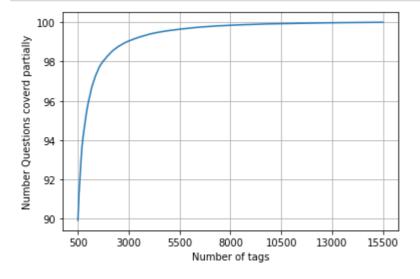
In [49]:

```
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
```

In [50]:

```
questions_explained
Out[50]:
[89.927,
91.247,
92.125,
92.974,
93.673,
 94.148,
 94.535,
 94.892,
 95.279,
95.592,
95.87,
 96.115,
96.324,
96.571,
96.765,
96.915,
97.074,
97.242.
In [51]:
```

```
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
print("with ",5500,"tags we are covering ",questions_explained[50],"% of questions")
```



with 5500 tags we are covering 99.048 % of questions

```
In [52]:
```

```
multilabel_yx = tags_to_choose(5500)
print("number of questions that are not covered :", questions_explained_fn(5500),"out of ",
number of questions that are not covered : 4760 out of 499999
```

In [53]:

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

```
Number of tags in sample : 30508
number of tags taken : 5500 ( 18.028058214238886 %)
```

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

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

In [54]:

```
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 [55]:

```
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: (399999, 5500)
Number of data points in test data: (100000, 5500)
```

4.3 Featurizing data

In [56]:

Time taken to run this cell: 0:02:36.205814

```
In [57]:
```

```
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)
Dimensions of train data X: (399999, 89487) Y: (399999, 5500)
Dimensions of test data X: (100000, 89487) Y: (100000, 5500)
In [58]:
# https://www.analyticsvidhya.com/bloq/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))
.....
# 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[58]:
"\nfrom skmultilearn.adapt import MLkNN\nclassifier = MLkNN(k=21)\n\n# train
\nclassifier.fit(x_train_multilabel, y_train)\n\n# predict\npredictions = cl
assifier.predict(x_test_multilabel)\nprint(accuracy_score(y_test,prediction
s))\nprint(metrics.f1_score(y_test, predictions, average = 'macro'))\nprint
(metrics.f1_score(y_test, predictions, average = 'micro'))\nprint(metrics.ha
mming_loss(y_test,predictions))\n\n"
In [59]:
x train multilabel.shape
Out[59]:
(399999, 89487)
In [60]:
x test multilabel.shape
Out[60]:
```

(100000, 89487)

4.4 Applying Logistic Regression with OneVsRest Classifier

```
In [64]:
```

```
classifier = OneVsRestClassifier(SGDClassifier(loss='log', alpha=0.00001, penalty='l1'), n
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 : 0.08126
macro f1 score : 0.09291293571750123
micro f1 scoore : 0.37398320216916875
hamming loss: 0.00041306181818182
Precision recall report :
               precision
                            recall f1-score
                                                support
           0
                   0.61
                             0.22
                                        0.32
                                                  7927
                   0.79
                             0.44
                                        0.56
           1
                                                  7236
           2
                   0.82
                             0.54
                                        0.65
                                                  6682
           3
                   0.74
                             0.44
                                        0.55
                                                  6298
           4
                   0.95
                             0.76
                                        0.84
                                                  5729
           5
                                        0.75
                   0.87
                             0.67
                                                  5244
           6
                   0.73
                             0.31
                                        0.43
                                                  3520
           7
                   0.88
                             0.60
                                        0.71
                                                  3150
           8
                   0.69
                             0.38
                                        0.49
                                                  3063
           9
                   0.79
                             0.41
                                        0.54
                                                  3055
          10
                   0.84
                             0.60
                                        0.70
                                                  2893
          11
                   0.56
                             0.18
                                        0.27
                                                  2889
```

In [65]:

```
from sklearn.externals import joblib
joblib.dump(classifier, 'saved_models/lr_with_equal_weight_1.pkl')
```

Out[65]:

['saved_models/lr_with_equal_weight_1.pkl']

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

```
In [66]:
```

```
sql_create_table = """CREATE TABLE IF NOT EXISTS QuestionsProcessed (question text NOT NULL
create_database_table("Titlemoreweight.db", sql_create_table)
```

Tables in the databse:

QuestionsProcessed

```
In [67]:
```

```
# http://www.sqlitetutorial.net/sqlite-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
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

```
 Remove stop words (Except 'C') 
 Remove HTML Tags 
 Convert all the characters into small letters 
 Use SnowballStemmer to stem the words
```

In [68]:

```
#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])
    if '<code>' in question:
        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
    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,
    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_pr
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
Time taken to run this cell: 0:13:05.637048
```

In [69]:

```
# never forget to close the conections or else we will end up with database locks
conn_r.commit()
conn_w.commit()
conn_r.close()
conn_w.close()
```

__ Sample quesitons after preprocessing of data ___

In [70]:

```
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 data grid bind silverlight bind datagrid dynam code wrote code debug code block s eem bind correct grid come column form come grid column although necessari b ind nthank repli advance..',)

('java.lang.noclassdeffounderror javax servlet jsp tagext taglibraryvalid ja va.lang.noclassdeffounderror javax servlet jsp tagext taglibraryvalid java.l ang.noclassdeffounderror javax servlet jsp tagext taglibraryvalid follow gui d link instal jstl got follow error tri launch jsp page java.lang.noclassdef founderror javax servlet jsp tagext taglibraryvalid taglib declar instal jst l 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 ja va.sql.sqlexcept microsoft odbc driver manag invalid descriptor index java.s ql.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 way use curl someth like way better',)

('btnadd click event open two window record ad btnadd click event open two window record ad btnadd click event open two window record ad open window sea rch.aspx use code hav add button search.aspx nwhen 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 che ck everyth think make sure input field safe type sql inject good news safe b ad 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 foru m field post problem use someth titl field none data get post current use pr int post see submit noth work flawless statement though also mention script work flawless local machin use host come across problem state list input tes t 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 s et sigma algebra mathcal think use monoton properti somewher proof start app reci littl help nthank ad han answer make follow addit construct given han a nswer clear bigcup bigcup cap emptyset neq left bigcup right left bigcup right sum left right also construct subset monoton left right leq left right fi nal would sum leq sum result follow',)

('hql equival sql queri hql equival sql queri hql equival sql queri hql quer i replac name class properti name error occur hql error',)

('undefin symbol architectur i386 objc class skpsmtpmessag referenc error un defin symbol architectur i386 objc class skpsmtpmessag referenc error undefin symbol architectur i386 objc class skpsmtpmessag referenc error import fra mework send email applic background import framework i.e skpsmtpmessag someb odi 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 copi nthat',)

__ Saving Preprocessed data to a Database ___

In [71]:

```
#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 QuestionsProces
conn_r.commit()
conn_r.close()
```

In [72]:

preprocessed_data.head()

Out[72]:

	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 api facebook-php-sdk

In [73]:

```
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 [74]:

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

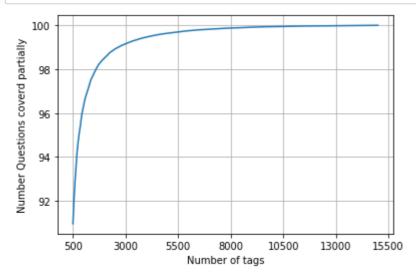
__ Selecting 500 Tags __

In [75]:

```
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
```

In [76]:

```
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 500(it covers
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 [77]:

```
# 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 ",
```

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

In [78]:

```
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 [79]:

```
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 [80]:

Time taken to run this cell: 0:03:58.877800

In [81]:

```
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)
```

```
Dimensions of train data X: (400000, 94927) Y: (400000, 500) Dimensions of test data X: (100000, 94927) Y: (100000, 500)
```

4.5.3 Applying Logistic Regression with OneVsRest Classifier

```
In [82]:
```

```
start = datetime.now()
classifier = OneVsRestClassifier(SGDClassifier(loss='log', alpha=0.00001, penalty='l1'), n
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 : 0.23614
Hamming loss 0.00278378
Micro-average quality numbers
Precision: 0.7208, Recall: 0.3251, F1-measure: 0.4481
Macro-average quality numbers
Precision: 0.5467, Recall: 0.2564, F1-measure: 0.3331
                           recall f1-score
              precision
                                              support
           0
                   0.94
                             0.64
                                       0.76
                                                  5519
           1
                   0.69
                             0.26
                                       0.38
                                                  8190
           2
                   0.81
                             0.38
                                       0.52
                                                  6529
                   0.81
           3
                             0.43
                                       0.56
                                                  3231
           4
                   0.81
                             0.41
                                       0.54
                                                  6430
           5
                   0.82
                             0.34
                                       0.48
                                                  2879
           6
                   0.87
                             0.50
                                       0.63
                                                  5086
           7
                   0.88
                             0.54
                                       0.67
                                                  4533
           8
                                       0.22
                                                  3000
                   0.60
                             0.13
           q
                                                  2765
                   0.81
                             0.52
                                       0.63
          10
                   0.59
                             0.16
                                       0.25
                                                  3051
In [83]:
joblib.dump(classifier, 'saved_models/lr_with_more_title_weight.pkl')
Out[83]:
```

Task 1:Use bag of words upto 4 grams and compute the micro f1 score with Logistic regression(OvR)

['lr_with_more_title_weight.pkl']

```
In [85]:
```

```
start = datetime.now()
vectorizer = CountVectorizer(min_df=0.00009, max_features=200000, ngram_range=(1,4),tokeniz
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: 4:04:07.910078

In [86]:

```
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)
```

```
Dimensions of train data X: (400000, 95585) Y: (400000, 500) Dimensions of test data X: (100000, 95585) Y: (100000, 500)
```

Logistic regression (BoW (1,4) gram without hyperparameter tunning

```
In [87]:
```

```
# this will be taking so much time try not to run it, download the lr_with_equal_weight.pkl
# This takes about 6-7 hours to run.
classifier = OneVsRestClassifier(SGDClassifier(loss='log', alpha=0.00001, penalty='l1'))
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)
         478
                                        0.45
                   0.30
                             0.88
                                                    24
         479
                   0.07
                                        0.11
                             0.26
                                                    92
         480
                   0.31
                             0.60
                                        0.41
                                                   100
         481
                   0.26
                             0.49
                                        0.33
                                                   103
                                                    74
                             0.32
                                        0.13
         482
                   0.08
         483
                   0.49
                             0.67
                                        0.57
                                                   105
         484
                   0.06
                             0.19
                                        0.09
                                                    83
         485
                   0.03
                             0.11
                                        0.04
                                                    82
         486
                   0.06
                             0.25
                                        0.10
                                                    71
         487
                   0.17
                             0.32
                                        0.22
                                                   120
         488
                   0.04
                                        0.06
                                                   105
                             0.11
                                        0.29
                                                    87
         489
                   0.21
                             0.52
         490
                   0.29
                             0.78
                                        0.42
                                                    32
         491
                             0.09
                                        0.04
                   0.02
                                                    69
         492
                   0.05
                             0.14
                                        0.07
                                                    49
         493
                   0.04
                             0.15
                                        0.07
                                                   117
         494
                   0.09
                             0.23
                                        0.13
                                                    61
         495
                   0.84
                             0.86
                                        0.85
                                                   344
         496
                             0.23
                   0.12
                                        0.16
                                                    52
         407
                   A 10
                             A 12
                                        0 26
                                                   107
In [88]:
```

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

Out[88]:

['lr_bow_without_hyp_tuned.pkl']

Task2: Hyperparameter tunning (GridsearchCV)

In [123]:

```
#https://stackoverflow.com/questions/12632992/gridsearch-for-an-estimator-inside-a-onevsres
from sklearn.multiclass import OneVsRestClassifier
from sklearn.model_selection import GridSearchCV

clf = OneVsRestClassifier(SGDClassifier(loss='log',penalty='l1'))

parameters = {
    "estimator__alpha": [0.0001,0.001,0.1,1,10]
}

model_tunning = GridSearchCV(clf, param_grid=parameters)

model_tunning.fit(x_train_multilabel, y_train)
```

Out[123]:

```
GridSearchCV(cv='warn', error_score='raise-deprecating',
              estimator=OneVsRestClassifier(estimator=SGDClassifier(alpha=0.0
001,
                                                                        average=F
alse,
                                                                        class_wei
ght=None,
                                                                        early_sto
pping=False,
                                                                        epsilon=
0.1,
                                                                        eta0=0.0,
                                                                       fit_inter
cept=True,
                                                                        l1_ratio=
0.15,
                                                                        learning
rate='optimal',
                                                                        loss='lo
g',
                                                                       max iter=
1000,
                                                                        n_iter_no
_change=5,
                                                                        n_jobs=No
ne,
                                                                        penalty
='11',
                                                                        power_t=
0.5,
                                                                        random_st
ate=None,
                                                                        shuffle=T
rue,
                                                                        tol=0.00
1,
                                                                       validatio
n fraction=0.1,
                                                                       verbose=
0,
                                                                       warm star
t=False),
                                              n_jobs=None),
```

In [125]:

```
print (model_tunning.best_score_)
print (model_tunning.best_params_)
```

0.189835
{'estimator__alpha': 0.001}

In [126]:

```
classifier = OneVsRestClassifier(SGDClassifier(loss='log', alpha=0.001, penalty='l1'))
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 : 0.19534
Hamming loss 0.00310584
Micro-average quality numbers
Precision: 0.6068, Recall: 0.3028, F1-measure: 0.4040
Macro-average quality numbers
Precision: 0.4187, Recall: 0.2349, F1-measure: 0.2821
              precision
                            recall f1-score
                                                support
           0
                    0.88
                              0.62
                                         0.73
                                                    5519
           1
                    0.58
                              0.18
                                         0.28
                                                    8190
           2
                                         0.44
                    0.77
                              0.31
                                                    6529
           3
                    0.79
                              0.36
                                         0.49
                                                    3231
           4
                                         0.51
                    0.78
                              0.38
                                                    6430
           5
                              0.30
                                         0.43
                                                    2879
                    0.77
           6
                    0.86
                              0.47
                                         0.61
                                                    5086
           7
                    0.87
                              0.51
                                         0.65
                                                    4533
           8
                    0.55
                                         0.22
                                                    3000
                              0.14
           9
                    0.79
                              0.48
                                         0.60
                                                    2765
          10
                    0.59
                              0.14
                                         0.23
                                                    3051
```

```
In [127]:
```

```
joblib.dump(classifier, 'lr_bow_with_hyp_tuned.pkl')
Out[127]:
['lr_bow_with_hyp_tuned.pkl']
```

Task3:Try OneVsRestClassifier with Linear-SVM (SGDClassifier with loss-hinge)

In [128]:

```
clf = OneVsRestClassifier(SGDClassifier(loss='hinge',penalty='l1'))
parameters = {
    "estimator__alpha": [0.0001,0.001,0.1,1,10]
}
model_tunning = GridSearchCV(clf, param_grid=parameters)
model_tunning.fit(x_train_multilabel, y_train)
```

Out[128]:

```
GridSearchCV(cv='warn', error_score='raise-deprecating',
             estimator=OneVsRestClassifier(estimator=SGDClassifier(alpha=
0.0001,
                                                                       average
=False,
                                                                       class_w
eight=None,
                                                                       early_s
topping=False,
                                                                       epsilon
=0.1,
                                                                       eta0=0.
0,
                                                                       fit_int
ercept=True,
                                                                       l1_rati
0=0.15,
                                                                       learnin
g_rate='optimal',
                                                                       loss='h
inge',
                                                                       max_ite
r=1000,
                                                                       n_iter_
no_change=5,
                                                                       n_jobs=
None,
                                                                       penalty
='11',
                                                                       power t
=0.5,
                                                                       random
state=None,
                                                                       shuffle
=True,
                                                                       tol=0.0
01,
                                                                       validat
ion_fraction=0.1,
                                                                       verbose
=0,
                                                                       warm_st
art=False),
                                             n_jobs=None),
             iid='warn', n_jobs=None,
             param_grid={'estimator__alpha': [0.0001, 0.001, 0.01, 0.1, 1,
10]},
```

e,

```
pre_dispatch='2*n_jobs', refit=True, return_train_score=Fals
scoring=None, verbose=0)
```

In [129]:

```
print (model_tunning.best_score_)
print (model_tunning.best_params_)
```

```
0.1796775
{'estimator__alpha': 0.001}
```

In [130]:

```
classifier = OneVsRestClassifier(SGDClassifier(loss='hinge', alpha=0.001, penalty='l1'))
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 : 0.19507
Hamming loss 0.00310382
Micro-average quality numbers
Precision: 0.6090, Recall: 0.2994, F1-measure: 0.4014
Macro-average quality numbers
Precision: 0.3197, Recall: 0.2267, F1-measure: 0.2492
              precision
                            recall f1-score
                                                support
                    0.84
                              0.63
                                         0.72
           0
                                                   5519
                    0.52
                                         0.26
           1
                              0.18
                                                   8190
           2
                                         0.44
                    0.80
                              0.30
                                                   6529
                                         0.52
           3
                    0.72
                              0.40
                                                   3231
           4
                    0.83
                              0.32
                                         0.46
                                                   6430
           5
                                         0.51
                    0.71
                              0.40
                                                   2879
           6
                    0.79
                              0.54
                                         0.64
                                                   5086
           7
                    0.81
                              0.60
                                         0.69
                                                   4533
           8
                    0.52
                                         0.27
                              0.18
                                                   3000
           9
                    0.71
                              0.53
                                         0.60
                                                   2765
                                         0.02
                                                   3051
          10
                    0.32
                              0.01
```

2000

```
In [131]:
joblib.dump(classifier, 'lr_svm.pkl')
Out[131]:
['lr_svm.pkl']
```

Conclusion:

In [133]:

```
from prettytable import PrettyTable
t = PrettyTable()
t.field_names= ("Featurization method","Model"," Macro Averaged F1 Score","micro f1 scoore"
t.add_row(["tf-idf", "Logistic Regression",0.09,0.37, 0.0004])
t.add_row(["tf-idf", "Logistic Regression",0.33,0.44,0.002])
t.add_row(["Bow", "Logistic Regression",0.27,0.36,0.005])
t.add_row(["Bow", "Logistic Regression",0.28,0.40,0.003])
t.add_row(["Bow", "Liner SVM classifier", 0.24,0.40,0.003])
print(t.get_string(titles = "Obeservations"))
```

```
------+
| Featurization method |
                      | Macro Averaged F1 Score | m
                Model
icro f1 scoore | hamming loss |
tf-idf
            | Logistic Regression |
                               0.09
    0.37
       0.0004
    tf-idf | Logistic Regression | 0.33
1
    0.002
0.44
     Bow
            | Logistic Regression |
                               0.27
        0.005
    0.36
            | Logistic Regression |
    Bow
    - 1
        0.003
0.4
            | Liner SVM classifier |
     Bow
0.4
        0.003
```

+-----

Procedure:

- · Load the data to the sqlite database.
- · Removing the duplicates rows and loading the data in a new database.
- · Analysis on tags (Frequecny of each tag).
- Text preprocessing removing HTML tags, stopwords except "c" and stemming.
- Finding tags the covers 99.08% questions --5500 tags.
- · Applied on Models:
 - M1. Logistic regression(OvR) tfidf featurization (5500 tags)
 - M2. Logistic regression(OvR) tfidf featurization and (more weight to title & 5500 tags)
- Adding more weight to title and Finding tags the covers 90.0% questions-- 500 tags
- Applied on Models:
 - M3. Logistic regression(OvR) BoW featurization (more weight to title & 500 tags)

- M4. Logistic regression(OvR) BOW featurization with Hyperparameter tuning (more weight to title & 500 tags)
- M5. Linear SVM BOW featurization with Hyperparameter tuning (more weight to title & 500 tags)
- · Summerizing all the models using pretty table