

## 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 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

(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: <a href="https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/data">https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/data</a> (<a href="https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/data">https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/data</a>)

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).

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## **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 \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
```

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}\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
                                99\n
   50
                50
                                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'
```

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**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)
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. <a href="https://www.kaggle.com/wiki/HammingLoss">https://www.kaggle.com/wiki/HammingLoss</a> (<a href="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)
```

```
180000 rows
360000 rows
540000 rows
720000 rows
900000 rows
1080000 rows
1260000 rows
1440000 rows
1620000 rows
1800000 rows
1980000 rows
2160000 rows
2340000 rows
2520000 rows
2700000 rows
2880000 rows
3060000 rows
3240000 rows
3420000 rows
3600000 rows
3780000 rows
3960000 rows
4140000 rows
4320000 rows
4500000 rows
4680000 rows
4860000 rows
5040000 rows
5220000 rows
5400000 rows
5580000 rows
5760000 rows
5940000 rows
6120000 rows
Time taken to run this cell: 0:04:22.415364
```

## 3.1.2 Counting the number of rows

## In [2]:

```
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 the start of the start
```

Number of rows in the database : 6034196

Time taken to count the number of rows: 0:00:00.055915

## 3.1.3 Checking for duplicates

### In [3]:

```
#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: 0:02:39.453849

#### In [4]:

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

#### Out[4]:

|   | Title   | Body  | Tags   | cnt_dup |
|---|---|---|--|---------|
| 0 | Implementing Boundary Value<br>Analysis of S  | <pre><pre><code>#include&amp;Itiostream&gt;\n#include&amp;</code></pre></pre> | c++ c  | 1       |
| 1 | Dynamic Datagrid Binding in<br>Silverlight?   | I should do binding for datagrid dynamicall                                   | c#<br>silverlight<br>data-<br>binding            | 1       |
| 2 | Dynamic Datagrid Binding in<br>Silverlight?   | I should do binding for datagrid dynamicall                                   | c#<br>silverlight<br>data-<br>binding<br>columns | 1       |
| 3 | java.lang.NoClassDefFoundError:<br>javax/serv | I followed the guide in <a<br>href="http://sta</a<br>                         | 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 [5]:

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

number of duplicate questions : 1827881 ( 30.292038906260256 % )

### In [6]:

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

## Out[6]:

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

Name: cnt\_dup, dtype: int64

#### In [7]:

```
start = datetime.now()
df_no_dup["tag_count"] = df_no_dup["Tags"].str.split().str.len()
# 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: 0:00:14.228219

#### Out[7]:

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

# distribution of number of tags per question

df\_no\_dup.tag\_count.value\_counts()

```
In [8]:
```

```
Out[8]:
       1206157
3.0
       1111706
2.0
        814996
4.0
        568291
1.0
5.0
        505158
Name: tag_count, dtype: int64
In [9]:
#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 [10]:

```
#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: 0:00:50.513593

## 3.2 Analysis of Tags

## 3.2.1 Total number of unique tags

```
In [11]:
```

```
tag_data["Tags"].fillna(value='NaN', inplace=True)
```

### In [12]:

```
# 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(),lowercase= False)
# 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 [13]:

```
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 : 42050

### In [14]:

```
#'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 [15]:

```
# 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 [16]:

```
#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[16]:

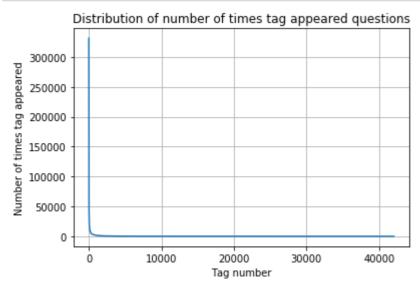
|   | Tags                     | Counts |
|---|--------------------------|--------|
| 0 | lwp                      | 118    |
| 1 | jobeet                   | 23     |
| 2 | windows-live-movie-maker | 12     |
| 3 | connection-timeout       | 95     |
| 4 | flowdocumentreader       | 27     |

### In [17]:

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

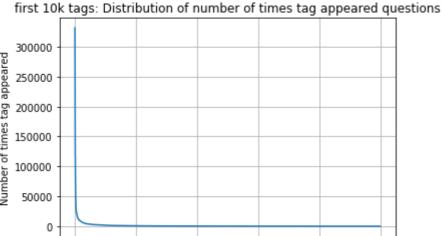
## In [18]:

```
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 [19]:

```
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])
```



Number of times tag appeared 8000 2000 4000 6000 10000 Tag number

400 [331505 44829 22429 17728 13364 11162 10029

| 6466 | 5865 | 5370 | 4983 | 4526 | 4281 | 4144 | 3929 | 3750 | 3593 |  |
|------|------|------|------|------|------|------|------|------|------|--|
| 3453 | 3299 | 3123 | 2986 | 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  |  |
|      |      |      |      |      |      |      |      |      |      |  |

9148

8054

7151

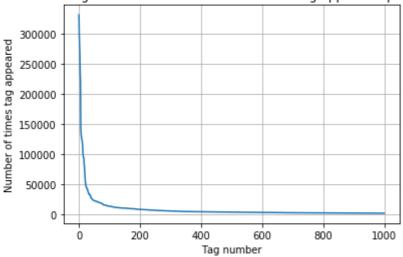
30/11/2019 11

| 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  | 75  |   |
| 75  | 74  | 74  | 74  | 73  | 73  | 73  | 73  | 72  | 72] | ~ |
|     |     |     |     |     |     |     |     |     |     |   |

## In [20]:

```
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])
```



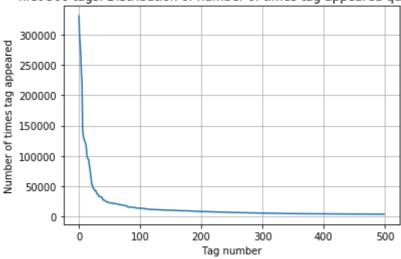


| 200 [331 | 505 221 | 533 122 | 769 95 | 160 62 | 023 44 | 829 37 | 170 31 | 897 26 | 925 24537 |
|----------|---------|---------|--------|--------|--------|--------|--------|--------|-----------|
| 22429    | 21820   | 20957   | 19758  | 18905  | 17728  | 15533  | 15097  | 14884  | 13703     |
| 13364    | 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   | 3482      |
| 3453     | 3427    | 3396    | 3363   | 3326   | 3299   | 3272   | 3232   | 3196   | 3168      |
| 3123     | 3094    | 3073    | 3050   | 3012   | 2986   | 2983   | 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 [21]:

```
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])
```





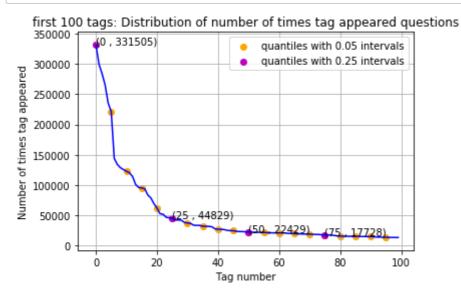
| 100 [331 | .505 221 | 533 122 | 769 95 | 160 62 | 023 44 | 829 37 | 170 31 | 897 26 | 925 245 | 37 |
|----------|----------|---------|--------|--------|--------|--------|--------|--------|---------|----|
| 22429    | 21820    | 20957   | 19758  | 18905  | 17728  | 15533  | 15097  | 14884  | 13703   |    |
| 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   | 3482]   |    |

## In [22]:

```
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 [23]:

```
# 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 [24]:

```
#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])
```

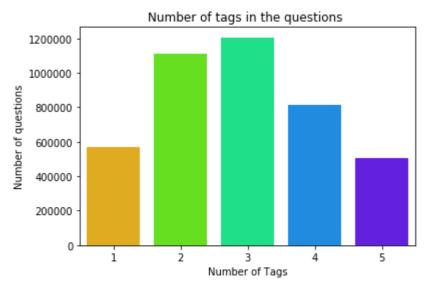
We have total 4206314 datapoints. [3, 4, 2, 2, 3]

#### In [25]:

```
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 [26]:

```
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()
```



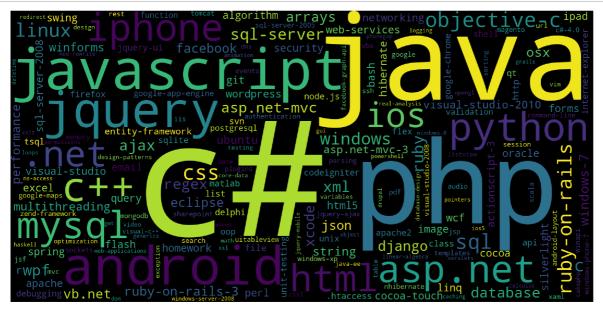
## **Observations:**

- 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 [27]:

```
# 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: 0:00:05.616900

#### Observations:

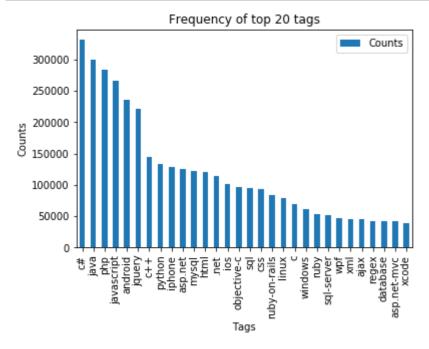
A look at the word cloud shows that "c#", "java", "php", "asp.net", "javascript", "c++" are some of the most frequent tags.

## 3.2.6 The top 20 tags

## In [28]:

```
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()
```

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### **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

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## In [7]:

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

### In [2]:

```
#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)
```

Tables in the databse: OuestionsProcessed

```
In [31]:
```

```
# 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: 0:03:14.073526

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

#### In [32]:

```
import nltk
nltk.download('punkt')
```

[nltk\_data] Downloading package punkt to /home/shanud6711/nltk\_data...
[nltk\_data] Package punkt is already up-to-date!

### Out[32]:

True

### In [33]:

```
#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
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: 0:24:39.138315
```

## In [34]:

```
# 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 [35]:
```

```
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

\_\_\_\_\_\_

('directori structur project servlet api read strut action say accord serv let specif top level directori except web inf document root web applic pag e third paragraph sampl project web inf still document root project creat servlet project eclips web inf sit insid web content directori thought eclips specif kind let know servlet specif say',)

-----

------

('qawc gsl could integr function got piec code thought deal kind function get could integr function error',)

-----

-----

('xpath multipl negat follow simplifi xml map target xml came follow xslt appli xslt get follow unwant output solv tri around xpath express get want result wrong xpath thank idea nbest regard npeter',)

-----

-----

('want result use subqueri want row maximum without use subqueri result ob tain run queri want appli filter queri',)

------

-----

('transit latex plain tex sever reason need typeset plain tex howev alway work latex hard basic need someth like book report documentclass toc chapt er section theorem appendic index bib thing learn latex come exampl somebo di direct complet exampl book typeset plain tex',)

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\_\_\_\_\_\_

\_\_\_\_\_

('passiv get new access token access token expir work simpl facebook app a llow post client websit client facebook fan page right set access token cr eat applic instanc php sdk use day long live extend access token client lo gin facebook time want post fan page point access token becom invalid numb er reason ideal circumst would everi day time would nice could continu pub

```
lish fan page websit app day arriv user de author app anyway get new acces s token expir without ask user login facebook',)

('delet share folder use c code directori open anoth machin tri delet empt i share directori open anoth machin nif direct delet directori right click delet remov delet directori directori open machin differ machin use share path found nbut understand code proper anyon suggest better method nthank advanc',)

('releas view stack applic use follow code convent open default screen mov e screen default screen default screen still exist even use self dismissmo delviewcontrol yes dimiss default screen view wrong want default screen co mplet remov view way call default screen actual applic npleas help thank a dvanc',)
```

## In [3]:

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

## In [4]:

```
preprocessed_data.head()
```

#### Out[4]:

|   | question                                       | tags                                     |
|---|--|--|
| 0 | xml net replac specif node mani one load raw x | .net xml xmldocument xmlnode xmlnodelist |
| 1 | directori structur project servlet api read st | jsp servlets                             |
| 2 | qawc gsl could integr function got piec code t | gsl                                      |
| 3 | xpath multipl negat follow simplifi xml map ta | xslt xpath foreach                       |
| 4 | want result use subqueri want row maximum with | sql sql-server-2008                      |
|   |  |  |

#### In [5]:

```
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
```

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## In [6]:

```
preprocessed_data['tags'].fillna(value='NaN', inplace=True)
```

## In [40]:

```
# binary='true' will give a binary vectorizer
vectorizer = CountVectorizer(tokenizer = lambda x: x.split(), binary='true',lowercase=False
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 [11]:

```
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

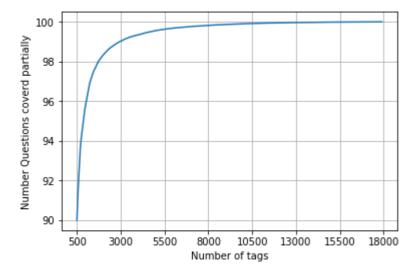
def questions_explained_fn(n):
    multilabel_yn = tags_to_choose(n)
    x= multilabel_yn.sum(axis=1)
    return (np.count_nonzero(x==0))
```

## In [42]:

```
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 [43]:
```

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



with 5500 tags we are covering 99.024 % of questions

```
In [44]:
```

```
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 : 9756 out of 999999

```
In [45]:
```

```
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],"(",(multilabel_yx.shape[1]/multilabel_yx.shape[1],"(",(multilabel_yx.shape[1]/multilabel_yx.shape[1])
number of tags in sample : 35391
number of tags taken : 5500 ( 15.540674182701816 %)
```

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

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

```
In [46]:
```

```
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 [47]:

```
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 [50]:

Time taken to run this cell: 0:15:22.046699

#### In [51]:

```
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: (799999, 88238) Y: (799999, 5500) Dimensions of test data X: (200000, 88238) Y: (200000, 5500)
```

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

```
In [52]:
```

```
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 [53]:

```
# 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 [54]:

```
#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: 0:19:08.365734
```

## In [55]:

```
# 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 [56]:
```

```
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 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 queri 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 [57]:

```
#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 [58]:

preprocessed\_data.head()

#### Out[58]:

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

```
In [59]:
```

```
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 [60]:

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

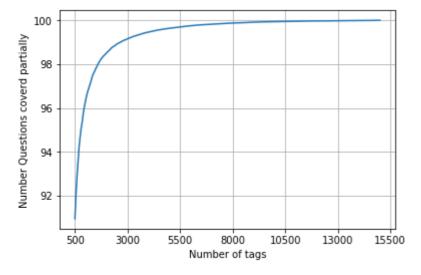
\_\_ Selecting 500 Tags \_\_

#### In [61]:

```
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 [62]:

```
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 [63]:
```

```
# 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 [64]:

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 [65]:

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

# 4.5.2 Featurizing data with Tfldf vectorizer

# In [ ]:

#### 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)
```

# 4.5.3 Applying Logistic Regression with OneVsRest Classifier

#### In [67]:

```
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)
           1
                   0.69
                             0.26
                                        0.38
                                                  8190
           2
                   0.81
                             0.37
                                        0.51
                                                  6529
           3
                                        0.56
                   0.81
                             0.43
                                                  3231
           4
                                        0.54
                   0.80
                             0.41
                                                  6430
           5
                                        0.48
                   0.82
                             0.34
                                                  2879
           6
                   0.87
                             0.50
                                        0.63
                                                  5086
           7
                   0.88
                             0.54
                                        0.67
                                                  4533
           8
                   0.60
                             0.13
                                        0.22
                                                  3000
           9
                   0.81
                             0.52
                                        0.63
                                                  2765
          10
                   0.59
                             0.16
                                        0.26
                                                  3051
                   0.69
                             0.33
                                        0.45
                                                  3009
          11
                                        0.35
          12
                   0.65
                             0.24
                                                  2630
          13
                   0.70
                             0.22
                                        0.34
                                                  1426
                   0.90
                             0.53
                                        0.67
                                                  2548
          14
          15
                   0.68
                             0.18
                                        0.28
                                                  2371
                                        0.34
          16
                   0.65
                             0.23
                                                   873
          17
                   0.89
                             0.61
                                        0.72
                                                  2151
          18
                   0.63
                             0.23
                                        0.34
                                                  2204
          19
                   0.72
                             0.40
                                        0.52
                                                   831
          20
                   0.77
                             0.41
                                        0.53
                                                  1860
```

#### In [68]:

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

/home/shanud6711/.local/lib/python3.5/site-packages/sklearn/externals/jobli b/\_\_init\_\_.py:15: DeprecationWarning: sklearn.externals.joblib is deprecated in 0.21 and will be removed in 0.23. Please import this functionality direct ly from joblib, which can be installed with: pip install joblib. If this war ning is raised when loading pickled models, you may need to re-serialize tho se models with scikit-learn 0.21+.

warnings.warn(msg, category=DeprecationWarning)

#### Out[68]:

['lr\_with\_more\_title\_weight.pkl']

```
In [67]:
```

```
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)
         4/3
                   0.67
                             0.3/
                                        0.4/
                                                   158
         474
                   0.48
                             0.10
                                        0.17
                                                    138
         475
                   0.00
                             0.00
                                        0.00
                                                    59
         476
                                        0.43
                                                    88
                   0.63
                             0.33
                                                   176
         477
                   0.83
                             0.65
                                        0.73
         478
                   0.95
                             0.79
                                        0.86
                                                    24
                                                    92
         479
                   0.22
                                        0.07
                             0.04
         480
                   0.79
                             0.50
                                        0.61
                                                   100
         481
                   0.51
                             0.28
                                        0.36
                                                   103
         482
                   0.40
                             0.22
                                        0.28
                                                    74
         483
                   0.78
                             0.63
                                        0.69
                                                   105
         484
                                        0.04
                                                    83
                   0.20
                             0.02
         485
                   0.20
                             0.02
                                        0.04
                                                    82
                                        0.23
                                                    71
         486
                   0.48
                             0.15
         487
                   0.45
                             0.21
                                        0.29
                                                   120
         488
                   0.50
                             0.06
                                        0.10
                                                   105
         489
                   0.73
                             0.37
                                        0.49
                                                    87
                                        0.90
                                                    32
         490
                   1.00
                             0.81
         491
                   0.33
                             0.03
                                        0.05
                                                    69
         492
                   0.33
                             0.02
                                        0.04
                                                    49
```

# 5. Assignments

- 1. Use bag of words upto 4 grams and compute the micro f1 score with Logistic regression(OvR)
- 2. Perform hyperparam tuning on alpha (or lambda) for Logistic regression to improve the performance using GridSearch
- 3. Try OneVsRestClassifier with Linear-SVM (SGDClassifier with loss-hinge)

```
In [7]:
```

```
preprocessed_data.head()
```

#### Out[7]:

question tags 0 xml net replac specif node mani one load raw x... .net xml xmldocument xmlnode xmlnodelist 1 directori structur project servlet api read st... jsp servlets 2 qawc gsl could integr function got piec code t... gsl 3 xpath multipl negat follow simplifi xml map ta... xslt xpath foreach want result use subqueri want row maximum with... sql sql-server-2008

## In [8]:

```
preprocessed_data_100k = preprocessed_data[:100000]
print("number of data points in sample :", preprocessed_data_100k.shape[0])
print("number of dimensions :", preprocessed_data_100k.shape[1])
number of data points in sample : 100000
number of dimensions : 2
```

### In [9]:

```
vectorizer = CountVectorizer(tokenizer = lambda x: x.split(), binary='true')
multilabel_y = vectorizer.fit_transform(preprocessed_data_100k['tags'])
```

#### In [12]:

```
# we will be taking 500 tags
multilabel_yx = tags_to_choose(500)
```

#### In [13]:

```
train_datasize = 70000
x_train=preprocessed_data.head(train_datasize)
x_test=preprocessed_data.tail(preprocessed_data_100k.shape[0] - train_datasize)

y_train = multilabel_yx[0:train_datasize,:]
y_test = multilabel_yx[train_datasize:preprocessed_data_100k.shape[0],:]
```

#### In [14]:

```
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: (70000, 500)
Number of data points in test data: (30000, 500)
```

# Bag of Words Upto 4 grams with Logistic regression(OvR)

```
In [15]:
```

```
start = datetime.now()
vectorizer = CountVectorizer(min_df=0.00009,tokenizer = lambda x: x.split(), ngram_range=(1
x_train_bow = vectorizer.fit_transform(x_train['question'])
x_test_bow = vectorizer.transform(x_test['question'])
print("Time taken to run this cell :", datetime.now() - start)
```

Time taken to run this cell: 0:01:29.338349

#### In [16]:

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

Dimensions of train data X: (70000, 40000) Y: (70000, 500) Dimensions of test data X: (30000, 40000) Y: (30000, 500)

### In [17]:

```
from sklearn.externals import joblib
joblib.dump(x_train_bow, 'x_trainbow.pkl')

joblib.dump(x_test_bow, 'x_testbow.pkl')

joblib.dump(y_train, 'y_trainbow.pkl')

joblib.dump(y_test, 'y_testbow.pkl')
```

/home/shanud6711/.local/lib/python3.5/site-packages/sklearn/externals/jobli b/\_\_init\_\_.py:15: DeprecationWarning: sklearn.externals.joblib is deprecated in 0.21 and will be removed in 0.23. Please import this functionality direct ly from joblib, which can be installed with: pip install joblib. If this war ning is raised when loading pickled models, you may need to re-serialize tho se models with scikit-learn 0.21+.

warnings.warn(msg, category=DeprecationWarning)

#### Out[17]:

['y\_testbow.pkl']

#### In [18]:

```
from sklearn.externals import joblib
x_train_bow=joblib.load( 'x_trainbow.pkl')
x_test_bow=joblib.load( 'x_testbow.pkl')

y_train=joblib.load( 'y_trainbow.pkl')

y_test=joblib.load('y_testbow.pkl')
```

#### In [22]:

```
import warnings
warnings.filterwarnings("ignore")
from sklearn.model_selection import GridSearchCV

start = datetime.now()

parameters={'estimator__C': [10**-5, 10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1]}
Logistic_clf = OneVsRestClassifier(LogisticRegression(class_weight='balanced', penalty='l1'
Gridsearch_clf = GridSearchCV(estimator = Logistic_clf, param_grid=parameters, cv=2, verbos
Gridsearch_clf.fit(x_train_bow, y_train)
best_alpha = Gridsearch_clf.best_estimator_.get_params()['estimator__C']

print('value of alpha after hyperparameter tuning : ',best_alpha)

print("Time taken to run this cell :", datetime.now() - start)
```

#### Fitting 2 folds for each of 7 candidates, totalling 14 fits

```
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done
                            3 out of 14 | elapsed: 2.5min remaining: 9.
0min
[Parallel(n_jobs=-1)]: Done
                             5 out of 14 | elapsed: 6.6min remaining: 11.
8min
[Parallel(n jobs=-1)]: Done
                             7 out of 14 | elapsed: 11.1min remaining: 11.
1min
[Parallel(n_jobs=-1)]: Done
                            9 out of 14 | elapsed: 17.9min remaining:
9min
[Parallel(n_jobs=-1)]: Done 11 out of 14 | elapsed: 22.2min remaining: 6.
0min
[Parallel(n jobs=-1)]: Done 14 out of 14 | elapsed: 25.7min finished
value of alpha after hyperparameter tuning: 10
Time taken to run this cell: 1:46:57.921464
```

```
In [24]:
```

```
start = datetime.now()
classifier1 = OneVsRestClassifier(LogisticRegression(penalty='11', C=best_alpha, class_weig
classifier1.fit(x_train_bow, y_train)
predictions = classifier1.predict(x_test_bow)
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: 0.01436666666666666
Hamming loss 0.0078594
Micro-average quality numbers
Precision: 0.0209, Recall: 0.0258, F1-measure: 0.0231
Macro-average quality numbers
Precision: 0.0036, Recall: 0.0043, F1-measure: 0.0039
                           recall f1-score
              precision
                                              support
           a
                   0.08
                             0.10
                                       0.09
                                                  2443
           1
                   0.07
                             0.08
                                       0.08
                                                  2171
           2
                                       0.06
                   0.06
                             0.07
                                                  1965
           3
                             0.07
                                       0.07
                                                  1854
                   0.06
           4
                   0.06
                             0.06
                                       0.06
                                                  1669
           5
                   0.05
                             0.06
                                       0.06
                                                  1565
           6
                   0.03
                             0.04
                                       0.03
                                                  1048
           7
                                                   953
                   0.04
                             0.04
                                       0.04
           8
                   0.04
                             0.04
                                       0.04
                                                   915
           9
                   0.03
                             0.03
                                       0.03
                                                   887
          10
                   0.02
                             0.03
                                       0.03
                                                   903
```

# OneVsRestClassifier with Linear-SVM (SGDClassifier with loss-hinge)

#### In [26]:

```
import warnings
warnings.filterwarnings("ignore")
from sklearn.model_selection import GridSearchCV
start = datetime.now()
parameters={'estimator__alpha': [10**-5, 10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1]}
svm_clf = OneVsRestClassifier(SGDClassifier(loss='hinge', penalty='l1'))
svm_gd_clf = GridSearchCV(estimator = svm_clf, param_grid=parameters, cv=2, verbose=10, scc
svm_gd_clf.fit(x_train_bow, y_train)
svm_best_alpha = svm_gd_clf.best_estimator_.get_params()['estimator__alpha']
print('value of alpha after hyperparameter tuning : ',best_alpha)
print("Time taken to run this cell :", datetime.now() - start)
```

Fitting 2 folds for each of 7 candidates, totalling 14 fits

```
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n jobs=-1)]: Done
                           3 out of 14 | elapsed: 3.9min remaining: 14.
5min
[Parallel(n_jobs=-1)]: Done
                            5 out of 14 | elapsed: 8.2min remaining: 14.
7min
[Parallel(n_jobs=-1)]: Done
                            7 out of 14 | elapsed: 9.4min remaining:
4min
[Parallel(n_jobs=-1)]: Done
                             9 out of 14 | elapsed: 10.2min remaining:
7min
[Parallel(n_jobs=-1)]: Done 11 out of 14 | elapsed: 10.6min remaining:
9min
[Parallel(n_jobs=-1)]: Done 14 out of 14 | elapsed: 11.0min finished
value of alpha after hyperparameter tuning: 10
Time taken to run this cell: 0:15:44.839545
```

#### In [27]:

```
start = datetime.now()
classifier2 = OneVsRestClassifier(SGDClassifier(loss='hinge', alpha=svm_best_alpha, penalty
classifier2.fit(x_train_bow, y_train)
predictions2 = classifier2.predict (x_test_bow)
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)
Hamming loss 0.0078594
Micro-average quality numbers
Precision: 0.0209, Recall: 0.0258, F1-measure: 0.0231
Macro-average quality numbers
Precision: 0.0036, Recall: 0.0043, F1-measure: 0.0039
              precision
                          recall f1-score
                                             support
           0
                  0.08
                            0.10
                                      0.09
                                                2443
           1
                  0.07
                            0.08
                                      0.08
                                                2171
           2
                  0.06
                            0.07
                                      0.06
                                                1965
           3
                            0.07
                                      0.07
                  0.06
                                                1854
           4
                                      0.06
                  0.06
                            0.06
                                                1669
           5
                  0.05
                            0.06
                                      0.06
                                                1565
           6
                                      0.03
                                                1048
                  0.03
                            0.04
           7
                  0.04
                            0.04
                                      0.04
                                                 953
          8
                                      0.04
                                                 915
                  0.04
                            0.04
          9
                  0.03
                            0.03
                                      0.03
                                                  887
          10
                  0.02
                            0.03
                                      0.03
                                                 903
```

```
In [29]:
```

```
from prettytable import PrettyTable
print("TF-IDF with 0.5 million dataset")
x = PrettyTable()
x.field_names = ["Model", "Vectorizer", "Accuracy", "Hamming loss", "Precision", "Recall", "Mi

x.add_row(["SDG with loss - log ", 'TF-IDF ', 0.23623, 0.0027, 0.7216, 0.3252, 0.4483])
x.add_row(["LogisticRegression", 'TF-IDF ', 0.25111, 0.002702, 0.7172, 0.3673, 0.4858])
print(x)
```

```
TF-IDF with 0.5 million dataset

+-----+

| Model | Vectorizer | Accuracy | Hamming loss | Precision |
Recall | Micro f1 |

+-----+

| SDG with loss - log | TF-IDF | 0.23623 | 0.0027 | 0.7216 |
0.3252 | 0.4483 |
| LogisticRegression | TF-IDF | 0.25111 | 0.002702 | 0.7172 |
0.3673 | 0.4858 |

+-----+
```

#### In [31]:

```
print("Hyperparameter Tunning of BOW with 100k dataset")
x = PrettyTable()
x.field_names = ["Model", "Vectorizer", "Accuracy", "Hamming loss", "Precision", "Recall", "Mi
x.add_row(["Logistic Regression", 'BOW', 0.014366, 0.0078594, 0.0209, 0.0258, 0.0231])
x.add_row(["Linear SVM", 'BOW', 0.014366, 0.0078594, 0.0209, 0.0258, 0.0231])
print(x)
```

# In [ ]: