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

 $\underline{(https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/data)}$

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:

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

Title - The question's title

Body - The body of the question

Tags - The tags associated with the question in a space-seperated format (all 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
```

}\n

 $n\n$

```
The answer should come in the form of a table like\n\n
       <code>
       1
                                    50\n
                    50
                    50
       2
                                    50\n
       99
                                    50\n
                    50
       100
                    50
                                    50\n
       50
                                    50\n
                    1
       50
                    2
                                    50\n
       50
                    99
                                    50\n
       50
                    100
                                    50\n
       50
                    50
                                    1\n
       50
                    50
                                    2\n
       50
                                    99\n
                    50
       50
                    50
                                    100\n
       </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 [0]:

```
#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=chunksize, iterator=T
    ):
        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

```
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: 0:01:15.750352

3.1.3 Checking for duplicates

In [0]:

```
#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: 0:04:33.560122

In [0]:

```
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		

```
print(
    "number of duplicate questions :", num_rows['count(*)'].values[0]- df_no_dup.shape[0],
    "(",(1-((df_no_dup.shape[0])/(num_rows['count(*)'].values[0])))*100,"% )"
)
```

number of duplicate questions : 1827881 (30.2920389063 %)

In [0]:

```
# 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

In [0]:

```
start = datetime.now()
df_no_dup["tag_count"] = df_no_dup["Tags"].apply(lambda text: len(text.split(" ")))
# adding a new feature number of tags per question
print("Time taken to run this cell :", datetime.now() - start)
df_no_dup.head()
```

Time taken to run this cell: 0:00:03.169523

Out[9]:

	Title	Body	Tags	cnt_dup t
0	Implementing Boundary Value Analysis of S	<pre><pre><code>#include<iostream>\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		

```
# distribution of number of tags per question
df_no_dup.tag_count.value_counts()
Out[10]:
3
     1206157
2
     1111706
4
      814996
1
      568298
      505158
Name: tag_count, dtype: int64
In [0]:
#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 [0]:

```
#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: 0:00:52.992676

3.2 Analysis of Tags

3.2.1 Total number of unique tags

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

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

In [0]:

```
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 [0]:

```
#'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 tages we have : ['.a', '.app', '.asp.net-mvc', '.aspxauth', '.ba sh-profile', '.class-file', '.cs-file', '.doc', '.drv', '.ds-store']
```

3.2.3 Number of times a tag appeared

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

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

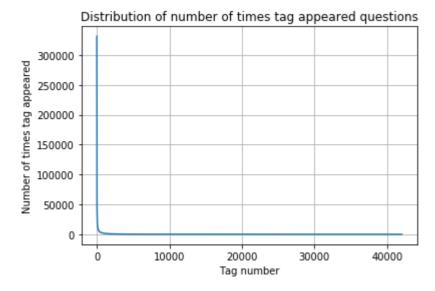
Out[17]:

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

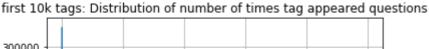
In [0]:

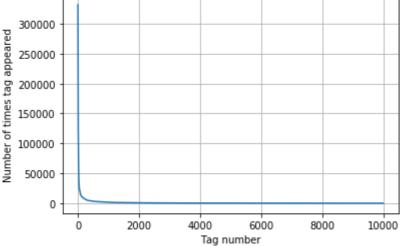
```
tag_df_sorted = tag_df.sort_values(['Counts'], ascending=False)
tag_counts = tag_df_sorted['Counts'].values
```

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



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



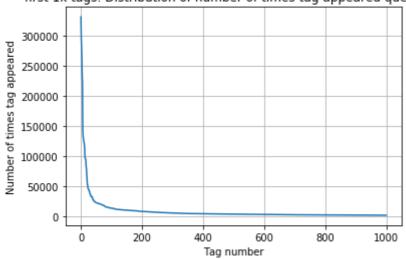


400	[3315	05	44829	224	29	17728	13	3364	11	162	10	029	9:	148	8	8054	7151	
	6466	586	55 53	370	498	3 4	526	42	81	41	44	392	29	375	50	359	93	
	3453	329	9 31	123	298	9 2	891	27	38	26	47	252	27	243	31	233	31	
	2259	218	6 26	997	202	0 1	959	19	00	18	28	17	70	172	23	167	73	
	1631	157	'4 15	532	147	9 1	448	14	06	13	65	132	28	136	90	126	56	
	1245	122	2 12	197	118	1 1	158	11	39	11	21	110	ð1	107	76	105	56	
	1038	102	23 10	906	98	3	966	9	52	9	38	92	26	91	11	89	91	
	882	86		356	84		830		16		04		39	77	79	77	70	
	752	74		733	72		712		02		88		78		71	65	58	
	650	64	ŀ3 €	534	62	7	616	6	07	5	98	58	39	58	33	57	77	
	568	55		552	54		540		33		26		18		12	56	96	
	500	49		190	48		480		77		69	46	55		57	45	50	
	447	44		437	43		426		22		18		13		98	46		
	398	39		388	38		381		78		74		70		57		55	
	361	35		354	35		347		44		42		39		36	33		
	330	32		323	31		315		12		09		ð7		94		91	
	299	29		293	29		289		86		84		31		78		76	
	275	27		270	26		265		62		60		58		56	25		
	252	25		249	24		245		43		41		39		38	23		
	234	23		232	23		228		26		24		22		20		L9	
	217	21		214	21		210		09		.07		ð5		94	26		
	201	26		199	19		196		94		93		92		91	18		
	188	18		185	18		182		81		.80		79		78	17		
	175	17		172	17		170		69		68		57		56	16		
	164	16		161	16		159		58		57		56		56	15		
	154	15		152	15		150		49		49		48		47	14		
	145	14		143	14		142		41		40		39		38	13		
	137	13		135	13		134		33		32		31		30	13		
	129	12		128	12		126		26		25		24		24	12		
	123	12		122	12		120		20		19		18		18	11		
	117	11		116	11		115		14		13		13		12	11		
	111	11		109	10		108		80		07		96		96	16		
	105	16		104	10		103		03		02		ð2		ð1	16		
	100	16	00	99	9	9	98		98		97	9	97	9	96	9	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]

```
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])
```

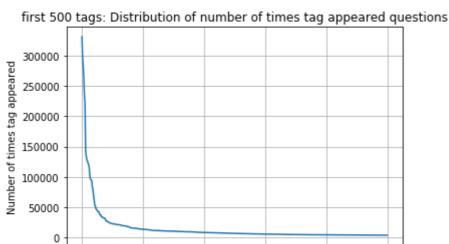




200 [331	505 221	533 122	769 95	160 62	023 44	829 37	170 31	897 26	925 24537
22429	21820	20957	19758	18905	17728	15533	15097	14884	13703
13364	13157	12407	11658	11228	11162	10863	10600	10350	10224
10029	9884	9719	9411	9252	9148	9040	8617	8361	8163
8054	7867	7702	7564	7274	7151	7052	6847	6656	6553
6466	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	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]

Ó

```
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])
```



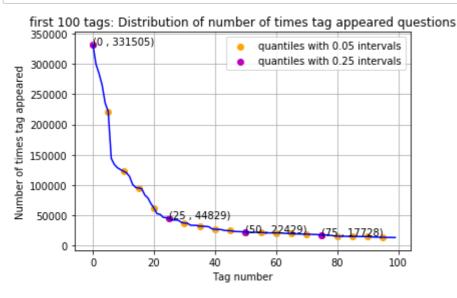
Tag number

```
100 [331505 221533 122769 95160
                                     62023
                                            44829
                                                    37170
                                                            31897
                                                                    26925
                                                                           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
                                                         6847
   8054
           7867
                  7702
                          7564
                                  7274
                                         7151
                                                 7052
                                                                6656
                                                                        6553
   6466
           6291
                  6183
                          6093
                                  5971
                                         5865
                                                 5760
                                                         5577
                                                                5490
                                                                        5411
           5283
                  5207
                          5107
                                  5066
                                         4983
                                                 4891
                                                         4785
                                                                4658
                                                                        4549
   5370
   4526
           4487
                  4429
                          4335
                                  4310
                                         4281
                                                 4239
                                                         4228
                                                                4195
                                                                        4159
   4144
           4088
                  4050
                          4002
                                  3957
                                         3929
                                                 3874
                                                         3849
                                                                3818
                                                                        3797
   3750
           3703
                  3685
                          3658
                                  3615
                                         3593
                                                 3564
                                                         3521
                                                                3505
                                                                        3483]
```

```
plt.plot(tag_counts[0:100], c='b')
plt.scatter(x=list(range(0,100,5)), y=tag_counts[0:100:5], c='orange', label="quantiles 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 [0]:

```
# 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 [0]:

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

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

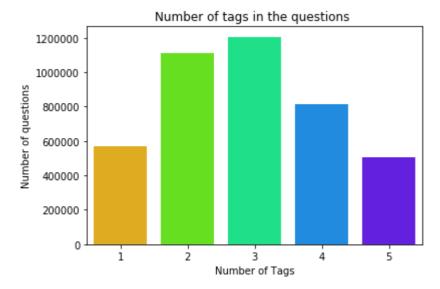
In [0]:

```
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 [0]:

```
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()
```



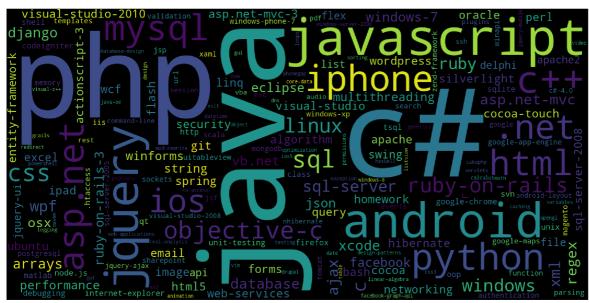
Observations:

- 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 [0]:

```
# 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)
```



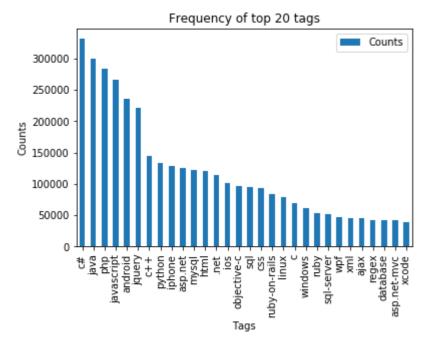
Time taken to run this cell : 0:00:05.470788

Observations:

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

3.2.6 The top 20 tags

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



Observations:

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

3.3 Cleaning and preprocessing of Questions

3.3.1 Preprocessing

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

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

```
#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, code text, tags text, words_pre integer, words_post integer, i
create_database_table("Processed.db", sql_create_table)
```

Tables in the databse: QuestionsProcessed

```
# 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 : 0:06:32.806567

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

```
#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: 0:47:05.946582
```

```
# 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()
```

```
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

('ef code first defin one mani relationship differ key troubl defin one zero mani relationship entiti ef object model look like use fluent api object com posit pk defin batch id batch detail id use fluent api object composit pk de fin batch detail id compani id map exist databas tpt basic idea submittedtra nsact zero mani submittedsplittransact associ navig realli need one way subm ittedtransact submittedsplittransact need dbcontext class onmodelcr overrid map class lazi load occur submittedtransact submittedsplittransact help woul d much appreci edit taken advic made follow chang dbcontext class ad follow onmodelcr overrid must miss someth get follow except thrown submittedtransact key batch id batch detail id zero one mani submittedsplittransact key batch detail id compani id rather assum convent creat relationship two object co nfigur requir sinc obvious wrong',)

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

('error function notat function solv logic riddl iloczyni list structur list possibl candid solut list possibl coordin matrix wan na choos one candid com par possibl candid element equal wan na delet coordin call function skasuj l ook like ni knowledg haskel cant see what wrong',)

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

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

('magento unit test problem magento site recent look way check integr magent o site given point unit test jump one method would assum would big job write whole lot test check everyth site work anyon involv unit test magento advis follow possibl test whole site custom modul nis exampl test would amaz given site heavili link databas would nbe possibl fulli test site without disturb databas better way automaticlli check integr magento site say integr realli

```
#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 [0]:

```
preprocessed_data.head()
```

Out[47]:

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

In [0]:

```
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
```

```
# 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) ___

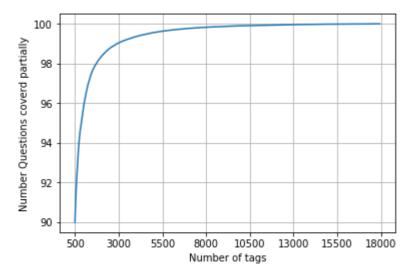
In [11]:

```
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))
```

```
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
```

```
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 99.04 % of questions

In [0]:

```
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: 9599 out of 999999

In [0]:

```
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 : 35422 number of tags taken : 5500 (15.527073570097679 %)

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

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

```
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 [0]:

```
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 [0]:

Time taken to run this cell: 0:09:50.460431

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

```
Diamensions of train data X: (799999, 88244) Y: (799999, 5500) Diamensions of test data X: (200000, 88244) Y: (200000, 5500)
```

```
# 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[92]:

"\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"

4.4 Applying Logistic Regression with OneVsRest Classifier

```
# 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'), 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: 0.081965
macro f1 score : 0.0963020140154
micro f1 scoore: 0.374270748817
hamming loss: 0.00041225090909090907
Precision recall report :
              precision
                           recall f1-score
                                               support
                            0.23
          0
                  0.62
                                      0.33
                                                15760
          1
                  0.79
                            0.43
                                      0.56
                                                14039
          2
                  0.82
                            0.55
                                      0.66
                                                13446
          3
                  0.76
                            0.42
                                      0.54
                                                12730
          4
                  0.94
                            0.76
                                      0.84
                                                11229
          5
                                      0.73
                  0.85
                            0.64
                                                10561
                  0.70
                            0.30
                                      0.42
                                                 6958
          6
          7
                  0.87
                            0.61
                                      0.72
                                                 6309
                            0.40
          8
                  0.70
                                      0.50
                                                 6032
          9
                            0.43
                  0.78
                                      0.55
                                                 6020
         10
                  0.86
                            0.62
                                      0.72
                                                 5707
                            0.17
         11
                  0.52
                                      0.25
                                                 5723
In [0]:
```

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

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

In [0]:

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

Tables in the databse: QuestionsProcessed

```
# 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 Special 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
```

```
#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: 0:23:12.329039
```

```
# 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 ____

```
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 [0]:

```
#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 [0]:

preprocessed_data.head()

Out[100]:

	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

```
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 [0]:

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

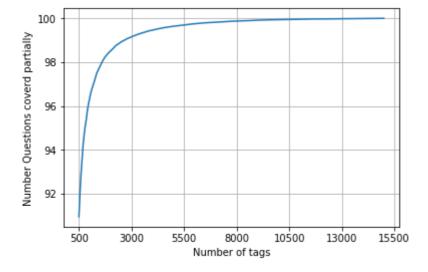
Selecting 500 Tags ___

In [0]:

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

In [0]:

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

```
# 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 [0]:

```
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 [0]:

```
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 [0]:

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

In [0]:

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

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

4.5.3 Applying Logistic Regression with OneVsRest Classifier

```
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 : 0.23623
Hamming loss 0.00278088
Micro-average quality numbers
Precision: 0.7216, Recall: 0.3256, F1-measure: 0.4488
Macro-average quality numbers
Precision: 0.5473, Recall: 0.2572, F1-measure: 0.3339
             precision
                          recall f1-score
                  0.94
                            0.64
          0
                                      0.76
                                                 5519
          1
                  0.69
                            0.26
                                      0.38
                                                 8190
          2
                  0.81
                            0.37
                                      0.51
                                                 6529
                            0.43
          3
                  0.81
                                      0.56
                                                 3231
          4
                  0.81
                            0.40
                                      0.54
                                                 6430
          5
                            0.33
                                      0.47
                                                 2879
                  0.82
          6
                  0.87
                            0.50
                                      0.63
                                                 5086
          7
                            0.54
                  0.87
                                      0.67
                                                 4533
          8
                  0.60
                            0.13
                                      0.22
                                                 3000
          9
                  0.81
                            0.53
                                      0.64
                                                 2765
         10
                  0.59
                            0.17
                                      0.26
                                                 3051
In [0]:
joblib.dump(classifier, 'lr_with_more_title_weight.pkl')
Out[113]:
['lr_with_more_title_weight.pkl']
```

```
In [0]:
```

```
start = datetime.now()
classifier_2 = OneVsRestClassifier(LogisticRegression(penalty='l1'), n_jobs=-1)
classifier_2.fit(x_train_multilabel, y_train)
predictions_2 = classifier_2.predict(x_test_multilabel)
print("Accuracy :",metrics.accuracy_score(y_test, predictions_2))
print("Hamming loss ",metrics.hamming_loss(y_test,predictions_2))
precision = precision_score(y_test, predictions_2, average='micro')
recall = recall score(y test, predictions 2, average='micro')
f1 = f1_score(y_test, predictions_2, average='micro')
print("Micro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1)
precision = precision_score(y_test, predictions_2, average='macro')
recall = recall_score(y_test, predictions_2, average='macro')
f1 = f1_score(y_test, predictions_2, average='macro')
print("Macro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1)
print (metrics.classification_report(y_test, predictions_2))
print("Time taken to run this cell :", datetime.now() - start)
Accuracy : 0.25108
Hamming loss 0.00270302
Micro-average quality numbers
Precision: 0.7172, Recall: 0.3672, F1-measure: 0.4858
Macro-average quality numbers
Precision: 0.5570, Recall: 0.2950, F1-measure: 0.3710
                          recall f1-score
             precision
          0
                  0.94
                            0.72
                                      0.82
                                                 5519
          1
                  0.70
                            0.34
                                      0.45
                                                 8190
          2
                  0.80
                            0.42
                                      0.55
                                                 6529
          3
                  0.82
                            0.49
                                      0.61
                                                 3231
          4
                  0.80
                            0.44
                                      0.57
                                                 6430
          5
                            0.38
                  0.82
                                      0.52
                                                 2879
          6
                  0.86
                            0.53
                                      0.66
                                                 5086
          7
                            0.58
                                      0.70
                                                 4533
                  0.87
          8
                  0.60
                            0.13
                                      0.22
                                                 3000
          9
                            0.57
                                                 2765
                  0.82
                                      0.67
                                                 3051
         10
                  0.60
                            0.20
                                      0.30
```

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 [2]:

```
# setting path
par_path = os.path.normpath(os.getcwd() + os.sep + os.pardir)
dir_path = os.path.join(par_path, 'stackoverflow-tag-predictor', 'datasets')
dir_path
```

Out[2]:

'/home/shekhar_bavanari/notebooks/stackoverflow-tag-predictor/datasets'

In [3]:

```
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)
        print("Error! cannot create the database connection.")
    conn.close()
```

```
In [4]:
```

```
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 [5]:

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

In [6]:

```
data.head()
```

Out[6]:

	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 [7]:

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

In [8]:

```
total_qs=data.shape[0]
```

In [9]:

```
# 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 [10]:
```

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

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

In [11]:

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

```
Dimensions of train data X: (400000, 2) Y: (400000, 500) Dimensions of test data X: (100000, 2) Y: (100000, 500)
```

In [12]:

```
vectorizer = CountVectorizer(
    min_df=0.00009, max_features=200000, tokenizer = lambda x: x.split(),
    ngram_range=(1,4)
)
x_train_multilabel = vectorizer.fit_transform(x_train['question'])
x_test_multilabel = vectorizer.transform(x_test['question'])
```

In [13]:

```
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)
```

In [14]:

```
from sklearn.model_selection import GridSearchCV
from sklearn.svm import SVC
```

[5.1] Logistic Regression

```
In [24]:
```

```
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 : 0.1084
Hamming loss 0.00603892
Micro-average quality numbers
Precision: 0.2827, Recall: 0.4797, F1-measure: 0.3558
Macro-average quality numbers
Precision: 0.2056, Recall: 0.4107, F1-measure: 0.2658
              precision
                           recall f1-score
                             0.80
                                       0.76
           0
                   0.72
                                                 5519
                                       0.45
           1
                   0.44
                             0.46
                                                 8190
           2
                   0.50
                             0.53
                                       0.52
                                                 6529
           3
                                       0.55
                   0.50
                             0.61
                                                 3231
           4
                   0.55
                             0.53
                                       0.54
                                                 6430
           5
                   0.40
                             0.51
                                       0.45
                                                 2879
           6
                   0.59
                             0.61
                                       0.60
                                                 5086
           7
                   0.59
                                       0.63
                                                 4533
                             0.67
           8
                   0.20
                             0.23
                                       0.22
                                                 3000
           9
                   0.58
                             0.62
                                       0.60
                                                 2765
          10
                   0.32
                             0.33
                                       0.33
                                                 3051
```

[5.2] Hyperparameter tuning Logistic Regression

In [15]:

```
start = datetime.now()
classifier = OneVsRestClassifier(LogisticRegression(penalty='l1'))
params = {
    'estimator C': [
        10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4
}
grid = GridSearchCV(classifier, params, cv=2, verbose=2, n_jobs=-2, scoring='f1_micro')
grid.fit(x_train_multilabel, y_train)
print("Grid best params:", grid.best params )
classifier = OneVsRestClassifier(
    LogisticRegression(C=grid.best_params_['estimator__C'], penalty='11'), n_jobs=-2
)
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)
Fitting 2 folds for each of 9 candidates, totalling 18 fits
[Parallel(n jobs=-2)]: Using backend LokyBackend with 7 concurrent worker
[Parallel(n_jobs=-2)]: Done 15 out of 18 | elapsed: 430.2min remaining:
86.0min
[Parallel(n jobs=-2)]: Done 18 out of 18 | elapsed: 525.4min finished
```

[5.3] Linear SVM

```
In [20]:
```

```
start = datetime.now()
classifier = OneVsRestClassifier(SGDClassifier(loss='hinge', alpha=0.00001, penalty='11'),
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: 0.10885
Hamming loss 0.0059297
Micro-average quality numbers
Precision: 0.2888, Recall: 0.4826, F1-measure: 0.3613
Macro-average quality numbers
Precision: 0.2097, Recall: 0.4082, F1-measure: 0.2694
                           recall f1-score
              precision
                                              support
           0
                   0.72
                             0.80
                                       0.76
                                                 5519
           1
                   0.44
                             0.44
                                       0.44
                                                 8190
           2
                   0.49
                             0.53
                                       0.51
                                                 6529
           3
                   0.49
                                       0.54
                             0.61
                                                 3231
           4
                   0.53
                             0.55
                                       0.54
                                                 6430
           5
                                       0.46
                   0.42
                             0.50
                                                 2879
           6
                   0.57
                                       0.60
                                                 5086
                             0.63
           7
                   0.59
                             0.67
                                       0.63
                                                 4533
           8
                   0.21
                             0.23
                                       0.22
                                                 3000
           9
                   0.54
                             0.67
                                       0.59
                                                 2765
                                       0.33
          10
                   0.32
                             0.36
                                                 3051
```

[6] Conclusion

Procedure followed

1. Understood Business problem

- 2. Understood Machine learning problem Understood data and its fields
- Did exploratory data analysis
 Data loading with Pandas and SQLite
 Counting total number of datapoints
 Checking for duplicates
- 4. Analysis of Tags

Total number of unique tags

Number of times a tag appeared

Tags Per Question

Most Frequent Tags

The top 20 tags

5. Cleaning and preprocessing of Questions

Separate Code from Body

Remove Special characters from Question title and description

Give more weightage to title: Add title three times to the question

Selected 0.5M datapoints

500 tags to represent questions

- 6. Converting tags for multilabel problems
- 7. Building machine Learning Models

Featurizing data with BOW vectorizer

Applying Logistic Regression with OneVsRest Classifier and BOW vectorizer

Applying Logistic Regression with OneVsRest Classifier, hyperparameter tuning and BOW vectorizer

Applying Linear SVM with OneVsRest Classifier and BOW vectorizer

In [1]:

```
from prettytable import PrettyTable
```

In [5]:

```
x = PrettyTable()
x.field_names = ["Algorithm", "Vectorizer", "Hyperparameters", "Micro F1"]
```

In [6]:

```
x.add_row(["Logistic regression", "BoW", "alpha->0.00001", 0.4763])
x.add_row(["Logistic regression", "BoW", "C->1", 0.4762])
x.add_row(["Linear SVM", "BoW", "alpha->0.00001", 0.3558])
print(x)
```

Algorithm	Vectorizer	+ Hyperparameters	++ Micro F1
Logistic regression	BoW	alpha->0.00001	0.4763
Logistic regression	BoW	C->1	0.4762
Linear SVM	BoW	alpha->0.00001	0.3558

Observations

- Logistic regression seems to out perform Linear SVM.
- C=1 -> Tuned hyperparameter for Logistic Regression using GridSearchCV.