

bold text

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 sklearnlearn.adapt import mlknn
from sklearnlearn.problem_transform import ClassifierChain
from sklearnlearn.problem_transform import BinaryRelevance
from sklearnlearn.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 Statement

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/nNDqbUhtIRg> (<https://youtu.be/nNDqbUhtIRg>)

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 Explanation

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-separated format (all lowercase, should not contain tabs '\t' or ampersands '&')

2.1.2 Example Data point

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

Body :

```

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

```

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

2.2.1 Type of Machine Learning Problem

It is a multi-label classification problem

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://www.kaggle.com/wiki/HammingLoss>).

3. Exploratory Data Analysis

3.1 Data Loading and Cleaning

3.1.1 Using Pandas with SQLite to Load the data

In [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'], chunks
size=chunksize, iterator=True, encoding='utf-8', ):
        df.index += index_start
        j+=1
        print('{} rows'.format(j*chunksize))
        df.to_sql('data', disk_engine, if_exists='append')
        index_start = df.index[-1] + 1
    print("Time taken to run this cell :", datetime.now() - start)
```

3.1.2 Counting the number of rows

In [3]:

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

Number of rows in the database :

6034196

Time taken to count the number of rows : 0:05:14.063909

3.1.3 Checking for duplicates

In [4]:

```
#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 FRO
M data GROUP BY Title, Body, Tags', con)
    con.close()
    print("Time taken to run this cell :", datetime.now() - start)
else:
    print("Please download the train.db file from drive or run the first to generate
train.db file")
```

Time taken to run this cell : 0:02:44.967777

In [5]:

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

Out[5]:

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

In [6]:

```
print("number of duplicate questions :", num_rows['count(*)'].values[0]- df_no_dup.s  
hape[0], "(", (1-((df_no_dup.shape[0])/(num_rows['count(*)'].values[0]))) *100, "% )")  
number of duplicate questions : 1827881 ( 30.292038906260256 % )
```

In [7]:

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

Out[7]:

```
1    2656284  
2    1272336  
3     277575  
4         90  
5         25  
6          5  
Name: cnt_dup, dtype: int64
```


In [11]:

```
#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 generate train.db file")
```

Time taken to run this cell : 0:00:54.166267

3.2 Analysis of Tags

3.2.1 Total number of unique tags

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())
# 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 : 42048

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',
' .bash-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]:

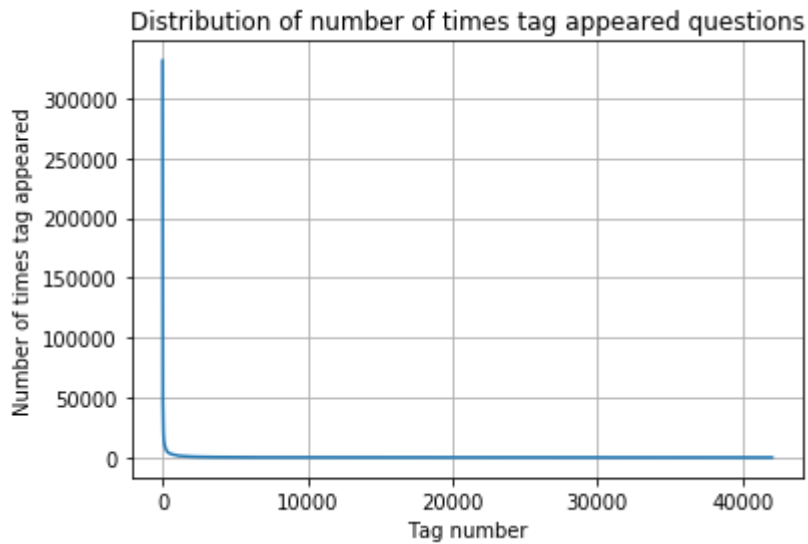
	Tags	Counts
0	.a	18
1	.app	37
2	.asp.net-mvc	1
3	.aspxauth	21
4	.bash-profile	138

In [17]:

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

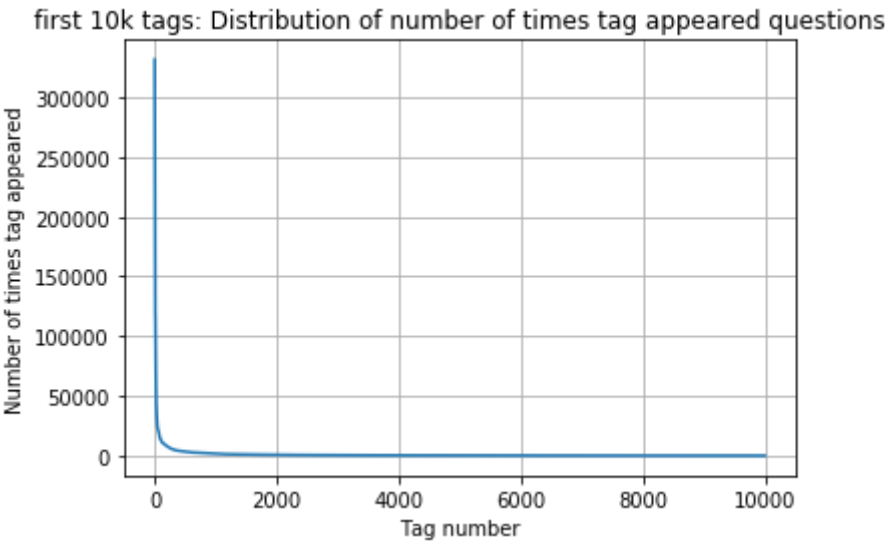
In [18]:

```
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])
```

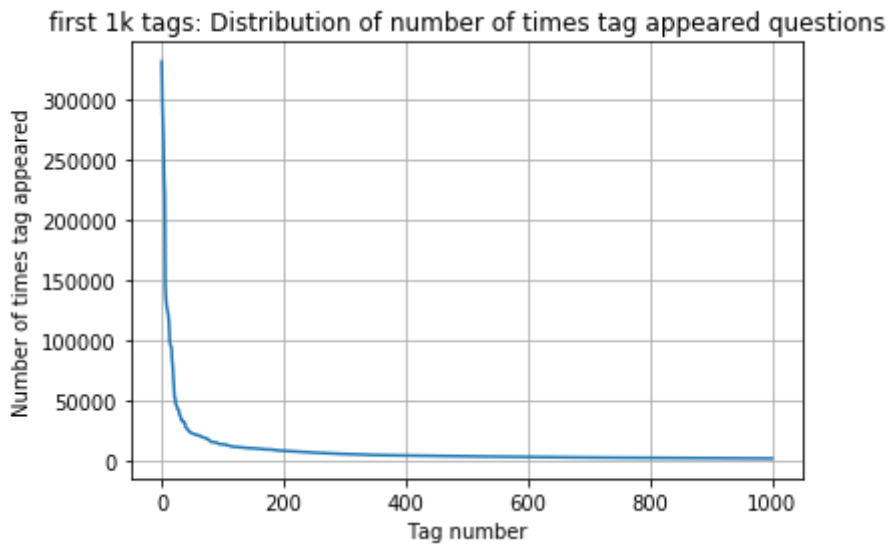


```

400 [331505 44829 22429 17728 13364 11162 10029 9148 8054 71
51
6466 5865 5370 4983 4526 4281 4144 3929 3750 3593
3453 3299 3123 2989 2891 2738 2647 2527 2431 2331
2259 2186 2097 2020 1959 1900 1828 1770 1723 1673
1631 1574 1532 1479 1448 1406 1365 1328 1300 1266
1245 1222 1197 1181 1158 1139 1121 1101 1076 1056
1038 1023 1006 983 966 952 938 926 911 891
882 869 856 841 830 816 804 789 779 770
752 743 733 725 712 702 688 678 671 658
650 643 634 627 616 607 598 589 583 577
568 559 552 545 540 533 526 518 512 506
500 495 490 485 480 477 469 465 457 450
447 442 437 432 426 422 418 413 408 403
398 393 388 385 381 378 374 370 367 365
361 357 354 350 347 344 342 339 336 332
330 326 323 319 315 312 309 307 304 301
299 296 293 291 289 286 284 281 278 276
275 272 270 268 265 262 260 258 256 254
252 250 249 247 245 243 241 239 238 236
234 233 232 230 228 226 224 222 220 219
217 215 214 212 210 209 207 205 204 203
201 200 199 198 196 194 193 192 191 189
188 186 185 183 182 181 180 179 178 177
175 174 172 171 170 169 168 167 166 165
164 162 161 160 159 158 157 156 156 155
154 153 152 151 150 149 149 148 147 146
145 144 143 142 142 141 140 139 138 137
137 136 135 134 134 133 132 131 130 130
129 128 128 127 126 126 125 124 124 123
123 122 122 121 120 120 119 118 118 117
117 116 116 115 115 114 113 113 112 111
111 110 109 109 108 108 107 106 106 106
105 105 104 104 103 103 102 102 101 101
100 100 99 99 98 98 97 97 96 96
95 95 94 94 93 93 93 92 92 91
91 90 90 89 89 88 88 87 87 86
86 86 85 85 84 84 83 83 83 82
82 82 81 81 80 80 80 79 79 78
78 78 78 77 77 76 76 76 75 75
75 74 74 74 73 73 73 73 72 72]
```

In [20]:

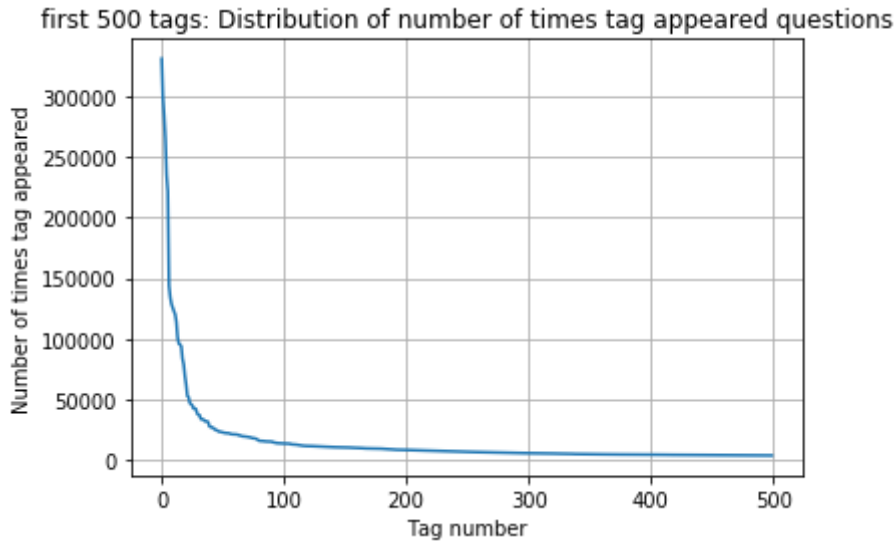
```
plt.plot(tag_counts[0:1000])
plt.title('first 1k tags: Distribution of number of times tag appeared questions')
plt.grid()
plt.xlabel("Tag number")
plt.ylabel("Number of times tag appeared")
plt.show()
print(len(tag_counts[0:1000:5]), tag_counts[0:1000:5])
```



```
200 [331505 221533 122769 95160 62023 44829 37170 31897 26925 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 3483
3453 3427 3396 3363 3326 3299 3272 3232 3196 3168
3123 3094 3073 3050 3012 2989 2984 2953 2934 2903
2891 2844 2819 2784 2754 2738 2726 2708 2681 2669
2647 2621 2604 2594 2556 2527 2510 2482 2460 2444
2431 2409 2395 2380 2363 2331 2312 2297 2290 2281
2259 2246 2222 2211 2198 2186 2162 2142 2132 2107
2097 2078 2057 2045 2036 2020 2011 1994 1971 1965
1959 1952 1940 1932 1912 1900 1879 1865 1855 1841
1828 1821 1813 1801 1782 1770 1760 1747 1741 1734
1723 1707 1697 1688 1683 1673 1665 1656 1646 1639]
```


In [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 [331505 221533 122769 95160 62023 44829 37170 31897 26925 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 3483]
```

In [22]:

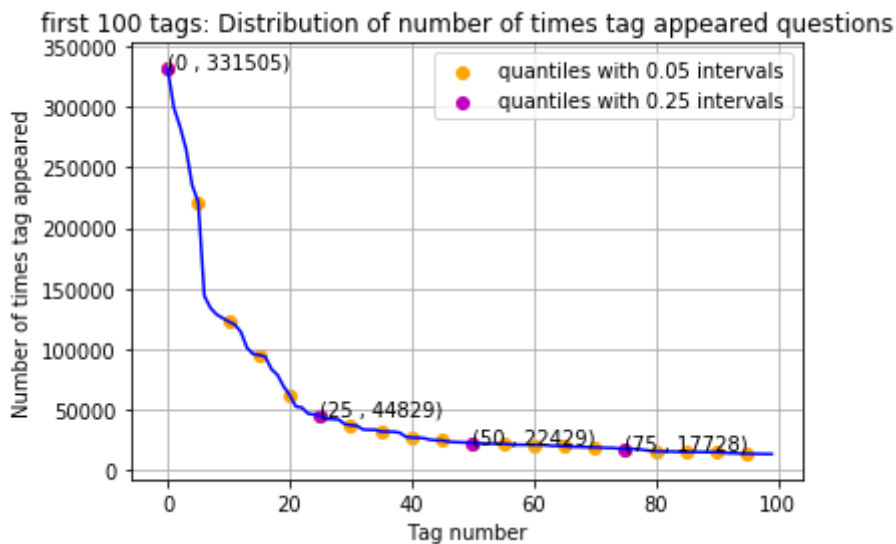
```

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

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

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

```



```

20 [331505 221533 122769 95160 62023 44829 37170 31897 26925 2453
7
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 frequently than others, Micro-averaged F1-score is the appropriate metric for this problem.

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 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 [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)))
```

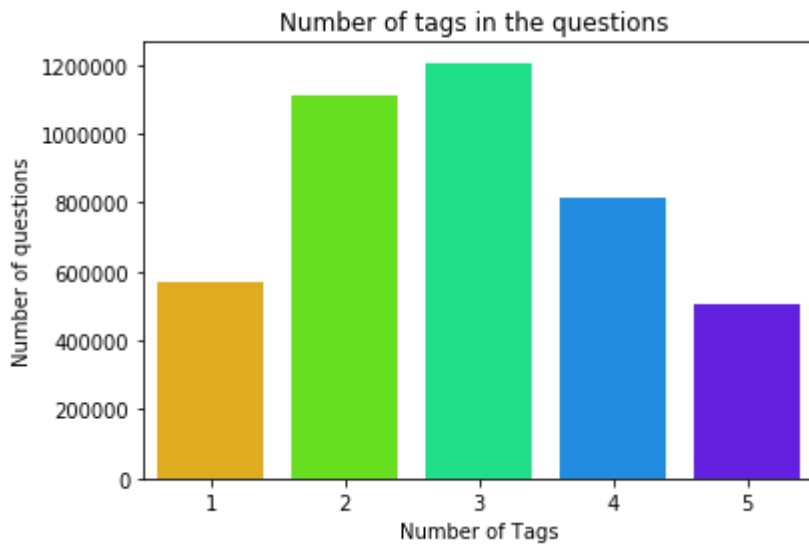
Maximum number of tags per question: 5

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

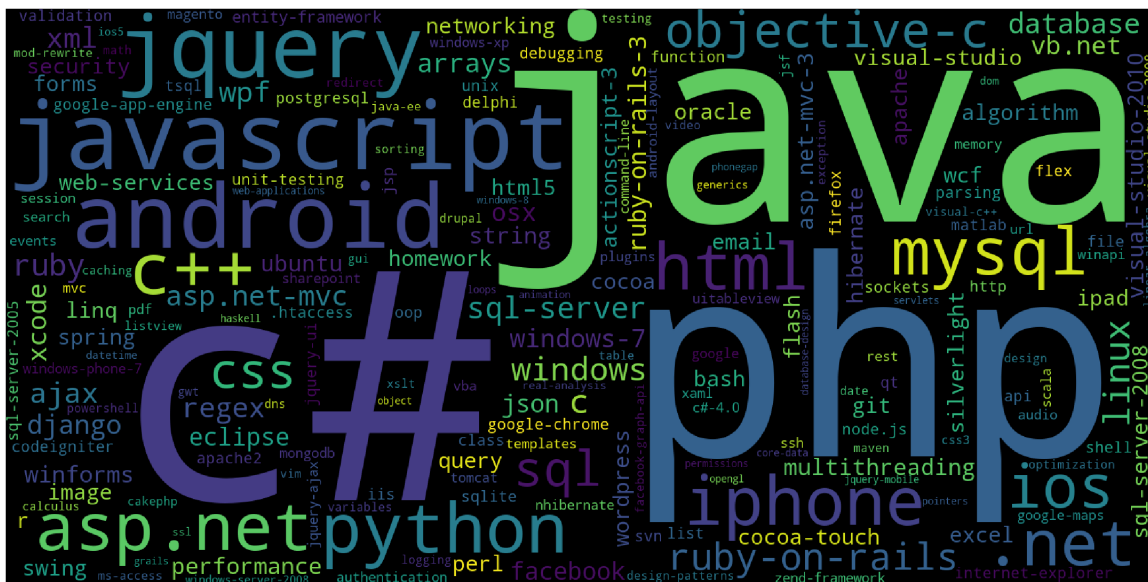
3.2.5 Most Frequent Tags

```
# Plotting 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)
```

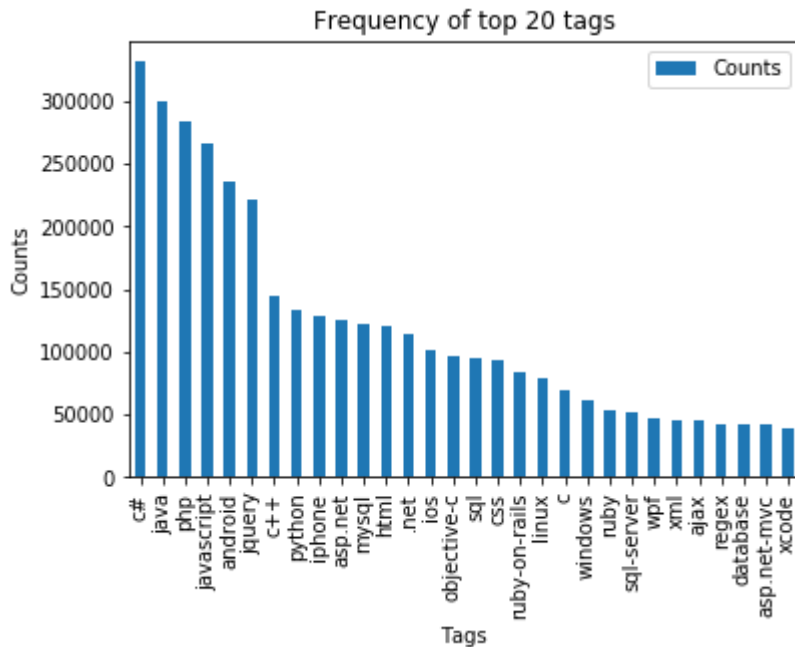


Observations:

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()
```



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

In [29]:

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

```
[nltk_data] Downloading package stopwords to
[nltk_data]   /home/shalini3316/nltk_data...
[nltk_data]   Package stopwords is already up-to-date!
```

Out[29]:

True

In [30]:

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

In [31]:

```

#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, is_code integer);"""
create_database_table("Processed.db", sql_create_table)

```

Tables in the databse:
QuestionsProcessed

In [32]:

```
# 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 1000000;")

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 database:

QuestionsProcessed

Cleared All the rows

Time taken to run this cell : 0:02:53.398210

In [33]:

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

```
[nltk_data] Downloading package punkt to
[nltk_data]      /home/shalini3316/nltk_data...
[nltk_data]   Package punkt is already up-to-date!
```

Out[33]:

True

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

In [34]:

```
#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 Lett
er 'c'
    question=' '.join(str(stemmer.stem(j)) for j in words if j not in stop_words and
(len(j)!=1 or j=='c'))

    len_post+=len(question)
    tup = (question,code,tags,x,len(question),is_code)
    questions_proccesed += 1
    writer.execute("insert into QuestionsProcessed(question,code,tags,words_pre,word
s_post,is_code) values (?,?,?,?,?,?)",tup)
    if (questions_proccesed%100000==0):
        print("number of questions completed=",questions_proccesed)

no_dup_avg_len_pre=(len_pre*1.0)/questions_proccesed
no_dup_avg_len_post=(len_post*1.0)/questions_proccesed

print( "Avg. length of questions(Title+Body) before processing: %d"%no_dup_avg_len_p
re)
print( "Avg. length of questions(Title+Body) after processing: %d"%no_dup_avg_len_po
st)
print ( "Percent of questions containing code: %d"%((questions_with_code*100.0)/quest
ions_proccesed))

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

```
number of questions completed= 100000  
number of questions completed= 200000  
number of questions completed= 300000  
number of questions completed= 400000  
number of questions completed= 500000  
number of questions completed= 600000  
number of questions completed= 700000  
number of questions completed= 800000  
number of questions completed= 900000  
Avg. length of questions(Title+Body) before processing: 1172  
Avg. length of questions(Title+Body) after processing: 326  
Percent of questions containing code: 57  
Time taken to run this cell : 0:24:02.585513
```

In [35]:

```
# 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 [36]:

```
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

```
=====
('advantag disadvantag paramet return type declar languag type infer wou
ld like know opinion declar hand paramet return type languag type infer
like scala reason compil infer type',)
-----
```

```
-----
('want close statement automat want close statement automat want achiev
use technolog follow packag java spring framework seem close statement a
utomat default set though think transact manag spring automat shut state
ment want call statement much possibl method maintain method ensur state
ment close nthe amount code increas come method call close statement pro
blem possibl omiss generat goe moreov make framework want make restrict
much possibl reason make littl method moreov offici site document show r
eason close method ni japanes pleas explain use simpl statement',)
-----
```

```
-----
('creat ado net entiti primari key less view store procedur use either v
iew store procedur return result execut mdx queri sql analysi servic use
openrowset link ssas server sql unfortun resultset includ pk field matte
r even column could act pk hack model anyon think hacktacular trick use
add field result insert auto increment valu act pk candid somehow set fi
eld pk easili build entiti sure answer point already get data back ssas
use ssrs report odata render creat type manual use play linq someon nail
amp may potenti better way access data sinc model would creat type entit
i thank',)
-----
```

```
-----
('get process belong particular session want get process meant particula
r session ni two person use web applic want get process use user process
use user anyon pleas help',)
-----
```

```
-----
('custom rule browser dictionari mani compani organ combin two word capi
t form name ex superus stackexchang word individu spell correct howev sp
ace browser get confus underlin let us know think misspel get googl chro
me mozilla firefox even microsoft internet explor microsoft offic recogn
capit letter middl word begin new word therefor mark word misspel unless
nd word misspel ex superusr',)
-----
```

```
-----
('possibl authent activ directori user expir password web form use ad au
thent user want abl authent user expir password redirect password chang
page authent instanc site admin reset user password use method make user
reset password next logon make user password expir user tri logon new pa
ssword authent fail logon failur unknown usernam bad password auth metho
d way around thank help',)
-----
```

```
-----
('initi memcach jdbc resourc jax rs servlet servic want maintain data pe
rsist mysql db use jdbc experi build jdbc app jax rs app isol never comb
in two question build tear requir jdbc type stuff go ordinarily put buil
d static block constructor id cleanup method get call final doesnt seem
work jax rs framework constructor get call everi invoc place knowledg pu
t clean method unfortun spars exampl combin two technolog onlin someth f
ind surpris guy help',)
-----
```

```
-----
('onnewint lifecycl regist listen use singletop activ receiv intent sear
ch dialog via notic call afterward call visual search dialog initi searc
```

```
h intent fire activ problem listen regist get remov need insid call stan
dard way make listen avail',)
```

```
('wrong url redirect index page type xyz com jargon take xyz com index p
hp use apach mod rewr it rule index page even default page setup director
yindex puzzl takesto index page',)
```

In [37]:

```
#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 Question
sProcessed""", conn_r)
    conn_r.commit()
    conn_r.close()
```

In [38]:

```
preprocessed_data.head()
```

Out[38]:

	question	tags
0	fail call design initi nsmanagedobject class b...	objective-c ios xcode parsing xcode4
1	advantag disadvantag paramet return type decla...	programming-languages
2	want close statement automat want close statem...	java spring statement
3	creat ado net entiti primari key less view sto...	sql sql-server wcf ssas ado.net-entity-data-model
4	get process belong particular session want get...	c# asp.net session process

In [39]:

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

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

4. Machine Learning Models

4.1 Converting tags for multilabel problems

X	y1	y2	y3	y4
x1	0	1	1	0
x1	1	0	0	0
x1	0	1	0	0

In [40]:

```
# 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 [41]:

```
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

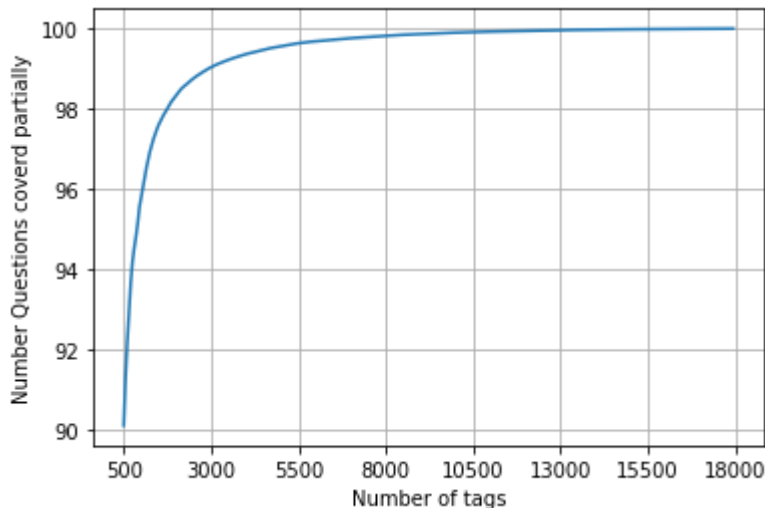
def questions_explained_fn(n):
    multilabel_yn = tags_to_choose(n)
    x= multilabel_yn.sum(axis=1)
    return (np.count_nonzero(x==0))
```

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)*100,3))
```

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 covered partially")
plt.grid()
plt.show()
# you can choose any number of tags based on your computing power, minimum is 50(it covers 90% of the tags)
print("with ",5500,"tags we are covering ",questions_explained[50],"% of questions")
```



with 5500 tags we are covering 99.035 % of questions

In [44]:

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

number of questions that are not covered : 9645 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_y.shape[1])*100, "%")
```

Number of tags in sample : 35455

number of tags taken : 5500 (15.512621633055987 %)

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)

In [48]:

```
x_train = x_train[:200000]
x_test = x_test[:40000]
y_train = y_train[:200000]
y_test = y_test[:40000]
```

4.3 Featurizing data

In [49]:

```
start = datetime.now()
vectorizer = TfidfVectorizer(min_df=0.00009, max_features=200000, smooth_idf=True, norm="l2", \
                             tokenizer = lambda x: x.split(), sublinear_tf=False, ngram_range=(1,3))
x_train_multilabel = vectorizer.fit_transform(x_train['question'])
x_test_multilabel = vectorizer.transform(x_test['question'])
print("Time taken to run this cell :", datetime.now() - start)
```

Time taken to run this cell : 0:01:55.017595

In [50]:

```
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: (200000, 92365) Y : (200000, 5500)
 Dimensions of test data X: (40000, 92365) Y: (40000, 5500)

In [51]:

```
# https://www.analyticsvidhya.com/blog/2017/08/introduction-to-multi-label-classification/
#https://stats.stackexchange.com/questions/117796/scikit-multi-label-classification
# classifier = LabelPowerset(GaussianNB())
"""
from skmultilearn.adapt import MLkNN
classifier = MLkNN(k=21)

# train
classifier.fit(x_train_multilabel, y_train)

# predict
predictions = classifier.predict(x_test_multilabel)
print(accuracy_score(y_test, predictions))
print(metrics.f1_score(y_test, predictions, average = 'macro'))
print(metrics.f1_score(y_test, predictions, average = 'micro'))
print(metrics.hamming_loss(y_test, predictions))

"""
# 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[51]:

```
"\nfrom skmultilearn.adapt import MLkNN\nnclassifier = MLkNN(k=21)\n\n# t
rain\nnclassifier.fit(x_train_multilabel, y_train)\n\n# predict\nnpredicti
ons = classifier.predict(x_test_multilabel)\n\nprint(accuracy_score(y_tes
t, predictions))\n\nprint(metrics.f1_score(y_test, predictions, average =
'macro'))\n\nprint(metrics.f1_score(y_test, predictions, average = 'micr
o'))\n\nprint(metrics.hamming_loss(y_test, predictions))\n\n"
```

4.4 Applying Logistic Regression with OneVsRest Classifier

In [52]:

```
# this will be taking so much time try not to run it, download the lr_with_equal_weight.pkl file and use to predict
# This takes about 6-7 hours to run.
classifier = OneVsRestClassifier(SGDClassifier(loss='log', alpha=0.00001, penalty='l1'), n_jobs=-1)
classifier.fit(x_train_multilabel, y_train)
predictions = classifier.predict(x_test_multilabel)

print("accuracy :",metrics.accuracy_score(y_test,predictions))
print("macro f1 score :",metrics.f1_score(y_test, predictions, average = 'macro'))
print("micro f1 score :",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 : 0.0803

macro f1 score : 0.08432376291016838

micro f1 scoore : 0.37235972797407163

hamming loss : 0.000411954545454543

Precision recall report :

	precision	recall	f1-score	support
0	0.63	0.23	0.34	3208
1	0.79	0.44	0.57	2833
2	0.82	0.56	0.66	2755
3	0.76	0.43	0.55	2611
4	0.94	0.75	0.83	2215
5	0.85	0.63	0.73	2096
6	0.67	0.30	0.42	1346
7	0.88	0.60	0.71	1256
8	0.69	0.42	0.52	1179
9	0.81	0.44	0.57	1188
10	0.85	0.61	0.71	1239
11	0.53	0.15	0.24	1110
12	0.48	0.09	0.16	1089
13	0.63	0.28	0.39	947
14	0.59	0.28	0.38	965
15	0.60	0.21	0.31	896
16	0.76	0.53	0.62	879
17	0.79	0.54	0.64	758
18	0.62	0.24	0.35	742
19	0.57	0.16	0.25	644
20	0.35	0.05	0.09	625
21	0.84	0.37	0.52	510
22	0.59	0.25	0.35	523
23	0.84	0.59	0.70	430
24	0.64	0.42	0.50	428
25	0.68	0.34	0.45	441
26	0.86	0.67	0.76	406
27	0.25	0.04	0.07	418
28	0.62	0.26	0.37	423
29	0.69	0.24	0.35	372
30	0.93	0.76	0.84	343
31	0.49	0.25	0.34	375
32	0.56	0.23	0.33	328
33	0.42	0.12	0.19	328
34	0.80	0.29	0.43	330
35	0.74	0.52	0.61	291
36	0.75	0.52	0.62	300
37	0.76	0.61	0.68	318
38	0.35	0.13	0.19	264
39	0.47	0.16	0.24	297
40	0.66	0.38	0.48	254
41	0.41	0.10	0.16	236
42	0.67	0.29	0.40	245
43	0.42	0.08	0.14	246
44	0.44	0.13	0.20	237
45	0.65	0.31	0.42	236
46	0.30	0.08	0.13	223
47	0.50	0.02	0.04	243
48	0.56	0.18	0.27	234
49	0.57	0.10	0.17	228
50	0.56	0.16	0.25	225
51	0.65	0.32	0.43	222
52	0.79	0.49	0.61	210
53	0.48	0.10	0.17	222

54	0.58	0.18	0.28	230
55	0.68	0.39	0.50	218
56	0.39	0.12	0.18	180
57	0.85	0.72	0.78	210
58	0.79	0.51	0.62	203
59	0.37	0.07	0.11	222
60	0.10	0.01	0.02	200
61	0.80	0.44	0.57	208
62	0.89	0.74	0.81	190
63	0.93	0.61	0.74	198
64	0.70	0.49	0.58	159
65	0.78	0.19	0.31	197
66	0.71	0.37	0.48	188
67	0.43	0.14	0.21	205
68	0.51	0.14	0.22	182
69	0.72	0.25	0.37	190
70	0.77	0.45	0.57	172
71	0.82	0.39	0.53	180
72	0.75	0.19	0.30	161
73	0.75	0.02	0.03	182
74	0.52	0.35	0.42	156
75	0.89	0.57	0.69	148
76	0.54	0.23	0.33	209
77	0.71	0.45	0.55	175
78	0.25	0.01	0.01	162
79	0.48	0.09	0.15	166
80	0.74	0.30	0.42	152
81	0.24	0.05	0.08	158
82	0.60	0.24	0.35	139
83	0.49	0.23	0.31	134
84	0.93	0.60	0.73	146
85	0.82	0.40	0.54	141
86	0.91	0.59	0.72	147
87	0.47	0.24	0.32	126
88	0.79	0.58	0.67	150
89	0.89	0.49	0.63	148
90	0.73	0.43	0.54	169
91	0.53	0.06	0.10	138
92	0.67	0.50	0.57	122
93	0.58	0.37	0.45	130
94	0.32	0.08	0.12	130
95	0.94	0.67	0.78	140
96	0.65	0.32	0.43	143
97	0.89	0.53	0.66	144
98	0.93	0.70	0.80	138
99	0.65	0.10	0.17	135
100	0.95	0.71	0.81	124
101	0.31	0.03	0.05	140
102	0.59	0.20	0.30	129
103	0.84	0.58	0.69	124
104	0.16	0.03	0.04	115
105	0.53	0.07	0.13	120
106	0.33	0.12	0.17	103
107	0.47	0.23	0.31	137
108	0.29	0.04	0.06	111
109	0.74	0.37	0.49	121
110	0.67	0.06	0.11	100
111	0.51	0.19	0.27	123
112	0.69	0.40	0.50	116
113	0.73	0.34	0.46	118
114	0.54	0.28	0.37	89

115	0.61	0.26	0.36	108
116	0.50	0.18	0.27	126
117	0.94	0.62	0.75	126
118	0.68	0.25	0.36	106
119	0.91	0.64	0.75	99
120	0.86	0.49	0.62	113
121	0.40	0.18	0.25	93
122	0.55	0.24	0.33	101
123	0.37	0.13	0.19	115
124	0.15	0.03	0.05	93
125	0.42	0.09	0.15	106
126	0.71	0.48	0.57	117
127	0.59	0.27	0.37	86
128	0.99	0.81	0.89	115
129	0.28	0.10	0.15	88
130	0.20	0.02	0.03	113
131	0.46	0.11	0.18	98
132	0.24	0.06	0.09	90
133	0.39	0.11	0.17	104
134	0.55	0.39	0.46	94
135	0.69	0.52	0.59	89
136	0.66	0.38	0.48	104
137	0.21	0.04	0.07	101
138	0.21	0.06	0.09	99
139	0.51	0.30	0.38	105
140	0.51	0.25	0.34	99
141	0.76	0.60	0.68	86
142	0.78	0.47	0.59	108
143	0.65	0.43	0.52	84
144	0.52	0.24	0.33	98
145	0.62	0.25	0.36	102
146	0.27	0.04	0.07	105
147	0.00	0.00	0.00	97
148	0.48	0.13	0.20	86
149	0.62	0.21	0.31	96
150	0.25	0.02	0.04	104
151	0.30	0.08	0.12	93
152	0.14	0.04	0.06	100
153	0.55	0.07	0.13	82
154	0.93	0.57	0.71	74
155	0.86	0.56	0.68	114
156	0.08	0.01	0.02	81
157	0.31	0.06	0.10	86
158	0.59	0.12	0.20	111
159	0.32	0.15	0.20	94
160	0.38	0.14	0.21	77
161	0.92	0.60	0.72	109
162	0.46	0.31	0.37	77
163	0.74	0.23	0.35	108
164	0.47	0.09	0.16	95
165	0.10	0.02	0.04	86
166	0.50	0.02	0.04	87
167	0.78	0.22	0.34	82
168	0.58	0.23	0.33	97
169	0.52	0.16	0.25	93
170	0.89	0.54	0.67	78
171	0.53	0.12	0.19	76
172	0.86	0.49	0.63	87
173	0.55	0.13	0.22	82
174	0.86	0.60	0.71	95
175	0.46	0.19	0.27	83

176	0.73	0.41	0.52	108
177	0.91	0.64	0.75	75
178	0.64	0.46	0.54	78
179	0.51	0.25	0.34	87
180	0.43	0.07	0.12	85
181	0.53	0.34	0.42	73
182	0.43	0.11	0.18	81
183	0.44	0.10	0.16	80
184	0.70	0.52	0.59	85
185	0.14	0.01	0.02	81
186	0.00	0.00	0.00	85
187	0.29	0.06	0.10	82
188	0.50	0.01	0.02	85
189	0.64	0.29	0.40	73
190	0.53	0.10	0.16	83
191	0.92	0.57	0.71	84
192	0.88	0.51	0.65	70
193	0.46	0.23	0.31	74
194	0.36	0.18	0.24	71
195	0.73	0.37	0.49	81
196	0.77	0.53	0.62	57
197	0.55	0.37	0.44	65
198	0.89	0.56	0.69	89
199	0.14	0.04	0.06	74
200	0.54	0.25	0.34	87
201	0.00	0.00	0.00	84
202	0.91	0.49	0.63	80
203	0.25	0.01	0.03	74
204	0.74	0.38	0.50	73
205	0.62	0.27	0.37	75
206	0.65	0.39	0.49	71
207	0.39	0.10	0.15	73
208	0.44	0.09	0.15	79
209	0.72	0.39	0.51	59
210	0.21	0.04	0.07	71
211	0.33	0.03	0.05	78
212	0.22	0.03	0.06	61
213	0.25	0.04	0.07	70
214	0.48	0.18	0.27	71
215	0.79	0.40	0.53	65
216	0.89	0.73	0.80	56
217	0.37	0.10	0.15	72
218	0.70	0.53	0.60	66
219	0.42	0.12	0.18	69
220	0.79	0.40	0.53	77
221	0.93	0.67	0.78	64
222	0.63	0.39	0.48	70
223	0.96	0.71	0.81	68
224	0.74	0.29	0.41	59
225	0.37	0.18	0.24	61
226	0.54	0.33	0.41	63
227	0.41	0.12	0.18	59
228	0.41	0.19	0.26	64
229	0.33	0.03	0.06	64
230	0.69	0.38	0.49	58
231	0.64	0.15	0.24	60
232	0.30	0.05	0.08	62
233	0.69	0.46	0.55	68
234	0.00	0.00	0.00	59
235	0.44	0.10	0.16	80
236	0.69	0.38	0.49	82

237	0.00	0.00	0.00	61
238	0.50	0.08	0.14	63
239	0.60	0.37	0.46	73
240	0.19	0.05	0.07	64
241	0.81	0.57	0.67	68
242	0.53	0.34	0.42	67
243	0.77	0.44	0.56	61
244	0.40	0.12	0.19	66
245	0.50	0.11	0.18	73
246	0.33	0.22	0.26	55
247	0.35	0.10	0.16	60
248	0.71	0.33	0.45	73
249	0.64	0.32	0.43	56
250	0.20	0.02	0.03	63
251	0.81	0.48	0.60	60
252	0.52	0.18	0.26	68
253	0.25	0.03	0.06	60
254	0.89	0.53	0.66	59
255	0.20	0.02	0.03	61
256	0.40	0.11	0.17	54
257	0.79	0.15	0.26	71
258	0.33	0.05	0.09	57
259	0.72	0.40	0.51	53
260	0.89	0.50	0.64	62
261	0.18	0.03	0.05	63
262	0.85	0.65	0.74	69
263	0.33	0.02	0.04	53
264	0.00	0.00	0.00	57
265	0.25	0.04	0.06	56
266	0.00	0.00	0.00	55
267	0.62	0.36	0.46	55
268	0.79	0.25	0.38	59
269	0.25	0.02	0.03	54
270	0.29	0.03	0.06	60
271	0.18	0.03	0.06	60
272	0.49	0.24	0.32	71
273	0.64	0.47	0.54	60
274	0.84	0.55	0.67	65
275	0.45	0.25	0.33	51
276	0.65	0.22	0.33	50
277	0.75	0.33	0.46	45
278	0.39	0.14	0.21	49
279	0.98	0.75	0.85	55
280	0.56	0.26	0.35	58
281	0.67	0.04	0.07	52
282	0.00	0.00	0.00	51
283	0.44	0.24	0.31	51
284	0.33	0.02	0.04	42
285	0.69	0.47	0.56	47
286	0.29	0.08	0.13	48
287	0.84	0.67	0.74	54
288	0.50	0.02	0.03	57
289	0.27	0.07	0.11	60
290	0.41	0.14	0.21	50
291	0.15	0.04	0.07	48
292	0.48	0.23	0.31	43
293	0.00	0.00	0.00	61
294	0.11	0.02	0.03	63
295	0.29	0.05	0.09	38
296	0.89	0.53	0.67	58
297	0.33	0.04	0.07	52

298	0.81	0.56	0.66	63
299	0.62	0.08	0.14	64
300	0.17	0.02	0.04	50
301	0.33	0.06	0.11	48
302	0.52	0.21	0.30	53
303	0.00	0.00	0.00	46
304	0.58	0.17	0.26	42
305	0.78	0.12	0.21	59
306	0.64	0.13	0.22	52
307	0.90	0.72	0.80	53
308	0.33	0.09	0.14	45
309	0.00	0.00	0.00	51
310	0.00	0.00	0.00	56
311	0.50	0.18	0.26	45
312	0.81	0.54	0.64	54
313	0.50	0.12	0.20	57
314	0.77	0.24	0.37	41
315	0.62	0.27	0.37	49
316	0.00	0.00	0.00	52
317	0.43	0.22	0.29	46
318	0.71	0.35	0.47	34
319	0.29	0.12	0.17	34
320	0.57	0.07	0.12	58
321	0.53	0.17	0.26	47
322	0.46	0.13	0.20	47
323	0.12	0.02	0.04	49
324	0.52	0.23	0.32	52
325	0.50	0.27	0.35	45
326	0.12	0.02	0.03	50
327	0.27	0.07	0.12	41
328	0.00	0.00	0.00	39
329	0.14	0.03	0.05	34
330	0.57	0.33	0.42	48
331	0.20	0.02	0.04	44
332	0.00	0.00	0.00	47
333	0.56	0.29	0.38	48
334	0.37	0.16	0.23	43
335	0.38	0.19	0.25	43
336	0.10	0.02	0.03	48
337	0.50	0.10	0.16	51
338	0.84	0.36	0.50	45
339	0.71	0.33	0.45	45
340	0.00	0.00	0.00	50
341	0.71	0.33	0.45	46
342	0.47	0.21	0.29	39
343	0.69	0.35	0.47	51
344	0.65	0.28	0.39	46
345	0.60	0.06	0.11	49
346	0.36	0.14	0.20	36
347	0.53	0.21	0.30	43
348	0.83	0.42	0.56	45
349	0.69	0.22	0.33	41
350	1.00	0.03	0.06	33
351	0.20	0.04	0.07	45
352	0.10	0.03	0.05	32
353	0.76	0.44	0.56	43
354	0.59	0.25	0.35	52
355	0.00	0.00	0.00	33
356	0.29	0.05	0.09	39
357	0.30	0.15	0.20	48
358	0.45	0.12	0.20	40

359	0.93	0.68	0.79	38
360	0.90	0.44	0.59	41
361	0.54	0.17	0.26	41
362	0.29	0.06	0.09	36
363	0.12	0.03	0.05	36
364	0.10	0.02	0.04	45
365	0.65	0.45	0.53	38
366	0.91	0.56	0.69	36
367	0.38	0.05	0.10	55
368	0.42	0.20	0.27	40
369	0.71	0.43	0.53	47
370	0.92	0.58	0.71	38
371	0.50	0.19	0.28	42
372	0.00	0.00	0.00	45
373	0.60	0.11	0.18	28
374	0.62	0.18	0.28	44
375	0.48	0.24	0.32	42
376	0.80	0.08	0.15	49
377	0.76	0.25	0.38	52
378	0.60	0.47	0.53	32
379	0.30	0.08	0.13	37
380	0.00	0.00	0.00	34
381	0.18	0.05	0.08	39
382	0.67	0.04	0.07	56
383	0.30	0.09	0.13	35
384	0.90	0.64	0.75	44
385	0.25	0.07	0.11	29
386	0.22	0.05	0.09	37
387	0.33	0.03	0.05	39
388	0.50	0.03	0.05	35
389	0.83	0.44	0.58	34
390	0.50	0.04	0.07	28
391	1.00	0.04	0.08	48
392	0.00	0.00	0.00	33
393	0.23	0.07	0.11	42
394	0.94	0.31	0.47	55
395	0.48	0.21	0.29	47
396	0.25	0.09	0.13	44
397	0.73	0.32	0.45	34
398	1.00	0.43	0.60	35
399	0.67	0.45	0.54	31
400	0.27	0.07	0.11	46
401	0.24	0.10	0.14	42
402	0.46	0.31	0.37	35
403	0.14	0.02	0.04	44
404	0.74	0.54	0.62	37
405	0.00	0.00	0.00	49
406	0.00	0.00	0.00	28
407	0.50	0.03	0.06	34
408	0.56	0.12	0.20	42
409	0.33	0.10	0.16	29
410	0.58	0.32	0.42	34
411	0.54	0.19	0.28	37
412	0.00	0.00	0.00	33
413	0.65	0.43	0.52	30
414	1.00	0.03	0.05	38
415	0.50	0.05	0.09	39
416	0.00	0.00	0.00	36
417	0.00	0.00	0.00	36
418	0.42	0.15	0.22	34
419	0.20	0.02	0.04	44

420	0.36	0.14	0.20	36
421	0.00	0.00	0.00	41
422	0.43	0.09	0.14	35
423	0.76	0.44	0.56	36
424	0.23	0.06	0.09	52
425	0.35	0.21	0.26	33
426	0.86	0.38	0.52	32
427	0.70	0.33	0.45	42
428	0.20	0.03	0.05	33
429	0.00	0.00	0.00	42
430	0.75	0.32	0.45	28
431	0.36	0.17	0.23	29
432	0.88	0.51	0.65	41
433	0.57	0.16	0.25	50
434	0.50	0.18	0.27	38
435	0.00	0.00	0.00	27
436	0.33	0.05	0.09	37
437	1.00	0.51	0.68	35
438	0.86	0.14	0.24	43
439	0.50	0.07	0.12	44
440	0.33	0.03	0.05	34
441	0.40	0.17	0.24	36
442	0.67	0.17	0.27	48
443	0.40	0.10	0.16	40
444	0.41	0.17	0.24	41
445	0.00	0.00	0.00	38
446	1.00	0.03	0.06	33
447	0.73	0.24	0.36	33
448	0.47	0.21	0.29	33
449	0.00	0.00	0.00	27
450	0.79	0.31	0.45	35
451	0.00	0.00	0.00	31
452	0.20	0.03	0.05	33
453	0.27	0.15	0.19	27
454	1.00	0.54	0.70	46
455	0.78	0.48	0.60	29
456	0.00	0.00	0.00	37
457	0.50	0.38	0.43	26
458	0.00	0.00	0.00	34
459	0.33	0.07	0.11	29
460	0.38	0.08	0.13	39
461	1.00	0.04	0.08	25
462	0.78	0.55	0.64	33
463	0.52	0.36	0.43	33
464	0.21	0.10	0.14	29
465	0.50	0.19	0.27	27
466	0.92	0.40	0.56	30
467	0.89	0.46	0.61	37
468	0.00	0.00	0.00	34
469	0.50	0.18	0.27	33
470	0.10	0.03	0.05	29
471	0.56	0.31	0.40	29
472	0.18	0.06	0.10	31
473	0.57	0.28	0.37	29
474	0.65	0.44	0.53	34
475	0.56	0.40	0.47	25
476	0.48	0.30	0.37	33
477	0.81	0.55	0.66	47
478	0.00	0.00	0.00	29
479	0.79	0.46	0.58	41
480	0.67	0.40	0.50	20

481	0.64	0.25	0.36	28
482	0.94	0.53	0.68	32
483	0.38	0.06	0.11	48
484	0.00	0.00	0.00	27
485	0.56	0.14	0.22	36
486	1.00	0.39	0.57	33
487	0.72	0.67	0.69	27
488	0.88	0.50	0.64	30
489	0.00	0.00	0.00	32
490	0.50	0.40	0.44	25
491	0.00	0.00	0.00	31
492	0.00	0.00	0.00	38
493	0.17	0.03	0.06	30
494	0.00	0.00	0.00	32
495	0.00	0.00	0.00	30
496	0.25	0.03	0.05	35
497	0.50	0.27	0.35	41
498	0.71	0.14	0.23	37
499	0.67	0.24	0.36	41
500	0.44	0.20	0.27	35
501	0.17	0.02	0.04	51
502	0.50	0.16	0.24	37
503	0.68	0.51	0.58	37
504	0.88	0.51	0.65	41
505	0.33	0.10	0.15	31
506	0.50	0.03	0.05	38
507	0.00	0.00	0.00	41
508	1.00	0.47	0.64	34
509	0.58	0.21	0.31	33
510	0.77	0.57	0.66	35
511	0.90	0.72	0.80	25
512	0.29	0.06	0.10	34
513	0.31	0.20	0.24	25
514	0.29	0.05	0.09	39
515	0.93	0.50	0.65	28
516	0.23	0.10	0.14	29
517	0.00	0.00	0.00	21
518	0.67	0.19	0.29	32
519	0.30	0.10	0.15	31
520	0.85	0.47	0.61	36
521	0.50	0.15	0.23	46
522	0.95	0.60	0.74	35
523	0.38	0.10	0.16	30
524	0.57	0.16	0.25	25
525	0.57	0.17	0.27	46
526	0.88	0.58	0.70	24
527	0.87	0.28	0.43	46
528	1.00	0.48	0.65	21
529	0.71	0.15	0.24	34
530	0.36	0.12	0.19	32
531	0.71	0.50	0.59	30
532	0.00	0.00	0.00	34
533	0.67	0.30	0.41	47
534	0.75	0.21	0.33	28
535	0.00	0.00	0.00	20
536	0.70	0.23	0.34	31
537	0.90	0.30	0.45	30
538	0.00	0.00	0.00	22
539	0.75	0.09	0.17	32
540	0.00	0.00	0.00	20
541	0.00	0.00	0.00	30

542	1.00	0.29	0.45	34
543	0.00	0.00	0.00	33
544	0.33	0.09	0.14	33
545	0.50	0.06	0.11	31
546	1.00	0.13	0.23	31
547	0.86	0.65	0.74	37
548	0.75	0.15	0.26	39
549	0.60	0.16	0.25	38
550	0.00	0.00	0.00	25
551	0.44	0.11	0.18	36
552	0.80	0.11	0.20	35
553	0.53	0.38	0.44	26
554	0.59	0.32	0.42	31
555	0.56	0.19	0.28	27
556	0.56	0.14	0.22	37
557	0.43	0.11	0.18	27
558	0.33	0.18	0.24	22
559	0.00	0.00	0.00	30
560	0.79	0.33	0.47	33
561	0.50	0.10	0.16	42
562	0.36	0.14	0.20	37
563	0.14	0.04	0.06	25
564	0.00	0.00	0.00	36
565	0.40	0.09	0.15	22
566	0.50	0.09	0.15	34
567	0.33	0.07	0.11	29
568	0.73	0.31	0.44	35
569	1.00	0.30	0.46	27
570	0.00	0.00	0.00	25
571	0.22	0.07	0.11	27
572	0.85	0.50	0.63	22
573	0.44	0.12	0.19	34
574	1.00	0.24	0.38	34
575	0.66	0.40	0.49	48
576	0.00	0.00	0.00	26
577	0.00	0.00	0.00	37
578	0.56	0.24	0.33	21
579	0.77	0.34	0.48	29
580	0.60	0.19	0.29	31
581	0.00	0.00	0.00	26
582	0.82	0.42	0.56	33
583	0.30	0.15	0.20	20
584	0.59	0.27	0.37	37
585	0.75	0.53	0.62	34
586	0.40	0.15	0.22	26
587	0.80	0.22	0.34	37
588	0.75	0.25	0.38	24
589	0.95	0.55	0.70	38
590	0.18	0.09	0.12	22
591	0.29	0.08	0.13	24
592	0.50	0.15	0.23	33
593	0.25	0.12	0.17	24
594	0.74	0.39	0.51	36
595	0.56	0.24	0.33	21
596	0.00	0.00	0.00	23
597	0.00	0.00	0.00	28
598	0.50	0.17	0.25	30
599	0.57	0.17	0.27	23
600	0.62	0.31	0.41	26
601	0.00	0.00	0.00	33
602	0.00	0.00	0.00	24

603	0.44	0.12	0.20	32
604	0.44	0.18	0.26	22
605	0.00	0.00	0.00	28
606	0.00	0.00	0.00	27
607	0.33	0.05	0.09	20
608	0.00	0.00	0.00	20
609	0.54	0.30	0.39	23
610	0.00	0.00	0.00	31
611	0.17	0.03	0.05	34
612	1.00	0.41	0.58	22
613	0.14	0.07	0.10	14
614	0.87	0.57	0.68	23
615	0.00	0.00	0.00	27
616	0.00	0.00	0.00	29
617	0.69	0.30	0.42	30
618	0.57	0.14	0.22	29
619	0.83	0.15	0.26	33
620	0.47	0.35	0.40	20
621	0.00	0.00	0.00	42
622	0.59	0.43	0.50	23
623	0.89	0.62	0.73	26
624	0.17	0.03	0.05	31
625	0.33	0.08	0.12	26
626	0.33	0.09	0.14	22
627	0.73	0.34	0.47	32
628	0.80	0.16	0.27	25
629	0.73	0.28	0.40	29
630	0.89	0.64	0.74	25
631	0.17	0.05	0.08	19
632	0.77	0.50	0.61	34
633	0.80	0.18	0.30	22
634	0.75	0.11	0.19	27
635	0.94	0.48	0.64	31
636	1.00	0.26	0.41	27
637	0.20	0.06	0.10	16
638	0.92	0.50	0.65	24
639	0.00	0.00	0.00	27
640	0.72	0.45	0.55	29
641	1.00	0.55	0.71	22
642	0.43	0.14	0.21	21
643	0.00	0.00	0.00	27
644	0.46	0.19	0.27	32
645	0.67	0.34	0.45	29
646	1.00	0.04	0.07	26
647	0.21	0.11	0.15	27
648	0.38	0.12	0.18	25
649	0.33	0.25	0.29	20
650	0.33	0.07	0.12	28
651	0.60	0.10	0.17	31
652	0.33	0.04	0.07	27
653	0.71	0.17	0.27	30
654	1.00	0.03	0.06	30
655	0.50	0.05	0.09	21
656	0.50	0.14	0.22	21
657	0.92	0.52	0.67	21
658	0.50	0.12	0.19	26
659	0.14	0.03	0.05	30
660	0.00	0.00	0.00	18
661	0.79	0.46	0.58	24
662	0.43	0.20	0.27	15
663	0.83	0.52	0.64	29

664	0.83	0.42	0.56	24
665	0.00	0.00	0.00	32
666	0.86	0.23	0.36	26
667	0.12	0.04	0.06	25
668	0.14	0.05	0.07	20
669	0.00	0.00	0.00	23
670	0.00	0.00	0.00	28
671	0.12	0.06	0.08	18
672	0.81	0.43	0.57	30
673	0.00	0.00	0.00	28
674	0.00	0.00	0.00	27
675	0.33	0.04	0.07	24
676	0.00	0.00	0.00	25
677	0.50	0.24	0.32	21
678	0.00	0.00	0.00	28
679	0.50	0.30	0.37	27
680	0.00	0.00	0.00	17
681	0.50	0.09	0.15	23
682	0.00	0.00	0.00	20
683	0.71	0.26	0.38	19
684	0.00	0.00	0.00	29
685	0.38	0.09	0.14	35
686	1.00	0.12	0.22	16
687	0.00	0.00	0.00	19
688	0.53	0.31	0.39	26
689	1.00	0.05	0.10	19
690	0.00	0.00	0.00	16
691	0.00	0.00	0.00	20
692	0.00	0.00	0.00	26
693	0.25	0.16	0.19	19
694	0.00	0.00	0.00	24
695	1.00	0.04	0.07	26
696	0.29	0.06	0.10	34
697	1.00	0.05	0.09	21
698	0.00	0.00	0.00	30
699	1.00	0.19	0.32	21
700	0.50	0.08	0.14	24
701	0.50	0.04	0.07	25
702	0.67	0.27	0.39	22
703	0.59	0.50	0.54	20
704	0.20	0.05	0.07	22
705	0.50	0.21	0.29	24
706	1.00	0.05	0.10	19
707	0.00	0.00	0.00	17
708	0.00	0.00	0.00	23
709	0.00	0.00	0.00	21
710	0.33	0.08	0.13	24
711	0.20	0.07	0.11	14
712	0.12	0.05	0.07	19
713	1.00	0.53	0.69	17
714	0.00	0.00	0.00	22
715	0.29	0.14	0.19	14
716	0.80	0.15	0.25	27
717	0.17	0.05	0.08	20
718	0.25	0.05	0.08	21
719	0.54	0.33	0.41	21
720	1.00	0.09	0.17	22
721	0.50	0.04	0.08	24
722	0.47	0.25	0.33	28
723	0.44	0.18	0.26	22
724	0.00	0.00	0.00	20

725	1.00	0.36	0.53	22
726	0.40	0.17	0.24	24
727	0.00	0.00	0.00	15
728	0.75	0.24	0.36	25
729	0.00	0.00	0.00	17
730	0.93	0.67	0.78	21
731	0.56	0.24	0.33	21
732	0.00	0.00	0.00	33
733	0.57	0.24	0.33	17
734	1.00	0.31	0.48	16
735	0.40	0.08	0.13	26
736	0.00	0.00	0.00	23
737	0.40	0.19	0.26	21
738	0.59	0.39	0.47	33
739	0.75	0.19	0.30	16
740	1.00	0.50	0.67	26
741	0.83	0.43	0.57	23
742	1.00	0.05	0.10	20
743	0.60	0.16	0.25	19
744	0.55	0.33	0.41	18
745	0.50	0.04	0.08	23
746	0.33	0.09	0.14	22
747	0.80	0.17	0.29	23
748	0.60	0.12	0.19	26
749	0.00	0.00	0.00	24
750	0.00	0.00	0.00	21
751	1.00	0.26	0.41	23
752	1.00	0.38	0.55	16
753	0.20	0.09	0.13	11
754	0.00	0.00	0.00	23
755	0.00	0.00	0.00	27
756	0.75	0.14	0.24	21
757	0.38	0.18	0.24	28
758	0.25	0.05	0.08	20
759	0.50	0.04	0.08	23
760	0.00	0.00	0.00	21
761	0.33	0.11	0.17	18
762	0.14	0.05	0.07	22
763	0.50	0.03	0.06	31
764	1.00	0.21	0.35	28
765	0.50	0.12	0.20	33
766	0.43	0.13	0.20	23
767	1.00	0.19	0.32	26
768	1.00	0.32	0.48	22
769	0.67	0.19	0.30	21
770	0.22	0.08	0.12	25
771	0.86	0.26	0.40	23
772	0.33	0.10	0.15	21
773	0.00	0.00	0.00	27
774	0.50	0.05	0.10	19
775	0.50	0.17	0.25	24
776	0.00	0.00	0.00	24
777	1.00	0.47	0.64	19
778	0.67	0.13	0.22	15
779	0.50	0.06	0.11	16
780	0.25	0.12	0.17	16
781	0.80	0.22	0.35	18
782	0.00	0.00	0.00	20
783	0.71	0.19	0.30	26
784	0.00	0.00	0.00	19
785	0.00	0.00	0.00	16

786	0.80	0.16	0.27	25
787	0.50	0.05	0.10	19
788	0.62	0.21	0.31	24
789	0.00	0.00	0.00	21
790	0.47	0.29	0.36	24
791	0.00	0.00	0.00	29
792	0.60	0.46	0.52	13
793	0.20	0.06	0.09	18
794	0.00	0.00	0.00	17
795	0.29	0.11	0.16	18
796	0.62	0.36	0.45	14
797	0.50	0.04	0.08	23
798	0.50	0.05	0.09	20
799	0.40	0.12	0.18	17
800	0.00	0.00	0.00	16
801	0.00	0.00	0.00	17
802	0.00	0.00	0.00	17
803	0.00	0.00	0.00	12
804	0.83	0.22	0.34	23
805	0.00	0.00	0.00	23
806	0.00	0.00	0.00	24
807	0.00	0.00	0.00	18
808	0.25	0.07	0.11	14
809	0.89	0.31	0.46	26
810	0.50	0.14	0.22	14
811	0.33	0.06	0.10	17
812	0.57	0.17	0.26	24
813	0.75	0.13	0.22	23
814	0.50	0.05	0.10	19
815	0.33	0.11	0.16	19
816	1.00	0.11	0.20	18
817	0.00	0.00	0.00	22
818	0.00	0.00	0.00	17
819	0.00	0.00	0.00	19
820	0.43	0.14	0.21	21
821	1.00	0.35	0.52	17
822	0.86	0.46	0.60	13
823	0.00	0.00	0.00	18
824	0.58	0.41	0.48	17
825	1.00	0.31	0.48	16
826	0.40	0.12	0.19	16
827	0.00	0.00	0.00	16
828	1.00	0.05	0.09	22
829	0.00	0.00	0.00	22
830	0.60	0.19	0.29	16
831	0.67	0.24	0.35	17
832	0.50	0.12	0.20	16
833	0.00	0.00	0.00	20
834	0.00	0.00	0.00	18
835	0.25	0.08	0.12	12
836	0.00	0.00	0.00	20
837	0.00	0.00	0.00	14
838	1.00	0.35	0.52	17
839	0.00	0.00	0.00	24
840	0.75	0.23	0.35	13
841	0.80	0.20	0.32	20
842	0.50	0.12	0.20	16
843	0.50	0.08	0.13	13
844	0.00	0.00	0.00	25
845	0.25	0.06	0.10	17
846	0.00	0.00	0.00	16

847	0.00	0.00	0.00	22
848	1.00	0.05	0.10	20
849	1.00	0.46	0.63	24
850	0.00	0.00	0.00	17
851	0.17	0.07	0.10	14
852	0.50	0.15	0.23	20
853	0.75	0.43	0.55	14
854	1.00	0.20	0.33	15
855	0.56	0.36	0.43	14
856	0.00	0.00	0.00	12
857	0.87	0.54	0.67	24
858	0.00	0.00	0.00	13
859	0.75	0.19	0.30	16
860	0.67	0.18	0.29	22
861	0.20	0.05	0.08	20
862	0.00	0.00	0.00	15
863	0.33	0.05	0.08	21
864	0.89	0.57	0.70	14
865	0.00	0.00	0.00	17
866	0.00	0.00	0.00	14
867	0.00	0.00	0.00	21
868	1.00	0.47	0.64	17
869	0.00	0.00	0.00	21
870	0.20	0.05	0.08	21
871	0.00	0.00	0.00	15
872	0.00	0.00	0.00	17
873	0.00	0.00	0.00	21
874	0.00	0.00	0.00	15
875	1.00	0.09	0.17	11
876	1.00	0.28	0.43	18
877	0.43	0.11	0.18	27
878	1.00	0.35	0.52	23
879	0.57	0.16	0.25	25
880	0.83	0.29	0.43	17
881	0.50	0.04	0.08	23
882	0.00	0.00	0.00	15
883	0.00	0.00	0.00	21
884	0.83	0.33	0.48	15
885	0.00	0.00	0.00	18
886	1.00	0.58	0.73	19
887	1.00	0.08	0.14	13
888	0.40	0.10	0.16	20
889	0.00	0.00	0.00	26
890	0.00	0.00	0.00	18
891	0.00	0.00	0.00	21
892	0.00	0.00	0.00	20
893	1.00	0.64	0.78	14
894	0.00	0.00	0.00	24
895	0.69	0.41	0.51	22
896	0.00	0.00	0.00	14
897	0.89	0.57	0.70	14
898	0.60	0.27	0.37	22
899	0.50	0.21	0.30	19
900	0.50	0.25	0.33	8
901	0.89	0.40	0.55	20
902	0.00	0.00	0.00	20
903	0.67	0.08	0.14	26
904	0.00	0.00	0.00	18
905	0.00	0.00	0.00	17
906	0.00	0.00	0.00	17
907	0.67	0.31	0.42	13

908	0.00	0.00	0.00	11
909	0.80	0.40	0.53	20
910	1.00	0.36	0.53	11
911	0.86	0.35	0.50	17
912	0.00	0.00	0.00	13
913	0.92	0.85	0.88	13
914	0.43	0.25	0.32	12
915	0.86	0.43	0.57	14
916	0.00	0.00	0.00	19
917	1.00	0.06	0.12	16
918	0.00	0.00	0.00	14
919	0.00	0.00	0.00	13
920	1.00	0.47	0.64	17
921	0.00	0.00	0.00	17
922	0.75	0.21	0.33	14
923	0.33	0.05	0.09	19
924	0.33	0.04	0.07	24
925	0.00	0.00	0.00	16
926	0.00	0.00	0.00	18
927	0.43	0.20	0.27	15
928	0.83	0.67	0.74	15
929	0.88	0.41	0.56	17
930	0.67	0.10	0.17	21
931	0.00	0.00	0.00	13
932	1.00	0.06	0.12	16
933	0.64	0.44	0.52	16
934	0.00	0.00	0.00	20
935	0.00	0.00	0.00	14
936	0.00	0.00	0.00	13
937	0.75	0.20	0.32	15
938	0.00	0.00	0.00	11
939	1.00	0.14	0.25	14
940	0.00	0.00	0.00	18
941	0.38	0.18	0.24	17
942	0.00	0.00	0.00	14
943	0.33	0.19	0.24	16
944	0.00	0.00	0.00	15
945	1.00	0.25	0.40	16
946	0.00	0.00	0.00	20
947	0.11	0.07	0.08	15
948	0.67	0.10	0.17	20
949	0.50	0.21	0.30	14
950	0.00	0.00	0.00	14
951	1.00	0.20	0.33	15
952	1.00	0.18	0.31	11
953	1.00	0.08	0.14	13
954	0.00	0.00	0.00	13
955	0.00	0.00	0.00	15
956	0.00	0.00	0.00	16
957	1.00	0.17	0.29	12
958	0.00	0.00	0.00	17
959	0.33	0.14	0.20	14
960	0.00	0.00	0.00	16
961	0.67	0.33	0.44	6
962	0.33	0.06	0.10	17
963	0.50	0.08	0.13	13
964	0.88	0.58	0.70	12
965	0.00	0.00	0.00	12
966	1.00	0.27	0.42	15
967	0.75	0.27	0.40	11
968	0.50	0.08	0.13	13

969	0.00	0.00	0.00	13
970	0.85	0.52	0.65	21
971	0.00	0.00	0.00	16
972	0.00	0.00	0.00	17
973	1.00	0.08	0.15	12
974	0.80	0.22	0.35	18
975	1.00	0.46	0.63	24
976	0.00	0.00	0.00	24
977	0.00	0.00	0.00	18
978	0.00	0.00	0.00	12
979	0.00	0.00	0.00	16
980	1.00	0.05	0.10	19
981	0.00	0.00	0.00	27
982	1.00	0.14	0.25	14
983	0.57	0.31	0.40	13
984	0.80	0.42	0.55	19
985	0.00	0.00	0.00	16
986	0.20	0.06	0.09	17
987	1.00	0.40	0.57	10
988	0.50	0.10	0.16	21
989	0.00	0.00	0.00	15
990	0.00	0.00	0.00	18
991	0.50	0.06	0.11	16
992	0.50	0.17	0.25	12
993	0.00	0.00	0.00	15
994	0.00	0.00	0.00	15
995	0.00	0.00	0.00	13
996	1.00	0.41	0.58	17
997	0.00	0.00	0.00	17
998	0.00	0.00	0.00	16
999	0.40	0.15	0.22	13
1000	0.67	0.11	0.18	19
1001	1.00	0.25	0.40	8
1002	0.11	0.08	0.09	13
1003	0.00	0.00	0.00	16
1004	0.00	0.00	0.00	15
1005	0.00	0.00	0.00	9
1006	0.00	0.00	0.00	23
1007	0.00	0.00	0.00	11
1008	0.88	0.37	0.52	19
1009	0.00	0.00	0.00	19
1010	0.00	0.00	0.00	15
1011	0.50	0.20	0.29	10
1012	0.00	0.00	0.00	12
1013	0.00	0.00	0.00	15
1014	0.67	0.14	0.24	14
1015	1.00	0.38	0.55	16
1016	0.00	0.00	0.00	12
1017	0.50	0.19	0.27	16
1018	0.00	0.00	0.00	17
1019	1.00	0.25	0.40	12
1020	0.67	0.20	0.31	10
1021	0.75	0.23	0.35	13
1022	0.33	0.11	0.16	19
1023	0.00	0.00	0.00	11
1024	0.25	0.09	0.13	11
1025	0.00	0.00	0.00	16
1026	0.50	0.21	0.30	19
1027	0.78	0.33	0.47	21
1028	0.80	0.33	0.47	12
1029	0.00	0.00	0.00	13

1030	0.00	0.00	0.00	11
1031	1.00	0.06	0.11	17
1032	0.50	0.10	0.17	10
1033	0.67	0.14	0.24	14
1034	0.00	0.00	0.00	19
1035	0.25	0.14	0.18	7
1036	0.80	0.31	0.44	13
1037	0.33	0.06	0.10	18
1038	1.00	0.13	0.24	15
1039	0.80	0.22	0.35	18
1040	0.00	0.00	0.00	11
1041	0.00	0.00	0.00	14
1042	0.10	0.05	0.07	19
1043	0.00	0.00	0.00	11
1044	0.00	0.00	0.00	15
1045	0.00	0.00	0.00	19
1046	0.00	0.00	0.00	14
1047	0.00	0.00	0.00	16
1048	0.00	0.00	0.00	7
1049	0.00	0.00	0.00	18
1050	0.00	0.00	0.00	22
1051	0.33	0.07	0.11	15
1052	0.00	0.00	0.00	10
1053	0.83	0.29	0.43	17
1054	0.00	0.00	0.00	22
1055	0.00	0.00	0.00	18
1056	0.67	0.36	0.47	11
1057	0.62	0.29	0.40	17
1058	0.00	0.00	0.00	17
1059	0.00	0.00	0.00	19
1060	0.00	0.00	0.00	15
1061	1.00	0.50	0.67	10
1062	1.00	0.17	0.29	18
1063	0.40	0.15	0.22	13
1064	0.00	0.00	0.00	8
1065	0.50	0.06	0.10	18
1066	0.00	0.00	0.00	10
1067	0.75	0.35	0.48	17
1068	0.00	0.00	0.00	16
1069	0.71	0.31	0.43	16
1070	0.75	0.16	0.26	19
1071	0.67	0.15	0.25	13
1072	1.00	0.33	0.50	9
1073	1.00	0.07	0.12	15
1074	1.00	0.16	0.27	19
1075	0.62	0.28	0.38	18
1076	0.50	0.08	0.14	12
1077	0.90	0.47	0.62	19
1078	1.00	0.36	0.53	11
1079	0.00	0.00	0.00	16
1080	0.00	0.00	0.00	12
1081	0.00	0.00	0.00	16
1082	0.80	0.24	0.36	17
1083	0.00	0.00	0.00	16
1084	0.33	0.06	0.10	17
1085	0.00	0.00	0.00	11
1086	0.00	0.00	0.00	11
1087	0.50	0.07	0.12	15
1088	1.00	0.50	0.67	8
1089	0.00	0.00	0.00	14
1090	0.75	0.25	0.38	12

1091	1.00	0.22	0.36	9
1092	0.00	0.00	0.00	16
1093	0.20	0.08	0.11	13
1094	0.00	0.00	0.00	14
1095	1.00	0.43	0.60	14
1096	0.00	0.00	0.00	16
1097	0.00	0.00	0.00	6
1098	0.50	0.09	0.15	22
1099	0.00	0.00	0.00	16
1100	1.00	0.11	0.20	9
1101	0.00	0.00	0.00	17
1102	0.67	0.29	0.40	14
1103	1.00	0.05	0.09	22
1104	0.00	0.00	0.00	9
1105	0.00	0.00	0.00	14
1106	0.00	0.00	0.00	12
1107	0.80	0.22	0.35	18
1108	0.00	0.00	0.00	14
1109	0.20	0.07	0.10	15
1110	0.00	0.00	0.00	13
1111	0.29	0.18	0.22	11
1112	0.00	0.00	0.00	10
1113	0.00	0.00	0.00	4
1114	0.00	0.00	0.00	14
1115	0.00	0.00	0.00	14
1116	0.00	0.00	0.00	15
1117	0.00	0.00	0.00	18
1118	0.00	0.00	0.00	20
1119	1.00	0.06	0.12	16
1120	0.00	0.00	0.00	10
1121	0.00	0.00	0.00	17
1122	0.50	0.08	0.14	12
1123	0.67	0.20	0.31	10
1124	0.00	0.00	0.00	11
1125	0.50	0.17	0.25	6
1126	0.00	0.00	0.00	9
1127	0.00	0.00	0.00	12
1128	0.50	0.21	0.30	14
1129	0.60	0.38	0.46	8
1130	0.60	0.23	0.33	13
1131	0.00	0.00	0.00	12
1132	0.12	0.09	0.11	11
1133	0.00	0.00	0.00	15
1134	0.00	0.00	0.00	10
1135	0.33	0.06	0.11	16
1136	1.00	0.06	0.12	16
1137	0.00	0.00	0.00	17
1138	0.00	0.00	0.00	15
1139	1.00	0.85	0.92	13
1140	0.00	0.00	0.00	10
1141	0.14	0.12	0.13	8
1142	0.00	0.00	0.00	12
1143	0.00	0.00	0.00	8
1144	0.50	0.06	0.11	17
1145	0.00	0.00	0.00	13
1146	0.00	0.00	0.00	14
1147	1.00	0.31	0.47	13
1148	0.00	0.00	0.00	11
1149	1.00	0.12	0.22	16
1150	0.00	0.00	0.00	15
1151	0.20	0.09	0.13	11

1152	0.67	0.17	0.27	12
1153	1.00	0.25	0.40	12
1154	0.57	0.31	0.40	13
1155	0.00	0.00	0.00	13
1156	0.00	0.00	0.00	7
1157	0.00	0.00	0.00	10
1158	0.00	0.00	0.00	16
1159	0.00	0.00	0.00	16
1160	0.67	0.18	0.29	11
1161	1.00	0.44	0.62	18
1162	0.75	0.19	0.30	16
1163	0.50	0.07	0.12	14
1164	0.00	0.00	0.00	18
1165	1.00	0.62	0.76	13
1166	0.50	0.10	0.17	10
1167	0.33	0.33	0.33	6
1168	0.83	0.45	0.59	11
1169	0.67	0.22	0.33	18
1170	1.00	0.27	0.43	11
1171	0.00	0.00	0.00	14
1172	0.00	0.00	0.00	13
1173	0.67	0.50	0.57	8
1174	1.00	0.13	0.24	15
1175	0.88	0.70	0.78	10
1176	0.00	0.00	0.00	5
1177	0.75	0.40	0.52	15
1178	0.00	0.00	0.00	5
1179	1.00	0.33	0.50	12
1180	0.00	0.00	0.00	12
1181	1.00	0.15	0.27	13
1182	0.00	0.00	0.00	16
1183	0.67	0.18	0.29	11
1184	0.00	0.00	0.00	13
1185	0.00	0.00	0.00	12
1186	0.83	0.50	0.62	10
1187	0.00	0.00	0.00	14
1188	0.86	0.55	0.67	22
1189	0.00	0.00	0.00	8
1190	0.00	0.00	0.00	8
1191	1.00	0.33	0.50	9
1192	0.50	0.17	0.25	12
1193	0.00	0.00	0.00	11
1194	0.33	0.07	0.12	14
1195	0.83	0.26	0.40	19
1196	1.00	0.11	0.20	9
1197	1.00	0.04	0.08	23
1198	0.00	0.00	0.00	7
1199	1.00	0.12	0.22	8
1200	0.00	0.00	0.00	11
1201	0.67	0.13	0.22	15
1202	0.00	0.00	0.00	17
1203	0.00	0.00	0.00	10
1204	0.50	0.14	0.22	7
1205	0.67	0.36	0.47	11
1206	0.50	0.14	0.22	14
1207	1.00	0.42	0.59	12
1208	0.33	0.07	0.12	14
1209	0.00	0.00	0.00	13
1210	0.00	0.00	0.00	7
1211	0.75	0.30	0.43	10
1212	1.00	0.11	0.20	9

1213	1.00	0.56	0.71	9
1214	0.33	0.07	0.12	14
1215	0.67	0.12	0.21	16
1216	0.33	0.14	0.20	14
1217	0.25	0.08	0.12	12
1218	0.00	0.00	0.00	14
1219	0.50	0.06	0.11	16
1220	0.00	0.00	0.00	11
1221	1.00	0.27	0.43	11
1222	0.00	0.00	0.00	9
1223	0.00	0.00	0.00	12
1224	0.60	0.30	0.40	10
1225	1.00	0.17	0.29	12
1226	0.67	0.25	0.36	8
1227	0.00	0.00	0.00	12
1228	0.50	0.08	0.13	13
1229	0.00	0.00	0.00	10
1230	0.25	0.14	0.18	7
1231	0.00	0.00	0.00	11
1232	0.83	0.56	0.67	9
1233	0.33	0.07	0.11	15
1234	0.25	0.11	0.15	9
1235	0.50	0.17	0.25	6
1236	0.00	0.00	0.00	15
1237	0.00	0.00	0.00	10
1238	1.00	0.40	0.57	10
1239	0.83	0.45	0.59	11
1240	0.00	0.00	0.00	13
1241	0.00	0.00	0.00	9
1242	0.75	0.38	0.50	8
1243	1.00	0.25	0.40	12
1244	0.00	0.00	0.00	10
1245	0.00	0.00	0.00	6
1246	0.00	0.00	0.00	4
1247	0.00	0.00	0.00	13
1248	1.00	0.05	0.10	19
1249	0.00	0.00	0.00	13
1250	0.00	0.00	0.00	7
1251	0.00	0.00	0.00	5
1252	0.00	0.00	0.00	5
1253	1.00	0.13	0.24	15
1254	0.00	0.00	0.00	12
1255	1.00	0.12	0.22	8
1256	0.86	0.60	0.71	10
1257	0.00	0.00	0.00	17
1258	0.00	0.00	0.00	8
1259	0.00	0.00	0.00	6
1260	1.00	0.73	0.84	11
1261	0.00	0.00	0.00	14
1262	0.00	0.00	0.00	22
1263	0.83	0.50	0.62	10
1264	0.33	0.12	0.18	8
1265	0.00	0.00	0.00	14
1266	0.00	0.00	0.00	10
1267	0.33	0.08	0.13	12
1268	0.67	0.29	0.40	7
1269	1.00	0.17	0.29	12
1270	0.00	0.00	0.00	12
1271	0.00	0.00	0.00	13
1272	0.67	0.12	0.20	17
1273	0.75	0.20	0.32	15

1274	0.00	0.00	0.00	17
1275	0.00	0.00	0.00	9
1276	0.00	0.00	0.00	9
1277	0.00	0.00	0.00	14
1278	0.00	0.00	0.00	14
1279	0.00	0.00	0.00	6
1280	0.00	0.00	0.00	9
1281	0.50	0.29	0.36	7
1282	0.00	0.00	0.00	10
1283	0.50	0.15	0.24	13
1284	0.00	0.00	0.00	9
1285	1.00	0.09	0.17	11
1286	0.00	0.00	0.00	7
1287	0.33	0.10	0.15	10
1288	0.33	0.14	0.20	7
1289	0.50	0.18	0.27	11
1290	0.00	0.00	0.00	18
1291	0.00	0.00	0.00	11
1292	0.00	0.00	0.00	7
1293	0.00	0.00	0.00	12
1294	0.00	0.00	0.00	11
1295	0.00	0.00	0.00	13
1296	1.00	0.08	0.15	12
1297	0.00	0.00	0.00	16
1298	1.00	0.08	0.15	12
1299	0.00	0.00	0.00	8
1300	0.67	0.12	0.21	16
1301	1.00	0.44	0.61	16
1302	0.50	0.11	0.18	9
1303	1.00	0.45	0.62	11
1304	0.00	0.00	0.00	10
1305	1.00	0.25	0.40	8
1306	1.00	0.08	0.14	13
1307	0.50	0.20	0.29	10
1308	0.00	0.00	0.00	16
1309	0.50	0.06	0.11	16
1310	0.00	0.00	0.00	15
1311	1.00	0.25	0.40	16
1312	0.00	0.00	0.00	15
1313	1.00	0.30	0.46	10
1314	0.00	0.00	0.00	9
1315	0.00	0.00	0.00	12
1316	0.00	0.00	0.00	12
1317	0.00	0.00	0.00	14
1318	1.00	0.11	0.20	9
1319	0.00	0.00	0.00	15
1320	1.00	0.45	0.62	11
1321	0.70	0.54	0.61	13
1322	0.00	0.00	0.00	12
1323	0.00	0.00	0.00	11
1324	0.00	0.00	0.00	17
1325	1.00	0.19	0.32	16
1326	0.00	0.00	0.00	14
1327	1.00	0.67	0.80	9
1328	0.00	0.00	0.00	8
1329	0.00	0.00	0.00	10
1330	0.00	0.00	0.00	8
1331	1.00	0.17	0.29	18
1332	0.50	0.20	0.29	10
1333	0.00	0.00	0.00	10
1334	0.00	0.00	0.00	10

1335	0.25	0.08	0.12	12
1336	0.00	0.00	0.00	8
1337	0.50	0.11	0.18	9
1338	0.00	0.00	0.00	15
1339	1.00	0.25	0.40	12
1340	0.00	0.00	0.00	11
1341	0.00	0.00	0.00	10
1342	0.00	0.00	0.00	17
1343	0.50	0.50	0.50	6
1344	0.57	0.33	0.42	12
1345	0.50	0.12	0.20	8
1346	1.00	0.08	0.14	13
1347	0.00	0.00	0.00	10
1348	0.00	0.00	0.00	10
1349	0.75	0.40	0.52	15
1350	1.00	0.08	0.14	13
1351	0.00	0.00	0.00	7
1352	1.00	0.47	0.64	15
1353	0.00	0.00	0.00	8
1354	1.00	0.10	0.18	10
1355	0.00	0.00	0.00	9
1356	1.00	0.10	0.18	10
1357	0.50	0.31	0.38	13
1358	0.00	0.00	0.00	9
1359	0.67	0.25	0.36	8
1360	0.00	0.00	0.00	9
1361	0.00	0.00	0.00	8
1362	1.00	0.08	0.14	13
1363	1.00	0.10	0.18	10
1364	0.50	0.08	0.13	13
1365	0.00	0.00	0.00	9
1366	1.00	0.09	0.17	11
1367	1.00	0.36	0.53	11
1368	0.00	0.00	0.00	6
1369	0.00	0.00	0.00	6
1370	1.00	0.09	0.17	11
1371	0.00	0.00	0.00	14
1372	0.00	0.00	0.00	9
1373	1.00	0.08	0.15	12
1374	0.00	0.00	0.00	6
1375	0.00	0.00	0.00	12
1376	0.67	0.18	0.29	11
1377	0.00	0.00	0.00	13
1378	0.00	0.00	0.00	10
1379	0.50	0.07	0.12	14
1380	0.00	0.00	0.00	15
1381	0.00	0.00	0.00	12
1382	0.00	0.00	0.00	11
1383	1.00	0.57	0.73	14
1384	0.00	0.00	0.00	3
1385	0.00	0.00	0.00	15
1386	0.00	0.00	0.00	15
1387	0.00	0.00	0.00	5
1388	0.00	0.00	0.00	15
1389	0.00	0.00	0.00	13
1390	0.50	0.08	0.14	12
1391	0.00	0.00	0.00	11
1392	0.00	0.00	0.00	14
1393	1.00	0.11	0.20	9
1394	1.00	0.09	0.17	11
1395	1.00	0.21	0.35	19

1396	0.00	0.00	0.00	11
1397	0.00	0.00	0.00	6
1398	0.00	0.00	0.00	10
1399	0.25	0.10	0.14	10
1400	0.00	0.00	0.00	15
1401	0.00	0.00	0.00	7
1402	0.00	0.00	0.00	12
1403	1.00	0.33	0.50	6
1404	0.00	0.00	0.00	4
1405	0.00	0.00	0.00	6
1406	0.75	0.43	0.55	7
1407	0.00	0.00	0.00	9
1408	0.00	0.00	0.00	10
1409	0.25	0.20	0.22	5
1410	0.00	0.00	0.00	11
1411	0.00	0.00	0.00	7
1412	1.00	0.36	0.53	11
1413	0.67	0.25	0.36	8
1414	0.67	0.12	0.20	17
1415	1.00	0.38	0.56	13
1416	1.00	0.12	0.22	16
1417	0.25	0.09	0.13	11
1418	0.00	0.00	0.00	11
1419	0.00	0.00	0.00	15
1420	0.00	0.00	0.00	9
1421	0.00	0.00	0.00	10
1422	0.50	0.14	0.22	7
1423	0.00	0.00	0.00	17
1424	0.00	0.00	0.00	13
1425	0.00	0.00	0.00	14
1426	0.00	0.00	0.00	16
1427	0.00	0.00	0.00	8
1428	1.00	0.06	0.12	16
1429	0.00	0.00	0.00	14
1430	0.00	0.00	0.00	10
1431	0.00	0.00	0.00	13
1432	0.00	0.00	0.00	8
1433	0.88	0.47	0.61	15
1434	0.50	0.12	0.19	17
1435	0.00	0.00	0.00	5
1436	0.00	0.00	0.00	7
1437	0.00	0.00	0.00	9
1438	0.00	0.00	0.00	3
1439	1.00	0.12	0.22	8
1440	0.50	0.17	0.25	12
1441	0.00	0.00	0.00	13
1442	1.00	0.29	0.44	7
1443	0.50	0.10	0.17	10
1444	0.00	0.00	0.00	9
1445	1.00	0.18	0.30	17
1446	0.50	0.10	0.16	21
1447	0.00	0.00	0.00	6
1448	0.00	0.00	0.00	8
1449	0.00	0.00	0.00	8
1450	1.00	0.10	0.18	10
1451	1.00	0.13	0.24	15
1452	0.00	0.00	0.00	15
1453	0.00	0.00	0.00	7
1454	0.00	0.00	0.00	14
1455	0.00	0.00	0.00	17
1456	1.00	0.33	0.50	9

1457	0.33	0.08	0.13	12
1458	0.80	0.36	0.50	11
1459	0.00	0.00	0.00	7
1460	1.00	0.25	0.40	4
1461	0.00	0.00	0.00	11
1462	0.00	0.00	0.00	6
1463	0.50	0.12	0.20	8
1464	0.00	0.00	0.00	8
1465	0.00	0.00	0.00	12
1466	0.00	0.00	0.00	9
1467	0.50	0.11	0.18	9
1468	0.00	0.00	0.00	13
1469	0.00	0.00	0.00	8
1470	0.00	0.00	0.00	9
1471	1.00	0.27	0.42	15
1472	0.50	0.20	0.29	10
1473	1.00	0.38	0.55	8
1474	0.00	0.00	0.00	12
1475	0.00	0.00	0.00	13
1476	0.00	0.00	0.00	11
1477	0.00	0.00	0.00	9
1478	0.00	0.00	0.00	13
1479	1.00	0.44	0.62	9
1480	1.00	0.08	0.14	13
1481	0.00	0.00	0.00	9
1482	0.00	0.00	0.00	7
1483	1.00	0.75	0.86	4
1484	0.86	0.40	0.55	15
1485	0.00	0.00	0.00	16
1486	0.00	0.00	0.00	10
1487	1.00	0.18	0.31	11
1488	0.00	0.00	0.00	16
1489	0.00	0.00	0.00	11
1490	0.50	0.10	0.17	10
1491	0.00	0.00	0.00	7
1492	0.00	0.00	0.00	9
1493	1.00	0.29	0.44	7
1494	0.50	0.33	0.40	6
1495	0.33	0.25	0.29	8
1496	0.00	0.00	0.00	14
1497	1.00	0.20	0.33	10
1498	1.00	0.36	0.53	14
1499	1.00	0.17	0.29	6
1500	0.00	0.00	0.00	9
1501	0.00	0.00	0.00	13
1502	0.00	0.00	0.00	5
1503	0.00	0.00	0.00	8
1504	0.00	0.00	0.00	7
1505	0.00	0.00	0.00	9
1506	0.00	0.00	0.00	8
1507	0.00	0.00	0.00	6
1508	0.50	0.10	0.17	10
1509	0.00	0.00	0.00	7
1510	0.64	0.58	0.61	12
1511	0.50	0.10	0.17	10
1512	0.00	0.00	0.00	8
1513	1.00	0.38	0.55	8
1514	0.00	0.00	0.00	12
1515	0.00	0.00	0.00	9
1516	1.00	0.29	0.44	7
1517	0.00	0.00	0.00	10

1518	0.00	0.00	0.00	10
1519	0.00	0.00	0.00	9
1520	1.00	0.11	0.20	9
1521	0.00	0.00	0.00	8
1522	0.50	0.08	0.13	13
1523	0.00	0.00	0.00	14
1524	0.00	0.00	0.00	8
1525	0.00	0.00	0.00	12
1526	1.00	0.40	0.57	5
1527	0.00	0.00	0.00	12
1528	0.00	0.00	0.00	8
1529	0.67	0.33	0.44	6
1530	0.75	0.33	0.46	9
1531	1.00	0.30	0.46	10
1532	1.00	0.12	0.22	8
1533	0.80	0.25	0.38	16
1534	0.00	0.00	0.00	9
1535	0.00	0.00	0.00	10
1536	0.00	0.00	0.00	13
1537	1.00	0.36	0.53	11
1538	1.00	0.08	0.15	12
1539	0.00	0.00	0.00	8
1540	0.00	0.00	0.00	9
1541	0.00	0.00	0.00	7
1542	0.80	0.50	0.62	8
1543	0.00	0.00	0.00	7
1544	0.50	0.19	0.27	16
1545	0.00	0.00	0.00	7
1546	1.00	0.08	0.15	12
1547	0.00	0.00	0.00	6
1548	0.00	0.00	0.00	11
1549	0.00	0.00	0.00	8
1550	0.50	0.12	0.20	8
1551	0.00	0.00	0.00	8
1552	0.60	0.38	0.46	8
1553	1.00	0.40	0.57	10
1554	1.00	0.17	0.29	12
1555	0.00	0.00	0.00	10
1556	0.00	0.00	0.00	11
1557	0.00	0.00	0.00	9
1558	0.33	0.14	0.20	7
1559	0.88	0.58	0.70	12
1560	0.00	0.00	0.00	11
1561	0.00	0.00	0.00	10
1562	0.00	0.00	0.00	8
1563	0.00	0.00	0.00	12
1564	0.00	0.00	0.00	9
1565	0.00	0.00	0.00	10
1566	1.00	0.22	0.36	9
1567	1.00	0.29	0.44	14
1568	0.00	0.00	0.00	9
1569	0.00	0.00	0.00	10
1570	1.00	0.50	0.67	10
1571	0.00	0.00	0.00	14
1572	0.00	0.00	0.00	6
1573	0.50	0.14	0.22	7
1574	0.00	0.00	0.00	9
1575	0.00	0.00	0.00	9
1576	0.00	0.00	0.00	11
1577	0.00	0.00	0.00	7
1578	0.00	0.00	0.00	8

1579	0.00	0.00	0.00	11
1580	1.00	0.17	0.29	6
1581	0.00	0.00	0.00	12
1582	0.50	0.17	0.25	6
1583	0.00	0.00	0.00	4
1584	0.00	0.00	0.00	15
1585	1.00	0.25	0.40	8
1586	0.00	0.00	0.00	12
1587	0.00	0.00	0.00	14
1588	0.00	0.00	0.00	7
1589	0.50	0.14	0.22	7
1590	0.00	0.00	0.00	11
1591	0.00	0.00	0.00	8
1592	0.50	0.14	0.22	7
1593	0.00	0.00	0.00	7
1594	0.00	0.00	0.00	4
1595	0.00	0.00	0.00	9
1596	0.00	0.00	0.00	8
1597	1.00	0.11	0.19	19
1598	0.00	0.00	0.00	12
1599	1.00	0.43	0.60	7
1600	0.50	0.11	0.18	9
1601	0.00	0.00	0.00	11
1602	0.00	0.00	0.00	8
1603	0.00	0.00	0.00	7
1604	1.00	0.27	0.43	11
1605	0.00	0.00	0.00	5
1606	1.00	0.20	0.33	5
1607	0.00	0.00	0.00	7
1608	0.00	0.00	0.00	7
1609	0.00	0.00	0.00	9
1610	0.00	0.00	0.00	8
1611	0.00	0.00	0.00	13
1612	0.00	0.00	0.00	9
1613	0.00	0.00	0.00	5
1614	0.00	0.00	0.00	11
1615	0.00	0.00	0.00	8
1616	0.00	0.00	0.00	6
1617	1.00	0.12	0.21	17
1618	0.00	0.00	0.00	8
1619	0.67	0.13	0.22	15
1620	1.00	0.38	0.55	8
1621	0.00	0.00	0.00	8
1622	0.00	0.00	0.00	8
1623	0.00	0.00	0.00	4
1624	0.00	0.00	0.00	5
1625	0.00	0.00	0.00	11
1626	0.00	0.00	0.00	2
1627	0.00	0.00	0.00	14
1628	0.00	0.00	0.00	8
1629	0.67	0.15	0.25	13
1630	1.00	0.55	0.71	11
1631	1.00	0.12	0.22	8
1632	0.00	0.00	0.00	13
1633	0.00	0.00	0.00	9
1634	0.75	0.60	0.67	5
1635	0.00	0.00	0.00	11
1636	0.50	0.33	0.40	3
1637	0.00	0.00	0.00	6
1638	1.00	0.08	0.15	12
1639	0.00	0.00	0.00	12

1640	0.00	0.00	0.00	6
1641	0.00	0.00	0.00	10
1642	0.00	0.00	0.00	9
1643	1.00	0.10	0.18	10
1644	1.00	0.54	0.70	13
1645	0.86	0.75	0.80	8
1646	0.00	0.00	0.00	10
1647	0.00	0.00	0.00	4
1648	0.00	0.00	0.00	8
1649	0.00	0.00	0.00	16
1650	0.00	0.00	0.00	11
1651	1.00	0.42	0.59	12
1652	1.00	0.14	0.25	7
1653	0.00	0.00	0.00	4
1654	0.50	0.11	0.18	9
1655	0.50	0.12	0.20	8
1656	0.50	0.14	0.22	7
1657	0.00	0.00	0.00	11
1658	0.00	0.00	0.00	8
1659	0.50	0.11	0.18	9
1660	0.00	0.00	0.00	9
1661	1.00	0.50	0.67	4
1662	0.00	0.00	0.00	8
1663	0.00	0.00	0.00	15
1664	0.00	0.00	0.00	14
1665	0.00	0.00	0.00	9
1666	0.00	0.00	0.00	6
1667	0.75	0.33	0.46	9
1668	0.00	0.00	0.00	5
1669	1.00	0.12	0.22	8
1670	0.80	0.40	0.53	10
1671	0.67	0.12	0.21	16
1672	0.00	0.00	0.00	4
1673	0.00	0.00	0.00	9
1674	1.00	0.56	0.71	9
1675	0.00	0.00	0.00	8
1676	0.50	0.25	0.33	4
1677	0.00	0.00	0.00	10
1678	1.00	0.08	0.14	13
1679	0.00	0.00	0.00	14
1680	1.00	0.33	0.50	6
1681	1.00	0.09	0.17	11
1682	0.00	0.00	0.00	10
1683	0.20	0.20	0.20	5
1684	0.86	0.60	0.71	10
1685	0.00	0.00	0.00	10
1686	1.00	0.10	0.18	10
1687	0.00	0.00	0.00	10
1688	0.00	0.00	0.00	9
1689	1.00	0.09	0.17	11
1690	0.00	0.00	0.00	8
1691	0.67	0.17	0.27	12
1692	0.00	0.00	0.00	7
1693	0.00	0.00	0.00	5
1694	0.00	0.00	0.00	7
1695	1.00	0.07	0.12	15
1696	0.00	0.00	0.00	10
1697	0.00	0.00	0.00	8
1698	1.00	0.11	0.20	9
1699	0.00	0.00	0.00	9
1700	0.00	0.00	0.00	9

1701	0.00	0.00	0.00	10
1702	0.00	0.00	0.00	12
1703	0.00	0.00	0.00	6
1704	0.00	0.00	0.00	6
1705	1.00	0.20	0.33	15
1706	1.00	0.17	0.29	6
1707	0.00	0.00	0.00	8
1708	0.00	0.00	0.00	7
1709	0.00	0.00	0.00	10
1710	0.00	0.00	0.00	11
1711	0.80	0.80	0.80	5
1712	0.00	0.00	0.00	8
1713	1.00	0.25	0.40	8
1714	0.00	0.00	0.00	9
1715	0.00	0.00	0.00	7
1716	0.00	0.00	0.00	6
1717	0.00	0.00	0.00	11
1718	1.00	0.54	0.70	13
1719	0.00	0.00	0.00	10
1720	1.00	0.14	0.25	7
1721	0.00	0.00	0.00	5
1722	0.00	0.00	0.00	4
1723	0.33	0.25	0.29	4
1724	1.00	0.12	0.22	8
1725	0.00	0.00	0.00	10
1726	0.00	0.00	0.00	7
1727	0.00	0.00	0.00	7
1728	0.00	0.00	0.00	7
1729	0.00	0.00	0.00	12
1730	0.00	0.00	0.00	6
1731	1.00	0.08	0.14	13
1732	0.00	0.00	0.00	6
1733	0.00	0.00	0.00	8
1734	0.33	0.20	0.25	5
1735	0.00	0.00	0.00	10
1736	0.00	0.00	0.00	8
1737	0.00	0.00	0.00	9
1738	0.00	0.00	0.00	6
1739	0.00	0.00	0.00	13
1740	1.00	0.31	0.47	13
1741	0.00	0.00	0.00	6
1742	0.00	0.00	0.00	1
1743	0.00	0.00	0.00	10
1744	0.00	0.00	0.00	5
1745	0.00	0.00	0.00	12
1746	1.00	0.22	0.36	9
1747	0.75	0.23	0.35	13
1748	0.00	0.00	0.00	9
1749	0.00	0.00	0.00	4
1750	0.00	0.00	0.00	11
1751	0.00	0.00	0.00	8
1752	0.00	0.00	0.00	9
1753	0.40	0.20	0.27	10
1754	0.00	0.00	0.00	5
1755	0.00	0.00	0.00	8
1756	0.75	0.50	0.60	6
1757	0.00	0.00	0.00	9
1758	0.83	0.50	0.62	10
1759	0.00	0.00	0.00	11
1760	0.25	0.09	0.13	11
1761	0.00	0.00	0.00	7

1762	0.00	0.00	0.00	11
1763	0.00	0.00	0.00	10
1764	0.00	0.00	0.00	11
1765	0.00	0.00	0.00	5
1766	0.00	0.00	0.00	7
1767	0.40	0.25	0.31	8
1768	0.00	0.00	0.00	11
1769	0.00	0.00	0.00	15
1770	0.50	0.14	0.22	7
1771	0.00	0.00	0.00	6
1772	0.00	0.00	0.00	4
1773	1.00	0.07	0.12	15
1774	0.00	0.00	0.00	8
1775	0.00	0.00	0.00	5
1776	0.00	0.00	0.00	8
1777	1.00	0.29	0.44	7
1778	1.00	0.08	0.15	12
1779	0.00	0.00	0.00	10
1780	0.50	0.10	0.17	10
1781	1.00	0.08	0.14	13
1782	0.00	0.00	0.00	5
1783	0.60	0.27	0.37	11
1784	0.00	0.00	0.00	9
1785	1.00	0.20	0.33	5
1786	0.00	0.00	0.00	5
1787	0.00	0.00	0.00	8
1788	1.00	0.78	0.88	9
1789	0.00	0.00	0.00	12
1790	0.00	0.00	0.00	9
1791	0.00	0.00	0.00	9
1792	0.80	0.44	0.57	9
1793	1.00	0.60	0.75	5
1794	0.50	0.29	0.36	7
1795	1.00	0.18	0.31	11
1796	0.75	0.18	0.29	17
1797	1.00	0.44	0.62	9
1798	0.50	0.07	0.12	15
1799	0.50	0.40	0.44	5
1800	0.00	0.00	0.00	6
1801	0.00	0.00	0.00	15
1802	0.00	0.00	0.00	9
1803	0.00	0.00	0.00	7
1804	0.00	0.00	0.00	7
1805	0.00	0.00	0.00	10
1806	0.00	0.00	0.00	6
1807	0.00	0.00	0.00	7
1808	0.00	0.00	0.00	14
1809	0.00	0.00	0.00	8
1810	0.00	0.00	0.00	8
1811	0.67	0.29	0.40	7
1812	0.00	0.00	0.00	9
1813	0.00	0.00	0.00	5
1814	1.00	0.10	0.18	10
1815	0.00	0.00	0.00	9
1816	1.00	0.20	0.33	5
1817	0.33	0.11	0.17	9
1818	0.33	0.20	0.25	5
1819	0.00	0.00	0.00	11
1820	0.00	0.00	0.00	8
1821	1.00	0.18	0.31	11
1822	0.00	0.00	0.00	8

1823	0.00	0.00	0.00	10
1824	0.00	0.00	0.00	6
1825	0.50	0.22	0.31	9
1826	0.00	0.00	0.00	11
1827	0.00	0.00	0.00	10
1828	0.00	0.00	0.00	2
1829	0.00	0.00	0.00	9
1830	0.00	0.00	0.00	5
1831	1.00	0.38	0.55	8
1832	0.00	0.00	0.00	2
1833	1.00	0.12	0.22	8
1834	0.00	0.00	0.00	5
1835	0.00	0.00	0.00	4
1836	1.00	0.11	0.20	9
1837	0.00	0.00	0.00	9
1838	0.00	0.00	0.00	7
1839	0.00	0.00	0.00	6
1840	0.00	0.00	0.00	8
1841	1.00	0.33	0.50	6
1842	1.00	0.11	0.20	9
1843	0.00	0.00	0.00	6
1844	0.00	0.00	0.00	7
1845	1.00	0.10	0.18	10
1846	0.00	0.00	0.00	11
1847	0.00	0.00	0.00	15
1848	0.75	0.38	0.50	8
1849	0.00	0.00	0.00	11
1850	0.00	0.00	0.00	8
1851	0.75	0.33	0.46	9
1852	0.00	0.00	0.00	4
1853	0.00	0.00	0.00	10
1854	0.00	0.00	0.00	7
1855	1.00	0.30	0.46	10
1856	0.00	0.00	0.00	3
1857	0.00	0.00	0.00	13
1858	0.00	0.00	0.00	6
1859	0.00	0.00	0.00	12
1860	0.00	0.00	0.00	11
1861	0.00	0.00	0.00	7
1862	0.00	0.00	0.00	11
1863	0.00	0.00	0.00	8
1864	0.00	0.00	0.00	7
1865	0.00	0.00	0.00	11
1866	0.00	0.00	0.00	7
1867	0.00	0.00	0.00	10
1868	0.00	0.00	0.00	7
1869	0.00	0.00	0.00	9
1870	0.00	0.00	0.00	6
1871	0.00	0.00	0.00	10
1872	1.00	0.44	0.62	9
1873	0.00	0.00	0.00	8
1874	0.00	0.00	0.00	9
1875	0.00	0.00	0.00	13
1876	0.00	0.00	0.00	10
1877	0.00	0.00	0.00	5
1878	0.00	0.00	0.00	6
1879	1.00	0.08	0.14	13
1880	0.00	0.00	0.00	8
1881	0.75	0.60	0.67	5
1882	0.00	0.00	0.00	12
1883	0.00	0.00	0.00	5

1884	1.00	0.10	0.18	10
1885	0.00	0.00	0.00	10
1886	1.00	0.40	0.57	10
1887	0.00	0.00	0.00	10
1888	0.00	0.00	0.00	13
1889	0.00	0.00	0.00	7
1890	1.00	0.36	0.53	11
1891	0.00	0.00	0.00	6
1892	0.00	0.00	0.00	10
1893	0.00	0.00	0.00	5
1894	1.00	0.11	0.20	9
1895	1.00	0.20	0.33	10
1896	1.00	0.50	0.67	2
1897	0.00	0.00	0.00	5
1898	1.00	0.33	0.50	9
1899	0.00	0.00	0.00	6
1900	0.00	0.00	0.00	3
1901	0.00	0.00	0.00	7
1902	1.00	0.38	0.55	8
1903	0.00	0.00	0.00	6
1904	0.00	0.00	0.00	6
1905	1.00	0.44	0.62	9
1906	1.00	0.22	0.36	9
1907	0.00	0.00	0.00	6
1908	1.00	0.14	0.25	7
1909	0.00	0.00	0.00	10
1910	0.00	0.00	0.00	10
1911	1.00	0.38	0.55	8
1912	1.00	0.09	0.17	11
1913	0.00	0.00	0.00	6
1914	0.00	0.00	0.00	4
1915	0.00	0.00	0.00	5
1916	0.00	0.00	0.00	13
1917	0.80	0.50	0.62	8
1918	0.50	0.12	0.20	8
1919	0.00	0.00	0.00	9
1920	0.00	0.00	0.00	8
1921	0.00	0.00	0.00	4
1922	0.00	0.00	0.00	7
1923	0.00	0.00	0.00	9
1924	0.00	0.00	0.00	5
1925	0.00	0.00	0.00	5
1926	0.00	0.00	0.00	10
1927	0.75	0.43	0.55	7
1928	0.00	0.00	0.00	5
1929	0.00	0.00	0.00	7
1930	0.00	0.00	0.00	4
1931	1.00	0.43	0.60	14
1932	0.00	0.00	0.00	7
1933	0.00	0.00	0.00	6
1934	0.00	0.00	0.00	9
1935	1.00	0.38	0.55	8
1936	0.00	0.00	0.00	9
1937	0.00	0.00	0.00	8
1938	0.00	0.00	0.00	9
1939	0.00	0.00	0.00	9
1940	0.50	0.50	0.50	2
1941	0.00	0.00	0.00	8
1942	0.00	0.00	0.00	19
1943	0.00	0.00	0.00	6
1944	0.00	0.00	0.00	5

1945	0.50	0.25	0.33	4
1946	0.00	0.00	0.00	8
1947	0.00	0.00	0.00	5
1948	1.00	0.17	0.29	6
1949	0.00	0.00	0.00	9
1950	0.00	0.00	0.00	10
1951	0.00	0.00	0.00	8
1952	0.00	0.00	0.00	7
1953	0.00	0.00	0.00	7
1954	0.00	0.00	0.00	12
1955	0.00	0.00	0.00	6
1956	0.00	0.00	0.00	8
1957	1.00	0.23	0.38	13
1958	0.00	0.00	0.00	5
1959	0.00	0.00	0.00	1
1960	0.00	0.00	0.00	10
1961	0.00	0.00	0.00	4
1962	0.00	0.00	0.00	10
1963	0.75	0.50	0.60	6
1964	0.33	0.20	0.25	5
1965	0.33	0.11	0.17	9
1966	0.75	0.75	0.75	4
1967	0.00	0.00	0.00	2
1968	0.00	0.00	0.00	5
1969	0.50	0.11	0.18	9
1970	0.00	0.00	0.00	10
1971	0.00	0.00	0.00	6
1972	0.00	0.00	0.00	9
1973	0.00	0.00	0.00	9
1974	0.00	0.00	0.00	10
1975	0.00	0.00	0.00	10
1976	0.00	0.00	0.00	10
1977	0.00	0.00	0.00	9
1978	0.00	0.00	0.00	8
1979	1.00	0.22	0.36	9
1980	1.00	0.17	0.29	6
1981	0.00	0.00	0.00	6
1982	1.00	0.20	0.33	5
1983	0.00	0.00	0.00	10
1984	0.00	0.00	0.00	8
1985	0.00	0.00	0.00	9
1986	1.00	0.20	0.33	10
1987	0.00	0.00	0.00	8
1988	0.00	0.00	0.00	8
1989	0.00	0.00	0.00	4
1990	0.00	0.00	0.00	4
1991	0.50	0.14	0.22	7
1992	0.00	0.00	0.00	8
1993	0.00	0.00	0.00	9
1994	1.00	0.14	0.25	7
1995	0.00	0.00	0.00	5
1996	0.00	0.00	0.00	8
1997	1.00	0.12	0.22	8
1998	0.00	0.00	0.00	8
1999	0.00	0.00	0.00	4
2000	1.00	0.50	0.67	2
2001	0.00	0.00	0.00	9
2002	1.00	0.12	0.22	8
2003	0.33	0.08	0.12	13
2004	1.00	0.43	0.60	7
2005	1.00	0.12	0.22	8

2006	0.00	0.00	0.00	11
2007	0.00	0.00	0.00	8
2008	0.00	0.00	0.00	8
2009	0.00	0.00	0.00	8
2010	1.00	0.38	0.55	8
2011	0.00	0.00	0.00	3
2012	1.00	0.09	0.17	11
2013	0.00	0.00	0.00	12
2014	0.00	0.00	0.00	9
2015	0.00	0.00	0.00	10
2016	0.00	0.00	0.00	6
2017	0.00	0.00	0.00	7
2018	0.00	0.00	0.00	2
2019	1.00	0.20	0.33	5
2020	1.00	0.25	0.40	4
2021	0.00	0.00	0.00	5
2022	0.00	0.00	0.00	9
2023	0.00	0.00	0.00	10
2024	0.00	0.00	0.00	9
2025	0.00	0.00	0.00	7
2026	0.00	0.00	0.00	5
2027	0.00	0.00	0.00	10
2028	0.00	0.00	0.00	3
2029	0.00	0.00	0.00	5
2030	0.00	0.00	0.00	8
2031	1.00	0.56	0.71	9
2032	0.00	0.00	0.00	10
2033	0.00	0.00	0.00	9
2034	0.00	0.00	0.00	5
2035	1.00	0.14	0.25	7
2036	0.00	0.00	0.00	9
2037	0.00	0.00	0.00	6
2038	0.00	0.00	0.00	8
2039	0.00	0.00	0.00	9
2040	0.00	0.00	0.00	3
2041	1.00	0.12	0.22	8
2042	0.00	0.00	0.00	3
2043	1.00	0.12	0.22	8
2044	1.00	0.20	0.33	10
2045	0.00	0.00	0.00	5
2046	1.00	0.20	0.33	5
2047	1.00	0.20	0.33	5
2048	0.00	0.00	0.00	8
2049	0.00	0.00	0.00	4
2050	0.00	0.00	0.00	3
2051	0.00	0.00	0.00	8
2052	0.50	0.40	0.44	5
2053	0.00	0.00	0.00	4
2054	0.00	0.00	0.00	10
2055	0.00	0.00	0.00	6
2056	0.00	0.00	0.00	4
2057	0.00	0.00	0.00	7
2058	0.00	0.00	0.00	6
2059	0.00	0.00	0.00	3
2060	1.00	0.14	0.25	7
2061	1.00	0.80	0.89	5
2062	0.00	0.00	0.00	2
2063	0.00	0.00	0.00	6
2064	0.20	0.20	0.20	5
2065	1.00	0.40	0.57	5
2066	0.00	0.00	0.00	11

2067	0.00	0.00	0.00	6
2068	0.67	0.33	0.44	6
2069	0.00	0.00	0.00	7
2070	0.00	0.00	0.00	8
2071	1.00	0.20	0.33	5
2072	0.00	0.00	0.00	4
2073	0.00	0.00	0.00	4
2074	0.00	0.00	0.00	10
2075	0.00	0.00	0.00	8
2076	0.00	0.00	0.00	12
2077	0.00	0.00	0.00	8
2078	0.00	0.00	0.00	6
2079	0.00	0.00	0.00	6
2080	0.00	0.00	0.00	5
2081	1.00	0.75	0.86	4
2082	0.00	0.00	0.00	8
2083	0.00	0.00	0.00	4
2084	0.00	0.00	0.00	12
2085	0.00	0.00	0.00	3
2086	0.00	0.00	0.00	9
2087	0.00	0.00	0.00	8
2088	0.00	0.00	0.00	5
2089	0.00	0.00	0.00	9
2090	1.00	0.50	0.67	4
2091	1.00	0.29	0.44	7
2092	0.00	0.00	0.00	7
2093	0.00	0.00	0.00	8
2094	0.00	0.00	0.00	8
2095	0.00	0.00	0.00	5
2096	0.00	0.00	0.00	9
2097	1.00	0.12	0.22	8
2098	0.00	0.00	0.00	4
2099	0.00	0.00	0.00	7
2100	1.00	0.83	0.91	6
2101	1.00	0.43	0.60	7
2102	1.00	0.17	0.29	6
2103	1.00	0.50	0.67	10
2104	0.00	0.00	0.00	10
2105	0.00	0.00	0.00	8
2106	0.75	0.43	0.55	7
2107	0.00	0.00	0.00	8
2108	0.00	0.00	0.00	5
2109	1.00	0.20	0.33	10
2110	0.00	0.00	0.00	6
2111	0.00	0.00	0.00	7
2112	0.00	0.00	0.00	7
2113	0.00	0.00	0.00	7
2114	0.00	0.00	0.00	6
2115	0.00	0.00	0.00	5
2116	1.00	0.38	0.55	8
2117	0.00	0.00	0.00	5
2118	0.00	0.00	0.00	6
2119	0.00	0.00	0.00	7
2120	0.25	0.14	0.18	7
2121	0.00	0.00	0.00	3
2122	0.50	0.17	0.25	6
2123	0.00	0.00	0.00	3
2124	0.00	0.00	0.00	6
2125	1.00	0.12	0.22	8
2126	1.00	0.08	0.15	12
2127	0.00	0.00	0.00	7

2128	0.50	0.20	0.29	5
2129	0.00	0.00	0.00	4
2130	0.00	0.00	0.00	6
2131	1.00	0.14	0.25	7
2132	0.00	0.00	0.00	9
2133	0.33	0.20	0.25	5
2134	0.00	0.00	0.00	13
2135	0.00	0.00	0.00	11
2136	0.00	0.00	0.00	7
2137	0.00	0.00	0.00	9
2138	0.00	0.00	0.00	2
2139	1.00	0.29	0.44	7
2140	0.67	0.40	0.50	5
2141	0.00	0.00	0.00	7
2142	0.00	0.00	0.00	5
2143	0.00	0.00	0.00	8
2144	0.00	0.00	0.00	6
2145	0.00	0.00	0.00	8
2146	0.00	0.00	0.00	9
2147	0.00	0.00	0.00	7
2148	0.00	0.00	0.00	7
2149	0.00	0.00	0.00	0
2150	1.00	0.43	0.60	7
2151	0.00	0.00	0.00	5
2152	0.00	0.00	0.00	11
2153	1.00	0.20	0.33	5
2154	0.00	0.00	0.00	8
2155	0.00	0.00	0.00	4
2156	0.00	0.00	0.00	3
2157	1.00	0.25	0.40	4
2158	0.00	0.00	0.00	6
2159	0.00	0.00	0.00	5
2160	0.00	0.00	0.00	7
2161	0.00	0.00	0.00	10
2162	0.00	0.00	0.00	5
2163	0.00	0.00	0.00	3
2164	0.00	0.00	0.00	11
2165	1.00	0.33	0.50	3
2166	0.00	0.00	0.00	5
2167	0.00	0.00	0.00	4
2168	1.00	0.17	0.29	6
2169	0.00	0.00	0.00	5
2170	0.00	0.00	0.00	5
2171	1.00	0.14	0.25	7
2172	0.00	0.00	0.00	9
2173	0.00	0.00	0.00	13
2174	0.00	0.00	0.00	9
2175	0.00	0.00	0.00	4
2176	0.00	0.00	0.00	10
2177	0.00	0.00	0.00	5
2178	0.00	0.00	0.00	10
2179	0.00	0.00	0.00	9
2180	0.00	0.00	0.00	7
2181	0.00	0.00	0.00	5
2182	0.00	0.00	0.00	7
2183	1.00	0.10	0.18	10
2184	0.00	0.00	0.00	8
2185	0.00	0.00	0.00	8
2186	0.00	0.00	0.00	9
2187	0.00	0.00	0.00	7
2188	0.00	0.00	0.00	3

2189	1.00	0.50	0.67	6
2190	1.00	0.67	0.80	9
2191	0.00	0.00	0.00	5
2192	0.00	0.00	0.00	7
2193	0.00	0.00	0.00	6
2194	0.00	0.00	0.00	7
2195	1.00	0.08	0.15	12
2196	0.00	0.00	0.00	7
2197	0.00	0.00	0.00	4
2198	0.00	0.00	0.00	3
2199	0.00	0.00	0.00	3
2200	0.00	0.00	0.00	6
2201	1.00	0.20	0.33	5
2202	0.67	0.20	0.31	10
2203	0.00	0.00	0.00	6
2204	1.00	0.20	0.33	5
2205	0.00	0.00	0.00	9
2206	0.00	0.00	0.00	7
2207	0.00	0.00	0.00	6
2208	0.00	0.00	0.00	4
2209	0.00	0.00	0.00	4
2210	0.00	0.00	0.00	5
2211	0.00	0.00	0.00	6
2212	0.00	0.00	0.00	2
2213	0.00	0.00	0.00	10
2214	0.00	0.00	0.00	9
2215	0.00	0.00	0.00	7
2216	0.00	0.00	0.00	5
2217	0.00	0.00	0.00	7
2218	0.00	0.00	0.00	6
2219	1.00	0.17	0.29	6
2220	0.00	0.00	0.00	8
2221	0.00	0.00	0.00	5
2222	1.00	0.29	0.44	7
2223	0.00	0.00	0.00	8
2224	0.00	0.00	0.00	8
2225	1.00	0.14	0.25	7
2226	0.00	0.00	0.00	6
2227	0.00	0.00	0.00	10
2228	0.00	0.00	0.00	6
2229	0.00	0.00	0.00	8
2230	0.00	0.00	0.00	5
2231	0.00	0.00	0.00	5
2232	0.00	0.00	0.00	4
2233	0.00	0.00	0.00	5
2234	1.00	0.33	0.50	6
2235	0.00	0.00	0.00	3
2236	1.00	0.25	0.40	4
2237	0.00	0.00	0.00	8
2238	0.00	0.00	0.00	3
2239	0.00	0.00	0.00	4
2240	0.00	0.00	0.00	6
2241	0.00	0.00	0.00	3
2242	0.00	0.00	0.00	6
2243	0.00	0.00	0.00	11
2244	0.00	0.00	0.00	3
2245	0.00	0.00	0.00	9
2246	1.00	0.14	0.25	7
2247	0.00	0.00	0.00	6
2248	0.00	0.00	0.00	4
2249	0.00	0.00	0.00	7

2250	0.00	0.00	0.00	2
2251	0.00	0.00	0.00	11
2252	0.00	0.00	0.00	1
2253	0.00	0.00	0.00	3
2254	0.00	0.00	0.00	9
2255	1.00	0.08	0.14	13
2256	1.00	0.11	0.20	9
2257	0.00	0.00	0.00	4
2258	0.00	0.00	0.00	8
2259	0.00	0.00	0.00	10
2260	0.00	0.00	0.00	9
2261	0.00	0.00	0.00	5
2262	0.00	0.00	0.00	4
2263	0.00	0.00	0.00	7
2264	0.00	0.00	0.00	11
2265	0.50	0.33	0.40	6
2266	0.00	0.00	0.00	4
2267	0.00	0.00	0.00	5
2268	0.00	0.00	0.00	8
2269	0.00	0.00	0.00	6
2270	0.00	0.00	0.00	6
2271	0.33	0.14	0.20	7
2272	0.00	0.00	0.00	5
2273	0.00	0.00	0.00	4
2274	0.50	0.33	0.40	3
2275	0.00	0.00	0.00	7
2276	0.00	0.00	0.00	3
2277	0.00	0.00	0.00	7
2278	0.00	0.00	0.00	6
2279	0.00	0.00	0.00	7
2280	0.00	0.00	0.00	3
2281	0.00	0.00	0.00	6
2282	0.00	0.00	0.00	7
2283	0.00	0.00	0.00	3
2284	0.00	0.00	0.00	6
2285	0.00	0.00	0.00	5
2286	0.00	0.00	0.00	5
2287	0.25	0.20	0.22	5
2288	0.00	0.00	0.00	6
2289	0.00	0.00	0.00	9
2290	0.00	0.00	0.00	16
2291	0.00	0.00	0.00	2
2292	0.00	0.00	0.00	4
2293	0.00	0.00	0.00	5
2294	0.00	0.00	0.00	5
2295	0.00	0.00	0.00	9
2296	0.00	0.00	0.00	9
2297	0.00	0.00	0.00	3
2298	0.00	0.00	0.00	6
2299	0.00	0.00	0.00	3
2300	0.00	0.00	0.00	7
2301	0.00	0.00	0.00	11
2302	0.00	0.00	0.00	4
2303	0.00	0.00	0.00	10
2304	0.00	0.00	0.00	1
2305	0.00	0.00	0.00	6
2306	1.00	0.14	0.25	7
2307	0.00	0.00	0.00	4
2308	0.00	0.00	0.00	6
2309	0.00	0.00	0.00	3
2310	0.00	0.00	0.00	5

2311	0.00	0.00	0.00	11
2312	0.00	0.00	0.00	7
2313	0.00	0.00	0.00	9
2314	0.00	0.00	0.00	6
2315	0.00	0.00	0.00	8
2316	0.00	0.00	0.00	7
2317	0.00	0.00	0.00	8
2318	0.00	0.00	0.00	3
2319	0.00	0.00	0.00	4
2320	0.00	0.00	0.00	7
2321	0.00	0.00	0.00	6
2322	0.00	0.00	0.00	8
2323	0.00	0.00	0.00	4
2324	0.00	0.00	0.00	8
2325	0.00	0.00	0.00	7
2326	0.00	0.00	0.00	5
2327	0.33	0.25	0.29	4
2328	0.00	0.00	0.00	9
2329	0.00	0.00	0.00	6
2330	0.00	0.00	0.00	5
2331	0.00	0.00	0.00	6
2332	1.00	0.67	0.80	6
2333	0.00	0.00	0.00	4
2334	0.00	0.00	0.00	8
2335	0.00	0.00	0.00	7
2336	0.00	0.00	0.00	9
2337	0.00	0.00	0.00	3
2338	0.00	0.00	0.00	4
2339	0.00	0.00	0.00	4
2340	1.00	1.00	1.00	1
2341	0.00	0.00	0.00	3
2342	0.00	0.00	0.00	6
2343	0.00	0.00	0.00	4
2344	0.00	0.00	0.00	9
2345	0.00	0.00	0.00	3
2346	0.00	0.00	0.00	3
2347	0.00	0.00	0.00	6
2348	0.00	0.00	0.00	5
2349	0.00	0.00	0.00	7
2350	0.00	0.00	0.00	4
2351	0.00	0.00	0.00	8
2352	0.00	0.00	0.00	6
2353	0.00	0.00	0.00	5
2354	0.00	0.00	0.00	5
2355	1.00	0.20	0.33	10
2356	0.00	0.00	0.00	9
2357	0.00	0.00	0.00	7
2358	0.00	0.00	0.00	6
2359	0.00	0.00	0.00	4
2360	1.00	0.17	0.29	6
2361	0.00	0.00	0.00	2
2362	0.50	0.10	0.17	10
2363	0.00	0.00	0.00	8
2364	0.00	0.00	0.00	4
2365	0.00	0.00	0.00	6
2366	0.00	0.00	0.00	6
2367	1.00	0.30	0.46	10
2368	0.00	0.00	0.00	3
2369	0.00	0.00	0.00	3
2370	0.00	0.00	0.00	9
2371	0.00	0.00	0.00	9

2372	0.00	0.00	0.00	4
2373	0.00	0.00	0.00	9
2374	0.00	0.00	0.00	8
2375	0.00	0.00	0.00	5
2376	0.00	0.00	0.00	7
2377	0.50	0.14	0.22	7
2378	0.00	0.00	0.00	7
2379	0.00	0.00	0.00	8
2380	0.00	0.00	0.00	6
2381	1.00	0.17	0.29	6
2382	0.00	0.00	0.00	10
2383	0.00	0.00	0.00	5
2384	0.00	0.00	0.00	5
2385	0.00	0.00	0.00	4
2386	0.00	0.00	0.00	3
2387	0.00	0.00	0.00	3
2388	0.00	0.00	0.00	11
2389	1.00	0.20	0.33	5
2390	0.00	0.00	0.00	8
2391	0.00	0.00	0.00	4
2392	0.00	0.00	0.00	8
2393	0.00	0.00	0.00	5
2394	0.00	0.00	0.00	4
2395	0.00	0.00	0.00	2
2396	0.00	0.00	0.00	5
2397	0.00	0.00	0.00	7
2398	0.00	0.00	0.00	4
2399	0.00	0.00	0.00	5
2400	0.00	0.00	0.00	5
2401	1.00	0.33	0.50	6
2402	0.00	0.00	0.00	4
2403	0.00	0.00	0.00	5
2404	0.00	0.00	0.00	7
2405	1.00	0.12	0.22	8
2406	0.00	0.00	0.00	2
2407	0.00	0.00	0.00	3
2408	0.00	0.00	0.00	4
2409	0.00	0.00	0.00	6
2410	0.00	0.00	0.00	7
2411	0.00	0.00	0.00	1
2412	0.00	0.00	0.00	6
2413	0.00	0.00	0.00	2
2414	0.00	0.00	0.00	7
2415	1.00	0.14	0.25	7
2416	0.00	0.00	0.00	3
2417	0.00	0.00	0.00	5
2418	0.00	0.00	0.00	7
2419	0.00	0.00	0.00	6
2420	1.00	0.17	0.29	6
2421	0.00	0.00	0.00	6
2422	0.00	0.00	0.00	4
2423	0.00	0.00	0.00	3
2424	0.00	0.00	0.00	5
2425	0.00	0.00	0.00	9
2426	0.00	0.00	0.00	9
2427	0.50	0.17	0.25	6
2428	0.00	0.00	0.00	3
2429	0.00	0.00	0.00	3
2430	0.00	0.00	0.00	4
2431	0.00	0.00	0.00	5
2432	0.00	0.00	0.00	7

2433	0.00	0.00	0.00	5
2434	0.00	0.00	0.00	5
2435	0.00	0.00	0.00	7
2436	0.00	0.00	0.00	8
2437	1.00	0.14	0.25	7
2438	0.00	0.00	0.00	7
2439	0.00	0.00	0.00	10
2440	0.00	0.00	0.00	3
2441	0.00	0.00	0.00	4
2442	0.00	0.00	0.00	7
2443	0.00	0.00	0.00	6
2444	1.00	0.40	0.57	5
2445	0.00	0.00	0.00	2
2446	0.00	0.00	0.00	9
2447	0.00	0.00	0.00	2
2448	0.00	0.00	0.00	9
2449	0.00	0.00	0.00	5
2450	0.00	0.00	0.00	7
2451	0.00	0.00	0.00	9
2452	0.00	0.00	0.00	4
2453	0.00	0.00	0.00	7
2454	0.00	0.00	0.00	6
2455	0.00	0.00	0.00	3
2456	0.00	0.00	0.00	2
2457	0.00	0.00	0.00	1
2458	0.67	0.33	0.44	6
2459	0.00	0.00	0.00	6
2460	0.00	0.00	0.00	6
2461	0.00	0.00	0.00	4
2462	0.00	0.00	0.00	8
2463	1.00	0.29	0.44	7
2464	0.00	0.00	0.00	5
2465	0.00	0.00	0.00	12
2466	0.00	0.00	0.00	3
2467	0.00	0.00	0.00	4
2468	0.00	0.00	0.00	7
2469	0.00	0.00	0.00	5
2470	0.00	0.00	0.00	2
2471	0.00	0.00	0.00	7
2472	0.00	0.00	0.00	3
2473	0.00	0.00	0.00	5
2474	0.00	0.00	0.00	7
2475	0.00	0.00	0.00	4
2476	0.00	0.00	0.00	5
2477	0.00	0.00	0.00	7
2478	0.00	0.00	0.00	4
2479	0.00	0.00	0.00	3
2480	0.00	0.00	0.00	7
2481	0.00	0.00	0.00	6
2482	0.00	0.00	0.00	3
2483	0.00	0.00	0.00	4
2484	1.00	0.17	0.29	6
2485	0.00	0.00	0.00	7
2486	0.00	0.00	0.00	3
2487	0.00	0.00	0.00	4
2488	0.00	0.00	0.00	4
2489	0.00	0.00	0.00	4
2490	0.00	0.00	0.00	4
2491	0.00	0.00	0.00	8
2492	0.00	0.00	0.00	6
2493	0.00	0.00	0.00	3

2494	0.00	0.00	0.00	5
2495	0.00	0.00	0.00	5
2496	0.00	0.00	0.00	5
2497	0.00	0.00	0.00	3
2498	0.00	0.00	0.00	4
2499	0.00	0.00	0.00	8
2500	0.00	0.00	0.00	5
2501	1.00	0.60	0.75	5
2502	0.50	0.20	0.29	5
2503	0.00	0.00	0.00	5
2504	0.00	0.00	0.00	5
2505	0.00	0.00	0.00	5
2506	0.00	0.00	0.00	9
2507	0.00	0.00	0.00	1
2508	0.00	0.00	0.00	7
2509	0.00	0.00	0.00	5
2510	0.00	0.00	0.00	4
2511	0.00	0.00	0.00	4
2512	0.00	0.00	0.00	4
2513	0.00	0.00	0.00	7
2514	0.00	0.00	0.00	6
2515	0.00	0.00	0.00	5
2516	0.00	0.00	0.00	4
2517	0.00	0.00	0.00	5
2518	0.00	0.00	0.00	2
2519	0.00	0.00	0.00	5
2520	0.00	0.00	0.00	7
2521	0.00	0.00	0.00	4
2522	0.00	0.00	0.00	3
2523	1.00	0.14	0.25	7
2524	0.00	0.00	0.00	1
2525	0.00	0.00	0.00	4
2526	0.00	0.00	0.00	4
2527	0.00	0.00	0.00	5
2528	0.00	0.00	0.00	3
2529	0.00	0.00	0.00	3
2530	1.00	0.12	0.22	8
2531	0.00	0.00	0.00	6
2532	1.00	0.40	0.57	10
2533	0.00	0.00	0.00	9
2534	0.00	0.00	0.00	6
2535	0.00	0.00	0.00	5
2536	0.00	0.00	0.00	2
2537	0.00	0.00	0.00	3
2538	0.00	0.00	0.00	4
2539	0.00	0.00	0.00	1
2540	0.00	0.00	0.00	3
2541	0.00	0.00	0.00	8
2542	0.00	0.00	0.00	9
2543	0.00	0.00	0.00	7
2544	0.00	0.00	0.00	5
2545	0.00	0.00	0.00	7
2546	0.00	0.00	0.00	4
2547	0.00	0.00	0.00	7
2548	0.00	0.00	0.00	6
2549	0.00	0.00	0.00	4
2550	0.00	0.00	0.00	2
2551	1.00	0.50	0.67	4
2552	0.00	0.00	0.00	4
2553	1.00	0.20	0.33	5
2554	0.00	0.00	0.00	4

2555	0.00	0.00	0.00	7
2556	0.00	0.00	0.00	8
2557	0.00	0.00	0.00	5
2558	0.00	0.00	0.00	5
2559	0.00	0.00	0.00	5
2560	0.00	0.00	0.00	2
2561	0.00	0.00	0.00	4
2562	0.00	0.00	0.00	4
2563	0.00	0.00	0.00	5
2564	0.33	0.33	0.33	3
2565	0.50	0.12	0.20	8
2566	0.00	0.00	0.00	8
2567	0.75	0.43	0.55	7
2568	1.00	0.17	0.29	6
2569	0.00	0.00	0.00	3
2570	0.00	0.00	0.00	1
2571	0.00	0.00	0.00	7
2572	0.00	0.00	0.00	5
2573	0.00	0.00	0.00	3
2574	0.00	0.00	0.00	2
2575	1.00	0.50	0.67	6
2576	0.00	0.00	0.00	2
2577	0.00	0.00	0.00	3
2578	0.00	0.00	0.00	5
2579	0.00	0.00	0.00	1
2580	0.00	0.00	0.00	2
2581	0.00	0.00	0.00	5
2582	0.00	0.00	0.00	1
2583	0.00	0.00	0.00	4
2584	0.00	0.00	0.00	11
2585	0.00	0.00	0.00	5
2586	0.00	0.00	0.00	4
2587	0.00	0.00	0.00	1
2588	0.00	0.00	0.00	7
2589	0.00	0.00	0.00	6
2590	0.00	0.00	0.00	5
2591	0.00	0.00	0.00	7
2592	0.00	0.00	0.00	5
2593	0.00	0.00	0.00	6
2594	1.00	0.25	0.40	4
2595	0.00	0.00	0.00	4
2596	0.00	0.00	0.00	2
2597	0.00	0.00	0.00	9
2598	0.00	0.00	0.00	6
2599	0.00	0.00	0.00	5
2600	0.00	0.00	0.00	3
2601	0.00	0.00	0.00	3
2602	0.00	0.00	0.00	4
2603	0.00	0.00	0.00	5
2604	0.00	0.00	0.00	8
2605	0.00	0.00	0.00	6
2606	0.00	0.00	0.00	7
2607	0.00	0.00	0.00	7
2608	0.00	0.00	0.00	7
2609	0.00	0.00	0.00	4
2610	0.00	0.00	0.00	2
2611	0.00	0.00	0.00	10
2612	0.00	0.00	0.00	4
2613	0.00	0.00	0.00	3
2614	0.00	0.00	0.00	6
2615	0.00	0.00	0.00	5

2616	0.00	0.00	0.00	3
2617	0.00	0.00	0.00	5
2618	0.00	0.00	0.00	7
2619	0.00	0.00	0.00	4
2620	1.00	0.40	0.57	5
2621	0.00	0.00	0.00	3
2622	0.00	0.00	0.00	5
2623	0.00	0.00	0.00	5
2624	0.00	0.00	0.00	5
2625	0.00	0.00	0.00	2
2626	0.00	0.00	0.00	7
2627	0.00	0.00	0.00	6
2628	0.00	0.00	0.00	7
2629	0.00	0.00	0.00	4
2630	0.00	0.00	0.00	4
2631	0.00	0.00	0.00	2
2632	0.00	0.00	0.00	8
2633	0.00	0.00	0.00	1
2634	0.00	0.00	0.00	5
2635	0.00	0.00	0.00	6
2636	0.00	0.00	0.00	3
2637	0.00	0.00	0.00	4
2638	0.00	0.00	0.00	8
2639	1.00	0.40	0.57	5
2640	0.00	0.00	0.00	3
2641	0.00	0.00	0.00	8
2642	1.00	0.17	0.29	6
2643	0.00	0.00	0.00	3
2644	0.00	0.00	0.00	7
2645	0.00	0.00	0.00	4
2646	0.00	0.00	0.00	4
2647	0.00	0.00	0.00	3
2648	0.00	0.00	0.00	3
2649	0.00	0.00	0.00	10
2650	0.00	0.00	0.00	3
2651	1.00	0.25	0.40	4
2652	0.00	0.00	0.00	9
2653	0.00	0.00	0.00	4
2654	0.00	0.00	0.00	4
2655	0.00	0.00	0.00	3
2656	0.00	0.00	0.00	5
2657	0.00	0.00	0.00	7
2658	0.00	0.00	0.00	5
2659	0.00	0.00	0.00	4
2660	0.00	0.00	0.00	4
2661	0.00	0.00	0.00	3
2662	0.00	0.00	0.00	8
2663	0.00	0.00	0.00	8
2664	0.00	0.00	0.00	10
2665	0.00	0.00	0.00	5
2666	0.00	0.00	0.00	10
2667	1.00	0.25	0.40	4
2668	0.00	0.00	0.00	5
2669	1.00	0.20	0.33	5
2670	0.00	0.00	0.00	6
2671	0.00	0.00	0.00	8
2672	0.00	0.00	0.00	5
2673	0.00	0.00	0.00	7
2674	0.00	0.00	0.00	4
2675	0.00	0.00	0.00	3
2676	0.00	0.00	0.00	3

2677	0.00	0.00	0.00	5
2678	0.00	0.00	0.00	7
2679	0.00	0.00	0.00	5
2680	0.00	0.00	0.00	0
2681	0.00	0.00	0.00	8
2682	0.00	0.00	0.00	5
2683	0.00	0.00	0.00	11
2684	0.00	0.00	0.00	7
2685	0.00	0.00	0.00	7
2686	0.00	0.00	0.00	5
2687	0.00	0.00	0.00	1
2688	0.00	0.00	0.00	6
2689	0.00	0.00	0.00	5
2690	1.00	0.17	0.29	6
2691	0.00	0.00	0.00	5
2692	1.00	0.17	0.29	6
2693	0.00	0.00	0.00	7
2694	0.00	0.00	0.00	6
2695	0.60	0.33	0.43	9
2696	0.00	0.00	0.00	3
2697	0.00	0.00	0.00	3
2698	0.00	0.00	0.00	2
2699	0.00	0.00	0.00	1
2700	0.00	0.00	0.00	6
2701	0.00	0.00	0.00	5
2702	0.00	0.00	0.00	4
2703	0.00	0.00	0.00	5
2704	0.00	0.00	0.00	3
2705	0.00	0.00	0.00	7
2706	0.00	0.00	0.00	4
2707	0.00	0.00	0.00	8
2708	0.00	0.00	0.00	7
2709	0.00	0.00	0.00	4
2710	0.00	0.00	0.00	4
2711	0.00	0.00	0.00	2
2712	0.00	0.00	0.00	4
2713	0.00	0.00	0.00	6
2714	0.00	0.00	0.00	3
2715	0.00	0.00	0.00	10
2716	1.00	0.33	0.50	3
2717	0.00	0.00	0.00	6
2718	0.00	0.00	0.00	5
2719	0.00	0.00	0.00	9
2720	0.00	0.00	0.00	6
2721	0.00	0.00	0.00	3
2722	0.00	0.00	0.00	3
2723	0.00	0.00	0.00	4
2724	1.00	0.29	0.44	7
2725	0.00	0.00	0.00	4
2726	0.00	0.00	0.00	4
2727	0.00	0.00	0.00	4
2728	0.00	0.00	0.00	6
2729	0.00	0.00	0.00	8
2730	0.00	0.00	0.00	7
2731	0.00	0.00	0.00	7
2732	0.00	0.00	0.00	5
2733	0.00	0.00	0.00	7
2734	0.00	0.00	0.00	7
2735	0.00	0.00	0.00	3
2736	0.00	0.00	0.00	7
2737	0.00	0.00	0.00	3

2738	0.00	0.00	0.00	5
2739	0.00	0.00	0.00	8
2740	1.00	0.20	0.33	5
2741	0.00	0.00	0.00	7
2742	0.00	0.00	0.00	8
2743	0.00	0.00	0.00	7
2744	0.00	0.00	0.00	8
2745	0.00	0.00	0.00	5
2746	0.00	0.00	0.00	7
2747	1.00	1.00	1.00	4
2748	0.00	0.00	0.00	6
2749	0.00	0.00	0.00	4
2750	0.00	0.00	0.00	4
2751	0.00	0.00	0.00	5
2752	0.00	0.00	0.00	5
2753	0.00	0.00	0.00	2
2754	0.00	0.00	0.00	5
2755	0.00	0.00	0.00	4
2756	0.00	0.00	0.00	5
2757	0.00	0.00	0.00	4
2758	0.00	0.00	0.00	10
2759	0.00	0.00	0.00	3
2760	0.00	0.00	0.00	2
2761	0.00	0.00	0.00	3
2762	0.00	0.00	0.00	1
2763	1.00	0.33	0.50	3
2764	0.00	0.00	0.00	3
2765	0.00	0.00	0.00	6
2766	0.00	0.00	0.00	5
2767	0.00	0.00	0.00	4
2768	0.00	0.00	0.00	9
2769	0.00	0.00	0.00	6
2770	0.00	0.00	0.00	6
2771	0.00	0.00	0.00	4
2772	0.67	0.25	0.36	8
2773	0.00	0.00	0.00	6
2774	0.00	0.00	0.00	7
2775	0.00	0.00	0.00	7
2776	0.00	0.00	0.00	4
2777	0.00	0.00	0.00	5
2778	0.00	0.00	0.00	5
2779	0.00	0.00	0.00	2
2780	0.00	0.00	0.00	9
2781	0.00	0.00	0.00	3
2782	0.00	0.00	0.00	7
2783	0.00	0.00	0.00	7
2784	0.00	0.00	0.00	5
2785	0.00	0.00	0.00	8
2786	0.00	0.00	0.00	5
2787	0.00	0.00	0.00	5
2788	1.00	0.25	0.40	4
2789	0.00	0.00	0.00	4
2790	0.00	0.00	0.00	3
2791	0.00	0.00	0.00	4
2792	0.00	0.00	0.00	8
2793	0.00	0.00	0.00	4
2794	0.00	0.00	0.00	7
2795	0.00	0.00	0.00	2
2796	0.00	0.00	0.00	1
2797	1.00	0.29	0.44	7
2798	0.00	0.00	0.00	3

2799	0.00	0.00	0.00	3
2800	0.00	0.00	0.00	8
2801	0.00	0.00	0.00	1
2802	0.00	0.00	0.00	4
2803	0.00	0.00	0.00	7
2804	0.00	0.00	0.00	4
2805	0.00	0.00	0.00	3
2806	0.00	0.00	0.00	3
2807	0.00	0.00	0.00	5
2808	0.00	0.00	0.00	5
2809	0.00	0.00	0.00	2
2810	0.00	0.00	0.00	8
2811	0.00	0.00	0.00	4
2812	0.00	0.00	0.00	3
2813	0.00	0.00	0.00	3
2814	0.00	0.00	0.00	3
2815	0.00	0.00	0.00	5
2816	0.00	0.00	0.00	3
2817	0.00	0.00	0.00	5
2818	0.00	0.00	0.00	2
2819	0.00	0.00	0.00	7
2820	0.00	0.00	0.00	3
2821	0.00	0.00	0.00	10
2822	0.00	0.00	0.00	3
2823	0.00	0.00	0.00	3
2824	0.00	0.00	0.00	7
2825	0.00	0.00	0.00	4
2826	0.00	0.00	0.00	3
2827	1.00	0.17	0.29	6
2828	0.00	0.00	0.00	4
2829	0.00	0.00	0.00	9
2830	0.00	0.00	0.00	4
2831	0.00	0.00	0.00	6
2832	0.00	0.00	0.00	5
2833	0.00	0.00	0.00	8
2834	0.00	0.00	0.00	7
2835	0.00	0.00	0.00	4
2836	0.00	0.00	0.00	4
2837	0.00	0.00	0.00	4
2838	0.00	0.00	0.00	6
2839	0.00	0.00	0.00	7
2840	0.00	0.00	0.00	6
2841	0.00	0.00	0.00	6
2842	0.00	0.00	0.00	5
2843	0.00	0.00	0.00	4
2844	0.00	0.00	0.00	4
2845	0.00	0.00	0.00	2
2846	0.00	0.00	0.00	3
2847	0.00	0.00	0.00	3
2848	0.00	0.00	0.00	8
2849	0.00	0.00	0.00	3
2850	0.00	0.00	0.00	3
2851	0.00	0.00	0.00	4
2852	0.00	0.00	0.00	1
2853	0.00	0.00	0.00	3
2854	0.00	0.00	0.00	4
2855	0.00	0.00	0.00	8
2856	0.00	0.00	0.00	8
2857	0.00	0.00	0.00	9
2858	0.00	0.00	0.00	1
2859	0.00	0.00	0.00	0

2860	0.00	0.00	0.00	4
2861	0.00	0.00	0.00	8
2862	0.00	0.00	0.00	4
2863	0.00	0.00	0.00	2
2864	0.00	0.00	0.00	5
2865	1.00	0.33	0.50	3
2866	0.00	0.00	0.00	2
2867	0.00	0.00	0.00	5
2868	0.00	0.00	0.00	1
2869	0.00	0.00	0.00	4
2870	0.00	0.00	0.00	1
2871	0.00	0.00	0.00	3
2872	0.00	0.00	0.00	1
2873	0.00	0.00	0.00	6
2874	0.00	0.00	0.00	4
2875	0.00	0.00	0.00	2
2876	0.00	0.00	0.00	4
2877	0.00	0.00	0.00	7
2878	0.00	0.00	0.00	4
2879	0.00	0.00	0.00	4
2880	0.00	0.00	0.00	4
2881	0.00	0.00	0.00	4
2882	0.00	0.00	0.00	5
2883	0.00	0.00	0.00	4
2884	0.00	0.00	0.00	4
2885	0.00	0.00	0.00	4
2886	0.00	0.00	0.00	4
2887	0.00	0.00	0.00	4
2888	0.00	0.00	0.00	5
2889	0.00	0.00	0.00	3
2890	0.00	0.00	0.00	5
2891	0.00	0.00	0.00	3
2892	0.00	0.00	0.00	2
2893	0.00	0.00	0.00	3
2894	0.00	0.00	0.00	6
2895	0.00	0.00	0.00	3
2896	0.00	0.00	0.00	5
2897	0.00	0.00	0.00	3
2898	0.00	0.00	0.00	5
2899	0.00	0.00	0.00	9
2900	0.00	0.00	0.00	3
2901	0.00	0.00	0.00	2
2902	0.00	0.00	0.00	6
2903	0.00	0.00	0.00	3
2904	0.00	0.00	0.00	2
2905	0.00	0.00	0.00	3
2906	0.00	0.00	0.00	6
2907	0.00	0.00	0.00	4
2908	0.00	0.00	0.00	2
2909	0.00	0.00	0.00	4
2910	0.00	0.00	0.00	2
2911	0.00	0.00	0.00	1
2912	1.00	0.25	0.40	4
2913	1.00	0.25	0.40	4
2914	0.00	0.00	0.00	4
2915	1.00	0.20	0.33	5
2916	0.00	0.00	0.00	4
2917	0.00	0.00	0.00	6
2918	0.00	0.00	0.00	7
2919	0.00	0.00	0.00	5
2920	0.00	0.00	0.00	7

2921	0.00	0.00	0.00	7
2922	0.00	0.00	0.00	6
2923	0.00	0.00	0.00	6
2924	0.00	0.00	0.00	5
2925	0.00	0.00	0.00	5
2926	0.00	0.00	0.00	4
2927	0.00	0.00	0.00	6
2928	0.00	0.00	0.00	3
2929	0.00	0.00	0.00	6
2930	0.00	0.00	0.00	2
2931	0.00	0.00	0.00	3
2932	0.00	0.00	0.00	6
2933	0.00	0.00	0.00	4
2934	0.00	0.00	0.00	2
2935	0.00	0.00	0.00	5
2936	0.00	0.00	0.00	4
2937	0.00	0.00	0.00	4
2938	0.00	0.00	0.00	4
2939	0.00	0.00	0.00	4
2940	0.00	0.00	0.00	6
2941	0.50	0.20	0.29	5
2942	0.00	0.00	0.00	4
2943	0.00	0.00	0.00	5
2944	0.00	0.00	0.00	4
2945	0.00	0.00	0.00	2
2946	0.00	0.00	0.00	6
2947	0.00	0.00	0.00	6
2948	0.00	0.00	0.00	5
2949	0.00	0.00	0.00	1
2950	0.00	0.00	0.00	3
2951	0.00	0.00	0.00	8
2952	0.00	0.00	0.00	6
2953	0.00	0.00	0.00	4
2954	0.00	0.00	0.00	5
2955	0.00	0.00	0.00	1
2956	0.00	0.00	0.00	3
2957	0.00	0.00	0.00	1
2958	1.00	0.40	0.57	5
2959	0.00	0.00	0.00	6
2960	0.00	0.00	0.00	2
2961	0.00	0.00	0.00	3
2962	0.00	0.00	0.00	7
2963	0.00	0.00	0.00	0
2964	0.00	0.00	0.00	1
2965	0.00	0.00	0.00	4
2966	0.00	0.00	0.00	6
2967	0.00	0.00	0.00	8
2968	0.00	0.00	0.00	8
2969	0.00	0.00	0.00	3
2970	0.00	0.00	0.00	3
2971	0.00	0.00	0.00	5
2972	0.00	0.00	0.00	5
2973	0.00	0.00	0.00	2
2974	0.00	0.00	0.00	0
2975	0.00	0.00	0.00	5
2976	0.00	0.00	0.00	2
2977	0.00	0.00	0.00	5
2978	0.00	0.00	0.00	7
2979	0.00	0.00	0.00	3
2980	0.00	0.00	0.00	3
2981	0.00	0.00	0.00	4

2982	0.00	0.00	0.00	4
2983	0.00	0.00	0.00	2
2984	0.00	0.00	0.00	3
2985	0.00	0.00	0.00	1
2986	0.00	0.00	0.00	3
2987	0.00	0.00	0.00	4
2988	1.00	1.00	1.00	1
2989	0.00	0.00	0.00	8
2990	0.00	0.00	0.00	3
2991	0.00	0.00	0.00	3
2992	1.00	0.67	0.80	3
2993	0.00	0.00	0.00	4
2994	0.00	0.00	0.00	2
2995	0.00	0.00	0.00	0
2996	0.00	0.00	0.00	6
2997	0.00	0.00	0.00	6
2998	0.00	0.00	0.00	6
2999	0.00	0.00	0.00	6
3000	0.00	0.00	0.00	7
3001	0.00	0.00	0.00	2
3002	0.00	0.00	0.00	4
3003	0.00	0.00	0.00	3
3004	0.00	0.00	0.00	3
3005	0.00	0.00	0.00	4
3006	0.00	0.00	0.00	4
3007	0.00	0.00	0.00	5
3008	0.00	0.00	0.00	3
3009	0.00	0.00	0.00	2
3010	0.00	0.00	0.00	2
3011	0.00	0.00	0.00	1
3012	0.00	0.00	0.00	8
3013	0.00	0.00	0.00	2
3014	0.00	0.00	0.00	5
3015	0.00	0.00	0.00	5
3016	1.00	0.25	0.40	4
3017	0.00	0.00	0.00	8
3018	0.00	0.00	0.00	4
3019	0.00	0.00	0.00	4
3020	0.00	0.00	0.00	5
3021	0.00	0.00	0.00	2
3022	0.00	0.00	0.00	5
3023	0.00	0.00	0.00	3
3024	0.00	0.00	0.00	5
3025	0.00	0.00	0.00	4
3026	0.00	0.00	0.00	1
3027	0.00	0.00	0.00	8
3028	0.00	0.00	0.00	2
3029	0.00	0.00	0.00	1
3030	0.00	0.00	0.00	2
3031	0.00	0.00	0.00	3
3032	0.00	0.00	0.00	3
3033	0.00	0.00	0.00	4
3034	0.00	0.00	0.00	2
3035	0.00	0.00	0.00	4
3036	0.00	0.00	0.00	6
3037	0.00	0.00	0.00	4
3038	0.00	0.00	0.00	5
3039	0.00	0.00	0.00	2
3040	0.00	0.00	0.00	5
3041	0.00	0.00	0.00	4
3042	0.00	0.00	0.00	0

3043	0.00	0.00	0.00	2
3044	0.00	0.00	0.00	5
3045	0.00	0.00	0.00	2
3046	0.00	0.00	0.00	2
3047	0.00	0.00	0.00	6
3048	0.00	0.00	0.00	2
3049	0.00	0.00	0.00	6
3050	0.00	0.00	0.00	5
3051	0.00	0.00	0.00	5
3052	0.00	0.00	0.00	1
3053	1.00	0.33	0.50	6
3054	0.00	0.00	0.00	3
3055	0.00	0.00	0.00	7
3056	0.00	0.00	0.00	4
3057	0.00	0.00	0.00	1
3058	0.00	0.00	0.00	2
3059	0.00	0.00	0.00	5
3060	0.00	0.00	0.00	3
3061	0.00	0.00	0.00	1
3062	0.00	0.00	0.00	3
3063	0.00	0.00	0.00	2
3064	0.00	0.00	0.00	4
3065	0.00	0.00	0.00	6
3066	0.00	0.00	0.00	2
3067	0.00	0.00	0.00	1
3068	0.00	0.00	0.00	6
3069	0.00	0.00	0.00	6
3070	0.00	0.00	0.00	3
3071	0.00	0.00	0.00	5
3072	0.00	0.00	0.00	2
3073	0.00	0.00	0.00	4
3074	0.00	0.00	0.00	2
3075	0.00	0.00	0.00	5
3076	0.00	0.00	0.00	2
3077	0.00	0.00	0.00	0
3078	0.00	0.00	0.00	3
3079	0.00	0.00	0.00	4
3080	0.00	0.00	0.00	6
3081	0.00	0.00	0.00	8
3082	0.00	0.00	0.00	4
3083	0.00	0.00	0.00	0
3084	0.00	0.00	0.00	2
3085	0.00	0.00	0.00	6
3086	0.00	0.00	0.00	1
3087	0.00	0.00	0.00	5
3088	0.00	0.00	0.00	1
3089	0.00	0.00	0.00	8
3090	0.00	0.00	0.00	4
3091	0.00	0.00	0.00	7
3092	0.00	0.00	0.00	2
3093	0.00	0.00	0.00	3
3094	0.00	0.00	0.00	9
3095	0.50	1.00	0.67	1
3096	0.00	0.00	0.00	4
3097	0.00	0.00	0.00	3
3098	0.00	0.00	0.00	4
3099	0.00	0.00	0.00	3
3100	0.00	0.00	0.00	3
3101	0.00	0.00	0.00	4
3102	0.00	0.00	0.00	4
3103	0.00	0.00	0.00	2

3104	0.00	0.00	0.00	3
3105	0.00	0.00	0.00	3
3106	0.00	0.00	0.00	2
3107	0.00	0.00	0.00	4
3108	0.00	0.00	0.00	2
3109	0.00	0.00	0.00	6
3110	0.00	0.00	0.00	1
3111	0.00	0.00	0.00	5
3112	0.00	0.00	0.00	2
3113	0.00	0.00	0.00	5
3114	0.00	0.00	0.00	2
3115	0.00	0.00	0.00	3
3116	0.00	0.00	0.00	4
3117	0.00	0.00	0.00	5
3118	1.00	0.25	0.40	4
3119	0.00	0.00	0.00	1
3120	0.00	0.00	0.00	4
3121	1.00	0.67	0.80	3
3122	0.00	0.00	0.00	2
3123	0.00	0.00	0.00	5
3124	0.00	0.00	0.00	2
3125	0.00	0.00	0.00	4
3126	0.00	0.00	0.00	1
3127	0.00	0.00	0.00	6
3128	0.00	0.00	0.00	3
3129	0.00	0.00	0.00	0
3130	0.00	0.00	0.00	4
3131	0.00	0.00	0.00	6
3132	0.00	0.00	0.00	4
3133	0.00	0.00	0.00	2
3134	0.00	0.00	0.00	2
3135	0.00	0.00	0.00	2
3136	0.00	0.00	0.00	6
3137	0.00	0.00	0.00	5
3138	1.00	0.17	0.29	6
3139	0.00	0.00	0.00	2
3140	0.00	0.00	0.00	1
3141	0.00	0.00	0.00	2
3142	0.00	0.00	0.00	3
3143	0.00	0.00	0.00	2
3144	0.00	0.00	0.00	3
3145	0.00	0.00	0.00	7
3146	0.00	0.00	0.00	4
3147	0.00	0.00	0.00	4
3148	0.00	0.00	0.00	3
3149	0.00	0.00	0.00	8
3150	0.00	0.00	0.00	2
3151	0.00	0.00	0.00	2
3152	0.00	0.00	0.00	8
3153	0.00	0.00	0.00	6
3154	0.00	0.00	0.00	4
3155	0.00	0.00	0.00	4
3156	1.00	0.25	0.40	4
3157	0.00	0.00	0.00	2
3158	0.00	0.00	0.00	2
3159	0.00	0.00	0.00	8
3160	0.00	0.00	0.00	3
3161	0.00	0.00	0.00	3
3162	1.00	0.20	0.33	5
3163	0.00	0.00	0.00	3
3164	1.00	0.50	0.67	4

3165	0.00	0.00	0.00	4
3166	0.00	0.00	0.00	0
3167	0.00	0.00	0.00	2
3168	0.00	0.00	0.00	4
3169	0.00	0.00	0.00	3
3170	0.00	0.00	0.00	3
3171	0.00	0.00	0.00	4
3172	0.00	0.00	0.00	3
3173	0.00	0.00	0.00	4
3174	0.00	0.00	0.00	4
3175	0.00	0.00	0.00	7
3176	0.00	0.00	0.00	5
3177	0.00	0.00	0.00	5
3178	0.00	0.00	0.00	2
3179	0.00	0.00	0.00	2
3180	0.00	0.00	0.00	5
3181	0.00	0.00	0.00	8
3182	0.00	0.00	0.00	6
3183	0.00	0.00	0.00	4
3184	0.00	0.00	0.00	3
3185	0.00	0.00	0.00	4
3186	0.00	0.00	0.00	2
3187	0.00	0.00	0.00	0
3188	0.00	0.00	0.00	2
3189	0.00	0.00	0.00	5
3190	0.00	0.00	0.00	4
3191	0.00	0.00	0.00	4
3192	0.00	0.00	0.00	4
3193	0.00	0.00	0.00	2
3194	0.00	0.00	0.00	6
3195	0.00	0.00	0.00	4
3196	0.00	0.00	0.00	1
3197	0.00	0.00	0.00	3
3198	0.00	0.00	0.00	5
3199	1.00	0.17	0.29	6
3200	0.00	0.00	0.00	1
3201	0.00	0.00	0.00	2
3202	0.00	0.00	0.00	2
3203	0.00	0.00	0.00	5
3204	0.00	0.00	0.00	5
3205	0.00	0.00	0.00	2
3206	0.00	0.00	0.00	2
3207	0.00	0.00	0.00	5
3208	0.00	0.00	0.00	2
3209	0.00	0.00	0.00	7
3210	0.00	0.00	0.00	7
3211	0.00	0.00	0.00	3
3212	0.00	0.00	0.00	1
3213	0.00	0.00	0.00	4
3214	0.00	0.00	0.00	7
3215	0.00	0.00	0.00	7
3216	0.00	0.00	0.00	3
3217	0.00	0.00	0.00	2
3218	0.00	0.00	0.00	3
3219	0.00	0.00	0.00	7
3220	0.00	0.00	0.00	3
3221	0.00	0.00	0.00	3
3222	0.00	0.00	0.00	4
3223	0.00	0.00	0.00	8
3224	0.00	0.00	0.00	6
3225	0.00	0.00	0.00	6

3226	0.00	0.00	0.00	1
3227	0.00	0.00	0.00	7
3228	0.00	0.00	0.00	0
3229	0.00	0.00	0.00	1
3230	0.00	0.00	0.00	2
3231	0.00	0.00	0.00	3
3232	0.00	0.00	0.00	5
3233	0.00	0.00	0.00	4
3234	0.00	0.00	0.00	2
3235	0.00	0.00	0.00	2
3236	0.00	0.00	0.00	1
3237	0.00	0.00	0.00	2
3238	0.00	0.00	0.00	2
3239	0.00	0.00	0.00	3
3240	0.00	0.00	0.00	4
3241	0.00	0.00	0.00	0
3242	0.00	0.00	0.00	2
3243	0.00	0.00	0.00	5
3244	0.00	0.00	0.00	3
3245	0.00	0.00	0.00	4
3246	0.00	0.00	0.00	4
3247	0.00	0.00	0.00	4
3248	0.00	0.00	0.00	3
3249	1.00	0.25	0.40	4
3250	0.00	0.00	0.00	2
3251	0.00	0.00	0.00	3
3252	0.00	0.00	0.00	2
3253	0.00	0.00	0.00	3
3254	0.00	0.00	0.00	1
3255	0.00	0.00	0.00	5
3256	0.00	0.00	0.00	3
3257	0.00	0.00	0.00	2
3258	0.00	0.00	0.00	5
3259	0.00	0.00	0.00	6
3260	0.00	0.00	0.00	2
3261	0.20	0.25	0.22	4
3262	0.00	0.00	0.00	5
3263	0.00	0.00	0.00	2
3264	0.00	0.00	0.00	5
3265	0.00	0.00	0.00	2
3266	0.00	0.00	0.00	1
3267	0.00	0.00	0.00	6
3268	0.00	0.00	0.00	2
3269	0.00	0.00	0.00	4
3270	0.00	0.00	0.00	2
3271	0.00	0.00	0.00	4
3272	0.00	0.00	0.00	4
3273	0.00	0.00	0.00	4
3274	0.00	0.00	0.00	3
3275	0.00	0.00	0.00	3
3276	0.00	0.00	0.00	2
3277	0.00	0.00	0.00	3
3278	0.00	0.00	0.00	3
3279	0.00	0.00	0.00	5
3280	0.00	0.00	0.00	4
3281	0.00	0.00	0.00	2
3282	0.00	0.00	0.00	5
3283	0.00	0.00	0.00	4
3284	0.00	0.00	0.00	4
3285	0.00	0.00	0.00	4
3286	1.00	0.20	0.33	5

3287	0.00	0.00	0.00	4
3288	0.00	0.00	0.00	2
3289	0.00	0.00	0.00	7
3290	0.00	0.00	0.00	6
3291	1.00	0.33	0.50	6
3292	0.00	0.00	0.00	4
3293	0.00	0.00	0.00	0
3294	0.00	0.00	0.00	4
3295	0.00	0.00	0.00	5
3296	0.00	0.00	0.00	3
3297	0.00	0.00	0.00	3
3298	0.00	0.00	0.00	4
3299	0.00	0.00	0.00	0
3300	0.00	0.00	0.00	4
3301	0.00	0.00	0.00	0
3302	0.00	0.00	0.00	6
3303	0.00	0.00	0.00	3
3304	0.00	0.00	0.00	3
3305	0.00	0.00	0.00	8
3306	0.00	0.00	0.00	1
3307	0.00	0.00	0.00	4
3308	0.00	0.00	0.00	4
3309	0.00	0.00	0.00	3
3310	0.00	0.00	0.00	3
3311	0.00	0.00	0.00	3
3312	0.00	0.00	0.00	5
3313	0.00	0.00	0.00	3
3314	0.00	0.00	0.00	3
3315	0.00	0.00	0.00	1
3316	0.00	0.00	0.00	1
3317	0.00	0.00	0.00	4
3318	0.00	0.00	0.00	3
3319	0.00	0.00	0.00	5
3320	1.00	0.25	0.40	4
3321	0.00	0.00	0.00	4
3322	0.00	0.00	0.00	4
3323	0.00	0.00	0.00	5
3324	0.00	0.00	0.00	8
3325	0.00	0.00	0.00	4
3326	0.00	0.00	0.00	6
3327	0.00	0.00	0.00	2
3328	0.00	0.00	0.00	3
3329	0.00	0.00	0.00	3
3330	0.00	0.00	0.00	4
3331	0.00	0.00	0.00	2
3332	0.00	0.00	0.00	4
3333	0.00	0.00	0.00	3
3334	0.00	0.00	0.00	2
3335	0.00	0.00	0.00	4
3336	0.50	0.14	0.22	7
3337	0.00	0.00	0.00	5
3338	0.00	0.00	0.00	4
3339	0.00	0.00	0.00	3
3340	0.00	0.00	0.00	3
3341	0.00	0.00	0.00	1
3342	0.00	0.00	0.00	5
3343	0.00	0.00	0.00	0
3344	0.00	0.00	0.00	6
3345	0.00	0.00	0.00	7
3346	0.00	0.00	0.00	3
3347	0.00	0.00	0.00	2

3348	0.00	0.00	0.00	3
3349	0.00	0.00	0.00	3
3350	0.00	0.00	0.00	5
3351	0.00	0.00	0.00	2
3352	0.00	0.00	0.00	2
3353	0.00	0.00	0.00	2
3354	0.00	0.00	0.00	3
3355	0.00	0.00	0.00	2
3356	0.00	0.00	0.00	5
3357	0.00	0.00	0.00	6
3358	0.00	0.00	0.00	6
3359	0.00	0.00	0.00	6
3360	0.00	0.00	0.00	2
3361	0.00	0.00	0.00	4
3362	0.00	0.00	0.00	2
3363	0.00	0.00	0.00	3
3364	0.00	0.00	0.00	6
3365	0.00	0.00	0.00	1
3366	0.00	0.00	0.00	5
3367	0.00	0.00	0.00	1
3368	0.00	0.00	0.00	2
3369	0.00	0.00	0.00	3
3370	0.00	0.00	0.00	3
3371	0.00	0.00	0.00	5
3372	0.00	0.00	0.00	2
3373	0.00	0.00	0.00	4
3374	0.00	0.00	0.00	2
3375	0.00	0.00	0.00	6
3376	0.00	0.00	0.00	2
3377	0.00	0.00	0.00	3
3378	0.00	0.00	0.00	5
3379	0.00	0.00	0.00	6
3380	0.00	0.00	0.00	3
3381	0.00	0.00	0.00	1
3382	0.00	0.00	0.00	2
3383	0.00	0.00	0.00	6
3384	0.00	0.00	0.00	7
3385	0.00	0.00	0.00	1
3386	0.00	0.00	0.00	5
3387	0.00	0.00	0.00	5
3388	0.00	0.00	0.00	4
3389	0.00	0.00	0.00	2
3390	0.00	0.00	0.00	2
3391	0.00	0.00	0.00	5
3392	0.00	0.00	0.00	4
3393	0.00	0.00	0.00	0
3394	0.00	0.00	0.00	2
3395	0.00	0.00	0.00	3
3396	0.00	0.00	0.00	3
3397	0.00	0.00	0.00	1
3398	0.00	0.00	0.00	4
3399	0.00	0.00	0.00	2
3400	0.00	0.00	0.00	4
3401	0.00	0.00	0.00	8
3402	0.00	0.00	0.00	3
3403	0.00	0.00	0.00	3
3404	0.00	0.00	0.00	4
3405	0.00	0.00	0.00	3
3406	0.00	0.00	0.00	8
3407	0.00	0.00	0.00	3
3408	0.00	0.00	0.00	7

3409	0.00	0.00	0.00	5
3410	0.00	0.00	0.00	4
3411	0.00	0.00	0.00	2
3412	0.00	0.00	0.00	5
3413	0.00	0.00	0.00	4
3414	0.00	0.00	0.00	4
3415	0.00	0.00	0.00	5
3416	0.00	0.00	0.00	3
3417	0.00	0.00	0.00	2
3418	0.00	0.00	0.00	3
3419	0.00	0.00	0.00	3
3420	0.00	0.00	0.00	4
3421	0.00	0.00	0.00	5
3422	0.00	0.00	0.00	6
3423	0.00	0.00	0.00	2
3424	0.00	0.00	0.00	4
3425	0.00	0.00	0.00	4
3426	0.00	0.00	0.00	3
3427	0.00	0.00	0.00	2
3428	0.00	0.00	0.00	2
3429	0.00	0.00	0.00	3
3430	0.00	0.00	0.00	3
3431	0.00	0.00	0.00	2
3432	0.00	0.00	0.00	3
3433	0.00	0.00	0.00	8
3434	0.00	0.00	0.00	5
3435	0.00	0.00	0.00	3
3436	0.00	0.00	0.00	4
3437	0.00	0.00	0.00	7
3438	0.00	0.00	0.00	2
3439	0.00	0.00	0.00	3
3440	0.00	0.00	0.00	4
3441	0.00	0.00	0.00	8
3442	0.00	0.00	0.00	3
3443	0.00	0.00	0.00	1
3444	0.00	0.00	0.00	4
3445	0.00	0.00	0.00	1
3446	0.00	0.00	0.00	0
3447	0.00	0.00	0.00	2
3448	0.00	0.00	0.00	1
3449	0.00	0.00	0.00	4
3450	0.00	0.00	0.00	2
3451	0.00	0.00	0.00	3
3452	0.00	0.00	0.00	3
3453	0.00	0.00	0.00	5
3454	0.00	0.00	0.00	2
3455	0.00	0.00	0.00	7
3456	0.00	0.00	0.00	4
3457	0.00	0.00	0.00	4
3458	0.00	0.00	0.00	4
3459	0.00	0.00	0.00	4
3460	0.00	0.00	0.00	1
3461	0.00	0.00	0.00	3
3462	0.00	0.00	0.00	5
3463	0.00	0.00	0.00	3
3464	0.00	0.00	0.00	2
3465	0.00	0.00	0.00	2
3466	0.00	0.00	0.00	2
3467	1.00	0.33	0.50	3
3468	0.00	0.00	0.00	2
3469	0.00	0.00	0.00	3

3470	0.00	0.00	0.00	4
3471	0.00	0.00	0.00	1
3472	0.00	0.00	0.00	4
3473	0.00	0.00	0.00	2
3474	0.00	0.00	0.00	6
3475	0.00	0.00	0.00	4
3476	0.00	0.00	0.00	3
3477	0.00	0.00	0.00	4
3478	0.00	0.00	0.00	4
3479	0.00	0.00	0.00	2
3480	0.00	0.00	0.00	6
3481	0.00	0.00	0.00	4
3482	0.00	0.00	0.00	5
3483	0.00	0.00	0.00	4
3484	0.00	0.00	0.00	1
3485	0.00	0.00	0.00	3
3486	0.00	0.00	0.00	2
3487	0.00	0.00	0.00	0
3488	0.00	0.00	0.00	1
3489	0.00	0.00	0.00	0
3490	0.00	0.00	0.00	4
3491	0.00	0.00	0.00	3
3492	0.00	0.00	0.00	2
3493	0.00	0.00	0.00	3
3494	0.00	0.00	0.00	6
3495	0.00	0.00	0.00	3
3496	0.00	0.00	0.00	1
3497	0.00	0.00	0.00	3
3498	0.00	0.00	0.00	3
3499	0.00	0.00	0.00	5
3500	0.00	0.00	0.00	0
3501	0.00	0.00	0.00	3
3502	0.00	0.00	0.00	4
3503	0.00	0.00	0.00	3
3504	0.00	0.00	0.00	6
3505	0.00	0.00	0.00	1
3506	0.00	0.00	0.00	2
3507	0.00	0.00	0.00	5
3508	0.00	0.00	0.00	4
3509	0.00	0.00	0.00	4
3510	0.00	0.00	0.00	4
3511	0.00	0.00	0.00	5
3512	0.00	0.00	0.00	3
3513	0.00	0.00	0.00	3
3514	0.00	0.00	0.00	1
3515	0.00	0.00	0.00	5
3516	0.00	0.00	0.00	5
3517	0.00	0.00	0.00	3
3518	0.00	0.00	0.00	3
3519	0.00	0.00	0.00	4
3520	0.00	0.00	0.00	5
3521	0.00	0.00	0.00	7
3522	0.00	0.00	0.00	2
3523	0.00	0.00	0.00	3
3524	0.00	0.00	0.00	2
3525	0.00	0.00	0.00	7
3526	0.00	0.00	0.00	2
3527	0.00	0.00	0.00	1
3528	0.00	0.00	0.00	4
3529	0.00	0.00	0.00	8
3530	0.00	0.00	0.00	2

3531	0.00	0.00	0.00	4
3532	0.00	0.00	0.00	3
3533	0.00	0.00	0.00	6
3534	0.00	0.00	0.00	0
3535	0.00	0.00	0.00	3
3536	0.00	0.00	0.00	4
3537	0.00	0.00	0.00	2
3538	0.00	0.00	0.00	2
3539	0.00	0.00	0.00	0
3540	0.00	0.00	0.00	3
3541	0.00	0.00	0.00	2
3542	0.00	0.00	0.00	5
3543	0.00	0.00	0.00	5
3544	0.00	0.00	0.00	7
3545	0.00	0.00	0.00	8
3546	0.00	0.00	0.00	3
3547	0.00	0.00	0.00	3
3548	0.00	0.00	0.00	4
3549	0.00	0.00	0.00	3
3550	0.00	0.00	0.00	3
3551	0.00	0.00	0.00	5
3552	0.00	0.00	0.00	2
3553	0.00	0.00	0.00	4
3554	0.00	0.00	0.00	2
3555	0.00	0.00	0.00	1
3556	1.00	0.33	0.50	3
3557	0.00	0.00	0.00	5
3558	0.00	0.00	0.00	6
3559	0.00	0.00	0.00	2
3560	0.00	0.00	0.00	4
3561	1.00	0.50	0.67	2
3562	0.00	0.00	0.00	5
3563	0.00	0.00	0.00	5
3564	0.00	0.00	0.00	2
3565	0.00	0.00	0.00	5
3566	0.00	0.00	0.00	5
3567	0.00	0.00	0.00	3
3568	0.00	0.00	0.00	5
3569	0.00	0.00	0.00	3
3570	0.00	0.00	0.00	5
3571	0.00	0.00	0.00	4
3572	0.00	0.00	0.00	6
3573	0.00	0.00	0.00	5
3574	0.00	0.00	0.00	3
3575	0.00	0.00	0.00	3
3576	0.00	0.00	0.00	6
3577	0.00	0.00	0.00	3
3578	0.00	0.00	0.00	4
3579	0.00	0.00	0.00	4
3580	0.00	0.00	0.00	5
3581	0.00	0.00	0.00	3
3582	0.00	0.00	0.00	3
3583	0.00	0.00	0.00	3
3584	0.00	0.00	0.00	3
3585	0.00	0.00	0.00	2
3586	0.00	0.00	0.00	4
3587	0.00	0.00	0.00	3
3588	0.00	0.00	0.00	2
3589	0.00	0.00	0.00	1
3590	0.00	0.00	0.00	2
3591	0.00	0.00	0.00	2

3592	0.00	0.00	0.00	3
3593	1.00	0.50	0.67	4
3594	0.00	0.00	0.00	2
3595	0.00	0.00	0.00	1
3596	0.00	0.00	0.00	0
3597	0.00	0.00	0.00	2
3598	0.00	0.00	0.00	3
3599	0.00	0.00	0.00	2
3600	0.00	0.00	0.00	3
3601	0.00	0.00	0.00	2
3602	0.00	0.00	0.00	1
3603	0.00	0.00	0.00	3
3604	0.00	0.00	0.00	1
3605	0.00	0.00	0.00	7
3606	0.00	0.00	0.00	6
3607	0.00	0.00	0.00	0
3608	0.00	0.00	0.00	1
3609	0.00	0.00	0.00	2
3610	0.00	0.00	0.00	6
3611	0.00	0.00	0.00	4
3612	0.00	0.00	0.00	4
3613	0.00	0.00	0.00	6
3614	0.00	0.00	0.00	7
3615	0.00	0.00	0.00	3
3616	0.00	0.00	0.00	2
3617	0.00	0.00	0.00	2
3618	0.00	0.00	0.00	3
3619	0.00	0.00	0.00	4
3620	0.00	0.00	0.00	5
3621	0.00	0.00	0.00	4
3622	0.00	0.00	0.00	2
3623	0.00	0.00	0.00	3
3624	0.00	0.00	0.00	1
3625	0.00	0.00	0.00	3
3626	0.00	0.00	0.00	4
3627	0.00	0.00	0.00	1
3628	0.00	0.00	0.00	4
3629	0.00	0.00	0.00	3
3630	0.00	0.00	0.00	2
3631	0.00	0.00	0.00	5
3632	0.00	0.00	0.00	1
3633	0.00	0.00	0.00	4
3634	0.00	0.00	0.00	3
3635	0.00	0.00	0.00	4
3636	0.00	0.00	0.00	1
3637	0.00	0.00	0.00	8
3638	0.00	0.00	0.00	1
3639	0.00	0.00	0.00	3
3640	0.00	0.00	0.00	3
3641	0.00	0.00	0.00	2
3642	0.00	0.00	0.00	6
3643	0.00	0.00	0.00	2
3644	0.00	0.00	0.00	3
3645	0.00	0.00	0.00	2
3646	0.00	0.00	0.00	1
3647	0.00	0.00	0.00	3
3648	0.00	0.00	0.00	0
3649	0.00	0.00	0.00	6
3650	0.00	0.00	0.00	3
3651	0.00	0.00	0.00	3
3652	0.00	0.00	0.00	6

3653	0.00	0.00	0.00	3
3654	0.00	0.00	0.00	1
3655	0.00	0.00	0.00	3
3656	0.00	0.00	0.00	4
3657	0.00	0.00	0.00	2
3658	0.00	0.00	0.00	3
3659	0.00	0.00	0.00	2
3660	0.00	0.00	0.00	5
3661	0.00	0.00	0.00	5
3662	0.00	0.00	0.00	2
3663	0.00	0.00	0.00	4
3664	0.00	0.00	0.00	5
3665	0.00	0.00	0.00	6
3666	0.00	0.00	0.00	4
3667	0.00	0.00	0.00	5
3668	0.00	0.00	0.00	1
3669	0.00	0.00	0.00	3
3670	0.00	0.00	0.00	3
3671	0.00	0.00	0.00	3
3672	0.00	0.00	0.00	4
3673	0.00	0.00	0.00	5
3674	0.00	0.00	0.00	4
3675	0.00	0.00	0.00	2
3676	0.00	0.00	0.00	7
3677	0.00	0.00	0.00	4
3678	0.00	0.00	0.00	3
3679	0.00	0.00	0.00	5
3680	0.00	0.00	0.00	3
3681	0.00	0.00	0.00	5
3682	0.00	0.00	0.00	1
3683	0.00	0.00	0.00	7
3684	0.00	0.00	0.00	3
3685	0.00	0.00	0.00	6
3686	0.00	0.00	0.00	4
3687	0.00	0.00	0.00	1
3688	0.00	0.00	0.00	4
3689	0.00	0.00	0.00	3
3690	0.00	0.00	0.00	4
3691	0.00	0.00	0.00	3
3692	0.00	0.00	0.00	7
3693	0.00	0.00	0.00	6
3694	0.00	0.00	0.00	2
3695	0.00	0.00	0.00	3
3696	0.00	0.00	0.00	2
3697	0.00	0.00	0.00	1
3698	0.00	0.00	0.00	4
3699	0.00	0.00	0.00	4
3700	0.00	0.00	0.00	4
3701	0.00	0.00	0.00	2
3702	0.00	0.00	0.00	4
3703	0.00	0.00	0.00	2
3704	0.00	0.00	0.00	2
3705	0.00	0.00	0.00	4
3706	0.00	0.00	0.00	5
3707	0.00	0.00	0.00	4
3708	0.00	0.00	0.00	3
3709	0.00	0.00	0.00	3
3710	0.00	0.00	0.00	2
3711	0.00	0.00	0.00	4
3712	0.00	0.00	0.00	2
3713	0.00	0.00	0.00	8

3714	0.00	0.00	0.00	2
3715	0.00	0.00	0.00	3
3716	0.00	0.00	0.00	2
3717	0.00	0.00	0.00	1
3718	0.00	0.00	0.00	2
3719	0.00	0.00	0.00	5
3720	0.00	0.00	0.00	3
3721	0.00	0.00	0.00	1
3722	0.00	0.00	0.00	4
3723	0.00	0.00	0.00	3
3724	0.00	0.00	0.00	2
3725	0.00	0.00	0.00	1
3726	0.00	0.00	0.00	1
3727	0.00	0.00	0.00	4
3728	0.00	0.00	0.00	3
3729	0.00	0.00	0.00	3
3730	0.00	0.00	0.00	1
3731	0.00	0.00	0.00	1
3732	0.00	0.00	0.00	1
3733	0.00	0.00	0.00	3
3734	0.00	0.00	0.00	1
3735	0.00	0.00	0.00	2
3736	0.00	0.00	0.00	3
3737	0.00	0.00	0.00	3
3738	0.00	0.00	0.00	2
3739	0.00	0.00	0.00	2
3740	0.00	0.00	0.00	1
3741	0.00	0.00	0.00	1
3742	0.00	0.00	0.00	3
3743	0.00	0.00	0.00	2
3744	0.00	0.00	0.00	2
3745	0.00	0.00	0.00	0
3746	0.00	0.00	0.00	5
3747	0.00	0.00	0.00	4
3748	0.00	0.00	0.00	2
3749	0.00	0.00	0.00	6
3750	0.00	0.00	0.00	1
3751	0.00	0.00	0.00	2
3752	0.00	0.00	0.00	3
3753	0.00	0.00	0.00	3
3754	0.00	0.00	0.00	1
3755	0.00	0.00	0.00	2
3756	0.00	0.00	0.00	3
3757	0.00	0.00	0.00	2
3758	0.00	0.00	0.00	3
3759	0.00	0.00	0.00	2
3760	0.00	0.00	0.00	3
3761	0.00	0.00	0.00	1
3762	0.00	0.00	0.00	3
3763	0.00	0.00	0.00	4
3764	0.00	0.00	0.00	1
3765	0.00	0.00	0.00	6
3766	0.00	0.00	0.00	4
3767	0.00	0.00	0.00	3
3768	0.00	0.00	0.00	2
3769	0.00	0.00	0.00	3
3770	0.00	0.00	0.00	2
3771	0.00	0.00	0.00	2
3772	0.00	0.00	0.00	2
3773	0.00	0.00	0.00	3
3774	0.00	0.00	0.00	1

3775	0.00	0.00	0.00	3
3776	0.00	0.00	0.00	3
3777	0.00	0.00	0.00	1
3778	0.00	0.00	0.00	1
3779	0.00	0.00	0.00	9
3780	0.00	0.00	0.00	4
3781	0.00	0.00	0.00	4
3782	0.00	0.00	0.00	2
3783	0.00	0.00	0.00	5
3784	0.00	0.00	0.00	2
3785	0.00	0.00	0.00	1
3786	0.00	0.00	0.00	5
3787	0.00	0.00	0.00	4
3788	0.00	0.00	0.00	0
3789	0.00	0.00	0.00	1
3790	0.00	0.00	0.00	3
3791	0.00	0.00	0.00	4
3792	1.00	0.11	0.20	9
3793	0.00	0.00	0.00	5
3794	0.00	0.00	0.00	6
3795	0.00	0.00	0.00	2
3796	0.00	0.00	0.00	2
3797	0.00	0.00	0.00	2
3798	0.00	0.00	0.00	3
3799	0.00	0.00	0.00	6
3800	0.00	0.00	0.00	5
3801	0.00	0.00	0.00	4
3802	0.00	0.00	0.00	5
3803	0.00	0.00	0.00	4
3804	1.00	0.33	0.50	3
3805	0.00	0.00	0.00	3
3806	0.00	0.00	0.00	2
3807	0.00	0.00	0.00	2
3808	0.00	0.00	0.00	1
3809	0.00	0.00	0.00	3
3810	0.00	0.00	0.00	5
3811	0.00	0.00	0.00	4
3812	0.00	0.00	0.00	3
3813	0.00	0.00	0.00	2
3814	0.00	0.00	0.00	5
3815	0.00	0.00	0.00	1
3816	0.00	0.00	0.00	3
3817	0.00	0.00	0.00	3
3818	0.00	0.00	0.00	3
3819	0.00	0.00	0.00	6
3820	0.00	0.00	0.00	1
3821	0.00	0.00	0.00	1
3822	0.00	0.00	0.00	4
3823	0.00	0.00	0.00	5
3824	0.00	0.00	0.00	3
3825	0.00	0.00	0.00	1
3826	0.00	0.00	0.00	2
3827	0.00	0.00	0.00	5
3828	0.00	0.00	0.00	2
3829	0.00	0.00	0.00	5
3830	0.00	0.00	0.00	3
3831	0.00	0.00	0.00	5
3832	0.00	0.00	0.00	3
3833	0.00	0.00	0.00	0
3834	0.00	0.00	0.00	1
3835	0.00	0.00	0.00	1

3836	0.00	0.00	0.00	4
3837	0.00	0.00	0.00	2
3838	0.00	0.00	0.00	0
3839	0.00	0.00	0.00	1
3840	0.00	0.00	0.00	2
3841	0.00	0.00	0.00	4
3842	0.00	0.00	0.00	2
3843	0.00	0.00	0.00	1
3844	0.00	0.00	0.00	1
3845	0.00	0.00	0.00	3
3846	0.00	0.00	0.00	1
3847	0.00	0.00	0.00	4
3848	0.00	0.00	0.00	4
3849	0.00	0.00	0.00	3
3850	0.00	0.00	0.00	4
3851	0.00	0.00	0.00	1
3852	0.00	0.00	0.00	4
3853	0.00	0.00	0.00	3
3854	0.00	0.00	0.00	2
3855	0.00	0.00	0.00	4
3856	0.00	0.00	0.00	1
3857	0.00	0.00	0.00	0
3858	0.00	0.00	0.00	0
3859	0.00	0.00	0.00	1
3860	0.00	0.00	0.00	3
3861	0.00	0.00	0.00	5
3862	0.00	0.00	0.00	0
3863	0.00	0.00	0.00	4
3864	0.00	0.00	0.00	3
3865	0.00	0.00	0.00	6
3866	0.00	0.00	0.00	2
3867	0.00	0.00	0.00	2
3868	0.00	0.00	0.00	4
3869	0.00	0.00	0.00	4
3870	0.00	0.00	0.00	2
3871	0.00	0.00	0.00	1
3872	0.00	0.00	0.00	5
3873	0.00	0.00	0.00	3
3874	0.00	0.00	0.00	2
3875	0.00	0.00	0.00	4
3876	0.00	0.00	0.00	3
3877	0.00	0.00	0.00	2
3878	0.00	0.00	0.00	3
3879	0.00	0.00	0.00	4
3880	0.00	0.00	0.00	3
3881	0.00	0.00	0.00	2
3882	0.00	0.00	0.00	2
3883	0.00	0.00	0.00	2
3884	0.00	0.00	0.00	2
3885	0.00	0.00	0.00	2
3886	0.00	0.00	0.00	3
3887	0.00	0.00	0.00	2
3888	0.00	0.00	0.00	6
3889	0.00	0.00	0.00	2
3890	0.00	0.00	0.00	2
3891	0.00	0.00	0.00	6
3892	0.00	0.00	0.00	3
3893	0.00	0.00	0.00	1
3894	0.00	0.00	0.00	2
3895	0.00	0.00	0.00	5
3896	0.00	0.00	0.00	2

3897	0.00	0.00	0.00	2
3898	0.00	0.00	0.00	1
3899	0.00	0.00	0.00	1
3900	0.00	0.00	0.00	1
3901	0.00	0.00	0.00	3
3902	0.00	0.00	0.00	2
3903	0.00	0.00	0.00	2
3904	0.00	0.00	0.00	1
3905	0.00	0.00	0.00	8
3906	0.00	0.00	0.00	1
3907	0.00	0.00	0.00	7
3908	0.00	0.00	0.00	3
3909	0.00	0.00	0.00	3
3910	0.00	0.00	0.00	0
3911	0.00	0.00	0.00	1
3912	0.00	0.00	0.00	4
3913	0.00	0.00	0.00	2
3914	0.00	0.00	0.00	2
3915	0.00	0.00	0.00	5
3916	0.00	0.00	0.00	4
3917	0.00	0.00	0.00	4
3918	0.00	0.00	0.00	3
3919	0.00	0.00	0.00	2
3920	0.00	0.00	0.00	0
3921	0.00	0.00	0.00	3
3922	0.00	0.00	0.00	0
3923	0.00	0.00	0.00	3
3924	0.00	0.00	0.00	2
3925	0.00	0.00	0.00	2
3926	0.00	0.00	0.00	2
3927	0.00	0.00	0.00	4
3928	0.00	0.00	0.00	4
3929	0.00	0.00	0.00	4
3930	0.00	0.00	0.00	2
3931	0.00	0.00	0.00	2
3932	0.00	0.00	0.00	0
3933	0.00	0.00	0.00	3
3934	0.00	0.00	0.00	1
3935	0.00	0.00	0.00	4
3936	0.00	0.00	0.00	4
3937	0.00	0.00	0.00	1
3938	0.00	0.00	0.00	3
3939	0.00	0.00	0.00	2
3940	0.00	0.00	0.00	1
3941	0.00	0.00	0.00	2
3942	0.00	0.00	0.00	4
3943	0.00	0.00	0.00	1
3944	0.00	0.00	0.00	2
3945	0.00	0.00	0.00	3
3946	0.00	0.00	0.00	3
3947	0.00	0.00	0.00	2
3948	0.00	0.00	0.00	2
3949	0.00	0.00	0.00	5
3950	0.00	0.00	0.00	3
3951	0.00	0.00	0.00	3
3952	0.00	0.00	0.00	3
3953	0.00	0.00	0.00	3
3954	0.00	0.00	0.00	3
3955	1.00	0.50	0.67	2
3956	0.00	0.00	0.00	2
3957	0.00	0.00	0.00	3

3958	0.00	0.00	0.00	5
3959	0.00	0.00	0.00	3
3960	0.00	0.00	0.00	0
3961	0.00	0.00	0.00	2
3962	0.00	0.00	0.00	3
3963	0.00	0.00	0.00	0
3964	0.00	0.00	0.00	1
3965	0.00	0.00	0.00	0
3966	0.00	0.00	0.00	2
3967	0.00	0.00	0.00	4
3968	0.00	0.00	0.00	2
3969	0.00	0.00	0.00	3
3970	0.00	0.00	0.00	4
3971	0.00	0.00	0.00	2
3972	0.00	0.00	0.00	2
3973	0.00	0.00	0.00	2
3974	0.00	0.00	0.00	6
3975	0.00	0.00	0.00	3
3976	0.00	0.00	0.00	1
3977	0.00	0.00	0.00	4
3978	0.00	0.00	0.00	1
3979	0.00	0.00	0.00	3
3980	0.00	0.00	0.00	1
3981	0.00	0.00	0.00	0
3982	0.00	0.00	0.00	2
3983	0.00	0.00	0.00	6
3984	0.00	0.00	0.00	1
3985	0.00	0.00	0.00	1
3986	0.00	0.00	0.00	6
3987	0.00	0.00	0.00	2
3988	0.00	0.00	0.00	3
3989	0.00	0.00	0.00	3
3990	0.00	0.00	0.00	3
3991	0.00	0.00	0.00	1
3992	0.00	0.00	0.00	3
3993	0.00	0.00	0.00	5
3994	0.00	0.00	0.00	3
3995	0.00	0.00	0.00	3
3996	0.00	0.00	0.00	3
3997	0.00	0.00	0.00	8
3998	0.00	0.00	0.00	1
3999	0.00	0.00	0.00	0
4000	0.00	0.00	0.00	2
4001	0.00	0.00	0.00	4
4002	0.00	0.00	0.00	3
4003	0.00	0.00	0.00	1
4004	0.00	0.00	0.00	2
4005	0.00	0.00	0.00	1
4006	0.00	0.00	0.00	0
4007	0.00	0.00	0.00	3
4008	0.00	0.00	0.00	3
4009	0.00	0.00	0.00	4
4010	0.00	0.00	0.00	1
4011	0.00	0.00	0.00	5
4012	0.00	0.00	0.00	3
4013	0.00	0.00	0.00	2
4014	0.00	0.00	0.00	1
4015	0.00	0.00	0.00	1
4016	0.00	0.00	0.00	4
4017	0.00	0.00	0.00	2
4018	0.00	0.00	0.00	3

4019	0.00	0.00	0.00	3
4020	0.00	0.00	0.00	2
4021	0.00	0.00	0.00	1
4022	0.00	0.00	0.00	3
4023	0.00	0.00	0.00	1
4024	0.00	0.00	0.00	1
4025	0.00	0.00	0.00	0
4026	0.00	0.00	0.00	2
4027	0.00	0.00	0.00	3
4028	0.00	0.00	0.00	1
4029	0.00	0.00	0.00	2
4030	0.00	0.00	0.00	1
4031	0.00	0.00	0.00	3
4032	0.00	0.00	0.00	3
4033	0.00	0.00	0.00	3
4034	0.00	0.00	0.00	4
4035	0.00	0.00	0.00	3
4036	0.00	0.00	0.00	4
4037	0.00	0.00	0.00	1
4038	0.00	0.00	0.00	2
4039	0.00	0.00	0.00	4
4040	0.00	0.00	0.00	3
4041	0.00	0.00	0.00	4
4042	1.00	0.50	0.67	2
4043	0.00	0.00	0.00	2
4044	0.00	0.00	0.00	5
4045	0.00	0.00	0.00	2
4046	0.00	0.00	0.00	3
4047	0.00	0.00	0.00	1
4048	0.00	0.00	0.00	1
4049	0.00	0.00	0.00	5
4050	0.00	0.00	0.00	3
4051	0.00	0.00	0.00	0
4052	0.00	0.00	0.00	1
4053	0.00	0.00	0.00	3
4054	0.00	0.00	0.00	2
4055	0.00	0.00	0.00	3
4056	0.00	0.00	0.00	6
4057	0.00	0.00	0.00	2
4058	0.00	0.00	0.00	1
4059	0.00	0.00	0.00	4
4060	0.00	0.00	0.00	2
4061	0.00	0.00	0.00	3
4062	0.00	0.00	0.00	1
4063	0.00	0.00	0.00	4
4064	0.00	0.00	0.00	2
4065	0.00	0.00	0.00	4
4066	0.00	0.00	0.00	4
4067	0.00	0.00	0.00	3
4068	0.00	0.00	0.00	2
4069	0.00	0.00	0.00	4
4070	0.00	0.00	0.00	2
4071	0.00	0.00	0.00	2
4072	0.00	0.00	0.00	2
4073	1.00	0.43	0.60	7
4074	0.00	0.00	0.00	2
4075	0.00	0.00	0.00	4
4076	0.00	0.00	0.00	3
4077	0.00	0.00	0.00	1
4078	0.00	0.00	0.00	5
4079	0.00	0.00	0.00	1

4080	0.00	0.00	0.00	5
4081	0.00	0.00	0.00	3
4082	0.00	0.00	0.00	4
4083	0.00	0.00	0.00	2
4084	0.00	0.00	0.00	2
4085	0.00	0.00	0.00	3
4086	0.00	0.00	0.00	3
4087	0.00	0.00	0.00	4
4088	0.00	0.00	0.00	2
4089	0.00	0.00	0.00	1
4090	0.00	0.00	0.00	4
4091	0.00	0.00	0.00	2
4092	0.00	0.00	0.00	6
4093	0.00	0.00	0.00	2
4094	0.00	0.00	0.00	1
4095	0.00	0.00	0.00	2
4096	0.00	0.00	0.00	1
4097	0.00	0.00	0.00	3
4098	0.00	0.00	0.00	2
4099	0.00	0.00	0.00	4
4100	0.00	0.00	0.00	1
4101	0.00	0.00	0.00	3
4102	0.00	0.00	0.00	3
4103	0.00	0.00	0.00	6
4104	0.00	0.00	0.00	4
4105	0.00	0.00	0.00	2
4106	0.00	0.00	0.00	7
4107	0.00	0.00	0.00	2
4108	0.00	0.00	0.00	4
4109	0.00	0.00	0.00	2
4110	0.00	0.00	0.00	2
4111	0.00	0.00	0.00	3
4112	0.00	0.00	0.00	0
4113	0.00	0.00	0.00	0
4114	0.00	0.00	0.00	3
4115	0.00	0.00	0.00	2
4116	0.00	0.00	0.00	3
4117	0.00	0.00	0.00	2
4118	0.00	0.00	0.00	3
4119	0.00	0.00	0.00	2
4120	0.00	0.00	0.00	2
4121	0.00	0.00	0.00	3
4122	0.00	0.00	0.00	4
4123	0.00	0.00	0.00	1
4124	0.00	0.00	0.00	3
4125	0.00	0.00	0.00	5
4126	0.00	0.00	0.00	4
4127	0.00	0.00	0.00	4
4128	0.00	0.00	0.00	3
4129	0.00	0.00	0.00	4
4130	0.00	0.00	0.00	2
4131	0.00	0.00	0.00	5
4132	0.00	0.00	0.00	0
4133	0.00	0.00	0.00	1
4134	0.00	0.00	0.00	2
4135	0.00	0.00	0.00	6
4136	0.00	0.00	0.00	2
4137	0.00	0.00	0.00	2
4138	0.00	0.00	0.00	4
4139	0.00	0.00	0.00	4
4140	0.00	0.00	0.00	2

4141	0.00	0.00	0.00	2
4142	0.00	0.00	0.00	0
4143	0.00	0.00	0.00	4
4144	0.00	0.00	0.00	2
4145	0.00	0.00	0.00	6
4146	0.00	0.00	0.00	2
4147	0.00	0.00	0.00	5
4148	0.00	0.00	0.00	2
4149	0.00	0.00	0.00	2
4150	0.00	0.00	0.00	2
4151	0.00	0.00	0.00	8
4152	0.00	0.00	0.00	3
4153	0.00	0.00	0.00	1
4154	0.00	0.00	0.00	1
4155	0.00	0.00	0.00	4
4156	0.00	0.00	0.00	3
4157	0.00	0.00	0.00	0
4158	0.00	0.00	0.00	3
4159	0.00	0.00	0.00	3
4160	0.00	0.00	0.00	1
4161	0.00	0.00	0.00	3
4162	0.00	0.00	0.00	2
4163	0.00	0.00	0.00	3
4164	0.00	0.00	0.00	2
4165	0.00	0.00	0.00	3
4166	0.00	0.00	0.00	3
4167	0.00	0.00	0.00	0
4168	0.00	0.00	0.00	3
4169	0.00	0.00	0.00	4
4170	0.00	0.00	0.00	1
4171	0.00	0.00	0.00	2
4172	0.00	0.00	0.00	1
4173	0.00	0.00	0.00	7
4174	0.00	0.00	0.00	2
4175	0.00	0.00	0.00	3
4176	0.00	0.00	0.00	2
4177	0.00	0.00	0.00	3
4178	0.00	0.00	0.00	1
4179	0.00	0.00	0.00	2
4180	0.00	0.00	0.00	1
4181	0.00	0.00	0.00	4
4182	0.00	0.00	0.00	1
4183	0.00	0.00	0.00	3
4184	0.00	0.00	0.00	0
4185	0.00	0.00	0.00	3
4186	0.00	0.00	0.00	3
4187	0.00	0.00	0.00	3
4188	0.00	0.00	0.00	6
4189	0.00	0.00	0.00	1
4190	0.00	0.00	0.00	2
4191	0.00	0.00	0.00	3
4192	0.00	0.00	0.00	1
4193	0.00	0.00	0.00	5
4194	0.00	0.00	0.00	4
4195	0.00	0.00	0.00	2
4196	0.00	0.00	0.00	2
4197	0.00	0.00	0.00	1
4198	0.00	0.00	0.00	6
4199	0.00	0.00	0.00	1
4200	0.00	0.00	0.00	4
4201	0.00	0.00	0.00	3

4202	0.00	0.00	0.00	1
4203	0.00	0.00	0.00	3
4204	0.00	0.00	0.00	1
4205	0.00	0.00	0.00	2
4206	0.00	0.00	0.00	8
4207	0.00	0.00	0.00	3
4208	0.00	0.00	0.00	4
4209	0.00	0.00	0.00	2
4210	0.00	0.00	0.00	4
4211	0.00	0.00	0.00	1
4212	0.00	0.00	0.00	4
4213	0.00	0.00	0.00	3
4214	0.00	0.00	0.00	2
4215	0.00	0.00	0.00	5
4216	0.00	0.00	0.00	4
4217	0.00	0.00	0.00	1
4218	0.00	0.00	0.00	3
4219	0.00	0.00	0.00	0
4220	0.00	0.00	0.00	4
4221	0.00	0.00	0.00	6
4222	0.00	0.00	0.00	3
4223	0.00	0.00	0.00	1
4224	0.00	0.00	0.00	3
4225	0.00	0.00	0.00	5
4226	0.00	0.00	0.00	4
4227	0.00	0.00	0.00	2
4228	0.00	0.00	0.00	0
4229	0.00	0.00	0.00	2
4230	0.00	0.00	0.00	3
4231	0.00	0.00	0.00	4
4232	0.00	0.00	0.00	1
4233	0.00	0.00	0.00	3
4234	0.00	0.00	0.00	2
4235	0.00	0.00	0.00	2
4236	0.00	0.00	0.00	5
4237	0.00	0.00	0.00	2
4238	0.00	0.00	0.00	6
4239	0.00	0.00	0.00	4
4240	0.00	0.00	0.00	2
4241	0.00	0.00	0.00	3
4242	0.00	0.00	0.00	4
4243	0.00	0.00	0.00	2
4244	0.00	0.00	0.00	2
4245	0.00	0.00	0.00	2
4246	0.00	0.00	0.00	3
4247	0.00	0.00	0.00	1
4248	0.00	0.00	0.00	3
4249	0.00	0.00	0.00	6
4250	0.00	0.00	0.00	4
4251	0.00	0.00	0.00	2
4252	0.00	0.00	0.00	6
4253	0.00	0.00	0.00	2
4254	0.00	0.00	0.00	1
4255	0.00	0.00	0.00	0
4256	0.00	0.00	0.00	2
4257	0.00	0.00	0.00	0
4258	0.00	0.00	0.00	3
4259	0.00	0.00	0.00	2
4260	0.00	0.00	0.00	3
4261	0.00	0.00	0.00	4
4262	0.00	0.00	0.00	3

4263	0.00	0.00	0.00	4
4264	0.00	0.00	0.00	3
4265	0.00	0.00	0.00	4
4266	0.00	0.00	0.00	4
4267	0.00	0.00	0.00	1
4268	0.00	0.00	0.00	1
4269	0.00	0.00	0.00	4
4270	0.00	0.00	0.00	2
4271	0.00	0.00	0.00	1
4272	0.00	0.00	0.00	3
4273	0.00	0.00	0.00	5
4274	0.00	0.00	0.00	2
4275	0.00	0.00	0.00	6
4276	0.00	0.00	0.00	2
4277	0.00	0.00	0.00	3
4278	0.00	0.00	0.00	5
4279	0.00	0.00	0.00	3
4280	0.00	0.00	0.00	2
4281	0.00	0.00	0.00	1
4282	0.00	0.00	0.00	2
4283	0.00	0.00	0.00	6
4284	0.00	0.00	0.00	2
4285	0.00	0.00	0.00	3
4286	0.00	0.00	0.00	1
4287	0.00	0.00	0.00	2
4288	0.00	0.00	0.00	5
4289	0.00	0.00	0.00	4
4290	0.00	0.00	0.00	5
4291	0.00	0.00	0.00	1
4292	0.00	0.00	0.00	3
4293	0.00	0.00	0.00	1
4294	0.00	0.00	0.00	4
4295	0.00	0.00	0.00	3
4296	0.00	0.00	0.00	1
4297	0.00	0.00	0.00	2
4298	0.00	0.00	0.00	3
4299	0.00	0.00	0.00	2
4300	0.00	0.00	0.00	3
4301	0.00	0.00	0.00	2
4302	0.00	0.00	0.00	4
4303	0.00	0.00	0.00	4
4304	0.00	0.00	0.00	3
4305	0.00	0.00	0.00	2
4306	0.00	0.00	0.00	2
4307	0.00	0.00	0.00	5
4308	0.00	0.00	0.00	2
4309	0.00	0.00	0.00	1
4310	0.00	0.00	0.00	3
4311	0.00	0.00	0.00	6
4312	0.00	0.00	0.00	2
4313	0.00	0.00	0.00	1
4314	0.00	0.00	0.00	5
4315	0.00	0.00	0.00	2
4316	0.00	0.00	0.00	5
4317	0.00	0.00	0.00	3
4318	0.00	0.00	0.00	1
4319	0.00	0.00	0.00	0
4320	0.00	0.00	0.00	1
4321	0.00	0.00	0.00	0
4322	0.00	0.00	0.00	4
4323	0.00	0.00	0.00	5

4324	0.00	0.00	0.00	1
4325	0.00	0.00	0.00	2
4326	0.00	0.00	0.00	4
4327	0.00	0.00	0.00	1
4328	0.00	0.00	0.00	4
4329	0.00	0.00	0.00	2
4330	0.00	0.00	0.00	3
4331	0.00	0.00	0.00	3
4332	0.00	0.00	0.00	1
4333	0.00	0.00	0.00	2
4334	0.00	0.00	0.00	3
4335	0.00	0.00	0.00	3
4336	0.00	0.00	0.00	2
4337	0.00	0.00	0.00	4
4338	0.00	0.00	0.00	6
4339	0.00	0.00	0.00	2
4340	0.00	0.00	0.00	1
4341	0.00	0.00	0.00	3
4342	0.00	0.00	0.00	2
4343	0.00	0.00	0.00	0
4344	0.00	0.00	0.00	1
4345	0.00	0.00	0.00	4
4346	0.00	0.00	0.00	4
4347	0.00	0.00	0.00	1
4348	0.00	0.00	0.00	5
4349	0.00	0.00	0.00	3
4350	0.00	0.00	0.00	2
4351	0.00	0.00	0.00	4
4352	0.00	0.00	0.00	2
4353	0.00	0.00	0.00	2
4354	0.00	0.00	0.00	5
4355	0.00	0.00	0.00	1
4356	0.00	0.00	0.00	5
4357	0.00	0.00	0.00	1
4358	0.00	0.00	0.00	2
4359	0.00	0.00	0.00	1
4360	0.00	0.00	0.00	1
4361	0.00	0.00	0.00	2
4362	0.00	0.00	0.00	4
4363	0.00	0.00	0.00	2
4364	0.00	0.00	0.00	2
4365	0.00	0.00	0.00	1
4366	0.00	0.00	0.00	2
4367	0.00	0.00	0.00	4
4368	0.00	0.00	0.00	2
4369	0.00	0.00	0.00	6
4370	0.00	0.00	0.00	1
4371	0.00	0.00	0.00	3
4372	0.00	0.00	0.00	2
4373	0.00	0.00	0.00	2
4374	0.00	0.00	0.00	1
4375	0.00	0.00	0.00	0
4376	0.00	0.00	0.00	3
4377	0.00	0.00	0.00	5
4378	0.00	0.00	0.00	2
4379	0.00	0.00	0.00	5
4380	0.00	0.00	0.00	2
4381	0.00	0.00	0.00	2
4382	0.00	0.00	0.00	2
4383	0.00	0.00	0.00	3
4384	0.00	0.00	0.00	0

4385	0.00	0.00	0.00	1
4386	0.00	0.00	0.00	3
4387	0.00	0.00	0.00	6
4388	0.00	0.00	0.00	5
4389	0.00	0.00	0.00	0
4390	0.00	0.00	0.00	4
4391	0.00	0.00	0.00	1
4392	0.00	0.00	0.00	1
4393	0.00	0.00	0.00	3
4394	0.00	0.00	0.00	0
4395	0.00	0.00	0.00	3
4396	0.00	0.00	0.00	4
4397	0.00	0.00	0.00	3
4398	0.00	0.00	0.00	3
4399	0.00	0.00	0.00	1
4400	0.00	0.00	0.00	1
4401	0.00	0.00	0.00	2
4402	0.00	0.00	0.00	0
4403	0.00	0.00	0.00	5
4404	0.00	0.00	0.00	3
4405	0.00	0.00	0.00	1
4406	0.00	0.00	0.00	3
4407	0.00	0.00	0.00	1
4408	0.00	0.00	0.00	0
4409	0.00	0.00	0.00	2
4410	0.00	0.00	0.00	3
4411	0.00	0.00	0.00	2
4412	0.00	0.00	0.00	5
4413	0.00	0.00	0.00	3
4414	0.00	0.00	0.00	3
4415	0.00	0.00	0.00	3
4416	0.00	0.00	0.00	1
4417	0.00	0.00	0.00	3
4418	0.00	0.00	0.00	2
4419	0.00	0.00	0.00	2
4420	0.00	0.00	0.00	4
4421	0.00	0.00	0.00	2
4422	0.00	0.00	0.00	0
4423	0.00	0.00	0.00	1
4424	0.00	0.00	0.00	7
4425	0.00	0.00	0.00	6
4426	0.00	0.00	0.00	2
4427	0.00	0.00	0.00	2
4428	0.00	0.00	0.00	2
4429	0.00	0.00	0.00	2
4430	0.00	0.00	0.00	2
4431	0.00	0.00	0.00	3
4432	0.00	0.00	0.00	1
4433	0.00	0.00	0.00	6
4434	0.00	0.00	0.00	2
4435	0.00	0.00	0.00	3
4436	0.00	0.00	0.00	1
4437	0.00	0.00	0.00	2
4438	0.00	0.00	0.00	2
4439	0.00	0.00	0.00	3
4440	0.00	0.00	0.00	4
4441	0.00	0.00	0.00	4
4442	0.00	0.00	0.00	2
4443	0.00	0.00	0.00	1
4444	0.00	0.00	0.00	5
4445	0.00	0.00	0.00	2

4446	0.00	0.00	0.00	2
4447	0.00	0.00	0.00	1
4448	0.00	0.00	0.00	3
4449	0.00	0.00	0.00	1
4450	0.00	0.00	0.00	0
4451	0.00	0.00	0.00	0
4452	0.00	0.00	0.00	3
4453	0.00	0.00	0.00	2
4454	0.00	0.00	0.00	0
4455	0.00	0.00	0.00	1
4456	0.00	0.00	0.00	2
4457	0.00	0.00	0.00	2
4458	0.00	0.00	0.00	1
4459	0.00	0.00	0.00	4
4460	0.00	0.00	0.00	3
4461	0.00	0.00	0.00	1
4462	0.00	0.00	0.00	1
4463	0.00	0.00	0.00	2
4464	0.00	0.00	0.00	2
4465	0.00	0.00	0.00	3
4466	0.00	0.00	0.00	1
4467	0.00	0.00	0.00	3
4468	0.00	0.00	0.00	0
4469	0.00	0.00	0.00	1
4470	0.00	0.00	0.00	2
4471	0.00	0.00	0.00	4
4472	0.00	0.00	0.00	5
4473	0.00	0.00	0.00	2
4474	0.00	0.00	0.00	1
4475	0.00	0.00	0.00	0
4476	0.00	0.00	0.00	4
4477	0.00	0.00	0.00	6
4478	0.00	0.00	0.00	5
4479	0.00	0.00	0.00	2
4480	0.00	0.00	0.00	2
4481	0.00	0.00	0.00	2
4482	0.00	0.00	0.00	1
4483	0.00	0.00	0.00	2
4484	0.00	0.00	0.00	4
4485	0.00	0.00	0.00	2
4486	0.00	0.00	0.00	3
4487	0.00	0.00	0.00	5
4488	0.00	0.00	0.00	3
4489	0.00	0.00	0.00	1
4490	0.00	0.00	0.00	2
4491	0.00	0.00	0.00	3
4492	0.00	0.00	0.00	4
4493	0.00	0.00	0.00	0
4494	0.00	0.00	0.00	3
4495	0.00	0.00	0.00	2
4496	0.00	0.00	0.00	3
4497	0.00	0.00	0.00	4
4498	0.00	0.00	0.00	2
4499	0.00	0.00	0.00	1
4500	0.00	0.00	0.00	3
4501	0.00	0.00	0.00	4
4502	0.00	0.00	0.00	4
4503	0.00	0.00	0.00	3
4504	0.00	0.00	0.00	2
4505	0.00	0.00	0.00	2
4506	0.00	0.00	0.00	4

4507	0.00	0.00	0.00	1
4508	0.00	0.00	0.00	2
4509	0.00	0.00	0.00	3
4510	0.00	0.00	0.00	3
4511	0.00	0.00	0.00	3
4512	0.00	0.00	0.00	1
4513	0.00	0.00	0.00	1
4514	0.00	0.00	0.00	5
4515	0.00	0.00	0.00	1
4516	0.00	0.00	0.00	1
4517	0.00	0.00	0.00	2
4518	0.00	0.00	0.00	4
4519	0.00	0.00	0.00	1
4520	0.00	0.00	0.00	0
4521	0.00	0.00	0.00	2
4522	0.00	0.00	0.00	0
4523	0.00	0.00	0.00	3
4524	0.00	0.00	0.00	4
4525	0.00	0.00	0.00	1
4526	0.00	0.00	0.00	2
4527	0.00	0.00	0.00	1
4528	0.00	0.00	0.00	2
4529	0.00	0.00	0.00	1
4530	0.00	0.00	0.00	3
4531	0.00	0.00	0.00	2
4532	0.00	0.00	0.00	5
4533	0.00	0.00	0.00	4
4534	0.00	0.00	0.00	1
4535	0.00	0.00	0.00	3
4536	0.00	0.00	0.00	5
4537	0.00	0.00	0.00	2
4538	0.00	0.00	0.00	3
4539	0.00	0.00	0.00	3
4540	0.00	0.00	0.00	1
4541	0.00	0.00	0.00	4
4542	0.00	0.00	0.00	6
4543	0.00	0.00	0.00	5
4544	0.00	0.00	0.00	2
4545	0.00	0.00	0.00	1
4546	0.00	0.00	0.00	3
4547	0.00	0.00	0.00	4
4548	0.00	0.00	0.00	4
4549	0.00	0.00	0.00	2
4550	0.00	0.00	0.00	3
4551	0.00	0.00	0.00	2
4552	0.00	0.00	0.00	1
4553	0.00	0.00	0.00	1
4554	0.00	0.00	0.00	3
4555	0.00	0.00	0.00	2
4556	0.00	0.00	0.00	2
4557	0.00	0.00	0.00	1
4558	0.00	0.00	0.00	3
4559	0.00	0.00	0.00	3
4560	0.00	0.00	0.00	0
4561	0.00	0.00	0.00	1
4562	0.00	0.00	0.00	1
4563	0.00	0.00	0.00	0
4564	0.00	0.00	0.00	0
4565	0.00	0.00	0.00	3
4566	0.00	0.00	0.00	3
4567	0.00	0.00	0.00	2

4568	0.00	0.00	0.00	5
4569	0.00	0.00	0.00	2
4570	0.00	0.00	0.00	3
4571	0.00	0.00	0.00	4
4572	0.00	0.00	0.00	4
4573	0.00	0.00	0.00	0
4574	0.00	0.00	0.00	6
4575	0.00	0.00	0.00	2
4576	0.00	0.00	0.00	1
4577	0.00	0.00	0.00	3
4578	0.00	0.00	0.00	3
4579	0.00	0.00	0.00	2
4580	0.00	0.00	0.00	5
4581	0.00	0.00	0.00	3
4582	0.00	0.00	0.00	1
4583	0.00	0.00	0.00	1
4584	0.00	0.00	0.00	3
4585	0.00	0.00	0.00	3
4586	0.00	0.00	0.00	1
4587	0.00	0.00	0.00	2
4588	0.00	0.00	0.00	0
4589	0.00	0.00	0.00	2
4590	0.00	0.00	0.00	2
4591	0.00	0.00	0.00	2
4592	0.00	0.00	0.00	4
4593	0.00	0.00	0.00	8
4594	0.00	0.00	0.00	3
4595	0.00	0.00	0.00	2
4596	0.00	0.00	0.00	4
4597	0.00	0.00	0.00	5
4598	0.00	0.00	0.00	2
4599	0.00	0.00	0.00	1
4600	0.00	0.00	0.00	2
4601	0.00	0.00	0.00	2
4602	0.00	0.00	0.00	2
4603	0.00	0.00	0.00	1
4604	0.00	0.00	0.00	2
4605	0.00	0.00	0.00	1
4606	0.00	0.00	0.00	2
4607	0.00	0.00	0.00	4
4608	0.00	0.00	0.00	3
4609	0.00	0.00	0.00	1
4610	0.00	0.00	0.00	4
4611	0.00	0.00	0.00	2
4612	0.00	0.00	0.00	1
4613	0.00	0.00	0.00	1
4614	0.00	0.00	0.00	3
4615	0.00	0.00	0.00	1
4616	0.00	0.00	0.00	1
4617	0.00	0.00	0.00	4
4618	0.00	0.00	0.00	3
4619	0.00	0.00	0.00	3
4620	0.00	0.00	0.00	4
4621	0.00	0.00	0.00	4
4622	0.00	0.00	0.00	1
4623	0.00	0.00	0.00	2
4624	0.00	0.00	0.00	2
4625	0.00	0.00	0.00	3
4626	0.00	0.00	0.00	2
4627	0.00	0.00	0.00	3
4628	0.00	0.00	0.00	3

4629	0.00	0.00	0.00	3
4630	0.00	0.00	0.00	1
4631	0.00	0.00	0.00	2
4632	0.00	0.00	0.00	2
4633	0.00	0.00	0.00	5
4634	0.00	0.00	0.00	2
4635	0.00	0.00	0.00	2
4636	0.00	0.00	0.00	1
4637	0.00	0.00	0.00	3
4638	0.00	0.00	0.00	3
4639	0.00	0.00	0.00	4
4640	0.00	0.00	0.00	3
4641	0.00	0.00	0.00	1
4642	0.00	0.00	0.00	4
4643	0.00	0.00	0.00	2
4644	0.00	0.00	0.00	2
4645	0.00	0.00	0.00	3
4646	0.00	0.00	0.00	0
4647	0.00	0.00	0.00	4
4648	0.00	0.00	0.00	2
4649	0.00	0.00	0.00	0
4650	0.00	0.00	0.00	5
4651	0.00	0.00	0.00	5
4652	0.00	0.00	0.00	5
4653	0.00	0.00	0.00	2
4654	0.00	0.00	0.00	4
4655	0.00	0.00	0.00	3
4656	0.00	0.00	0.00	2
4657	0.00	0.00	0.00	2
4658	0.00	0.00	0.00	1
4659	0.00	0.00	0.00	3
4660	0.00	0.00	0.00	5
4661	0.00	0.00	0.00	2
4662	0.00	0.00	0.00	4
4663	0.00	0.00	0.00	4
4664	0.00	0.00	0.00	3
4665	0.00	0.00	0.00	2
4666	0.00	0.00	0.00	3
4667	0.00	0.00	0.00	1
4668	0.00	0.00	0.00	5
4669	0.00	0.00	0.00	3
4670	0.00	0.00	0.00	0
4671	0.00	0.00	0.00	0
4672	0.00	0.00	0.00	1
4673	0.00	0.00	0.00	1
4674	0.00	0.00	0.00	3
4675	0.00	0.00	0.00	1
4676	0.00	0.00	0.00	3
4677	0.00	0.00	0.00	1
4678	0.00	0.00	0.00	4
4679	0.00	0.00	0.00	4
4680	0.00	0.00	0.00	5
4681	0.00	0.00	0.00	1
4682	0.00	0.00	0.00	0
4683	0.00	0.00	0.00	1
4684	0.00	0.00	0.00	4
4685	0.00	0.00	0.00	1
4686	0.00	0.00	0.00	1
4687	0.00	0.00	0.00	2
4688	0.00	0.00	0.00	2
4689	0.00	0.00	0.00	1

4690	0.00	0.00	0.00	2
4691	0.00	0.00	0.00	2
4692	0.00	0.00	0.00	2
4693	0.00	0.00	0.00	1
4694	0.00	0.00	0.00	1
4695	0.00	0.00	0.00	2
4696	0.00	0.00	0.00	1
4697	0.00	0.00	0.00	4
4698	0.00	0.00	0.00	2
4699	0.00	0.00	0.00	1
4700	0.00	0.00	0.00	1
4701	0.00	0.00	0.00	3
4702	0.00	0.00	0.00	1
4703	0.00	0.00	0.00	1
4704	0.00	0.00	0.00	1
4705	0.00	0.00	0.00	1
4706	0.00	0.00	0.00	3
4707	0.00	0.00	0.00	1
4708	0.00	0.00	0.00	2
4709	0.00	0.00	0.00	1
4710	0.00	0.00	0.00	1
4711	0.00	0.00	0.00	3
4712	0.00	0.00	0.00	2
4713	0.00	0.00	0.00	0
4714	0.00	0.00	0.00	3
4715	0.00	0.00	0.00	3
4716	0.00	0.00	0.00	3
4717	0.00	0.00	0.00	3
4718	0.00	0.00	0.00	6
4719	0.00	0.00	0.00	1
4720	0.00	0.00	0.00	3
4721	0.00	0.00	0.00	3
4722	0.00	0.00	0.00	3
4723	0.00	0.00	0.00	4
4724	0.00	0.00	0.00	2
4725	0.00	0.00	0.00	3
4726	0.00	0.00	0.00	0
4727	0.00	0.00	0.00	1
4728	0.00	0.00	0.00	4
4729	0.00	0.00	0.00	2
4730	0.00	0.00	0.00	5
4731	0.00	0.00	0.00	3
4732	0.00	0.00	0.00	2
4733	0.00	0.00	0.00	2
4734	0.00	0.00	0.00	0
4735	0.00	0.00	0.00	2
4736	0.00	0.00	0.00	2
4737	0.00	0.00	0.00	0
4738	0.00	0.00	0.00	4
4739	0.00	0.00	0.00	1
4740	0.00	0.00	0.00	5
4741	0.00	0.00	0.00	3
4742	0.00	0.00	0.00	1
4743	0.00	0.00	0.00	2
4744	0.00	0.00	0.00	2
4745	0.00	0.00	0.00	1
4746	0.00	0.00	0.00	3
4747	0.00	0.00	0.00	1
4748	0.00	0.00	0.00	2
4749	0.00	0.00	0.00	2
4750	0.00	0.00	0.00	4

4751	0.00	0.00	0.00	2
4752	0.00	0.00	0.00	2
4753	0.00	0.00	0.00	1
4754	0.00	0.00	0.00	1
4755	0.00	0.00	0.00	1
4756	0.00	0.00	0.00	5
4757	0.00	0.00	0.00	2
4758	0.00	0.00	0.00	4
4759	0.00	0.00	0.00	1
4760	0.00	0.00	0.00	1
4761	0.00	0.00	0.00	0
4762	0.00	0.00	0.00	3
4763	0.00	0.00	0.00	0
4764	0.00	0.00	0.00	5
4765	0.00	0.00	0.00	2
4766	0.00	0.00	0.00	4
4767	0.00	0.00	0.00	4
4768	0.00	0.00	0.00	4
4769	0.00	0.00	0.00	0
4770	0.00	0.00	0.00	1
4771	0.00	0.00	0.00	5
4772	0.00	0.00	0.00	3
4773	0.00	0.00	0.00	3
4774	0.00	0.00	0.00	3
4775	0.00	0.00	0.00	1
4776	0.00	0.00	0.00	3
4777	0.00	0.00	0.00	0
4778	0.00	0.00	0.00	1
4779	0.00	0.00	0.00	3
4780	0.00	0.00	0.00	4
4781	0.00	0.00	0.00	1
4782	0.00	0.00	0.00	0
4783	0.00	0.00	0.00	0
4784	0.00	0.00	0.00	3
4785	0.00	0.00	0.00	0
4786	0.00	0.00	0.00	3
4787	0.00	0.00	0.00	2
4788	0.00	0.00	0.00	3
4789	0.00	0.00	0.00	1
4790	0.00	0.00	0.00	3
4791	0.00	0.00	0.00	3
4792	0.00	0.00	0.00	4
4793	0.00	0.00	0.00	3
4794	0.00	0.00	0.00	0
4795	0.00	0.00	0.00	1
4796	0.00	0.00	0.00	3
4797	0.00	0.00	0.00	1
4798	0.00	0.00	0.00	2
4799	0.00	0.00	0.00	2
4800	0.00	0.00	0.00	1
4801	0.00	0.00	0.00	3
4802	0.00	0.00	0.00	2
4803	0.00	0.00	0.00	2
4804	0.00	0.00	0.00	4
4805	0.00	0.00	0.00	1
4806	0.00	0.00	0.00	0
4807	0.00	0.00	0.00	4
4808	0.00	0.00	0.00	4
4809	0.00	0.00	0.00	1
4810	0.00	0.00	0.00	4
4811	0.00	0.00	0.00	3

4812	0.00	0.00	0.00	2
4813	0.00	0.00	0.00	3
4814	0.00	0.00	0.00	0
4815	0.00	0.00	0.00	1
4816	0.00	0.00	0.00	2
4817	0.00	0.00	0.00	2
4818	0.00	0.00	0.00	4
4819	0.00	0.00	0.00	1
4820	0.00	0.00	0.00	3
4821	0.00	0.00	0.00	3
4822	0.00	0.00	0.00	1
4823	0.00	0.00	0.00	2
4824	0.00	0.00	0.00	1
4825	0.00	0.00	0.00	1
4826	0.00	0.00	0.00	6
4827	0.00	0.00	0.00	4
4828	0.00	0.00	0.00	0
4829	0.00	0.00	0.00	3
4830	0.00	0.00	0.00	1
4831	0.00	0.00	0.00	3
4832	0.00	0.00	0.00	4
4833	0.00	0.00	0.00	4
4834	0.00	0.00	0.00	2
4835	0.00	0.00	0.00	2
4836	0.00	0.00	0.00	5
4837	0.00	0.00	0.00	2
4838	0.00	0.00	0.00	2
4839	0.00	0.00	0.00	2
4840	0.00	0.00	0.00	2
4841	0.00	0.00	0.00	1
4842	0.00	0.00	0.00	4
4843	0.00	0.00	0.00	5
4844	0.00	0.00	0.00	2
4845	0.00	0.00	0.00	1
4846	0.00	0.00	0.00	3
4847	0.00	0.00	0.00	3
4848	0.00	0.00	0.00	1
4849	0.00	0.00	0.00	2
4850	0.00	0.00	0.00	0
4851	0.00	0.00	0.00	1
4852	0.00	0.00	0.00	7
4853	0.00	0.00	0.00	4
4854	0.00	0.00	0.00	1
4855	0.00	0.00	0.00	3
4856	0.00	0.00	0.00	1
4857	0.00	0.00	0.00	2
4858	0.00	0.00	0.00	2
4859	0.00	0.00	0.00	5
4860	0.00	0.00	0.00	1
4861	0.00	0.00	0.00	1
4862	0.00	0.00	0.00	4
4863	0.00	0.00	0.00	3
4864	0.00	0.00	0.00	3
4865	0.00	0.00	0.00	2
4866	0.00	0.00	0.00	2
4867	0.00	0.00	0.00	4
4868	0.00	0.00	0.00	1
4869	0.00	0.00	0.00	2
4870	0.00	0.00	0.00	2
4871	0.00	0.00	0.00	1
4872	0.00	0.00	0.00	2

4873	0.00	0.00	0.00	2
4874	0.00	0.00	0.00	4
4875	0.00	0.00	0.00	4
4876	0.00	0.00	0.00	4
4877	0.00	0.00	0.00	4
4878	0.00	0.00	0.00	2
4879	0.00	0.00	0.00	3
4880	0.00	0.00	0.00	2
4881	0.00	0.00	0.00	4
4882	0.00	0.00	0.00	1
4883	0.00	0.00	0.00	2
4884	0.00	0.00	0.00	2
4885	0.00	0.00	0.00	0
4886	0.00	0.00	0.00	2
4887	0.00	0.00	0.00	2
4888	0.00	0.00	0.00	2
4889	0.00	0.00	0.00	3
4890	0.00	0.00	0.00	1
4891	0.00	0.00	0.00	0
4892	0.00	0.00	0.00	0
4893	0.00	0.00	0.00	3
4894	0.00	0.00	0.00	0
4895	0.00	0.00	0.00	0
4896	0.00	0.00	0.00	0
4897	0.00	0.00	0.00	7
4898	0.00	0.00	0.00	1
4899	0.00	0.00	0.00	5
4900	0.00	0.00	0.00	4
4901	0.00	0.00	0.00	2
4902	0.00	0.00	0.00	1
4903	0.00	0.00	0.00	0
4904	0.00	0.00	0.00	4
4905	0.00	0.00	0.00	4
4906	0.00	0.00	0.00	3
4907	0.00	0.00	0.00	1
4908	0.00	0.00	0.00	0
4909	0.00	0.00	0.00	4
4910	0.00	0.00	0.00	3
4911	0.00	0.00	0.00	2
4912	0.00	0.00	0.00	3
4913	0.00	0.00	0.00	1
4914	0.00	0.00	0.00	5
4915	0.00	0.00	0.00	4
4916	1.00	0.67	0.80	3
4917	0.00	0.00	0.00	2
4918	0.00	0.00	0.00	2
4919	0.00	0.00	0.00	2
4920	0.00	0.00	0.00	2
4921	0.00	0.00	0.00	3
4922	0.00	0.00	0.00	1
4923	0.00	0.00	0.00	4
4924	0.00	0.00	0.00	2
4925	0.00	0.00	0.00	1
4926	0.00	0.00	0.00	1
4927	0.00	0.00	0.00	1
4928	0.00	0.00	0.00	2
4929	0.00	0.00	0.00	2
4930	0.00	0.00	0.00	1
4931	0.00	0.00	0.00	3
4932	0.00	0.00	0.00	2
4933	0.00	0.00	0.00	1

4934	0.00	0.00	0.00	0
4935	0.00	0.00	0.00	2
4936	0.00	0.00	0.00	5
4937	0.00	0.00	0.00	5
4938	0.00	0.00	0.00	2
4939	0.00	0.00	0.00	2
4940	0.00	0.00	0.00	3
4941	0.00	0.00	0.00	1
4942	0.00	0.00	0.00	4
4943	0.00	0.00	0.00	1
4944	0.00	0.00	0.00	4
4945	0.00	0.00	0.00	2
4946	0.00	0.00	0.00	5
4947	0.00	0.00	0.00	2
4948	0.00	0.00	0.00	1
4949	0.00	0.00	0.00	0
4950	0.00	0.00	0.00	1
4951	0.00	0.00	0.00	1
4952	0.00	0.00	0.00	0
4953	0.00	0.00	0.00	1
4954	0.00	0.00	0.00	2
4955	0.00	0.00	0.00	2
4956	0.00	0.00	0.00	3
4957	0.00	0.00	0.00	2
4958	0.00	0.00	0.00	1
4959	0.00	0.00	0.00	1
4960	0.00	0.00	0.00	2
4961	0.00	0.00	0.00	3
4962	0.00	0.00	0.00	1
4963	0.00	0.00	0.00	2
4964	0.00	0.00	0.00	5
4965	0.00	0.00	0.00	0
4966	0.00	0.00	0.00	5
4967	0.00	0.00	0.00	3
4968	0.00	0.00	0.00	2
4969	0.00	0.00	0.00	0
4970	0.00	0.00	0.00	3
4971	0.00	0.00	0.00	2
4972	0.00	0.00	0.00	1
4973	0.00	0.00	0.00	1
4974	0.00	0.00	0.00	2
4975	0.00	0.00	0.00	3
4976	0.00	0.00	0.00	1
4977	0.00	0.00	0.00	2
4978	0.00	0.00	0.00	0
4979	0.00	0.00	0.00	5
4980	0.00	0.00	0.00	2
4981	0.00	0.00	0.00	1
4982	0.00	0.00	0.00	4
4983	0.00	0.00	0.00	3
4984	0.00	0.00	0.00	0
4985	0.00	0.00	0.00	3
4986	0.00	0.00	0.00	4
4987	0.00	0.00	0.00	0
4988	0.00	0.00	0.00	1
4989	0.00	0.00	0.00	0
4990	0.00	0.00	0.00	5
4991	0.00	0.00	0.00	0
4992	0.00	0.00	0.00	2
4993	0.00	0.00	0.00	4
4994	0.00	0.00	0.00	4

4995	0.00	0.00	0.00	2
4996	0.00	0.00	0.00	3
4997	0.00	0.00	0.00	2
4998	0.00	0.00	0.00	3
4999	0.00	0.00	0.00	3
5000	0.00	0.00	0.00	2
5001	0.00	0.00	0.00	0
5002	0.00	0.00	0.00	5
5003	0.00	0.00	0.00	6
5004	0.00	0.00	0.00	1
5005	0.00	0.00	0.00	1
5006	0.00	0.00	0.00	3
5007	0.00	0.00	0.00	2
5008	0.00	0.00	0.00	1
5009	0.00	0.00	0.00	0
5010	0.00	0.00	0.00	5
5011	0.00	0.00	0.00	4
5012	0.00	0.00	0.00	6
5013	0.00	0.00	0.00	6
5014	0.00	0.00	0.00	4
5015	0.00	0.00	0.00	0
5016	0.00	0.00	0.00	1
5017	0.00	0.00	0.00	2
5018	0.00	0.00	0.00	1
5019	0.00	0.00	0.00	4
5020	0.00	0.00	0.00	3
5021	0.00	0.00	0.00	4
5022	0.00	0.00	0.00	4
5023	0.00	0.00	0.00	1
5024	0.00	0.00	0.00	2
5025	0.00	0.00	0.00	2
5026	0.00	0.00	0.00	4
5027	0.00	0.00	0.00	1
5028	0.00	0.00	0.00	1
5029	0.00	0.00	0.00	0
5030	0.00	0.00	0.00	4
5031	0.00	0.00	0.00	1
5032	0.00	0.00	0.00	2
5033	0.00	0.00	0.00	4
5034	0.00	0.00	0.00	1
5035	0.00	0.00	0.00	2
5036	0.00	0.00	0.00	2
5037	0.00	0.00	0.00	2
5038	0.00	0.00	0.00	2
5039	0.00	0.00	0.00	1
5040	0.00	0.00	0.00	1
5041	0.00	0.00	0.00	2
5042	0.00	0.00	0.00	1
5043	0.00	0.00	0.00	2
5044	0.00	0.00	0.00	4
5045	0.00	0.00	0.00	2
5046	0.00	0.00	0.00	3
5047	0.00	0.00	0.00	2
5048	0.00	0.00	0.00	1
5049	0.00	0.00	0.00	4
5050	0.00	0.00	0.00	7
5051	0.00	0.00	0.00	2
5052	0.00	0.00	0.00	1
5053	0.00	0.00	0.00	4
5054	0.00	0.00	0.00	2
5055	0.00	0.00	0.00	4

5056	0.00	0.00	0.00	1
5057	0.00	0.00	0.00	3
5058	0.00	0.00	0.00	3
5059	0.00	0.00	0.00	2
5060	0.00	0.00	0.00	2
5061	0.00	0.00	0.00	3
5062	0.00	0.00	0.00	1
5063	0.00	0.00	0.00	4
5064	0.00	0.00	0.00	2
5065	0.00	0.00	0.00	2
5066	0.00	0.00	0.00	1
5067	0.00	0.00	0.00	4
5068	0.00	0.00	0.00	3
5069	0.00	0.00	0.00	1
5070	0.00	0.00	0.00	0
5071	0.00	0.00	0.00	1
5072	0.00	0.00	0.00	1
5073	0.00	0.00	0.00	2
5074	0.00	0.00	0.00	2
5075	0.00	0.00	0.00	4
5076	0.00	0.00	0.00	4
5077	0.00	0.00	0.00	3
5078	0.00	0.00	0.00	2
5079	0.00	0.00	0.00	2
5080	0.00	0.00	0.00	3
5081	0.00	0.00	0.00	2
5082	0.00	0.00	0.00	1
5083	0.00	0.00	0.00	2
5084	0.00	0.00	0.00	0
5085	0.00	0.00	0.00	0
5086	0.00	0.00	0.00	4
5087	0.00	0.00	0.00	1
5088	0.00	0.00	0.00	1
5089	0.00	0.00	0.00	1
5090	0.00	0.00	0.00	3
5091	0.00	0.00	0.00	4
5092	0.00	0.00	0.00	3
5093	0.00	0.00	0.00	2
5094	0.00	0.00	0.00	1
5095	0.00	0.00	0.00	3
5096	0.00	0.00	0.00	0
5097	0.00	0.00	0.00	4
5098	0.00	0.00	0.00	1
5099	0.00	0.00	0.00	3
5100	0.00	0.00	0.00	2
5101	0.00	0.00	0.00	3
5102	0.00	0.00	0.00	3
5103	0.00	0.00	0.00	0
5104	0.00	0.00	0.00	2
5105	0.00	0.00	0.00	1
5106	0.00	0.00	0.00	3
5107	0.00	0.00	0.00	1
5108	0.00	0.00	0.00	0
5109	0.00	0.00	0.00	0
5110	0.00	0.00	0.00	1
5111	0.00	0.00	0.00	4
5112	0.00	0.00	0.00	5
5113	0.00	0.00	0.00	3
5114	0.00	0.00	0.00	3
5115	0.00	0.00	0.00	2
5116	0.00	0.00	0.00	3

5117	0.00	0.00	0.00	3
5118	0.00	0.00	0.00	3
5119	0.00	0.00	0.00	1
5120	0.00	0.00	0.00	2
5121	0.00	0.00	0.00	2
5122	0.00	0.00	0.00	2
5123	0.00	0.00	0.00	0
5124	0.00	0.00	0.00	3
5125	0.00	0.00	0.00	4
5126	0.00	0.00	0.00	4
5127	0.00	0.00	0.00	3
5128	0.00	0.00	0.00	3
5129	0.00	0.00	0.00	5
5130	0.00	0.00	0.00	2
5131	0.00	0.00	0.00	3
5132	0.00	0.00	0.00	2
5133	0.00	0.00	0.00	1
5134	0.00	0.00	0.00	1
5135	0.00	0.00	0.00	0
5136	0.00	0.00	0.00	1
5137	0.00	0.00	0.00	1
5138	0.00	0.00	0.00	2
5139	0.00	0.00	0.00	2
5140	0.00	0.00	0.00	2
5141	0.00	0.00	0.00	1
5142	0.00	0.00	0.00	2
5143	0.00	0.00	0.00	2
5144	0.00	0.00	0.00	2
5145	0.00	0.00	0.00	2
5146	0.00	0.00	0.00	2
5147	0.00	0.00	0.00	2
5148	0.00	0.00	0.00	1
5149	0.00	0.00	0.00	3
5150	0.00	0.00	0.00	2
5151	0.00	0.00	0.00	1
5152	0.00	0.00	0.00	0
5153	0.00	0.00	0.00	1
5154	0.00	0.00	0.00	1
5155	0.00	0.00	0.00	0
5156	0.00	0.00	0.00	1
5157	0.00	0.00	0.00	3
5158	0.00	0.00	0.00	4
5159	0.00	0.00	0.00	1
5160	0.00	0.00	0.00	2
5161	0.00	0.00	0.00	5
5162	0.00	0.00	0.00	0
5163	0.00	0.00	0.00	3
5164	0.00	0.00	0.00	1
5165	0.00	0.00	0.00	3
5166	0.00	0.00	0.00	2
5167	0.00	0.00	0.00	2
5168	0.00	0.00	0.00	1
5169	0.00	0.00	0.00	2
5170	0.00	0.00	0.00	2
5171	0.00	0.00	0.00	2
5172	0.00	0.00	0.00	2
5173	0.00	0.00	0.00	2
5174	0.00	0.00	0.00	1
5175	0.00	0.00	0.00	2
5176	0.00	0.00	0.00	2
5177	0.00	0.00	0.00	2

5178	0.00	0.00	0.00	5
5179	0.00	0.00	0.00	0
5180	0.00	0.00	0.00	3
5181	0.00	0.00	0.00	2
5182	0.00	0.00	0.00	0
5183	0.00	0.00	0.00	0
5184	0.00	0.00	0.00	1
5185	0.00	0.00	0.00	2
5186	0.00	0.00	0.00	0
5187	0.00	0.00	0.00	3
5188	0.00	0.00	0.00	1
5189	0.00	0.00	0.00	1
5190	0.00	0.00	0.00	1
5191	0.00	0.00	0.00	1
5192	0.00	0.00	0.00	3
5193	0.00	0.00	0.00	4
5194	0.00	0.00	0.00	4
5195	0.00	0.00	0.00	1
5196	0.00	0.00	0.00	1
5197	0.00	0.00	0.00	2
5198	0.00	0.00	0.00	5
5199	0.00	0.00	0.00	1
5200	0.00	0.00	0.00	1
5201	0.00	0.00	0.00	0
5202	0.00	0.00	0.00	0
5203	0.00	0.00	0.00	3
5204	0.00	0.00	0.00	3
5205	0.00	0.00	0.00	1
5206	0.00	0.00	0.00	2
5207	0.00	0.00	0.00	1
5208	0.00	0.00	0.00	0
5209	0.00	0.00	0.00	3
5210	0.00	0.00	0.00	1
5211	0.00	0.00	0.00	2
5212	0.00	0.00	0.00	1
5213	0.00	0.00	0.00	5
5214	0.00	0.00	0.00	3
5215	0.00	0.00	0.00	2
5216	0.00	0.00	0.00	1
5217	0.00	0.00	0.00	2
5218	0.00	0.00	0.00	2
5219	0.00	0.00	0.00	0
5220	0.00	0.00	0.00	4
5221	0.00	0.00	0.00	1
5222	0.00	0.00	0.00	3
5223	0.00	0.00	0.00	1
5224	0.00	0.00	0.00	0
5225	0.00	0.00	0.00	2
5226	0.00	0.00	0.00	0
5227	0.00	0.00	0.00	1
5228	0.00	0.00	0.00	2
5229	0.00	0.00	0.00	2
5230	0.00	0.00	0.00	0
5231	0.00	0.00	0.00	2
5232	0.00	0.00	0.00	1
5233	0.00	0.00	0.00	2
5234	0.00	0.00	0.00	1
5235	0.00	0.00	0.00	1
5236	0.00	0.00	0.00	3
5237	0.00	0.00	0.00	1
5238	0.00	0.00	0.00	2

5239	0.00	0.00	0.00	2
5240	0.00	0.00	0.00	4
5241	0.00	0.00	0.00	2
5242	0.00	0.00	0.00	1
5243	0.00	0.00	0.00	0
5244	0.00	0.00	0.00	2
5245	0.00	0.00	0.00	2
5246	0.00	0.00	0.00	4
5247	0.00	0.00	0.00	0
5248	0.00	0.00	0.00	3
5249	0.00	0.00	0.00	3
5250	0.00	0.00	0.00	2
5251	0.00	0.00	0.00	1
5252	0.00	0.00	0.00	4
5253	0.00	0.00	0.00	4
5254	0.00	0.00	0.00	0
5255	0.00	0.00	0.00	0
5256	0.00	0.00	0.00	0
5257	0.00	0.00	0.00	1
5258	0.00	0.00	0.00	2
5259	0.00	0.00	0.00	2
5260	0.00	0.00	0.00	3
5261	0.00	0.00	0.00	2
5262	0.00	0.00	0.00	1
5263	0.00	0.00	0.00	3
5264	0.00	0.00	0.00	3
5265	0.00	0.00	0.00	1
5266	0.00	0.00	0.00	5
5267	0.00	0.00	0.00	3
5268	0.00	0.00	0.00	2
5269	0.00	0.00	0.00	1
5270	0.00	0.00	0.00	0
5271	0.00	0.00	0.00	2
5272	0.00	0.00	0.00	3
5273	0.00	0.00	0.00	3
5274	0.00	0.00	0.00	3
5275	0.00	0.00	0.00	3
5276	0.00	0.00	0.00	2
5277	0.00	0.00	0.00	1
5278	0.00	0.00	0.00	3
5279	0.00	0.00	0.00	1
5280	0.00	0.00	0.00	0
5281	0.00	0.00	0.00	1
5282	0.00	0.00	0.00	2
5283	0.00	0.00	0.00	1
5284	0.00	0.00	0.00	1
5285	0.00	0.00	0.00	0
5286	0.00	0.00	0.00	2
5287	0.00	0.00	0.00	2
5288	0.00	0.00	0.00	2
5289	0.00	0.00	0.00	4
5290	0.00	0.00	0.00	0
5291	0.00	0.00	0.00	3
5292	0.00	0.00	0.00	0
5293	0.00	0.00	0.00	5
5294	0.00	0.00	0.00	1
5295	0.00	0.00	0.00	2
5296	0.00	0.00	0.00	1
5297	0.00	0.00	0.00	3
5298	0.00	0.00	0.00	0
5299	0.00	0.00	0.00	3

5300	0.00	0.00	0.00	1
5301	0.00	0.00	0.00	1
5302	0.00	0.00	0.00	0
5303	0.00	0.00	0.00	2
5304	0.00	0.00	0.00	0
5305	0.00	0.00	0.00	1
5306	0.00	0.00	0.00	1
5307	0.00	0.00	0.00	2
5308	0.00	0.00	0.00	0
5309	0.00	0.00	0.00	3
5310	0.00	0.00	0.00	2
5311	0.00	0.00	0.00	1
5312	0.00	0.00	0.00	0
5313	0.00	0.00	0.00	1
5314	0.00	0.00	0.00	2
5315	0.00	0.00	0.00	2
5316	0.00	0.00	0.00	1
5317	0.00	0.00	0.00	0
5318	0.00	0.00	0.00	0
5319	0.00	0.00	0.00	4
5320	0.00	0.00	0.00	2
5321	0.00	0.00	0.00	2
5322	0.00	0.00	0.00	2
5323	0.00	0.00	0.00	1
5324	0.00	0.00	0.00	0
5325	0.00	0.00	0.00	1
5326	0.00	0.00	0.00	2
5327	0.00	0.00	0.00	1
5328	0.00	0.00	0.00	1
5329	0.00	0.00	0.00	1
5330	0.00	0.00	0.00	3
5331	0.00	0.00	0.00	0
5332	0.00	0.00	0.00	2
5333	0.00	0.00	0.00	4
5334	0.00	0.00	0.00	3
5335	0.00	0.00	0.00	0
5336	0.00	0.00	0.00	4
5337	0.00	0.00	0.00	5
5338	0.00	0.00	0.00	2
5339	0.00	0.00	0.00	2
5340	0.00	0.00	0.00	1
5341	0.00	0.00	0.00	2
5342	0.00	0.00	0.00	2
5343	0.00	0.00	0.00	1
5344	0.00	0.00	0.00	1
5345	0.00	0.00	0.00	3
5346	0.00	0.00	0.00	2
5347	0.00	0.00	0.00	2
5348	0.00	0.00	0.00	2
5349	0.00	0.00	0.00	0
5350	0.00	0.00	0.00	2
5351	0.00	0.00	0.00	1
5352	0.00	0.00	0.00	2
5353	0.00	0.00	0.00	3
5354	0.00	0.00	0.00	2
5355	0.00	0.00	0.00	1
5356	0.00	0.00	0.00	2
5357	0.00	0.00	0.00	4
5358	0.00	0.00	0.00	2
5359	0.00	0.00	0.00	1
5360	0.00	0.00	0.00	1

5361	0.00	0.00	0.00	2
5362	0.00	0.00	0.00	0
5363	0.00	0.00	0.00	2
5364	0.00	0.00	0.00	3
5365	0.00	0.00	0.00	3
5366	0.00	0.00	0.00	0
5367	0.00	0.00	0.00	2
5368	0.00	0.00	0.00	2
5369	0.00	0.00	0.00	5
5370	0.00	0.00	0.00	0
5371	0.00	0.00	0.00	1
5372	0.00	0.00	0.00	0
5373	0.00	0.00	0.00	1
5374	0.00	0.00	0.00	1
5375	0.00	0.00	0.00	4
5376	0.00	0.00	0.00	4
5377	0.00	0.00	0.00	3
5378	0.00	0.00	0.00	2
5379	0.00	0.00	0.00	4
5380	0.00	0.00	0.00	2
5381	0.00	0.00	0.00	4
5382	0.00	0.00	0.00	3
5383	0.00	0.00	0.00	2
5384	0.00	0.00	0.00	0
5385	0.00	0.00	0.00	3
5386	0.00	0.00	0.00	2
5387	0.00	0.00	0.00	1
5388	0.00	0.00	0.00	2
5389	0.00	0.00	0.00	2
5390	0.00	0.00	0.00	2
5391	0.00	0.00	0.00	3
5392	0.00	0.00	0.00	2
5393	0.00	0.00	0.00	3
5394	0.00	0.00	0.00	0
5395	0.00	0.00	0.00	1
5396	0.00	0.00	0.00	2
5397	0.00	0.00	0.00	0
5398	0.00	0.00	0.00	0
5399	0.00	0.00	0.00	4
5400	0.00	0.00	0.00	2
5401	0.00	0.00	0.00	1
5402	0.00	0.00	0.00	2
5403	0.00	0.00	0.00	1
5404	1.00	0.50	0.67	2
5405	0.00	0.00	0.00	0
5406	0.00	0.00	0.00	0
5407	0.00	0.00	0.00	3
5408	0.00	0.00	0.00	4
5409	0.00	0.00	0.00	1
5410	0.00	0.00	0.00	1
5411	0.00	0.00	0.00	1
5412	0.00	0.00	0.00	1
5413	0.00	0.00	0.00	2
5414	0.00	0.00	0.00	1
5415	0.00	0.00	0.00	3
5416	0.00	0.00	0.00	2
5417	0.00	0.00	0.00	2
5418	0.00	0.00	0.00	3
5419	0.00	0.00	0.00	2
5420	0.00	0.00	0.00	2
5421	0.00	0.00	0.00	3

5422	0.00	0.00	0.00	0
5423	0.00	0.00	0.00	1
5424	0.00	0.00	0.00	1
5425	0.00	0.00	0.00	3
5426	0.00	0.00	0.00	2
5427	0.00	0.00	0.00	1
5428	0.00	0.00	0.00	4
5429	0.00	0.00	0.00	3
5430	0.00	0.00	0.00	0
5431	0.00	0.00	0.00	0
5432	0.00	0.00	0.00	2
5433	0.00	0.00	0.00	2
5434	0.00	0.00	0.00	3
5435	0.00	0.00	0.00	3
5436	0.00	0.00	0.00	2
5437	0.00	0.00	0.00	3
5438	0.00	0.00	0.00	1
5439	0.00	0.00	0.00	2
5440	0.00	0.00	0.00	0
5441	0.00	0.00	0.00	2
5442	0.00	0.00	0.00	2
5443	0.00	0.00	0.00	2
5444	0.00	0.00	0.00	1
5445	0.00	0.00	0.00	3
5446	0.00	0.00	0.00	3
5447	0.00	0.00	0.00	1
5448	0.00	0.00	0.00	2
5449	0.00	0.00	0.00	2
5450	0.00	0.00	0.00	3
5451	0.00	0.00	0.00	5
5452	0.00	0.00	0.00	1
5453	0.00	0.00	0.00	1
5454	0.00	0.00	0.00	3
5455	0.00	0.00	0.00	2
5456	0.00	0.00	0.00	0
5457	0.00	0.00	0.00	3
5458	0.00	0.00	0.00	2
5459	0.00	0.00	0.00	1
5460	0.00	0.00	0.00	1
5461	0.00	0.00	0.00	1
5462	0.00	0.00	0.00	1
5463	0.00	0.00	0.00	2
5464	0.00	0.00	0.00	4
5465	0.00	0.00	0.00	1
5466	0.00	0.00	0.00	1
5467	0.00	0.00	0.00	2
5468	0.00	0.00	0.00	3
5469	0.00	0.00	0.00	3
5470	0.00	0.00	0.00	0
5471	0.00	0.00	0.00	0
5472	0.00	0.00	0.00	8
5473	0.00	0.00	0.00	2
5474	0.00	0.00	0.00	4
5475	0.00	0.00	0.00	2
5476	0.00	0.00	0.00	4
5477	0.00	0.00	0.00	2
5478	0.00	0.00	0.00	3
5479	0.00	0.00	0.00	2
5480	0.00	0.00	0.00	3
5481	0.00	0.00	0.00	3
5482	0.00	0.00	0.00	3

5483	0.00	0.00	0.00	1
5484	0.00	0.00	0.00	2
5485	0.00	0.00	0.00	3
5486	0.00	0.00	0.00	2
5487	0.00	0.00	0.00	3
5488	0.00	0.00	0.00	0
5489	0.00	0.00	0.00	3
5490	0.00	0.00	0.00	0
5491	0.00	0.00	0.00	7
5492	0.00	0.00	0.00	1
5493	0.00	0.00	0.00	1
5494	0.00	0.00	0.00	0
5495	0.00	0.00	0.00	2
5496	0.00	0.00	0.00	0
5497	0.00	0.00	0.00	1
5498	0.00	0.00	0.00	0
5499	0.00	0.00	0.00	2
micro avg	0.70	0.25	0.37	105988
macro avg	0.17	0.06	0.08	105988
weighted avg	0.52	0.25	0.33	105988
samples avg	0.46	0.28	0.32	105988

In [53]:

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

Out[53]:

```
['lr_with_equal_weight.pkl']
```

4.5 Modeling with less data points (0.2M data points) and more weight to title and 500 tags only.

In [54]:

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

Tables in the database:
QuestionsProcessed

In [55]:

```
# 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 = 200000
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 0.2M rows
        reader.execute("SELECT Title, Body, Tags From no_dup_train LIMIT 500001;")
        # for selecting random points
        #reader.execute("SELECT Title, Body, Tags From no_dup_train ORDER BY RANDOM
        () LIMIT 500001;")

if os.path.isfile(write_db):
    conn_w = create_connection(write_db)
    if conn_w is not None:
        tables = checkTableExists(conn_w)
        writer = conn_w.cursor()
        if tables != 0:
            writer.execute("DELETE FROM QuestionsProcessed WHERE 1")
            print("Cleared All the rows")
```

Tables in the database:

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**
4. Remove stop words (Except 'C')
5. Remove HTML Tags
6. Convert all the characters into small letters
7. Use SnowballStemmer to stem the words

In [56]:

```

#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:
#         question=str(title)+" "+str(title)+" "+str(title)+" "+question+" "+str(tag
s)
#     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 lett
er 'c'
    question=' '.join(str(stemmer.stem(j)) for j in words if j not in stop_words and
(len(j)!=1 or j=='c'))

    len_post+=len(question)
    tup = (question,code,tags,x,len(question),is_code)
    questions_proccesed += 1
    writer.execute("insert into QuestionsProcessed(question,code,tags,words_pre,word
s_post,is_code) values (?,?,?,?,?,?)",tup)
    if (questions_proccesed%100000==0):
        print("number of questions completed=",questions_proccesed)

no_dup_avg_len_pre=(len_pre*1.0)/questions_proccesed
no_dup_avg_len_post=(len_post*1.0)/questions_proccesed

print( "Avg. length of questions(Title+Body) before processing: %d"%no_dup_avg_len_p

```



```
re)
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_processed))

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 : 0:17:48.671155
```

In [57]:

```
# never forget to close the connections 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 [58]:

```
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
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block seem bind correct grid come column form come grid column although
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```
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onfused.i find post feed api method like correct second way use curl som
eth like way better',)
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```
-----
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wo window record ad btnadd click event open two window record ad open wi
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btnadd click event open anoth window nafter insert record close windo
w',)
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```

```
-----
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nt correct form submiss php sql inject issu prevent correct form submiss
php check everyth think make sure input field safe type sql inject good
news safe bad news one tag mess form submiss place even touch life figur
exact html use templat file forgiv okay entir php script get execut see
data post none forum field post problem use someth titl field none data
get post current use print post see submit noth work flawless statement
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ss problem state list input test mess',)
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```
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measur defin set sigma algebra mathcal think use monoton properti somewh
er proof start appreci littl help nthank ad han answer make follow addit
construct given han answer clear bigcup bigcup cap emptyset neq left big
cup right left bigcup right sum left right also construct subset monoton
left right leq left right final would sum leq sum result follow',)
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```
-----
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r undefin symbol architectur i386 objc class skpsmtpmessag referenc erro
r undefin symbol architectur i386 objc class skpsmtpmessag referenc erro
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smtpmessag somebodi suggest get error collect2 ld return exit status imp
ort framework correct sorc taken framework follow mfmcomposeviewcontr
ol question lock field updat answer drag drop folder project click copi
nthat',)
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```

Saving Preprocessed data to a Database

In [59]:

```
#Taking 0.5 Million entries to a dataframe.
write_db = 'Titlmoreweight.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 Question
sProcessed""", conn_r)
    conn_r.commit()
    conn_r.close()
```

In [60]:

```
preprocessed_data.head()
```

Out[60]:

	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 [61]:

```
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 [62]:

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

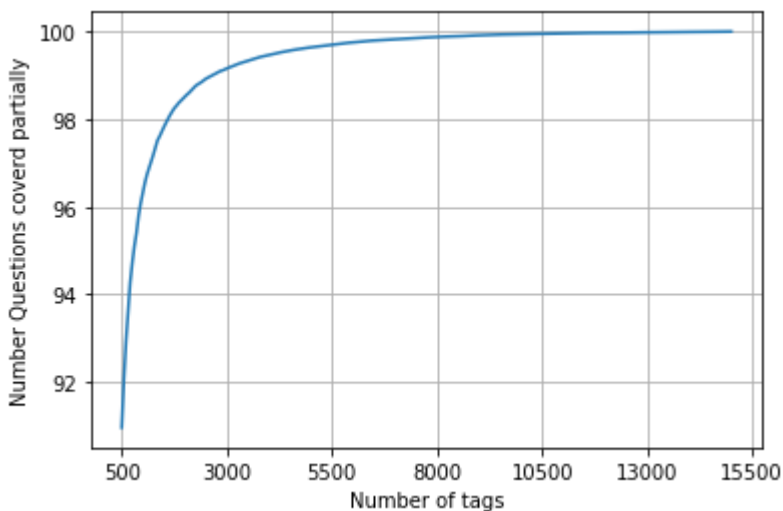
Selecting 500 Tags

In [63]:

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

In [64]:

```
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 covered partially")
plt.grid()
plt.show()
# you can choose any number of tags based on your computing power, minimum is 500(it
covers 90% of the tags)
print("with ",5500,"tags we are covering ",questions_explained[50],"% of questions")
print("with ",500,"tags we are covering ",questions_explained[0],"% of questions")
```



with 5500 tags we are covering 99.157 % of questions

with 500 tags we are covering 90.956 % of questions

In [65]:

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

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

In [66]:

```
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 [67]:

```
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 : (200000, 500)
 Number of data points in test data : (300000, 500)

4.5.2 Featurizing data with Tfidf vectorizer

In [68]:

```
start = datetime.now()
vectorizer = TfidfVectorizer(min_df=0.00009, max_features=100000, smooth_idf=True, norm="l2", \
                             tokenizer = lambda x: x.split(), sublinear_tf=False, ngram_range=(1,3))
x_train_multilabel = vectorizer.fit_transform(x_train['question'])
x_test_multilabel = vectorizer.transform(x_test['question'])
print("Time taken to run this cell :", datetime.now() - start)
```

Time taken to run this cell : 0:02:20.073092

In [69]:

```
y_test = y_test[:100000]
```

In [70]:

```
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: (200000, 98488) Y : (200000, 500)
 Dimensions of test data X: (100000, 98488) Y: (100000, 500)

4.5.3 Applying Logistic Regression with OneVsRest Classifier

In [71]:

```
start = datetime.now()
classifier = OneVsRestClassifier(SGDClassifier(loss='log', alpha=0.00001, penalty='l1'), n_jobs=-1)
classifier.fit(x_train_multilabel, y_train)
predictions = classifier.predict(x_test_multilabel)

print("Accuracy :", metrics.accuracy_score(y_test, predictions))
print("Hamming loss ", metrics.hamming_loss(y_test, predictions))

precision = precision_score(y_test, predictions, average='micro')
recall = recall_score(y_test, predictions, average='micro')
f1 = f1_score(y_test, predictions, average='micro')

print("Micro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1))

precision = precision_score(y_test, predictions, average='macro')
recall = recall_score(y_test, predictions, average='macro')
f1 = f1_score(y_test, predictions, average='macro')

print("Macro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1))

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

Accuracy : 0.05216

Hamming loss 0.00505718

Micro-average quality numbers

Precision: 0.0204, Recall: 0.0088, F1-measure: 0.0123

Macro-average quality numbers

Precision: 0.0036, Recall: 0.0015, F1-measure: 0.0019

	precision	recall	f1-score	support
0	0.05	0.04	0.04	4633
1	0.09	0.03	0.04	7549
2	0.08	0.03	0.04	7112
3	0.04	0.02	0.03	2888
4	0.05	0.03	0.04	5009
5	0.04	0.01	0.02	3919
6	0.06	0.03	0.04	5204
7	0.03	0.03	0.03	3229
8	0.03	0.01	0.01	3097
9	0.02	0.02	0.02	1829
10	0.02	0.01	0.01	2157
11	0.03	0.01	0.02	3142
12	0.03	0.01	0.02	2575
13	0.01	0.00	0.01	1431
14	0.02	0.01	0.01	2649
15	0.02	0.01	0.01	2094
16	0.02	0.00	0.01	1549
17	0.02	0.01	0.02	1869
18	0.02	0.01	0.01	2268
19	0.01	0.00	0.00	852
20	0.02	0.01	0.02	1614
21	0.03	0.01	0.01	1798
22	0.01	0.01	0.01	1350
23	0.01	0.01	0.01	888
24	0.01	0.00	0.00	1528
25	0.00	0.00	0.00	388
26	0.00	0.00	0.00	717
27	0.01	0.01	0.01	986
28	0.01	0.00	0.00	1046
29	0.04	0.00	0.01	2380
30	0.00	0.00	0.00	695
31	0.01	0.00	0.00	1095
32	0.04	0.00	0.00	2896
33	0.01	0.01	0.01	652
34	0.01	0.00	0.00	929
35	0.01	0.01	0.01	632
36	0.00	0.00	0.00	535
37	0.00	0.00	0.00	172
38	0.01	0.01	0.01	728
39	0.01	0.01	0.01	632
40	0.01	0.01	0.01	964
41	0.00	0.00	0.00	579
42	0.01	0.00	0.00	480
43	0.00	0.00	0.00	2408
44	0.00	0.00	0.00	719
45	0.01	0.01	0.01	455
46	0.01	0.01	0.01	385
47	0.02	0.00	0.01	517
48	0.01	0.00	0.00	516
49	0.01	0.01	0.01	531
50	0.00	0.00	0.00	410
51	0.00	0.00	0.00	790
52	0.00	0.00	0.00	151

53	0.00	0.00	0.00	702
54	0.00	0.00	0.00	480
55	0.00	0.00	0.00	423
56	0.00	0.00	0.00	485
57	0.00	0.00	0.00	848
58	0.00	0.00	0.00	473
59	0.00	0.00	0.00	640
60	0.01	0.00	0.00	499
61	0.00	0.00	0.00	234
62	0.01	0.01	0.01	546
63	0.01	0.01	0.01	492
64	0.00	0.00	0.00	138
65	0.00	0.00	0.00	350
66	0.00	0.00	0.00	233
67	0.00	0.00	0.00	203
68	0.00	0.00	0.00	401
69	0.00	0.00	0.00	283
70	0.00	0.00	0.00	444
71	0.00	0.00	0.00	839
72	0.00	0.00	0.00	281
73	0.00	0.00	0.00	1622
74	0.00	0.00	0.00	716
75	0.00	0.00	0.00	127
76	0.00	0.00	0.00	413
77	0.00	0.00	0.00	296
78	0.01	0.01	0.01	264
79	0.00	0.00	0.00	469
80	0.00	0.00	0.00	229
81	0.00	0.00	0.00	416
82	0.00	0.00	0.00	403
83	0.00	0.00	0.00	307
84	0.00	0.00	0.00	224
85	0.00	0.00	0.00	353
86	0.00	0.00	0.00	383
87	0.01	0.00	0.00	260
88	0.00	0.00	0.00	237
89	0.00	0.00	0.00	107
90	0.01	0.00	0.01	490
91	0.01	0.00	0.00	325
92	0.01	0.01	0.01	525
93	0.01	0.00	0.00	263
94	0.00	0.00	0.00	111
95	0.00	0.00	0.00	283
96	0.00	0.00	0.00	205
97	0.00	0.00	0.00	284
98	0.00	0.00	0.00	129
99	0.02	0.01	0.01	568
100	0.00	0.00	0.00	374
101	0.01	0.00	0.00	854
102	0.00	0.00	0.00	360
103	0.00	0.00	0.00	101
104	0.01	0.00	0.01	238
105	0.00	0.00	0.00	205
106	0.02	0.00	0.00	375
107	0.00	0.00	0.00	219
108	0.01	0.00	0.01	255
109	0.00	0.00	0.00	352
110	0.00	0.00	0.00	353
111	0.00	0.00	0.00	224
112	0.00	0.00	0.00	144
113	0.01	0.01	0.01	199

114	0.00	0.00	0.00	237
115	0.00	0.00	0.00	273
116	0.00	0.00	0.00	255
117	0.01	0.00	0.00	606
118	0.00	0.00	0.00	238
119	0.00	0.00	0.00	147
120	0.00	0.00	0.00	374
121	0.00	0.00	0.00	292
122	0.00	0.00	0.00	74
123	0.00	0.00	0.00	232
124	0.03	0.01	0.01	905
125	0.00	0.00	0.00	306
126	0.00	0.00	0.00	153
127	0.00	0.00	0.00	457
128	0.00	0.00	0.00	225
129	0.00	0.00	0.00	196
130	0.00	0.00	0.00	87
131	0.00	0.00	0.00	116
132	0.00	0.00	0.00	228
133	0.00	0.00	0.00	278
134	0.00	0.00	0.00	745
135	0.00	0.00	0.00	213
136	0.00	0.00	0.00	427
137	0.01	0.00	0.00	405
138	0.00	0.00	0.00	164
139	0.00	0.00	0.00	216
140	0.00	0.00	0.00	292
141	0.00	0.00	0.00	126
142	0.00	0.00	0.00	119
143	0.00	0.00	0.00	408
144	0.00	0.00	0.00	629
145	0.00	0.00	0.00	866
146	0.00	0.00	0.00	228
147	0.00	0.00	0.00	100
148	0.00	0.00	0.00	265
149	0.00	0.00	0.00	280
150	0.00	0.00	0.00	169
151	0.00	0.00	0.00	32
152	0.00	0.00	0.00	259
153	0.00	0.00	0.00	220
154	0.00	0.00	0.00	369
155	0.00	0.00	0.00	191
156	0.00	0.00	0.00	81
157	0.00	0.00	0.00	216
158	0.00	0.00	0.00	69
159	0.00	0.00	0.00	127
160	0.00	0.00	0.00	201
161	0.01	0.01	0.01	172
162	0.00	0.00	0.00	271
163	0.00	0.00	0.00	207
164	0.01	0.00	0.01	234
165	0.00	0.00	0.00	363
166	0.00	0.00	0.00	480
167	0.00	0.00	0.00	234
168	0.00	0.00	0.00	224
169	0.00	0.00	0.00	238
170	0.00	0.00	0.00	56
171	0.00	0.00	0.00	824
172	0.01	0.00	0.01	219
173	0.01	0.01	0.01	206
174	0.00	0.00	0.00	240

175	0.00	0.00	0.00	79
176	0.00	0.00	0.00	254
177	0.00	0.00	0.00	305
178	0.00	0.00	0.00	343
179	0.00	0.00	0.00	86
180	0.00	0.00	0.00	113
181	0.00	0.00	0.00	530
182	0.00	0.00	0.00	160
183	0.01	0.01	0.01	101
184	0.00	0.00	0.00	172
185	0.00	0.00	0.00	119
186	0.00	0.00	0.00	228
187	0.01	0.00	0.01	401
188	0.00	0.00	0.00	208
189	0.00	0.00	0.00	186
190	0.01	0.00	0.01	201
191	0.00	0.00	0.00	270
192	0.01	0.00	0.01	222
193	0.00	0.00	0.00	69
194	0.01	0.00	0.01	229
195	0.00	0.00	0.00	108
196	0.00	0.00	0.00	332
197	0.00	0.00	0.00	209
198	0.00	0.00	0.00	289
199	0.00	0.00	0.00	199
200	0.00	0.00	0.00	153
201	0.00	0.00	0.00	225
202	0.00	0.00	0.00	208
203	0.00	0.00	0.00	189
204	0.00	0.00	0.00	345
205	0.01	0.03	0.01	78
206	0.00	0.00	0.00	88
207	0.00	0.00	0.00	159
208	0.01	0.01	0.01	187
209	0.00	0.00	0.00	364
210	0.00	0.00	0.00	136
211	0.00	0.00	0.00	183
212	0.00	0.00	0.00	246
213	0.00	0.00	0.00	192
214	0.00	0.00	0.00	76
215	0.00	0.00	0.00	125
216	0.00	0.00	0.00	158
217	0.00	0.00	0.00	42
218	0.00	0.00	0.00	207
219	0.00	0.00	0.00	129
220	0.01	0.01	0.01	191
221	0.00	0.00	0.00	208
222	0.00	0.00	0.00	180
223	0.07	0.00	0.01	211
224	0.00	0.00	0.00	76
225	0.00	0.00	0.00	175
226	0.00	0.00	0.00	199
227	0.00	0.00	0.00	156
228	0.00	0.00	0.00	293
229	0.00	0.00	0.00	227
230	0.00	0.00	0.00	308
231	0.00	0.00	0.00	311
232	0.00	0.00	0.00	186
233	0.00	0.00	0.00	177
234	0.00	0.00	0.00	167
235	0.00	0.00	0.00	38

236	0.00	0.00	0.00	62
237	0.00	0.00	0.00	256
238	0.00	0.00	0.00	118
239	0.00	0.00	0.00	86
240	0.00	0.00	0.00	595
241	0.00	0.00	0.00	118
242	0.00	0.00	0.00	30
243	0.00	0.00	0.00	248
244	0.00	0.00	0.00	143
245	0.00	0.00	0.00	193
246	0.00	0.00	0.00	100
247	0.00	0.00	0.00	109
248	0.00	0.00	0.00	559
249	0.00	0.00	0.00	142
250	0.00	0.00	0.00	109
251	0.00	0.00	0.00	119
252	0.00	0.00	0.00	154
253	0.00	0.00	0.00	103
254	0.00	0.00	0.00	152
255	0.00	0.00	0.00	99
256	0.00	0.00	0.00	384
257	0.00	0.00	0.00	172
258	0.00	0.00	0.00	201
259	0.00	0.00	0.00	141
260	0.00	0.00	0.00	87
261	0.00	0.00	0.00	153
262	0.00	0.00	0.00	82
263	0.00	0.00	0.00	49
264	0.00	0.00	0.00	142
265	0.00	0.00	0.00	66
266	0.00	0.00	0.00	52
267	0.00	0.00	0.00	71
268	0.00	0.00	0.00	152
269	0.00	0.00	0.00	108
270	0.01	0.01	0.01	140
271	0.00	0.00	0.00	138
272	0.00	0.00	0.00	160
273	0.00	0.00	0.00	159
274	0.00	0.00	0.00	87
275	0.00	0.00	0.00	183
276	0.00	0.00	0.00	370
277	0.00	0.00	0.00	171
278	0.00	0.00	0.00	138
279	0.00	0.00	0.00	170
280	0.00	0.00	0.00	99
281	0.00	0.00	0.00	111
282	0.00	0.00	0.00	114
283	0.00	0.00	0.00	186
284	0.00	0.00	0.00	441
285	0.00	0.00	0.00	169
286	0.01	0.01	0.01	157
287	0.00	0.00	0.00	115
288	0.00	0.00	0.00	80
289	0.00	0.00	0.00	319
290	0.00	0.00	0.00	114
291	0.00	0.00	0.00	224
292	0.00	0.00	0.00	259
293	0.00	0.00	0.00	79
294	0.00	0.00	0.00	95
295	0.00	0.00	0.00	62
296	0.00	0.00	0.00	65

297	0.00	0.00	0.00	113
298	0.00	0.00	0.00	129
299	0.00	0.00	0.00	99
300	0.00	0.00	0.00	127
301	0.00	0.00	0.00	167
302	0.00	0.00	0.00	116
303	0.00	0.00	0.00	111
304	0.00	0.00	0.00	192
305	0.00	0.00	0.00	107
306	0.00	0.00	0.00	72
307	0.00	0.00	0.00	186
308	0.00	0.00	0.00	156
309	0.00	0.00	0.00	81
310	0.00	0.00	0.00	160
311	0.00	0.00	0.00	50
312	0.00	0.00	0.00	279
313	0.00	0.00	0.00	86
314	0.00	0.00	0.00	139
315	0.00	0.00	0.00	106
316	0.00	0.00	0.00	62
317	0.00	0.00	0.00	267
318	0.00	0.00	0.00	124
319	0.00	0.00	0.00	120
320	0.00	0.00	0.00	99
321	0.00	0.00	0.00	199
322	0.00	0.00	0.00	217
323	0.01	0.01	0.01	109
324	0.00	0.00	0.00	107
325	0.00	0.00	0.00	111
326	0.00	0.00	0.00	97
327	0.12	0.01	0.02	93
328	0.00	0.00	0.00	149
329	0.00	0.00	0.00	126
330	0.00	0.00	0.00	81
331	0.00	0.00	0.00	144
332	0.00	0.00	0.00	95
333	0.00	0.00	0.00	126
334	0.00	0.00	0.00	174
335	0.00	0.00	0.00	63
336	0.00	0.00	0.00	115
337	0.00	0.00	0.00	40
338	0.00	0.00	0.00	95
339	0.00	0.00	0.00	216
340	0.00	0.00	0.00	110
341	0.00	0.00	0.00	84
342	0.00	0.00	0.00	190
343	0.00	0.00	0.00	24
344	0.00	0.00	0.00	100
345	0.00	0.00	0.00	37
346	0.00	0.00	0.00	98
347	0.00	0.00	0.00	50
348	0.00	0.00	0.00	200
349	0.00	0.00	0.00	112
350	0.00	0.00	0.00	16
351	0.00	0.00	0.00	117
352	0.00	0.00	0.00	143
353	0.00	0.00	0.00	34
354	0.00	0.00	0.00	202
355	0.00	0.00	0.00	134
356	0.00	0.00	0.00	107
357	0.00	0.00	0.00	91

358	0.00	0.00	0.00	68
359	0.00	0.00	0.00	161
360	0.00	0.00	0.00	120
361	0.00	0.00	0.00	115
362	0.00	0.00	0.00	91
363	0.00	0.00	0.00	143
364	0.00	0.00	0.00	112
365	0.00	0.00	0.00	47
366	0.00	0.00	0.00	439
367	0.12	0.00	0.01	298
368	0.00	0.00	0.00	160
369	0.00	0.00	0.00	97
370	0.00	0.00	0.00	101
371	0.00	0.00	0.00	95
372	0.00	0.00	0.00	50
373	0.00	0.00	0.00	141
374	0.00	0.00	0.00	138
375	0.00	0.00	0.00	83
376	0.00	0.00	0.00	49
377	0.00	0.00	0.00	232
378	0.00	0.00	0.00	143
379	0.00	0.00	0.00	23
380	0.00	0.00	0.00	75
381	0.00	0.00	0.00	87
382	0.00	0.00	0.00	71
383	0.00	0.00	0.00	96
384	0.00	0.00	0.00	122
385	0.00	0.00	0.00	41
386	0.00	0.00	0.00	91
387	0.00	0.00	0.00	121
388	0.00	0.00	0.00	80
389	0.00	0.00	0.00	142
390	0.00	0.00	0.00	134
391	0.00	0.00	0.00	82
392	0.00	0.00	0.00	14
393	0.00	0.00	0.00	121
394	0.00	0.00	0.00	124
395	0.00	0.00	0.00	95
396	0.00	0.00	0.00	142
397	0.00	0.00	0.00	112
398	0.00	0.00	0.00	114
399	0.00	0.00	0.00	189
400	0.00	0.00	0.00	75
401	0.00	0.00	0.00	123
402	0.00	0.00	0.00	128
403	0.00	0.00	0.00	69
404	0.00	0.00	0.00	14
405	0.00	0.00	0.00	120
406	0.00	0.00	0.00	85
407	0.00	0.00	0.00	92
408	0.00	0.00	0.00	70
409	0.00	0.00	0.00	47
410	0.00	0.00	0.00	24
411	0.00	0.00	0.00	88
412	0.00	0.00	0.00	136
413	0.00	0.00	0.00	34
414	0.00	0.00	0.00	67
415	0.00	0.00	0.00	37
416	0.00	0.00	0.00	81
417	0.00	0.00	0.00	112
418	0.00	0.00	0.00	81

419	0.00	0.00	0.00	89
420	0.00	0.00	0.00	27
421	0.00	0.00	0.00	70
422	0.00	0.00	0.00	89
423	0.00	0.00	0.00	97
424	0.00	0.00	0.00	10
425	0.00	0.00	0.00	16
426	0.00	0.00	0.00	85
427	0.00	0.00	0.00	108
428	0.00	0.00	0.00	18
429	0.00	0.00	0.00	53
430	0.00	0.00	0.00	88
431	0.00	0.00	0.00	115
432	0.00	0.00	0.00	20
433	0.00	0.00	0.00	42
434	0.00	0.00	0.00	316
435	0.00	0.00	0.00	43
436	0.00	0.00	0.00	67
437	0.00	0.00	0.00	214
438	0.00	0.00	0.00	110
439	0.00	0.00	0.00	162
440	0.00	0.00	0.00	63
441	0.00	0.00	0.00	83
442	0.00	0.00	0.00	83
443	0.00	0.00	0.00	53
444	0.00	0.00	0.00	40
445	0.00	0.00	0.00	87
446	0.00	0.00	0.00	44
447	0.00	0.00	0.00	56
448	0.00	0.00	0.00	68
449	0.00	0.00	0.00	59
450	0.00	0.00	0.00	113
451	0.00	0.00	0.00	51
452	0.00	0.00	0.00	66
453	0.00	0.00	0.00	31
454	0.00	0.00	0.00	30
455	0.00	0.00	0.00	139
456	0.00	0.00	0.00	26
457	0.00	0.00	0.00	77
458	0.00	0.00	0.00	78
459	0.00	0.00	0.00	117
460	0.00	0.00	0.00	83
461	0.00	0.00	0.00	74
462	0.00	0.00	0.00	83
463	0.00	0.00	0.00	86
464	0.02	0.01	0.01	76
465	0.00	0.00	0.00	123
466	0.01	0.01	0.01	92
467	0.00	0.00	0.00	90
468	0.00	0.00	0.00	135
469	0.00	0.00	0.00	80
470	0.00	0.00	0.00	94
471	0.00	0.00	0.00	62
472	0.00	0.00	0.00	90
473	0.00	0.00	0.00	69
474	0.00	0.00	0.00	84
475	0.00	0.00	0.00	100
476	0.00	0.00	0.00	85
477	0.00	0.00	0.00	79
478	0.00	0.00	0.00	94
479	0.00	0.00	0.00	59

480	0.02	0.01	0.01	132
481	0.00	0.00	0.00	247
482	0.00	0.00	0.00	75
483	0.00	0.00	0.00	62
484	0.00	0.00	0.00	33
485	0.00	0.00	0.00	54
486	0.08	0.01	0.01	170
487	0.00	0.00	0.00	85
488	0.00	0.00	0.00	105
489	0.00	0.00	0.00	116
490	0.00	0.00	0.00	351
491	0.00	0.00	0.00	48
492	0.00	0.00	0.00	65
493	0.00	0.00	0.00	74
494	0.00	0.00	0.00	55
495	0.00	0.00	0.00	22
496	0.00	0.00	0.00	80
497	0.00	0.00	0.00	63
498	0.00	0.00	0.00	67
499	0.00	0.00	0.00	121
micro avg	0.02	0.01	0.01	178469
macro avg	0.00	0.00	0.00	178469
weighted avg	0.02	0.01	0.01	178469
samples avg	0.01	0.01	0.01	178469

Time taken to run this cell : 0:02:17.153715

In [72]:

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

Out[72]:

```
['lr_with_more_title_weight.pkl']
```


In [73]:

```
start = datetime.now()
classifier_2 = OneVsRestClassifier(LogisticRegression(penalty='l2'), 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 : 0.06345

Hamming loss 0.00472184

Micro-average quality numbers

Precision: 0.0240, Recall: 0.0081, F1-measure: 0.0122

Macro-average quality numbers

Precision: 0.0039, Recall: 0.0012, F1-measure: 0.0016

	precision	recall	f1-score	support
0	0.05	0.04	0.04	4633
1	0.09	0.03	0.04	7549
2	0.08	0.03	0.04	7112
3	0.04	0.02	0.03	2888
4	0.05	0.03	0.04	5009
5	0.04	0.01	0.02	3919
6	0.06	0.03	0.04	5204
7	0.03	0.03	0.03	3229
8	0.03	0.01	0.01	3097
9	0.02	0.01	0.01	1829
10	0.02	0.01	0.01	2157
11	0.03	0.01	0.01	3142
12	0.03	0.01	0.01	2575
13	0.02	0.00	0.01	1431
14	0.02	0.01	0.02	2649
15	0.03	0.01	0.01	2094
16	0.02	0.00	0.01	1549
17	0.02	0.01	0.02	1869
18	0.02	0.01	0.01	2268
19	0.00	0.00	0.00	852
20	0.02	0.01	0.01	1614
21	0.04	0.01	0.01	1798
22	0.01	0.01	0.01	1350
23	0.01	0.01	0.01	888
24	0.01	0.00	0.00	1528
25	0.00	0.00	0.00	388
26	0.00	0.00	0.00	717
27	0.01	0.01	0.01	986
28	0.01	0.00	0.01	1046
29	0.05	0.00	0.00	2380
30	0.00	0.00	0.00	695
31	0.01	0.00	0.00	1095
32	0.01	0.00	0.00	2896
33	0.00	0.00	0.00	652
34	0.01	0.00	0.00	929
35	0.01	0.01	0.01	632
36	0.00	0.00	0.00	535
37	0.00	0.00	0.00	172
38	0.01	0.01	0.01	728
39	0.02	0.01	0.01	632
40	0.01	0.00	0.00	964
41	0.00	0.00	0.00	579
42	0.01	0.00	0.00	480
43	0.00	0.00	0.00	2408
44	0.00	0.00	0.00	719
45	0.01	0.01	0.01	455
46	0.01	0.01	0.01	385
47	0.02	0.00	0.01	517
48	0.01	0.00	0.00	516
49	0.01	0.01	0.01	531
50	0.00	0.00	0.00	410
51	0.00	0.00	0.00	790
52	0.00	0.00	0.00	151

53	0.00	0.00	0.00	702
54	0.00	0.00	0.00	480
55	0.00	0.00	0.00	423
56	0.00	0.00	0.00	485
57	0.00	0.00	0.00	848
58	0.00	0.00	0.00	473
59	0.00	0.00	0.00	640
60	0.01	0.00	0.00	499
61	0.00	0.00	0.00	234
62	0.01	0.01	0.01	546
63	0.01	0.01	0.01	492
64	0.00	0.00	0.00	138
65	0.00	0.00	0.00	350
66	0.00	0.00	0.00	233
67	0.00	0.00	0.00	203
68	0.00	0.00	0.00	401
69	0.00	0.00	0.00	283
70	0.00	0.00	0.00	444
71	0.00	0.00	0.00	839
72	0.00	0.00	0.00	281
73	0.00	0.00	0.00	1622
74	0.00	0.00	0.00	716
75	0.00	0.00	0.00	127
76	0.00	0.00	0.00	413
77	0.00	0.00	0.00	296
78	0.01	0.01	0.01	264
79	0.00	0.00	0.00	469
80	0.00	0.00	0.00	229
81	0.00	0.00	0.00	416
82	0.00	0.00	0.00	403
83	0.00	0.00	0.00	307
84	0.00	0.00	0.00	224
85	0.01	0.00	0.00	353
86	0.00	0.00	0.00	383
87	0.00	0.00	0.00	260
88	0.00	0.00	0.00	237
89	0.00	0.00	0.00	107
90	0.01	0.00	0.00	490
91	0.01	0.00	0.00	325
92	0.01	0.01	0.01	525
93	0.01	0.00	0.01	263
94	0.00	0.00	0.00	111
95	0.00	0.00	0.00	283
96	0.00	0.00	0.00	205
97	0.00	0.00	0.00	284
98	0.00	0.00	0.00	129
99	0.01	0.00	0.00	568
100	0.00	0.00	0.00	374
101	0.01	0.00	0.00	854
102	0.00	0.00	0.00	360
103	0.00	0.00	0.00	101
104	0.02	0.00	0.01	238
105	0.00	0.00	0.00	205
106	0.05	0.00	0.01	375
107	0.00	0.00	0.00	219
108	0.00	0.00	0.00	255
109	0.00	0.00	0.00	352
110	0.00	0.00	0.00	353
111	0.00	0.00	0.00	224
112	0.00	0.00	0.00	144
113	0.00	0.00	0.00	199

114	0.00	0.00	0.00	237
115	0.01	0.00	0.01	273
116	0.00	0.00	0.00	255
117	0.00	0.00	0.00	606
118	0.00	0.00	0.00	238
119	0.00	0.00	0.00	147
120	0.00	0.00	0.00	374
121	0.00	0.00	0.00	292
122	0.00	0.00	0.00	74
123	0.00	0.00	0.00	232
124	0.05	0.01	0.01	905
125	0.00	0.00	0.00	306
126	0.00	0.00	0.00	153
127	0.00	0.00	0.00	457
128	0.00	0.00	0.00	225
129	0.00	0.00	0.00	196
130	0.00	0.00	0.00	87
131	0.00	0.00	0.00	116
132	0.00	0.00	0.00	228
133	0.01	0.00	0.00	278
134	0.00	0.00	0.00	745
135	0.00	0.00	0.00	213
136	0.00	0.00	0.00	427
137	0.01	0.00	0.00	405
138	0.00	0.00	0.00	164
139	0.00	0.00	0.00	216
140	0.00	0.00	0.00	292
141	0.00	0.00	0.00	126
142	0.00	0.00	0.00	119
143	0.00	0.00	0.00	408
144	0.00	0.00	0.00	629
145	0.00	0.00	0.00	866
146	0.00	0.00	0.00	228
147	0.00	0.00	0.00	100
148	0.00	0.00	0.00	265
149	0.00	0.00	0.00	280
150	0.00	0.00	0.00	169
151	0.00	0.00	0.00	32
152	0.00	0.00	0.00	259
153	0.00	0.00	0.00	220
154	0.00	0.00	0.00	369
155	0.00	0.00	0.00	191
156	0.00	0.00	0.00	81
157	0.00	0.00	0.00	216
158	0.00	0.00	0.00	69
159	0.00	0.00	0.00	127
160	0.00	0.00	0.00	201
161	0.00	0.00	0.00	172
162	0.00	0.00	0.00	271
163	0.00	0.00	0.00	207
164	0.01	0.00	0.01	234
165	0.00	0.00	0.00	363
166	0.00	0.00	0.00	480
167	0.00	0.00	0.00	234
168	0.00	0.00	0.00	224
169	0.00	0.00	0.00	238
170	0.00	0.00	0.00	56
171	0.00	0.00	0.00	824
172	0.00	0.00	0.00	219
173	0.00	0.00	0.00	206
174	0.00	0.00	0.00	240

175	0.00	0.00	0.00	79
176	0.00	0.00	0.00	254
177	0.00	0.00	0.00	305
178	0.00	0.00	0.00	343
179	0.00	0.00	0.00	86
180	0.00	0.00	0.00	113
181	0.00	0.00	0.00	530
182	0.00	0.00	0.00	160
183	0.01	0.01	0.01	101
184	0.00	0.00	0.00	172
185	0.00	0.00	0.00	119
186	0.00	0.00	0.00	228
187	0.01	0.00	0.01	401
188	0.00	0.00	0.00	208
189	0.00	0.00	0.00	186
190	0.01	0.00	0.01	201
191	0.00	0.00	0.00	270
192	0.00	0.00	0.00	222
193	0.00	0.00	0.00	69
194	0.01	0.00	0.01	229
195	0.00	0.00	0.00	108
196	0.00	0.00	0.00	332
197	0.00	0.00	0.00	209
198	0.00	0.00	0.00	289
199	0.00	0.00	0.00	199
200	0.00	0.00	0.00	153
201	0.00	0.00	0.00	225
202	0.00	0.00	0.00	208
203	0.00	0.00	0.00	189
204	0.00	0.00	0.00	345
205	0.01	0.01	0.01	78
206	0.00	0.00	0.00	88
207	0.00	0.00	0.00	159
208	0.01	0.01	0.01	187
209	0.00	0.00	0.00	364
210	0.00	0.00	0.00	136
211	0.00	0.00	0.00	183
212	0.00	0.00	0.00	246
213	0.00	0.00	0.00	192
214	0.00	0.00	0.00	76
215	0.00	0.00	0.00	125
216	0.00	0.00	0.00	158
217	0.00	0.00	0.00	42
218	0.00	0.00	0.00	207
219	0.00	0.00	0.00	129
220	0.01	0.01	0.01	191
221	0.00	0.00	0.00	208
222	0.00	0.00	0.00	180
223	0.00	0.00	0.00	211
224	0.00	0.00	0.00	76
225	0.00	0.00	0.00	175
226	0.00	0.00	0.00	199
227	0.00	0.00	0.00	156
228	0.00	0.00	0.00	293
229	0.00	0.00	0.00	227
230	0.00	0.00	0.00	308
231	0.00	0.00	0.00	311
232	0.00	0.00	0.00	186
233	0.00	0.00	0.00	177
234	0.00	0.00	0.00	167
235	0.00	0.00	0.00	38

236	0.00	0.00	0.00	62
237	0.00	0.00	0.00	256
238	0.00	0.00	0.00	118
239	0.00	0.00	0.00	86
240	0.00	0.00	0.00	595
241	0.00	0.00	0.00	118
242	0.00	0.00	0.00	30
243	0.00	0.00	0.00	248
244	0.00	0.00	0.00	143
245	0.00	0.00	0.00	193
246	0.00	0.00	0.00	100
247	0.00	0.00	0.00	109
248	0.00	0.00	0.00	559
249	0.00	0.00	0.00	142
250	0.00	0.00	0.00	109
251	0.00	0.00	0.00	119
252	0.00	0.00	0.00	154
253	0.00	0.00	0.00	103
254	0.00	0.00	0.00	152
255	0.00	0.00	0.00	99
256	0.00	0.00	0.00	384
257	0.00	0.00	0.00	172
258	0.00	0.00	0.00	201
259	0.00	0.00	0.00	141
260	0.00	0.00	0.00	87
261	0.00	0.00	0.00	153
262	0.00	0.00	0.00	82
263	0.00	0.00	0.00	49
264	0.00	0.00	0.00	142
265	0.00	0.00	0.00	66
266	0.00	0.00	0.00	52
267	0.00	0.00	0.00	71
268	0.00	0.00	0.00	152
269	0.00	0.00	0.00	108
270	0.01	0.01	0.01	140
271	0.00	0.00	0.00	138
272	0.00	0.00	0.00	160
273	0.00	0.00	0.00	159
274	0.00	0.00	0.00	87
275	0.00	0.00	0.00	183
276	0.00	0.00	0.00	370
277	0.00	0.00	0.00	171
278	0.00	0.00	0.00	138
279	0.00	0.00	0.00	170
280	0.00	0.00	0.00	99
281	0.00	0.00	0.00	111
282	0.00	0.00	0.00	114
283	0.00	0.00	0.00	186
284	0.00	0.00	0.00	441
285	0.00	0.00	0.00	169
286	0.00	0.00	0.00	157
287	0.00	0.00	0.00	115
288	0.00	0.00	0.00	80
289	0.00	0.00	0.00	319
290	0.00	0.00	0.00	114
291	0.00	0.00	0.00	224
292	0.00	0.00	0.00	259
293	0.00	0.00	0.00	79
294	0.00	0.00	0.00	95
295	0.00	0.00	0.00	62
296	0.00	0.00	0.00	65

297	0.00	0.00	0.00	113
298	0.00	0.00	0.00	129
299	0.00	0.00	0.00	99
300	0.00	0.00	0.00	127
301	0.00	0.00	0.00	167
302	0.00	0.00	0.00	116
303	0.00	0.00	0.00	111
304	0.00	0.00	0.00	192
305	0.00	0.00	0.00	107
306	0.00	0.00	0.00	72
307	0.00	0.00	0.00	186
308	0.00	0.00	0.00	156
309	0.00	0.00	0.00	81
310	0.00	0.00	0.00	160
311	0.00	0.00	0.00	50
312	0.00	0.00	0.00	279
313	0.00	0.00	0.00	86
314	0.00	0.00	0.00	139
315	0.00	0.00	0.00	106
316	0.00	0.00	0.00	62
317	0.00	0.00	0.00	267
318	0.00	0.00	0.00	124
319	0.00	0.00	0.00	120
320	0.00	0.00	0.00	99
321	0.00	0.00	0.00	199
322	0.00	0.00	0.00	217
323	0.01	0.01	0.01	109
324	0.00	0.00	0.00	107
325	0.00	0.00	0.00	111
326	0.00	0.00	0.00	97
327	0.00	0.00	0.00	93
328	0.00	0.00	0.00	149
329	0.00	0.00	0.00	126
330	0.00	0.00	0.00	81
331	0.00	0.00	0.00	144
332	0.00	0.00	0.00	95
333	0.00	0.00	0.00	126
334	0.00	0.00	0.00	174
335	0.00	0.00	0.00	63
336	0.00	0.00	0.00	115
337	0.00	0.00	0.00	40
338	0.00	0.00	0.00	95
339	0.00	0.00	0.00	216
340	0.00	0.00	0.00	110
341	0.00	0.00	0.00	84
342	0.00	0.00	0.00	190
343	0.00	0.00	0.00	24
344	0.00	0.00	0.00	100
345	0.00	0.00	0.00	37
346	0.00	0.00	0.00	98
347	0.00	0.00	0.00	50
348	0.00	0.00	0.00	200
349	0.00	0.00	0.00	112
350	0.00	0.00	0.00	16
351	0.00	0.00	0.00	117
352	0.00	0.00	0.00	143
353	0.00	0.00	0.00	34
354	0.00	0.00	0.00	202
355	0.00	0.00	0.00	134
356	0.00	0.00	0.00	107
357	0.00	0.00	0.00	91

358	0.00	0.00	0.00	68
359	0.00	0.00	0.00	161
360	0.00	0.00	0.00	120
361	0.00	0.00	0.00	115
362	0.00	0.00	0.00	91
363	0.00	0.00	0.00	143
364	0.00	0.00	0.00	112
365	0.00	0.00	0.00	47
366	0.00	0.00	0.00	439
367	0.00	0.00	0.00	298
368	0.00	0.00	0.00	160
369	0.00	0.00	0.00	97
370	0.00	0.00	0.00	101
371	0.00	0.00	0.00	95
372	0.00	0.00	0.00	50
373	0.00	0.00	0.00	141
374	0.00	0.00	0.00	138
375	0.00	0.00	0.00	83
376	0.00	0.00	0.00	49
377	0.00	0.00	0.00	232
378	0.00	0.00	0.00	143
379	0.00	0.00	0.00	23
380	0.00	0.00	0.00	75
381	0.00	0.00	0.00	87
382	0.00	0.00	0.00	71
383	0.00	0.00	0.00	96
384	0.00	0.00	0.00	122
385	0.00	0.00	0.00	41
386	0.00	0.00	0.00	91
387	0.00	0.00	0.00	121
388	0.00	0.00	0.00	80
389	0.00	0.00	0.00	142
390	0.00	0.00	0.00	134
391	0.00	0.00	0.00	82
392	0.00	0.00	0.00	14
393	0.00	0.00	0.00	121
394	0.00	0.00	0.00	124
395	0.00	0.00	0.00	95
396	0.00	0.00	0.00	142
397	0.00	0.00	0.00	112
398	0.00	0.00	0.00	114
399	0.00	0.00	0.00	189
400	0.00	0.00	0.00	75
401	0.00	0.00	0.00	123
402	0.00	0.00	0.00	128
403	0.00	0.00	0.00	69
404	0.00	0.00	0.00	14
405	0.00	0.00	0.00	120
406	0.00	0.00	0.00	85
407	0.00	0.00	0.00	92
408	0.00	0.00	0.00	70
409	0.00	0.00	0.00	47
410	0.00	0.00	0.00	24
411	0.00	0.00	0.00	88
412	0.00	0.00	0.00	136
413	0.00	0.00	0.00	34
414	0.00	0.00	0.00	67
415	0.00	0.00	0.00	37
416	0.00	0.00	0.00	81
417	0.00	0.00	0.00	112
418	0.00	0.00	0.00	81

419	0.00	0.00	0.00	89
420	0.00	0.00	0.00	27
421	0.00	0.00	0.00	70
422	0.00	0.00	0.00	89
423	0.00	0.00	0.00	97
424	0.00	0.00	0.00	10
425	0.00	0.00	0.00	16
426	0.00	0.00	0.00	85
427	0.00	0.00	0.00	108
428	0.00	0.00	0.00	18
429	0.00	0.00	0.00	53
430	0.00	0.00	0.00	88
431	0.00	0.00	0.00	115
432	0.00	0.00	0.00	20
433	0.00	0.00	0.00	42
434	0.00	0.00	0.00	316
435	0.00	0.00	0.00	43
436	0.00	0.00	0.00	67
437	0.00	0.00	0.00	214
438	0.00	0.00	0.00	110
439	0.00	0.00	0.00	162
440	0.00	0.00	0.00	63
441	0.00	0.00	0.00	83
442	0.00	0.00	0.00	83
443	0.00	0.00	0.00	53
444	0.00	0.00	0.00	40
445	0.00	0.00	0.00	87
446	0.00	0.00	0.00	44
447	0.00	0.00	0.00	56
448	0.00	0.00	0.00	68
449	0.00	0.00	0.00	59
450	0.00	0.00	0.00	113
451	0.00	0.00	0.00	51
452	0.00	0.00	0.00	66
453	0.00	0.00	0.00	31
454	0.00	0.00	0.00	30
455	0.00	0.00	0.00	139
456	0.00	0.00	0.00	26
457	0.00	0.00	0.00	77
458	0.00	0.00	0.00	78
459	0.00	0.00	0.00	117
460	0.00	0.00	0.00	83
461	0.00	0.00	0.00	74
462	0.00	0.00	0.00	83
463	0.00	0.00	0.00	86
464	0.03	0.01	0.02	76
465	0.00	0.00	0.00	123
466	0.03	0.01	0.02	92
467	0.00	0.00	0.00	90
468	0.00	0.00	0.00	135
469	0.00	0.00	0.00	80
470	0.00	0.00	0.00	94
471	0.00	0.00	0.00	62
472	0.00	0.00	0.00	90
473	0.00	0.00	0.00	69
474	0.00	0.00	0.00	84
475	0.00	0.00	0.00	100
476	0.00	0.00	0.00	85
477	0.00	0.00	0.00	79
478	0.00	0.00	0.00	94
479	0.00	0.00	0.00	59

480	0.00	0.00	0.00	132
481	0.00	0.00	0.00	247
482	0.00	0.00	0.00	75
483	0.00	0.00	0.00	62
484	0.00	0.00	0.00	33
485	0.00	0.00	0.00	54
486	0.50	0.01	0.01	170
487	0.00	0.00	0.00	85
488	0.00	0.00	0.00	105
489	0.00	0.00	0.00	116
490	0.00	0.00	0.00	351
491	0.00	0.00	0.00	48
492	0.00	0.00	0.00	65
493	0.00	0.00	0.00	74
494	0.00	0.00	0.00	55
495	0.00	0.00	0.00	22
496	0.00	0.00	0.00	80
497	0.00	0.00	0.00	63
498	0.00	0.00	0.00	67
499	0.00	0.00	0.00	121
micro avg	0.02	0.01	0.01	178469
macro avg	0.00	0.00	0.00	178469
weighted avg	0.02	0.01	0.01	178469
samples avg	0.01	0.01	0.01	178469

Time taken to run this cell : 0:07:01.812820

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 [74]:

```
from scipy.sparse import hstack
from sklearn.decomposition import SparsePCA
```

In [75]:

```
%%time
#Bag Of Words upto 4 grams
vectorizer = CountVectorizer(min_df=10, ngram_range=(1,4))
# Bow on Tags
train_t_bow = vectorizer.fit_transform(x_train['tags'])
test_t_bow = vectorizer.transform(x_test['tags'])
# Merging the above features
x_train_wbow = hstack((x_train_multilabel,train_t_bow))
x_test_wbow = hstack((x_test_multilabel,test_t_bow))
```

CPU times: user 9.97 s, sys: 552 ms, total: 10.5 s
Wall time: 8.84 s

In [76]:

```
from scipy.sparse import csr
x_train_wbow, y_train = x_train_wbow.tocsr(), y_train.tocsr()
x_test_wbow, y_test = x_test_wbow.tocsr(), y_test.tocsr()
```

In [77]:

```
print(x_train_wbow.shape, y_train.shape)
print(x_test_wbow.shape, y_test.shape)
```

```
(200000, 112283) (200000, 500)
(100000, 112283) (100000, 500)
```

In [130]:

```
%%time
# Hyperparameter tuning onevsrest logistic regression classifier using Gridsearchcv
from sklearn.model_selection import ParameterGrid, GridSearchCV
score = 'f1_micro'
params = [{'estimator__alpha': [0.00001, 0.0001, 0.001, 0.01, 0.1, 1]}]
gs = GridSearchCV(classifier, params, scoring=score, n_jobs=-1, verbose=True, return_train_score=True)
gfit = gs.fit(x_train_wbow, y_train)
```

Fitting 5 folds for each of 6 candidates, totalling 30 fits

[Parallel(n_jobs=-1)]: Using backend LokyBackend with 10 concurrent workers.

[Parallel(n_jobs=-1)]: Done 30 out of 30 | elapsed: 44.3min finished

CPU times: user 16min 38s, sys: 12.2 s, total: 16min 51s

Wall time: 52min 32s

In [116]:

```
gs.best_estimator_
```

Out[116]:

```
OneVsRestClassifier(estimator=SGDClassifier(alpha=1e-05, average=False,
class_weight=None,
early_stopping=False, epsilon=0.1,
eta0=0.0, fit_intercept=True,
l1_ratio=0.15,
learning_rate='optimal', loss='log',
max_iter=1000, n_iter_no_change=5,
n_jobs=-1, penalty='l2',
power_t=0.5, random_state=None,
shuffle=True, tol=0.001,
validation_fraction=0.1, verbose=0,
warm_start=False),
n_jobs=None)
```

In [126]:

```
gs.cv_results_
```

Out[126]:

```

{'mean_fit_time': array([ 615.16039138,  632.76341586,  715.72279239, 10
65.77075591,
      1131.75176678,  702.47776361]),
 'std_fit_time': array([ 6.66956259,  4.09864426,  3.96914896, 41.074257
32, 35.74754305,
      3.37426937]),
 'mean_score_time': array([9.19909186,  9.88324418,  9.10949764,  9.1438131
8,  5.68389053,
      7.75838442]),
 'std_score_time': array([0.42258927,  0.3179997 ,  0.32863077,  0.3773540
8,  1.32091021,
      1.47708891]),
 'param_estimator__alpha': masked_array(data=[1e-05, 0.0001, 0.001, 0.0
1, 0.1, 1],
      mask=[False, False, False, False, False, False],
      fill_value='?',
      dtype=object),
 'params': [{'estimator__alpha': 1e-05},
 {'estimator__alpha': 0.0001},
 {'estimator__alpha': 0.001},
 {'estimator__alpha': 0.01},
 {'estimator__alpha': 0.1},
 {'estimator__alpha': 1}],
 'split0_test_score': array([0.92487413, 0.85466329, 0.5760799 , 0.30065
135, 0.10453158,
      0.05567565]),
 'split1_test_score': array([0.89911867, 0.83786172, 0.61279988, 0.34913
309, 0.03796946,
      0.02980274]),
 'split2_test_score': array([0.92990352, 0.86665431, 0.64338855, 0.37878
904, 0.13851549,
      0.09102765]),
 'split3_test_score': array([0.94394956, 0.88646231, 0.66010298, 0.42134
772, 0.20240208,
      0.17158429]),
 'split4_test_score': array([0.97608059, 0.90469973, 0.74650318, 0.68988
367, 0.40244037,
      0.20502417]),
 'mean_test_score': array([0.93478529, 0.87006827, 0.6477749 , 0.4279609
7, 0.1771718 ,
      0.1106229 ]),
 'std_test_score': array([0.02523274, 0.02347055, 0.05706474, 0.1367382
1, 0.12452365,
      0.06717584]),
 'rank_test_score': array([1, 2, 3, 4, 5, 6], dtype=int32),
 'split0_train_score': array([0.96347751, 0.90374913, 0.71235582, 0.5014
4437, 0.35407945,
      0.30640236]),
 'split1_train_score': array([0.96794756, 0.90710614, 0.70437692, 0.4728
7082, 0.29407699,
      0.28077086]),
 'split2_train_score': array([0.96158213, 0.90126703, 0.70851509, 0.4802
7492, 0.32259883,
      0.27092399]),
 'split3_train_score': array([0.96010181, 0.89865283, 0.70060647, 0.4697
2819, 0.31338297,
      0.25925409]),
 'split4_train_score': array([0.95239213, 0.88703837, 0.67459252, 0.4018
6501, 0.14241467,

```

```
    0.04471279]),  
    'mean_train_score': array([0.96110023, 0.8995627 , 0.70008936, 0.465236  
66, 0.28531058,  
    0.23241282]),  
    'std_train_score': array([0.00509112, 0.00685596, 0.01334315, 0.0335606  
, 0.07403205,  
    0.09512919])}}
```

In [118]:

```
%%time
from sklearn.metrics import *
# micro f1 using logistic regression(OvR)
classifier1 = OneVsRestClassifier(SGDClassifier(loss='log',alpha=1e-05,penalty='l2',n_jobs=-1))
classifier1.fit(x_train_wbow,y_train)
predictions_2 = classifier1.predict(x_test_wbow)
print("Accuracy :",accuracy_score(y_test, predictions_2))
print("Hamming loss ",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(classification_report(y_test, predictions_2))
print("Time taken to run this cell :", datetime.now() - start)
```

Accuracy : 0.01771

Hamming loss 0.00681732

Micro-average quality numbers

Precision: 0.0186, Recall: 0.0176, F1-measure: 0.0181

Macro-average quality numbers

Precision: 0.0034, Recall: 0.0033, F1-measure: 0.0031

	precision	recall	f1-score	support
0	0.05	0.06	0.05	4633
1	0.08	0.05	0.06	7549
2	0.07	0.07	0.07	7112
3	0.04	0.03	0.03	2888
4	0.05	0.07	0.06	5009
5	0.04	0.03	0.03	3919
6	0.05	0.05	0.05	5204
7	0.04	0.05	0.04	3229
8	0.04	0.04	0.04	3097
9	0.02	0.03	0.02	1829
10	0.02	0.03	0.02	2157
11	0.03	0.03	0.03	3142
12	0.03	0.03	0.03	2575
13	0.02	0.01	0.01	1431
14	0.02	0.02	0.02	2649
15	0.02	0.03	0.02	2094
16	0.01	0.01	0.01	1549
17	0.02	0.02	0.02	1869
18	0.02	0.02	0.02	2268
19	0.00	0.00	0.00	852
20	0.02	0.02	0.02	1614
21	0.02	0.02	0.02	1798
22	0.02	0.02	0.02	1350
23	0.01	0.01	0.01	888
24	0.03	0.01	0.01	1528
25	0.00	0.00	0.00	388
26	0.01	0.01	0.01	717
27	0.01	0.01	0.01	986
28	0.01	0.01	0.01	1046
29	0.02	0.00	0.00	2380
30	0.01	0.01	0.01	695
31	0.01	0.01	0.01	1095
32	0.03	0.00	0.01	2896
33	0.00	0.01	0.01	652
34	0.01	0.01	0.01	929
35	0.01	0.02	0.01	632
36	0.00	0.00	0.00	535
37	0.00	0.00	0.00	172
38	0.01	0.01	0.01	728
39	0.01	0.01	0.01	632
40	0.01	0.01	0.01	964
41	0.01	0.00	0.00	579
42	0.00	0.00	0.00	480
43	0.01	0.00	0.00	2408
44	0.00	0.00	0.00	719
45	0.01	0.01	0.01	455
46	0.00	0.01	0.00	385
47	0.00	0.00	0.00	517
48	0.00	0.00	0.00	516
49	0.01	0.01	0.01	531
50	0.01	0.01	0.01	410
51	0.01	0.00	0.00	790
52	0.00	0.00	0.00	151

53	0.01	0.01	0.01	702
54	0.00	0.01	0.00	480
55	0.00	0.00	0.00	423
56	0.01	0.01	0.01	485
57	0.01	0.01	0.01	848
58	0.01	0.01	0.01	473
59	0.00	0.00	0.00	640
60	0.00	0.00	0.00	499
61	0.00	0.00	0.00	234
62	0.01	0.01	0.01	546
63	0.01	0.01	0.01	492
64	0.01	0.01	0.01	138
65	0.01	0.01	0.01	350
66	0.00	0.00	0.00	233
67	0.00	0.00	0.00	203
68	0.00	0.00	0.00	401
69	0.00	0.00	0.00	283
70	0.01	0.01	0.01	444
71	0.00	0.00	0.00	839
72	0.00	0.00	0.00	281
73	0.00	0.00	0.00	1622
74	0.01	0.00	0.00	716
75	0.00	0.00	0.00	127
76	0.01	0.01	0.01	413
77	0.01	0.00	0.00	296
78	0.01	0.01	0.01	264
79	0.01	0.01	0.01	469
80	0.00	0.00	0.00	229
81	0.00	0.00	0.00	416
82	0.00	0.00	0.00	403
83	0.00	0.00	0.00	307
84	0.01	0.00	0.00	224
85	0.00	0.01	0.00	353
86	0.00	0.00	0.00	383
87	0.01	0.01	0.01	260
88	0.00	0.00	0.00	237
89	0.00	0.01	0.01	107
90	0.01	0.01	0.01	490
91	0.01	0.01	0.01	325
92	0.01	0.01	0.01	525
93	0.00	0.00	0.00	263
94	0.00	0.00	0.00	111
95	0.00	0.00	0.00	283
96	0.00	0.00	0.00	205
97	0.00	0.00	0.00	284
98	0.00	0.00	0.00	129
99	0.02	0.01	0.01	568
100	0.00	0.00	0.00	374
101	0.00	0.00	0.00	854
102	0.00	0.01	0.00	360
103	0.00	0.00	0.00	101
104	0.00	0.00	0.00	238
105	0.00	0.00	0.00	205
106	0.01	0.00	0.00	375
107	0.01	0.01	0.01	219
108	0.00	0.00	0.00	255
109	0.00	0.00	0.00	352
110	0.01	0.00	0.00	353
111	0.00	0.00	0.00	224
112	0.00	0.00	0.00	144
113	0.00	0.00	0.00	199

114	0.00	0.00	0.00	237
115	0.00	0.00	0.00	273
116	0.00	0.00	0.00	255
117	0.01	0.00	0.01	606
118	0.00	0.01	0.00	238
119	0.00	0.00	0.00	147
120	0.00	0.00	0.00	374
121	0.00	0.00	0.00	292
122	0.00	0.00	0.00	74
123	0.00	0.00	0.00	232
124	0.03	0.01	0.01	905
125	0.00	0.00	0.00	306
126	0.00	0.01	0.00	153
127	0.02	0.00	0.01	457
128	0.01	0.01	0.01	225
129	0.00	0.00	0.00	196
130	0.00	0.00	0.00	87
131	0.00	0.00	0.00	116
132	0.00	0.00	0.00	228
133	0.01	0.01	0.01	278
134	0.00	0.00	0.00	745
135	0.01	0.00	0.01	213
136	0.00	0.00	0.00	427
137	0.01	0.00	0.01	405
138	0.00	0.00	0.00	164
139	0.00	0.00	0.00	216
140	0.00	0.00	0.00	292
141	0.00	0.00	0.00	126
142	0.00	0.01	0.00	119
143	0.00	0.00	0.00	408
144	0.01	0.00	0.00	629
145	0.00	0.00	0.00	866
146	0.00	0.00	0.00	228
147	0.00	0.00	0.00	100
148	0.00	0.00	0.00	265
149	0.00	0.00	0.00	280
150	0.01	0.01	0.01	169
151	0.00	0.00	0.00	32
152	0.00	0.00	0.00	259
153	0.00	0.00	0.00	220
154	0.00	0.00	0.00	369
155	0.00	0.00	0.00	191
156	0.00	0.00	0.00	81
157	0.00	0.00	0.00	216
158	0.00	0.00	0.00	69
159	0.00	0.00	0.00	127
160	0.00	0.00	0.00	201
161	0.00	0.01	0.01	172
162	0.00	0.00	0.00	271
163	0.00	0.00	0.00	207
164	0.00	0.00	0.00	234
165	0.00	0.00	0.00	363
166	0.00	0.00	0.00	480
167	0.00	0.00	0.00	234
168	0.00	0.00	0.00	224
169	0.00	0.00	0.00	238
170	0.00	0.00	0.00	56
171	0.00	0.00	0.00	824
172	0.00	0.00	0.00	219
173	0.00	0.00	0.00	206
174	0.00	0.00	0.00	240

175	0.00	0.00	0.00	79
176	0.00	0.00	0.00	254
177	0.00	0.00	0.00	305
178	0.00	0.00	0.00	343
179	0.00	0.00	0.00	86
180	0.00	0.00	0.00	113
181	0.00	0.00	0.00	530
182	0.00	0.01	0.01	160
183	0.00	0.01	0.01	101
184	0.00	0.00	0.00	172
185	0.00	0.00	0.00	119
186	0.00	0.00	0.00	228
187	0.01	0.00	0.01	401
188	0.00	0.00	0.00	208
189	0.00	0.00	0.00	186
190	0.00	0.00	0.00	201
191	0.00	0.00	0.00	270
192	0.00	0.00	0.00	222
193	0.00	0.00	0.00	69
194	0.00	0.00	0.00	229
195	0.00	0.00	0.00	108
196	0.00	0.00	0.00	332
197	0.00	0.00	0.00	209
198	0.00	0.00	0.00	289
199	0.01	0.01	0.01	199
200	0.00	0.00	0.00	153
201	0.01	0.01	0.01	225
202	0.00	0.00	0.00	208
203	0.00	0.00	0.00	189
204	0.00	0.00	0.00	345
205	0.00	0.01	0.00	78
206	0.00	0.00	0.00	88
207	0.00	0.00	0.00	159
208	0.00	0.01	0.00	187
209	0.00	0.00	0.00	364
210	0.00	0.00	0.00	136
211	0.00	0.00	0.00	183
212	0.00	0.00	0.00	246
213	0.00	0.00	0.00	192
214	0.00	0.00	0.00	76
215	0.00	0.00	0.00	125
216	0.01	0.01	0.01	158
217	0.00	0.00	0.00	42
218	0.00	0.00	0.00	207
219	0.00	0.00	0.00	129
220	0.00	0.01	0.00	191
221	0.00	0.00	0.00	208
222	0.00	0.00	0.00	180
223	0.00	0.00	0.00	211
224	0.01	0.01	0.01	76
225	0.00	0.00	0.00	175
226	0.01	0.01	0.01	199
227	0.00	0.01	0.00	156
228	0.00	0.00	0.00	293
229	0.01	0.01	0.01	227
230	0.00	0.00	0.00	308
231	0.00	0.00	0.00	311
232	0.00	0.00	0.00	186
233	0.00	0.00	0.00	177
234	0.01	0.01	0.01	167
235	0.00	0.00	0.00	38

236	0.00	0.00	0.00	62
237	0.00	0.00	0.00	256
238	0.00	0.00	0.00	118
239	0.00	0.00	0.00	86
240	0.00	0.00	0.00	595
241	0.00	0.00	0.00	118
242	0.00	0.00	0.00	30
243	0.00	0.00	0.00	248
244	0.00	0.00	0.00	143
245	0.00	0.00	0.00	193
246	0.00	0.00	0.00	100
247	0.00	0.00	0.00	109
248	0.00	0.00	0.00	559
249	0.00	0.00	0.00	142
250	0.00	0.00	0.00	109
251	0.00	0.00	0.00	119
252	0.00	0.00	0.00	154
253	0.00	0.00	0.00	103
254	0.00	0.01	0.01	152
255	0.00	0.00	0.00	99
256	0.00	0.00	0.00	384
257	0.00	0.00	0.00	172
258	0.00	0.00	0.00	201
259	0.00	0.00	0.00	141
260	0.00	0.00	0.00	87
261	0.00	0.00	0.00	153
262	0.00	0.00	0.00	82
263	0.00	0.00	0.00	49
264	0.00	0.00	0.00	142
265	0.00	0.00	0.00	66
266	0.00	0.00	0.00	52
267	0.00	0.00	0.00	71
268	0.01	0.01	0.01	152
269	0.00	0.00	0.00	108
270	0.01	0.01	0.01	140
271	0.00	0.00	0.00	138
272	0.00	0.00	0.00	160
273	0.00	0.00	0.00	159
274	0.00	0.00	0.00	87
275	0.00	0.00	0.00	183
276	0.01	0.00	0.00	370
277	0.00	0.00	0.00	171
278	0.00	0.00	0.00	138
279	0.00	0.00	0.00	170
280	0.00	0.00	0.00	99
281	0.00	0.00	0.00	111
282	0.00	0.00	0.00	114
283	0.00	0.01	0.00	186
284	0.00	0.00	0.00	441
285	0.00	0.00	0.00	169
286	0.00	0.01	0.01	157
287	0.00	0.00	0.00	115
288	0.00	0.00	0.00	80
289	0.03	0.01	0.01	319
290	0.00	0.00	0.00	114
291	0.00	0.00	0.00	224
292	0.00	0.00	0.00	259
293	0.00	0.00	0.00	79
294	0.00	0.00	0.00	95
295	0.00	0.00	0.00	62
296	0.00	0.00	0.00	65

297	0.00	0.00	0.00	113
298	0.00	0.00	0.00	129
299	0.00	0.00	0.00	99
300	0.00	0.00	0.00	127
301	0.00	0.00	0.00	167
302	0.00	0.00	0.00	116
303	0.01	0.01	0.01	111
304	0.00	0.00	0.00	192
305	0.00	0.00	0.00	107
306	0.00	0.00	0.00	72
307	0.01	0.01	0.01	186
308	0.00	0.00	0.00	156
309	0.00	0.00	0.00	81
310	0.00	0.00	0.00	160
311	0.00	0.00	0.00	50
312	0.00	0.00	0.00	279
313	0.00	0.00	0.00	86
314	0.00	0.00	0.00	139
315	0.00	0.00	0.00	106
316	0.00	0.00	0.00	62
317	0.00	0.00	0.00	267
318	0.00	0.00	0.00	124
319	0.01	0.01	0.01	120
320	0.00	0.00	0.00	99
321	0.00	0.00	0.00	199
322	0.00	0.00	0.00	217
323	0.00	0.01	0.01	109
324	0.00	0.00	0.00	107
325	0.00	0.00	0.00	111
326	0.00	0.00	0.00	97
327	0.00	0.00	0.00	93
328	0.00	0.00	0.00	149
329	0.00	0.00	0.00	126
330	0.00	0.00	0.00	81
331	0.00	0.00	0.00	144
332	0.00	0.00	0.00	95
333	0.00	0.00	0.00	126
334	0.00	0.00	0.00	174
335	0.00	0.00	0.00	63
336	0.00	0.00	0.00	115
337	0.00	0.00	0.00	40
338	0.00	0.00	0.00	95
339	0.00	0.00	0.00	216
340	0.00	0.00	0.00	110
341	0.00	0.00	0.00	84
342	0.00	0.00	0.00	190
343	0.00	0.00	0.00	24
344	0.00	0.00	0.00	100
345	0.00	0.00	0.00	37
346	0.01	0.01	0.01	98
347	0.00	0.00	0.00	50
348	0.00	0.00	0.00	200
349	0.00	0.00	0.00	112
350	0.00	0.00	0.00	16
351	0.00	0.00	0.00	117
352	0.00	0.00	0.00	143
353	0.00	0.00	0.00	34
354	0.00	0.00	0.00	202
355	0.00	0.00	0.00	134
356	0.00	0.00	0.00	107
357	0.00	0.01	0.01	91

358	0.00	0.00	0.00	68
359	0.00	0.00	0.00	161
360	0.00	0.00	0.00	120
361	0.00	0.00	0.00	115
362	0.00	0.00	0.00	91
363	0.00	0.00	0.00	143
364	0.00	0.00	0.00	112
365	0.01	0.02	0.02	47
366	0.00	0.00	0.00	439
367	0.01	0.00	0.00	298
368	0.00	0.00	0.00	160
369	0.00	0.00	0.00	97
370	0.00	0.00	0.00	101
371	0.00	0.00	0.00	95
372	0.00	0.00	0.00	50
373	0.00	0.00	0.00	141
374	0.00	0.00	0.00	138
375	0.00	0.00	0.00	83
376	0.00	0.00	0.00	49
377	0.00	0.00	0.00	232
378	0.00	0.00	0.00	143
379	0.00	0.00	0.00	23
380	0.00	0.00	0.00	75
381	0.01	0.01	0.01	87
382	0.00	0.00	0.00	71
383	0.00	0.00	0.00	96
384	0.00	0.00	0.00	122
385	0.00	0.00	0.00	41
386	0.00	0.00	0.00	91
387	0.00	0.00	0.00	121
388	0.00	0.00	0.00	80
389	0.00	0.00	0.00	142
390	0.00	0.00	0.00	134
391	0.00	0.00	0.00	82
392	0.00	0.00	0.00	14
393	0.01	0.02	0.01	121
394	0.00	0.00	0.00	124
395	0.00	0.00	0.00	95
396	0.01	0.01	0.01	142
397	0.00	0.00	0.00	112
398	0.00	0.00	0.00	114
399	0.00	0.00	0.00	189
400	0.00	0.00	0.00	75
401	0.00	0.00	0.00	123
402	0.00	0.00	0.00	128
403	0.00	0.00	0.00	69
404	0.00	0.00	0.00	14
405	0.00	0.00	0.00	120
406	0.00	0.00	0.00	85
407	0.00	0.00	0.00	92
408	0.00	0.00	0.00	70
409	0.00	0.00	0.00	47
410	0.00	0.00	0.00	24
411	0.00	0.00	0.00	88
412	0.00	0.00	0.00	136
413	0.00	0.00	0.00	34
414	0.00	0.00	0.00	67
415	0.00	0.00	0.00	37
416	0.00	0.00	0.00	81
417	0.00	0.00	0.00	112
418	0.00	0.00	0.00	81

419	0.00	0.00	0.00	89
420	0.00	0.00	0.00	27
421	0.00	0.00	0.00	70
422	0.00	0.00	0.00	89
423	0.00	0.00	0.00	97
424	0.00	0.00	0.00	10
425	0.00	0.00	0.00	16
426	0.00	0.00	0.00	85
427	0.00	0.00	0.00	108
428	0.00	0.00	0.00	18
429	0.00	0.00	0.00	53
430	0.00	0.00	0.00	88
431	0.01	0.01	0.01	115
432	0.00	0.00	0.00	20
433	0.00	0.00	0.00	42
434	0.00	0.00	0.00	316
435	0.00	0.00	0.00	43
436	0.01	0.01	0.01	67
437	0.00	0.00	0.00	214
438	0.00	0.00	0.00	110
439	0.00	0.00	0.00	162
440	0.00	0.00	0.00	63
441	0.00	0.00	0.00	83
442	0.00	0.00	0.00	83
443	0.00	0.00	0.00	53
444	0.00	0.00	0.00	40
445	0.00	0.00	0.00	87
446	0.00	0.00	0.00	44
447	0.00	0.00	0.00	56
448	0.00	0.00	0.00	68
449	0.00	0.00	0.00	59
450	0.00	0.00	0.00	113
451	0.00	0.00	0.00	51
452	0.00	0.00	0.00	66
453	0.00	0.00	0.00	31
454	0.01	0.03	0.02	30
455	0.00	0.00	0.00	139
456	0.00	0.00	0.00	26
457	0.00	0.00	0.00	77
458	0.00	0.00	0.00	78
459	0.00	0.00	0.00	117
460	0.00	0.00	0.00	83
461	0.00	0.00	0.00	74
462	0.00	0.00	0.00	83
463	0.00	0.00	0.00	86
464	0.03	0.04	0.03	76
465	0.00	0.00	0.00	123
466	0.00	0.00	0.00	92
467	0.00	0.00	0.00	90
468	0.00	0.00	0.00	135
469	0.00	0.00	0.00	80
470	0.00	0.00	0.00	94
471	0.00	0.00	0.00	62
472	0.00	0.00	0.00	90
473	0.00	0.00	0.00	69
474	0.00	0.00	0.00	84
475	0.00	0.00	0.00	100
476	0.00	0.00	0.00	85
477	0.00	0.00	0.00	79
478	0.00	0.00	0.00	94
479	0.00	0.00	0.00	59

480	0.02	0.02	0.02	132
481	0.00	0.00	0.00	247
482	0.00	0.00	0.00	75
483	0.00	0.00	0.00	62
484	0.00	0.00	0.00	33
485	0.00	0.00	0.00	54
486	0.00	0.00	0.00	170
487	0.00	0.00	0.00	85
488	0.01	0.01	0.01	105
489	0.00	0.00	0.00	116
490	0.00	0.00	0.00	351
491	0.00	0.00	0.00	48
492	0.00	0.00	0.00	65
493	0.00	0.00	0.00	74
494	0.00	0.00	0.00	55
495	0.00	0.00	0.00	22
496	0.00	0.00	0.00	80
497	0.00	0.00	0.00	63
498	0.00	0.00	0.00	67
499	0.00	0.00	0.00	121
micro avg	0.02	0.02	0.02	178469
macro avg	0.00	0.00	0.00	178469
weighted avg	0.02	0.02	0.02	178469
samples avg	0.02	0.02	0.01	178469

Time taken to run this cell : 8:23:03.546896

CPU times: user 16min 56s, sys: 8.96 s, total: 17min 5s

Wall time: 8min 32s

In [131]:

```
from sklearn.model_selection import ParameterGrid, GridSearchCV
score = 'f1_micro'
params = [{'estimator__alpha':[0.00001,0.0001,0.001,0.01,0.1,1]}]
classifierf = OneVsRestClassifier(SGDClassifier(loss='hinge', penalty='l2'))
gsf = GridSearchCV(classifierf,params,scoring=score,n_jobs=-1,verbose=True,return_train_score=True)
gfitf = gsf.fit(x_train_wbow,y_train)
```

Fitting 5 folds for each of 6 candidates, totalling 30 fits

[Parallel(n_jobs=-1)]: Using backend LokyBackend with 10 concurrent workers.

[Parallel(n_jobs=-1)]: Done 30 out of 30 | elapsed: 44.4min finished

In [132]:

```
gfitf.best_estimator_
```

Out[132]:

```
OneVsRestClassifier(estimator=SGDClassifier(alpha=1e-05, average=False,
                                             class_weight=None,
                                             early_stopping=False, epsilon
n=0.1,
                                             eta0=0.0, fit_intercept=True
e,
                                             l1_ratio=0.15,
                                             learning_rate='optimal', loss
s='log',
                                             max_iter=1000, n_iter_no_change
nge=5,
                                             n_jobs=-1, penalty='l2',
                                             power_t=0.5, random_state=None
ne,
                                             shuffle=True, tol=0.001,
                                             validation_fraction=0.1, ver
bose=0,
                                             warm_start=False),
n_jobs=None)
```

In [133]:

```
gfit.cv_results_
```

Out[133]:

```

{'mean_fit_time': array([ 616.95418673,  635.48110318,  718.49127617, 10
62.75535603,
      1139.3322104 ,  703.26464276])),
 'std_fit_time': array([ 5.61128433,  4.49199077,  4.21355524, 26.440214
94, 35.49924695,
      4.42560551])),
 'mean_score_time': array([9.24277344, 9.98951302, 9.18088212, 9.2514324
7, 5.61206827,
      8.24464612])),
 'std_score_time': array([0.50420647, 0.5177753 , 0.39461141, 0.4277663
, 1.26716757,
      0.72798833])),
 'param_estimator__alpha': masked_array(data=[1e-05, 0.0001, 0.001, 0.0
1, 0.1, 1],
      mask=[False, False, False, False, False, False],
      fill_value='?',
      dtype=object),
 'params': [{'estimator__alpha': 1e-05},
 {'estimator__alpha': 0.0001},
 {'estimator__alpha': 0.001},
 {'estimator__alpha': 0.01},
 {'estimator__alpha': 0.1},
 {'estimator__alpha': 1}],
 'split0_test_score': array([0.92612409, 0.85410127, 0.57621544, 0.30083
674, 0.09293081,
      0.05993031])),
 'split1_test_score': array([0.90167997, 0.83859613, 0.61285554, 0.34938
219, 0.03796993,
      0.02977798])),
 'split2_test_score': array([0.93043654, 0.86702821, 0.64357266, 0.37903
403, 0.1336102 ,
      0.09531864])),
 'split3_test_score': array([0.94367021, 0.88677088, 0.66001438, 0.42149
41 , 0.20276397,
      0.17250814])),
 'split4_test_score': array([0.97530996, 0.90524508, 0.74663231, 0.68988
367, 0.32152021,
      0.20661622])),
 'mean_test_score': array([0.93544416, 0.87034832, 0.64785807, 0.4281261
4, 0.15775902,
      0.11283026])),
 'std_test_score': array([0.02412016, 0.02354008, 0.05706195, 0.1366559
4, 0.0979576 ,
      0.06687301])),
 'rank_test_score': array([1, 2, 3, 4, 5, 6], dtype=int32),
 'split0_train_score': array([0.96239113, 0.90343433, 0.71244755, 0.5014
2015, 0.34688712,
      0.31555165])),
 'split1_train_score': array([0.96731837, 0.9076435 , 0.70428761, 0.4730
933 , 0.29399768,
      0.28077648])),
 'split2_train_score': array([0.96224599, 0.90172126, 0.70880031, 0.4804
5337, 0.32099976,
      0.27377205])),
 'split3_train_score': array([0.96071698, 0.89928164, 0.70065372, 0.4692
362 , 0.31344139,
      0.25955247])),
 'split4_train_score': array([0.95302023, 0.88801466, 0.67460012, 0.4016
9508, 0.13452168,

```

```
    0.0506589 ]),  
    'mean_train_score': array([0.96113854, 0.90001908, 0.70015786, 0.465179  
62, 0.28196953,  
    0.23606231]),  
    'std_train_score': array([0.00462754, 0.00659307, 0.01338826, 0.0336334  
1, 0.07564876,  
    0.09451576])}]}
```

In [135]:

```
%%time
classifierf = OneVsRestClassifier(SGDClassifier(loss='hinge', alpha=1e-05, penalty=
'12'))
classifierf.fit(x_train_wbow,y_train)
predictionsf = classifierf.predict(x_test_wbow)

print("Accuracy :",accuracy_score(y_test, predictionsf))
print("Hamming loss ",hamming_loss(y_test,predictionsf))

precision = precision_score(y_test, predictionsf, average='micro')
recall = recall_score(y_test, predictionsf, average='micro')
f1 = f1_score(y_test, predictionsf, average='micro')

print("Micro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1))

precision = precision_score(y_test, predictionsf, average='macro')
recall = recall_score(y_test, predictionsf, average='macro')
f1 = f1_score(y_test, predictionsf, average='macro')

print("Macro-average quality numbers")
print("Precision: {:.4f}, Recall: {:.4f}, F1-measure: {:.4f}".format(precision, recall, f1))

print (classification_report(y_test, predictionsf))
```

Accuracy : 0.01715

Hamming loss 0.00683364

Micro-average quality numbers

Precision: 0.0184, Recall: 0.0174, F1-measure: 0.0179

Macro-average quality numbers

Precision: 0.0034, Recall: 0.0033, F1-measure: 0.0031

	precision	recall	f1-score	support
0	0.05	0.06	0.05	4633
1	0.07	0.04	0.06	7549
2	0.07	0.07	0.07	7112
3	0.03	0.02	0.03	2888
4	0.05	0.07	0.06	5009
5	0.04	0.03	0.03	3919
6	0.05	0.05	0.05	5204
7	0.04	0.05	0.04	3229
8	0.04	0.04	0.04	3097
9	0.02	0.03	0.02	1829
10	0.02	0.03	0.02	2157
11	0.03	0.03	0.03	3142
12	0.03	0.03	0.03	2575
13	0.02	0.01	0.01	1431
14	0.02	0.02	0.02	2649
15	0.02	0.03	0.02	2094
16	0.02	0.01	0.01	1549
17	0.02	0.02	0.02	1869
18	0.02	0.02	0.02	2268
19	0.00	0.00	0.00	852
20	0.02	0.02	0.02	1614
21	0.02	0.02	0.02	1798
22	0.02	0.02	0.02	1350
23	0.01	0.01	0.01	888
24	0.03	0.01	0.01	1528
25	0.00	0.00	0.00	388
26	0.01	0.01	0.01	717
27	0.01	0.01	0.01	986
28	0.01	0.01	0.01	1046
29	0.01	0.00	0.00	2380
30	0.01	0.00	0.00	695
31	0.01	0.01	0.01	1095
32	0.03	0.00	0.01	2896
33	0.01	0.01	0.01	652
34	0.01	0.01	0.01	929
35	0.01	0.02	0.01	632
36	0.00	0.00	0.00	535
37	0.00	0.00	0.00	172
38	0.01	0.01	0.01	728
39	0.01	0.01	0.01	632
40	0.01	0.01	0.01	964
41	0.01	0.00	0.00	579
42	0.00	0.00	0.00	480
43	0.01	0.00	0.00	2408
44	0.00	0.00	0.00	719
45	0.01	0.01	0.01	455
46	0.00	0.01	0.00	385
47	0.00	0.00	0.00	517
48	0.00	0.00	0.00	516
49	0.01	0.01	0.01	531
50	0.01	0.01	0.01	410
51	0.01	0.00	0.00	790
52	0.00	0.00	0.00	151

53	0.01	0.01	0.01	702
54	0.00	0.01	0.00	480
55	0.00	0.00	0.00	423
56	0.01	0.01	0.01	485
57	0.01	0.01	0.01	848
58	0.01	0.01	0.01	473
59	0.00	0.00	0.00	640
60	0.00	0.00	0.00	499
61	0.00	0.00	0.00	234
62	0.01	0.01	0.01	546
63	0.01	0.01	0.01	492
64	0.01	0.01	0.01	138
65	0.01	0.01	0.01	350
66	0.00	0.00	0.00	233
67	0.00	0.00	0.00	203
68	0.00	0.00	0.00	401
69	0.00	0.00	0.00	283
70	0.01	0.01	0.01	444
71	0.00	0.00	0.00	839
72	0.00	0.00	0.00	281
73	0.00	0.00	0.00	1622
74	0.01	0.00	0.00	716
75	0.00	0.00	0.00	127
76	0.01	0.01	0.01	413
77	0.01	0.00	0.00	296
78	0.01	0.01	0.01	264
79	0.01	0.01	0.01	469
80	0.00	0.00	0.00	229
81	0.00	0.00	0.00	416
82	0.01	0.00	0.01	403
83	0.00	0.01	0.01	307
84	0.01	0.00	0.00	224
85	0.00	0.01	0.00	353
86	0.00	0.00	0.00	383
87	0.01	0.01	0.01	260
88	0.00	0.00	0.00	237
89	0.00	0.01	0.01	107
90	0.01	0.01	0.01	490
91	0.01	0.01	0.01	325
92	0.01	0.01	0.01	525
93	0.00	0.00	0.00	263
94	0.00	0.00	0.00	111
95	0.00	0.00	0.00	283
96	0.00	0.00	0.00	205
97	0.00	0.00	0.00	284
98	0.00	0.00	0.00	129
99	0.02	0.01	0.01	568
100	0.00	0.00	0.00	374
101	0.00	0.00	0.00	854
102	0.00	0.01	0.00	360
103	0.00	0.00	0.00	101
104	0.00	0.00	0.00	238
105	0.00	0.00	0.00	205
106	0.01	0.00	0.00	375
107	0.01	0.01	0.01	219
108	0.00	0.00	0.00	255
109	0.00	0.00	0.00	352
110	0.01	0.00	0.00	353
111	0.00	0.00	0.00	224
112	0.00	0.00	0.00	144
113	0.00	0.00	0.00	199

114	0.00	0.00	0.00	237
115	0.00	0.00	0.00	273
116	0.00	0.00	0.00	255
117	0.01	0.00	0.01	606
118	0.00	0.01	0.00	238
119	0.00	0.00	0.00	147
120	0.00	0.00	0.00	374
121	0.00	0.00	0.00	292
122	0.00	0.00	0.00	74
123	0.00	0.00	0.00	232
124	0.03	0.01	0.01	905
125	0.00	0.00	0.00	306
126	0.00	0.01	0.00	153
127	0.02	0.00	0.01	457
128	0.01	0.01	0.01	225
129	0.00	0.00	0.00	196
130	0.00	0.00	0.00	87
131	0.00	0.00	0.00	116
132	0.00	0.00	0.00	228
133	0.01	0.01	0.01	278
134	0.00	0.00	0.00	745
135	0.01	0.00	0.01	213
136	0.00	0.00	0.00	427
137	0.01	0.00	0.01	405
138	0.00	0.00	0.00	164
139	0.00	0.00	0.00	216
140	0.00	0.00	0.00	292
141	0.00	0.00	0.00	126
142	0.00	0.01	0.00	119
143	0.00	0.00	0.00	408
144	0.01	0.00	0.00	629
145	0.00	0.00	0.00	866
146	0.00	0.00	0.00	228
147	0.00	0.00	0.00	100
148	0.00	0.00	0.00	265
149	0.00	0.00	0.00	280
150	0.01	0.01	0.01	169
151	0.00	0.00	0.00	32
152	0.00	0.00	0.00	259
153	0.00	0.00	0.00	220
154	0.00	0.00	0.00	369
155	0.00	0.00	0.00	191
156	0.00	0.00	0.00	81
157	0.00	0.00	0.00	216
158	0.00	0.00	0.00	69
159	0.00	0.00	0.00	127
160	0.00	0.00	0.00	201
161	0.00	0.01	0.01	172
162	0.00	0.00	0.00	271
163	0.00	0.00	0.00	207
164	0.00	0.00	0.00	234
165	0.00	0.00	0.00	363
166	0.00	0.00	0.00	480
167	0.00	0.00	0.00	234
168	0.00	0.00	0.00	224
169	0.00	0.00	0.00	238
170	0.00	0.00	0.00	56
171	0.00	0.00	0.00	824
172	0.00	0.00	0.00	219
173	0.00	0.00	0.00	206
174	0.00	0.00	0.00	240

175	0.00	0.00	0.00	79
176	0.00	0.00	0.00	254
177	0.00	0.00	0.00	305
178	0.00	0.00	0.00	343
179	0.00	0.00	0.00	86
180	0.00	0.00	0.00	113
181	0.00	0.00	0.00	530
182	0.00	0.01	0.01	160
183	0.00	0.01	0.01	101
184	0.00	0.00	0.00	172
185	0.00	0.00	0.00	119
186	0.00	0.00	0.00	228
187	0.01	0.00	0.01	401
188	0.00	0.00	0.00	208
189	0.00	0.00	0.00	186
190	0.00	0.00	0.00	201
191	0.00	0.00	0.00	270
192	0.00	0.00	0.00	222
193	0.00	0.00	0.00	69
194	0.00	0.00	0.00	229
195	0.00	0.00	0.00	108
196	0.00	0.00	0.00	332
197	0.00	0.00	0.00	209
198	0.00	0.00	0.00	289
199	0.01	0.01	0.01	199
200	0.00	0.00	0.00	153
201	0.01	0.01	0.01	225
202	0.00	0.00	0.00	208
203	0.00	0.00	0.00	189
204	0.00	0.00	0.00	345
205	0.00	0.01	0.00	78
206	0.00	0.00	0.00	88
207	0.00	0.00	0.00	159
208	0.00	0.01	0.00	187
209	0.00	0.00	0.00	364
210	0.00	0.00	0.00	136
211	0.00	0.00	0.00	183
212	0.00	0.00	0.00	246
213	0.00	0.00	0.00	192
214	0.00	0.00	0.00	76
215	0.00	0.00	0.00	125
216	0.01	0.01	0.01	158
217	0.00	0.00	0.00	42
218	0.00	0.00	0.00	207
219	0.00	0.00	0.00	129
220	0.00	0.01	0.00	191
221	0.00	0.00	0.00	208
222	0.00	0.00	0.00	180
223	0.00	0.00	0.00	211
224	0.01	0.01	0.01	76
225	0.00	0.00	0.00	175
226	0.01	0.01	0.01	199
227	0.00	0.01	0.00	156
228	0.00	0.00	0.00	293
229	0.01	0.01	0.01	227
230	0.00	0.00	0.00	308
231	0.00	0.00	0.00	311
232	0.00	0.00	0.00	186
233	0.00	0.00	0.00	177
234	0.01	0.01	0.01	167
235	0.00	0.00	0.00	38

236	0.00	0.00	0.00	62
237	0.00	0.00	0.00	256
238	0.00	0.00	0.00	118
239	0.00	0.00	0.00	86
240	0.00	0.00	0.00	595
241	0.00	0.00	0.00	118
242	0.00	0.00	0.00	30
243	0.00	0.00	0.00	248
244	0.00	0.00	0.00	143
245	0.00	0.00	0.00	193
246	0.00	0.00	0.00	100
247	0.00	0.00	0.00	109
248	0.00	0.00	0.00	559
249	0.00	0.00	0.00	142
250	0.00	0.00	0.00	109
251	0.00	0.00	0.00	119
252	0.00	0.00	0.00	154
253	0.00	0.00	0.00	103
254	0.00	0.01	0.01	152
255	0.00	0.00	0.00	99
256	0.00	0.00	0.00	384
257	0.00	0.00	0.00	172
258	0.00	0.00	0.00	201
259	0.00	0.00	0.00	141
260	0.00	0.00	0.00	87
261	0.00	0.00	0.00	153
262	0.00	0.00	0.00	82
263	0.00	0.00	0.00	49
264	0.00	0.00	0.00	142
265	0.00	0.00	0.00	66
266	0.00	0.00	0.00	52
267	0.00	0.00	0.00	71
268	0.01	0.01	0.01	152
269	0.00	0.00	0.00	108
270	0.01	0.01	0.01	140
271	0.00	0.00	0.00	138
272	0.00	0.00	0.00	160
273	0.00	0.00	0.00	159
274	0.00	0.00	0.00	87
275	0.00	0.00	0.00	183
276	0.00	0.00	0.00	370
277	0.00	0.00	0.00	171
278	0.00	0.00	0.00	138
279	0.00	0.00	0.00	170
280	0.00	0.00	0.00	99
281	0.00	0.00	0.00	111
282	0.00	0.00	0.00	114
283	0.00	0.01	0.00	186
284	0.00	0.00	0.00	441
285	0.00	0.00	0.00	169
286	0.00	0.01	0.01	157
287	0.00	0.00	0.00	115
288	0.00	0.00	0.00	80
289	0.02	0.01	0.01	319
290	0.00	0.00	0.00	114
291	0.00	0.00	0.00	224
292	0.00	0.00	0.00	259
293	0.00	0.00	0.00	79
294	0.00	0.00	0.00	95
295	0.00	0.00	0.00	62
296	0.00	0.00	0.00	65

297	0.00	0.00	0.00	113
298	0.00	0.00	0.00	129
299	0.00	0.00	0.00	99
300	0.00	0.00	0.00	127
301	0.00	0.00	0.00	167
302	0.00	0.00	0.00	116
303	0.01	0.01	0.01	111
304	0.00	0.00	0.00	192
305	0.00	0.00	0.00	107
306	0.00	0.00	0.00	72
307	0.01	0.01	0.01	186
308	0.00	0.00	0.00	156
309	0.00	0.00	0.00	81
310	0.00	0.00	0.00	160
311	0.00	0.00	0.00	50
312	0.00	0.00	0.00	279
313	0.00	0.00	0.00	86
314	0.00	0.00	0.00	139
315	0.00	0.00	0.00	106
316	0.00	0.00	0.00	62
317	0.00	0.00	0.00	267
318	0.00	0.00	0.00	124
319	0.01	0.01	0.01	120
320	0.00	0.00	0.00	99
321	0.00	0.00	0.00	199
322	0.00	0.00	0.00	217
323	0.00	0.01	0.01	109
324	0.00	0.00	0.00	107
325	0.00	0.00	0.00	111
326	0.00	0.00	0.00	97
327	0.04	0.01	0.02	93
328	0.00	0.00	0.00	149
329	0.00	0.00	0.00	126
330	0.00	0.00	0.00	81
331	0.00	0.00	0.00	144
332	0.00	0.00	0.00	95
333	0.00	0.00	0.00	126
334	0.00	0.00	0.00	174
335	0.00	0.00	0.00	63
336	0.00	0.00	0.00	115
337	0.00	0.00	0.00	40
338	0.00	0.00	0.00	95
339	0.00	0.00	0.00	216
340	0.00	0.00	0.00	110
341	0.00	0.00	0.00	84
342	0.00	0.00	0.00	190
343	0.00	0.00	0.00	24
344	0.00	0.00	0.00	100
345	0.00	0.00	0.00	37
346	0.01	0.01	0.01	98
347	0.00	0.00	0.00	50
348	0.00	0.00	0.00	200
349	0.00	0.00	0.00	112
350	0.00	0.00	0.00	16
351	0.00	0.00	0.00	117
352	0.00	0.00	0.00	143
353	0.00	0.00	0.00	34
354	0.00	0.00	0.00	202
355	0.00	0.00	0.00	134
356	0.00	0.00	0.00	107
357	0.00	0.01	0.01	91

358	0.00	0.00	0.00	68
359	0.00	0.00	0.00	161
360	0.00	0.00	0.00	120
361	0.00	0.00	0.00	115
362	0.00	0.00	0.00	91
363	0.00	0.00	0.00	143
364	0.00	0.00	0.00	112
365	0.01	0.02	0.02	47
366	0.00	0.00	0.00	439
367	0.01	0.00	0.00	298
368	0.00	0.00	0.00	160
369	0.00	0.00	0.00	97
370	0.00	0.00	0.00	101
371	0.00	0.00	0.00	95
372	0.00	0.00	0.00	50
373	0.00	0.00	0.00	141
374	0.00	0.00	0.00	138
375	0.00	0.00	0.00	83
376	0.00	0.00	0.00	49
377	0.00	0.00	0.00	232
378	0.00	0.00	0.00	143
379	0.00	0.00	0.00	23
380	0.00	0.00	0.00	75
381	0.01	0.01	0.01	87
382	0.00	0.00	0.00	71
383	0.00	0.00	0.00	96
384	0.00	0.00	0.00	122
385	0.00	0.00	0.00	41
386	0.00	0.00	0.00	91
387	0.00	0.00	0.00	121
388	0.00	0.00	0.00	80
389	0.00	0.00	0.00	142
390	0.00	0.00	0.00	134
391	0.00	0.00	0.00	82
392	0.00	0.00	0.00	14
393	0.01	0.02	0.01	121
394	0.00	0.00	0.00	124
395	0.00	0.00	0.00	95
396	0.01	0.01	0.01	142
397	0.00	0.00	0.00	112
398	0.00	0.00	0.00	114
399	0.00	0.00	0.00	189
400	0.00	0.00	0.00	75
401	0.00	0.00	0.00	123
402	0.00	0.00	0.00	128
403	0.00	0.00	0.00	69
404	0.00	0.00	0.00	14
405	0.00	0.00	0.00	120
406	0.00	0.00	0.00	85
407	0.00	0.00	0.00	92
408	0.00	0.00	0.00	70
409	0.00	0.00	0.00	47
410	0.00	0.00	0.00	24
411	0.00	0.00	0.00	88
412	0.00	0.00	0.00	136
413	0.00	0.00	0.00	34
414	0.00	0.00	0.00	67
415	0.00	0.00	0.00	37
416	0.00	0.00	0.00	81
417	0.00	0.00	0.00	112
418	0.00	0.00	0.00	81

419	0.00	0.00	0.00	89
420	0.00	0.00	0.00	27
421	0.00	0.00	0.00	70
422	0.00	0.00	0.00	89
423	0.00	0.00	0.00	97
424	0.00	0.00	0.00	10
425	0.00	0.00	0.00	16
426	0.00	0.00	0.00	85
427	0.00	0.00	0.00	108
428	0.00	0.00	0.00	18
429	0.00	0.00	0.00	53
430	0.00	0.00	0.00	88
431	0.01	0.01	0.01	115
432	0.00	0.00	0.00	20
433	0.00	0.00	0.00	42
434	0.00	0.00	0.00	316
435	0.00	0.00	0.00	43
436	0.01	0.01	0.01	67
437	0.00	0.00	0.00	214
438	0.00	0.00	0.00	110
439	0.00	0.00	0.00	162
440	0.00	0.00	0.00	63
441	0.00	0.00	0.00	83
442	0.00	0.00	0.00	83
443	0.00	0.00	0.00	53
444	0.00	0.00	0.00	40
445	0.00	0.00	0.00	87
446	0.00	0.00	0.00	44
447	0.00	0.00	0.00	56
448	0.00	0.00	0.00	68
449	0.00	0.00	0.00	59
450	0.00	0.00	0.00	113
451	0.00	0.00	0.00	51
452	0.00	0.00	0.00	66
453	0.00	0.00	0.00	31
454	0.01	0.03	0.02	30
455	0.00	0.00	0.00	139
456	0.00	0.00	0.00	26
457	0.00	0.00	0.00	77
458	0.00	0.00	0.00	78
459	0.00	0.00	0.00	117
460	0.00	0.00	0.00	83
461	0.00	0.00	0.00	74
462	0.00	0.00	0.00	83
463	0.00	0.00	0.00	86
464	0.03	0.04	0.03	76
465	0.00	0.00	0.00	123
466	0.00	0.00	0.00	92
467	0.00	0.00	0.00	90
468	0.00	0.00	0.00	135
469	0.00	0.00	0.00	80
470	0.00	0.00	0.00	94
471	0.00	0.00	0.00	62
472	0.00	0.00	0.00	90
473	0.00	0.00	0.00	69
474	0.00	0.00	0.00	84
475	0.00	0.00	0.00	100
476	0.00	0.00	0.00	85
477	0.00	0.00	0.00	79
478	0.00	0.00	0.00	94
479	0.00	0.00	0.00	59

480	0.02	0.02	0.02	132
481	0.00	0.00	0.00	247
482	0.00	0.00	0.00	75
483	0.00	0.00	0.00	62
484	0.00	0.00	0.00	33
485	0.00	0.00	0.00	54
486	0.00	0.00	0.00	170
487	0.00	0.00	0.00	85
488	0.01	0.01	0.01	105
489	0.00	0.00	0.00	116
490	0.00	0.00	0.00	351
491	0.00	0.00	0.00	48
492	0.00	0.00	0.00	65
493	0.00	0.00	0.00	74
494	0.00	0.00	0.00	55
495	0.00	0.00	0.00	22
496	0.00	0.00	0.00	80
497	0.00	0.00	0.00	63
498	0.00	0.00	0.00	67
499	0.00	0.00	0.00	121
micro avg	0.02	0.02	0.02	178469
macro avg	0.00	0.00	0.00	178469
weighted avg	0.02	0.02	0.02	178469
samples avg	0.02	0.02	0.01	178469

CPU times: user 14min 30s, sys: 9.92 s, total: 14min 40s
 Wall time: 6min 5s

In [143]:

```
# Summary of the total models
from prettytable import PrettyTable
pt = PrettyTable()
pt.field_names = ["No", "MODEL", "FEATURIZATION", "ALPHA", 'LOSS', 'MICRO_F1_SCORE']
```

In [144]:

```
pt.add_row(["1", 'OneVsRest+SGD Classifier', "Tf-idf",0.0001,"log",0.00505718])
pt.add_row(["2", 'OneVsRest+SGD(log)=LR', "Bag-of-words",0.001,"log",0.00472184 ])
pt.add_row(["3", 'OneVsRest+SGD(log)=LR', "Bag-of-words",1e-05,"log",0.01771])
pt.add_row(["4", 'OneVsRest+SGD Classifier', "Bag-of-words",1e-05,"Hinge",0.01715])
```

In [145]:

```
print(pt)
```

+-----+-----+-----+-----+-----+-----+-----+ -----+						
No	MODEL	FEATURIZATION	ALPHA	LOSS	MICRO	
_F1_SCORE						
+-----+-----+-----+-----+-----+-----+-----+ -----+						
1	OneVsRest+SGD Classifier	Tf-idf	0.0001	log	0.0	
0505718						
2	OneVsRest+SGD(log)=LR	Bag-of-words	0.001	log	0.0	
0472184						
3	OneVsRest+SGD(log)=LR	Bag-of-words	1e-05	log	0.	
01771						
4	OneVsRest+SGD Classifier	Bag-of-words	1e-05	Hinge	0.	
01715						
+-----+-----+-----+-----+-----+-----+-----+ -----+						