



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

▼ Stack Overflow: Tag Prediction

1. Business Problem

1.1 Description

Description

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

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

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>

Youtube : <https://youtu.be/nNDqbUhtIRg>

Research paper : <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>

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>

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 program 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, 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 variables";\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
    system("PAUSE");\n
    return 0;    \n
}\n

```

\n\n

<p>The answer should come in the form of a table like</p>\n\n

<pre><code>

```

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

```

```

50          2          50\n
50          99         50\n
50          100        50\n
50          50         1\n
50          50         2\n
50          50         99\n
50          50        100\n
</code></pre>\n\n
<p>if the no of inputs is 3 and their ranges are\n
1,100\n
1,100\n
1,100\n
(could be varied too)</p>\n\n
<p>The output is not coming,can anyone correct the code or tell me what\'s wrong?<

```

Tags : 'c++ c'

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

2.2.1 Type of Machine Learning Problem

It is a multi-label classification problem

Multi-label Classification: Multilabel classification assigns to each sample a set of target labels. This can be thought of as a data-point that are not mutually exclusive, such as topics that are relevant for a document. A question on Stack Overflow asks for 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. The score reaches its best value at 1 and worst score at 0. The relative contribution of precision and recall to the F1 score 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 for imbalanced data.

'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://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>

3. Exploratory Data Analysis

3.1 Data Loading and Cleaning

3.1.1 Using Pandas with SQLite to Load the data

```
#Creating db file from csv
#Learn SQL: https://www.w3schools.com/sql/default.asp
if not os.path.isfile('train.db'):
    start = datetime.now()
    disk_engine = create_engine('sqlite:///train.db')
    start = dt.datetime.now()
    chunksize = 180000
    j = 0
    index_start = 1
    for df in pd.read_csv('Train.csv', names=['Id', 'Title', 'Body', 'Tags'], chunksize=chunksize):
        df.index += index_start
        j+=1
        print('{} rows'.format(j*chunksize))
        df.to_sql('data', disk_engine, if_exists='append')
        index_start = df.index[-1] + 1
    print("Time taken to run this cell :", datetime.now() - start)
```

3.1.2 Counting the number of rows

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



Number of rows in the database :


6034196

Time taken to count the number of rows : 0:01:15.750352


3.1.3 Checking for duplicates

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

```
if os.path.isfile('train.db'):
    start = datetime.now()
    con = sqlite3.connect('train.db')
    df_no_dup = pd.read_sql_query('SELECT Title, Body, Tags, COUNT(*) as cnt_dup FROM data GR
    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")
```


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

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




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

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

 number of duplicate questions : 1827881 (30.2920389063 %)


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



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


```
start = datetime.now()
df_no_dup["tag count"] = df_no_dup["Tags"].apply(lambda text: len(text.split(" ")))
```

```
# adding a new feature number of tags per question
print("Time taken to run this cell :", datetime.now() - start)
df_no_dup.head()
```

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

| | Title | Body | Tag |
|---|---|--|--------------------------------------|
| 0 | Implementing Boundary Value Analysis of S... | <pre>#include< iostream>\n#include<... | c++ |
| 1 | Dynamic Datagrid Binding in Silverlight? | <p>I should do binding for datagrid dynamicall... | c# silverlight dat
bindir |
| 2 | Dynamic Datagrid Binding in Silverlight? | <p>I should do binding for datagrid dynamicall... | c# silverlight dat
binding columr |
| 3 | java.lang.NoClassDefFoundError: javax/serv... | <p>I followed the guide in <a href="http://sta... | jsp js |
| 4 | java.sql.SQLException:[Microsoft][ODBC Dri... | <p>I use the following code</p>\n\n<pre>...<code>... | java jdt |

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

 3 1206157
2 1111706
4 814996
1 568298
5 505158
Name: tag_count, dtype: int64

```
#Creating a new database with no duplicates
if not os.path.isfile('train_no_dup.db'):
    disk_dup = create_engine("sqlite:///train_no_dup.db")
    no_dup = pd.DataFrame(df_no_dup, columns=['Title', 'Body', 'Tags'])
    no_dup.to_sql('no_dup_train',disk_dup)


#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 tr
```

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


3.2 Analysis of Tags

3.2.1 Total number of unique tags


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

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

```
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

```
#'get_feature_name()' gives us the vocabulary.
tags = vectorizer.get_feature_names()
#Lets look at the tags we have.
print("Some of the tags we have :", tags[:10])
```

 Some of the tages we have : ['.a', '.app', '.asp.net-mvc', '.aspxauth', '.bash-profile',

3.2.3 Number of times a tag appeared

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

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

```
tag_df = pd.read_csv("tag_counts_dict_dtm.csv", names=['Tags', 'Counts'])
```

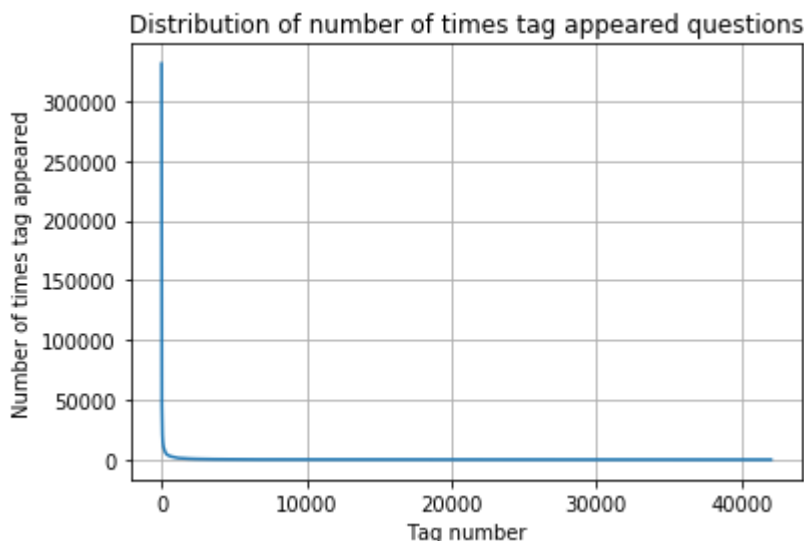
```
tag_df = pd.read_csv('tag_counts_with_urls.csv', names=[ 'tags', 'counts' ])
tag_df.head()
```



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

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

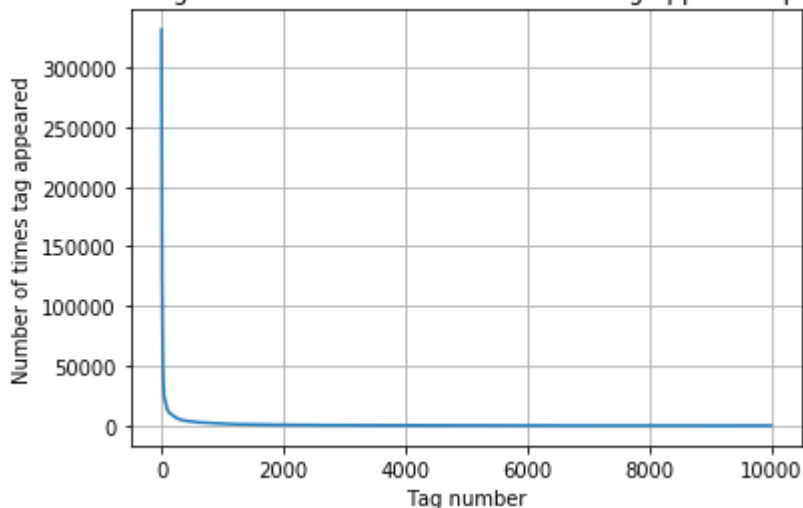
```
plt.plot(tag_counts)
plt.title("Distribution of number of times tag appeared questions")
plt.grid()
plt.xlabel("Tag number")
plt.ylabel("Number of times tag appeared")
plt.show()
```



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

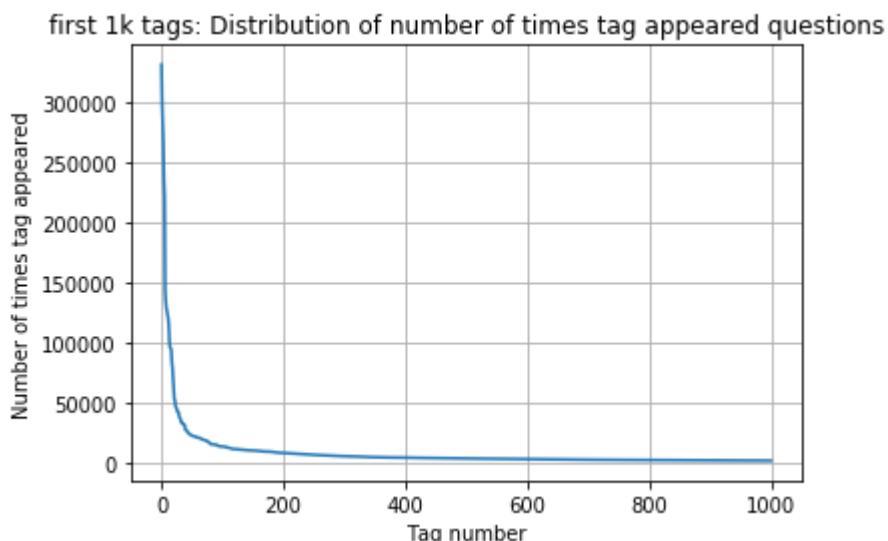


first 10k tags: Distribution of number of times tag appeared questions



| | | | | | | | | | | |
|------|---------|-------|-------|-------|-------|-------|-------|------|------|------|
| 400 | [331505 | 44829 | 22429 | 17728 | 13364 | 11162 | 10029 | 9148 | 8054 | 7151 |
| 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] | |

```
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])
```



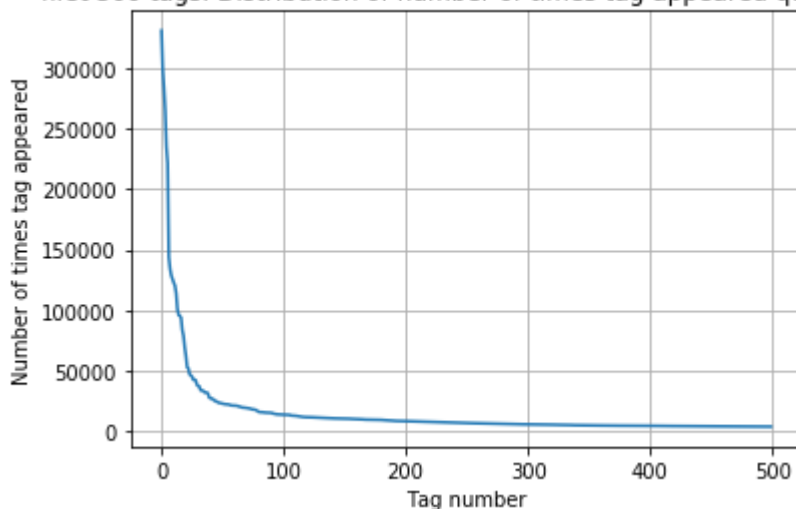
```
200 [331505 221533 122769 95160 62023 44829 37170 31897 26925 24537
22429 21820 20957 19758 18905 17728 15533 15097 14884 13703
13364 13157 12407 11658 11228 11162 10863 10600 10350 10224
10029 9884 9719 9411 9252 9148 9040 8617 8361 8163
8054 7867 7702 7564 7274 7151 7052 6847 6656 6553
6466 6291 6183 6093 5971 5865 5760 5577 5490 5411
5370 5283 5207 5107 5066 4983 4891 4785 4658 4549
4526 4487 4429 4335 4310 4281 4239 4228 4195 4159
4144 4088 4050 4002 3957 3929 3874 3849 3818 3797
3750 3703 3685 3658 3615 3593 3564 3521 3505 3483
3453 3427 3396 3363 3326 3299 3272 3232 3196 3168
3123 3094 3073 3050 3012 2989 2984 2953 2934 2903
2891 2844 2819 2784 2754 2738 2726 2708 2681 2669
2647 2621 2604 2594 2556 2527 2510 2482 2460 2444
2431 2409 2395 2380 2363 2331 2312 2297 2290 2281
2259 2246 2222 2211 2198 2186 2162 2142 2132 2107
2097 2078 2057 2045 2036 2020 2011 1994 1971 1965
1959 1952 1940 1932 1912 1900 1879 1865 1855 1841
1828 1821 1813 1801 1782 1770 1760 1747 1741 1734
1723 1707 1697 1688 1683 1673 1665 1656 1646 1639]
```

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

```
print(len(tag_counts[0:500:5]), tag_counts[0:500:5])
```



first 500 tags: Distribution of number of times tag appeared questions



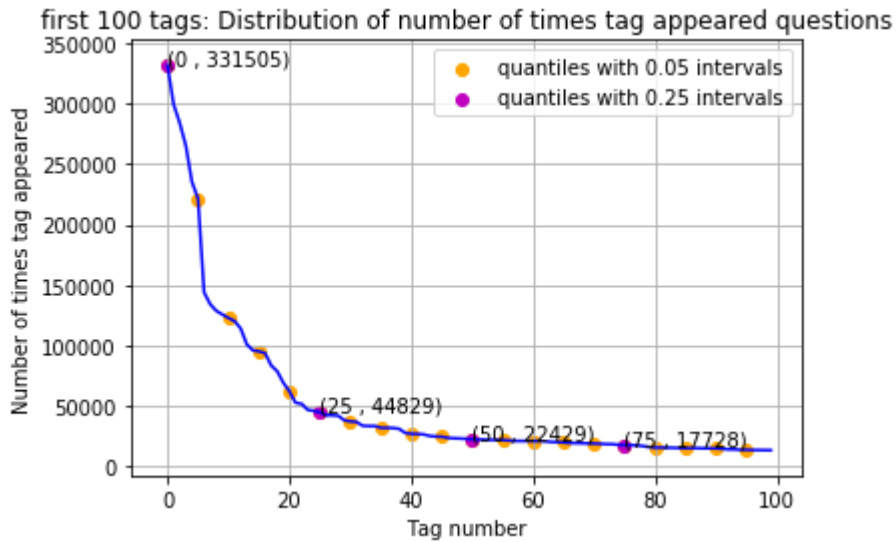
```
100 [331505 221533 122769 95160 62023 44829 37170 31897 26925 24537
22429 21820 20957 19758 18905 17728 15533 15097 14884 13703
13364 13157 12407 11658 11228 11162 10863 10600 10350 10224
10029 9884 9719 9411 9252 9148 9040 8617 8361 8163
8054 7867 7702 7564 7274 7151 7052 6847 6656 6553
6466 6291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

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

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

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





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

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



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

Observations:

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

3.2.4 Tags Per Question

```
#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
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])
```

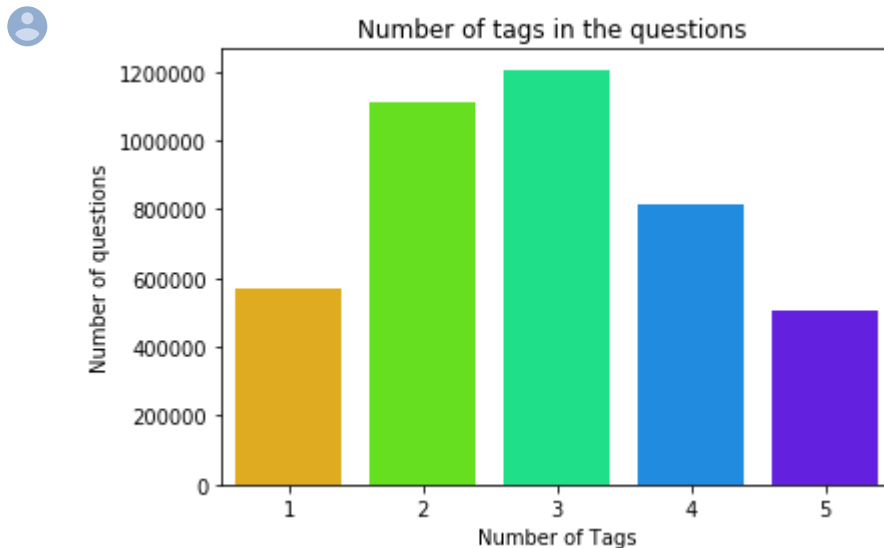


We have total 4206314 datapoints.
[3, 4, 2, 2, 3]

```
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_coun
```

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

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



Observations:

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

3.2.5 Most Frequent Tags

```
# Plotting word cloud
start = datetime.now()

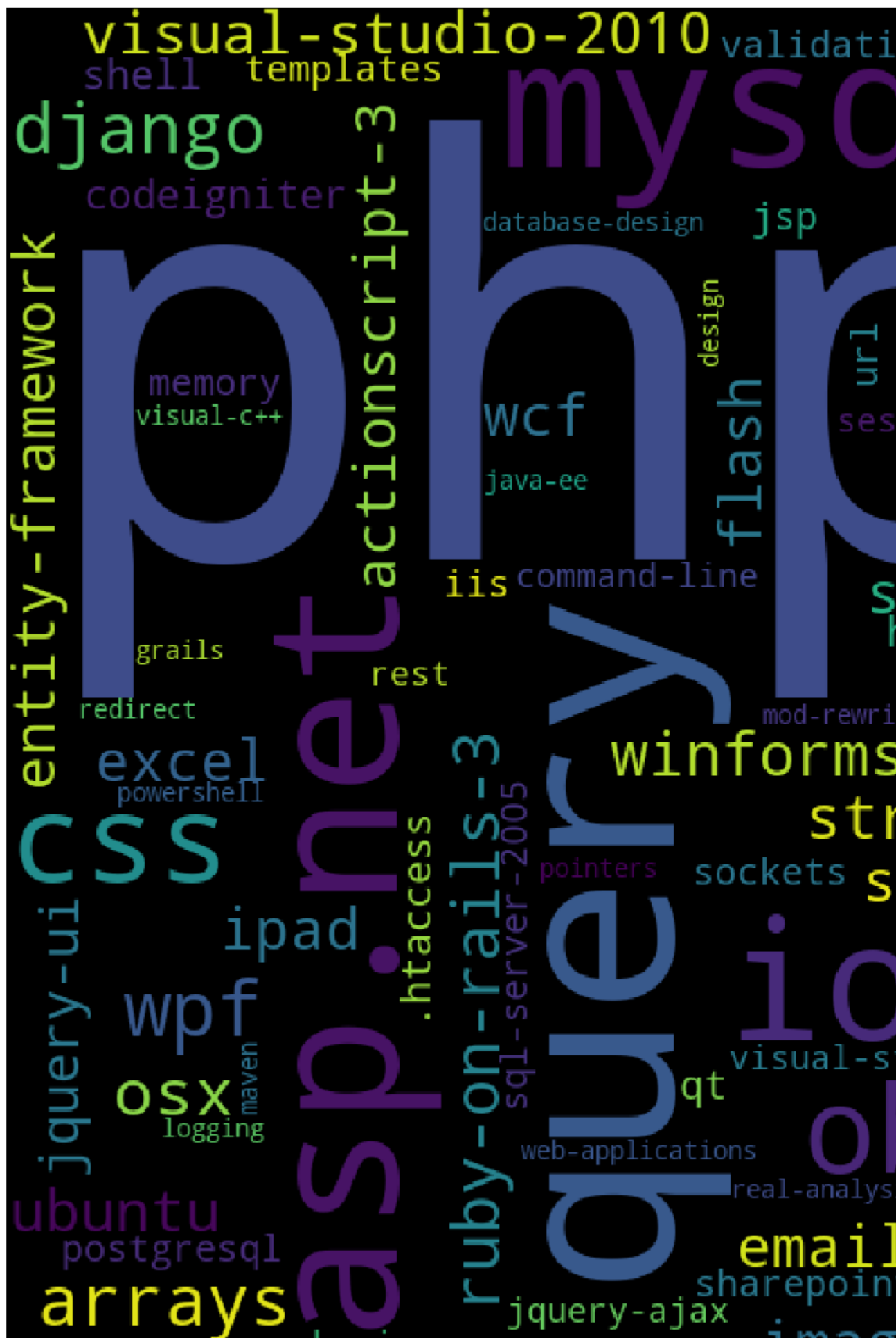
# Lets first convert the 'result' dictionary to 'list of tuples'
tup = dict(result.items())

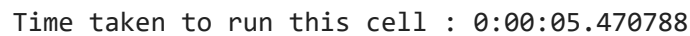
#Initializing WordCloud using frequencies of tags.
```

```
wordcloud = WordCloud(    background_color='black',
                          width=1600,
                          height=800,
                          ).generate_from_frequencies(tup)

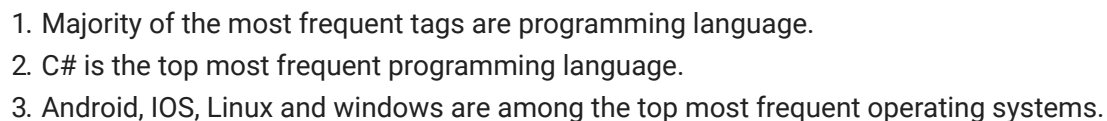
fig = plt.figure(figsize=(30,20))
plt.imshow(wordcloud)
plt.axis('off')
plt.tight_layout(pad=0)
fig.savefig("tag.png")
plt.show()
print("Time taken to run this cell :", datetime.now() - start)
```







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



3.3 Cleaning and preprocessing of Questions

3.3.1 Preprocessing

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

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

#http://www.sqlitetutorial.net/sqlite-python/create-tables/
def create_connection(db_file):
    """ create a database connection to the SQLite database
        specified by db_file
    :param db_file: database file
    :return: Connection object or None
    """
    try:
        conn = sqlite3.connect(db_file)
        return conn
    except Error as e:
        print(e)

    return None

def create_table(conn, create_table_sql):
    """ create a table from the create_table_sql statement
    :param conn: Connection object
    :param create_table_sql: a CREATE TABLE statement
    :return:
    """
    try:
        c = conn.cursor()
        c.execute(create_table_sql)
    except Error as e:
        print(e)
```

```
def checkTableExists(dbcon):
```

```
    cursor = dbcon.cursor()
```

```

cursr = conn.cursor()
str = "select name from sqlite_master where type='table'"
table_names = cursr.execute(str)
print("Tables in the database:")
tables = table_names.fetchall()
print(tables[0][0])
return(len(tables))

```

```

def create_database_table(database, query):
    conn = create_connection(database)
    if conn is not None:
        create_table(conn, query)
        checkTableExists(conn)
    else:
        print("Error! cannot create the database connection.")
    conn.close()

```

```

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

```



Tables in the database:
QuestionsProcessed

```

# http://www.sqlitetutorial.net/sqlite-delete/
# https://stackoverflow.com/questions/2279706/select-random-row-from-a-sqlite-table
start = datetime.now()
read_db = 'train_no_dup.db'
write_db = 'Processed.db'
if os.path.isfile(read_db):
    conn_r = create_connection(read_db)
    if conn_r is not None:
        reader = conn_r.cursor()
        reader.execute("SELECT Title, Body, Tags From no_dup_train ORDER BY RANDOM() LIMIT 10")

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 : 0:06:32.806567

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

<http://www.bernzilla.com/2008/05/13/selecting-a-random-row-from-an-sqlite-table/>

```

start = datetime.now()
preprocessed_data_list=[]
reader.fetchone()
questions_with_code=0
len_pre=0
len_post=0
questions_proccesed = 0
for row in reader:

    is_code = 0

    title, question, tags = row[0], row[1], row[2]

    if '<code>' in question:
        questions_with_code+=1
        is_code = 1
    x = len(question)+len(title)
    len_pre+=x

    code = str(re.findall(r'<code>(.*?)</code>', question, flags=re.DOTALL))

    question=re.sub('<code>(.*?)</code>', '', question, flags=re.MULTILINE|re.DOTALL)
    question=striphtml(question.encode('utf-8'))

    title=title.encode('utf-8')

    question=str(title)+" "+str(question)
    question=re.sub(r'^[A-Za-z]+',' ',question)
    words=word_tokenize(str(question.lower()))

    #Removing all single letter and and stopwords from question exceptt for the letter 'c'
    question=' '.join(str(stemmer.stem(j)) for j in words if j not in stop_words and (len(j)!

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

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

print( "Avg. length of questions(Title+Body) before processing: %d"%no_dup_avg_len_pre)
print( "Avg. length of questions(Title+Body) after processing: %d"%no_dup_avg_len_post)
print( "Percent of questions containing code: %d"%((questions_with_code*100.0)/questions_proc

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

```



```
number of questions completed= 100000
number of questions completed= 200000
number of questions completed= 300000
number of questions completed= 400000
number of questions completed= 500000
number of questions completed= 600000
number of questions completed= 700000
number of questions completed= 800000
number of questions completed= 900000
Avg. length of questions(Title+Body) before processing: 1169
Avg. length of questions(Title+Body) after processing: 327
Percent of questions containing code: 57
Time taken to run this cell : 0:47:05.946582
```

```
# dont forget to close the connections, or else you will end up with locks
```

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

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



Questions after preprocessed

```
=====
('ef code first defin one mani relationship differ key troubl defin one zero mani relati
-----
('explan new statement review section c code came accross statement block come accross n
-----
('error function notat function solv logic riddl iloczyni list structur list possibl can
-----
('step plan move one isp anoth one work busi plan switch isp realli soon need chang lot
-----
('use ef migrat creat databas googl migrat tutori af first run applic creat databas ef e
-----
('magento unit test problem magento site recent look way check integr magento site given
-----
('find network devic without bonjour write mac applic need discov mac pcs iphon ipad con
-----
('send multipl row mysql databas want send user mysql databas column user skill time nno
-----
('insert data mysql php powerpoint event powerpoint present run continu way updat slide
-----
```

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

```
preprocessed_data.head()
```



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

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



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

4. Machine Learning Models

4.1 Converting tags for multilabel problems

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

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

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

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

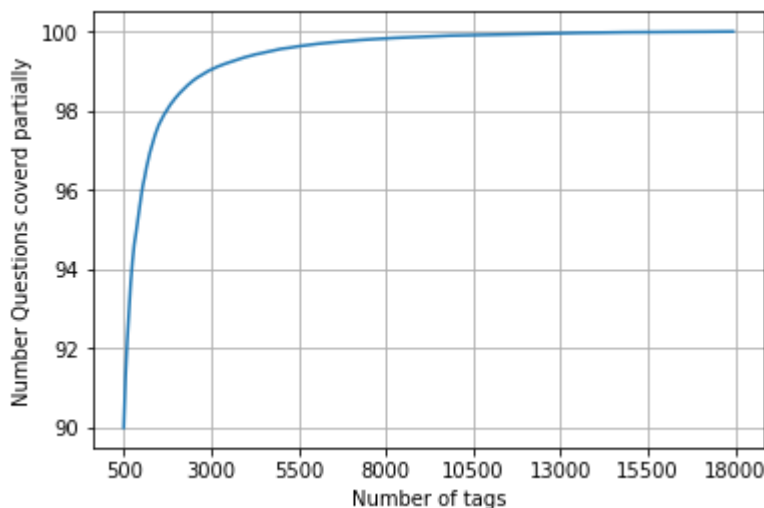
def questions_explained_fn(n):
    multilabel_yn = tags_to_choose(n)
    x= multilabel_yn.sum(axis=1)
    return (np.count_nonzero(x==0))

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

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



with 5500 tags we are covering 99.04 % of questions

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



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

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



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

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

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

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

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

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

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



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

4.3 Featurizing data

```
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
x_train_multilabel = vectorizer.fit_transform(x_train['question'])
x_test_multilabel = vectorizer.transform(x_test['question'])
print("Time taken to run this cell :", datetime.now() - start)
```



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

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



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

```
# 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)
```



```
"\nfrom skmultilearn.adapt import MLkNN\nnclassifier = MLkNN(k=21)\n\n# train\nclassifier
```

4.4 Applying Logistic Regression with OneVsRest Classifier

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

print("accuracy :",metrics.accuracy_score(y_test,predictions))
print("macro f1 score :",metrics.f1_score(y_test, predictions, average = 'macro'))
print("micro f1 scoore :",metrics.f1_score(y_test, predictions, average = 'micro'))
print("hamming loss :",metrics.hamming_loss(y_test,predictions))
print("Precision recall report :\n",metrics.classification_report(y_test, predictions))
```



accuracy : 0.081965
 macro f1 score : 0.0963020140154
 micro f1 score : 0.374270748817
 hamming loss : 0.000412250909090907
 Precision recall report :

| | precision | recall | f1-score | support |
|----|-----------|--------|----------|---------|
| 0 | 0.62 | 0.23 | 0.33 | 15760 |
| 1 | 0.79 | 0.43 | 0.56 | 14039 |
| 2 | 0.82 | 0.55 | 0.66 | 13446 |
| 3 | 0.76 | 0.42 | 0.54 | 12730 |
| 4 | 0.94 | 0.76 | 0.84 | 11229 |
| 5 | 0.85 | 0.64 | 0.73 | 10561 |
| 6 | 0.70 | 0.30 | 0.42 | 6958 |
| 7 | 0.87 | 0.61 | 0.72 | 6309 |
| 8 | 0.70 | 0.40 | 0.50 | 6032 |
| 9 | 0.78 | 0.43 | 0.55 | 6020 |
| 10 | 0.86 | 0.62 | 0.72 | 5707 |
| 11 | 0.52 | 0.17 | 0.25 | 5723 |
| 12 | 0.55 | 0.10 | 0.16 | 5521 |
| 13 | 0.59 | 0.25 | 0.35 | 4722 |
| 14 | 0.61 | 0.22 | 0.32 | 4468 |
| 15 | 0.79 | 0.52 | 0.63 | 4536 |
| 16 | 0.58 | 0.27 | 0.37 | 4545 |
| 17 | 0.80 | 0.53 | 0.64 | 4069 |
| 18 | 0.61 | 0.24 | 0.35 | 3638 |
| 19 | 0.57 | 0.18 | 0.27 | 3218 |
| 20 | 0.33 | 0.06 | 0.10 | 3000 |
| 21 | 0.73 | 0.34 | 0.46 | 2585 |
| 22 | 0.59 | 0.29 | 0.38 | 2439 |
| 23 | 0.88 | 0.61 | 0.72 | 2199 |
| 24 | 0.64 | 0.39 | 0.48 | 2157 |
| 25 | 0.67 | 0.39 | 0.49 | 2123 |
| 26 | 0.86 | 0.65 | 0.74 | 1948 |
| 27 | 0.35 | 0.07 | 0.12 | 2027 |
| 28 | 0.59 | 0.29 | 0.39 | 2013 |
| 29 | 0.61 | 0.20 | 0.30 | 1801 |
| 30 | 0.48 | 0.24 | 0.32 | 1728 |
| 31 | 0.94 | 0.75 | 0.84 | 1725 |
| 32 | 0.60 | 0.26 | 0.36 | 1581 |
| 33 | 0.49 | 0.14 | 0.22 | 1533 |
| 34 | 0.81 | 0.33 | 0.47 | 1565 |
| 35 | 0.75 | 0.62 | 0.68 | 1568 |
| 36 | 0.76 | 0.50 | 0.60 | 1542 |
| 37 | 0.74 | 0.50 | 0.59 | 1536 |
| 38 | 0.37 | 0.12 | 0.19 | 1524 |
| 39 | 0.40 | 0.12 | 0.19 | 1345 |
| 40 | 0.65 | 0.38 | 0.48 | 1292 |
| 41 | 0.41 | 0.11 | 0.17 | 1264 |
| 42 | 0.69 | 0.25 | 0.37 | 1265 |
| 43 | 0.59 | 0.29 | 0.38 | 1171 |
| 44 | 0.41 | 0.15 | 0.22 | 1173 |
| 45 | 0.38 | 0.10 | 0.16 | 1137 |
| 46 | 0.62 | 0.12 | 0.20 | 1125 |
| 47 | 0.26 | 0.07 | 0.11 | 1116 |
| 48 | 0.44 | 0.15 | 0.22 | 1042 |
| 49 | 0.40 | 0.02 | 0.03 | 1096 |

| | | | | |
|-----|------|------|------|------|
| 50 | 0.63 | 0.38 | 0.48 | 1031 |
| 51 | 0.47 | 0.14 | 0.22 | 1033 |
| 52 | 0.87 | 0.68 | 0.76 | 1042 |
| 53 | 0.32 | 0.09 | 0.14 | 1027 |
| 54 | 0.53 | 0.14 | 0.22 | 1063 |
| 55 | 0.63 | 0.34 | 0.44 | 1048 |
| 56 | 0.78 | 0.42 | 0.54 | 1054 |
| 57 | 0.91 | 0.77 | 0.83 | 1058 |
| 58 | 0.37 | 0.10 | 0.16 | 1000 |
| 59 | 0.26 | 0.03 | 0.05 | 973 |
| 60 | 0.76 | 0.42 | 0.54 | 978 |
| 61 | 0.74 | 0.43 | 0.54 | 977 |
| 62 | 0.27 | 0.06 | 0.10 | 957 |
| 63 | 0.81 | 0.22 | 0.34 | 958 |
| 64 | 0.88 | 0.63 | 0.73 | 944 |
| 65 | 0.76 | 0.49 | 0.60 | 923 |
| 66 | 0.67 | 0.36 | 0.47 | 959 |
| 67 | 0.55 | 0.15 | 0.24 | 951 |
| 68 | 0.38 | 0.13 | 0.20 | 924 |
| 69 | 0.71 | 0.25 | 0.37 | 897 |
| 70 | 0.78 | 0.47 | 0.59 | 900 |
| 71 | 0.82 | 0.40 | 0.54 | 893 |
| 72 | 0.21 | 0.01 | 0.01 | 836 |
| 73 | 0.74 | 0.16 | 0.26 | 850 |
| 74 | 0.58 | 0.37 | 0.45 | 838 |
| 75 | 0.88 | 0.64 | 0.74 | 855 |
| 76 | 0.47 | 0.28 | 0.35 | 837 |
| 77 | 0.68 | 0.41 | 0.52 | 824 |
| 78 | 0.14 | 0.01 | 0.01 | 793 |
| 79 | 0.34 | 0.09 | 0.14 | 751 |
| 80 | 0.31 | 0.08 | 0.13 | 793 |
| 81 | 0.71 | 0.33 | 0.45 | 758 |
| 82 | 0.60 | 0.28 | 0.38 | 764 |
| 83 | 0.82 | 0.59 | 0.69 | 710 |
| 84 | 0.82 | 0.48 | 0.61 | 734 |
| 85 | 0.79 | 0.42 | 0.55 | 723 |
| 86 | 0.44 | 0.23 | 0.30 | 708 |
| 87 | 0.93 | 0.58 | 0.72 | 714 |
| 88 | 0.91 | 0.53 | 0.67 | 683 |
| 89 | 0.58 | 0.20 | 0.30 | 711 |
| 90 | 0.71 | 0.42 | 0.53 | 699 |
| 91 | 0.44 | 0.03 | 0.06 | 725 |
| 92 | 0.71 | 0.47 | 0.57 | 676 |
| 93 | 0.47 | 0.10 | 0.16 | 672 |
| 94 | 0.66 | 0.40 | 0.50 | 645 |
| 95 | 0.86 | 0.66 | 0.75 | 691 |
| 96 | 0.57 | 0.09 | 0.15 | 664 |
| 97 | 0.91 | 0.59 | 0.72 | 633 |
| 98 | 0.64 | 0.38 | 0.48 | 615 |
| 99 | 0.53 | 0.19 | 0.29 | 667 |
| 100 | 0.89 | 0.71 | 0.79 | 656 |
| 101 | 0.22 | 0.03 | 0.05 | 648 |
| 102 | 0.64 | 0.13 | 0.22 | 654 |
| 103 | 0.92 | 0.63 | 0.75 | 653 |
| 104 | 0.87 | 0.52 | 0.65 | 656 |
| 105 | 0.20 | 0.02 | 0.04 | 607 |
| 106 | 0.68 | 0.34 | 0.45 | 635 |
| 107 | 0.23 | 0.03 | 0.05 | 594 |

| | | | | |
|-----|------|------|------|-----|
| 108 | 0.40 | 0.18 | 0.25 | 592 |
| 109 | 0.32 | 0.07 | 0.12 | 604 |
| 110 | 0.46 | 0.21 | 0.29 | 606 |
| 111 | 0.70 | 0.39 | 0.50 | 567 |
| 112 | 0.68 | 0.27 | 0.38 | 571 |
| 113 | 0.61 | 0.36 | 0.45 | 578 |
| 114 | 0.47 | 0.18 | 0.26 | 564 |
| 115 | 0.35 | 0.13 | 0.19 | 537 |
| 116 | 0.93 | 0.66 | 0.77 | 583 |
| 117 | 0.59 | 0.09 | 0.15 | 534 |
| 118 | 0.66 | 0.35 | 0.46 | 566 |
| 119 | 0.20 | 0.04 | 0.07 | 567 |
| 120 | 0.48 | 0.16 | 0.24 | 497 |
| 121 | 0.55 | 0.19 | 0.29 | 536 |
| 122 | 0.24 | 0.05 | 0.08 | 528 |
| 123 | 0.81 | 0.53 | 0.64 | 550 |
| 124 | 0.50 | 0.21 | 0.29 | 563 |
| 125 | 0.35 | 0.06 | 0.10 | 545 |
| 126 | 0.49 | 0.18 | 0.27 | 544 |
| 127 | 0.95 | 0.76 | 0.84 | 549 |
| 128 | 0.63 | 0.34 | 0.44 | 495 |
| 129 | 0.94 | 0.59 | 0.73 | 509 |
| 130 | 0.34 | 0.11 | 0.16 | 501 |
| 131 | 0.28 | 0.04 | 0.07 | 524 |
| 132 | 0.48 | 0.26 | 0.34 | 485 |
| 133 | 0.55 | 0.37 | 0.45 | 515 |
| 134 | 0.32 | 0.04 | 0.08 | 536 |
| 135 | 0.77 | 0.38 | 0.51 | 526 |
| 136 | 0.67 | 0.34 | 0.45 | 493 |
| 137 | 0.40 | 0.08 | 0.14 | 501 |
| 138 | 0.31 | 0.05 | 0.09 | 501 |
| 139 | 0.29 | 0.02 | 0.04 | 523 |
| 140 | 0.88 | 0.64 | 0.74 | 508 |
| 141 | 0.33 | 0.11 | 0.16 | 490 |
| 142 | 0.77 | 0.50 | 0.60 | 482 |
| 143 | 0.49 | 0.25 | 0.33 | 461 |
| 144 | 0.74 | 0.48 | 0.58 | 496 |
| 145 | 0.62 | 0.17 | 0.26 | 521 |
| 146 | 0.39 | 0.13 | 0.19 | 481 |
| 147 | 0.00 | 0.00 | 0.00 | 486 |
| 148 | 0.37 | 0.09 | 0.14 | 497 |
| 149 | 0.54 | 0.09 | 0.16 | 470 |
| 150 | 0.37 | 0.11 | 0.17 | 459 |
| 151 | 0.74 | 0.45 | 0.56 | 464 |
| 152 | 0.50 | 0.24 | 0.32 | 482 |
| 153 | 0.46 | 0.09 | 0.15 | 507 |
| 154 | 0.29 | 0.04 | 0.07 | 503 |
| 155 | 0.90 | 0.59 | 0.71 | 456 |
| 156 | 0.50 | 0.27 | 0.35 | 480 |
| 157 | 0.54 | 0.26 | 0.35 | 443 |
| 158 | 0.92 | 0.70 | 0.80 | 457 |
| 159 | 0.57 | 0.08 | 0.13 | 478 |
| 160 | 0.16 | 0.03 | 0.05 | 470 |
| 161 | 0.37 | 0.18 | 0.24 | 468 |
| 162 | 0.24 | 0.05 | 0.09 | 428 |
| 163 | 0.40 | 0.08 | 0.13 | 462 |
| 164 | 0.73 | 0.32 | 0.45 | 493 |
| 165 | 0.02 | 0.68 | 0.70 | 127 |

| | | | | |
|-----|------|------|------|-----|
| 165 | 0.35 | 0.08 | 0.75 | 437 |
| 166 | 0.40 | 0.20 | 0.26 | 435 |
| 167 | 0.30 | 0.02 | 0.03 | 448 |
| 168 | 0.53 | 0.16 | 0.25 | 436 |
| 169 | 0.36 | 0.10 | 0.15 | 437 |
| 170 | 0.38 | 0.09 | 0.15 | 410 |
| 171 | 0.59 | 0.32 | 0.41 | 450 |
| 172 | 0.69 | 0.39 | 0.50 | 435 |
| 173 | 0.91 | 0.67 | 0.77 | 427 |
| 174 | 0.45 | 0.16 | 0.24 | 427 |
| 175 | 0.43 | 0.17 | 0.24 | 424 |
| 176 | 0.64 | 0.43 | 0.52 | 410 |
| 177 | 0.67 | 0.29 | 0.40 | 426 |
| 178 | 0.74 | 0.49 | 0.59 | 459 |
| 179 | 0.52 | 0.13 | 0.20 | 433 |
| 180 | 0.71 | 0.36 | 0.48 | 452 |
| 181 | 0.91 | 0.62 | 0.74 | 427 |
| 182 | 0.46 | 0.13 | 0.20 | 410 |
| 183 | 0.28 | 0.02 | 0.04 | 404 |
| 184 | 0.69 | 0.42 | 0.52 | 406 |
| 185 | 0.68 | 0.41 | 0.52 | 411 |
| 186 | 0.22 | 0.02 | 0.03 | 394 |
| 187 | 0.90 | 0.65 | 0.75 | 414 |
| 188 | 0.64 | 0.10 | 0.18 | 430 |
| 189 | 0.16 | 0.04 | 0.06 | 389 |
| 190 | 0.28 | 0.03 | 0.05 | 418 |
| 191 | 0.36 | 0.16 | 0.22 | 371 |
| 192 | 0.83 | 0.57 | 0.68 | 363 |
| 193 | 0.91 | 0.55 | 0.69 | 389 |
| 194 | 0.44 | 0.04 | 0.07 | 411 |
| 195 | 0.49 | 0.22 | 0.31 | 383 |
| 196 | 0.95 | 0.74 | 0.83 | 423 |
| 197 | 0.91 | 0.54 | 0.68 | 378 |
| 198 | 0.69 | 0.38 | 0.49 | 382 |
| 199 | 0.12 | 0.01 | 0.02 | 344 |
| 200 | 0.71 | 0.31 | 0.44 | 383 |
| 201 | 0.77 | 0.34 | 0.47 | 390 |
| 202 | 0.18 | 0.02 | 0.04 | 405 |
| 203 | 0.43 | 0.07 | 0.11 | 365 |
| 204 | 0.42 | 0.14 | 0.21 | 346 |
| 205 | 0.21 | 0.05 | 0.08 | 378 |
| 206 | 0.67 | 0.27 | 0.39 | 390 |
| 207 | 0.33 | 0.07 | 0.11 | 379 |
| 208 | 0.39 | 0.11 | 0.17 | 386 |
| 209 | 0.42 | 0.15 | 0.22 | 339 |
| 210 | 0.27 | 0.07 | 0.12 | 382 |
| 211 | 0.37 | 0.05 | 0.08 | 374 |
| 212 | 0.62 | 0.38 | 0.47 | 364 |
| 213 | 0.94 | 0.76 | 0.84 | 372 |
| 214 | 0.96 | 0.63 | 0.76 | 350 |
| 215 | 0.76 | 0.38 | 0.50 | 352 |
| 216 | 0.00 | 0.00 | 0.00 | 351 |
| 217 | 0.64 | 0.29 | 0.40 | 329 |
| 218 | 0.72 | 0.31 | 0.44 | 341 |
| 219 | 0.94 | 0.71 | 0.81 | 331 |
| 220 | 0.49 | 0.27 | 0.35 | 342 |
| 221 | 0.76 | 0.39 | 0.52 | 339 |
| 222 | 0.29 | 0.04 | 0.06 | 332 |

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|-----|------|------|------|-----|
| 223 | 0.43 | 0.12 | 0.18 | 327 |
| 224 | 0.31 | 0.06 | 0.11 | 324 |
| 225 | 0.51 | 0.21 | 0.30 | 352 |
| 226 | 0.65 | 0.30 | 0.41 | 317 |
| 227 | 0.54 | 0.12 | 0.20 | 355 |
| 228 | 0.57 | 0.19 | 0.29 | 341 |
| 229 | 0.58 | 0.37 | 0.46 | 334 |
| 230 | 0.64 | 0.49 | 0.56 | 304 |
| 231 | 0.43 | 0.04 | 0.07 | 321 |
| 232 | 0.77 | 0.50 | 0.61 | 311 |
| 233 | 0.32 | 0.10 | 0.15 | 312 |
| 234 | 0.09 | 0.01 | 0.02 | 306 |
| 235 | 0.03 | 0.00 | 0.01 | 305 |
| 236 | 0.16 | 0.02 | 0.04 | 340 |
| 237 | 0.58 | 0.30 | 0.40 | 316 |
| 238 | 0.65 | 0.23 | 0.34 | 297 |
| 239 | 0.35 | 0.13 | 0.19 | 305 |
| 240 | 0.73 | 0.44 | 0.55 | 310 |
| 241 | 0.67 | 0.36 | 0.47 | 307 |
| 242 | 0.58 | 0.16 | 0.25 | 316 |
| 243 | 0.26 | 0.07 | 0.11 | 314 |
| 244 | 0.51 | 0.12 | 0.19 | 316 |
| 245 | 0.67 | 0.46 | 0.55 | 313 |
| 246 | 0.79 | 0.46 | 0.58 | 325 |
| 247 | 0.60 | 0.36 | 0.45 | 291 |
| 248 | 0.33 | 0.01 | 0.02 | 311 |
| 249 | 0.57 | 0.24 | 0.33 | 314 |
| 250 | 0.38 | 0.05 | 0.09 | 309 |
| 251 | 0.30 | 0.08 | 0.13 | 300 |
| 252 | 0.55 | 0.27 | 0.36 | 325 |
| 253 | 0.76 | 0.51 | 0.61 | 316 |
| 254 | 0.43 | 0.09 | 0.15 | 306 |
| 255 | 0.54 | 0.19 | 0.28 | 289 |
| 256 | 0.49 | 0.11 | 0.18 | 304 |
| 257 | 0.16 | 0.02 | 0.04 | 268 |
| 258 | 0.85 | 0.58 | 0.69 | 266 |
| 259 | 0.06 | 0.00 | 0.01 | 298 |
| 260 | 0.55 | 0.36 | 0.43 | 292 |
| 261 | 0.25 | 0.05 | 0.08 | 289 |
| 262 | 0.50 | 0.01 | 0.01 | 305 |
| 263 | 0.00 | 0.00 | 0.00 | 281 |
| 264 | 0.59 | 0.25 | 0.35 | 295 |
| 265 | 0.16 | 0.02 | 0.04 | 281 |
| 266 | 0.83 | 0.52 | 0.64 | 269 |
| 267 | 0.45 | 0.12 | 0.19 | 312 |
| 268 | 0.75 | 0.40 | 0.52 | 294 |
| 269 | 0.34 | 0.05 | 0.09 | 285 |
| 270 | 0.56 | 0.33 | 0.42 | 279 |
| 271 | 0.50 | 0.28 | 0.36 | 269 |
| 272 | 0.59 | 0.38 | 0.46 | 277 |
| 273 | 0.69 | 0.31 | 0.43 | 272 |
| 274 | 0.36 | 0.01 | 0.03 | 285 |
| 275 | 0.94 | 0.69 | 0.80 | 295 |
| 276 | 0.46 | 0.19 | 0.27 | 283 |
| 277 | 0.65 | 0.29 | 0.40 | 250 |
| 278 | 0.57 | 0.20 | 0.30 | 281 |
| 279 | 0.86 | 0.58 | 0.69 | 270 |
| 280 | 0.62 | 0.35 | 0.44 | 272 |

| | | | | |
|-----|------|------|------|-----|
| 281 | 0.32 | 0.07 | 0.11 | 278 |
| 282 | 0.00 | 0.00 | 0.00 | 264 |
| 283 | 0.85 | 0.59 | 0.70 | 281 |
| 284 | 0.78 | 0.53 | 0.63 | 261 |
| 285 | 0.33 | 0.09 | 0.14 | 283 |
| 286 | 0.00 | 0.00 | 0.00 | 275 |
| 287 | 0.29 | 0.03 | 0.05 | 274 |
| 288 | 0.37 | 0.04 | 0.06 | 284 |
| 289 | 0.00 | 0.00 | 0.00 | 260 |
| 290 | 0.54 | 0.24 | 0.34 | 245 |
| 291 | 0.07 | 0.00 | 0.01 | 267 |
| 292 | 0.33 | 0.07 | 0.11 | 263 |
| 293 | 0.30 | 0.09 | 0.14 | 268 |
| 294 | 0.33 | 0.11 | 0.16 | 270 |
| 295 | 0.48 | 0.06 | 0.10 | 261 |
| 296 | 0.84 | 0.59 | 0.69 | 240 |
| 297 | 0.43 | 0.22 | 0.29 | 250 |
| 298 | 0.81 | 0.51 | 0.63 | 245 |
| 299 | 0.11 | 0.01 | 0.01 | 283 |
| 300 | 0.51 | 0.21 | 0.30 | 236 |
| 301 | 0.78 | 0.51 | 0.62 | 267 |
| 302 | 0.19 | 0.02 | 0.04 | 243 |
| 303 | 0.26 | 0.04 | 0.06 | 276 |
| 304 | 0.89 | 0.71 | 0.79 | 280 |
| 305 | 0.37 | 0.14 | 0.20 | 249 |
| 306 | 0.24 | 0.02 | 0.04 | 258 |
| 307 | 0.00 | 0.00 | 0.00 | 262 |
| 308 | 0.53 | 0.20 | 0.29 | 248 |
| 309 | 0.58 | 0.25 | 0.35 | 244 |
| 310 | 0.33 | 0.06 | 0.09 | 254 |
| 311 | 0.41 | 0.10 | 0.16 | 263 |
| 312 | 0.52 | 0.25 | 0.33 | 232 |
| 313 | 0.75 | 0.55 | 0.63 | 235 |
| 314 | 0.61 | 0.11 | 0.19 | 248 |
| 315 | 0.49 | 0.16 | 0.25 | 263 |
| 316 | 0.33 | 0.08 | 0.12 | 264 |
| 317 | 0.61 | 0.06 | 0.12 | 216 |
| 318 | 0.05 | 0.00 | 0.01 | 230 |
| 319 | 0.53 | 0.27 | 0.36 | 230 |
| 320 | 0.00 | 0.00 | 0.00 | 239 |
| 321 | 0.45 | 0.08 | 0.13 | 265 |
| 322 | 0.69 | 0.32 | 0.44 | 253 |
| 323 | 0.23 | 0.04 | 0.06 | 238 |
| 324 | 0.72 | 0.37 | 0.49 | 232 |
| 325 | 0.22 | 0.05 | 0.08 | 239 |
| 326 | 0.49 | 0.18 | 0.26 | 261 |
| 327 | 0.64 | 0.14 | 0.23 | 261 |
| 328 | 0.67 | 0.47 | 0.55 | 231 |
| 329 | 0.46 | 0.13 | 0.20 | 264 |
| 330 | 0.18 | 0.02 | 0.03 | 242 |
| 331 | 0.80 | 0.37 | 0.50 | 231 |
| 332 | 0.63 | 0.28 | 0.39 | 234 |
| 333 | 0.50 | 0.32 | 0.39 | 212 |
| 334 | 0.26 | 0.05 | 0.09 | 221 |
| 335 | 0.15 | 0.03 | 0.05 | 242 |
| 336 | 0.57 | 0.30 | 0.40 | 211 |
| 337 | 0.20 | 0.01 | 0.03 | 212 |
| 338 | 0.00 | 0.00 | 0.00 | 222 |

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|-----|------|------|------|-----|
| 338 | 0.00 | 0.00 | 0.00 | 222 |
| 339 | 0.22 | 0.02 | 0.04 | 227 |
| 340 | 0.66 | 0.30 | 0.41 | 216 |
| 341 | 0.57 | 0.26 | 0.36 | 231 |
| 342 | 0.45 | 0.22 | 0.29 | 233 |
| 343 | 0.17 | 0.03 | 0.04 | 232 |
| 344 | 0.28 | 0.02 | 0.04 | 209 |
| 345 | 0.37 | 0.11 | 0.17 | 216 |
| 346 | 0.27 | 0.09 | 0.13 | 222 |
| 347 | 0.48 | 0.19 | 0.28 | 243 |
| 348 | 0.51 | 0.26 | 0.35 | 222 |
| 349 | 0.57 | 0.12 | 0.20 | 228 |
| 350 | 0.44 | 0.12 | 0.18 | 205 |
| 351 | 0.58 | 0.30 | 0.39 | 177 |
| 352 | 0.77 | 0.39 | 0.52 | 234 |
| 353 | 0.96 | 0.57 | 0.71 | 230 |
| 354 | 0.47 | 0.21 | 0.29 | 195 |
| 355 | 0.90 | 0.42 | 0.57 | 209 |
| 356 | 0.06 | 0.00 | 0.01 | 205 |
| 357 | 0.50 | 0.11 | 0.18 | 211 |
| 358 | 0.43 | 0.16 | 0.23 | 230 |
| 359 | 0.27 | 0.08 | 0.12 | 211 |
| 360 | 0.39 | 0.09 | 0.14 | 221 |
| 361 | 0.24 | 0.04 | 0.08 | 200 |
| 362 | 0.82 | 0.15 | 0.25 | 219 |
| 363 | 0.36 | 0.07 | 0.12 | 222 |
| 364 | 0.62 | 0.27 | 0.38 | 213 |
| 365 | 0.94 | 0.36 | 0.52 | 199 |
| 366 | 0.80 | 0.37 | 0.51 | 200 |
| 367 | 0.76 | 0.29 | 0.42 | 199 |
| 368 | 0.57 | 0.26 | 0.36 | 212 |
| 369 | 0.93 | 0.71 | 0.80 | 214 |
| 370 | 0.10 | 0.02 | 0.03 | 197 |
| 371 | 0.20 | 0.03 | 0.05 | 212 |
| 372 | 0.41 | 0.14 | 0.21 | 210 |
| 373 | 0.43 | 0.03 | 0.05 | 211 |
| 374 | 0.41 | 0.15 | 0.22 | 213 |
| 375 | 0.00 | 0.00 | 0.00 | 216 |
| 376 | 0.87 | 0.53 | 0.66 | 195 |
| 377 | 0.95 | 0.67 | 0.79 | 187 |
| 378 | 0.15 | 0.03 | 0.04 | 191 |
| 379 | 0.17 | 0.02 | 0.04 | 178 |
| 380 | 0.79 | 0.48 | 0.60 | 193 |
| 381 | 0.13 | 0.02 | 0.04 | 187 |
| 382 | 0.67 | 0.03 | 0.06 | 193 |
| 383 | 0.17 | 0.04 | 0.06 | 204 |
| 384 | 0.28 | 0.15 | 0.19 | 193 |
| 385 | 0.12 | 0.02 | 0.04 | 207 |
| 386 | 0.84 | 0.45 | 0.59 | 211 |
| 387 | 0.06 | 0.00 | 0.01 | 210 |
| 388 | 0.31 | 0.04 | 0.06 | 223 |
| 389 | 0.24 | 0.09 | 0.13 | 203 |
| 390 | 0.72 | 0.24 | 0.36 | 199 |
| 391 | 0.40 | 0.08 | 0.13 | 200 |
| 392 | 0.22 | 0.05 | 0.09 | 183 |
| 393 | 0.62 | 0.31 | 0.41 | 189 |
| 394 | 0.96 | 0.66 | 0.78 | 194 |
| 395 | 0.53 | 0.18 | 0.27 | 183 |

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|-----|------|------|------|-----|
| 396 | 0.43 | 0.21 | 0.28 | 189 |
| 397 | 0.71 | 0.34 | 0.46 | 191 |
| 398 | 0.34 | 0.06 | 0.11 | 206 |
| 399 | 0.33 | 0.01 | 0.03 | 221 |
| 400 | 0.28 | 0.04 | 0.07 | 196 |
| 401 | 0.28 | 0.09 | 0.14 | 179 |
| 402 | 0.28 | 0.08 | 0.12 | 187 |
| 403 | 0.51 | 0.22 | 0.31 | 203 |
| 404 | 0.46 | 0.12 | 0.19 | 205 |
| 405 | 0.35 | 0.08 | 0.13 | 218 |
| 406 | 0.19 | 0.04 | 0.06 | 196 |
| 407 | 0.72 | 0.35 | 0.47 | 206 |
| 408 | 0.31 | 0.06 | 0.10 | 203 |
| 409 | 0.70 | 0.43 | 0.53 | 187 |
| 410 | 0.85 | 0.54 | 0.66 | 208 |
| 411 | 0.83 | 0.45 | 0.58 | 193 |
| 412 | 0.33 | 0.02 | 0.03 | 192 |
| 413 | 0.66 | 0.36 | 0.46 | 182 |
| 414 | 0.45 | 0.19 | 0.27 | 175 |
| 415 | 0.64 | 0.49 | 0.55 | 181 |
| 416 | 0.00 | 0.00 | 0.00 | 202 |
| 417 | 0.92 | 0.44 | 0.60 | 202 |
| 418 | 0.17 | 0.01 | 0.02 | 195 |
| 419 | 0.78 | 0.25 | 0.38 | 177 |
| 420 | 0.26 | 0.07 | 0.11 | 168 |
| 421 | 0.80 | 0.45 | 0.58 | 187 |
| 422 | 0.92 | 0.46 | 0.62 | 209 |
| 423 | 0.66 | 0.16 | 0.26 | 177 |
| 424 | 0.35 | 0.06 | 0.10 | 182 |
| 425 | 0.52 | 0.14 | 0.23 | 187 |
| 426 | 0.22 | 0.04 | 0.07 | 185 |
| 427 | 0.43 | 0.13 | 0.20 | 185 |
| 428 | 0.42 | 0.18 | 0.25 | 185 |
| 429 | 0.92 | 0.46 | 0.61 | 175 |
| 430 | 0.90 | 0.49 | 0.64 | 190 |
| 431 | 0.31 | 0.03 | 0.05 | 185 |
| 432 | 0.71 | 0.03 | 0.05 | 189 |
| 433 | 0.60 | 0.20 | 0.30 | 184 |
| 434 | 0.79 | 0.36 | 0.49 | 200 |
| 435 | 0.20 | 0.01 | 0.01 | 167 |
| 436 | 0.21 | 0.01 | 0.03 | 209 |
| 437 | 0.50 | 0.07 | 0.12 | 200 |
| 438 | 0.29 | 0.09 | 0.14 | 169 |
| 439 | 0.44 | 0.15 | 0.23 | 170 |
| 440 | 0.25 | 0.04 | 0.07 | 182 |
| 441 | 0.62 | 0.34 | 0.44 | 156 |
| 442 | 0.20 | 0.02 | 0.03 | 170 |
| 443 | 0.00 | 0.00 | 0.00 | 189 |
| 444 | 0.00 | 0.00 | 0.00 | 172 |
| 445 | 0.33 | 0.11 | 0.16 | 180 |
| 446 | 0.21 | 0.06 | 0.10 | 175 |
| 447 | 0.48 | 0.12 | 0.19 | 187 |
| 448 | 0.00 | 0.00 | 0.00 | 170 |
| 449 | 0.41 | 0.24 | 0.30 | 170 |
| 450 | 0.35 | 0.10 | 0.16 | 176 |
| 451 | 0.62 | 0.15 | 0.24 | 194 |
| 452 | 0.61 | 0.31 | 0.41 | 175 |
| 453 | 0.19 | 0.04 | 0.07 | 187 |

| | | | | |
|-----|------|------|------|-----|
| 454 | 0.11 | 0.01 | 0.01 | 181 |
| 455 | 0.62 | 0.14 | 0.23 | 177 |
| 456 | 0.50 | 0.18 | 0.26 | 170 |
| 457 | 0.24 | 0.03 | 0.05 | 182 |
| 458 | 0.68 | 0.37 | 0.48 | 172 |
| 459 | 0.00 | 0.00 | 0.00 | 190 |
| 460 | 0.43 | 0.16 | 0.23 | 183 |
| 461 | 0.94 | 0.63 | 0.75 | 182 |
| 462 | 0.35 | 0.16 | 0.22 | 173 |
| 463 | 0.91 | 0.69 | 0.79 | 171 |
| 464 | 0.58 | 0.27 | 0.37 | 173 |
| 465 | 0.77 | 0.41 | 0.53 | 184 |
| 466 | 0.72 | 0.22 | 0.34 | 175 |
| 467 | 0.43 | 0.19 | 0.26 | 162 |
| 468 | 0.12 | 0.01 | 0.02 | 176 |
| 469 | 0.91 | 0.46 | 0.61 | 177 |
| 470 | 0.52 | 0.07 | 0.13 | 167 |
| 471 | 0.27 | 0.06 | 0.10 | 192 |
| 472 | 0.50 | 0.32 | 0.39 | 168 |
| 473 | 0.32 | 0.05 | 0.09 | 188 |
| 474 | 0.31 | 0.05 | 0.08 | 163 |
| 475 | 0.44 | 0.17 | 0.24 | 160 |
| 476 | 0.89 | 0.56 | 0.69 | 180 |
| 477 | 0.92 | 0.46 | 0.61 | 182 |
| 478 | 0.49 | 0.27 | 0.35 | 171 |
| 479 | 0.57 | 0.18 | 0.27 | 174 |
| 480 | 0.96 | 0.52 | 0.68 | 162 |
| 481 | 0.21 | 0.04 | 0.06 | 169 |
| 482 | 0.33 | 0.03 | 0.06 | 157 |
| 483 | 0.77 | 0.48 | 0.59 | 200 |
| 484 | 0.58 | 0.21 | 0.31 | 177 |
| 485 | 0.51 | 0.26 | 0.34 | 175 |
| 486 | 0.64 | 0.51 | 0.57 | 185 |
| 487 | 0.96 | 0.52 | 0.67 | 167 |
| 488 | 0.00 | 0.00 | 0.00 | 192 |
| 489 | 0.30 | 0.09 | 0.14 | 176 |
| 490 | 0.00 | 0.00 | 0.00 | 167 |
| 491 | 0.33 | 0.01 | 0.01 | 177 |
| 492 | 0.47 | 0.26 | 0.33 | 160 |
| 493 | 0.46 | 0.22 | 0.30 | 159 |
| 494 | 0.15 | 0.03 | 0.04 | 159 |
| 495 | 0.31 | 0.10 | 0.15 | 162 |
| 496 | 0.82 | 0.46 | 0.59 | 167 |
| 497 | 0.17 | 0.02 | 0.03 | 168 |
| 498 | 0.40 | 0.12 | 0.19 | 154 |
| 499 | 0.00 | 0.00 | 0.00 | 184 |
| 500 | 0.14 | 0.03 | 0.05 | 167 |
| 501 | 0.41 | 0.20 | 0.27 | 153 |
| 502 | 0.78 | 0.55 | 0.65 | 143 |
| 503 | 0.22 | 0.07 | 0.10 | 177 |
| 504 | 0.69 | 0.32 | 0.44 | 177 |
| 505 | 0.90 | 0.50 | 0.64 | 152 |
| 506 | 0.80 | 0.40 | 0.54 | 179 |
| 507 | 0.60 | 0.12 | 0.20 | 171 |
| 508 | 0.61 | 0.28 | 0.39 | 151 |
| 509 | 0.51 | 0.23 | 0.32 | 162 |
| 510 | 0.63 | 0.24 | 0.35 | 158 |
| 511 | 0.18 | 0.03 | 0.05 | 164 |

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|-----|------|------|------|-----|
| 511 | 0.18 | 0.05 | 0.05 | 164 |
| 512 | 0.00 | 0.00 | 0.00 | 149 |
| 513 | 0.78 | 0.60 | 0.68 | 174 |
| 514 | 0.51 | 0.15 | 0.23 | 172 |
| 515 | 0.34 | 0.14 | 0.20 | 144 |
| 516 | 0.57 | 0.15 | 0.23 | 164 |
| 517 | 0.88 | 0.67 | 0.76 | 152 |
| 518 | 0.60 | 0.02 | 0.03 | 175 |
| 519 | 0.29 | 0.04 | 0.06 | 168 |
| 520 | 0.52 | 0.11 | 0.18 | 145 |
| 521 | 0.89 | 0.38 | 0.53 | 165 |
| 522 | 0.91 | 0.55 | 0.69 | 151 |
| 523 | 0.93 | 0.57 | 0.71 | 171 |
| 524 | 0.89 | 0.53 | 0.66 | 160 |
| 525 | 0.59 | 0.41 | 0.49 | 139 |
| 526 | 0.57 | 0.19 | 0.29 | 165 |
| 527 | 0.57 | 0.22 | 0.31 | 148 |
| 528 | 0.64 | 0.21 | 0.32 | 178 |
| 529 | 0.31 | 0.06 | 0.10 | 152 |
| 530 | 0.11 | 0.01 | 0.01 | 143 |
| 531 | 0.57 | 0.20 | 0.30 | 174 |
| 532 | 0.63 | 0.20 | 0.30 | 135 |
| 533 | 0.35 | 0.05 | 0.09 | 179 |
| 534 | 0.26 | 0.04 | 0.08 | 135 |
| 535 | 0.29 | 0.09 | 0.14 | 157 |
| 536 | 0.88 | 0.53 | 0.66 | 163 |
| 537 | 0.79 | 0.39 | 0.53 | 127 |
| 538 | 0.34 | 0.13 | 0.19 | 130 |
| 539 | 0.55 | 0.20 | 0.29 | 155 |
| 540 | 0.43 | 0.18 | 0.25 | 165 |
| 541 | 0.35 | 0.11 | 0.16 | 139 |
| 542 | 0.38 | 0.05 | 0.09 | 159 |
| 543 | 0.44 | 0.18 | 0.25 | 140 |
| 544 | 0.76 | 0.17 | 0.28 | 143 |
| 545 | 0.44 | 0.12 | 0.19 | 147 |
| 546 | 0.47 | 0.18 | 0.26 | 153 |
| 547 | 0.76 | 0.28 | 0.41 | 165 |
| 548 | 0.35 | 0.10 | 0.16 | 149 |
| 549 | 0.62 | 0.26 | 0.37 | 123 |
| 550 | 0.82 | 0.06 | 0.11 | 148 |
| 551 | 0.68 | 0.41 | 0.51 | 145 |
| 552 | 0.50 | 0.04 | 0.07 | 157 |
| 553 | 0.46 | 0.23 | 0.31 | 151 |
| 554 | 0.50 | 0.01 | 0.01 | 152 |
| 555 | 0.43 | 0.17 | 0.24 | 147 |
| 556 | 0.72 | 0.35 | 0.47 | 143 |
| 557 | 0.47 | 0.20 | 0.28 | 139 |
| 558 | 0.92 | 0.54 | 0.68 | 165 |
| 559 | 0.37 | 0.10 | 0.16 | 147 |
| 560 | 0.27 | 0.13 | 0.17 | 139 |
| 561 | 0.29 | 0.08 | 0.12 | 152 |
| 562 | 0.45 | 0.26 | 0.33 | 132 |
| 563 | 0.41 | 0.17 | 0.24 | 150 |
| 564 | 0.30 | 0.08 | 0.13 | 165 |
| 565 | 0.73 | 0.38 | 0.50 | 147 |
| 566 | 0.27 | 0.05 | 0.08 | 151 |
| 567 | 0.52 | 0.24 | 0.33 | 153 |
| 568 | 0.48 | 0.19 | 0.27 | 148 |

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|-----|------|------|------|-----|
| 569 | 0.17 | 0.04 | 0.06 | 142 |
| 570 | 0.11 | 0.02 | 0.04 | 140 |
| 571 | 0.07 | 0.01 | 0.01 | 149 |
| 572 | 1.00 | 0.02 | 0.04 | 146 |
| 573 | 0.51 | 0.29 | 0.37 | 135 |
| 574 | 0.73 | 0.24 | 0.36 | 137 |
| 575 | 0.50 | 0.11 | 0.18 | 142 |
| 576 | 0.24 | 0.10 | 0.14 | 145 |
| 577 | 0.82 | 0.25 | 0.38 | 145 |
| 578 | 0.72 | 0.33 | 0.45 | 131 |
| 579 | 0.40 | 0.15 | 0.22 | 142 |
| 580 | 0.00 | 0.00 | 0.00 | 143 |
| 581 | 0.38 | 0.09 | 0.15 | 139 |
| 582 | 0.57 | 0.15 | 0.24 | 150 |
| 583 | 0.00 | 0.00 | 0.00 | 121 |
| 584 | 0.57 | 0.28 | 0.38 | 148 |
| 585 | 0.61 | 0.41 | 0.49 | 134 |
| 586 | 0.64 | 0.37 | 0.47 | 151 |
| 587 | 0.74 | 0.11 | 0.20 | 150 |
| 588 | 0.48 | 0.11 | 0.18 | 141 |
| 589 | 0.20 | 0.03 | 0.05 | 137 |
| 590 | 0.79 | 0.36 | 0.50 | 154 |
| 591 | 0.52 | 0.22 | 0.31 | 126 |
| 592 | 0.85 | 0.49 | 0.62 | 144 |
| 593 | 0.29 | 0.06 | 0.10 | 130 |
| 594 | 0.46 | 0.15 | 0.22 | 148 |
| 595 | 0.13 | 0.02 | 0.03 | 115 |
| 596 | 0.64 | 0.46 | 0.53 | 142 |
| 597 | 0.95 | 0.46 | 0.62 | 123 |
| 598 | 0.63 | 0.21 | 0.32 | 150 |
| 599 | 0.00 | 0.00 | 0.00 | 134 |
| 600 | 0.24 | 0.04 | 0.07 | 154 |
| 601 | 0.36 | 0.08 | 0.14 | 165 |
| 602 | 0.50 | 0.02 | 0.04 | 150 |
| 603 | 0.49 | 0.15 | 0.23 | 137 |
| 604 | 0.89 | 0.53 | 0.67 | 133 |
| 605 | 0.38 | 0.14 | 0.21 | 146 |
| 606 | 0.88 | 0.12 | 0.21 | 129 |
| 607 | 0.17 | 0.03 | 0.05 | 151 |
| 608 | 0.86 | 0.55 | 0.67 | 138 |
| 609 | 0.36 | 0.13 | 0.19 | 124 |
| 610 | 0.40 | 0.01 | 0.03 | 144 |
| 611 | 0.00 | 0.00 | 0.00 | 150 |
| 612 | 0.00 | 0.00 | 0.00 | 130 |
| 613 | 0.21 | 0.05 | 0.08 | 127 |
| 614 | 0.41 | 0.17 | 0.24 | 141 |
| 615 | 0.10 | 0.02 | 0.03 | 133 |
| 616 | 0.54 | 0.29 | 0.38 | 132 |
| 617 | 0.67 | 0.02 | 0.03 | 131 |
| 618 | 0.21 | 0.03 | 0.06 | 125 |
| 619 | 0.63 | 0.37 | 0.46 | 123 |
| 620 | 0.00 | 0.00 | 0.00 | 148 |
| 621 | 0.12 | 0.01 | 0.02 | 117 |
| 622 | 0.72 | 0.47 | 0.57 | 129 |
| 623 | 0.36 | 0.04 | 0.06 | 113 |
| 624 | 0.88 | 0.51 | 0.64 | 110 |
| 625 | 0.92 | 0.63 | 0.75 | 121 |
| 626 | 0.22 | 0.08 | 0.12 | 125 |

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|-----|------|------|------|-----|
| 627 | 0.95 | 0.59 | 0.73 | 132 |
| 628 | 0.67 | 0.30 | 0.42 | 116 |
| 629 | 0.81 | 0.38 | 0.52 | 126 |
| 630 | 0.29 | 0.04 | 0.07 | 126 |
| 631 | 0.28 | 0.06 | 0.10 | 148 |
| 632 | 0.91 | 0.61 | 0.74 | 140 |
| 633 | 0.50 | 0.02 | 0.03 | 128 |
| 634 | 0.40 | 0.16 | 0.22 | 128 |
| 635 | 0.00 | 0.00 | 0.00 | 140 |
| 636 | 0.95 | 0.41 | 0.57 | 130 |
| 637 | 0.62 | 0.23 | 0.34 | 126 |
| 638 | 0.75 | 0.08 | 0.15 | 143 |
| 639 | 0.67 | 0.31 | 0.42 | 121 |
| 640 | 0.16 | 0.04 | 0.07 | 117 |
| 641 | 0.36 | 0.12 | 0.19 | 112 |
| 642 | 0.46 | 0.14 | 0.21 | 137 |
| 643 | 0.96 | 0.61 | 0.74 | 141 |
| 644 | 0.71 | 0.37 | 0.49 | 127 |
| 645 | 0.28 | 0.06 | 0.10 | 128 |
| 646 | 0.10 | 0.01 | 0.01 | 124 |
| 647 | 0.11 | 0.03 | 0.05 | 138 |
| 648 | 0.13 | 0.03 | 0.04 | 119 |
| 649 | 0.00 | 0.00 | 0.00 | 137 |
| 650 | 0.33 | 0.01 | 0.02 | 121 |
| 651 | 0.07 | 0.02 | 0.03 | 108 |
| 652 | 0.72 | 0.41 | 0.52 | 122 |
| 653 | 0.61 | 0.26 | 0.36 | 139 |
| 654 | 0.40 | 0.02 | 0.03 | 112 |
| 655 | 0.53 | 0.14 | 0.22 | 125 |
| 656 | 0.64 | 0.19 | 0.29 | 124 |
| 657 | 0.30 | 0.08 | 0.12 | 117 |
| 658 | 0.50 | 0.20 | 0.28 | 116 |
| 659 | 0.37 | 0.08 | 0.14 | 130 |
| 660 | 0.15 | 0.02 | 0.03 | 121 |
| 661 | 0.75 | 0.35 | 0.48 | 124 |
| 662 | 0.48 | 0.12 | 0.19 | 121 |
| 663 | 0.84 | 0.63 | 0.72 | 126 |
| 664 | 0.00 | 0.00 | 0.00 | 118 |
| 665 | 0.18 | 0.06 | 0.09 | 113 |
| 666 | 0.00 | 0.00 | 0.00 | 128 |
| 667 | 0.53 | 0.12 | 0.20 | 139 |
| 668 | 0.29 | 0.04 | 0.07 | 131 |
| 669 | 0.26 | 0.05 | 0.08 | 127 |
| 670 | 0.47 | 0.07 | 0.12 | 125 |
| 671 | 0.33 | 0.02 | 0.03 | 111 |
| 672 | 0.55 | 0.37 | 0.44 | 127 |
| 673 | 0.72 | 0.48 | 0.57 | 130 |
| 674 | 0.19 | 0.02 | 0.04 | 130 |
| 675 | 0.60 | 0.20 | 0.30 | 126 |
| 676 | 0.15 | 0.02 | 0.03 | 104 |
| 677 | 0.53 | 0.14 | 0.22 | 127 |
| 678 | 0.57 | 0.15 | 0.24 | 130 |
| 679 | 0.26 | 0.10 | 0.14 | 112 |
| 680 | 0.43 | 0.09 | 0.15 | 131 |
| 681 | 0.00 | 0.00 | 0.00 | 140 |
| 682 | 0.53 | 0.35 | 0.42 | 114 |
| 683 | 0.78 | 0.12 | 0.22 | 112 |
| 684 | 0.35 | 0.05 | 0.10 | 115 |

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|-----|------|------|------|-----|
| 684 | 0.35 | 0.06 | 0.10 | 115 |
| 685 | 0.66 | 0.15 | 0.24 | 128 |
| 686 | 0.57 | 0.10 | 0.17 | 122 |
| 687 | 0.25 | 0.03 | 0.05 | 109 |
| 688 | 0.29 | 0.02 | 0.03 | 108 |
| 689 | 0.00 | 0.00 | 0.00 | 125 |
| 690 | 0.50 | 0.01 | 0.02 | 117 |
| 691 | 0.36 | 0.09 | 0.15 | 127 |
| 692 | 0.80 | 0.35 | 0.49 | 129 |
| 693 | 0.42 | 0.16 | 0.23 | 118 |
| 694 | 0.72 | 0.37 | 0.49 | 151 |
| 695 | 0.67 | 0.29 | 0.41 | 112 |
| 696 | 0.81 | 0.22 | 0.34 | 119 |
| 697 | 0.19 | 0.05 | 0.07 | 109 |
| 698 | 0.58 | 0.33 | 0.42 | 122 |
| 699 | 0.96 | 0.49 | 0.65 | 102 |
| 700 | 0.29 | 0.07 | 0.11 | 102 |
| 701 | 0.46 | 0.26 | 0.33 | 107 |
| 702 | 0.25 | 0.03 | 0.05 | 105 |
| 703 | 0.25 | 0.01 | 0.02 | 113 |
| 704 | 0.62 | 0.27 | 0.37 | 98 |
| 705 | 0.21 | 0.05 | 0.08 | 100 |
| 706 | 0.72 | 0.33 | 0.45 | 131 |
| 707 | 0.45 | 0.21 | 0.29 | 112 |
| 708 | 0.44 | 0.03 | 0.06 | 119 |
| 709 | 0.28 | 0.07 | 0.11 | 105 |
| 710 | 0.18 | 0.03 | 0.04 | 117 |
| 711 | 0.39 | 0.14 | 0.21 | 115 |
| 712 | 0.41 | 0.10 | 0.16 | 129 |
| 713 | 0.68 | 0.27 | 0.38 | 101 |
| 714 | 0.57 | 0.10 | 0.17 | 122 |
| 715 | 0.00 | 0.00 | 0.00 | 97 |
| 716 | 0.38 | 0.16 | 0.23 | 116 |
| 717 | 0.43 | 0.08 | 0.14 | 110 |
| 718 | 0.38 | 0.04 | 0.08 | 113 |
| 719 | 0.75 | 0.49 | 0.59 | 110 |
| 720 | 0.78 | 0.05 | 0.10 | 130 |
| 721 | 0.00 | 0.00 | 0.00 | 104 |
| 722 | 0.89 | 0.66 | 0.75 | 119 |
| 723 | 0.00 | 0.00 | 0.00 | 108 |
| 724 | 0.43 | 0.22 | 0.29 | 112 |
| 725 | 0.32 | 0.05 | 0.08 | 126 |
| 726 | 0.93 | 0.67 | 0.78 | 120 |
| 727 | 0.30 | 0.05 | 0.09 | 130 |
| 728 | 0.67 | 0.02 | 0.04 | 103 |
| 729 | 0.70 | 0.17 | 0.28 | 111 |
| 730 | 0.33 | 0.03 | 0.05 | 110 |
| 731 | 0.00 | 0.00 | 0.00 | 96 |
| 732 | 0.55 | 0.05 | 0.10 | 112 |
| 733 | 0.39 | 0.08 | 0.13 | 90 |
| 734 | 0.28 | 0.11 | 0.15 | 95 |
| 735 | 0.80 | 0.39 | 0.52 | 116 |
| 736 | 0.40 | 0.02 | 0.03 | 128 |
| 737 | 0.25 | 0.09 | 0.13 | 93 |
| 738 | 0.89 | 0.15 | 0.26 | 107 |
| 739 | 0.58 | 0.29 | 0.39 | 99 |
| 740 | 0.40 | 0.04 | 0.07 | 105 |
| 741 | 0.46 | 0.05 | 0.09 | 116 |

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|-----|------|------|------|-----|
| 742 | 0.68 | 0.43 | 0.53 | 105 |
| 743 | 0.40 | 0.19 | 0.26 | 84 |
| 744 | 0.44 | 0.14 | 0.21 | 102 |
| 745 | 0.69 | 0.23 | 0.34 | 111 |
| 746 | 0.36 | 0.10 | 0.15 | 104 |
| 747 | 0.44 | 0.14 | 0.21 | 110 |
| 748 | 0.58 | 0.21 | 0.30 | 92 |
| 749 | 0.87 | 0.57 | 0.69 | 106 |
| 750 | 0.00 | 0.00 | 0.00 | 116 |
| 751 | 0.28 | 0.09 | 0.14 | 109 |
| 752 | 0.85 | 0.54 | 0.66 | 104 |
| 753 | 1.00 | 0.01 | 0.02 | 119 |
| 754 | 0.27 | 0.06 | 0.10 | 96 |
| 755 | 0.17 | 0.04 | 0.06 | 104 |
| 756 | 0.00 | 0.00 | 0.00 | 101 |
| 757 | 0.50 | 0.19 | 0.28 | 114 |
| 758 | 0.00 | 0.00 | 0.00 | 112 |
| 759 | 0.67 | 0.04 | 0.08 | 95 |
| 760 | 0.00 | 0.00 | 0.00 | 102 |
| 761 | 0.31 | 0.11 | 0.17 | 105 |
| 762 | 0.57 | 0.25 | 0.35 | 109 |
| 763 | 0.09 | 0.01 | 0.02 | 112 |
| 764 | 0.94 | 0.40 | 0.56 | 116 |
| 765 | 0.60 | 0.31 | 0.41 | 109 |
| 766 | 0.00 | 0.00 | 0.00 | 96 |
| 767 | 0.50 | 0.09 | 0.15 | 114 |
| 768 | 0.00 | 0.00 | 0.00 | 99 |
| 769 | 0.65 | 0.15 | 0.25 | 98 |
| 770 | 0.48 | 0.21 | 0.30 | 107 |
| 771 | 0.00 | 0.00 | 0.00 | 103 |
| 772 | 0.00 | 0.00 | 0.00 | 96 |
| 773 | 0.00 | 0.00 | 0.00 | 106 |
| 774 | 0.76 | 0.33 | 0.46 | 97 |
| 775 | 0.27 | 0.03 | 0.06 | 91 |
| 776 | 0.00 | 0.00 | 0.00 | 101 |
| 777 | 0.76 | 0.38 | 0.50 | 109 |
| 778 | 0.00 | 0.00 | 0.00 | 104 |
| 779 | 0.33 | 0.08 | 0.13 | 116 |
| 780 | 0.00 | 0.00 | 0.00 | 102 |
| 781 | 0.85 | 0.26 | 0.40 | 106 |
| 782 | 0.64 | 0.15 | 0.24 | 108 |
| 783 | 0.80 | 0.08 | 0.15 | 95 |
| 784 | 0.91 | 0.36 | 0.52 | 108 |
| 785 | 0.94 | 0.43 | 0.59 | 113 |
| 786 | 0.40 | 0.06 | 0.10 | 109 |
| 787 | 0.78 | 0.41 | 0.54 | 112 |
| 788 | 0.00 | 0.00 | 0.00 | 104 |
| 789 | 0.43 | 0.17 | 0.25 | 92 |
| 790 | 0.44 | 0.06 | 0.11 | 116 |
| 791 | 0.29 | 0.04 | 0.07 | 96 |
| 792 | 0.58 | 0.15 | 0.24 | 118 |
| 793 | 0.64 | 0.27 | 0.38 | 106 |
| 794 | 0.26 | 0.06 | 0.10 | 93 |
| 795 | 0.80 | 0.31 | 0.45 | 103 |
| 796 | 0.39 | 0.12 | 0.18 | 104 |
| 797 | 0.57 | 0.09 | 0.16 | 89 |
| 798 | 0.55 | 0.06 | 0.11 | 97 |
| 799 | 0.00 | 0.00 | 0.00 | 92 |

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|-----|------|------|------|-----|
| 800 | 0.55 | 0.14 | 0.22 | 85 |
| 801 | 1.00 | 0.04 | 0.08 | 93 |
| 802 | 0.79 | 0.28 | 0.41 | 93 |
| 803 | 0.36 | 0.13 | 0.19 | 102 |
| 804 | 0.65 | 0.12 | 0.20 | 108 |
| 805 | 0.87 | 0.37 | 0.52 | 111 |
| 806 | 0.61 | 0.14 | 0.23 | 98 |
| 807 | 0.20 | 0.03 | 0.06 | 94 |
| 808 | 0.15 | 0.02 | 0.04 | 84 |
| 809 | 0.84 | 0.32 | 0.46 | 100 |
| 810 | 0.22 | 0.02 | 0.04 | 92 |
| 811 | 0.37 | 0.11 | 0.17 | 88 |
| 812 | 0.39 | 0.13 | 0.20 | 104 |
| 813 | 0.50 | 0.04 | 0.08 | 90 |
| 814 | 0.38 | 0.07 | 0.12 | 109 |
| 815 | 0.23 | 0.04 | 0.06 | 81 |
| 816 | 0.70 | 0.22 | 0.33 | 96 |
| 817 | 0.98 | 0.53 | 0.69 | 88 |
| 818 | 0.56 | 0.24 | 0.33 | 101 |
| 819 | 0.94 | 0.45 | 0.61 | 103 |
| 820 | 0.00 | 0.00 | 0.00 | 94 |
| 821 | 0.72 | 0.17 | 0.27 | 108 |
| 822 | 0.29 | 0.06 | 0.09 | 90 |
| 823 | 0.81 | 0.44 | 0.57 | 97 |
| 824 | 0.50 | 0.02 | 0.04 | 90 |
| 825 | 0.52 | 0.23 | 0.32 | 102 |
| 826 | 0.12 | 0.01 | 0.02 | 85 |
| 827 | 0.20 | 0.02 | 0.03 | 109 |
| 828 | 0.30 | 0.03 | 0.05 | 103 |
| 829 | 0.98 | 0.40 | 0.56 | 106 |
| 830 | 0.88 | 0.26 | 0.40 | 108 |
| 831 | 0.50 | 0.04 | 0.07 | 84 |
| 832 | 0.00 | 0.00 | 0.00 | 98 |
| 833 | 0.77 | 0.26 | 0.39 | 92 |
| 834 | 0.50 | 0.10 | 0.17 | 91 |
| 835 | 0.87 | 0.28 | 0.43 | 92 |
| 836 | 0.28 | 0.07 | 0.11 | 104 |
| 837 | 0.63 | 0.24 | 0.34 | 102 |
| 838 | 0.22 | 0.07 | 0.11 | 111 |
| 839 | 0.00 | 0.00 | 0.00 | 96 |
| 840 | 0.41 | 0.15 | 0.22 | 86 |
| 841 | 0.34 | 0.10 | 0.16 | 105 |
| 842 | 0.20 | 0.01 | 0.02 | 92 |
| 843 | 0.39 | 0.16 | 0.23 | 86 |
| 844 | 0.00 | 0.00 | 0.00 | 108 |
| 845 | 0.45 | 0.06 | 0.11 | 82 |
| 846 | 0.22 | 0.04 | 0.07 | 101 |
| 847 | 0.97 | 0.60 | 0.74 | 94 |
| 848 | 1.00 | 0.41 | 0.58 | 101 |
| 849 | 0.39 | 0.14 | 0.20 | 88 |
| 850 | 0.88 | 0.36 | 0.51 | 81 |
| 851 | 0.79 | 0.10 | 0.18 | 109 |
| 852 | 0.45 | 0.13 | 0.20 | 101 |
| 853 | 0.25 | 0.03 | 0.06 | 91 |
| 854 | 0.29 | 0.06 | 0.10 | 95 |
| 855 | 0.20 | 0.01 | 0.02 | 99 |
| 856 | 0.14 | 0.01 | 0.02 | 79 |

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|-----|------|------|------|-----|
| 857 | 0.67 | 0.32 | 0.43 | 91 |
| 858 | 0.00 | 0.00 | 0.00 | 89 |
| 859 | 0.42 | 0.09 | 0.15 | 91 |
| 860 | 0.49 | 0.19 | 0.28 | 88 |
| 861 | 0.32 | 0.07 | 0.11 | 101 |
| 862 | 0.51 | 0.30 | 0.37 | 81 |
| 863 | 0.69 | 0.20 | 0.31 | 101 |
| 864 | 0.28 | 0.11 | 0.16 | 80 |
| 865 | 0.00 | 0.00 | 0.00 | 97 |
| 866 | 0.88 | 0.46 | 0.60 | 94 |
| 867 | 0.00 | 0.00 | 0.00 | 97 |
| 868 | 0.29 | 0.07 | 0.11 | 91 |
| 869 | 0.35 | 0.09 | 0.14 | 88 |
| 870 | 0.53 | 0.25 | 0.34 | 112 |
| 871 | 0.93 | 0.57 | 0.71 | 94 |
| 872 | 0.00 | 0.00 | 0.00 | 84 |
| 873 | 0.89 | 0.53 | 0.66 | 74 |
| 874 | 0.91 | 0.53 | 0.67 | 80 |
| 875 | 0.46 | 0.23 | 0.31 | 79 |
| 876 | 0.56 | 0.07 | 0.12 | 71 |
| 877 | 0.77 | 0.26 | 0.39 | 92 |
| 878 | 1.00 | 0.08 | 0.15 | 99 |
| 879 | 0.56 | 0.14 | 0.23 | 98 |
| 880 | 0.37 | 0.18 | 0.24 | 82 |
| 881 | 0.70 | 0.35 | 0.47 | 80 |
| 882 | 0.91 | 0.55 | 0.69 | 94 |
| 883 | 0.07 | 0.01 | 0.02 | 102 |
| 884 | 0.88 | 0.22 | 0.35 | 95 |
| 885 | 0.91 | 0.57 | 0.70 | 87 |
| 886 | 0.20 | 0.01 | 0.02 | 88 |
| 887 | 0.41 | 0.08 | 0.13 | 90 |
| 888 | 0.84 | 0.46 | 0.60 | 104 |
| 889 | 0.20 | 0.01 | 0.02 | 93 |
| 890 | 0.14 | 0.02 | 0.04 | 83 |
| 891 | 0.00 | 0.00 | 0.00 | 92 |
| 892 | 0.58 | 0.17 | 0.26 | 88 |
| 893 | 0.00 | 0.00 | 0.00 | 74 |
| 894 | 1.00 | 0.40 | 0.57 | 98 |
| 895 | 0.47 | 0.22 | 0.30 | 73 |
| 896 | 0.00 | 0.00 | 0.00 | 87 |
| 897 | 0.29 | 0.03 | 0.05 | 73 |
| 898 | 0.58 | 0.22 | 0.32 | 86 |
| 899 | 0.24 | 0.08 | 0.12 | 100 |
| 900 | 0.43 | 0.14 | 0.21 | 93 |
| 901 | 0.82 | 0.36 | 0.50 | 86 |
| 902 | 0.38 | 0.07 | 0.12 | 107 |
| 903 | 0.43 | 0.03 | 0.06 | 97 |
| 904 | 0.52 | 0.17 | 0.26 | 88 |
| 905 | 0.00 | 0.00 | 0.00 | 94 |
| 906 | 0.14 | 0.02 | 0.04 | 83 |
| 907 | 0.00 | 0.00 | 0.00 | 85 |
| 908 | 0.00 | 0.00 | 0.00 | 90 |
| 909 | 0.14 | 0.01 | 0.02 | 83 |
| 910 | 0.60 | 0.07 | 0.13 | 83 |
| 911 | 0.19 | 0.03 | 0.06 | 87 |
| 912 | 0.94 | 0.38 | 0.54 | 87 |
| 913 | 0.56 | 0.10 | 0.18 | 86 |
| 914 | 0.52 | 0.16 | 0.25 | 91 |

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|-----|------|------|------|----|
| 915 | 0.25 | 0.02 | 0.04 | 87 |
| 916 | 0.00 | 0.00 | 0.00 | 92 |
| 917 | 0.00 | 0.00 | 0.00 | 92 |
| 918 | 0.81 | 0.37 | 0.51 | 78 |
| 919 | 0.44 | 0.10 | 0.16 | 81 |
| 920 | 0.00 | 0.00 | 0.00 | 87 |
| 921 | 0.00 | 0.00 | 0.00 | 95 |
| 922 | 0.85 | 0.27 | 0.41 | 82 |
| 923 | 0.33 | 0.02 | 0.04 | 89 |
| 924 | 0.00 | 0.00 | 0.00 | 73 |
| 925 | 0.41 | 0.09 | 0.14 | 82 |
| 926 | 0.43 | 0.03 | 0.06 | 91 |
| 927 | 0.38 | 0.10 | 0.15 | 83 |
| 928 | 0.33 | 0.03 | 0.05 | 79 |
| 929 | 0.55 | 0.07 | 0.12 | 89 |
| 930 | 0.29 | 0.07 | 0.11 | 85 |
| 931 | 0.00 | 0.00 | 0.00 | 95 |
| 932 | 0.25 | 0.01 | 0.02 | 80 |
| 933 | 0.50 | 0.07 | 0.12 | 72 |
| 934 | 0.64 | 0.29 | 0.40 | 79 |
| 935 | 0.52 | 0.15 | 0.23 | 75 |
| 936 | 0.70 | 0.22 | 0.34 | 85 |
| 937 | 0.47 | 0.09 | 0.16 | 75 |
| 938 | 0.23 | 0.09 | 0.13 | 69 |
| 939 | 0.00 | 0.00 | 0.00 | 85 |
| 940 | 0.11 | 0.01 | 0.02 | 72 |
| 941 | 0.00 | 0.00 | 0.00 | 69 |
| 942 | 0.44 | 0.09 | 0.14 | 94 |
| 943 | 0.00 | 0.00 | 0.00 | 85 |
| 944 | 0.94 | 0.36 | 0.52 | 89 |
| 945 | 0.19 | 0.04 | 0.06 | 77 |
| 946 | 0.78 | 0.15 | 0.25 | 93 |
| 947 | 0.00 | 0.00 | 0.00 | 81 |
| 948 | 0.95 | 0.50 | 0.66 | 78 |
| 949 | 0.00 | 0.00 | 0.00 | 75 |
| 950 | 0.00 | 0.00 | 0.00 | 80 |
| 951 | 0.12 | 0.01 | 0.02 | 88 |
| 952 | 0.29 | 0.03 | 0.05 | 80 |
| 953 | 1.00 | 0.71 | 0.83 | 85 |
| 954 | 0.83 | 0.55 | 0.66 | 71 |
| 955 | 0.00 | 0.00 | 0.00 | 80 |
| 956 | 0.81 | 0.37 | 0.51 | 68 |
| 957 | 0.87 | 0.52 | 0.65 | 75 |
| 958 | 0.43 | 0.13 | 0.20 | 90 |
| 959 | 0.81 | 0.15 | 0.25 | 87 |
| 960 | 0.89 | 0.38 | 0.53 | 87 |
| 961 | 0.74 | 0.29 | 0.42 | 68 |
| 962 | 0.65 | 0.26 | 0.37 | 86 |
| 963 | 0.57 | 0.19 | 0.28 | 85 |
| 964 | 0.43 | 0.15 | 0.23 | 78 |
| 965 | 0.76 | 0.44 | 0.56 | 88 |
| 966 | 0.93 | 0.46 | 0.61 | 85 |
| 967 | 0.52 | 0.23 | 0.32 | 70 |
| 968 | 0.33 | 0.04 | 0.07 | 82 |
| 969 | 0.88 | 0.47 | 0.61 | 92 |
| 970 | 0.31 | 0.05 | 0.09 | 73 |
| 971 | 0.00 | 0.00 | 0.00 | 77 |
| 972 | 0.46 | 0.16 | 0.24 | 82 |

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|------|------|------|------|----|
| 972 | 0.75 | 0.10 | 0.27 | 82 |
| 973 | 0.80 | 0.10 | 0.18 | 80 |
| 974 | 0.12 | 0.01 | 0.02 | 83 |
| 975 | 0.98 | 0.58 | 0.73 | 76 |
| 976 | 0.00 | 0.00 | 0.00 | 85 |
| 977 | 0.00 | 0.00 | 0.00 | 65 |
| 978 | 0.57 | 0.11 | 0.19 | 72 |
| 979 | 0.33 | 0.02 | 0.04 | 85 |
| 980 | 0.23 | 0.05 | 0.08 | 64 |
| 981 | 0.25 | 0.03 | 0.05 | 76 |
| 982 | 0.58 | 0.07 | 0.13 | 96 |
| 983 | 0.94 | 0.31 | 0.46 | 94 |
| 984 | 0.29 | 0.02 | 0.04 | 87 |
| 985 | 0.33 | 0.01 | 0.03 | 75 |
| 986 | 0.00 | 0.00 | 0.00 | 79 |
| 987 | 0.00 | 0.00 | 0.00 | 86 |
| 988 | 0.50 | 0.01 | 0.02 | 88 |
| 989 | 0.00 | 0.00 | 0.00 | 84 |
| 990 | 0.52 | 0.14 | 0.22 | 95 |
| 991 | 0.37 | 0.15 | 0.22 | 71 |
| 992 | 0.57 | 0.38 | 0.46 | 68 |
| 993 | 0.00 | 0.00 | 0.00 | 75 |
| 994 | 0.00 | 0.00 | 0.00 | 90 |
| 995 | 0.95 | 0.43 | 0.60 | 83 |
| 996 | 0.89 | 0.43 | 0.58 | 79 |
| 997 | 0.71 | 0.08 | 0.14 | 64 |
| 998 | 0.27 | 0.04 | 0.07 | 74 |
| 999 | 0.81 | 0.36 | 0.50 | 81 |
| 1000 | 0.00 | 0.00 | 0.00 | 74 |
| 1001 | 0.14 | 0.02 | 0.03 | 62 |
| 1002 | 0.67 | 0.25 | 0.37 | 71 |
| 1003 | 0.00 | 0.00 | 0.00 | 72 |
| 1004 | 0.50 | 0.08 | 0.14 | 75 |
| 1005 | 0.93 | 0.53 | 0.67 | 72 |
| 1006 | 0.52 | 0.15 | 0.23 | 81 |
| 1007 | 0.00 | 0.00 | 0.00 | 74 |
| 1008 | 0.17 | 0.01 | 0.03 | 72 |
| 1009 | 0.00 | 0.00 | 0.00 | 75 |
| 1010 | 0.47 | 0.16 | 0.24 | 91 |
| 1011 | 0.59 | 0.18 | 0.27 | 90 |
| 1012 | 0.62 | 0.25 | 0.36 | 80 |
| 1013 | 0.00 | 0.00 | 0.00 | 88 |
| 1014 | 0.80 | 0.06 | 0.11 | 71 |
| 1015 | 0.57 | 0.11 | 0.18 | 74 |
| 1016 | 0.88 | 0.22 | 0.35 | 68 |
| 1017 | 0.70 | 0.39 | 0.50 | 71 |
| 1018 | 0.65 | 0.21 | 0.32 | 80 |
| 1019 | 0.00 | 0.00 | 0.00 | 83 |
| 1020 | 0.46 | 0.08 | 0.14 | 74 |
| 1021 | 0.93 | 0.49 | 0.64 | 78 |
| 1022 | 0.86 | 0.32 | 0.47 | 77 |
| 1023 | 0.12 | 0.01 | 0.02 | 78 |
| 1024 | 0.68 | 0.31 | 0.43 | 67 |
| 1025 | 0.50 | 0.01 | 0.02 | 80 |
| 1026 | 0.69 | 0.23 | 0.35 | 77 |
| 1027 | 0.80 | 0.32 | 0.46 | 88 |
| 1028 | 0.24 | 0.06 | 0.09 | 70 |
| 1029 | 0.00 | 0.00 | 0.00 | 79 |

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|------|------|------|------|----|
| 1030 | 0.33 | 0.07 | 0.12 | 67 |
| 1031 | 0.88 | 0.47 | 0.61 | 75 |
| 1032 | 0.56 | 0.28 | 0.38 | 64 |
| 1033 | 0.88 | 0.21 | 0.34 | 70 |
| 1034 | 0.17 | 0.06 | 0.09 | 69 |
| 1035 | 0.44 | 0.10 | 0.16 | 72 |
| 1036 | 0.30 | 0.04 | 0.07 | 79 |
| 1037 | 0.24 | 0.05 | 0.08 | 84 |
| 1038 | 0.00 | 0.00 | 0.00 | 87 |
| 1039 | 0.68 | 0.35 | 0.46 | 65 |
| 1040 | 0.72 | 0.36 | 0.48 | 73 |
| 1041 | 0.00 | 0.00 | 0.00 | 77 |
| 1042 | 0.27 | 0.05 | 0.09 | 77 |
| 1043 | 0.16 | 0.07 | 0.09 | 60 |
| 1044 | 0.00 | 0.00 | 0.00 | 73 |
| 1045 | 0.00 | 0.00 | 0.00 | 67 |
| 1046 | 0.43 | 0.04 | 0.07 | 83 |
| 1047 | 1.00 | 0.40 | 0.57 | 70 |
| 1048 | 1.00 | 0.02 | 0.03 | 65 |
| 1049 | 0.62 | 0.14 | 0.22 | 74 |
| 1050 | 0.50 | 0.02 | 0.03 | 62 |
| 1051 | 0.58 | 0.16 | 0.25 | 70 |
| 1052 | 0.00 | 0.00 | 0.00 | 69 |
| 1053 | 0.25 | 0.08 | 0.12 | 72 |
| 1054 | 0.44 | 0.15 | 0.23 | 72 |
| 1055 | 0.90 | 0.52 | 0.66 | 73 |
| 1056 | 0.74 | 0.34 | 0.46 | 92 |
| 1057 | 0.67 | 0.05 | 0.10 | 73 |
| 1058 | 0.31 | 0.12 | 0.17 | 68 |
| 1059 | 0.00 | 0.00 | 0.00 | 71 |
| 1060 | 0.33 | 0.10 | 0.16 | 69 |
| 1061 | 0.85 | 0.24 | 0.37 | 72 |
| 1062 | 0.44 | 0.29 | 0.35 | 66 |
| 1063 | 0.14 | 0.01 | 0.02 | 84 |
| 1064 | 0.00 | 0.00 | 0.00 | 78 |
| 1065 | 0.81 | 0.45 | 0.58 | 66 |
| 1066 | 0.21 | 0.04 | 0.07 | 69 |
| 1067 | 0.11 | 0.01 | 0.02 | 80 |
| 1068 | 1.00 | 0.01 | 0.03 | 71 |
| 1069 | 0.52 | 0.18 | 0.27 | 60 |
| 1070 | 0.20 | 0.01 | 0.02 | 77 |
| 1071 | 0.88 | 0.29 | 0.43 | 80 |
| 1072 | 0.25 | 0.06 | 0.10 | 80 |
| 1073 | 0.00 | 0.00 | 0.00 | 74 |
| 1074 | 0.21 | 0.04 | 0.07 | 69 |
| 1075 | 0.44 | 0.07 | 0.12 | 56 |
| 1076 | 0.32 | 0.13 | 0.18 | 63 |
| 1077 | 0.58 | 0.19 | 0.29 | 58 |
| 1078 | 0.00 | 0.00 | 0.00 | 63 |
| 1079 | 0.83 | 0.24 | 0.37 | 85 |
| 1080 | 0.52 | 0.15 | 0.24 | 78 |
| 1081 | 0.00 | 0.00 | 0.00 | 84 |
| 1082 | 0.74 | 0.42 | 0.54 | 73 |
| 1083 | 0.09 | 0.02 | 0.03 | 55 |
| 1084 | 0.51 | 0.26 | 0.34 | 70 |
| 1085 | 0.69 | 0.26 | 0.38 | 85 |
| 1086 | 0.00 | 0.00 | 0.00 | 68 |
| 1087 | 0.40 | 0.02 | 0.05 | 82 |

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|------|------|------|------|----|
| 1088 | 0.00 | 0.00 | 0.00 | 67 |
| 1089 | 0.81 | 0.44 | 0.57 | 78 |
| 1090 | 0.70 | 0.11 | 0.19 | 64 |
| 1091 | 0.35 | 0.09 | 0.15 | 75 |
| 1092 | 0.38 | 0.16 | 0.23 | 61 |
| 1093 | 0.65 | 0.17 | 0.28 | 63 |
| 1094 | 0.00 | 0.00 | 0.00 | 77 |
| 1095 | 0.36 | 0.13 | 0.19 | 70 |
| 1096 | 0.86 | 0.34 | 0.48 | 71 |
| 1097 | 0.44 | 0.12 | 0.18 | 69 |
| 1098 | 0.58 | 0.22 | 0.32 | 63 |
| 1099 | 0.80 | 0.49 | 0.61 | 67 |
| 1100 | 0.57 | 0.06 | 0.11 | 68 |
| 1101 | 0.00 | 0.00 | 0.00 | 57 |
| 1102 | 0.90 | 0.54 | 0.67 | 69 |
| 1103 | 0.14 | 0.01 | 0.03 | 70 |
| 1104 | 0.40 | 0.05 | 0.09 | 75 |
| 1105 | 0.21 | 0.05 | 0.08 | 62 |
| 1106 | 0.25 | 0.01 | 0.03 | 72 |
| 1107 | 0.00 | 0.00 | 0.00 | 76 |
| 1108 | 0.00 | 0.00 | 0.00 | 72 |
| 1109 | 0.00 | 0.00 | 0.00 | 86 |
| 1110 | 0.85 | 0.43 | 0.57 | 82 |
| 1111 | 0.00 | 0.00 | 0.00 | 70 |
| 1112 | 0.50 | 0.01 | 0.03 | 72 |
| 1113 | 0.65 | 0.24 | 0.35 | 70 |
| 1114 | 0.20 | 0.02 | 0.03 | 57 |
| 1115 | 0.25 | 0.04 | 0.07 | 68 |
| 1116 | 0.00 | 0.00 | 0.00 | 64 |
| 1117 | 0.29 | 0.03 | 0.05 | 66 |
| 1118 | 0.50 | 0.11 | 0.18 | 81 |
| 1119 | 0.68 | 0.24 | 0.35 | 63 |
| 1120 | 0.15 | 0.06 | 0.09 | 62 |
| 1121 | 0.00 | 0.00 | 0.00 | 79 |
| 1122 | 0.80 | 0.21 | 0.34 | 56 |
| 1123 | 0.24 | 0.06 | 0.09 | 71 |
| 1124 | 0.00 | 0.00 | 0.00 | 78 |
| 1125 | 0.80 | 0.06 | 0.11 | 66 |
| 1126 | 0.00 | 0.00 | 0.00 | 62 |
| 1127 | 0.75 | 0.18 | 0.29 | 66 |
| 1128 | 0.00 | 0.00 | 0.00 | 70 |
| 1129 | 0.94 | 0.46 | 0.62 | 65 |
| 1130 | 0.85 | 0.37 | 0.51 | 63 |
| 1131 | 0.89 | 0.52 | 0.66 | 79 |
| 1132 | 0.38 | 0.07 | 0.12 | 67 |
| 1133 | 0.00 | 0.00 | 0.00 | 64 |
| 1134 | 0.20 | 0.03 | 0.05 | 67 |
| 1135 | 0.73 | 0.21 | 0.32 | 78 |
| 1136 | 0.44 | 0.07 | 0.13 | 54 |
| 1137 | 0.00 | 0.00 | 0.00 | 64 |
| 1138 | 0.39 | 0.09 | 0.15 | 76 |
| 1139 | 0.00 | 0.00 | 0.00 | 64 |
| 1140 | 0.00 | 0.00 | 0.00 | 67 |
| 1141 | 0.06 | 0.01 | 0.02 | 70 |
| 1142 | 0.44 | 0.06 | 0.11 | 66 |
| 1143 | 0.74 | 0.40 | 0.52 | 62 |
| 1144 | 0.00 | 0.00 | 0.00 | 67 |
| 1145 | 0.12 | 0.06 | 0.11 | 17 |

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|------|------|------|------|----|
| 1145 | 0.43 | 0.00 | 0.11 | 47 |
| 1146 | 0.35 | 0.09 | 0.14 | 69 |
| 1147 | 0.71 | 0.40 | 0.51 | 63 |
| 1148 | 0.37 | 0.10 | 0.16 | 70 |
| 1149 | 0.41 | 0.13 | 0.19 | 55 |
| 1150 | 0.57 | 0.33 | 0.42 | 49 |
| 1151 | 0.57 | 0.07 | 0.12 | 58 |
| 1152 | 0.00 | 0.00 | 0.00 | 65 |
| 1153 | 0.00 | 0.00 | 0.00 | 67 |
| 1154 | 0.00 | 0.00 | 0.00 | 66 |
| 1155 | 0.94 | 0.52 | 0.67 | 62 |
| 1156 | 0.62 | 0.07 | 0.12 | 72 |
| 1157 | 0.90 | 0.42 | 0.57 | 62 |
| 1158 | 0.00 | 0.00 | 0.00 | 60 |
| 1159 | 0.43 | 0.16 | 0.23 | 64 |
| 1160 | 0.30 | 0.05 | 0.09 | 59 |
| 1161 | 0.10 | 0.02 | 0.03 | 55 |
| 1162 | 0.51 | 0.29 | 0.37 | 63 |
| 1163 | 0.77 | 0.36 | 0.49 | 64 |
| 1164 | 0.00 | 0.00 | 0.00 | 54 |
| 1165 | 0.32 | 0.10 | 0.15 | 62 |
| 1166 | 0.00 | 0.00 | 0.00 | 73 |
| 1167 | 0.46 | 0.21 | 0.29 | 56 |
| 1168 | 0.33 | 0.03 | 0.06 | 60 |
| 1169 | 0.35 | 0.11 | 0.17 | 63 |
| 1170 | 0.80 | 0.05 | 0.10 | 73 |
| 1171 | 0.60 | 0.31 | 0.41 | 58 |
| 1172 | 0.29 | 0.03 | 0.06 | 59 |
| 1173 | 0.23 | 0.04 | 0.07 | 68 |
| 1174 | 0.45 | 0.14 | 0.22 | 63 |
| 1175 | 0.98 | 0.60 | 0.74 | 70 |
| 1176 | 0.87 | 0.42 | 0.57 | 62 |
| 1177 | 0.00 | 0.00 | 0.00 | 62 |
| 1178 | 0.00 | 0.00 | 0.00 | 45 |
| 1179 | 0.97 | 0.37 | 0.53 | 79 |
| 1180 | 0.70 | 0.12 | 0.21 | 58 |
| 1181 | 0.88 | 0.30 | 0.44 | 71 |
| 1182 | 0.12 | 0.02 | 0.03 | 56 |
| 1183 | 0.00 | 0.00 | 0.00 | 63 |
| 1184 | 0.00 | 0.00 | 0.00 | 72 |
| 1185 | 0.33 | 0.04 | 0.06 | 56 |
| 1186 | 0.82 | 0.19 | 0.30 | 75 |
| 1187 | 0.17 | 0.02 | 0.03 | 57 |
| 1188 | 0.45 | 0.08 | 0.14 | 60 |
| 1189 | 0.25 | 0.02 | 0.03 | 65 |
| 1190 | 0.50 | 0.01 | 0.03 | 68 |
| 1191 | 0.59 | 0.16 | 0.25 | 62 |
| 1192 | 0.00 | 0.00 | 0.00 | 68 |
| 1193 | 0.00 | 0.00 | 0.00 | 66 |
| 1194 | 0.40 | 0.04 | 0.06 | 57 |
| 1195 | 0.11 | 0.01 | 0.03 | 67 |
| 1196 | 0.88 | 0.10 | 0.18 | 69 |
| 1197 | 0.36 | 0.06 | 0.10 | 66 |
| 1198 | 0.40 | 0.03 | 0.06 | 62 |
| 1199 | 0.33 | 0.08 | 0.14 | 59 |
| 1200 | 0.92 | 0.21 | 0.34 | 57 |
| 1201 | 1.00 | 0.31 | 0.47 | 62 |
| 1202 | 0.87 | 0.47 | 0.61 | 58 |

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|------|------|------|------|----|
| 1203 | 0.00 | 0.00 | 0.00 | 67 |
| 1204 | 0.63 | 0.35 | 0.45 | 74 |
| 1205 | 0.50 | 0.02 | 0.04 | 55 |
| 1206 | 0.55 | 0.09 | 0.16 | 65 |
| 1207 | 0.47 | 0.11 | 0.17 | 75 |
| 1208 | 0.63 | 0.20 | 0.30 | 61 |
| 1209 | 0.69 | 0.39 | 0.49 | 62 |
| 1210 | 0.14 | 0.02 | 0.03 | 59 |
| 1211 | 0.50 | 0.19 | 0.28 | 47 |
| 1212 | 0.00 | 0.00 | 0.00 | 59 |
| 1213 | 0.95 | 0.36 | 0.52 | 59 |
| 1214 | 1.00 | 0.03 | 0.05 | 74 |
| 1215 | 0.25 | 0.02 | 0.03 | 65 |
| 1216 | 0.00 | 0.00 | 0.00 | 60 |
| 1217 | 0.53 | 0.19 | 0.27 | 54 |
| 1218 | 0.00 | 0.00 | 0.00 | 62 |
| 1219 | 0.93 | 0.68 | 0.79 | 78 |
| 1220 | 0.85 | 0.57 | 0.68 | 72 |
| 1221 | 0.75 | 0.35 | 0.48 | 60 |
| 1222 | 0.43 | 0.14 | 0.21 | 63 |
| 1223 | 0.00 | 0.00 | 0.00 | 66 |
| 1224 | 0.56 | 0.14 | 0.23 | 69 |
| 1225 | 0.00 | 0.00 | 0.00 | 69 |
| 1226 | 0.80 | 0.18 | 0.29 | 68 |
| 1227 | 0.53 | 0.17 | 0.26 | 58 |
| 1228 | 0.00 | 0.00 | 0.00 | 51 |
| 1229 | 0.00 | 0.00 | 0.00 | 59 |
| 1230 | 0.00 | 0.00 | 0.00 | 75 |
| 1231 | 0.50 | 0.11 | 0.18 | 64 |
| 1232 | 0.00 | 0.00 | 0.00 | 66 |
| 1233 | 0.29 | 0.03 | 0.06 | 58 |
| 1234 | 0.00 | 0.00 | 0.00 | 63 |
| 1235 | 0.06 | 0.02 | 0.03 | 62 |
| 1236 | 0.00 | 0.00 | 0.00 | 57 |
| 1237 | 1.00 | 0.01 | 0.03 | 77 |
| 1238 | 0.81 | 0.40 | 0.54 | 52 |
| 1239 | 0.86 | 0.30 | 0.45 | 63 |
| 1240 | 0.90 | 0.40 | 0.55 | 48 |
| 1241 | 0.00 | 0.00 | 0.00 | 71 |
| 1242 | 0.79 | 0.18 | 0.29 | 62 |
| 1243 | 0.43 | 0.10 | 0.16 | 61 |
| 1244 | 0.00 | 0.00 | 0.00 | 53 |
| 1245 | 0.09 | 0.01 | 0.02 | 75 |
| 1246 | 0.38 | 0.05 | 0.10 | 55 |
| 1247 | 0.50 | 0.02 | 0.04 | 55 |
| 1248 | 0.00 | 0.00 | 0.00 | 49 |
| 1249 | 0.33 | 0.05 | 0.09 | 74 |
| 1250 | 0.97 | 0.47 | 0.64 | 59 |
| 1251 | 0.38 | 0.14 | 0.21 | 56 |
| 1252 | 0.33 | 0.10 | 0.15 | 63 |
| 1253 | 0.59 | 0.21 | 0.31 | 48 |
| 1254 | 0.95 | 0.60 | 0.73 | 62 |
| 1255 | 0.00 | 0.00 | 0.00 | 69 |
| 1256 | 0.30 | 0.05 | 0.08 | 65 |
| 1257 | 0.00 | 0.00 | 0.00 | 62 |
| 1258 | 0.39 | 0.14 | 0.20 | 51 |
| 1259 | 0.62 | 0.12 | 0.21 | 64 |
| 1260 | 0.00 | 0.00 | 0.00 | 64 |

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|------|------|------|------|----|
| 1261 | 0.00 | 0.00 | 0.00 | 63 |
| 1262 | 0.93 | 0.22 | 0.36 | 58 |
| 1263 | 0.36 | 0.07 | 0.12 | 54 |
| 1264 | 0.00 | 0.00 | 0.00 | 62 |
| 1265 | 0.00 | 0.00 | 0.00 | 59 |
| 1266 | 0.90 | 0.46 | 0.60 | 57 |
| 1267 | 0.14 | 0.02 | 0.03 | 51 |
| 1268 | 0.25 | 0.04 | 0.07 | 46 |
| 1269 | 0.97 | 0.53 | 0.68 | 55 |
| 1270 | 0.88 | 0.10 | 0.18 | 69 |
| 1271 | 0.60 | 0.14 | 0.22 | 65 |
| 1272 | 0.38 | 0.08 | 0.14 | 60 |
| 1273 | 0.35 | 0.10 | 0.16 | 59 |
| 1274 | 0.25 | 0.05 | 0.08 | 62 |
| 1275 | 0.00 | 0.00 | 0.00 | 52 |
| 1276 | 0.40 | 0.07 | 0.12 | 57 |
| 1277 | 0.29 | 0.03 | 0.06 | 61 |
| 1278 | 0.70 | 0.11 | 0.19 | 62 |
| 1279 | 0.93 | 0.57 | 0.71 | 47 |
| 1280 | 0.25 | 0.03 | 0.06 | 63 |
| 1281 | 0.58 | 0.11 | 0.19 | 61 |
| 1282 | 0.60 | 0.18 | 0.28 | 50 |
| 1283 | 0.27 | 0.08 | 0.12 | 52 |
| 1284 | 0.68 | 0.23 | 0.35 | 56 |
| 1285 | 0.67 | 0.04 | 0.07 | 57 |
| 1286 | 0.71 | 0.10 | 0.18 | 49 |
| 1287 | 0.57 | 0.14 | 0.23 | 56 |
| 1288 | 0.57 | 0.27 | 0.36 | 49 |
| 1289 | 0.00 | 0.00 | 0.00 | 55 |
| 1290 | 0.00 | 0.00 | 0.00 | 68 |
| 1291 | 0.90 | 0.50 | 0.64 | 52 |
| 1292 | 0.29 | 0.03 | 0.05 | 73 |
| 1293 | 0.88 | 0.43 | 0.58 | 67 |
| 1294 | 0.00 | 0.00 | 0.00 | 54 |
| 1295 | 0.25 | 0.06 | 0.10 | 34 |
| 1296 | 1.00 | 0.34 | 0.51 | 56 |
| 1297 | 0.00 | 0.00 | 0.00 | 66 |
| 1298 | 1.00 | 0.03 | 0.06 | 68 |
| 1299 | 0.57 | 0.06 | 0.11 | 64 |
| 1300 | 0.91 | 0.50 | 0.65 | 64 |
| 1301 | 0.00 | 0.00 | 0.00 | 48 |
| 1302 | 0.00 | 0.00 | 0.00 | 63 |
| 1303 | 0.00 | 0.00 | 0.00 | 62 |
| 1304 | 0.50 | 0.02 | 0.04 | 54 |
| 1305 | 0.23 | 0.10 | 0.14 | 51 |
| 1306 | 0.22 | 0.07 | 0.11 | 55 |
| 1307 | 0.00 | 0.00 | 0.00 | 53 |
| 1308 | 0.61 | 0.31 | 0.41 | 54 |
| 1309 | 0.67 | 0.16 | 0.26 | 61 |
| 1310 | 0.00 | 0.00 | 0.00 | 42 |
| 1311 | 0.25 | 0.02 | 0.03 | 55 |
| 1312 | 0.00 | 0.00 | 0.00 | 64 |
| 1313 | 0.00 | 0.00 | 0.00 | 58 |
| 1314 | 0.90 | 0.36 | 0.51 | 50 |
| 1315 | 0.00 | 0.00 | 0.00 | 57 |
| 1316 | 0.59 | 0.22 | 0.32 | 46 |
| 1317 | 1.00 | 0.05 | 0.09 | 42 |
| 1318 | 0.50 | 0.22 | 0.30 | 74 |