

**CSE3019 - DATA MINING**

**PROJECT REPORT**

**SUBMITTED TO – Prof. MEENAKSHI SP**

**TOPIC – ROUGHSET NEIGHBOURHOOD CLASSIFICATION**

**GROUP MEMBERS**

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**ABSTRACT**

Medical datasets consume enormous amount of information about the patients, diseases and the physicians. Diseases diagnosis required many expensive tests to predict the diseases. Cost of disease prediction and diagnosis can be reduced by applying machine learning and data mining methods. Disease prediction and decision making plays asignificant role in medical diagnosis. In this study, a novel neighborhood rough set classification approach is presented to deal with medical datasets. Five benchmarked medical datasets have been used in this research work for studying the impact of proposed work in decision making.Experimental resultof the proposed classification algorithm is compared with other existing approaches such as rough set, Kth–nearest neighbor, support vector machine, Back propagation algorithm and multilayer perceptron to conclude that the proposed approach is cheaper way for disease prediction and decision making. The performance of classification algorithms measured based on various classification accuracy measures.

**INTRODUCTION**

Rough set theory was originally proposed by Pawlak [27,28] as a tool for dealing with uncertain knowledge in information systems. So far, rough set theory has been successfully applied in fields such as knowledge discovery [6,32,33], machine learning [21], knowledge acquisition [18,20,43], decision analysis [12,35,23,44,45] and many other areas [2,10,11,19,36,37]. As is well-known, equivalence relation plays an important role in rough set theory. However, such an equivalence relation is still restrictive for many applications [13,47,50,51]. To address this issue, many meaningful extensions of the equivalence relation have been proposed, such as tolerance relations [24,25,34,39], similarity relations [1,38,40], pre-order relations [14,17] and arbitrary binary relations [16,48,49]. Most extensions of the equivalence relation are imposed to be reflexive.

However, the reflexive property could be relaxed in some cases as shown in [5]. In this paper, we consider a new binary relation, which is irreflexive and induced from a graph. A graph is called to be bipartite if the vertex set can be partitioned into two independent sets. Testing bipartiteness of a graph is an important problem in graph theory. Traditionally, the Breadth First Search (BFS) algorithm is often used to check whether or not a graph is bipartite [4]. However, the running time of this algorithm is determined by the representations of graphs. For example, if the adjacency list representation is used, the average running time of this algorithm is O(jUj + jEj) [15], and in the worst case the running time is O(jUj2). Here U is a set of vertices and E is a set of edges. Also, if the adjacency matrix representation is used, the running time is O(jUj2).

An important concept for binary relations is represented by the visual form of digraphs [14]. Conversely, for a simple undirected graph, a corresponding binary relation may be used to represent it. Thus the method of generalized rough set theory with neighborhoods defined over such relations is used to study some problems of graph theory. In this paper, we focus mainly on the study of the problem of testing bipartiteness of simple undirected graphs based on generalized rough sets. To deal with this problem, the modified concept of definability called quasi-outer definable sets is introduced, and an algorithm is designed for this purpose. Experimental results show that the proposed algorithm is more efficient than the traditional ones when the adjacency matrix representation is used. The main contribution of the work is that we provide an approach for testing bipartiteness of graph theory from the viewpoint of generalized rough sets. The remainder of this paper is organized as follows: In Section 2, some basic notions related to generalized rough sets and graph theory are introduced. In Section 3, a new binary relation is proposed. In Section 4, the concept of quasi-outer definable sets is presented, and its basic properties are studied. In Section 5, a new method for testing bipartiteness of a simple undirected graph is given.

**NEIGHBOURHOOD ROUGH SET**

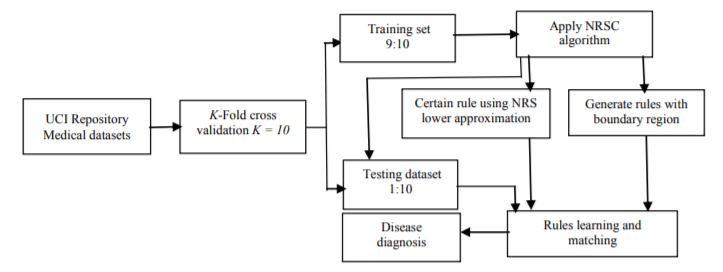
The Neighborhood Rough Set (NRS) is used to replace the equivalent approximation of traditional rough set model with neighborhood relation, which supports both continuous and discrete datasets. In this section, we present various essential concepts in NRS used in this work [9, 10, 11, 12, 13].

The complete NRS model works based on neighbourhood relation which uses distance metric functions (Euclidean distance) based neighborhood relation to replace the equivalence relation in traditional rough set. The Euclidean distance metric function is defined as:

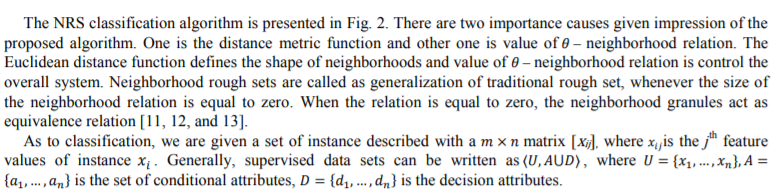


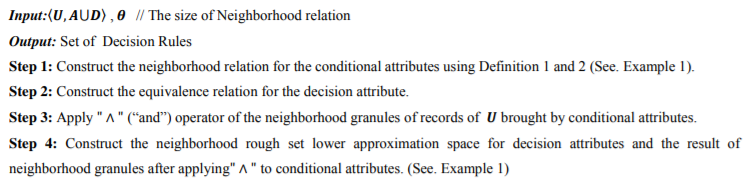
**PROPOSED METHODOLGY**

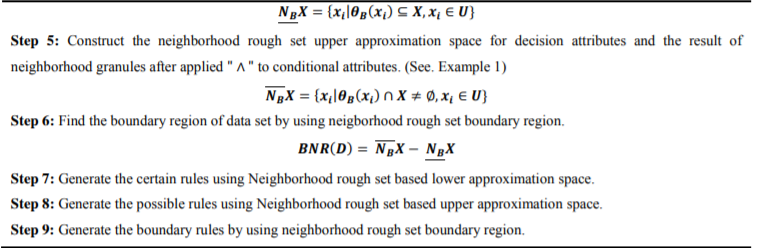
The complete methodology adopted in this work for the diagnosis of medical data sets is exposed in Fig. 1. First stage of the proposed system is the process of medical data sets gathered using various computerized devices.Real time medical data acquisition is quite difficult and extraordinary domain in data mining and machine learning. In this work, we acquired medical data sets form UCI machine learning repository. The basic information of the medical data sets is explained in section 5. In second stage, data sets are separated into training and testing based on K-fold cross validation. After the separation of data sets, neighborhood rough set is applied to training data set based medical data sets classification, NRSC algorithm is explained in section 4. In the next step, generated decision rules are matched with test data set for validating the proposed classifier algorithm. The various classification measures are applied to evaluate the performance of the proposed classification algorithm and the proposed approach is also compared with various classification algorithms for medical diagnosis.



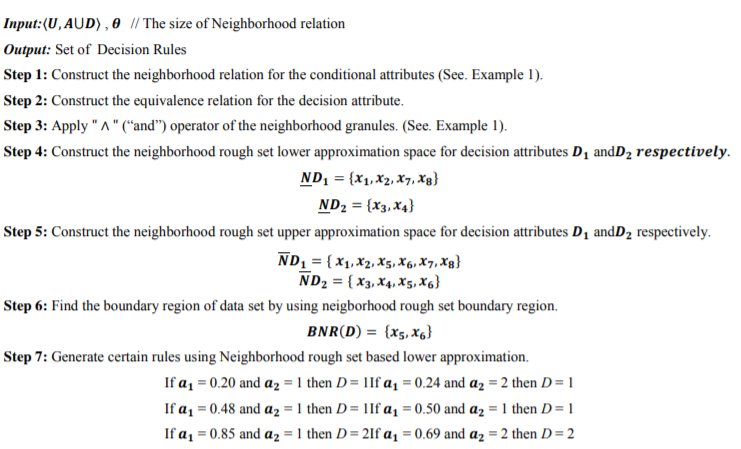
**Proposed Neighbourhood rough set based classification algorithm**







The initial steps of the proposed classification algorithm is compute the neighbourhood relation of the conditional attributes by using Definition 1, 2 and compute equivalence classes of decision attributes. Distance metrics values are calculated based on Euclidean distance and the size of the neighbourhood is “and” operator to neighbourhood granules of conditional attributes. Furthermore, construct NRS lower and upper approximations to the result of after applied “and” operator and decision classes. The NRS lower approximation space of the decision attributes is definite as the union of the lower approximation space of each decision class. Find the boundary region, boundary region of the subset is come from more than one decision class. With the help of neighbourhood approximation we generated two types of decision rules: certain rules (deterministic rules) and possible rules (non-deterministic rules). The certain rules are generated by applying lower approximation and possible rules are generated by applying upper approximation of neighbourhood rough set. The proposed algorithm is explained with example.



**EXPERIMAENTAL RESULTS AND DISCUSSION**

The simulation of proposed algorithm and benchmark algorithms used for comparison were performed using an Intel (R) Core (TM) i3 CPU 2330M–2.20 GHz machine with 4 GB RAM and a Microsoft Windows 7 64-bit operating system. The essential of the NRSC algorithm calculations was implemented using the MATLAB software package (MATLAB R2013b).

**MEDICAL DATASET DESCRIPTION**

The applicability of NRSC classification is validated in publicly available real-world medical data sets. In this paper, we used five different medical data sets acquired from the well-known UCI repository. The medical data sets are Pima diabetes database [32, 34], heart disease (echocardiogram data) [33], Breast cancer data set [28, 32], liver disorder [28] and Hepatitis’s [28, 32]. Pima diabetes database contains 768 instance, 7 attributes and 2 decision classes (negative and positive). In this diabetes dataset, all patients were females with its disturbed with appearance or nonappearance of diabetes. Echocardiogram dataset contains 132 instance, 13 attributes and 2 decision classes (Alive or Not alive). All samples of echocardiogram affected by heart attack, some patients are died and some are still alive. The major problem statement of this dataset is to exactly classify the NOT Alivepatient’s samples. The Wisconsin breast cancer data set contains 699 samples, 11 attributes and 2 decision classes (Benign and Malignant). Liver disorder data set covers 345 instances, 7 attributes and two decision classes (Sick or Normal). The major cause of the liver disorder is alcohol intake, so initially patient’s bloods are need for further diagnosis purpose. In this liver disorder data set initially 5 types of blood test are taken and recorded in first 5 attributes. The hepatitis’s contains 155 instance, 19 attributes and two classes (die and live). The k-fold cross validation (CV) method is applied for evaluate the classification accuracy of test results. The k-fold CV method is extensively used by many researchers with the purpose of random sampling of the training. Initially, all the data’s form database randomly separated to k equally select and almost same size subsets. Furthermore, the classification algorithm is trained and tested k times. In this paper, we defined k as 10, so data is divided into ten subsets.

**PERFORMANCE ANALYSIS**

The performance of the proposed Neighborhood rough set based classification algorithm is compared with traditional Pawlak’s rough set (RS), K-nearest neighbor algorithm (KNN), Back propagation algorithm (BPN), Multilayer perceptron (MLP) and support vector machine (SVM). The obtained results of above classification algorithms are validated based on classification validation accuracy measures. Validation is important for the classification of medical data sets because accurate classification and decision making system is very important in data mining and medical diagnosis. There are many validation methods available for evaluating the accuracy of classification algorithm. In this paper, we demonstrated the performance of classification algorithm using the most familiar metrics such as Precision, Recall, F-Measure, and some other validation measures such as Folke-mallows Index, Kulcznski Index and rand Index [35]. These various measures are presented in Table 2. Sensitivity or recall is a measure of the ability of a prediction model to select instances of a certain class from a dataset. Positive predictive value is the amount of positive test results that are true positives (correct diagnoses). It is a critical measure of the performance of an analytical method, as it reproduces the probability that a positive test reflects the underlying conditions. Table 3shows the performance of proposed neighborhood rough set classification algorithm and comparative algorithms for the five medical data sets.

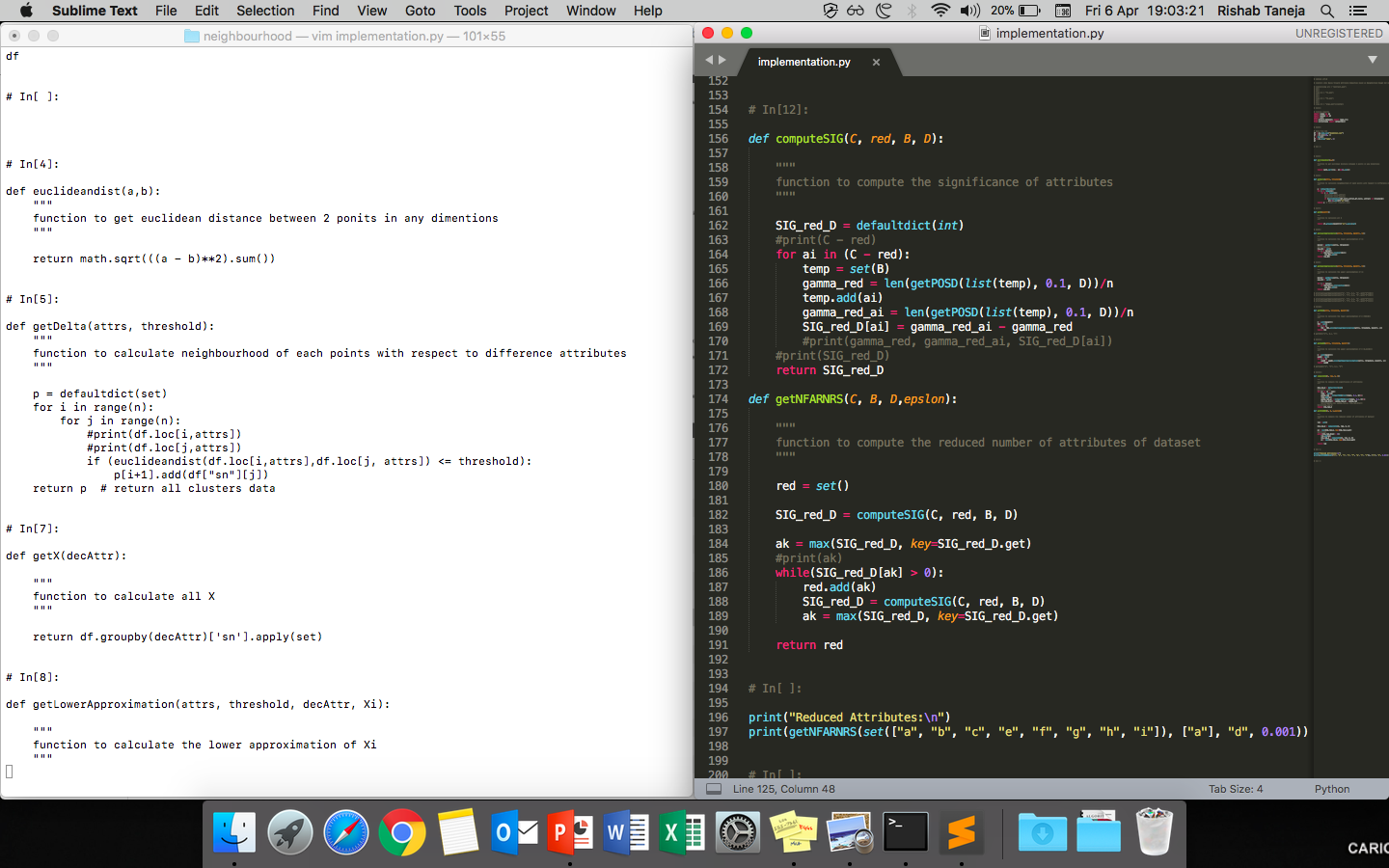
**DISCUSSION**

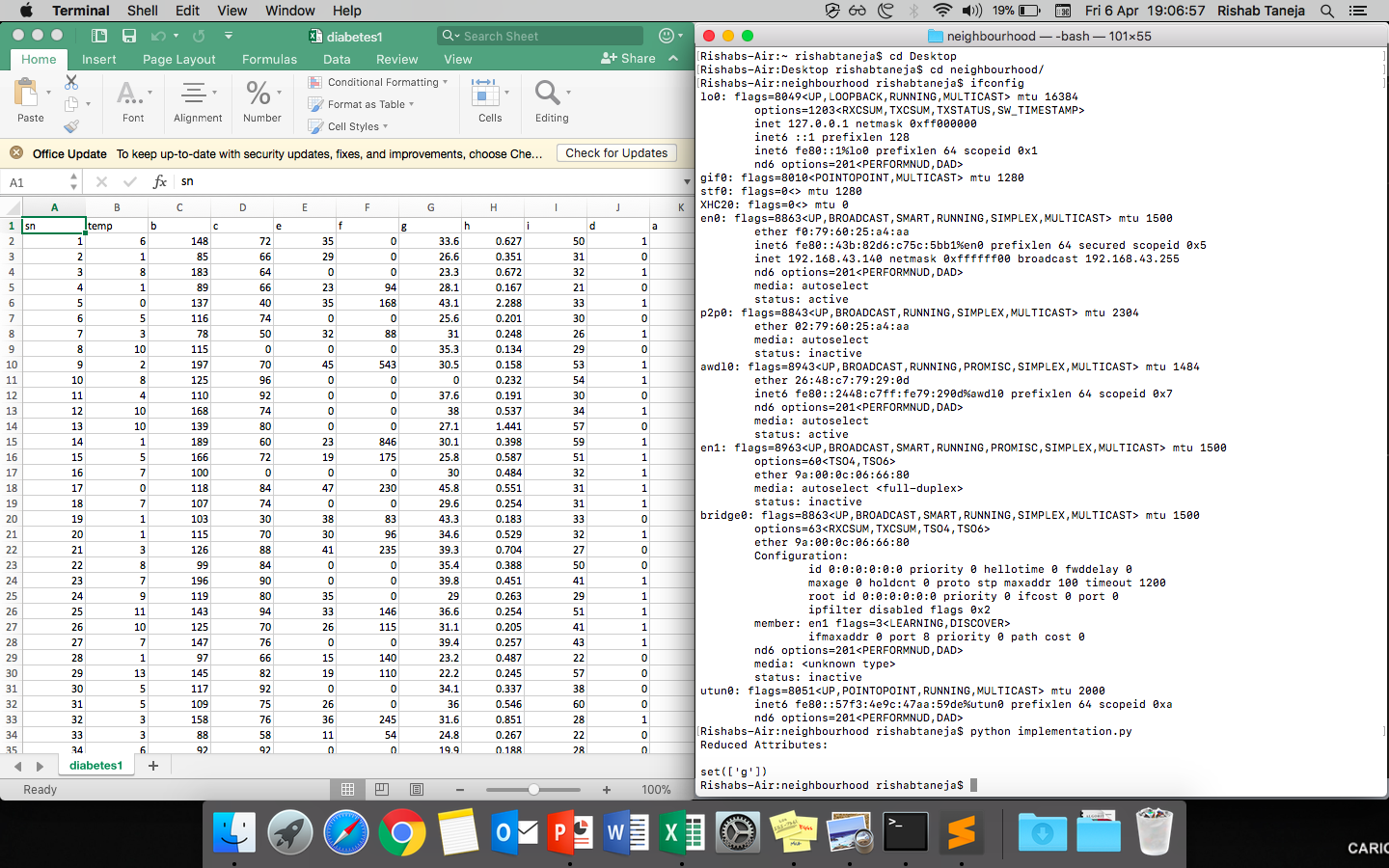
Medical data classification is a major element of the many decision-making tasks. Decision-making tasks are instances of classification problem that can be easily formulated into a prediction tasks, diagnosis and pattern recognition.To avoid the risks of decision-making, we need computerized decision making techniques. In this work, we proposed the neighborhood rough set based classification for medical diagnosis. In previous study, neighborhood rough set theory was widely applied for feature selection and none of the approaches have been adoptedcompletely for medical diagnosis. The proposed work is applied for five medical data sets and evaluated the efficiency of NRSC compared with five different classification algorithms. The efficiency of the classification algorithm is validated from six performance measures. The performance resultsconfidently demonstrated that the neighborhood rough set based classification method is very effective for medical data classification. Furthermore, the NRSC delivered good results over Pawlak’s rough set (ߠBPN, MLP, SVM, KNN. The result is intensely important in decision support tasks not only for exact prognostic or diagnostic prediction, but also would like to be convinced that the prediction is based on reasonable justifications; hence accepting the use of proposed classification system in clinical practice.

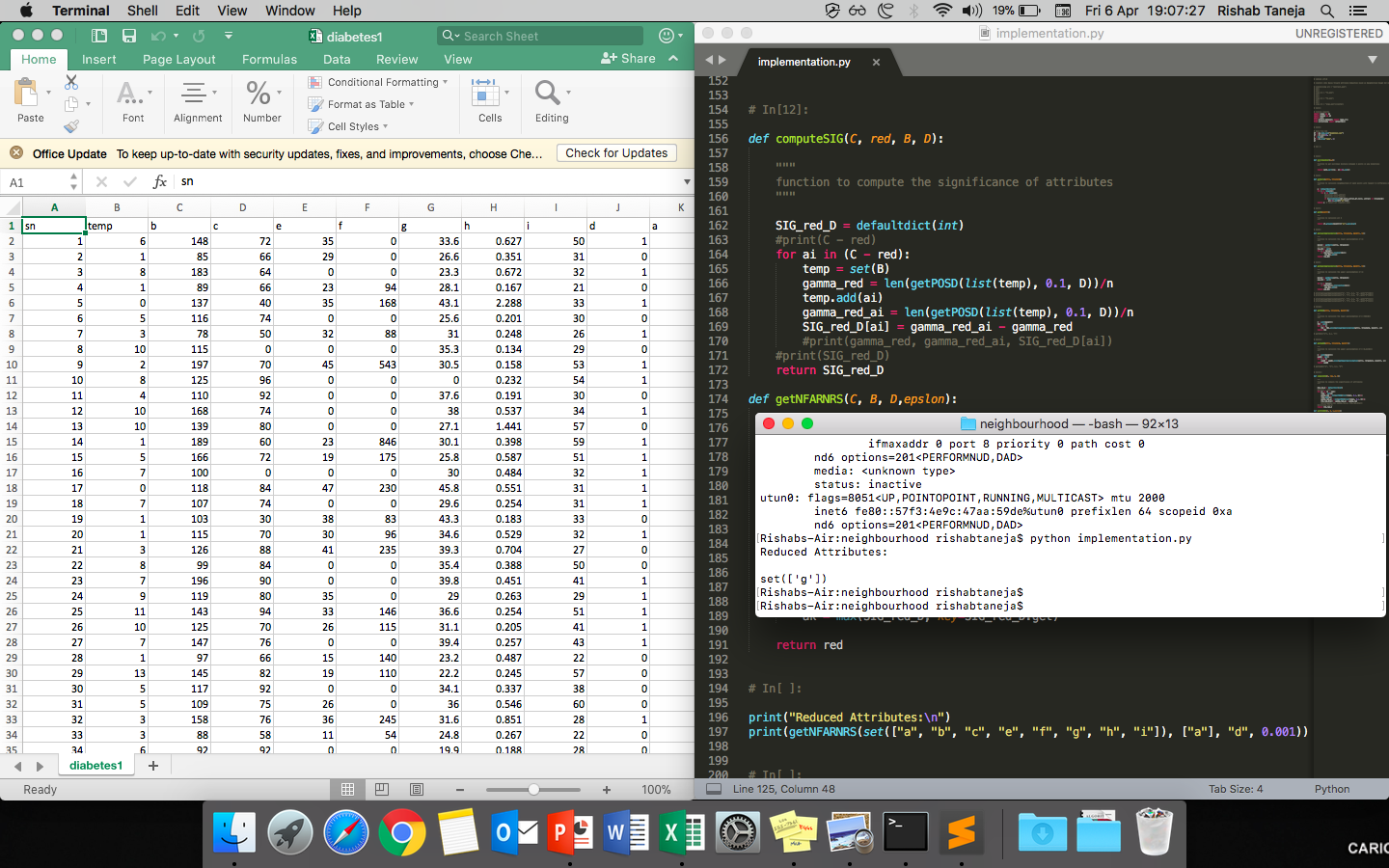
**CODE**

*# coding: utf-8  
  
# <center> <h1> Naive Forward Attribute Reduction based on Neighborhood Rough Set model Implementation (NFARNRS) </h1> </center>  
  
# <center><img src = "abstract.png">  
# <br>  
# <br>  
# <img src = "f1.png">  
# <br>  
# <br>  
# <img src = "f2.png">  
# <br>  
# <br>  
# <img src = "algo.png"></center>  
  
# In[1]:  
  
# library imports***import** numpy **as** np  
**import** pandas **as** pd  
**import** math  
**from** sklearn.datasets **import** load\_iris  
**from** collections **import** defaultdict  
  
  
*# In[3]:  
  
# data from CSV*df = pd.read\_csv(**"diabetes1.csv"**)  
df = df.loc[:20, :]  
n = len(df)  
df = df.drop(**"temp"**, 1)  
df  
  
  
*# In[ ]:  
  
# In[4]:***def** euclideandist(a,b):  
 *"""  
 function to get euclidean distance between 2 ponits in any dimentions  
 """* **return** math.sqrt(((a - b)\*\*2).sum())  
  
  
*# In[5]:***def** getDelta(attrs, threshold):  
 *"""  
 function to calculate neighbourhood of each points with respect to difference attributes  
 """* p = defaultdict(set)  
 **for** i **in** range(n):  
 **for** j **in** range(n):  
 *#print(df.loc[i,attrs])  
 #print(df.loc[j,attrs])* **if** (euclideandist(df.loc[i,attrs],df.loc[j, attrs]) <= threshold):  
 p[i+1].add(df[**"sn"**][j])  
 **return** p *# return all clusters data  
  
  
# In[7]:***def** getX(decAttr):  
   
 *"""  
 function to calculate all X  
 """* **return** df.groupby(decAttr)[**'sn'**].apply(set)  
  
  
*# In[8]:***def** getLowerApproximation(attrs, threshold, decAttr, Xi):  
   
 *"""  
 function to calculate the lower approximation of Xi  
 """* delset = getDelta(attrs, threshold)  
 *#print(delset)* low\_NXi = set()  
 **for** i **in** delset:  
 **if** (delset[i].issubset(Xi)):  
 low\_NXi.add(i)  
 **return** low\_NXi  
  
  
*# In[9]:***def** getUpperApproximation(attrs, threshold, decAttr, Xi):  
   
 *"""  
 function to calculate the upper approximation of Xi  
 """* delset = getDelta(attrs, threshold)  
 upp\_NXi = set()  
   
 **for** i **in** delset:  
 **if** (delset[i].intersection(Xi)):  
 upp\_NXi.add(i)  
 **return** upp\_NXi  
  
*# print(getLowerApproximation(["a", "b"], 0.1, "d", getX("d")[0]))  
# print(getUpperApproximation(["a", "b"], 0.1, "d", getX("d")[0]))  
  
# print(getLowerApproximation(["a", "b"], 0.1, "d", getX("d")[1]))  
# print(getUpperApproximation(["a", "b"], 0.1, "d", getX("d")[1]))  
  
  
# In[10]:***def** getPOSD(attrs, threshold, decAttr):  
   
 *"""  
 function to calculate the lower approximation of D (POS(D))  
 """* X = getX(decAttr)  
 POS = set()  
 **for** i **in** X:  
 POS = POS.union(getLowerApproximation(attrs, threshold, decAttr, i))  
 **return** POS  
  
*# getPOSD(["a"], 0.1, "d")  
  
  
# In[11]:***def** getUppND(attrs, threshold, decAttr):  
   
 *"""  
 function to calculate the upper approximation of D (N\_Xi(D)))  
 """* X = getX(decAttr)  
 uppND = set()  
 **for** i **in** X:  
 uppND = uppND.union(getUpperApproximation(attrs, threshold, decAttr, i))  
 **return** uppND  
  
*# getUppND(["a", "b"], 0.1, "d")  
  
  
# In[12]:***def** computeSIG(C, red, B, D):  
   
 *"""  
 function to compute the significance of attributes  
 """* SIG\_red\_D = defaultdict(int)  
 *#print(C - red)* **for** ai **in** (C - red):  
 temp = set(B)  
 gamma\_red = len(getPOSD(list(temp), 0.1, D))/n  
 temp.add(ai)  
 gamma\_red\_ai = len(getPOSD(list(temp), 0.1, D))/n  
 SIG\_red\_D[ai] = gamma\_red\_ai - gamma\_red  
 *#print(gamma\_red, gamma\_red\_ai, SIG\_red\_D[ai])   
 #print(SIG\_red\_D)* **return** SIG\_red\_D  
  
**def** getNFARNRS(C, B, D,epslon):  
   
 *"""  
 function to compute the reduced number of attributes of dataset  
 """* red = set()  
  
 SIG\_red\_D = computeSIG(C, red, B, D)  
   
 ak = max(SIG\_red\_D, key=SIG\_red\_D.get)  
 *#print(ak)* **while**(SIG\_red\_D[ak] > 0):  
 red.add(ak)  
 SIG\_red\_D = computeSIG(C, red, B, D)  
 ak = max(SIG\_red\_D, key=SIG\_red\_D.get)  
   
 **return** red   
  
  
*# In[ ]:*print(**"Reduced Attributes:\n"**)  
print(getNFARNRS(set([**"a"**, **"b"**, **"c"**, **"e"**, **"f"**, **"g"**, **"h"**, **"i"**]), [**"a"**], **"d"**, 0.001))

**SCREENSHOTS**







**DATASET**

|  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **sn** | **a** | **b** | **c** | **e** | **f** | **g** | **h** | **i** | **d** |
| 1 | 6 | 148 | 72 | 35 | 0 | 33.6 | 0.627 | 50 | 1 |
| 2 | 1 | 85 | 66 | 29 | 0 | 26.6 | 0.351 | 31 | 0 |
| 3 | 8 | 183 | 64 | 0 | 0 | 23.3 | 0.672 | 32 | 1 |
| 4 | 1 | 89 | 66 | 23 | 94 | 28.1 | 0.167 | 21 | 0 |
| 5 | 0 | 137 | 40 | 35 | 168 | 43.1 | 2.288 | 33 | 1 |
| 6 | 5 | 116 | 74 | 0 | 0 | 25.6 | 0.201 | 30 | 0 |
| 7 | 3 | 78 | 50 | 32 | 88 | 31 | 0.248 | 26 | 1 |
| 8 | 10 | 115 | 0 | 0 | 0 | 35.3 | 0.134 | 29 | 0 |
| 9 | 2 | 197 | 70 | 45 | 543 | 30.5 | 0.158 | 53 | 1 |
| 10 | 8 | 125 | 96 | 0 | 0 | 0 | 0.232 | 54 | 1 |
| 11 | 4 | 110 | 92 | 0 | 0 | 37.6 | 0.191 | 30 | 0 |
| 12 | 10 | 168 | 74 | 0 | 0 | 38 | 0.537 | 34 | 1 |
| 13 | 10 | 139 | 80 | 0 | 0 | 27.1 | 1.441 | 57 | 0 |
| 14 | 1 | 189 | 60 | 23 | 846 | 30.1 | 0.398 | 59 | 1 |
| 15 | 5 | 166 | 72 | 19 | 175 | 25.8 | 0.587 | 51 | 1 |
| 16 | 7 | 100 | 0 | 0 | 0 | 30 | 0.484 | 32 | 1 |
| 17 | 0 | 118 | 84 | 47 | 230 | 45.8 | 0.551 | 31 | 1 |
| 18 | 7 | 107 | 74 | 0 | 0 | 29.6 | 0.254 | 31 | 1 |
| 19 | 1 | 103 | 30 | 38 | 83 | 43.3 | 0.183 | 33 | 0 |
| 20 | 1 | 115 | 70 | 30 | 96 | 34.6 | 0.529 | 32 | 1 |
| 21 | 3 | 126 | 88 | 41 | 235 | 39.3 | 0.704 | 27 | 0 |
| 22 | 8 | 99 | 84 | 0 | 0 | 35.4 | 0.388 | 50 | 0 |
| 23 | 7 | 196 | 90 | 0 | 0 | 39.8 | 0.451 | 41 | 1 |
| 24 | 9 | 119 | 80 | 35 | 0 | 29 | 0.263 | 29 | 1 |
| 25 | 11 | 143 | 94 | 33 | 146 | 36.6 | 0.254 | 51 | 1 |
| 26 | 10 | 125 | 70 | 26 | 115 | 31.1 | 0.205 | 41 | 1 |
| 27 | 7 | 147 | 76 | 0 | 0 | 39.4 | 0.257 | 43 | 1 |
| 28 | 1 | 97 | 66 | 15 | 140 | 23.2 | 0.487 | 22 | 0 |
| 29 | 13 | 145 | 82 | 19 | 110 | 22.2 | 0.245 | 57 | 0 |
| 30 | 5 | 117 | 92 | 0 | 0 | 34.1 | 0.337 | 38 | 0 |
| 31 | 5 | 109 | 75 | 26 | 0 | 36 | 0.546 | 60 | 0 |
| 32 | 3 | 158 | 76 | 36 | 245 | 31.6 | 0.851 | 28 | 1 |
| 33 | 3 | 88 | 58 | 11 | 54 | 24.8 | 0.267 | 22 | 0 |
| 34 | 6 | 92 | 92 | 0 | 0 | 19.9 | 0.188 | 28 | 0 |
| 35 | 10 | 122 | 78 | 31 | 0 | 27.6 | 0.512 | 45 | 0 |
| 36 | 4 | 103 | 60 | 33 | 192 | 24 | 0.966 | 33 | 0 |
| 37 | 11 | 138 | 76 | 0 | 0 | 33.2 | 0.42 | 35 | 0 |
| 38 | 9 | 102 | 76 | 37 | 0 | 32.9 | 0.665 | 46 | 1 |
| 39 | 2 | 90 | 68 | 42 | 0 | 38.2 | 0.503 | 27 | 1 |
| 40 | 4 | 111 | 72 | 47 | 207 | 37.1 | 1.39 | 56 | 1 |
| 41 | 3 | 180 | 64 | 25 | 70 | 34 | 0.271 | 26 | 0 |
| 42 | 7 | 133 | 84 | 0 | 0 | 40.2 | 0.696 | 37 | 0 |
| 43 | 7 | 106 | 92 | 18 | 0 | 22.7 | 0.235 | 48 | 0 |
| 44 | 9 | 171 | 110 | 24 | 240 | 45.4 | 0.721 | 54 | 1 |
| 45 | 7 | 159 | 64 | 0 | 0 | 27.4 | 0.294 | 40 | 0 |
| 46 | 0 | 180 | 66 | 39 | 0 | 42 | 1.893 | 25 | 1 |
| 47 | 1 | 146 | 56 | 0 | 0 | 29.7 | 0.564 | 29 | 0 |
| 48 | 2 | 71 | 70 | 27 | 0 | 28 | 0.586 | 22 | 0 |
| 49 | 7 | 103 | 66 | 32 | 0 | 39.1 | 0.344 | 31 | 1 |
| 50 | 7 | 105 | 0 | 0 | 0 | 0 | 0.305 | 24 | 0 |
| 51 | 1 | 103 | 80 | 11 | 82 | 19.4 | 0.491 | 22 | 0 |
| 52 | 1 | 101 | 50 | 15 | 36 | 24.2 | 0.526 | 26 | 0 |
| 53 | 5 | 88 | 66 | 21 | 23 | 24.4 | 0.342 | 30 | 0 |
| 54 | 8 | 176 | 90 | 34 | 300 | 33.7 | 0.467 | 58 | 1 |
| 55 | 7 | 150 | 66 | 42 | 342 | 34.7 | 0.718 | 42 | 0 |
| 56 | 1 | 73 | 50 | 10 | 0 | 23 | 0.248 | 21 | 0 |
| 57 | 7 | 187 | 68 | 39 | 304 | 37.7 | 0.254 | 41 | 1 |
| 58 | 0 | 100 | 88 | 60 | 110 | 46.8 | 0.962 | 31 | 0 |
| 59 | 0 | 146 | 82 | 0 | 0 | 40.5 | 1.781 | 44 | 0 |
| 60 | 0 | 105 | 64 | 41 | 142 | 41.5 | 0.173 | 22 | 0 |
| 61 | 2 | 84 | 0 | 0 | 0 | 0 | 0.304 | 21 | 0 |
| 62 | 8 | 133 | 72 | 0 | 0 | 32.9 | 0.27 | 39 | 1 |
| 63 | 5 | 44 | 62 | 0 | 0 | 25 | 0.587 | 36 | 0 |
| 64 | 2 | 141 | 58 | 34 | 128 | 25.4 | 0.699 | 24 | 0 |
| 65 | 7 | 114 | 66 | 0 | 0 | 32.8 | 0.258 | 42 | 1 |
| 66 | 5 | 99 | 74 | 27 | 0 | 29 | 0.203 | 32 | 0 |
| 67 | 0 | 109 | 88 | 30 | 0 | 32.5 | 0.855 | 38 | 1 |
| 68 | 2 | 109 | 92 | 0 | 0 | 42.7 | 0.845 | 54 | 0 |
| 69 | 1 | 95 | 66 | 13 | 38 | 19.6 | 0.334 | 25 | 0 |
| 70 | 4 | 146 | 85 | 27 | 100 | 28.9 | 0.189 | 27 | 0 |
| 71 | 2 | 100 | 66 | 20 | 90 | 32.9 | 0.867 | 28 | 1 |
| 72 | 5 | 139 | 64 | 35 | 140 | 28.6 | 0.411 | 26 | 0 |
| 73 | 13 | 126 | 90 | 0 | 0 | 43.4 | 0.583 | 42 | 1 |
| 74 | 4 | 129 | 86 | 20 | 270 | 35.1 | 0.231 | 23 | 0 |
| 75 | 1 | 79 | 75 | 30 | 0 | 32 | 0.396 | 22 | 0 |
| 76 | 1 | 0 | 48 | 20 | 0 | 24.7 | 0.14 | 22 | 0 |
| 77 | 7 | 62 | 78 | 0 | 0 | 32.6 | 0.391 | 41 | 0 |
| 78 | 5 | 95 | 72 | 33 | 0 | 37.7 | 0.37 | 27 | 0 |
| 79 | 0 | 131 | 0 | 0 | 0 | 43.2 | 0.27 | 26 | 1 |
| 80 | 2 | 112 | 66 | 22 | 0 | 25 | 0.307 | 24 | 0 |
| 81 | 3 | 113 | 44 | 13 | 0 | 22.4 | 0.14 | 22 | 0 |
| 82 | 2 | 74 | 0 | 0 | 0 | 0 | 0.102 | 22 | 0 |
| 83 | 7 | 83 | 78 | 26 | 71 | 29.3 | 0.767 | 36 | 0 |
| 84 | 0 | 101 | 65 | 28 | 0 | 24.6 | 0.237 | 22 | 0 |
| 85 | 5 | 137 | 108 | 0 | 0 | 48.8 | 0.227 | 37 | 1 |
| 86 | 2 | 110 | 74 | 29 | 125 | 32.4 | 0.698 | 27 | 0 |
| 87 | 13 | 106 | 72 | 54 | 0 | 36.6 | 0.178 | 45 | 0 |
| 88 | 2 | 100 | 68 | 25 | 71 | 38.5 | 0.324 | 26 | 0 |
| 89 | 15 | 136 | 70 | 32 | 110 | 37.1 | 0.153 | 43 | 1 |
| 90 | 1 | 107 | 68 | 19 | 0 | 26.5 | 0.165 | 24 | 0 |
| 91 | 1 | 80 | 55 | 0 | 0 | 19.1 | 0.258 | 21 | 0 |
| 92 | 4 | 123 | 80 | 15 | 176 | 32 | 0.443 | 34 | 0 |
| 93 | 7 | 81 | 78 | 40 | 48 | 46.7 | 0.261 | 42 | 0 |
| 94 | 4 | 134 | 72 | 0 | 0 | 23.8 | 0.277 | 60 | 1 |
| 95 | 2 | 142 | 82 | 18 | 64 | 24.7 | 0.761 | 21 | 0 |
| 96 | 6 | 144 | 72 | 27 | 228 | 33.9 | 0.255 | 40 | 0 |
| 97 | 2 | 92 | 62 | 28 | 0 | 31.6 | 0.13 | 24 | 0 |
| 98 | 1 | 71 | 48 | 18 | 76 | 20.4 | 0.323 | 22 | 0 |
| 99 | 6 | 93 | 50 | 30 | 64 | 28.7 | 0.356 | 23 | 0 |
| 100 | 1 | 122 | 90 | 51 | 220 | 49.7 | 0.325 | 31 | 1 |
| 101 | 1 | 163 | 72 | 0 | 0 | 39 | 1.222 | 33 | 1 |
| 102 | 1 | 151 | 60 | 0 | 0 | 26.1 | 0.179 | 22 | 0 |
| 103 | 0 | 125 | 96 | 0 | 0 | 22.5 | 0.262 | 21 | 0 |
| 104 | 1 | 81 | 72 | 18 | 40 | 26.6 | 0.283 | 24 | 0 |
| 105 | 2 | 85 | 65 | 0 | 0 | 39.6 | 0.93 | 27 | 0 |
| 106 | 1 | 126 | 56 | 29 | 152 | 28.7 | 0.801 | 21 | 0 |
| 107 | 1 | 96 | 122 | 0 | 0 | 22.4 | 0.207 | 27 | 0 |
| 108 | 4 | 144 | 58 | 28 | 140 | 29.5 | 0.287 | 37 | 0 |
| 109 | 3 | 83 | 58 | 31 | 18 | 34.3 | 0.336 | 25 | 0 |
| 110 | 0 | 95 | 85 | 25 | 36 | 37.4 | 0.247 | 24 | 1 |
| 111 | 3 | 171 | 72 | 33 | 135 | 33.3 | 0.199 | 24 | 1 |
| 112 | 8 | 155 | 62 | 26 | 495 | 34 | 0.543 | 46 | 1 |
| 113 | 1 | 89 | 76 | 34 | 37 | 31.2 | 0.192 | 23 | 0 |
| 114 | 4 | 76 | 62 | 0 | 0 | 34 | 0.391 | 25 | 0 |
| 115 | 7 | 160 | 54 | 32 | 175 | 30.5 | 0.588 | 39 | 1 |
| 116 | 4 | 146 | 92 | 0 | 0 | 31.2 | 0.539 | 61 | 1 |
| 117 | 5 | 124 | 74 | 0 | 0 | 34 | 0.22 | 38 | 1 |
| 118 | 5 | 78 | 48 | 0 | 0 | 33.7 | 0.654 | 25 | 0 |
| 119 | 4 | 97 | 60 | 23 | 0 | 28.2 | 0.443 | 22 | 0 |
| 120 | 4 | 99 | 76 | 15 | 51 | 23.2 | 0.223 | 21 | 0 |
| 121 | 0 | 162 | 76 | 56 | 100 | 53.2 | 0.759 | 25 | 1 |
| 122 | 6 | 111 | 64 | 39 | 0 | 34.2 | 0.26 | 24 | 0 |
| 123 | 2 | 107 | 74 | 30 | 100 | 33.6 | 0.404 | 23 | 0 |
| 124 | 5 | 132 | 80 | 0 | 0 | 26.8 | 0.186 | 69 | 0 |
| 125 | 0 | 113 | 76 | 0 | 0 | 33.3 | 0.278 | 23 | 1 |
| 126 | 1 | 88 | 30 | 42 | 99 | 55 | 0.496 | 26 | 1 |
| 127 | 3 | 120 | 70 | 30 | 135 | 42.9 | 0.452 | 30 | 0 |
| 128 | 1 | 118 | 58 | 36 | 94 | 33.3 | 0.261 | 23 | 0 |
| 129 | 1 | 117 | 88 | 24 | 145 | 34.5 | 0.403 | 40 | 1 |
| 130 | 0 | 105 | 84 | 0 | 0 | 27.9 | 0.741 | 62 | 1 |
| 131 | 4 | 173 | 70 | 14 | 168 | 29.7 | 0.361 | 33 | 1 |
| 132 | 9 | 122 | 56 | 0 | 0 | 33.3 | 1.114 | 33 | 1 |
| 133 | 3 | 170 | 64 | 37 | 225 | 34.5 | 0.356 | 30 | 1 |
| 134 | 8 | 84 | 74 | 31 | 0 | 38.3 | 0.457 | 39 | 0 |
| 135 | 2 | 96 | 68 | 13 | 49 | 21.1 | 0.647 | 26 | 0 |
| 136 | 2 | 125 | 60 | 20 | 140 | 33.8 | 0.088 | 31 | 0 |
| 137 | 0 | 100 | 70 | 26 | 50 | 30.8 | 0.597 | 21 | 0 |
| 138 | 0 | 93 | 60 | 25 | 92 | 28.7 | 0.532 | 22 | 0 |
| 139 | 0 | 129 | 80 | 0 | 0 | 31.2 | 0.703 | 29 | 0 |
| 140 | 5 | 105 | 72 | 29 | 325 | 36.9 | 0.159 | 28 | 0 |
| 141 | 3 | 128 | 78 | 0 | 0 | 21.1 | 0.268 | 55 | 0 |
| 142 | 5 | 106 | 82 | 30 | 0 | 39.5 | 0.286 | 38 | 0 |
| 143 | 2 | 108 | 52 | 26 | 63 | 32.5 | 0.318 | 22 | 0 |
| 144 | 10 | 108 | 66 | 0 | 0 | 32.4 | 0.272 | 42 | 1 |
| 145 | 4 | 154 | 62 | 31 | 284 | 32.8 | 0.237 | 23 | 0 |
| 146 | 0 | 102 | 75 | 23 | 0 | 0 | 0.572 | 21 | 0 |
| 147 | 9 | 57 | 80 | 37 | 0 | 32.8 | 0.096 | 41 | 0 |
| 148 | 2 | 106 | 64 | 35 | 119 | 30.5 | 1.4 | 34 | 0 |
| 149 | 5 | 147 | 78 | 0 | 0 | 33.7 | 0.218 | 65 | 0 |
| 150 | 2 | 90 | 70 | 17 | 0 | 27.3 | 0.085 | 22 | 0 |
| 151 | 1 | 136 | 74 | 50 | 204 | 37.4 | 0.399 | 24 | 0 |
| 152 | 4 | 114 | 65 | 0 | 0 | 21.9 | 0.432 | 37 | 0 |
| 153 | 9 | 156 | 86 | 28 | 155 | 34.3 | 1.189 | 42 | 1 |
| 154 | 1 | 153 | 82 | 42 | 485 | 40.6 | 0.687 | 23 | 0 |
| 155 | 8 | 188 | 78 | 0 | 0 | 47.9 | 0.137 | 43 | 1 |
| 156 | 7 | 152 | 88 | 44 | 0 | 50 | 0.337 | 36 | 1 |
| 157 | 2 | 99 | 52 | 15 | 94 | 24.6 | 0.637 | 21 | 0 |
| 158 | 1 | 109 | 56 | 21 | 135 | 25.2 | 0.833 | 23 | 0 |
| 159 | 2 | 88 | 74 | 19 | 53 | 29 | 0.229 | 22 | 0 |
| 160 | 17 | 163 | 72 | 41 | 114 | 40.9 | 0.817 | 47 | 1 |
| 161 | 4 | 151 | 90 | 38 | 0 | 29.7 | 0.294 | 36 | 0 |
| 162 | 7 | 102 | 74 | 40 | 105 | 37.2 | 0.204 | 45 | 0 |
| 163 | 0 | 114 | 80 | 34 | 285 | 44.2 | 0.167 | 27 | 0 |
| 164 | 2 | 100 | 64 | 23 | 0 | 29.7 | 0.368 | 21 | 0 |
| 165 | 0 | 131 | 88 | 0 | 0 | 31.6 | 0.743 | 32 | 1 |
| 166 | 6 | 104 | 74 | 18 | 156 | 29.9 | 0.722 | 41 | 1 |
| 167 | 3 | 148 | 66 | 25 | 0 | 32.5 | 0.256 | 22 | 0 |
| 168 | 4 | 120 | 68 | 0 | 0 | 29.6 | 0.709 | 34 | 0 |
| 169 | 4 | 110 | 66 | 0 | 0 | 31.9 | 0.471 | 29 | 0 |
| 170 | 3 | 111 | 90 | 12 | 78 | 28.4 | 0.495 | 29 | 0 |
| 171 | 6 | 102 | 82 | 0 | 0 | 30.8 | 0.18 | 36 | 1 |
| 172 | 6 | 134 | 70 | 23 | 130 | 35.4 | 0.542 | 29 | 1 |
| 173 | 2 | 87 | 0 | 23 | 0 | 28.9 | 0.773 | 25 | 0 |
| 174 | 1 | 79 | 60 | 42 | 48 | 43.5 | 0.678 | 23 | 0 |
| 175 | 2 | 75 | 64 | 24 | 55 | 29.7 | 0.37 | 33 | 0 |
| 176 | 8 | 179 | 72 | 42 | 130 | 32.7 | 0.719 | 36 | 1 |
| 177 | 6 | 85 | 78 | 0 | 0 | 31.2 | 0.382 | 42 | 0 |
| 178 | 0 | 129 | 110 | 46 | 130 | 67.1 | 0.319 | 26 | 1 |
| 179 | 5 | 143 | 78 | 0 | 0 | 45 | 0.19 | 47 | 0 |
| 180 | 5 | 130 | 82 | 0 | 0 | 39.1 | 0.956 | 37 | 1 |
| 181 | 6 | 87 | 80 | 0 | 0 | 23.2 | 0.084 | 32 | 0 |
| 182 | 0 | 119 | 64 | 18 | 92 | 34.9 | 0.725 | 23 | 0 |
| 183 | 1 | 0 | 74 | 20 | 23 | 27.7 | 0.299 | 21 | 0 |
| 184 | 5 | 73 | 60 | 0 | 0 | 26.8 | 0.268 | 27 | 0 |
| 185 | 4 | 141 | 74 | 0 | 0 | 27.6 | 0.244 | 40 | 0 |
| 186 | 7 | 194 | 68 | 28 | 0 | 35.9 | 0.745 | 41 | 1 |
| 187 | 8 | 181 | 68 | 36 | 495 | 30.1 | 0.615 | 60 | 1 |
| 188 | 1 | 128 | 98 | 41 | 58 | 32 | 1.321 | 33 | 1 |
| 189 | 8 | 109 | 76 | 39 | 114 | 27.9 | 0.64 | 31 | 1 |
| 190 | 5 | 139 | 80 | 35 | 160 | 31.6 | 0.361 | 25 | 1 |
| 191 | 3 | 111 | 62 | 0 | 0 | 22.6 | 0.142 | 21 | 0 |
| 192 | 9 | 123 | 70 | 44 | 94 | 33.1 | 0.374 | 40 | 0 |
| 193 | 7 | 159 | 66 | 0 | 0 | 30.4 | 0.383 | 36 | 1 |
| 194 | 11 | 135 | 0 | 0 | 0 | 52.3 | 0.578 | 40 | 1 |
| 195 | 8 | 85 | 55 | 20 | 0 | 24.4 | 0.136 | 42 | 0 |
| 196 | 5 | 158 | 84 | 41 | 210 | 39.4 | 0.395 | 29 | 1 |
| 197 | 1 | 105 | 58 | 0 | 0 | 24.3 | 0.187 | 21 | 0 |
| 198 | 3 | 107 | 62 | 13 | 48 | 22.9 | 0.678 | 23 | 1 |
| 199 | 4 | 109 | 64 | 44 | 99 | 34.8 | 0.905 | 26 | 1 |
| 200 | 4 | 148 | 60 | 27 | 318 | 30.9 | 0.15 | 29 | 1 |
| 201 | 0 | 113 | 80 | 16 | 0 | 31 | 0.874 | 21 | 0 |
| 202 | 1 | 138 | 82 | 0 | 0 | 40.1 | 0.236 | 28 | 0 |
| 203 | 0 | 108 | 68 | 20 | 0 | 27.3 | 0.787 | 32 | 0 |
| 204 | 2 | 99 | 70 | 16 | 44 | 20.4 | 0.235 | 27 | 0 |
| 205 | 6 | 103 | 72 | 32 | 190 | 37.7 | 0.324 | 55 | 0 |
| 206 | 5 | 111 | 72 | 28 | 0 | 23.9 | 0.407 | 27 | 0 |
| 207 | 8 | 196 | 76 | 29 | 280 | 37.5 | 0.605 | 57 | 1 |
| 208 | 5 | 162 | 104 | 0 | 0 | 37.7 | 0.151 | 52 | 1 |
| 209 | 1 | 96 | 64 | 27 | 87 | 33.2 | 0.289 | 21 | 0 |
| 210 | 7 | 184 | 84 | 33 | 0 | 35.5 | 0.355 | 41 | 1 |
| 211 | 2 | 81 | 60 | 22 | 0 | 27.7 | 0.29 | 25 | 0 |
| 212 | 0 | 147 | 85 | 54 | 0 | 42.8 | 0.375 | 24 | 0 |
| 213 | 7 | 179 | 95 | 31 | 0 | 34.2 | 0.164 | 60 | 0 |
| 214 | 0 | 140 | 65 | 26 | 130 | 42.6 | 0.431 | 24 | 1 |
| 215 | 9 | 112 | 82 | 32 | 175 | 34.2 | 0.26 | 36 | 1 |
| 216 | 12 | 151 | 70 | 40 | 271 | 41.8 | 0.742 | 38 | 1 |
| 217 | 5 | 109 | 62 | 41 | 129 | 35.8 | 0.514 | 25 | 1 |
| 218 | 6 | 125 | 68 | 30 | 120 | 30 | 0.464 | 32 | 0 |
| 219 | 5 | 85 | 74 | 22 | 0 | 29 | 1.224 | 32 | 1 |
| 220 | 5 | 112 | 66 | 0 | 0 | 37.8 | 0.261 | 41 | 1 |
| 221 | 0 | 177 | 60 | 29 | 478 | 34.6 | 1.072 | 21 | 1 |
| 222 | 2 | 158 | 90 | 0 | 0 | 31.6 | 0.805 | 66 | 1 |
| 223 | 7 | 119 | 0 | 0 | 0 | 25.2 | 0.209 | 37 | 0 |
| 224 | 7 | 142 | 60 | 33 | 190 | 28.8 | 0.687 | 61 | 0 |
| 225 | 1 | 100 | 66 | 15 | 56 | 23.6 | 0.666 | 26 | 0 |
| 226 | 1 | 87 | 78 | 27 | 32 | 34.6 | 0.101 | 22 | 0 |
| 227 | 0 | 101 | 76 | 0 | 0 | 35.7 | 0.198 | 26 | 0 |
| 228 | 3 | 162 | 52 | 38 | 0 | 37.2 | 0.652 | 24 | 1 |
| 229 | 4 | 197 | 70 | 39 | 744 | 36.7 | 2.329 | 31 | 0 |
| 230 | 0 | 117 | 80 | 31 | 53 | 45.2 | 0.089 | 24 | 0 |
| 231 | 4 | 142 | 86 | 0 | 0 | 44 | 0.645 | 22 | 1 |
| 232 | 6 | 134 | 80 | 37 | 370 | 46.2 | 0.238 | 46 | 1 |
| 233 | 1 | 79 | 80 | 25 | 37 | 25.4 | 0.583 | 22 | 0 |
| 234 | 4 | 122 | 68 | 0 | 0 | 35 | 0.394 | 29 | 0 |
| 235 | 3 | 74 | 68 | 28 | 45 | 29.7 | 0.293 | 23 | 0 |
| 236 | 4 | 171 | 72 | 0 | 0 | 43.6 | 0.479 | 26 | 1 |
| 237 | 7 | 181 | 84 | 21 | 192 | 35.9 | 0.586 | 51 | 1 |
| 238 | 0 | 179 | 90 | 27 | 0 | 44.1 | 0.686 | 23 | 1 |
| 239 | 9 | 164 | 84 | 21 | 0 | 30.8 | 0.831 | 32 | 1 |
| 240 | 0 | 104 | 76 | 0 | 0 | 18.4 | 0.582 | 27 | 0 |
| 241 | 1 | 91 | 64 | 24 | 0 | 29.2 | 0.192 | 21 | 0 |
| 242 | 4 | 91 | 70 | 32 | 88 | 33.1 | 0.446 | 22 | 0 |
| 243 | 3 | 139 | 54 | 0 | 0 | 25.6 | 0.402 | 22 | 1 |
| 244 | 6 | 119 | 50 | 22 | 176 | 27.1 | 1.318 | 33 | 1 |
| 245 | 2 | 146 | 76 | 35 | 194 | 38.2 | 0.329 | 29 | 0 |
| 246 | 9 | 184 | 85 | 15 | 0 | 30 | 1.213 | 49 | 1 |
| 247 | 10 | 122 | 68 | 0 | 0 | 31.2 | 0.258 | 41 | 0 |
| 248 | 0 | 165 | 90 | 33 | 680 | 52.3 | 0.427 | 23 | 0 |
| 249 | 9 | 124 | 70 | 33 | 402 | 35.4 | 0.282 | 34 | 0 |
| 250 | 1 | 111 | 86 | 19 | 0 | 30.1 | 0.143 | 23 | 0 |
| 251 | 9 | 106 | 52 | 0 | 0 | 31.2 | 0.38 | 42 | 0 |
| 252 | 2 | 129 | 84 | 0 | 0 | 28 | 0.284 | 27 | 0 |
| 253 | 2 | 90 | 80 | 14 | 55 | 24.4 | 0.249 | 24 | 0 |
| 254 | 0 | 86 | 68 | 32 | 0 | 35.8 | 0.238 | 25 | 0 |
| 255 | 12 | 92 | 62 | 7 | 258 | 27.6 | 0.926 | 44 | 1 |
| 256 | 1 | 113 | 64 | 35 | 0 | 33.6 | 0.543 | 21 | 1 |
| 257 | 3 | 111 | 56 | 39 | 0 | 30.1 | 0.557 | 30 | 0 |
| 258 | 2 | 114 | 68 | 22 | 0 | 28.7 | 0.092 | 25 | 0 |
| 259 | 1 | 193 | 50 | 16 | 375 | 25.9 | 0.655 | 24 | 0 |
| 260 | 11 | 155 | 76 | 28 | 150 | 33.3 | 1.353 | 51 | 1 |
| 261 | 3 | 191 | 68 | 15 | 130 | 30.9 | 0.299 | 34 | 0 |
| 262 | 3 | 141 | 0 | 0 | 0 | 30 | 0.761 | 27 | 1 |
| 263 | 4 | 95 | 70 | 32 | 0 | 32.1 | 0.612 | 24 | 0 |
| 264 | 3 | 142 | 80 | 15 | 0 | 32.4 | 0.2 | 63 | 0 |
| 265 | 4 | 123 | 62 | 0 | 0 | 32 | 0.226 | 35 | 1 |
| 266 | 5 | 96 | 74 | 18 | 67 | 33.6 | 0.997 | 43 | 0 |
| 267 | 0 | 138 | 0 | 0 | 0 | 36.3 | 0.933 | 25 | 1 |
| 268 | 2 | 128 | 64 | 42 | 0 | 40 | 1.101 | 24 | 0 |
| 269 | 0 | 102 | 52 | 0 | 0 | 25.1 | 0.078 | 21 | 0 |
| 270 | 2 | 146 | 0 | 0 | 0 | 27.5 | 0.24 | 28 | 1 |
| 271 | 10 | 101 | 86 | 37 | 0 | 45.6 | 1.136 | 38 | 1 |
| 272 | 2 | 108 | 62 | 32 | 56 | 25.2 | 0.128 | 21 | 0 |
| 273 | 3 | 122 | 78 | 0 | 0 | 23 | 0.254 | 40 | 0 |
| 274 | 1 | 71 | 78 | 50 | 45 | 33.2 | 0.422 | 21 | 0 |
| 275 | 13 | 106 | 70 | 0 | 0 | 34.2 | 0.251 | 52 | 0 |
| 276 | 2 | 100 | 70 | 52 | 57 | 40.5 | 0.677 | 25 | 0 |
| 277 | 7 | 106 | 60 | 24 | 0 | 26.5 | 0.296 | 29 | 1 |
| 278 | 0 | 104 | 64 | 23 | 116 | 27.8 | 0.454 | 23 | 0 |
| 279 | 5 | 114 | 74 | 0 | 0 | 24.9 | 0.744 | 57 | 0 |
| 280 | 2 | 108 | 62 | 10 | 278 | 25.3 | 0.881 | 22 | 0 |
| 281 | 0 | 146 | 70 | 0 | 0 | 37.9 | 0.334 | 28 | 1 |
| 282 | 10 | 129 | 76 | 28 | 122 | 35.9 | 0.28 | 39 | 0 |
| 283 | 7 | 133 | 88 | 15 | 155 | 32.4 | 0.262 | 37 | 0 |
| 284 | 7 | 161 | 86 | 0 | 0 | 30.4 | 0.165 | 47 | 1 |
| 285 | 2 | 108 | 80 | 0 | 0 | 27 | 0.259 | 52 | 1 |
| 286 | 7 | 136 | 74 | 26 | 135 | 26 | 0.647 | 51 | 0 |
| 287 | 5 | 155 | 84 | 44 | 545 | 38.7 | 0.619 | 34 | 0 |
| 288 | 1 | 119 | 86 | 39 | 220 | 45.6 | 0.808 | 29 | 1 |
| 289 | 4 | 96 | 56 | 17 | 49 | 20.8 | 0.34 | 26 | 0 |
| 290 | 5 | 108 | 72 | 43 | 75 | 36.1 | 0.263 | 33 | 0 |
| 291 | 0 | 78 | 88 | 29 | 40 | 36.9 | 0.434 | 21 | 0 |
| 292 | 0 | 107 | 62 | 30 | 74 | 36.6 | 0.757 | 25 | 1 |
| 293 | 2 | 128 | 78 | 37 | 182 | 43.3 | 1.224 | 31 | 1 |
| 294 | 1 | 128 | 48 | 45 | 194 | 40.5 | 0.613 | 24 | 1 |
| 295 | 0 | 161 | 50 | 0 | 0 | 21.9 | 0.254 | 65 | 0 |
| 296 | 6 | 151 | 62 | 31 | 120 | 35.5 | 0.692 | 28 | 0 |
| 297 | 2 | 146 | 70 | 38 | 360 | 28 | 0.337 | 29 | 1 |
| 298 | 0 | 126 | 84 | 29 | 215 | 30.7 | 0.52 | 24 | 0 |
| 299 | 14 | 100 | 78 | 25 | 184 | 36.6 | 0.412 | 46 | 1 |
| 300 | 8 | 112 | 72 | 0 | 0 | 23.6 | 0.84 | 58 | 0 |
| 301 | 0 | 167 | 0 | 0 | 0 | 32.3 | 0.839 | 30 | 1 |
| 302 | 2 | 144 | 58 | 33 | 135 | 31.6 | 0.422 | 25 | 1 |
| 303 | 5 | 77 | 82 | 41 | 42 | 35.8 | 0.156 | 35 | 0 |
| 304 | 5 | 115 | 98 | 0 | 0 | 52.9 | 0.209 | 28 | 1 |
| 305 | 3 | 150 | 76 | 0 | 0 | 21 | 0.207 | 37 | 0 |
| 306 | 2 | 120 | 76 | 37 | 105 | 39.7 | 0.215 | 29 | 0 |
| 307 | 10 | 161 | 68 | 23 | 132 | 25.5 | 0.326 | 47 | 1 |
| 308 | 0 | 137 | 68 | 14 | 148 | 24.8 | 0.143 | 21 | 0 |
| 309 | 0 | 128 | 68 | 19 | 180 | 30.5 | 1.391 | 25 | 1 |
| 310 | 2 | 124 | 68 | 28 | 205 | 32.9 | 0.875 | 30 | 1 |
| 311 | 6 | 80 | 66 | 30 | 0 | 26.2 | 0.313 | 41 | 0 |
| 312 | 0 | 106 | 70 | 37 | 148 | 39.4 | 0.605 | 22 | 0 |
| 313 | 2 | 155 | 74 | 17 | 96 | 26.6 | 0.433 | 27 | 1 |
| 314 | 3 | 113 | 50 | 10 | 85 | 29.5 | 0.626 | 25 | 0 |
| 315 | 7 | 109 | 80 | 31 | 0 | 35.9 | 1.127 | 43 | 1 |
| 316 | 2 | 112 | 68 | 22 | 94 | 34.1 | 0.315 | 26 | 0 |
| 317 | 3 | 99 | 80 | 11 | 64 | 19.3 | 0.284 | 30 | 0 |
| 318 | 3 | 182 | 74 | 0 | 0 | 30.5 | 0.345 | 29 | 1 |
| 319 | 3 | 115 | 66 | 39 | 140 | 38.1 | 0.15 | 28 | 0 |
| 320 | 6 | 194 | 78 | 0 | 0 | 23.5 | 0.129 | 59 | 1 |
| 321 | 4 | 129 | 60 | 12 | 231 | 27.5 | 0.527 | 31 | 0 |
| 322 | 3 | 112 | 74 | 30 | 0 | 31.6 | 0.197 | 25 | 1 |
| 323 | 0 | 124 | 70 | 20 | 0 | 27.4 | 0.254 | 36 | 1 |
| 324 | 13 | 152 | 90 | 33 | 29 | 26.8 | 0.731 | 43 | 1 |
| 325 | 2 | 112 | 75 | 32 | 0 | 35.7 | 0.148 | 21 | 0 |
| 326 | 1 | 157 | 72 | 21 | 168 | 25.6 | 0.123 | 24 | 0 |
| 327 | 1 | 122 | 64 | 32 | 156 | 35.1 | 0.692 | 30 | 1 |
| 328 | 10 | 179 | 70 | 0 | 0 | 35.1 | 0.2 | 37 | 0 |
| 329 | 2 | 102 | 86 | 36 | 120 | 45.5 | 0.127 | 23 | 1 |
| 330 | 6 | 105 | 70 | 32 | 68 | 30.8 | 0.122 | 37 | 0 |
| 331 | 8 | 118 | 72 | 19 | 0 | 23.1 | 1.476 | 46 | 0 |
| 332 | 2 | 87 | 58 | 16 | 52 | 32.7 | 0.166 | 25 | 0 |
| 333 | 1 | 180 | 0 | 0 | 0 | 43.3 | 0.282 | 41 | 1 |
| 334 | 12 | 106 | 80 | 0 | 0 | 23.6 | 0.137 | 44 | 0 |
| 335 | 1 | 95 | 60 | 18 | 58 | 23.9 | 0.26 | 22 | 0 |
| 336 | 0 | 165 | 76 | 43 | 255 | 47.9 | 0.259 | 26 | 0 |
| 337 | 0 | 117 | 0 | 0 | 0 | 33.8 | 0.932 | 44 | 0 |
| 338 | 5 | 115 | 76 | 0 | 0 | 31.2 | 0.343 | 44 | 1 |
| 339 | 9 | 152 | 78 | 34 | 171 | 34.2 | 0.893 | 33 | 1 |
| 340 | 7 | 178 | 84 | 0 | 0 | 39.9 | 0.331 | 41 | 1 |
| 341 | 1 | 130 | 70 | 13 | 105 | 25.9 | 0.472 | 22 | 0 |
| 342 | 1 | 95 | 74 | 21 | 73 | 25.9 | 0.673 | 36 | 0 |
| 343 | 1 | 0 | 68 | 35 | 0 | 32 | 0.389 | 22 | 0 |
| 344 | 5 | 122 | 86 | 0 | 0 | 34.7 | 0.29 | 33 | 0 |
| 345 | 8 | 95 | 72 | 0 | 0 | 36.8 | 0.485 | 57 | 0 |
| 346 | 8 | 126 | 88 | 36 | 108 | 38.5 | 0.349 | 49 | 0 |
| 347 | 1 | 139 | 46 | 19 | 83 | 28.7 | 0.654 | 22 | 0 |
| 348 | 3 | 116 | 0 | 0 | 0 | 23.5 | 0.187 | 23 | 0 |
| 349 | 3 | 99 | 62 | 19 | 74 | 21.8 | 0.279 | 26 | 0 |
| 350 | 5 | 0 | 80 | 32 | 0 | 41 | 0.346 | 37 | 1 |
| 351 | 4 | 92 | 80 | 0 | 0 | 42.2 | 0.237 | 29 | 0 |
| 352 | 4 | 137 | 84 | 0 | 0 | 31.2 | 0.252 | 30 | 0 |
| 353 | 3 | 61 | 82 | 28 | 0 | 34.4 | 0.243 | 46 | 0 |
| 354 | 1 | 90 | 62 | 12 | 43 | 27.2 | 0.58 | 24 | 0 |
| 355 | 3 | 90 | 78 | 0 | 0 | 42.7 | 0.559 | 21 | 0 |
| 356 | 9 | 165 | 88 | 0 | 0 | 30.4 | 0.302 | 49 | 1 |
| 357 | 1 | 125 | 50 | 40 | 167 | 33.3 | 0.962 | 28 | 1 |
| 358 | 13 | 129 | 0 | 30 | 0 | 39.9 | 0.569 | 44 | 1 |
| 359 | 12 | 88 | 74 | 40 | 54 | 35.3 | 0.378 | 48 | 0 |
| 360 | 1 | 196 | 76 | 36 | 249 | 36.5 | 0.875 | 29 | 1 |
| 361 | 5 | 189 | 64 | 33 | 325 | 31.2 | 0.583 | 29 | 1 |
| 362 | 5 | 158 | 70 | 0 | 0 | 29.8 | 0.207 | 63 | 0 |
| 363 | 5 | 103 | 108 | 37 | 0 | 39.2 | 0.305 | 65 | 0 |
| 364 | 4 | 146 | 78 | 0 | 0 | 38.5 | 0.52 | 67 | 1 |
| 365 | 4 | 147 | 74 | 25 | 293 | 34.9 | 0.385 | 30 | 0 |
| 366 | 5 | 99 | 54 | 28 | 83 | 34 | 0.499 | 30 | 0 |
| 367 | 6 | 124 | 72 | 0 | 0 | 27.6 | 0.368 | 29 | 1 |
| 368 | 0 | 101 | 64 | 17 | 0 | 21 | 0.252 | 21 | 0 |
| 369 | 3 | 81 | 86 | 16 | 66 | 27.5 | 0.306 | 22 | 0 |
| 370 | 1 | 133 | 102 | 28 | 140 | 32.8 | 0.234 | 45 | 1 |
| 371 | 3 | 173 | 82 | 48 | 465 | 38.4 | 2.137 | 25 | 1 |
| 372 | 0 | 118 | 64 | 23 | 89 | 0 | 1.731 | 21 | 0 |
| 373 | 0 | 84 | 64 | 22 | 66 | 35.8 | 0.545 | 21 | 0 |
| 374 | 2 | 105 | 58 | 40 | 94 | 34.9 | 0.225 | 25 | 0 |
| 375 | 2 | 122 | 52 | 43 | 158 | 36.2 | 0.816 | 28 | 0 |
| 376 | 12 | 140 | 82 | 43 | 325 | 39.2 | 0.528 | 58 | 1 |
| 377 | 0 | 98 | 82 | 15 | 84 | 25.2 | 0.299 | 22 | 0 |
| 378 | 1 | 87 | 60 | 37 | 75 | 37.2 | 0.509 | 22 | 0 |
| 379 | 4 | 156 | 75 | 0 | 0 | 48.3 | 0.238 | 32 | 1 |
| 380 | 0 | 93 | 100 | 39 | 72 | 43.4 | 1.021 | 35 | 0 |
| 381 | 1 | 107 | 72 | 30 | 82 | 30.8 | 0.821 | 24 | 0 |
| 382 | 0 | 105 | 68 | 22 | 0 | 20 | 0.236 | 22 | 0 |
| 383 | 1 | 109 | 60 | 8 | 182 | 25.4 | 0.947 | 21 | 0 |
| 384 | 1 | 90 | 62 | 18 | 59 | 25.1 | 1.268 | 25 | 0 |
| 385 | 1 | 125 | 70 | 24 | 110 | 24.3 | 0.221 | 25 | 0 |
| 386 | 1 | 119 | 54 | 13 | 50 | 22.3 | 0.205 | 24 | 0 |
| 387 | 5 | 116 | 74 | 29 | 0 | 32.3 | 0.66 | 35 | 1 |
| 388 | 8 | 105 | 100 | 36 | 0 | 43.3 | 0.239 | 45 | 1 |
| 389 | 5 | 144 | 82 | 26 | 285 | 32 | 0.452 | 58 | 1 |
| 390 | 3 | 100 | 68 | 23 | 81 | 31.6 | 0.949 | 28 | 0 |
| 391 | 1 | 100 | 66 | 29 | 196 | 32 | 0.444 | 42 | 0 |
| 392 | 5 | 166 | 76 | 0 | 0 | 45.7 | 0.34 | 27 | 1 |
| 393 | 1 | 131 | 64 | 14 | 415 | 23.7 | 0.389 | 21 | 0 |
| 394 | 4 | 116 | 72 | 12 | 87 | 22.1 | 0.463 | 37 | 0 |
| 395 | 4 | 158 | 78 | 0 | 0 | 32.9 | 0.803 | 31 | 1 |
| 396 | 2 | 127 | 58 | 24 | 275 | 27.7 | 1.6 | 25 | 0 |
| 397 | 3 | 96 | 56 | 34 | 115 | 24.7 | 0.944 | 39 | 0 |
| 398 | 0 | 131 | 66 | 40 | 0 | 34.3 | 0.196 | 22 | 1 |
| 399 | 3 | 82 | 70 | 0 | 0 | 21.1 | 0.389 | 25 | 0 |
| 400 | 3 | 193 | 70 | 31 | 0 | 34.9 | 0.241 | 25 | 1 |
| 401 | 4 | 95 | 64 | 0 | 0 | 32 | 0.161 | 31 | 1 |
| 402 | 6 | 137 | 61 | 0 | 0 | 24.2 | 0.151 | 55 | 0 |
| 403 | 5 | 136 | 84 | 41 | 88 | 35 | 0.286 | 35 | 1 |
| 404 | 9 | 72 | 78 | 25 | 0 | 31.6 | 0.28 | 38 | 0 |
| 405 | 5 | 168 | 64 | 0 | 0 | 32.9 | 0.135 | 41 | 1 |
| 406 | 2 | 123 | 48 | 32 | 165 | 42.1 | 0.52 | 26 | 0 |
| 407 | 4 | 115 | 72 | 0 | 0 | 28.9 | 0.376 | 46 | 1 |
| 408 | 0 | 101 | 62 | 0 | 0 | 21.9 | 0.336 | 25 | 0 |
| 409 | 8 | 197 | 74 | 0 | 0 | 25.9 | 1.191 | 39 | 1 |
| 410 | 1 | 172 | 68 | 49 | 579 | 42.4 | 0.702 | 28 | 1 |
| 411 | 6 | 102 | 90 | 39 | 0 | 35.7 | 0.674 | 28 | 0 |
| 412 | 1 | 112 | 72 | 30 | 176 | 34.4 | 0.528 | 25 | 0 |
| 413 | 1 | 143 | 84 | 23 | 310 | 42.4 | 1.076 | 22 | 0 |
| 414 | 1 | 143 | 74 | 22 | 61 | 26.2 | 0.256 | 21 | 0 |
| 415 | 0 | 138 | 60 | 35 | 167 | 34.6 | 0.534 | 21 | 1 |
| 416 | 3 | 173 | 84 | 33 | 474 | 35.7 | 0.258 | 22 | 1 |
| 417 | 1 | 97 | 68 | 21 | 0 | 27.2 | 1.095 | 22 | 0 |
| 418 | 4 | 144 | 82 | 32 | 0 | 38.5 | 0.554 | 37 | 1 |
| 419 | 1 | 83 | 68 | 0 | 0 | 18.2 | 0.624 | 27 | 0 |
| 420 | 3 | 129 | 64 | 29 | 115 | 26.4 | 0.219 | 28 | 1 |
| 421 | 1 | 119 | 88 | 41 | 170 | 45.3 | 0.507 | 26 | 0 |
| 422 | 2 | 94 | 68 | 18 | 76 | 26 | 0.561 | 21 | 0 |
| 423 | 0 | 102 | 64 | 46 | 78 | 40.6 | 0.496 | 21 | 0 |
| 424 | 2 | 115 | 64 | 22 | 0 | 30.8 | 0.421 | 21 | 0 |
| 425 | 8 | 151 | 78 | 32 | 210 | 42.9 | 0.516 | 36 | 1 |
| 426 | 4 | 184 | 78 | 39 | 277 | 37 | 0.264 | 31 | 1 |
| 427 | 0 | 94 | 0 | 0 | 0 | 0 | 0.256 | 25 | 0 |
| 428 | 1 | 181 | 64 | 30 | 180 | 34.1 | 0.328 | 38 | 1 |
| 429 | 0 | 135 | 94 | 46 | 145 | 40.6 | 0.284 | 26 | 0 |
| 430 | 1 | 95 | 82 | 25 | 180 | 35 | 0.233 | 43 | 1 |
| 431 | 2 | 99 | 0 | 0 | 0 | 22.2 | 0.108 | 23 | 0 |
| 432 | 3 | 89 | 74 | 16 | 85 | 30.4 | 0.551 | 38 | 0 |
| 433 | 1 | 80 | 74 | 11 | 60 | 30 | 0.527 | 22 | 0 |
| 434 | 2 | 139 | 75 | 0 | 0 | 25.6 | 0.167 | 29 | 0 |
| 435 | 1 | 90 | 68 | 8 | 0 | 24.5 | 1.138 | 36 | 0 |
| 436 | 0 | 141 | 0 | 0 | 0 | 42.4 | 0.205 | 29 | 1 |
| 437 | 12 | 140 | 85 | 33 | 0 | 37.4 | 0.244 | 41 | 0 |
| 438 | 5 | 147 | 75 | 0 | 0 | 29.9 | 0.434 | 28 | 0 |
| 439 | 1 | 97 | 70 | 15 | 0 | 18.2 | 0.147 | 21 | 0 |
| 440 | 6 | 107 | 88 | 0 | 0 | 36.8 | 0.727 | 31 | 0 |
| 441 | 0 | 189 | 104 | 25 | 0 | 34.3 | 0.435 | 41 | 1 |
| 442 | 2 | 83 | 66 | 23 | 50 | 32.2 | 0.497 | 22 | 0 |
| 443 | 4 | 117 | 64 | 27 | 120 | 33.2 | 0.23 | 24 | 0 |
| 444 | 8 | 108 | 70 | 0 | 0 | 30.5 | 0.955 | 33 | 1 |
| 445 | 4 | 117 | 62 | 12 | 0 | 29.7 | 0.38 | 30 | 1 |
| 446 | 0 | 180 | 78 | 63 | 14 | 59.4 | 2.42 | 25 | 1 |
| 447 | 1 | 100 | 72 | 12 | 70 | 25.3 | 0.658 | 28 | 0 |
| 448 | 0 | 95 | 80 | 45 | 92 | 36.5 | 0.33 | 26 | 0 |
| 449 | 0 | 104 | 64 | 37 | 64 | 33.6 | 0.51 | 22 | 1 |
| 450 | 0 | 120 | 74 | 18 | 63 | 30.5 | 0.285 | 26 | 0 |
| 451 | 1 | 82 | 64 | 13 | 95 | 21.2 | 0.415 | 23 | 0 |
| 452 | 2 | 134 | 70 | 0 | 0 | 28.9 | 0.542 | 23 | 1 |
| 453 | 0 | 91 | 68 | 32 | 210 | 39.9 | 0.381 | 25 | 0 |
| 454 | 2 | 119 | 0 | 0 | 0 | 19.6 | 0.832 | 72 | 0 |
| 455 | 2 | 100 | 54 | 28 | 105 | 37.8 | 0.498 | 24 | 0 |
| 456 | 14 | 175 | 62 | 30 | 0 | 33.6 | 0.212 | 38 | 1 |
| 457 | 1 | 135 | 54 | 0 | 0 | 26.7 | 0.687 | 62 | 0 |
| 458 | 5 | 86 | 68 | 28 | 71 | 30.2 | 0.364 | 24 | 0 |
| 459 | 10 | 148 | 84 | 48 | 237 | 37.6 | 1.001 | 51 | 1 |
| 460 | 9 | 134 | 74 | 33 | 60 | 25.9 | 0.46 | 81 | 0 |
| 461 | 9 | 120 | 72 | 22 | 56 | 20.8 | 0.733 | 48 | 0 |
| 462 | 1 | 71 | 62 | 0 | 0 | 21.8 | 0.416 | 26 | 0 |
| 463 | 8 | 74 | 70 | 40 | 49 | 35.3 | 0.705 | 39 | 0 |
| 464 | 5 | 88 | 78 | 30 | 0 | 27.6 | 0.258 | 37 | 0 |
| 465 | 10 | 115 | 98 | 0 | 0 | 24 | 1.022 | 34 | 0 |
| 466 | 0 | 124 | 56 | 13 | 105 | 21.8 | 0.452 | 21 | 0 |
| 467 | 0 | 74 | 52 | 10 | 36 | 27.8 | 0.269 | 22 | 0 |
| 468 | 0 | 97 | 64 | 36 | 100 | 36.8 | 0.6 | 25 | 0 |
| 469 | 8 | 120 | 0 | 0 | 0 | 30 | 0.183 | 38 | 1 |
| 470 | 6 | 154 | 78 | 41 | 140 | 46.1 | 0.571 | 27 | 0 |
| 471 | 1 | 144 | 82 | 40 | 0 | 41.3 | 0.607 | 28 | 0 |
| 472 | 0 | 137 | 70 | 38 | 0 | 33.2 | 0.17 | 22 | 0 |
| 473 | 0 | 119 | 66 | 27 | 0 | 38.8 | 0.259 | 22 | 0 |
| 474 | 7 | 136 | 90 | 0 | 0 | 29.9 | 0.21 | 50 | 0 |
| 475 | 4 | 114 | 64 | 0 | 0 | 28.9 | 0.126 | 24 | 0 |
| 476 | 0 | 137 | 84 | 27 | 0 | 27.3 | 0.231 | 59 | 0 |
| 477 | 2 | 105 | 80 | 45 | 191 | 33.7 | 0.711 | 29 | 1 |
| 478 | 7 | 114 | 76 | 17 | 110 | 23.8 | 0.466 | 31 | 0 |
| 479 | 8 | 126 | 74 | 38 | 75 | 25.9 | 0.162 | 39 | 0 |
| 480 | 4 | 132 | 86 | 31 | 0 | 28 | 0.419 | 63 | 0 |
| 481 | 3 | 158 | 70 | 30 | 328 | 35.5 | 0.344 | 35 | 1 |
| 482 | 0 | 123 | 88 | 37 | 0 | 35.2 | 0.197 | 29 | 0 |
| 483 | 4 | 85 | 58 | 22 | 49 | 27.8 | 0.306 | 28 | 0 |
| 484 | 0 | 84 | 82 | 31 | 125 | 38.2 | 0.233 | 23 | 0 |
| 485 | 0 | 145 | 0 | 0 | 0 | 44.2 | 0.63 | 31 | 1 |
| 486 | 0 | 135 | 68 | 42 | 250 | 42.3 | 0.365 | 24 | 1 |
| 487 | 1 | 139 | 62 | 41 | 480 | 40.7 | 0.536 | 21 | 0 |
| 488 | 0 | 173 | 78 | 32 | 265 | 46.5 | 1.159 | 58 | 0 |
| 489 | 4 | 99 | 72 | 17 | 0 | 25.6 | 0.294 | 28 | 0 |
| 490 | 8 | 194 | 80 | 0 | 0 | 26.1 | 0.551 | 67 | 0 |
| 491 | 2 | 83 | 65 | 28 | 66 | 36.8 | 0.629 | 24 | 0 |
| 492 | 2 | 89 | 90 | 30 | 0 | 33.5 | 0.292 | 42 | 0 |
| 493 | 4 | 99 | 68 | 38 | 0 | 32.8 | 0.145 | 33 | 0 |
| 494 | 4 | 125 | 70 | 18 | 122 | 28.9 | 1.144 | 45 | 1 |
| 495 | 3 | 80 | 0 | 0 | 0 | 0 | 0.174 | 22 | 0 |
| 496 | 6 | 166 | 74 | 0 | 0 | 26.6 | 0.304 | 66 | 0 |
| 497 | 5 | 110 | 68 | 0 | 0 | 26 | 0.292 | 30 | 0 |
| 498 | 2 | 81 | 72 | 15 | 76 | 30.1 | 0.547 | 25 | 0 |
| 499 | 7 | 195 | 70 | 33 | 145 | 25.1 | 0.163 | 55 | 1 |
| 500 | 6 | 154 | 74 | 32 | 193 | 29.3 | 0.839 | 39 | 0 |
| 501 | 2 | 117 | 90 | 19 | 71 | 25.2 | 0.313 | 21 | 0 |
| 502 | 3 | 84 | 72 | 32 | 0 | 37.2 | 0.267 | 28 | 0 |
| 503 | 6 | 0 | 68 | 41 | 0 | 39 | 0.727 | 41 | 1 |
| 504 | 7 | 94 | 64 | 25 | 79 | 33.3 | 0.738 | 41 | 0 |
| 505 | 3 | 96 | 78 | 39 | 0 | 37.3 | 0.238 | 40 | 0 |
| 506 | 10 | 75 | 82 | 0 | 0 | 33.3 | 0.263 | 38 | 0 |
| 507 | 0 | 180 | 90 | 26 | 90 | 36.5 | 0.314 | 35 | 1 |
| 508 | 1 | 130 | 60 | 23 | 170 | 28.6 | 0.692 | 21 | 0 |
| 509 | 2 | 84 | 50 | 23 | 76 | 30.4 | 0.968 | 21 | 0 |
| 510 | 8 | 120 | 78 | 0 | 0 | 25 | 0.409 | 64 | 0 |
| 511 | 12 | 84 | 72 | 31 | 0 | 29.7 | 0.297 | 46 | 1 |
| 512 | 0 | 139 | 62 | 17 | 210 | 22.1 | 0.207 | 21 | 0 |
| 513 | 9 | 91 | 68 | 0 | 0 | 24.2 | 0.2 | 58 | 0 |
| 514 | 2 | 91 | 62 | 0 | 0 | 27.3 | 0.525 | 22 | 0 |
| 515 | 3 | 99 | 54 | 19 | 86 | 25.6 | 0.154 | 24 | 0 |
| 516 | 3 | 163 | 70 | 18 | 105 | 31.6 | 0.268 | 28 | 1 |
| 517 | 9 | 145 | 88 | 34 | 165 | 30.3 | 0.771 | 53 | 1 |
| 518 | 7 | 125 | 86 | 0 | 0 | 37.6 | 0.304 | 51 | 0 |
| 519 | 13 | 76 | 60 | 0 | 0 | 32.8 | 0.18 | 41 | 0 |
| 520 | 6 | 129 | 90 | 7 | 326 | 19.6 | 0.582 | 60 | 0 |
| 521 | 2 | 68 | 70 | 32 | 66 | 25 | 0.187 | 25 | 0 |
| 522 | 3 | 124 | 80 | 33 | 130 | 33.2 | 0.305 | 26 | 0 |
| 523 | 6 | 114 | 0 | 0 | 0 | 0 | 0.189 | 26 | 0 |
| 524 | 9 | 130 | 70 | 0 | 0 | 34.2 | 0.652 | 45 | 1 |
| 525 | 3 | 125 | 58 | 0 | 0 | 31.6 | 0.151 | 24 | 0 |
| 526 | 3 | 87 | 60 | 18 | 0 | 21.8 | 0.444 | 21 | 0 |
| 527 | 1 | 97 | 64 | 19 | 82 | 18.2 | 0.299 | 21 | 0 |
| 528 | 3 | 116 | 74 | 15 | 105 | 26.3 | 0.107 | 24 | 0 |
| 529 | 0 | 117 | 66 | 31 | 188 | 30.8 | 0.493 | 22 | 0 |
| 530 | 0 | 111 | 65 | 0 | 0 | 24.6 | 0.66 | 31 | 0 |
| 531 | 2 | 122 | 60 | 18 | 106 | 29.8 | 0.717 | 22 | 0 |
| 532 | 0 | 107 | 76 | 0 | 0 | 45.3 | 0.686 | 24 | 0 |
| 533 | 1 | 86 | 66 | 52 | 65 | 41.3 | 0.917 | 29 | 0 |
| 534 | 6 | 91 | 0 | 0 | 0 | 29.8 | 0.501 | 31 | 0 |
| 535 | 1 | 77 | 56 | 30 | 56 | 33.3 | 1.251 | 24 | 0 |
| 536 | 4 | 132 | 0 | 0 | 0 | 32.9 | 0.302 | 23 | 1 |
| 537 | 0 | 105 | 90 | 0 | 0 | 29.6 | 0.197 | 46 | 0 |
| 538 | 0 | 57 | 60 | 0 | 0 | 21.7 | 0.735 | 67 | 0 |
| 539 | 0 | 127 | 80 | 37 | 210 | 36.3 | 0.804 | 23 | 0 |
| 540 | 3 | 129 | 92 | 49 | 155 | 36.4 | 0.968 | 32 | 1 |
| 541 | 8 | 100 | 74 | 40 | 215 | 39.4 | 0.661 | 43 | 1 |
| 542 | 3 | 128 | 72 | 25 | 190 | 32.4 | 0.549 | 27 | 1 |
| 543 | 10 | 90 | 85 | 32 | 0 | 34.9 | 0.825 | 56 | 1 |
| 544 | 4 | 84 | 90 | 23 | 56 | 39.5 | 0.159 | 25 | 0 |
| 545 | 1 | 88 | 78 | 29 | 76 | 32 | 0.365 | 29 | 0 |
| 546 | 8 | 186 | 90 | 35 | 225 | 34.5 | 0.423 | 37 | 1 |
| 547 | 5 | 187 | 76 | 27 | 207 | 43.6 | 1.034 | 53 | 1 |
| 548 | 4 | 131 | 68 | 21 | 166 | 33.1 | 0.16 | 28 | 0 |
| 549 | 1 | 164 | 82 | 43 | 67 | 32.8 | 0.341 | 50 | 0 |
| 550 | 4 | 189 | 110 | 31 | 0 | 28.5 | 0.68 | 37 | 0 |
| 551 | 1 | 116 | 70 | 28 | 0 | 27.4 | 0.204 | 21 | 0 |
| 552 | 3 | 84 | 68 | 30 | 106 | 31.9 | 0.591 | 25 | 0 |
| 553 | 6 | 114 | 88 | 0 | 0 | 27.8 | 0.247 | 66 | 0 |
| 554 | 1 | 88 | 62 | 24 | 44 | 29.9 | 0.422 | 23 | 0 |
| 555 | 1 | 84 | 64 | 23 | 115 | 36.9 | 0.471 | 28 | 0 |
| 556 | 7 | 124 | 70 | 33 | 215 | 25.5 | 0.161 | 37 | 0 |
| 557 | 1 | 97 | 70 | 40 | 0 | 38.1 | 0.218 | 30 | 0 |
| 558 | 8 | 110 | 76 | 0 | 0 | 27.8 | 0.237 | 58 | 0 |
| 559 | 11 | 103 | 68 | 40 | 0 | 46.2 | 0.126 | 42 | 0 |
| 560 | 11 | 85 | 74 | 0 | 0 | 30.1 | 0.3 | 35 | 0 |
| 561 | 6 | 125 | 76 | 0 | 0 | 33.8 | 0.121 | 54 | 1 |
| 562 | 0 | 198 | 66 | 32 | 274 | 41.3 | 0.502 | 28 | 1 |
| 563 | 1 | 87 | 68 | 34 | 77 | 37.6 | 0.401 | 24 | 0 |
| 564 | 6 | 99 | 60 | 19 | 54 | 26.9 | 0.497 | 32 | 0 |
| 565 | 0 | 91 | 80 | 0 | 0 | 32.4 | 0.601 | 27 | 0 |
| 566 | 2 | 95 | 54 | 14 | 88 | 26.1 | 0.748 | 22 | 0 |
| 567 | 1 | 99 | 72 | 30 | 18 | 38.6 | 0.412 | 21 | 0 |
| 568 | 6 | 92 | 62 | 32 | 126 | 32 | 0.085 | 46 | 0 |
| 569 | 4 | 154 | 72 | 29 | 126 | 31.3 | 0.338 | 37 | 0 |
| 570 | 0 | 121 | 66 | 30 | 165 | 34.3 | 0.203 | 33 | 1 |
| 571 | 3 | 78 | 70 | 0 | 0 | 32.5 | 0.27 | 39 | 0 |
| 572 | 2 | 130 | 96 | 0 | 0 | 22.6 | 0.268 | 21 | 0 |
| 573 | 3 | 111 | 58 | 31 | 44 | 29.5 | 0.43 | 22 | 0 |
| 574 | 2 | 98 | 60 | 17 | 120 | 34.7 | 0.198 | 22 | 0 |
| 575 | 1 | 143 | 86 | 30 | 330 | 30.1 | 0.892 | 23 | 0 |
| 576 | 1 | 119 | 44 | 47 | 63 | 35.5 | 0.28 | 25 | 0 |
| 577 | 6 | 108 | 44 | 20 | 130 | 24 | 0.813 | 35 | 0 |
| 578 | 2 | 118 | 80 | 0 | 0 | 42.9 | 0.693 | 21 | 1 |
| 579 | 10 | 133 | 68 | 0 | 0 | 27 | 0.245 | 36 | 0 |
| 580 | 2 | 197 | 70 | 99 | 0 | 34.7 | 0.575 | 62 | 1 |
| 581 | 0 | 151 | 90 | 46 | 0 | 42.1 | 0.371 | 21 | 1 |
| 582 | 6 | 109 | 60 | 27 | 0 | 25 | 0.206 | 27 | 0 |
| 583 | 12 | 121 | 78 | 17 | 0 | 26.5 | 0.259 | 62 | 0 |
| 584 | 8 | 100 | 76 | 0 | 0 | 38.7 | 0.19 | 42 | 0 |
| 585 | 8 | 124 | 76 | 24 | 600 | 28.7 | 0.687 | 52 | 1 |
| 586 | 1 | 93 | 56 | 11 | 0 | 22.5 | 0.417 | 22 | 0 |
| 587 | 8 | 143 | 66 | 0 | 0 | 34.9 | 0.129 | 41 | 1 |
| 588 | 6 | 103 | 66 | 0 | 0 | 24.3 | 0.249 | 29 | 0 |
| 589 | 3 | 176 | 86 | 27 | 156 | 33.3 | 1.154 | 52 | 1 |
| 590 | 0 | 73 | 0 | 0 | 0 | 21.1 | 0.342 | 25 | 0 |
| 591 | 11 | 111 | 84 | 40 | 0 | 46.8 | 0.925 | 45 | 1 |
| 592 | 2 | 112 | 78 | 50 | 140 | 39.4 | 0.175 | 24 | 0 |
| 593 | 3 | 132 | 80 | 0 | 0 | 34.4 | 0.402 | 44 | 1 |
| 594 | 2 | 82 | 52 | 22 | 115 | 28.5 | 1.699 | 25 | 0 |
| 595 | 6 | 123 | 72 | 45 | 230 | 33.6 | 0.733 | 34 | 0 |
| 596 | 0 | 188 | 82 | 14 | 185 | 32 | 0.682 | 22 | 1 |
| 597 | 0 | 67 | 76 | 0 | 0 | 45.3 | 0.194 | 46 | 0 |
| 598 | 1 | 89 | 24 | 19 | 25 | 27.8 | 0.559 | 21 | 0 |
| 599 | 1 | 173 | 74 | 0 | 0 | 36.8 | 0.088 | 38 | 1 |
| 600 | 1 | 109 | 38 | 18 | 120 | 23.1 | 0.407 | 26 | 0 |
| 601 | 1 | 108 | 88 | 19 | 0 | 27.1 | 0.4 | 24 | 0 |
| 602 | 6 | 96 | 0 | 0 | 0 | 23.7 | 0.19 | 28 | 0 |
| 603 | 1 | 124 | 74 | 36 | 0 | 27.8 | 0.1 | 30 | 0 |
| 604 | 7 | 150 | 78 | 29 | 126 | 35.2 | 0.692 | 54 | 1 |
| 605 | 4 | 183 | 0 | 0 | 0 | 28.4 | 0.212 | 36 | 1 |
| 606 | 1 | 124 | 60 | 32 | 0 | 35.8 | 0.514 | 21 | 0 |
| 607 | 1 | 181 | 78 | 42 | 293 | 40 | 1.258 | 22 | 1 |
| 608 | 1 | 92 | 62 | 25 | 41 | 19.5 | 0.482 | 25 | 0 |
| 609 | 0 | 152 | 82 | 39 | 272 | 41.5 | 0.27 | 27 | 0 |
| 610 | 1 | 111 | 62 | 13 | 182 | 24 | 0.138 | 23 | 0 |
| 611 | 3 | 106 | 54 | 21 | 158 | 30.9 | 0.292 | 24 | 0 |
| 612 | 3 | 174 | 58 | 22 | 194 | 32.9 | 0.593 | 36 | 1 |
| 613 | 7 | 168 | 88 | 42 | 321 | 38.2 | 0.787 | 40 | 1 |
| 614 | 6 | 105 | 80 | 28 | 0 | 32.5 | 0.878 | 26 | 0 |
| 615 | 11 | 138 | 74 | 26 | 144 | 36.1 | 0.557 | 50 | 1 |
| 616 | 3 | 106 | 72 | 0 | 0 | 25.8 | 0.207 | 27 | 0 |
| 617 | 6 | 117 | 96 | 0 | 0 | 28.7 | 0.157 | 30 | 0 |
| 618 | 2 | 68 | 62 | 13 | 15 | 20.1 | 0.257 | 23 | 0 |
| 619 | 9 | 112 | 82 | 24 | 0 | 28.2 | 1.282 | 50 | 1 |
| 620 | 0 | 119 | 0 | 0 | 0 | 32.4 | 0.141 | 24 | 1 |
| 621 | 2 | 112 | 86 | 42 | 160 | 38.4 | 0.246 | 28 | 0 |
| 622 | 2 | 92 | 76 | 20 | 0 | 24.2 | 1.698 | 28 | 0 |
| 623 | 6 | 183 | 94 | 0 | 0 | 40.8 | 1.461 | 45 | 0 |
| 624 | 0 | 94 | 70 | 27 | 115 | 43.5 | 0.347 | 21 | 0 |
| 625 | 2 | 108 | 64 | 0 | 0 | 30.8 | 0.158 | 21 | 0 |
| 626 | 4 | 90 | 88 | 47 | 54 | 37.7 | 0.362 | 29 | 0 |
| 627 | 0 | 125 | 68 | 0 | 0 | 24.7 | 0.206 | 21 | 0 |
| 628 | 0 | 132 | 78 | 0 | 0 | 32.4 | 0.393 | 21 | 0 |
| 629 | 5 | 128 | 80 | 0 | 0 | 34.6 | 0.144 | 45 | 0 |
| 630 | 4 | 94 | 65 | 22 | 0 | 24.7 | 0.148 | 21 | 0 |
| 631 | 7 | 114 | 64 | 0 | 0 | 27.4 | 0.732 | 34 | 1 |
| 632 | 0 | 102 | 78 | 40 | 90 | 34.5 | 0.238 | 24 | 0 |
| 633 | 2 | 111 | 60 | 0 | 0 | 26.2 | 0.343 | 23 | 0 |
| 634 | 1 | 128 | 82 | 17 | 183 | 27.5 | 0.115 | 22 | 0 |
| 635 | 10 | 92 | 62 | 0 | 0 | 25.9 | 0.167 | 31 | 0 |
| 636 | 13 | 104 | 72 | 0 | 0 | 31.2 | 0.465 | 38 | 1 |
| 637 | 5 | 104 | 74 | 0 | 0 | 28.8 | 0.153 | 48 | 0 |
| 638 | 2 | 94 | 76 | 18 | 66 | 31.6 | 0.649 | 23 | 0 |
| 639 | 7 | 97 | 76 | 32 | 91 | 40.9 | 0.871 | 32 | 1 |
| 640 | 1 | 100 | 74 | 12 | 46 | 19.5 | 0.149 | 28 | 0 |
| 641 | 0 | 102 | 86 | 17 | 105 | 29.3 | 0.695 | 27 | 0 |
| 642 | 4 | 128 | 70 | 0 | 0 | 34.3 | 0.303 | 24 | 0 |
| 643 | 6 | 147 | 80 | 0 | 0 | 29.5 | 0.178 | 50 | 1 |
| 644 | 4 | 90 | 0 | 0 | 0 | 28 | 0.61 | 31 | 0 |
| 645 | 3 | 103 | 72 | 30 | 152 | 27.6 | 0.73 | 27 | 0 |
| 646 | 2 | 157 | 74 | 35 | 440 | 39.4 | 0.134 | 30 | 0 |
| 647 | 1 | 167 | 74 | 17 | 144 | 23.4 | 0.447 | 33 | 1 |
| 648 | 0 | 179 | 50 | 36 | 159 | 37.8 | 0.455 | 22 | 1 |
| 649 | 11 | 136 | 84 | 35 | 130 | 28.3 | 0.26 | 42 | 1 |
| 650 | 0 | 107 | 60 | 25 | 0 | 26.4 | 0.133 | 23 | 0 |
| 651 | 1 | 91 | 54 | 25 | 100 | 25.2 | 0.234 | 23 | 0 |
| 652 | 1 | 117 | 60 | 23 | 106 | 33.8 | 0.466 | 27 | 0 |
| 653 | 5 | 123 | 74 | 40 | 77 | 34.1 | 0.269 | 28 | 0 |
| 654 | 2 | 120 | 54 | 0 | 0 | 26.8 | 0.455 | 27 | 0 |
| 655 | 1 | 106 | 70 | 28 | 135 | 34.2 | 0.142 | 22 | 0 |
| 656 | 2 | 155 | 52 | 27 | 540 | 38.7 | 0.24 | 25 | 1 |
| 657 | 2 | 101 | 58 | 35 | 90 | 21.8 | 0.155 | 22 | 0 |
| 658 | 1 | 120 | 80 | 48 | 200 | 38.9 | 1.162 | 41 | 0 |
| 659 | 11 | 127 | 106 | 0 | 0 | 39 | 0.19 | 51 | 0 |
| 660 | 3 | 80 | 82 | 31 | 70 | 34.2 | 1.292 | 27 | 1 |
| 661 | 10 | 162 | 84 | 0 | 0 | 27.7 | 0.182 | 54 | 0 |
| 662 | 1 | 199 | 76 | 43 | 0 | 42.9 | 1.394 | 22 | 1 |
| 663 | 8 | 167 | 106 | 46 | 231 | 37.6 | 0.165 | 43 | 1 |
| 664 | 9 | 145 | 80 | 46 | 130 | 37.9 | 0.637 | 40 | 1 |
| 665 | 6 | 115 | 60 | 39 | 0 | 33.7 | 0.245 | 40 | 1 |
| 666 | 1 | 112 | 80 | 45 | 132 | 34.8 | 0.217 | 24 | 0 |
| 667 | 4 | 145 | 82 | 18 | 0 | 32.5 | 0.235 | 70 | 1 |
| 668 | 10 | 111 | 70 | 27 | 0 | 27.5 | 0.141 | 40 | 1 |
| 669 | 6 | 98 | 58 | 33 | 190 | 34 | 0.43 | 43 | 0 |
| 670 | 9 | 154 | 78 | 30 | 100 | 30.9 | 0.164 | 45 | 0 |
| 671 | 6 | 165 | 68 | 26 | 168 | 33.6 | 0.631 | 49 | 0 |
| 672 | 1 | 99 | 58 | 10 | 0 | 25.4 | 0.551 | 21 | 0 |
| 673 | 10 | 68 | 106 | 23 | 49 | 35.5 | 0.285 | 47 | 0 |
| 674 | 3 | 123 | 100 | 35 | 240 | 57.3 | 0.88 | 22 | 0 |
| 675 | 8 | 91 | 82 | 0 | 0 | 35.6 | 0.587 | 68 | 0 |
| 676 | 6 | 195 | 70 | 0 | 0 | 30.9 | 0.328 | 31 | 1 |
| 677 | 9 | 156 | 86 | 0 | 0 | 24.8 | 0.23 | 53 | 1 |
| 678 | 0 | 93 | 60 | 0 | 0 | 35.3 | 0.263 | 25 | 0 |
| 679 | 3 | 121 | 52 | 0 | 0 | 36 | 0.127 | 25 | 1 |
| 680 | 2 | 101 | 58 | 17 | 265 | 24.2 | 0.614 | 23 | 0 |
| 681 | 2 | 56 | 56 | 28 | 45 | 24.2 | 0.332 | 22 | 0 |
| 682 | 0 | 162 | 76 | 36 | 0 | 49.6 | 0.364 | 26 | 1 |
| 683 | 0 | 95 | 64 | 39 | 105 | 44.6 | 0.366 | 22 | 0 |
| 684 | 4 | 125 | 80 | 0 | 0 | 32.3 | 0.536 | 27 | 1 |
| 685 | 5 | 136 | 82 | 0 | 0 | 0 | 0.64 | 69 | 0 |
| 686 | 2 | 129 | 74 | 26 | 205 | 33.2 | 0.591 | 25 | 0 |
| 687 | 3 | 130 | 64 | 0 | 0 | 23.1 | 0.314 | 22 | 0 |
| 688 | 1 | 107 | 50 | 19 | 0 | 28.3 | 0.181 | 29 | 0 |
| 689 | 1 | 140 | 74 | 26 | 180 | 24.1 | 0.828 | 23 | 0 |
| 690 | 1 | 144 | 82 | 46 | 180 | 46.1 | 0.335 | 46 | 1 |
| 691 | 8 | 107 | 80 | 0 | 0 | 24.6 | 0.856 | 34 | 0 |
| 692 | 13 | 158 | 114 | 0 | 0 | 42.3 | 0.257 | 44 | 1 |
| 693 | 2 | 121 | 70 | 32 | 95 | 39.1 | 0.886 | 23 | 0 |
| 694 | 7 | 129 | 68 | 49 | 125 | 38.5 | 0.439 | 43 | 1 |
| 695 | 2 | 90 | 60 | 0 | 0 | 23.5 | 0.191 | 25 | 0 |
| 696 | 7 | 142 | 90 | 24 | 480 | 30.4 | 0.128 | 43 | 1 |
| 697 | 3 | 169 | 74 | 19 | 125 | 29.9 | 0.268 | 31 | 1 |
| 698 | 0 | 99 | 0 | 0 | 0 | 25 | 0.253 | 22 | 0 |
| 699 | 4 | 127 | 88 | 11 | 155 | 34.5 | 0.598 | 28 | 0 |
| 700 | 4 | 118 | 70 | 0 | 0 | 44.5 | 0.904 | 26 | 0 |
| 701 | 2 | 122 | 76 | 27 | 200 | 35.9 | 0.483 | 26 | 0 |
| 702 | 6 | 125 | 78 | 31 | 0 | 27.6 | 0.565 | 49 | 1 |
| 703 | 1 | 168 | 88 | 29 | 0 | 35 | 0.905 | 52 | 1 |
| 704 | 2 | 129 | 0 | 0 | 0 | 38.5 | 0.304 | 41 | 0 |
| 705 | 4 | 110 | 76 | 20 | 100 | 28.4 | 0.118 | 27 | 0 |
| 706 | 6 | 80 | 80 | 36 | 0 | 39.8 | 0.177 | 28 | 0 |
| 707 | 10 | 115 | 0 | 0 | 0 | 0 | 0.261 | 30 | 1 |
| 708 | 2 | 127 | 46 | 21 | 335 | 34.4 | 0.176 | 22 | 0 |
| 709 | 9 | 164 | 78 | 0 | 0 | 32.8 | 0.148 | 45 | 1 |
| 710 | 2 | 93 | 64 | 32 | 160 | 38 | 0.674 | 23 | 1 |
| 711 | 3 | 158 | 64 | 13 | 387 | 31.2 | 0.295 | 24 | 0 |
| 712 | 5 | 126 | 78 | 27 | 22 | 29.6 | 0.439 | 40 | 0 |
| 713 | 10 | 129 | 62 | 36 | 0 | 41.2 | 0.441 | 38 | 1 |
| 714 | 0 | 134 | 58 | 20 | 291 | 26.4 | 0.352 | 21 | 0 |
| 715 | 3 | 102 | 74 | 0 | 0 | 29.5 | 0.121 | 32 | 0 |
| 716 | 7 | 187 | 50 | 33 | 392 | 33.9 | 0.826 | 34 | 1 |
| 717 | 3 | 173 | 78 | 39 | 185 | 33.8 | 0.97 | 31 | 1 |
| 718 | 10 | 94 | 72 | 18 | 0 | 23.1 | 0.595 | 56 | 0 |
| 719 | 1 | 108 | 60 | 46 | 178 | 35.5 | 0.415 | 24 | 0 |
| 720 | 5 | 97 | 76 | 27 | 0 | 35.6 | 0.378 | 52 | 1 |
| 721 | 4 | 83 | 86 | 19 | 0 | 29.3 | 0.317 | 34 | 0 |
| 722 | 1 | 114 | 66 | 36 | 200 | 38.1 | 0.289 | 21 | 0 |
| 723 | 1 | 149 | 68 | 29 | 127 | 29.3 | 0.349 | 42 | 1 |
| 724 | 5 | 117 | 86 | 30 | 105 | 39.1 | 0.251 | 42 | 0 |
| 725 | 1 | 111 | 94 | 0 | 0 | 32.8 | 0.265 | 45 | 0 |
| 726 | 4 | 112 | 78 | 40 | 0 | 39.4 | 0.236 | 38 | 0 |
| 727 | 1 | 116 | 78 | 29 | 180 | 36.1 | 0.496 | 25 | 0 |
| 728 | 0 | 141 | 84 | 26 | 0 | 32.4 | 0.433 | 22 | 0 |
| 729 | 2 | 175 | 88 | 0 | 0 | 22.9 | 0.326 | 22 | 0 |
| 730 | 2 | 92 | 52 | 0 | 0 | 30.1 | 0.141 | 22 | 0 |
| 731 | 3 | 130 | 78 | 23 | 79 | 28.4 | 0.323 | 34 | 1 |
| 732 | 8 | 120 | 86 | 0 | 0 | 28.4 | 0.259 | 22 | 1 |
| 733 | 2 | 174 | 88 | 37 | 120 | 44.5 | 0.646 | 24 | 1 |
| 734 | 2 | 106 | 56 | 27 | 165 | 29 | 0.426 | 22 | 0 |
| 735 | 2 | 105 | 75 | 0 | 0 | 23.3 | 0.56 | 53 | 0 |
| 736 | 4 | 95 | 60 | 32 | 0 | 35.4 | 0.284 | 28 | 0 |
| 737 | 0 | 126 | 86 | 27 | 120 | 27.4 | 0.515 | 21 | 0 |
| 738 | 8 | 65 | 72 | 23 | 0 | 32 | 0.6 | 42 | 0 |
| 739 | 2 | 99 | 60 | 17 | 160 | 36.6 | 0.453 | 21 | 0 |
| 740 | 1 | 102 | 74 | 0 | 0 | 39.5 | 0.293 | 42 | 1 |
| 741 | 11 | 120 | 80 | 37 | 150 | 42.3 | 0.785 | 48 | 1 |
| 742 | 3 | 102 | 44 | 20 | 94 | 30.8 | 0.4 | 26 | 0 |
| 743 | 1 | 109 | 58 | 18 | 116 | 28.5 | 0.219 | 22 | 0 |
| 744 | 9 | 140 | 94 | 0 | 0 | 32.7 | 0.734 | 45 | 1 |
| 745 | 13 | 153 | 88 | 37 | 140 | 40.6 | 1.174 | 39 | 0 |
| 746 | 12 | 100 | 84 | 33 | 105 | 30 | 0.488 | 46 | 0 |
| 747 | 1 | 147 | 94 | 41 | 0 | 49.3 | 0.358 | 27 | 1 |
| 748 | 1 | 81 | 74 | 41 | 57 | 46.3 | 1.096 | 32 | 0 |
| 749 | 3 | 187 | 70 | 22 | 200 | 36.4 | 0.408 | 36 | 1 |
| 750 | 6 | 162 | 62 | 0 | 0 | 24.3 | 0.178 | 50 | 1 |
| 751 | 4 | 136 | 70 | 0 | 0 | 31.2 | 1.182 | 22 | 1 |
| 752 | 1 | 121 | 78 | 39 | 74 | 39 | 0.261 | 28 | 0 |
| 753 | 3 | 108 | 62 | 24 | 0 | 26 | 0.223 | 25 | 0 |
| 754 | 0 | 181 | 88 | 44 | 510 | 43.3 | 0.222 | 26 | 1 |
| 755 | 8 | 154 | 78 | 32 | 0 | 32.4 | 0.443 | 45 | 1 |
| 756 | 1 | 128 | 88 | 39 | 110 | 36.5 | 1.057 | 37 | 1 |
| 757 | 7 | 137 | 90 | 41 | 0 | 32 | 0.391 | 39 | 0 |
| 758 | 0 | 123 | 72 | 0 | 0 | 36.3 | 0.258 | 52 | 1 |
| 759 | 1 | 106 | 76 | 0 | 0 | 37.5 | 0.197 | 26 | 0 |
| 760 | 6 | 190 | 92 | 0 | 0 | 35.5 | 0.278 | 66 | 1 |
| 761 | 2 | 88 | 58 | 26 | 16 | 28.4 | 0.766 | 22 | 0 |
| 762 | 9 | 170 | 74 | 31 | 0 | 44 | 0.403 | 43 | 1 |
| 763 | 9 | 89 | 62 | 0 | 0 | 22.5 | 0.142 | 33 | 0 |
| 764 | 10 | 101 | 76 | 48 | 180 | 32.9 | 0.171 | 63 | 0 |
| 765 | 2 | 122 | 70 | 27 | 0 | 36.8 | 0.34 | 27 | 0 |
| 766 | 5 | 121 | 72 | 23 | 112 | 26.2 | 0.245 | 30 | 0 |
| 767 | 1 | 126 | 60 | 0 | 0 | 30.1 | 0.349 | 47 | 1 |
| 768 | 1 | 93 | 70 | 31 | 0 | 30.4 | 0.315 | 23 | 0 |