test_2_notebook

July 14, 2021

1 Y-Data Test 2

1.1 D2: Implement

You are given a chess board with n rows and n columns. There are m chess rooks located on the board. Determine a number of cells that are free* and not under attack by any rook. (See illustration below – places threatened by the rook are marked with 'x') * free = not occupied by a rook

```
input:
```

```
N - size of the chess board
rooks - 2D array of integers, rooks[i][0] is a row and rooks[i][1] is a column
  1. board - empty [1..N]x[1..N] array of integers, initialized with 0
  2.
  3. for id from 0 to length(rooks) - 1:
      for row from 1 to N:
  5.
          for column from 1 to N:
  6.
               if row == rooks[id][0] and column == rooks[id][1]:
  7.
                    board[row][column] = 1 // the rook
  8.
               if row == rooks[id][0] and column != rooks[id][1]:
                   board[row][column] = 2 // attacked cell
  9.
 10.
              if row != rooks[id][0] and column == rooks[id][1]:
                  board[row] [column] = 2 // attacked cell
 11.
 12.
 13.
 14. unattacked_cells = N * N // number of unattacked cells
```

15. for row from 1 to N:

```
16. for column from 1 to N:
17.    if board[row][column] == 2:
18.         unattacked_cells -= 1
19.
20. return unattacked_cells
```

In this question you are required to write code in the window below or submit a source file implementing a solution to the problem presented in the previous question (D1). The first input line contains two integers n and m ($1\,$ n $\,1\,0\,0\,$, $1\,$ m $\,1\,0\,0\,$). Each of the m following lines contains two integers r i and c i ($1\,$ r i $\,$ n , $1\,$ c i $\,$ n). Print one integer, the number of cells that are free* and not under attack of any rook.

```
[9]: import numpy as np
import os
os.chdir(r'C:\Users\sfrie\Ydata\ydata_test_2\test2_attempt2')

input = open('d_input.txt').read().strip().split('\n')
input = [x.split(' ') for x in input]
rooks = [list(map(int, x)) for x in input]
print('Input: ', rooks)
n= rooks[0][0]
rooks.pop(0)
print('Cleaned input: ', rooks)
```

Input: [[5, 4], [2, 2], [1, 1], [4, 2], [1, 3]]
Cleaned input: [[2, 2], [1, 1], [4, 2], [1, 3]]

```
[15]: N = n*n
      board = np.reshape(list(np.zeros(N)), [5,5])
      for id in range(0, len(rooks)):
          for row in range(1, n+1):
              for column in range(1, n+1): ##When just leaving it as range(1, n), u
       → this gives a range of 1,2,3. However, we also need 4, in order to cover the
       \rightarrow last rows and columns.
                  if row == rooks[id][0] and column == rooks[id][1]: ## Additonally,
       →each conditional here needs -1 offsets in the change, to convert from the
       \rightarrow input (1, for example) to an index (0)
                      board[row-1] [column-1] = 1 ## the rook
                  if row == rooks[id][0] and column != rooks[id][1]:
                      board[row-1] [column-1] = 2 ## attacked cell
                  if row != rooks[id][0] and column == rooks[id][1]:
                      board[row-1] [column-1] = 2 ## attacked cell
          print('round', id, ': \n', board)
```

```
round 0 : [[0. 2. 0. 0. 0.]
```

```
[2. 1. 2. 2. 2.]
      [0. 2. 0. 0. 0.]
      [0. 2. 0. 0. 0.]
      [0. 2. 0. 0. 0.]]
     round 1:
      [[1. 2. 2. 2. 2.]
      [2. 1. 2. 2. 2.]
      [2. 2. 0. 0. 0.]
      [2. 2. 0. 0. 0.]
      [2. 2. 0. 0. 0.]]
     round 2:
      [[1. 2. 2. 2. 2.]
      [2. 2. 2. 2. 2.]
      [2. 2. 0. 0. 0.]
      [2. 1. 2. 2. 2.]
      [2. 2. 0. 0. 0.]]
     round 3 :
      [[2. 2. 1. 2. 2.]
      [2. 2. 2. 2. 2.]
      [2. 2. 2. 0. 0.]
      [2. 1. 2. 2. 2.]
      [2. 2. 2. 0. 0.]]
[16]: unattacked_cells = N ## number of unattacked cells
      for row in range(0, n):
          for column in range(0, n):
              if board[row] [column] == 2 or board[row] [column] ==1 :
                  unattacked_cells -= 1
      print(unattacked_cells)
```

1.2 D3: Optimize

4

The pseudocode given in D1 has sub-optimal complexity. Optimize your solution for the previous problem (D2) if necessary, allowing it to run efficiently. Your solution will run on tests with larger constraints:

```
rows = set()
cols = set()
for x in rooks:
    cols.add(x[0])
    rows.add(x[1])
missing_rows = (n- len(rows))
missing_cols = (n- len(cols))
sum = missing_rows * missing_cols
print(sum)
```

```
[['5', '4'], ['2', '2'], ['1', '1'], ['4', '2'], ['1', '3']]
[[5, 4], [2, 2], [1, 1], [4, 2], [1, 3]]
4
```

1.3 E1: Shopping, Find Common

In this question (E1-E3) you are asked to assist a store in building a series of tools to understand its customers' behavior and improve future sales. You are provided with a file, purchase_data.csv, containing a dataset of past purchases by customers in the store's loyalty club. Each row in the dataset represents a single past purchase by a club member of a specific product. In this question you are asked to write a code implementing a function find_common, which receives as input the IDs of two products and returns the list of all members who bought both products.

Input format Two integers - product_1, product_2, corresponding to product IDs in the datafile (product_1 product_2).

Output format Print a sequence of integers corresponding to memberIDs of members who bought both products, each number in a single line. Return 0 if no members bought both products.

```
[19]: import pandas as pd
df = pd.read_csv('purchase_data (1).csv')
df.head(5)
```

```
[19]:
                    memberID productID productScore
         timestamp
      0 827500641
                         215
                                      10
                                                      4
      1 827500641
                         215
                                      11
      2 827500641
                                      16
                                                      3
                         215
                                      21
                                                      4
      3 827500641
                         215
                                                      3
      4 827500641
                         215
                                      40
```

```
[21]: file = open('e_input.txt').read().split(' ')
print(file)
prod1, prod2 = [int(x) for x in file]
```

['59', '72']

```
[35]: # def find_common(prod1, prod2):
# prod1_frame = df[df.productID == prod1]['memberID']
```

```
# prod2_frame = df[df.productID == prod2]['memberID']
    # result = set(prod1_frame) & set(prod2_frame)
    # if len(result) == 0:
          return [0]
    # return result
#memberID is int64 type. We need to convert our str file into int.
def find_common(prod1, prod2):
    checked customers = set()
    customers = []
    for id in df['memberID'].unique():
        if id not in customers:
            check frame = df[df['memberID'] == id]
            if prod1 in check_frame['productID'].to_list() and prod2 in_
 →check_frame['productID'].to_list():
                customers.append(id)
            checked_customers.add(id)
    if len(customers) == 0:
        return [0]
    else:
        return sorted(customers)
print(find_common(prod1, prod2))
# for x in find_common(prod1, prod2):
      print(x)
```

[1, 36, 50, 73, 78, 80, 91, 96, 104, 119, 126, 131, 132, 156, 174, 196, 201, 205, 208, 247]

1.4 E2: Shopping, Similarity

In this question, you are asked to implement a measurement of similarity of purchasing history between two products. To do this, you must find the common customers who bought both products using the function you wrote in the previous question, and count how many times each of them bought each of the two products (count[memberID,productID]).

The formula for the measurement is as follows:

```
similarity (product_1, product_2):
    common = find_common(product_1, product_2)
    for member in common:
        sum_mult += (count[member,product_1] - productScore[product_1]) *
            (count[member,product_2] - productScore[product_2])
        sum_sq_1 += (count[member,product_1] - productScore[product_1])^2
        sum_sq_2 += (count[member,product_2] - productScore[product_2])^2
        similarity = sum_mult / (sqrt(sum_sq_1) * sqrt(sum_sq_2))
```

You are requested to write code implementing the measurement described above, so when given a

pair of products (product_1, product_2), it will return the score resulting from the formula.

Note: The data is the same as in question E1. In addition to columns used in E1 (memberID, productID), the formula makes use of productScore: a score given by the store to a specific product based on past sales performance. productScore is constant for all rows with the same productID.

Input format Two integers - product_1, product_2, corresponding to product IDs in the datafile (product_1 product_2).

Output format Print one float value, the score resulting from the running the comparison formula on the two products. Return NA if there are <2 common members or the similarity cannot be computed. Round your output to two decimal digits.

```
[30]: import pandas as pd
      import math
      def similarity(prod1, prod2):
          common = find_common(prod1, prod2)
          if len(common) >=2:
              sum mult =0
              sum sq 1 = 0
              sum_sq_2 = 0
              for member in common:
                   prod1_count = len(df[(df['memberID'] == member) & (df['productID']_
       \rightarrow == prod1)])
                   prod2_count = len(df[(df['memberID'] == member) & (df['productID']_
       \rightarrow == prod2)
                   prod1_score = int(df[df['productID'] == prod1]['productScore'].
       →unique())
                   prod2_score = int(df[df['productID'] == prod2]['productScore'].
       →unique())
                   sum_mult += (prod1_count - prod1_score ) * (prod2_count -__
       →prod2_score)
                   sum_sq_1 += (prod1_count - prod1_score)**2
                   sum sq 2 += (prod2 count - prod2 score)**2
              similarity = round( sum_mult / (math.sqrt(sum_sq_1) * math.
       \rightarrowsqrt(sum_sq_2)), 2)
              if similarity != 0:
                   return similarity
              else:
                  return 'NA'
          else:
              return 'NA'
      print(similarity(prod1, prod2))
      #Return NA if there are <2 common members or the similarity cannot be computed.
      #Round your output to two decimal digits
```

1.4.1 E3:

In this question, you are asked to use your work in the two previous questions to offer relevant products to a new member who has just made their first purchase at the store. Given a list of 3 product IDs and their counts representing the client's current history, you should find and return the ID of the product with the highest predicted purchase count. To do this, you should use the similarity function (from E2) to find the most similar products to the ones in the new client's purchase among all other products, predict how much of each product is the new customer likely to buy in the future using the formula below, and return the one among them with the highest predicted count.

```
predicted_count (new_member, candidate_product):
    for bought_product:
        sum_mult += similarity(candidate_product, bought_product) * count(new_member, bought_product)
        sum_similarities += abs(similarity(candidate_product, bought_product))
        predicted_count = (sum_mult / sum_similarities)
```

Calculate only for products whose similarity with target is not NA If predicted_count cannot be calculated, treat it as 0 If several products tie for highest predicted count, return the one with lowest ID.

Input format Three lines, each containing two integers separated with space - productID and count, corresponding to Product IDs in the datafile, representing products purchased by new member and the count for each product.

Output format Print one integer, corresponding to the product_ID of the product with the highest predicted count.

```
[39]: import pandas as pd
      import math
      df = pd.read_csv('purchase_data (1).csv')
      #In order to solve this question, we need to optimize our code by making our
      dataframe into a less time-consuming dictionaries and lists.
      csv = open("purchase_data (1).csv").read().strip().split('\n')
      members = \Pi
      products = []
      scores = []
      productCount = {}
      productScore = {}
      for line in csv[1:]:
          line = line.split(',')
          memid = int(line[1])
          prodid = int(line[2])
          productScore[prodid] = int(line[3])
```

```
count = productCount.get((memid, prodid), 0)
   productCount[(memid, prodid)] = count+1
   members.append(memid)
   products.append(prodid)
   scores.append(scores)
#Running this new 'Find_common' saves the program 60 seconds on a small input,
→let alone the time it saves on the rest of the code.
def find_common(prod1, prod2):
   customers = []
   for id in set(members):
        if (id, prod1) in productCount.keys() and (id, prod2) in productCount.
→keys():
            customers.append(id)
   if len(customers) == 0:
       return [0]
   else:
        return sorted(customers)
def similarity(prod1, prod2):
    common = find_common(prod1, prod2)
    if len(common) >=2:
       sum_mult =0
       sum_sq_1 = 0
       sum_sq_2 = 0
        for member in sorted(common):
            sum_mult += (productCount[(member, prod1)] - productScore[prod1]) *__
→(productCount[(member, prod2)] - productScore[prod2])
            sum_sq_1 += (productCount[(member, prod1)] - productScore[prod1])**2
            sum_sq_2 += (productCount[(member, prod2)] - productScore[prod2])**2
        similarity = sum_mult / (math.sqrt(sum_sq 1) * math.sqrt(sum_sq 2))
        if similarity != 0:
            return similarity
        else:
           return 'NA'
   else:
       return 'NA'
def input(new_member):
   bought_products = []
   for line in new member:
       line = line.split(' ')
       bought_product, prod_count = [int(x) for x in line]
        bought_products.append(bought_product)
   return bought_products
```

```
def predict_count(bought_products):
    for candidate_product in products:
        sum_mult = 0
        sum_similarities = 0
        for line in new_member:
            line = line.split(' ')
            bought_product, prod_count = [int(x) for x in line]
            similar result = similarity(candidate product, bought product)
            if similar result == 'NA':
                break
            if bought_product != candidate_product:
                sum_mult += similar_result * prod_count
                sum_similarities += abs(similar_result)
        if sum_similarities == 0:
            predicted_count == 0
        else:
            predicted_count = (sum_mult / sum_similarities)
        if predicted_count >0 and candidate_product not in bought_products:
            all_counts[candidate_product] = predicted_count
    return max(all_counts, key = all_counts.get)
new_member = open('e_input.txt').read().strip().split('\n')
products = set(list(df['productID']))
all counts = {}
print(predict_count(input(new_member)))
```

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1.5 F1: Pre-Process

In this question you are provided with a dataset of school statistics and performance on standardised national tests. You are required to perform a series of manipulations and calculations on the dataset. Clean the dataset and correct errors in the following manner (in the listed order):

Remove entries (rows) with any of the following data entry errors:

Student Attendance Rate 0

Rigorous Instruction 0

Students Chronically Absent 100%

Remove entries (rows) with N/A in "Student Achievement Rating" OR "Average Proficiency Score"

Convert "Student Achievement Rating" column from text into numeric values as follows:

Not Meeting Target = 0.0, Approaching Target = 0.5, Meeting Target = 0.8, Exceeding Target = 1.0

Input format schools.csv (link to download it is located in 'Notes' section below) #### Output format Submit the resulting csv file: schools.csv. Do not reorder columns or rows in the file.

```
[52]: import pandas as pd
      import numpy as np
      df = pd.read_csv('schools (1).csv')
      df= df[df['Student Attendance Rate'].str.rstrip('%').astype('int')/100 > 0]
      df = df[df['Rigorous Instruction'].str.rstrip('%').astype('int')/100 > 0]
      df = df[df['Students Chronically Absent'].str.rstrip('%').astype('int')/100 < 1]</pre>
      df = df.dropna(subset = ['Student Achievement Rating', 'Average Proficiency⊔

¬Score'])
      def change(n):
          if n == 'Exceeding Target': return 1.0
          if n == 'Meeting Target': return 0.8
          if n == 'Approaching Target': return 0.5
          if n == 'Not Meeting Target': return 0.0
      df['Student Achievement Rating'] = df['Student Achievement Rating'].
       →apply(change)
      # df.loc[df['Student Achievement Rating'] == 'Exceeding Target', ['Student⊔
      → Achievement Rating']] = 1
      # df.loc[df['Student Achievement Rating'] == 'Meeting Target', ['Student_1]
      → Achievement Rating']] = 0.8
      # df.loc[df['Student Achievement Rating'] == 'Approaching Target', ['Student⊔
      → Achievement Rating']] = 0.5
      # df.loc[df['Student Achievement Rating'] == 'Not Meeting Target', ['Student_
      → Achievement Rating']] = 0.0
      result = df.to_csv('result.csv', index = False)
      df.head(4)
[52]:
         School Index District Lowest Year Highest Year School Income Estimate \
      0
               310001
                              5
                                                                             NaN
                              2
                                                        5
                                                                     $26,114.78
      1
               310002
                                         PΚ
               310003
                              2
                                         PΚ
                                                        5
                                                                    $103,399.19
      3
               310004
                                         PΚ
                                                        5
                                                                     $34,684.21
        Average Proficiency Score Student Attendance Rate
                                                       94%
      0
                              65%
      1
                              69%
                                                       98%
      2
                              77%
                                                       95%
      3
                              56%
                                                       92%
```

Students Chronically Absent Rigorous Instruction \

```
19%
                                                   97%
0
                             6%
                                                   93%
1
2
                            13%
                                                   97%
3
                            33%
                                                   91%
   Year 3 - All Students Tested
                                   Year 4 - All Students Tested
0
                                0
1
                              125
                                                               106
2
                                                               142
                              105
3
                              103
                                                               100
   Year 5 - All Students Tested
                                    Year 6 - All Students Tested
0
1
                              104
                                                                 0
2
                                                                 0
                               61
3
                               81
                                                                 0
                          Year 4 - Top Scoring Year 5 - Top Scoring
   Year 3 - Top Scoring
0
                       8
                                               13
                                                                       11
1
2
                       9
                                               44
                                                                       18
3
                        3
                                               7
                                                                        1
   Year 6 - Top Scoring
                          Student Achievement Rating
0
                                                    1.0
1
                        0
                                                    0.8
2
                        0
                                                    0.8
3
                                                    0.5
```

1.6 F2: Best Year

Write code solving the following problem: given a school index, find and return the grade year with the highest ratio of students who achieved the top score out of the total tested. Exclude grade years, for which performance data is not available (i.e. years where Year X All Students Tested = 0).

Return the numeric value corresponding to the grade Year. In case of a tie, return the lowest grade Year applicable. If no applicable year groups exists, return 0.

Input format One integer corresponding to School Index. #### Output format Print one integer corresponding to the grade year with maximum calculated ratio.

```
[53]: ratio3 = df['Year 3 - All Students Tested']/ df['Year 3 - Top Scoring']
ratio4 = df['Year 4 - All Students Tested']/ df['Year 4 - Top Scoring']
ratio5 = df['Year 5 - All Students Tested']/ df['Year 5 - Top Scoring']
ratio6 = df['Year 6 - All Students Tested']/ df['Year 6 - Top Scoring']
```

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1.7 F3: District Score

Write code which calculates performance by district based on the criteria provided below. Given a district code and Average Proficiency Score range (min and max), find all the schools corresponding to the criteria, and return the average Student Achievement Rating for the schools selected. For Average Proficiency Score you are given a minimum and a maximum value. Find all schools with min Average Proficiency Score (school) max. Return the average value for Student Achievement Rating.

Round your answer to two decimal digits

If no relevant schools are available, return 0.

Input format A list of 3 numbers separated by spaces: district, minimum threshold, maximum threshold. #### Output format Print out a single float value, rounded to two decimal digits corresponding to the average Student Achievement Rating among schools fulfilling the criteria described above.

```
df.head()
print(round(df['Student Achievement Rating'].mean(), 2))
# def find_best(i):
      bests = 0
      result = 0
      valid = []
      six = df.loc[i]['Year 6 - Top Scoring']/df.loc[i]['Year 6 - All_{\square}]
→Students Tested']
      if df.loc[i]['Year 6 - All Students Tested'] >= 0:
          if six >= bests:
#
              bests = six
#
              result = 6
#
              valid.append(6)
      five = df.loc[i]['Year 5 - Top Scoring']/ df.loc[i]['Year 5 - All_{\sqcup}]
→Students Tested'
      if df.loc[i]['Year 5 - All Students Tested'] >= 0:
#
          if five >= bests:
              bests = five
#
              result = 5
#
              valid.append(5)
      four = df.loc[i]['Year 4 - Top Scoring']/ df.loc[i]['Year 4 - All
→Students Tested']
      if df.loc[i]['Year 4 - All Students Tested'] >= 0:
#
          if four >= bests:
              bests = four
#
#
              result = 4
#
              valid.append(4)
      three = df.loc[i]['Year 3 - Top Scoring']/df.loc[i, 'Year 3 - All_u']
→Students Tested']
#
      if df.loc[i]['Year 3 - All Students Tested'] != 0:
          if three >= bests:
#
#
              bests = three
#
              result = 3
#
              valid.append(3)
      if len(valid) == 0:
          return 0
      return result
# file = open('f2_input.txt').read().strip()
\# i = int(file)
# df = df.set_index('School Index')
# print(find_best(i))
```

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