2.1. PROBLEM STATEMENT: NUMPY

Problem Statement 1:

Write a function so that the columns of the output matrix are powers of the input vector.

The order of the powers is determined by the increasing boolean argument. Specifically, when increasing is False, the i-th output column is the input vector raised element-wise power of N-i-1.

HINT: Such a matrix with a geometric progression in each row is named for Alexandre Theophile Vandermonde

```
In [1]:
```

```
vander_matrix(x)
Out[2]:
array([[ 1,
                 1,
                     1],
            1,
      [ 8,
            4,
                 2,
                      1],
                3,
      [ 27,
            9,
                      1],
      [ 64,
            16,
                  4,
                      1],
      [125, 25,
                 5,
                     1]], dtype=int32)
```

Problem Statement 2:

Write a function to find moving average in an array over a window:

Test it over [3,5,7,2,8,10,11,65,72,81,99,100,150] and window of 3

```
In [3]:
```

```
import numpy as np

seq=[3,5,7,2,8,10,11,65,72,81,99,100,150] # here seq is the given input list of elements
w=3

def moving_average(seq,w):
    n=len(seq) # n is the length of the list

for i in range(n-w+1): #iterating throught the list to find the average for a window of 3

till n-w+1,ie., 11 values should be displayed

    x=sum(seq[i:i+w]) #find the sum of the elements in seq for index equal to window,ie., dur
ing first iteration, 3 elements from the list will be taken--> (3+5+7)will be value of x

mavg=x/w #finding moving average by dividing sum by window.With respect to the ak
```

```
ove step, for 1st iteration it becomes (3+5+7)/3=15/3=5.0 which is the first output
        print(mavg)
4
In [4]:
moving average(seq,w)
5.0
4.66666666666667
5.66666666666667
6.66666666666667
9.66666666666666
28.6666666666668
49.333333333333336
72.66666666666667
84.0
93.3333333333333
116.33333333333333
```

2.2. PROBLEM STATEMENT: PANDAS

Problem Statement 1:

Qn.1. How-to-count-the-distance-to-the-previous-zero

For each value, count the difference of the distance from the previous zero(or the start of the Series, whichever is closer) and if there are no previous zeros, print the position

Consider a DataFrame df where there is an integer column{'X':[7,2,0,3,4,2,5,0,3,4]}

The values should therefore be [1,2,0,1,2,3,4,0,1,2]. Make this a new column 'Y'/

```
In [5]:
import numpy as np
import pandas as pd
In [6]:
df=pd.DataFrame({'X':[7,2,0,3,4,2,5,0,3,4]})
In [7]:
df
Out[7]:
   Х
1 2
2 0
3 3
```

```
5 2
X
6 5
7 0
8 3
9 4
In [8]:
#Counting the distance to the previous zero
X=[7,2,0,3,4,2,5,0,3,4]
count=1
Y=[]
for items in X:
    if items==0:
       count=0
   Y.append(count)
   count+=1
print('Y = ',Y)
Y = [1, 2, 0, 1, 2, 3, 4, 0, 1, 2]
In [9]:
df=pd.DataFrame({'X':[7,2,0,3,4,2,5,0,3,4],'Y':[1, 2, 0, 1, 2, 3, 4, 0, 1, 2]})
In [10]:
df
Out[10]:
  X Y
1 2 2
2 0 0
3 3 1
4 4 2
5 2 3
6 5 4
7 0 0
8 3 1
9 4 2
Qn.2. Create a Datetimeindex that contains each business day of 2015 and use it to index a
Series of random numbers
In [11]:
DtIndex = pd.date_range('2015-01-01', '2015-12-31',freq='B') #freq='B' stands for 'Business day'
starting from 1st Jan to 31st Dec of 2015
In [12]:
DtIndex
Out[12]:
DatetimeIndex(['2015-01-01', '2015-01-02', '2015-01-05', '2015-01-06',
```

```
'2015-01-07', '2015-01-08', '2015-01-09', '2015-01-12', '2015-01-13', '2015-01-14', ...
'2015-12-18', '2015-12-21', '2015-12-22', '2015-12-23', '2015-12-24', '2015-12-25', '2015-12-28', '2015-12-29', '2015-12-30', '2015-12-31'], dtype='datetime64[ns]', length=261, freq='B')
```

In [13]:

```
#making s as the series of random numbers and using my DtIndex to index s
```

In [14]:

```
s=pd.Series(np.random.rand(261),index=DtIndex)
#created a DatetimeIndex that contains each businessday of 2015 and that has been used
to index to a series of random numbers
```

In [15]:

```
s
```

Out[15]:

```
2015-01-01
            0.180522
2015-01-02
            0.384560
2015-01-05
            0.837433
2015-01-06
             0.404452
2015-01-07
             0.581959
2015-01-08
             0.483476
2015-01-09
             0.128267
2015-01-12
            0.965145
2015-01-13
             0.203820
2015-01-14
             0.201965
2015-01-15
             0.238237
2015-01-16
             0.483211
2015-01-19
             0.489988
2015-01-20
            0.604303
2015-01-21
             0.743370
2015-01-22
             0.747126
2015-01-23
             0.855448
2015-01-26
             0.135125
2015-01-27
            0.967107
2015-01-28
             0.359201
2015-01-29
             0.332033
2015-01-30
             0.339583
2015-02-02
             0.256298
2015-02-03
            0.974069
2015-02-04
             0.127657
2015-02-05
             0.445769
2015-02-06
             0.682649
2015-02-09
             0.718425
2015-02-10
            0.894220
2015-02-11
           0.997396
               . . .
2015-11-20
             0.349118
2015-11-23
             0.219259
2015-11-24
             0.484825
2015-11-25
            0.183250
2015-11-26
             0.580114
2015-11-27
             0.219244
2015-11-30
             0.940591
2015-12-01
             0.058943
2015-12-02
             0.030501
2015-12-03
            0.654721
2015-12-04
             0.168405
2015-12-07
             0.904779
2015-12-08
             0.387871
2015-12-09
             0.917678
2015-12-10
            0.434431
2015-12-11
            0.857675
2015-12-14
             0.094920
2015-12-15
             0.081586
```

2015-12-16

N 889119

```
0.406439
~UIU IC IU
2015-12-17
2015-12-18 0.601203
2015-12-21 0.901803
2015-12-22 0.496089
2015-12-23
              0.549698
2015-12-24
              0.087471
2015-12-25
            0.365286
2015-12-28 0.046043
            0.562544
2015-12-29
2015-12-30 0.427023
2015-12-31 0.129502
Freq: B, Length: 261, dtype: float64
In [16]:
BD = pd.DataFrame(data = s, index= DtIndex)
#Created a dataframe with DtIndex as 'index' and 's' as attribute
In [17]:
BD.columns=['s'] #naming column as 's'
In [18]:
BD.head()
Out[18]:
2015-01-01 0.180522
2015-01-02 0.384560
2015-01-05 0.837433
2015-01-06 0.404452
2015-01-07 0.581959
In [19]:
BD.shape
Out[19]:
(261, 1)
Qn.3 Find the sum of the values in s for every Wednesday
In [20]:
#I have created the dataframe BD with attribute 's'. So taking sum of all wednesdays in 2015 from
BD
In [21]:
BD.head()
Out[21]:
2015-01-01 0.180522
2015-01-02 0.384560
2015-01-05 0.837433
2015-01-06 0 404452
```

```
401J-01-00 0.404404
2015-01-07 0.581959
In [22]:
BD['Day']=DtIndex.day_name() #creating a column for showing the corresponding days for the date
In [23]:
BD.head(7)
Out[23]:
              s
                        Day
2015-01-01 0.180522
                    Thursday
 2015-01-02 0.384560
                      Friday
 2015-01-05 0.837433
                     Monday
 2015-01-06 0.404452
                     Tuesday
2015-01-07 0.581959 Wednesday
 2015-01-08 0.483476
                    Thursday
2015-01-09 0.128267
                      Friday
In [24]:
BD1=BD.copy()
In [25]:
BD1=BD1.groupby(['Day']).sum()
In [26]:
BD1
Out[26]:
     Day
  Friday 24.784556
   Monday 28.025872
  Thursday 25.100930
   Tuesday 26.532080
Wednesday 25.892034
In [27]:
BD1.iloc[4] #gives the sum of all wednesdays
Out[27]:
s 25.892034
Name: Wednesday, dtype: float64
In [28]:
\# Another\ method\ for\ finding\ the\ sum\ of\ s\ for\ all\ Wednesday
In [29]:
```

```
BD2=BD.copy() #copying the data to another dataframe
In [30]:
sum_of_s=BD2[BD2.Day=="Wednesday"].s.sum()
In [31]:
sum_of_s
Out[31]:
25.892033950143148
Qn.4.Average for each calendar month
```

```
In [32]:
```

```
BD.resample('M').mean() #finding the average for each month in 2015
```

Out[32]:

 2015-01-31
 0.484833

 2015-02-28
 0.529426

 2015-03-31
 0.487405

 2015-04-30
 0.597809

 2015-05-31
 0.457258

 2015-06-30
 0.575066

 2015-07-31
 0.464190

 2015-08-31
 0.505761

 2015-10-31
 0.472586

 2015-11-30
 0.472767

 2015-12-31
 0.437119

Qn.5. For each group of four consecutive calendar months in s, find the date on which the highest value occurred

```
In [33]:
```

```
BD.head()
```

Out[33]:

	s	Day
2015-01-01	0.180522	Thursday
2015-01-02	0.384560	Friday
2015-01-05	0.837433	Monday
2015-01-06	0.404452	Tuesday
2015-01-07	0.581959	Wednesday

In [34]:

 ${\tt BD.groupby(pd.Grouper(freq='M')).max()[:4]} \ \# finding \ date \ on \ which \ s \ was \ highest \ for \ 4 \ consecutive \ months$

```
Out[34]:
                          Day
2015-01-31 0.967107 Wednesday
 2015-02-28 0.997396 Wednesday
 2015-03-31 0.911611 Wednesday
 2015-04-30 0.983615 Wednesday
```

Problem Statement 2:

Read the dataset from the below link

https://raw.githubusercontent.com/guipsamora/pandas_exercises/master/06_Stats/US_Baby_Names/US_Baby_Names_right.csv

```
In [1]:
```

```
import numpy as np
import pandas as pd
In [2]:
```

```
df=pd.read csv('https://raw.githubusercontent.com/guipsamora/pandas exercises/master/06 Stats/US Ba
Names/US_Baby_Names_right.csv')
4
```

```
In [3]:
```

```
df.shape #to know the number of rows and columns for the data in the dataframe
Out[3]:
(1016395, 7)
```

In [4]:

```
df.head() #Gives the first 5 rows in the dataframe. Here df is my dataframe
```

Out[4]:

	Unnamed: 0	ld	Name	Year	Gender	State	Count
0	11349	11350	Emma	2004	F	AK	62
1	11350	11351	Madison	2004	F	AK	48
2	11351	11352	Hannah	2004	F	AK	46
3	11352	11353	Grace	2004	F	AK	44
4	11353	11354	Emily	2004	F	AK	41

Questions Answered:

Qn.1. Delete unnamed columns

```
In [5]:
```

```
df.columns
                            #To know the unnamed column by knowing the column names
Out[5]:
Index(['Unnamed: 0', 'Id', 'Name', 'Year', 'Gender', 'State', 'Count'], dtype='object')
In [6]:
```

```
dfl=df.copy() #Since the original dataframe should not be changed copying the df to
another dataframe named dfl
In [7]:
df1.drop("Unnamed: 0",axis=1,inplace=True) #Dropping the unnamed column using drop method.
In [8]:
dfl.head() #Printing the first 5 rows to check if unnamed has been removed
Out[8]:
     ld
         Name Year Gender State Count
0 11350 Emma 2004
                             ΑK
1 11351 Madison 2004
                            ΑK
                                   48
2 11352 Hannah 2004
                             ΑK
                                   46
3 11353
                             ΑK
          Grace 2004
                                   44
4 11354
          Emily 2004
                             ΑK
                                   41
Qn.2. Show the distribution of male and female
In [9]:
both genders=df1['Gender'].value counts()
In [10]:
both genders
Out[10]:
F
   558846
   457549
Name: Gender, dtype: int64
In [11]:
genders=df1.groupby(['Gender'])
In [12]:
genders.describe()
Out[12]:
       Count
                                                                                                 Year
                                                                           ... 75%
                       std
       count mean
                                 min 25% 50% 75% max
                                                         count
                                                                 mean
                                                                                        max
                                                                                                 count
Gender
    F 558846.0 29.310925 75.962992 5.0 6.0 10.0 23.0 3634.0 558846.0 2.793178e+06 ... 4333051.75 5634124.0 558846.0
    M 457549.0 41.615650 118.074308 5.0 7.0 12.0 29.0 4167.0 457549.0 2.877176e+06 ... 4241792.00 5647426.0 457549.0
2 rows × 24 columns
                                                                                                       Þ
Qn.3 Show the top 5 most preferred names
In [13]:
preferred name=df1['Name'].value counts()
```

```
preferred name.head()
                                         #knowing the 5 most preferred name by knowing the val
lue counts for the 'Name 'column in dfl dataframe
4
                                                                                                      •
Out[13]:
Riley
          1112
         1080
Avery
       1073
Jordan
Peyton 1064
Hayden 1049
Name: Name, dtype: int64
Qn.4. What is the median name occurence in the dataset
In [14]:
df1.head()
Out[14]:
     ld
          Name Year Gender State Count
0 11350
         Emma 2004
                         F
                             AK
                                    62
1 11351 Madison 2004
                             ΑK
                                    48
                             ΑK
                                    46
2 11352
         Hannah 2004
3 11353
          Grace 2004
                             \mathsf{AK}
                                    44
4 11354
          Emily 2004
                             \mathsf{AK}
                                    41
In [15]:
df1.Id.median()
Out[15]:
2811921.0
In [16]:
median_name=df1[df1['Id']==df1.Id.median()]
In [17]:
median_name
Out[17]:
           ld Name Year Gender State Count
508197 2811921 Kasey 2010
Qn.5.Distribution of male and female born count by states
In [18]:
distribution=df1.groupby(['Gender','State']).count()
#Showing the distribution of male and female with respect to each State
In [19]:
distribution.loc[: , 'Count']
Out[19]:
Gender State
```

```
AL
                  9878
        AR
                  7171
        ΑZ
                 14518
        CA
                 45144
                 11424
        CO
        СТ
                 6575
        DC
                 3053
        DE
                  2549
        FL
                 25781
        GΑ
                 19385
                  3255
        ΗI
                  7131
        ΙA
                  4918
        ID
        ΙL
                 21268
        IN
                 13056
                  7753
        KS
        ΚY
                 8817
        LA
                 10510
                 10580
        MA
        MD
                 11276
        ME
                  2976
                 16038
        MI
        MN
                 10677
        MO
                 11948
        MS
                  7235
        MT
                  2690
       NC
                 17357
        ND
                 2399
        NE
                  5370
        ME
                  2777
        ΜI
                 13243
        MN
                  9004
        MO
                  9917
        MS
                  6862
        МТ
                  2986
        NC
                 13530
        ND
                  2581
                  5029
        NE
        NH
                  2659
                 12274
        NJ
        NM
                  4966
                  6024
        NV
                 22585
        NY
        ОН
                 14318
        OK
                  8138
        OR
                  7333
        PΑ
                 14171
        RI
                  2468
        SC
                  8195
        SD
                  2908
        TN
                 10588
        TX
                 27791
        UT
                  8233
                 11997
        VA
        VT
                  1618
        WA
                 11049
        WI
                  8940
        WV
                  3733
        WY
                  1904
Name: Count, Length: 102, dtype: int64
```

In [20]:

F

AK

2404

pd.crosstab(index=df1.Gender,columns=df1.State) #distribution in a more formatted way

Out[20]:

State Gender	AK	AL	AR	AZ	CA	со	СТ	DC	DE	FL	 SD	TN	TX	UT	VA	VT	WA	WI
	2404	0878	7171	14518	15111	11/12/	6575	3053	25/10	25781	2838	13063	30760	0515	1/1750	1308	13320	10549
			6475							20070		10588			11997		11049	8940

Value at index 3

2.3. PROBLEM STATEMENT: USE CASES OF NUMPY AND PANDAS

Qn.1.Write a Python program which accepts a list named: randomList=['a',0,2]Use exception handling using try-catch

```
In [1]:
import sys
randomList=['a',0,2]
for item in randomList:
        print("The entry is",item)
        r=1/it.em
    except:
        print("Oops!", sys.exc info()[0], "occurred")
        print("Next entry")
        print()
print("The reciprocal of",item,"is",r)
The entry is a
Oops! <class 'TypeError'> occurred
Next entry
The entry is 0
Oops! <class 'ZeroDivisionError'> occurred
Next entry
The entry is 2
The reciprocal of 2 is 0.5
```

Qn.2.Write a Python program to give exception "Array Out of Bound" if the user wants to access the elements beyond the list size(use try and except)

```
import sys
l=[1,2,3,4]
i=0
for i in 1:
    try:
        print("Value at index",i)
        print("is",1[i])

except:
        print("\nYour list does not contain anymore elements")
        print("Aray out of bound ",sys.exc_info()[0])
Value at index 1
is 2
Value at index 2
```

```
is 4
Value at index 4
Your list does not contain anymore elements
Aray out of bound <class 'IndexError'>
```

Qn.3 Write a python module script that contains fib2() method to calculate the Fibonacci series till 1000 and save it as fibo.py

In [3]:

```
def fib2(n):
    a=0
    b=1
    print(a)
    while b<n:
        print(b)
        a, b=b,a+b</pre>
```

In [4]:

```
#To run this program as a module, I need to import the module as
#import fibo
#fibo.fib2(1000)
fib2(1000)
0
1
1
2
3
5
8
13
21
34
55
89
144
233
377
610
987
```

Qn.4.Write a Python module script that contains is palindrome() method to calculate the input string as palindrome string or not and save it as palindrome.py

```
In [5]:
```

```
def ispalindrome(s):
    s=str(s)
    s=s.lower()
    rev=reversed(s)
    if list(s) == list(rev):
        return True
    return False
```

In [6]:

```
Out[6]:
True

In [7]:
    ispalindrome(121)

Out[7]:
    True

In [8]:
    ispalindrome('any')
Out[8]:
False
```

Qn.5.Write a program in Python with one class called Cipher.

In [13]:

```
import numpy as np
 class Cipher:
             L2I = dict(zip("ABCDEFGHIJKLMNOPQRSTUVWXYZabcdefghijklmnopqrstuvwxyz0123456789",range(62)))
              I2L = dict(zip(range(62), "ABCDEFGHIJKLMNOPQRSTUVWXYZabcdefghijklmnopqrstuvwxyz0123456789"))
              def __init_ (self,Instr=""):
                          self.Instr=str(input("Enter the input string "))
              def encrypt(self,key):
                          ciphertext = ""
                          Instr=self.Instr
                          for c in Instr:
                                       if c.isalnum():
                                                    ciphertext += self.I2L[ (self.L2I[c] + key) ]
                                       else:
                                                    ciphertext += c
                          return ciphertext
              def decrypt(self,Enstr,key):
                       plaintext2 = ""
                          for c in Enstr:
                                       if c.isalnum(): plaintext2 += self.I2L[ (self.L2I[c] - key)]
                                       else: plaintext2 += c
                          return plaintext2
 k=np.random.randint(1,50,1)
 key=k[0]
 c=Cipher()
encryptstr=c.encrypt(key)
decryptstr=c.decrypt(encryptstr,key)
print("\n Input String is :\t"+c.Instr)
print("\nEncryption value of given string is :\t"+encryptstr)
print("\ndering) where the constant of the c
Enter the input string acadgild
```

```
Input String is : acadgild

Encryption value of given string is : oqoruwzr

Decrypted value is: acadgild
```

Qn.6. Get Data from the following link:

```
In [1]:
```

```
import numpy as np
import pandas as pd
```

Qn1.Read the dataset using pandas

```
In [2]:
```

```
import os
import pandas as pd
os.chdir("c:\data") #since the data is stored in my system importing it to the jupyter notebook
```

In [3]:

```
movies=pd.read_csv("movies.csv")
```

In [4]:

```
tags= pd.read_csv("tags.csv")
```

In [5]:

```
ratings=pd.read_csv("ratings.csv")
```

In [6]:

```
movies.head()
```

Out[6]:

	movield	title	genres
0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy
1	2	Jumanji (1995)	Adventure Children Fantasy
2	3	Grumpier Old Men (1995)	Comedy Romance
3	4	Waiting to Exhale (1995)	Comedy Drama Romance
4	5	Father of the Bride Part II (1995)	Comedy

In [7]:

```
tags.head()
```

Out[7]:

	userld	movield	tag	timestamp
0	18	4141	Mark Waters	1240597180
1	65	208	dark hero	1368150078
2	65	353	dark hero	1368150079
3	65	521	noir thriller	1368149983
4	65	592	dark hero	1368150078

In [8]:

```
ratings.head()
```

Out[8]:

0	userlø	movield	ratiṇg	111724818021
1	1	29	3.5	1112484676
2	1	32	3.5	1112484819
3	1	47	3.5	1112484727
4	1	50	3.5	1112484580

Qn.2.Extract first row from tags and print its type

```
In [9]:
```

```
tagsfirstrow=tags.iloc[0]
```

In [10]:

```
tagsfirstrow
```

Out[10]:

userId 18
movieId 4141
tag Mark Waters
timestamp 1240597180
Name: 0, dtype: object

In [11]:

```
print(type(tagsfirstrow))
```

<class 'pandas.core.series.Series'>

Qn.3.Extract row 0,11,2000 from tags dataframe

In [12]:

```
tagsrows= tags.iloc[[0,11,2000]]
```

In [13]:

```
tagsrows #displaying tag rows
```

Out[13]:

tag timestam	tag	movield	userld	
ters 124059718	Mark Waters	4141	18	0
riller 136814998	noir thriller	65 1783 910 68554		11
racy eory 136804394	conspiracy theory			2000

Qn.4. Print index, columns of the dataframe

In [14]:

```
ratings.columns
```

Out[14]:

```
Index(['userId', 'movieId', 'rating', 'timestamp'], dtype='object')
```

In [15]:

```
movies.columns
Out[15]:
Index(['movieId', 'title', 'genres'], dtype='object')
In [16]:
tags.columns
Out[16]:
Index(['userId', 'movieId', 'tag', 'timestamp'], dtype='object')
Qn.5.Calculate the descriptive statistics for the 'rating' column of the ratings dataframe.Verify using describe
In [17]:
ratings.head()
Out[17]:
   userld movield rating timestamp
0
                  3.5 1112486027
                  3.5 1112484676
1
      1
             29
2
             32
                3.5 1112484819
3
      1
             47
                 3.5 1112484727
             50
                3.5 1112484580
In [18]:
print("The descriptive statistics for the rating column is:")
print("count= ", ratings["rating"].count())
print("mean= ",ratings["rating"].mean())
print("standard deviation=", ratings["rating"].std())
print("min", ratings["rating"].min())
print("median= ",ratings["rating"].median())
print("max= ",ratings["rating"].max())
The descriptive statistics for the rating column is:
count= 20000263
mean= 3.5255285642993797
standard deviation= 1.051988919275684
min 0.5
median= 3.5
max = 5.0
In [19]:
\verb"ratings"]. describe() \textit{\#verifying using desscribe}
Out[19]:
count 2.000026e+07
mean
        3.525529e+00
         1.051989e+00
std
         5.000000e-01
        3.000000e+00
2.5%
50%
        3.500000e+00
75%
        4.000000e+00
        5.000000e+00
max
Name: rating, dtype: float64
```

Qn.6. Filter out ratings with rating >5

```
In [20]:
r=ratings[ratings["rating"]>5]
In [21]:
print(r)
Empty DataFrame
Columns: [userId, movieId, rating, timestamp]
Index: []
Qn.7.Find how many null values,missing values are present.Deal with them.Print out how many rows have been
modified
In [22]:
ratings.isnull().sum() #finding null values in ratings column
Out[22]:
            0
userId
movieId
             0
             Ω
rating
timestamp
             0
dtype: int64
In [23]:
movies.isnull().sum() #finding null values in movies column
Out[23]:
movieId
         0
title
genres
          0
dtype: int64
In [24]:
tags.isnull().sum()#finding null values in tags column
Out[24]:
userId
             0
movieId
              0
tag
timestamp
dtype: int64
In [25]:
\#Only\ tags\ data\ has\ null\ values\ for'tag'\ and\ there\ are\ 16\ null\ values
In [26]:
tags.fillna(method = "ffill", inplace=True) # using forward fill to fill in the missing null values
in tags
In [27]:
tags.isnull().sum() #thus null values removed using forward fill
Out[27]:
userId
```

morriald

tag 0
timestamp 0
dtype: int64

Qn.8.Filter out movies from the movies dataframe that are of type'Animation'

In [28]:

animatedmovies=movies[movies.genres =="Animation"]

In [29]:

animatedmovies

Out[29]:

	movield	title	genres	
2503	2588	Cloudland (1998)	Animation	
4906	5002	Fritz the Cat (1972)	Animation	
4907	5003	Nine Lives of Fritz the Cat, The (1974)	Animation	
9455	27738	Cathedral, The (Katedra) (2002)	Animation	
9989	32840	Vincent (1982)	Animation	
13444	66335	Afro Samurai: Resurrection (2009)	Animation	
13858	69469	Garfield's Pet Force (2009)	Animation	
14184	71158	Immigrants (L.A. Dolce Vita) (2008)	Animation	
14492	72603	Merry Madagascar (2009)	Animation	
14578	72927	Donkey Xote (2007)	Animation	
14580	72931	Tango (1981)	Animation	
14876	74486	\$9.99 (2008)	Animation	
14938	74791	Town Called Panic, A (Panique au village) (2009)	Animation	
15085	76709	Spider-Man: The Ultimate Villain Showdown (2002)	Animation	
15860	80469	Superman/Batman: Apocalypse (2010)	Animation	
15990	81018	Illusionist, The (L'illusionniste) (2010)	Animation	
16551	83603	Fern flowers (Fleur de fougère) (1949)	Animation	
16649	84192	Corto Maltese: Ballad of the Salt Sea (Corto M	Animation	
18144	90843	B Lavatory Lovestory (Ubornaya istoriya - lyubov		
18145	90845	Fall of the House of Usher, The (Zánik domu Us		
18224	91187	Millhaven (2010)	Animation	
18695	93083	Live Music (2009)	Animation	
18980	94423	Disney Princess Collection: Jasmine's Enchante	Animation	
19562	96897	Bleach: Fade to Black (Burīchi Fade to Bl	Animation	
20310	99820	Pokémon the Movie: Black - Victini and Reshira	Animation	
20311	99822	Pokémon the Movie: White - Victini and Zekrom	Animation	
20314	99832	Hand, The (Ruka) (1966)	Animation	
20527	100509	Tale of Tales (Skazka skazok) (1979)	Animation	
20871	102007	Invincible Iron Man, The (2007)	Animation	
20872	102009	Thor: Tales of Asgard (2011)	Animation	
25207	118948	Blackbird (1959)	Animation	
25208	118950	A Phantasy (1952)	Animation	
25211	118956	Lines: Horizontal (1962)	Animation	
25212	118958	Mosaic (1966)	Animation	
25213	118960	Begone Dull Care (1949)	Animation	

25214	mqvield	Synchromy (1 9^{††}	An genties
25538	120853	Fresh Guacamole (2012)	Animation
25539	120855	Adam and Dog (2012)	Animation
25663	121302	Someone's Gaze (2013)	Animation
25718	121600	Bosko's Parlor Pranks (1934)	Animation
26052	124889	The Adventures of Tom Thumb & Thumbelina (2002)	Animation
26163	125924	The Periwig-Maker (1999)	Animation
26169	125936	Crac (1981)	Animation
26171	125940	Syrinx (1966)	Animation
26181	125960	The Trip to Squash Land (1967)	Animation
26185	125968	It's Christmastime Again, Charlie Brown (1992)	Animation
26189	125976	Nocturna Artificialia (1979)	Animation
26192	125982	Trick or Treat (1952)	Animation
26207	126012	The Fat Albert Halloween Special (1977)	Animation
26226	126050	Stille Nacht I: Dramolet (1988)	Animation
26230	126058	Rehearsals for Extinct Anatomies (1987)	Animation
26236	126070	Chainsaw Maid (2007)	Animation
26237	126072	The Little Matchgirl (2006)	Animation
26246	126090	Hedgehog in the Fog (1975)	Animation
26247	126092	The Cat's Out (1931)	Animation
26248	126094	Claymation Comedy of Horrors (1991)	Animation
26319	126405	The Adventures of André and Wally B. (1984)	Animation
26809	128864	Four Sahibjade (2014)	Animation
27103	130394	The Mascot (1934)	Animation
27155	130644	The Garden of Sinners - Chapter 5: Paradox Par	Animation

83 rows × 3 columns

Qn.9.Find the average ratings of movies

In [30]:

```
ratings.rating.mean()
```

Out[30]:

3.5255285642993797

Qn.10.Perform an inner join of movies and tags based on movield

In [31]:

```
result=pd.merge(movies, tags, how = "inner", on = "movieId")
```

In [32]:

```
result.head()
```

Out[32]:

	movield	title	genres	userld	tag	timestamp
0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1644	Watched	1417736680
1	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	computer animation	1183903155
2	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	Disney animated feature	1183933307

```
3 movield Toy Story (1996) Adventure|Animation|Children|Comedy|Fgentees useAdd Pixar animating 116339546776

4 1 Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy 1741 Tĩa Leoni does not star in this movie 1245093573
```

Qn.11.Print out the 5 movies that belong to the Comedy genre and having rating greater than 4

```
In [33]:
```

```
\label{eq:dfl}  dfl=pd.merge \mbox{(movies, ratings, how = "inner", on = "movieId")}  \#performing inner join based on movieid and creating a dataframe dfl
```

In [34]:

```
c1=df1[(df1["genres"] =="Comedy") & (df1["rating"] >4)] # applying condition for df1
```

In [35]:

```
c1.head()
```

Out[35]:

	movield	title	genres	userld	rating	timestamp
87435	5	Father of the Bride Part II (1995)	Comedy	117	5.0	861553146
87437	5	Father of the Bride Part II (1995)	Comedy	127	5.0	847127740
87455	5	Father of the Bride Part II (1995)	Comedy	350	5.0	1360209812
87460	5	Father of the Bride Part II (1995)	Comedy	390	5.0	836139583
87462	5	Father of the Bride Part II (1995)	Comedy	401	5.0	847049988

Qn.12.Split 'genres' into multiple columns

```
In [36]:
```

```
new = movies["genres"].str.split("|", expand = True)
```

In [37]:

```
new.head()
```

Out[37]:

	0	1	2	3	4	5	6	7	8	9
0	Adventure	Animation	Children	Comedy	Fantasy	None	None	None	None	None
1	Adventure	Children	Fantasy	None	None	None	None	None	None	None
2	Comedy	Romance	None	None	None	None	None	None	None	None
3	Comedy	Drama	Romance	None	None	None	None	None	None	None
4	Comedy	None	None	None	None	None	None	None	None	None

Qn.13. Extract year from title, e.g.(1995)

```
In [38]:
```

```
movies['year'] = movies['title']
movies['year'] = movies.year.str[-6:]
```

```
In [39]:
```

```
movies.head()
```

Out[39]:

	movield	title	genres	year
0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	(1995)
1	2	Jumanji (1995)	Adventure Children Fantasy	(1995)
2	3	Grumpier Old Men (1995)	Comedy Romance	(1995)
3	4	Waiting to Exhale (1995)	Comedy Drama Romance	(1995)
4	5	Father of the Bride Part II (1995)	Comedy	(1995)

Qn.14. Select rows based on timestamps later than 2015-02-01

```
In [40]:
```

```
tags["times"]=pd.to_datetime(tags["timestamp"], format = '%Y-%m-%d', exact = True)
```

In [41]:

```
ratings["times"]=pd.to_datetime(ratings["timestamp"], format = '%Y-%m-%d', exact = True)
```

In [42]:

```
print(tags[tags["times"]> '2015-02-01'])
```

Empty DataFrame
Columns: [userId, movieId, tag, timestamp, times]
Index: []

In [43]:

```
print(ratings[ratings["times"]> '2015-02-01'])
```

Empty DataFrame
Columns: [userId, movieId, rating, times

Columns: [userId, movieId, rating, timestamp, times]
Index: []

Qn.15. Sort the tags dataframe based on timestamp

In [44]:

```
sorted_tags=tags.sort_values(by=['timestamp'])
```

In [45]:

```
sorted_tags.head()
```

Out[45]:

	userld	movield	tag	timestamp	times
333932	100371	2788	monty python	1135429210	1970-01-01 00:00:01.135429210
333927	100371	1732	coen brothers	1135429236	1970-01-01 00:00:01.135429236
333924	100371	1206	stanley kubrick	1135429248	1970-01-01 00:00:01.135429248
333923	100371	1193	jack nicholson	1135429371	1970-01-01 00:00:01.135429371
333939	100371	5004	peter sellers	1135429399	1970-01-01 00:00:01.135429399

2.4. PROBLEM STATEMENT: SCIPY AND MATPLOTLIB

Scipy:

We have the min and max temperatures in a city in India for each months of the year. We would like to find a function to describe this and show it graphically, the dataset given below:

Task:

- 1. fitting it to the periodic function
- 2. plot the fit

Data

Max=39,41,43,47,49,51,45,38,37,29,27,25

Min=21,23,27,28,32,35,31,28,21,19,17,18

```
In [1]:
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

In [2]:

```
%matplotlib inline
```

In [3]:

```
Max=np.array([39,41,43,47,49,51,45,38,37,29,27,25])
Min=np.array([21,23,27,28,32,35,31,28,21,19,17,18])
Months=np.arange(12)
```

In [4]:

```
Max
```

Out[4]:

```
array([39, 41, 43, 47, 49, 51, 45, 38, 37, 29, 27, 25])
```

In [5]:

```
Min
```

Out[5]:

```
array([21, 23, 27, 28, 32, 35, 31, 28, 21, 19, 17, 18])
```

In [6]:

```
fig=plt.figure()
axes=fig.add_axes([0.1,0.1,1,1])

axes.plot(Months,Max,'ro',label='Max Temperature')
axes.plot(Months,Min,'bo',label='Min Temperature')

axes.set_xlabel('Months')
axes.set_ylabel('Temperature($^\circ$C)')
axes.set_title('Temperature Graph for a city in India')
```

```
axes.sec_crite( remperature Graph for a crty in india )
axes.legend(loc=0)

plt.show()
```


Months

In [7]:

```
#Writing the periodic function for curve fitting
```

In [8]:

```
from scipy import optimize
```

In [9]:

```
def yearly_temp(times, avg, ampl, time_offset):
    return (avg+ ampl * np.cos((times + time_offset) * 2 * np.pi / times.max()))
```

In [10]:

```
res_max, cov_max = optimize.curve_fit(yearly_temp, Months,Max, [30, 20, 0])
res_min, cov_min = optimize.curve_fit(yearly_temp, Months,Min, [50, 10, 0])
optimize.curve_fit(yearly_temp, Months,Min, [40, 10, 0])
```

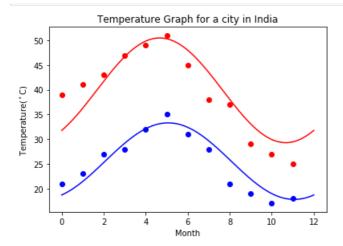
Out[10]:

In [11]:

```
#to plot the fit
```

In [12]:

```
days=np.linspace(0,12,num=365)
fig=plt.figure()
axes=fig.add_axes([0.1,0.1,0.8,0.8])
axes.plot(Months, Max, 'ro')
axes.plot(days, yearly_temp(days, *res_max), 'r-')
axes.plot(Months, Min, 'bo')
axes.plot(days, yearly_temp(days, *res_min), 'b-')
axes.set_xlabel('Month')
axes.set_ylabel('Temperature($^\circ$C)')
axes.set_title('Temperature Graph for a city in India')
plt.show()
```



Matplotlib

Qn. This assignment is for visualization using matplotlib:

data to use:

 $\textbf{url=} \underline{https://raw.githubusercontent.com/Geoyi/Cleaning-Titanic-Data/master/titanic_original.csv}$

titanic=pd.read_csv(url)

Charts to plot:

- 1. Create a pie chart representing the male/female proportion
- 2. Create a scatterplot with the Fare paid and the Age, differ the plot color by gender

```
In [13]:
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

In [14]:

```
%matplotlib inline
```

In [15]:

```
titanic=pd.read_csv("https://raw.githubusercontent.com/Geoyi/Cleaning-Titanic-
Data/master/titanic_original.csv")
```

In [16]:

```
titanic.head()
```

Out[16]:

	pclass	survived	name	sex	age	sibsp	parch	ticket	fare	cabin	embarked	boat	body	home.dest
0	1.0	1.0	Allen, Miss. Elisabeth Walton	female	29.0000	0.0	0.0	24160	211.3375	B5	S	2	NaN	St Louis, MO
1	1.0	1.0	Allison, Master. Hudson Trevor	male	0.9167	1.0	2.0	113781	151.5500	C22 C26	S	11	NaN	Montreal, PQ / Chesterville, ON
2	1.0	0.0	Allison, Miss. Helen Loraine	female	2.0000	1.0	2.0	113781	151.5500	C22 C26	S	NaN	NaN	Montreal, PQ / Chesterville, ON

	pclas	s	survived	name	sex	age	sibsp	parch	ticket	fare	cabin	embarked	boat	body	homoendesit
3	1.	.0	0.0	Joshua Creighton	male	30.0000	1.0	2.0	113781	151.5500	C22 C26	S	NaN	135.0	PQ / Chesterville, ON
4	. 1.	.0	0.0	Allison, Mrs. Hudson J C (Bessie Waldo Daniels)	female	25.0000	1.0	2.0	113781	151.5500	C22 C26	S	NaN	NaN	Montreal, PQ / Chesterville, ON

In [17]:

```
titanic.shape
```

Out[17]:

(1310, 14)

In [18]:

```
gender=titanic['sex'].value_counts() #gives the male female proportion
```

In [19]:

```
gender
```

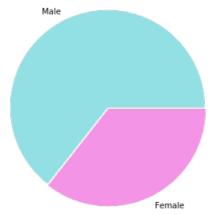
Out[19]:

male 843 female 466

Name: sex, dtype: int64

In [20]:

Male/Female Proportion



```
In [21]:
```

titanic.head()

Out[21]:

ı	pclass	survived	name	sex	age	sibsp	parch	ticket	fare	cabin	embarked	boat	body	home.dest
0	1.0	1.0	Allen, Miss. Elisabeth Walton	female	29.0000	0.0	0.0	24160	211.3375	B5	S	2	NaN	St Louis, MO
1	1.0	1.0	Allison, Master. Hudson Trevor	male	0.9167	1.0	2.0	113781	151.5500	C22 C26	S	11	NaN	Montreal, PQ / Chesterville, ON
2	1.0	0.0	Allison, Miss. Helen Loraine	female	2.0000	1.0	2.0	113781	151.5500	C22 C26	S	NaN	NaN	Montreal, PQ / Chesterville, ON
3	1.0	0.0	Allison, Mr. Hudson Joshua Creighton	male	30.0000	1.0	2.0	113781	151.5500	C22 C26	S	NaN	135.0	Montreal, PQ / Chesterville, ON
4	1.0	0.0	Allison, Mrs. Hudson J C (Bessie Waldo Daniels)	female	25.0000	1.0	2.0	113781	151.5500	C22 C26	S	NaN	NaN	Montreal, PQ / Chesterville, ON

In [22]:

titanic.columns

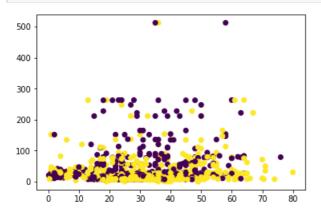
Out[22]:

In [23]:

```
#have to get color from titanic['sex']. so have to split the male and female to labels of 1 and 0
and hence using factorize

fig, ax = plt.subplots()
ax.scatter(titanic['age'], titanic['fare'], c=pd.factorize(titanic['sex']) [0])

plt.show()
```



2.5. PROBLEM STATEMENT: DATA CLEANING

Qn1. Some Values in the FlightNumber column are missing. These numbers are meant to increase by 10 with each row so 10055 and 10075 need to be put in place. Fill in these missing numbers and make the column an integer column(instead of a float column).

```
In [1]:
```

```
import numpy as np
import pandas as pd
```

In [2]:

In [3]:

df

Out[3]:

	From_To	Flight Number	RecentDelays	Airline
0	LoNDon_paris	10045.0	[23, 47]	KLM(!)
1	MAdrid_miLAn	NaN	0	<air france=""> (12)</air>
2	londON_StockhOlm	10065.0	[24, 43, 87]	(BritishAirways.)
3	Budapest_PaRis	NaN	[13]	12.Air France
4	Brussels_londOn	10085.0	[67, 32]	'Swiss Air'

In [4]:

```
df['Flight Number']=df['Flight Number'].interpolate().astype(int)
#interpolate increases the number by 10 with each rows in Flight Number column
#astype changes the float value to int
```

In [5]:

df

Out[5]:

	From_To	Flight Number	RecentDelays	Airline
0	LoNDon_paris	10045	[23, 47]	KLM(!)
1	MAdrid_miLAn	10055	D	<air france=""> (12)</air>
2	londON_StockhOlm	10065	[24, 43, 87]	(BritishAirways.)
3	Budapest_PaRis	10075	[13]	12.Air France
4	Brussels_londOn	10085	[67, 32]	'Swiss Air'

Qn2. The From To column would be better as two separate columns! Split each string on the underscore delimiter to give a new temporary DataFrame with the correct values. Assign the correct column names to this temporary DataFrame.

```
In [6]:
```

```
df['From'], df['To'] = df['From_To'].str.split('_').str
```

#splitting From_To to 2 columns and so df is the new temporary dataframe with correct column names 'From' and 'To'

In [7]:

df

Out[7]:

	From_To	Flight Number	RecentDelays	Airline	From	То
0	LoNDon_paris	10045	[23, 47]	KLM(!)	LoNDon	paris
1	MAdrid_miLAn	10055	О	<air france=""> (12)</air>	MAdrid	miLAn
2	londON_StockhOlm	10065	[24, 43, 87]	(BritishAirways.)	londON	StockhOlm
3	Budapest_PaRis	10075	[13]	12.Air France	Budapest	PaRis
4	Brussels_londOn	10085	[67, 32]	'Swiss Air'	Brussels	londOn

Qn.3. Notice how the capitalisation of the city names is all mixed up in this temporary DataFrame. Standardise the strings so that only the first letter is uppercase(e.g., "londOn" should become "London".)

```
In [8]:
```

```
df['From'] = df['From'].str.capitalize()
```

In [9]:

```
df['To'] = df['To'].str.capitalize()
```

In [10]:

df

Out[10]:

	From_To	Flight Number	RecentDelays	Airline	From	То
0	LoNDon_paris	10045	[23, 47]	KLM(!)	London	Paris
1	MAdrid_miLAn	10055	0	<air france=""> (12)</air>	Madrid	Milan
2	londON_StockhOlm	10065	[24, 43, 87]	(BritishAirways.)	London	Stockholm
3	Budapest_PaRis	10075	[13]	12.Air France	Budapest	Paris
4	Brussels_londOn	10085	[67, 32]	'Swiss Air'	Brussels	London

Qn.4. Delete the From_To column from df and attach the temporary DataFrame from the previous questions

In [11]:

```
df.drop(["From_To"],axis=1,inplace=True) #deleting the From_To column
```

In [12]:

df

Out[12]:

	Flight Number	RecentDelays	Airline	From	То
0	10045	[23, 47]	KLM(!)	London	Paris

-		[==,]			
1	Flight Nupopogs	RecentDelays []	<air f<b="">Airlies (12)</air>	From Madrid	To Milan
2	10065	[24, 43, 87]	(BritishAirways.)	London	Stockholm
3	10075	[13]	12.Air France	Budapest	Paris
4	10085	[67, 32]	'Swiss Air'	Brussels	London

Qn.5. In the RecentDelays column, the values have been entered into the DataFrame as a list. We would like each first value in its column, each second value in its column, and so on. If there isn't an Nth value, the value should be NaN.

Expand the Series of lists into a DataFrame named delays, rename the columns delay_1, delay_2, etc. and replace the unwanted RecentDdelays column in df with delays.

In [13]:

```
df['RecentDelays'] #is the column in the dataframe which is represented as a list
```

Out[13]:

```
0 [23, 47]
1 []
2 [24, 43, 87]
3 [13]
4 [67, 32]
```

Name: RecentDelays, dtype: object

In [14]:

```
df[['delay_1','delay_2','delay_3']]=pd.DataFrame(df['RecentDelays'].values.tolist(), index=
df.index)
# using the function .values.tolist() to convert list to different columns and storing it in df
```

In [15]:

df

Out[15]:

	Flight Number	RecentDelays	Airline	From	То	delay_1	delay_2	delay_3
0	10045	[23, 47]	KLM(!)	London	Paris	23.0	47.0	NaN
1	10055	П	<air france=""> (12)</air>	Madrid	Milan	NaN	NaN	NaN
2	10065	[24, 43, 87]	(BritishAirways.)	London	Stockholm	24.0	43.0	87.0
3	10075	[13]	12.Air France	Budapest	Paris	13.0	NaN	NaN
4	10085	[67, 32]	'Swiss Air'	Brussels	London	67.0	32.0	NaN

In [16]:

```
df.drop(["RecentDelays"],axis=1,inplace=True) #deleting the 'RecentDelays' column
```

In [17]:

df

Out[17]:

	Flight Number	Airline	From	То	delay_1	delay_2	delay_3
0	10045	KLM(!)	London	Paris	23.0	47.0	NaN
1	10055	<air france=""> (12)</air>	Madrid	Milan	NaN	NaN	NaN

3 10075 12.Air France Budapest Paris 13.0 NaN NaN 4 10085 'Swiss Air' Brussels London 67.0 32.0 NaN	2	#0g65 Number	(BritishAirways.)	London From	Stockholm	24.0 delay_1	43.0 delay_2	87.0 delay_3
4 10085 'Swiss Air' Brussels London 67.0 32.0 NaN	_3		12 Air France	Budapest	Paris	13.0	NaN	NaN
	4	10085	'Swiss Air'	Brussels	London	67.0	32.0	NaN

In []: