```
Q 19,20
import numpy as np
import math
import pandas as pd
Default title text
#@title Default title text
data=[12,4,5,21,7,14,15,18,20]
print(data)
x=np.sort(data)
print(x)
    [12, 4, 5, 21, 7, 14, 15, 18, 20]
    [ 4 5 7 12 14 15 18 20 21]
bin1=np.zeros((3,3))
bin2=np.zeros((3,3))
bin3=np.zeros((3,3))
MEAN
#MEAN MEAN MEAN
for i in range (0,9,3):
 k=int(i/3)
 mean=(x[i] + x[i+1] + x[i+2])/3
 for j in range(3):
   bin1[k,j]=mean
print("----- \n",bin1)
          -----Bean Bin:-----
     [[ 5.33333333 5.33333333 5.33333333]
     [13.66666667 13.66666667 13.66666667]
     [19.66666667 19.66666667 19.66666667]]
#MEDIAN MEDIAN MEDIAN
for i in range (0,9,3):
 k=int(i/3)
 for j in range (3):
   bin2[k,j]=x[i+1]
print("----- \n",bin2)
               -----Median Bin :-----
     [[ 5. 5. 5.]
     [14. 14. 14.]
     [20. 20. 20.]]
```

```
Q
```

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

```
x=[27,23,34,45,49,69,62,55,21,18]
print(x)
```

```
np.digitize(x,bins=[15,30,45,60,75])
```

```
array([1, 1, 2, 3, 3, 4, 4, 3, 1, 1])
```

```
df = pd.DataFrame({"age":x})
df.head()
```

age	9
-----	---

- 0 27
- 1 23
- **2** 34
- **3** 45
- 4 49

df['binned']=pd.cut(x=df['age'], bins=[0,15,30,45,60,75],labels=['child','young','adult','

## df.head()

	age	binned
0	27	young
1	23	young
2	34	adult
3	45	adult
4	49	senior
•	49	Senic

import pandas as pd
import numpy as np

```
x=[12,23,34,45,50,110,80,60,150,125]
print(x)

   [12, 23, 34, 45, 50, 110, 80, 60, 150, 125]

np.digitize(x,bins=[30,60,90,120,150,180])

   array([0, 0, 1, 1, 1, 3, 2, 2, 5, 4])

df=pd.DataFrame({'salary':x})
   df.head()
```

```
    salary

    0
    12

    1
    23

    2
    34

    3
    45

    4
    50
```

df['discritized']=pd.cut(x=df['salary'],bins=[0,30,60,90,120,150,180],labels=['boom','shak
df.head(15)

	salary	discritized
0	12	boom
1	23	boom
2	34	shake
3	45	shake
4	50	shake
5	110	room
6	80	the
7	60	shake
8	150	da
9	125	da

```
df['discritized']=pd.cut(x=df['salary'],bins=[0,30,60,90,120,150,180])
df.head(15)
```

```
salary discritized
0
        12
                    (0, 30]
        23
                    (0, 30]
2
        34
                   (30, 60]
        45
                   (30, 60]
        50
                   (30, 60]
                 (90, 120]
5
       110
6
        80
                   (60, 90]
7
        60
                   (30, 60]
```

```
points = [ [2, 3], [2, 4], [4, 1], [3, 2] ]

dis_matrix =[]
for p in points:
    list = []
    for j in points:
        list.append((abs(p[0]-j[0])+abs(p[1]-j[1])))
        dis_matrix.append(list)

for i in range(len(points)):
    for j in range(len(points)):
        if i<j: dis_matrix[i][j] = 0

# (dis_matrix)
print(dis_matrix)

        [[0, 0, 0, 0], [1, 0, 0, 0], [4, 5, 0, 0], [2, 3, 2, 0]]

Double-click (or enter) to edit</pre>
```

```
points=[[2,3],[2,4],[4,1],[3,2]]

dis_matrix=[]
for p in points:
    list=[]
    for j in points:
        list.append((abs(p[0]-j[0])+abs(p[1]-j[1])))
        dis_matrix.append(list)

for i in range(len(points)):
    for j in range(len(points)):
        if i<j: dis_matrix[i][j]=0
        (dis_matrix)</pre>

[[0, 0, 0, 0], [1, 0, 0, 0], [4, 5, 0, 0], [2, 3, 2, 0]]
```

min maxxmin max

```
import pandas as pd

df=pd.DataFrame([12000,145000,320000,45000,90000,134000,200000])

display(df)
min_max=df.copy()
# print(df.copy())

for i in min_max:
    min_max[i]=(min_max[i]-min_max[i].min())/(min_max[i].max()-min_max[i].min())
print(min_max)
```

```
0
   12000
1 145000
2 320000
3
   45000
   90000
5 134000
6 200000
0.000000
1 0.431818
2 1.000000
3 0.107143
4 0.253247
5 0.396104
6 0.610390
```

for i in zscore.columns:

display(zscore)

```
09/11/2022, 02:30
                                               Untitled4.ipynb - Colaboratory
            Stud Id Stud Braanch
                                  Score
        0
                  1
                              ΑI
                                   17.0
                            COMP
                  2
                                    NaN
        1
                            COMP
                                    15.0
                 4
                            COMP
                                   16.0
                              ΑI
                                    18.0
                  6
                              ΑI
                                   15.0
        6
                              ΑI
                                    NaN
        0
              False
        1
              True
        2
             False
              False
        4
             False
              False
        6
               True
        Name: Score, dtype: bool
            Stud Id Stud Braanch Score
        0
                              ΑI
                                   17.0
        1
                  2
                            COMP
                                    0.0
        2
                            COMP 15.0
                            COMP 16.0
                  4
        4
                              ΑI
                                   18.0
                              AI 15.0
                              ΑI
        6
                                    0.0
   import pandas as pd
   import numpy as np
   df=pd.DataFrame([[101,120000],
                     [102,145000],
                     [103,320000],
                     [104,45000],
                     [105,90000],
                     [106,134000],
                     [107,200000]],
                    columns=['EMP ID','Salary'])
   display(df)
   zscore=df.copy()
```

zscore[i]=(zscore[i]-zscore[i].mean())/zscore[i].std()

```
EMP ID Salary
0
      101
           120000
      102
           145000
2
      103
           320000
      104
            45000
      105
            90000
5
      106
          134000
      107 200000
    EMP ID
               Salary
            -0.344670
  -1.38873
  -0.92582
            -0.062814
```

## pip install efficient-apriori

Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/p</a> Requirement already satisfied: efficient-apriori in /usr/local/lib/python3.7/dist-page 1.00 (a) the content of the co

```
from efficient_apriori import apriori
transactions=[(1,3,4,2),
                                             (2,3,5,1),
                                             (1,2,3,4),
                                             (2,5,3),
                                             (1,2,3,4)
min_support=0.7
min confidence = 0
itemsets, rules = apriori(transactions, min_support=min_support, min_confidence=min_confid
print(itemsets)
for rule in rules:
      print(rule)
                 \{1: \{(1,): 4, (3,): 5, (2,): 5\}, 2: \{(1, 2): 4, (1, 3): 4, (2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5\}, 3: \{(1, 2, 3): 5
                 {2} -> {1} (conf: 0.800, supp: 0.800, lift: 1.000, conv: 1.000)
                 {1} -> {2} (conf: 1.000, supp: 0.800, lift: 1.000, conv: 0.000)
                 {3} -> {1} (conf: 0.800, supp: 0.800, lift: 1.000, conv: 1.000)
                 {1} -> {3} (conf: 1.000, supp: 0.800, lift: 1.000, conv: 0.000)
                {3} -> {2} (conf: 1.000, supp: 1.000, lift: 1.000, conv: 0.000)
                 {2} -> {3} (conf: 1.000, supp: 1.000, lift: 1.000, conv: 0.000)
                {2, 3} -> {1} (conf: 0.800, supp: 0.800, lift: 1.000, conv: 1.000)
                 {1, 3} -> {2} (conf: 1.000, supp: 0.800, lift: 1.000, conv: 0.000)
                 {1, 2} -> {3} (conf: 1.000, supp: 0.800, lift: 1.000, conv: 0.000)
                 {3} -> {1, 2} (conf: 0.800, supp: 0.800, lift: 1.000, conv: 1.000)
```

{2} -> {1, 3} (conf: 0.800, supp: 0.800, lift: 1.000, conv: 1.000) {1} -> {2, 3} (conf: 1.000, supp: 0.800, lift: 1.000, conv: 0.000)

```
import pandas as pd
import numpy as np
def mean(arr):
 m=0
 for i in arr:
    m+=i
  res=m//len(arr)
  return res
def k_mean(arr,m1,m2):
  cluster1=[]
  cluster2=[]
  for i in arr:
    if(abs(m1-i)<abs(m2-i)):</pre>
      cluster1.append(i)
    elif(abs(m1-i)==abs(m2-i)):
      if(len(cluster1)>len(cluster2)):
        cluster2.append(i)
      else:
        cluster1.append(i)
    else:
        cluster2.append(i)
  c1_mean=mean(cluster1)
  c2_mean=mean(cluster2)
  print('C1 :',c1_mean)
  print('C2 :',c2_mean)
  print('Cluster1',cluster1)
  print('Cluster2',cluster2)
  if(c1_mean==m1 and c2_mean==m2):
    print('-----
    print('final cluster are')
    print('Cluster1',cluster1)
    print('Cluster2',cluster2)
  else:
    k_mean(arr,c1_mean,c2_mean)
arr=[2,5,8,12,14,16,22,24,30,32,35]
k=2
k_mean(arr,5,14)
     C1 : 5
     C2: 23
     Cluster1 [2, 5, 8]
     Cluster2 [12, 14, 16, 22, 24, 30, 32, 35]
     C1 : 6
     C2 : 24
     Cluster1 [2, 5, 8, 12]
```

```
Cluster2 [14, 16, 22, 24, 30, 32, 35]
     C1:8
     C2 : 26
     Cluster1 [2, 5, 8, 12, 14]
     Cluster2 [16, 22, 24, 30, 32, 35]
     C1 : 9
     C2: 28
     Cluster1 [2, 5, 8, 12, 14, 16]
     Cluster2 [22, 24, 30, 32, 35]
     C1:9
     C2 : 28
     Cluster1 [2, 5, 8, 12, 14, 16]
     Cluster2 [22, 24, 30, 32, 35]
     final cluster are
     Cluster1 [2, 5, 8, 12, 14, 16]
     Cluster2 [22, 24, 30, 32, 35]
#binning
data = [[1,120000],
        [2,23000],
        [3,34000],
        [4,45000],
        [5,50000],
        [6,110000],
        [7,80000],
        [9,150000],
        [10,125000]]
def binning(data):
  data1= data.copy()
  bin1 = []
  bin2 = []
  bin3 = []
  #create bins
  for i in range (len(data)):
    if ((data1[i][1] >10000) and data1[i][1] < 55001):</pre>
      bin1.append(data[i][1])
      data1[i][1] = 1
    if ((data1[i][1] >55000) and data1[i][1] < 90001):</pre>
      bin2.append(data[i][1])
      data1[i][1] = 2
    if ((data1[i][1] >90000) and data1[i][1] < 1250001):</pre>
      bin3.append(data[i][1])
      data1[i][1] = 3
  print(data1)
  print("bin1 = ", bin1)
  print("bin2 = ", bin2)
```

```
print("bin3 = ", bin3)
binning(data)
      [[1, 3], [2, 1], [3, 1], [4, 1], [5, 1], [6, 3], [7, 2], [9, 3], [10, 3]]
bin1 = [23000, 34000, 45000, 50000]
      bin2 = [80000]
      bin3 = [120000, 110000, 150000, 125000]
                              Colab paid products - Cancel contracts here
```