Zadanie 1

```
import pandas as pd
import numpy as np

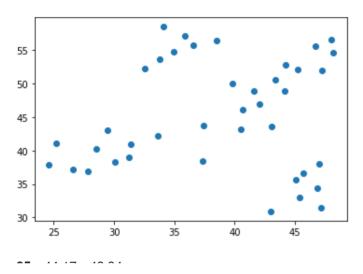
dataEx1 = pd.read_csv('k_means_data.csv')
dataEx1
```

	X	Y
0	25.23	41.09
1	45.10	35.69
2	26.59	37.21
3	27.80	36.93
4	28.56	40.21
5	29.49	43.05
6	30.04	38.33
7	31.25	39.03
8	31.35	40.98
9	42.95	30.91
10	37.30	38.42
11	37.39	43.69
12	32.53	52.18
13	33.60	42.20
14	24.60	37.88
15	33.76	53.60
16	34.97	54.72
17	35.84	57.04
18	47.16	31.52
19	36.58	55.76
20	46.82	34.33
21	46.98	38.03
າາ	<i>1</i> 5 3 <i>1</i>	33 US

```
import matplotlib.pyplot as plt
```

```
24 48.12 54.53
```

```
plt.scatter(dataEx1.iloc[:,0],dataEx1.iloc[:,1])
plt.show()
```



from sklearn.cluster import AgglomerativeClustering

k = 3

model = AgglomerativeClustering(linkage='single',n_clusters=k,affinity='euclidean', distance_threshold=None)
model.fit(dataEx1)

AgglomerativeClustering(linkage='single', n_clusters=3)

model.labels_

```
array([0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 2, 0, 0, 2, 2, 2, 1, 2, 1, 1, 1, 1, 0, 0, 2, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0])
```

from sklearn.cluster import KMeans

```
model.labels
    array([2, 1, 2, 2, 2, 2, 2, 2, 2, 1, 2, 2, 0, 2, 2, 0, 0, 0, 1, 0, 1, 1,
           1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, dtype=int32)
K=model.cluster centers
K
    array([[40.91210526, 52.77789474],
           [44.84
                      , 36.31444444],
           [30.26666667, 39.91833333]])
Zadanie 2
```

```
data = pd.read csv('k means data.csv')
print(data)
data = np.array(data)
           X
                  Y
       25.23 41.09
    0
    1
       45.10 35.69
       26.59 37.21
       27.80 36.93
       28.56 40.21
       29.49 43.05
       30.04 38.33
       31.25 39.03
       31.35 40.98
       42.95 30.91
    10 37.30 38.42
    11 37.39 43.69
    12 32.53 52.18
    13 33.60 42.20
    14 24.60 37.88
    15 33.76 53.60
    16 34.97 54.72
    17 35.84 57.04
```

18 47.16 31.52

```
19 36.58 55.76
       46.82 34.33
    20
    21 46.98 38.03
    22
       45.34 33.02
    23
       45.70 36.63
       48.12 54.53
       47.25 51.99
    25
       38.48 56.41
    26
        34.10 58.45
    27
       39.81 49.98
        40.47 43.18
       40.66
              46.02
    30
    31 41.59 48.86
       42.05 46.89
    32
    33
       43.04 43.52
       43.34 50.49
       44.17 48.94
       44.23 52.81
       45.19 52.05
    37
       46.68 55.59
    38
    39 47.98 56.47
Label = np.zeros((len(data),1),dtype=np.float64)
Label
    array([[0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
           [0.],
```

```
[0.],
           [0.],
           [0.],
           [0.],
          [0.],
           [0.],
          [0.],
          [0.],
           [0.],
          [0.],
           [0.],
          [0.],
          [0.],
           [0.],
           [0.],
           [0.],
          [0.],
           [0.],
          [0.],
           [0.],
          [0.],
          [0.]])
NewData = np.concatenate([data,Label],axis=1)
NewData
    array([[25.23, 41.09, 0. ],
          [45.1 , 35.69, 0. ],
          [26.59, 37.21, 0. ],
          [27.8 , 36.93, 0. ],
          [28.56, 40.21, 0. ],
           [29.49, 43.05, 0. ],
           [30.04, 38.33, 0. ],
          [31.25, 39.03, 0. ],
          [31.35, 40.98, 0. ],
          [42.95, 30.91, 0. ],
          [37.3 , 38.42, 0. ],
          [37.39, 43.69, 0. ],
          [32.53, 52.18, 0. ],
          [33.6 , 42.2 , 0. ],
          [24.6 , 37.88, 0. ],
          [33.76, 53.6, 0.],
           [34.97, 54.72, 0. ],
```

```
[35.84, 57.04, 0. ],
         [47.16, 31.52, 0. ],
         [36.58, 55.76, 0. ],
         [46.82, 34.33, 0. ],
         [46.98, 38.03, 0. ],
         [45.34, 33.02, 0. ],
         [45.7, 36.63, 0.],
         [48.12, 54.53, 0. ],
         [47.25, 51.99, 0. ],
         [38.48, 56.41, 0. ],
         [34.1 , 58.45 , 0. ],
         [39.81, 49.98, 0. ],
         [40.47, 43.18, 0. ],
         [40.66, 46.02, 0. ],
         [41.59, 48.86, 0. ],
         [42.05, 46.89, 0. ],
         [43.04, 43.52, 0. ],
         [43.34, 50.49, 0. ],
         [44.17, 48.94, 0.],
         [44.23, 52.81, 0. ],
         [45.19, 52.05, 0. ],
         [46.68, 55.59, 0. ],
         [47.98, 56.47, 0. ]])
NewData[:,2]
   0., 0., 0., 0., 0., 0.1
minX = int(np.min(data[:,0]))
maxX = int(np.max(data[:,0]))
minY = int(np.min(data[:,1]))
maxY = int(np.max(data[:,1]))
print(minX," ",maxX," ",minY," ",maxY)
   24 48 30 58
import random
CentersX = []
CentersY = []
```

https://colab.research.google.com/drive/1gbDLqgzYsrLh-RMgVZNC66ghyUGPt8r3#printMode=true

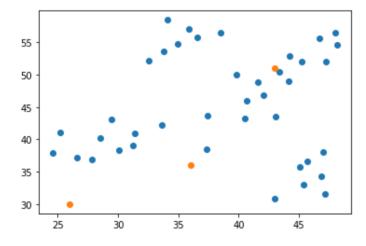
```
for _ in range(0,3):
    CentersX.append(random.randrange(minX,maxX))
    CentersY.append(random.randrange(minY,maxY))

CentersX = np.array(CentersX).reshape(3,1)
CentersY = np.array(CentersY).reshape(3,1)

Centers = np.concatenate([CentersX,CentersY],axis=1).astype(np.float64)

Centers
    array([[36., 36.],
        [43., 51.],
        [26., 30.]])

import matplotlib.pyplot as plt
plt.scatter(data[:,0],data[:,1])
plt.scatter(Centers[:,0],Centers[:,1])
plt.show()
```



import math

```
for point in NewData:
   print(point[0],point[1])
```

```
i = -1
j = -1
print("
for cent in Centers:
 print("Cent",cent[0],cent[1])
 i = i + 1
 euclidesian=math.sqrt(((cent[1]-point[1])*(cent[1]-point[1]))+((cent[0]-point[0])*(cent[0]-point[0])))
 if(i==0):
   min=euclidesian
   j=i
  if(min>euclidesian):
   min=euclidesian
   j=i
  print(euclidesian)
 print("Max"," i ", min,i,j)
  point[2]=j
  25.23 41.09
  Cent 36.0 36.0
  11.912220615821385
  Max i 11.912220615821385 0 0
  Cent 43.0 51.0
  20.34652304449092
  Max i 11.912220615821385 1 0
  Cent 26.0 30.0
  11.11669915037733
  Max i 11.11669915037733 2 2
  45.1 35.69
  Cent 36.0 36.0
  9.105278688760714
  Max i 9.105278688760714 0 0
  Cent 43.0 51.0
  15.453352387103585
  Max i 9.105278688760714 1 0
  Cent 26.0 30.0
  19.929528343641252
  Max i 9.105278688760714 2 0
  26.59 37.21
  Cent 36.0 36.0
```

11.12.2022, 11:10

```
9.487475955173748
Max i 9.487475955173748 0 0
Cent 43.0 51.0
21.434836131867208
Max i 9.487475955173748 1 0
Cent 26.0 30.0
7.234099805780953
Max i 7.234099805780953 2 2
27.8 36.93
Cent 36.0 36.0
8.252569296891727
Max i 8.252569296891727 0 0
Cent 43.0 51.0
20.712433463984862
Max i 8.252569296891727 1 0
Cent 26.0 30.0
7.159951117151569
Max i 7.159951117151569 2 2
28.56 40.21
Cent 36.0 36.0
8.548549584578662
Max i 8.548549584578662 0 0
Cent 43.0 51.0
18.02602840339491
Max i 8.548549584578662 1 0
Cent 26.0 30.0
10.526048641346856
Max i 8.548549584578662 2 0
29.49 43.05
- . . . . . . .
```

NewData

array([[25.23, 41.09, 2.],
[45.1, 35.69, 0.],
[26.59, 37.21, 2.],
[27.8, 36.93, 2.],
[28.56, 40.21, 0.],
[29.49, 43.05, 0.],
[30.04, 38.33, 0.],
[31.25, 39.03, 0.],

x0=0

```
[31.35, 40.98, 0. ],
         [42.95, 30.91, 0. ],
         [37.3 , 38.42 , 0. ],
         [37.39, 43.69, 0. ],
         [32.53, 52.18, 1. ],
         [33.6, 42.2, 0.],
         [24.6 , 37.88, 2. ],
         [33.76, 53.6, 1.],
         [34.97, 54.72, 1. ],
         [35.84, 57.04, 1.],
         [47.16, 31.52, 0. ],
         [36.58, 55.76, 1. ],
         [46.82, 34.33, 0. ],
         [46.98, 38.03, 0. ],
         [45.34, 33.02, 0. ],
         [45.7, 36.63, 0. ],
         [48.12, 54.53, 1.],
         [47.25, 51.99, 1.],
         [38.48, 56.41, 1. ],
         [34.1, 58.45, 1.],
         [39.81, 49.98, 1.],
         [40.47, 43.18, 1.],
         [40.66, 46.02, 1.],
         [41.59, 48.86, 1.],
         [42.05, 46.89, 1.],
         [43.04, 43.52, 1. ],
         [43.34, 50.49, 1.],
         [44.17, 48.94, 1.],
         [44.23, 52.81, 1.],
         [45.19, 52.05, 1.],
         [46.68, 55.59, 1.],
         [47.98, 56.47, 1. ]])
NewData[:,2]
   array([2., 0., 2., 2., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 2., 1., 1.,
         1., 1., 1., 1., 1., 1.]
OldCenters = Centers
while True:
```

```
11.12.2022, 11:10
     y0 = 0
     i0=0
     x1=0
     y1 = 0
     i1=0
     x2 = 0
     y2 = 0
     i2=0
     Cluster1X=[]
     Cluster1Y=[]
     Cluster2X=[]
     Cluster2Y=[]
     Cluster3X=[]
     Cluster3Y=[]
     for point in NewData:
       print(point[0],point[1],point[2])
       if(point[2]==0):
         x0=x0+point[0]
         Cluster1X.append(point[0])
         y0=y0+point[1]
         Cluster1Y.append(point[1])
         i0=i0+1
       if(point[2]==1):
         x1=x1+point[0]
         Cluster2X.append(point[0])
         y1=y1+point[1]
         Cluster2Y.append(point[1])
         i1=i1+1
       if(point[2]==2):
         x2=x2+point[0]
         Cluster3X.append(point[0])
         y2=y2+point[1]
         Cluster3Y.append(point[1])
         i2=i2+1
     NewCentersX=[]
     NewCentersY=[]
     #NewCentersX.append(x0/i0)
```

```
#NewCentersY.append(y0/i0)
#NewCentersX.append(x1/i1)
#NewCentersY.append(y1/i1)
#NewCentersX.append(x2/i2)
#NewCentersY.append(y2/i2)
NewCentersX.append(np.mean(Cluster1X))
NewCentersX.append(np.mean(Cluster2X))
NewCentersX.append(np.mean(Cluster3X))
NewCentersY.append(np.mean(Cluster1Y))
NewCentersY.append(np.mean(Cluster2Y))
NewCentersY.append(np.mean(Cluster3Y))
NewCentersX = np.array(NewCentersX).reshape(3,1)
NewCentersY = np.array(NewCentersY).reshape(3,1)
print(np.mean(Cluster1X))
print(np.mean(Cluster1Y))
print(np.mean(Cluster2X))
print(np.mean(Cluster2Y))
print(np.mean(Cluster3X))
print(np.mean(Cluster3Y))
NewCenters = np.concatenate([NewCentersX,NewCentersY],axis=1).astype(np.float64)
print(x0,y0,i0)
print(x1,y1,i1)
print(x2,y2,i2)
plt.scatter(NewData[:,0],NewData[:,1])
plt.scatter(NewCenters[:,0],NewCenters[:,1])
plt.show()
print(OldCenters)
print(NewCenters)
print("Space")
if np.array equal(OldCenters, NewCenters) == True:
  break
OldCenters = NewCenters
#NewCenters
```

₽

```
for point in NewData:
 #print(point[0],point[1])
 i = -1
 j = -1
              ")
 #print("
  for cent in NewCenters:
   print("Cent",cent[0],cent[1])
   i= i+1
   #euclidesian=math.sqrt(((cent[1]-point[1])*(cent[1]-point[1]))+((cent[0]-point[0])*(cent[0]-point[0])))
   euclidesian=math.sqrt(((cent[0]-point[0])*(cent[0]-point[0]))+((cent[1]-point[1])*(cent[1]-point[1])))
   if(i==0):
     min=euclidesian
      j=i
   if(min>euclidesian):
      min=euclidesian
     j=i
   #print(euclidesian)
   #print("Max"," i ", min,i,j)
   point[2]=j
```

https://colab.research.google.com/drive/1gbDLqgzYsrLh-RMgVZNC66ghyUGPt8r3#printMode=true

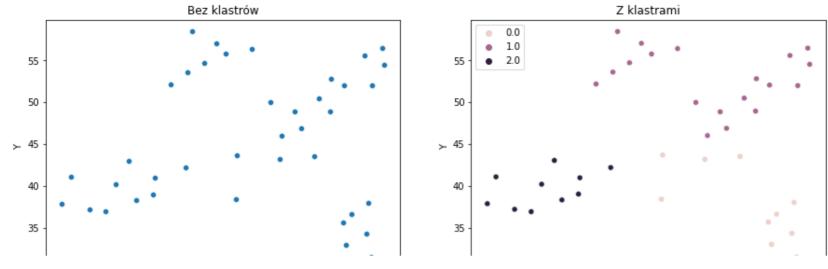
- 25.23 41.09 2.0 45.1 35.69 0.0 26.59 37.21 2.0 27.8 36.93 2.0 28.56 40.21 0.0 29.49 43.05 0.0 30.04 38.33 0.0 31.25 39.03 0.0 31.35 40.98 0.0 42.95 30.91 0.0 37.3 38.42 0.0 37.39 43.69 0.0 32.53 52.18 1.0 33.6 42.2 0.0 24.6 37.88 2.0 33.76 53.6 1.0 34.97 54.72 1.0 35.84 57.04 1.0 47.16 31.52 0.0 36.58 55.76 1.0 46.82 34.33 0.0 46.98 38.03 0.0 45.34 33.02 0.0 45.7 36.63 0.0 48.12 54.53 1.0 47.25 51.99 1.0 38.48 56.41 1.0 34.1 58.45 1.0 39.81 49.98 1.0 40.47 43.18 1.0 40.66 46.02 1.0 41.59 48.86 1.0 42.05 46.89 1.0 43.04 43.52 1.0 43.34 50.49 1.0 44.17 48.94 1.0 44.23 52.81 1.0 45.19 52.05 1.0 46.68 55.59 1.0 47.98 56.47 1.0 38.602 37.736 40.99238095238094 51.87999999999999
- 31.073333333333

26.055

NewData

```
array([[25.23, 41.09, 2. ],
      [45.1, 35.69, 0.],
      [26.59, 37.21, 2. ],
      [27.8, 36.93, 2. ],
      [28.56, 40.21, 2. ],
      [29.49, 43.05, 2. ],
      [30.04, 38.33, 2. ],
      [31.25, 39.03, 2. ],
      [31.35, 40.98, 2. ],
      [42.95, 30.91, 0. ],
      [37.3 , 38.42 , 0. ],
      [37.39, 43.69, 0. ],
      [32.53, 52.18, 1.],
      [33.6, 42.2, 2.],
      [24.6 , 37.88 , 2. ],
      [33.76, 53.6, 1.],
      [34.97, 54.72, 1.],
      [35.84, 57.04, 1.],
      [47.16, 31.52, 0. ],
      [36.58, 55.76, 1. ],
      [46.82, 34.33, 0. ],
      [46.98, 38.03, 0. ],
      [45.34, 33.02, 0. ],
      [45.7, 36.63, 0.],
      [48.12, 54.53, 1. ],
      [47.25, 51.99, 1. ],
      [38.48, 56.41, 1. ],
      [34.1 , 58.45 , 1. ],
      [39.81, 49.98, 1.],
      [40.47, 43.18, 0. ],
      [40.66, 46.02, 1. ],
      [41.59, 48.86, 1.],
      [42.05, 46.89, 1.],
      [43.04, 43.52, 0. ],
      [43.34, 50.49, 1.],
      [44.17, 48.94, 1.],
      [44.23, 52.81, 1.],
      [45.19, 52.05, 1. ],
      [46.68, 55.59, 1.],
      [47.98, 56.47, 1. ]])
```

```
Con+ 30 602 37 736
test=NewData[:,2]
    Comt 20 602 27 726
X=np.asarray(TEST[0])
#print(X)
a=np.asarray(X)
print(a)
    [1. 0. 1. 1. 1. 1. 1. 1. 1. 0. 0. 0. 2. 1. 1. 2. 2. 2. 0. 2. 0. 0. 0. 0.
     2. 2. 2. 2. 2. 0. 2. 2. 0. 2. 2. 2. 2. 2. 2. 2. ]
    Cent 38.602 37.736
df = pd.DataFrame(NewData, columns = ['X', 'Y', 'L'])
    Cent 38,602 37,736
df.head()
           X
                 Y L
     0 25.23 41.09 2.0
     1 45.10 35.69 0.0
     2 26.59 37.21 2.0
     3 27.80 36.93 2.0
     4 28.56 40.21 2.0
    Cen+ 38 602 37 736
dataset = df[["X", "Y"]]
    Con+ 20 602 27 726
import seaborn as sns
fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(15,5))
sns.scatterplot(ax=axes[0], data=df, x='X', y='Y').set_title('Bez klastrów')
sns.scatterplot(ax=axes[1], data=df, x='X', y='Y', hue=test).set title('Z klastrami');
```



NewData

```
array([[25.23, 41.09, 2. ],
      [45.1 , 35.69, 0. ],
      [26.59, 37.21, 2. ],
      [27.8 , 36.93 , 2. ],
      [28.56, 40.21, 2. ],
      [29.49, 43.05, 2. ],
      [30.04, 38.33, 2. ],
      [31.25, 39.03, 2. ],
      [31.35, 40.98, 2. ],
      [42.95, 30.91, 0. ],
      [37.3 , 38.42, 0. ],
      [37.39, 43.69, 0. ],
      [32.53, 52.18, 1. ],
      [33.6 , 42.2 , 2. ],
      [24.6 , 37.88, 2. ],
      [33.76, 53.6, 1.],
      [34.97, 54.72, 1.],
      [35.84, 57.04, 1.],
      [47.16, 31.52, 0. ],
      [36.58, 55.76, 1. ],
      [46.82, 34.33, 0. ],
      [46.98, 38.03, 0. ],
      [45.34, 33.02, 0. ],
      [45.7, 36.63, 0.],
      [48.12, 54.53, 1. ],
```

```
[47.25, 51.99, 1.],
           [38.48, 56.41, 1. ],
           [34.1 , 58.45 , 1. ],
           [39.81, 49.98, 1. ],
           [40.47, 43.18, 0. ],
           [40.66, 46.02, 1.],
           [41.59, 48.86, 1.],
           [42.05, 46.89, 1.],
           [43.04, 43.52, 0. ],
           [43.34, 50.49, 1.],
           [44.17, 48.94, 1.],
           [44.23, 52.81, 1.],
           [45.19, 52.05, 1.],
           [46.68, 55.59, 1.],
           [47.98, 56.47, 1. ]])
    27 0 26 02 2 0
Aby edytować zawartość komórki, kliknij ją dwukrotnie (lub naciśnij klawisz Enter)
    30.04 38.33 2.0
Zadanie 3
    42.95 30.91 0.0
def lcs(X, Y, m, n):
    if m == 0 or n == 0:
      return 0;
    elif X[m-1] == Y[n-1]:
      return 1 + lcs(X, Y, m-1, n-1);
   else:
      return \max(lcs(X, Y, m, n-1), lcs(X, Y, m-1, n));
X = "abcde"
Y = "bcdxy"
print ("Length of LCS is ", lcs(X, Y, len(X), len(Y)))
    Length of LCS is 3
    34.1 58.45 1.0
Zadanie 4
    40.00 40.02 I.U
```

dataEx4.describe()

dataEx4 = pd.read_csv('iris.csv')
dataEx4

1 4.9 3.0 1.4 0.2 seto 2 4.7 3.2 1.3 0.2 seto 3 4.6 3.1 1.5 0.2 seto 4 5.0 3.6 1.4 0.2 seto 145 6.7 3.0 5.2 2.3 virgini 146 6.3 2.5 5.0 1.9 virgini 147 6.5 3.0 5.2 2.0 virgini 148 6.2 3.4 5.4 2.3 virgini		sepal_length	sepal_width	petal_length	petal_width	species
2 4.7 3.2 1.3 0.2 seto 3 4.6 3.1 1.5 0.2 seto 4 5.0 3.6 1.4 0.2 seto 145 6.7 3.0 5.2 2.3 virgini 146 6.3 2.5 5.0 1.9 virgini 147 6.5 3.0 5.2 2.0 virgini 148 6.2 3.4 5.4 2.3 virgini 149 5.9 3.0 5.1 1.8 virgini 150 rows × 5 columns	0	5.1	3.5	1.4	0.2	setosa
3 4.6 3.1 1.5 0.2 seto 4 5.0 3.6 1.4 0.2 seto 145 6.7 3.0 5.2 2.3 virgini 146 6.3 2.5 5.0 1.9 virgini 147 6.5 3.0 5.2 2.0 virgini 148 6.2 3.4 5.4 2.3 virgini 149 5.9 3.0 5.1 1.8 virgini 150 rows × 5 columns	1	4.9	3.0	1.4	0.2	setosa
4 5.0 3.6 1.4 0.2 seto 145 6.7 3.0 5.2 2.3 virgini 146 6.3 2.5 5.0 1.9 virgini 147 6.5 3.0 5.2 2.0 virgini 148 6.2 3.4 5.4 2.3 virgini 149 5.9 3.0 5.1 1.8 virgini 150 rows × 5 columns	2	4.7	3.2	1.3	0.2	setosa
	3	4.6	3.1	1.5	0.2	setosa
145 6.7 3.0 5.2 2.3 virgini 146 6.3 2.5 5.0 1.9 virgini 147 6.5 3.0 5.2 2.0 virgini 148 6.2 3.4 5.4 2.3 virgini 149 5.9 3.0 5.1 1.8 virgini 150 rows × 5 columns	4	5.0	3.6	1.4	0.2	setosa
146 6.3 2.5 5.0 1.9 virgini 147 6.5 3.0 5.2 2.0 virgini 148 6.2 3.4 5.4 2.3 virgini 149 5.9 3.0 5.1 1.8 virgini 150 rows × 5 columns						
147 6.5 3.0 5.2 2.0 virgini 148 6.2 3.4 5.4 2.3 virgini 149 5.9 3.0 5.1 1.8 virgini 150 rows × 5 columns	145	6.7	3.0	5.2	2.3	virginica
148 6.2 3.4 5.4 2.3 virgini 149 5.9 3.0 5.1 1.8 virgini 150 rows × 5 columns	146	6.3	2.5	5.0	1.9	virginica
149 5.9 3.0 5.1 1.8 virgini 150 rows × 5 columns	147	6.5	3.0	5.2	2.0	virginica
150 rows × 5 columns	148	6.2	3.4	5.4	2.3	virginica
	149	5.9	3.0	5.1	1.8	virginica
¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬¬	150 ro	ws × 5 columns				
	₹]	-		•*		

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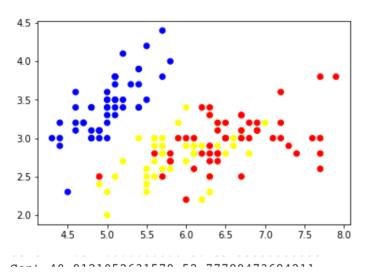
 sepal_length
 sepal_width
 petal_length
 petal_width

 150.000000
 150.000000
 150.000000
 150.000000

```
colors = {'virginica':'red', 'setosa':'blue','versicolor':'yellow'}
plt.scatter(dataEx4.iloc[:,0],dataEx4.iloc[:,1],c=dataEx4.iloc[:,4].map(colors))
```

plt.show()

count



cor = dataEx4.iloc[:,:].corr()
cor.style.background gradient(cmap='coolwarm')

sepal_length	sepal_width	petal_length	petal_width

sepal_length	1.000000	-0.109369	0.871754	0.817954
sepal_width	-0.109369	1.000000	-0.420516	-0.356544
petal_length	0.871754	-0.420516	1.000000	0.962757
petal_width	0.817954	-0.356544	0.962757	1.000000

Can+ 12 65/16666666666 37 50/10000000000

```
colors = {'virginica':'red', 'setosa':'blue','versicolor':'yellow'}
plt.scatter(dataEx4.iloc[:,2],dataEx4.iloc[:,3],c=dataEx4.iloc[:,4].map(colors))
```

```
plt.show
     2.0
     1.5
     1.0
    Cent 42.00410000000000 3/.0747777777777
5.TAK
```

test=dataEx4.copy()

CETTC 47.0041000000000 21.074777777777

test.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 150 entries, 0 to 149 Data columns (total 5 columns):

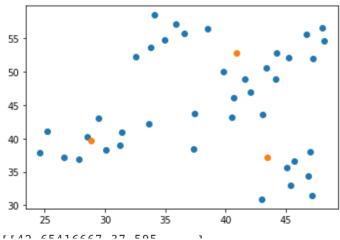
Data	cordina (cocar		rumins).	
#	Column	Non-N	Mull Count	Dtype
0	sepal_length	150 n	on-null	float64
1	sepal_width	150 n	on-null	float64
2	petal_length	150 n	on-null	float64
3	petal_width	150 n	on-null	float64
4	species	150 n	on-null	object
<pre>dtypes: float64(4), object(1)</pre>				
memory usage: 6.0+ KB				

CCIIC 40.71210320313/7 32.///074/3004211

trainingData=pd.DataFrame(data=dataEx4,columns=['sepal_length','sepal_width',],copy=True) CEHL 40.71210320313/7 32.///074/3004211

```
k = 3
model2 = AgglomerativeClustering(linkage='average', n clusters=k, affinity='euclidean', distance threshold=None)
model2.fit(trainingData)
   AgglomerativeClustering(linkage='average', n clusters=3)
model2.labels
   0, 0, 0, 0, 0, 0, 2, 2, 2, 2, 2, 2, 2, 0, 2, 0, 0, 2, 2, 2, 2, 2,
        2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 0, 2, 2, 2, 1, 2, 2, 1, 0, 1, 2, 1,
        2, 2, 2, 2, 2, 2, 2, 1, 1, 2, 2, 2, 1, 2, 2, 1, 2, 2, 1, 1, 1, 1,
        2, 2, 2, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2)
k = 3
model2 = KMeans(n clusters=3).fit(trainingData)
model2.labels
   0, 0, 0, 0, 0, 0, 1, 1, 1, 2, 1, 2, 1, 2, 1, 2, 2, 2, 2, 2, 2, 1,
        2, 2, 2, 2, 2, 2, 2, 2, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 1, 2,
        2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 1, 2, 1, 1, 1, 1, 2, 1, 1, 1,
        1, 1, 1, 2, 2, 1, 1, 1, 1, 2, 1, 2, 1, 2, 1, 1, 2, 2, 1, 1, 1, 1,
        1, 2, 2, 1, 1, 1, 2, 1, 1, 1, 2, 1, 1, 1, 2, 1, 1, 2], dtype=int32)
   29.49 43.05 2.0
   30.04 38.33 2.0
   31.25 39.03 2.0
   31.35 40.98 2.0
   42.95 30.91 0.0
   37.3 38.42 0.0
   37.39 43.69 0.0
   32.53 52.18 1.0
   33.6 42.2 2.0
   24.6 37.88 2.0
   33.76 53.6 1.0
   34.97 54.72 1.0
   35.84 57.04 1.0
   47.16 31.52 0.0
```

36.58 55.76 1.0 46.82 34.33 0.0 46.98 38.03 0.0 45.34 33.02 0.0 45.7 36.63 0.0 48.12 54.53 1.0 47.25 51.99 1.0 38.48 56.41 1.0 34.1 58.45 1.0 39.81 49.98 1.0 40.47 43.18 0.0 40.66 46.02 1.0 41.59 48.86 1.0 42.05 46.89 1.0 43.04 43.52 0.0 43.34 50.49 1.0 44.17 48.94 1.0 44.23 52.81 1.0 45.19 52.05 1.0 46.68 55.59 1.0 47.98 56.47 1.0 43.47727272727272 37.17636363636364 40.9121052631579 52.77789473684211 28.851000000000006 39.69099999999995 478.25000000000006 408.94 11 777.33 1002.779999999999 19 288.5100000000005 396.91 10



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- 26.59 37.21 2.0
- 27.8 36.93 2.0
- 28.56 40.21 2.0
- 29.49 43.05 2.0
- 30.04 38.33 2.0

31.25 39.03 2.0 31.35 40.98 2.0 42.95 30.91 0.0 37.3 38.42 0.0 37.39 43.69 0.0 32.53 52.18 1.0 33.6 42.2 2.0 24.6 37.88 2.0 33.76 53.6 1.0 34.97 54.72 1.0 35.84 57.04 1.0 47.16 31.52 0.0 36.58 55.76 1.0 46.82 34.33 0.0 46.98 38.03 0.0 45.34 33.02 0.0 45.7 36.63 0.0 48.12 54.53 1.0 47.25 51.99 1.0 38.48 56.41 1.0 34.1 58.45 1.0 39.81 49.98 1.0 40.47 43.18 0.0 40.66 46.02 1.0 41.59 48.86 1.0 42.05 46.89 1.0 43.04 43.52 0.0 43.34 50.49 1.0 44.17 48.94 1.0 44.23 52.81 1.0 45.19 52.05 1.0 46.68 55.59 1.0 47.98 56.47 1.0 43.47727272727272 37.17636363636364 40.9121052631579 52.77789473684211 28.851000000000006 39.69099999999999 478.25000000000006 408.94 11 777.33 1002.779999999999 19 288.5100000000005 396.91 10



