

```
elif (i == '0' and y_pred[i] == '1'):
```

```
    FN += 1
```

```
    j += 1
```

```
Confusion-matrix = [TP, TN, FP, FN]
```

```
print('Confusion-matrix', Confusion-matrix)
```

```
ACC = (TP + TN) / (TP + TN + FP + FN)
```

```
print("Accuracy: ", ACC)
```

```
PREC = (TP) / (TP + FP)
```

```
print("Precision: ", PREC)
```

```
REC = TP / (TP + FN)
```

```
print("Recall: ", REC)
```

```
SN = TP / (TP + FN)
```

```
print("Sensitivity: ", SN)
```

```
SP = TN / (TN + FP)
```

```
print("Specificity: ", SP)
```

```
print("Misclassification Error: ", MCE)
```

```
MCE = 1 - ACC
```

Result:

The above program evaluating the results of machine learning is executed successfully.

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22/3/22

```
[2] df = pd.DataFrame()
df['refund'] = ['yes', 'no', 'no', 'yes', 'no', 'no', 'yes', 'no', 'no', 'no']
df['marital_status'] = ['single', 'married', 'single', 'married', 'divorced', 'married', 'divorced', 'single', 'married', 'single']
df['taxable_income'] = [125000, 100000, 70000, 120000, 95000, 60000, 220000, 85000, 75000, 90000]
df['evade'] = ['no', 'no', 'no', 'no', 'yes', 'no', 'no', 'yes', 'no', 'yes']
df
```

	refund	marital_status	taxable_income	evade
0	yes	single	125000	no
1	no	married	100000	no
2	no	single	70000	no
3	yes	married	120000	no
4	no	divorced	95000	yes
5	no	married	60000	no
6	yes	divorced	220000	no
7	no	single	85000	yes
8	no	married	75000	no
9	no	single	90000	yes

```
for i in range(len(df)):
    df.loc[i, 'taxable_income'] = str(ceil(df.loc[i, 'taxable_income']/100000))
df
```

	refund	marital_status	taxable_income	evade
0	yes	single	2	no
1	no	married	1	no
2	no	single	1	no
3	yes	married	2	no
4	no	divorced	1	yes
5	no	married	1	no
6	yes	divorced	3	no
7	no	single	1	yes
8	no	married	1	no
9	no	single	1	yes

Aim:

To write a python program implementing classification algorithm

Algorithm:

step 1: start

step 2: import the necessary packages

step 3: Calculate the dataframe using df define the dataframe
refund, flexible income, evade.

step 4: Using for loop find the length of df

step 5: Using ceil function calculate 'flexable-income'

step 6: Calculate the strna, strnb, prby (1, prbn[1])

step 7: Calculate the dummy values in df

step 8: print the evade of x

step 9: print the prb + na(data)

step 10: stop

Program:

```
import pandas as pd
```

```
import numpy as np
```

```
from math import *
```

```
df = pd.DataFrame()
```

```
df['refund'] = ['yes', 'no', 'no', 'yes', 'no', 'no', 'yes', 'no', 'no', 'no']
```

```
df['marital_status'] = ['single', 'married', 'single', 'married', 'divorced',  
                        'married', 'divorced', 'single', 'married', 'single']
```

```
df['flexable_income'] = [125000, 100000, 70000, 120000, 95000, 60000,  
                        220000, 85000, 75000, 90000]
```

```
df['evade'] = ['no', 'no', 'no', 'no', 'yes', 'no', 'yes', 'no', 'yes']
```

```
df
```

```
data = pd.get_dummies(df[df.columns])
data
```

	refund_no	refund_yes	marital_status_divorced	marital_status_married	marital_status_single	taxable_income_1	taxable_income_2	taxable_income_3	evade_no	evade_yes
0	0	1	0	1	0	0	1	0	1	0
1	1	0	0	0	1	1	0	0	1	0
2	1	0	0	1	0	0	1	0	1	0
3	0	1	0	0	0	1	0	0	0	1
4	1	0	1	1	0	1	0	0	1	0
5	1	0	0	0	0	0	0	1	1	0
6	0	1	1	0	1	1	0	0	0	1
7	1	0	0	1	0	1	0	0	1	0
8	1	0	0	0	1	1	0	0	0	1
9	1	0	0	0	0	1	1	0	0	1

```
x=['no','married',140000]
x[2]=str(ceil(x[2]/100000))
x
```

```
['no', 'married', '2']
```

```
print('no : ',prb*pa(data,'evade','no'))
print('Evade of X is No')
```

```
yes : 0.0
no : 0.08163265306122447
Evade of X is No
```



```

for i in range(len(df)):
    df.loc[i, 'taxable-income'] = str(ceil(df.loc[i, 'taxable-income'] / 100000))
df
data = pd.get_dummies(df[df.columns])
data
for i in range(1, 4):
    if ('taxable-income' + str(i) not in data.columns):
        data['taxable-income' + str(i)] = [0 in range(10)]
x = ['no', 'married', 14000]
x[2] = str(ceil(x[2] / 100000))
x
def pa(data, cls, sbcls):
    strn = str(cls) + '-' + str(sbcls)
    return (sum(data[strn]) / len(strn))
def pa-b(data, clsa, sbcls, clsb, sbclsb):
    strna = str(clsa) + '-' + str(sbcls)
    strnb = str(clsb) + '-' + str(sbclsb)
    anb = 0
    for i in range(len(data)):
        if (data.loc[i, strna] & data.loc[i, strnb]):
            anb += 1
    return (anb / sum(data[strnb]))
prby = []
prbn = []
Col = df.columns
for i in range(len(x)):
    prby.append(pa-b(data, Col[i], x[i], 'evade', 'yes'))
for i in prby:
    prb *= i

```

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```
print("Yes: ", prb * pa(data, 'evade', 'Yes'))  
for i in range(len(z)):  
    prb.append(pa_b(data, col[i], x[i], 'evade', 'no'))  
prb = 1  
for i in prbn:  
    prb *= i  
print("no: ", prb * pa(data, 'evade', 'no'))  
print('Evade of x is No')
```

Result:

The above program of Implementing Classification Algorithm is successfully executed.

Yash 22/3/22

```

[ ] y=[]
    for i in p1:
        x.append(i[0])
        y.append(i[1])
    plt.scatter(x,y,c='b',label='Cluster 1')
x=[]
y=[]
for i in p2:
    x.append(i[0])
    y.append(i[1])
plt.scatter(x,y,c='r',label='Cluster 2')
plt.legend()
plt.show()

```

After Epoch 1 ERROR: 7.5

After Epoch 2 ERROR: 2.6666666666666667

Data Objects:

('x1', [1, 0])

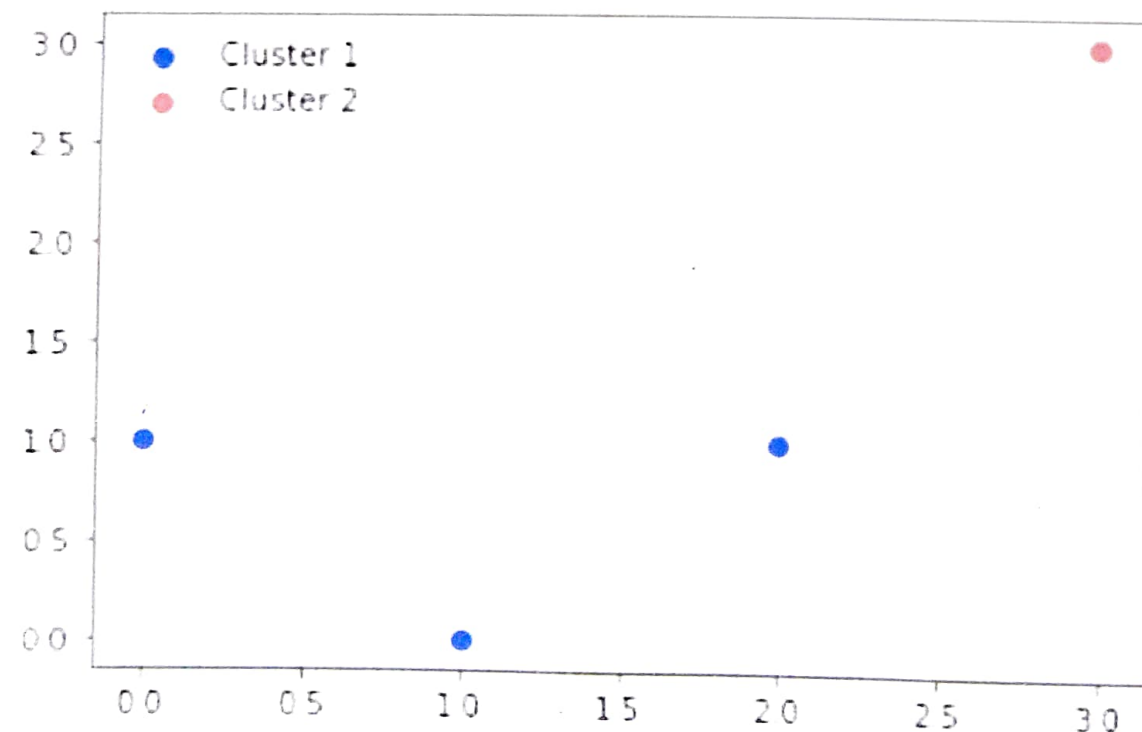
('x2', [0, 1])

('x3', [2, 1])

('x4', [3, 3])

Cluster 1: ['x1', 'x2', 'x3']

Cluster 2: ['x4']



Aim:

To write a python program implementing k-means algorithm

Algorithm:

- step 1: start
- step 2: select the number k to decide the number of cluster
- step 3: select random k points or centroid
- step 4: Calculate the variance and place a new centroid of each cluster
- step 5: Assign each data point to their closest centroid will assign k cluster
- step 6: Repeat step 3 untill it nears to centroid of cluster
- step 7: if any reassignment occurs, then go to step 4, goto finish
- step 8: stop

Program:

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

```
class k-meansalg:
```

```
    def __init__(self):
```

```
        self.m1 = None
```

```
        self.m2 = None
```

```
        self.c1, self.c2, self.nc1, self.nc2 = [], [], [], []
```

```
        self.pts = {}
```

```
        self.err = 0
```

```
    def error(self):
```

```
        e1 = 0
```

```
        e2 = 0
```

```
        for i in self.c1:
```



```
e1 += (i[0] - self.m1[0])**2 + (i[1] - self.m1[1])**2)
for i in self.c2:
    e2 += (i[0] - self.m2[0])**2 + (i[1] - self.m2[1])**2)
return (e1 + e2)

def fitting(self, c1, c2, *args):
    def mean(c1):
        n = len(c1)
        a = 0
        b = 0
        for i in c1:
            a += i[0]/n
            b += i[1]/n
        return [a, b]

    def dist(a, b):
        return ((a[0] - b[0])**2 + (a[1] - b[1])**2)**(1/2)

    self.c1 = c1
    self.c2 = c2
    con = 1
    for i in args:
        self.pts['x' + str(con)] = i
        con += 1

    ep0 = 1
    while (True):
        self.m1 = mean(self.c1)
        self.m2 = mean(self.c2)
        for i in args:
            p = dist(i, self.m1)
            q = dist(i, self.m2)
            if (p > q):
```



```
        self.nc2.append(i)
    else:
        self.nc1.append(i)
    print('After Epoch', epo, 'Error: ', self.Error())
    epo += 1
    if (self.c1 == self.nc1 and self.c2 == self.nc2):
        break;
    else:
        self.c1, self.c2 = self.nc1.copy(), self.nc2.copy()
        self.nc1, self.nc2 = [], []
    for i, j in sorted(self.pts.items()):
        for k in range(len(self.c1)):
            if (j == self.c1[k]):
                self.c1[k] = i
        for l in range(len(self.c2)):
            if (j == self.c2[l]):
                self.c2[l] = i
    def display(self):
        print("Cluster 1: ", self.c1, "\nCluster 2: ", self.c2)

x1 = [1, 0]
x2 = [0, 1]
x3 = [2, 1]
x4 = [3, 3]
c1 = [x1, x3]
c2 = [x2, x4]
model = k_means()
model.fitting(c1, c2, x1, x2, x3, x4)
print("Data Objects: ")
```



```
for i in sorted(model.pts.items()):  
    print(i)  
model.display()
```

```
p1 = [model.pts[i] for i in model.c1]  
p2 = [model.pts[i] for i in model.c2]  
x = []
```

```
y = []
```

```
for i in p1:  
    x.append(i[0])  
    y.append(i[1])
```

```
plt.scatter(x, y, c='b', label='cluster 1')  
x = []
```

```
y = []
```

```
for i in p2:  
    x.append(i[0])  
    y.append(i[1])
```

```
plt.scatter(x, y, c='r', label='cluster 2')
```

```
plt.legend()
```

```
plt.show()
```

Result:

The above program implementing of k-means algorithm is executed successfully.