

Question 1

```
In [457]: import numpy as np  
import pandas as pd
```

```
In [458]: data=pd.read_csv("C:/Users/91888/Desktop/Assignment/PerformanceMetric Assignment/5_a.csv")
```

```
In [459]: data.head()
```

```
Out[459]:
```

	y	proba
0	1.0	0.637387
1	1.0	0.635165
2	1.0	0.766586
3	1.0	0.724564
4	1.0	0.889199

```
In [460]: data.shape
```

```
Out[460]: (10100, 2)
```

```
In [461]: def predict(data,proba,y_score):  
    y_pred=[]  
    for i in data['proba']:  
        if i<y_score:  
            y_pred.append(0)  
        else:  
            y_pred.append(1)  
    return y_pred
```

```
In [462]: y_score=0.5
data['y_pred']=predict(data,'proba',y_score)
```

```
In [463]: data.head()
```

```
Out[463]:
```

	y	proba	y_pred
0	1.0	0.637387	1
1	1.0	0.635165	1
2	1.0	0.766586	1
3	1.0	0.724564	1
4	1.0	0.889199	1

```
In [464]: def confusion_matrix(y_pred,y):

    fp = np.sum((y_pred == 1) & (y == 0))
    tp = np.sum((y_pred == 1) & (y == 1))

    fn = np.sum((y_pred == 0) & (y == 1))
    tn = np.sum((y_pred == 0) & (y == 0))
    return {'tn':tn,'tp':tp,'fn':fn,'fp':fp}
cm=confusion_matrix(data.y_pred,data.y)
cm
```

```
Out[464]: {'tn': 0, 'tp': 10000, 'fn': 0, 'fp': 100}
```

```
In [465]: def f1_score(TP,FP):
    P=(data['y']== 1).sum()
    precision=TP/(TP+FP)
    Recall=TP/P
    f1=2*(precision*Recall/(precision+Recall))
    return f1
f1=f1_score(cm['tp'],cm['fp'])
```

```
In [466]: print("f1 score : ",f1)
```

```
f1 score : 0.9950248756218906
```

```
In [467]: def AUC_score(data):  
    P=(data['y']== 1).sum()  
    N=(data['y']== 0).sum()  
    tpr=[]  
    fpr=[]  
    thresholds = np.arange(1, 0, -0.1)  
    for y_score in thresholds:  
        y_score = round(y_score,2)  
        data['y_pred']=predict(data,'proba',y_score)  
        cm=confusion_matrix(data.y_pred,data.y)  
        tpr.append(cm['tp']/P)  
        fpr.append(cm['fp']/N)  
        data.drop(columns=['y_pred'])  
    return np.trapz(tpr,fpr)  
AUC=AUC_score(data)
```

```
In [468]: print("AUC Score is : ",AUC)
```

```
AUC Score is : 0.48897750000000006
```

```
In [469]: def Accuracy_score(data,tp,tn):  
    total_pts=data['y'].count()  
    acc=(tp+tn)/total_pts  
    return acc  
Accuracy=Accuracy_score(data,cm['tp'],cm['tn'])
```

```
In [470]: print("Accuracy is : ", Accuracy)
```

```
Accuracy is : 0.9900990099009901
```

Question 2

```
In [176]: data=pd.read_csv("C:/Users/91888/Desktop/Assignment/PerformanceMetric Assignment/5_b.csv")
```

```
In [177]: data.shape
```

```
Out[177]: (10100, 2)
```

```
In [178]: data.head()
```

```
Out[178]:
```

	y	proba
0	0.0	0.281035
1	0.0	0.465152
2	0.0	0.352793
3	0.0	0.157818
4	0.0	0.276648

```
In [179]: def predict(data,proba,y_score):  
    y_pred=[]  
    for i in data['proba']:  
        if i<y_score:  
            y_pred.append(0)  
        else:  
            y_pred.append(1)  
    return y_pred
```

```
In [180]: y_score=0.5  
data['y_pred']=predict(data,'proba',y_score)
```

In [95]: data.head()

Out[95]:

	y	proba	y_pred
0	0.0	0.281035	0
1	0.0	0.465152	0
2	0.0	0.352793	0
3	0.0	0.157818	0
4	0.0	0.276648	0

```
In [181]: def confusion_matrix(y_pred,y):

    fp = np.sum((y_pred == 1) & (y == 0))
    tp = np.sum((y_pred == 1) & (y == 1))

    fn = np.sum((y_pred == 0) & (y == 1))
    tn = np.sum((y_pred == 0) & (y == 0))
    return {'tn':tn,'tp':tp,'fn':fn,'fp':fp}
cm=confusion_matrix(data.y_pred,data.y)
cm
```

Out[181]: {'tn': 9761, 'tp': 55, 'fn': 45, 'fp': 239}

```
In [182]: def f1_score(TP,FP):
    P=(data['y']== 1).sum()
    precision=TP/(TP+FP)
    Recall=TP/P
    f1=2*(precision*Recall/(precision+Recall))
    return f1
f1=f1_score(cm['tp'],cm['fp'])
print("f1 score : ",f1)
```

f1 score : 0.2791878172588833

```
In [183]: def AUC_score(data):  
    P=(data['y']== 1).sum()  
    N=(data['y']== 0).sum()  
    tpr=[]  
    fpr=[]  
    thresholds = np.arange(1,0,-0.1)  
    for y_score in thresholds:  
        y_score = round(y_score,2)  
        data['y_pred']=predict(data,'proba',y_score)  
        cm=confusion_matrix(data.y_pred,data.y)  
        tpr.append(cm['tp']/P)  
        fpr.append(cm['fp']/N)  
        data.drop(columns=['y_pred'])  
    return np.trapz(tpr,fpr)  
AUC=AUC_score(data)
```

```
In [184]: print("AUC score is : ", AUC)
```

AUC score is : 0.9276825

```
In [185]: def Accuracy_score(data,tp,tn):  
    total_pts=data['y'].count()  
    acc=(tp+tn)/total_pts  
    return acc  
Accuracy=Accuracy_score(data,cm['tp'],cm['tn'])
```

```
In [186]: print("Accuracy is : ",Accuracy)
```

Accuracy is : 0.9718811881188119

Question 3

```
In [471]: data=pd.read_csv("C:/Users/91888/Desktop/Assignment/PerformanceMetric Assignment/5_c.csv")
```

In [136]: `data.shape`

Out[136]: (2852, 2)

In [188]: `data.head()`

Out[188]:

	y	prob
0	0	0.458521
1	0	0.505037
2	0	0.418652
3	0	0.412057
4	0	0.375579

```
In [472]: def predict(data,proba,y_score):  
            y_pred=[]  
            for i in data['prob']:  
                if i<y_score:  
                    y_pred.append(0)  
                else:  
                    y_pred.append(1)  
            return y_pred  
data['y_pred']=predict(data,'prob',y_score)
```

```
In [473]: def confusion_matrix(y_pred,y):  
  
            fp = np.sum((y_pred == 1) & (y == 0))  
            tp = np.sum((y_pred == 1) & (y == 1))  
  
            fn = np.sum((y_pred == 0) & (y == 1))  
            tn = np.sum((y_pred == 0) & (y == 0))  
            return {'tn':tn,'tp':tp,'fn':fn,'fp':fp}
```

```
In [474]: def metric_value(data):  
          A={}  
          thresholds = np.arange(1,0,-0.01)  
          for y_score in thresholds:  
              y_score = round(y_score,2)  
              data['y_pred']=predict(data,'prob',y_score)  
              cm=confusion_matrix(data.y_pred,data.y)  
              metric_val=(500*cm['fn'])+(100*cm['fp'])  
              A[y_score]=metric_val  
  
              data.drop(columns=['y_pred'])  
          return A  
metric=metric_value(data)
```

```
In [475]: minv = min(metric.values())  
          key = [k for k, v in metric.items() if v==minv]
```

```
In [476]: print("Threshold giving lowest values of metric A",key)  
          print("Minimum value of metric A is ",minv)
```

```
Threshold giving lowest values of metric A [0.23]  
Minimum value of metric A is 141000
```

Question 4

```
In [477]: data=pd.read_csv("C:/Users/91888/Desktop/Assignment/PerformanceMetric Assignment/5_d.csv")
```

```
In [478]: data.shape
```

```
Out[478]: (157200, 2)
```


In [340]: data.head()

Out[340]:

	y	pred
0	101.0	100.0
1	120.0	100.0
2	131.0	113.0
3	164.0	125.0
4	154.0	152.0

```
In [479]: def error(y,pred):  
            val=[]  
            for i in range(len(y)):  
                e=y[i]-pred[i]  
                val.append(e)  
            return val  
err=error(data['y'],data['pred'])
```

In [480]: data['error']=err

```
In [481]: #converting errors to absolute errors  
A=data.error.abs()  
data['abs2']=A
```

In [482]: data.head()

Out[482]:

	y	pred	error	abs2
0	101.0	100.0	1.0	1.0
1	120.0	100.0	20.0	20.0
2	131.0	113.0	18.0	18.0
3	164.0	125.0	39.0	39.0
4	154.0	152.0	2.0	2.0

```
In [483]: def MSE(error):  
            sum=0  
            for i in range(len(error)):  
                sum=sum+(np.square(error[i]))  
            mse=sum/len(error)  
  
            return mse  
MSE=MSE(data['error'])
```

```
In [484]: print("Mean Square Error is : ",MSE)
```

Mean Square Error is : 177.16569974554707

```
In [485]: def MAPE(absolute,y):  
            mape=np.sum(data.abs2)/np.sum(data.y)  
            return mape
```

```
In [486]: M=MAPE(data.abs2,data.y)  
print("MAPE is : ",M)
```

MAPE is : 0.1291202994009687

```
In [487]: def st(y):  
          s=0  
          mean=data.y.mean()  
          for i in range(len(y)):  
              s=s+(y[i]-mean)*(y[i]-mean)  
  
          return s  
stotal=st(data.y)
```

```
In [488]: def s_res(error):  
          sum1=0  
          for i in range(len(error)):  
              sum1=sum1+(np.square(error[i]))  
  
          return sum1  
s_res=s_res(data['error'])
```

```
In [489]: RSQUARE=1-(s_res/stotal)
```

```
In [490]: print(" R square error is : ",RSQUARE)  
  
R square error is :  0.9563582786990964
```