

Assignment 1

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Roll Number: 21CS30020

```
# import all the necessary libraries here
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from numpy.linalg import inv
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
```

```
df =
pd.read_excel('../..../dataset/logistic-regression/Pumpkin_Seeds_Dataset
.xlsx')
print(df.shape)
```

```
(2500, 13)
```

```
df.head()
```

	Area	Perimeter	Major_Axis_Length	Minor_Axis_Length	Convex_Area
0	56276	888.242	326.1485	220.2388	56831
1	76631	1068.146	417.1932	234.2289	77280
2	71623	1082.987	435.8328	211.0457	72663
3	66458	992.051	381.5638	222.5322	67118
4	66107	998.146	383.8883	220.4545	67117

	Equiv_Diameter	Eccentricity	Solidity	Extent	Roundness
Aspect_Ration \					
0	267.6805	0.7376	0.9902	0.7453	0.8963
1.4809					
1	312.3614	0.8275	0.9916	0.7151	0.8440
1.7811					
2	301.9822	0.8749	0.9857	0.7400	0.7674
2.0651					
3	290.8899	0.8123	0.9902	0.7396	0.8486
1.7146					
4	290.1207	0.8187	0.9850	0.6752	0.8338
1.7413					

Compactness

Class

0	0.8207	Çerçevelik
1	0.7487	Çerçevelik
2	0.6929	Çerçevelik
3	0.7624	Çerçevelik
4	0.7557	Çerçevelik

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 2500 entries, 0 to 2499
```

```
Data columns (total 13 columns):
```

#	Column	Non-Null Count	Dtype
0	Area	2500 non-null	int64
1	Perimeter	2500 non-null	float64
2	Major_Axis_Length	2500 non-null	float64
3	Minor_Axis_Length	2500 non-null	float64
4	Convex_Area	2500 non-null	int64
5	Equiv_Diameter	2500 non-null	float64
6	Eccentricity	2500 non-null	float64
7	Solidity	2500 non-null	float64
8	Extent	2500 non-null	float64
9	Roundness	2500 non-null	float64
10	Aspect_Ration	2500 non-null	float64
11	Compactness	2500 non-null	float64
12	Class	2500 non-null	object

```
dtypes: float64(10), int64(2), object(1)
```

```
memory usage: 254.0+ KB
```

```
df.describe()
```

	Area	Perimeter	Major_Axis_Length
Minor_Axis_Length \			
count	2500.000000	2500.000000	2500.000000
mean	80658.220800	1130.279015	456.601840
std	13664.510228	109.256418	56.235704
min	47939.000000	868.485000	320.844600
25%	70765.000000	1048.829750	414.957850
50%	79076.000000	1123.672000	449.496600
75%	89757.500000	1203.340500	492.737650
max	136574.000000	1559.450000	661.911300

	Convex_Area	Equiv_Diameter	Eccentricity	Solidity
Extent \				
count	2500.000000	2500.000000	2500.000000	2500.000000
mean	81508.084400	319.334230	0.860879	0.989492
std	13764.092788	26.891920	0.045167	0.003494
min	48366.000000	247.058400	0.492100	0.918600
25%	71512.000000	300.167975	0.831700	0.988300
50%	79872.000000	317.305350	0.863700	0.990300
75%	90797.750000	338.057375	0.897025	0.991500
max	138384.000000	417.002900	0.948100	0.994400

	Roundness	Aspect_Ration	Compactness
count	2500.000000	2500.000000	2500.000000
mean	0.791533	2.041702	0.704121
std	0.055924	0.315997	0.053067
min	0.554600	1.148700	0.560800
25%	0.751900	1.801050	0.663475
50%	0.797750	1.984200	0.707700
75%	0.834325	2.262075	0.743500
max	0.939600	3.144400	0.904900

```
print(df["Class"].unique())
```

```
['Çerçvelik' 'Ürgüp Sivrisi']
```

```
mapping = {'Çerçvelik': 0 , 'Ürgüp Sivrisi' : 1}
df.replace({'Class': mapping} , inplace=True)
df.head()
```

	Area	Perimeter	Major_Axis_Length	Minor_Axis_Length	Convex_Area
0	56276	888.242	326.1485	220.2388	56831
1	76631	1068.146	417.1932	234.2289	77280
2	71623	1082.987	435.8328	211.0457	72663
3	66458	992.051	381.5638	222.5322	67118
4	66107	998.146	383.8883	220.4545	67117

Equiv_Diameter	Eccentricity	Solidity	Extent	Roundness
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Aspect_Ration \					
0	267.6805	0.7376	0.9902	0.7453	0.8963
1.4809					
1	312.3614	0.8275	0.9916	0.7151	0.8440
1.7811					
2	301.9822	0.8749	0.9857	0.7400	0.7674
2.0651					
3	290.8899	0.8123	0.9902	0.7396	0.8486
1.7146					
4	290.1207	0.8187	0.9850	0.6752	0.8338
1.7413					

	Compactness	Class
0	0.8207	0
1	0.7487	0
2	0.6929	0
3	0.7624	0
4	0.7557	0

```
X = df.iloc[:, :-1].values
Y = df.iloc[:, -1].values
X_train, X, Y_train, Y =
train_test_split(X, Y, test_size=0.5, random_state=0)
X_val, X_test, Y_val, Y_test = train_test_split(X, Y, test_size =
0.4, random_state = 0)
```

```
from sklearn.preprocessing import StandardScaler
st_x= StandardScaler()
X_train= st_x.fit_transform(X_train)
X_test= st_x.transform(X_test)
```

```
print(X_train.shape, Y_train.shape)
print(X_val.shape, Y_val.shape)
print(X_test.shape, Y_test.shape)
```

```
(1250, 12) (1250,)
(750, 12) (750,)
(500, 12) (500,)
```

```
class logistic_regression():
    def __init__(self, epoch= 15000, learning_rate = 0.001 ):
        self.epoch = epoch
        self.learning_rate = learning_rate
        self.cost = []
        self.init_weight = None
        self.final_weight = None

    def initialize_weight(self, n_feature):
        limit = np.sqrt(1/n_feature)
        weight = np.random.uniform(-limit, limit, (n_feature, 1))
```

```

b = 0
self.init_weight = np.insert(weight,0,b,axis = 0)

def train(self, X,Y,X_val,Y_val):
    n_sample ,n_feature = X.shape
    X = np.insert(X,0,1,axis = 1)
    Y = np.reshape(Y,(n_sample,1))
    nv_sample = X_val.shape[0];
    X_val = np.insert(X_val,0,1,axis = 1);
    Y_val = np.reshape(Y_val,(nv_sample,1));
    self.initialize_weight(n_feature)
    self.fit(X,Y,X_val,Y_val)

def fit(self,X,Y,X_val,Y_val):
    _weight = self.init_weight.copy()
    y_pred = self.sigmoid(np.dot(X,_weight))
    self.cost.append(self.gradient_cost(X,Y,_weight))

    for iter in range(self.epoch):
        y_pred = self.sigmoid(np.dot(X,_weight))
        grad = np.dot(X.T, y_pred - Y)
        _weight = _weight - self.learning_rate*grad
        self.cost.append(self.gradient_cost(X,Y,_weight))
        if iter%100 ==0:
            print(f"The training cost for iteration ::{iter} is
{np.squeeze(self.cost[
-1]))}", "\n.....")
        self.final_weight = _weight
    return

def predict(self,X):
    out = np.dot(X,self.final_weight)
    out = self.sigmoid(out)
    out = (out >= 0.5)*1
    return out

def sigmoid(self,Y):
    sig = 1 + np.exp(-1*Y)
    sig = 1/sig
    return sig

def gradient_cost(self,X,Y,_weight):
    y_pred = self.sigmoid(np.dot(X,_weight))
    return np.mean(-1*(Y*np.log(y_pred) + (1-Y)*np.log(1 -
y_pred)))

```

```

def visualize_loss(self):
    figure, ax = plt.subplots()
    nums = np.arange(len(self.cost))
    ax.plot(nums, np.array(self.cost).reshape((len(self.cost),)))
    ax.set_xlabel('Epoch')
    ax.set_ylabel('Cost')
    plt.show()

def metrics_loss(self,X,Y):
    n_sample,n_feature = X.shape
    X = np.insert(X,0,1,axis = 1)
    Y = np.reshape(Y,(n_sample,1))
    y_pred = self.sigmoid(np.dot(X,self.final_weight))
    y_pred = (y_pred >= 0.5)
    con_matrix = confusion_matrix(Y,y_pred)
    cm_display = ConfusionMatrixDisplay(confusion_matrix =
con_matrix, display_labels = [False, True])
    cm_display.plot()
    plt.show()
    recall = con_matrix[1][1]/(con_matrix[1][0] + con_matrix[1]
[1])
    precision = con_matrix[1][1] /(con_matrix[1][1] + con_matrix[0]
[1])
    accuracy = (con_matrix[0][0] + con_matrix[1]
[1])/(con_matrix[0][0] + con_matrix[0][1] + con_matrix[1][0] +
con_matrix[1][1])
    df = pd.DataFrame([[recall, precision, accuracy]],
columns=['Recall', 'Precision', 'Mean Accuracy'])
    return df

def print_loss(self):
    print(self.cost)

```

```
logistic_regressor = logistic_regression()
```

```
logistic_regressor.train(X_train,Y_train,X_val,Y_val)
```

```
The training cost for iteration ::0 is
_____0.33958563353374965
```

```
.....
The training cost for iteration ::100 is
_____0.3143659029205259
```

```
.....
The training cost for iteration ::200 is
_____0.31298912246293287
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```
.....
The training cost for iteration ::300 is
_____0.3117200765150373
```

```
.....
The training cost for iteration ::400 is
_____0.3105483882430749
.....
The training cost for iteration ::500 is
_____0.30946511667783855
.....
The training cost for iteration ::600 is
_____0.3084623582810819
.....
The training cost for iteration ::700 is
_____0.30753310998426475
.....
The training cost for iteration ::800 is
_____0.30667124084556024
.....
The training cost for iteration ::900 is
_____0.3058715389732966
.....
The training cost for iteration ::1000 is
_____0.30512982070815686
.....
The training cost for iteration ::1100 is
_____0.3044430559703882
.....
The training cost for iteration ::1200 is
_____0.3038093738368832
.....
The training cost for iteration ::1300 is
_____0.3032277022295546
.....
The training cost for iteration ::1400 is
_____0.3026968552707426
.....
The training cost for iteration ::1500 is
_____0.3022143535503374
.....
The training cost for iteration ::1600 is
_____0.3017758178904056
.....
The training cost for iteration ::1700 is
_____0.3013754791752596
.....
The training cost for iteration ::1800 is
_____0.3010073253986448
.....
The training cost for iteration ::1900 is
_____0.30066603092866834
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The training cost for iteration ::2000 is
_____0.300347326031182
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The training cost for iteration ::2100 is
_____0.30004795802148
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The training cost for iteration ::2200 is
_____0.299765488233756
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The training cost for iteration ::2300 is
_____0.29949807315037835
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The training cost for iteration ::2400 is
_____0.29924428565791966
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The training cost for iteration ::2500 is
_____0.2990029852625272
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The training cost for iteration ::2600 is
_____0.298773229330753
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The training cost for iteration ::2700 is
_____0.2985542142813959
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The training cost for iteration ::2800 is
_____0.29834523719020983
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The training cost for iteration ::2900 is
_____0.2981456707769319
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The training cost for iteration ::3000 is
_____0.29795494696814223
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The training cost for iteration ::3100 is
_____0.2977725458877133
.....
The training cost for iteration ::3200 is
_____0.2975979882664914
.....
The training cost for iteration ::3300 is
_____0.29743083001129883
.....
The training cost for iteration ::3400 is
_____0.2972706581509636
.....
The training cost for iteration ::3500 is
_____0.2971170876761652
.....
The training cost for iteration ::3600 is
```


	0.2969697589748651
.....	
The training cost for iteration ::3700 is	0.29682833567848427
.....	
The training cost for iteration ::3800 is	0.2966925028032118
.....	
The training cost for iteration ::3900 is	0.296561965112941
.....	
The training cost for iteration ::4000 is	0.29643644565600474
.....	
The training cost for iteration ::4100 is	0.29631568444361
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The training cost for iteration ::4200 is	0.29619943724755177
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The training cost for iteration ::4300 is	0.29608747450089967
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The training cost for iteration ::4400 is	0.29597958028922067
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The training cost for iteration ::4500 is	0.295875551422466
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The training cost for iteration ::4600 is	0.2957751965794106
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The training cost for iteration ::4700 is	0.2956783355177691
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The training cost for iteration ::4800 is	0.2955847983440341
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The training cost for iteration ::4900 is	0.29549442483781174
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The training cost for iteration ::5000 is	0.2954070638260047
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The training cost for iteration ::5100 is	0.2953225726026603
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The training cost for iteration ::5200 is	0.295240816390739

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The training cost for iteration ::5300 is
_____0.29516166784239484
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The training cost for iteration ::5400 is
_____0.2950850065746761
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The training cost for iteration ::5500 is
_____0.2950107187378407
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The training cost for iteration ::5600 is
_____0.29493869661372574
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The training cost for iteration ::5700 is
_____0.2948688382418357
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The training cost for iteration ::5800 is
_____0.29480104707101035
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The training cost for iteration ::5900 is
_____0.2947352316347255
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The training cost for iteration ::6000 is
_____0.29467130524823393
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The training cost for iteration ::6100 is
_____0.2946091857259151
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The training cost for iteration ::6200 is
_____0.29454879511732585
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The training cost for iteration ::6300 is
_____0.2944900594605785
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The training cost for iteration ::6400 is
_____0.29443290855177545
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The training cost for iteration ::6500 is
_____0.294377275729341
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The training cost for iteration ::6600 is
_____0.29432309767217923
.....
The training cost for iteration ::6700 is
_____0.294270314210668
.....
The training cost for iteration ::6800 is
_____0.2942188681495869
.....
```

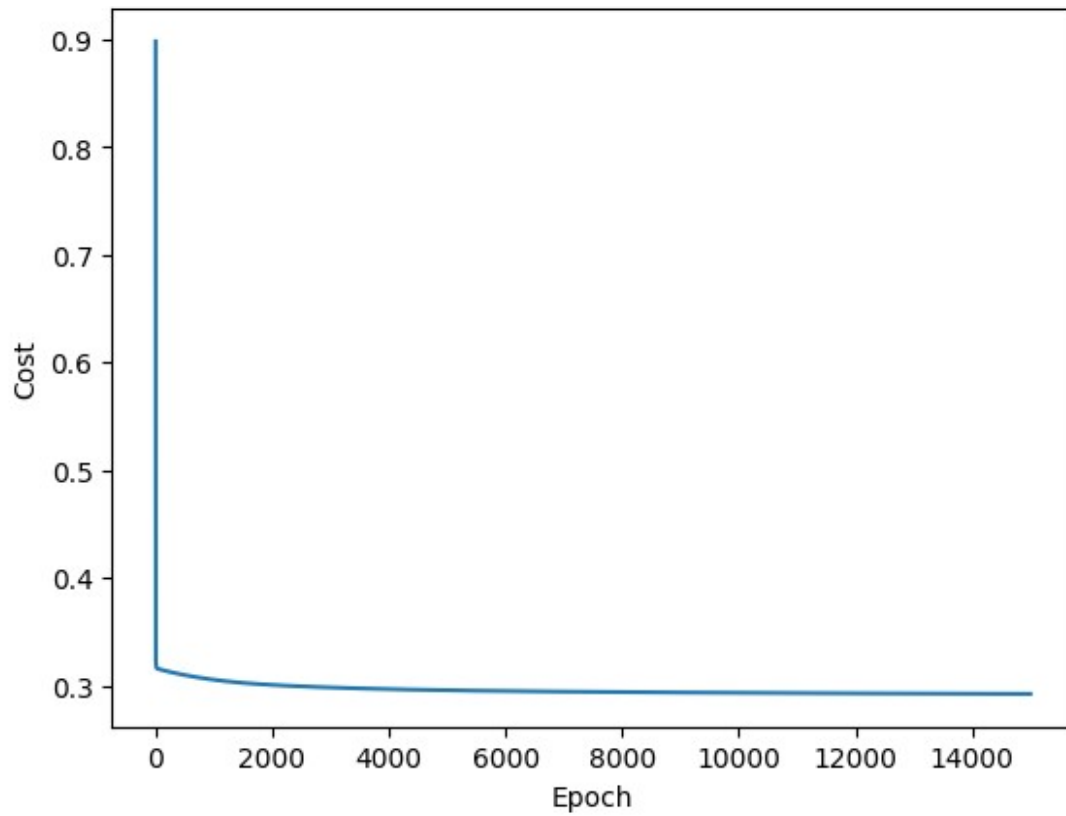
```
The training cost for iteration ::6900 is
_____0.29416870510213905
.....
The training cost for iteration ::7000 is
_____0.2941197733342938
.....
The training cost for iteration ::7100 is
_____0.29407202361874263
.....
The training cost for iteration ::7200 is
_____0.2940254090978021
.....
The training cost for iteration ::7300 is
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The training cost for iteration ::7400 is
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The training cost for iteration ::7500 is
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The training cost for iteration ::7600 is
_____0.29384944174697386
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The training cost for iteration ::7700 is
_____0.29380787467947783
.....
The training cost for iteration ::7800 is
_____0.29376720472897894
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The training cost for iteration ::7900 is
_____0.2937273984213862
.....
The training cost for iteration ::8000 is
_____0.29368842381340626
.....
The training cost for iteration ::8100 is
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The training cost for iteration ::8200 is
_____0.293612849106746
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The training cost for iteration ::8300 is
_____0.2935761920855168
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The training cost for iteration ::8400 is
_____0.2935402527820469
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The training cost for iteration ::8500 is
```

	0.2935050058063021
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The training cost for iteration ::8600 is	0.29347042688671754
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The training cost for iteration ::8700 is	0.29343649281433154
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The training cost for iteration ::9000 is	0.2933383424414113
.....	
The training cost for iteration ::9100 is	0.2933067751324231
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The training cost for iteration ::9200 is	0.2932757508152218
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The training cost for iteration ::9300 is	0.293245251643514
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The training cost for iteration ::9400 is	0.29321526051923247
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The training cost for iteration ::9500 is	0.2931857610567382
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The training cost for iteration ::9600 is	0.29315673754891314
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The training cost for iteration ::9700 is	0.29312817493503435
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The training cost for iteration ::9800 is	0.29310005877032264
.....	
The training cost for iteration ::9900 is	0.29307237519707824
.....	
The training cost for iteration ::10000 is	0.2930451109173031
.....	
The training cost for iteration ::10100 is	0.29301825316673746

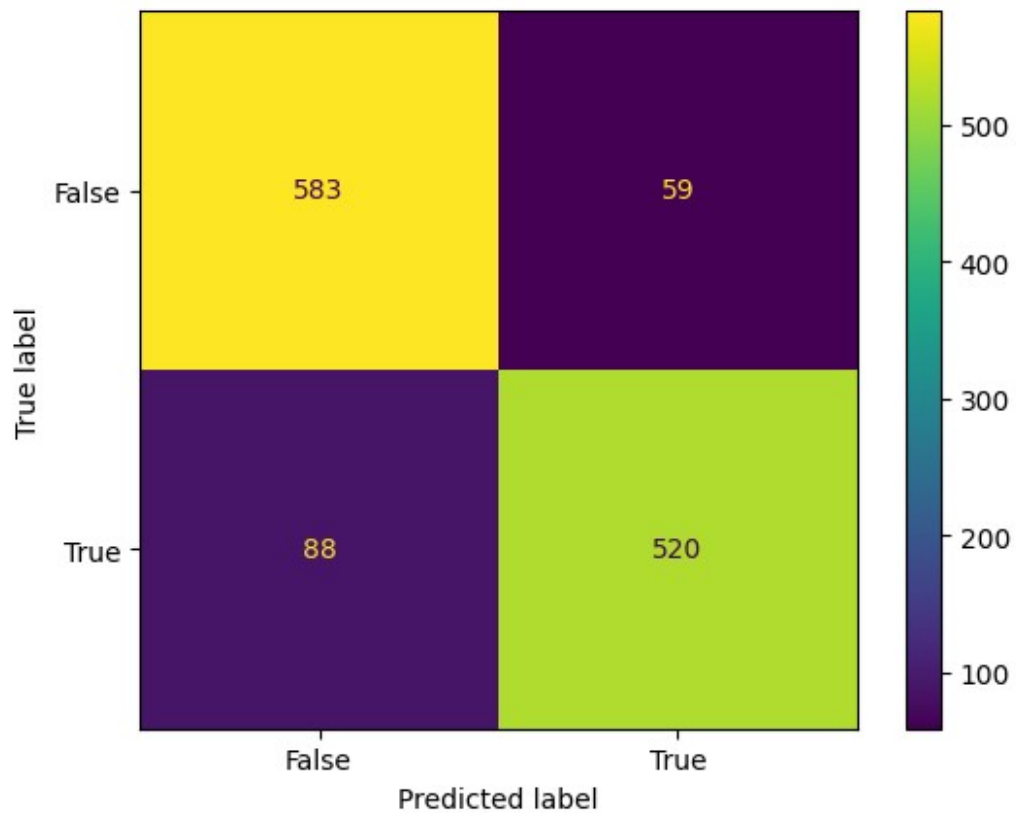
```
.....
The training cost for iteration ::10200 is
_____0.2929917896902186
.....
The training cost for iteration ::10300 is
_____0.29296570871829963
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The training cost for iteration ::10400 is
_____0.2929399989450493
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The training cost for iteration ::10500 is
_____0.29291464950697427
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The training cost for iteration ::10600 is
_____0.29288964996299516
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The training cost for iteration ::10700 is
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The training cost for iteration ::10800 is
_____0.29284066079190163
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The training cost for iteration ::10900 is
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The training cost for iteration ::11000 is
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The training cost for iteration ::11100 is
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The training cost for iteration ::11200 is
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The training cost for iteration ::11300 is
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The training cost for iteration ::11400 is
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The training cost for iteration ::11500 is
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The training cost for iteration ::11600 is
_____0.2926568682788987
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The training cost for iteration ::11700 is
_____0.29263513197774627
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The training cost for iteration ::11800 is
_____0.292613647154614
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The training cost for iteration ::11900 is
_____0.29259240742726295
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The training cost for iteration ::12000 is
_____0.29257140663964065
.....
The training cost for iteration ::12100 is
_____0.2925506388523657
.....
The training cost for iteration ::12200 is
_____0.29253009833365834
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The training cost for iteration ::12300 is
_____0.2925097795506923
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The training cost for iteration ::12400 is
_____0.29248967716134705
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The training cost for iteration ::12500 is
_____0.29246978600634
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The training cost for iteration ::12600 is
_____0.2924501011017225
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The training cost for iteration ::12700 is
_____0.2924306176317169
.....
The training cost for iteration ::12800 is
_____0.2924113309418809
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The training cost for iteration ::12900 is
_____0.2923922365325834
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The training cost for iteration ::13000 is
_____0.29237333005277316
.....
The training cost for iteration ::13100 is
_____0.29235460729402946
.....
The training cost for iteration ::13200 is
_____0.2923360641848809
.....
The training cost for iteration ::13300 is
_____0.2923176967853741
.....
The training cost for iteration ::13400 is
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```
0.29229950128188786
.....
The training cost for iteration ::13500 is
0.29228147398217613
.....
The training cost for iteration ::13600 is
0.29226361131062856
.....
The training cost for iteration ::13700 is
0.292245909803741
.....
The training cost for iteration ::13800 is
0.29222836610578373
.....
The training cost for iteration ::13900 is
0.2922109769646587
.....
The training cost for iteration ::14000 is
0.2921937392279365
.....
The training cost for iteration ::14100 is
0.29217664983906677
.....
The training cost for iteration ::14200 is
0.2921597058337487
.....
The training cost for iteration ::14300 is
0.29214290433645945
.....
The training cost for iteration ::14400 is
0.2921262425571334
.....
The training cost for iteration ::14500 is
0.29210971778797695
.....
The training cost for iteration ::14600 is
0.29209332740042393
.....
The training cost for iteration ::14700 is
0.29207706884221796
.....
The training cost for iteration ::14800 is
0.2920609396346169
.....
The training cost for iteration ::14900 is
0.29204493736971576
.....
logistic_regressor.viswalize_loss()
```



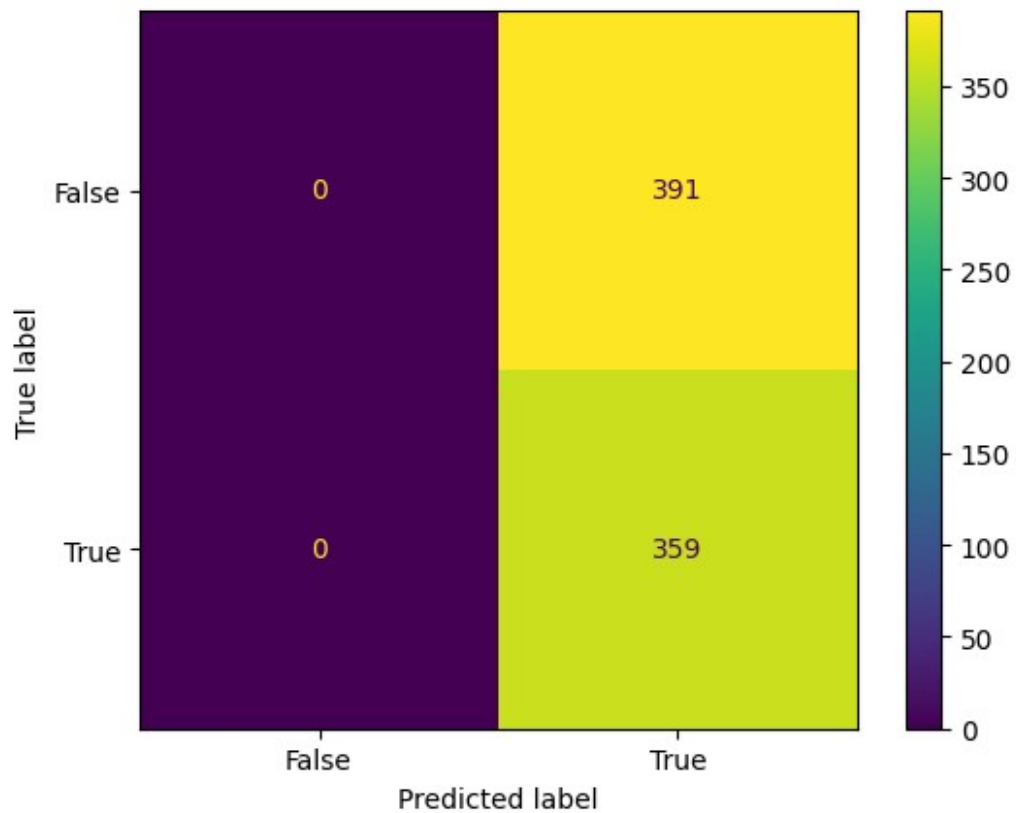
```
mat_loss_df = logistic_regressor.metrics_loss(X_train,Y_train)
mat_loss_df.rename(index={0:'Train_data'},inplace=True)
```

```
mat_loss_df
```

```
      Recall  Precision  Mean Accuracy  
Train_data  0.855263    0.8981         0.8824
```

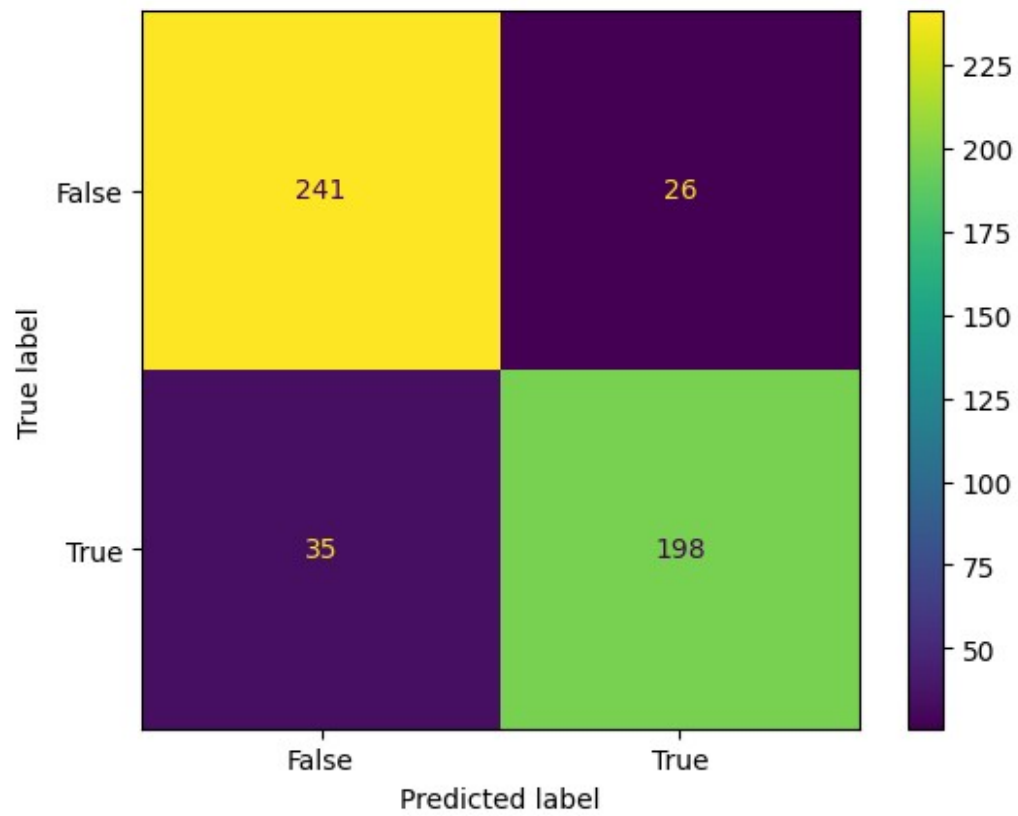
```
mat_loss_df = logistic_regressor.metrics_loss(X_val,Y_val)  
mat_loss_df.rename(index={0:'Validation_data'},inplace=True)
```



```
mat_loss_df
```

	Recall	Precision	Mean Accuracy
Validation_data	1.0	0.478667	0.478667

```
mat_loss_df = logistic_regressor.metrics_loss(X_test,Y_test)  
mat_loss_df.rename(index={0:'Test_data'},inplace=True)
```



```
mat_loss_df
```

	Recall	Precision	Mean Accuracy
Test_data	0.849785	0.883929	0.878