https://github.com/Abhishek841428/ETE-18SCSE1010334.git

```
In [1]:
#2.Write a program to demonstrate the working of the Logistic Regression.
#Use an appropriate data set the implementation.
print('Abhishek Kumar 18SCSE1010334')
Abhishek Kumar 18SCSE1010334
In [2]:
#example is related to a single-variate binary classification problem.
#This is the most straightforward kind of classification problem.
import matplotlib.pyplot as plt
import numpy as np
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import classification_report, confusion_matrix
In [3]:
x = np.arange(10).reshape(-1, 1)
y = np.array([0, 0, 0, 0, 1, 1, 1, 1, 1, 1])
In [4]:
Х
Out[4]:
array([[0],
       [1],
       [2],
       [3],
       [4],
       [5],
       [6],
       [7],
       [8],
       [9]])
In [5]:
У
Out[5]:
array([0, 0, 0, 0, 1, 1, 1, 1, 1, 1])
In [6]:
model = LogisticRegression(solver='liblinear', random_state=0)
```

```
In [7]:
model.fit(x, y)
Out[7]:
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept
=True,
                   intercept_scaling=1, l1_ratio=None, max_iter=100,
                   multi_class='auto', n_jobs=None, penalty='12',
                   random_state=0, solver='liblinear', tol=0.0001, ver
bose=0,
                   warm_start=False)
In [8]:
model = LogisticRegression(solver='liblinear', random_state=0).fit(x, y)
In [9]:
model.classes_
Out[9]:
array([0, 1])
In [10]:
model.intercept_
Out[10]:
array([-1.04608067])
In [11]:
model.coef_
Out[11]:
```

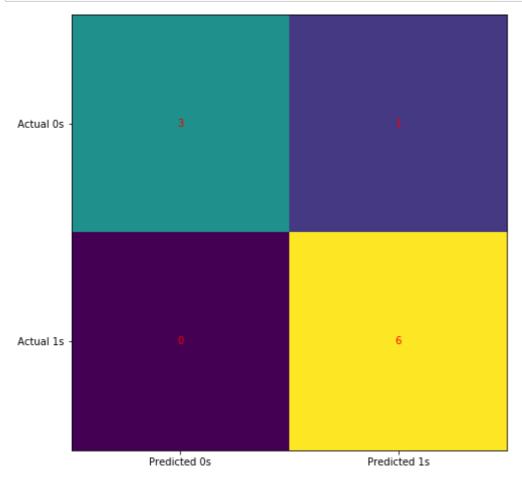
array([[0.51491375]])

```
In [12]:
model.predict_proba(x)
Out[12]:
array([[0.74002157, 0.25997843],
       [0.62975524, 0.37024476],
       [0.5040632 , 0.4959368 ], [0.37785549, 0.62214451],
       [0.26628093, 0.73371907],
       [0.17821501, 0.82178499],
       [0.11472079, 0.88527921],
       [0.07186982, 0.92813018],
       [0.04422513, 0.95577487],
       [0.02690569, 0.97309431]])
In [13]:
model.predict(x)
Out[13]:
array([0, 0, 0, 1, 1, 1, 1, 1, 1])
In [14]:
model.score(x, y)
Out[14]:
0.9
In [15]:
confusion_matrix(y, model.predict(x))
Out[15]:
array([[3, 1],
       [0, 6]], dtype=int64)
```

In [16]:

```
cm = confusion_matrix(y, model.predict(x))

fig, ax = plt.subplots(figsize=(8, 8))
ax.imshow(cm)
ax.grid(False)
ax.xaxis.set(ticks=(0, 1), ticklabels=('Predicted 0s', 'Predicted 1s'))
ax.yaxis.set(ticks=(0, 1), ticklabels=('Actual 0s', 'Actual 1s'))
ax.set_ylim(1.5, -0.5)
for i in range(2):
    for j in range(2):
        ax.text(j, i, cm[i, j], ha='center', va='center', color='red')
plt.show()
```



In [17]:

```
print(classification_report(y, model.predict(x)))
```

	precision	recall	f1-score	support	
0	1.00	0.75	0.86	4	
1	0.86	1.00	0.92	6	
accuracy			0.90	10	
macro avg	0.93	0.88	0.89	10	
weighted avg	0.91	0.90	0.90	10	

In [18]:

```
model = LogisticRegression(solver='liblinear', C=10.0, random_state=0)
model.fit(x, y)
```

Out[18]:

LogisticRegression(C=10.0, class_weight=None, dual=False, fit_intercep
t=True,

intercept_scaling=1, l1_ratio=None, max_iter=100,
multi_class='auto', n_jobs=None, penalty='l2',
random_state=0, solver='liblinear', tol=0.0001, ver

bose=0,

warm_start=False)

In [19]:

```
model.intercept_
```

Out[19]:

array([-3.51335372])

In [20]:

```
model.coef
```

Out[20]:

```
array([[1.12066084]])
```

```
In [21]:
model.predict_proba(x)
Out[21]:
array([[0.97106534, 0.02893466],
       [0.9162684, 0.0837316],
       [0.7810904 , 0.2189096 ],
       [0.53777071, 0.46222929],
       [0.27502212, 0.72497788],
       [0.11007743, 0.88992257],
       [0.03876835, 0.96123165],
       [0.01298011, 0.98701989],
       [0.0042697, 0.9957303],
       [0.00139621, 0.99860379]])
In [22]:
model.predict(x)
Out[22]:
array([0, 0, 0, 0, 1, 1, 1, 1, 1])
In [23]:
model.score(x, y)
Out[23]:
1.0
In [24]:
confusion_matrix(y, model.predict(x))
Out[24]:
array([[4, 0],
       [0, 6]], dtype=int64)
In [25]:
print(classification_report(y, model.predict(x)))
                            recall
              precision
                                    f1-score
                                                support
                              1.00
                                        1.00
                                                      4
           0
                    1.00
           1
                    1.00
                              1.00
                                        1.00
                                                      6
                                        1.00
                                                     10
    accuracy
                    1.00
                                        1.00
   macro avg
                              1.00
                                                     10
weighted avg
                    1.00
                              1.00
                                        1.00
                                                     10
```

In []:			