

## EXPN.O:3. A PYTHON PROGRAM TO IMPLEMENT

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Aim:  $\left(\left[\frac{1}{2}, \frac{1}{2}\right] \text{ is closed} \iff \forall x \in \left[\frac{1}{2}, \frac{1}{2}\right], \exists \delta > 0 \text{ such that } \forall y \in B(x, \delta) \cap \left[\frac{1}{2}, \frac{1}{2}\right] \Rightarrow y \in \left[\frac{1}{2}, \frac{1}{2}\right]\right)$

To implement python program for the logistic model using sun car dataset.

Algorithm: The algorithm of the program of the library function  $\text{libc}_\text{float}(\text{x}, \text{y})$  is as follows:

Step 1 : Import necessary library  
+ pandas for data manipulation.

\* Sklearn model\_selection for train-test split

\* Sklearn & preprocessing for data preprocessing

\* `sklearn.linear_model` for logistic regression.

Step 8: Read The dataset :

use pandas to read the sun-cars.csv dataset into a  
(or boston.csv)

## Dataframe

### Step 3: Split The Data

\* Split the dataset into training and testing sets using train-test split.

## Step 4: Feature scaling

Standardize the features using StandardScaler to ensure they have the same scale.

Step 5: Create and train the model

Create a logistic regression model using Logit Regression from sklearn module.

## Step 7: Make Prediction:

+ Use the trained model to make prediction on the test data using the predict method.

\* Use the "predict()" function to predict the values of the testing data and assign the values to "y-pred" variable.

### Step 8 : Evaluate The Model :

\* Calculate the accuracy of the model on the test data using the slow method (Accuracy =  $(tp+tn) / (tp+tn-fp+fn)$ )

Step 9: Visualize the results:

Plot the decision boundary of the logistic regression model (optional).

[[-1.05714987 0.53420426]]

## Confusion Matrix:

[EB31 1 7]

[1 77].

Accuracy: 0.95

[1.04388575 0.47576806]).

0.7583333333333334

0.823529411764706

Accuracy is : 0.925

## Result:

Thus the Python program for the given sun-care dataset is developed and the logistic regression model is classified successfully. The performance of the developed model is measured using F1-score and Accuracy.

```

import pandas as pd
import numpy as np
from numpy import log, dot, exp, shape
from sklearn.metrics import confusion_matrix
data = pd.read_csv('../input/survived/surv_data.csv')
print(data.head())
n = data.iloc[:, [2, 3]].values
y = data.iloc[:, 4].values
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
print(X_train[0:10, :])
classifier = LogisticRegression(random_state=0)
classifier.fit(X_train, y_train)
LogisticRegression(random_state=0)
y_pred = classifier.predict(X_test)
print(y_pred)

cm = confusion_matrix(y_test, y_pred)
print("Confusion Matrix: \n", cm)
print("Accuracy: ", accuracy_score(y_test, y_pred))

def load_data():
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.1)
    return X_train, X_test, y_train, y_test

```

```

def std (n):
    mean0 = np.mean (input_data[:, 0])
    sd0 = np.std (input_data[:, 0])
    mean1 = np.mean (input_data[:, 1])
    sd1 = np.std (input_data[:, 1])
    my_std = std (n)
    my_std (n - train [0])

```

```

def predict (self; x):
    z = dot (self.initialize (n) [1], self.weights)

```

```

lis = []
for i in self.sigmoid (z):
    if i > 0.5:
        lis.append (1)
    else:
        lis.append (0)

```

return lis

Standardize (x-train)

Standardize (x-test)

Obj 1 = Logistic Regression()

model = Obj 1.fit (x-train, y-train)

y-pred = Obj 1.predict (x-test)

y-train = Obj 1.predict (x-train)

f1-score-tr = f1-score (y-train, y-pred)

f1-score-te = f1-score (y-train, y-pred)

print ("Accuracy is : ", accuracy\_score (y-train, y-pred))

	User ID	Gender	Age	Salary	Purchased
0	15024510	Male	19	19000	0
1	...	...	...	21 ~	...
2	...	...	...	...	...
3	...	...	...	...	...
4	...	...	...	...	...