

17. Data Science – Machine Learning – Logistic Regression

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17. Data Science – Machine Learning – Logistic Regression

1. Logistic Regression

- ✓ Logistic regression comes under supervised Learning.
- ✓ It is a technique that is used to solve for classification problems.
- ✓ It is used for predicting the categorical dependent variable using a given set of independent variables.
- ✓ Examples
 - Email spam or not
 - Customer will buy product or not

2. Types of logistic regression

- ✓ Binary classification
 - This is having two classes
- ✓ Multiclass classification
 - This is having more than two classes

3. Binary classification

- ✓ In binary classification, there can be only two possible types of the dependent variables, such as,
 - 0 or 1
 - Pass or Fail
 - Yes or No etc.

4. Multiclass classification

- ✓ In multiclass classification, there can be 3 or more possible unordered types of the dependent variable, such as,
 - Ok, good, best
 - Cat, dog, sheep etc

5. Data set

- ✓ Its insurance dataset
 - ZERO means didn't buy the insurance
 - ONE means will buy the insurance
- ✓ We can understand one pattern here like, young people not buying the insurance
- ✓ Whereas person age increasing then that person more likely to buy the insurance

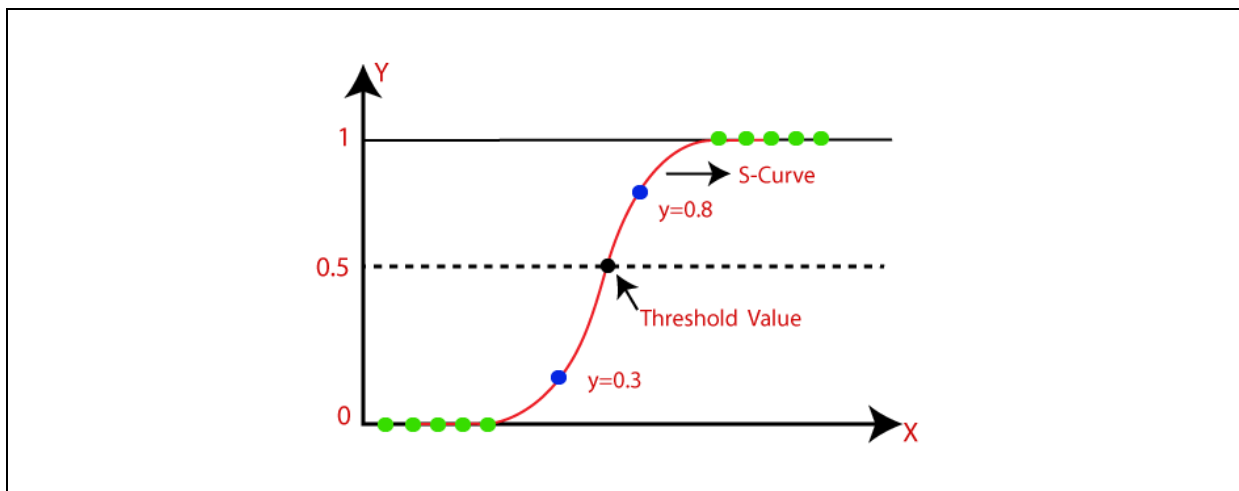
6. Problem statement

- ✓ Based on the age, we wanted to predict for persons will chose insurance or not.

age	Insurance status
22	0
25	0
47	1
52	0
46	1
56	1
55	0
60	1
62	1
61	1

7. Logistic function or Sigmoid

- ✓ The logistic function, also called the sigmoid function.
- ✓ It maps any real value into another value within a range of 0 and 1.
- ✓ The value of the logistic regression must be between 0 and 1.
- ✓ In logistic regression, we use the concept of the threshold value, which defines the probability of either 0 or 1.



Formula

$$\text{sigmoid}(z) = \frac{1}{1 + e^{-z}}$$

e = Euler's number ~ 2.71828

Sigmoid function converts input into range 0 to 1

Program Name Loading insurance dataset
demo1.py

```
import pandas as pd
```

```
# Loading the dataset  
df = pd.read_csv("insurance_data.csv")
```

```
print(df.head(10))
```

Output

```
   age  bought_insurance  
0   22                0  
1   25                0  
2   47                1  
3   52                0  
4   46                1  
5   56                1  
6   55                0  
7   60                1  
8   62                1  
9   61                1
```

Program Name Plotting the dataset
demo2.py

```
import pandas as pd
from matplotlib import pyplot as plt

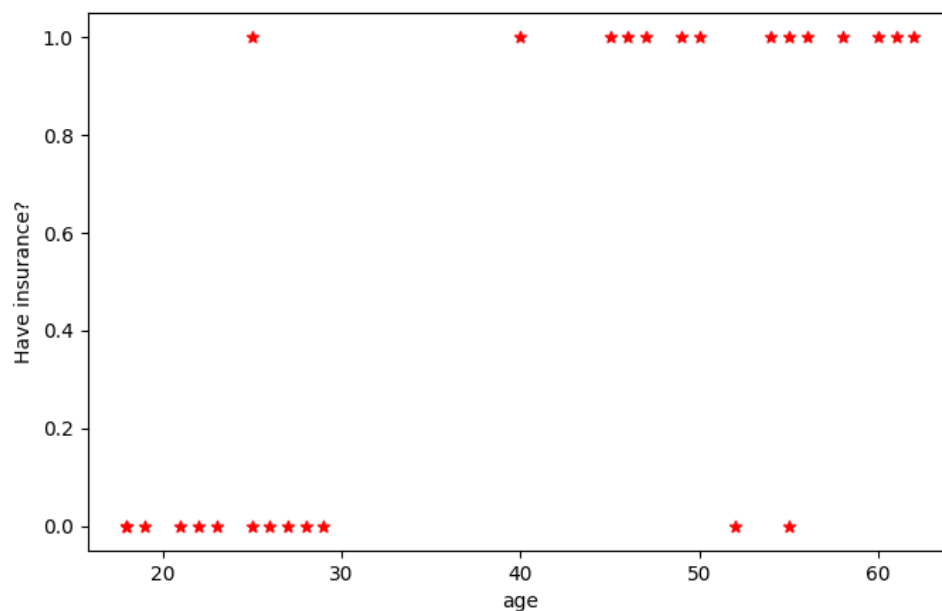
# Loading the dataset
df = pd.read_csv("insurance_data.csv")

# plotting the data
plt.scatter(df.age, df.bought_insurance, marker = '*', color = 'red')

plt.xlabel('age')
plt.ylabel('Have insurance?')

plt.show()
```

Output



Program Name Splitting the dataset
demo3.py

```
import pandas as pd
from sklearn.model_selection import train_test_split

# Loading the dataset
df = pd.read_csv("insurance_data.csv")
X = df[['age']]
y = df.bought_insurance

# Splitting the data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.1, random_state=52)

print("X_train", '\n')
print(X_train, '\n')

print("X_test", '\n')
print(X_test)
```

Output

```
X_train
```

```
      age
```

```
14    49
```

```
4     46
```

```
12    27
```

```
2     47
```

```
8     62
```

```
6     55
```

```
19    18
```

```
26    23
```

```
24    50
```

```
20    21
```

```
15    55
```

```
16    25
```

```
1     25
```

```
17    58
```

```
3     52
```

```
25    54
```

```
10    18
```

```
5     56
```

```
0     22
```

```
22    40
```

```
23    45
```

```
13    29
```

```
11    28
```

```
21    26
```

```
X_test
```

```
      age
```

```
7     60
```

```
9     61
```

```
18    19
```


Program Name Splitting the dataset
demo4.py

```
import pandas as pd
from sklearn.model_selection import train_test_split

# Loading the dataset
df = pd.read_csv("insurance_data.csv")

X = df[['age']]
y = df.bought_insurance

# Splitting the data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.1, random_state = 52)

print("y_train", '\n')
print(y_train, '\n')

print("y_test", '\n')
print(y_test)
```

Output

```
y_train
14    1
4     1
12    0
2     1
8     1
6     0
19    0
26    0
24    1
20    0
15    1
16    1
1     0
17    1
3     0
25    1
10    0
5     1
0     0
22    1
23    1
13    0
11    0
21    0
Name: bought_insurance, dtype: int64

y_test
7     1
9     1
18    0
Name: bought_insurance, dtype: int64
```

Program Name Training the model
demo5.py

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression

# Loading the dataset
df = pd.read_csv("insurance_data.csv")

X = df[['age']]
y = df.bought_insurance

# Splitting the data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.1, random_state = 52)

# Creating and training the model
model = LogisticRegression()
model.fit(X_train, y_train)

print("Model got trained")
```

Output

Model got trained

Program Name Prediction with single value
demo6.py

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression

# Loading the dataset
df = pd.read_csv("insurance_data.csv")

X = df[['age']]
y = df.bought_insurance

# Splitting the data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.1, random_state = 52)

# Creating and training the model
model = LogisticRegression()
model.fit(X_train, y_train)

# Prediction
print(model.predict([[50]]))
print(model.predict([[25]]))
```

Output

```
[1]
[0]
```

Program Name Prediction the result
demo7.py

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression

# Loading the dataset
df = pd.read_csv("insurance_data.csv")

X = df[['age']]
y = df.bought_insurance

# Splitting the data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.1, random_state = 52)

# Creating and training the model
model = LogisticRegression()
model.fit(X_train, y_train)

# Prediction
y_predicted = model.predict(X_test)

print("X_test data is \n")
print(X_test, '\n')

print("Prediction for X_test data \n")
print(y_predicted)
```

Output

```
X_test data is
   age
7    60
9    61
18   19

Prediction for X_test data
[1 1 0]
```

Prediction

- ✓ The one who has 19 years age he will not buy the insurance
- ✓ Both who has 60 and 61 years age persons will buy the insurance

Program Prediction score
Name demo8.py

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression

# Loading the dataset
df=pd.read_csv("insurance_data.csv")

X = df[['age']]
y = df.bought_insurance

# Splitting the data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.1, random_state = 52)

# Creating and training the model
model = LogisticRegression()
model.fit(X_train, y_train)

# Prediction score
print(model.score(X_test, y_test))
```

Output

1.0

Program Name Prediction probability
demo9.py

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression

# Loading the dataset
df = pd.read_csv("insurance_data.csv")

X = df[['age']]
y = df.bought_insurance

# Splitting the data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.1, random_state = 52)

# Creating and training the model
model = LogisticRegression()
model.fit(X_train, y_train)

print("X_test")
print(X_test)

print()

# Prediction probability
print(model.predict_proba(X_test))
```

Output

```
X_test
  age
7   60
9   61
18  19

[[0.06266647 0.93733353]
 [0.05553734 0.94446266]
 [0.92804604 0.07195396]]
```