

logisticpynb

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Roll : **21X05A6701**

Stream : **DataScience**

College : **Narsimha Reddy Engineering College**

Project Title : To Predict the Heartattack disease for organization (WHO:World Health Organization) using machine learning algorithm rate of heart attack disease will increasing manner or decreasing manner

Problem Statement: A WHO estimated 12 millions death records. One of them half off the death result is found in US. There search scholars point out the relevant risk factor of heart attack as a DataScience engineer predict the overall risk using ML algorithm(Logistic Regression)

Task : 1) Import the libraries required for prediction.

2) Import the Dataset using workspace.

3) Use the appropriate argument of sklearn Library to train,test and split the dataset

4) Fit your values with a range function using feature scalling

5) Check your model accuracy and precison using confusion matrix

Conclusion : According to the model analysis the Logistic Regression Algorithm works successfully with 0.6 accuracy.

The accuracy shows that building model is successfull

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[2]: data = pd.read_csv("framingham.csv")
data
```

```
[2]:      male  age  education  currentSmoker  cigsPerDay  BPMeds  \
0         1   39         4.0              0          0.0     0.0
1         0   46         2.0              0          0.0     0.0
2         1   48         1.0              1         20.0     0.0
3         0   61         3.0              1         30.0     0.0
```

4	0	46	3.0	1	23.0	0.0
...
4233	1	50	1.0	1	1.0	0.0
4234	1	51	3.0	1	43.0	0.0
4235	0	48	2.0	1	20.0	NaN
4236	0	44	1.0	1	15.0	0.0
4237	0	52	2.0	0	0.0	0.0

	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	\
0	0	0	0	195.0	106.0	70.0	26.97	
1	0	0	0	250.0	121.0	81.0	28.73	
2	0	0	0	245.0	127.5	80.0	25.34	
3	0	1	0	225.0	150.0	95.0	28.58	
4	0	0	0	285.0	130.0	84.0	23.10	
...	
4233	0	1	0	313.0	179.0	92.0	25.97	
4234	0	0	0	207.0	126.5	80.0	19.71	
4235	0	0	0	248.0	131.0	72.0	22.00	
4236	0	0	0	210.0	126.5	87.0	19.16	
4237	0	0	0	269.0	133.5	83.0	21.47	

	heartRate	glucose	TenYearCHD
0	80.0	77.0	0
1	95.0	76.0	0
2	75.0	70.0	0
3	65.0	103.0	1
4	85.0	85.0	0
...
4233	66.0	86.0	1
4234	65.0	68.0	0
4235	84.0	86.0	0
4236	86.0	NaN	0
4237	80.0	107.0	0

[4238 rows x 16 columns]

```
[15]: from sklearn.model_selection import train_test_split
X=data[["age"]]
y=data["currentSmoker"]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
↳random_state=42)
```

```
[16]: print(X_train)
```

	age
3252	40
3946	57

```

1261    47
2536    41
4089    64
...    ...
3444    36
466     57
3092    60
3772    39
860     35

```

```
[3390 rows x 1 columns]
```

```
[17]: print(y_train)
```

```

3252    1
3946    0
1261    0
2536    1
4089    0
...
3444    1
466     1
3092    0
3772    1
860     0

```

```
Name: currentSmoker, Length: 3390, dtype: int64
```

```
[18]: print(X_test)
```

```

      age
3188    63
764     45
3264    51
1967    45
2185    45
...    ...
3303    47
4056    44
4210    50
3971    64
2540    55

```

```
[848 rows x 1 columns]
```

```
[19]: print(y_test)
```

```

3188    0
764     0

```

```

3264    1
1967    1
2185    1
..
3303    0
4056    0
4210    0
3971    0
2540    1
Name: currentSmoker, Length: 848, dtype: int64

```

```

[20]: from sklearn.preprocessing import StandardScaler
      sc = StandardScaler()
      X_train = sc.fit_transform(X_train)
      X_test = sc.transform(X_test)

```

```

[21]: print(X_train)

```

```

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 ...
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[22]: print(X_test)

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```

```
[23]: from sklearn.linear_model import LogisticRegression
      classifier = LogisticRegression(random_state = 0)
      classifier.fit(X_train, y_train)
```

```
[23]: LogisticRegression(random_state=0)
```

```
[25]: y_pred = classifier.predict(X_test)
```

```
[26]: from sklearn.metrics import confusion_matrix, accuracy_score
      cm = confusion_matrix(y_test, y_pred)
      print(cm)
      accuracy_score(y_test, y_pred)
```

```
[[263 183]
 [155 247]]
```

```
[26]: 0.6014150943396226
```