

logisticregression

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##COLLEGE : NRCM

#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 : The world health organisation estimated 12 millions death records.one of them half of the death results is found in us. the research scholar point-out the most relevant risk factor of heart attack, as the data science engineer predict the overall risk using machine learning algorithm called as logistic regression

###Task1: Import the required library which is required for prediction ###Task2: Import the dataset using our workspace. ###Task3: Use the appropriate of sklearn library to train,test and split the dataset ###Task4: fit your values with a range function using future scaling ###Task5: check your model accuracy and precision using confusion matrix

#conclusion : According to the model analysis the LogisticRegression algorithm works successfully with 0.6 accuracy. the accuracy shows that building the model is successful.

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

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

```
[ ]:      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]

```
[ ]: 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)
```

```
[ ]: 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]
```

```
[ ]: 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
```

```
[ ]: 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]
```

```
[ ]: 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
```

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

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

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

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

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

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

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

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```

```
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```

```
[ ]:
```