HEART DISEASE PREDICTION USING ML

1.IMPORT THE DEPENDENCIES

In [5]: import numpy as np
 import pandas as pd
 from sklearn.model_selection import train_test_split
 from sklearn.linear_model import LogisticRegression
 from sklearn.metrics import accuracy_score

data processing

In [7]: heart_data=pd.read_csv("C://Users//91903//Downloads//heart_disease_data.csv")

In [8]: heart_data

Out[8]:

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	targe
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1	
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	•
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	•
3	56	1	1	120	236	0	1	178	0	0.8	2	0	2	
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	
298	57	0	0	140	241	0	1	123	1	0.2	1	0	3	(
299	45	1	3	110	264	0	1	132	0	1.2	1	0	3	(
300	68	1	0	144	193	1	1	141	0	3.4	1	2	3	(
301	57	1	0	130	131	0	1	115	1	1.2	1	1	3	(
302	57	0	1	130	236	0	0	174	0	0.0	1	1	2	(

303 rows × 14 columns

```
In [9]: heart_data.head()
```

Out[9]:

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	са	thal	target
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1	1
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	1
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	1
3	56	1	1	120	236	0	1	178	0	8.0	2	0	2	1
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	1

In [10]: heart_data.tail()

Out[10]:

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	са	thal	targe
298	57	0	0	140	241	0	1	123	1	0.2	1	0	3	(
299	45	1	3	110	264	0	1	132	0	1.2	1	0	3	(
300	68	1	0	144	193	1	1	141	0	3.4	1	2	3	(
301	57	1	0	130	131	0	1	115	1	1.2	1	1	3	(
302	57	0	1	130	236	0	0	174	0	0.0	1	1	2	(

In [12]: heart_data.shape

Out[12]: (303, 14)

In [13]: heart_data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 303 entries, 0 to 302
Data columns (total 14 columns):

#	Column	Non-	-Null Count	Dtype
0	age	303	non-null	int64
1	sex	303	non-null	int64
2	ср	303	non-null	int64
3	trestbps	303	non-null	int64
4	chol	303	non-null	int64
5	fbs	303	non-null	int64
6	restecg	303	non-null	int64
7	thalach	303	non-null	int64
8	exang	303	non-null	int64
9	oldpeak	303	non-null	float64
10	slope	303	non-null	int64
11	ca	303	non-null	int64
12	thal	303	non-null	int64
13	target	303	non-null	int64
4+,,,,	oc. <u>floot</u> 6	1/11	in+61/12)	

dtypes: float64(1), int64(13)

memory usage: 33.3 KB

```
In [17]: #checking for missing values
         heart_data.isnull().sum()
Out[17]: age
                      0
         sex
                      0
                      0
         ср
         trestbps
                      0
         chol
                      0
         fbs
                      0
         restecg
                      0
         thalach
                      0
         exang
                      0
         oldpeak
         slope
                      0
         ca
                      0
         thal
                      0
         target
         dtype: int64
In [18]: #statistical measures of the data
```

heart_data.describe()

Out[18]:

је	sex	ср	trestbps	chol	fbs	restecg	thalach	exan
00	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.00000
37	0.683168	0.966997	131.623762	246.264026	0.148515	0.528053	149.646865	0.32673
)1	0.466011	1.032052	17.538143	51.830751	0.356198	0.525860	22.905161	0.46979
00	0.000000	0.000000	94.000000	126.000000	0.000000	0.000000	71.000000	0.00000
00	0.000000	0.000000	120.000000	211.000000	0.000000	0.000000	133.500000	0.00000
00	1.000000	1.000000	130.000000	240.000000	0.000000	1.000000	153.000000	0.00000
)0	1.000000	2.000000	140.000000	274.500000	0.000000	1.000000	166.000000	1.00000
00	1.000000	3.000000	200.000000	564.000000	1.000000	2.000000	202.000000	1.00000
4								>

In [21]: #checking the distribution of target variable heart_data['target'].value_counts()

Out[21]: 1 165 138

Name: target, dtype: int64

1-->defective heart

0-->healthy heart

```
In [23]: #splitting the features and target
X=heart_data.drop(columns="target",axis=1)
Y=heart_data['target']
```

In [27]: X

Out[27]:

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	са	thal
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2
3	56	1	1	120	236	0	1	178	0	0.8	2	0	2
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2
298	57	0	0	140	241	0	1	123	1	0.2	1	0	3
299	45	1	3	110	264	0	1	132	0	1.2	1	0	3
300	68	1	0	144	193	1	1	141	0	3.4	1	2	3
301	57	1	0	130	131	0	1	115	1	1.2	1	1	3
302	57	0	1	130	236	0	0	174	0	0.0	1	1	2

303 rows × 13 columns

```
In [28]:
Out[28]: 0
                  1
          1
                  1
          2
                  1
          3
                  1
                  1
          298
                 0
          299
                 0
          300
                 0
          301
                 0
          302
          Name: target, Length: 303, dtype: int64
```

Splitting the data into training data and test data

```
In [31]: X_train,X_test,Y_train,Y_test=train_test_split(X,Y,test_size=0.2,stratify=Y,ra
In [32]: print(X.shape,X_train.shape,X_test.shape)
```

(303, 13) (242, 13) (61, 13)

Model training(LOGISTIC REGRESSION)

```
In [33]: |model=LogisticRegression()
In [35]: #train the ML model with training data
         model.fit(X train,Y train)
         C:\Users\91903\anaconda3\lib\site-packages\sklearn\linear model\ logistic.py:
         814: ConvergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html (https://sciki
         t-learn.org/stable/modules/preprocessing.html)
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear model.html#logistic-regres
         sion (https://scikit-learn.org/stable/modules/linear model.html#logistic-regr
         ession)
           n_iter_i = _check_optimize_result(
Out[35]: LogisticRegression()
         model evaluation accuracy score
In [39]: #accuracy on training data
         X train prediction=model.predict(X train)
         training data accuracy=accuracy score(X train prediction,Y train)
In [41]: |print("Accuracy on training data:",training_data_accuracy)
         Accuracy on training data: 0.8512396694214877
In [42]: #accuracy on test data
         X_test_prediction=model.predict(X_test)
         test_data_accuracy=accuracy_score(X_test_prediction,Y_test)
In [43]: |print("Accuracy on test data",test_data_accuracy)
         Accuracy on test data 0.819672131147541
         BUILDING PREDICTIVE SYSTEM
In [44]: input_data=(41,0,1,130,204,0,0,172,0,1.4,2,0,2)
```

```
In [60]: #change input_data to numpy array
         input_data_as_numpy_array=np.asarray(input_data)
         #reshape the numpy array as we are predicting for only one instance
         input data reshaped=input data as numpy array.reshape(1,-1)
         prediction=model.predict(input_data_reshaped)
         print(prediction)
         if (prediction[0]==0):
             print("the person does not have heart disease")
         else:
             print("The person has heart disease")
         [1]
         The person has heart disease
         C:\Users\91903\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning:
         X does not have valid feature names, but LogisticRegression was fitted with f
         eature names
           warnings.warn(
In [ ]:
```