]:	+ Importing libraries import numpy as np import matplotlib.pyplot as plt import pandas as pd
	+ Studying the Dataset . age . sex: 1 = Male, 0 = Female (Binary)
	. (cp) chest pain [type (4 values, Ordinal)] . (trestbps) resting blood pressure . (chol) serum cholestoral in mg/dl
	. (fbs) fasting blood sugar > 120 mg/dl (Binary) [1 = true; 0 = false] . (restecg) resting electrocardiographic results [values 0,1,2] . (thalach) maximum heart rate achieved
	. (exang) exercise induced angina (Binary) [1 = yes; 0 = no] . (oldpeak) = ST depression induced by exercise relative to rest . (slope) of the peak exercise ST segment (Ordinal) [1: upsloping, 2: flat , 3: downsloping)
	. (ca) number of major vessels (0-3, Ordinal) colored by fluoroscopy . (thal) maximum heart rate achieved (Ordinal) [3 = normal; 6 = fixed defect; 7 = reversable defect] + Importing dataset
]:]:	<pre>dataset1=pd.read_csv('heart.csv') dataset2=pd.read_csv('o2saturation.csv') dataset1 age sex cp trtbps chol fbs restecg thalachh exng oldpeak slp caa thall output</pre>
	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 0.8 2 0 2 1 4 57 0 0 120 354 0 1 163 1 0.6 2 0 2 1
	302 57 0 1 130 236 0 0 174 0 0.0 1 1 2 0 303 rows × 14 columns + Exploring Type of data Dataset1 contains
L]: L]:	age int64 sex int64 cp int64 trtbps int64 chol int64 fbs int64
	restecg int64 thalachh int64 exng int64 oldpeak float64 slp int64 caa int64 thall int64 output int64 dtype: object
]:	98.6 0 98.6 1 98.6
	 2 98.6 3 98.1 4 97.5 3580 98.6
	3581 98.6 3582 98.6 3583 98.6 3584 98.6 3585 rows × 1 columns
]:	+ Dividing the dataset into feature matrix and dependent variable vector X = dataset1.iloc[:,:-1].values Y = dataset1.iloc[:,-1].values
	<pre>array([[63., 1., 3.,, 0., 0., 1.], [37., 1., 2.,, 0., 0., 2.], [41., 0., 1.,, 2., 0., 2.], , [68., 1., 0.,, 1., 2., 3.],</pre>
	[57., 1., 0.,, 1., 1., 3.], [57., 0., 1.,, 1., 1., 2.]]) X.reshape(1,-1) array([[63., 1., 3.,, 1., 1., 2.]])
	<pre>array([1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1</pre>
	1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
6]:	<pre>+ Replacing missing data from sklearn.impute import SimpleImputer imputer=SimpleImputer(missing_values=np.nan, strategy='mean') imputer.fit(X[:,:]) X[:,:]=imputer.transform(X[:,:])</pre>
7]:	+ Spliting the data into Training and Testing Data from sklearn.model_selection import train_test_split Xtrain, Xtest, Ytrain, Ytest=train_test_split(X,Y,test_size=0.2, random_state=1)
)]:	+ Feature scaling from sklearn.preprocessing import StandardScaler sc=StandardScaler() Xtrain=sc.fit_transform(Xtrain) Xtest=sc.fit_transform(Xtest)
1]: 1]:	<pre>Xtrain array([[-0.27090572,</pre>
	0.27160724, 1.16707438],, [-2.78820331, 0.6636838, 0.02021114,, 0.96628239, -0.72428597, -0.47497213], [-0.38035344, 0.6636838, -0.95800789,, 0.96628239, -0.72428597, 1.16707438], [-0.05201028, 0.6636838, 0.99843017,, 0.96628239, -0.72428597, 1.16707438]])
	array([[-0.27090572,
5]:	+ Taining the model from sklearn.neighbors import KNeighborsClassifier KC=KNeighborsClassifier(n_neighbors=5, weights='uniform', p=2) KC.fit(Xtrain, Ytrain)
5]: 6]:	<pre>KNeighborsClassifier() + Testing the model Y_estimated=KC.predict(Xtest) Y_estimated</pre>
	array([0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
7]:	<pre>from sklearn.metrics import confusion_matrix,accuracy_score,precision_score,recall_score cm=confusion_matrix(Ytest,Y_estimated) print(cm) print('\n') print('KNN algorithm:-') print('\n') print(f"Accuracy score : {accuracy_score(Ytest,Y_estimated)}") print(f"Precision score : {precision_score(Ytest,Y_estimated)}")</pre>
	<pre>print(f"Recall score : {recall_score(Ytest,Y_estimated)}") [[21 9] [6 25]] KNN algorithm:-</pre>
5]:	Accuracy score : 0.7540983606557377 Precision score : 0.7352941176470589 Recall score : 0.8064516129032258 error_rate=[] for i in range(1,30): KNN=KNeighborsClassifier(n_neighbors=i) KNN_fit(Xtrain, Ytrain)
;]:	<pre>ypred_i=KNN.predict(Xtest) error_rate.append(np.mean(ypred_i!=Ytest)) + visualization of Dataset import seaborn as sns</pre>
	<pre>corrmat = dataset1.corr() f, ax = plt.subplots(figsize=(12, 9)) sns.heatmap(corrmat, vmax=.8, square=True, annot=True); age - 1</pre>
	sex -0.098 1 0.049 -0.057 0.2 0.045 -0.058 -0.044 0.14 0.096 -0.031 0.12 0.21 0.28 qp -0.069 -0.069 -0.049 1 0.048 -0.077 0.094 0.044 0.3 0.39 -0.15 0.12 -0.18 -0.16 0.43 trtbps - 0.28 0.057 0.048 1 0.12 0.18 -0.11 -0.047 0.068 0.19 -0.12 0.1 0.062 -0.14 chol - 0.21 0.2 0.077 0.12 1 0.013 -0.15 0.0099 0.067 0.054 0.004 0.071 0.099 0.085
	fbs - 0.12 0.045 0.094 0.18 0.013 1 0.084 0.0086 0.026 0.0057 -0.06 0.14 0.032 0.028 restecg - 0.12 0.058 0.044 0.11 0.15 0.084 1 0.044 0.071 0.059 0.093 0.072 0.012 0.14 thalachh - 0.4 0.097 0.14 0.39 0.068 0.067 0.026 0.071 0.38 1 0.29 0.26 0.12 0.14 0.099 - 0.097 0.14 0.39 0.068 0.067 0.026 0.071 0.38 1 0.29 0.26 0.12 0.14
	oldpeak - 0.21 0.096 0.15 0.19 0.054 0.0057 0.059 0.34 0.29 1 0.58 0.22 0.21 0.43 slp - 0.17 0.031 0.12 0.12 0.004 0.05 0.093 0.39 0.39 0.39 0.39 0.39 0.39 0.3
	output - 0.23
]:	+ Ploting Error Rate vs K value Graph to Max Accuracy plt.plot(range(1,30),error_rate,marker='o',markerfacecolor='red',markersize=5) plt.xlabel('K value') plt.ylabel('Error Rate') Text(0, 0.5, 'Error Rate')
	0.30 - 0.28 - $\frac{8}{2}$ 0.26 - $\frac{1}{2}$ 0.24 -
	0.22 - 0.20 - 5 10 15 20 25 30 K value
)]:	~ Since Accuracy is higher at 7 and 10 from sklearn.neighbors import KNeighborsClassifier KC=KNeighborsClassifier(n_neighbors=7, weights='uniform', p=2) KC.fit(Xtrain,Ytrain)
]:	<pre>KNeighborsClassifier(n_neighbors=7) Y_estimated=KC.predict(Xtest) from sklearn.metrics import confusion_matrix,accuracy_score,precision_score,recall_score cm=confusion_matrix(Ytest,Y_estimated) print(cm) print('\n')</pre>
	<pre>print('KNN algorithm:-') print('\n') print(f"Accuracy score : {accuracy_score(Ytest,Y_estimated)}") print(f"Precision score : {precision_score(Ytest,Y_estimated)}") print(f"Recall score : {recall_score(Ytest,Y_estimated)}") [[22 8] [4 27]]</pre>
	KNN algorithm:- Accuracy score : 0.8032786885245902 Precision score : 0.7714285714285715 Recall score : 0.8709677419354839
	Prediction that a person will have heart attack from random data to the final model:- age 24 sex 1 cp 2 trtbps 140 chol 200
	chol 200 fbs 1 restecg 0 thalachh 130 exng 0 oldpeak 2 slp 0 caa 0
	thall 1 KC.predict([[24,1,2,140,200,1,0,130,0,2,0,0,1]]) array([1], dtype=int64)
	+ Result As the predicted value is " 1 " which means More chance of heart attack for the person details which we have provided BY:-
	Name :- Aman Asish Gupta

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