

```
In [ ]: #Experiment no.9
```

```
In [ ]: #Aim :To perform and analysis of KNN
```

```
In [ ]: #Name:Janvi R.Kale  
#Roll no.:29  
#sec:A  
#sub:ET 1  
#date:30-09-2025
```

## Importing the Libraries

```
In [1]: import pandas as pd  
import numpy as np
```

## Data acquisition using Pandas

```
In [2]: import os
```

```
In [3]: os.getcwd()
```

```
Out[3]: 'C:\\Users\\This PC'
```

```
In [4]: os.chdir('C:\\Users\\This PC\\OneDrive\\Desktop\\dss practical datasets')
```

```
In [5]: data=pd.read_csv("heart.csv")
```

```
In [6]: data.head()
```

```
Out[6]:
```

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	targ
0	52	1	0	125	212	0	1	168	0	1.0	2	2	3	
1	53	1	0	140	203	1	0	155	1	3.1	0	0	3	
2	70	1	0	145	174	0	1	125	1	2.6	0	0	3	
3	61	1	0	148	203	0	1	161	0	0.0	2	1	3	
4	62	0	0	138	294	1	1	106	0	1.9	1	3	2	



In [7]: data.tail()

Out[7]:

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	t
1020	59	1	1	140	221	0	1	164	1	0.0	2	0	2	
1021	60	1	0	125	258	0	0	141	1	2.8	1	1	3	
1022	47	1	0	110	275	0	0	118	1	1.0	1	1	2	
1023	50	0	0	110	254	0	0	159	0	0.0	2	0	2	
1024	54	1	0	120	188	0	1	113	0	1.4	1	1	3	

In [8]: data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1025 entries, 0 to 1024
Data columns (total 14 columns):
#   Column      Non-Null Count  Dtype
---  -
0    age         1025 non-null   int64
1    sex         1025 non-null   int64
2    cp          1025 non-null   int64
3    trestbps    1025 non-null   int64
4    chol        1025 non-null   int64
5    fbs         1025 non-null   int64
6    restecg     1025 non-null   int64
7    thalach     1025 non-null   int64
8    exang       1025 non-null   int64
9    oldpeak     1025 non-null   float64
10   slope       1025 non-null   int64
11   ca          1025 non-null   int64
12   thal        1025 non-null   int64
13   target      1025 non-null   int64
dtypes: float64(1), int64(13)
memory usage: 112.2 KB
```

In [9]: data.describe()

Out[9]:

	age	sex	cp	trestbps	chol	fbs	res
count	1025.000000	1025.000000	1025.000000	1025.000000	1025.000000	1025.000000	1025.00
mean	54.434146	0.695610	0.942439	131.611707	246.000000	0.149268	0.52
std	9.072290	0.460373	1.029641	17.516718	51.59251	0.356527	0.52
min	29.000000	0.000000	0.000000	94.000000	126.000000	0.000000	0.00
25%	48.000000	0.000000	0.000000	120.000000	211.000000	0.000000	0.00
50%	56.000000	1.000000	1.000000	130.000000	240.000000	0.000000	1.00
75%	61.000000	1.000000	2.000000	140.000000	275.000000	0.000000	1.00
max	77.000000	1.000000	3.000000	200.000000	564.000000	1.000000	2.00

```
In [10]: data.shape
```

```
Out[10]: (1025, 14)
```

```
In [11]: data.size
```

```
Out[11]: 14350
```

```
In [12]: data.ndim
```

```
Out[12]: 2
```

## Data preprocessing \_ data cleaning \_ missing value treatment

```
In [13]: # check Missing Value by record
```

```
data.isna()
```

```
Out[13]:
```

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	
0	False	False	False	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	False	False	False
...	...	...	...	...	...	...	...	...	...	...	...	...
1020	False	False	False	False	False	False	False	False	False	False	False	False
1021	False	False	False	False	False	False	False	False	False	False	False	False
1022	False	False	False	False	False	False	False	False	False	False	False	False
1023	False	False	False	False	False	False	False	False	False	False	False	False
1024	False	False	False	False	False	False	False	False	False	False	False	False

1025 rows × 14 columns



```
In [14]: data.isna().any()
```

```
Out[14]: age          False
sex          False
cp           False
trestbps     False
chol         False
fbs          False
restecg      False
thalach      False
exang        False
oldpeak      False
slope        False
ca           False
thal         False
target       False
dtype: bool
```

```
In [15]: data.isna().sum()
```

```
Out[15]: age          0
sex          0
cp           0
trestbps     0
chol         0
fbs          0
restecg      0
thalach      0
exang        0
oldpeak      0
slope        0
ca           0
thal         0
target       0
dtype: int64
```

## Independent and Dependent Variables

```
In [16]: x=data.drop("target", axis=1)
y=data["target"]
```

## Splitting of DataSet into train and Test

```
In [17]: #splitting the data into training and testing data sets
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2 ,random_st
```

## KNN Classifier

```
In [18]: from sklearn.neighbors import KNeighborsClassifier  
from sklearn.metrics import accuracy_score
```

```
In [19]: knn=KNeighborsClassifier()
```

```
In [20]: knn.fit(x_train, y_train)
```

```
Out[20]: KNeighborsClassifier()
```

```
In [21]: y_pred2=knn.predict(x_test)
```

```
In [22]: accuracy = accuracy_score(y_test, y_pred2)
```

```
In [23]: accuracy
```

```
Out[23]: 0.7317073170731707
```

◆ Conclusion: In this practical, we implemented the K-Nearest Neighbors (KNN) algorithm for classification. We learned how KNN predicts the class of a data point based on the majority of its nearest neighbors. This practical helped us understand the importance of the value of K in achieving accurate predictions for machine learning tasks.

```
In [ ]:
```