

# K-Nearest Neighbor Algorithm

```
In [1]: #Exp no.:9
```

```
In [2]: #Aim : Understanding K-Nearest Neighbor Algorithm
```

```
In [3]: #Exp no.:6  
#Name:Sahil A. Bankar  
#Roll no:04  
#Sec:B  
#Subject:  
#Date:25/09/2025
```

## Importing The Libraries

```
In [4]: import pandas as pd  
import numpy as np
```

Data Acquisition using Pandas

```
In [5]: import os
```

```
In [6]: os.getcwd()
```

```
Out[6]: 'C:\\Users\\DELL'
```

```
In [7]: os.chdir('C:\\Users\\DELL\\Desktop')
```

```
In [8]: data=pd.read_csv("heart.csv")
```

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

```
Out[9]:
```

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

```
In [11]: data.tail()
```

```
Out[11]:
```

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

```
In [12]: 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 [13]: data.describe()
```

```
Out[13]:
```

	age	sex	cp	trestbps	chol	fbs	restecg	
count	1025.000000	1025.000000	1025.000000	1025.000000	1025.000000	1025.000000	1025.000000	1025.000000
mean	54.434146	0.695610	0.942439	131.611707	246.000000	0.149268	0.529756	1.000000
std	9.072290	0.460373	1.029641	17.516718	51.59251	0.356527	0.527878	1.000000
min	29.000000	0.000000	0.000000	94.000000	126.000000	0.000000	0.000000	1.000000
25%	48.000000	0.000000	0.000000	120.000000	211.000000	0.000000	0.000000	1.000000
50%	56.000000	1.000000	1.000000	130.000000	240.000000	0.000000	1.000000	1.000000
75%	61.000000	1.000000	2.000000	140.000000	275.000000	0.000000	1.000000	1.000000
max	77.000000	1.000000	3.000000	200.000000	564.000000	1.000000	2.000000	2.000000

```
In [15]: data.shape
```

```
Out[15]: (1025, 14)
```

```
In [16]: data.size
```

```
Out[16]: 14350
```

```
In [17]: data.ndim
```

```
Out[17]: 2
```

## Data preprocessing *data cleaning* missing value treatment

```
In [18]: # check Missing Value by record  
data.isna()
```

```
Out[18]:
```

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal
0	False	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	False
2	False	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	False
4	False	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	False
1021	False	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	False
1023	False	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	False

1025 rows × 14 columns



```
In [19]: # check Missing Value by record  
data.isna().any()
```

```
Out[19]: 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
```

## Independent and Dependent Variables

```
In [20]: x=data.drop("target", axis=1)  
y=data["target"]
```

# Splitting of DataSet into train and Test

```
In [21]: 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_stat
```

## KNN Classifier

```
In [22]: from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score
```

```
In [23]: knn=KNeighborsClassifier()
```

```
In [24]: knn.fit(x_train, y_train)
```

```
Out[24]: KNeighborsClassifier()
```

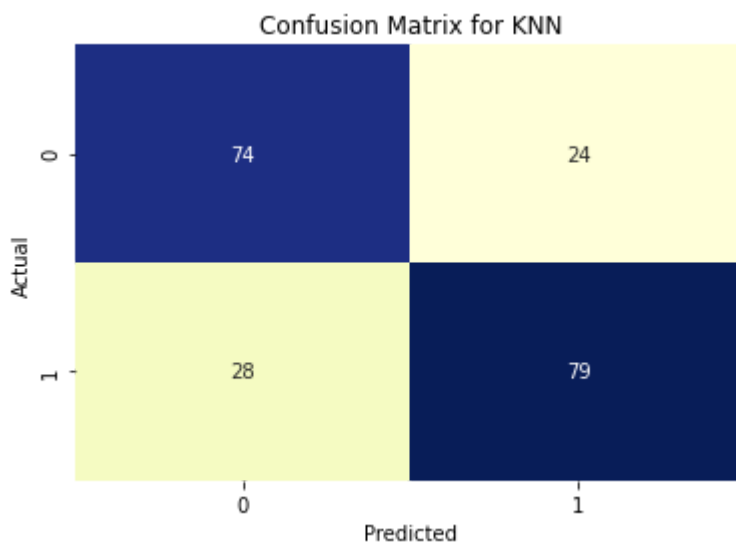
```
In [25]: y_pred2=knn.predict(x_test)
```

```
In [26]: accuracy = accuracy_score(y_test, y_pred2)
```

```
In [27]: accuracy
```

```
Out[27]: 0.7463414634146341
```

```
In [29]: import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import accuracy_score, confusion_matrix, classification
cm = confusion_matrix(y_test, y_pred2)
plt.figure(figsize=(6,4))
sns.heatmap(cm, annot=True, fmt="d", cmap="YlGnBu", cbar=False)
plt.title("Confusion Matrix for KNN")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.show()
```



```
In [35]: from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score
import matplotlib.pyplot as plt
```

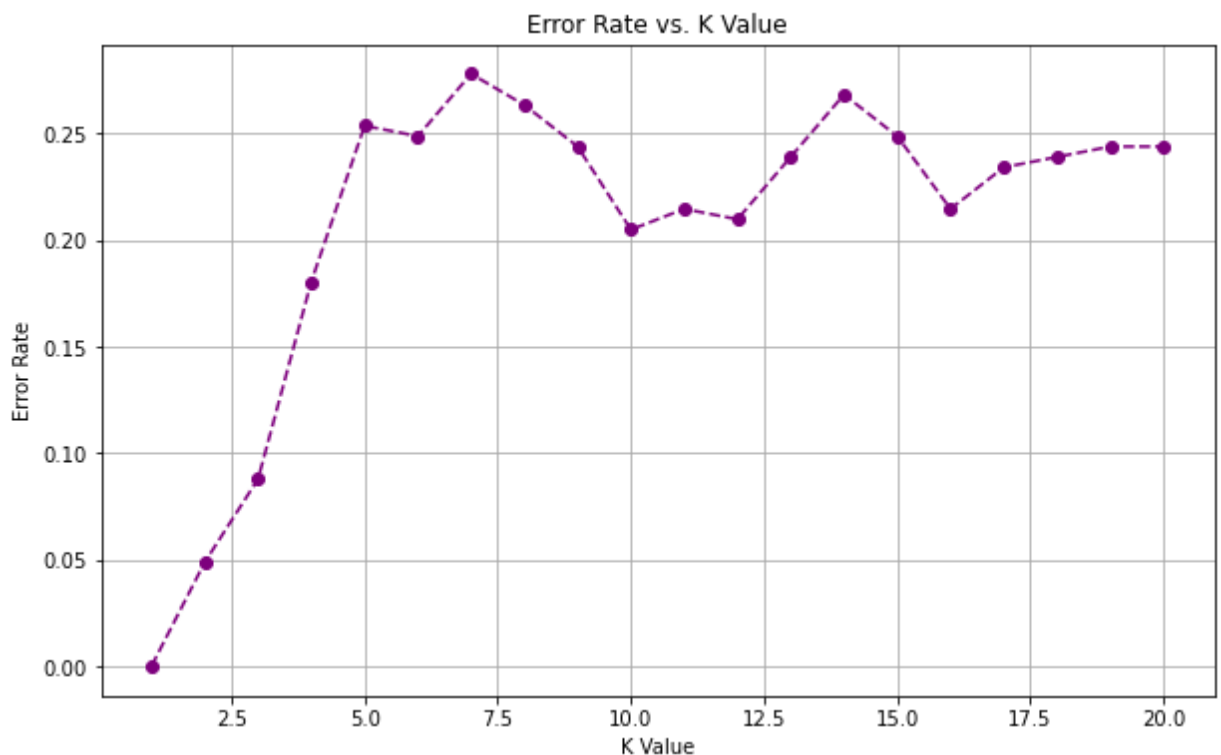
```

error_rates = []

for k in range(1, 21):
    knn_temp = KNeighborsClassifier(n_neighbors=k)
    knn_temp.fit(x_train, y_train)
    pred_k = knn_temp.predict(x_test)
    error_rates.append(1 - accuracy_score(y_test, pred_k))

plt.figure(figsize=(10, 6))
plt.plot(range(1, 21), error_rates, marker='o', linestyle='--', color='purple') # <
plt.title("Error Rate vs. K Value")
plt.xlabel("K Value")
plt.ylabel("Error Rate")
plt.grid(True)
plt.show()

```



## Conclusion :

The experiment successfully implemented the KNearest Neighbour (KNN) algorithm, showcasing its simplicity and effectiveness in classification tasks. The analysis confirmed that the model's accuracy depends significantly on the choice of 'K' and distance metrics.

In [ ]: