

K-Nearest Neighbor Algorithm

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

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

```
In [3]: #Name:Vedant M.Padole  
#Roll no:42  
#Sec:C  
#Subject:ET1  
#Date:
```

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	fb	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	fbstest	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  
--- 
age            1025 non-null    int64  
sex            1025 non-null    int64  
cp             1025 non-null    int64  
trestbps       1025 non-null    int64  
fbs            1025 non-null    int64  
restecg        1025 non-null    int64  
thalach        1025 non-null    int64  
exang          1025 non-null    int64  
oldpeak        1025 non-null    float64
slope          1025 non-null    int64  
ca             1025 non-null    int64  
thal           1025 non-null    int64  
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	fbstest	restecg
count	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
std	9.072290	0.460373	1.029641	17.516718	51.59251	0.356527	0.527878
min	29.000000	0.000000	0.000000	94.000000	126.000000	0.000000	0.000000
25%	48.000000	0.000000	0.000000	120.000000	211.000000	0.000000	0.000000
50%	56.000000	1.000000	1.000000	130.000000	240.000000	0.000000	1.000000
75%	61.000000	1.000000	2.000000	140.000000	275.000000	0.000000	1.000000
max	77.000000	1.000000	3.000000	200.000000	564.000000	1.000000	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	fb	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
fb	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()
```

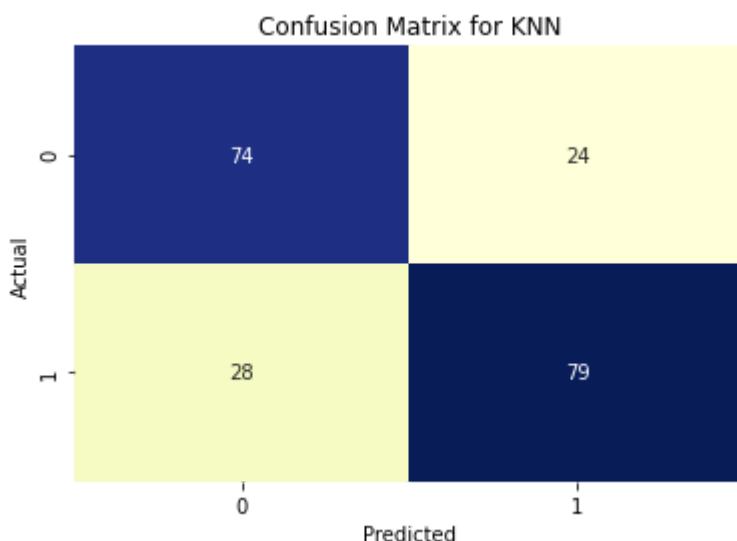
```
In [25]: y_pred2=knn.predict(x_test)
```

```
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
```

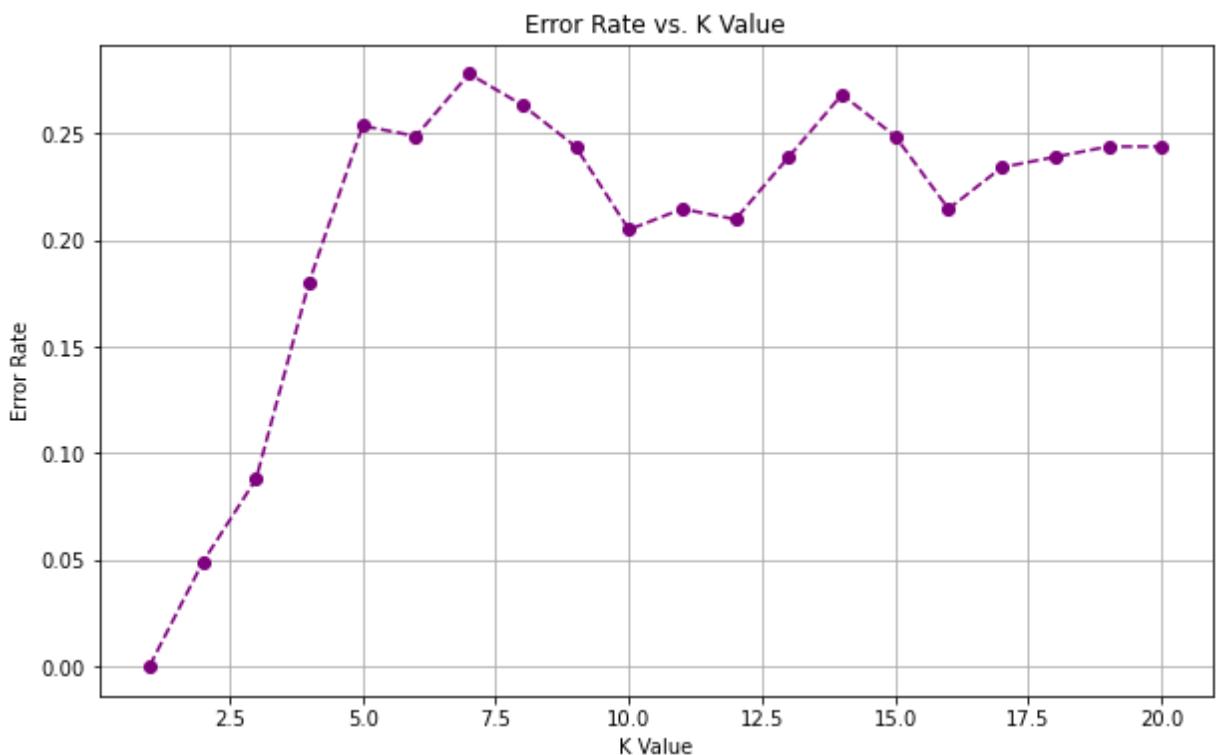
```

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 []: