

Lab-03

January 25, 2026

1 KNN as Classifier [As Regressor in another program]

```
[1]: #Assigning Feature and local variables
#First Feature
weather = ['Sunny', 'Sunny',
'Overcast', 'Rainy', 'Rainy', 'Rainy', 'Overcast', 'Sunny', 'Sunny', 'Rainy',
'Sunny', 'Overcast', 'Overcast', 'Rainy']

[2]: #Second Feature
temp =
↳ ['Hot', 'Hot', 'Hot', 'Mild', 'Cool', 'Cool', 'Cool', 'Mild', 'Cool', 'Mild', 'Mild', 'Mild', 'Hot', 'Mi

[3]: # Label or Target Variable
play = ['No', 'No', 'Yes', 'Yes', 'Yes', 'No', 'Yes', 'No', 'Yes', 'Yes', 'Yes', 'Yes',
'Yes', 'No' ]

[4]: # import Label Encoder
from sklearn import preprocessing

[5]: #creating Label Encoder
le = preprocessing.LabelEncoder()

[6]: #converting Strigns Labels to numbers
weather_encoded = le.fit_transform(weather)

[7]: weather_encoded

[7]: array([2, 2, 0, 1, 1, 1, 0, 2, 2, 1, 2, 0, 0, 1])

[8]: #converting Strigns Labels to numbers
temp_encoded = le.fit_transform(temp)
label_col = le.fit_transform(play)
temp_encoded

[8]: array([1, 1, 1, 2, 0, 0, 0, 2, 0, 2, 2, 2, 1, 2])

[9]: #combining weather and temp into single list of tuples
features= list(zip(weather_encoded,temp_encoded))
```

```
features
```

```
[9]: [(np.int64(2), np.int64(1)),
      (np.int64(2), np.int64(1)),
      (np.int64(0), np.int64(1)),
      (np.int64(1), np.int64(2)),
      (np.int64(1), np.int64(0)),
      (np.int64(1), np.int64(0)),
      (np.int64(0), np.int64(0)),
      (np.int64(2), np.int64(2)),
      (np.int64(2), np.int64(0)),
      (np.int64(1), np.int64(2)),
      (np.int64(2), np.int64(2)),
      (np.int64(0), np.int64(2)),
      (np.int64(0), np.int64(1)),
      (np.int64(1), np.int64(2))]
```

```
[10]: from sklearn.neighbors import KNeighborsClassifier
      model = KNeighborsClassifier(n_neighbors=3)
      model.fit(features, label_col)
      KNeighborsClassifier(n_neighbors=3)
```

```
[10]: KNeighborsClassifier(n_neighbors=3)
```

```
[11]: #predict output
      predicted = model.predict([[1,0]])
      predicted
```

```
[11]: array([1])
```

```
[12]: print(predicted)
```

```
[1]
```

```
[ ]:
```

KNN Classifier

January 25, 2026

1 KNN Classifier in Iris Dataset

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score, confusion_matrix, \
↳classification_report
```

```
[2]: iris = load_iris()
X = iris.data
y = iris.target

df = pd.DataFrame(X, columns=iris.feature_names)
df['target'] = y
df.head()
```

```
[2]:      sepal length (cm)  sepal width (cm)  petal length (cm)  petal width (cm)  \
0                5.1             3.5             1.4             0.2
1                4.9             3.0             1.4             0.2
2                4.7             3.2             1.3             0.2
3                4.6             3.1             1.5             0.2
4                5.0             3.6             1.4             0.2

      target
0         0
1         0
2         0
3         0
4         0
```

```
[3]: X_train, X_test, y_train, y_test = train_test_split( X, y, test_size=0.2, \
↳random_state=42 )
```

```
[4]: scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

```
[5]: knn = KNeighborsClassifier(n_neighbors=5)
knn.fit(X_train, y_train)
```

```
[5]: KNeighborsClassifier()
```

```
[6]: y_pred = knn.predict(X_test)
y_pred
```

```
[6]: array([1, 0, 2, 1, 1, 0, 1, 2, 1, 1, 2, 0, 0, 0, 0, 1, 2, 1, 1, 2, 0, 2,
         0, 2, 2, 2, 2, 2, 0, 0])
```

```
[7]: accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)

print("\nConfusion Matrix:\n", confusion_matrix(y_test, y_pred))
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

Accuracy: 1.0

Confusion Matrix:

```
[[10  0  0]
 [ 0  9  0]
 [ 0  0 11]]
```

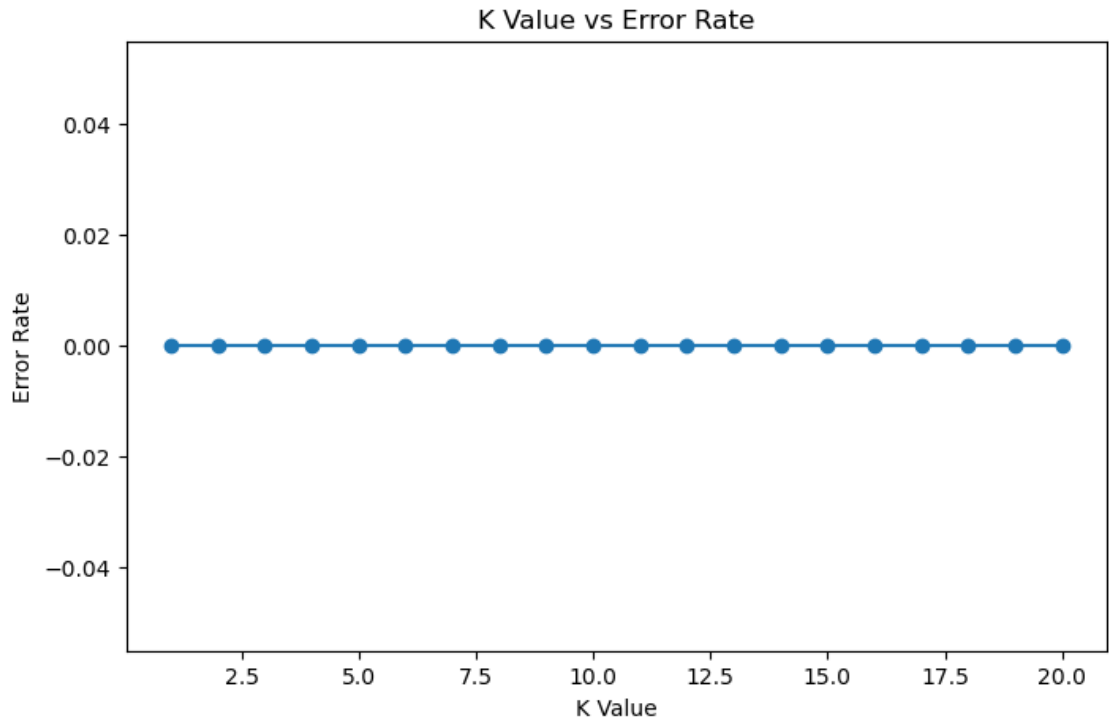
Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	10
1	1.00	1.00	1.00	9
2	1.00	1.00	1.00	11
accuracy			1.00	30
macro avg	1.00	1.00	1.00	30
weighted avg	1.00	1.00	1.00	30

```
[8]: error_rate = []

for k in range(1, 21):
    knn = KNeighborsClassifier(n_neighbors=k)
    knn.fit(X_train, y_train)
    pred_k = knn.predict(X_test)
    error_rate.append(np.mean(pred_k != y_test))
```

```
plt.figure(figsize=(8,5))
plt.plot(range(1,21), error_rate, marker='o')
plt.xlabel('K Value')
plt.ylabel('Error Rate')
plt.title('K Value vs Error Rate')
plt.show()
```



[]:

KNN as Regrassior

January 25, 2026

0.1 KNN AS REGRESSOR

```
[1]: #Step 1: Import Libraries
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsRegressor
from sklearn.metrics import mean_squared_error
```

```
[2]: #Step 2: Create the DataFrame
data = {
    'id': [1, 2, 3, 4, 5, 6, 7, 8, 9, 10],
    'age': [25, 30, 35, 40, 45, 50, 55, 60, 65, 70],
    'height': [170, 160, 180, 165, 175, 160, 170, 180, 165, 175],
    'weight': [65, 60, 80, 70, 75, 55, 70, 85, 60, 75]
}
df = pd.DataFrame(data)
print(df)
```

	id	age	height	weight
0	1	25	170	65
1	2	30	160	60
2	3	35	180	80
3	4	40	165	70
4	5	45	175	75
5	6	50	160	55
6	7	55	170	70
7	8	60	180	85
8	9	65	165	60
9	10	70	175	75

```
[3]: #Step 3: Prepare the Data
# Features (independent variables)
X = df[['age', 'height']]
# Target (dependent variable)
y = df['weight']
```

```
[4]: #Step 4: Split the Data
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)
```

```
[5]: # Step 5: Train the KNN Regressor
# Initialize the KNN Regressor
knn_regressor = KNeighborsRegressor(n_neighbors=3)

# Train the model
knn_regressor.fit(X_train, y_train)
KNeighborsRegressor(n_neighbors=3)
```

```
[5]: KNeighborsRegressor(n_neighbors=3)
```

```
[6]: #Step 6: Make Predictions
# Predict weights for the test set
y_pred = knn_regressor.predict(X_test)

# Display predictions
print("Predicted Weights:", y_pred)
```

```
Predicted Weights: [76.66666667 63.33333333]
```

```
[7]: #Step 7: Evaluate the Model

# Calculate Mean Squared Error
mse = mean_squared_error(y_test, y_pred)
print("Mean Squared Error:", mse)
```

```
Mean Squared Error: 144.444444444444454
```

```
[8]: #Step 8: Predict Weight for New Data

# New input data
new_data = [[32, 172]]

# Predict weight
predicted_weight = knn_regressor.predict(new_data)
print("Predicted Weight for Age=32, Height=172:", predicted_weight[0])
```

```
Predicted Weight for Age=32, Height=172: 71.66666666666667
```

```
/usr/lib/python3/dist-packages/sklearn/utils/validation.py:2749: UserWarning: X
does not have valid feature names, but KNeighborsRegressor was fitted with
feature names
```

```
warnings.warn(
```

```
[ ]:
```

Carprice-KNN Regrassor

January 25, 2026

1 KNN Regrassor in Carprice Dataset

```
[1]: #Step 1: Import Libraries  
import pandas as pd  
from sklearn.model_selection import train_test_split  
from sklearn.neighbors import KNeighborsRegressor  
from sklearn.metrics import mean_squared_error
```

```
[2]: #Step 2: Read Data  
df = pd.read_csv('carprices.csv')
```

```
[3]: df.head(16)
```

```
[3]:
```

	Car Model	Mileage	Sell Price	Age
0	BMW X5	69000	18000	6
1	BMW X5	35000	34000	3
2	BMW X5	57000	26100	5
3	BMW X5	22500	40000	2
4	BMW X5	46000	31500	4
5	Audi	59000	29400	5
6	Audi	52000	32000	5
7	Audi	72000	19300	6
8	Audi	91000	12000	8
9	Mercedez Benz	67000	22000	6
10	Mercedez Benz	83000	20000	7
11	Mercedez Benz	79000	21000	7
12	Mercedez Benz	59000	33000	5
13	Toyota	51000	42000	4
14	Toyota	65000	32000	7
15	Toyota	39000	55000	5

```
[4]: #Step 3: Prepare the Data  
  
# Features (independent variables)  
X = df[['Mileage', 'Age']]  
# Target (dependent variable)  
y = df['Sell Price']
```

```
[5]: #Step 4: Split the Data

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)
```

```
[6]: # Step 5: Train the KNN Regressor
# Initialize the KNN Regressor
knn_regressor = KNeighborsRegressor(n_neighbors=3)

# Train the model
knn_regressor.fit(X_train, y_train)
KNeighborsRegressor(n_neighbors=3)
```

```
[6]: KNeighborsRegressor(n_neighbors=3)
```

```
[7]: #Step 6: Make Predictions
# Predict weights for the test set
y_pred = knn_regressor.predict(X_test)

# Display predictions
print("Predicted Sell Price:", y_pred)
```

```
Predicted Sell Price: [20766.66666667 42166.66666667 30366.66666667
24766.66666667]
```

```
[8]: #Step 7: Evaluate the Model

# Calculate Mean Squared Error
mse = mean_squared_error(y_test, y_pred)
print("Mean Squared Error:", mse)
```

```
Mean Squared Error: 31901111.111111097
```

```
[9]: #Step 8: Predict Sell Price for New Data

# New input data
new_data = [[48000, 3]]

# Predict weight
predicted_price = knn_regressor.predict(new_data)
print("Predicted Sell Price for Using Age=3, Mileage=48000, Sell_Price=",
↪predicted_price[0])
```

```
Predicted Sell Price for Using Age=3, Mileage=48000, Sell_Price=
35166.666666666664
```

```
/usr/lib/python3/dist-packages/sklearn/utils/validation.py:2749: UserWarning: X
does not have valid feature names, but KNeighborsRegressor was fitted with
feature names
warnings.warn(
```

```
[10]: df.head(17)
```

```
[10]:
```

	Car Model	Mileage	Sell Price	Age
0	BMW X5	69000	18000	6
1	BMW X5	35000	34000	3
2	BMW X5	57000	26100	5
3	BMW X5	22500	40000	2
4	BMW X5	46000	31500	4
5	Audi	59000	29400	5
6	Audi	52000	32000	5
7	Audi	72000	19300	6
8	Audi	91000	12000	8
9	Mercedez Benz	67000	22000	6
10	Mercedez Benz	83000	20000	7
11	Mercedez Benz	79000	21000	7
12	Mercedez Benz	59000	33000	5
13	Toyota	51000	42000	4
14	Toyota	65000	32000	7
15	Toyota	39000	55000	5

```
[ ]:
```