

```
In [3]: #Import Libraries
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [4]: #Load the Dataset
df = pd.read_csv("glass.csv")
print(df.head())
```

	RI	Na	Mg	Al	Si	K	Ca	Ba	Fe	Type
0	1.52101	13.64	4.49	1.10	71.78	0.06	8.75	0.0	0.0	1
1	1.51761	13.89	3.60	1.36	72.73	0.48	7.83	0.0	0.0	1
2	1.51618	13.53	3.55	1.54	72.99	0.39	7.78	0.0	0.0	1
3	1.51766	13.21	3.69	1.29	72.61	0.57	8.22	0.0	0.0	1
4	1.51742	13.27	3.62	1.24	73.08	0.55	8.07	0.0	0.0	1

```
In [5]: #Check for Missing Values
print(df.isnull().sum())
```

	RI	Na	Mg	Al	Si	K	Ca	Ba	Fe	Type
RI	0									
Na	0									
Mg	0									
Al	0									
Si	0									
K	0									
Ca	0									
Ba	0									
Fe	0									
Type	0									

dtype: int64

```
In [6]: df = df.fillna(df.mean())
```

```
In [7]: #Split Features and Target
X = df.drop("Type", axis=1)      # input features
y = df["Type"]                  # target variable
#If the target variable is not numeric (e.g., "building_windows_float"), use LabelEncoder:
le = LabelEncoder()
y = le.fit_transform(y)
```

```
In [8]: #Train-Test Split
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42
)
```

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In [9]: #Data Scaling (Optional)
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

```
In [10]: #Train the Model
model = DecisionTreeClassifier(random_state=42)
model.fit(X_train, y_train)
```

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Out[10]: ▾ DecisionTreeClassifier ⓘ ?
```

► Parameters

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In [11]: #Make Predictions
y_pred = model.predict(X_test)
```

```
In [12]: #Evaluate the Model
print("Accuracy:", accuracy_score(y_test, y_pred))
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

Accuracy: 0.7209302325581395

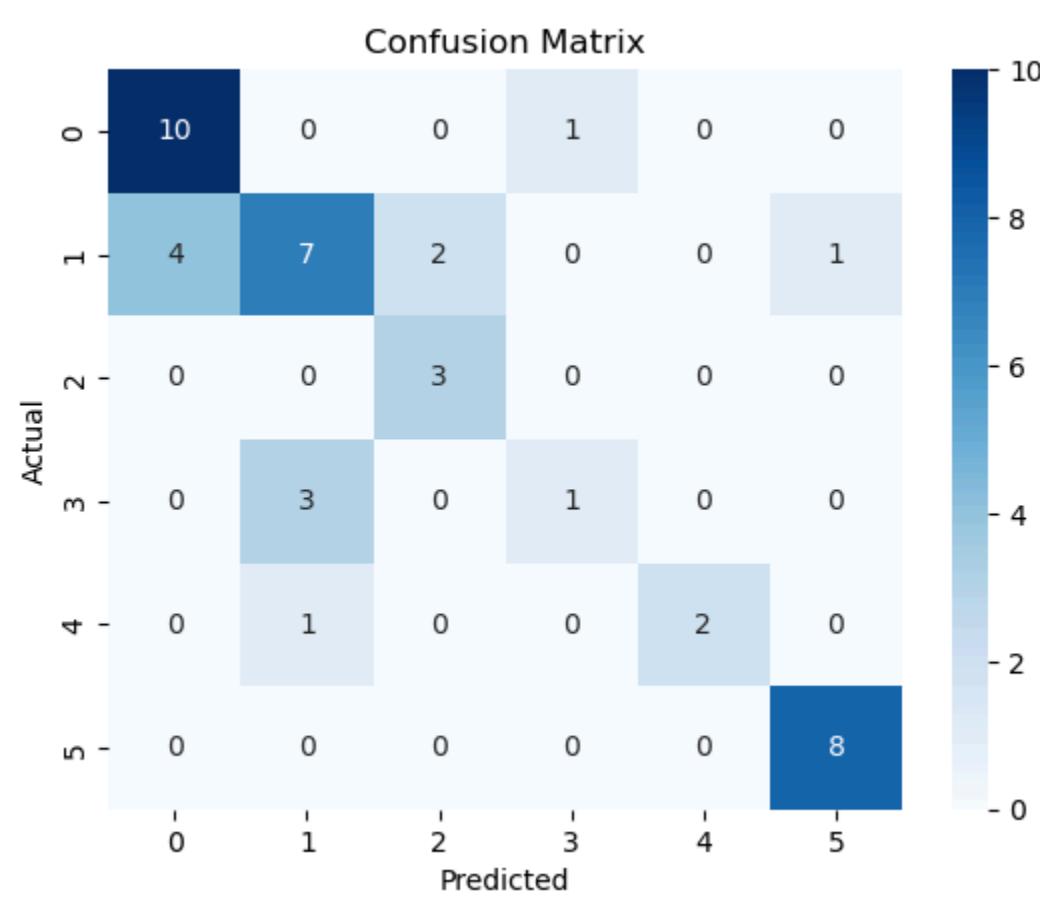
	precision	recall	f1-score	support
0	0.71	0.91	0.80	11
1	0.64	0.50	0.56	14
2	0.60	1.00	0.75	3
3	0.50	0.25	0.33	4
4	1.00	0.67	0.80	3
5	0.89	1.00	0.94	8

	accuracy			
accuracy			0.72	43
macro avg	0.72	0.72	0.70	43
weighted avg	0.71	0.72	0.70	43

```
In [13]: #Confusion Matrix
cm = confusion_matrix(y_test, y_pred)

sns.heatmap(cm, annot=True, cmap='Blues')
plt.title("Confusion Matrix")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.show()
```



In [ ]: