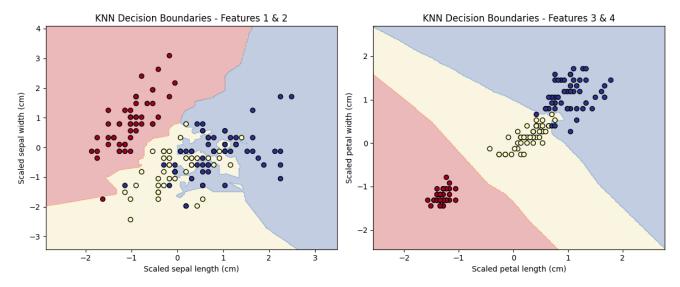
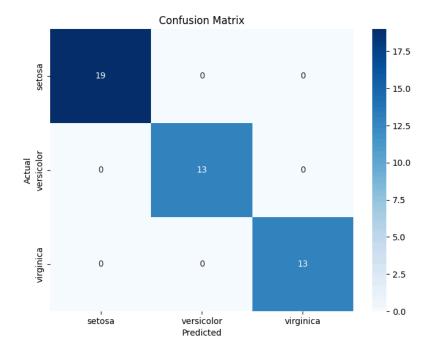
```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
{\color{red} \textbf{from sklearn.metrics import accuracy\_score, confusion\_matrix}}
from sklearn.preprocessing import StandardScaler
import seaborn as sns
iris = load iris()
X = iris.data
y = iris.target
feature_names = iris.feature_names
target_names = iris.target_names
 X\_train, \ X\_test, \ y\_train, \ y\_test = train\_test\_split(X, \ y, \ test\_size=0.3, \ random\_state=42) 
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
k_range = range(1, 20)
k_scores = []
for k in k_range:
    knn = KNeighborsClassifier(n_neighbors=k)
    knn.fit(X_train_scaled, y_train)
    y_pred = knn.predict(X_test_scaled)
    k\_scores.append(accuracy\_score(y\_test, y\_pred))
plt.figure(figsize=(10, 6))
\verb|plt.plot(k_range, k_scores, marker='o')|
plt.xlabel('Number of Neighbors (k)')
plt.ylabel('Accuracy')
plt.title('KNN Accuracy vs. k Value')
plt.grid(True)
plt.show()
optimal_k = k_range[np.argmax(k_scores)]
print(f"Optimal k: {optimal_k}")
knn = KNeighborsClassifier(n_neighbors=optimal_k)
knn.fit(X_train_scaled, y_train)
y_pred = knn.predict(X_test_scaled)
print(f"Accuracy with k={optimal_k}: {accuracy_score(y_test, y_pred):.4f}")
plt.figure(figsize=(8, 6))
cm = confusion_matrix(y_test, y_pred)
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=target_names, yticklabels=target_names)
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.title('Confusion Matrix')
plt.show()
plt.figure(figsize=(12, 5))
for i, pair in enumerate([(0, 1), (2, 3)]):
    plt.subplot(1, 2, i+1)
    x_idx, y_idx = pair
    X_{subset} = X[:, [x_{idx}, y_{idx}]]
    X_subset_scaled = scaler.fit_transform(X_subset)
    knn = KNeighborsClassifier(n_neighbors=optimal_k)
    knn.fit(X_subset_scaled, y)
    h = 0.02
   xx, yy = np.meshgrid(np.arange(x_min, x_max, h), np.arange(y_min, y_max, h))
    Z = knn.predict(np.c_[xx.ravel(), yy.ravel()])
    Z = Z.reshape(xx.shape)
    \verb|plt.contourf(xx, yy, Z, alpha=0.3, cmap=plt.cm.RdYlBu)|\\
    scatter = plt.scatter(X\_subset\_scaled[:, \ 0], \ X\_subset\_scaled[:, \ 1], \ c=y, \ edgecolors='k', \ cmap=plt.cm.RdYlBu)
    plt.xlabel(f'Scaled {feature_names[x_idx]}')
    plt.ylabel(f'Scaled {feature_names[y_idx]}')
```

```
\label{eq:plt.title} plt.title(f'KNN Decision Boundaries - Features \{x\_idx+1\} \ \& \ \{y\_idx+1\}') \\ \\ plt.tight\_layout() \\ plt.show() \\ \\ \\
```



Figure_3.png



Figure_2.png

• ishadpande@fedora:~/Music/Documents/College/Sem 6/Practicals\$ /bin/python "/hom Optimal k: 3 Accuracy with k=3: 1.0000

image-2.png