

```


import pandas as pd
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
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
from sklearn.datasets import load_iris, load_diabetes
import pandas as pd
from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
import seaborn as sns
import matplotlib.pyplot as plt

```

```

from google.colab import files
uploaded = files.upload()


```




 Choose Files iris.csv
 • iris.csv(text/csv) - 4617 bytes, last modified: 4/21/2025 - 100% done
 Saving iris.csv to iris.csv

```

df = pd.read_csv('iris.csv')
df

```



	sepal_length	sepal_width	petal_length	petal_width	species	
0	5.1	3.5	1.4	0.2	Iris-setosa	
1	4.9	3.0	1.4	0.2	Iris-setosa	
2	4.7	3.2	1.3	0.2	Iris-setosa	
3	4.6	3.1	1.5	0.2	Iris-setosa	
4	5.0	3.6	1.4	0.2	Iris-setosa	
...	
145	6.7	3.0	5.2	2.3	Iris-virginica	
146	6.3	2.5	5.0	1.9	Iris-virginica	
147	6.5	3.0	5.2	2.0	Iris-virginica	
148	6.2	3.4	5.4	2.3	Iris-virginica	
149	5.9	3.0	5.1	1.8	Iris-virginica	

150 rows × 5 columns

Next steps: [Generate code with df](#) [View recommended plots](#) [New interactive sheet](#)

```

X = df[['petal_length', 'petal_width']]

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```

scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

```

```

inertia = []
k_range = range(1, 11)
for k in k_range:
    kmeans = KMeans(n_clusters=k, random_state=42)
    kmeans.fit(X_scaled)
    inertia.append(kmeans.inertia_)

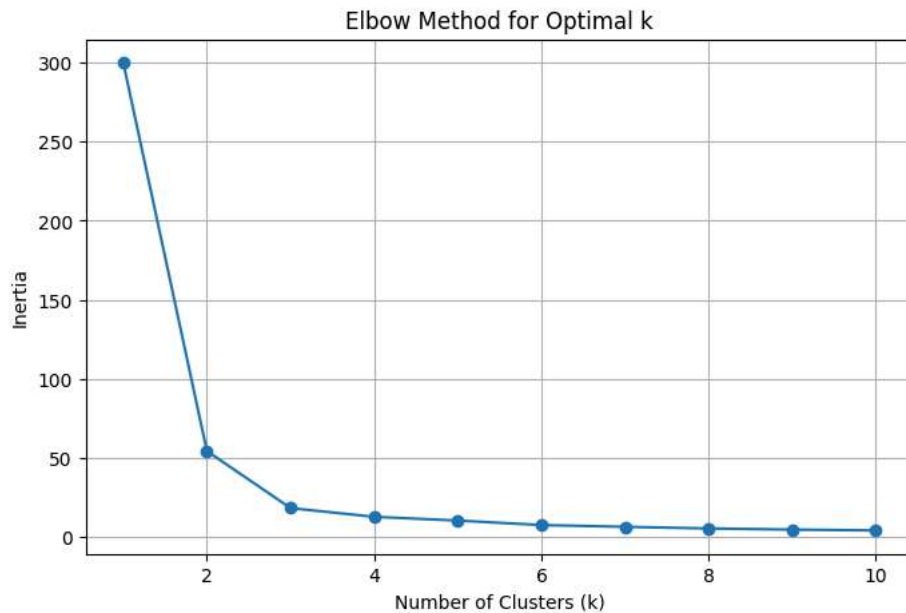
```

```

plt.figure(figsize=(8, 5))
plt.plot(k_range, inertia, marker='o')
plt.title("Elbow Method for Optimal k")

```

```
plt.xlabel("Number of Clusters (k)")
plt.ylabel("Inertia")
plt.grid(True)
plt.show()
```



```
optimal_k = 3
kmeans = KMeans(n_clusters=optimal_k, random_state=42)
clusters = kmeans.fit_predict(X_scaled)
df['cluster'] = clusters
```

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df['cluster'] = clusters
```

```
plt.scatter(kmeans.cluster_centers_[0, 0], kmeans.cluster_centers_[0, 1],
            s=200, c='black', marker='x', label='Centroids')
```

```
plt.title("K-Means Clusters on Petal Features")
plt.xlabel("Petal Length (scaled)")
plt.ylabel("Petal Width (scaled)")
plt.legend()
plt.grid(True)
plt.show()
```

