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problem statement - Implement K-Means clustering on Iris.csv dataset. Determine the number of clusters using the elbow method.

```
In [5]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
```

In [ ]:

```
In [7]: df = pd.read_csv('Iris.csv')
```

In [ ]:

```
In [8]: df
```

```
Out[8]:
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
<b>0</b>	1	5.1	3.5	1.4	0.2	Iris-setosa
<b>1</b>	2	4.9	3.0	1.4	0.2	Iris-setosa
<b>2</b>	3	4.7	3.2	1.3	0.2	Iris-setosa
<b>3</b>	4	4.6	3.1	1.5	0.2	Iris-setosa
<b>4</b>	5	5.0	3.6	1.4	0.2	Iris-setosa
...	...	...	...	...	...	...
<b>145</b>	146	6.7	3.0	5.2	2.3	Iris-virginica
<b>146</b>	147	6.3	2.5	5.0	1.9	Iris-virginica
<b>147</b>	148	6.5	3.0	5.2	2.0	Iris-virginica
<b>148</b>	149	6.2	3.4	5.4	2.3	Iris-virginica
<b>149</b>	150	5.9	3.0	5.1	1.8	Iris-virginica

150 rows × 6 columns

```
In [9]: df.drop('Id',axis=1,inplace=True)
```

```
In [10]: df
```

```
Out[10]:
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	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

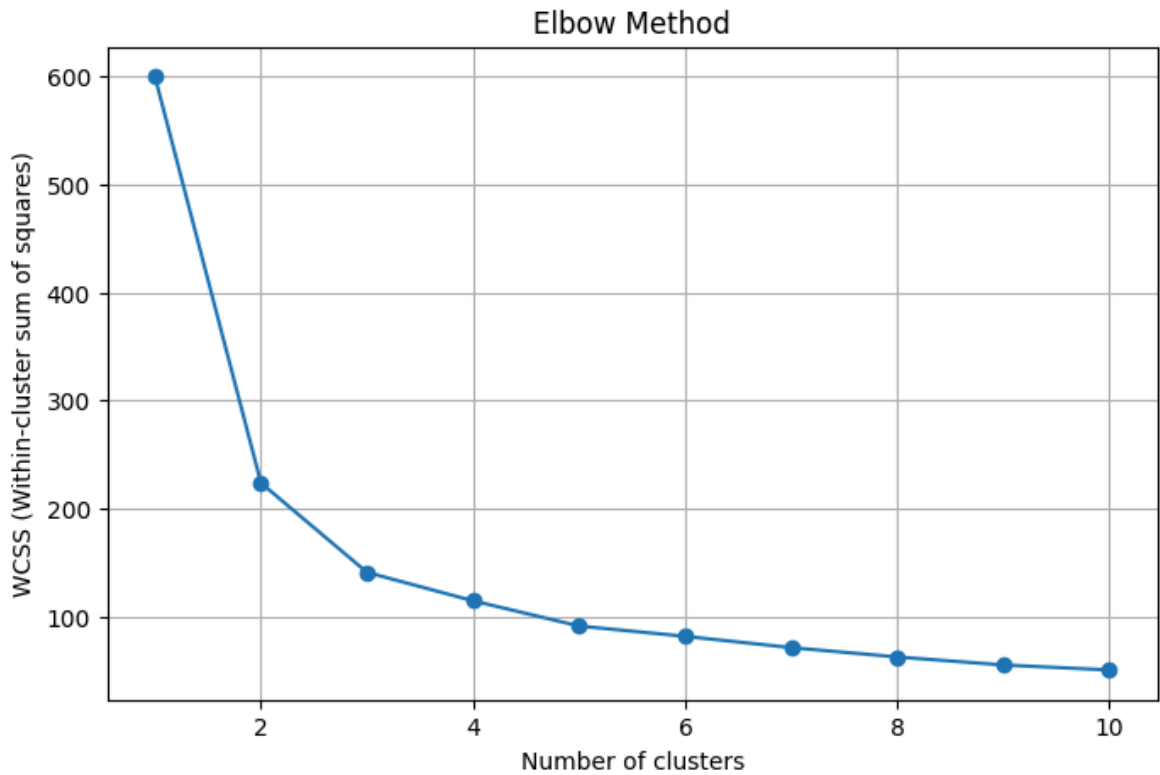
150 rows × 5 columns

```
In [11]: X = df[['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm']]
```

```
In [12]: scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
```

```
In [13]: wcss = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters=i, init='k-means++', max_iter=300, n_init=10, ran
    kmeans.fit(X_scaled)
    wcss.append(kmeans.inertia_)
```

```
In [14]: plt.figure(figsize=(8,5))
plt.plot(range(1, 11), wcss, marker='o')
plt.title('Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS (Within-cluster sum of squares)')
plt.grid(True)
plt.show()
```

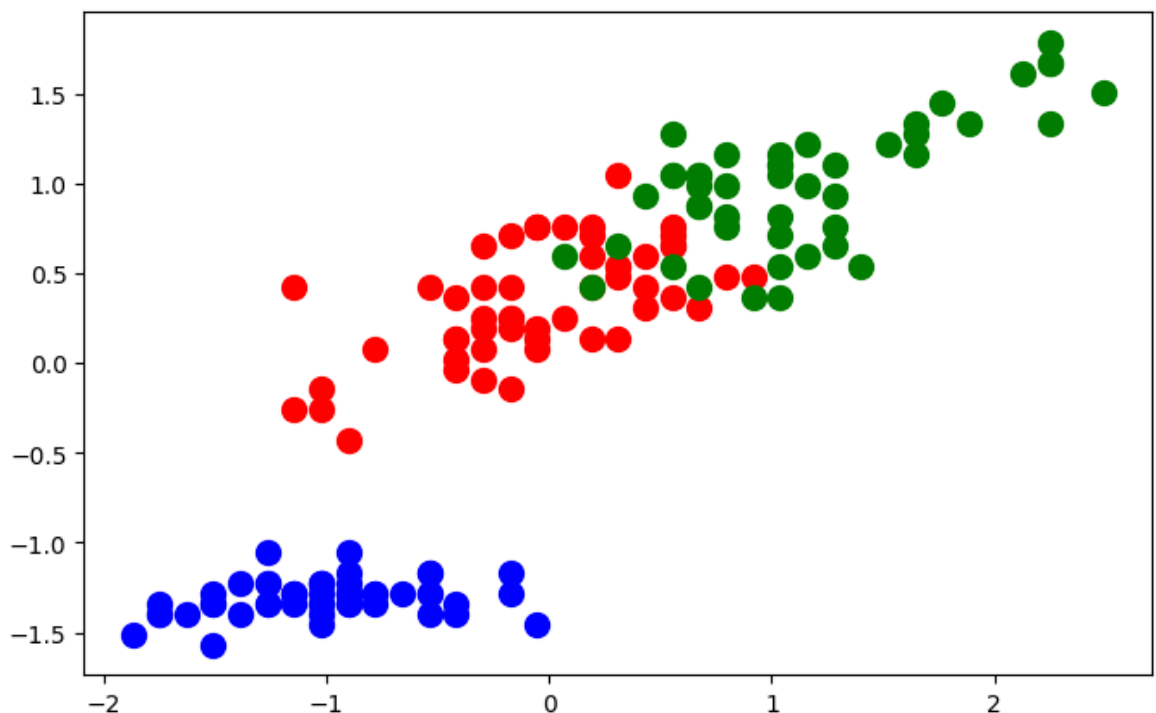


```
In [15]: kmeans = KMeans(n_clusters=3, init='k-means++', max_iter=300, n_init=10, random_state=0)
y_kmeans = kmeans.fit_predict(X_scaled)
```

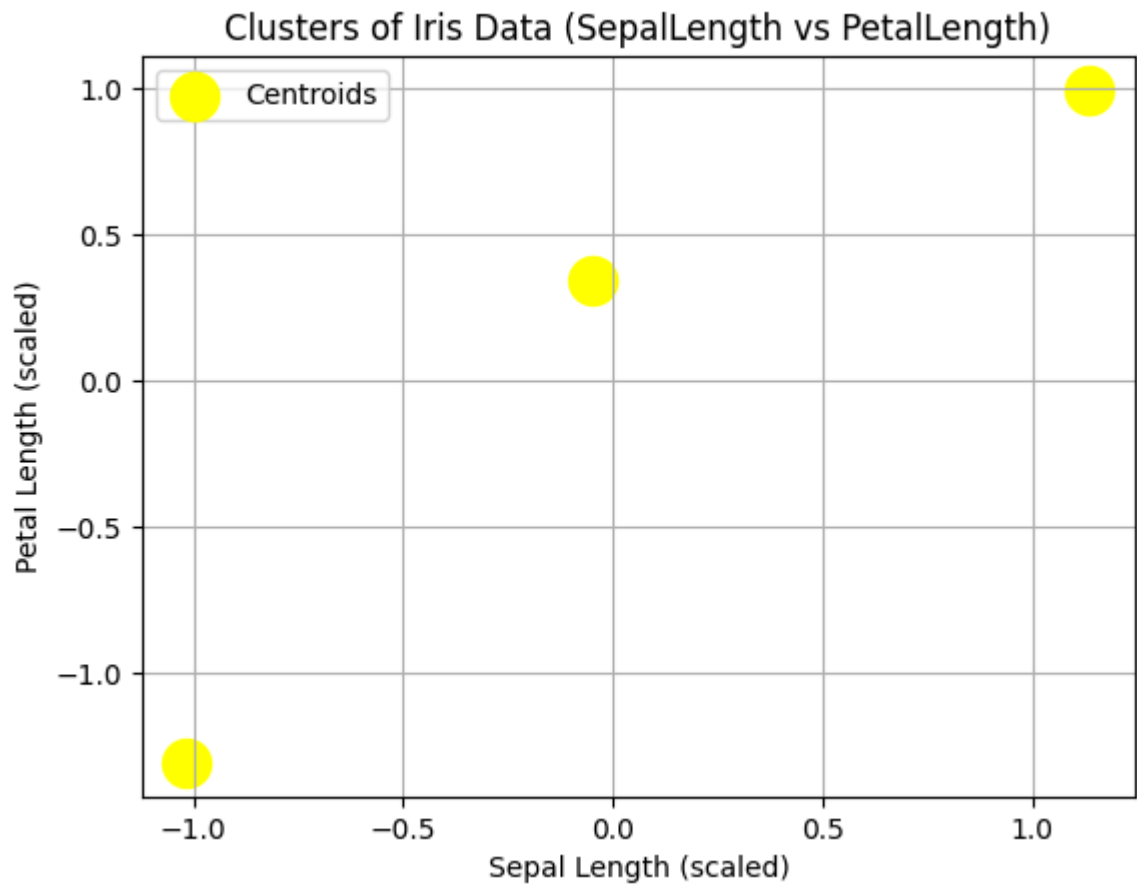
```
In [16]: df['Cluster'] = y_kmeans
```

```
In [17]: plt.figure(figsize=(8,5))
plt.scatter(X_scaled[y_kmeans == 0, 0], X_scaled[y_kmeans == 0, 2], s=100, c='red')
plt.scatter(X_scaled[y_kmeans == 1, 0], X_scaled[y_kmeans == 1, 2], s=100, c='blue')
plt.scatter(X_scaled[y_kmeans == 2, 0], X_scaled[y_kmeans == 2, 2], s=100, c='green')
```

```
Out[17]: <matplotlib.collections.PathCollection at 0x262d35952d0>
```



```
In [18]: plt.scatter(kmeans.cluster_centers_[0], kmeans.cluster_centers_[2], s=300,  
plt.title('Clusters of Iris Data (SepalLength vs PetalLength)')  
plt.xlabel('Sepal Length (scaled)')  
plt.ylabel('Petal Length (scaled)')  
plt.legend()  
plt.grid(True)  
plt.show()
```



In [ ]: