

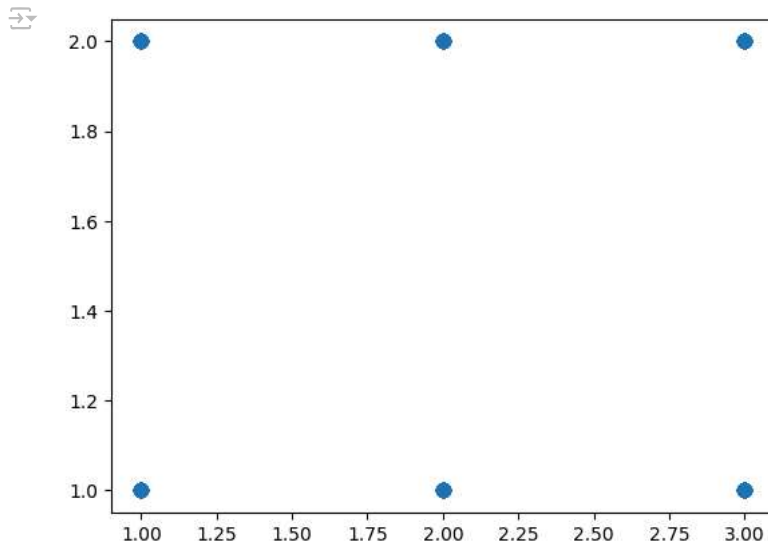
## K-Means Clustering

```
# Import necessary libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from sklearn.datasets import make_blobs

df = pd.read_csv("/content/titanic_cleaned.csv")

X = df.drop('Survived',axis=1)
y = df['Survived']

plt.scatter(X.iloc[:, 0], X.iloc[:, 1], s=50)
plt.show()
```



```
# Apply K-Means clustering
kmeans = KMeans(n_clusters=3)
kmeans.fit(X)
```

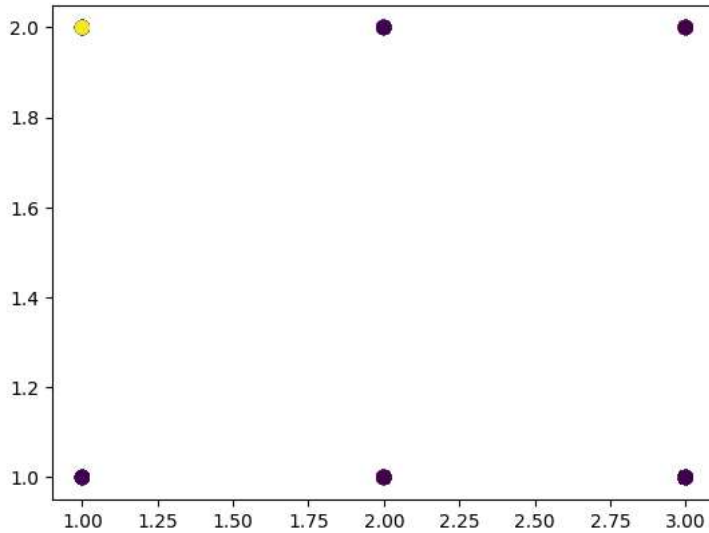
▼ KMeans ⓘ ?

KMeans(n\_clusters=3)

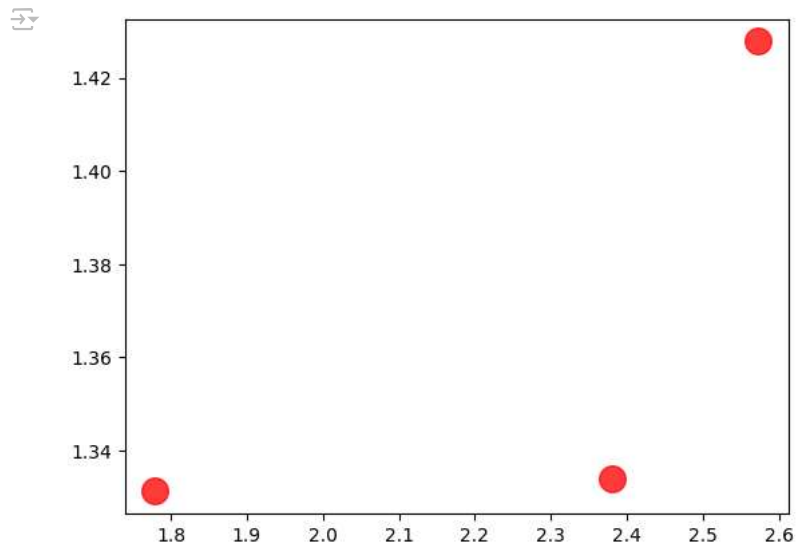
```
# Get the cluster centers and labels
centroids = kmeans.cluster_centers_
labels = kmeans.labels_
```

```
# Visualize the clusters
plt.scatter(X.iloc[:, 0], X.iloc[:, 1], c=labels, s=50, cmap='viridis')
```

 <matplotlib.collections.PathCollection at 0x7cafaeb2e8f0>



```
plt.scatter(centroids[:, 0], centroids[:, 1], c='red', s=200, alpha=0.75)
plt.show()
```



```
# Elbow method to determine the optimal number of clusters
```

```
inertia = []
```

```
K = range(1, 10)
```

```
for k in K:
```

```
    kmeans = KMeans(n_clusters=k)
```

```
    kmeans.fit(X)
```

```
    inertia.append(kmeans.inertia_)
```

```
# Plot the elbow curve
```

```
plt.plot(K, inertia, 'bx-')
```

```
plt.xlabel('Number of clusters')
```

```
plt.ylabel('Inertia')
```

```
plt.title('Elbow Method For Optimal k')
```

```
plt.show()
```

