

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

dataset = pd.read_csv('E:/sistem-cerdas/Customer.csv')
dataset.head()
```

```
Out[1]:
```

	IDPelanggan	Kelamin	Usia	Rating_belanja (1-100)	Pendapatan (juta Rp)
0	1	Perempuan	23	87	29
1	2	Laki	60	4	30
2	3	Perempuan	21	73	30
3	4	Laki	53	4	33
4	5	Laki	18	92	33

```
In [2]: X = dataset.iloc[:,10, [2,3,4]].values
X
```

```
Out[2]: array([[23, 87, 29],
               [60, 4, 30],
               [21, 73, 30],
               [53, 4, 33],
               [18, 92, 33],
               [49, 14, 33],
               [21, 81, 33],
               [42, 17, 34],
               [30, 73, 34],
               [36, 26, 37]], dtype=int64)
```

```
In [3]: from sklearn.cluster import KMeans

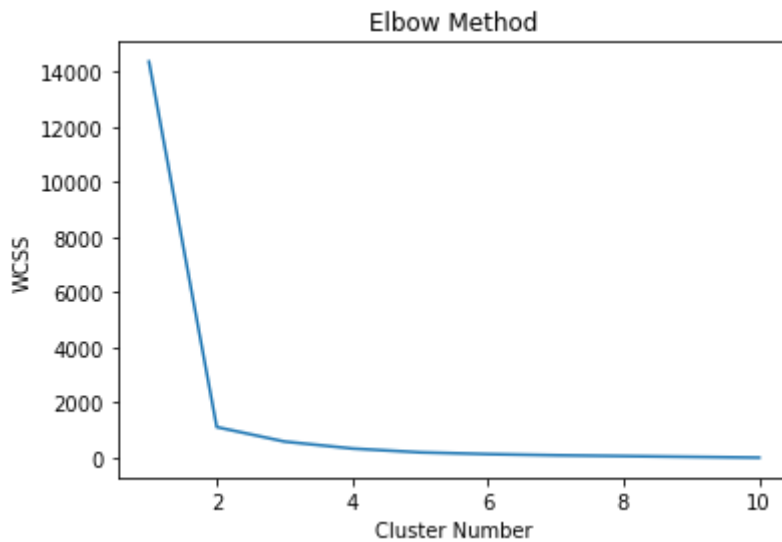
wcss = []

for i in range(1, 11):
    kmeans = KMeans(n_clusters = i, init = 'k-means++', random_state = 42)
    kmeans.fit(X)
    wcss.append(kmeans.inertia_)

plt.plot(range(1, 11), wcss)
plt.title('Elbow Method')
plt.xlabel('Cluster Number')
plt.ylabel('WCSS')
plt.show()
```

D:\Program Files 2\Anaconda\lib\site-packages\sklearn\cluster_kmeans.py:1038: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

```
warnings.warn(
```



```
In [4]: kmeans = KMeans(n_clusters = 1, init = 'k-means++', random_state = 42)
y_kmeans = kmeans.fit_predict(X)

plt.scatter(X[y_kmeans == 0, 0], X[y_kmeans == 0, 1], s = 100, c = 'blue', label = 'Cluster 1')
plt.scatter(kmeans.cluster_centers_[0, 0], kmeans.cluster_centers_[0, 1], s = 300, c = 'black', label = 'Centroids')
plt.title('Consumers Cluster')
plt.xlabel('Yearly Salary')
plt.ylabel('Yearly expense rating (1-100)')
plt.legend()
plt.show()
```

D:\Program Files 2\Anaconda\lib\site-packages\sklearn\cluster_kmeans.py:1038: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(



```
In [5]: kmeans = KMeans(n_clusters = 2, init = 'k-means++', random_state = 42)
y_kmeans = kmeans.fit_predict(X)

plt.scatter(X[y_kmeans == 0, 0], X[y_kmeans == 0, 1], s = 100, c = 'blue', label = 'Cluster 1')
plt.scatter(X[y_kmeans == 1, 0], X[y_kmeans == 1, 1], s = 100, c = 'red', label = 'Cluster 2')
plt.scatter(kmeans.cluster_centers_[0, 0], kmeans.cluster_centers_[0, 1], s = 300, c = 'black', label = 'Centroids')
plt.title('Consumers Cluster')
plt.xlabel('Yearly Salary')
plt.ylabel('Yearly expense rating (1-100)')
```

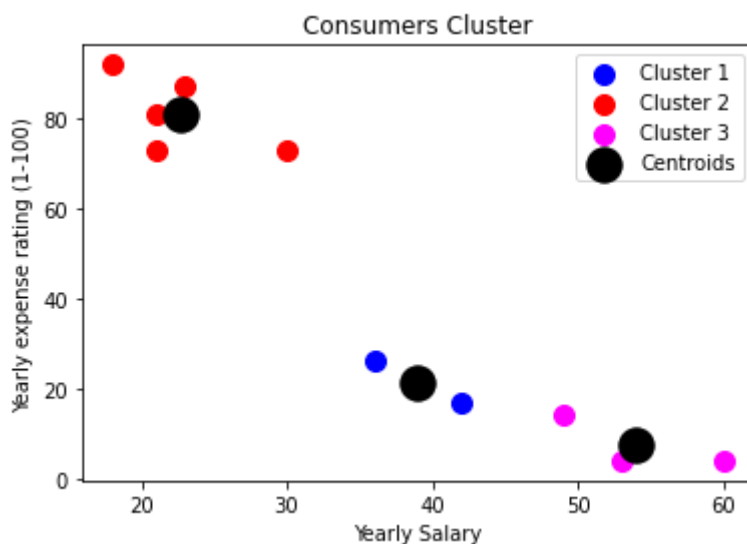
```
plt.legend()
plt.show()
```



In [6]:

```
kmeans = KMeans(n_clusters = 3, init = 'k-means++', random_state = 42)
y_kmeans = kmeans.fit_predict(X)

plt.scatter(X[y_kmeans == 0, 0], X[y_kmeans == 0, 1], s = 100, c = 'blue', label = 'Cluster 1')
plt.scatter(X[y_kmeans == 1, 0], X[y_kmeans == 1, 1], s = 100, c = 'red', label = 'Cluster 2')
plt.scatter(X[y_kmeans == 2, 0], X[y_kmeans == 2, 1], s = 100, c = 'magenta', label = 'Cluster 3')
plt.scatter(kmeans.cluster_centers[:, 0], kmeans.cluster_centers[:, 1], s = 300, c = 'black', label = 'Centroids')
plt.title('Consumers Cluster')
plt.xlabel('Yearly Salary')
plt.ylabel('Yearly expense rating (1-100)')
plt.legend()
plt.show()
```



In [7]:

```
kmeans = KMeans(n_clusters = 4, init = 'k-means++', random_state = 42)
y_kmeans = kmeans.fit_predict(X)

plt.scatter(X[y_kmeans == 0, 0], X[y_kmeans == 0, 1], s = 100, c = 'blue', label = 'Cluster 1')
plt.scatter(X[y_kmeans == 1, 0], X[y_kmeans == 1, 1], s = 100, c = 'red', label = 'Cluster 2')
plt.scatter(X[y_kmeans == 2, 0], X[y_kmeans == 2, 1], s = 100, c = 'magenta', label = 'Cluster 3')
plt.scatter(X[y_kmeans == 3, 0], X[y_kmeans == 3, 1], s = 100, c = 'cyan', label = 'Cluster 4')
plt.scatter(kmeans.cluster_centers[:, 0], kmeans.cluster_centers[:, 1], s = 300, c = 'black', label = 'Centroids')
plt.title('Consumers Cluster')
plt.xlabel('Yearly Salary')
```

```
plt.ylabel('Yearly expense rating (1-100)')
plt.legend()
plt.show()
```



In [8]:

```
kmeans = KMeans(n_clusters = 5, init = 'k-means++', random_state = 42)
y_kmeans = kmeans.fit_predict(X)

plt.scatter(X[y_kmeans == 0, 0], X[y_kmeans == 0, 1], s = 100, c = 'blue', label = 'Cluster 1')
plt.scatter(X[y_kmeans == 1, 0], X[y_kmeans == 1, 1], s = 100, c = 'red', label = 'Cluster 2')
plt.scatter(X[y_kmeans == 2, 0], X[y_kmeans == 2, 1], s = 100, c = 'magenta', label = 'Cluster 3')
plt.scatter(X[y_kmeans == 3, 0], X[y_kmeans == 3, 1], s = 100, c = 'cyan', label = 'Cluster 4')
plt.scatter(X[y_kmeans == 4, 0], X[y_kmeans == 4, 1], s = 100, c = 'green', label = 'Cluster 5')
plt.scatter(kmeans.cluster_centers[:, 0], kmeans.cluster_centers[:, 1], s = 300, c = 'black', label = 'Centroids')
plt.title('Consumers Cluster')
plt.xlabel('Yearly Salary')
plt.ylabel('Yearly expense rating (1-100)')
plt.legend()
plt.show()
```



In []: