Assignment 3 ME 793

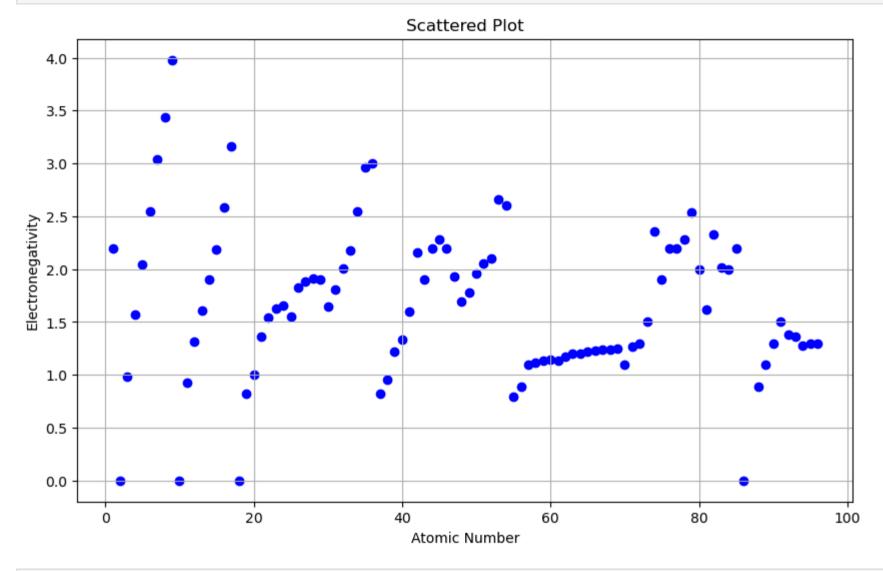
Roll No.: 19D100011

```
In [25]: import pandas as pd
          import numpy as np
          from sklearn.cluster import KMeans
          from scipy.spatial.distance import cdist
          import matplotlib.pyplot as plt
          import warnings
          warnings.filterwarnings("ignore", category=UserWarning)
          warnings.filterwarnings("ignore", category=FutureWarning)
In [26]: # Loading the data
          data = pd.read csv("material data.csv")
          data.head()
Out[26]:
             Elements Atomic_Number Electronegativity Atomic_Radius Thermal_Conductivity Density Crystal_System
          0
                   Н
                                                 2.20
                                                               0.25
                                                                                 0.1805
                                                                                           0.09
                                                                                                          HEX
                                                 0.00
                                                               1.20
                                                                                           0.18
                  He
                                                                                 0.1513
                                                                                                         HCP
          2
                   Li
                                   3
                                                 0.98
                                                               1.45
                                                                                85.0000
                                                                                         530.00
                                                                                                          BCC
          3
                  Be
                                                 1.57
                                                               1.05
                                                                               190.0000 1850.00
                                                                                                         HCP
          4
                                   5
                                                               0.85
                    В
                                                 2.04
                                                                                27.0000 2340.00
                                                                                                         RHO
```

Plotting electronegativity vs. atomic number.

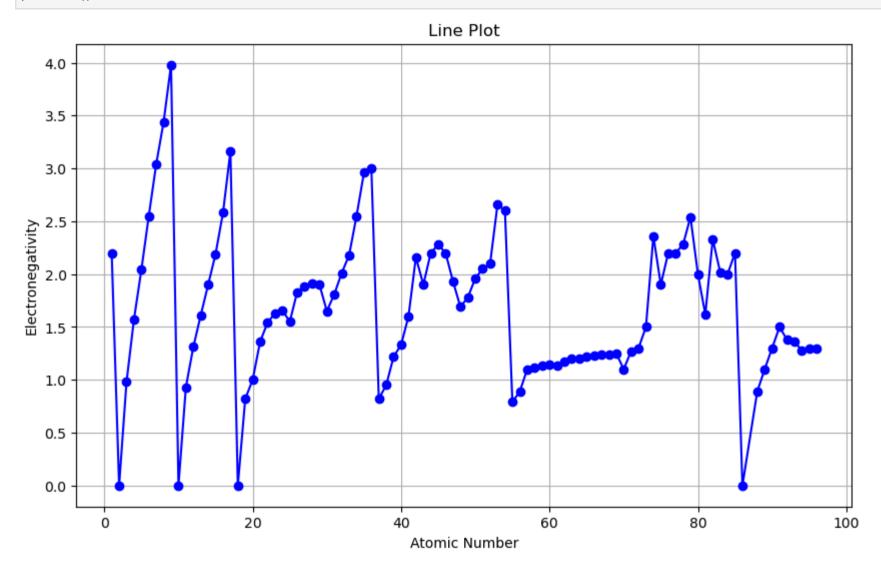
```
In [27]: plt.figure(figsize=(10, 6))
    plt.scatter(data['Atomic_Number'], data['Electronegativity'], color='blue')
    plt.title('Scattered Plot')
    plt.xlabel('Atomic Number')
```

```
plt.ylabel('Electronegativity')
plt.grid(True)
plt.show()
```



```
In [28]: plt.figure(figsize=(10, 6))
    plt.plot(data['Atomic_Number'], data['Electronegativity'], marker='o', color='blue')
    plt.title('Line Plot')
    plt.xlabel('Atomic Number')
    plt.ylabel('Electronegativity')
```

plt.grid(True)
plt.show()



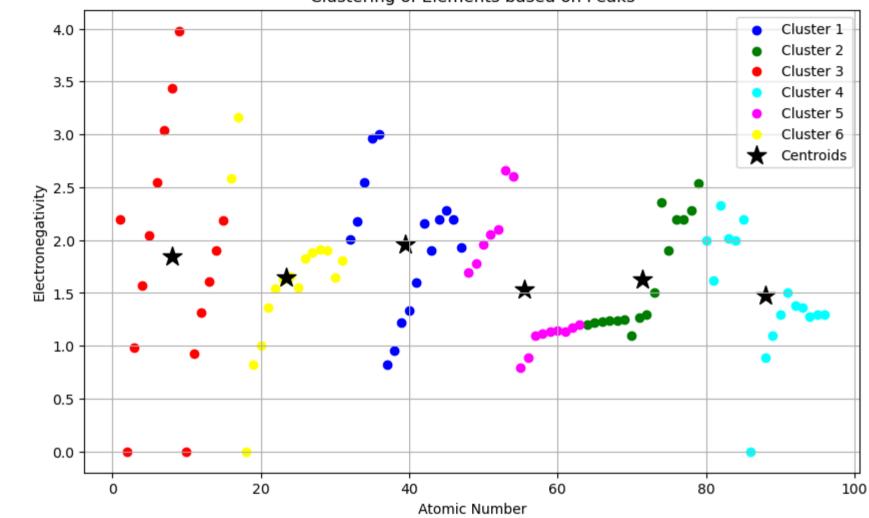
Number of Peaks: 6

Based on Electronegativity vs. Atomic Number graph there are 6 peaks

Performing K Means clustering

```
In [29]: # Select relevant features for clustering
         features = data[['Electronegativity', 'Atomic Number']]
          k = 6
          # Color map
          colors = ['blue', 'green', 'red', 'cyan', 'magenta', 'yellow', 'purple']
         # Perform KMeans clustering
          kmeans = KMeans(n clusters=k, random state=42)
         data['Cluster'] = kmeans.fit predict(features) + 1
          # PLot clusters
          plt.figure(figsize=(10, 6))
          # Plot each cluster
         for cluster num, color in zip(range(1, k+1), colors):
             cluster_data = data[data['Cluster'] == cluster_num]
             plt.scatter(cluster data['Atomic Number'], cluster data['Electronegativity'], color=color, label=f'Cluster {cluster num}')
          # Plot centroids
          centroids = kmeans.cluster centers
          plt.scatter(centroids[:, 1], centroids[:, 0], marker='*', s=200, c='black', label='Centroids')
         # Labels and Legend
         plt.xlabel('Atomic Number')
         plt.ylabel('Electronegativity')
         plt.title('Clustering of Elements based on Peaks')
          plt.legend()
         plt.grid(True)
          plt.show()
```





In [30]: data.to_csv('data.csv', index=False)

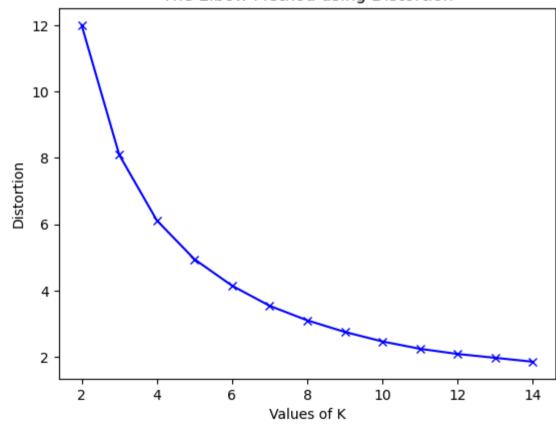
Elbow Method

Finding best k using the Elbow method using Distortion

```
In [31]: distortions = []
K = range(2, 15)
for k in K:
    k_means = KMeans(n_clusters=k, random_state=42)
    k_means.fit(features)
    distortions.append(sum(np.min(cdist(features, k_means.cluster_centers_,'euclidean'), axis=1)) / features.shape[0])

In [32]: plt.plot(K, distortions, 'bx-')
plt.xlabel('Values of K')
plt.ylabel('Distortion')
plt.title('The Elbow Method using Distortion')
plt.show()
```

The Elbow Method using Distortion

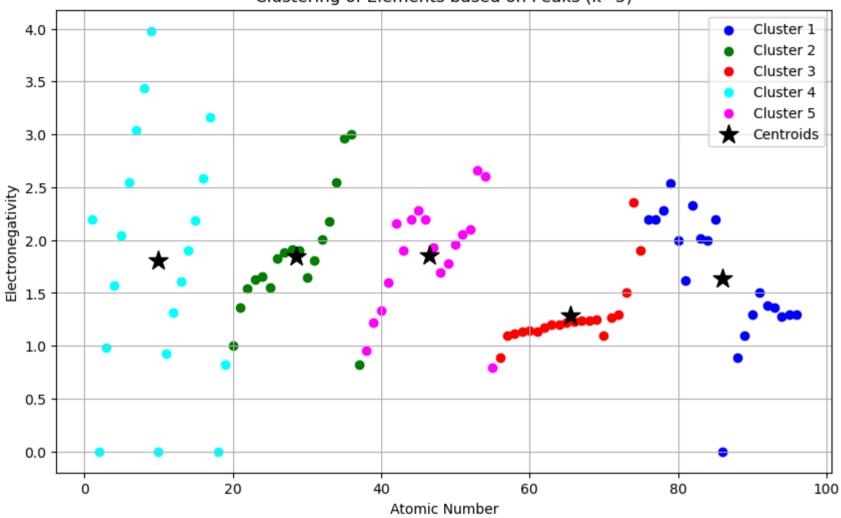


This plot doesn't have pointy elbow but we can see k=5 and k=6 are relatively appropriate values

Let's plot for k=5 and k=7

```
In [33]: # Perform KMeans clustering
          k = 5
          kmeans = KMeans(n clusters=k, random state=42)
          data['Cluster'] = kmeans.fit predict(features) + 1
          # PLot clusters
          plt.figure(figsize=(10, 6))
          # Plot each cluster
          for cluster num, color in zip(range(1, k+1), colors):
             cluster data = data[data['Cluster'] == cluster num]
             plt.scatter(cluster data['Atomic Number'], cluster data['Electronegativity'], color=color, label=f'Cluster {cluster num}')
          # Plot centroids
          centroids = kmeans.cluster centers
          plt.scatter(centroids[:, 1], centroids[:, 0], marker='*', s=200, c='black', label='Centroids')
          # Labels and Legend
          plt.xlabel('Atomic Number')
          plt.ylabel('Electronegativity')
          plt.title('Clustering of Elements based on Peaks (k=5)')
          plt.legend()
          plt.grid(True)
          plt.show()
```

Clustering of Elements based on Peaks (k=5)



```
In [34]: # Perform KMeans clustering
k = 7
kmeans = KMeans(n_clusters=k, random_state=42)
data['Cluster'] = kmeans.fit_predict(features) + 1

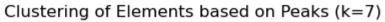
# Plot clusters
plt.figure(figsize=(10, 6))

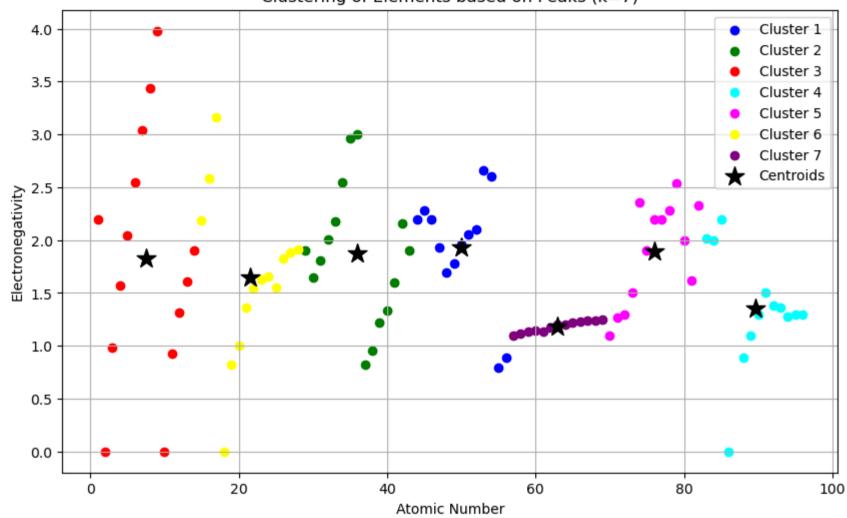
# Plot each cluster
```

```
for cluster_num, color in zip(range(1, k+1), colors):
        cluster_data = data[data['Cluster'] == cluster_num]
        plt.scatter(cluster_data['Atomic_Number'], cluster_data['Electronegativity'], color=color, label=f'Cluster {cluster_num}')

# Plot centroids
centroids = kmeans.cluster_centers_
plt.scatter(centroids[:, 1], centroids[:, 0], marker='*', s=200, c='black', label='Centroids')

# Labels and Legend
plt.xlabel('Atomic Number')
plt.ylabel('Electronegativity')
plt.title('Clustering of Elements based on Peaks (k=7)')
plt.legend()
plt.grid(True)
plt.show()
```





In []: