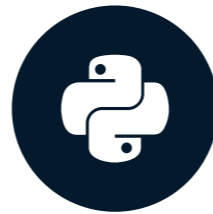


Unsupervised learning: basics

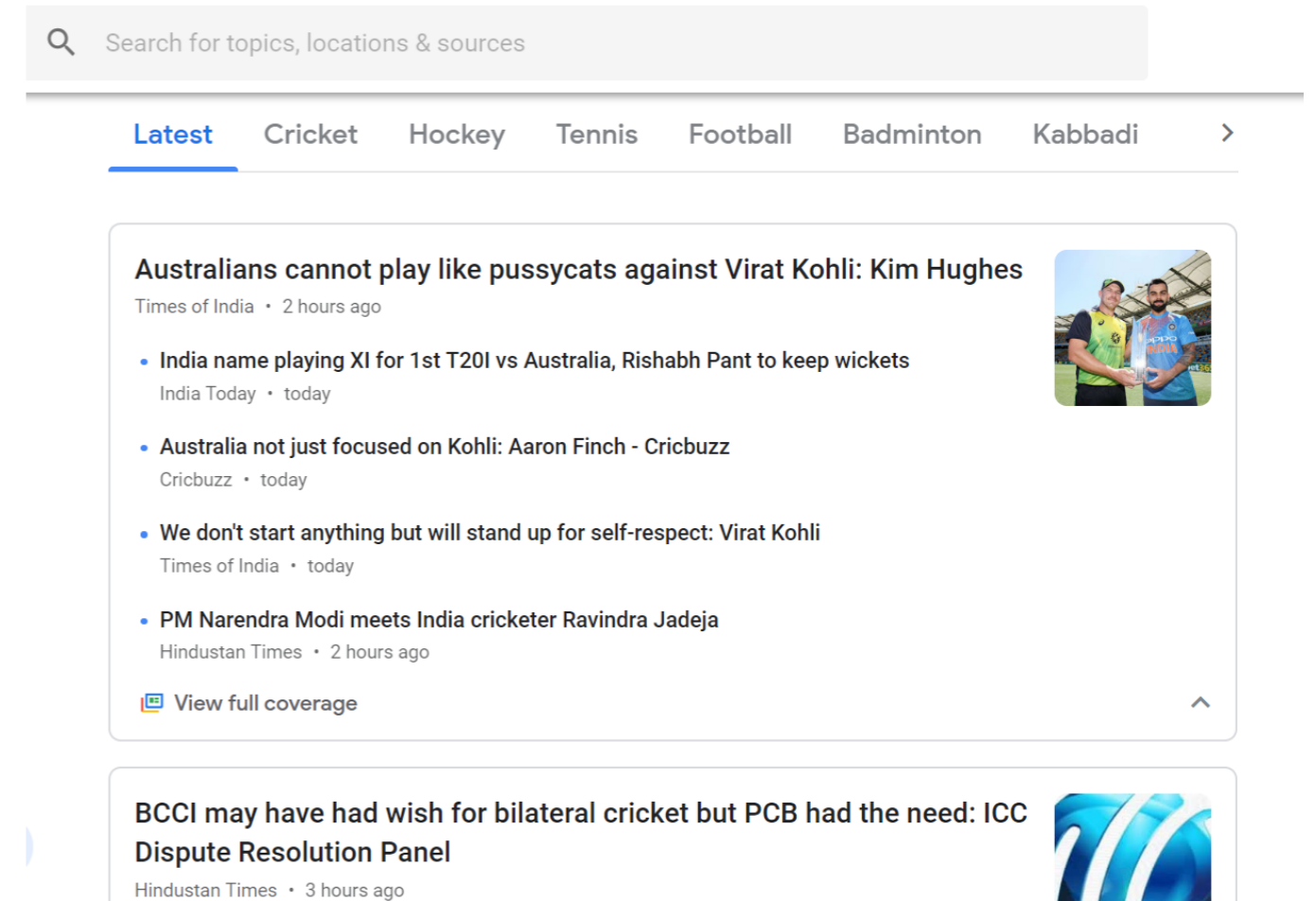
CLUSTER ANALYSIS IN PYTHON



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Everyday example: Google news

- How does Google News classify articles?
- Unsupervised Learning Algorithm:
Clustering
- Match frequent terms in articles to find
similarity



The screenshot shows a Google News search interface. At the top is a search bar with the placeholder text "Search for topics, locations & sources". Below the search bar is a navigation menu with tabs for "Latest", "Cricket", "Hockey", "Tennis", "Football", "Badminton", and "Kabbadi". The "Latest" tab is selected. The main content area displays two news articles. The first article is titled "Australians cannot play like pussycats against Virat Kohli: Kim Hughes" and is from "Times of India" published "2 hours ago". It includes a list of related headlines: "India name playing XI for 1st T20I vs Australia, Rishabh Pant to keep wickets" (India Today, today), "Australia not just focused on Kohli: Aaron Finch - Cricbuzz" (Cricbuzz, today), "We don't start anything but will stand up for self-respect: Virat Kohli" (Times of India, today), and "PM Narendra Modi meets India cricketer Ravindra Jadeja" (Hindustan Times, 2 hours ago). A "View full coverage" link is at the bottom of this article. The second article is titled "BCCI may have had wish for bilateral cricket but PCB had the need: ICC Dispute Resolution Panel" and is from "Hindustan Times" published "3 hours ago". It features a blue and white abstract graphic.

Labeled and unlabeled data

Data with no labels

- Point 1: (1, 2)
- Point 2: (2, 2)
- Point 3: (3, 1)

Data with labels

- Point 1: (1, 2), Label: Danger Zone
- Point 2: (2, 2), Label: Normal Zone
- Point 3: (3, 1), Label: Normal Zone

What is unsupervised learning?

- A group of machine learning algorithms that find patterns in data
- Data for algorithms has not been labeled, classified or characterized
- The objective of the algorithm is to interpret any structure in the data
- Common unsupervised learning algorithms: clustering, neural networks, anomaly detection

What is clustering?

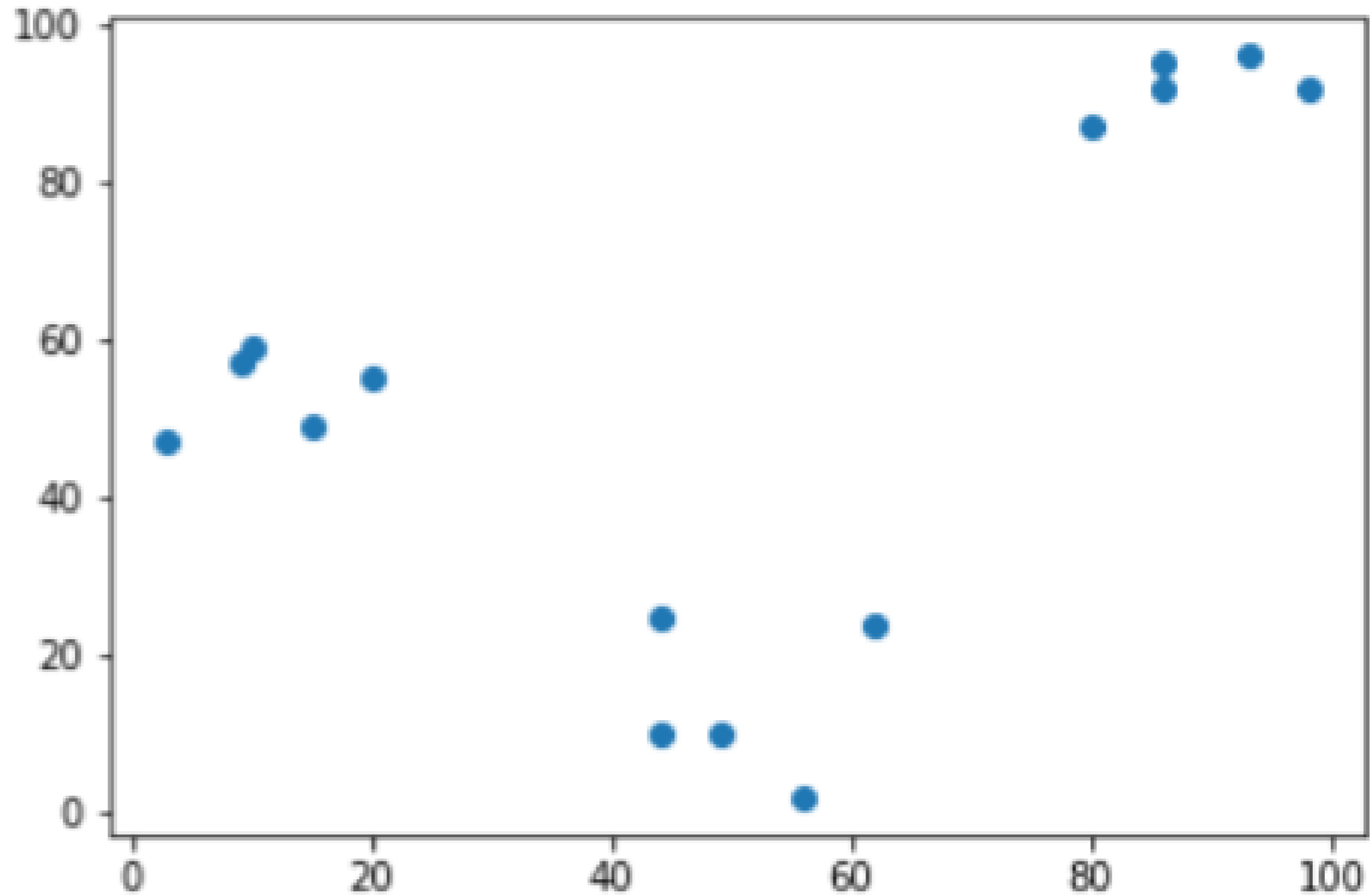
- The process of grouping items with similar characteristics
- Items in groups similar to each other than in other groups
- Example: distance between points on a 2D plane

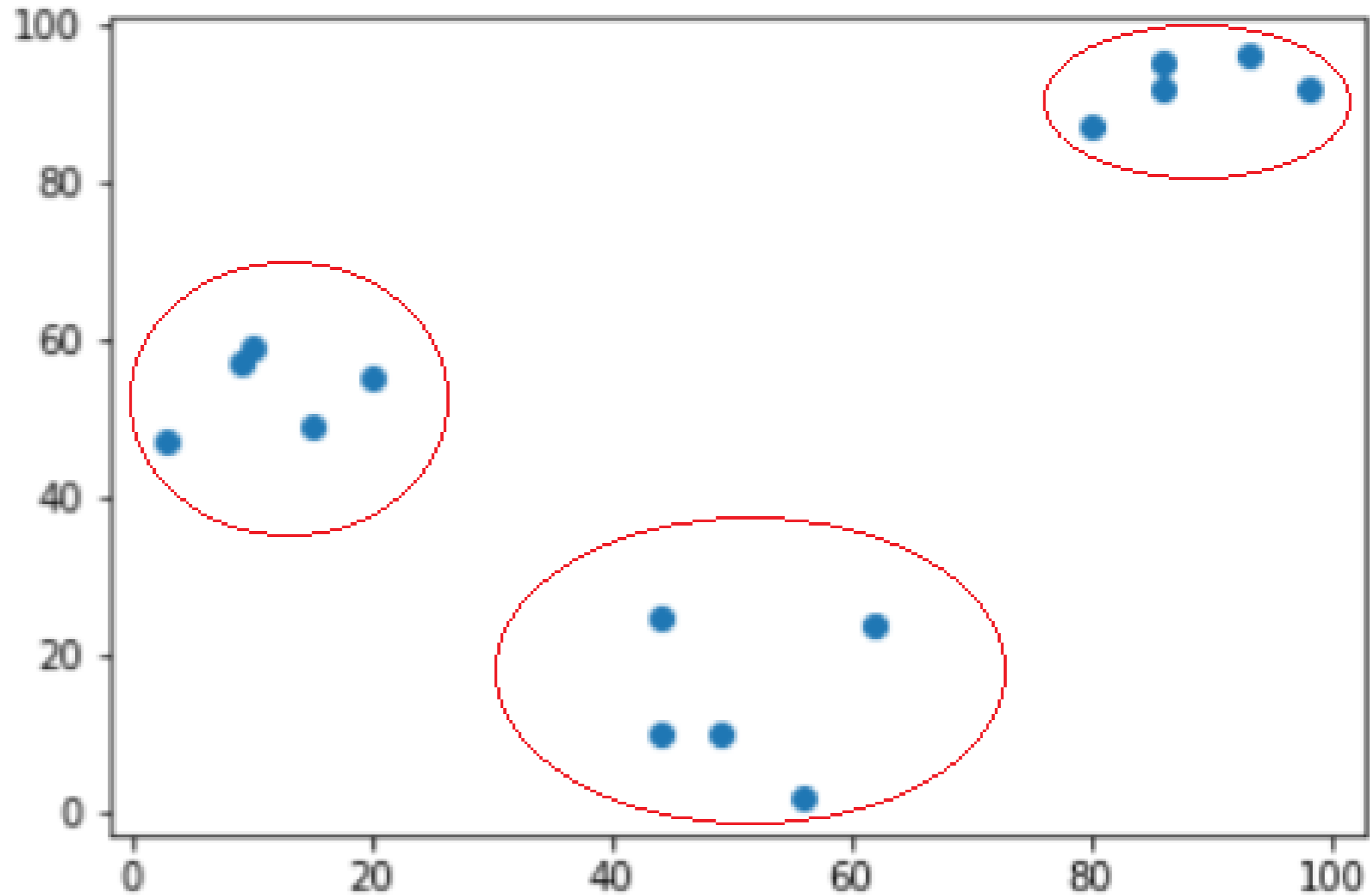
Plotting data for clustering - Pokemon sightings

```
from matplotlib import pyplot as plt
```

```
x_coordinates = [80, 93, 86, 98, 86, 9, 15, 3, 10, 20, 44, 56, 49, 62, 44]  
y_coordinates = [87, 96, 95, 92, 92, 57, 49, 47, 59, 55, 25, 2, 10, 24, 10]
```

```
plt.scatter(x_coordinates, y_coordinates)  
plt.show()
```





Up next - some practice

CLUSTER ANALYSIS IN PYTHON

Basics of cluster analysis

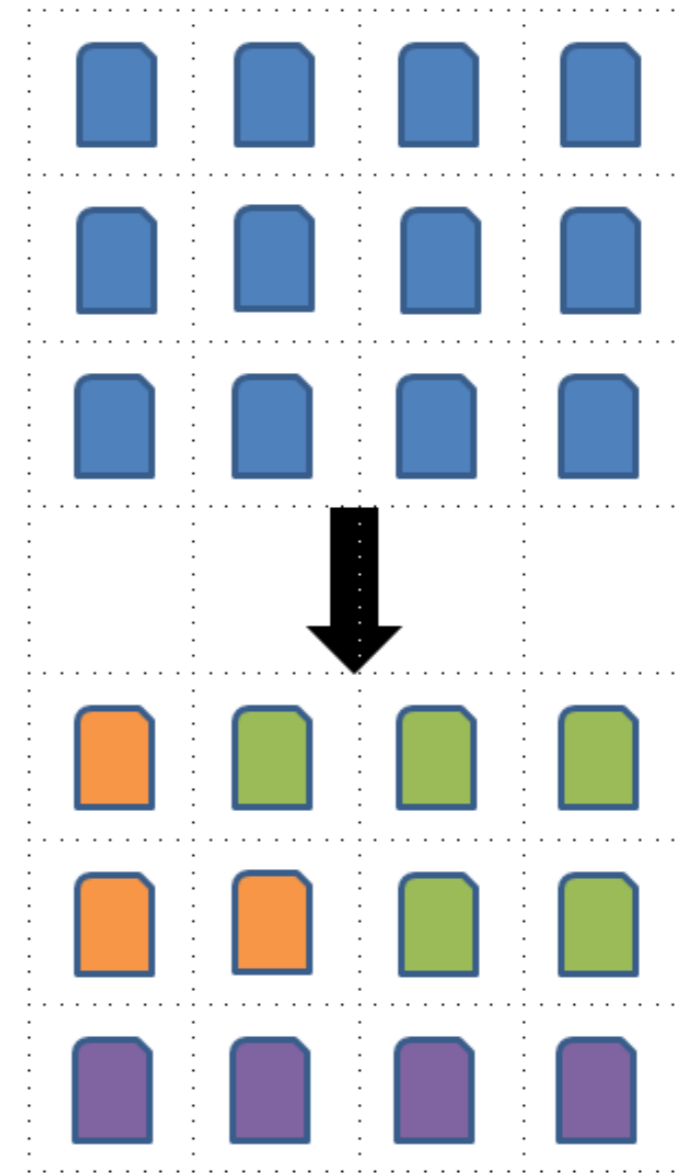
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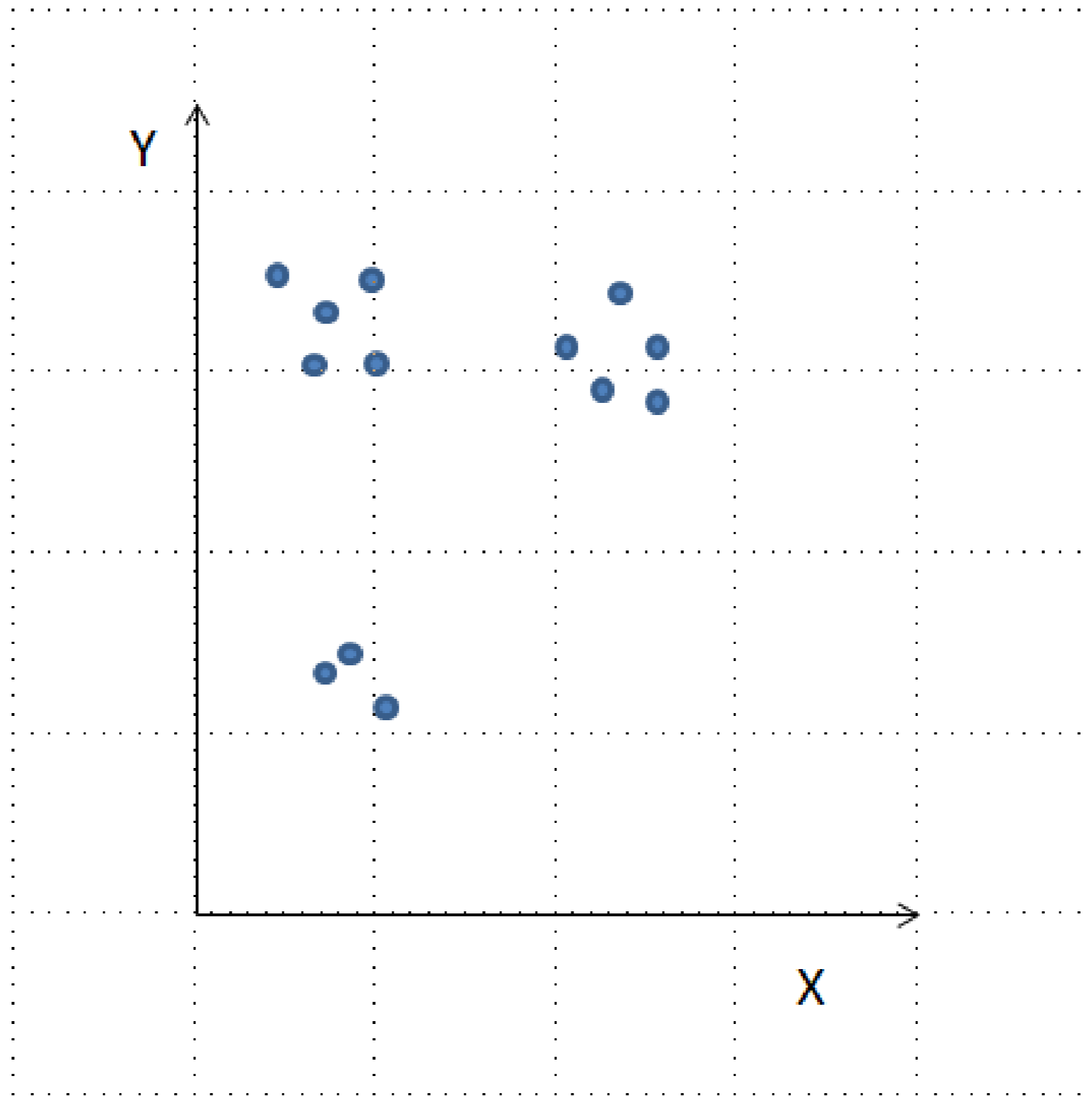
What is a cluster?

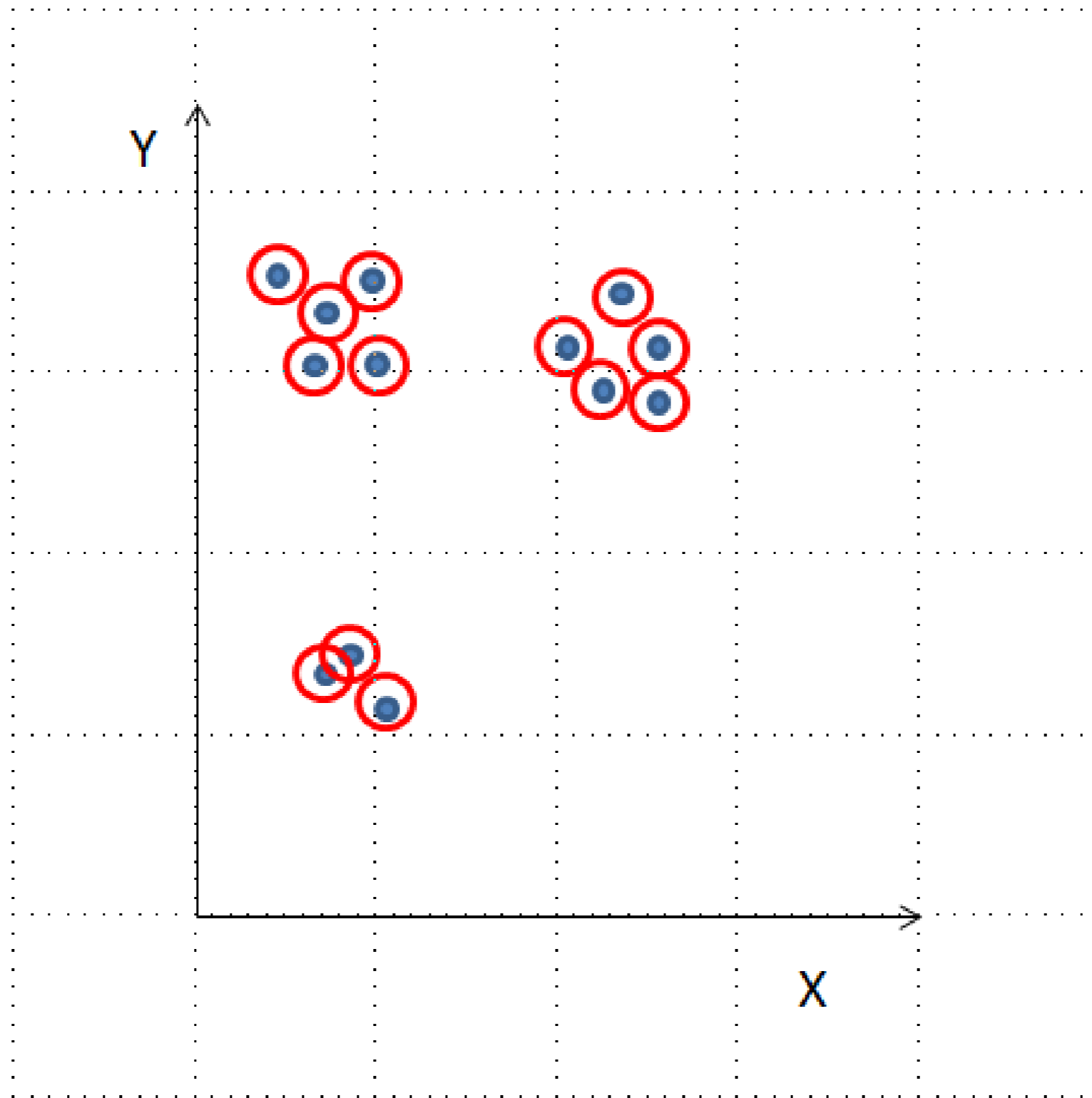
- A group of items with similar characteristics
- Google News: articles where similar words and word associations appear together
- Customer Segments

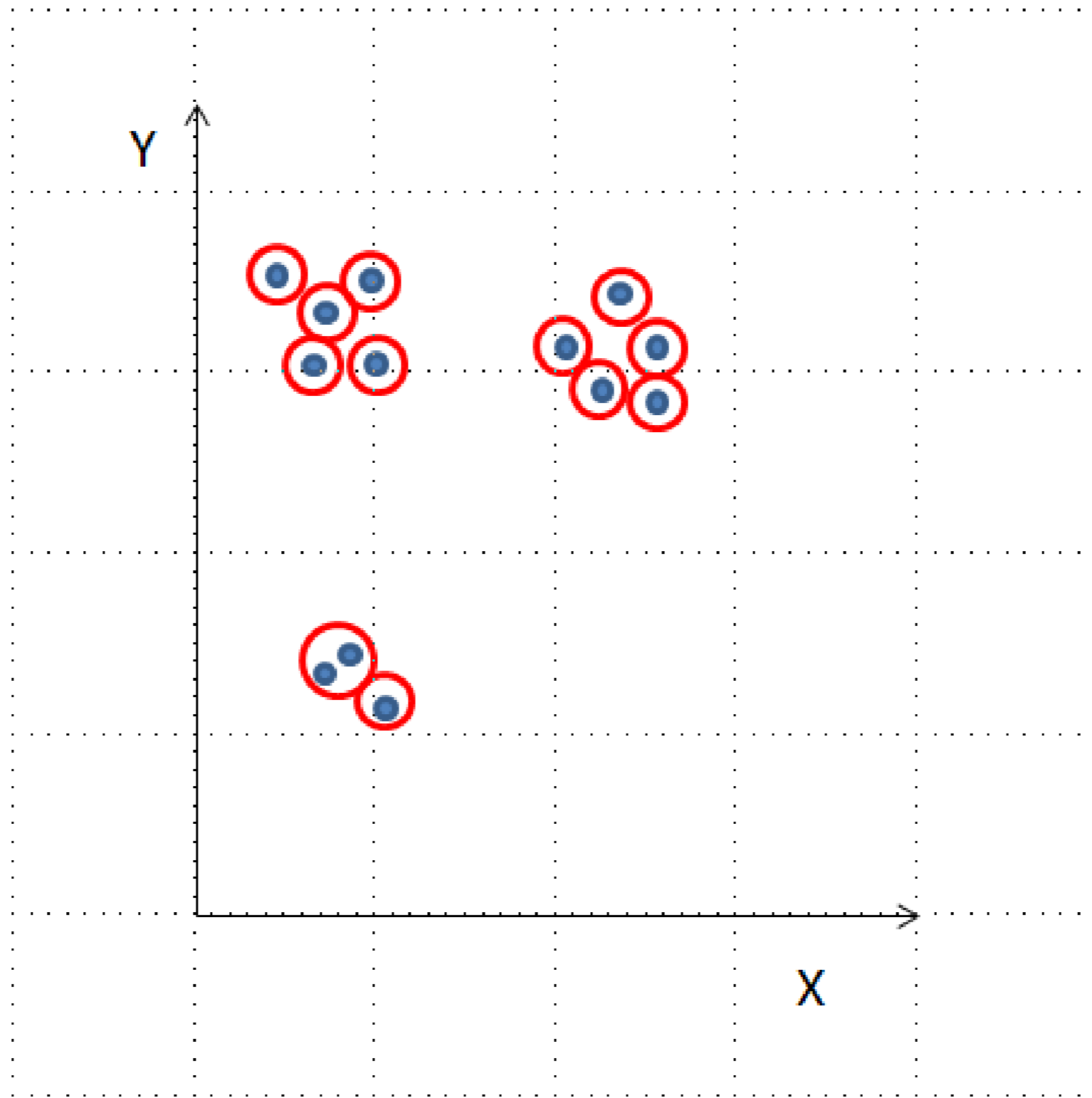


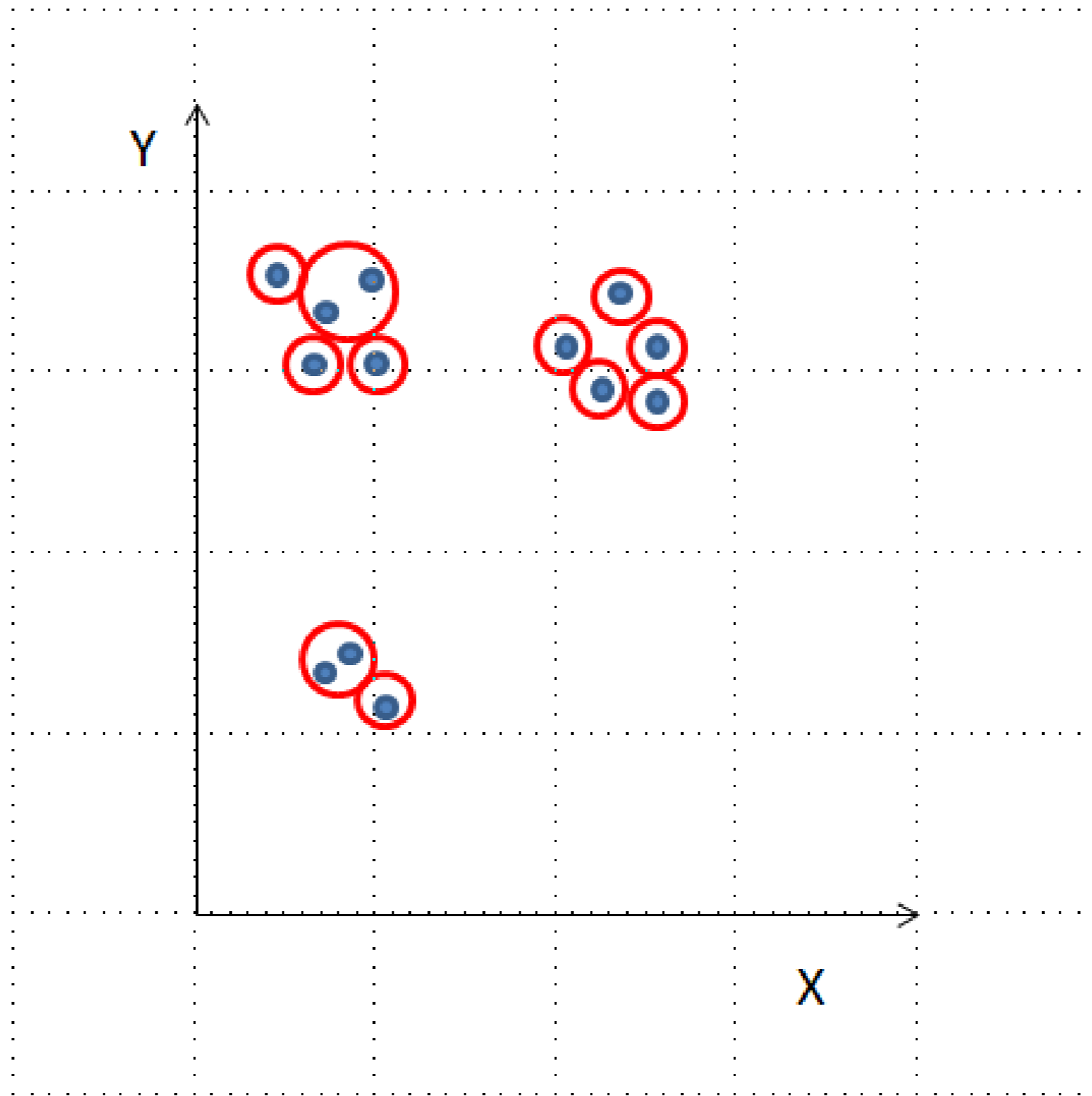
Clustering algorithms

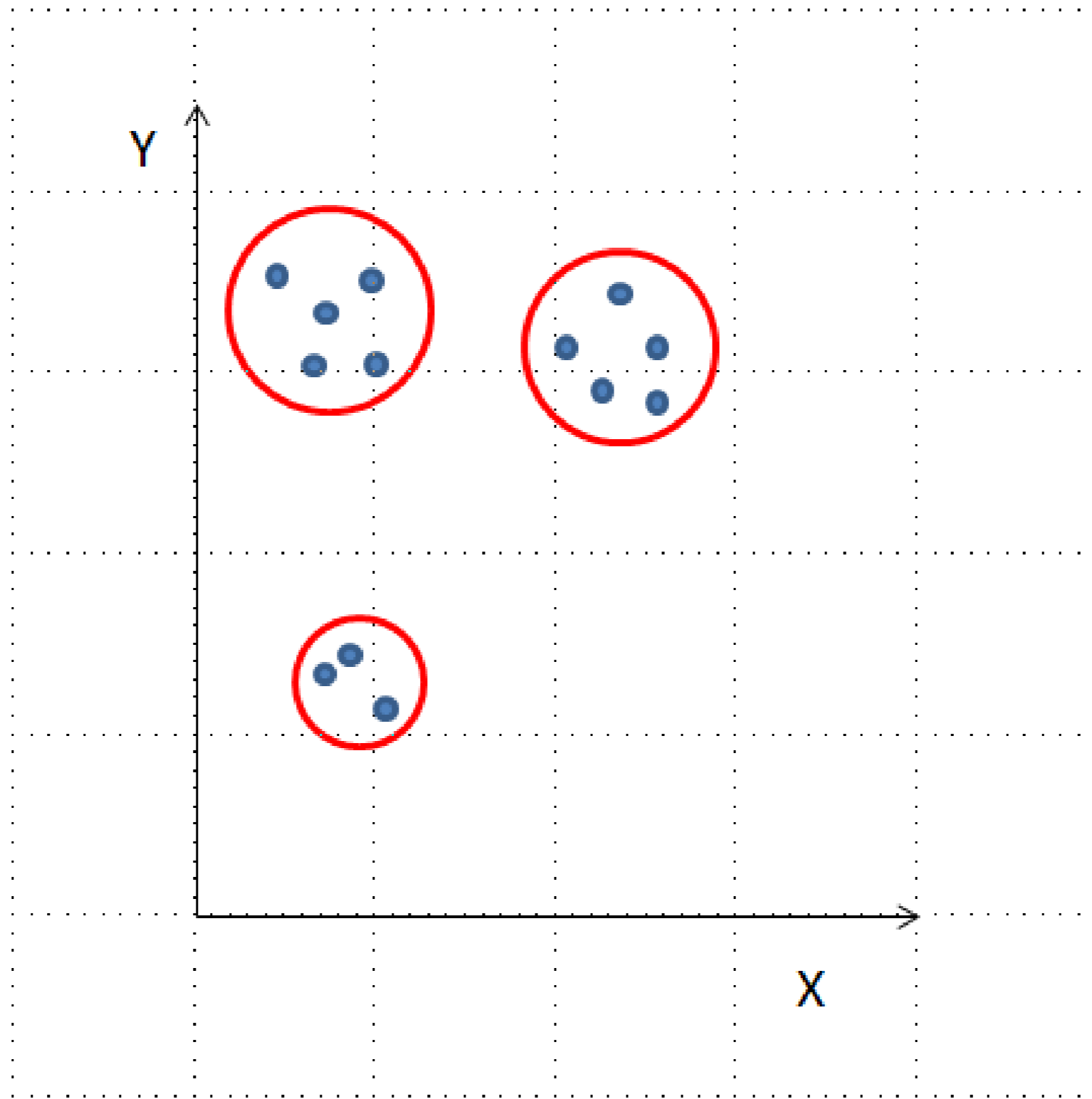
- Hierarchical clustering
- K means clustering
- Other clustering algorithms: DBSCAN, Gaussian Methods











Hierarchical clustering in SciPy

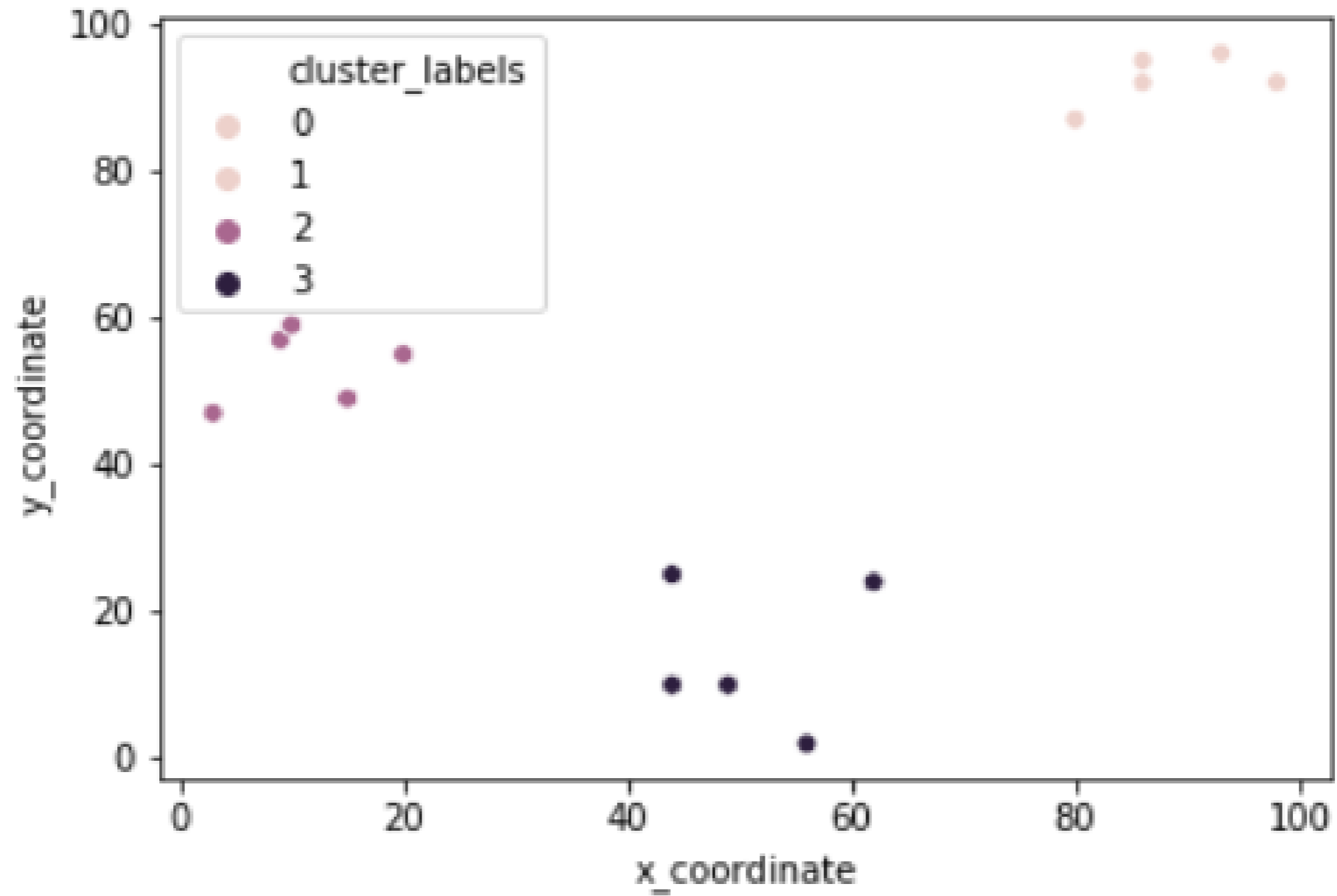
```
from scipy.cluster.hierarchy import linkage, fcluster
from matplotlib import pyplot as plt
import seaborn as sns, pandas as pd
```

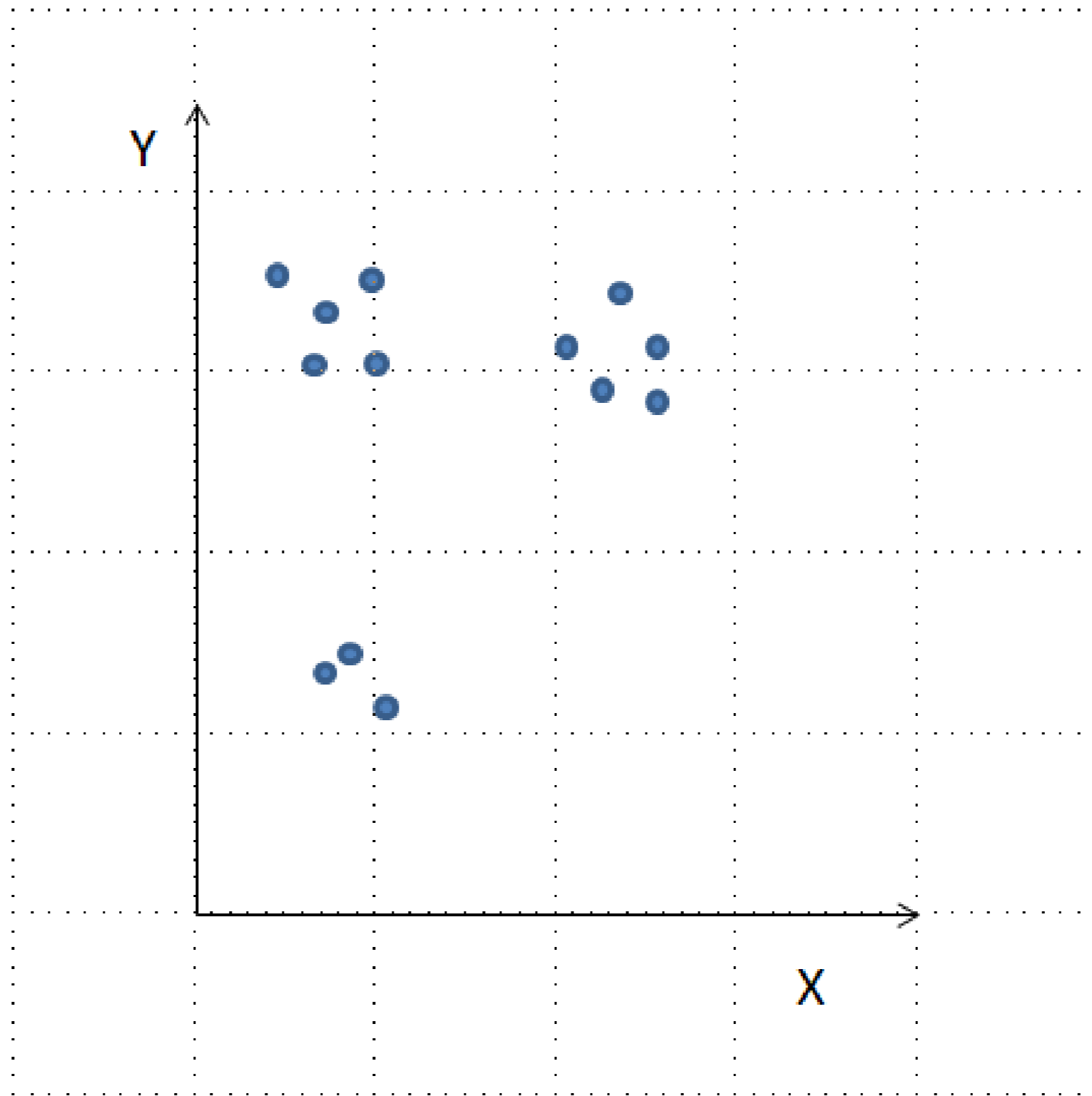
```
x_coordinates = [80.1, 93.1, 86.6, 98.5, 86.4, 9.5, 15.2, 3.4,
                 10.4, 20.3, 44.2, 56.8, 49.2, 62.5, 44.0]
y_coordinates = [87.2, 96.1, 95.6, 92.4, 92.4, 57.7, 49.4,
                 47.3, 59.1, 55.5, 25.6, 2.1, 10.9, 24.1, 10.3]
```

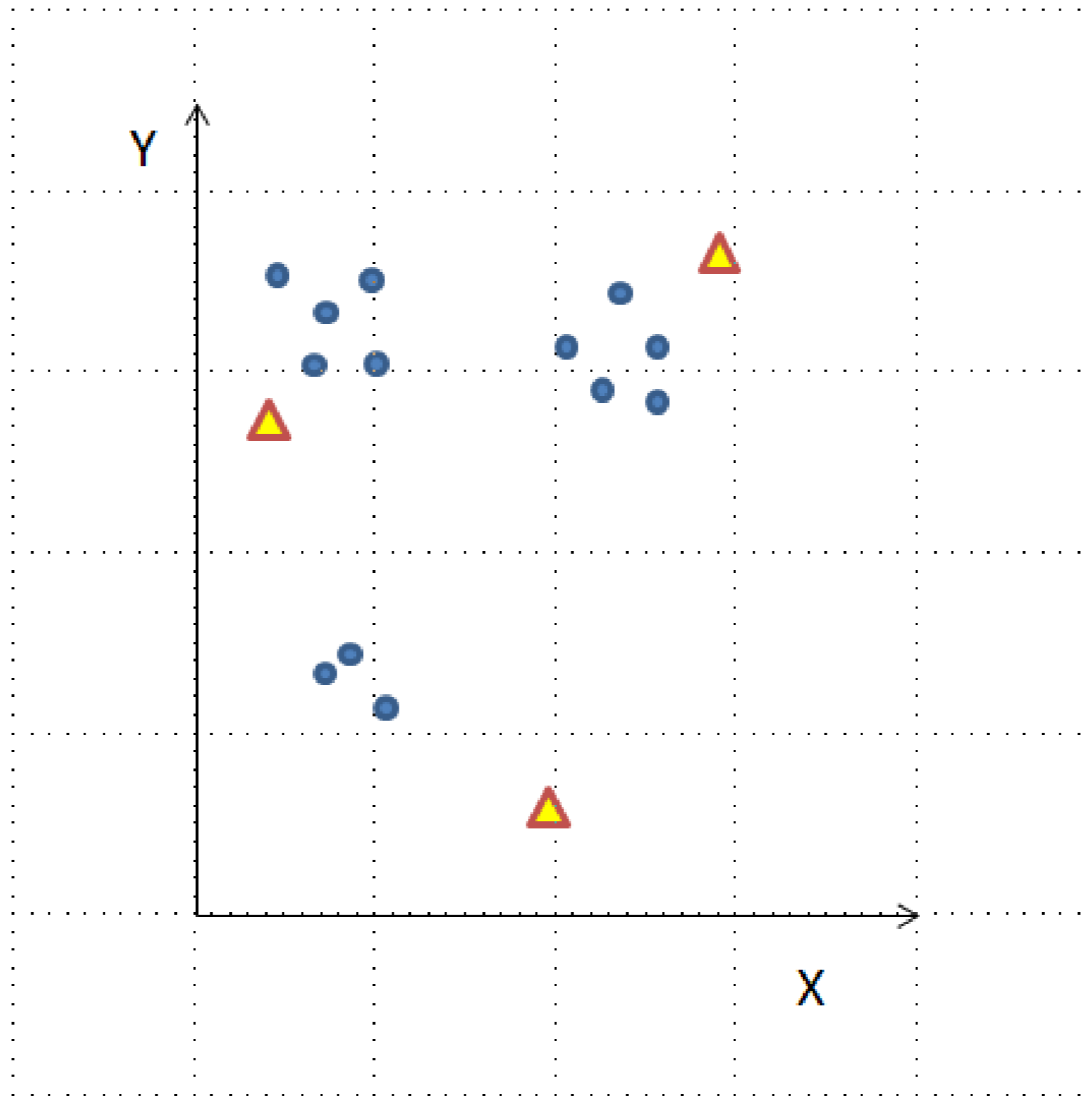
```
df = pd.DataFrame({'x_coordinate': x_coordinates,
                  'y_coordinate': y_coordinates})
```

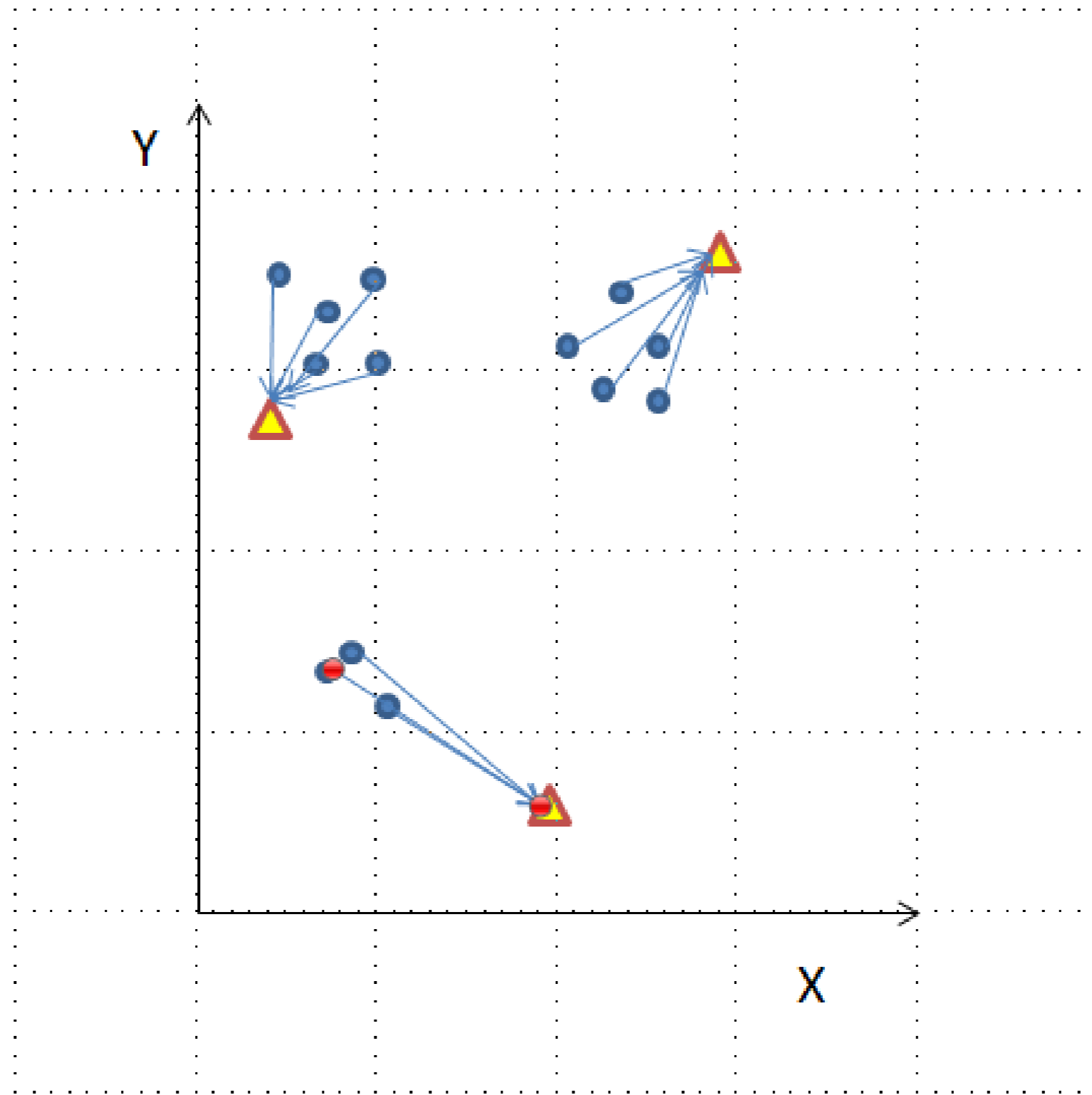
```
Z = linkage(df, 'ward')
df['cluster_labels'] = fcluster(Z, 3, criterion='maxclust')
```

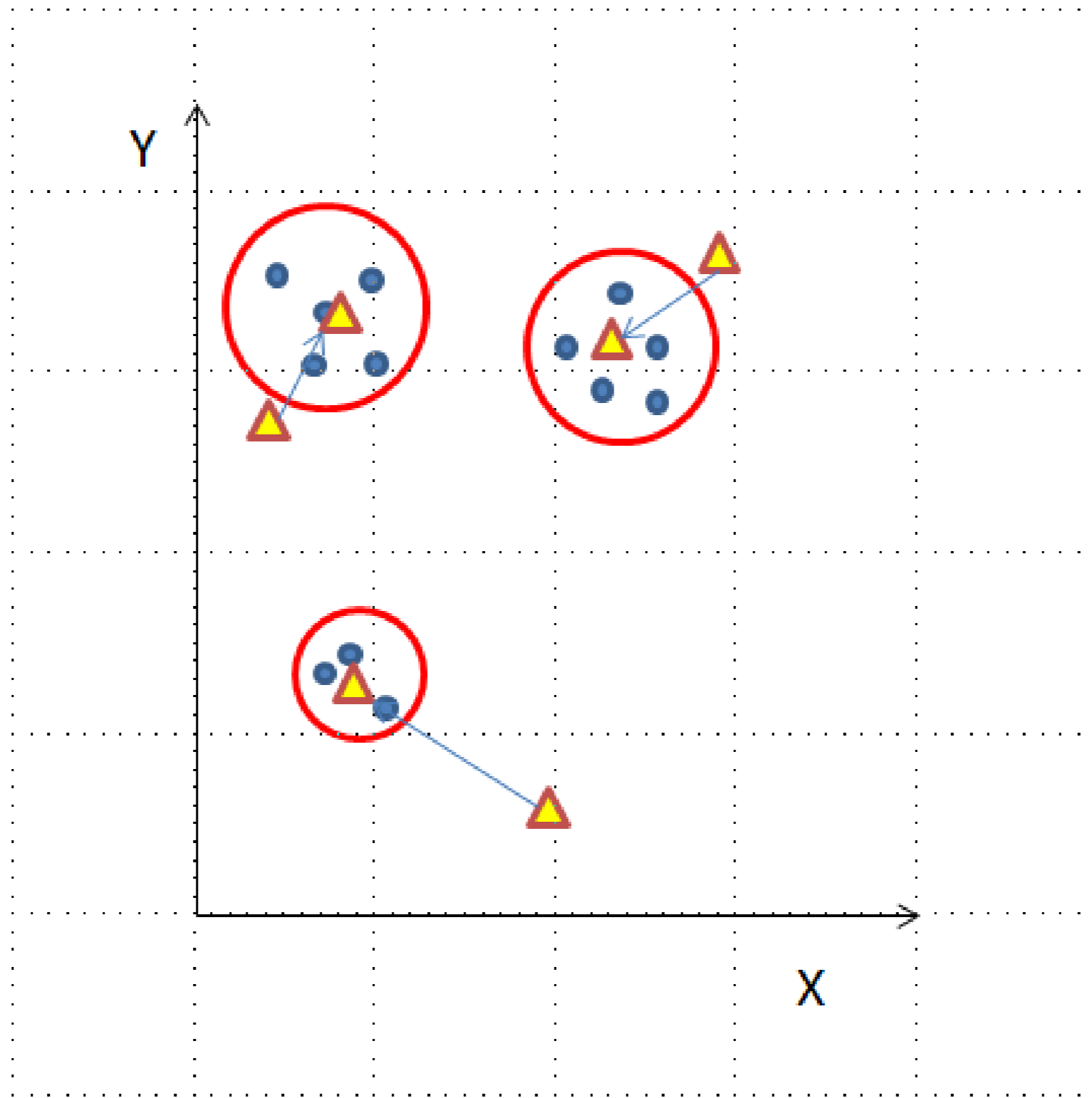
```
sns.scatterplot(x='x_coordinate', y='y_coordinate',
                hue='cluster_labels', data = df)
plt.show()
```











K-means clustering in SciPy

```
from scipy.cluster.vq import kmeans, vq
from matplotlib import pyplot as plt
import seaborn as sns, pandas as pd
```

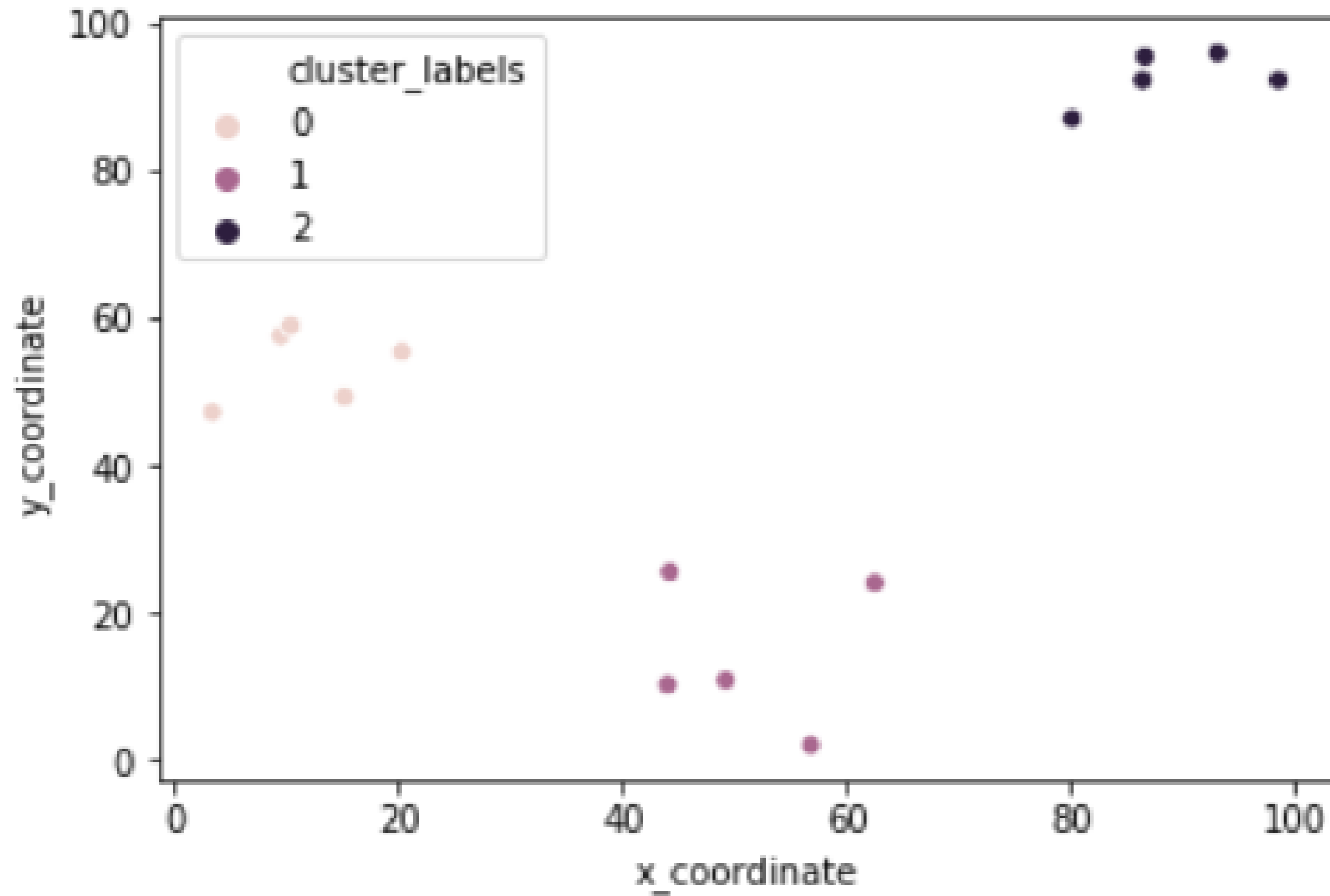
```
import random
random.seed((1000,2000))
```

```
x_coordinates = [80.1, 93.1, 86.6, 98.5, 86.4, 9.5, 15.2, 3.4,
                 10.4, 20.3, 44.2, 56.8, 49.2, 62.5, 44.0]
y_coordinates = [87.2, 96.1, 95.6, 92.4, 92.4, 57.7, 49.4,
                 47.3, 59.1, 55.5, 25.6, 2.1, 10.9, 24.1, 10.3]
```

```
df = pd.DataFrame({'x_coordinate': x_coordinates, 'y_coordinate': y_coordinates})
```

```
centroids,_ = kmeans(df, 3)
df['cluster_labels'], _ = vq(df, centroids)
```

```
sns.scatterplot(x='x_coordinate', y='y_coordinate',
                hue='cluster_labels', data = df)
plt.show()
```

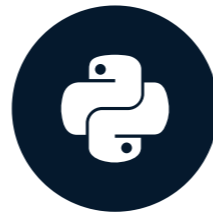



Next up: hands-on exercises

CLUSTER ANALYSIS IN PYTHON

Data preparation for cluster analysis

CLUSTER ANALYSIS IN PYTHON



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Why do we need to prepare data for clustering?

- Variables have incomparable units (product dimensions in cm, price in \$)
- Variables with same units have vastly different scales and variances (expenditures on cereals, travel)
- Data in raw form may lead to bias in clustering
- Clusters may be heavily dependent on one variable
- Solution: normalization of individual variables

Normalization of data

Normalization: process of rescaling data to a standard deviation of 1

$$x_{\text{new}} = x / \text{std_dev}(x)$$

```
from scipy.cluster.vq import whiten
```

```
data = [5, 1, 3, 3, 2, 3, 3, 8, 1, 2, 2, 3, 5]
```

```
scaled_data = whiten(data)  
print(scaled_data)
```

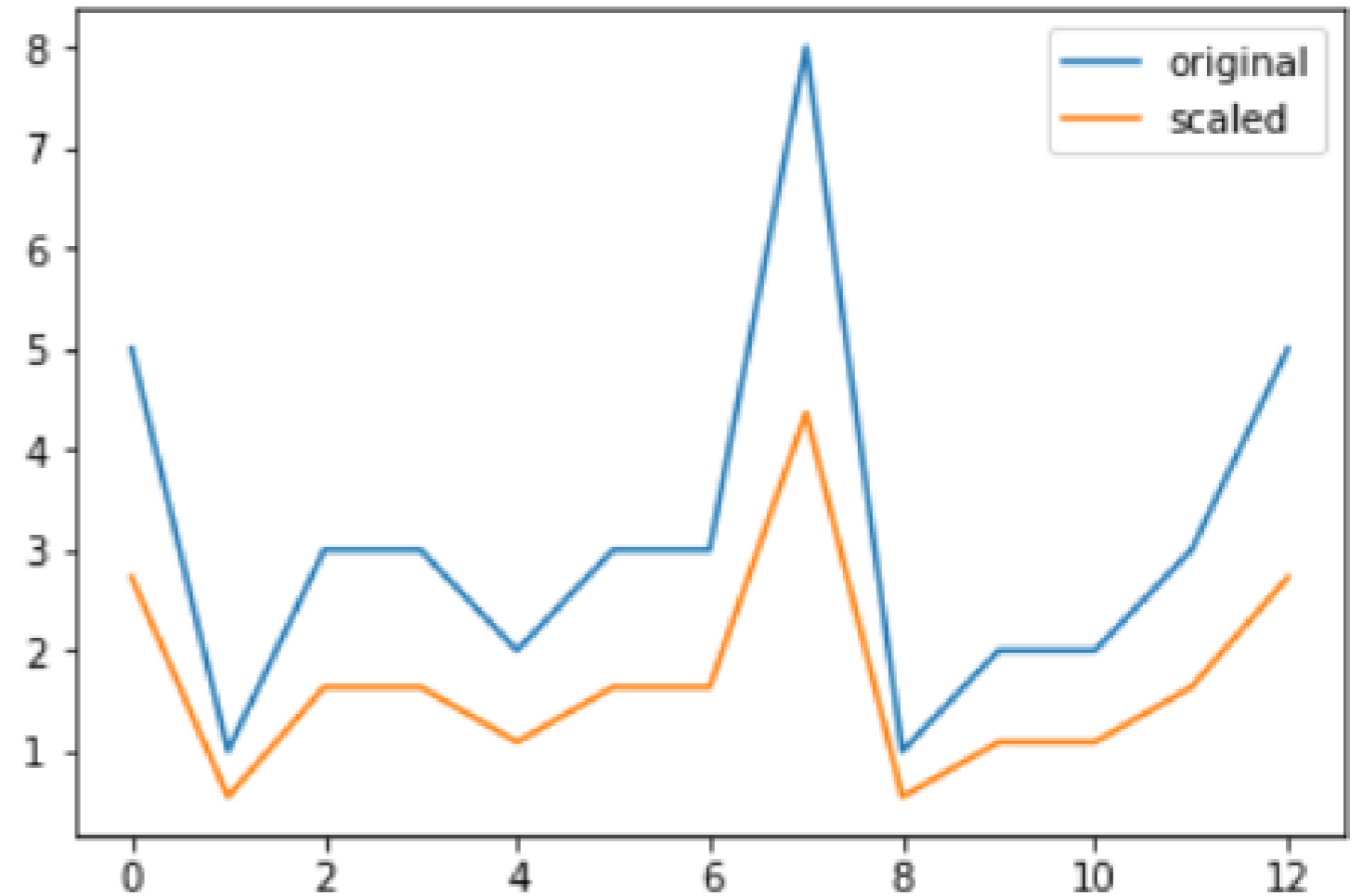
```
[2.73, 0.55, 1.64, 1.64, 1.09, 1.64, 1.64, 4.36, 0.55, 1.09, 1.09, 1.64, 2.73]
```

Illustration: normalization of data

```
# Import plotting library
from matplotlib import pyplot as plt

# Initialize original, scaled data
plt.plot(data,
         label="original")
plt.plot(scaled_data,
         label="scaled")

# Show legend and display plot
plt.legend()
plt.show()
```



Next up: some DIY exercises

CLUSTER ANALYSIS IN PYTHON