\*Affinity Propagation (AP) is a clustering algorithm that doesn't require the user to specify the number of clusters beforehand. Instead, it works by sending messages between data points to discover exemplars, which are representative points in the dataset.

### Similarity Matrix:

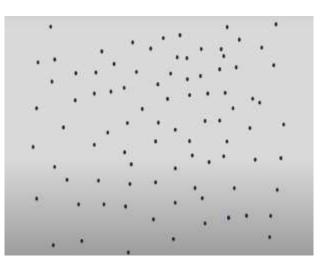
AP starts by computing a similarity matrix where each element represents the similarity between pairs of data points. The similarity metric could be any measure of similarity or dissimilarity, such as negative Euclidean distance or negative squared distance.

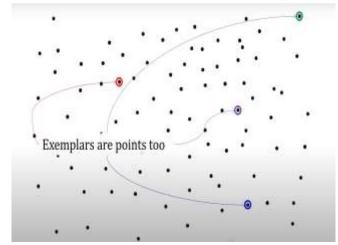
### **Responsibility Matrix**:

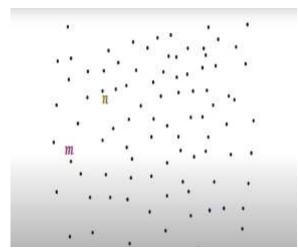
AP iteratively updates a matrix of responsibilities, denoted by r(i,k), which represents how well-suited one data point i is to be an exemplar for another data point K.

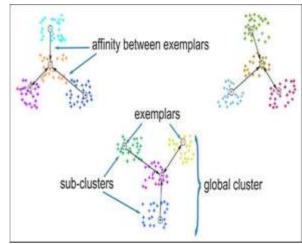
- Availability Matrix:
- Similarly, AP updates a matrix of availabilities, denoted by a(i,k), which represents the accumulated evidence that data point k should choose data point i as its exemplar.
- Message Passing:
- AP iterates between updating the responsibility and availability matrices based on current values. It updates these matrices by passing messages between data points according to specific rules.

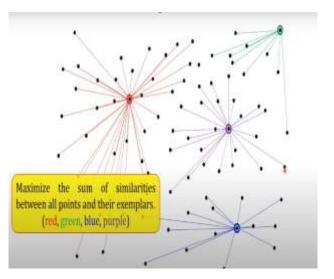
### **EVALUATION OF AFFINITY PROPAGATION**

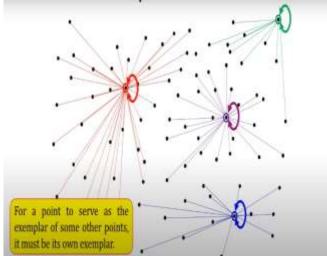


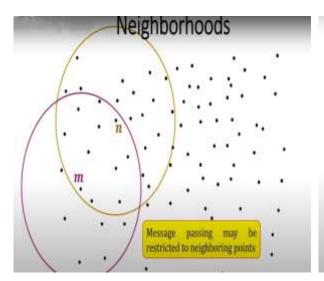


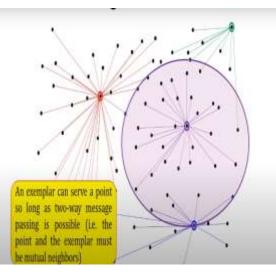












- Exemplar Selection:
- After a certain number of iterations or when the changes in the matrices become small, the algorithm identifies exemplars and assigns data points to these exemplars based on the values in the responsibility and availability matrices.
- Cluster Formation:
- Finally, clusters are formed based on the exemplars identified.
   Data points are assigned to clusters based on which exemplar they are most strongly associated with.

## **Affinity Propagation**

# (AP)

Here's a breakdown of the Python code, focusing on the application

### of Affinity Propagation

- import pandas as pd
- **import** numpy as np
- **import** matplotlib.pyplot as **plt**
- from sklearn.cluster import Affinity Propagation
- from sklearn.datasets import make blobs
- from **sklearn** import metrics

### • Imports:

 The necessary libraries are imported: pandas for data handling, NumPy for numerical computations, matplotlib for plotting, and various modules from scikit-learn for clustering and metrics evaluation.

Dataset = pd.read\_csv("Mall\_Customers.csv")

### **Dataset Loading:**

The code loads a dataset named "Mall\_Customers.csv" using pandas.

X = dataset.iloc[:, 3:5].values

### Feature Selection:

It selects specific columns (features) from the dataset, which will be used for clustering.

- centers = [[1, 1], [-1, -1], [1, -1]]
- X, labels true = make blobs(n\_samples=300, centers=centers, cluster\_std=0.5, random\_state=0)
- Data Generation:
- The code generates sample data for clustering using the make blobs function from scikit-learn. This data is used to demonstrate clustering
- af = AffinityPropagation(preference=-50, random\_state=0).fit(X).
- Affinity Propagation:
- It applies the Affinity Propagation algorithm to the generated data.

- cluster\_centers\_indices = af.cluster\_centers\_indices\_
- labels = af.labels
- n\_clusters\_ = len(cluster\_centers\_indices)
- Cluster Retrieval The code retrieves the cluster centers, labels, and the number of clusters identified by the algorithm.
- print("Estimated number of clusters: %d" % n\_clusters\_)
- print("Homogeneity: %0.3f" % metrics.homogeneity\_score(labels\_true, labels))
- Metrics Evaluation:
- Various clustering metrics are calculated and printed to evaluate the performance of the clustering algorithm.
- # Plotting clusters:
- plt.close("all")
- plt.figure(1)
- plt.clf()
- colors = plt.cycler("color", plt.cm.viridis(np.linspace(0, 1, 4)))

```
for k, col in zip(range(n_clusters_), colors):
    class_members = labels == k
    cluster_center = X[cluster_centers_indices[k]]
    plt.scatter(X[class_members, 0], X[class_members, 1], color=col["color"], marker=".")
    plt.scatter(cluster_center[0], cluster_center[1], s=14, color=col["color"], marker="o")
    for x in X[class members]:
       plt.plot([cluster_center[0], x[0]], [cluster_center[1], x[1]], color=col["color"])
  plt.title("Estimated number of clusters: %d" % n clusters )
  plt.show()
For Loop:
```

for k, col in zip(range(n\_clusters\_), colors): This loop iterates over the range of the number of clusters
 (n\_clusters\_) along with the colors generated by plt.cycler.

#### Cluster Members:

• class\_members = labels == k: This line identifies the data points that belong to the current cluster k. It creates a boolean array where True indicates that the corresponding data point belongs to the current cluster.

#### Cluster Center:

• cluster center = X[cluster centers indices[k]]: It retrieves the coordinates of the cluster center for the current cluster k from the data X.

#### Scatter Plot:

- plt.scatter(X[class\_members, 0], X[class\_members, 1], color=col["color"], marker="."): This line plots the data points belonging to the current cluster k using a scatter plot. It assigns the color col["color"] to the points and uses a dot marker.
- plt.scatter(cluster\_center[0], cluster\_center[1], s=14, color=col["color"], marker="o"): This line plots the cluster center using a larger circular marker.

### Connecting Lines:

• for x in X[class\_members]: plt.plot([cluster\_center[0], x[0]], [cluster\_center[1], x[1]], color=col["color"]): This loop plots lines connecting each data point in the current cluster k to the cluster center. It creates lines from the cluster center to each data point.

#### • Title and Display:

- plt.title("Estimated number of clusters: %d" % n\_clusters\_): It sets the title of the plot to indicate the estimated number of clusters.
- plt.show(): This command displays the plot with all the clusters and their respective center points, along with the connecting lines.



