

# OPTICS (Ordering Points To Identify the Clustering Structure)

- **OPTICS (Ordering Points To Identify the Clustering Structure)** is a density-based hierarchical clustering algorithm designed to efficiently cluster large datasets. It extends the concepts of DBSCAN by producing a reachability plot that provides a hierarchical view of the clustering structure.
- **Importing Libraries and Generating Data:**
  - `from sklearn.cluster import OPTICS`
  - `from sklearn.datasets import make_moons`
  - `Moons = make_moons(n_samples=200, noise=0.05, random_state=42)`
- This part of the code imports the OPTICS clustering algorithm from scikit-learn and generates synthetic moon-shaped data using the `make_moons` function.

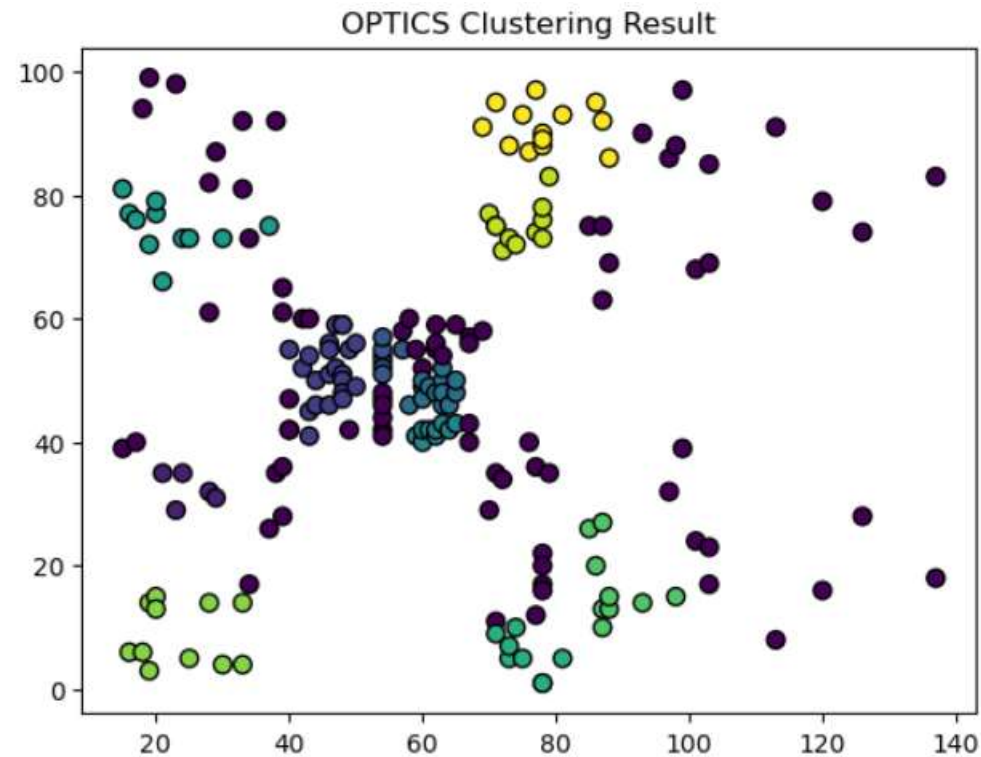
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- **Applying OPTICS:**
- **`optics = OPTICS(eps=0.3, min_samples=5, cluster_method='xi', xi=0.05)`**
- **`optics.fit(X)`**
- Here, an instance of the OPTICS clustering algorithm is created with parameters such as `eps` (maximum distance between two samples) and `min_samples` (minimum number of samples in a neighborhood).
- `cluster_method` specifies the method used to determine clusters, and `xi` is a parameter that influences the sensitivity to the cluster structure.

**`import matplotlib.pyplot as plt`**

- **`plt.scatter(X[:, 0], X[:, 1], c=optics.labels_, cmap='viridis', marker='o', s=50, edgecolor='k')`**
- **`plt.title('OPTICS Clustering Result')`**
- **`plt.show()`**
- This part of the code visualizes the clustering result by creating a scatter plot of the data points. Each point is colored according to its assigned cluster label.

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- **OPTICS Clustering:**
- OPTICS (Ordering Points To Identify the Clustering Structure) is a hierarchical density-based clustering algorithm designed to efficiently cluster large datasets. It extends DBSCAN by producing a reachability plot that provides a hierarchical view of the clustering structure. OPTICS can handle datasets with varying densities and shapes of clusters.
- **Key Concepts of OPTICS:**
- **Reachability Distance:**
- In OPTICS, each point is assigned a reachability distance, which is a measure of how close it is to its nearest neighbor. Reachability distance helps in identifying clusters of varying densities.
- **Core Distance:**
- The core distance of a point is the distance to its  $k$ -th nearest neighbor, where  $k$  is the minimum number of points specified by the `min_samples` parameter.

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## 3. Reachability Plot:

OPTICS generates a reachability plot that orders points based on their reachability distances. This plot provides insights into the hierarchical structure of the clusters.

## 4. Clustering Structure:

By analyzing the reachability plot, OPTICS identifies clusters and outliers in the dataset. The flat regions in the reachability plot represent clusters, while steep inclines indicate cluster boundaries.

```
from sklearn.cluster import OPTICS
from sklearn.datasets import make_moons
Moons = make_moons(n_samples=200, noise=0.05, random_state=42)
optics = OPTICS(eps=0.3, min_samples=5, cluster_method='xi', xi=0.05)
optics.fit(X)
import matplotlib.pyplot as plt
plt.scatter(X[:, 0], X[:, 1], c=optics.labels_, cmap='viridis', marker='o', s=50,
            edgecolor='k')
plt.title('OPTICS Clustering Result')
plt.show()
```

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- **Importing Libraries and Generating Data:**

- The code imports necessary libraries (OPTICS from scikit-learn and make\_moons from scikit-learn datasets) and generates synthetic moon-shaped data with 200 samples, a noise level of 0.05, and a random state of 42.

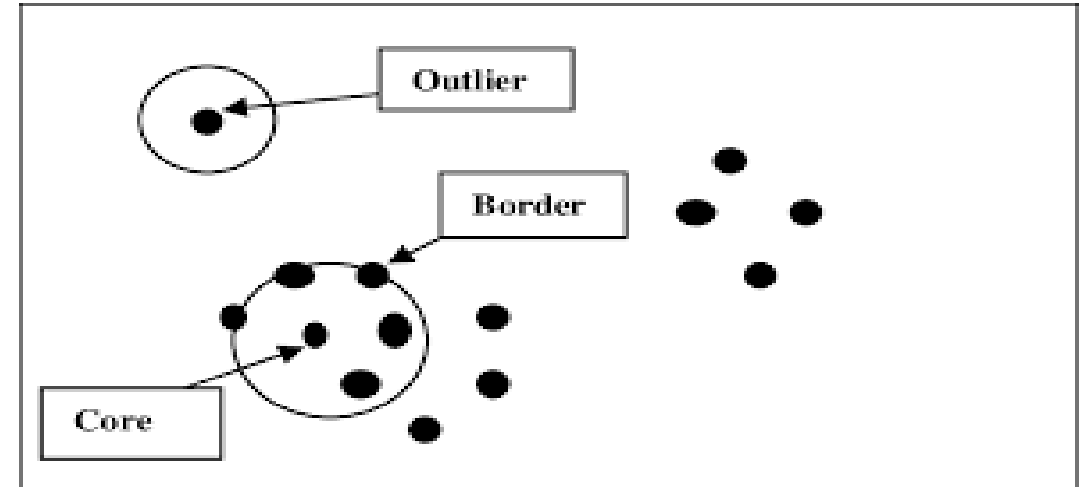
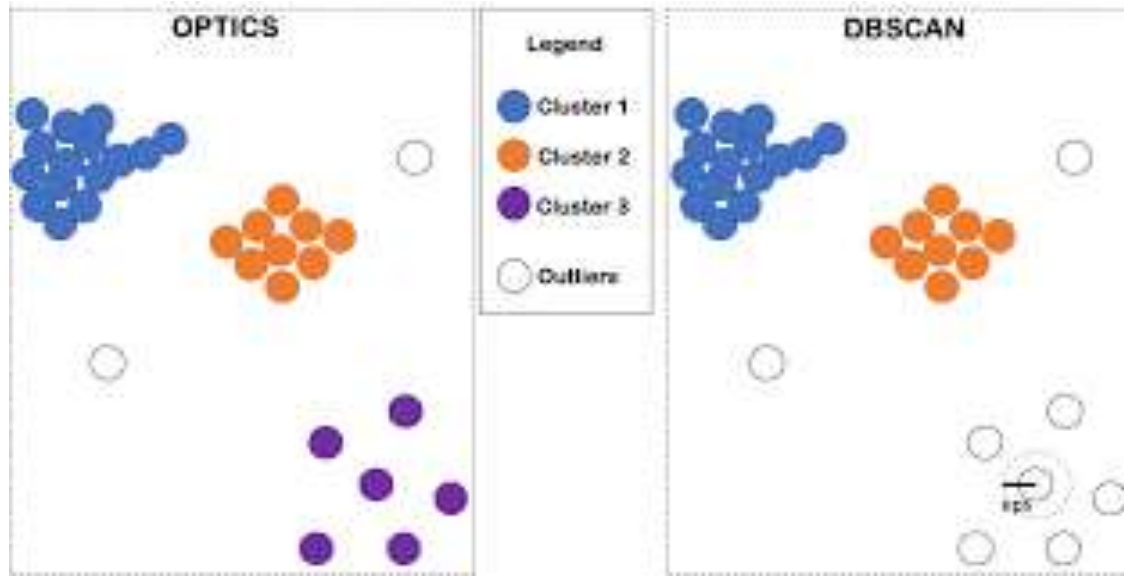
- **Applying OPTICS:**

- An instance of the OPTICS clustering algorithm is created with parameters such as eps (maximum distance between two samples), min\_samples (minimum number of samples in a neighborhood), cluster\_method (method used to determine clusters), and xi (a parameter that influences the sensitivity to the cluster structure).
- The fit() method is called to fit the OPTICS model to the data.

- **Visualizing the Clustering Result:**

- The code creates a scatter plot of the data points, where each point is colored according to its assigned cluster label obtained from OPTICS clustering.
- The title of the plot is set as 'OPTICS Clustering Result', and the plot is displayed using plt.show().

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## DENSITY BASED CLUSTERING

Flat Clustering:  
one level of clusters



e.g. density-based clustering algorithm  
DBSCAN [KDD 96]

Hierarchical Clustering:  
nested clusters



e.g. density-based clustering algorithm  
OPTICS [SIGMOD 99]

