



DICE
ANALYTICS

DATA SCIENCE & MACHINE LEARNING COURSE

<https://www.facebook.com/diceanalytics/>
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DBSCAN – Density Based Clustering

Density-Based Spatial Clustering of Applications with Noise

epsilon: 1.0
MinPoints: 4

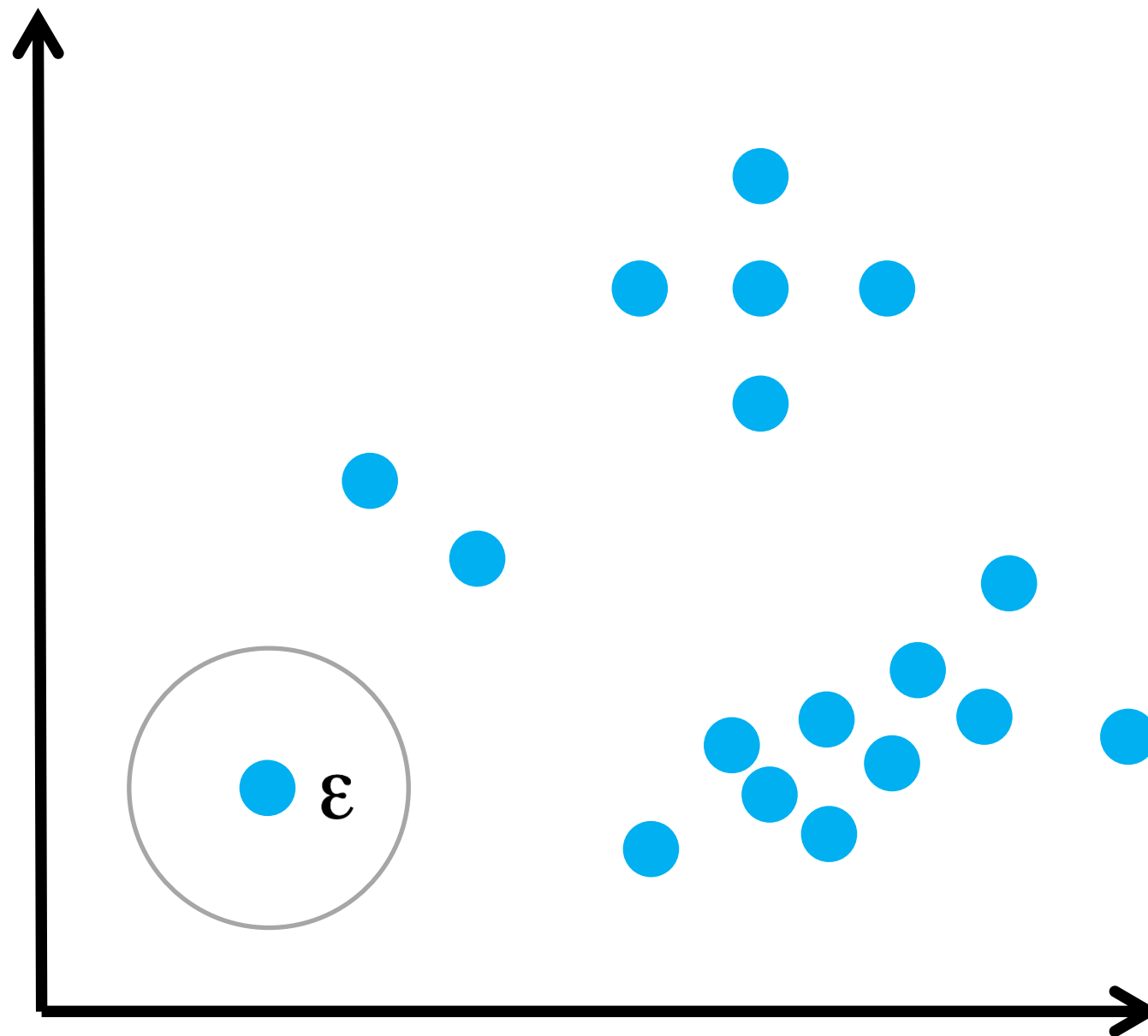


DBSCAN – Density Based Clustering

Inputs

Epsilon = 1.0

ϵ



DBSCAN – Density Based Clustering

Inputs

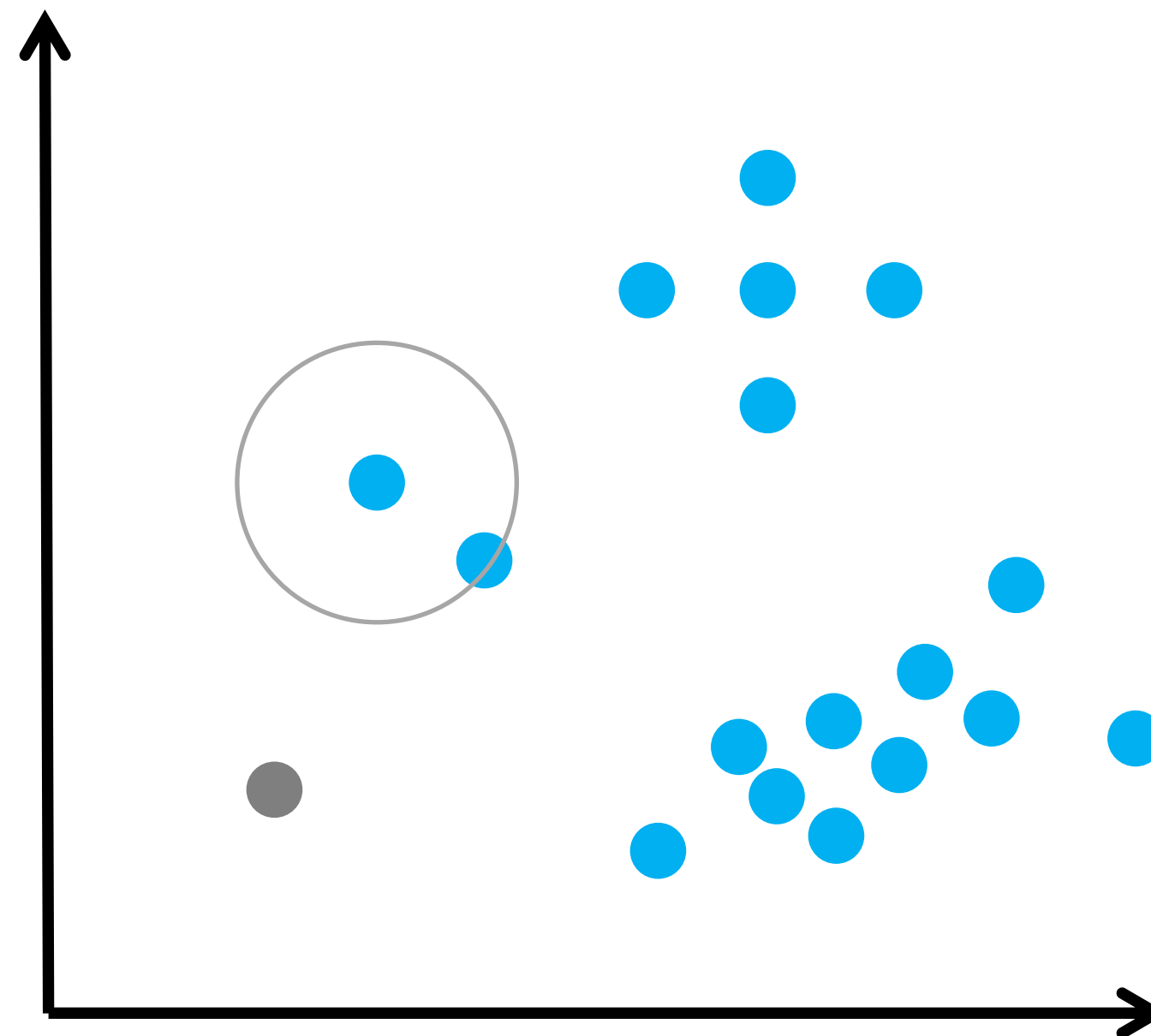
ϵ

Epsilon = 1.0

Search distance around points

Min Points = 5

Minimum points required to
form a density cluster



● Noise Point

DBSCAN – Density Based Clustering

Inputs

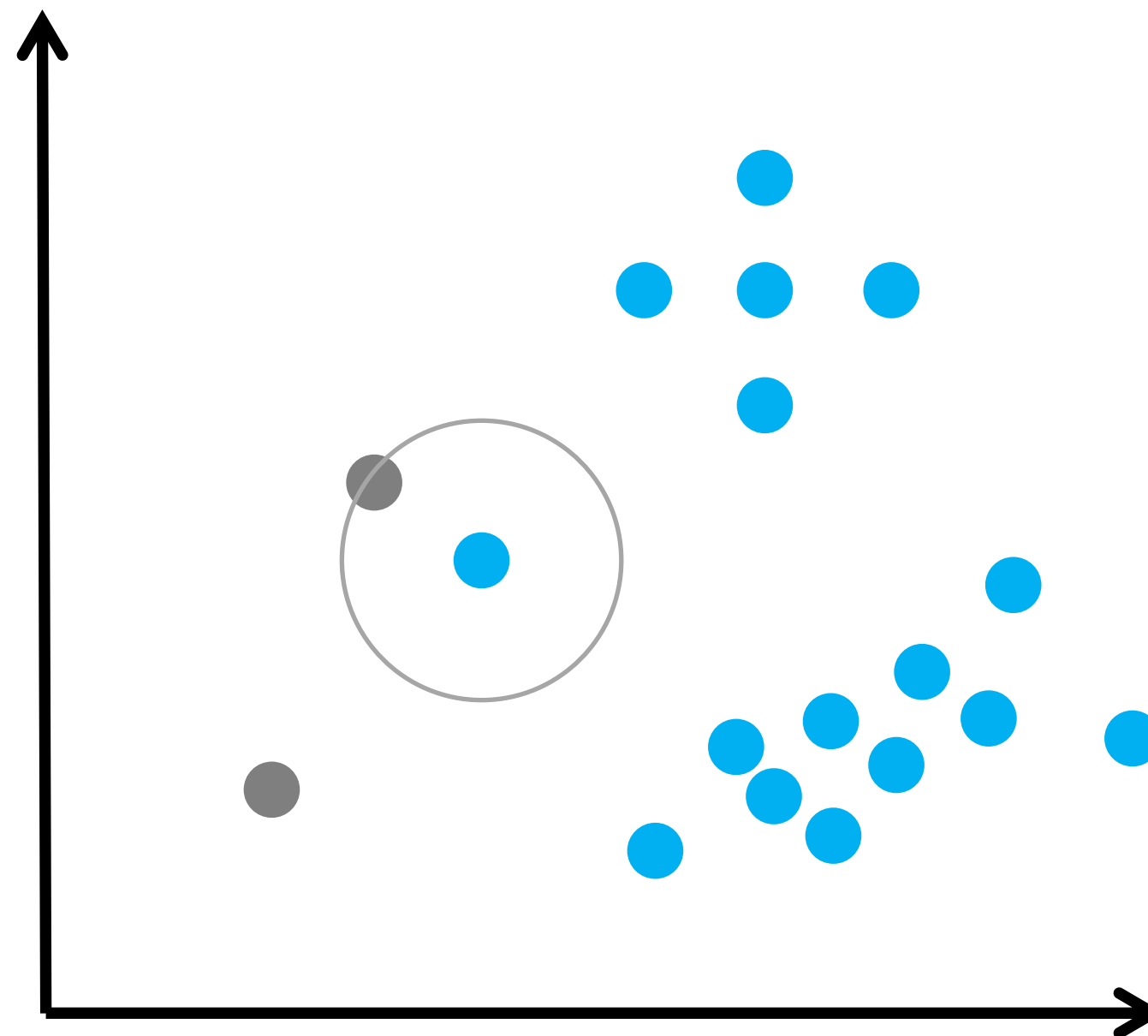
ϵ

Epsilon = 1.0

Search distance around points

Min Points = 5

Minimum points required to
form a density cluster



● Noise Point

DBSCAN – Density Based Clustering

Inputs

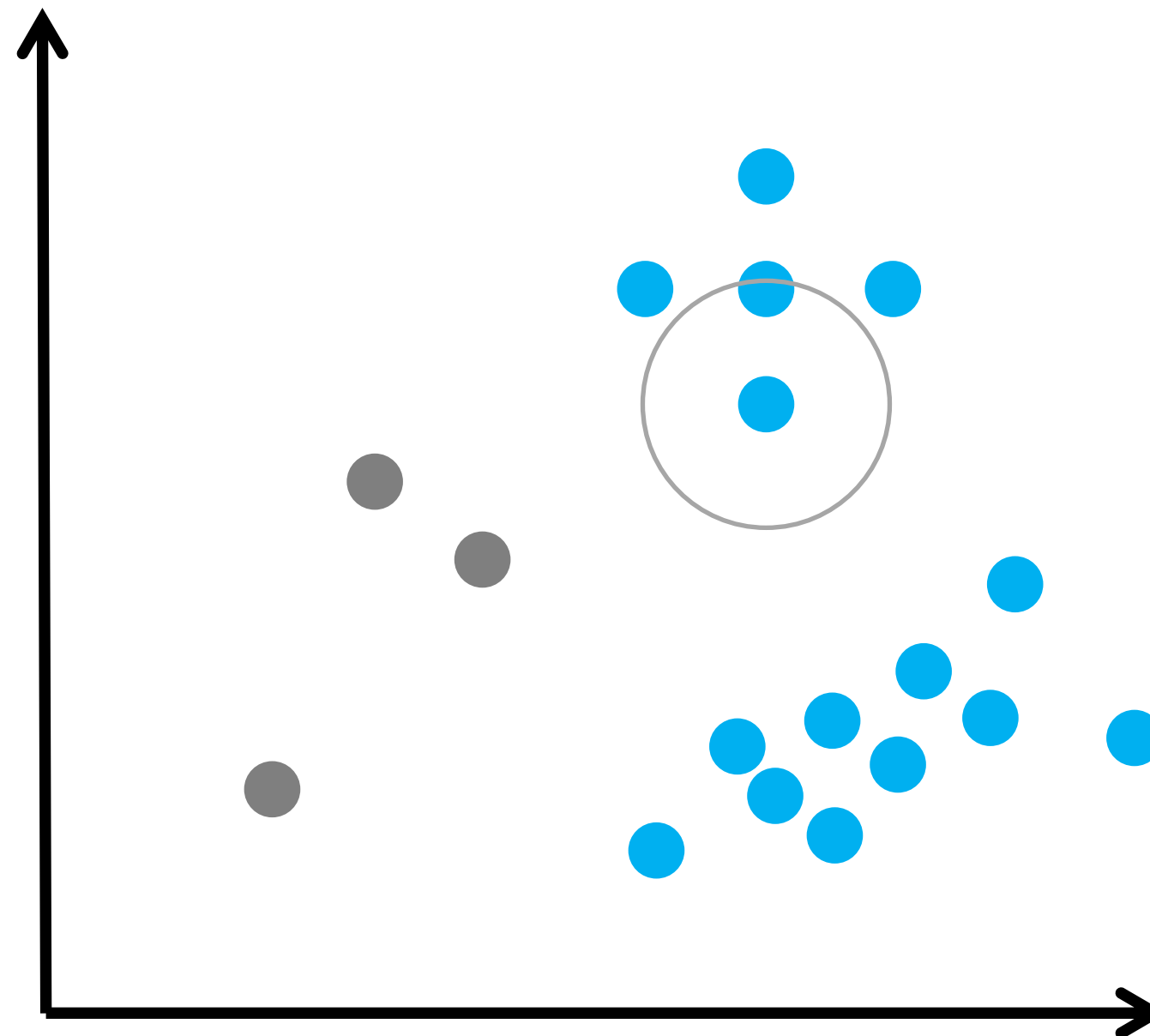
ϵ

Epsilon = 1.0

Search distance around points

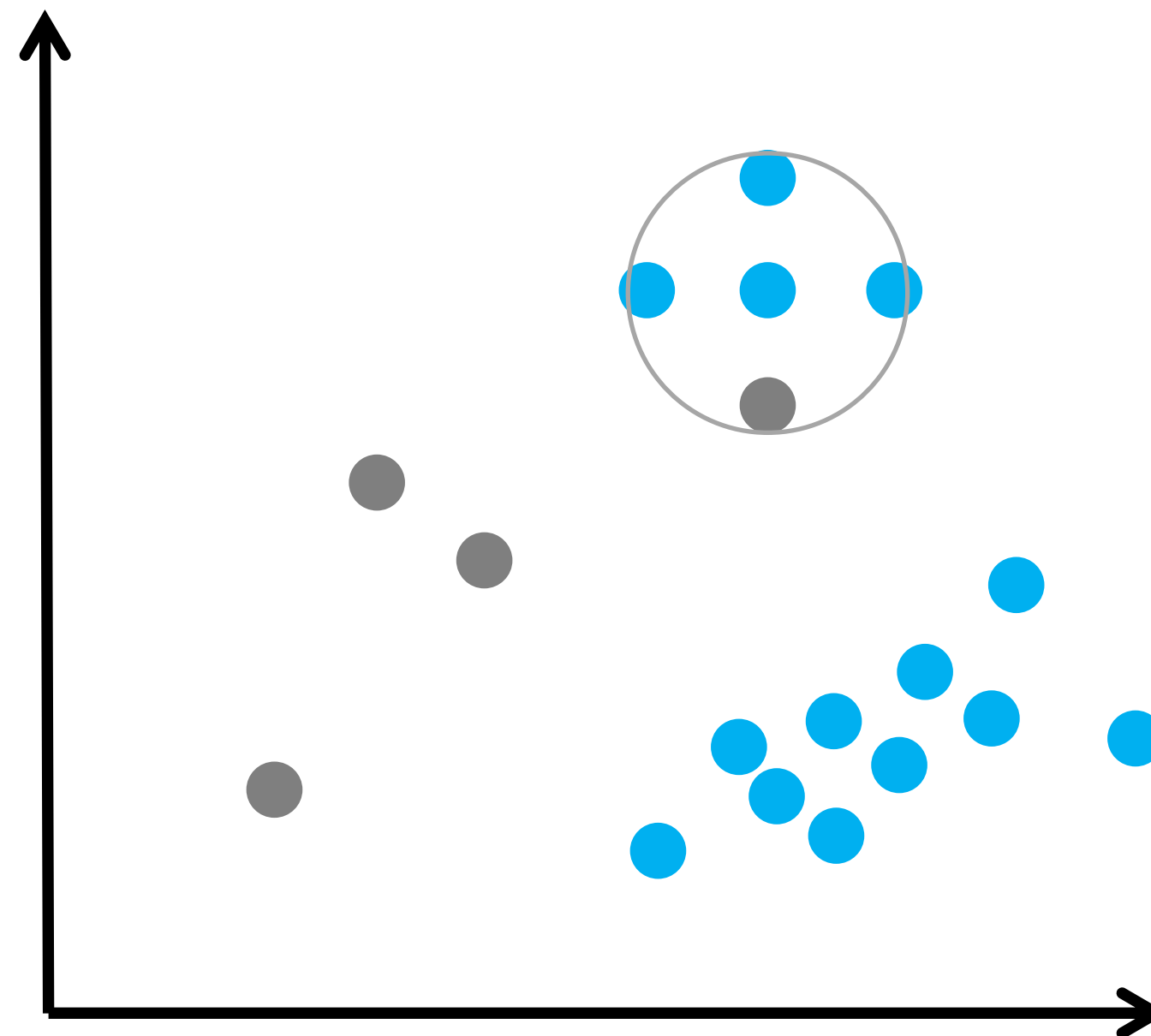
Min Points = 5

Minimum points required to
form a density cluster



● Noise Point

DBSCAN – Density Based Clustering



Inputs

ϵ

Epsilon = 1.0

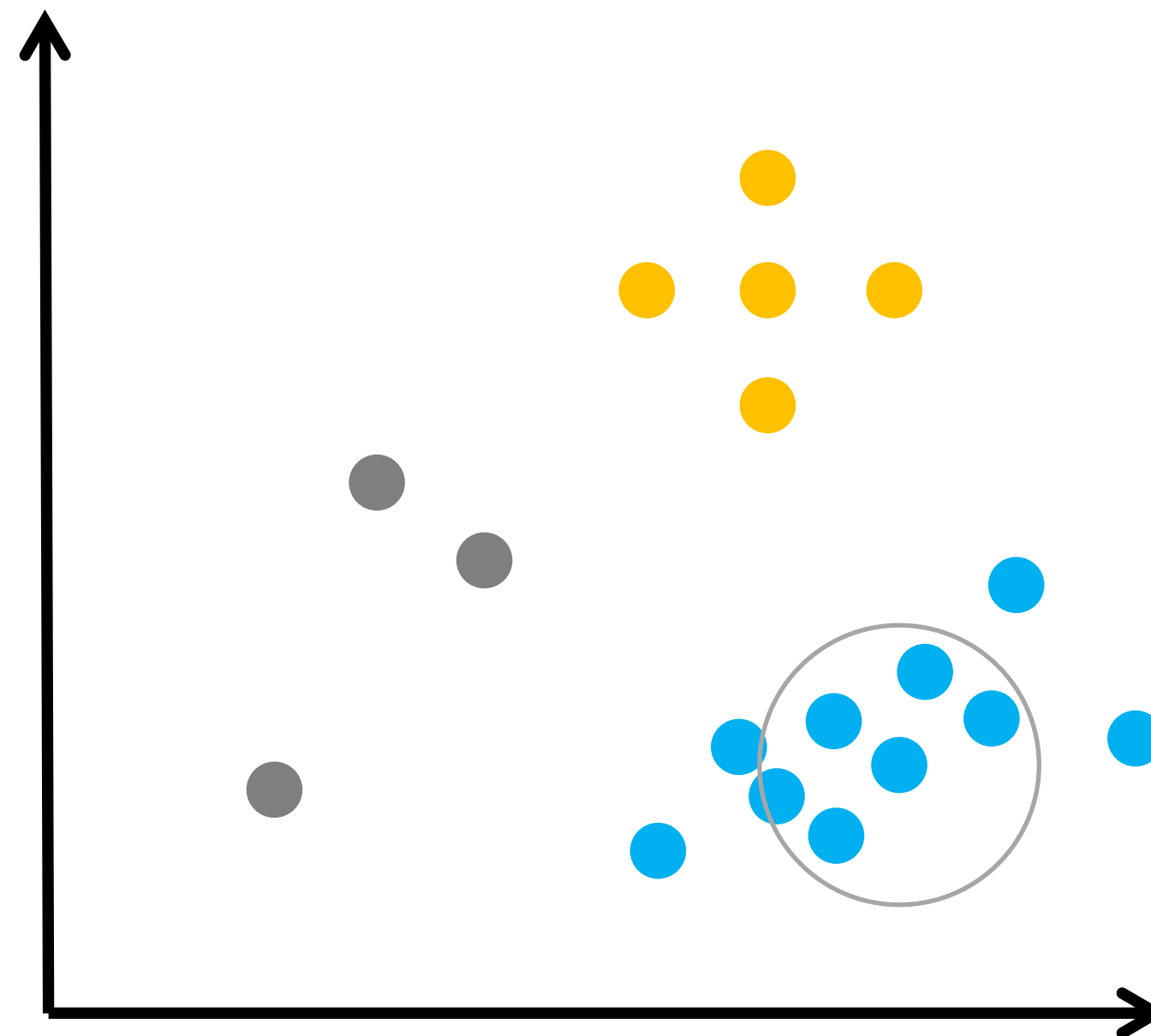
Search distance around points

Min Points = 5

Minimum points required to
form a density cluster

● Noise Point

DBSCAN – Density Based Clustering



Inputs

ϵ

Epsilon = 1.0

Search distance around points

Min Points = 5

Minimum points required to
form a density cluster

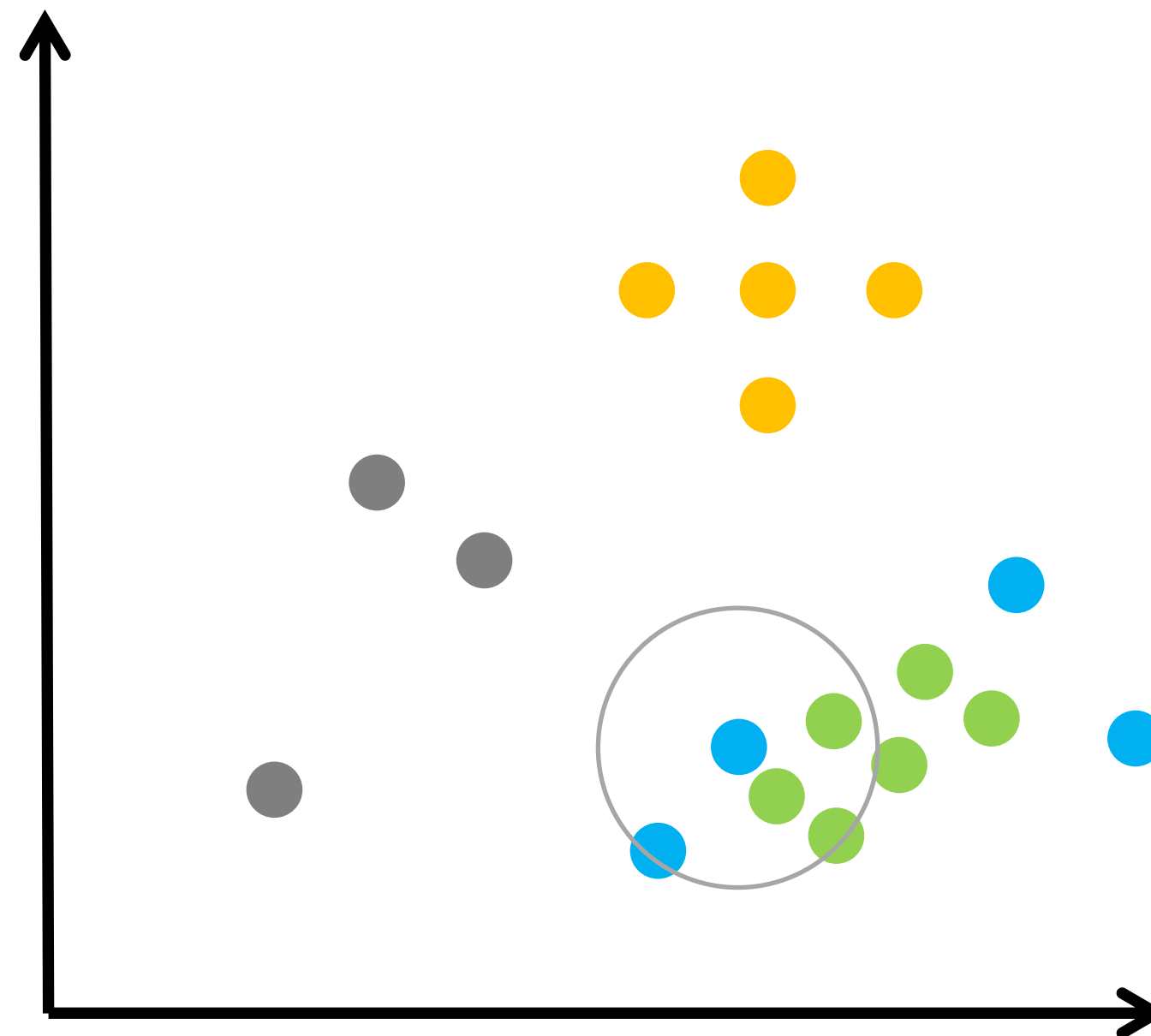


Cluster-1



Noise Point

DBSCAN – Density Based Clustering



Inputs

ϵ

Epsilon = 1.0

Search distance around points

Min Points = 5

Minimum points required to
form a density cluster

- Cluster-2
- Cluster-1
- Noise Point

DBSCAN – Density Based Clustering

Inputs

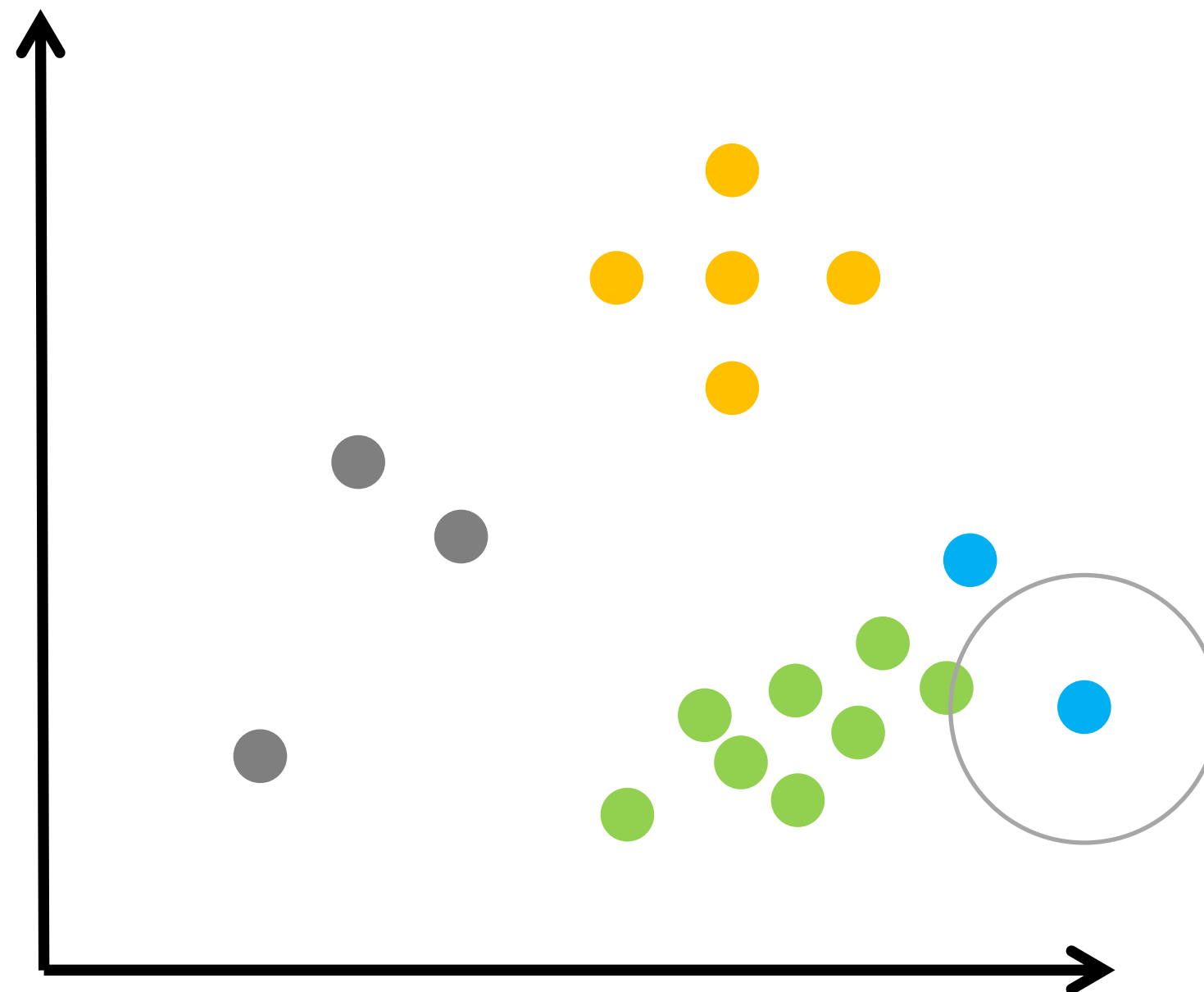
ϵ

Epsilon = 1.0

Search distance around points

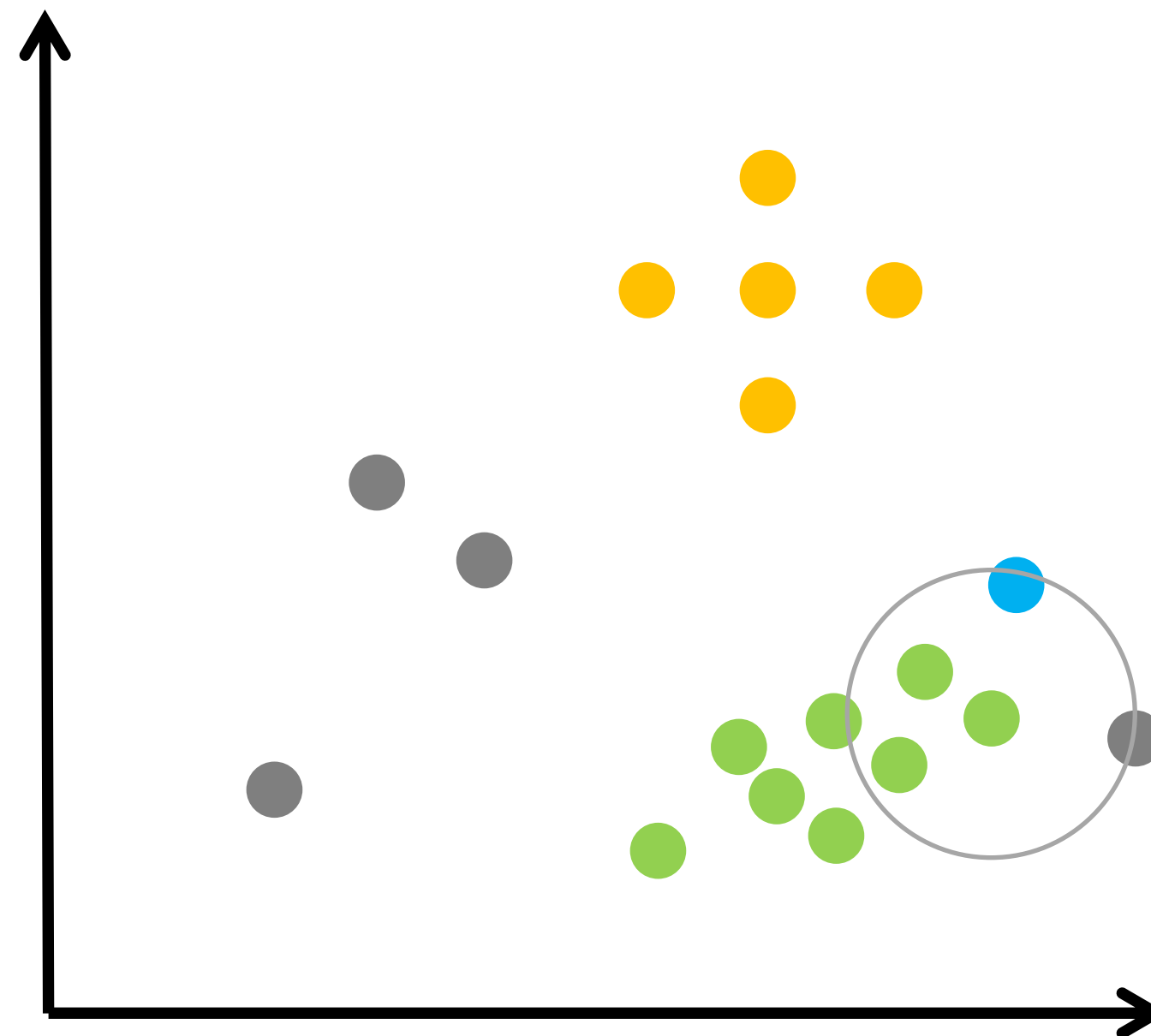
Min Points = 5

Minimum points required to
form a density cluster



- Cluster-2
- Cluster-1
- Noise Point

DBSCAN – Density Based Clustering



Inputs

ϵ

Epsilon = 1.0

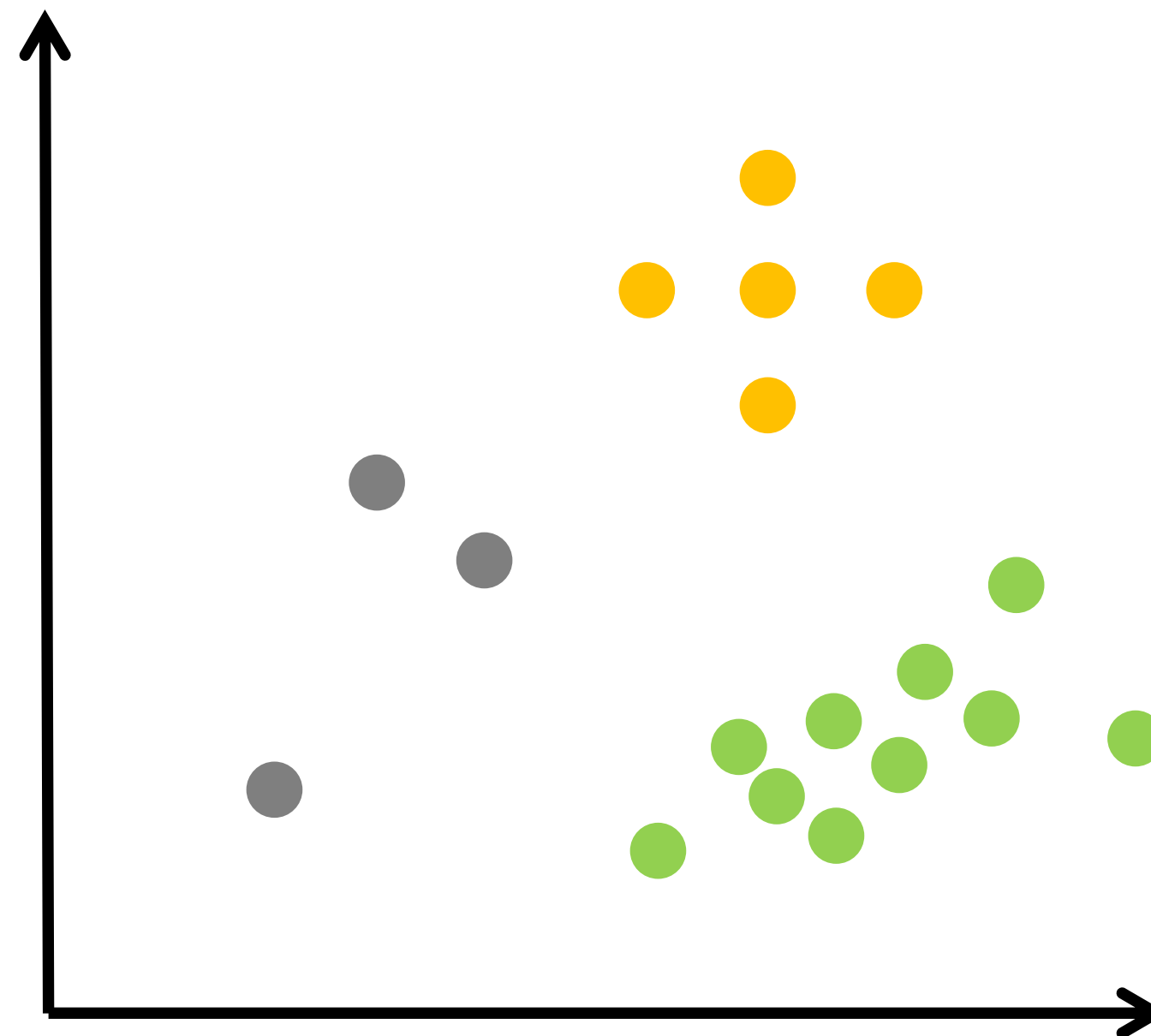
Search distance around points

Min Points = 5

Minimum points required to
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- Cluster-2
- Cluster-1
- Noise Point

DBSCAN – Density Based Clustering



Inputs

ϵ

Epsilon = 1.0

Search distance around points

Min Points = 5

Minimum points required to
form a density cluster

- Cluster-2
- Cluster-1
- Noise Point

DBSCAN Implementation

```
class sklearn.cluster.DBSCAN (eps=0.5, min_samples=5, metric='euclidean', metric_params=None, algorithm='auto',  
leaf_size=30, p=None, n_jobs=None) \[source\]
```

```
>>> from sklearn.cluster import DBSCAN  
>>> import numpy as np  
>>> X = np.array([[1, 2], [2, 2], [2, 3],  
...              [8, 7], [8, 8], [25, 80]])  
>>> clustering = DBSCAN(eps=3, min_samples=2).fit(X)  
>>> clustering.labels_  
array([ 0,  0,  0,  1,  1, -1])  
>>> clustering  
DBSCAN(algorithm='auto', eps=3, leaf_size=30, metric='euclidean',  
        metric_params=None, min_samples=2, n_jobs=None, p=None)
```

<https://scikit-learn.org/stable/modules/generated/sklearn.cluster.DBSCAN.html>

DBSCAN Applications

Color image segmentation using density-based clustering



Pepper Segmented Pepper Plane Segmented Plane



Mountain Segmented Mountain Hand Segmented Hand



Tiger (with texture) Segmented Tiger Cameraman (with noise) Segmented Cameraman

https://www.researchgate.net/publication/4028066_Color_image_segmentation_using_density-based_clustering

DBSCAN Applications

Density Based Clustering to Oil Spill Detection on Satellite Images



https://shodhganga.inflibnet.ac.in/bitstream/10603/25515/11/11_chapter%205.pdf

DBSCAN Applications

Evolution of Star Formation of Dwarf Galaxies within Extragalactic Cluster Substructures

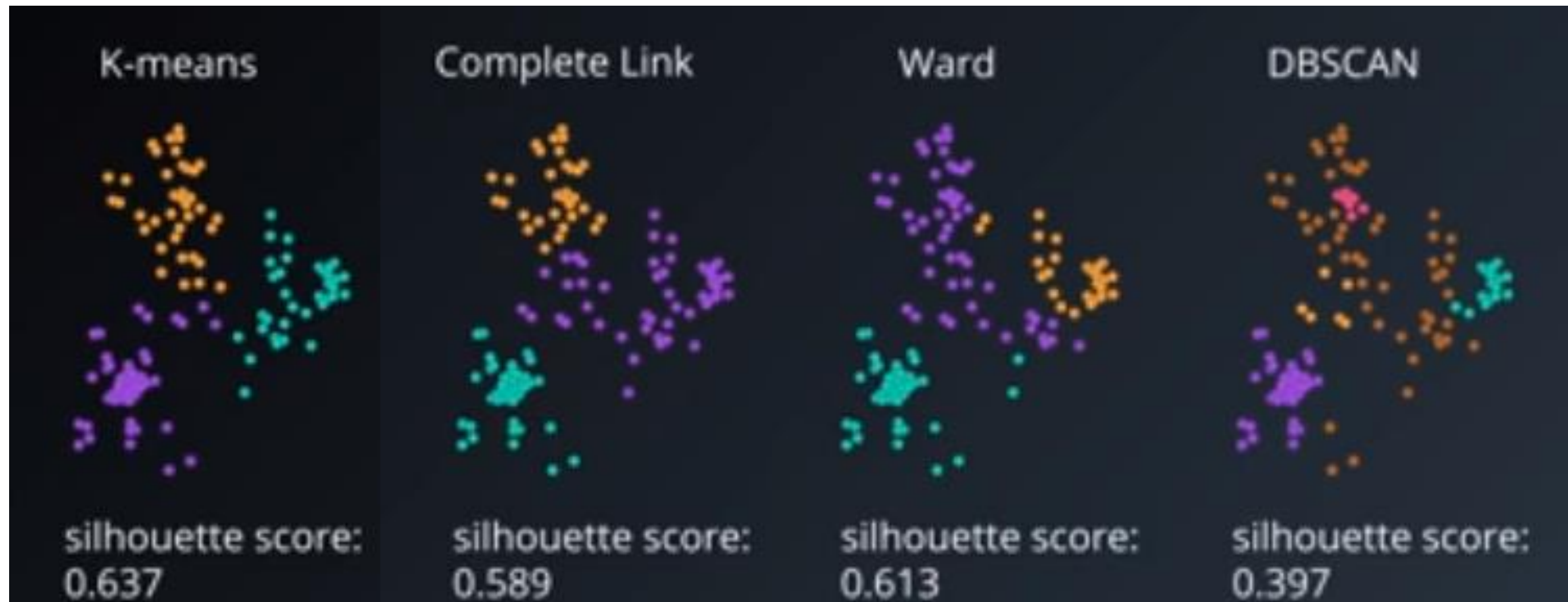


https://www.haystack.mit.edu/edu/reu/2016/files/2016_Archer_Presentation.pdf

Comparing Clustering Algos

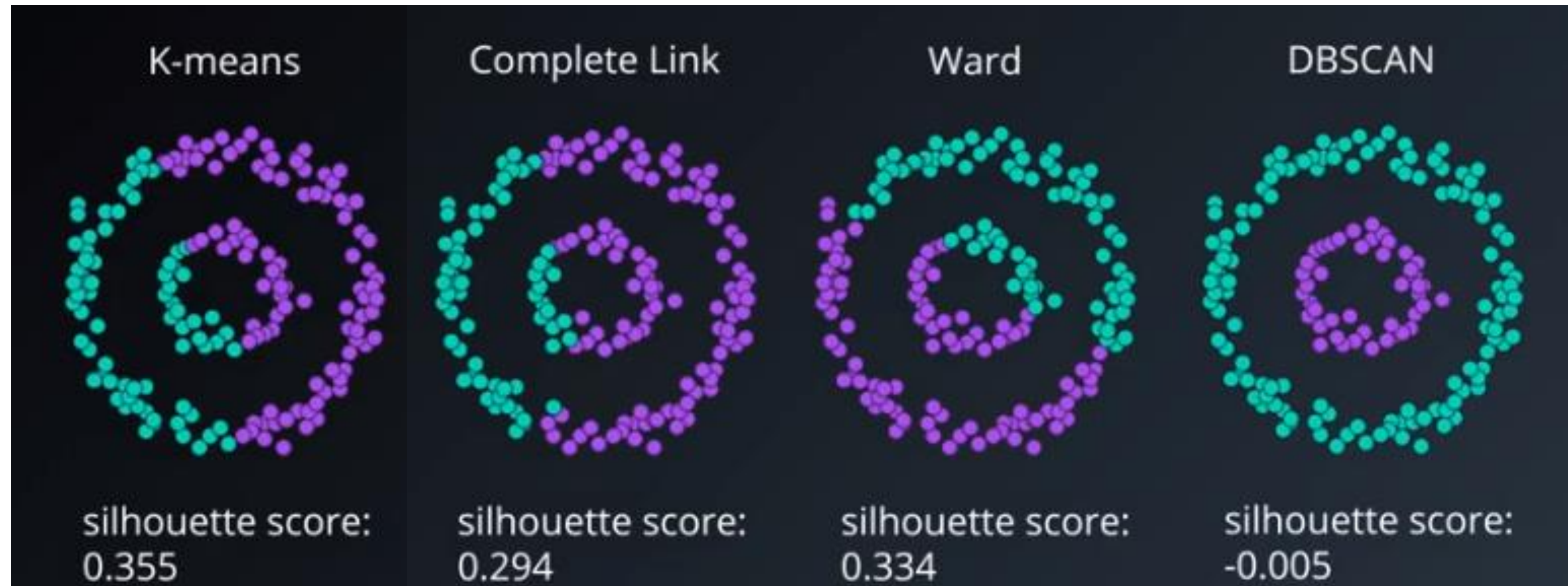


Comparing Clustering Algos



Comparing Clustering Algos

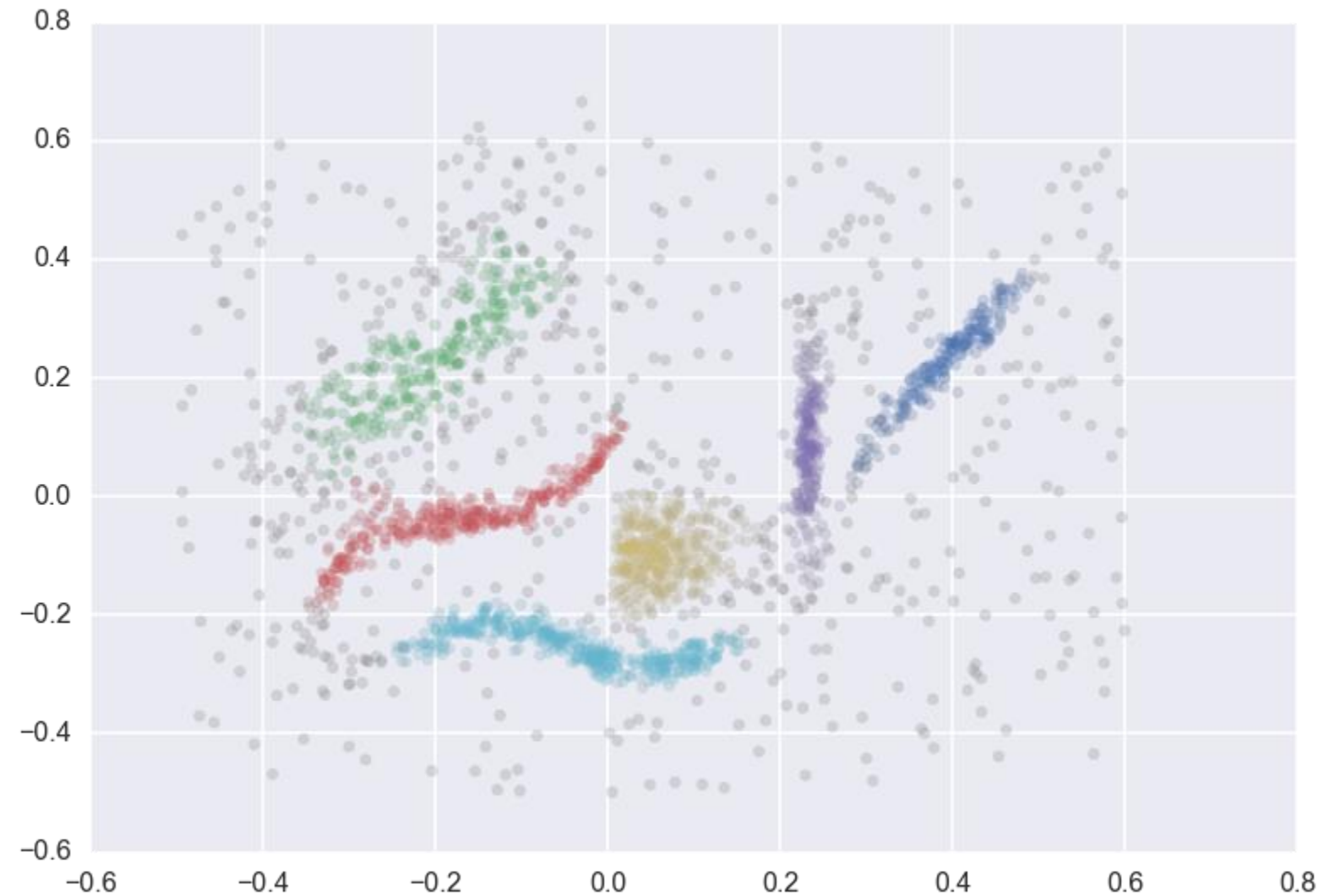
DBCV



DBCV – Density Based Cluster Validation

DBCV can validate clustering assignments on non-globular, arbitrarily shaped clusters. In essence, DBCV computes two values:

- The density **within** a cluster
- The density **between** clusters



<https://epubs.siam.org/doi/pdf/10.1137/1.9781611973440.96>