



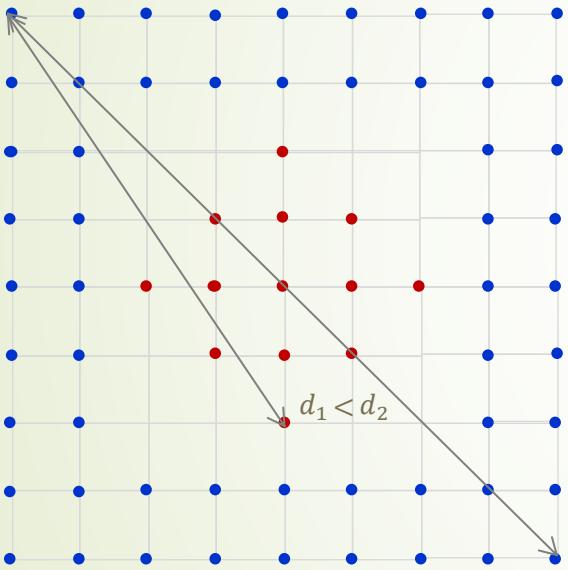
香港城市大學
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Cluster Analysis: Density-Based Methods

CS5483 Data Warehousing and Data Mining

How to identify the following clusters?



- Remedy?
- Why they should be clusters?
They are **d**_____ regions.

- ▶ Why centroid-based method fails?
 - ▶ Both clusters have the same c_____.
 - ▶ Bias towards s_____ cluster.
- ▶ Why single-linkage fails?
 - ▶ The c_____ distance of the two clusters is no larger than those between two points in the same cluster.
 - ▶ Chaining phenomenon
- ▶ Why complete-linkage method fails?
 - ▶ The f_____ distance of the two clusters is no larger than those between two points in one cluster.
 - ▶ Bias towards s_____ cluster.

Identify pillars of dense regions

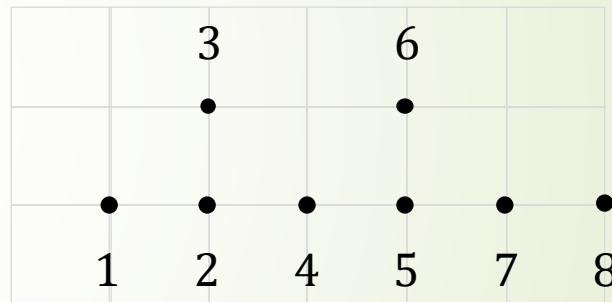
- ε -neighbourhood of $\mathbf{p} \in D$ is the region:

$$N_\varepsilon(\mathbf{p}) := \{\mathbf{q} \in \mathbb{R}^d \mid \text{dist}(\mathbf{p}, \mathbf{q}) \leq \varepsilon\}$$

- Within r_____ $\varepsilon > 0$
- from the c_____ \mathbf{p} .

- **C_____ point:** $\mathbf{p} \in D$ such that
 $|D \cap N_\varepsilon(\mathbf{p})| \geq \text{MinPts}$
- How to find clusters?

$$\varepsilon = 1 \quad \text{MinPts} = 4$$



Core points: _____

Density-reachability

- q is **directly density-reachable** from p , denoted as $p \rightarrow q$, if

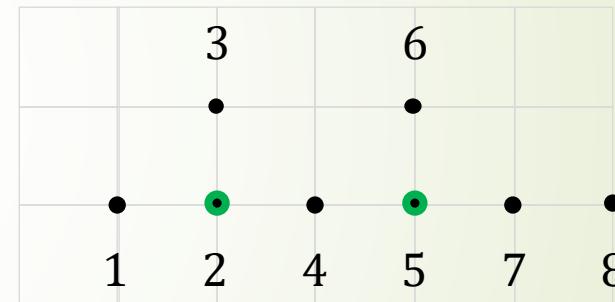
1. p is a core point and
2. $q \in N_\varepsilon(p)$

- q is **density-reachable** from p if there is a **path**

$$p \rightarrow \dots \rightarrow q.$$

- All points in the path must be core points except q .
- q is called a **border** point if it is not a core point.

$$\varepsilon = 1 \quad \text{MinPts} = 4$$

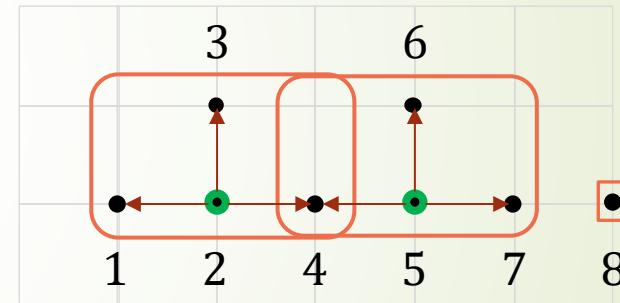


- $p_1 \rightarrow p_2$? _____

Density-connectedness

- q is **density-connected** with p , denoted as $p \sim q$, if p, q are reachable from a c_____ core point.
- Density-connected components as clusters? Y/N because _____

$\varepsilon = 1$ MinPts = 4



- $p_1 \sim p_2$? _____

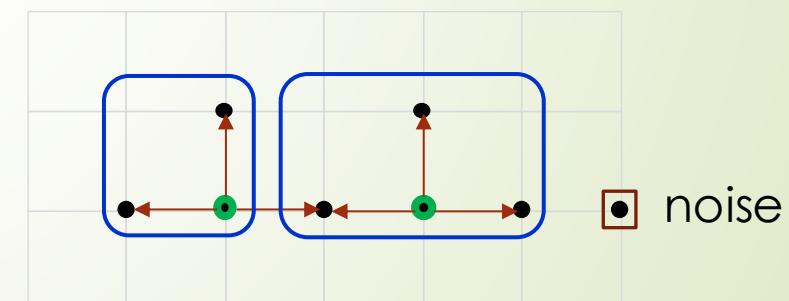
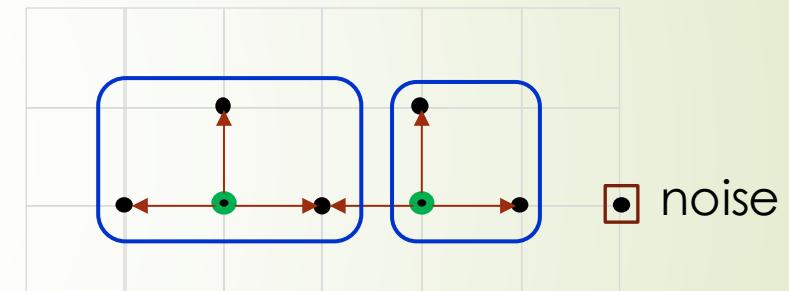
Border point: _____

DBSCAN

Density-Based Spatial Clustering of Applications with Noise

1. Identify core points and their density reachable points.
2. Return density-connected components of core points as clusters.
3. Assign border points to one of the cluster it is density-connected to.
4. Label the remaining points as noise.
► Clustering solution is not unique.

$$\varepsilon = 1 \quad \text{MinPts} = 4$$



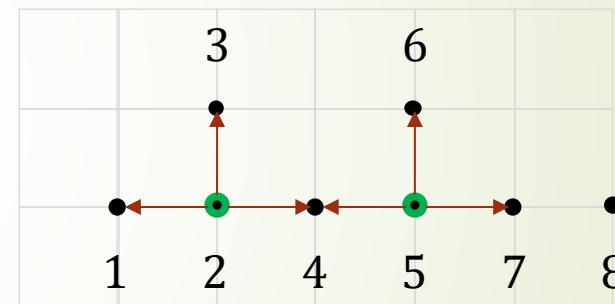
Uniqueness of clusters for core points

- DBSCAN give unique clusters for core points because density-connectedness is an equivalence relation on core points:

- $r \underline{\hspace{2cm}} : p \sim p$
- $s \underline{\hspace{2cm}} : p \sim q \Leftrightarrow q \sim p$
- $t \underline{\hspace{2cm}} : p \sim r, r \sim q \Rightarrow p \sim q$

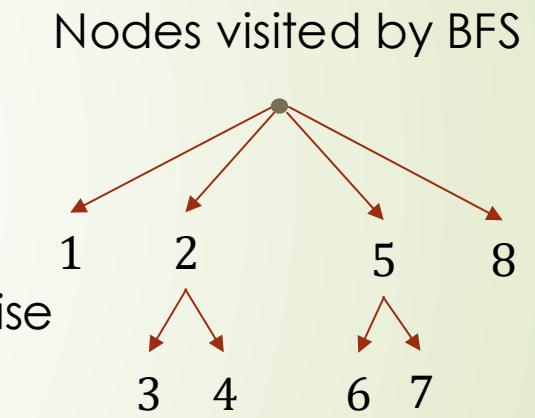
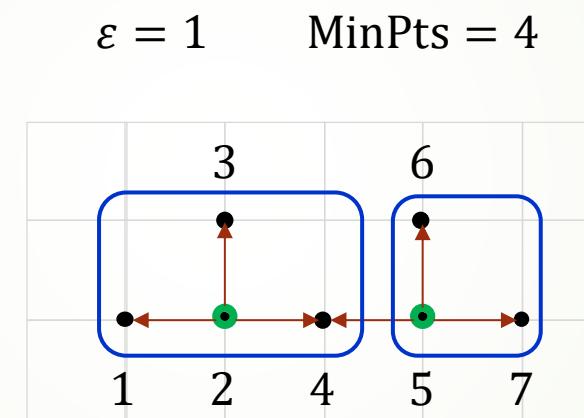
- However, transitivity can fail on border points:
 - E.g.,
 - see the last slide for the mistake in [Han11].

$$\varepsilon = 1 \quad \text{MinPts} = 4$$



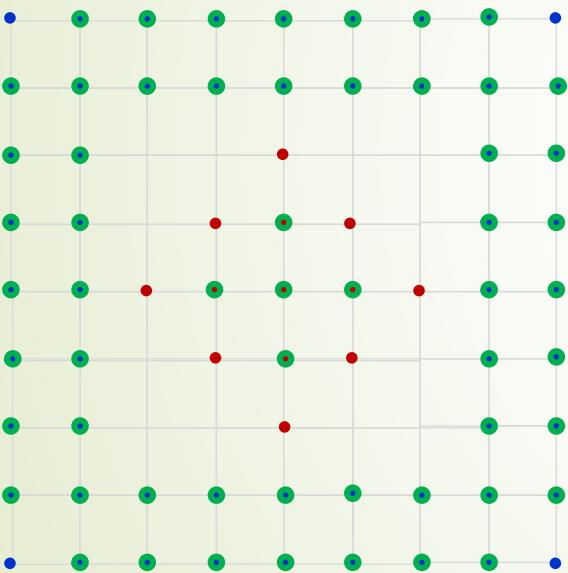
Implementation

- ▶ Repeatedly start a breadth first search (BFS) on a new point to add density-reachable points (not already assigned to another cluster) to a cluster.
- ▶ Label remaining points as noise.
- ▶ Complexity: _____
(or $O(n \log n)$ with spatial indexing.)



Clusters can be non-convex

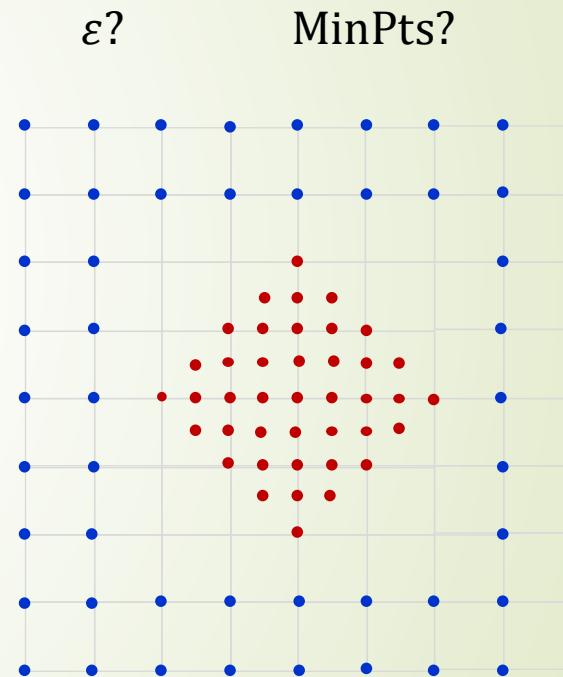
$\varepsilon = 1$ MinPts = 4



- ▶ Identify the core points, clusters, ambiguous border points and the noise.

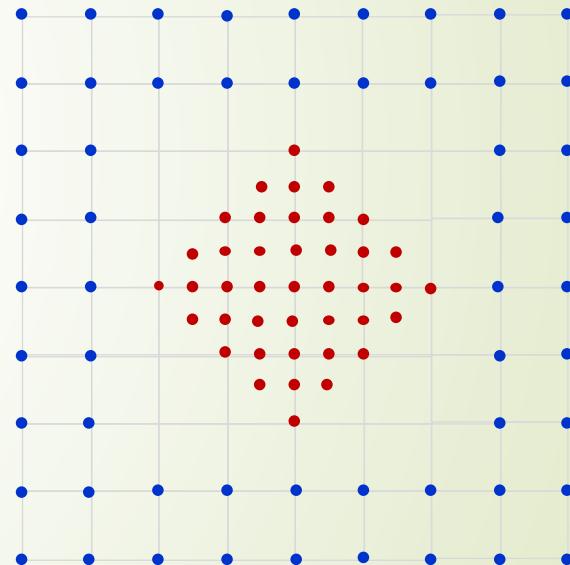
Limitations

- ▶ How to choose MinPts and ε ?
 - ▶ For the outer ring, want $\varepsilon \geq \underline{\hspace{2cm}}$.
 - ▶ But the corner point of the inner diamond will be _____.
- ▶ What about clusters with different densities?



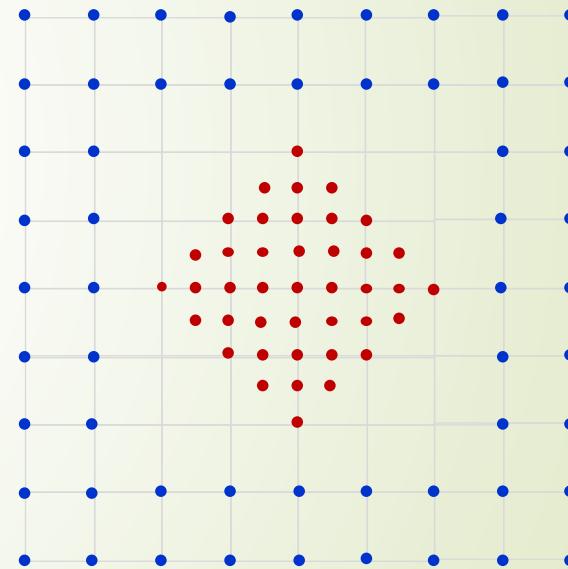
Density-based method

- ▶ Can DBSCAN recover the two clusters?
 - ▶ How to choose the parameters ε and MinPts?
- ▶ How to handle clusters with different densities?
 - ▶ Varying ε to obtain a h_____ of clusters.
 - ▶ How to do this efficiently?



OPTICS

- ▶ Ordering Points to Identify the Clustering Structure.
 - ▶ Idea: Apply DBSCAN but visit c_____ points first.
- ▶ Choice of parameters:
 - ▶ MinPts normally chosen as dimension + 1. Why?
[Optional] See [Carathéodory's theorem](#).
 - ▶ Choose the worst-case (largest) radius ε for the neighborhood.
- ▶ For the example, we can choose
 - ▶ MinPts =_____
 - ▶ $\varepsilon \geq$ _____, where all the points are density-connected.



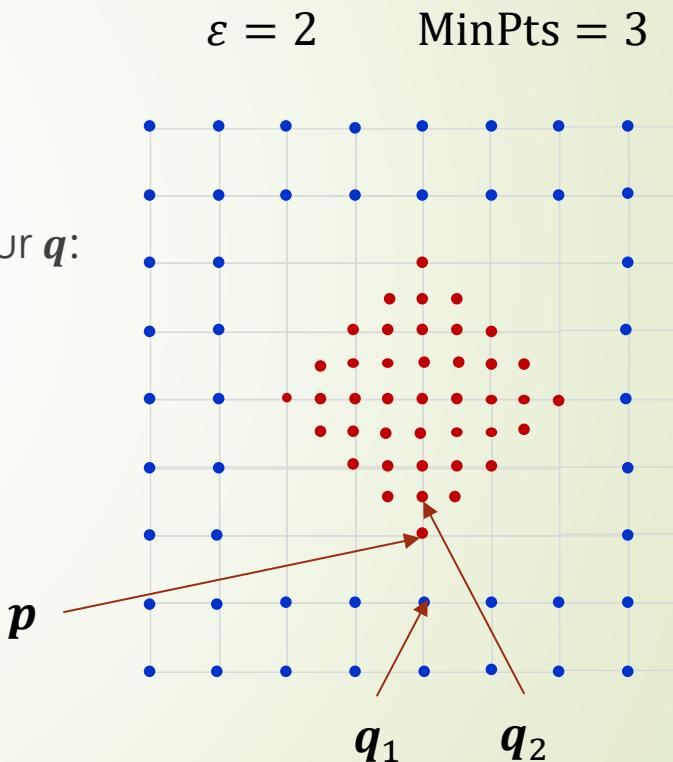
Core and reachability distances

- ▶ When a core point \mathbf{p} is reached,
 - ▶ Calculate the true density:
 $\text{core-distance}(\mathbf{p}) := \min\{0 \leq \varepsilon' \leq \varepsilon \mid |D \cap N_{\varepsilon'}(\mathbf{p})| \geq \text{MinPts}\}$
 - ▶ Calculate the distance to its density-reachable neighbour \mathbf{q} :
 $\text{reachability-distance}(\mathbf{q}) := \max\{\text{dist}(\mathbf{p}, \mathbf{q}), \text{core-distance}(\mathbf{p})\}$
 - ▶ Visit neighbors with smaller reachability distance first.
 - ▶ Why use reachability distance instead of distance?
-

$\text{core-distance}(\mathbf{p}) = \underline{\hspace{2cm}}$

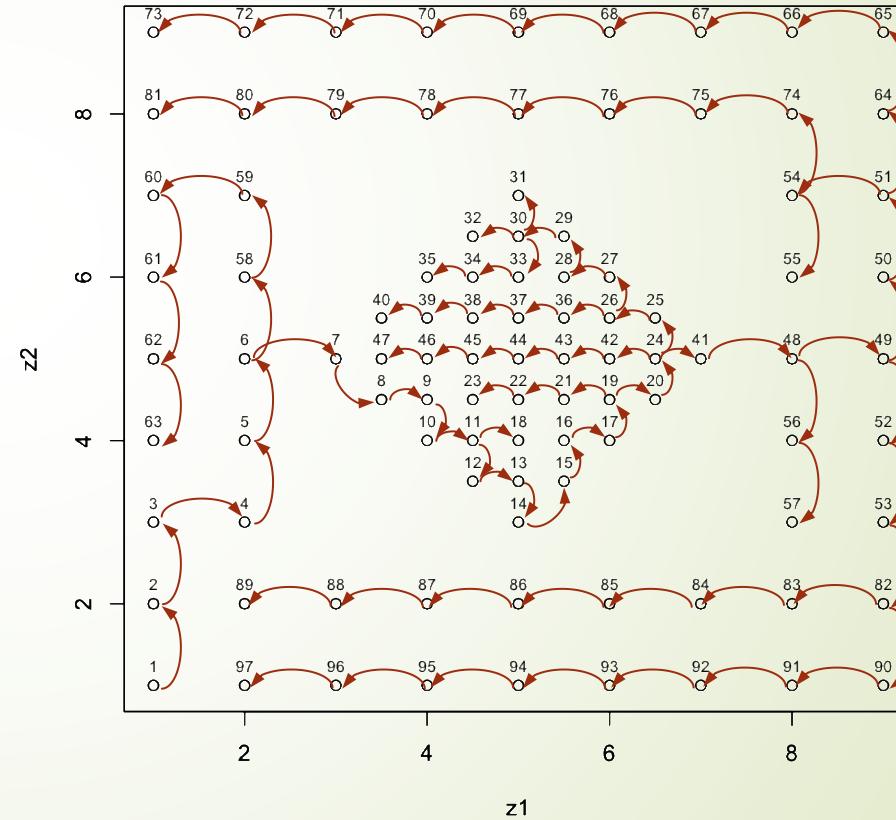
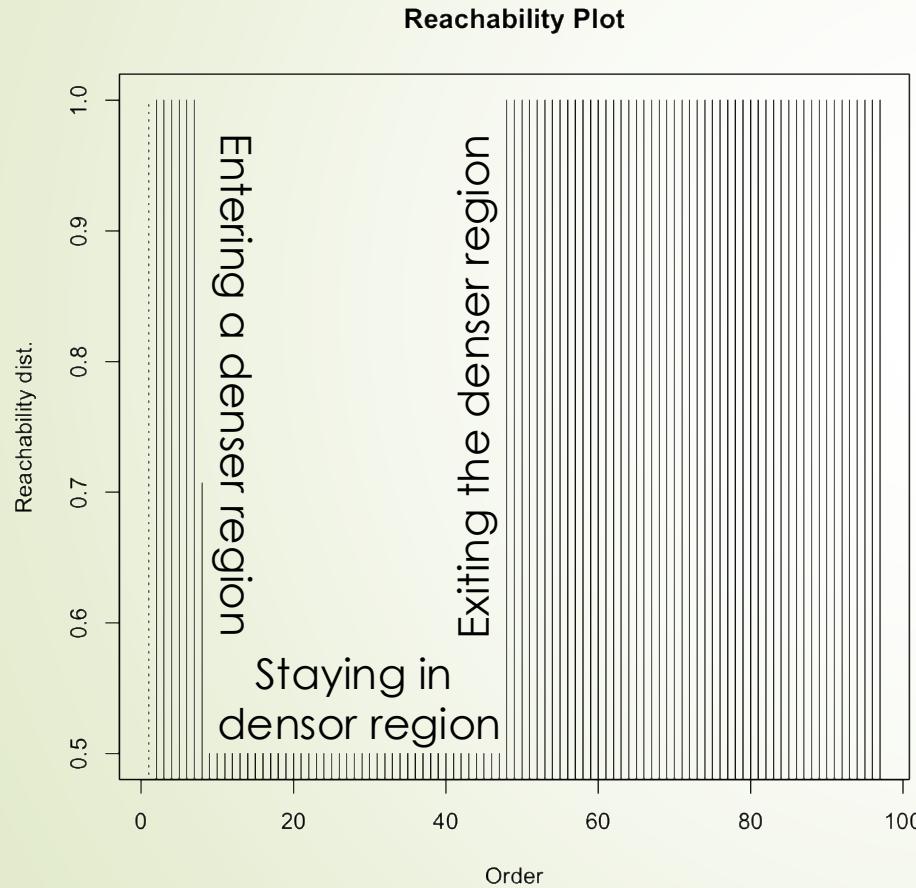
$\text{reachability-distance}(\mathbf{p}, \mathbf{q}_1) = \underline{\hspace{2cm}}$

$\text{reachability-distance}(\mathbf{p}, \mathbf{q}_2) = \underline{\hspace{2cm}}$

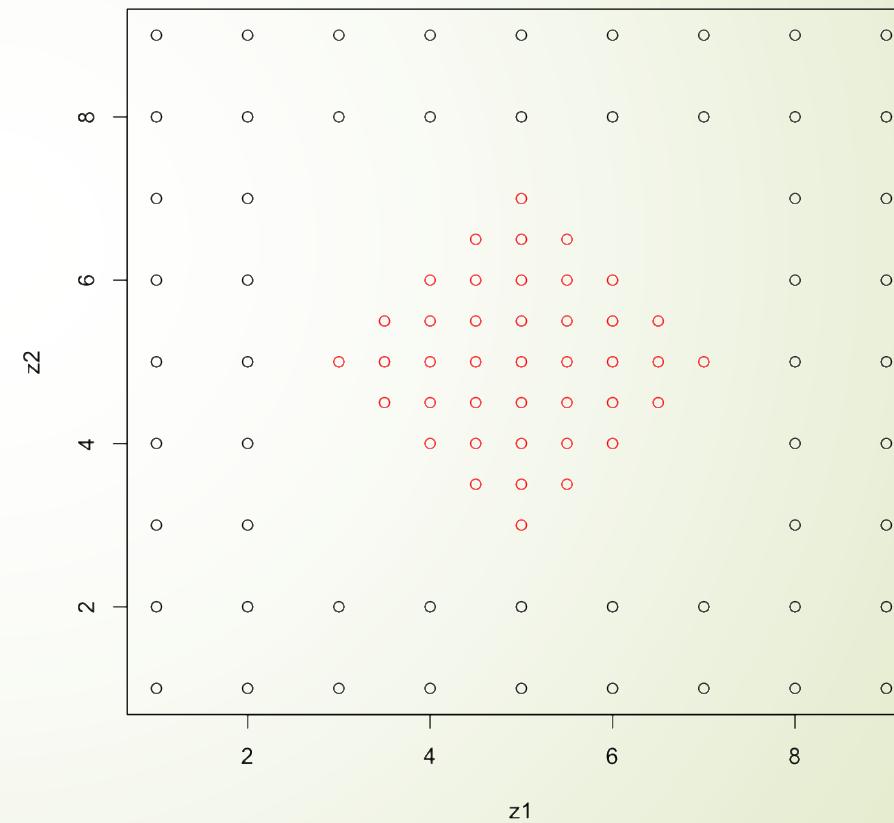


Reachability plot

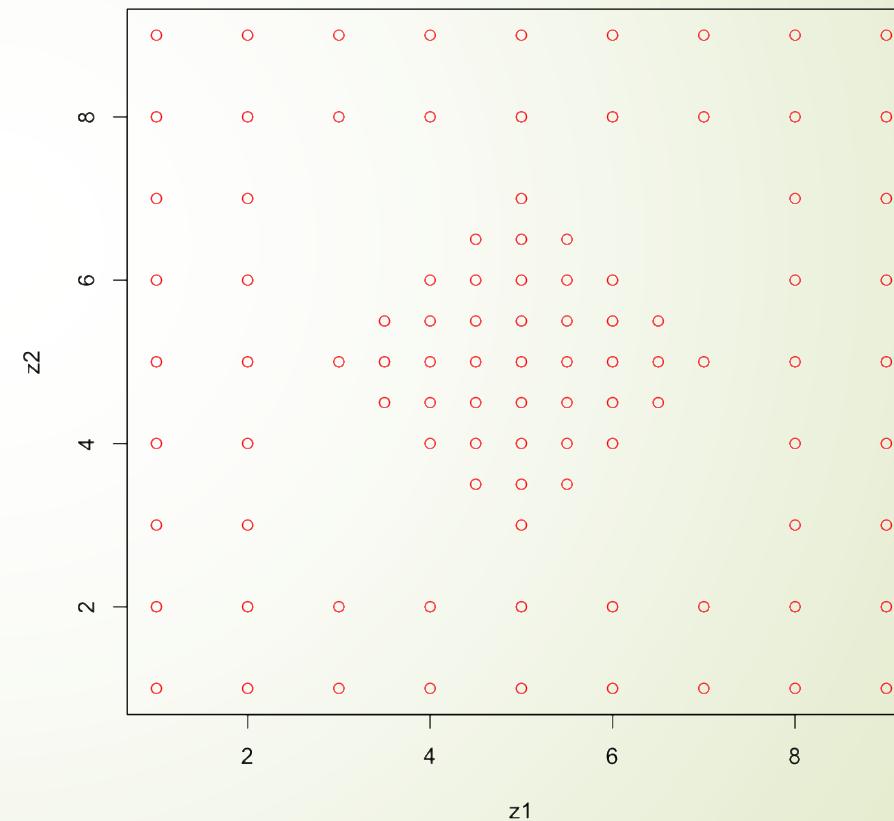
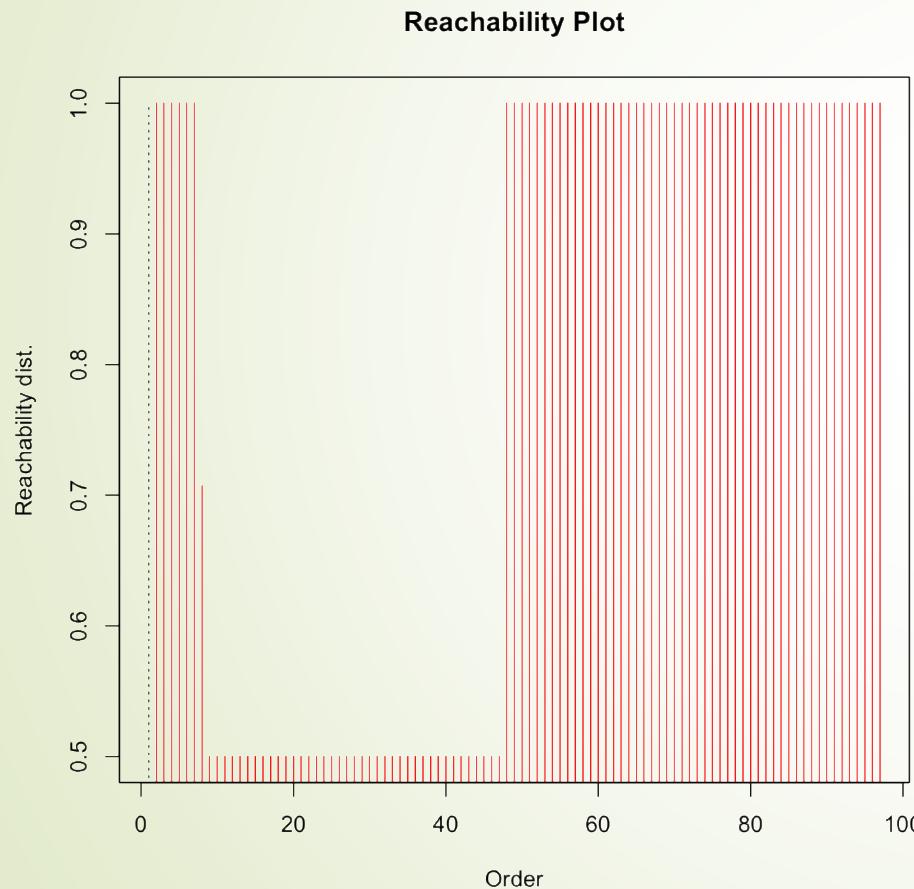
- Plot the reachability distances of the sequences of visited points.



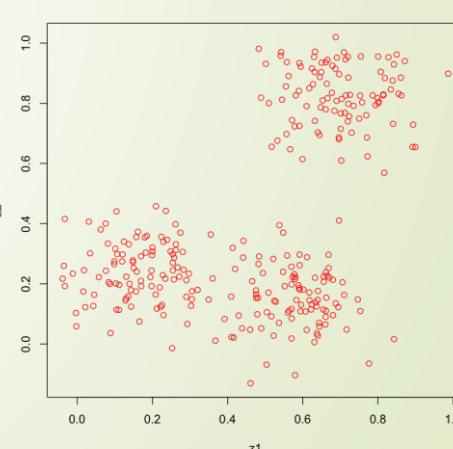
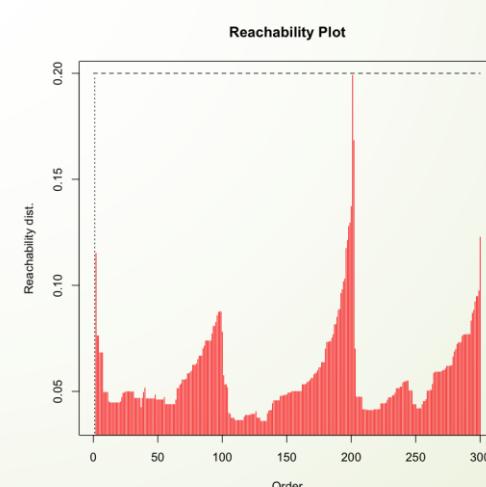
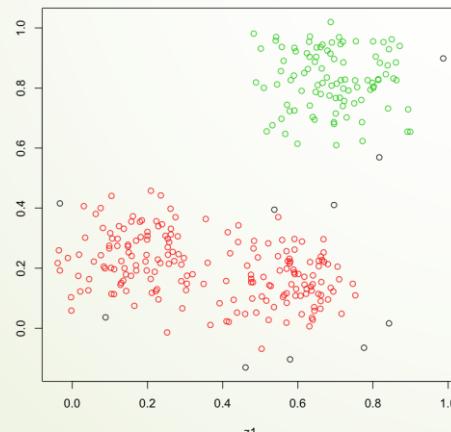
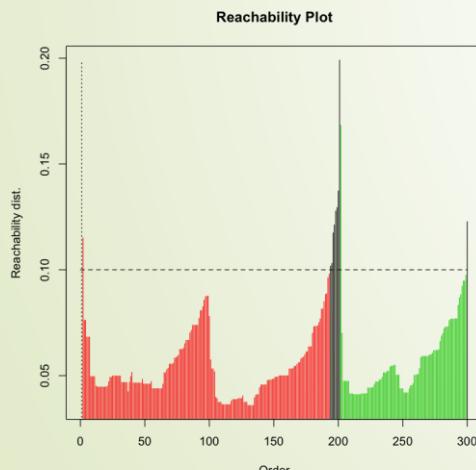
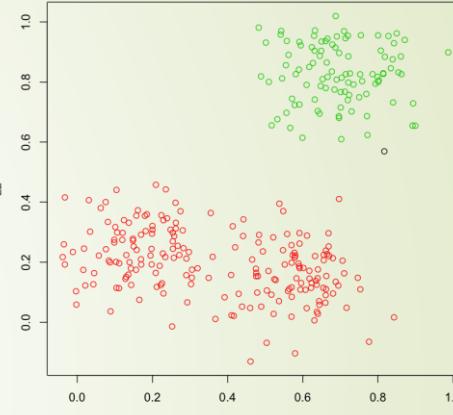
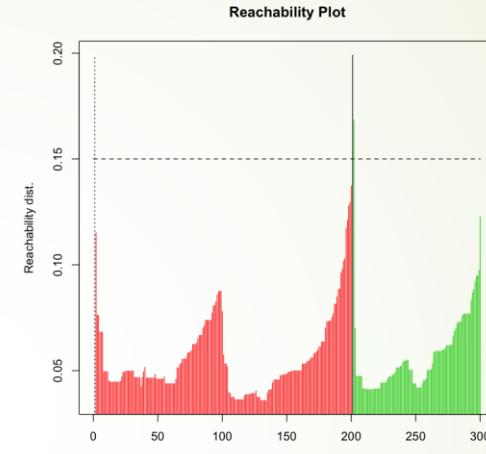
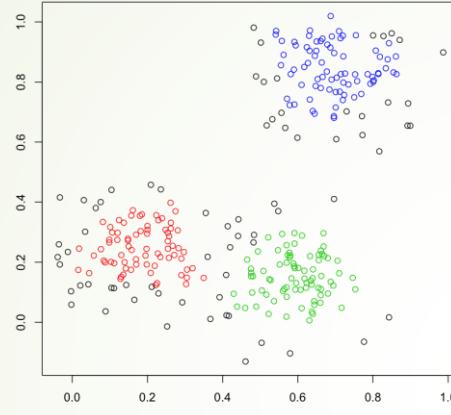
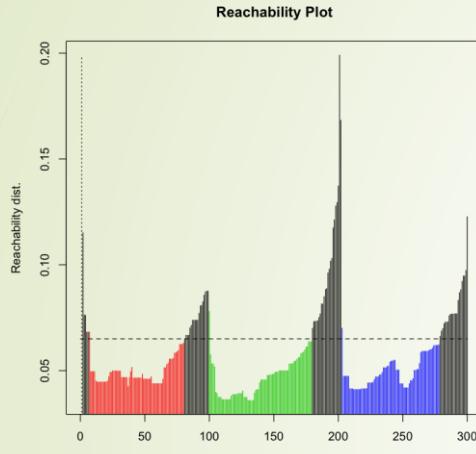
- ▶ A cluster at threshold ε is discovered as the valley of points with reachability distance below ε except for the first point.



- With the threshold of reachability raised to __ or above, the valley covers all the points.
- Note that the outer ring is not identified as a separate cluster from the inner diamond.



An example with multiple clusters



References

- ▶ 10.4.1 DBSCAN: Density-Based Clustering Based on Connected Regions with High Density
 - ▶ Errata on p.472: Density-connectedness is NOT an equivalence relation unless restricted to only core points.
- ▶ McInnes, Leland, and John Healy. "Accelerated Hierarchical Density Based Clustering." *Data Mining Workshops (ICDMW), 2017 IEEE International Conference on*. IEEE, 2017.