# OPTICS: Ordering Points To Identify the Clustering Structure

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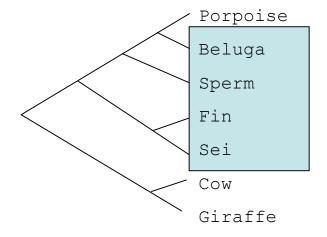
Presented by Chris Mueller November 4, 2004

# Clustering

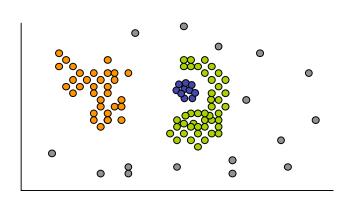
Goal: Group objects into meaningful subclasses as part of an exploratory process to insight into data or as a preprocessing step for other algorithms.

# **Clustering Strategies**

- Hierarchical
- Partitioning
  - k-means
  - Density Based



Density Based clustering requires a distance metric between points and works well on high dimensional data and data that forms irregular clusters.



## **DBSCAN: Density Based Clustering**

An object p is in the  $\varepsilon$ -neighborhood of q if the distance from p to q is less than  $\varepsilon$ .

A **core object** has at least *MinPts* in its  $\varepsilon$ -neighborhood.

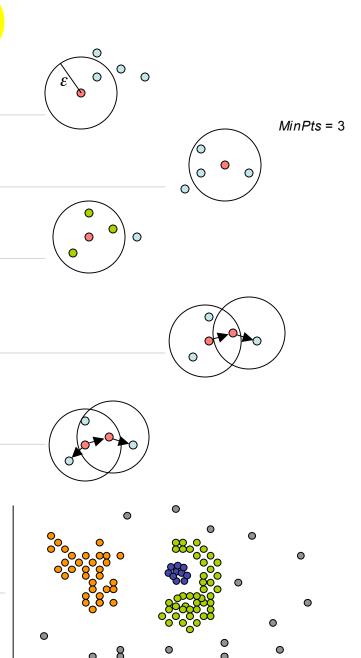
An object p is **directly density-reachable** from object q if q is a core object and p is in the  $\varepsilon$ -neighborhood of q.

An object p is **density-reachable** from object q if there is a chain of objects  $p_1, ..., p_n$ , where  $p_1 = q$  and  $p_n = p$  such that  $p_{i+1}$  is directly density reachable from  $p_i$ .

An object p is **density-connected** to object q if there is an object o such that both p and q are **density-reachable** from o.

A **cluster** is a set of density-connected objects which is maximal with respect to density-reachability.

**Noise** is the set of objects not contained in any cluster.



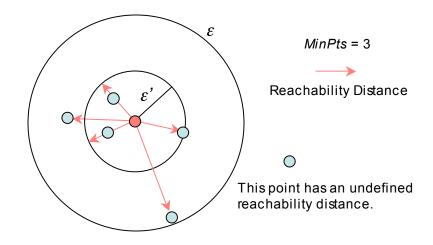
## **OPTICS: Density-Based Cluster Ordering**

OPTICS generalizes DB clustering by creating an ordering of the points that allows the extraction of clusters with arbitrary values for  $\varepsilon$ .

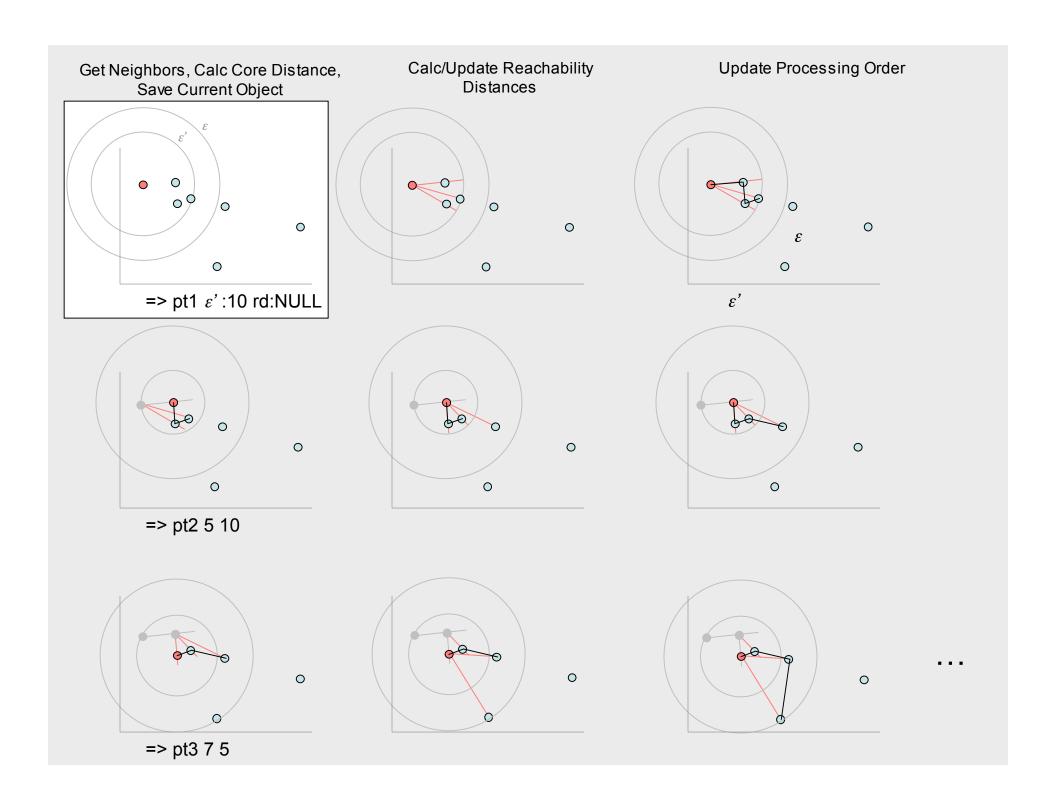
The **generating-distance**  $\varepsilon$  is the largest distance considered for clusters. Clusters can be extracted for all  $\varepsilon_i$  such that  $0 \le \varepsilon_i \le \varepsilon$ .

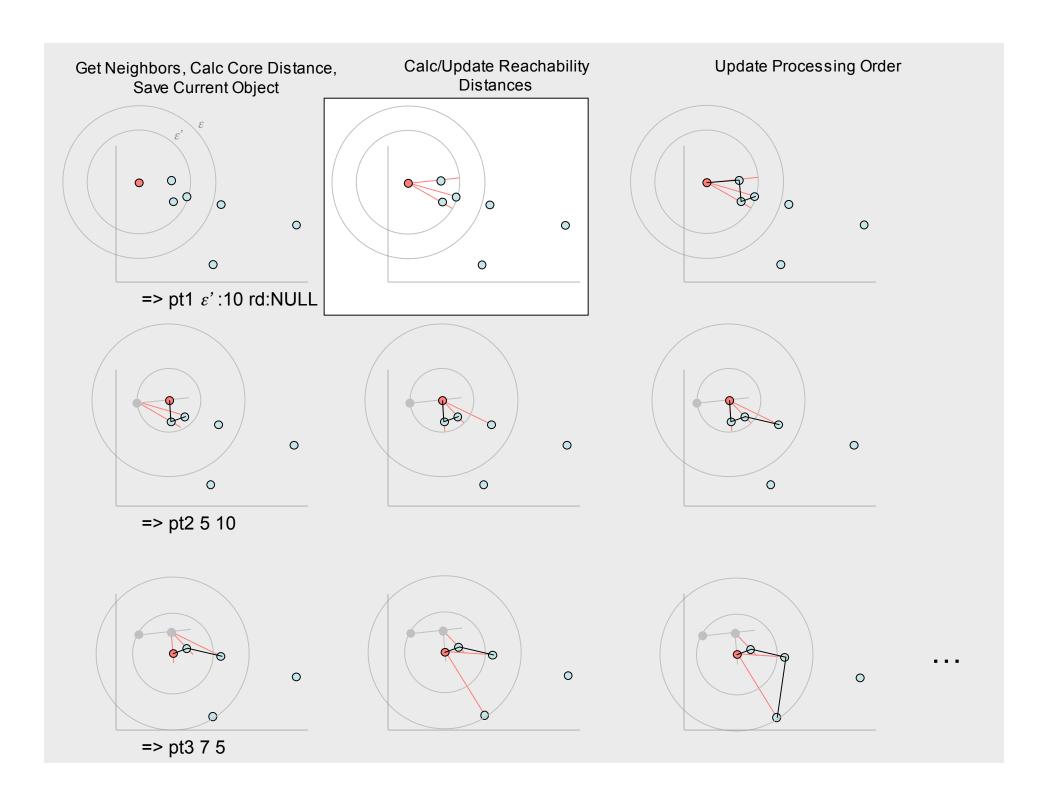
The **core-distance** is the smallest distance  $\varepsilon'$  between p and an object in its  $\varepsilon$ -neighborhood such that p would be a core object.

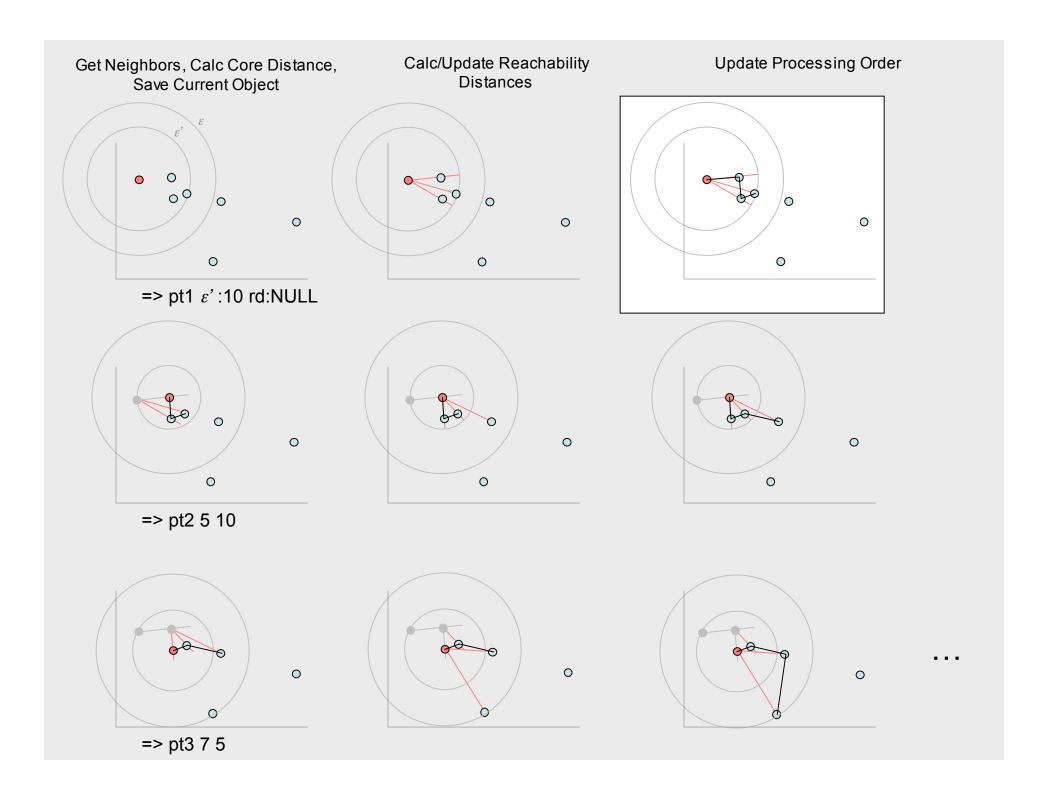
The **reachability-distance** of p is the smallest distance such that p is density-reachable from a core object o.

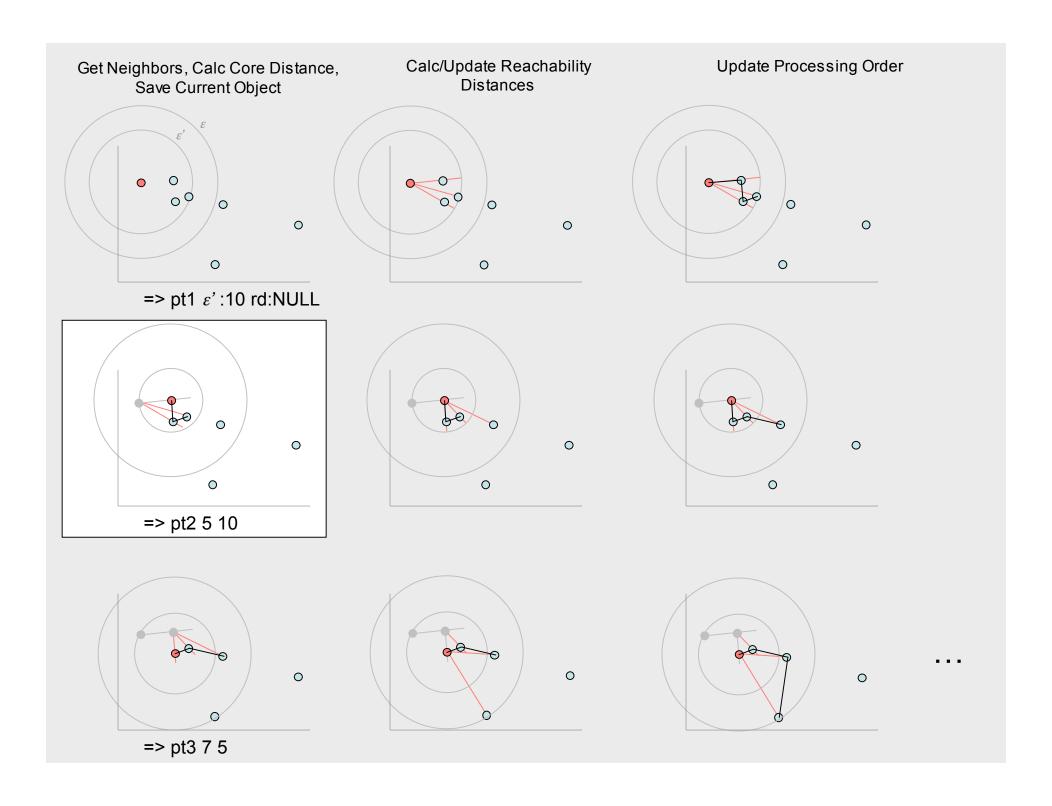


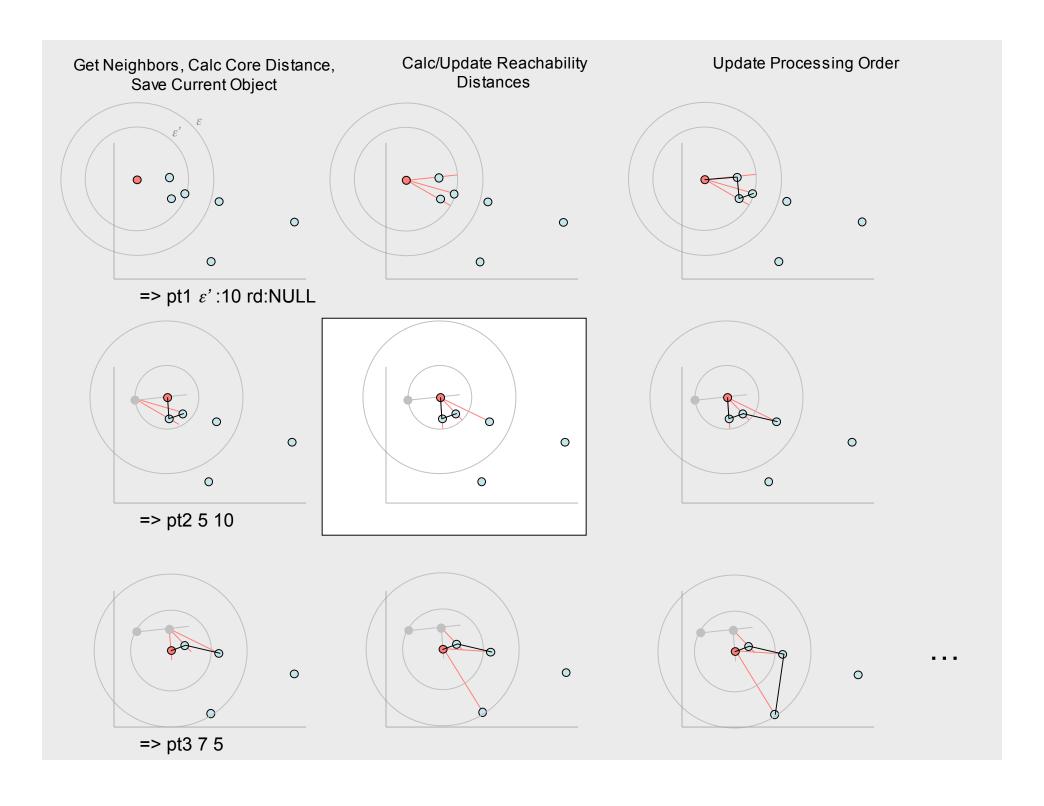
```
1 OPTICS(Objects, e, MinPts, OrderFile):
                                                         1 OrderSeeds::update(neighbors, centerObj):
    for each unprocessed obj in objects:
                                                             d = centerObj.coreDistance
3
      neighbors = Objects.getNeighbors(obj, e)
                                                             for each unprocessed obj in neighbors:
                                                         3
      obj.setCoreDistance(neighbors, e, MinPts)
                                                               newRdist = max(d, dist(obj, centerObj))
      OrderFile.write(obj)
5
                                                         5
                                                               if obj.reachability == NULL:
      if obj.coreDistance != NULL:
6
                                                                 obj.reachability = newRdist
        orderSeeds.update(neighbors, obj)
                                                                 insert(obj, newRdist)
8
        for obj in orderSeeds:
                                                         8
                                                               elif newRdist < obj.reachability:</pre>
          neighbors = Objects.getNeighbors(obj, e)
9
                                                                 obj.reachability = newRdist
          obj.setCoreDistance(neighbors, e, MinPts)
10
                                                         10
                                                                 decrease(obj, newRdist)
          OrderFile.write(obj)
11
12
          if obj.coreDistance != NULL:
            orderSeeds.update(neighbors, obj)
13
```

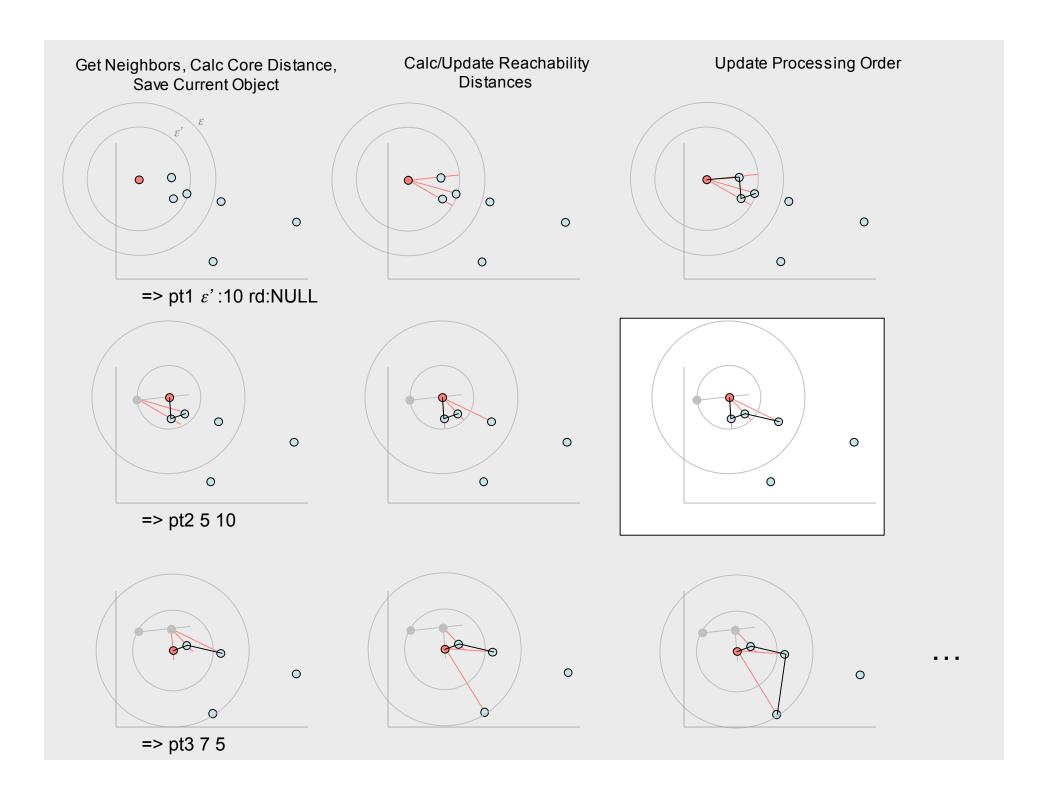


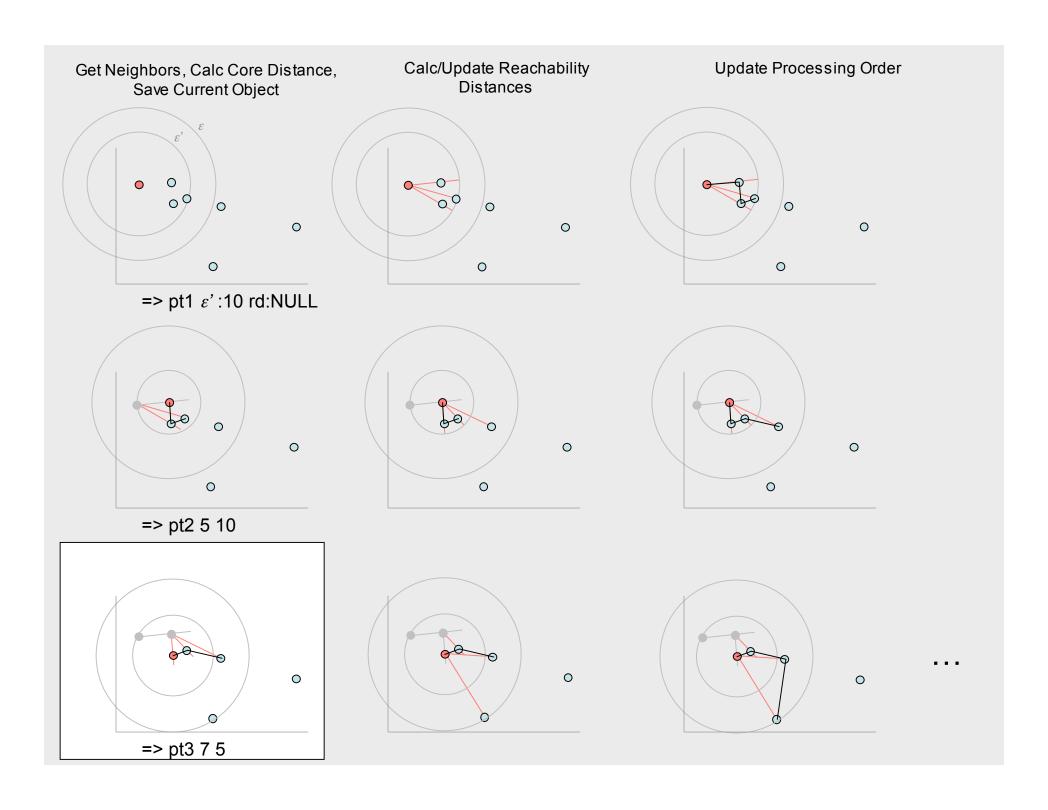


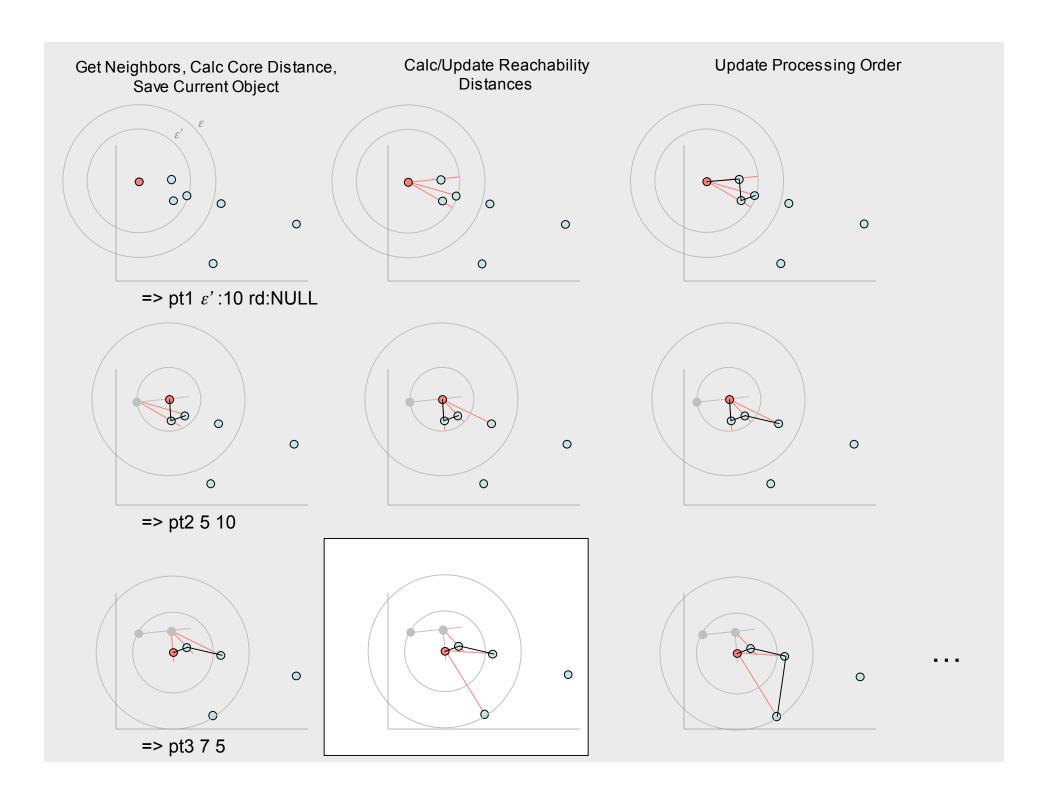


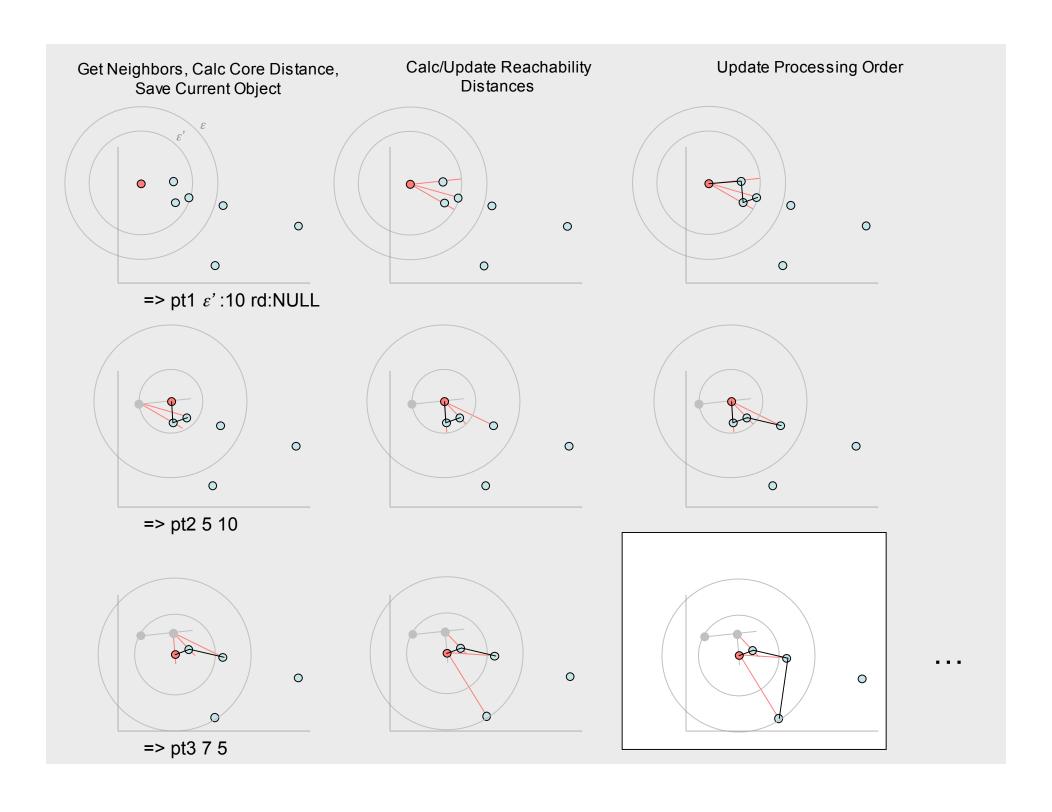


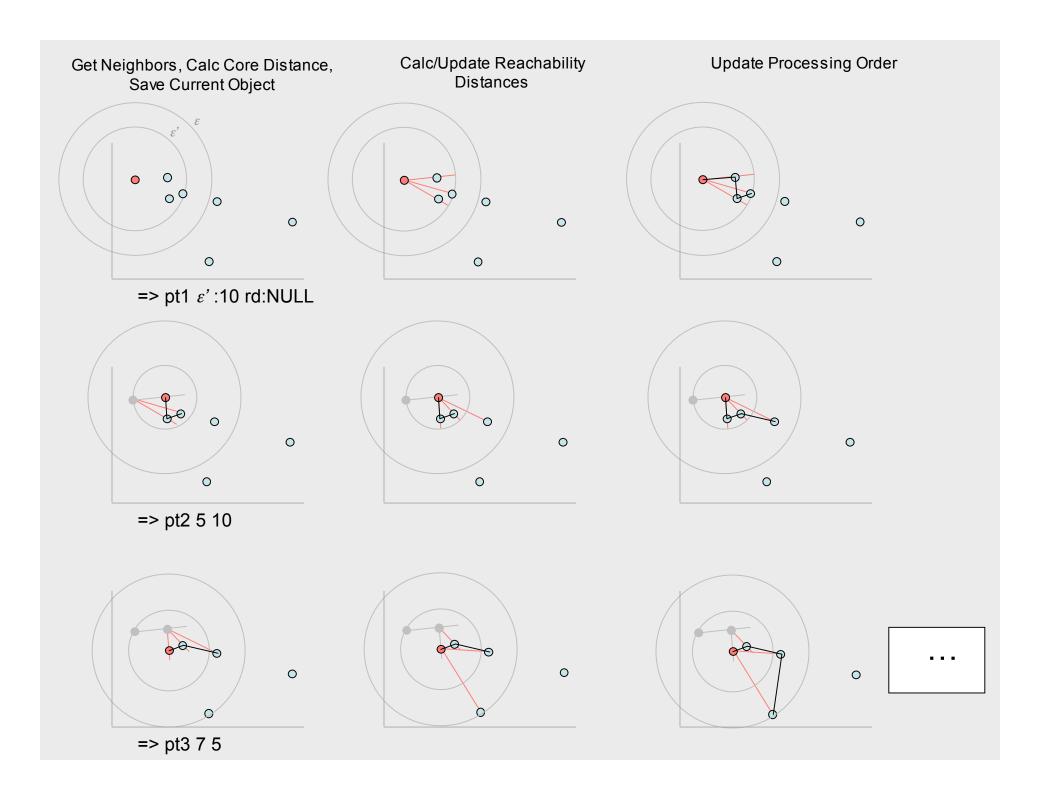






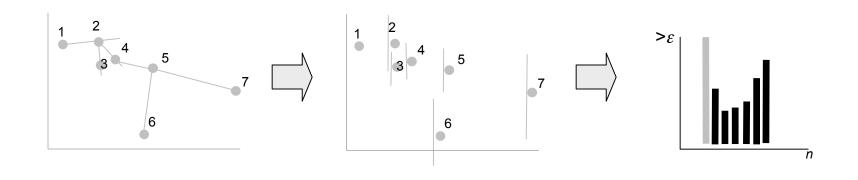


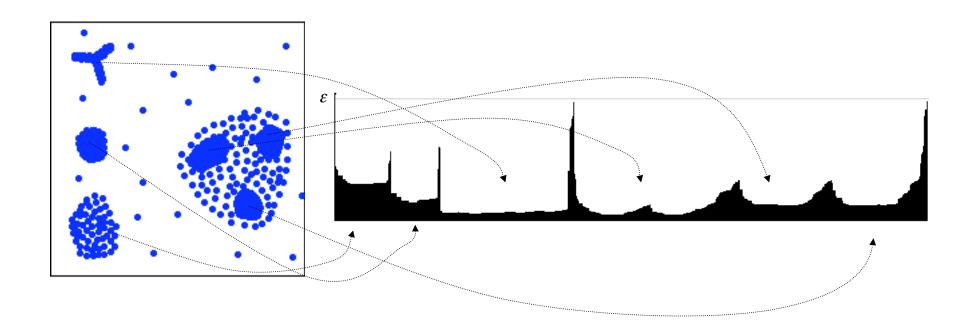




# Reachability Plots

A **reachability plot** is a bar chart that shows each object's reachability distance in the order the object was processed. These plots clearly show the cluster structure of the data.

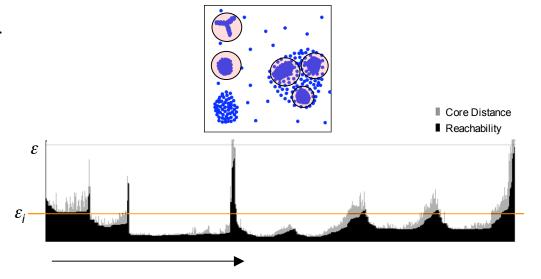




#### **Automatic Cluster Extraction**

#### Retrieving DBSCAN clusters

```
1 ExtractDBSCAN(OrderedPoints, ei, MinPts):
2   clusterId = NOISE
3   for each obj in OrderedPoints:
4    if obj.reachability > ei:
5     if obj.coreDistance <= ei:
6         clusterId = nextId(clusterId)
7         obj.clusterId = clusterId
8     else:
9         obj.clusterId = NOISE
10    else:
11    obj.clusterId = clusterId</pre>
```



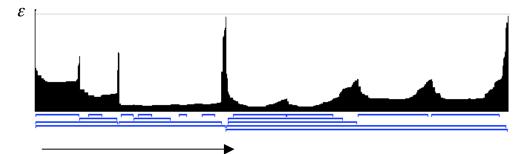
#### Extracting hierarchical clusters

A **steep upward point** is a point that is *t*% lower that its successor. A **steep downward point** is similarly defined.

A **steep upward area** is a region from [s, e] such that s and e are both steep upward points, each successive point is at least as high as its predecessors, and the region does not contain more than *MinPts* successive points that are not steep upward.

#### A cluster:

- Starts with a steep downward area
- Ends with a steep upward area
- Contains at least MinPts
- The reachability values in the cluster are at least *t*% lower than the first point in the cluster.



#### 1 HierachicalCluster(objects):

```
2 for each index:
3   if start of down area D:
4   add D to steep down areas
5   index = end of D
6   elif start of steep up area U:
7   index = end of U
8   for each steep down area D:
9   if D and U form a cluster:
10   add [start(D), end(U)] to set of clusters
```

## <u>References</u>

[DBSCAN] Ester M., Kriegel H.-P., Sander J., Xu X.: "A DensityBased Algorithm for Discovering Clusters in Large Spatial Databases with Noise", Proc. 2nd Int. Conf. on Knowledge Discovery and Data Mining, Portland, OR, AAAI Press, 1996, pp.226-231.

[OPTICS] Ankerst, M., Breunig, M., Kreigel, H.-P., and Sander, J. 1999. OPTICS: Ordering points to identify clustering structure. In Proceedings of the ACM SIGMOD Conference, 49-60, Philadelphia, PA.