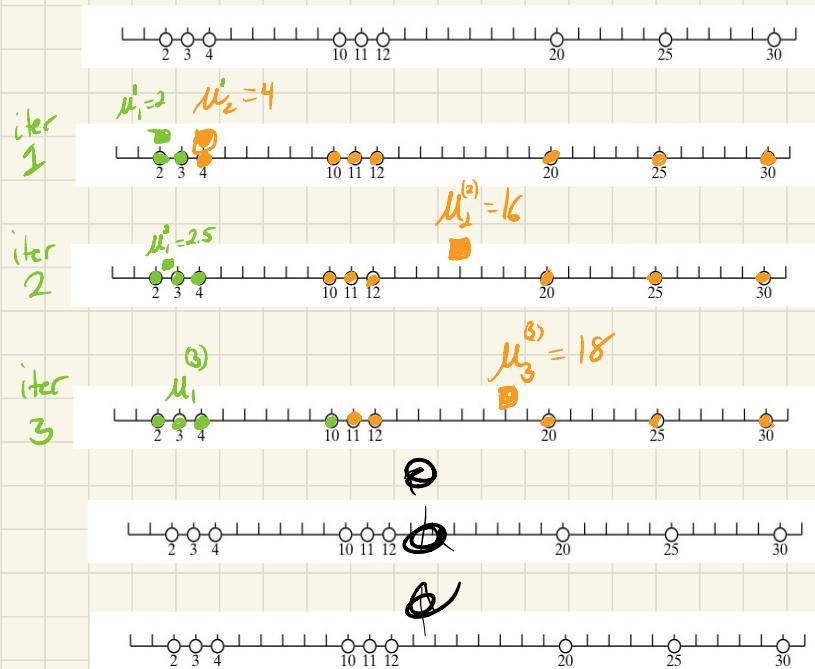


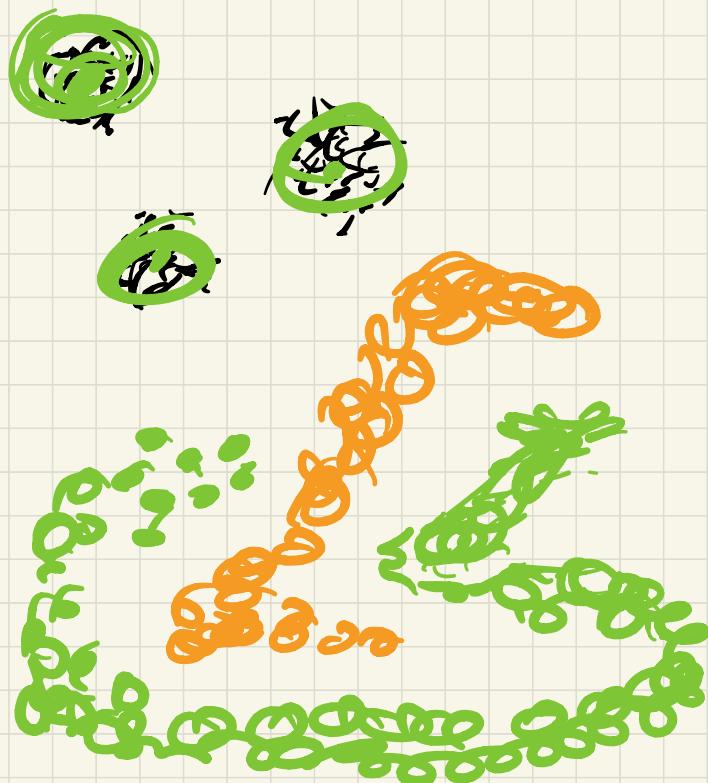
DBSCAN

Clustering So Far

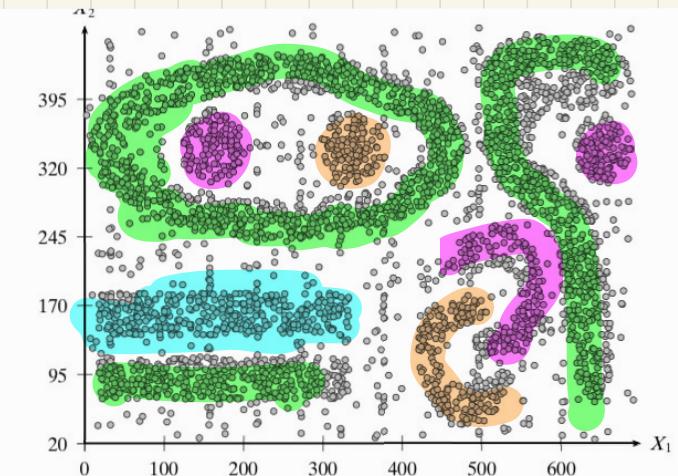
K-means



In 2D



Even More Complex



Soln: DBSCAN

Density
Based
Spatial
Clustering of
Applications w/
Noise

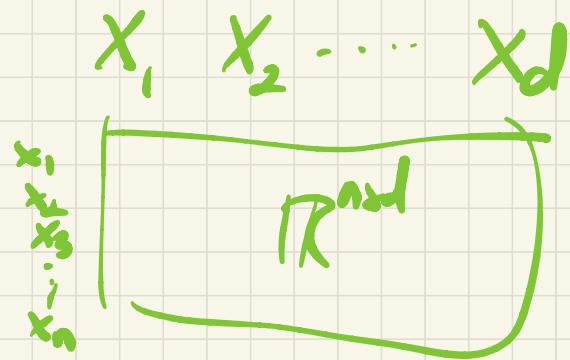
Defs

Given Dataset D

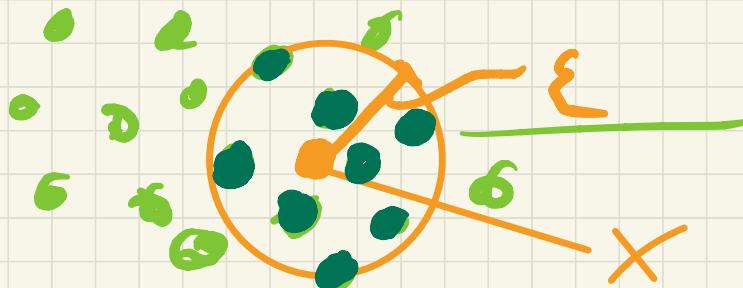
Let $\epsilon > 0 \quad x \in \mathbb{R}^d$

ϵ -neighborhood of x

$$N_\epsilon(x) = \{y \in D : \|y - x\| \leq \epsilon\}$$



all pts of
 D w/in
 ϵ of x



all dark green
are in $N_\epsilon(x)$

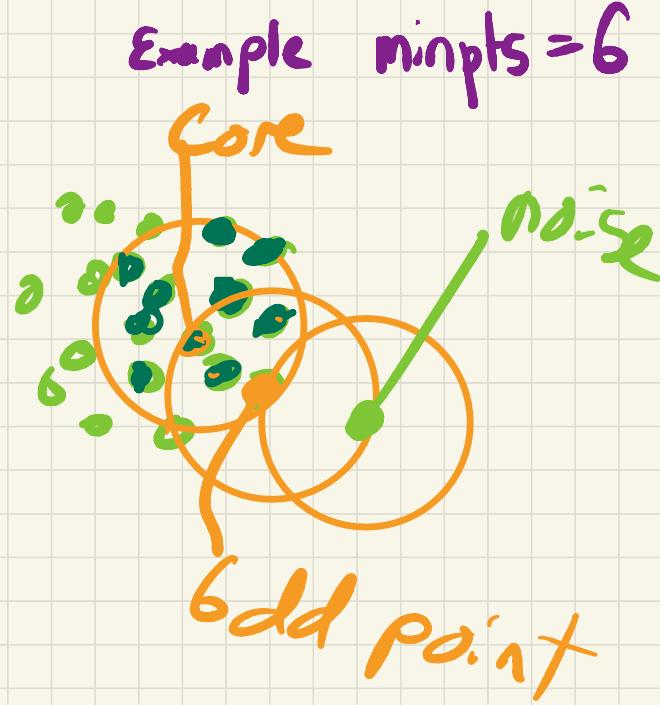
Given param $\text{minpts} \in \mathbb{Z}^+$

For any $x \in D$

x is a **core point** if
there are at least
 minpts in its ϵ -neighborhood

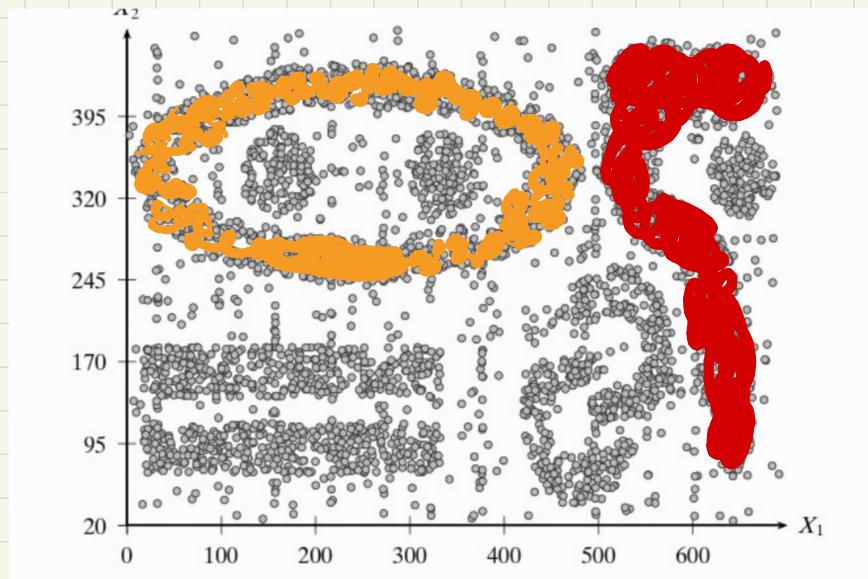
x is a **border point** if
not a core pt and in
 ϵ -neighborhood of core pt

x is **noise** if !border & !core



DBSCAN algo (high level)

1. $\forall x \in D$
 - compute $N_\epsilon(x)$
 - check if core pt
2. $\forall x \in \text{Core Points}$
 - recursively find all "density connected" points



Pseudo Code (Find Corepts)

Core = \emptyset

for $x \in D$

compute $N_\epsilon(x)$

$id(x) = \emptyset$

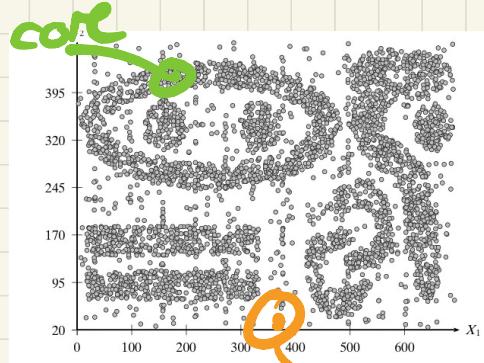
if $|N_\epsilon(x)| \geq \text{minpts}$

Core = Core $\cup \{x\}$

Compute
 $N_\epsilon(x)$

\in sets cluster id
to \emptyset

adds to
core
if enough
points in
neighborhood



Pseudo code (find clusters)

$k=0$

for $x \in \text{Core}$ w/ $\text{id}(x) = \emptyset$

For each unlabeled core pt

$k=k+1$

inc cluster label

$\text{id}(x)=k$

set label

$\text{Density Connected}(x, k)$

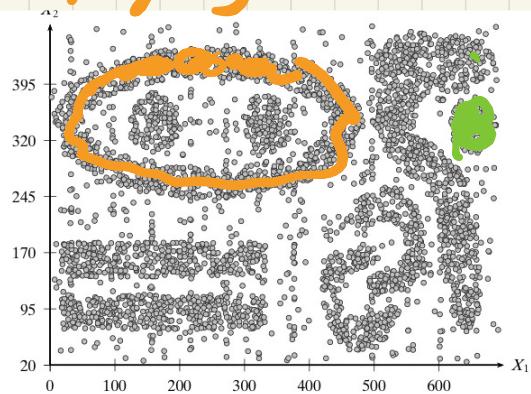
← Search for density connected neighbors

$C = [\text{id}(x_1), \text{id}(x_2), \dots, \text{id}(x_n)]$

$\text{Noise} = \{x \in D \mid \text{id}(x) = \emptyset\}$

$\text{Border} = D \setminus (\text{Core} \cup \text{Noise})$

return $C, \text{Core}, \text{Border}, \text{Noise}$



Pseudocode (Density Connected)

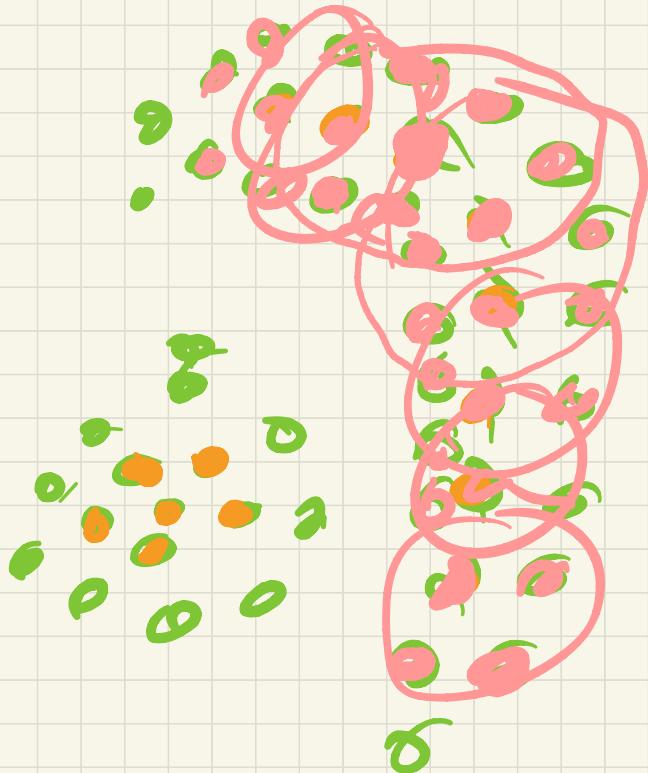
DensityConnected(x, k):

for $y \in N_\epsilon(x)$

:id(y) = k

if $y \in \text{Core}$:

 DensityConnected(y, k)



Full Pseudocode

$$|D|=n$$

Range Query
algs

Algorithm 15.1: Density-based Clustering Algorithm

DBSCAN ($D, \epsilon, minpts$):

```
1 Core  $\leftarrow \emptyset$ 
2 foreach  $x_i \in D$  do // Find the core points
3   Compute  $N_\epsilon(x_i)$ 
4    $id(x_i) \leftarrow \emptyset$  // cluster id for  $x_i$ 
5   if  $N_\epsilon(x_i) \geq minpts$  then  $Core \leftarrow Core \cup \{x_i\}$ 
6  $k \leftarrow 0$  // cluster id
7 foreach  $x_i \in Core$ , such that  $id(x_i) = \emptyset$  do
8    $k \leftarrow k + 1$ 
9    $id(x_i) \leftarrow k$  // assign  $x_i$  to cluster id  $k$ 
10  DENSITYCONNECTED ( $x_i, k$ )
11  $C \leftarrow \{C_i\}_{i=1}^k$ , where  $C_i \leftarrow \{x \in D \mid id(x) = i\}$ 
12  $Noise \leftarrow \{x \in D \mid id(x) = \emptyset\}$ 
13  $Border \leftarrow D \setminus (Core \cup Noise)$ 
14 return  $C, Core, Border, Noise$ 
```

DENSITYCONNECTED (x, k):

```
15 foreach  $y \in N_\epsilon(x)$  do
16    $id(y) \leftarrow k$  // assign  $y$  to cluster id  $k$ 
17   if  $y \in Core$  then DENSITYCONNECTED ( $y, k$ )
```

$O(n^2)$ / if we
as not
fancy

$O(n)$ ↗
basically
Depth
first
Search