



Means & Medians to Machine Learning: Spatial Statistics Basics and Innovations

Flora Vale

esriurl.com/spatialstats

2018 Esri Federal GIS Conference | Washington, DC

**What are
Spatial
Statistics?**

Spatial Statistics are a set of exploratory techniques for describing and modeling spatial distributions, patterns, processes, and relationships.

coincidence

area

connectivity

proximity

orientation

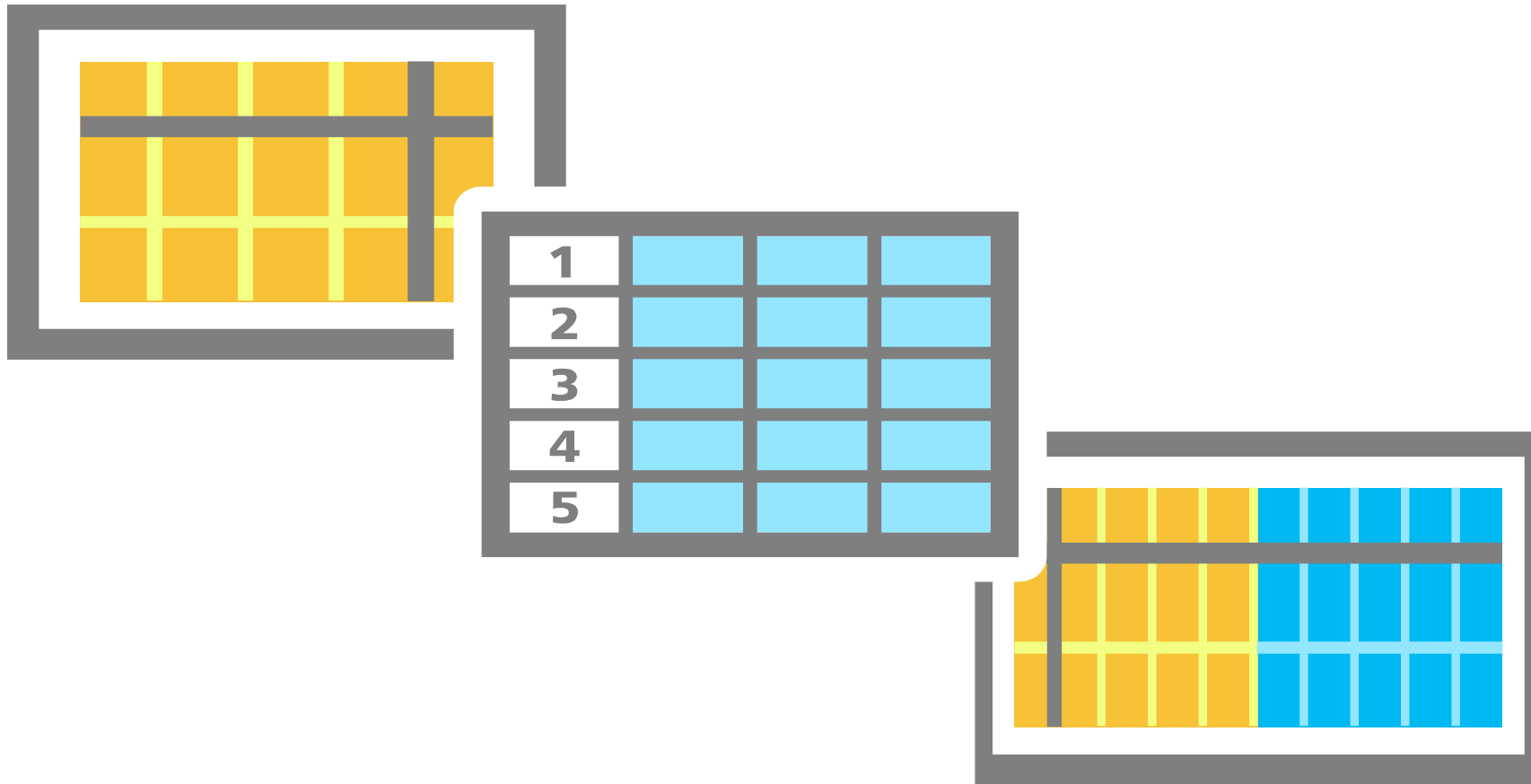
length

direction



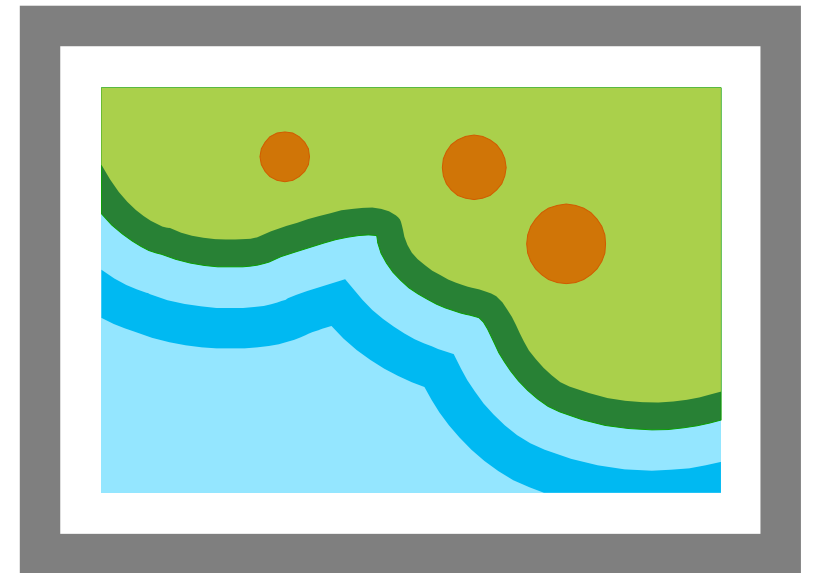
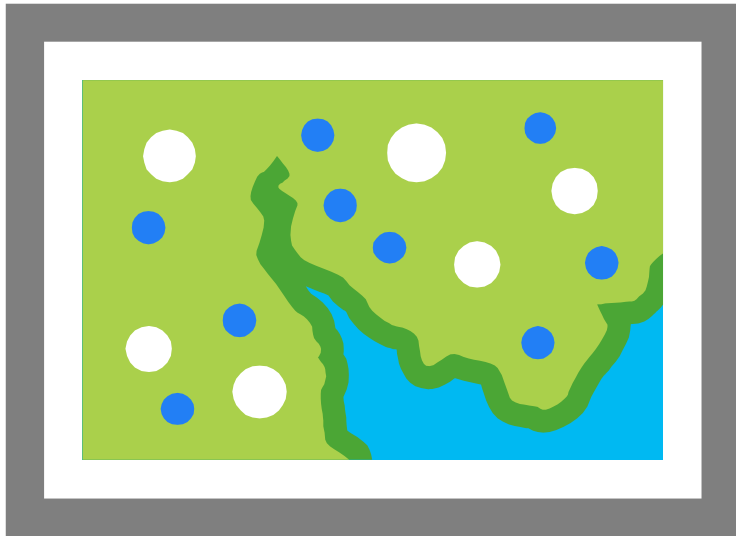
Spreadsheets

Data or Information?



Maps

Data or Information?



When you look at a spreadsheet...

1			
2			
3			
4			
5			

You ask for more

1			
2			
3			
4			
5			

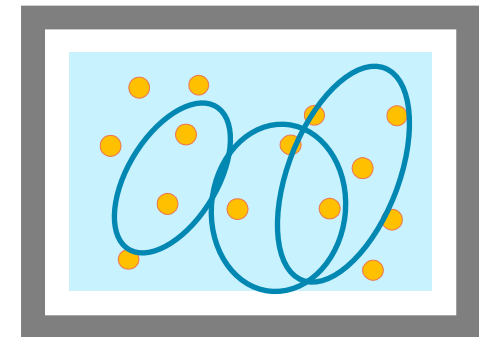
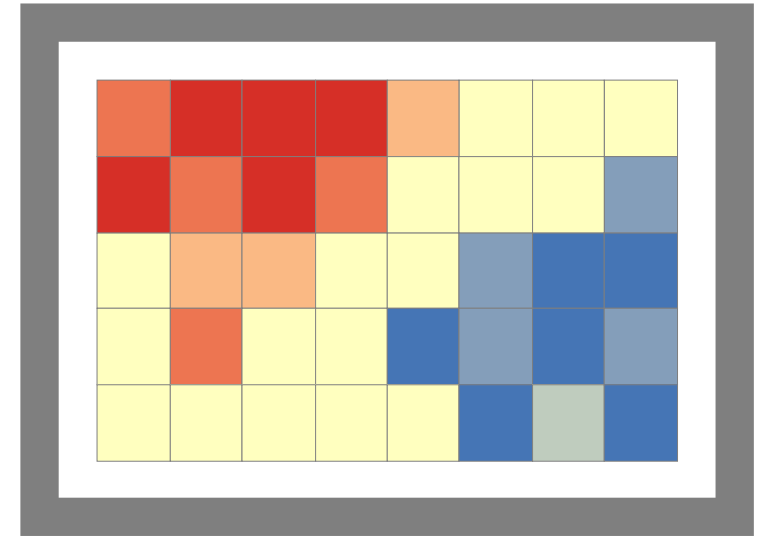
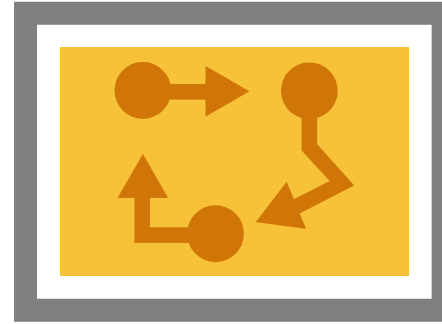


- Mean
- Standard Deviations
- Min and Max
- ...

Same goes for maps!



We can do more



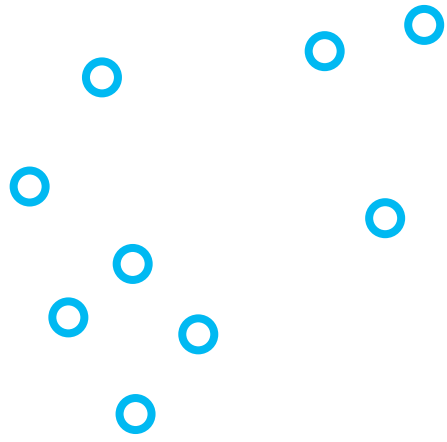
Means and Medians

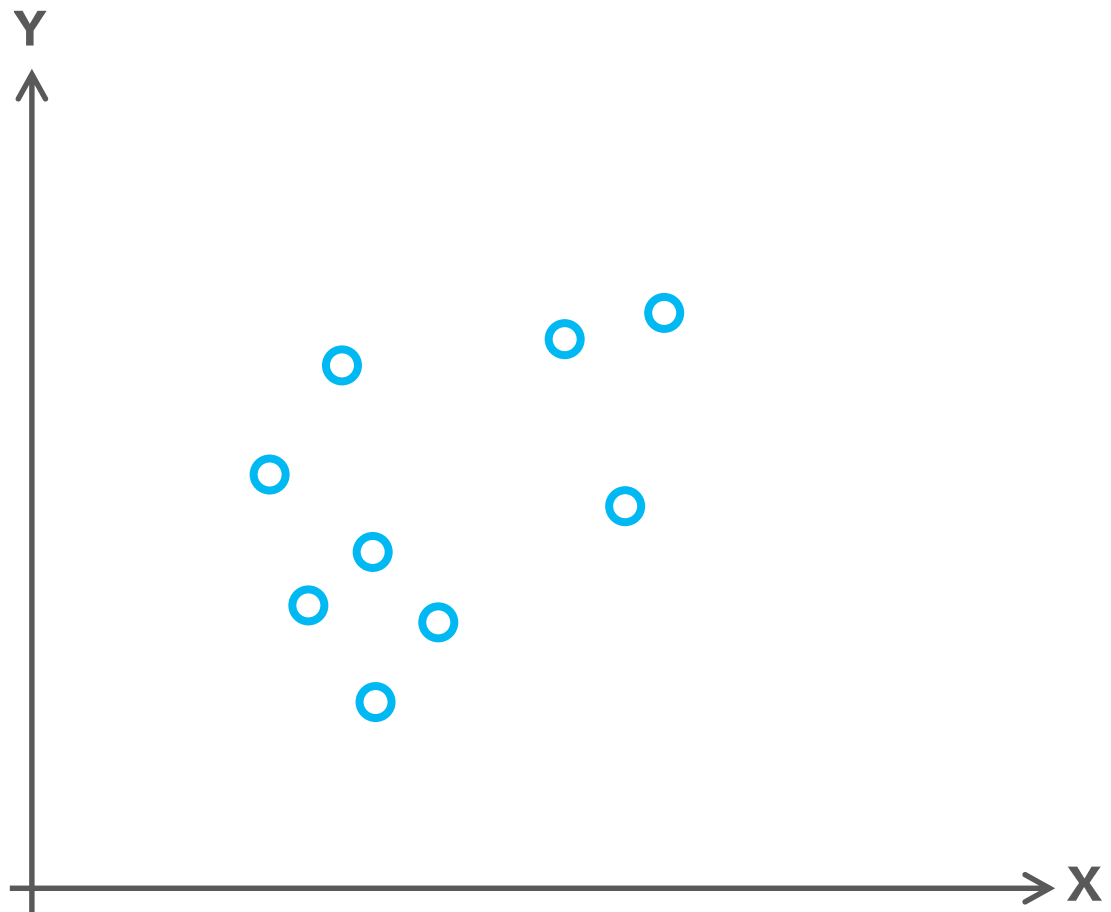
summarizing spatial distributions

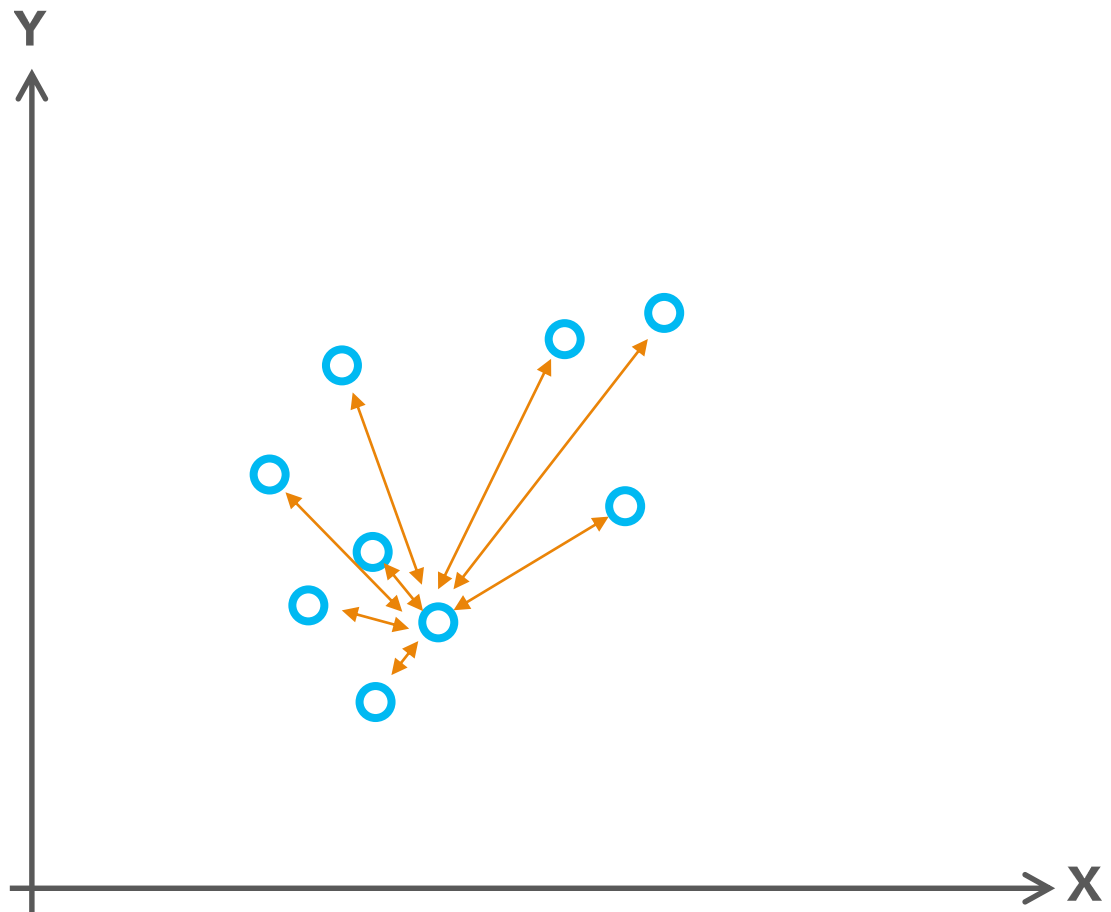


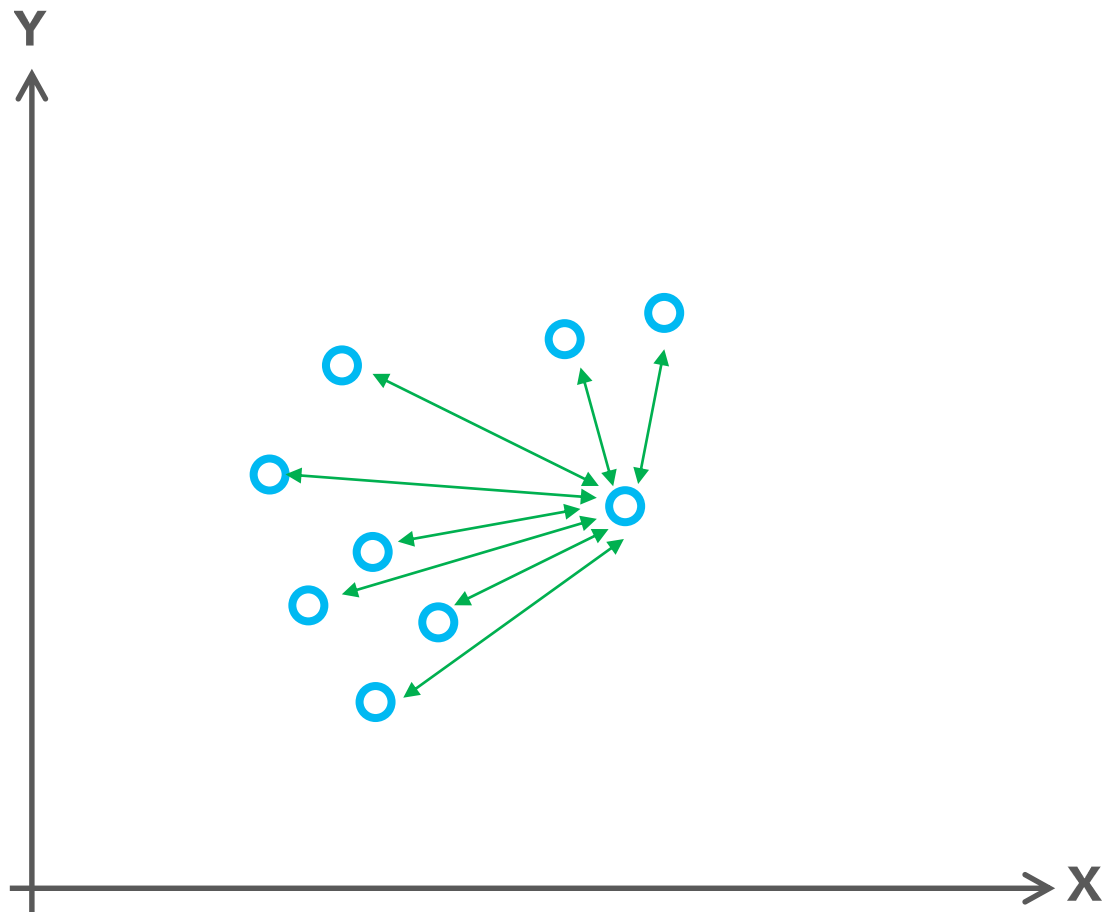
Central Feature

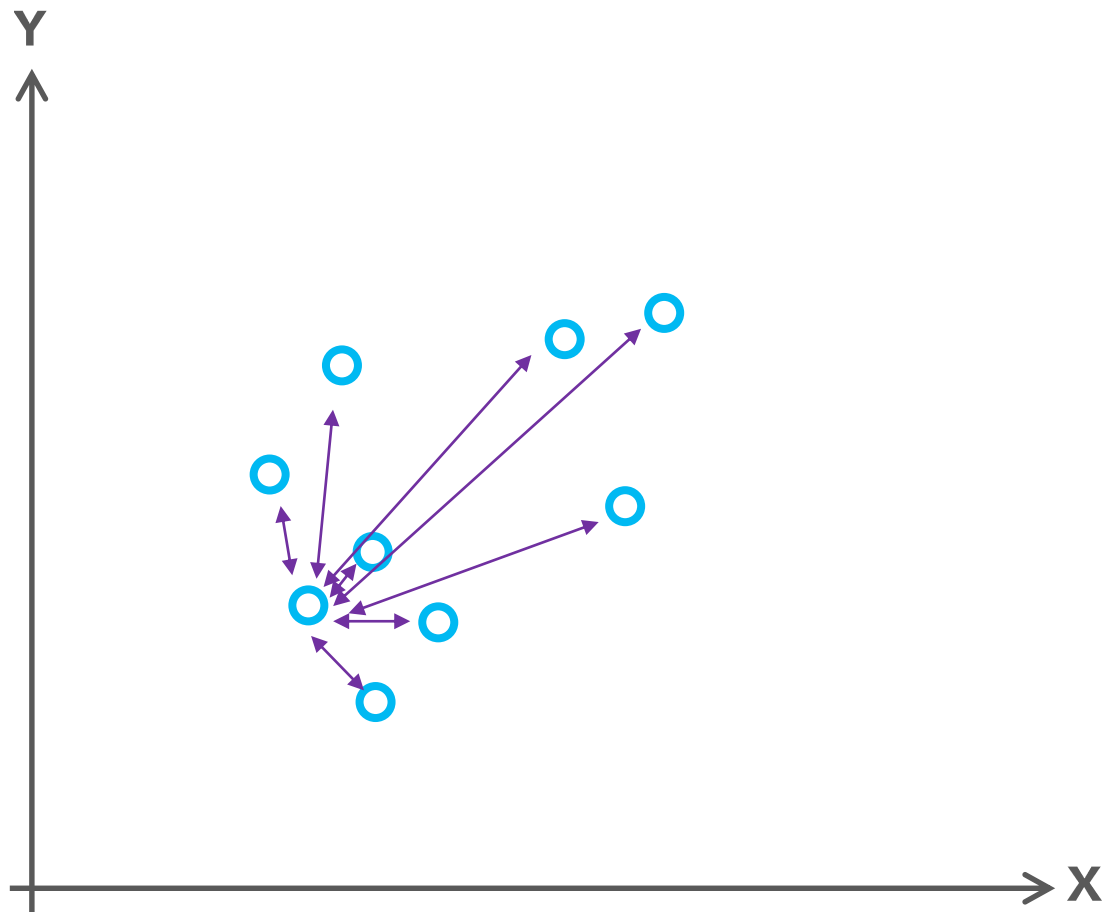
identifies the most centrally located feature
in a point, line, or polygon feature class

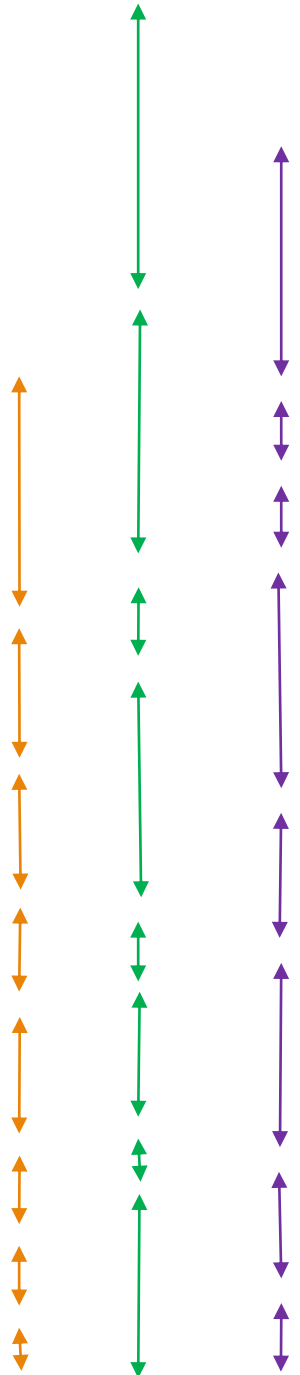
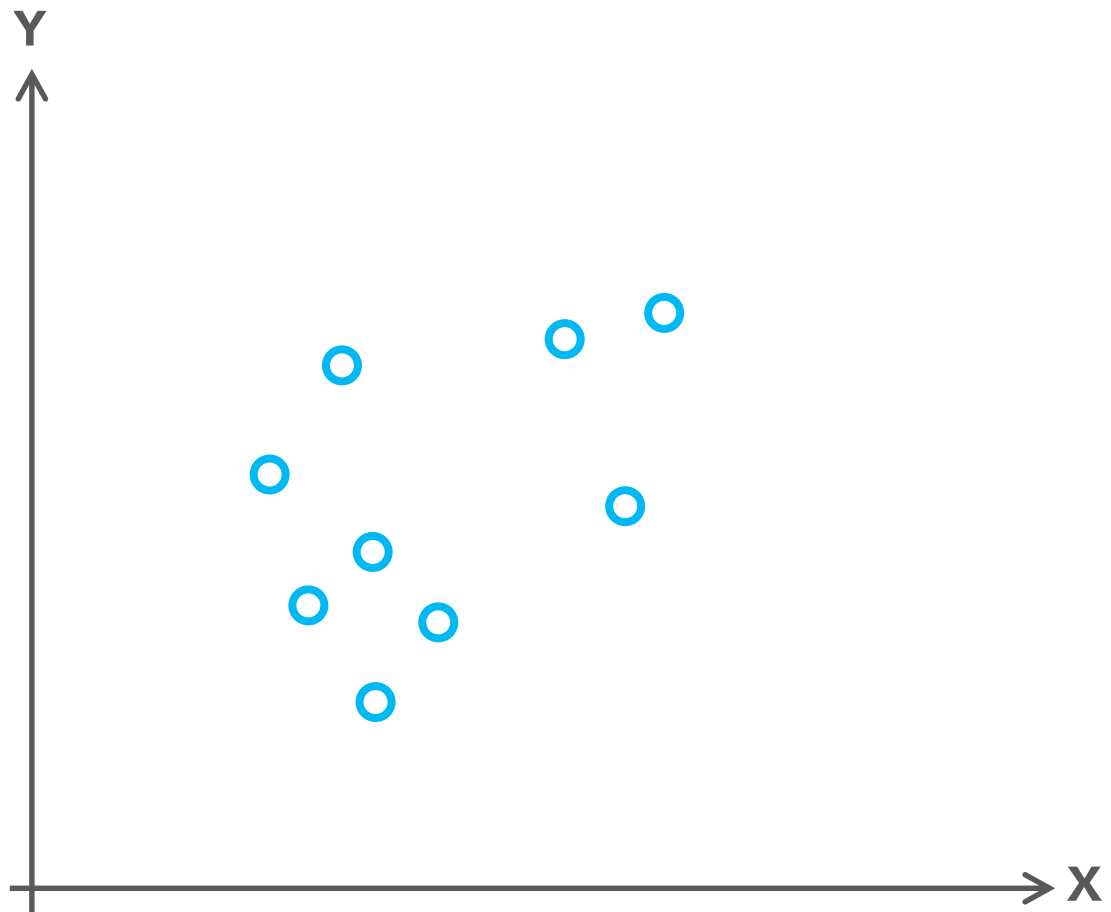


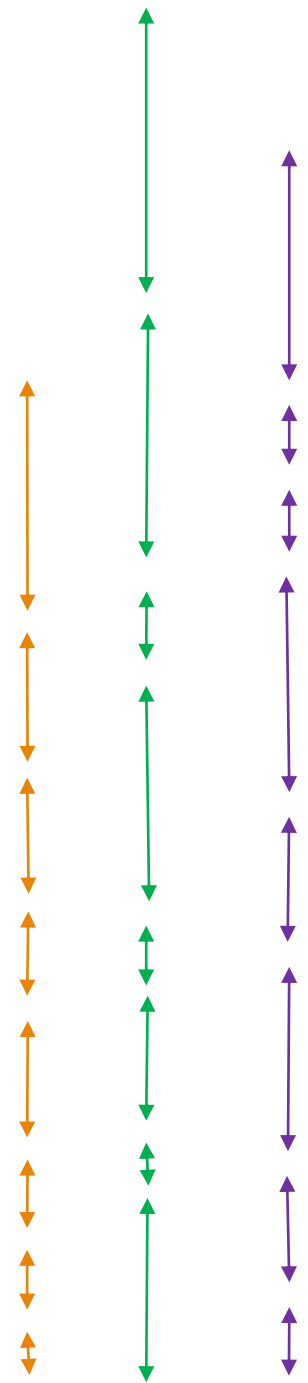
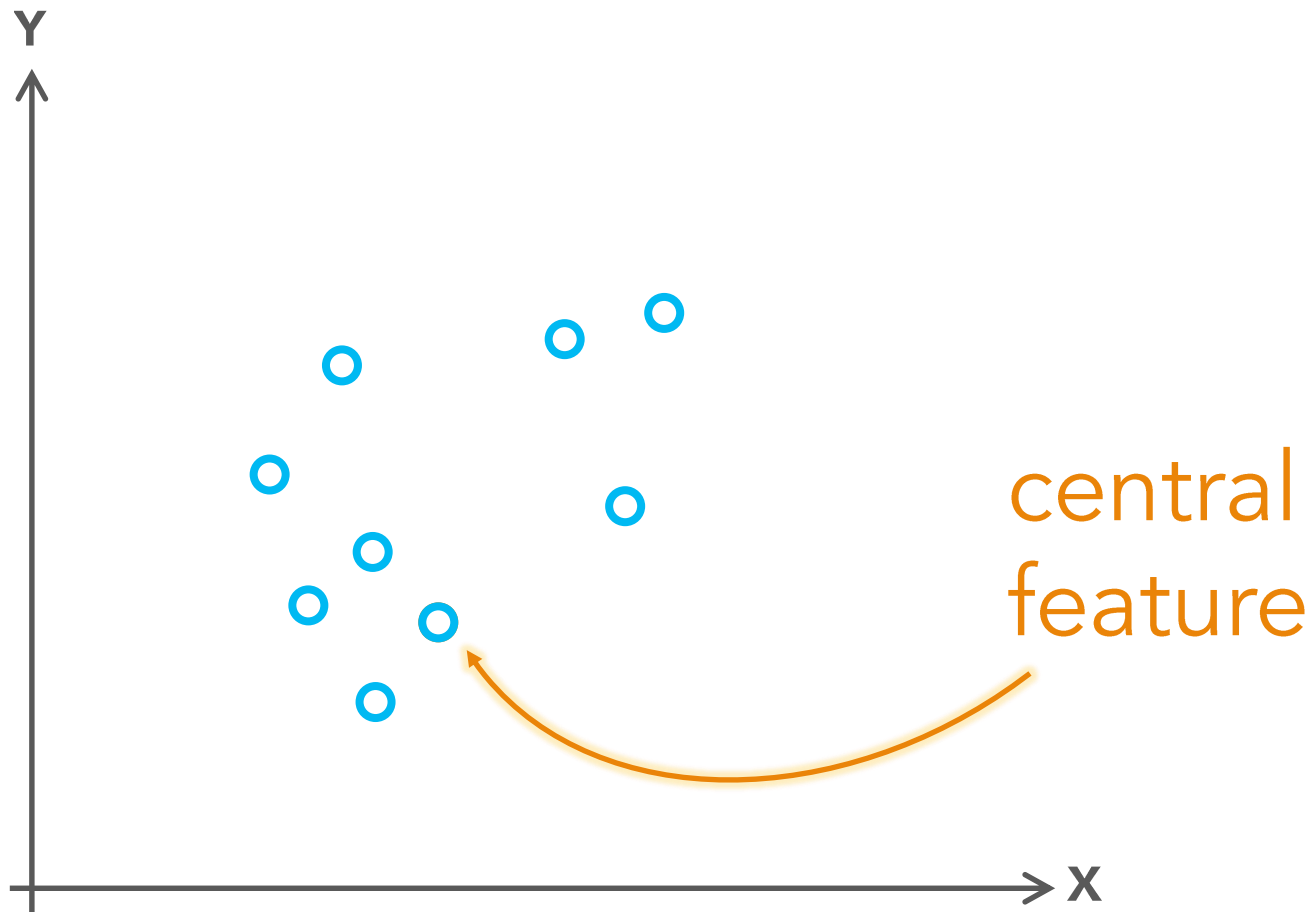






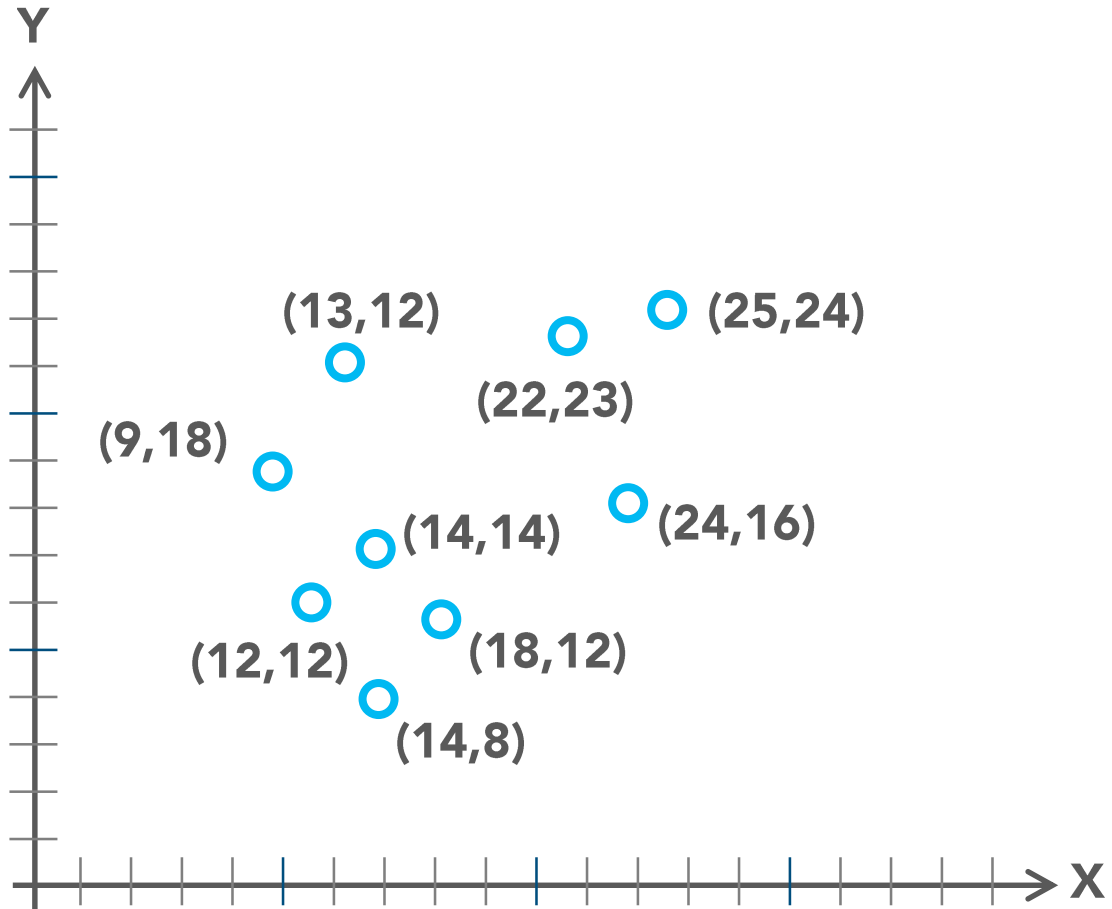


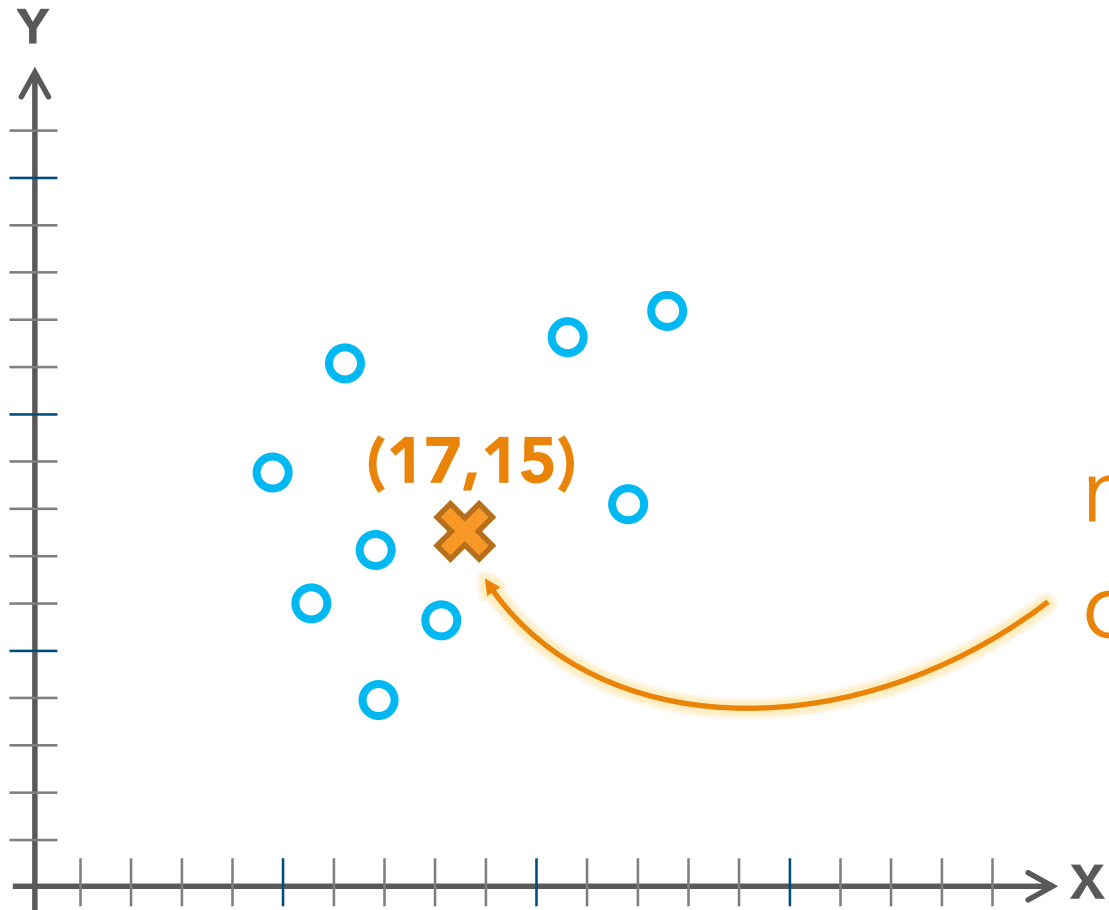




Mean Center

identifies the geographic center (or the center of concentration) for a set of features





(14,14)

(13,12)

(25,24)

(24,16)

(22,23)

(18,12)

(12,12)

(14,8)

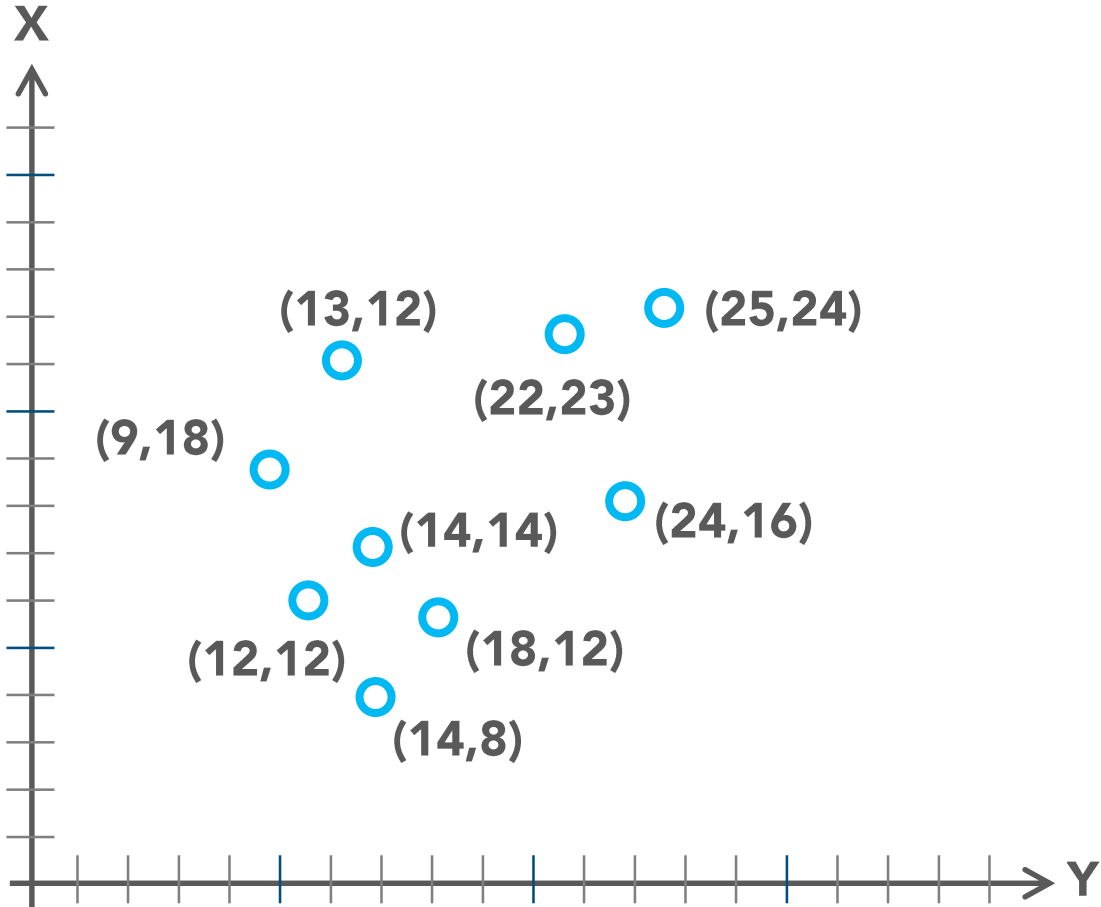
(9,18)

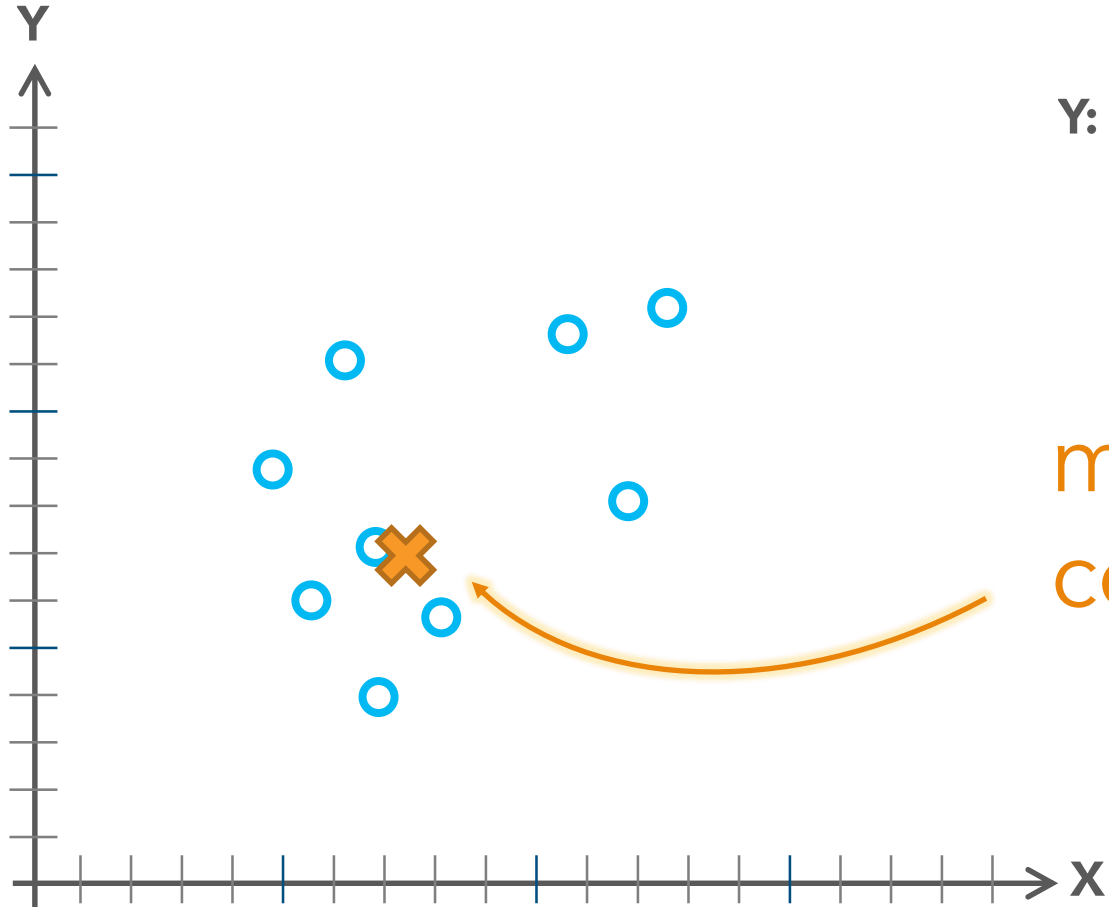
mean =

(17,15)

Median Center

identifies the location that minimizes overall
Euclidean distance to the features in a
dataset





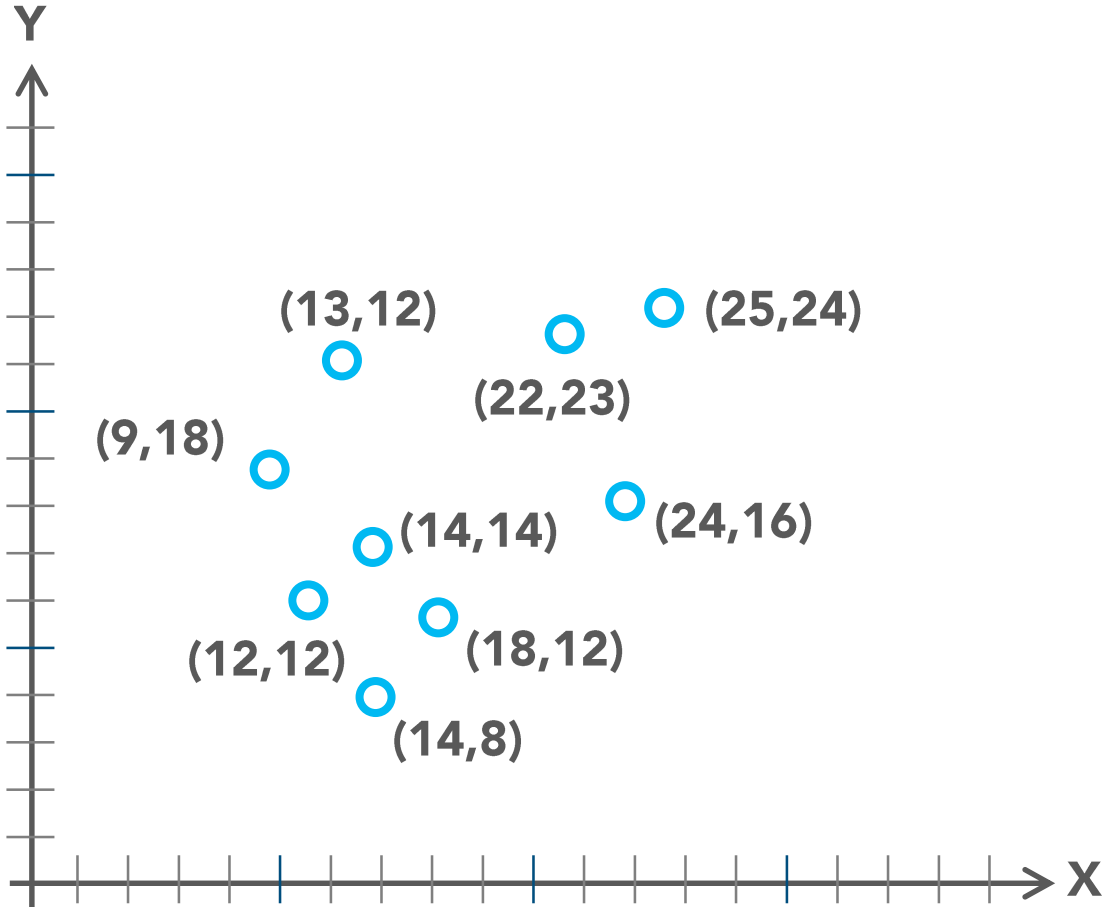
X: ~~25~~ . ~~24~~ . ~~22~~ . ~~18~~ . 14 . 14 . ~~13~~ . ~~12~~ . ~~9~~

Y: ~~24~~ . ~~23~~ . ~~18~~ . ~~16~~ . 14 . ~~12~~ . ~~12~~ . ~~12~~ . ~~8~~

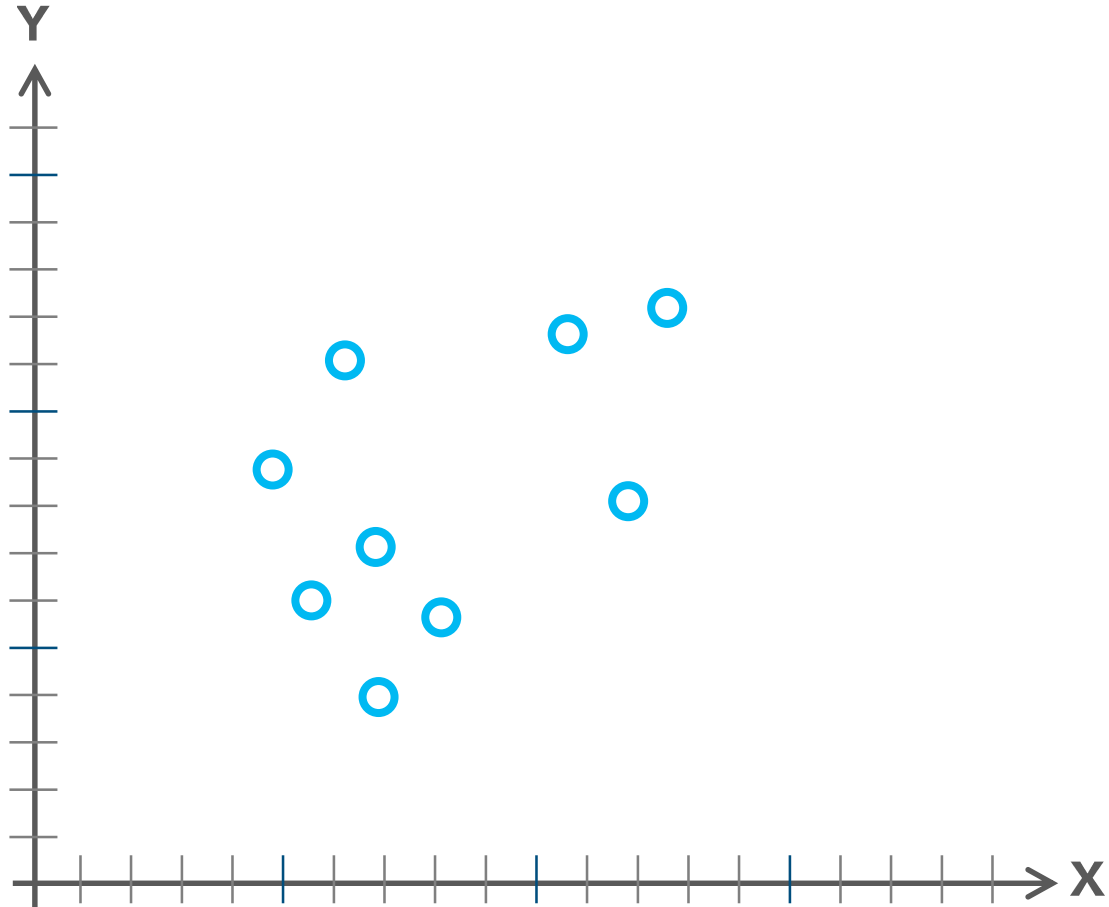
median = **(14,14)**

median
center

Mean vs Median?



(176,138)



(14,14)

(13,12)

(25,24)

(24,16)

(22,23)

(18,12)

(12,12)

(14,8)

(9,18)

(74,38)

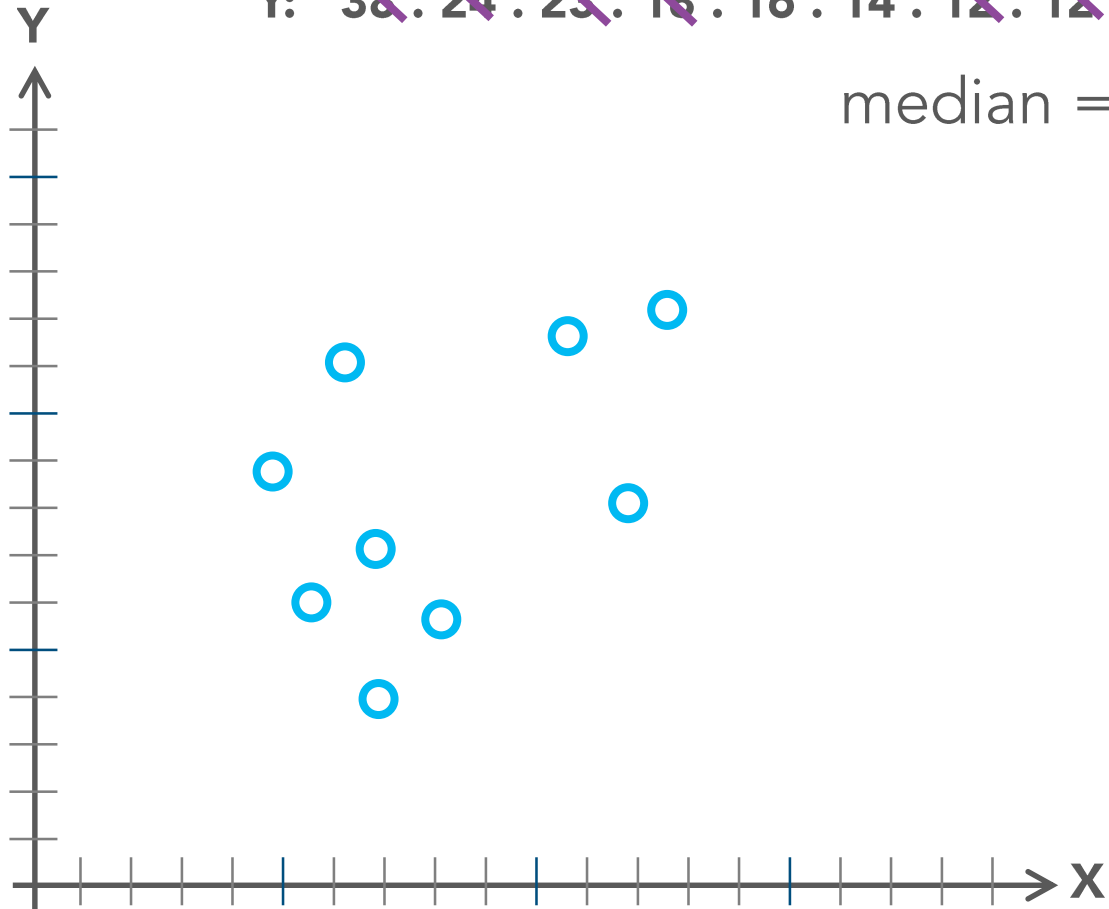
mean =

(33,28)

X: ~~74~~ . ~~25~~ . ~~24~~ . ~~22~~ . 18 . 14 . ~~14~~ . ~~13~~ . ~~12~~ . ~~9~~

Y: ~~38~~ . ~~24~~ . ~~23~~ . ~~18~~ . 16 . 14 . ~~12~~ . ~~12~~ . ~~12~~ . ~~8~~

median = **(16,15)**



(14,14)

(13,12)

(25,24)

(24,16)

(22,23)

(18,12)

(12,12)

(14,8)

(9,18)

(74,38)

mean =

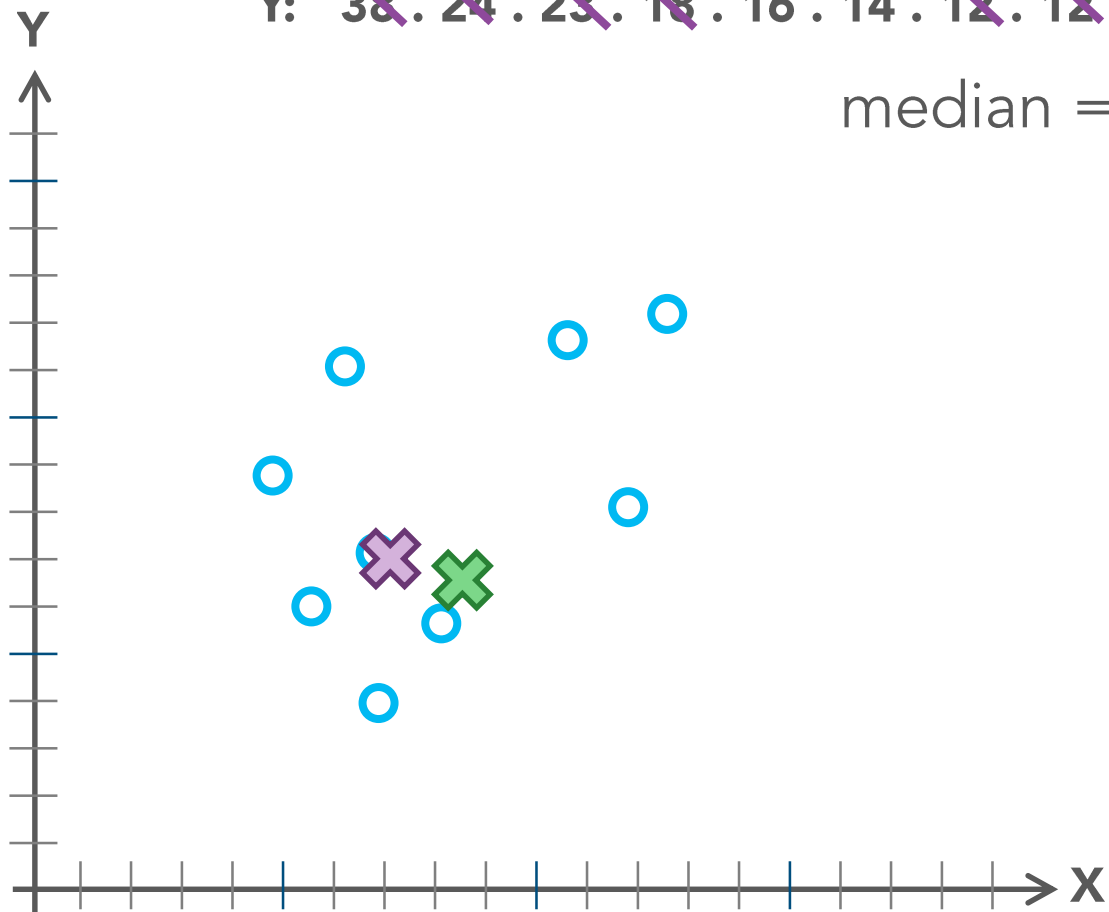
(33,28)



X: ~~74~~ . ~~25~~ . ~~24~~ . ~~22~~ . 18 . 14 . ~~14~~ . ~~13~~ . ~~12~~ . ~~9~~

Y: ~~38~~ . ~~24~~ . ~~23~~ . ~~18~~ . 16 . 14 . ~~12~~ . ~~12~~ . ~~12~~ . ~~8~~

median = **(16,15)**



(14,14)

(13,12)

(25,24)

(24,16)

(22,23)

(18,12)

(12,12)

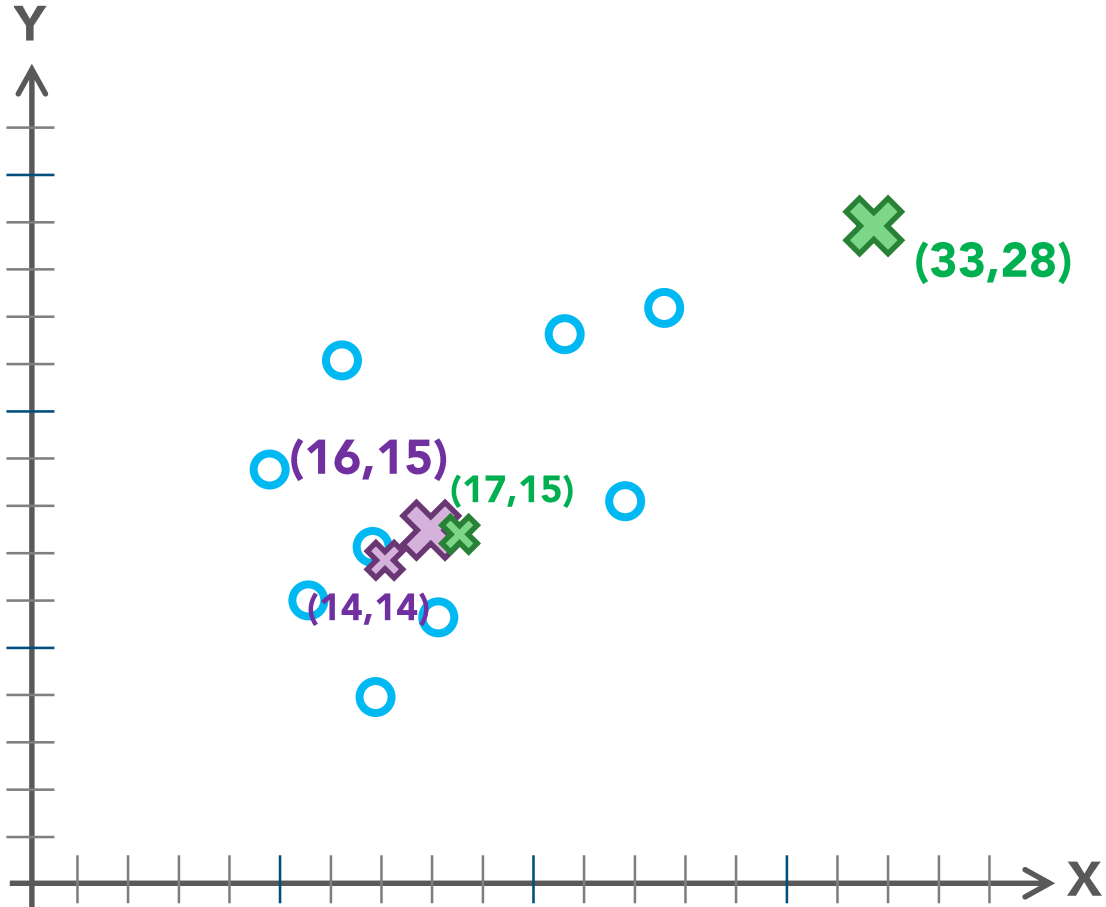
(14,8)

(9,18)

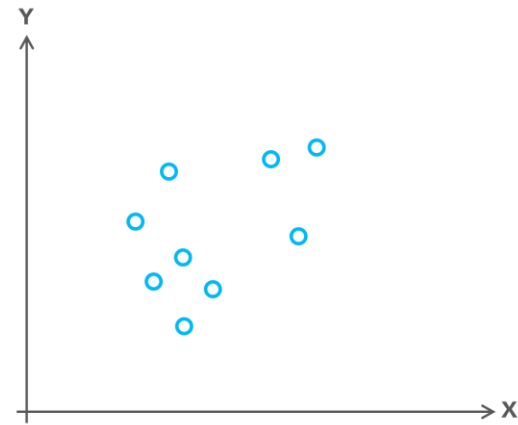
(74,38)

mean =

(33,28)

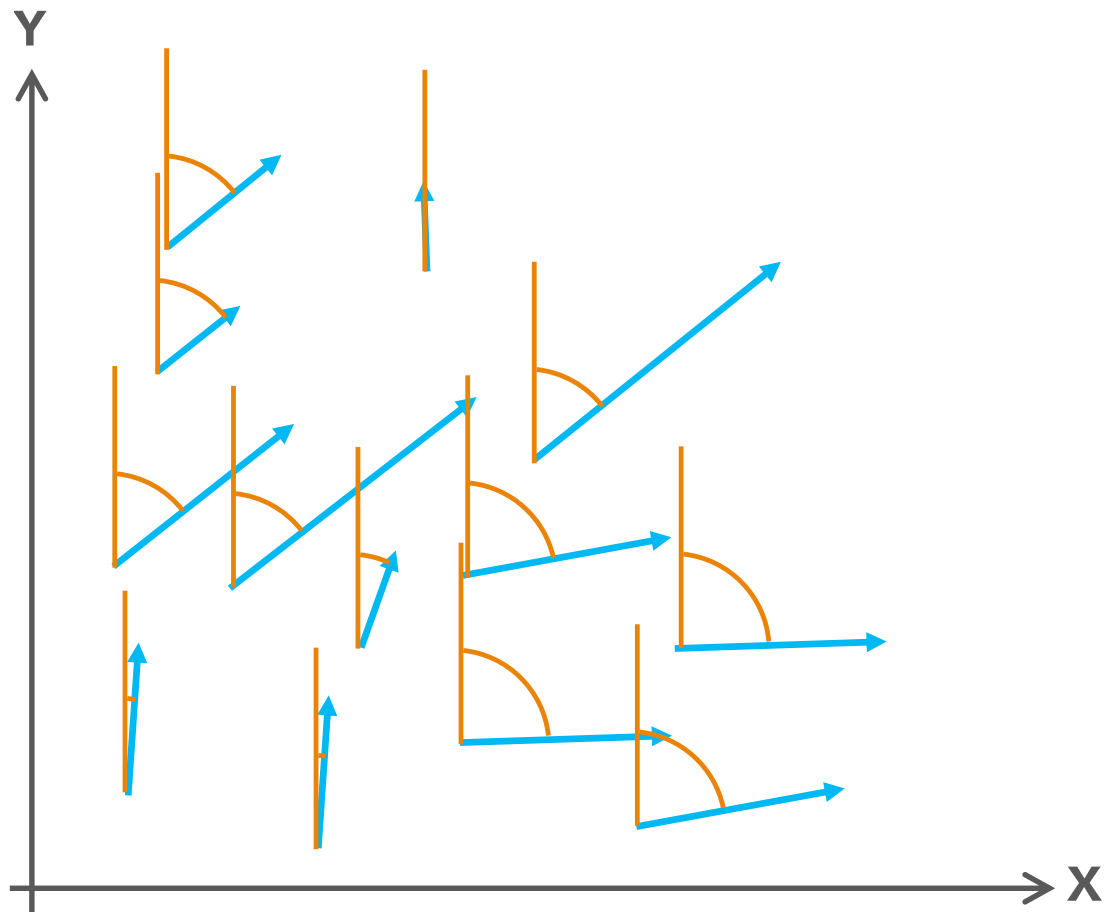


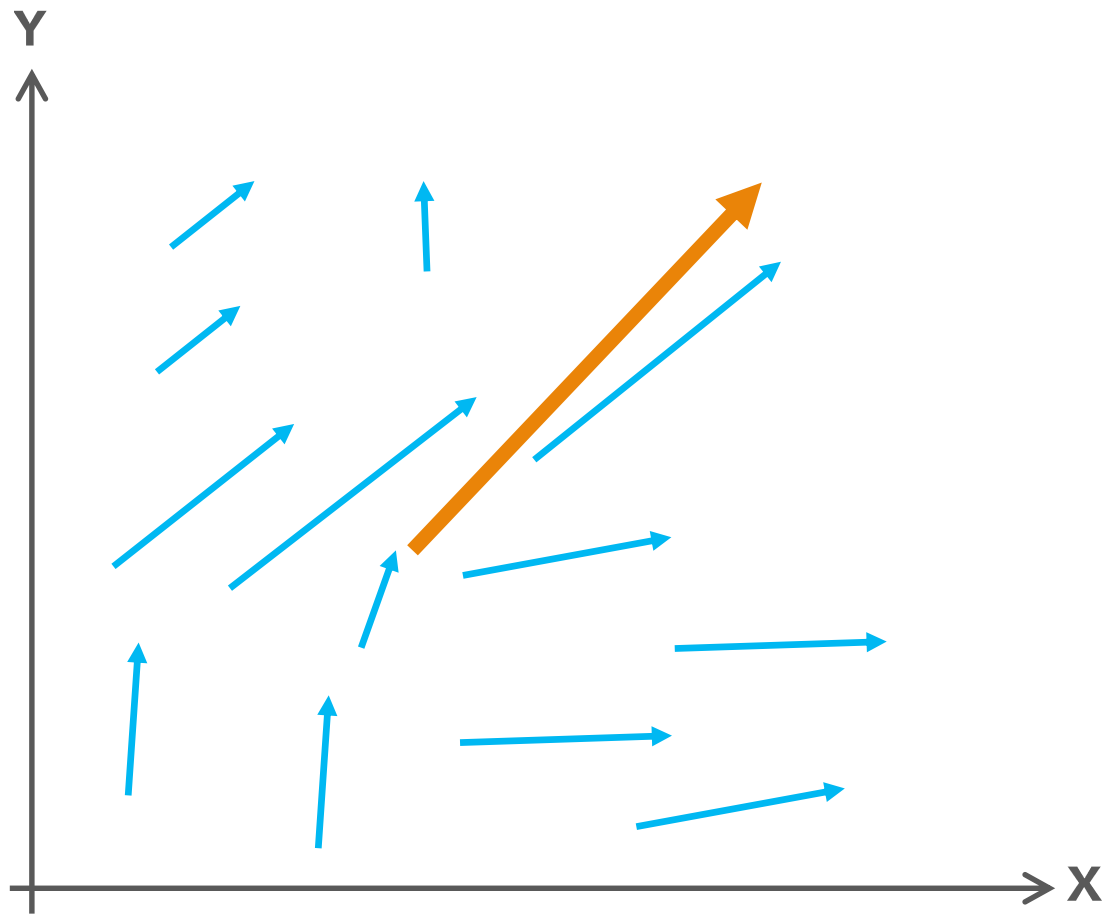
Demo



Linear Directional Mean

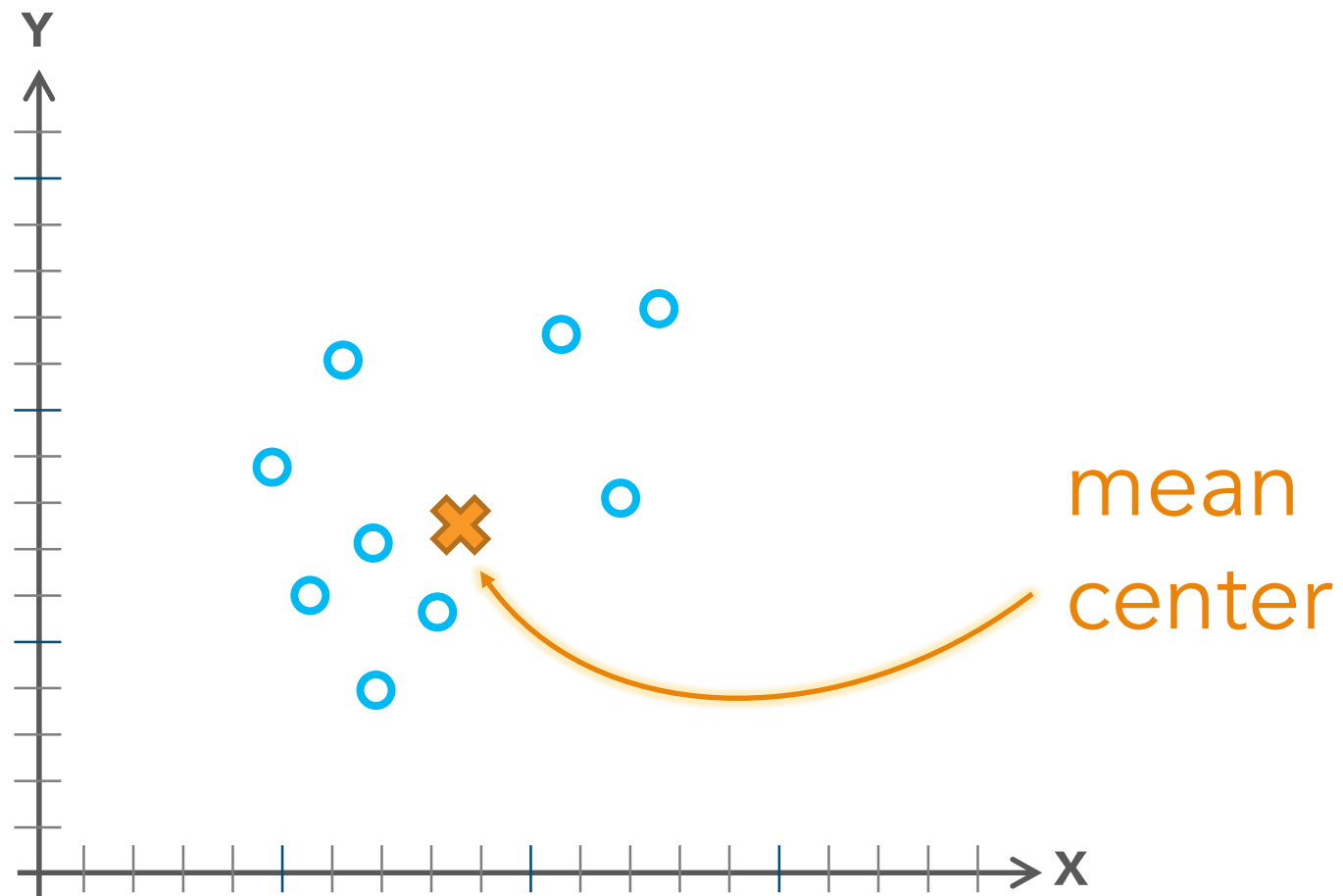
identifies the mean direction, length, and geographic center for a set of lines

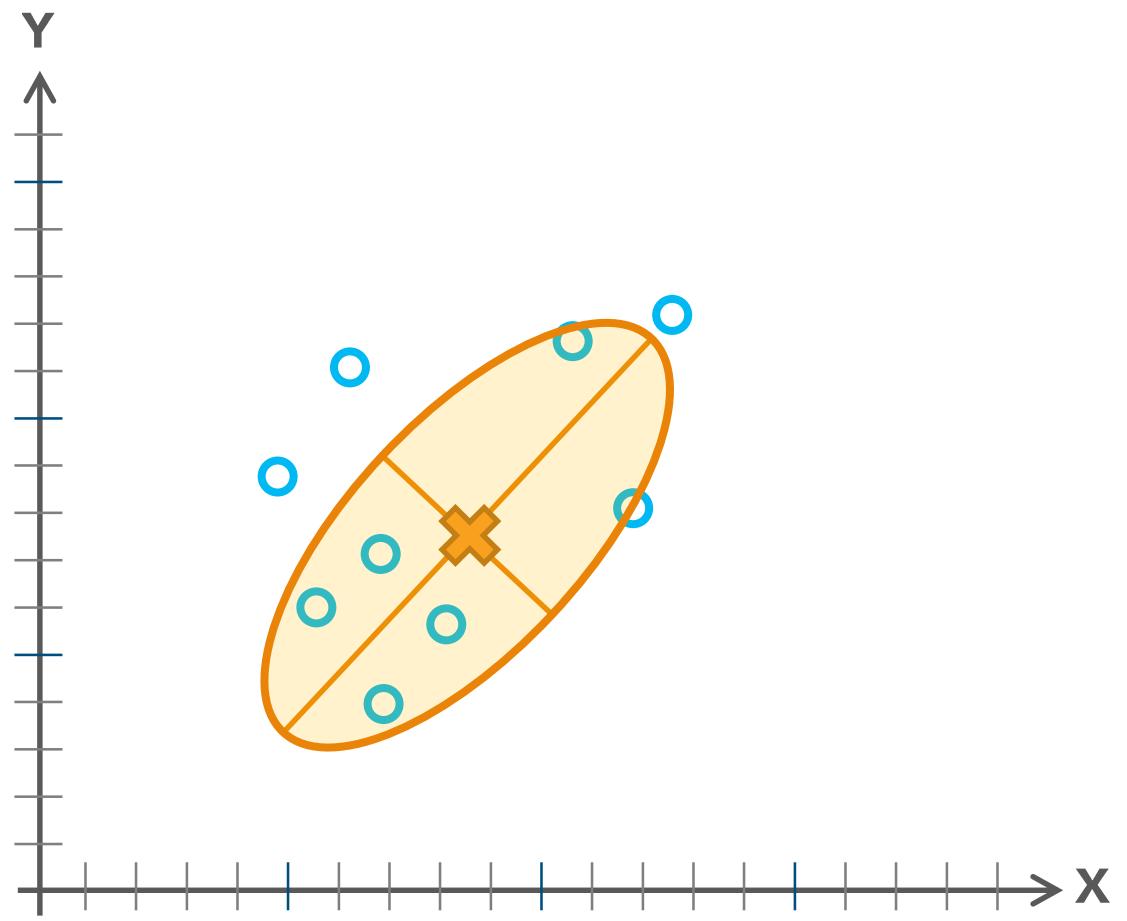




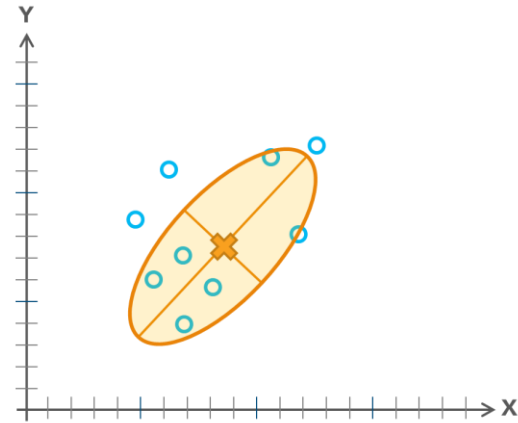
Directional Distribution (Standard Deviational Ellipse)

creates standard deviational ellipses to summarize the spatial characteristics of geographic features: central tendency, dispersion, and directional trends





Demo



Similarity Search

identifies which candidate features are most similar or most dissimilar to one or more input features based on feature attributes

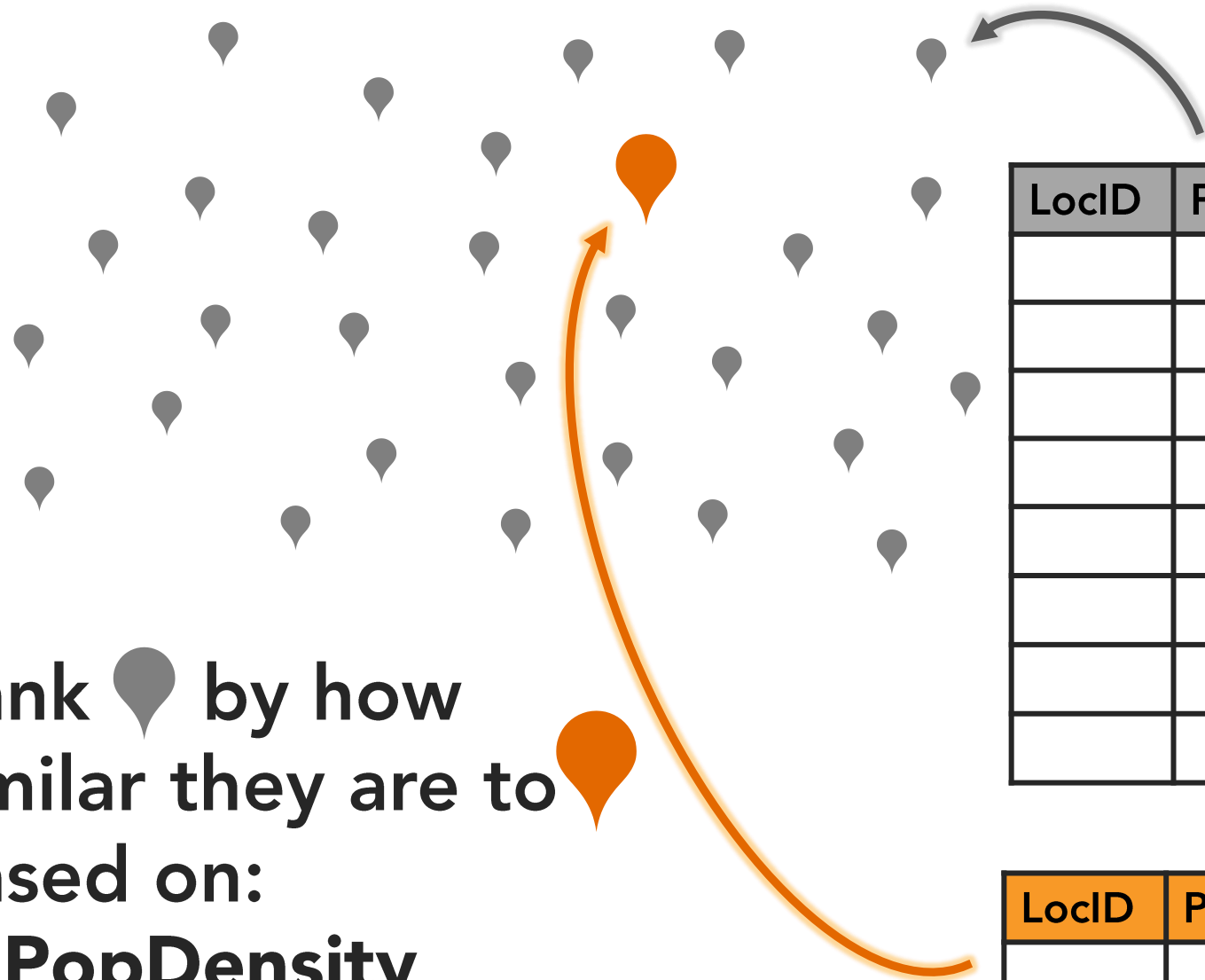


**potential
store
locations**



**high
performing
store**

**potential
store
locations**

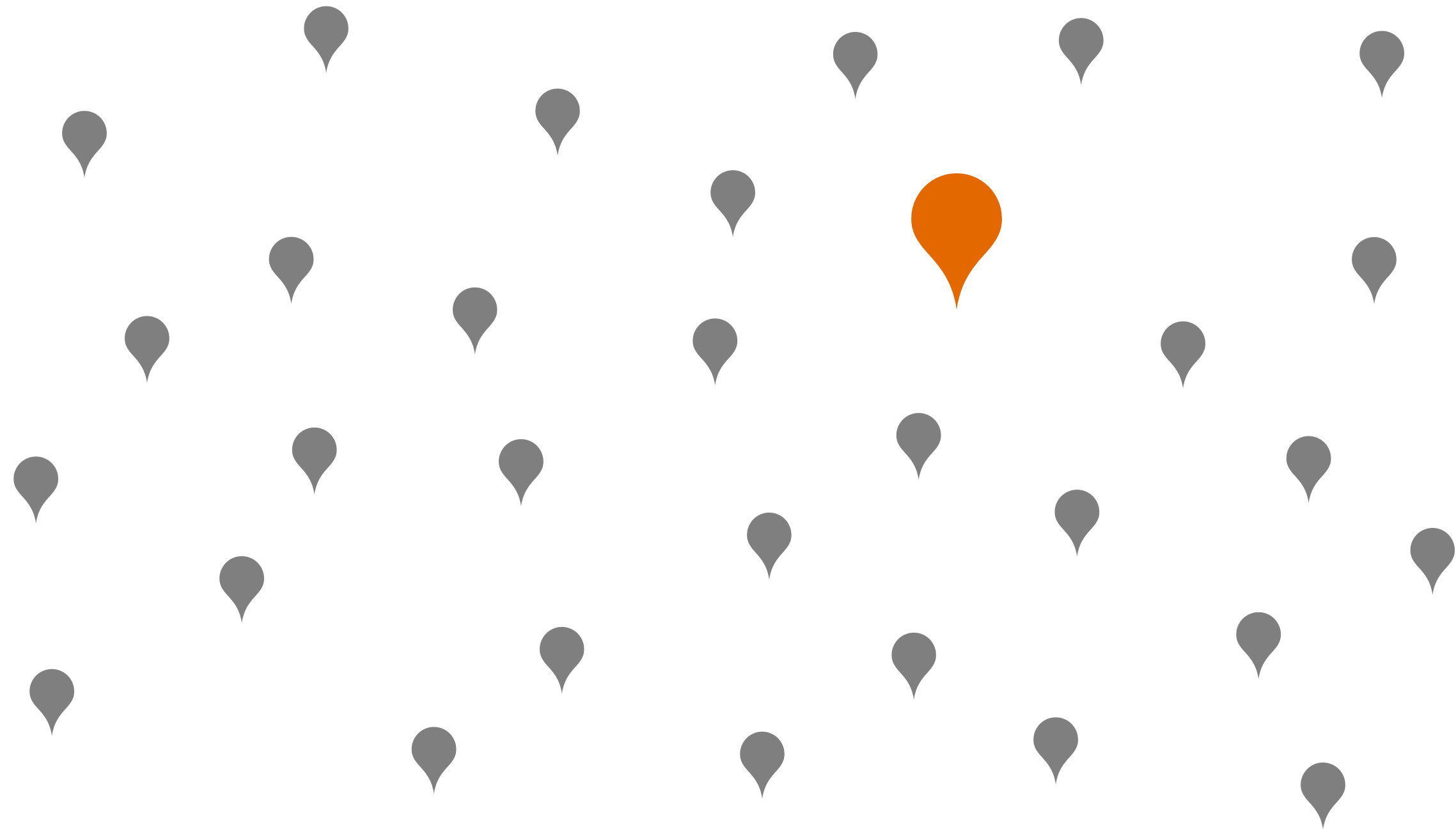


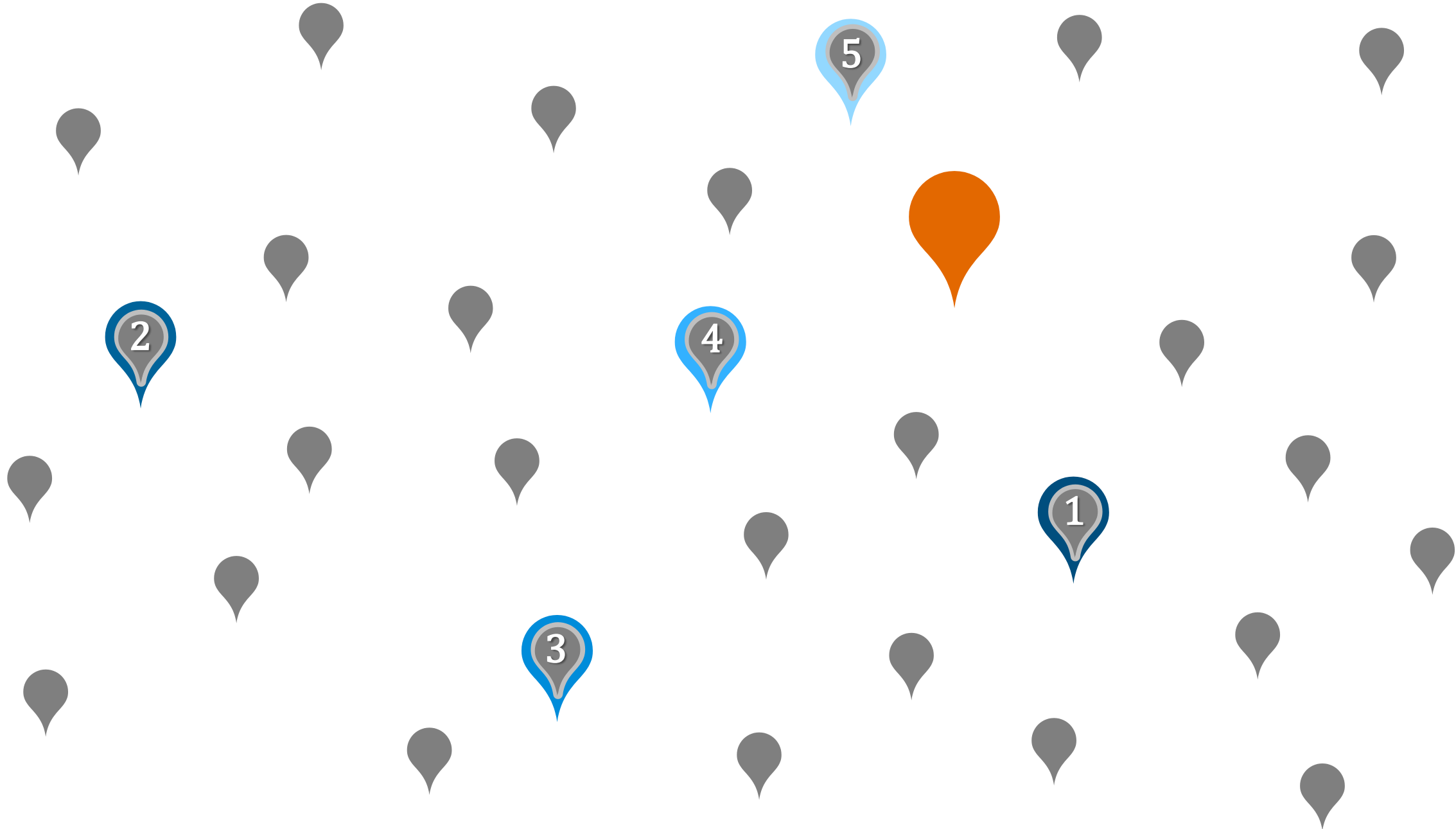
Rank 📍 by how similar they are to 📍 based on:

- **PopDensity**
- **AvIncome**
- **DistToCompetition**

LocID	PopDensity	AvIncome	DistToCompetition

LocID	PopDensity	AvIncome	DistToCompetition





Attribute Values



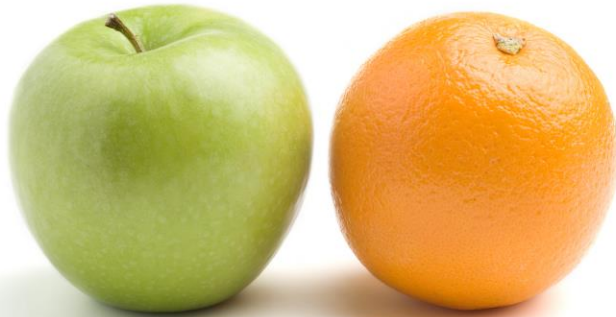
standardize
attributes

Z-transform:

$$(x - \bar{x}) / SD$$

Attribute Values

standardize
attributes



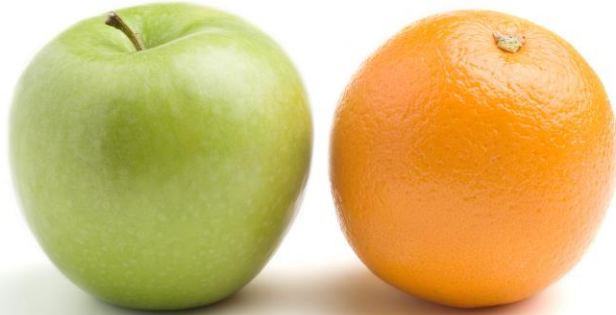
Population = 14,159

% Uninsured = .26

Distance (km) = 535.89

Attribute Values

standardize
attributes



Population = $-.7932$

% Uninsured = 3.8462

Distance (km) = $.6433$

Attribute Values

standardize
attributes

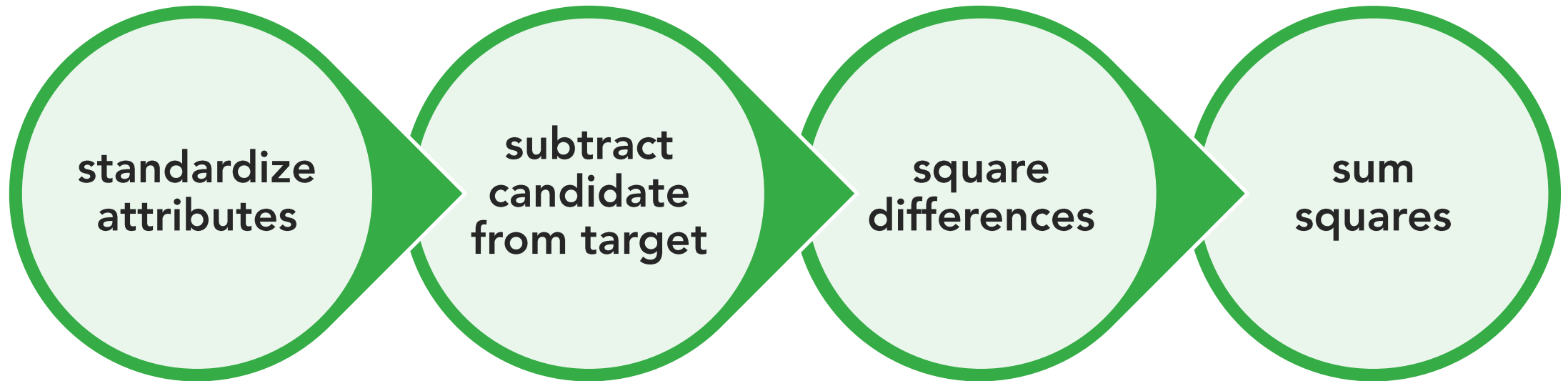


Population = $-.7932$

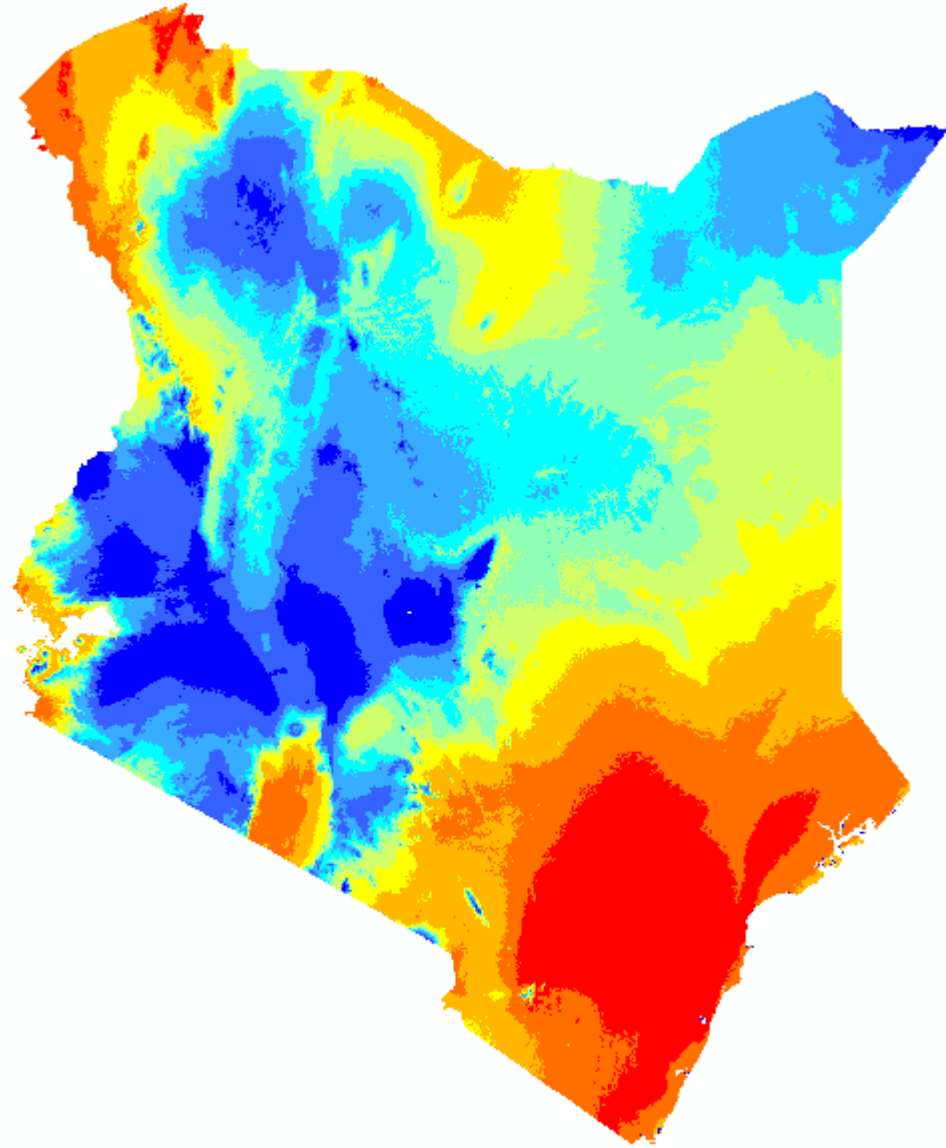
% Uninsured = 3.8462

Distance (km) = $.6433$

Attribute Values



Dengue Fever Risk in Kenya



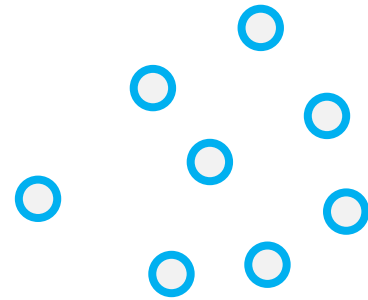
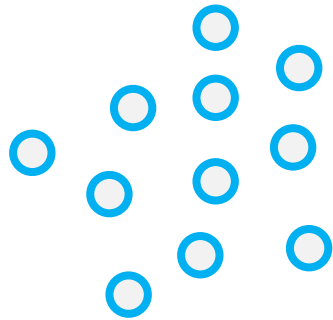
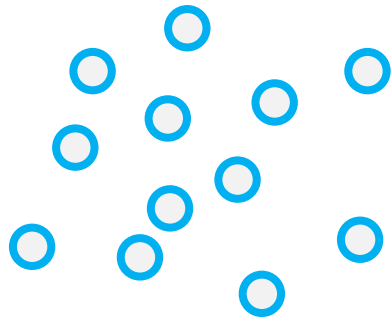
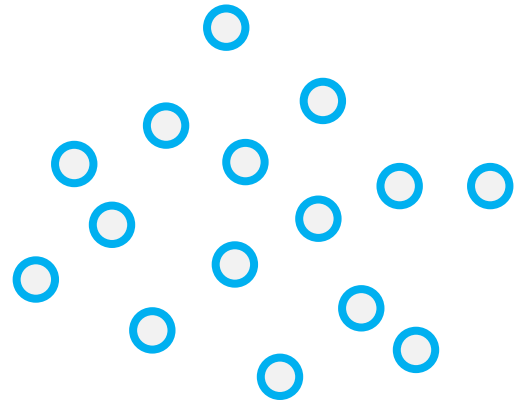
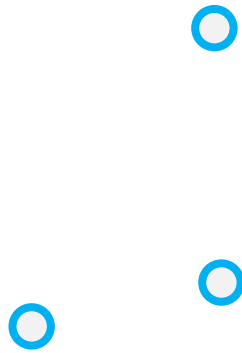
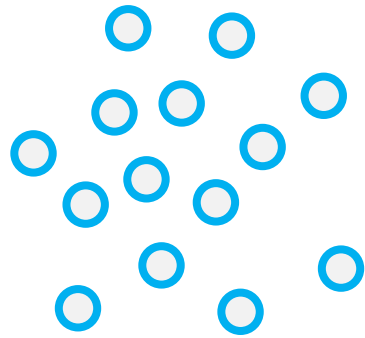
Machine Learning

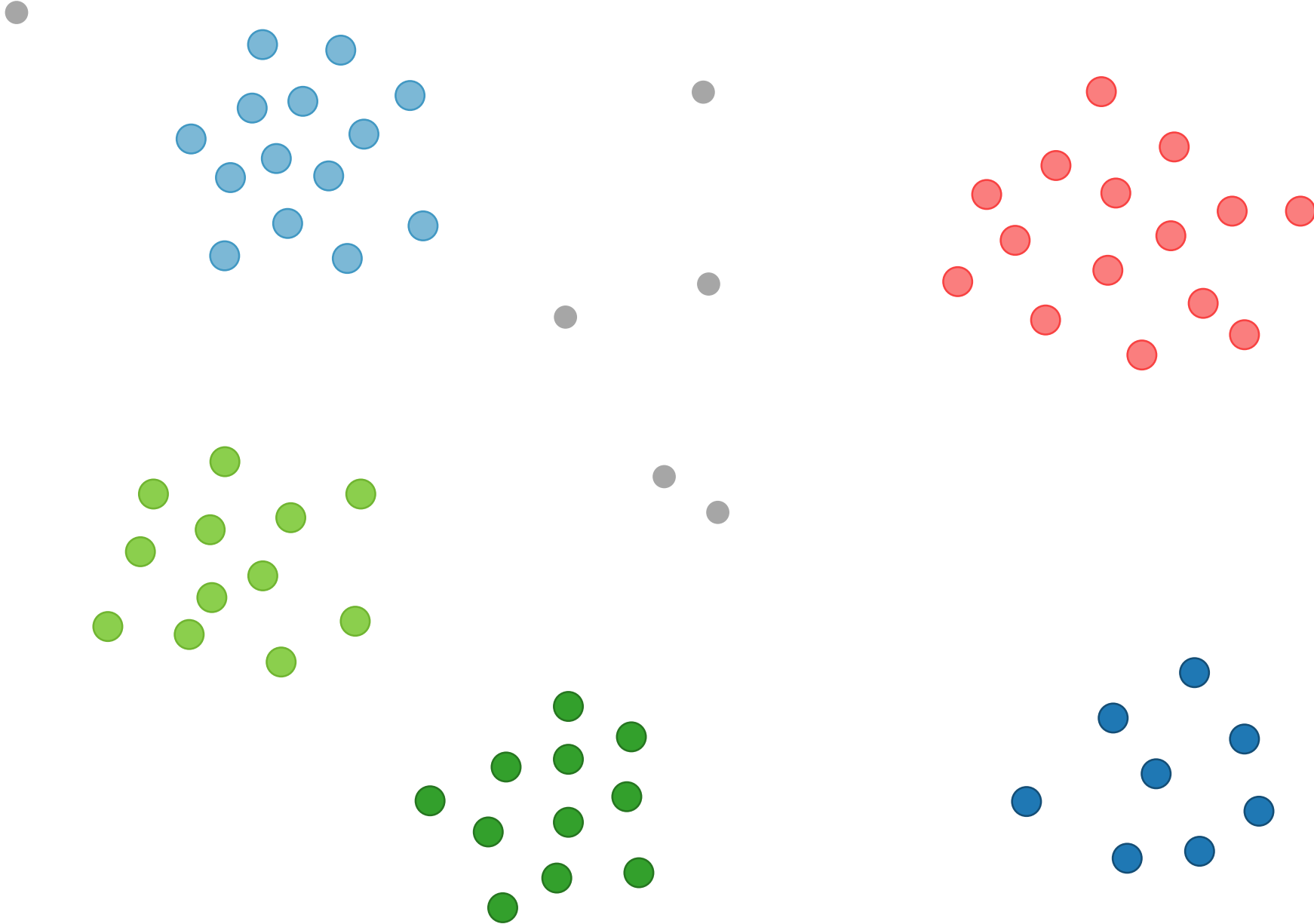
clustering methods



Density-based Clustering

finds clusters based on feature locations



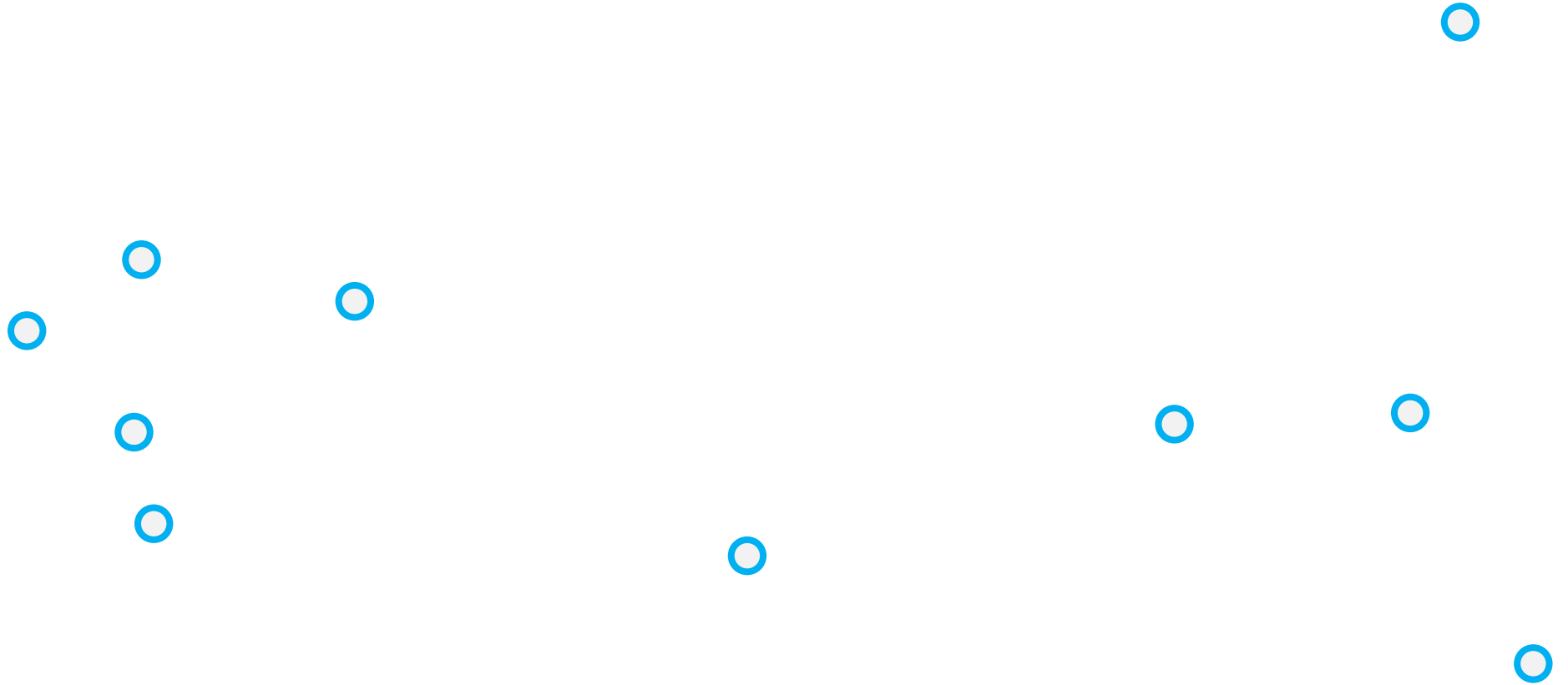


DBSCAN – defined distance

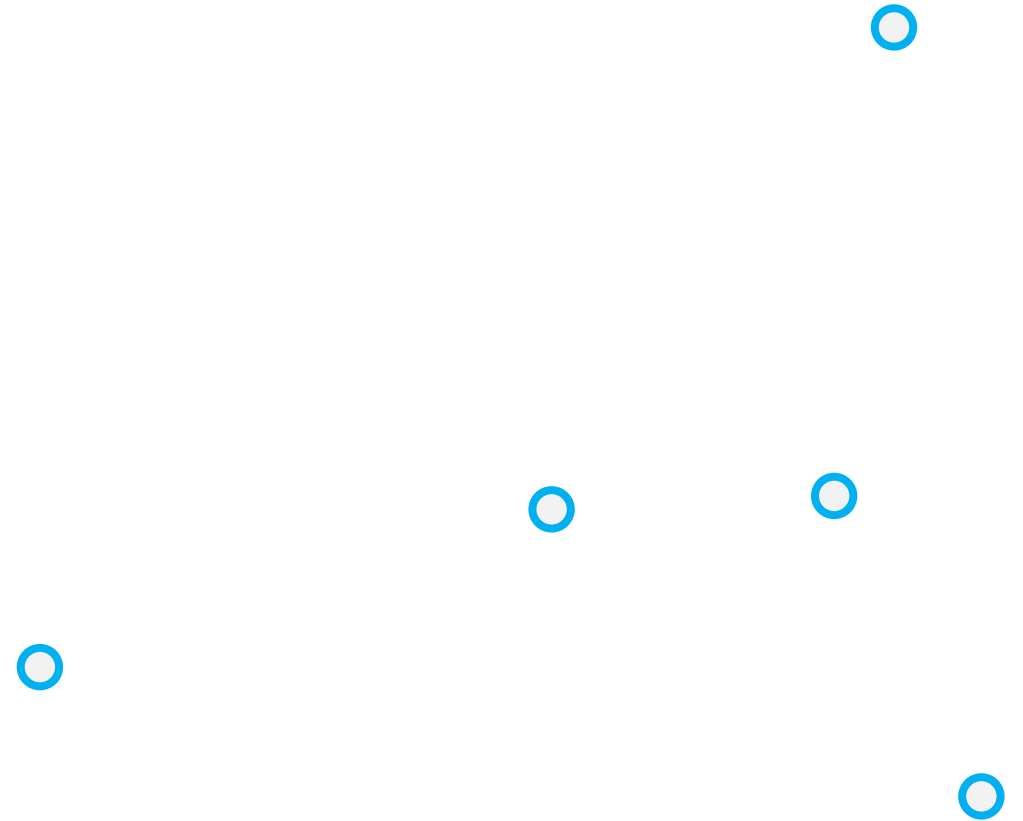
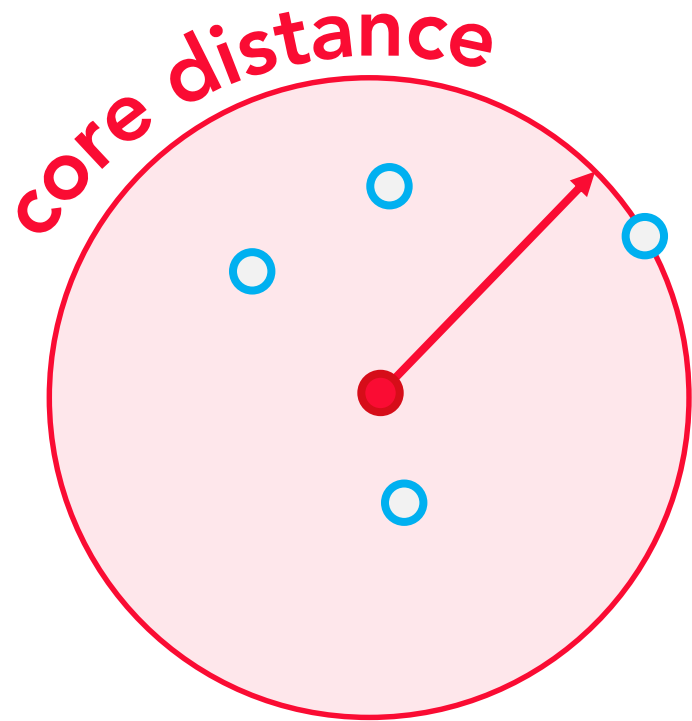
HDBSCAN – self adjusting

OPTICS – multi-scale

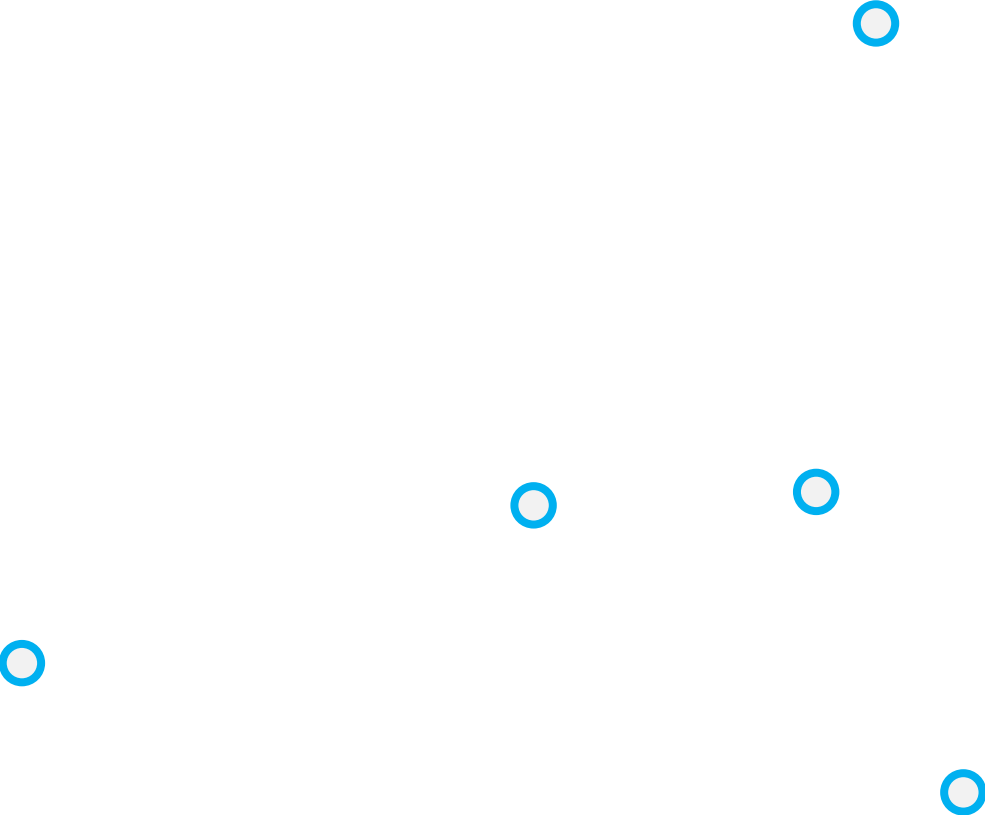
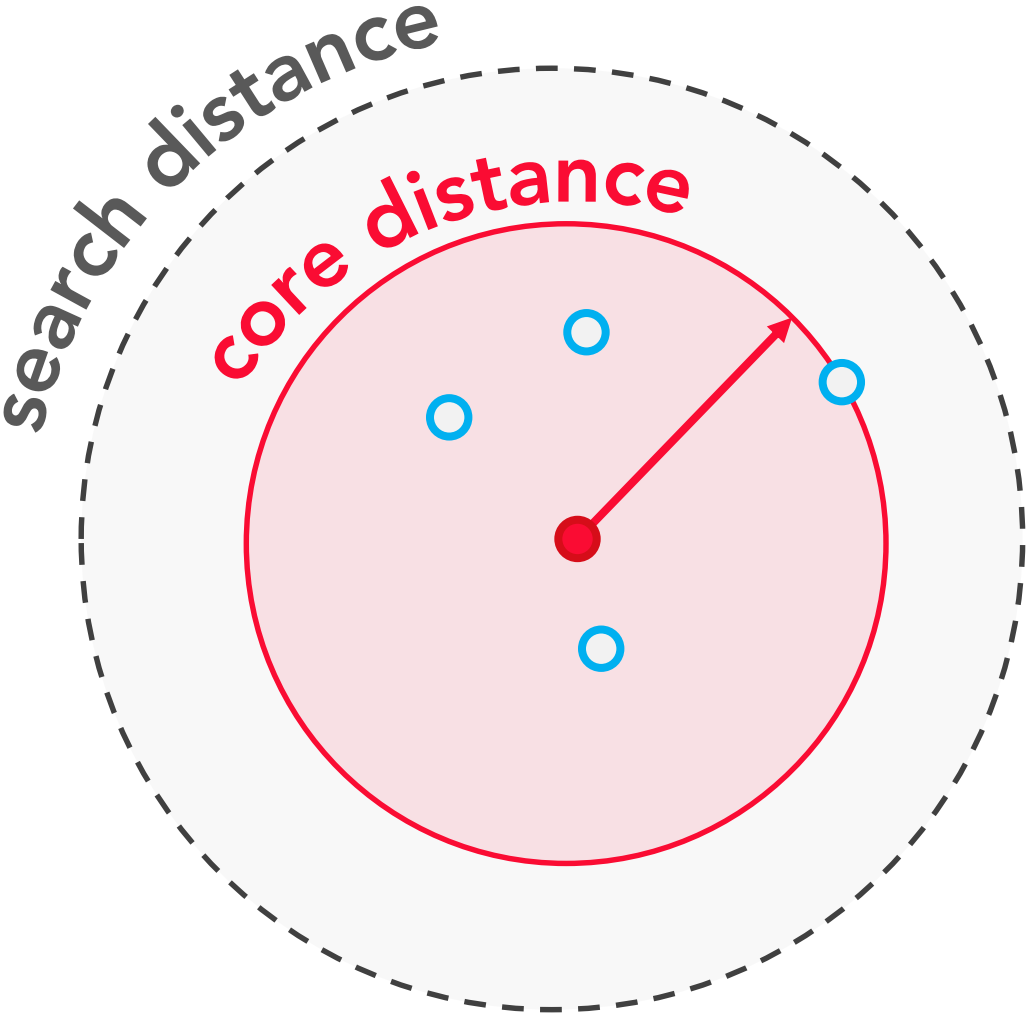
DBSCAN – defined distance



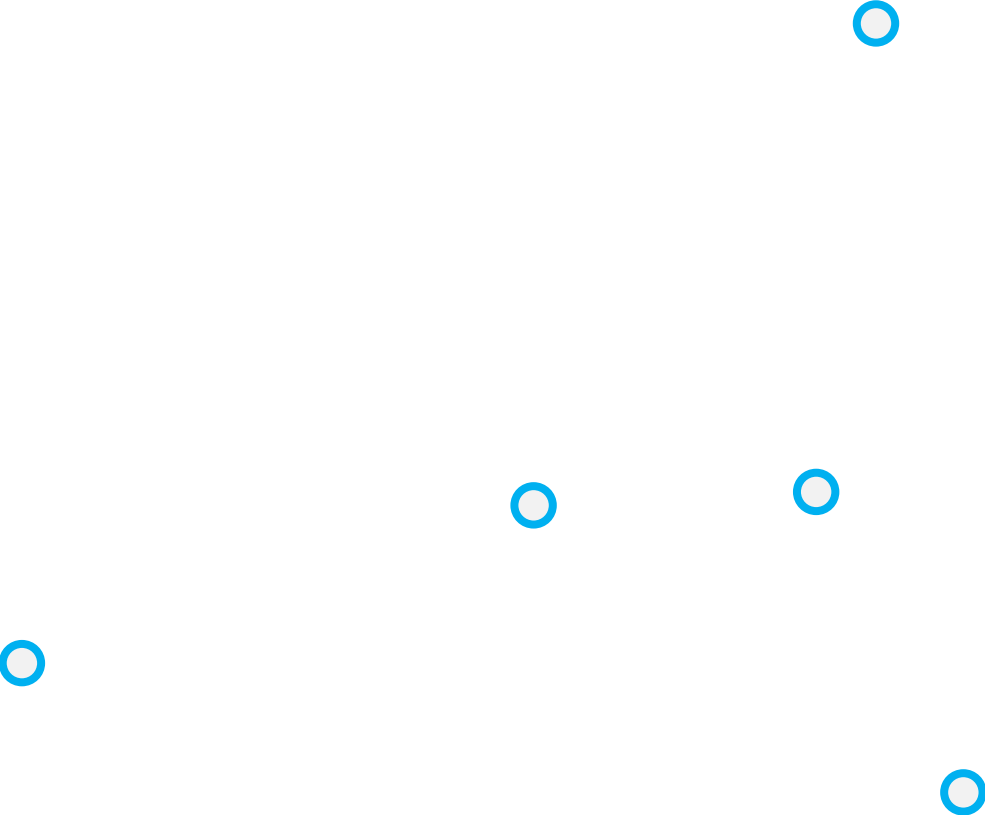
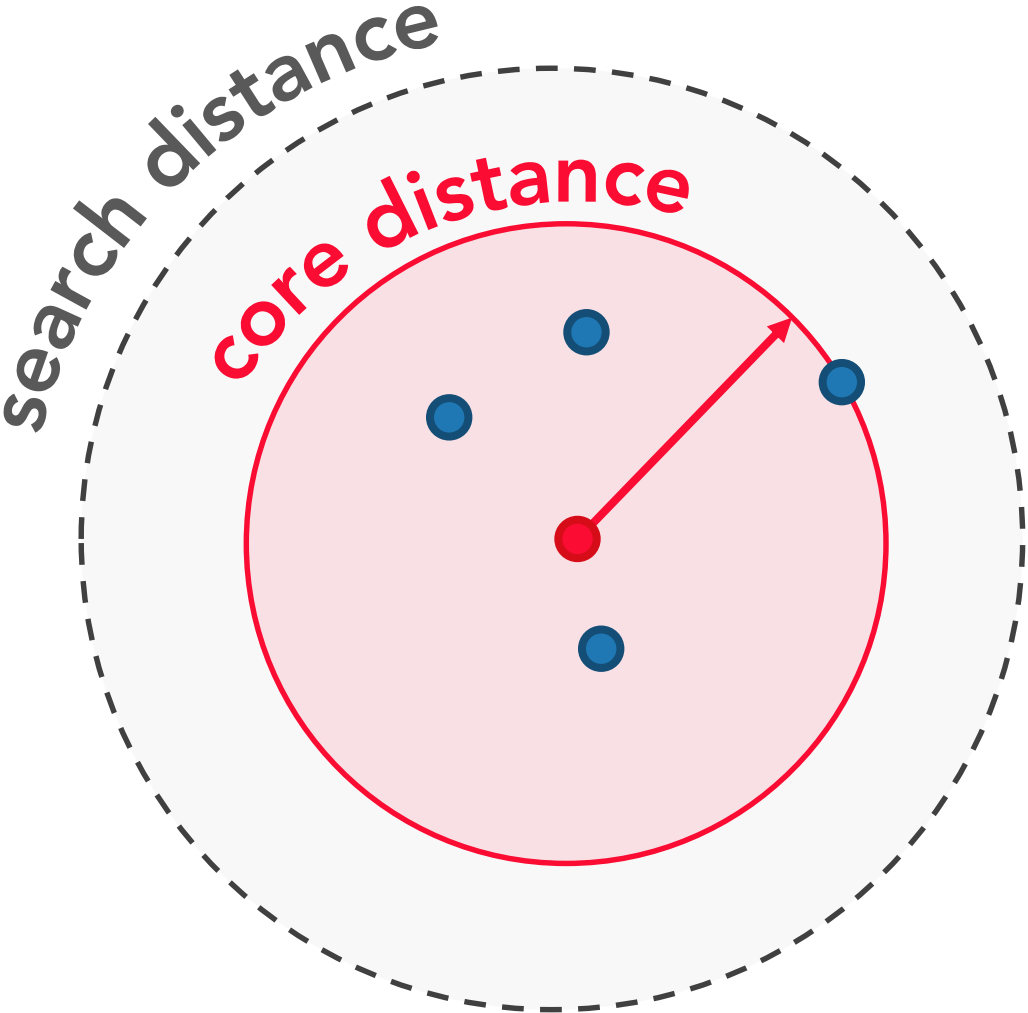
DBSCAN – defined distance



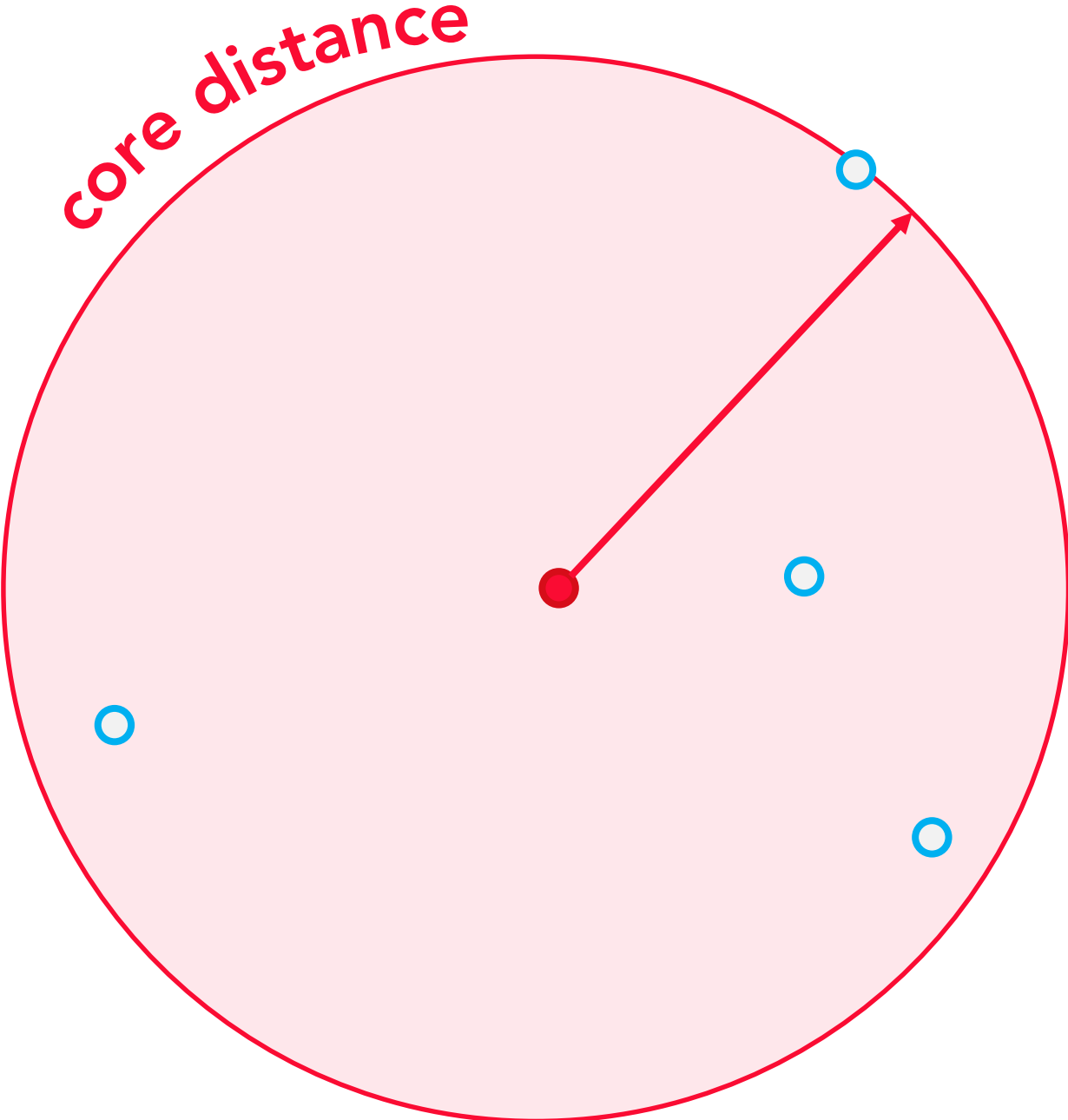
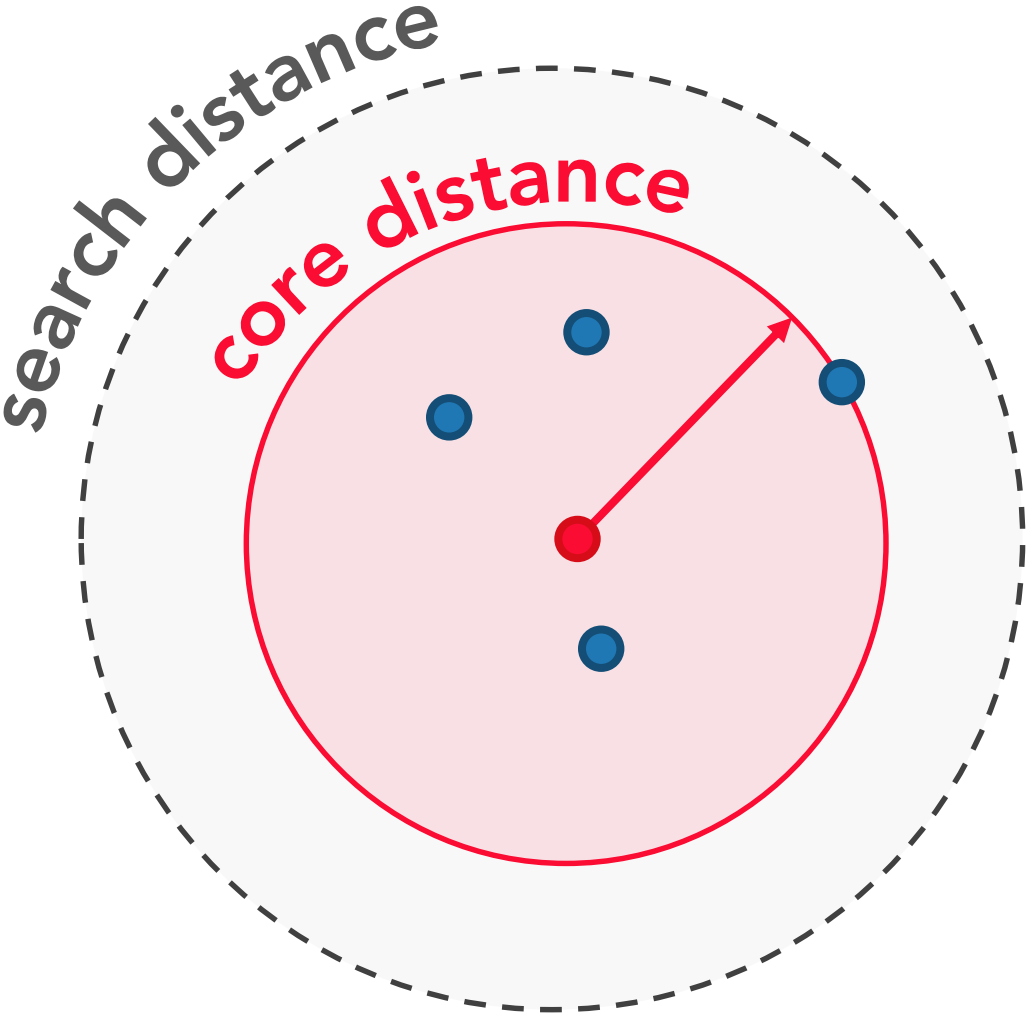
DBSCAN – defined distance



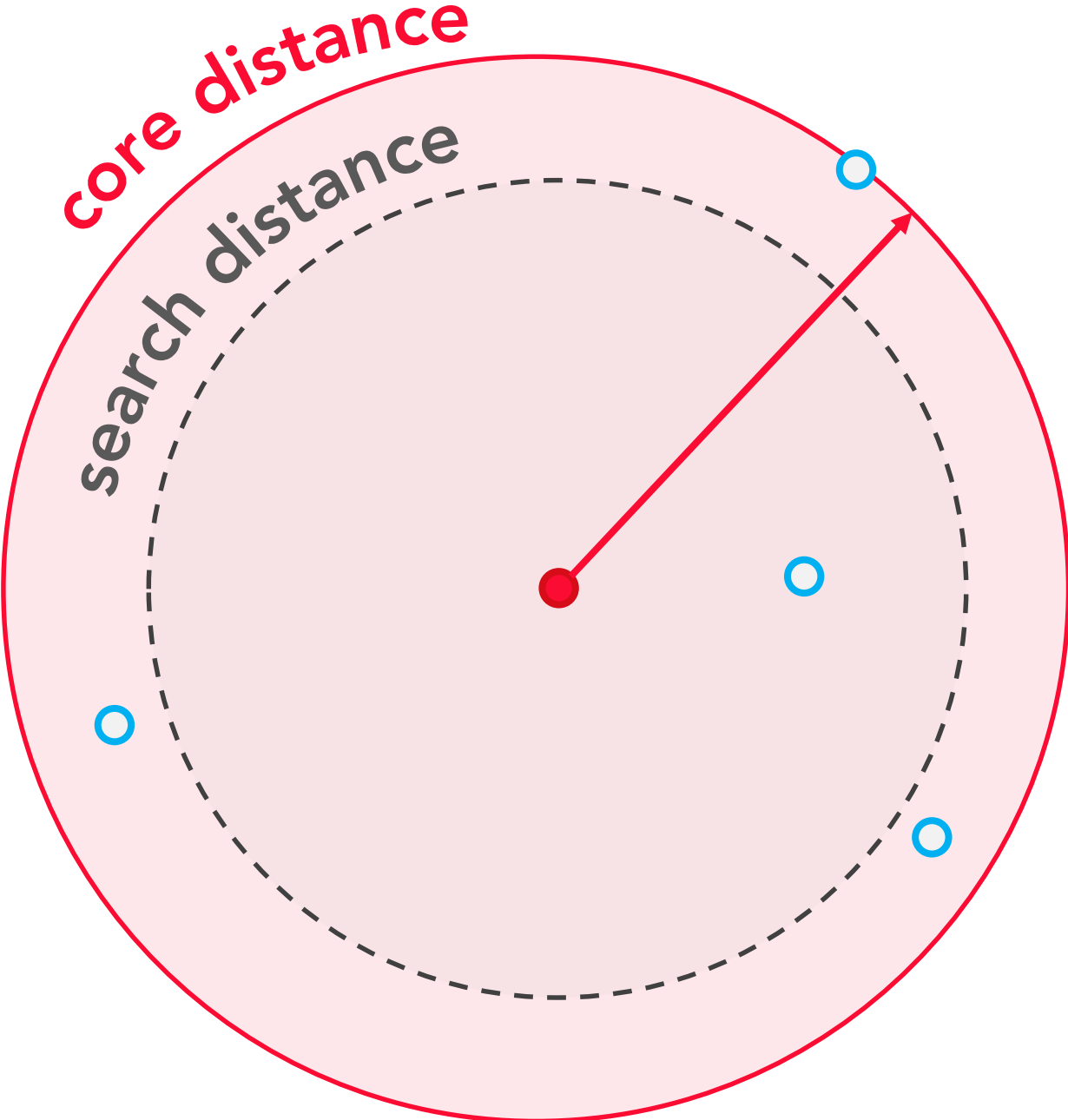
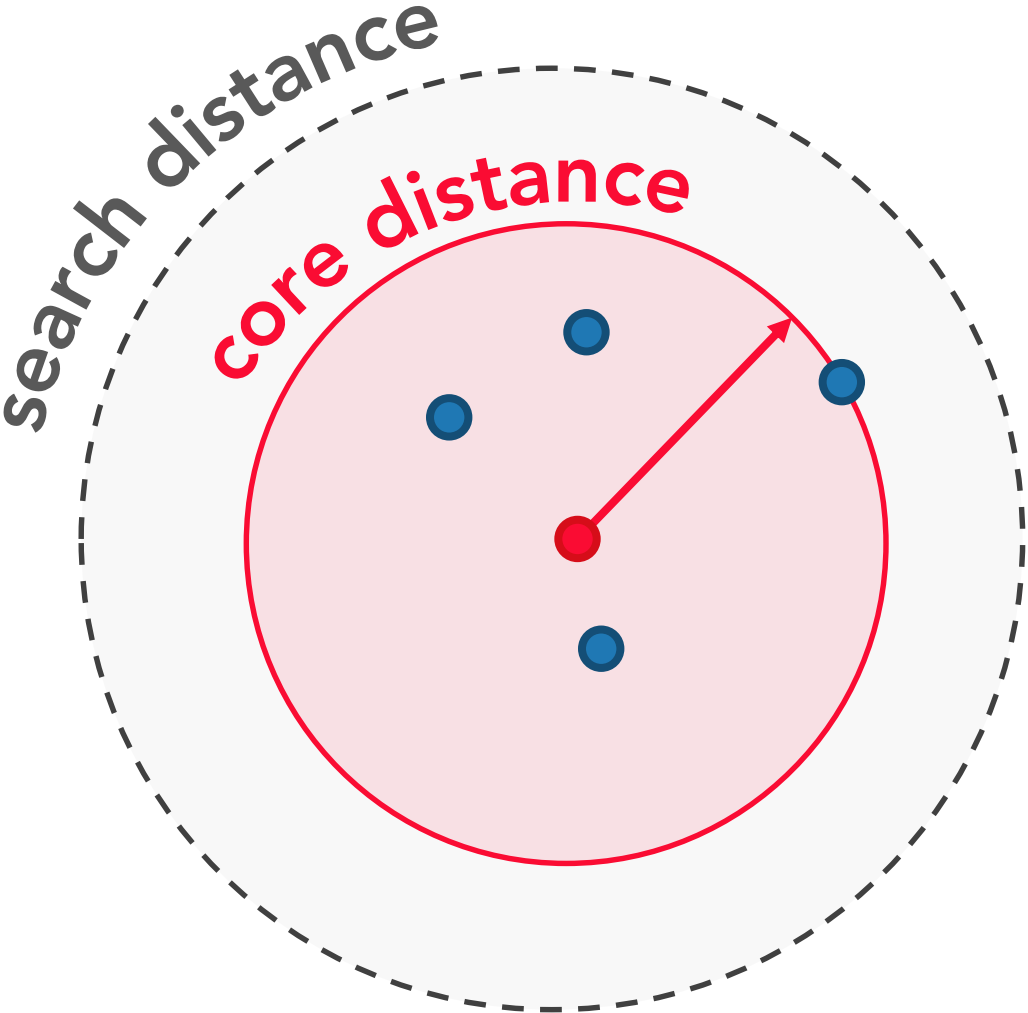
DBSCAN – defined distance



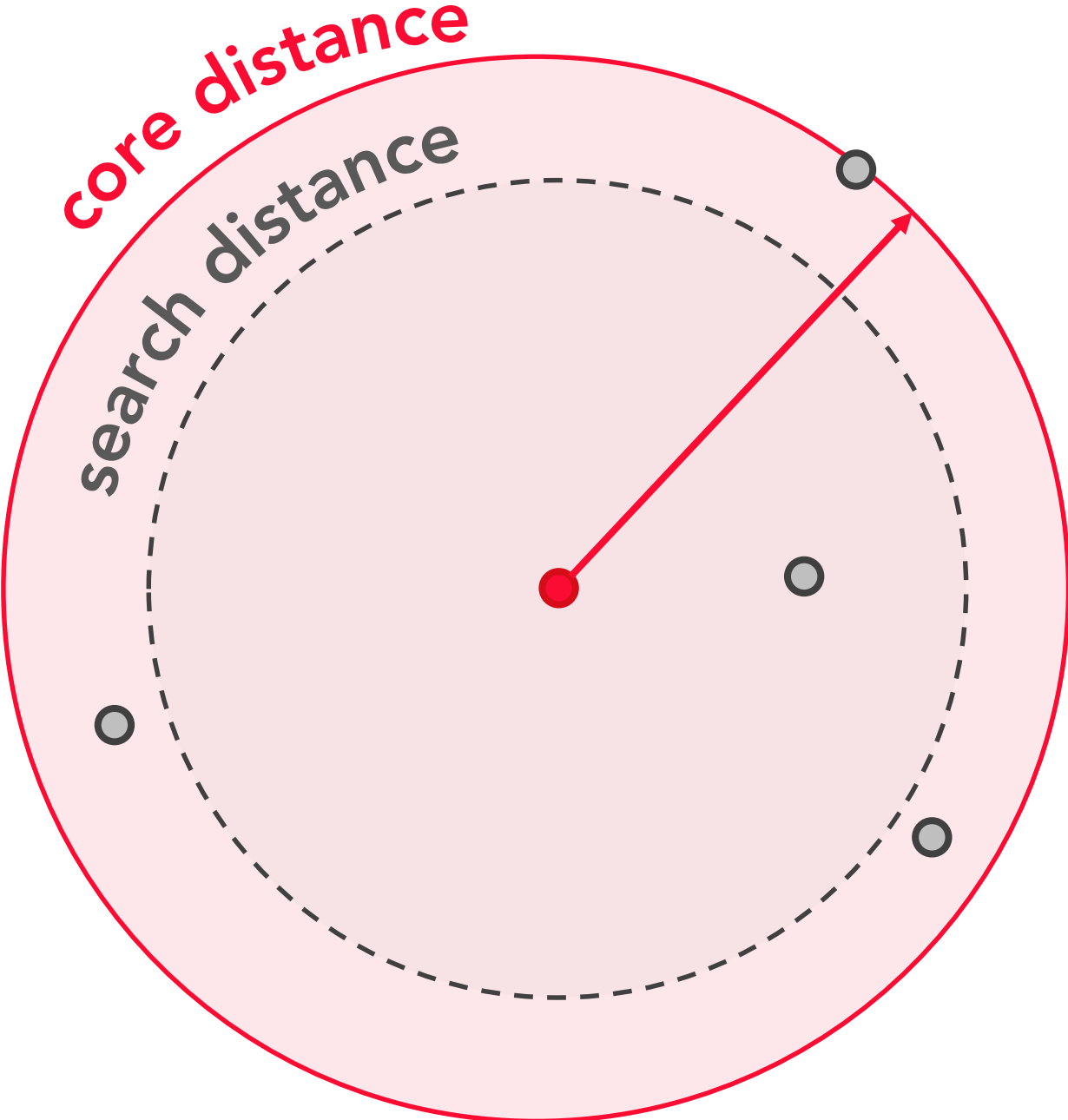
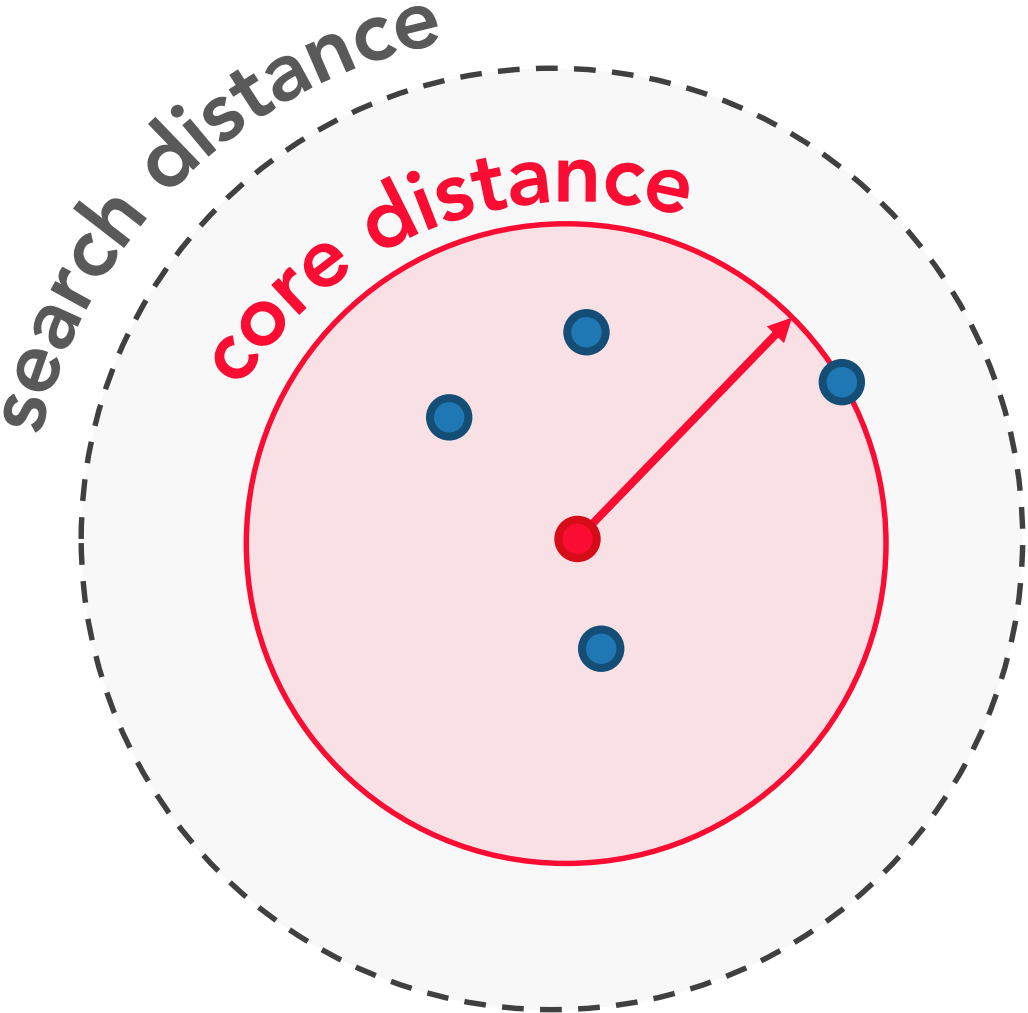
DBSCAN – defined distance



DBSCAN – defined distance

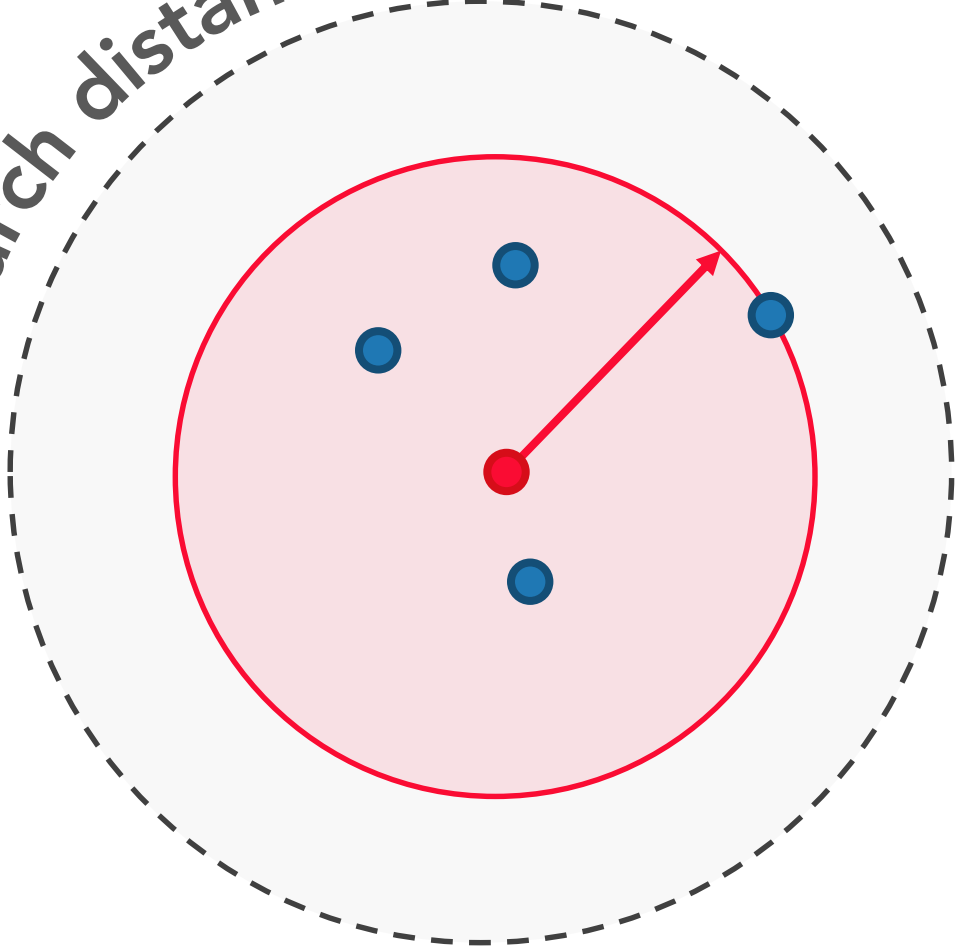


DBSCAN – defined distance

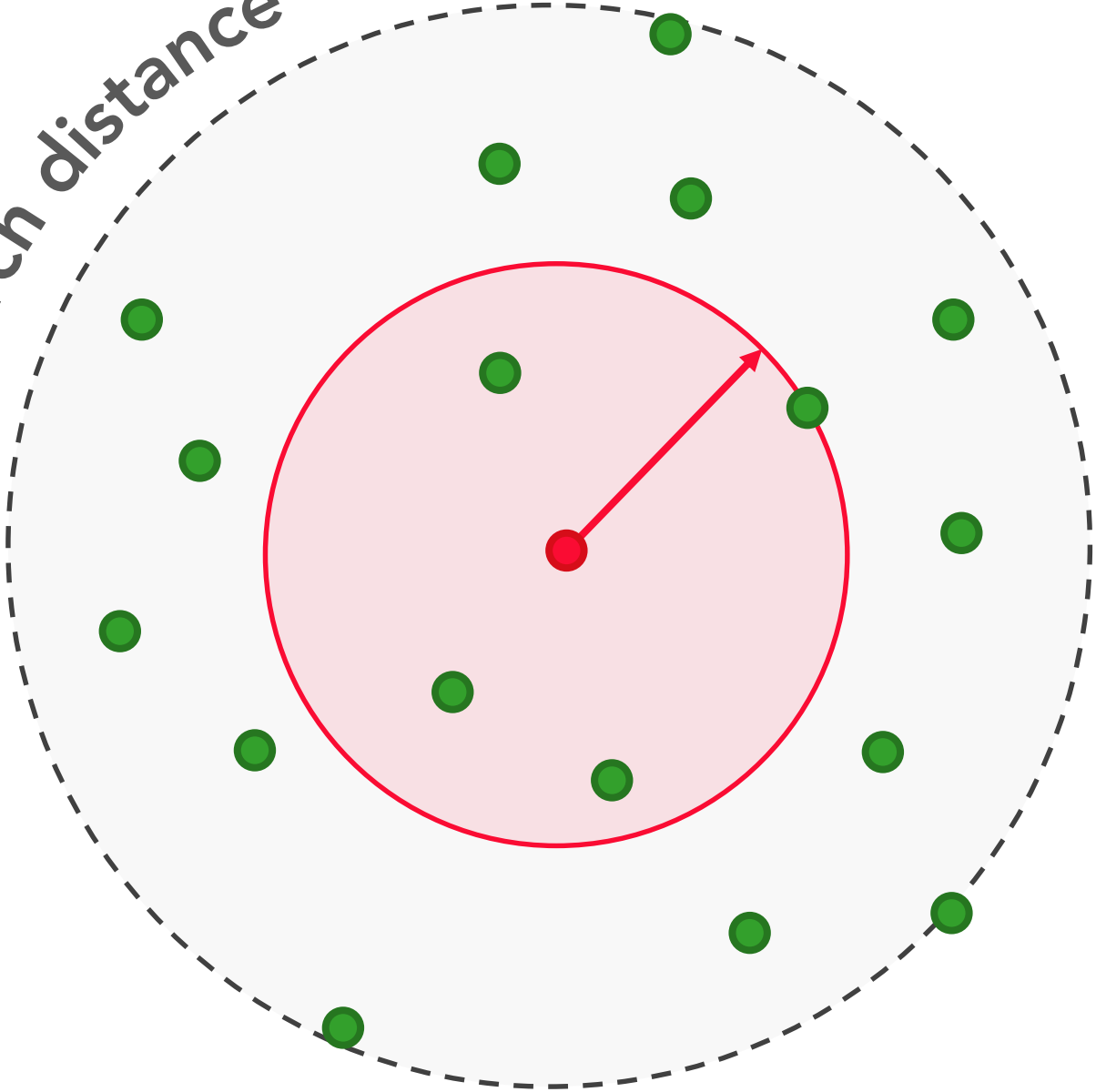


HDBSCAN – self adjusting

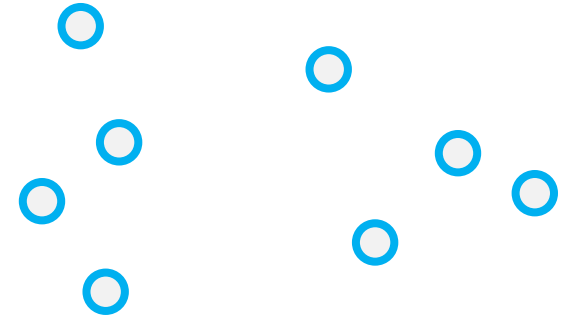
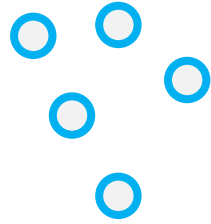
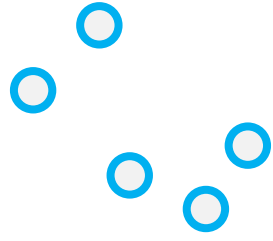
search distance



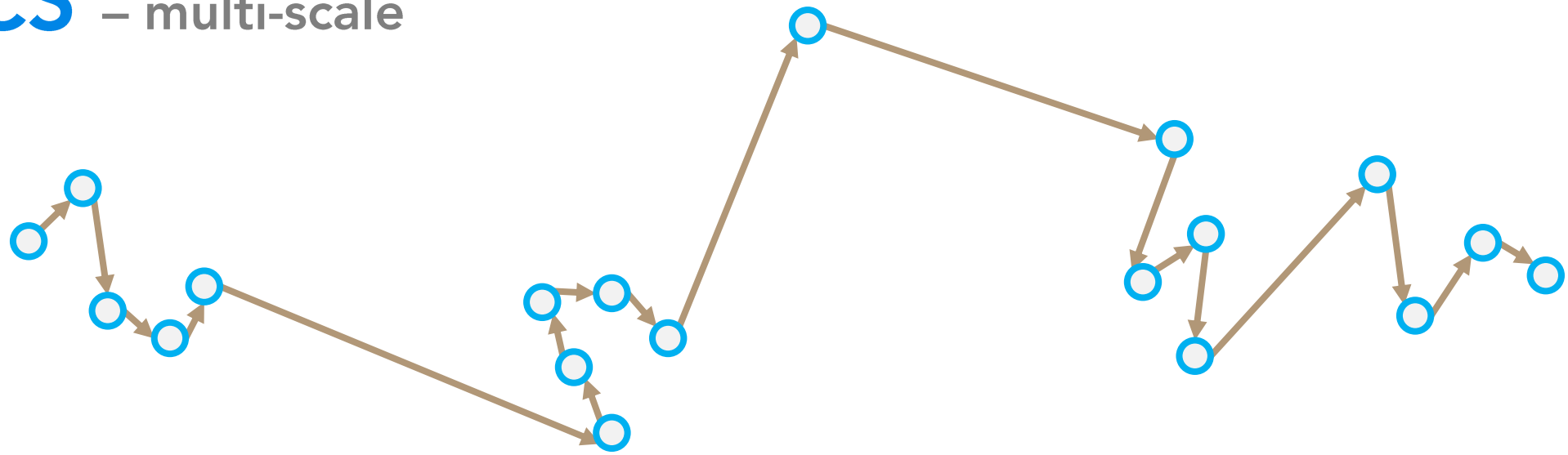
search distance



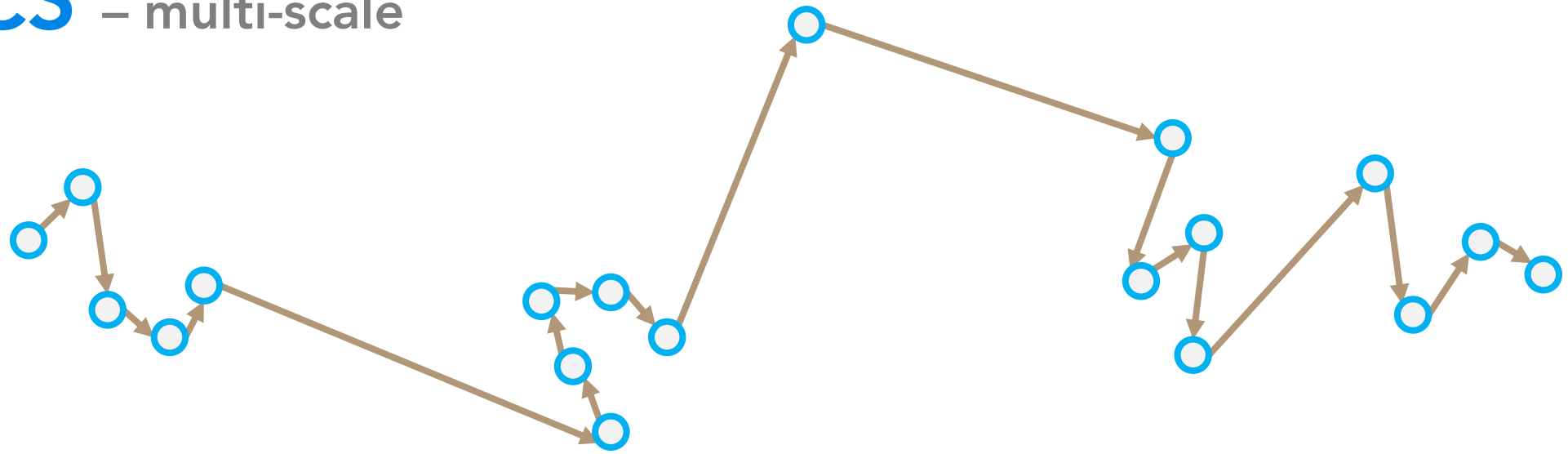
OPTICS – multi-scale



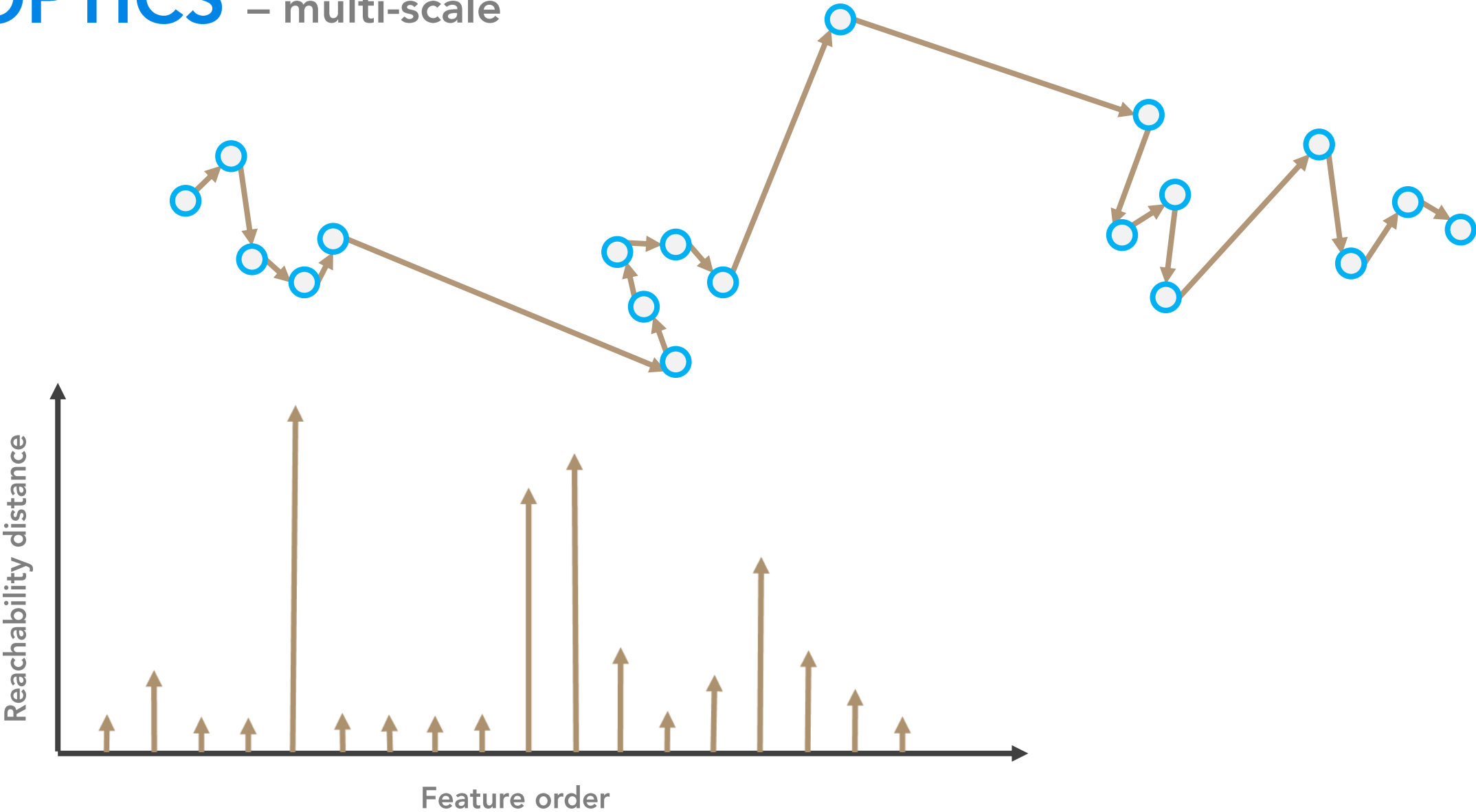
OPTICS – multi-scale



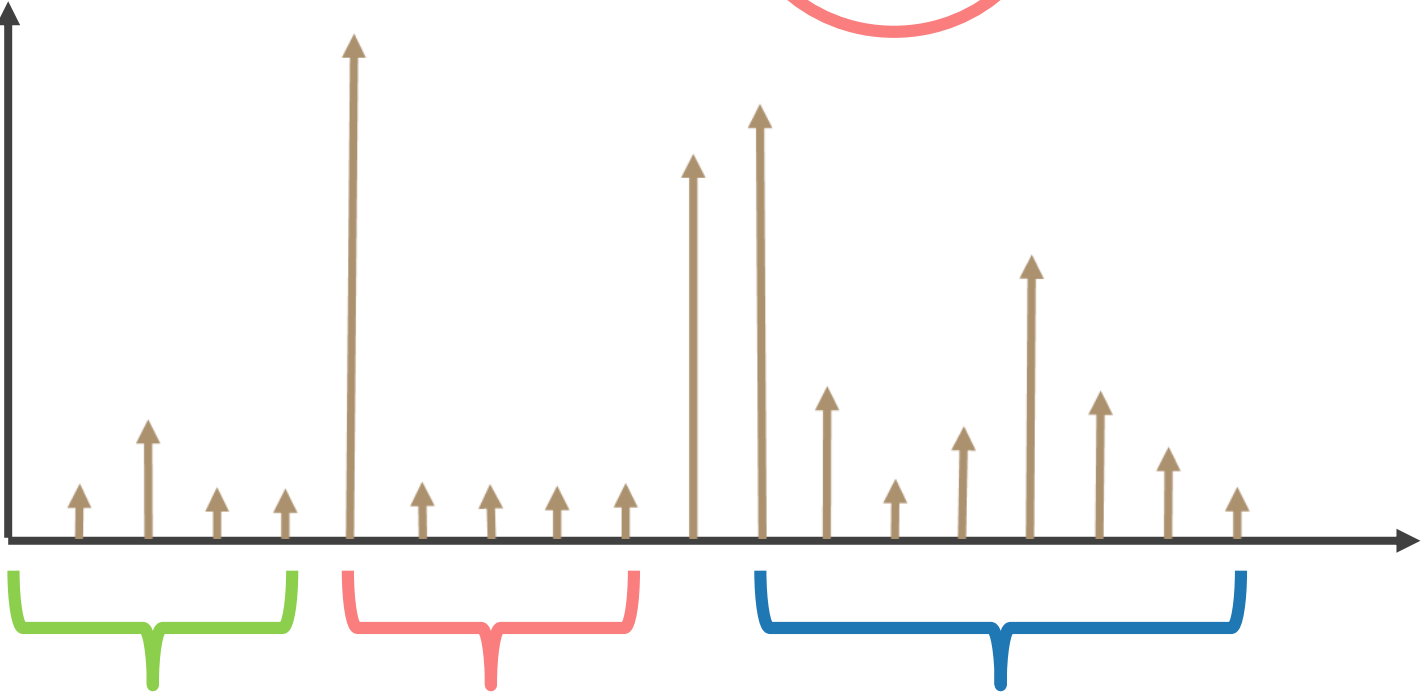
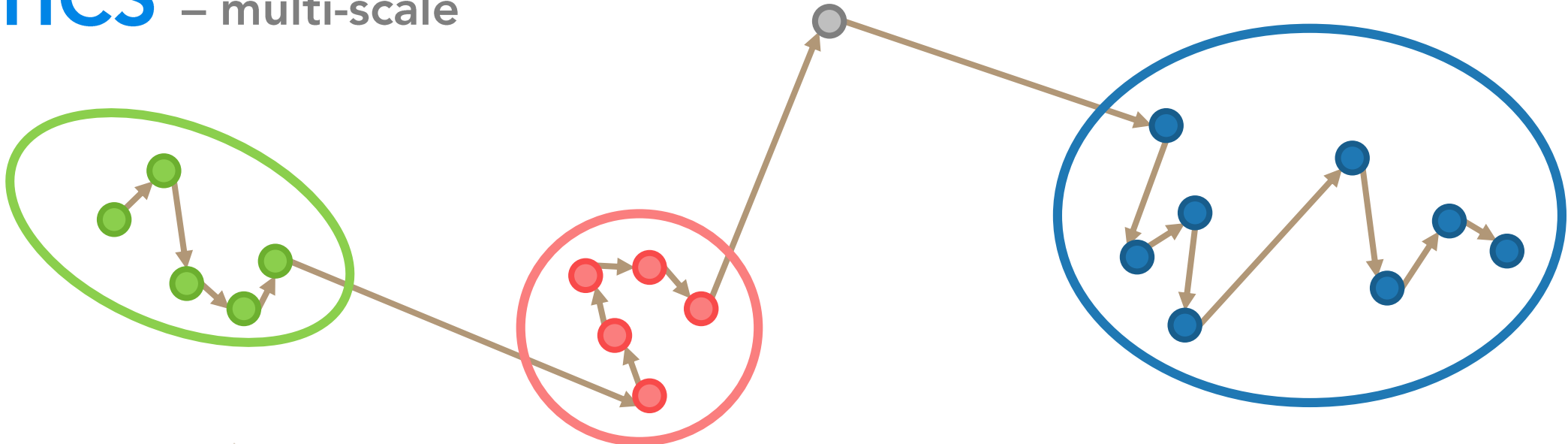
OPTICS – multi-scale



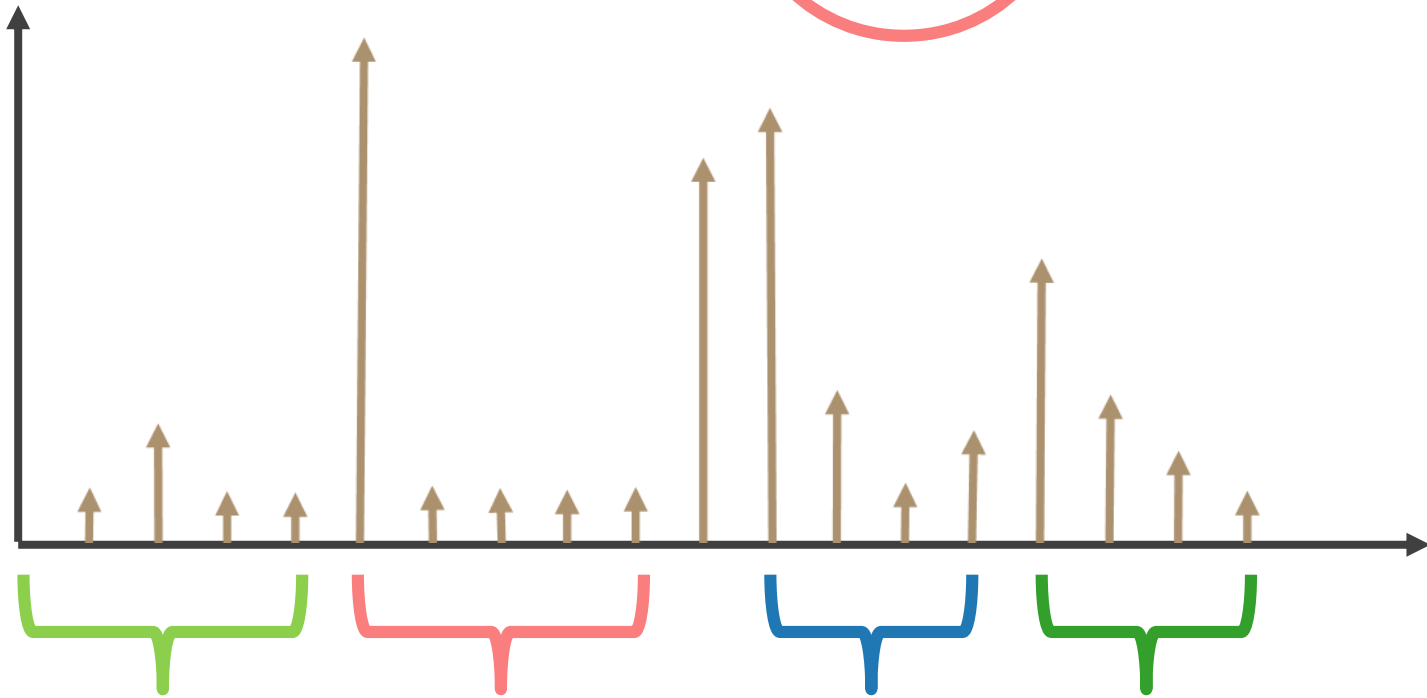
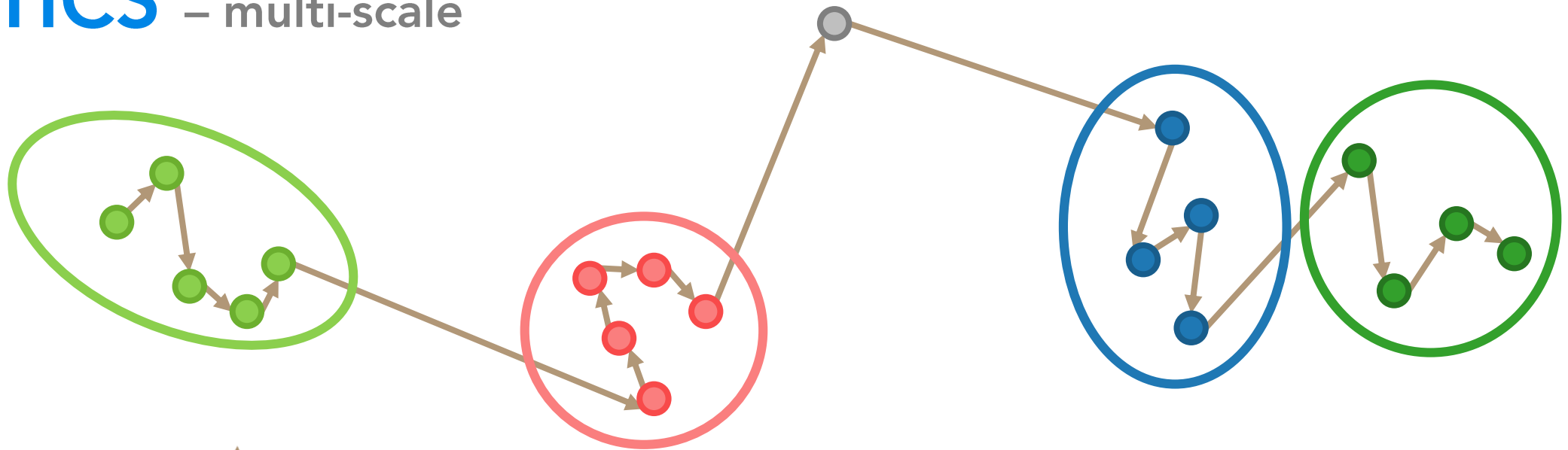
OPTICS – multi-scale



OPTICS – multi-scale



OPTICS – multi-scale



DBSCAN

- Uses fixed search distance
- Clusters of similar densities
- Fast

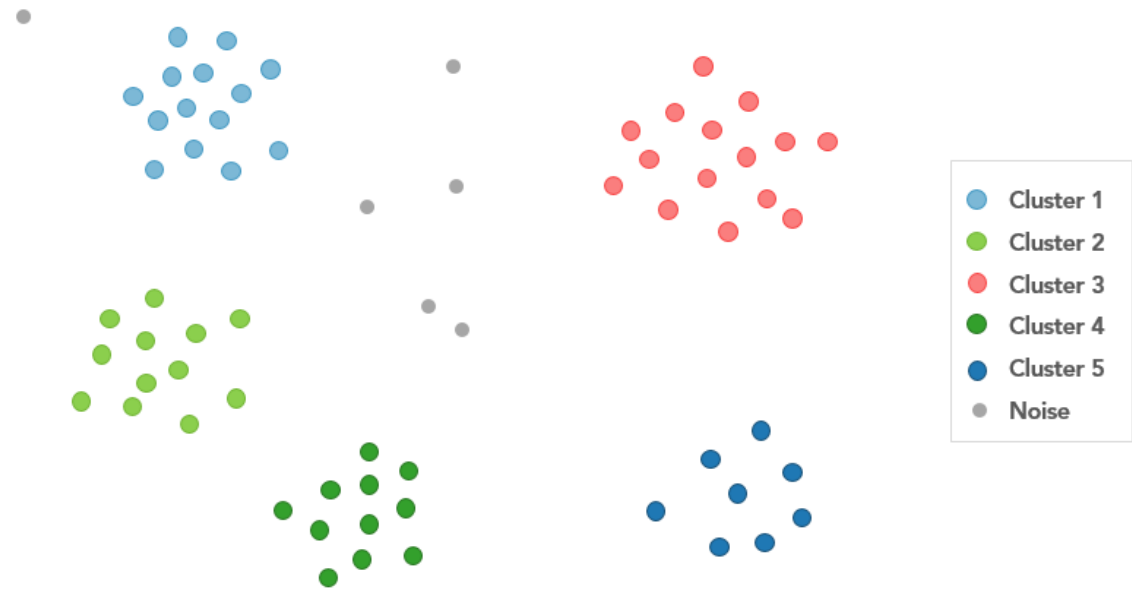
HDBSCAN

- Uses range of search distances to find clusters of varying densities
- Data driven, requires least user input

OPTICS

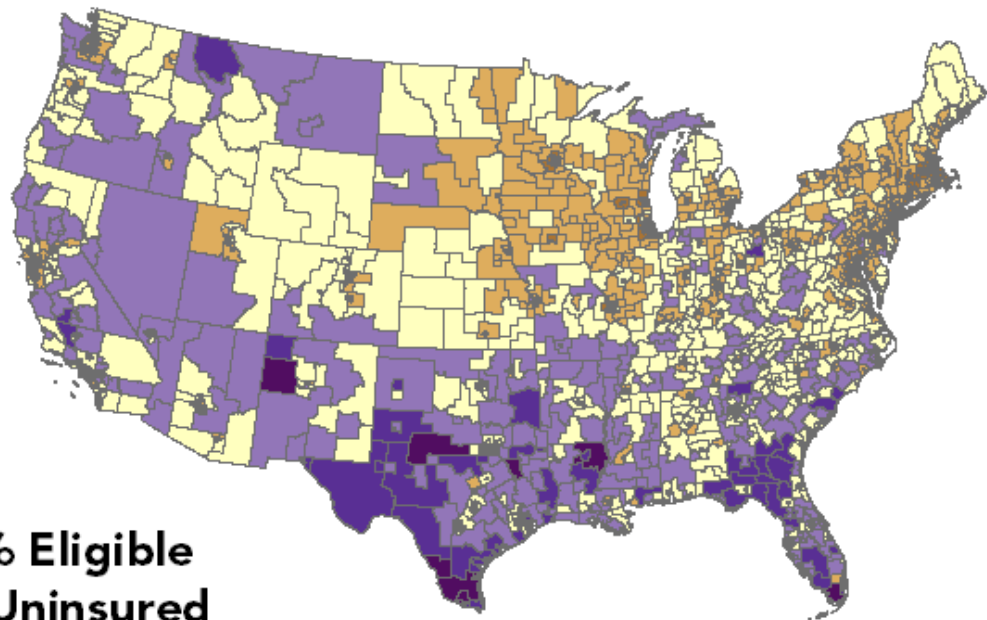
- Uses neighbor distances to create reachability plot
- Most flexibility for fine tuning
- Can be computationally intensive

Demo

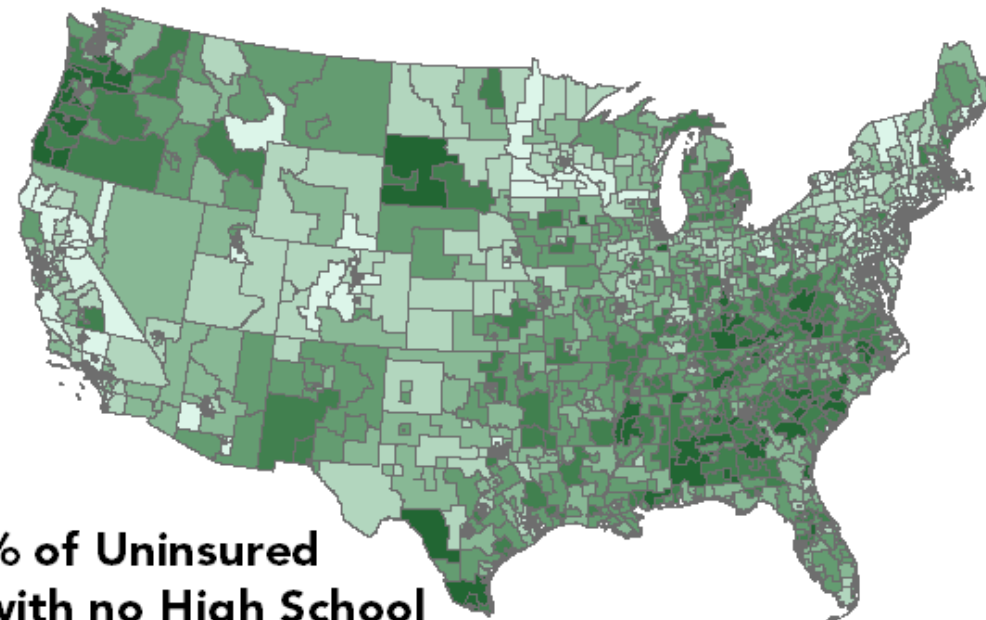


Multivariate Clustering

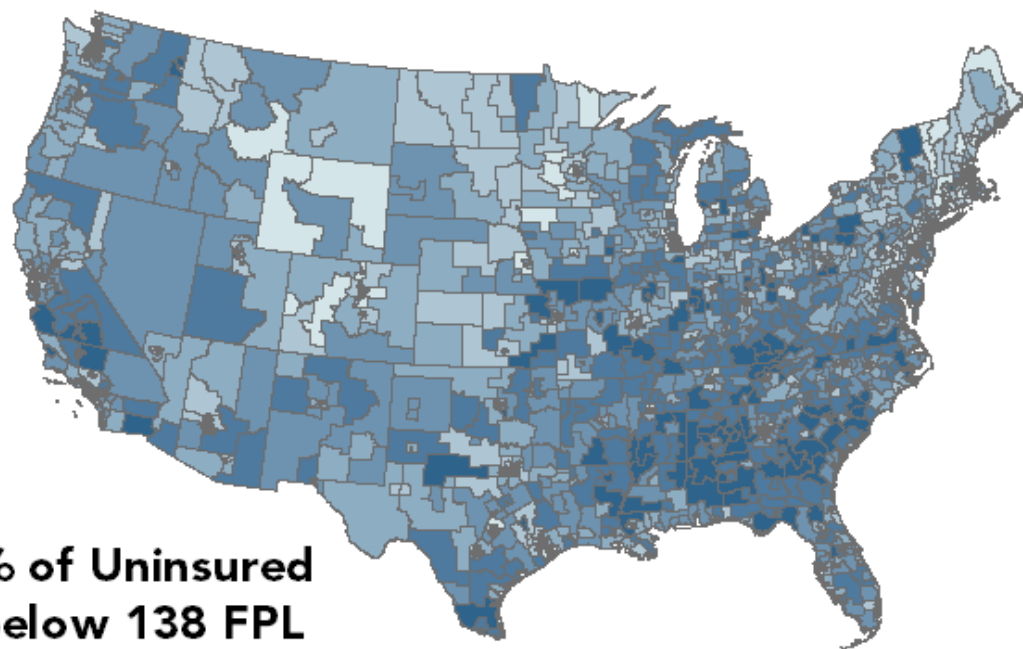
finds clusters based on feature attributes



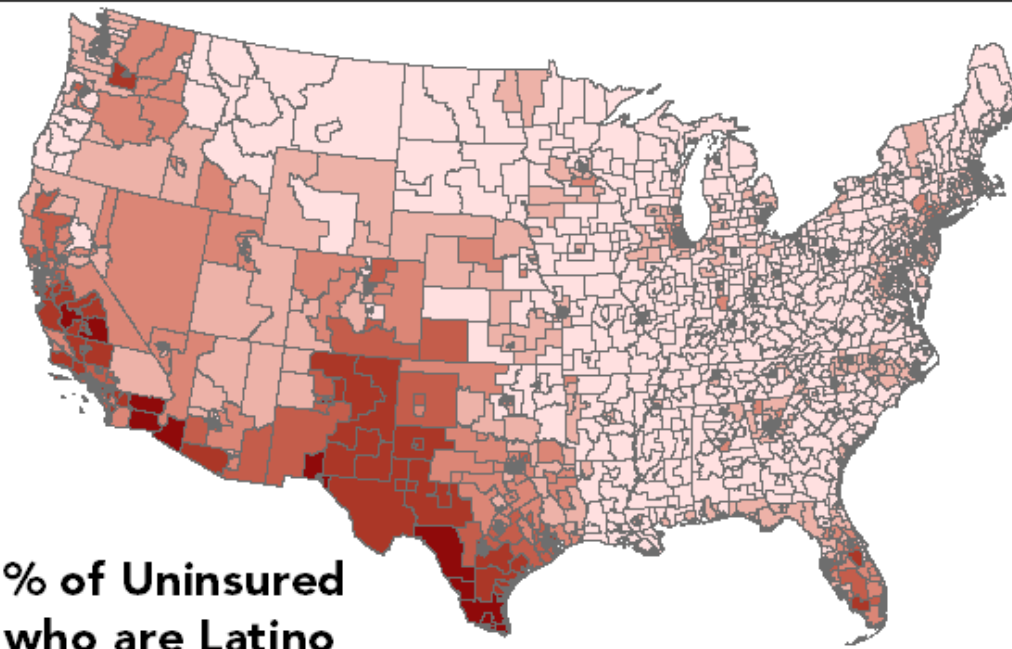
**% Eligible
Uninsured**



**% of Uninsured
with no High School**

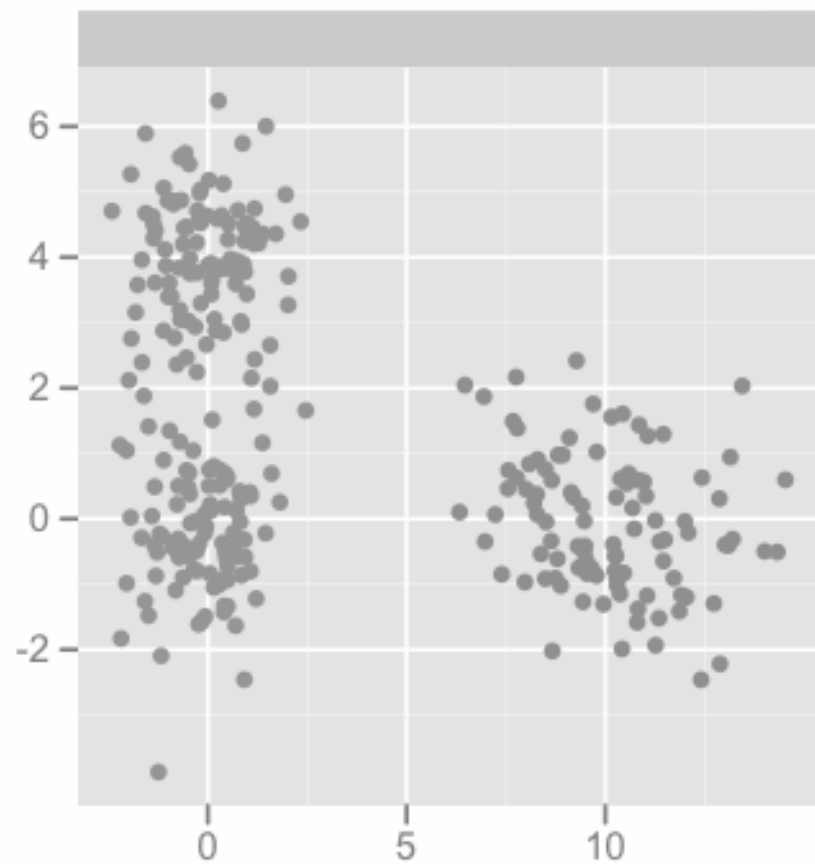


**% of Uninsured
below 138 FPL**



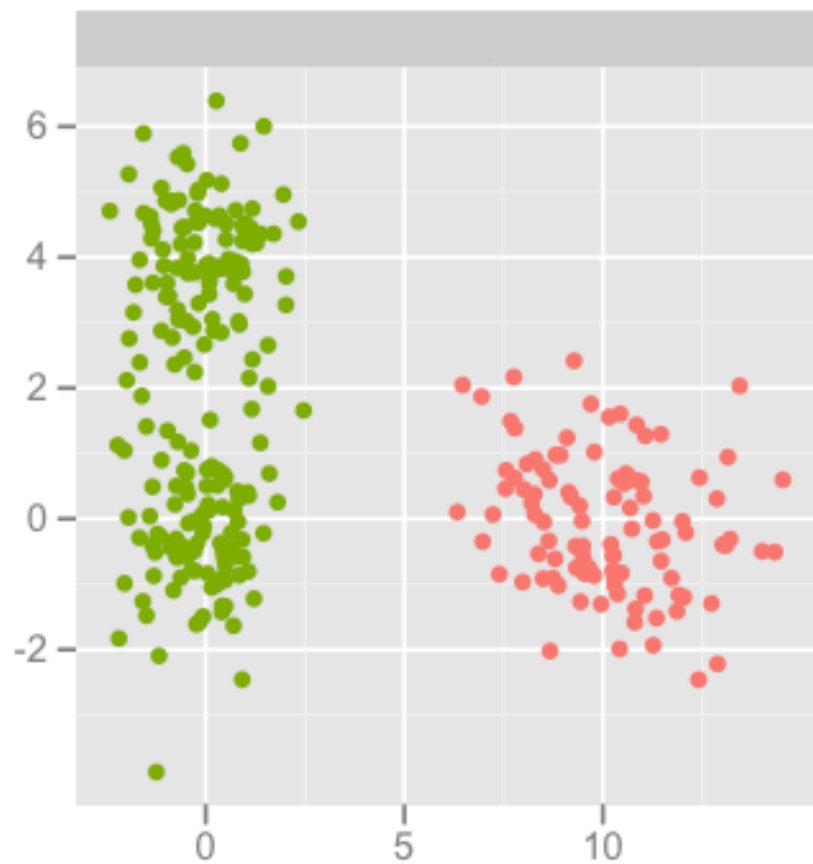
**% of Uninsured
who are Latino**

K Means



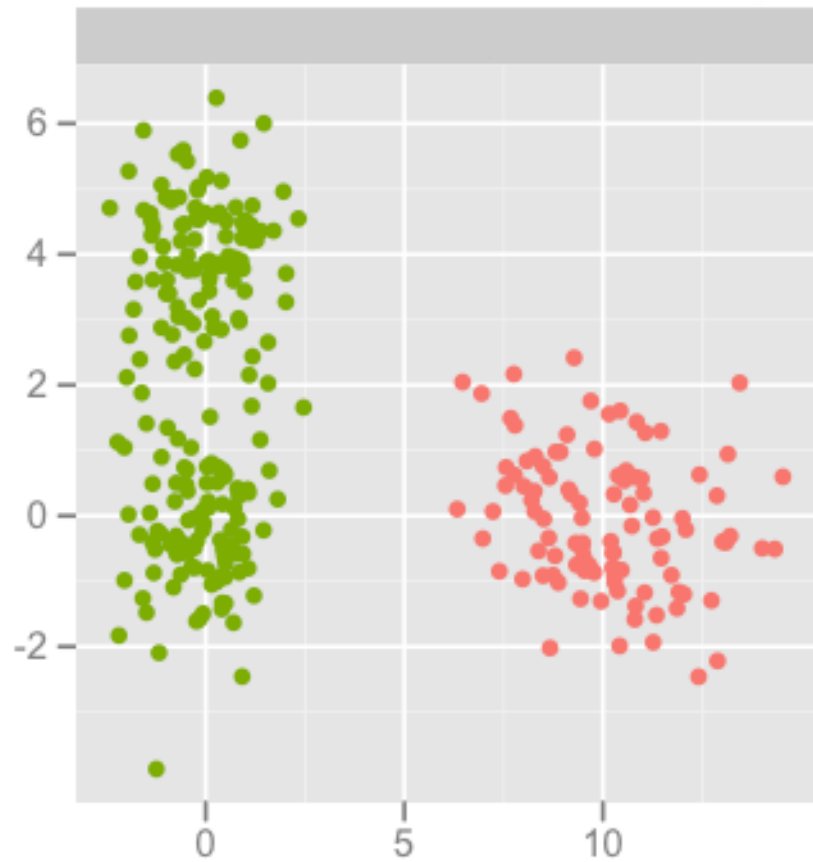
K Means

2 groups



K Means

2 groups

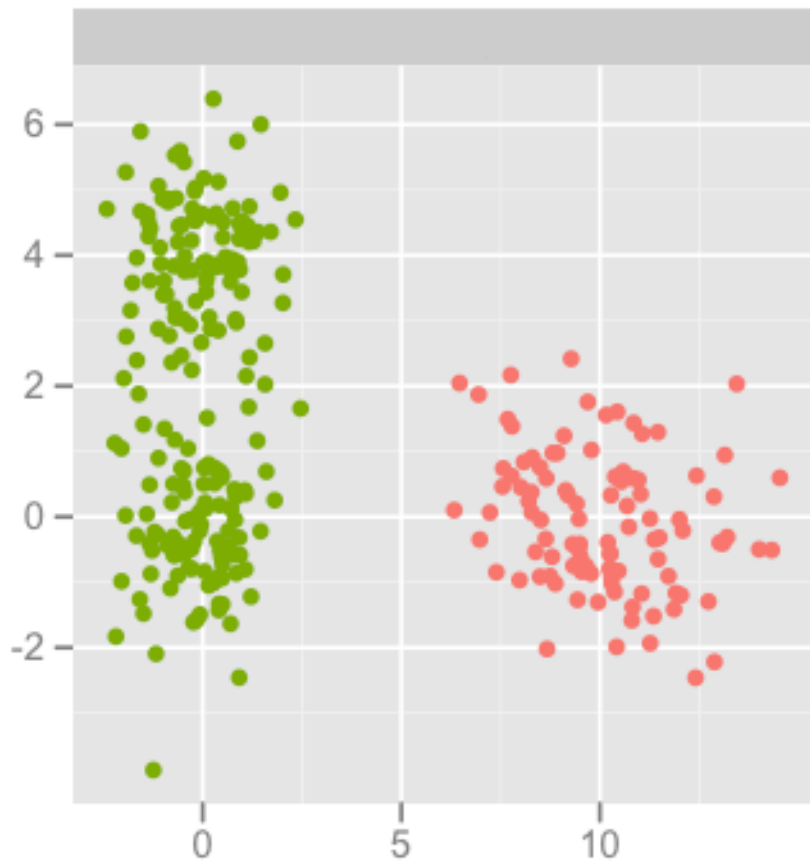


3 groups

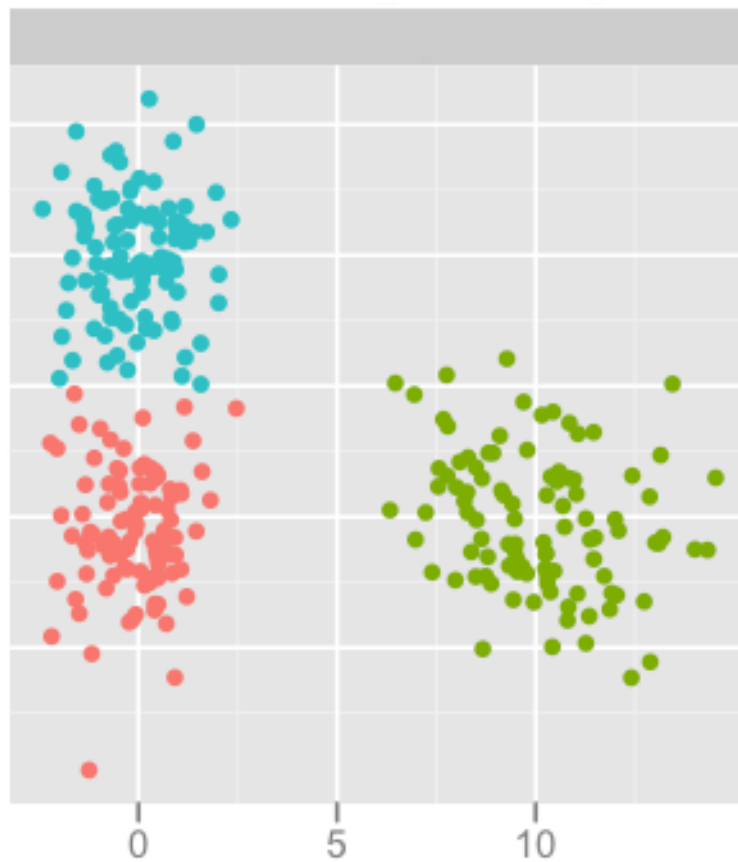


K Means

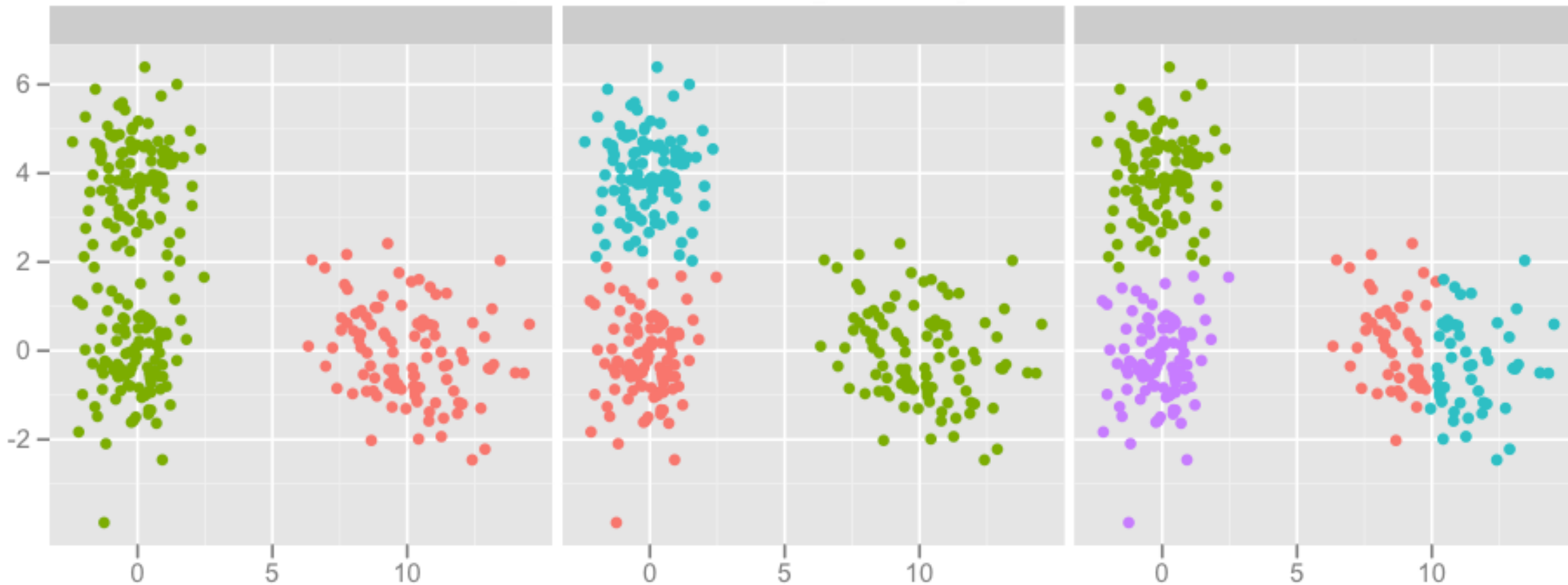
2 groups



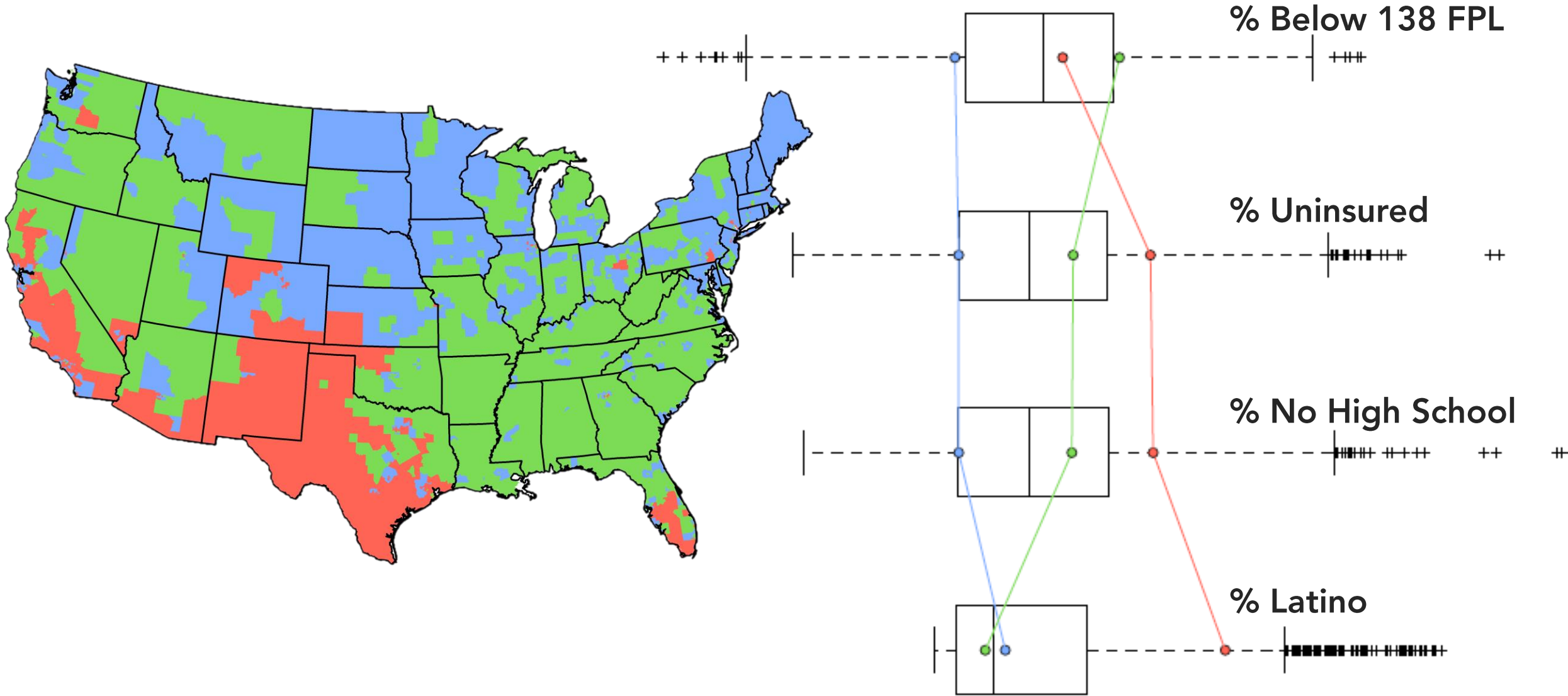
3 groups



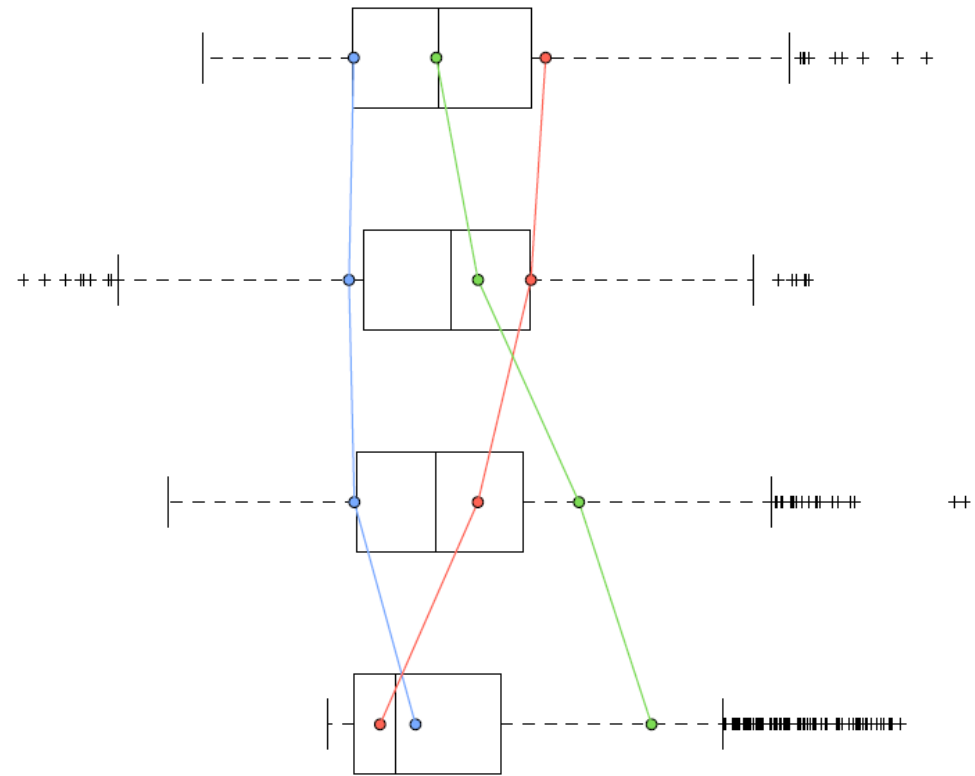
4 groups



Eligible Uninsured Americans



Demo



Spatially Constrained Multivariate Clustering

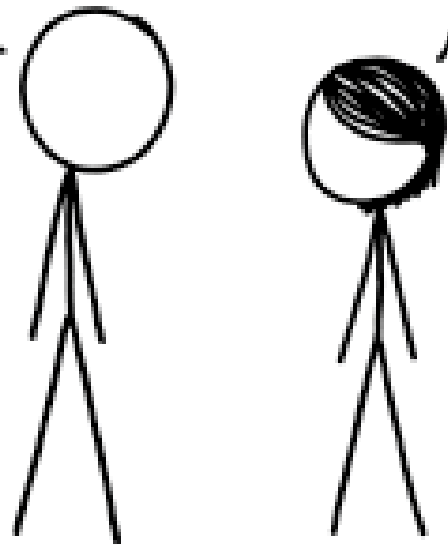
finds clusters based on feature attributes
and proximity

WHAT'VE YOU BEEN UP TO?

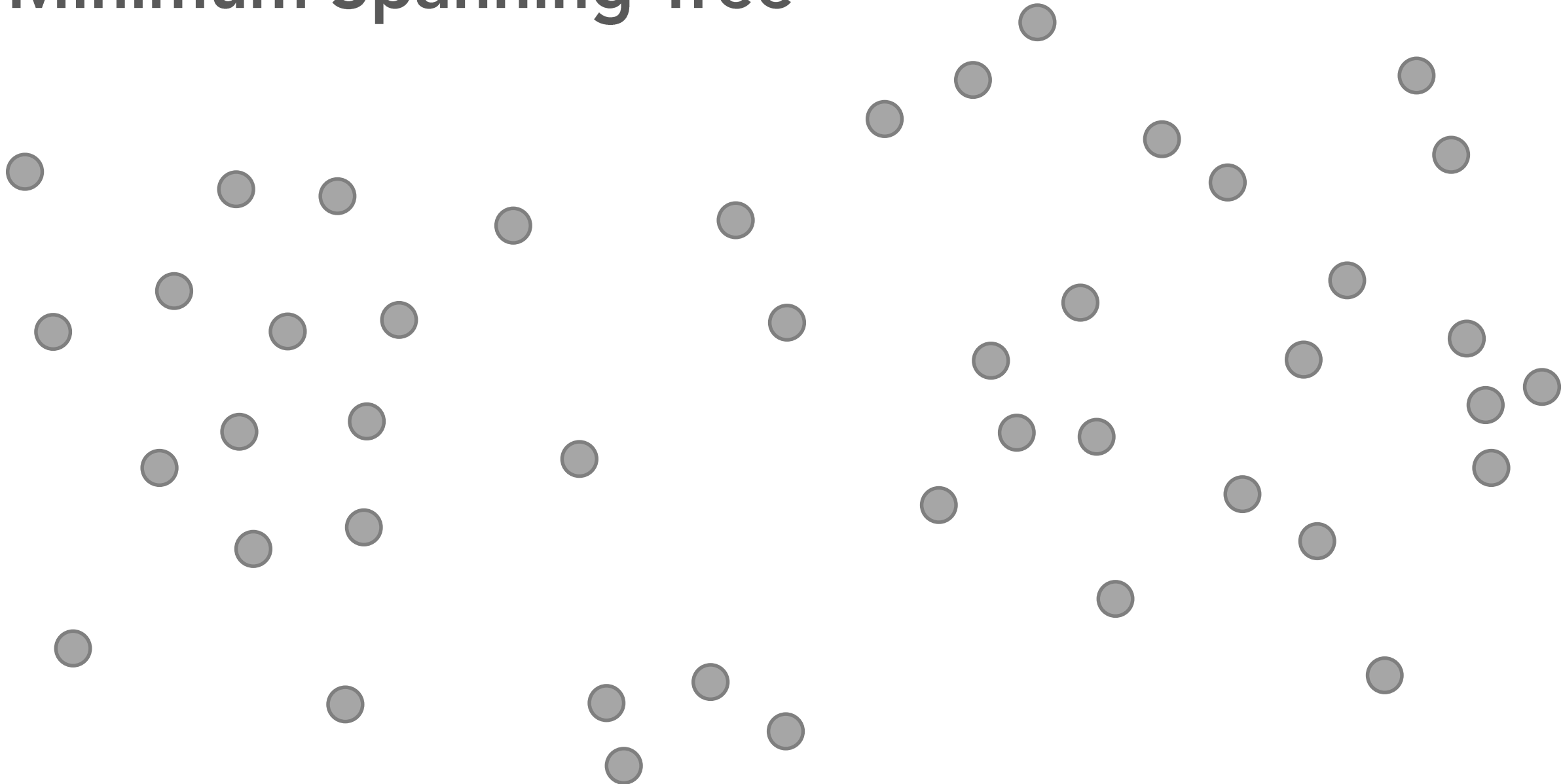
DOING TONS OF
MATH FOR MY THESIS.

CAN YOU EXPLAIN
IT LIKE I'M FIVE?

"OH MY GOD, WHERE
ARE YOUR PARENTS?"



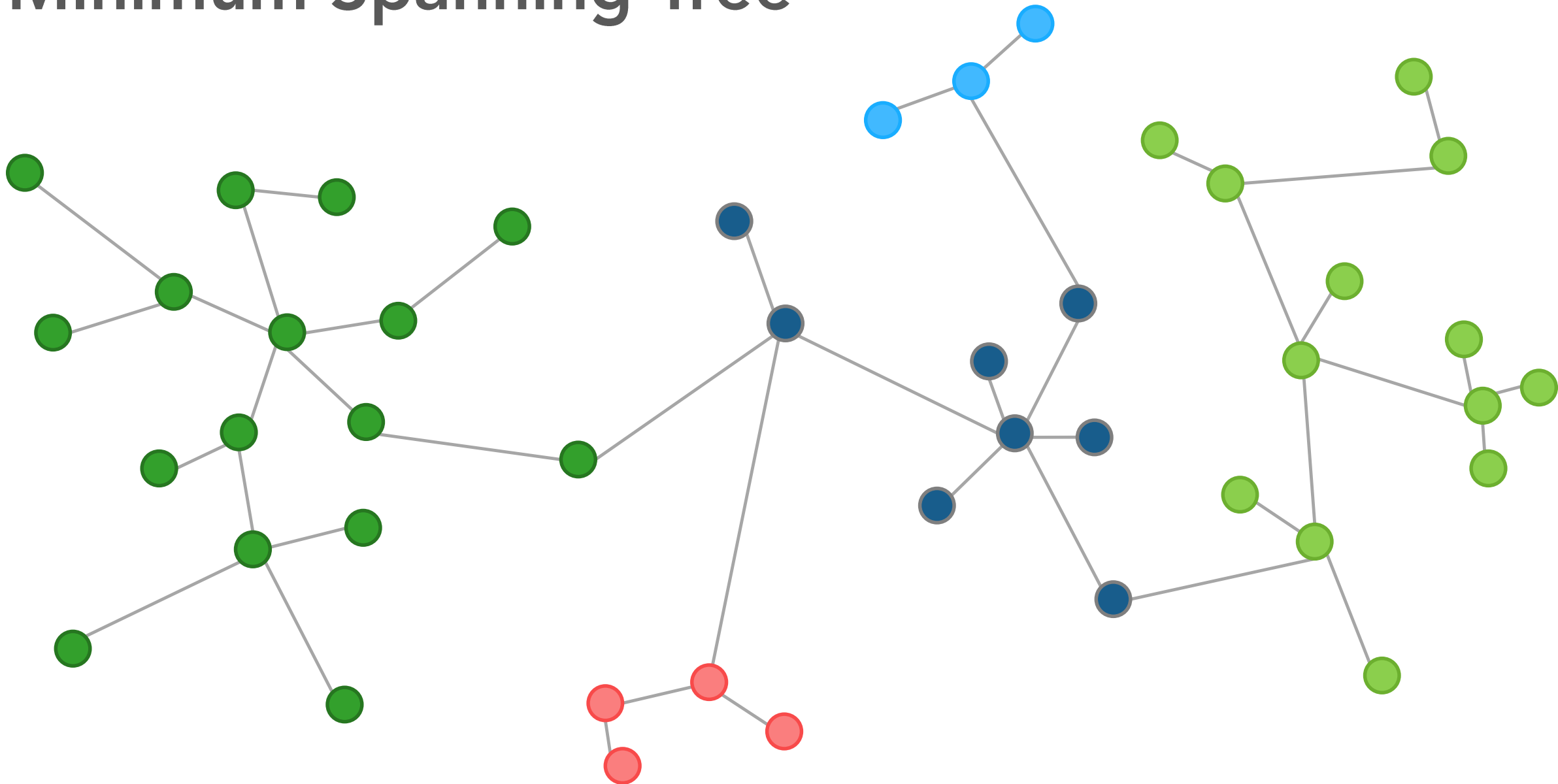
Minimum Spanning Tree



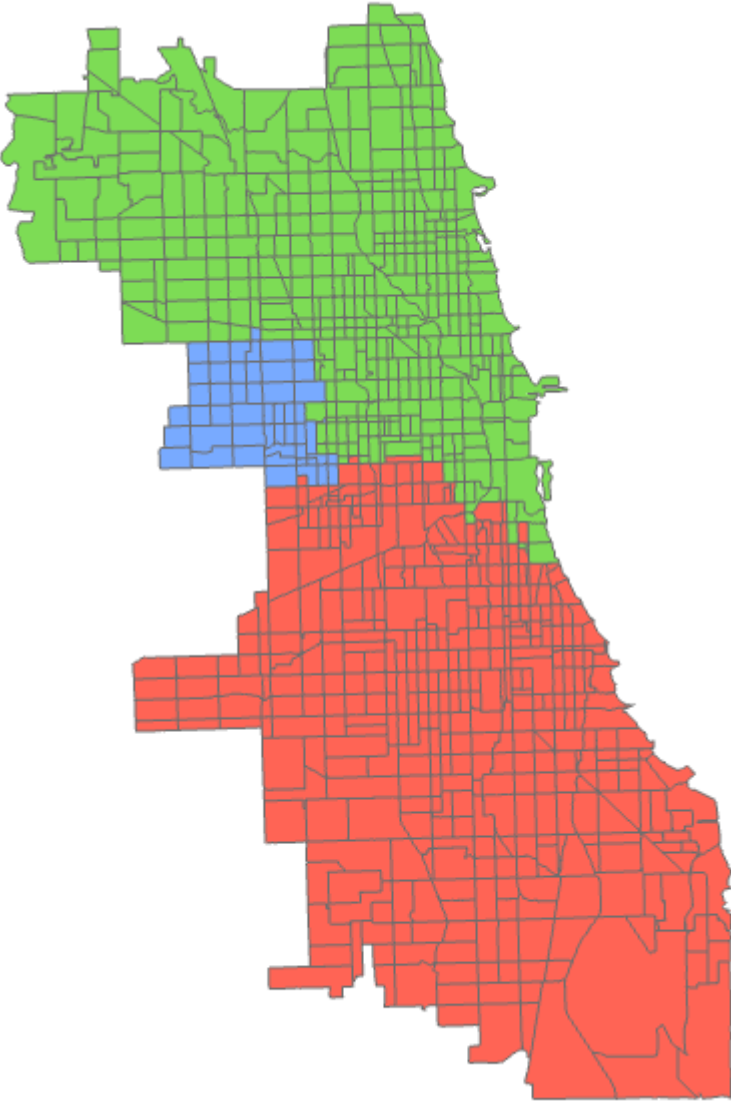
Minimum Spanning Tree



Minimum Spanning Tree



Crime in Chicago



Median Income



HS Dropout Rate



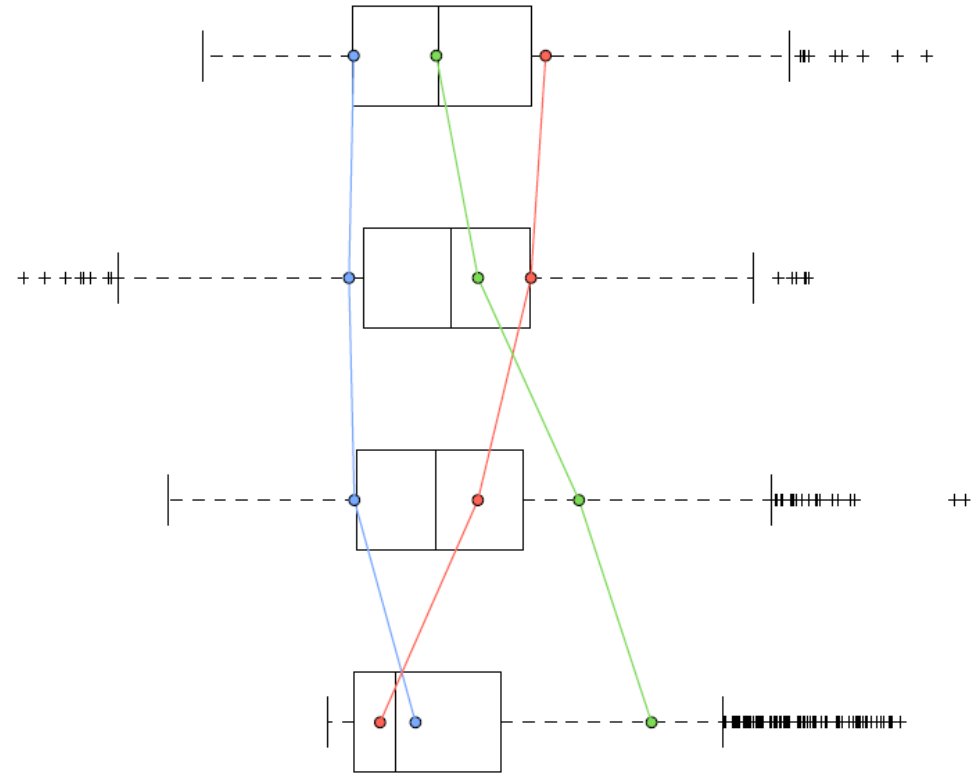
Unemployment



Crime Count



Demo





esri

THE
SCIENCE
OF
WHERE