

# Mutual $k$ -nearest neighbor graph for data analysis: Application to metric space clustering



**Proposal: Broaden the applicability of a theoretically sound and simple clustering algorithm**

Build the Mutual  $k$ -Nearest Neighbor graph of the data, for certain  $k$

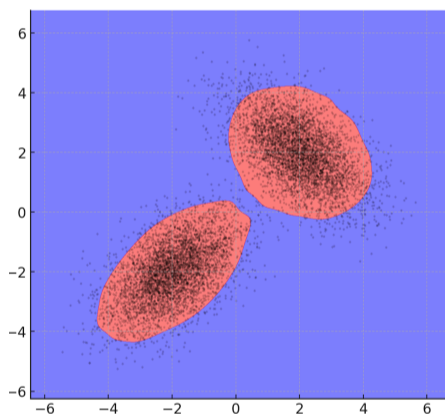
- ▶ Large connected components are clusters
- ▶ Small components are outliers



**Old algorithm requires a bounded-away from zero distribution**

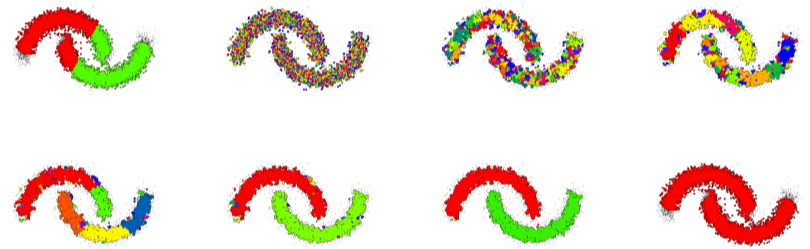
**Idea: Shave the distribution!**

- ▶ Filter out low density regions
- ▶ The remaining is bounded-away from zero by construction



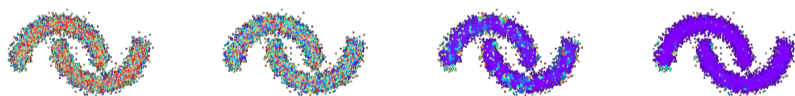
**The correct  $k$  can be found iteratively**

- ▶ We can use dendrograms



**Figure:** Two Moons partitions. [top left]  $k$ -means partition. [others] our procedure fixing  $\tau_{LOF} = 1.1$  and evolving  $k \in \{1, 4, 5, 6, 7, 10, 1000\}$  in reading order

**Without shaving**



**Figure:** Two Moons partitions without prior filtering and evolving  $k \in \{1, 2, 4, 5\}$  in reading order

**Compared to DBSCAN**



**Figure:** Adapted Two Moons dataset with varying densities. [left]  $k$ -means partition. [center] DBSCAN partition. [right]  $m$ -based partition

**Take out**

- ▶ The filtering step, to be useful, should retain most data from the original dataset
- ▶ Finding the proper filtering value is a problem in itself, it should be related to *fords* detection