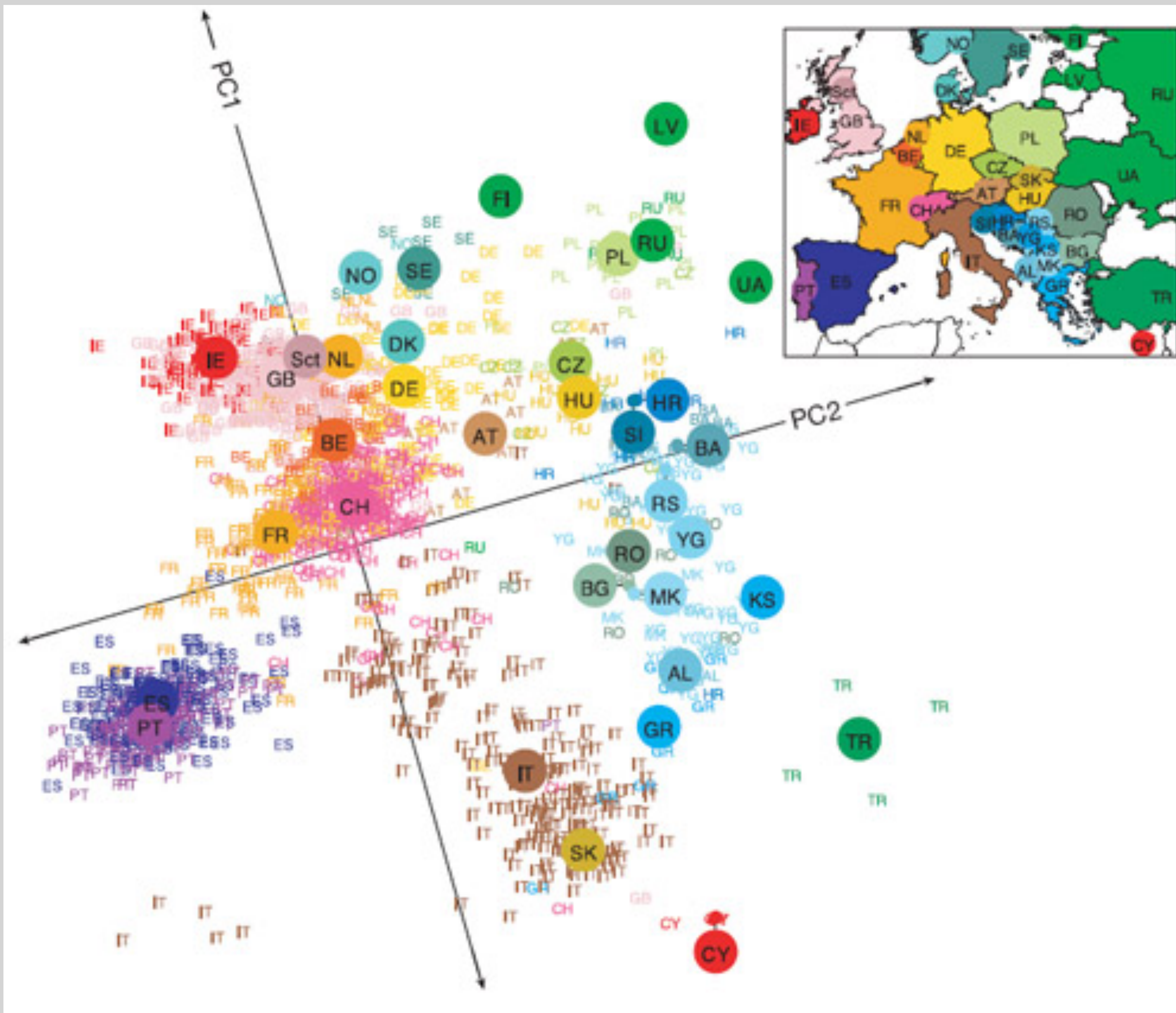


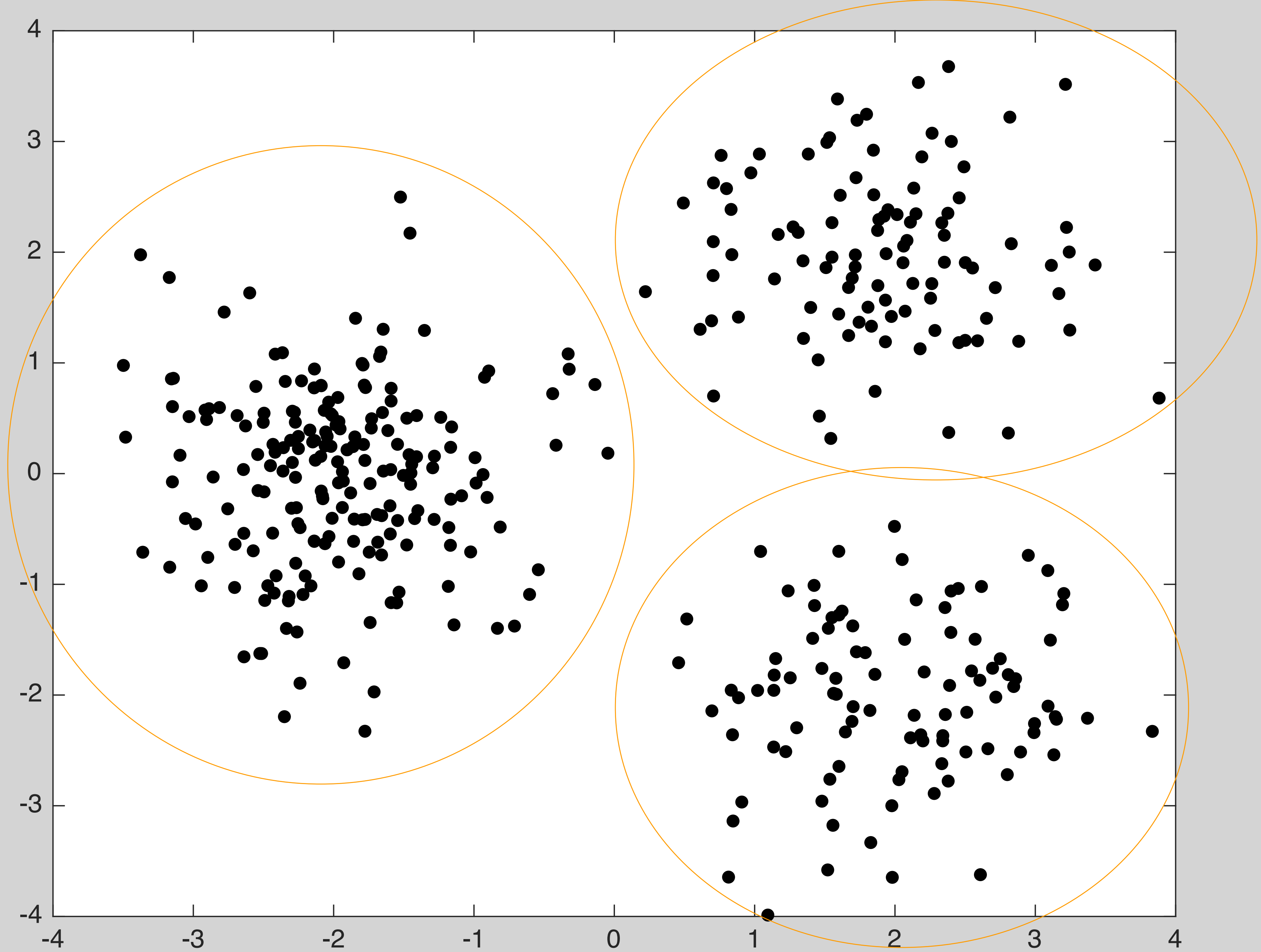
EE16B

Designing Information Devices and Systems II

Lecture 10A
k-means Clustering







k-means

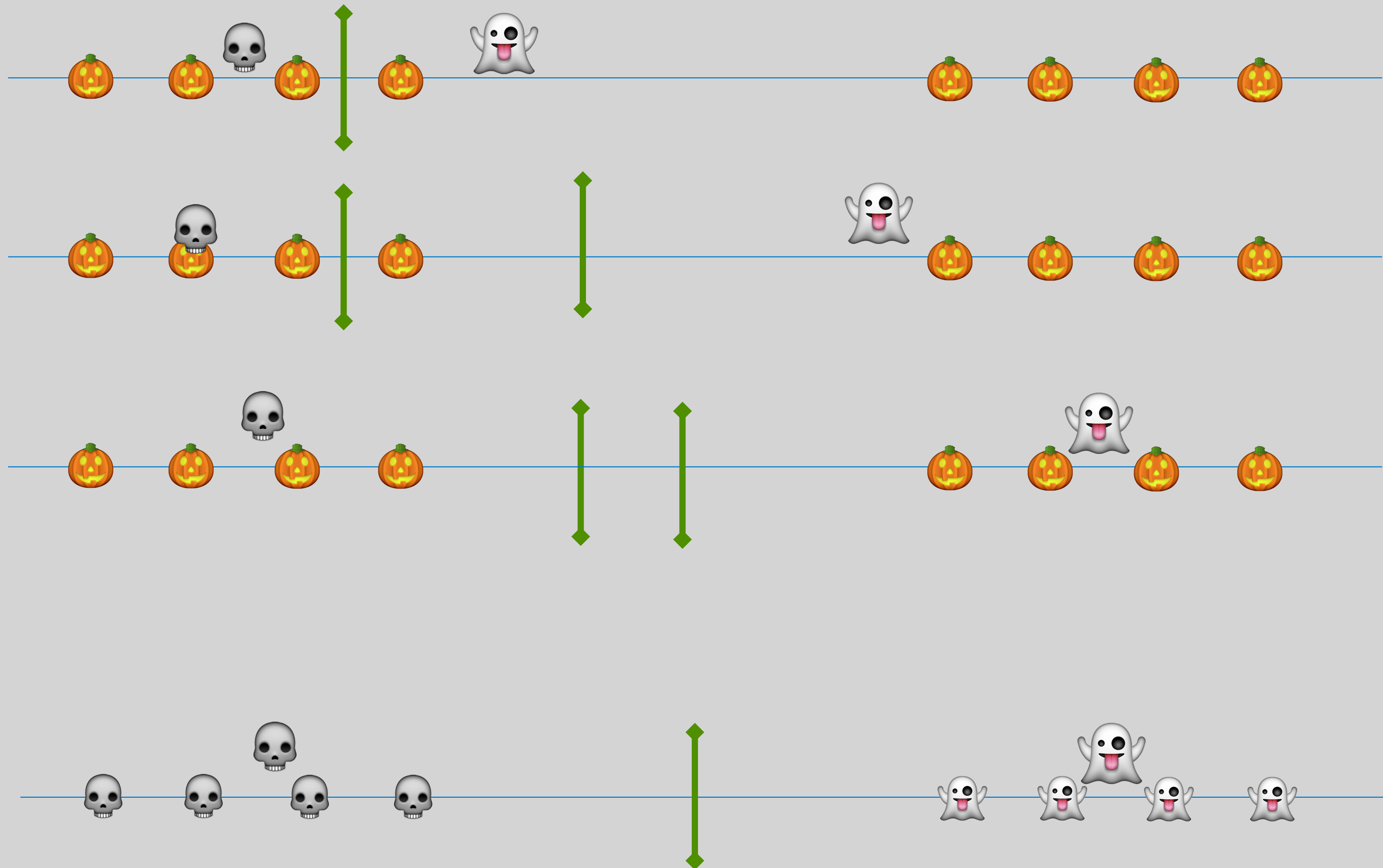
Given: $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_m \in \mathbb{R}^n$

Partition them into $k \ll m$ groups

- 0) Guess cluster centers to initialize
- 1) Group points around nearest center
- 2) Update cluster centers by averaging within group
- 3) If centers have changed, repeat 1-3

k-means 1D example

$$n = 1, m = 8, k = 2$$



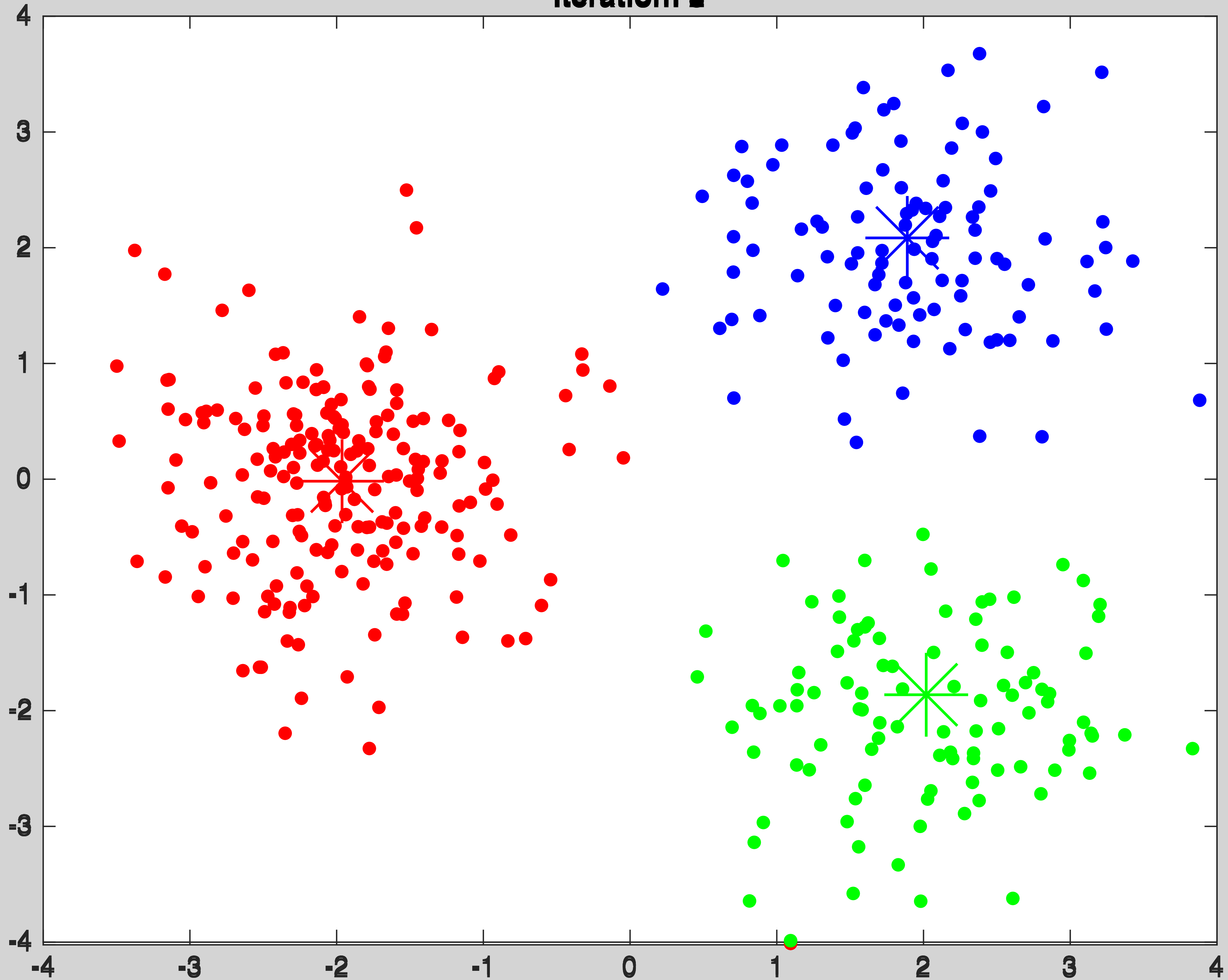
General k-means Algorithm

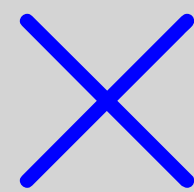
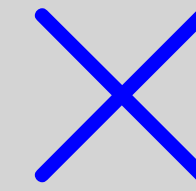
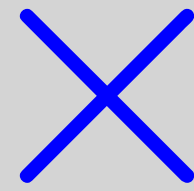
- 0) Initialize k cluster centers $\vec{m}_1, \vec{m}_2, \dots, \vec{m}_k$
- 1) Assign points to cluster: point \vec{x} goes to cluster i if,
- $$\|\vec{x} - \vec{m}_i\| < \|\vec{x} - \vec{m}_j\| \quad \forall j \neq i$$
- 2) Let S_i be the set of samples in cluster i
recompute cluster centers:

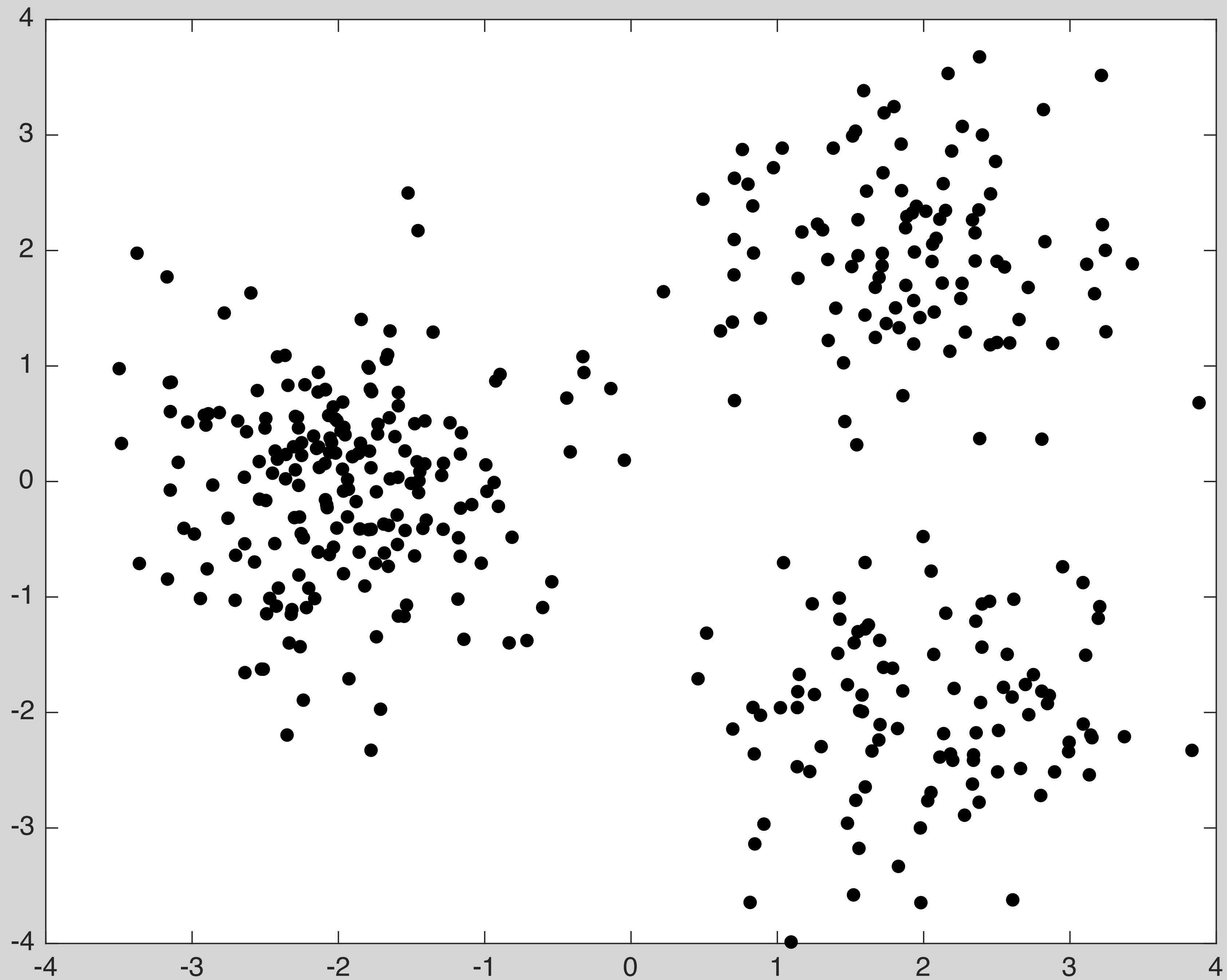
$$\vec{m}_i = \frac{1}{|S_i|} \sum_{\vec{x} \in S_i} \vec{x}$$

- 3) If any m_i has changed, repeat 1-3

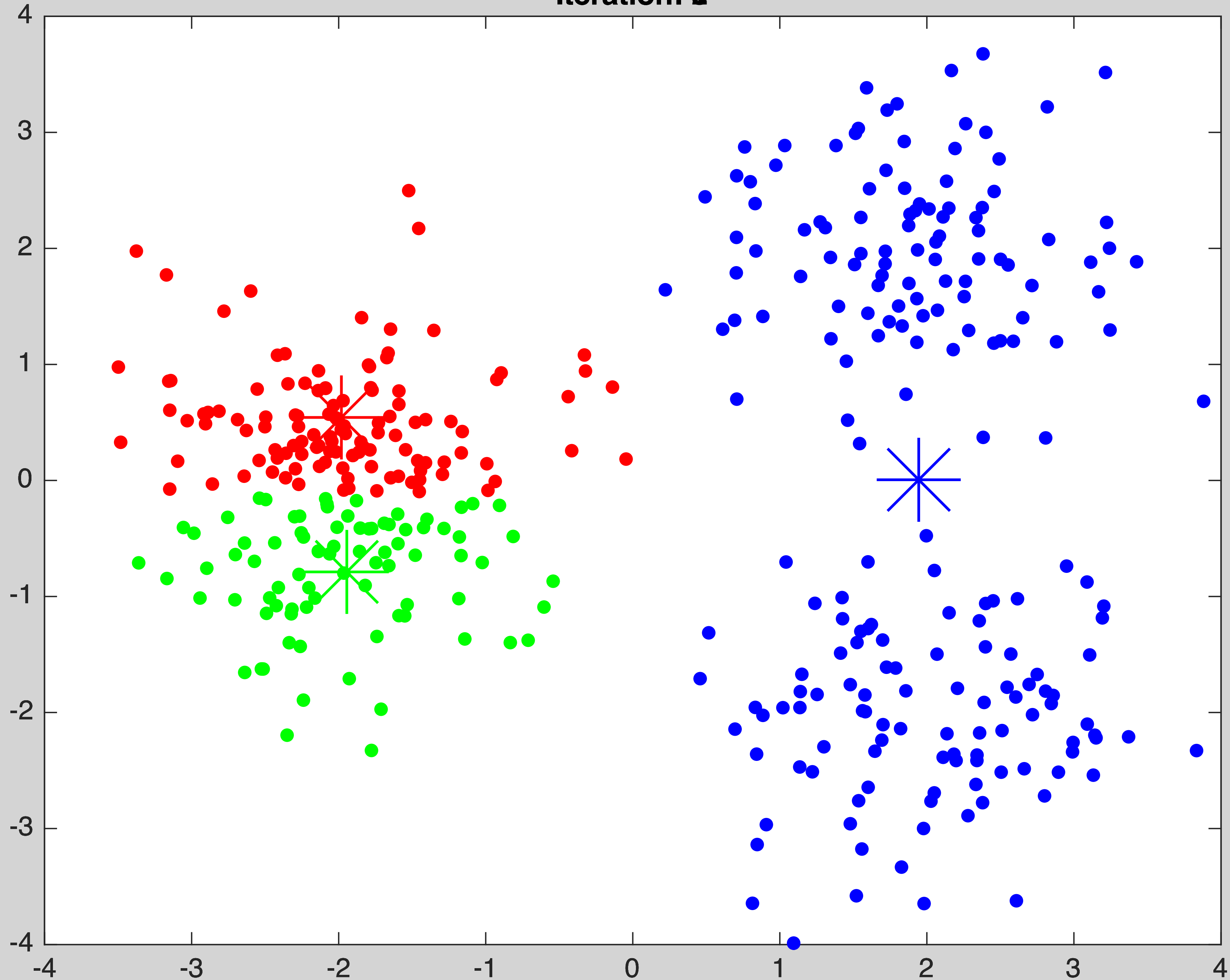
Iteration: 4







Iteration: 2



Objective Function

Find the clustering of $\vec{x}_1, \dots, \vec{x}_m$ into sets S_1, \dots, S_k which minimizes:

$$D = \sum_{i=1}^k \sum_{\vec{x} \in S_i} \|\vec{x} - \mu_i\|$$

$$\mu_i = \frac{1}{|S_i|} \sum_{x \in S_i} \vec{x}$$

While the algorithm decreases the objective, the objective is non-convex and can be stuck on local minima.

General problem is N-P Complete

Management of intersections with multi-modal high-resolution data ☆☆☆



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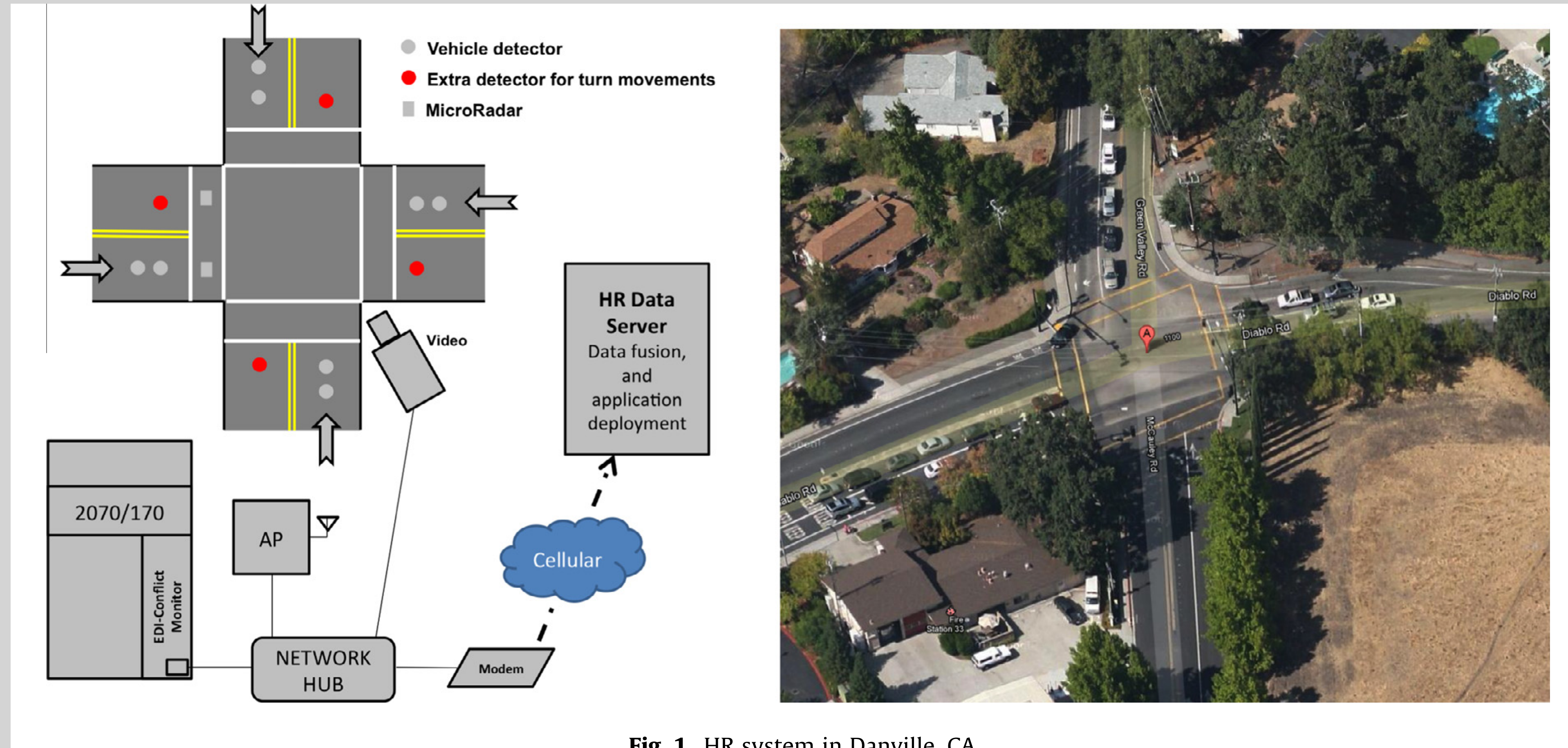
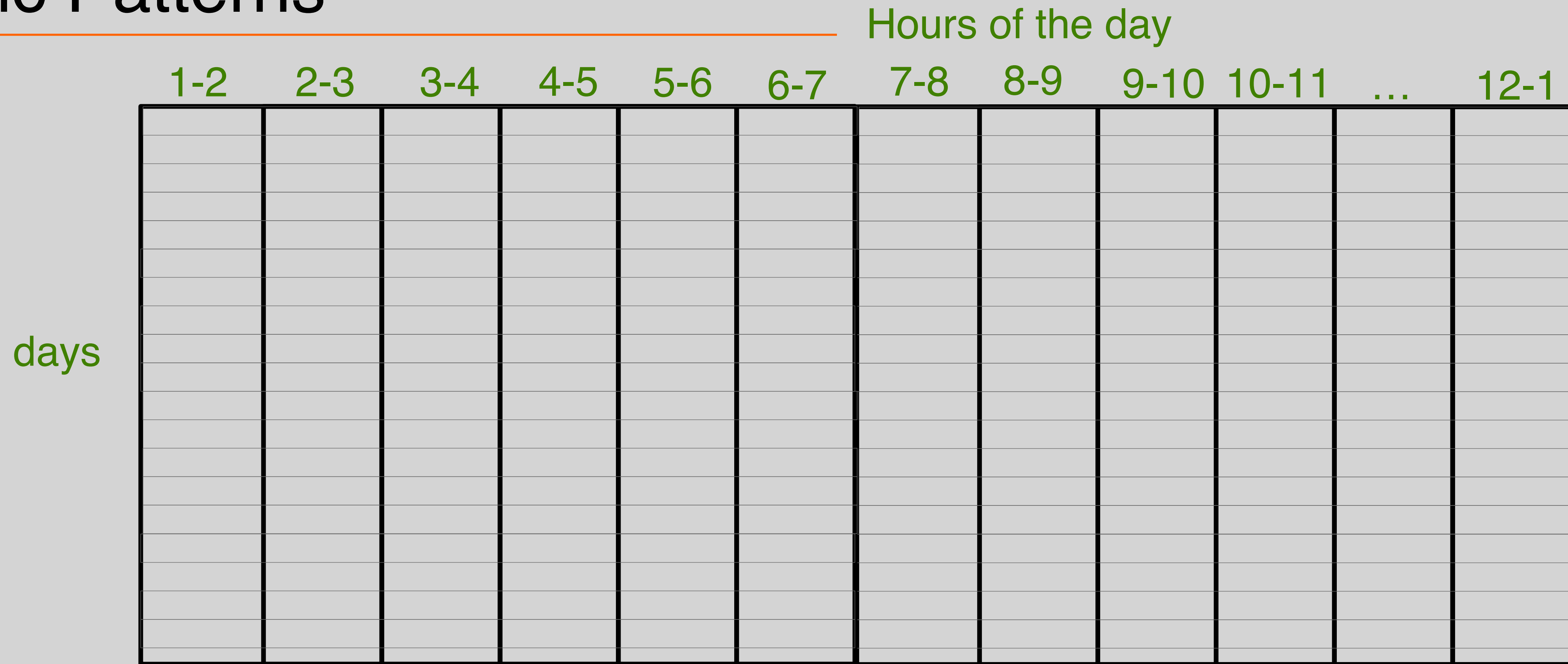


Fig. 1. HR system in Danville, CA.

Traffic Patterns



What would k-means cluster to?

$K = 2?$

$K = 4?$

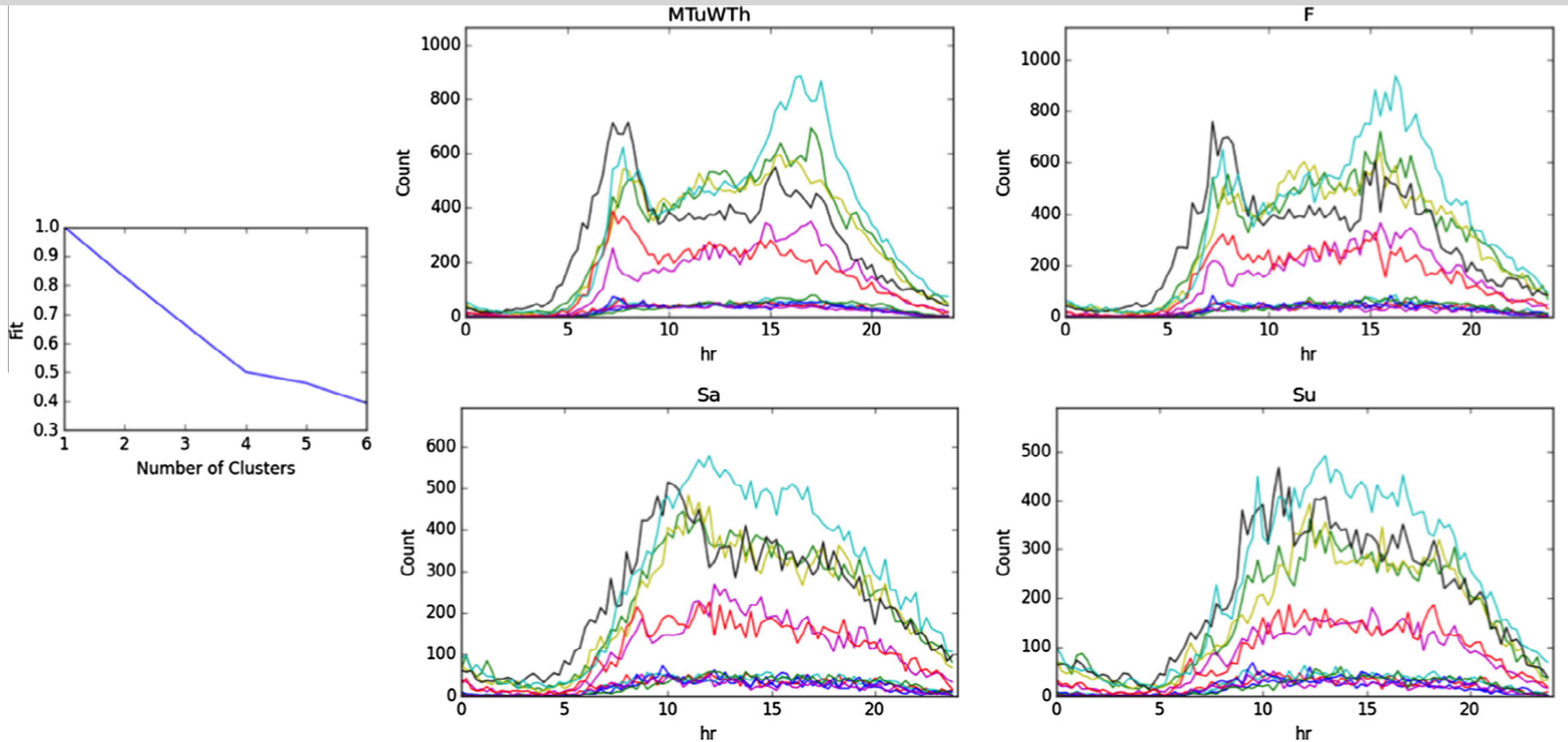


Fig. 5. Clustering of daily data for Dec 2014 to May 2015 in an intersection in Beaufort, SC.

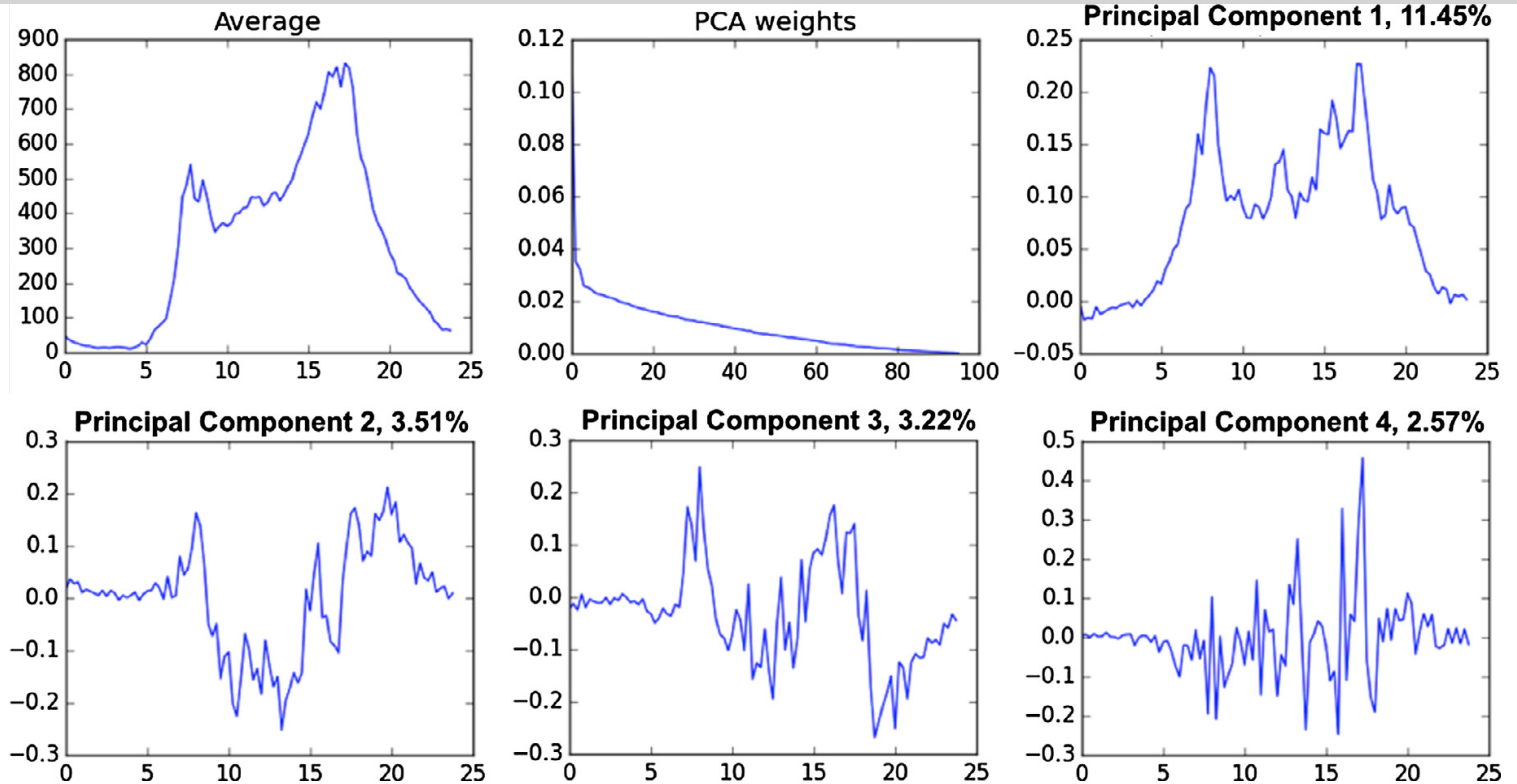


Fig. 8. Four PCA components of the North–South through movement and the average. The x -axis is hours.

