

Lecture 17: Clustering and density estimation

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Clustering using k-means

Clustering

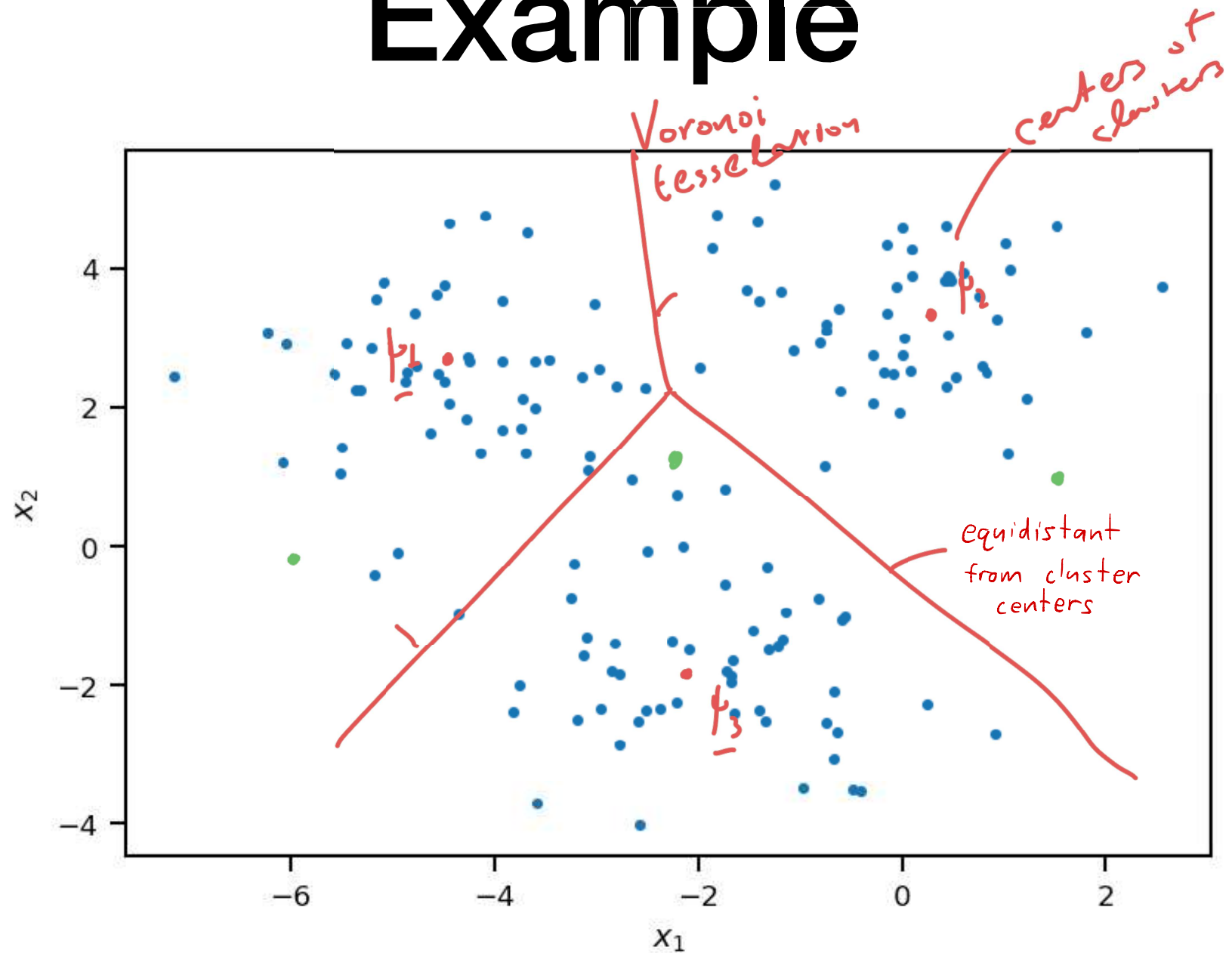
You are given n observations:

$$\mathbf{x}_{1:n} = \{\mathbf{x}_1, \dots, \mathbf{x}_n\}$$

(inputs, features, ...)

Problem: Separate the data into K groups? How many such groups exist?

Example



K-means objective

must specify

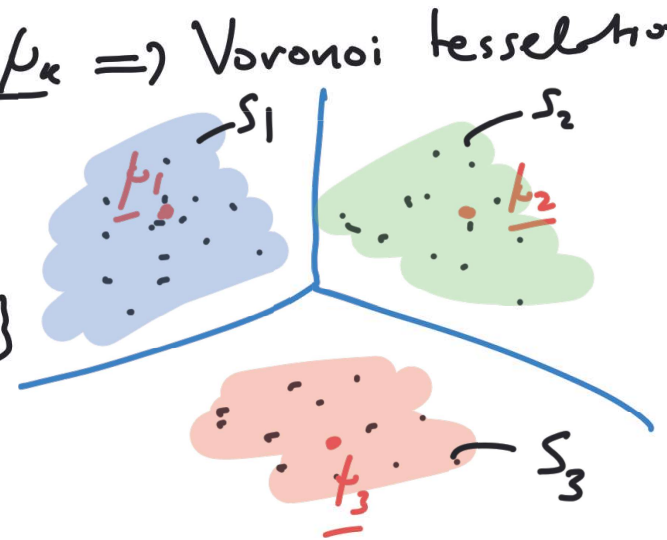
K clusters w/ centers $\underline{\mu}_1, \dots, \underline{\mu}_K \Rightarrow$ Voronoi tessellation

$$\underline{x}_{1:n} = (\underline{x}_1, \underline{x}_2, \dots, \underline{x}_n)$$

$$S_1 \subset \{\underline{x}_1, \dots, \underline{x}_n\}, S_2 \subset \{\underline{x}_1, \dots, \underline{x}_n\}$$

$$S_3, \dots, S_K$$

$$\min_{\underline{\mu}_1, \dots, \underline{\mu}_K} \sum_{i=1}^K \sum_{\underline{x} \in S_i} \|\underline{\mu}_i - \underline{x}\|^2$$



Standard k-means algorithm

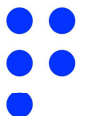
$$\min_{\mu_1, \dots, \mu_k} \sum_{i=1}^k \sum_{x \in S_i} \|x - \mu_i\|^2$$

1. Start by randomly choosing $\mu_1^{(1)}, \dots, \mu_k^{(1)}$. $t \leftarrow 1$.
2. Assignment step: $S_1^{(t)}, \dots, S_k^{(t)}$

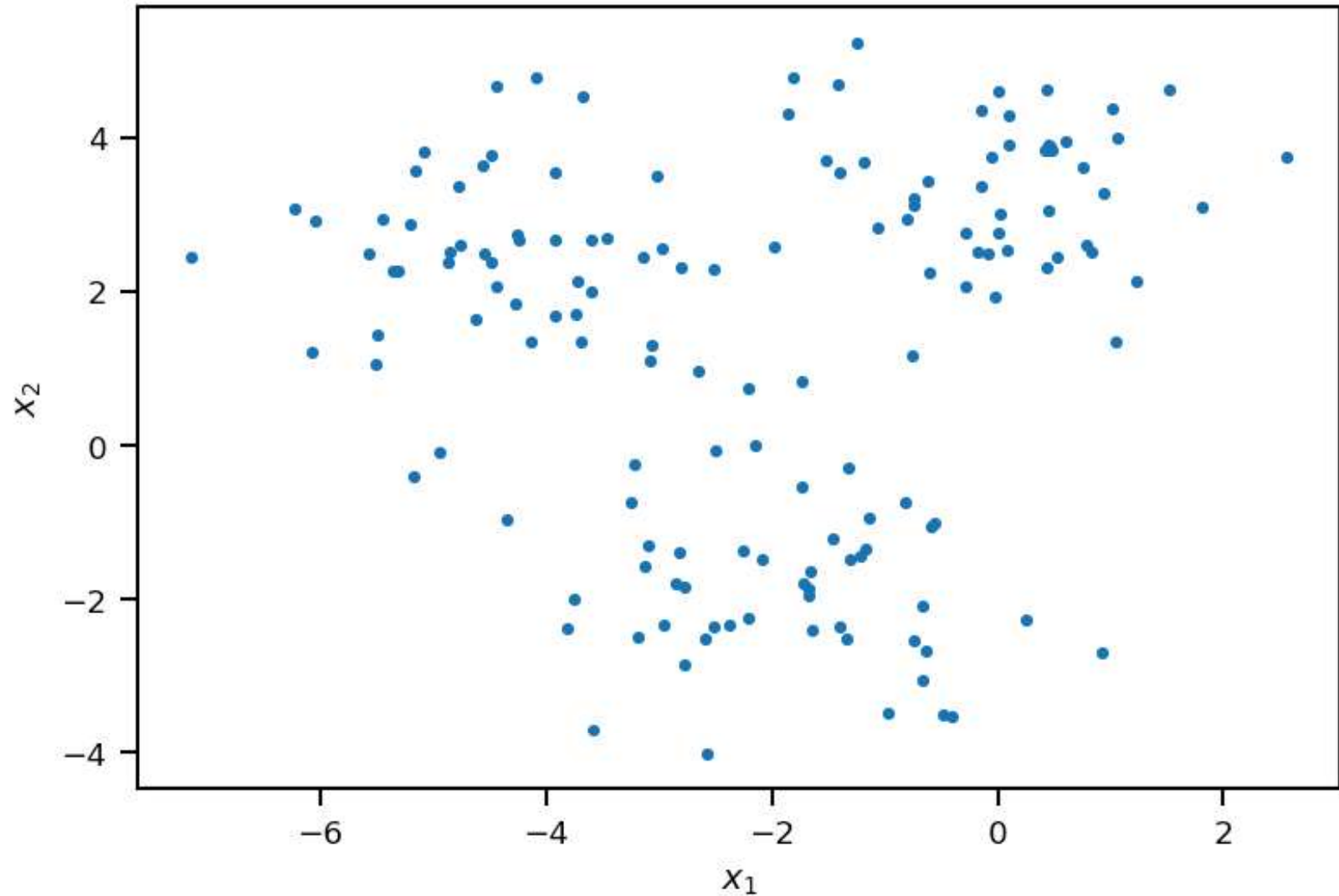
$$S_i^{(t)} = \{x_j \in \{x_1, \dots, x_n\} : \|x_j - \mu_i^{(t)}\| \leq \|x_j - \mu_r^{(t)}\|, r \neq i\}$$
3. Update step:

$$\mu_i^{(t+1)} = \frac{1}{|S_i^{(t)}|} \cdot \sum_{x \in S_i^{(t)}} x$$

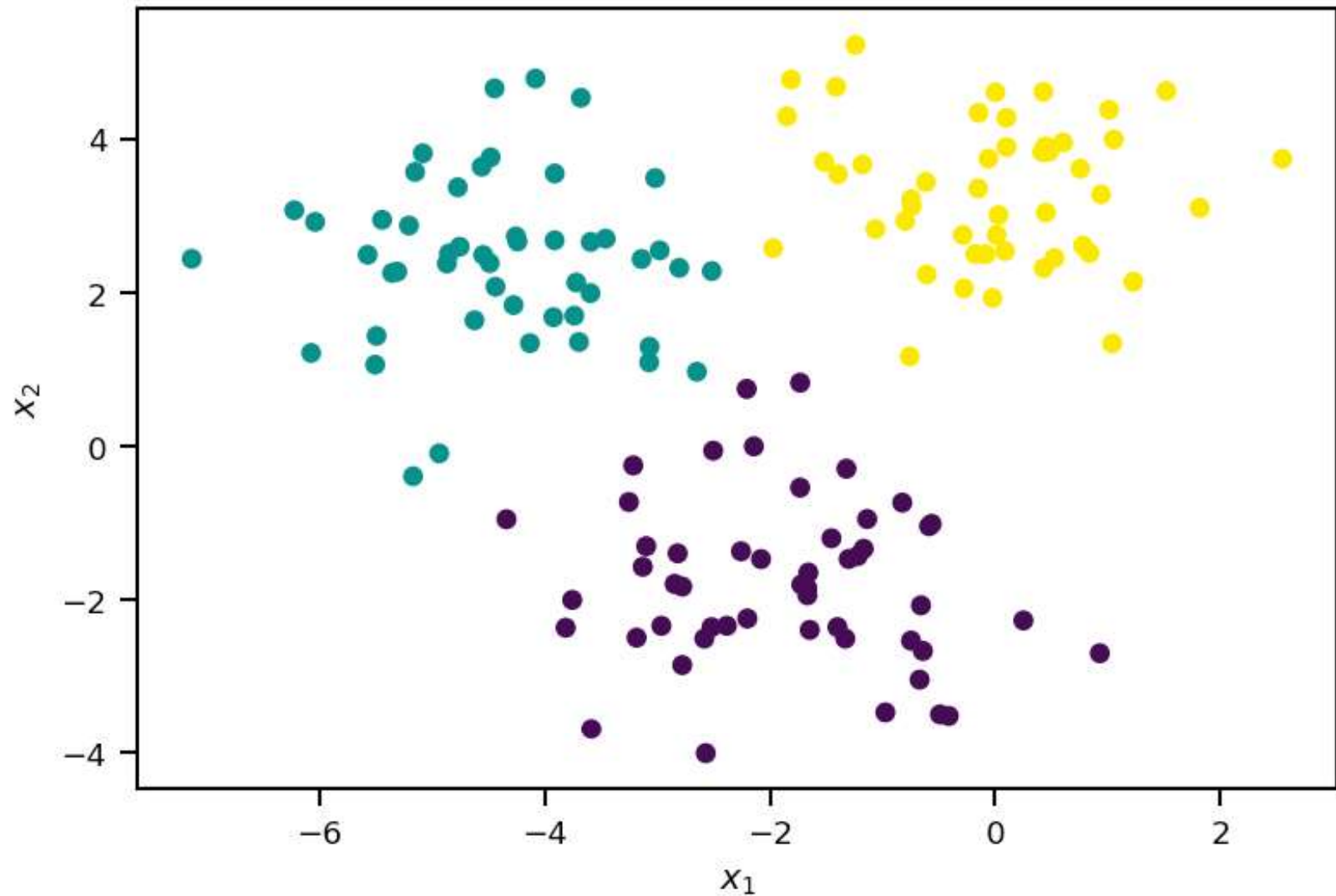
points in cluster empirical mean
4. Convergence test: If $\|\mu_i^{(t+1)} - \mu_i^{(t)}\| < \epsilon \forall i=1, \dots, k$, then STOP. Otherwise GOTO 2.
threshold



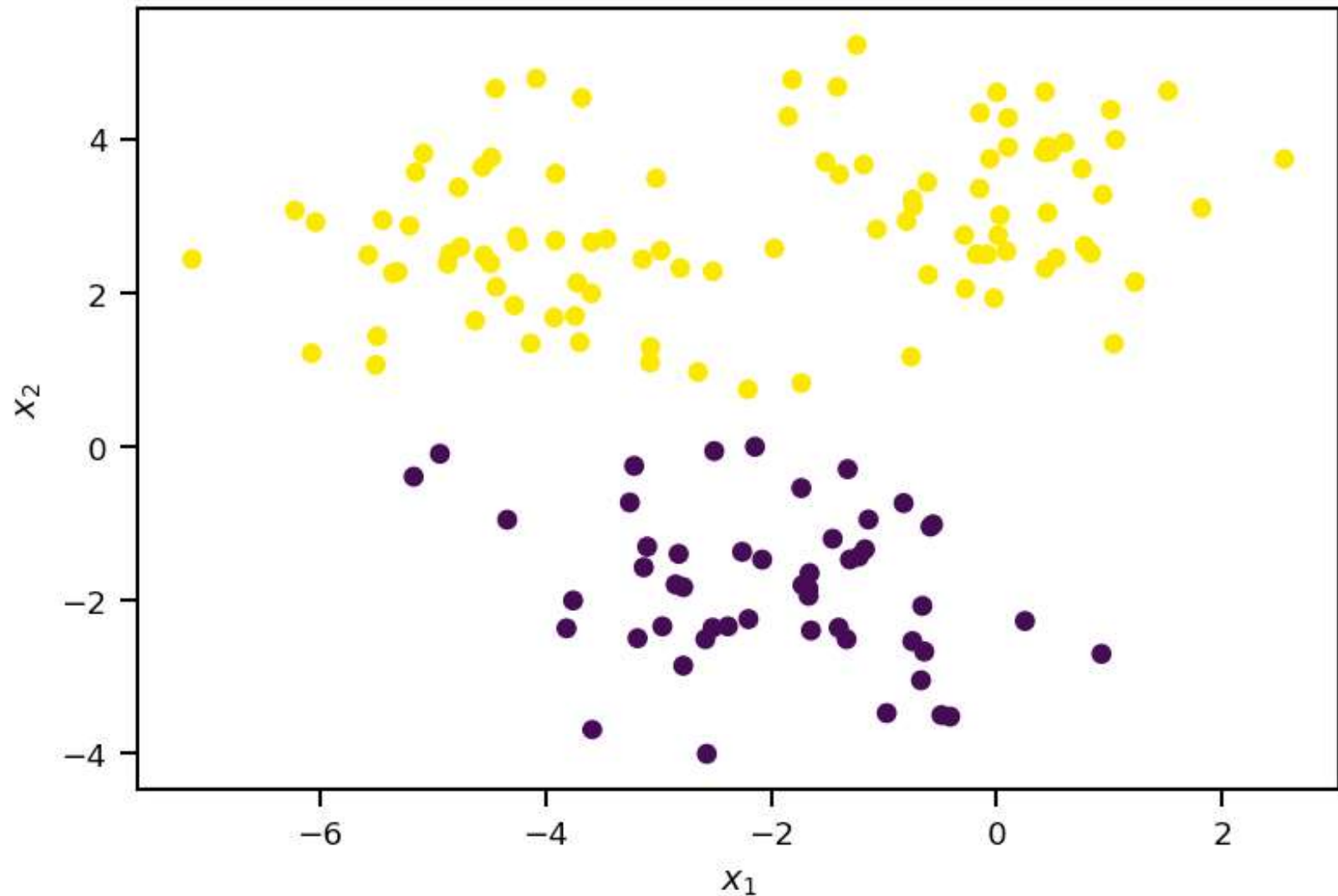
Example



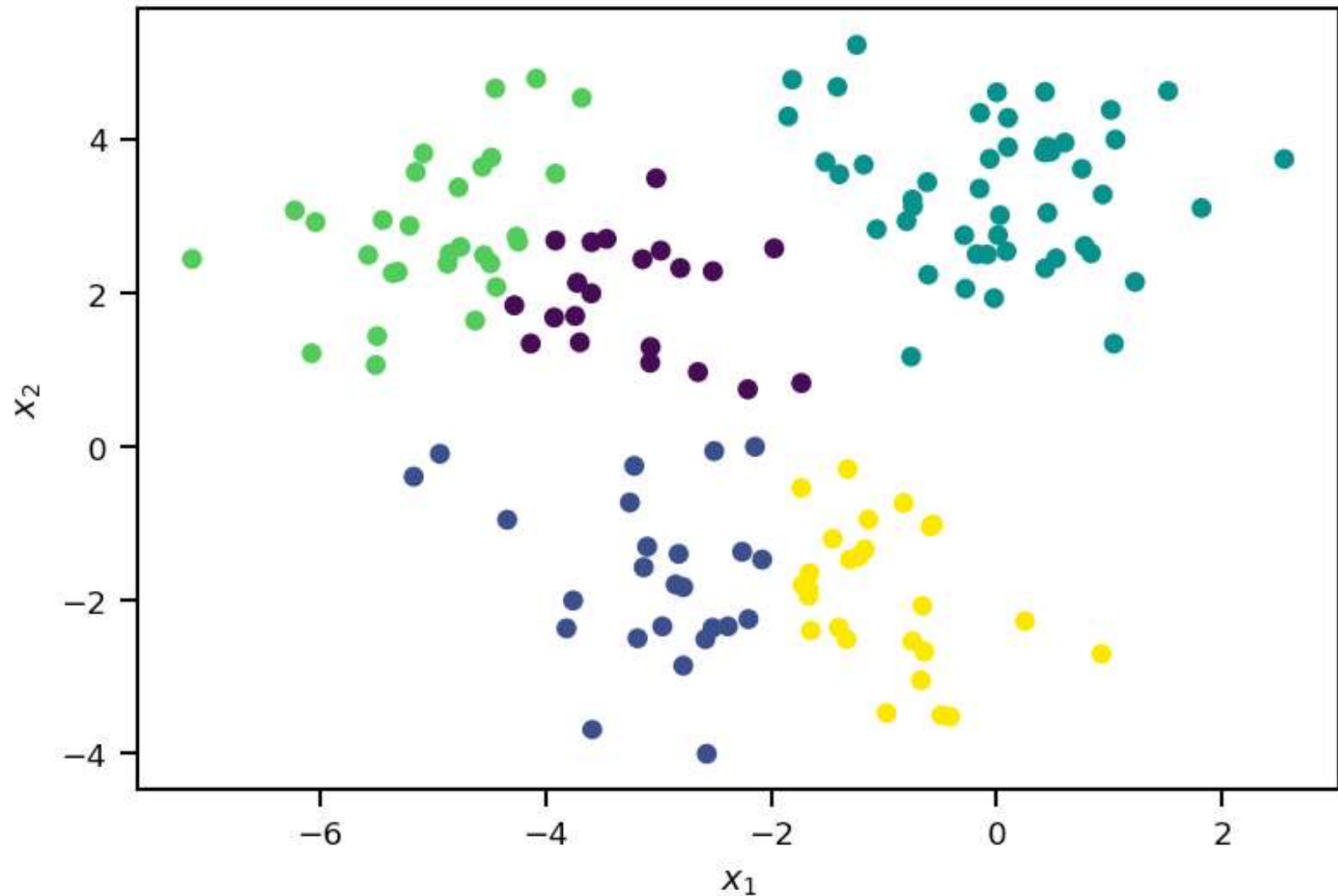
Example



What if I used two clusters?



What if I used five clusters?



Limitations of k-means

- How many clusters?
- Assumes spherical clusters.
- Cannot be applied to high-dimensional datasets, e.g., images.

Beyond k-means

- Clustering is related to density estimation.
- Idea:
 - Make hypothesis about how data are generated.
 - Train your model.
 - Let the structure arise naturally.