



Unsupervised Learning & Data Clustering



Unsupervised Learning – Part 1: Problem Set-up

Objective



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Define the set-up of
unsupervised learning

Learning from Unlabeled Data

- | Given a training set of n **unlabeled** samples $\{\mathbf{x}^{(i)}\}$
- | What can we learn from the samples?
- We could estimate the overall distribution of the data without knowing their label.
- We could figure out the groupings of the samples (if any).
- We could identify some features that may be more important than others.

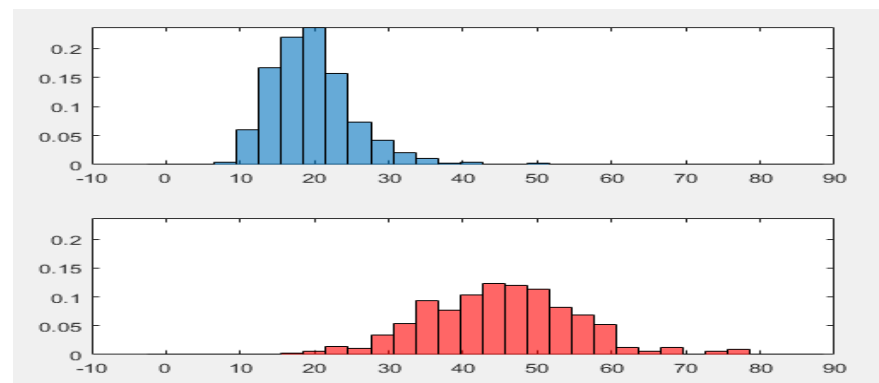
...

An Example

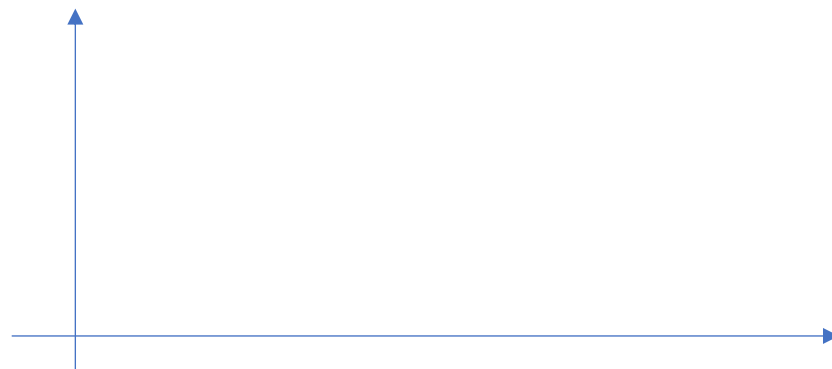
Illustrating structures/groupings of unlabeled samples may relate to the (unknown) labels of the samples



→ If we know the labels, we may find the densities of the classes →

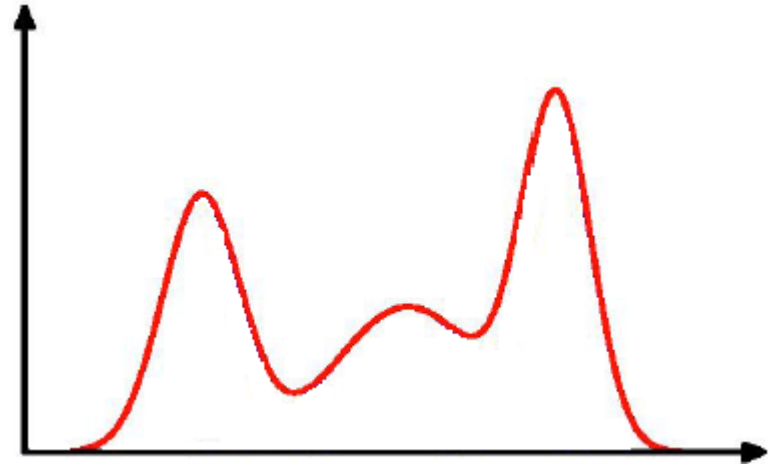


→ What may we see if we have no label for the data samples?



Another Example

| A density estimated from unlabeled samples may help us to identify densities of different classes



| If we know there are three classes in the data, each having a normal distribution ...

A Mixture-Density Model

| Assume a parametric model like this:

- The samples come from C classes.
- The prior probabilities $P(\omega_j)$ for each class are known, for $j = 1, \dots, C$.
- The form of $p(\mathbf{x} \mid \omega_j, \theta_j)$ ($j = 1, \dots, C$) are known.
- The C parameter vectors $\theta_1, \theta_2, \dots, \theta_C$ are unknown.

| Samples from this distribution are given, but the labels of the training samples are *unknown*.

A Mixture-Density Model (cont'd)

| What is the PDF of the unlabeled samples?

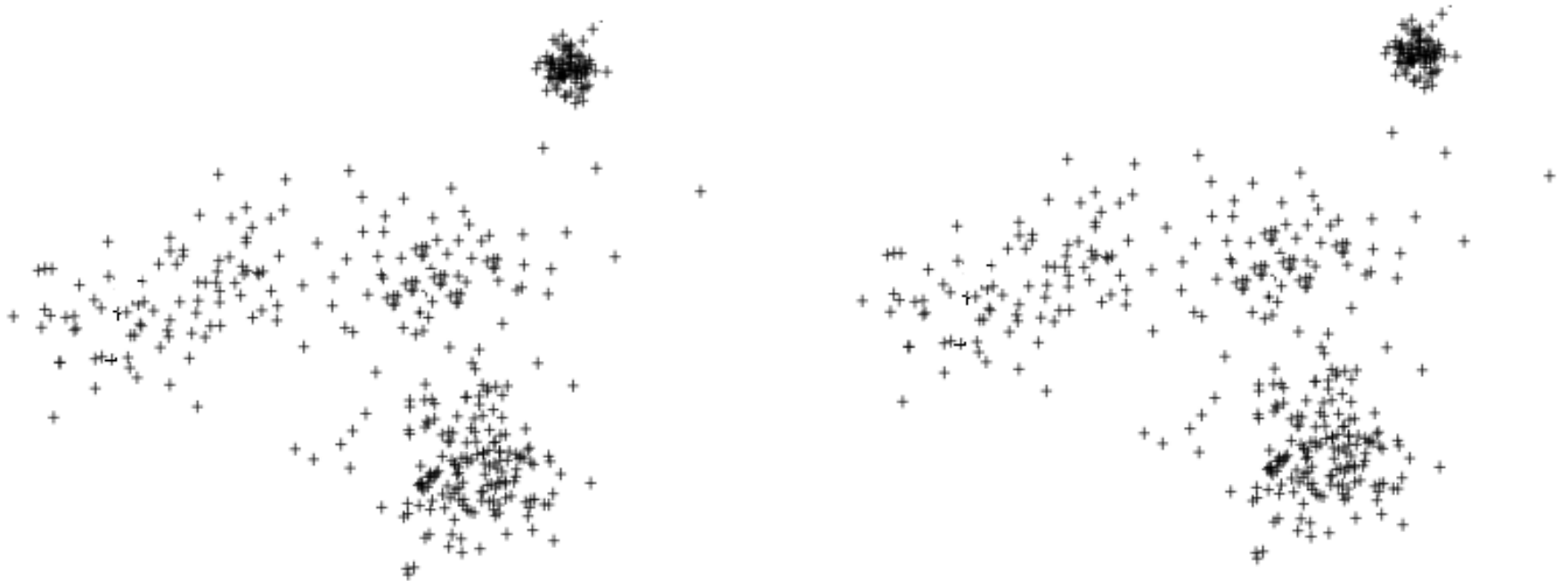
$$p(\mathbf{x} | \boldsymbol{\theta}) = \sum_{j=1}^c p(\mathbf{x} | \omega_j, \boldsymbol{\theta}_j) P(\omega_j)$$

where $\boldsymbol{\theta} = (\theta_1, \theta_2, \dots, \theta_c)$

| Can we learn $\boldsymbol{\theta}$ from unlabeled samples from this **mixture density**?

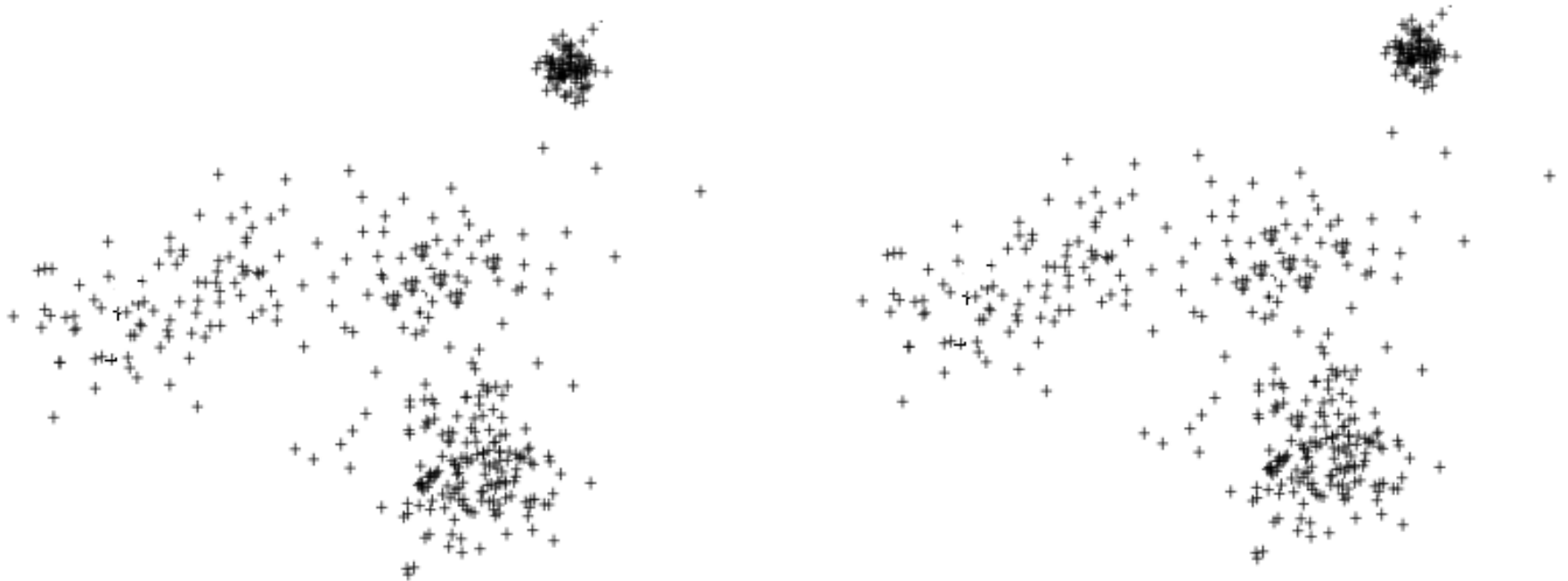
Illustrating Mixture-Density Model

| An example: with the assumption of 4 classes



Illustrating Mixture-Density Model

| An example: with the assumption of 2 classes



The Question of Identifiability

| Can we learn a unique θ from a set of unlabeled samples from a mixture density?

- For continuous features (with PDFs), the answer is often “Yes”.

| An example in discrete case (with PMF).

- Two coins with $P(\text{head})$ being p & q respectively.
- Randomly pick one and toss it; Record the outcome.
- With only the outcomes of N tosses, but not knowing which coin was used each time (\rightarrow unsupervised), can we figure out p and q ?