# Introduction to Big Data and Machine Learning Mixture Models 

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## Mixture Models

## Idea

- Data can be made of complicated distributions that are formed from simpler compoenents.




## Mixture Models

## In this lecture

- Consider the problem of finding clusters in a set of data points
- Non-probabilistic method, hard assignment
- Assumption is that the number of clusters is known
- K-Means algorithm
- A version of expectation maximization (EM) algorithm


## Mixture Models

## Problem Definition

- Consider the problem of identifying groups, or clusters, of data points in a multidimensional space
- Suppose the data set consists of $\left\{x_{1}, \ldots, x_{N}\right\}$, consisting of $N$ D-dimensional observations
- Our goal is to partition the data into a set of $K$, where we will assume the number $K$ is given (there are, however, ways to estimate it)


## Mixture Models

## Intuition

- A cluster, or group, is a set of data points where the inter-point distances within the group are small compared to the distances with points outside the cluster
- We formalize this by introducing a set of $D$-dimensional vectors $\mu_{k}$, where $k=1 \ldots k$, where $\mu_{k}$ is a prototype associated with the $k^{t h}$ cluster.
- $\mu_{k}$ can be thought of as the cluster centers


## Goal

- Find cluster centers $\mu_{k}$ as well as an assignment of the data points to clusters, such that the sum of squares of the distances of each data point to its closest vector $\mu_{k}$ is a minimum.


## Mixture Models

## Notation

- For each data point $x_{n}$, we introduce a corresponding set of binary indicator variables $r_{n k} \in\{0,1\}$, where $k=1, \ldots, K$ describing which of the $K$ clusters the data point $x_{n}$ is assigned to, so that if data point $x_{n}$ is assigned to cluster $k$ then $r_{n k}=1$, and $r_{n j}=0$ for $j \neq k$
- This is known as 1 -of- $K$ coding scheme


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## Optimization problem

- We can then define an objective function:

$$
\begin{equation*}
J=\sum_{n=1}^{N} \sum_{k=1}^{K} r_{n k}\left\|x_{n}-\mu_{k}\right\|^{2} \tag{1}
\end{equation*}
$$

## Mixture Models

Solving the optimization problem

$$
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- Goal is to find values for $\left\{r_{n k}\right\}$ and the $\left\{\mu_{k}\right\}$ so as to minimize $J$


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## Algorithm

First, choose initial values $\mu_{k}$. Then, iterate two steps (EM):
(1) Minimize $J$ with respect to $r_{n k}$, keeping $\mu_{k}$ fixed (Expectation)
(2) Minimize $J$ with respect to $\mu_{k}$, keeping $r_{n k}$ fixed (Maximization)

## Mixture Models

## First step

- Consider determination of $r_{n k}$
- Since $J$ is a linear function of $r_{n k}$, it has a closed-form solution
- We have $n$ independent terms, each can be found in linear time, choose $r_{n k}=1$ for whichever value of $k$ gives the minimum value of $\left\|x_{n}-\mu_{k}\right\|^{2}$
- More formally:

$$
r_{n k}= \begin{cases}1 & \text { if } k=\operatorname{argmin}_{j}\left\|x_{n}-\mu_{j}\right\|^{2}  \tag{2}\\ 0 & \text { otherwise }\end{cases}
$$

## Mixture Models

## Second step

- No consider optimizing $\mu_{k}$, with $r_{n k}$ fixed
- $J$ is a quadratic function of $\mu_{k}$, and it can be minimized by setting its derivative with respect to $\mu_{k}$ to zero, giving

$$
\begin{equation*}
2 \sum_{n=1}^{N} r_{n k}\left(x_{n}-\mu_{k}\right)=0 \tag{3}
\end{equation*}
$$

solving for $\mu_{k}$ gives:

$$
\begin{equation*}
\mu_{k}=\frac{\sum_{n} r_{n k} x_{n}}{\sum_{n} r_{n k}} \tag{4}
\end{equation*}
$$

## Mixture Models

EM algorithm

- The steps above are repeated until no change in assignment is seen, or after a fixed number of iterations

