## The EM algorithm for Gaussian mixture models

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## Topics we'll cover

- 1 Gaussian mixture models
- 2 The optimization problem
- 3 The EM algorithm
- 4 Examples

### *K*-means: the good and the bad

#### The good:

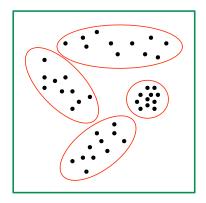
- Fast and easy.
- Effective in quantization.

#### The bad:

 Geared towards data in which the clusters are spherical, and of roughly the same radius.

Is there is a similarly-simple algorithm in which clusters of more general shape are accommodated?

#### Mixtures of Gaussians



Each of the k clusters is specified by:

- ullet a Gaussian distribution  $P_j = \mathcal{N}(\mu_j, \Sigma_j)$
- a mixing weight  $\pi_i$

Overall distribution over  $\mathbb{R}^d$ : a **mixture of Gaussians** 

$$Pr(x) = \pi_1 P_1(x) + \dots + \pi_k P_k(x)$$

### The clustering task

We are given data  $x_1, \ldots, x_n \in \mathbb{R}^d$ .

For any mixture model  $\pi_1, \ldots, \pi_k, \ P_1 = N(\mu_1, \Sigma_1), \ldots, P_k = N(\mu_k, \Sigma_k)$ ,

$$\Pr\left(\text{data} \mid \pi_{1} P_{1} + \dots + \pi_{k} P_{k}\right) \\
= \prod_{i=1}^{n} \left(\pi_{1} P_{1}(x_{i}) + \dots + \pi_{k} P_{k}(x_{i})\right) \\
= \prod_{i=1}^{n} \left(\sum_{j=1}^{k} \frac{\pi_{j}}{(2\pi)^{d/2} |\Sigma_{j}|^{1/2}} \exp\left(-\frac{1}{2}(x_{i} - \mu_{j})^{T} \Sigma_{j}^{-1}(x_{i} - \mu_{j})\right)\right)$$

Find the maximum-likelihood mixture of Gaussians:

the parameters  $\{\pi_j, \mu_j, \Sigma_j : j = 1 \dots k\}$  that maximize this function.

### **Optimization surface**

Minimize the negative log-likelihood,

$$L(\{\pi_j, \mu_j, \Sigma_j\}) = \sum_{i=1}^n \ln \left( \sum_{j=1}^k \frac{\pi_j}{(2\pi)^{d/2} |\Sigma_j|^{1/2}} \exp \left( -\frac{1}{2} (x_i - \mu_j)^T \Sigma_j^{-1} (x_i - \mu_j) \right) \right)$$

### The EM algorithm

- 1 Initialize  $\pi_1, \ldots, \pi_k$  and  $P_1 = N(\mu_1, \Sigma_1), \ldots, P_k = N(\mu_k, \Sigma_k)$ .
- 2 Repeat until convergence:
  - Assign each point  $x_i$  fractionally between the k clusters:

$$w_{ij} = \Pr(\text{cluster } j \mid x_i) = \frac{\pi_j P_j(x_i)}{\sum_{\ell} \pi_{\ell} P_{\ell}(x_i)}$$

• Update mixing weights, means, and covariances:

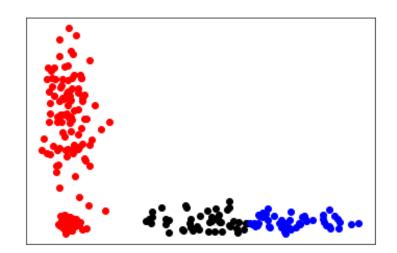
$$\pi_j = \frac{1}{n} \sum_{i=1}^n w_{ij}$$

$$\mu_j = \frac{1}{n\pi_j} \sum_{i=1}^n w_{ij} x_i$$

$$\Sigma_j = \frac{1}{n\pi_j} \sum_{i=1}^n w_{ij} (x_i - \mu_j) (x_i - \mu_j)^T$$

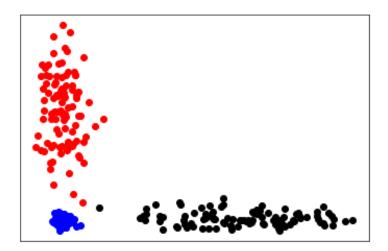
### **Example**

Data with 3 clusters, each with 100 points.



# **E**xample

Data with 3 clusters, each with 100 points.



EM for mixture of Gaussians