

MS1b: SDM - Problem Sheet 3

1. (Expectation-Maximization) Assume you are interested in clustering n binary images. Each binary image \mathbf{x} corresponds to a vector of p binary random variables; p being the number of pixels. We adopt a probabilistic approach and model the probability mass function of \mathbf{x} using a finite mixture model

$$f(\mathbf{x}|\theta) = \sum_{k=1}^K \pi_k f(\mathbf{x}|\phi_k)$$

where $f(\mathbf{x}|\phi_k)$ is a product of p Bernoulli distributions of parameters $\phi_k = (\phi_k^1, \dots, \phi_k^p) \in [0, 1]^p$ and π_1, \dots, π_K satisfy $\pi_k \geq 0 \forall k$ and $\sum_{k=1}^K \pi_k = 1$. We want to estimate the unknown parameters $\theta = \{\pi_k, \phi_k\}_{k=1}^K$ given $\mathbf{x}_{1:n}$ by maximizing the associated likelihood function using the EM algorithm. Establish explicitly the EM update; i.e. the expression of the estimate $\theta^{(t)}$ at iteration t as a function of $\theta^{(t-1)}$.

Answer: Each data $\mathbf{x} = (x^1, \dots, x^p) \in \{0, 1\}^p$ is modeled by

$$f(\mathbf{x}|\theta) = \sum_{k=1}^K \pi_k f(\mathbf{x}|\phi_k)$$

with $\phi_k = (\phi_k^1, \dots, \phi_k^p)$ and

$$f(\mathbf{x}|\phi_k) = \prod_{l=1}^p (\phi_k^l)^{x^l} (1 - \phi_k^l)^{1-x^l}.$$

The EM equations are simply

$$\begin{aligned} \pi_k^{(t)} &= \frac{\sum_{i=1}^n \Pr(z_i = k | \mathbf{x}_i, \phi_k^{(t-1)})}{n}, \\ (\phi_k^l)^{(t)} &= \frac{\sum_{i=1: x_i^l = 1}^n \Pr(z_i = k | \mathbf{x}_i, \phi_k^{(t-1)})}{\sum_{i=1}^n \Pr(z_i = k | \mathbf{x}_i, \phi_k^{(t-1)})} \end{aligned}$$

where

$$\Pr(z_i = k | \mathbf{x}_i, \phi_k^{(t-1)}) = \frac{\pi_k^{(t-1)} f(\mathbf{x}_i | \phi_k^{(t-1)})}{f(\mathbf{x}_i | \theta^{(t-1)})}.$$