## Mixture of Gaussian(Soft Clustering)

Give probabilities that an instance belongs to each cluster instead of assigning only one. Let  $g(x;m,\sigma)$  be the probability of a point x based on a Gaussian Distribution with mean m and variance  $\sigma$ . Suppose there is a distribution generated by randomly selecting one of K Gaussians. We randomly draw a point from this distribution. Let  $p_k$  be the probability of choosing the  $k^{\rm th}$  Gaussian. Then

$$p(x) = \sum_{k=1}^K p_k g(x; m_k, \sigma_k)$$

Now we want to find  $p_k$ ,  $\sigma_k$  and  $m_k$  that maximize p(x).

E step: compute the probability that point  $x_n$  is generated by distribution k,i stands for steps

$$p^{(i)}(k \mid x_n) = \frac{p_k^{(i)} g\!\left(x_n; m_k^{(i)}, \sigma_k^{(i)}\right)}{\sum_{j=1}^K p_j^{(i)} g\!\left(x_n; m_j^{(i)}, \sigma_j^{(i)}\right)}$$

M step: update  $p_k$ ,  $\sigma_k$  and  $m_k$ .

$$\begin{split} m_k^{(i+1)} &= \frac{\sum_{n=1}^N p^{(i)}(k \mid x_n) x_n}{\sum_{n=1}^N p^{(i)}(k \mid x_n)} \\ \sigma_k^{(i+1)} &= \sqrt{\frac{\sum_{n=1}^N p^{(i)}(k \mid x_n) \parallel x_n - n_k^{(i+1)} \parallel}{\sum_{n=1}^N p^{(i)}(k \mid x_n)}} \\ p_k^{(i+1)} &= \frac{1}{N} \sum_{n=1}^N p^{(i)}(k \mid x_n) \end{split}$$