>



Unit 4 Unsupervised Learning (2

Lecture 16. Mixture Models; EM

<u>Course</u> > <u>weeks</u>)

3. Introduction to Mixture Models

> <u>algorithm</u>

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3. Introduction to Mixture Models
Gaussian Mixture Model: Definitions





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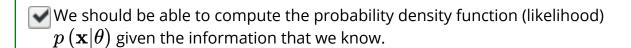
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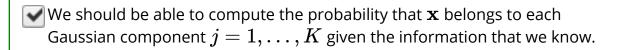
Gaussian Mixture Model: Definitions

1/1 point (graded)

Assume a Gaussian mixture model with K Gaussians such that we know all the means and variances. Assume that we also know the mixture weights p_1,\ldots,p_K . Let ${\bf x}$ be an observation obtained from the Gaussian mixture model. Let all of the parameters of the Gaussian mixture model be collectively represented as θ .

Which of the following are true?







Solution:

Both the statements are true. The generative Gaussian mixture model means that if we know all of the parameters of the K Gaussians and the mixture weights, the probability density function $p\left(\mathbf{x}|\theta\right)$ can be computed using the law of total probability as

$$p\left(\mathbf{x}| heta
ight) = \sum_{j=1}^{K} p_{j} \mathcal{N}\left(\mathbf{x}; \mathbf{\mu}^{(j)}, \sigma_{j}^{2}
ight).$$

The posterior probability that ${\bf x}$ belongs to a Gaussian component j can then be computed using Bayes rule.

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• Answers are displayed within the problem

Likelihood of Gaussian Mixture Model



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Gaussian Mixture Model: Definitions

A **Gaussian Mixture Model (GMM)** , which is a generative model for data $\mathbf{x} \in \mathbb{R}^d$, is defined by the following set of parameters:

- 1. K: Number of mixture components
- 2. A d-dimensional Gaussian $\mathcal{N}\left(\mu^{(j)},\sigma_{j}^{2}
 ight)$ for every $j=1,\ldots,K$
- 3. p_1, \dots, p_K : Mixture weights

The parameters of a K-component GMM can be collectively represented as $\theta=\left\{p_1,\ldots,p_K,\mu^{(1)},\ldots,\mu^{(K)},\sigma_1^2,\ldots,\sigma_K^2\right\}$. Note that we have assumed the same variance σ_j^2 across all components of the j^{th} Gaussian mixture component for $j=1,\ldots,K$. Further, every Gaussian component is assumed to have a

diagonal covariance matrix. These are two assumptions that are made only for simplicity and the methodology presented can be extended to the setting of a general covariance matrix. Also, note that $\mu^{(j)}$ is a d-dimensional vector for every $j=1,\ldots,K$.

The **likelihood** of a point \mathbf{x} in a GMM is given as

$$p\left(\mathbf{x}\mid heta
ight) = \sum_{j=1}^{K} p_{j} \mathcal{N}\left(\mathbf{x}, \mu^{(j)}, \sigma_{j}^{2}
ight).$$

The generative model can be thought of first selecting the component $j\in\{1,\ldots,K\}$, which is modeled using the multinomial distribution with parameters p_1,\ldots,p_K , and then selecting a point ${\bf x}$ from the Gaussian component ${\cal N}\,(\mu^{(j)},\sigma_j^2)$.

Discussion

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Meaning of "belonging to a Gaussian component".
 I don't understand the 2nd option in Q1 (Gaussian Mixture Model: Definitions): > We should ...

 Difference between some terms
 What is the difference between "mixture components" and Clusters? what is the difference b...

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