

Mixture of Gaussians [1] as A Marginalization [2]

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1. Introduction

The mixture of Gaussians model can be expressed as the marginalization of a joint probability distribution [4] between the observed data x and a discrete hidden variable h that takes values $h \in \{1 \dots K\}$ (Figure 1). If it defines in Equation 1.

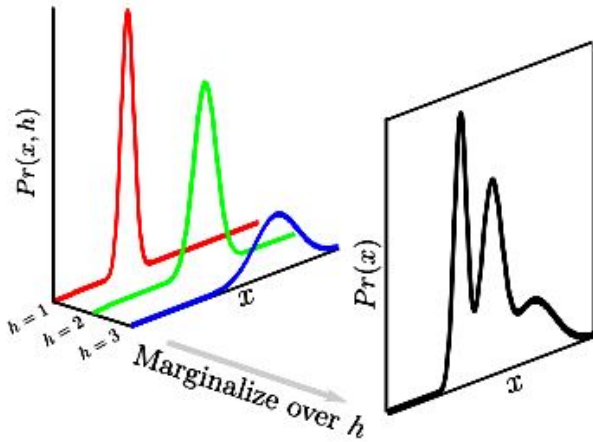


Figure 1. Mixture of Gaussians as a marginalization. The mixture of Gaussians can also be thought of in terms of a joint distribution $Pr(x, h)$ between the observed variable x and a discrete hidden variable h . To create the mixture density it marginalizes over h . The hidden variable has a straightforward interpretation: it is the index of the constituent normal distribution [5].

$$\begin{aligned} Pr(x|h, \theta) &= Norm_x[\mu_h, \Sigma_h] \\ Pr(h, \theta) &= Cat_h[\lambda] \end{aligned} \quad (1)$$

where $\lambda = [\lambda_1 \dots \lambda_K]$ are the parameters of the categorical distribution, then it can recover the original density us-

ing in Equation 2.

$$\begin{aligned} Pr(x|\theta) &= \sum_{k=1}^K Pr(x, h = k|\theta) \\ &= \sum_{k=1}^K Pr(x|h = k, \theta) Pr(h = k|\theta) \quad (2) \\ &= \sum_{k=1}^K \lambda_k Norm_x[\mu_k, \Sigma_k] \end{aligned}$$

2. Conclusions

Interpreting the model in this way also provides a method to draw samples from a mixture of Gaussians: it samples from the joint distribution $Pr(x, h)$, and then discard the hidden variable h to leave just a data sample x . To sample from the joint distribution $Pr(x, h)$ it first sample h from the categorical prior $Pr(h)$ [3], then sample x from the normal distribution $Pr(x|h)$ associated with the value of h . Notice that the hidden variable h has a clear interpretation in this procedure; it determines which of the constituent normal distributions is responsible for the observed data point x .

References

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