

Data assimilation practicals



Recapitulation of Bayes' Theorem

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Bayes' Theorem



Likelihood. Pdf of the observations given a value of the state variable.

$$p(\mathbf{x}|\mathbf{y}) = \frac{p(\mathbf{y}|\mathbf{x})p(\mathbf{x})}{p(\mathbf{y})}$$

Prior pdf. Pdf of the state variables coming from the model.

Posterior pdf. Pdf of the state variables given the observations.

Marginal pdf of the observations. It acts as a normalisation constant. Usually not necessary to compute explicitly.

$$p(\mathbf{y}) = \int_{\mathcal{X}} p(\mathbf{y}|\mathbf{x}) p(\mathbf{x}) d\mathbf{x}$$

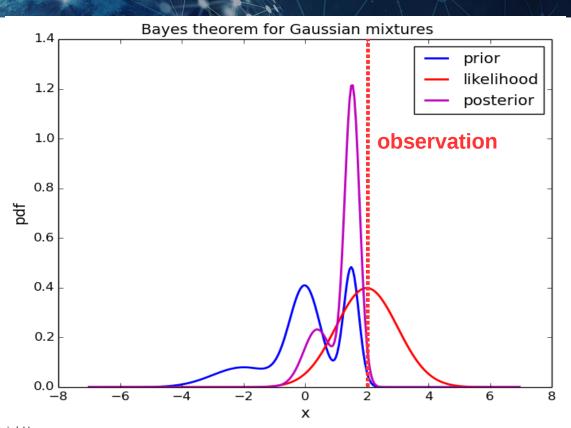
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Bayes' Theorem: illustration in 1D





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Gaussian mixtures



Gaussian (or normal) pdf:

$$\phi(x; \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

Gaussian mixture (convex linear combination of Gaussian pdf's):

$$p(x) = \sum_{n=1}^{N} \alpha_n \phi(x; \mu_n, \sigma_n)$$

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