



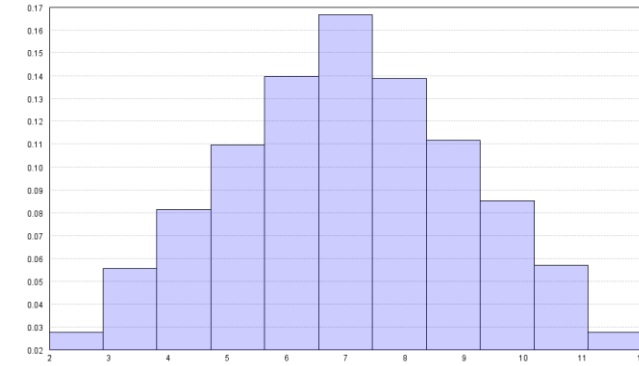
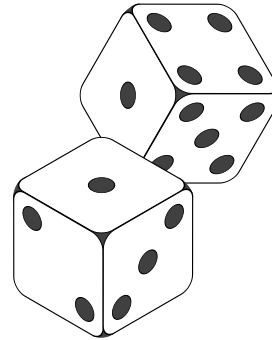
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Bayesian probability

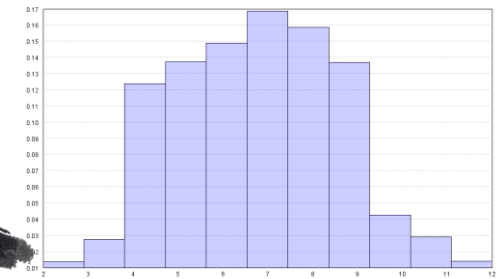
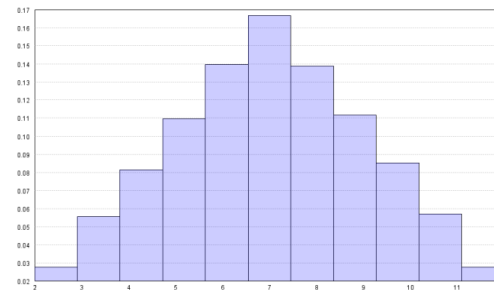
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Different interpretations of probabilities

Long-term frequencies



Degree of belief (Bayesian probability)



Bayesian probability in shape and image analysis

Sources of uncertainty in image and shape analysis

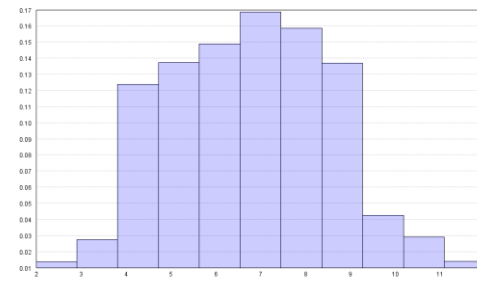
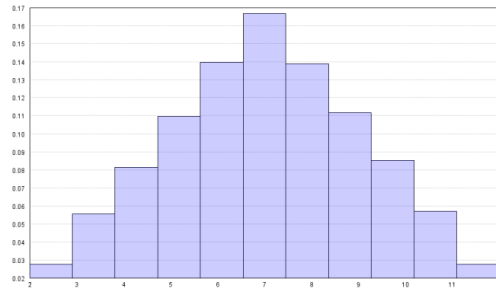
- Measurement noise
- Calibration of acquisition device
- Limited measurement accuracy
- Missing data

Repeating a measurement does not give us much more information.



Bayesian probability

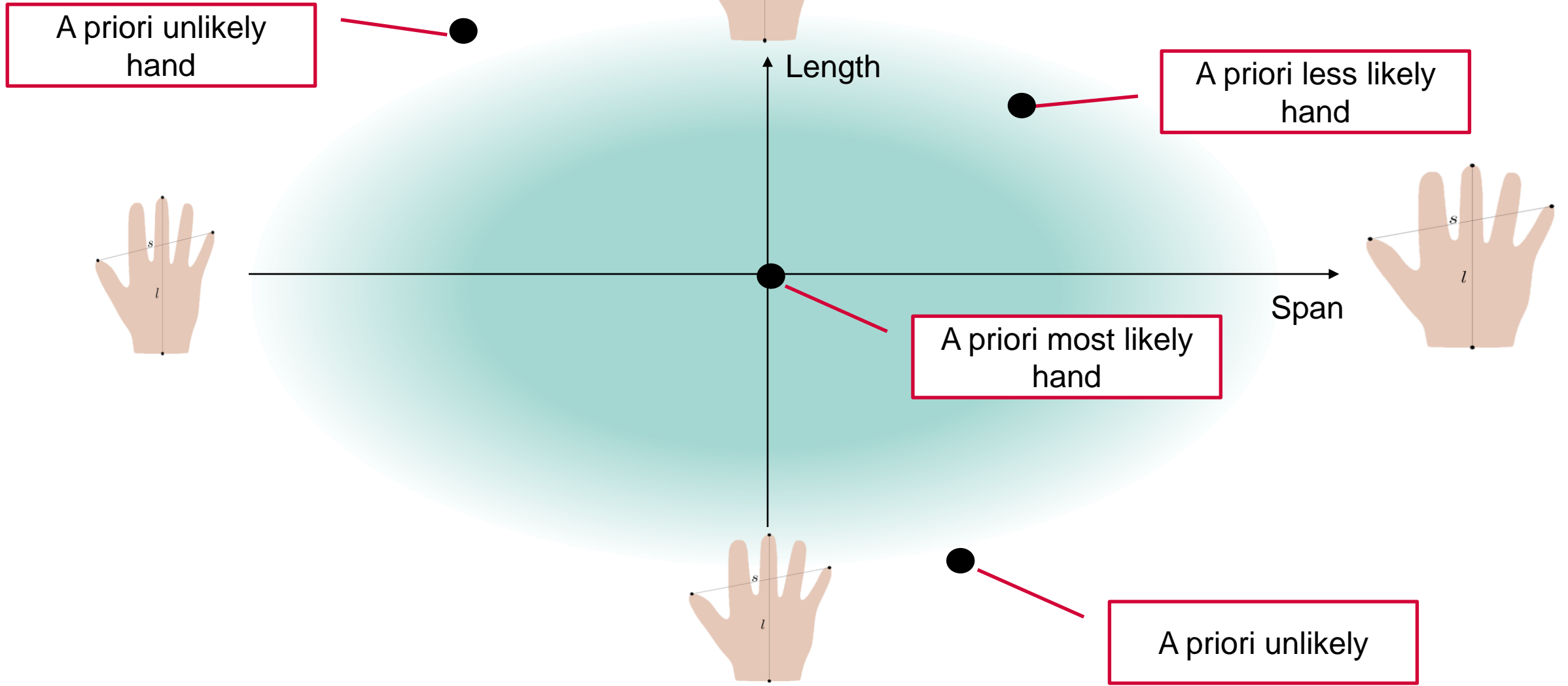
Bayesian probabilities rely on a *subjective* perspective:



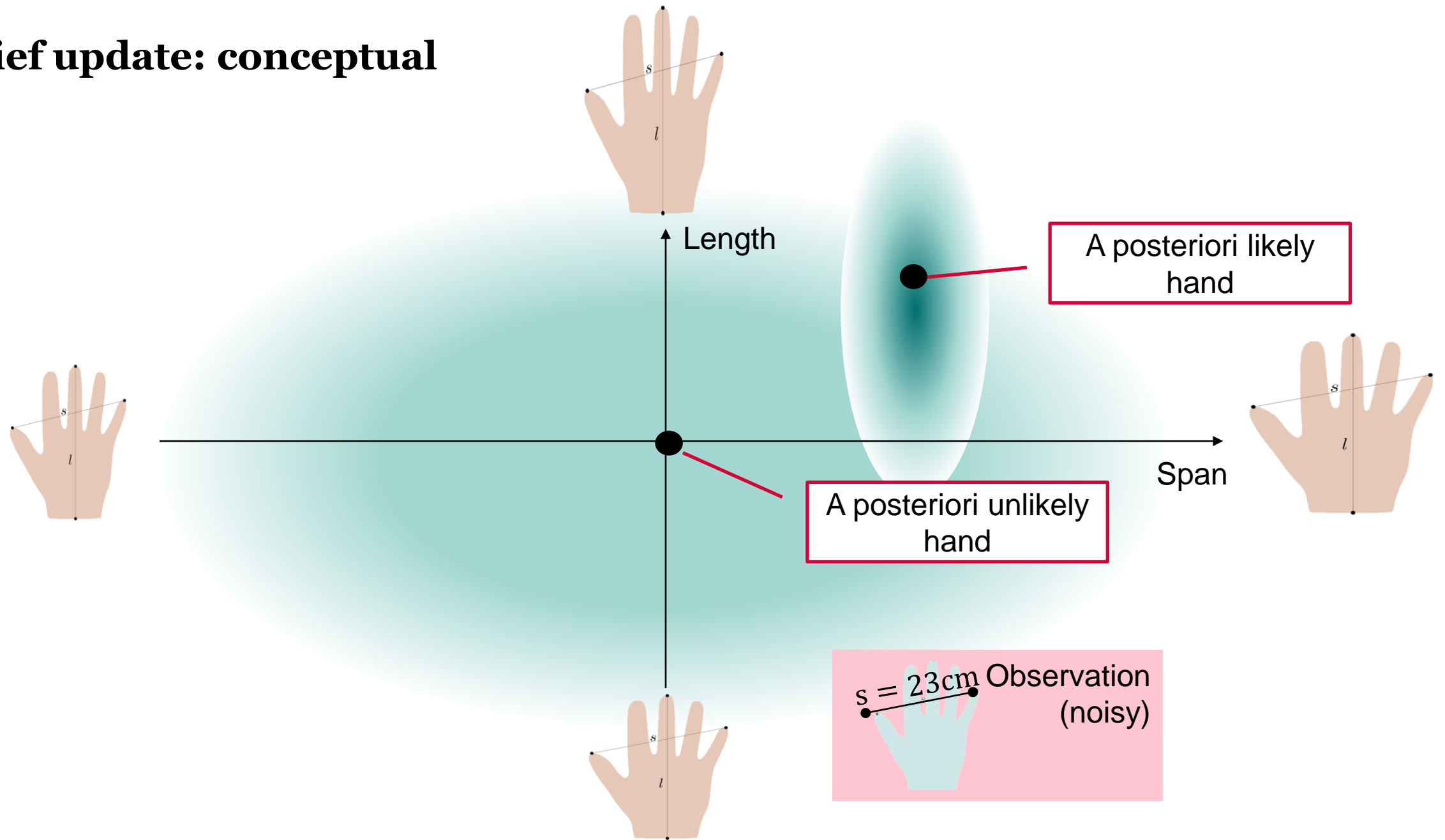
Subjective is not arbitrary!

- Beliefs are updated when new data becomes available
- Belief updates follow rules of probability theory.

Belief update: conceptual



Belief update: conceptual



Belief update: Conditioning and Bayes Rule

Starting point: Joint distribution (usually factorized)

$$p(s, l) = p(s|l)p(l)$$

Example:

- 1. Length of an average hand is approximately 24 centimeter
ca 2/3 of the hands are within a range of ± 2 cm*

$$p(l) = N(24, 2)$$

- 2. On average, length and span are approximately the same
The span varies more than the length. 2/3 of the hands are within a 4 cm range*

$$p(s|l) = N(l, 4)$$

Belief update: Conditioning and Bayes Rule

Starting point: Joint distribution (usually factorized)

$$p(s, l) = p(s|l)p(l)$$

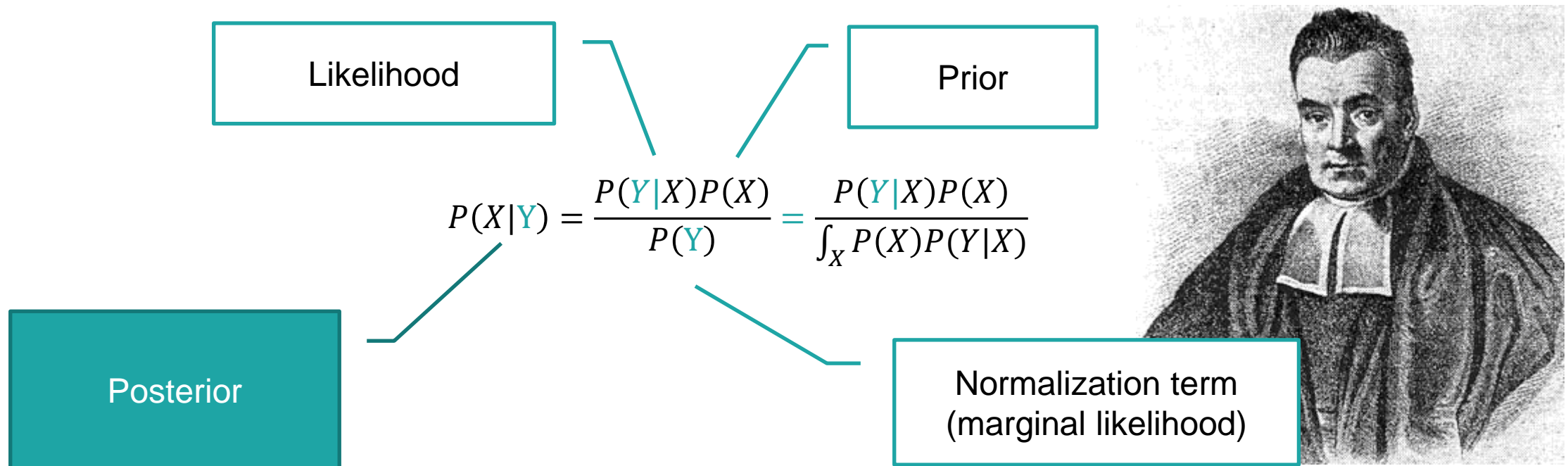
Belief update: Conditioning

$$p(l|s) = \frac{p(s, l)}{p(s)} = \frac{p(s|l)p(l)}{p(s)} = \frac{p(s|l)p(l)}{\int p(s|l)p(l)dl}$$

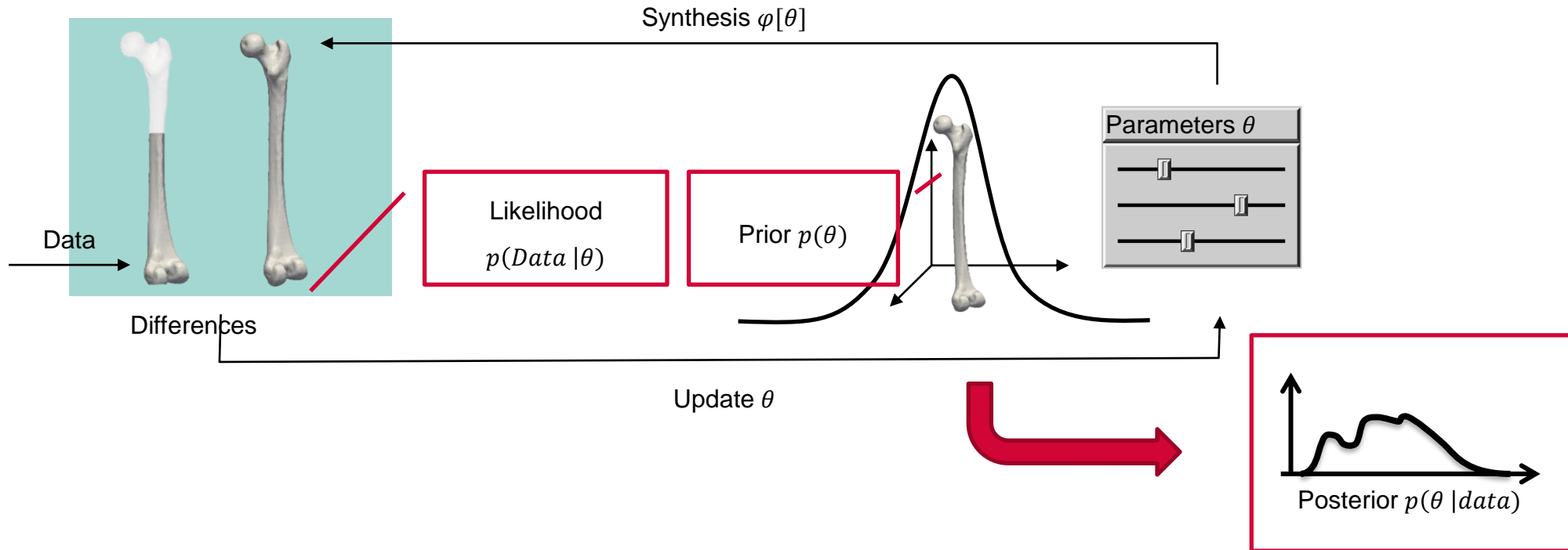
Belief update is just conditioning and marginalizing!

Bayes rule

Rule for updating our belief in X , after we have observed data Y



Bayesian Inference



Bayes rule

$$p(\theta|D) = \frac{p(D|\theta)p(\theta)}{p(D)} = \frac{p(D|\theta)p(\theta)}{\int p(D|\theta)p(\theta)d\theta}$$