Lecture 12: Analytical examples of Bayesian inference

Professor Ilias Bilionis

Bayesian parameter estimation



Example: Coin toss

- We run a coin toss experiment N times and we wish to figure out the probability of heads.
- The data we have observe are:

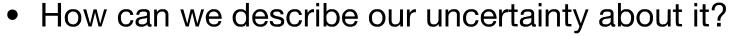
For notational convenience we will be writing:

$$x_{1:N} = (x_1, ..., x_N)$$

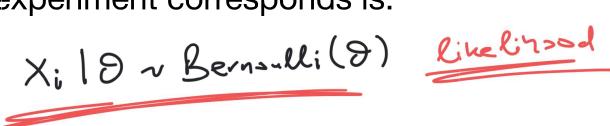


Example: Coin toss

The probability of success of the coin toss:



Each coin toss experiment corresponds is:





The likelihood of the data

$$\zeta_{i} | \theta \sim \text{Bernoulli}(9)
\rho(x; | \theta) = \theta^{x_{i}} (1-\theta)^{1-x_{i}} = \begin{cases} \theta , x_{i}=1 \\ 1-\theta , x_{i}=0 \end{cases}$$

$$x_{i} \text{ ind.}
\rho(x_{1}: N|\theta) = \rho(x_{1}|\theta) \cdot \rho(x_{2}|\theta) \cdot \dots \cdot \rho(x_{N}|\theta)$$

$$= \prod_{i=1}^{N} \rho(x_{i}|\theta)$$

$$= \prod_{i=1}^{N} \theta^{x_{i}} (1-\theta)$$

$$= \sum_{i=1}^{N} \lambda_{i} (1-\theta)$$

$$= \sum_{i=1}^{N} \lambda_{i} (1-\theta)$$

$$= \sum_{i=1}^{N} \lambda_{i} (1-\theta)$$



Bayes' rule applied

$$p(A \mid B) = \frac{p(AB)}{p(B)}.$$

A =the model parameters $= \theta$

$$B =$$
the data $= x_{1:N}$



The joint probability density

Posterior state of knowledge - posterior p(the model parameters | the data)

p(the parameters and the data) p(the data)



Simplifying the joint

$$p(AB) = p(B|A)p(A)$$

A =the model parameters $= \theta$

$$B =$$
the data $= x_{1:N}$



Joint

Likelihood

Prior

 $p(\text{the parameters and the data}) \neq p(\text{the data}|\text{the parameters})$

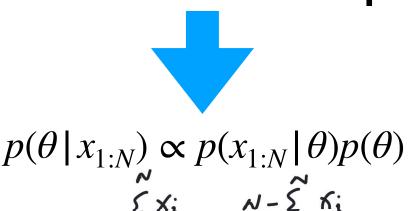


posterior ∝ likelihood × prior



The posterior for the coin toss example

posterior ∝ likelihood × prior

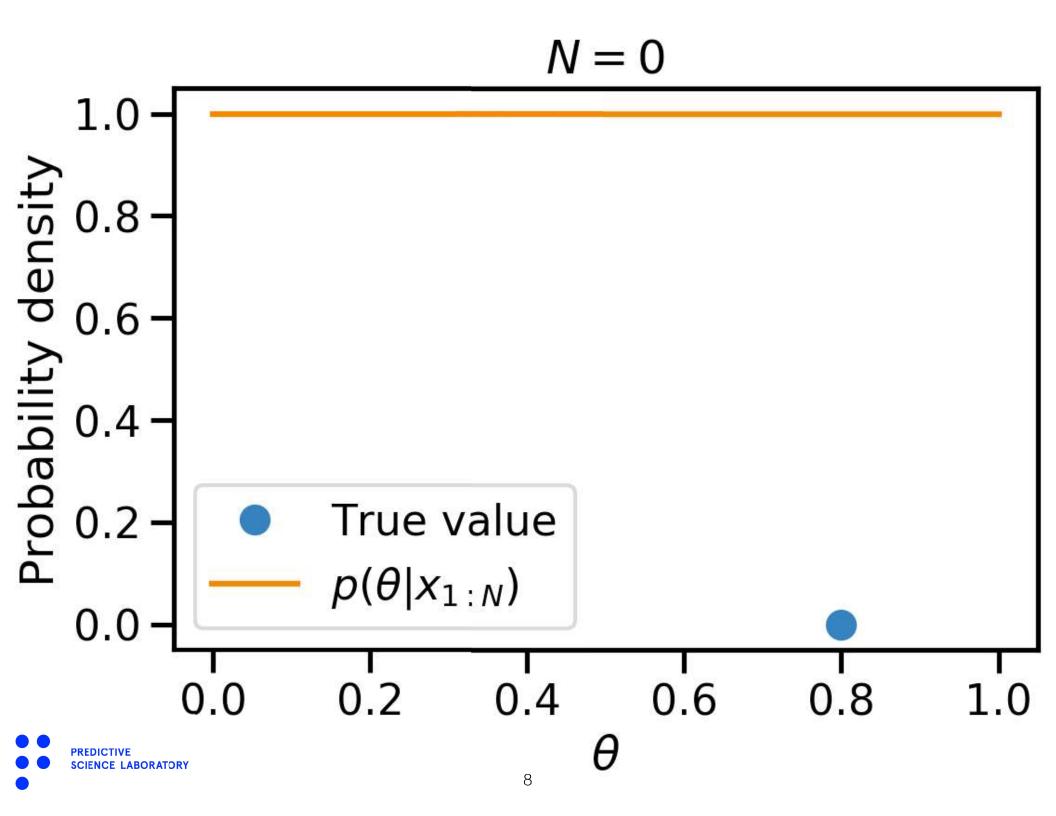


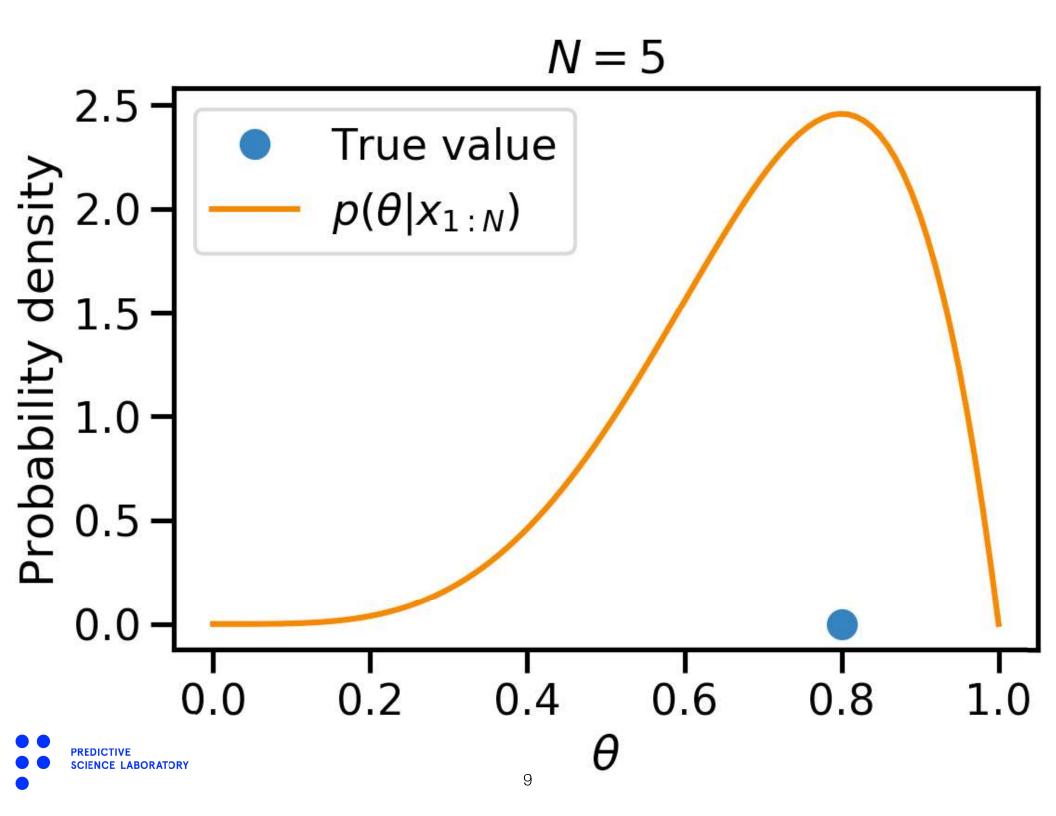
$$p(\theta \mid x_{1:N}) \propto p(x_{1:N} \mid \theta) p(\theta)$$

$$= 9^{\frac{\zeta}{1} \times i} (1-9)^{-\frac{\zeta}{1}} \kappa_{i}$$

$$= 5^{-\frac{\zeta}{1}} (9)$$







N = 20

