

Theorem (Bayes' Theorem). The probability that a system is well-described a model with parameter set \mathbf{X} , given a dataset \mathbf{D} and initial knowledge I , is:

$$\Pr(\mathbf{X}|\mathbf{D}, I) = \frac{\Pr(\mathbf{D}|\mathbf{X}, I) \cdot \Pr(\mathbf{X}|I)}{\Pr(\mathbf{D}|I)}.$$

$\Pr(\mathbf{X}|I)$ is called the *prior*, $\Pr(\mathbf{X}|\mathbf{D}, I)$ the *posterior*, and $\Pr(\mathbf{D}|\mathbf{X}, I)$ the *likelihood*.

1 Coin tossing

We consider a sequence of tosses of a biased coin. The coin can be modeled with a single parameter $H \in [0, 1]$, the probability of heads on a single toss. Thus our parameter set is $\mathbf{X} = \{H\}$.

The dataset \mathbf{D} can be reduced to a tuple (n, h) where h is the number of heads in n tosses. There is no other prior information: $I = \emptyset$. The probability of obtaining some data given the parameter H is:

$$\Pr(\mathbf{D}|\mathbf{X}, I) = \Pr((n, h)|H) = \binom{n}{h} H^h (1 - H)^{n-h},$$

since there are $\binom{n}{h}$ ways of getting h heads (each with probability H) and $n - h$ tails (each with probability $1 - H$).

2 Lighthouse problem

We have a lighthouse at a distance β from the shore and at a location α along the shore. The lighthouse emits flashes at random angles θ_k which arrive at the shore at locations x_k . Let:

$f_\theta(\theta) \equiv \Pr(\theta_k = \theta)$ be the pdf of θ_k ,

$F_\theta(\theta) \equiv \Pr(\theta_k \leq \theta)$ be the cdf of θ_k ,

$f_x(x) \equiv \Pr(x_k = x)$ be the pdf of x_k ,

$F_x(x) \equiv \Pr(x_k \leq x)$ be the cdf of x_k .

Since the lighthouse emits flashes at random angles, we assume θ_k is uniformly distributed:

$$f_\theta(\theta) = \frac{1}{2\pi}, \quad F_\theta(\theta) = \int_0^\theta f_\theta(\tilde{\theta}) d\tilde{\theta} = \frac{\theta}{2\pi}.$$

To derive f_x , we first derive F_x :

$$F_x(x) = \Pr(x_k \leq x) = \Pr(\theta_k \leq \hat{\theta}(x)) = F_\theta(\hat{\theta}(x)),$$

where $\hat{\theta}(x)$ is the angle that creates a flash at x . To find it, we use the geometry of the problem:

$$\hat{\theta}(x) = \arctan\left(\frac{x - \alpha}{\beta}\right).$$

Thus:

$$F_x(x) = \frac{1}{2\pi} \arctan\left(\frac{x - \alpha}{\beta}\right).$$

$$f_x(x) = \frac{d}{dx} F_x(x) = \frac{\beta/2\pi}{(x - \alpha)^2 + \beta^2}.$$

This makes intuitive sense: the distribution is a Lorentzian centered around α , the location of the the lighthouse along the shore. The further the lighthouse is, the wider the distribution is.