

MACHINE LEARNING 1: ASSIGNMENT 2

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Exercise 1

(a) With the definition $P(x_k | \sigma)$:

$$P(x | \sigma) = \begin{cases} \sigma & \text{if } x = \text{head} \\ 1 - \sigma & \text{if } x = \text{tail} \end{cases}$$

We can state the likelihood function $P(\mathcal{D} | \sigma)$:

$$\begin{aligned} P(\mathcal{D} | \sigma) &= \prod_{k=1}^n P(x_k | \sigma) \\ &= \sigma^5 \cdot (1 - \sigma)^2 \end{aligned}$$

(b) The maximum likelihood solution for σ is simply the sample mean:

$$\sigma = \frac{\#\{x = \text{head} \mid x \in \mathcal{D}\}}{\#\mathcal{D}} = \frac{5}{7}$$

With this we can compute $P(x_8 = \text{head}, x_9 = \text{head} | \sigma)$:

$$P(x_8 = \text{head}, x_9 = \text{head} | \sigma) = P(x_8 = \text{head} | \sigma)P(x_9 = \text{head} | \sigma) = \sigma^2 = \frac{25}{49} \approx 0.51$$

(c) With the definition of $p(\sigma) = 1$, the posterior is computed as:

$$\begin{aligned} p(\sigma | \mathcal{D}) &= \frac{p(\mathcal{D} | \sigma)p(\sigma)}{p(\mathcal{D})} \\ &= \frac{p(\mathcal{D} | \sigma)}{p(\mathcal{D})} \\ &= \frac{\sigma^5 \cdot (1 - \sigma)^2}{p(\mathcal{D})} \end{aligned}$$