

# Exercise 1.

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

with the definition  $P(X_k|\theta)$ , We state the likelihood function

$$P(D|\theta) = \prod_{k=1}^n P(X_k|\theta) = \theta^5 \cdot (1-\theta)^2$$

$$(b) f(x) = P(D|\theta) = \theta^5 \cdot (1-\theta)^2$$

$$L(D|\theta) = 5 \log \theta + 2 \log (1-\theta)$$

$$= \frac{5}{\theta} + \frac{2}{1-\theta} (-1)$$

$$= \frac{5}{\theta} - \frac{2}{1-\theta}$$

$$= \frac{5-5\theta-2\theta}{\theta(1-\theta)}$$

$$= \frac{5-7\theta}{\theta(1-\theta)}$$

$$5-7\theta=0$$

$$\hat{\theta} = \frac{5}{7}$$

Team = R C S K S K

Ramesha 387219

CHEN 387275

Shah 384926

Swamy 384418

Khalil 385402

Kao 387470

With this fact, each sample  $X_k$  is generated independently  
compute  $P(X_8 = \text{head}, X_9 = \text{head} | \hat{\theta})$

$$\begin{aligned} P(X_8 = \text{head}, X_9 = \text{head} | \hat{\theta}) &= P(X_8 = \text{head} | \hat{\theta}) P(X_9 = \text{head} | \hat{\theta}) = \hat{\theta}^2 \\ &= \frac{5}{7} \times \frac{5}{7} \\ &= \frac{25}{49} \end{aligned}$$

$$(c) P(\theta) = \begin{cases} 1 & \text{if } 0 \leq \theta \leq 1 \\ 0 & \text{else} \end{cases}$$

prior distribution

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

posterior distribution

$$\alpha = \frac{1}{\int_0^1 \prod_{k=1}^n P(X_k|\theta) P(\theta) d\theta}$$

$$= \frac{1}{\int_0^1 \prod_{k=1}^n P(X_k|\theta) d\theta}$$

$$= \frac{1}{\int_0^1 \theta^5 - 2\theta^6 + \theta^5 d\theta}$$

$$= \frac{1}{\left[ \frac{\theta^6}{6} - \frac{2\theta^7}{7} + \frac{\theta^6}{6} \right]_0^1} = \frac{1}{\frac{28}{168} - \frac{48}{168} + \frac{24}{168}} = \frac{1}{\frac{1}{168}} = 168$$

$$= \propto \prod_{k=1}^n P(X_k|\theta) P(\theta)$$

$$= \begin{cases} \propto \cdot \theta^5 \cdot (1-\theta)^2 & \text{if } 0 \leq \theta \leq 1 \\ 0 & \text{else} \end{cases}$$

$$\begin{aligned}
 \int P(X_8 = \text{head}, X_9 = \text{head} | \theta) P(\theta | D) d\theta &= \int P(X_8 = \text{head} | \theta) P(X_9 = \text{head} | \theta) P(\theta | D) d\theta \\
 &= \int_0^1 \theta^2 \cdot \alpha \cdot \theta^5 \cdot (1-\theta)^2 d\theta \\
 &= 168 \int_0^1 \theta^7 (1-\theta)^2 d\theta \\
 &= 168 \int_0^1 \theta^9 - 2\theta^8 + \theta^7 d\theta \\
 &= 168 \left[ \frac{\theta^{10}}{10} - \frac{2\theta^9}{9} + \frac{\theta^8}{8} \right]_0^1 \\
 &= 168 \cdot \frac{1}{360} = \frac{7}{15} \approx 0.4\bar{6}
 \end{aligned}$$

Exercise 2.

(a) To show  $\sigma_n^2 \leq \min\left(\frac{\sigma^2}{n}, \sigma_0^2\right)$

$$\frac{1}{\sigma_n^2} = \frac{n}{\sigma^2} + \frac{1}{\sigma_0^2}$$

$$\Leftrightarrow 1 = \frac{\sigma_0^2 n + \sigma^2}{\sigma^2 \sigma_0^2} \cdot \sigma_n^2$$

$$\Leftrightarrow \sigma_n^2 = \frac{\sigma^2 \sigma_0^2}{\sigma_0^2 n + \sigma^2}$$

We can show  $\sigma_n^2 \leq \min\left(\frac{\sigma^2}{n}, \sigma_0^2\right)$ . Assume  $\frac{\sigma^2}{n} \leq \sigma_0^2$

$$\sigma_n^2 \leq \min\left(\frac{\sigma^2}{n}, \sigma_0^2\right)$$

$$\Leftrightarrow \frac{\sigma^2 \sigma_0^2}{\sigma_0^2 n + \sigma^2} \leq \frac{\sigma^2}{n}$$

$$\Leftrightarrow \sigma^2 \sigma_0^2 \leq \sigma^2 \sigma_0^2 + \frac{\sigma^4}{n}$$

$$\Leftrightarrow \sigma^2 \sigma_0^2 \leq \sigma^2 \cdot \left(\sigma_0^2 + \frac{\sigma^2}{n}\right)$$

$$\Leftrightarrow \sigma_0^2 \leq \sigma_0^2 + \frac{\sigma^2}{n} \quad \text{This holds true since } n \text{ is the number of features}$$

$$\Leftrightarrow 0 \leq \frac{\sigma^2}{n} \Rightarrow N(\mu, \sigma^2), \text{ thus } n, \sigma > 0$$

$\sigma^2$  is the variance of Gaussian distribution

Another case:  $(\sigma_0^2 \leq \frac{\sigma^2}{n})$

$$\sigma_n^2 \leq \min\left(\frac{\sigma^2}{n}, \sigma_0^2\right)$$

$$\Leftrightarrow \frac{\sigma^2 \sigma_0^2}{\sigma_0^2 n + \sigma^2} \leq \sigma_0^2$$

$$\Leftrightarrow \sigma^2 \sigma_0^2 \leq \sigma_0^2 \cdot (\sigma_0^2 n + \sigma^2)$$

$$\Leftrightarrow \sigma^2 \leq \sigma_0^2 n + \sigma^2$$

$$\Leftrightarrow 0 \leq \sigma_0^2 n$$

This holds true since  $n > 0$  and  $\sigma_0^2 > 0$  since it's the variance of the Gaussian distribution

$N(\mu_0, \sigma_0^2)$

(b) To show

$$\min(M_0, \hat{M}_n) \leq M_n \leq \max(M_0, \hat{M}_n)$$

solve  $M_n$ , also using our  $\sigma_n^2$  from above:

$$\frac{M_n}{\sigma_n^2} = \frac{n}{\sigma^2} \hat{M}_n + \frac{M_0}{\sigma_0^2}$$

$$\Leftrightarrow M_n = \left( \frac{n}{\sigma^2} \hat{M}_n + \frac{M_0}{\sigma_0^2} \right) \cdot \sigma_n^2$$

$$\Leftrightarrow M_n = \left( \frac{n}{\sigma^2} \hat{M}_n + \frac{M_0}{\sigma_0^2} \right) \cdot \frac{\sigma^2 \sigma_0^2}{\sigma_0^2 n + \sigma^2}$$

$$\Leftrightarrow M_n = \frac{n \sigma_0^2}{n \sigma_0^2 + \sigma^2} \hat{M}_n + \frac{\sigma^2}{n \sigma_0^2 + \sigma^2} M_0$$

Assume  $M_0 < \hat{M}_n$  (1) and show first part of the inequality

$$\min(M_0, \hat{M}_n) \leq M_n$$

$$\Leftrightarrow M_0 \leq M_n$$

$$\Leftrightarrow M_0 \leq \frac{n \sigma_0^2}{n \sigma_0^2 + \sigma^2} \hat{M}_n + \frac{\sigma^2}{n \sigma_0^2 + \sigma^2} M_0$$

$$\Leftrightarrow M_0 \cdot (n \sigma_0^2 + \sigma^2) \leq n \sigma_0^2 \hat{M}_n + \sigma^2 M_0$$

$$\Leftrightarrow M_0 n \sigma_0^2 \leq n \sigma_0^2 \hat{M}_n$$

$$\Leftrightarrow M_0 \leq \hat{M}_n$$

This is true due to assumption (1) <sup>above</sup> Show the 2nd part of the inequality  
 $M_n \leq \max(M_0, \hat{M}_n)$

$$\Leftrightarrow M_n \leq \hat{M}_n$$

$$\Leftrightarrow \frac{n \sigma_0^2}{n \sigma_0^2 + \sigma^2} \hat{M}_n + \frac{\sigma^2}{n \sigma_0^2 + \sigma^2} M_0 \leq \hat{M}_n$$

$$\Leftrightarrow n \sigma_0^2 \hat{M}_n + \sigma^2 M_0 \leq \hat{M}_n \cdot (n \sigma_0^2 + \sigma^2)$$

$$\Leftrightarrow \sigma^2 M_0 \leq \hat{M}_n \sigma^2$$

$$\Leftrightarrow M_0 \leq \hat{M}_n$$

This is true due to assumption (1) <sup>above</sup> also  
Now Assume  $\hat{M}_n < M_0$  (2) and show that our inequality still holds

We'll start with the first part again:

$$\min(M_0, \hat{M}_n) \leq M_n$$

$$\Leftrightarrow \hat{M}_n \leq M_n$$

$$\Leftrightarrow \hat{M}_n \leq \frac{n\sigma_0^2}{n\sigma_0^2 + \sigma^2} \hat{M}_n + \frac{\sigma^2}{n\sigma_0^2 + \sigma^2} M_0$$

$$\Leftrightarrow \hat{M}_n (n\sigma_0^2 + \sigma^2) \leq n\sigma_0^2 \hat{M}_n + \sigma^2 M_0$$

$$\Leftrightarrow \hat{M}_n \sigma^2 \leq \sigma^2 M_0$$

$$\Leftrightarrow \hat{M}_n \leq M_0$$

This holds true due to assumption (2) above

Let's show the second part of the inequality

$$M_n \leq \max(M_0, \hat{M}_n)$$

$$\Leftrightarrow M_n \leq M_0$$

$$\Leftrightarrow \frac{n\sigma_0^2}{n\sigma_0^2 + \sigma^2} \hat{M}_n + \frac{\sigma^2}{n\sigma_0^2 + \sigma^2} M_0 \leq M_0$$

$$\Leftrightarrow n\sigma_0^2 \hat{M}_n + \sigma^2 M_0 \leq M_0 (n\sigma_0^2 + \sigma^2)$$

$$\Leftrightarrow n\sigma_0^2 \hat{M}_n \leq M_0 n\sigma_0^2$$

$$\Leftrightarrow \hat{M}_n \leq M_0$$

This holds true due to assumption (2) above