

ECE 485 Homework 4 - Jordan Carlson V00714886

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1 (3.7). Given an unknown μ and Σ

a) $\hat{\mu} = \Theta_1 = \frac{1}{n} \sum_{k=1}^n X_k = \frac{n}{n} \rightarrow \boxed{\hat{\mu} = 1}$

b) $P(x|\omega_1) \sim N(0, 1)$, $P(x|\omega_2) \sim N(\hat{\mu}, 1)$
 as $\hat{\mu} = 1$ from a, $\frac{1-0}{2} = \frac{1}{2} \rightarrow \boxed{x = 0.5}$

c) $P(x|\omega_1) = P(x|\omega_2)$
 $\frac{1}{2\pi\sqrt{2}} e^{-\frac{x^2}{2}} = \frac{1}{2\pi\sqrt{10^6}} e^{-\frac{(x-1)^2}{2(10^6)}}$ $\boxed{x = \pm 3.72}$

d) $\frac{1}{2\pi} e^{-\frac{3.72^2}{2}} = \frac{1}{2\pi} e^{-\frac{(\pm 3.72 - \mu)^2}{2}}$
 $\rightarrow 3.72^2 = -(\pm 3.72 - \mu)^2$
 $\rightarrow -13.84 = \pm 13.84 - 7.44\mu + \mu^2$
 $\rightarrow 0 = \mu^2 - 7.44\mu + 27.68 \rightarrow$ irreducible
 $\mu_1 = 0, \mu_2 = 7.44, \text{ so } \boxed{x = 3.72}$

e) The value approximated in part c is a much more valuable parameter, demonstrating the effectiveness of the Bayes decision boundary calculation above poorly modelled distributions from a priori conjectures

Q. (3.11.) Minimum DKL, so $\frac{d}{d\mu} D_{KL}(p_1(x), p_2(x)) = 0$

$$= \frac{d}{d\mu} \int p_1(x) \ln \frac{p_1(x)}{p_2(x)} dx = 0 = - \int p(x) \sum^{-1} (x-\mu) dx$$

$$0 = \int p(x)(x-\mu) dx \rightarrow E_2(x-\mu) = 0$$

$$\text{so } \boxed{E_2(x) = \mu}$$

$$\frac{d}{d\sum} D_{KL}(p_1(x), p_2(x)) = 0 = \int p_1(x) \left(\sum^{-1} (x-\mu)(x-\mu)^T \right)$$

$$p(x)(x-\mu)(x-\mu)^T = \boxed{\sum = E_2((x-\mu)(x-\mu)^T)}$$

3. (3.17.)

a) $P(D|\theta) = \prod_{k=1}^n (s_i, \dots, s_d)$ for $\prod_k P(K|\theta)$

$$= \prod_{k=1}^d \theta_i \sum_{k=1}^n (s_i, \dots, s_d) (1 - \theta_i) \sum_{k=1}^n (1 - (s_i, \dots, s_d))$$

$$= \boxed{\prod_{i=1}^d \theta_i^{s_i} (1 - \theta_i)^{n-s_i}}$$

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

$$P(D) = \prod_{i=1}^d \theta_i^{s_i} (1-\theta_i)^{n-s_i} d\theta_i$$

from identity $\rightarrow P(D) = \prod_{i=1}^d \frac{s_i!(n-s_i)!}{(n+1)!}$

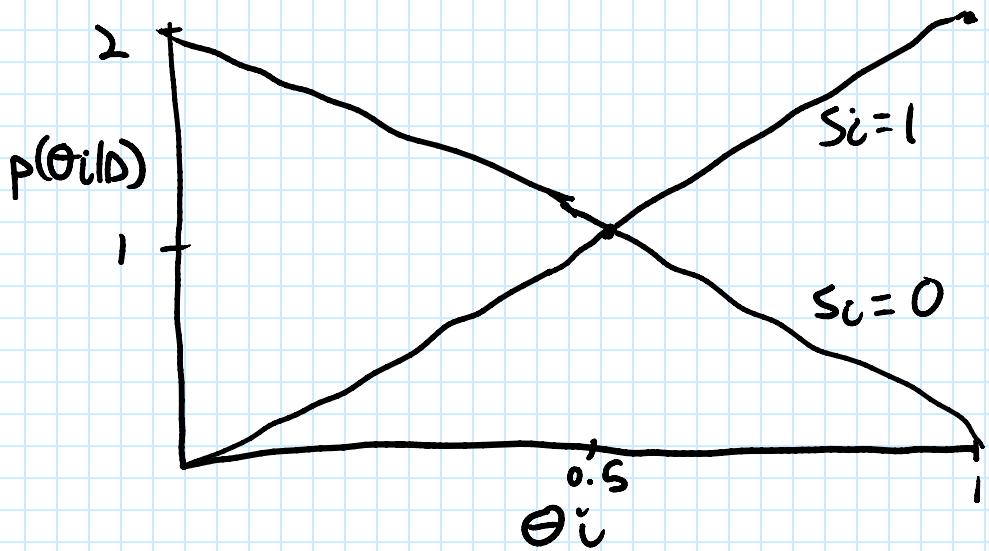
$$P(\theta|D) = \frac{P(D|\theta)P(\theta)}{P(D)} = \frac{\prod_{i=1}^d \theta_i^{s_i} (1-\theta_i)^{n-s_i}}{\prod_{i=1}^d \frac{s_i!(n-s_i)!}{(n+1)!}}$$

$\rightarrow P(\theta|D) = \prod_{i=1}^d \frac{(n+1)!}{s_i!(n-s_i)!} \theta_i^{s_i} (1-\theta_i)^{n-s_i}$

c) $P(\theta|D) = \frac{2}{s_i!(1-s_i)!} \theta_i^{s_i} (1-\theta_i)^{1-s_i}$

when $s_i=0, P(\theta_i|D) = 2 - 2\theta_i$

when $s_i=1, P(\theta_i|D) = 2\theta_i$



d)

$$\int p(x|\theta) p(\theta|D) d\theta$$

$$= \prod_{i=1}^n \left(\int_0^1 \theta_i^{x_i} (1-\theta_i)^{1-x_i} \right) \left(\prod_{i=1}^n \frac{(n+1)!}{s_i!(n-s_i)!} \theta_i^{s_i} (1-\theta_i)^{n-s_i} \right)$$

$$= \prod_{i=1}^n \int_0^1 \left(\theta_i^{x_i} (1-\theta_i)^{1-x_i} \right) \left(\frac{(n+1)!}{s_i!(n-s_i)!} \theta_i^{s_i} (1-\theta_i)^{n-s_i} \right)$$

$$= \prod_{i=1}^n \int_0^1 \frac{\left(\theta_i^{x_i} (1-\theta_i)^{1-x_i} \right) (n+1)! \theta_i^{s_i} (1-\theta_i)^{n-s_i}}{s_i!(n-s_i)!} d\theta_i$$

$$\rightarrow P(x|D) = \prod_{i=1}^n \left[\left(\frac{s_i+1}{n+2} \right)^{x_i} \left(1 - \frac{s_i+1}{n+2} \right)^{1-x_i} \right]$$

e) Using $P(X|D)$, $P(X|\Theta)$, and $P(\Theta|D)$, the estimate for Θ will include both conditional probability functions and thus will be inclusive of both conditional implications, giving valuable information if both conditions are present in the system.

4. (3.22.)

$$p(\theta|s, D)p(D,s) = p(\theta|s)(p(s))(p(D|s,\theta))$$

$$p(\theta|s, D) = p(\theta|s), \text{ so}$$

$$p(\theta|s)p(D|s,\theta) = p(\theta|s,D)p(D|s)$$

$$p(D|s,\theta) = p(D|s)$$

$$\boxed{p(D|s) \neq p(\theta)}$$

5. (3.29.)

$$a^{n+1} = a^n a' \times a^n a x = O(a^n)$$

Yes

b) $a^{bn} = (a^b)a^n \gg x a^n = O(a^n)$ No

c) $a^{n+b} = a^b a^n \propto x a^n = O(a^n)$ Yes

d) $x f(n) \propto O(f(n))$
for any $x \geq 0$, so $f(n) = O(f(n))$

6.(3.43)

$$\text{Eq 85. } = p(\omega_k | x) \propto \prod_{i=1}^d p(x_i | \omega_k)$$

