

# **SML - Assignment 3**

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## SML - Assignment-3

Q.1

$$P(x|\theta) = \prod_{i=1}^d \theta_i^{x_i} (1-\theta_i)^{1-x_i}$$

$$\theta = (\theta_1, \dots, \theta_d)^T$$

$$\text{To show : } \hat{\theta} = \frac{1}{n} \sum_{k=1}^n x_k$$

Consider the distribution of multivariate Bernoulli distribution :-

$$P(x|\theta) = \prod_{i=1}^d \theta_i^{x_i} (1-\theta_i)^{1-x_i}$$

Its likelihood function will be :  $\prod_{k=1}^n P(x_k|\theta)$

$$\text{So, } L = \prod_{k=1}^n P(x_k|\theta) = \prod_{k=1}^n \prod_{i=1}^d \theta_i^{x_i^k} (1-\theta_i)^{1-x_i^k}$$

(Note:  $x_i^k \Rightarrow i$  is for indexing of inner loop and  $k$  is for indexing outer loop).

Take Log on both sides

$$\log(L) = \log \left[ \prod_{k=1}^n \prod_{i=1}^d \theta_i^{x_i^k} (1-\theta_i)^{1-x_i^k} \right]$$

Next page  $\Rightarrow$

$$\log(L) = \sum_{k=1}^n \sum_{i=1}^d [x_i^k \log(\theta_i) + (1-x_i^k) \log(1-\theta_i)]$$

$[\because \text{log will change } \prod \text{ to } \Sigma]$

Differentiate above eq. wrt  $\theta$

$$\frac{\partial \log L}{\partial \theta} = \sum_{k=1}^n \left[ \frac{x^k}{\theta} - \frac{(1-x^k)}{1-\theta} \right] \quad \begin{array}{l} (\sum_{i=1}^d \text{ is dropped off}) \\ \Downarrow \text{ summed up.} \end{array}$$

$$= 0 \quad \theta = (\theta_1, \theta_2, \dots, \theta_n)$$

$$\frac{\sum_{k=1}^n x^k}{\theta} - \frac{n - \sum x^k}{1-\theta} = 0 \Rightarrow \frac{\sum x^k}{\theta} = \frac{n - \sum x^k}{1-\theta}$$

$$\frac{1}{\theta} - 1 = \frac{n - \sum x^k}{\sum x^k} \Rightarrow$$

$$\hat{\theta}_{ML} = \frac{1}{n} \sum_{k=1}^n x^k$$

Q.2

$$p(x|w_1) \sim N(0, 1)$$

$$p(x|w_2) \sim N(\mu, 1) \quad \& \quad p(x|w_3) \sim N(1, 10^6)$$

$$p(w_1) = p(w_2) = \frac{1}{2}$$

↓ poor model

- a) To get estimate of  $\mu$  in poor model, we know that for gaussian model :-
- $$\mu = \bar{x} = \frac{1}{n} \sum x_i$$

given large data we will have

$$\hat{\mu}_{ML} = 1$$

- b) Decision Boundary (DB) :- for  $p(x|w_1) \sim N(0, 1)$  and  $p(x|w_2) \sim N(1, 1)$

$$p(w_1) = p(w_2) = \frac{1}{2}$$

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$$DB : \frac{1}{2} \sqrt{\frac{1}{2\pi}} \exp\left[-\frac{(x-0)^2}{1^2}\right] = \frac{1}{2} \sqrt{\frac{1}{2\pi}} \exp\left[-\frac{(x-1)^2}{1^2}\right]$$

$x: DB$

$$DB = x = \frac{1}{2}$$

So, we have  $DB = 0.5$  for the specified 2 classes.

c) Assumption / Note: In the question we are taking about true distribution, but we are provided the distribution of assumed DB class (i.e.  $N(\mu, 1)$ ) which we have already computed in b) part. So, I am assuming distribution to be  $N(1, 10^6)$

Two classes we have are :-

$$p(x|w_1) \sim N(0, 1) \quad p(x|w_2) \sim N(1, 10^6)$$

$$p(w_1) = p(w_2) = \frac{1}{2}$$

for DB :-

$$DB : \frac{P(x|w_1) \cdot P(w_1)}{P(x)} = \frac{P(x|w_2) \cdot P(w_2)}{P(x)}$$

$$\frac{1}{\sqrt{2\pi}} \exp\left[-\frac{(x-0)^2}{1^2}\right] = \frac{1}{\sqrt{2\pi \cdot 10^6}} \exp\left[-\frac{(x-1)^2}{10^6}\right]$$

Take log :-

$$-x^2 = -3(\ln 10) - \frac{(x-1)^2}{10^6}$$

$$-10^6 x^2 = -3 \times 10^6 (\ln 10) - (x^2 + 1 - 2x)$$

$$x^2 (10^6 - 1) + 2x - (3 \times 10^6 \ln 10 + 1) = 0$$

Solving the above quadratic equation gives:-

$$DB : x = 8.311, -8.311$$

So, b/w these 2 points, we choose  $w_1$ ,

else  $w_2$  :

$$DB = \begin{cases} w_1 & \text{if } x \in [8.311, 8.311] \\ w_2 & \text{o/w.} \end{cases}$$

Spiral

Q.3

$$D = x_1, x_2, \dots, x_n \quad \mu' = x_1$$

The true mean of data  $D$  is given by :-

$$\mu = \frac{1}{n} \sum_{i=1}^n x_i$$

$\mu'$ : True mean

a) To show :-  $\mu' = x_1$  is unbiased

$\mu' = x_1$  To show that  $\mu' = x_1$  is unbiased, we need to show that  $E(\mu')$  is equal to the true mean of data. So,

So, it is unbiased

$$E(\mu') = E(x_1) = \mu$$

True  
mean  
of data

$\left[ \because \text{expectation of any data point} = \text{mean of data} \right]$

b) It is undesirable :-

Consider the data :  $D = 10^6, -10^6$  with 2 data points

$$\text{True mean of } D = \mu = \frac{1}{2} (10^6 - 10^6) = 0$$

But the mean from above part  $\mu' = 10^6 = x_1$

If we compute mean absolute error :-

$$e = |\mu - \mu'| = |0 - 10^6| = 10^6 \rightarrow \text{Very high error}$$

So, choosing the first point as mean is a very bad method.