

### **Problem 1**

Let the function be denoted by  $g(\theta)$

$$g(\theta) = \theta^t(1 - \theta)^h$$

Differentiating w.r.t to  $\theta$  we get the first derivative

$$g'(\theta) = t\theta^{t-1}(1 - \theta)^h - h\theta^t(1 - \theta)^{h-1}$$

Differentiating w.r.t to  $\theta$  again, we get the second derivate:

$$g''(\theta) = t(t-1)\theta^{t-2}(1 - \theta)^h - 2th\theta^{t-1}(1 - \theta)^{h-1} + h(h-1)\theta^t(1 - \theta)^{h-2}$$

Now we let  $f(\theta) = \log g(\theta)$  and find the 1st and 2nd derivatives

$$f(\theta) = \log(g(\theta)) = t \log(\theta) + h \log(1-\theta)$$

$$f'(\theta) = \frac{t}{\theta} - \frac{h}{1-\theta}$$

$$\text{and } f''(\theta) = \frac{h}{(1-\theta)^2} - \frac{t}{\theta^2}$$

### **Problem 2**

We find points on  $f(\theta) = \log g(\theta)$  by equating  $f'(\theta)$  to 0

$$\Rightarrow \frac{t}{\theta} - \frac{h}{1-\theta} = 0$$

$$\Rightarrow \frac{t-(t+h)\theta}{\theta(1-\theta)} = 0$$

$$\Rightarrow t - (t+h)\theta = 0$$

$$\Rightarrow \theta = \frac{t}{t+h} \quad (1)$$

We substitute the value of  $\theta$  in  $f''(\theta)$  with that in equation (1) and get

$$f''(\theta) = \frac{(t+h)^2}{h} - \frac{(t+h)^2}{t} = \frac{t-h}{th}(t+h)^2 < 0 \text{ if } h > t$$

Again replacing the value of  $\theta$  in  $g'(\theta)$  with that in equation (1) we get

$$g'(\theta) = t \left[ \frac{t}{t+h} \right]^{t-1} \left[ 1 - \frac{t}{t+h} \right]^h - h \left[ \frac{t}{t+h} \right]^t \left[ 1 - \frac{t}{t+h} \right]^{h-1}$$

$$g'(\theta) = \frac{t^t h^h}{(t+h)^{t+h-1}} - \frac{t^t h^{h-1}}{(t+h)^{t+h-1}} = 0 \quad (2)$$

Hence, the critical points are same.

Now we find if it is a maximum or not.

Now  $g''(\theta)$  is given by:

$$g''(\theta) = t(t-1)\theta^{t-2}(1-\theta)^h - 2th\theta^{t-1}(1-\theta)^{h-1} + h(h-1)\theta^t(1-\theta)^{h-2}$$

$$g''\left(\frac{t}{t+h}\right) = t(t-1)\left(\frac{t}{t+h}\right)^{t-2}\left(1-\frac{t}{t+h}\right)^h - 2th\left(\frac{t}{t+h}\right)^{t-1}\left(1-\frac{t}{t+h}\right)^{h-1} + h(h-1)\left(\frac{t}{t+h}\right)^t\left(1-\frac{t}{t+h}\right)^{h-2}$$

On solving further we get,

$$g''\left(\frac{t}{t+h}\right) = -\frac{t^{t-1}h^{h-1}}{(t+h)^{t+h-1}} < 0 \text{ as } t, h \in \mathbb{N}$$

Therefore, log function retains the critical points of the main function

### **Problem 3**

We know that  $\theta_{\text{MAP}}$  is the maximum of the  $p(\theta = x | D)$  and is given by:

$$\theta_{\text{MAP}} = \frac{N_T + a - 1}{N + a + b - 1} \text{ and } \theta_{\text{MLE}} = \frac{N_T}{N}$$

If  $\theta_{\text{MAP}} = \theta_{\text{MLE}}$  then  $a=b=1$

This means that the prior distribution is uniform i.e there exists a prior  $p(\theta)$  such that the result holds. Also, we know that such a prior will always exist. Hence, it is true that  $\theta_{\text{MLE}}$  is a special case of  $\theta_{\text{MAP}}$

### **Problem 4**

$$\text{Now } \theta_{\text{MLE}} = \frac{m}{m+l} \text{ and } E_{\text{PR}}[\theta | a, b] = \frac{a}{a+b}$$

The consequent posterior distribution is given by the Beta distribution  $\text{Beta}(x | a+m, b+l)$  and the mean of this is:

$$E_{\text{PS}}[X] = \frac{a+m}{a+b+m+l} = \frac{a}{a+b+m+l} + \frac{m}{a+b+m+l}$$

$$\text{Let } 0 \leq \lambda \leq 1 \text{ be } \frac{a+b}{a+b+m+l}$$

Then we get,

$$\lambda E_{\text{PR}} = \left[ \frac{a+b}{a+b+m+l} \right] \frac{a}{a+b} = \frac{a}{a+b+m+l} \quad (4)$$

$$\text{and } (1 - \lambda)\theta_{\text{MLE}} = \left[ \frac{m+l}{a+b+m+l} \right] \frac{m}{m+l} = \frac{m}{a+b+m+l} \quad (5)$$

On adding equation (4) and (5) we have,  $\lambda E_{\text{PR}} +$

$$(1 - \lambda)\theta_{\text{MLE}} = E_{\text{PS}}$$

### **Problem 5**

The Poisson Distribution of X is given by

$$P(X = k|\lambda) = \frac{\lambda^k e^{-\lambda}}{k!}$$

Now, for n i.i.d samples for X the probability is given by:

$$\begin{aligned} P(D|\lambda) &= \prod_{i=1}^n \frac{\lambda^{k_i} e^{-\lambda}}{k_i!} \\ \Rightarrow P(D|\lambda) &= e^{-n\lambda} \prod_{i=1}^n \frac{\lambda^{k_i}}{k_i!} \end{aligned} \quad (6)$$

Taking log of equation (6) and denoting it by  $f(\lambda)$  we have,

$$f(\lambda) = -n\lambda + \left( \sum_{i=1}^n k_i \right) \log \lambda + \log \left( \prod_{i=1}^n k_i! \right)$$

Differentiating w.r.t to  $\lambda$  and equating to 0 we get

$$f'(\lambda) = -n + \frac{\sum_{i=1}^n k_i}{\lambda} = 0$$

$$\Rightarrow \lambda = \frac{\sum_{i=1}^n k_i}{n}$$

$$\text{Therefore we have } \theta_{\text{MLE}} = \frac{\sum_{i=1}^n k_i}{n}$$

Now we let the prior have a Gamma distribution with constants  $\alpha$  and  $\beta$

$$P(\lambda|\alpha, \beta) = \frac{\beta^\alpha}{\Gamma(\alpha)} \lambda^{\alpha-1} e^{-\lambda\beta}$$

Now the posterior distribution is given by:

$$P(\lambda|D) = \frac{P(D|\lambda) \cdot P(\lambda)}{P(D)}$$

On replacing values we get,

$$P(\lambda|D) = \frac{1}{P(D)} \cdot \frac{e^{-n\lambda} \lambda^{\sum k_i}}{\prod k_i!} \cdot \frac{\beta^\alpha}{\Gamma(\alpha)} \lambda^{\alpha-1} e^{-\lambda\beta}$$

$$\Rightarrow P(\lambda|D) = c e^{(-n+\beta)\lambda} \lambda^{(\alpha-1+\sum k_i)} \text{ where } c \text{ is a constant}$$

Again taking log of the above function we get,

$$(-n + \beta)\lambda + (\alpha - 1 + \sum k_i) \log \lambda$$

On finding the derivative and equating it to 0 we get,

$$\lambda = \frac{\alpha-1+\sum k_i}{(-n+\beta)} = \theta_{\text{MAP}}$$