

## Exercise 1

1) likelihood function  $= \prod_{i=1}^n \frac{\mu_i^k}{k!} \exp(-\mu_i)$

$$\log(\text{likelihood function}) = L = \sum_{i=1}^n [k \log(\mu_i) - \log(k!) - \mu_i]$$

2) Poisson distribution belongs to discrete probability distribution.

So can use  $\mu_i$  to represent the expected number of events happened per unit time. So  $\mu_i$  is continuous and non-negative. Firstly, because it is continuous, we can let  $w^T x_i + b$  (this is also continuous) relate to  $\mu_i$ . Secondly,  $\mu_i$  is non-negative, so we need use a mathematical way to express this limitation.  $\log(\mu_i)$  is reasonable because true number in logarithmic function is non-negative. Above all, we can let  $\log(\mu_i) = w^T x_i + b$

3) Now it is known that  $\log(\mu_i) = w^T x_i + b$ .

so,

$$\hat{w} = \arg \max_w L = \arg \max_w \sum_{i=1}^n [k \log(w^T x_i + b) - \log(k!) - w^T x_i - b]$$

likelihood function used the multiply of probability that  $x_i$  equals to  $k$  subject to  $x_i$ . For each input  $x_i$ , we want probability is high (this means high accuracy). If each probability is high, then the multiply of probability is high. Therefore,  $\hat{w} = \arg \max_w L$  can give resulted parameter maximizing the label accuracy.

$$4) \frac{dL}{dw} = \sum_{i=1}^n \left( \frac{Kx_i}{w^T x_i + b} - x_i \right)$$

$$\frac{dL}{db} = \sum_{i=1}^n \left( \frac{K}{w^T x_i + b} - 1 \right)$$

Input:  $X \in \mathbb{R}^{n \times d}$ ,  $Y \in \mathbb{R}^n$ ,  $w_0 = \mathbf{0}_d$ ,  $b_0 = 0$ ,  $\text{max\_pass} \in \mathbb{N}$ ,  $\eta > 0$ ,  $\text{tol} > 0$   
 output:  $w, b$   
 for  $t = 1, 2, \dots, \text{max\_pass}$  do  
    $w_t \leftarrow w_{t-1} - \eta \frac{dL}{dw}$   
    $b_t \leftarrow b_{t-1} - \eta \frac{dL}{db}$   
   if  $\|w_t - w_{t-1}\| \leq \text{tol}$  then  
     break  
 $w \leftarrow w_t, b \leftarrow b_t$

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