

# Representation Learning

Q3(a) Binomial

$$\alpha(x; p) = \prod_{i=1}^N {}^n C_{x_i} p^{x_i} (1-p)^{n-x_i}$$

$$l(x; p) = \sum_{i=1}^N \log \left( {}^n C_{x_i} p^{x_i} (1-p)^{n-x_i} \right)$$

$$= \sum_{i=1}^N \left[ \log {}^n C_x + x_i \log p + (n-x_i) \log (1-p) \right]$$

$$\min_p l(x; p) \Rightarrow \frac{dl}{dp} \geq 0$$

$$\Rightarrow \sum_{i=1}^N \left[ \frac{x_i}{p} + \left( \frac{n-x_i}{1-p} \right) (-i) \right] = 0$$

$$\frac{\sum_{i=1}^N x_i}{p} - \frac{nN}{(1-p)} + \frac{\sum_{i=1}^N x_i}{(1-p)} = 0$$

$$\sum_{i=1}^N x_i (1-p) = \left( nN - \sum_{i=1}^N x_i \right) \left( \frac{p}{N} \right)$$

$$\sum_{i=1}^N x_i = \sum_{i=1}^N x_i \cdot p = Np = \sum_{i=1}^N x_i p$$

$$p = \frac{\sum_{i=1}^N x_i}{n \cdot N}$$

$$b) f(x; \lambda) = \lambda^x e^{-\lambda}$$

$$L(x; \lambda) = \prod_{i=1}^N \frac{\lambda^{x_i} e^{-\lambda}}{x_i!}$$

$$\ell(x; \lambda) = \sum_{i=1}^N \log \left( \frac{\lambda^{x_i} e^{-\lambda}}{x_i!} \right)$$

$$= \sum_{i=1}^N \left[ x_i \log \lambda - \lambda - \log(x_i!) \right]$$

$$\frac{\partial \ell}{\partial \lambda} = 0$$

$$\Rightarrow \sum_{i=1}^N \left[ \frac{x_i}{\lambda} - 1 \right] = 0$$

$$\lambda = \frac{\sum_{i=1}^N x_i}{N}$$

$$c) f(x; \lambda) = \begin{cases} \lambda e^{-\lambda x} & x \in \mathbb{R} \\ 0 & x \notin \mathbb{R} \end{cases}$$

$$L(x; \lambda) = \prod_{i=1}^N f(x_i; \lambda)$$

$$\begin{aligned} \ell(x; \lambda) &= \sum_{i=1}^N \log(\lambda e^{-\lambda x_i}) \\ &= \sum_{i=1}^N [\lambda - \lambda x_i] \end{aligned}$$

$$\frac{\partial L}{\partial \lambda} = 0$$

$$\Rightarrow \sum_{i=1}^N [l_i - \bar{x}_i] = 0$$

$$\bar{x} = \frac{1}{N} \sum_{i=1}^N x_i$$

$$\bar{x} = \frac{N}{\sum_{i=1}^N x_i}$$

d)  $f(x; \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$

$$L(x; \mu, \sigma) = \sum_{i=1}^N \left[ \log\left(\frac{1}{\sqrt{2\pi}\sigma}\right) - \frac{(x_i - \mu)^2}{2\sigma^2} \right]$$

$$\frac{\partial L}{\partial \mu} = 0$$

$$\Rightarrow \sum_{i=1}^N \frac{2(x_i - \mu)}{2\sigma^2} = 0$$

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

$$\frac{\partial L}{\partial \sigma} = 0$$

$$\Rightarrow \sum_{i=1}^N \left[ \frac{1}{\sigma} - \frac{(x_i - \mu)^2}{2\sigma^3} \cdot (-2) \right] = 0$$

$$\Rightarrow \sum_{i=1}^N \frac{(x_i - \mu)^2}{\sigma^2} = N$$

$$\sigma = \sqrt{\frac{\sum_{i=1}^N (x_i - \mu)^2}{N}}$$

$$e) f(x; \mu, b) = \frac{1}{2b} e^{-|x-\mu|/b}$$

$$L(x; \mu, b) = \prod_{i=1}^N \frac{1}{2b} e^{-|x_i-\mu|/b}$$

$$\ell(x; \mu, b) = \sum_{i=1}^N \left[ \log \frac{1}{2b} - \frac{|x_i - \mu|}{b} \right]$$

$$\frac{\partial \ell}{\partial \mu} = 0$$

$$\Rightarrow \frac{\partial \ell}{\partial \mu} = \frac{1}{b} \sum_{i=1}^N \frac{\partial |x_i - \mu|}{\partial \mu}$$

$$\therefore \frac{\partial |x|}{\partial x} = \frac{\partial \sqrt{x^2}}{\partial x} = x(x^2)^{-1/2} = \frac{x}{|x|} = \text{sgn}(x)$$

$$\Rightarrow \frac{\partial \ell}{\partial \mu} = \frac{\sum_{i=1}^N \text{sgn}(x_i - \mu)}{b} = 0$$

If N is odd &  $\underline{x}$  is sorted

$$\sum_{i=1}^N \text{sgn}(x_i - \mu) = 0 \text{ when } \mu_{MLE} = \frac{x_{N+1}}{2}$$

If N is even and  $\underline{x}$  is sorted

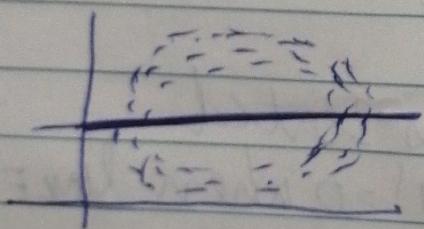
$$\sum_{i=1}^N \text{sgn}(x_i - \mu) = 0 \text{ when } \mu_{MLE} = \frac{(x_{N/2} + x_{N/2+1})}{2}$$

$$\frac{\partial L}{\partial b} = 0$$

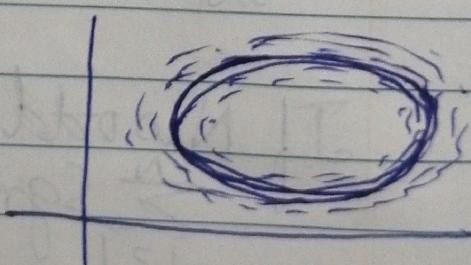
$$\Rightarrow \sum_{i=1}^N \left( -\frac{1}{b} \right) + \sum_{i=1}^N \frac{|x_i - \mu_{MLE}|}{b^2} = 0$$

$$\therefore b = \frac{\sum_{i=1}^N |x_i - \mu_{MLE}|}{N}$$

Q2 ① The Standard PCA always finds linear principal components to represent the data in lower dimension. Sometimes we need non-linear principal components.



PCA  
(linear)



PCA  
(nonlinear)

② Suppose a simple case with 3 independent variables  $x_1, x_2, x_3$  and the output  $y$  and suppose now that  $x_3 = y$ . and so you should be able to get a 0 error model.

Now suppose that in the training set the

Variation of  $y$  is very small and so also the variation of  $x_3$ .

Now if you run PCA and you decide to select only 2 variables you will obtain a combination of  $x_1$  and  $x_2$ , so the information of  $x_3$  that was the only variable able to explain  $y$  is lost.