

Q3)

$$p = \binom{n}{x} p^x (1-p)^{n-x}$$

$$L(x, p) = \prod_{i=1}^n \frac{n!}{x_i! (n-x_i)!} p^{x_i} (1-p)^{n-x_i}$$

$$l(x, p) = \log(L(x, p))$$

$$= \sum_{i=1}^n \log \left( \frac{n!}{x_i! (n-x_i)!} p^{x_i} (1-p)^{n-x_i} \right)$$

$$= \sum_{i=1}^n \log \left( \frac{n!}{x_i! (n-x_i)!} \right) + \sum_{i=1}^n x_i \log(p) + \sum_{i=1}^n (n-x_i) \log(1-p)$$

For min,

$$\frac{dl}{dp} = 0$$

$$\Rightarrow \sum_{i=1}^n x_i \frac{1}{p} - \sum_{i=1}^n (n-x_i) \frac{1}{(1-p)} = 0$$

$$\Rightarrow \left( \sum x_i - \sum x_i \right) \left[ \frac{1}{p} - \frac{1}{1-p} \right] \quad \text{Let } \sum_{i=1}^n x_i = x$$

$$\Rightarrow \frac{x}{p} - \frac{(n-x)}{1-p} = 0$$

$$\Rightarrow x(1-p) - (n-x)p = 0$$

$$\Rightarrow x - xp + xp - np = 0$$

$$\Rightarrow p = \frac{x}{n} = \frac{\sum_{i=1}^n x_i}{n}$$

$$b) f_x(x; \lambda) = \frac{\lambda^x e^{-\lambda}}{x!}$$

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

$$l(x; \lambda) = \log(L(x; \lambda))$$

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

$$= \sum_{i=1}^N x_i \log \lambda + \sum_{i=1}^N (-\lambda) - \sum_{i=1}^N \log(x_i!)$$

$$\frac{d}{d\lambda} l = 0$$

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

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

$$c) f_x(x) = \begin{cases} \lambda e^{-\lambda x} & x \in R_x \\ 0 & x \notin R_x \end{cases}$$

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

$$l(x, \lambda) = \sum_{i=1}^N \log(\lambda e^{-\lambda x_i})$$

$$= \sum_{i=1}^N \log(\lambda) - \sum_{i=1}^N \lambda x_i$$

for min,

$$\frac{d}{d\lambda} l = 0$$

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

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

$$d) f_X(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right];$$

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

For min,

$$\frac{\partial}{\partial \mu} \log[L(x, \theta, \mu, \sigma^2)] = 0 \quad \& \quad \frac{\partial}{\partial \sigma^2} \log[L(x, \theta, \mu, \sigma^2)] = 0$$

Solving, we get,  $\mu_{MLE} = \frac{\sum_{i=1}^N x_i}{N}$ ;  $\sigma^2 = \frac{\sum_{i=1}^N (x_i - \mu_{MLE})^2}{N}$

e)

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

$$L(x, \theta) = \prod_{i=1}^N \frac{1}{2b} e^{-\frac{|x_i - \mu|}{b}} \quad \theta \in [\mu, b]$$

$$\begin{aligned} \ell(L(x, \theta)) &= \sum_{i=1}^N \log\left(\frac{1}{2b} e^{-\frac{|x_i - \mu|}{b}}\right) \\ &= \sum_{i=1}^N \log\left(\frac{1}{2b}\right) - \sum_{i=1}^N \frac{|x_i - \mu|}{b} \end{aligned}$$

For min,

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

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

$$\Rightarrow -\frac{N}{b} + \frac{\sum_{i=1}^N |x_i - \mu|}{b^2} = 0$$

$$\Rightarrow b = \frac{\sum_{i=1}^N |x_i - \mu|}{N}; \quad \mu \text{ here is } \mu_{MLE}$$

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

$$\Rightarrow \sum_{i=1}^N \frac{\partial}{\partial \mu} |x_i - \mu| = 0$$

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

$\therefore$  for this to happen, there are 2 cases -

$\rightarrow$   $N$  is odd, then we choose

$$\mu_{MLE} = \text{median}(x_1, \dots, x_N)$$

In this case,  $\text{sgn}(x_{\text{median}} - \mu_{MLE}) = 0$

& other terms will cancel each other out

$\rightarrow$   $N$  is even, then we choose

$\mu_{MLE}$  to be a value bet<sup>n</sup> the 2

medians of the data.

This way all the terms to the left of  $\mu_{MLE}$  in ~~arranged~~ sorted order will cancel out the ones on the right.

Q2)

Ans - If we have similar, high variances along 2 perpendicular directions, then PCA might fail as ~~into~~ finding an axis system to reduce covariance will be difficult.