

Let $x^{(1)}, \dots, x^{(n)} \stackrel{\text{iid}}{\sim} p$

$x^{(1)}, \dots, x^{(n)} \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$

$x^{(1)}, \dots, x^{(n)} \stackrel{\text{iid}}{\sim} \text{Ber}(\theta)$

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

$x \sim \text{Ber}(\theta) \quad p(x=1) = \theta, \quad p(x=0) = 1-\theta$

Q. $\{x^{(1)}, \dots, x^{(n)}\}$ find parameters

$\hat{\theta} = \frac{n_1}{n}$ for $\text{Ber}(\theta)$

$\bar{x} = \frac{1}{n} \sum_{i=1}^n x^{(i)}, \quad s^2 = \frac{1}{n} \sum_{i=1}^n (x^{(i)} - \bar{x})^2$

$E(\theta) = \frac{E(n_1)}{n} = \frac{1}{n} \sum_{i=1}^n E(\mathbb{1}(x^{(i)}=1)) = \frac{n\theta}{n} = \theta$

$E(\bar{x}) = \mu, \quad s^2 = \frac{1}{n} \sum_{i=1}^n (z_i - \bar{z})^2$

$x_i = z_i + \mu$

$\bar{x} = \bar{z} + \mu$

$E(s^2) = \frac{1}{n} \left(\sum_{i=1}^n z_i^2 - \bar{z}^2 n \right)$

$\frac{\sigma^2 n}{n} = \sigma^2$

Maximum likelihood

$$\mathcal{L}_n(\theta) = \sum_{i=1}^n \log p(x^{(i)} | \theta)$$

$$\hat{\theta}_n = \max_{\theta} \mathcal{L}_n(\theta)$$

$$g(t) = t - \log t$$

$$\min_t g(t) = ?$$

$$g'(t) = 1 - \frac{1}{t}$$

$$g''(t) = \frac{1}{t^2}$$

$g(t)$ is convex over $t > 0$

$g'(1) = 0$ and hence it is the minimum

$$g(1) \leq t - \log t \quad t > 0$$

$$1 \leq t - \log t$$

$$1 \leq t - 1, \quad t > 0,$$

$$\text{Let } p_i > 0, \quad \sum_{i=1}^d p_i = 1$$

$$q_i > 0 \quad \sum_{i=1}^d q_i = 1$$

$$P(X = a_i) = p_i$$

$$\text{or } P(X = a_i) = q_i$$

$$\log \frac{q_i}{p_i} \leq \frac{q_i}{p_i} - 1$$

$$\sum_{i=1}^d p_i \log \frac{q_i}{p_i} \leq \sum_{i=1}^d q_i - \sum_{i=1}^d p_i = 0$$

$$KL(P, Q) = \sum_{i=1}^d P_i \log \frac{P_i}{Q_i}$$

$$KL(P, Q) \geq 0$$

Equality holds iff $P = Q$

P is X with p.m.f $\{P_1, \dots, P_d\}$

Q is X with p.m.f $\{Q_1, \dots, Q_d\}$

$$KL(P, Q) = 0 \quad \text{iff} \quad P_i = Q_i \quad i = \{1, \dots, d\}$$

$X \sim \text{Ber}(\theta) \quad \mathcal{X} = \{x^{(1)}, \dots, x^{(n)}\}$

$$p(x^{(i)} | \theta) = \theta^x (1 - \theta)^{(1-x)}$$

$$\mathcal{L}(\theta) = \sum_{i=1}^n \left(x^{(i)} \log \theta + (1 - x^{(i)}) \log (1 - \theta) \right)$$

$$\mathcal{L}(\theta) = n_1 \log \theta + n_0 \log (1 - \theta)$$

$$n_1 = \sum_{i=1}^n x^{(i)}, \quad n_0 = \sum_{i=1}^n (1 - x^{(i)})$$

$$\hat{\theta} = \frac{n_1}{n} \quad 1 - \hat{\theta} = \frac{n_0}{n}$$

$$n = n_1 + n_0$$

$$\mathcal{L}(\theta) = n (\hat{\theta} \log \theta + (1 - \hat{\theta}) \log (1 - \theta))$$

$$\mathcal{L}(\theta) = n \left(\hat{\theta} \log \frac{\theta}{\hat{\theta}} + (1 - \hat{\theta}) \log \frac{(1 - \theta)}{(1 - \hat{\theta})} \right) \\ + n (\hat{\theta} \log \hat{\theta} + (1 - \hat{\theta}) \log (1 - \hat{\theta}))$$

$$S(P) = - \sum_{i=1}^d p_i \log p_i \quad \text{Entropy}$$

$$\mathcal{L}(\theta) = -n \text{KL}(P, Q(\theta)) \\ = -n S(P)$$

$$d=2, \quad a_1=1, \quad a_2=0$$

$$\underline{P} \quad P(X=1) = \hat{\theta}$$

$$P(X=0) = (1-\hat{\theta})$$

$$\underline{Q} \quad P(X=1) = \theta$$

$$P(X=0) = 1-\theta$$

$$\theta^*, \arg\max_{\theta} \ell(\theta)$$

$$= \arg\min_{\theta} KL(P, Q(\theta))$$

$$KL(P, Q(\theta)) = 0 \quad \text{iff}$$

$$\theta_n = \hat{\theta}$$

$$X \sim N(\mu, C), \quad C \succ 0$$

$$P(X=x|\theta) = \frac{1}{(\sqrt{2\pi})^d |C|^{1/2}} e^{-\frac{1}{2}(x-\mu)^T C^{-1}(x-\mu)} \quad |C| = \det(C)$$

$$\ell(\theta) = \sum_{i=1}^n \log P(X=x^{(i)}|\theta)$$

$$= \sum_{i=1}^n \left(\log \frac{1}{(\sqrt{2\pi})^d} - \frac{1}{2} \log |C| - \frac{1}{2} (x^{(i)} - \mu)^T C^{-1} (x^{(i)} - \mu) \right)$$

For a fixed C

$$f(\mu) = -\frac{1}{2} \sum_{i=1}^n (x^{(i)} - \mu)^T C^{-1} (x^{(i)} - \mu)$$

$$\nabla f(\mu) = C^{-1} \sum_{i=1}^n (\mu - x^{(i)})$$

$$H(\mu) = n C^{-1}$$

$$f(\mu) \geq f(\mu^*) \quad \left| \quad C \geq 0 \right.$$

$$\text{st } \nabla f(\mu^*) = 0$$

$$\mu^* = \frac{1}{n} \sum x^{(i)} = \bar{x}$$

$$\mathcal{L}(\mu, C) \leq \mathcal{L}(\bar{x}, C)$$

$$S = \frac{1}{N} \sum_{i=1}^N (x^{(i)} - \bar{x})(x^{(i)} - \bar{x})^T$$

$$\mathcal{L}(\mu, C)$$

$$= \sum_{i=1}^n \left(\log \frac{1}{(\sqrt{2\pi})^d} - \frac{1}{2} \log |C| - \frac{1}{2} (x^{(i)} - \mu)^T C^{-1} (x^{(i)} - \mu) \right)$$

$$= n \left(\log \frac{1}{(\sqrt{2\pi})^d} - \frac{1}{2} \log |C| \right) - \frac{1}{2} \text{Tr} \left(C^{-1} \sum_{i=1}^n (x^{(i)} - \mu)(x^{(i)} - \mu)^T \right)$$

$$\mathcal{L}(\bar{x}, S) = n \left(\log \frac{1}{(\sqrt{2\pi})^d} - \frac{1}{2} \log |S| \right) - \frac{1}{2} \text{Tr} \left(S^{-1} \sum_{i=1}^n (x^{(i)} - \bar{x})(x^{(i)} - \bar{x})^T \right)$$

$$= n \left(\log \frac{1}{(\sqrt{2\pi})^d} - \frac{1}{2} \log |S| \right)$$

$$- \frac{n}{2} \text{Tr} (S^{-1} S)$$

$$= n \left(\log \frac{1}{(\sqrt{2\pi})^d} - \frac{1}{2} \log |S| \right) - \frac{n}{2} d$$

$$\mathcal{L}(\bar{x}, C) =$$

$$n \left(\log \frac{1}{(\sqrt{2\pi})^d} - \frac{1}{2} \log |C| \right) - \frac{1}{2} \text{Tr} \left(C^{-1} \sum_{i=1}^n (x^{(i)} - \bar{x})(x^{(i)} - \bar{x})^T \right)$$

$$= n \left(\log \frac{1}{(\sqrt{2\pi})^d} - \frac{1}{2} \log |C| \right)$$

$$- \frac{n}{2} \text{Tr} (C^{-1} S)$$

$$\mathcal{L}(\bar{x}, S) - \mathcal{L}(\bar{x}, C)$$

$$= -\frac{n}{2} \log |S| + \frac{n}{2} \log |C| - \frac{nd}{2} + \frac{n}{2} \text{Tr} (C^{-1} S)$$

$$= \frac{n}{2} \left(-\log \frac{|S|}{|C|} - d + \text{Tr}(C^{-1} S) \right)$$

$$A = C^{-1}S$$

$$\det(A) = \frac{\det(S)}{\det(C)}$$

$$\sum_{i=1}^d \lambda_i(A) = \det(A)$$

$$\text{Trace}(A) = \sum_{i=1}^d \lambda_i(A)$$

$$\mathcal{L}(\bar{X}, S) - \mathcal{L}(\bar{X}, C)$$

$$= \frac{n}{2} \left(-\log |A| - d + \text{Tr}(A) \right)$$

$$= \frac{n}{2} \sum_{i=1}^d \left(-\log \lambda_i(A) - 1 + \lambda_i(A) \right)$$

$$\geq 0$$

Equality holds $\iff \lambda_i(A) = 1$
 $\quad \quad \quad = A = I$

$$\therefore \mathcal{L}(\mu, c) \leq \mathcal{L}(\bar{x}, c) \leq \mathcal{L}(\bar{x}, s)$$

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