

1.3 两者均属于指数族函数, MLE 正确性无需二阶求导验证

$$a. p(x; \mu) = \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{1}{2}(x-\mu)^2\right\}$$

$$p(\vec{x}; \mu) = \frac{1}{(2\pi)^{N/2}} \exp\left\{-\frac{1}{2} \sum_{n=0}^{N-1} (x[n] - \mu)^2\right\}$$

$$\frac{d \ln p(\vec{x}; \mu)}{d\mu} = \sum_{n=0}^{N-1} (x[n] - \mu) = 0 \Rightarrow \hat{\mu} = \frac{1}{N} \sum_{n=0}^{N-1} x[n] = \bar{x} \quad \text{可看作 WGN 模型中的直流分量}$$

$$b. p(x; \lambda) = \lambda \exp\{-\lambda x\} u(x).$$

$$p(\vec{x}; \lambda) = \begin{cases} \lambda^N \exp\{-\lambda \sum_{n=0}^{N-1} x[n]\}, & x[n] > 0 \text{ for all } n \\ 0, & \text{otherwise.} \end{cases}$$

$$\frac{d \ln p(\vec{x}; \lambda)}{d\lambda} = \frac{\partial}{\partial \lambda} \left( \sum_{n=0}^{N-1} \ln \lambda - \lambda \sum_{n=0}^{N-1} x[n] \right) = \frac{N}{\lambda} - \sum_{n=0}^{N-1} x[n] = 0 \Rightarrow \hat{\lambda} = \frac{1}{\bar{x}} \quad \text{是对独立指数分布参数 } \lambda \text{ 的估计.}$$

指数分布中  $E[X] = \frac{1}{\lambda}$ , 因此该估计很合理

$$1.8. \quad p(x; p) = \prod_{n=0}^{N-1} p^{x[n]} (1-p)^{1-x[n]} = p^{\sum_{n=0}^{N-1} x[n]} (1-p)^{N - \sum_{n=0}^{N-1} x[n]}$$

$$\ln p(x; p) = \sum_{n=0}^{N-1} x[n] \ln p + (N - \sum_{n=0}^{N-1} x[n]) \ln(1-p).$$

$$\frac{d \ln p(x; p)}{dp} = \frac{\sum_{n=0}^{N-1} x[n]}{p} - \frac{N - \sum_{n=0}^{N-1} x[n]}{1-p} = 0 \Rightarrow \hat{p} = \bar{x}$$

$$1.9. \quad p(x; \theta) = \begin{cases} \frac{1}{\theta}, & 0 \leq x \leq \theta \\ 0, & \text{其他} \end{cases}$$

$$p(\vec{x}; \theta) = \begin{cases} \frac{1}{\theta^N}, & x[n] \in [0, \theta] \text{ for all } n \\ 0, & \text{其他} \end{cases}$$

当  $p(\vec{x}; \theta) \neq 0$  时,  $\theta \downarrow p(\vec{x}; \theta) \uparrow$ , 但是  $\theta \geq x[n], n=0, \dots, N-1$

所以  $\theta$  能取到的 min 为  $x_{\max}$ , 即  $\hat{\theta} = \max\{x[n]\}$

$$1.17 \quad x[n] \stackrel{\text{iid}}{\sim} N(0, \sigma^2)$$

$$\text{令 } \alpha = \frac{1}{\sigma^2}, \text{ 则 } \hat{\sigma} = \frac{1}{\alpha}$$

$$p(x; \alpha) = \frac{1}{(2\pi\alpha)^{N/2}} \exp\left\{-\frac{1}{2\alpha} \sum_{n=0}^{N-1} x^2[n]\right\}$$

$$0 = \frac{d \ln p(x; \alpha)}{d\alpha} = -\frac{N}{2} \cdot \frac{1}{\alpha} + \frac{1}{2\alpha^2} \sum_{n=0}^{N-1} x^2[n] \Rightarrow \hat{\alpha} = \frac{1}{N} \sum_{n=0}^{N-1} x^2[n]$$



$$17) \hat{\theta} = \frac{N}{\sum x[n]}$$

$$I(\alpha) = \frac{N}{2\alpha^2}$$

$$I'(\alpha) = I'(\theta) \left( \frac{\partial \theta}{\partial \alpha} \right)^2$$

$$\begin{aligned} I'(\theta) &= -\frac{2\alpha^2}{N} \left( \frac{\partial \theta}{\partial \alpha} \right)^2 \\ &= -\frac{2\alpha^2}{N} \left( -\frac{1}{\alpha^2} \right)^2 \\ &= -\frac{2}{N} \end{aligned}$$

$$\text{则 } \hat{\theta} \sim N(\theta, \frac{2}{N})$$

$$\text{记 } g(f_0) = \sum_n x[n] \cos 2\pi f_0 n$$

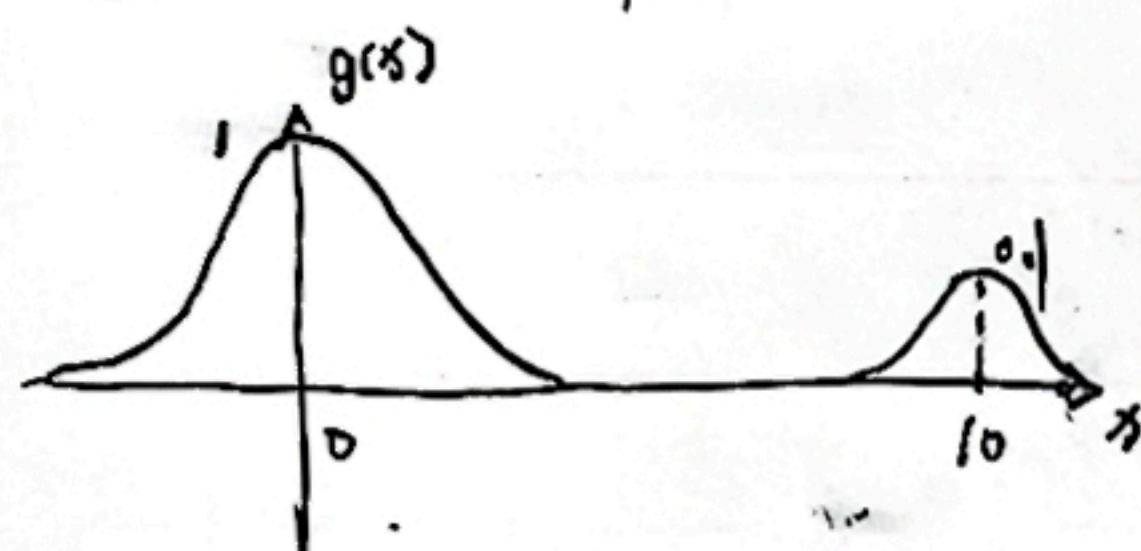
$$g'(f_0) = -\sum_n 2\pi n x[n] \sin 2\pi f_0 n$$

$$g''(f_0) = -\sum_n (2\pi n)^2 x[n] \cos 2\pi f_0 n$$

$$\text{NR: } f_{0,k+1} = f_{0,k} - \frac{g'(f_{0,k})}{g''(f_{0,k})}$$

数值求解略

$$7.18. \quad g(x) = e^{-\frac{x^2}{2}} + 0.1 e^{-\frac{1}{2}(x-10)^2}$$



$$g'(x) = -x e^{-\frac{x^2}{2}} - 0.1(x-10) e^{-\frac{1}{2}(x-10)^2}$$

$$g''(x) = x^2 e^{-\frac{x^2}{2}} + 0.1(x-10)^2 e^{-\frac{1}{2}(x-10)^2} - 0.1 e^{-\frac{1}{2}(x-10)^2} - e^{-\frac{x^2}{2}}$$

$$x_{k+1} = x_k - \frac{g'(x_k)}{g''(x_k)}$$

当  $x_0 = 0.5$  时收敛到  $x=0$

当  $x_0 = 9.5$  时收敛到  $x=10$

NR可迭代收敛至局部最优,不一定是全局最优,因此选择初值要尽可能接近真值

$$7.19. \quad x[n] = \cos 2\pi f_0 n + w[n]$$

$$p(x; f_0) = \frac{1}{(2\pi\sigma^2)^{N/2}} \exp\left\{-\frac{1}{2\sigma^2} \sum_n (x[n] - \cos 2\pi f_0 n)^2\right\}$$

$$\ln p(x; f_0) = -\frac{N}{2} \ln(2\pi\sigma^2) - \frac{1}{2\sigma^2} \sum_n (x[n] - \cos 2\pi f_0 n)^2$$

$$\text{只需 min } \sum_n (x[n] - \cos 2\pi f_0 n)^2 = \sum_n x^2[n] - 2 \sum_n x[n] \cos 2\pi f_0 n + \sum_n \cos^2 2\pi f_0 n$$

$$\approx \sum_n x^2[n] - 2 \sum_n x[n] \cos 2\pi f_0 n + \frac{N}{2} + \frac{1}{2} \sum_n \cos 4\pi f_0 n \approx 0 \text{ 当 } 0 < f_0 < \frac{1}{2}$$

$$\Rightarrow \text{只需 max } \sum_n x[n] \cos 2\pi f_0 n$$