

1. Let $Y = X - \mu$

$$\Rightarrow f_Y(y) = f_X(y+\mu) \left| \frac{dy+\mu}{dy} \right|$$

$$= \frac{1}{\sigma} \exp\left(-\frac{y}{\sigma}\right), \quad y \geq 0$$

is pdf of $\text{Neg}\left(\frac{1}{\sigma}\right)$

$$\Rightarrow Y \sim \text{Neg}\left(\frac{1}{\sigma}\right)$$

$$\Rightarrow EX = E(Y + \mu) = \sigma + \mu = \bar{X}$$

$$\text{Var } X = \text{Var}(Y + \mu) = \text{Var } Y = \sigma^2 = S^{*2} = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$$

$$\Rightarrow \hat{\sigma}_{\text{MOM}} = S^*$$

$$\hat{\mu}_{\text{MOM}} = \bar{X} - S^*$$

$$\Rightarrow \bigwedge_{\text{MOM}} = \widehat{e^{\sum \log(\hat{\sigma})_{\text{MOM}}}} = e^{-\hat{\mu}} \log(\hat{\sigma}) = e^{-(\bar{X} - S^*)} \log(S^*)$$

$$= \mu^2 + \frac{\sigma^2}{n} - \frac{\sigma^2}{n} = \mu^2$$

Hence $\bar{X} - \frac{S^2}{n}$ is UMVUE for μ^2 ✓

Goal: Check CRLB is attained

3. $X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$, $\mu \in \mathbb{R}, \sigma^2 > 0$ unknown, $n \geq 5$ (let $\theta = (\mu, \sigma^2)$)CRLB for μ^2 is $\frac{(\frac{\partial}{\partial \mu} \mu^2)^2}{I_n(\mu)}$

$$(a) f(x; \theta) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2\sigma^2}(x_i - \mu)^2\right)$$

$$= (\sqrt{2\pi}\sigma)^{-n} \exp\left(-\frac{1}{2\sigma^2} \sum (x_i - \mu)^2\right)$$

$$= (\sqrt{2\pi}\sigma)^{-n} \exp\left(-\frac{1}{2\sigma^2} (\sum x_i^2 - 2\mu \sum x_i + n\mu^2)\right)$$

$$= (\sqrt{2\pi}\sigma)^{-n} \exp\left(-\frac{1}{2\sigma^2} \sum x_i^2 + \frac{\mu}{\sigma^2} \sum x_i - \frac{1}{2\sigma^2} n\mu^2\right)$$

$$= C(\theta) \cdot \exp\left(\underbrace{\left(\frac{\mu}{\sigma^2} \sum x_i - \frac{1}{2\sigma^2} n\mu^2\right)}_{\downarrow \quad \downarrow} \underbrace{\left(\sum x_i - n\mu\right)}_{\downarrow} \right) h(x)$$

$$= \frac{1}{\sigma^4} E((X - \mu)^2) = \text{Var } X$$

$$\text{and } \mu = \mathbb{R}^1 \perp \theta$$

 $\Rightarrow f(x; \theta) \in 2\text{-par. exp. family}$

$$B = \{(\alpha_1(\theta), \alpha_2(\theta)) \mid \theta \in \mathbb{R} \times (0, \infty)\}$$

$$= \left\{ \left(-\frac{1}{2\sigma^2}, \frac{\mu}{\sigma^2} \right) \mid \mu \in \mathbb{R}, \sigma^2 > 0 \right\}$$

$$= (-\infty, 0) \times \mathbb{R}$$

$$\Rightarrow B^0 \neq \emptyset$$

$$\Rightarrow T' = (T_1, T_2) = \left(\sum_{i=1}^n X_i^2, \sum_{i=1}^n X_i \right)$$

is suff. stat. and complete

$$\therefore \left(\sum_{i=1}^n X_i^2, \sum_{i=1}^n X_i \right) \xrightarrow{1-1} (\bar{X}, S^2)$$

 $\therefore T = (\bar{X}, S^2)$ also is suff. stat.

and complete

this is easy

$$\text{Var}(\bar{X}^2 - \frac{1}{n} S^2) \quad \because \bar{X} \perp S^2 \quad \text{Basu's thm}$$

$$= \text{Var}(\bar{X}^2) + \frac{1}{n^2} \text{Var}(S^2) \quad \checkmark$$

$$\neq \text{Var}(\bar{X})$$

5. (a) $X_1, \dots, X_n \stackrel{iid}{\sim} f(x, \theta) = \theta e^{-\theta x}, \theta \in (0, \infty)$

$\therefore$ CRLB for θ is not attained

$$f(x; \theta) = \prod_{i=1}^n \theta e^{-\theta x_i}$$

$$S(b) L(\theta|x) = f(x; \theta)$$

$$= \theta^n e^{-\theta \sum x_i}$$

$$\log L(\theta|x) = n \log \theta - \theta \sum x_i$$

$$\frac{\partial}{\partial \theta} \log L(\theta|x) = \frac{n}{\theta} - \sum x_i > 0 \text{ if } \frac{n}{\theta} > \sum x_i \Leftrightarrow \theta < \frac{n}{\sum x_i}$$

$$w/ X = \{x | x_i > 0\} \cup \theta$$

$$= 0 \text{ if } \theta = \frac{n}{\sum x_i}$$

$\Rightarrow f(x; \theta) \in 1$ -par. exp. family

$$< 0 \text{ if } \theta > \frac{n}{\sum x_i}$$

$$\therefore B = \{ \eta(\theta) | \theta \in (0, \infty) \}$$

$$\therefore \hat{\theta}_{MLE} = \frac{n}{\sum X_i}$$

$$= (-\infty, 0)$$

$$\eta(\theta)_{MLE} = \widehat{\eta(\theta)}_{MLE} = \widehat{\eta(\theta)}_{MLE} = \widehat{\theta} e^{-\widehat{\theta} c}$$

$$\Rightarrow B^0 \neq \emptyset$$

$$= \frac{n}{\sum X_i} e^{-\frac{n}{\sum X_i} c}$$

$\Rightarrow T = \sum_{i=1}^n X_i$ is suff. stat. and complete

$$T \text{ is } E(T) = E(\sum x_i) = n \frac{1}{\theta}$$

$$b. n \geq 3, X_i \stackrel{iid}{\sim} P(c_i \theta), c_i > 0 \text{ known}$$

$$f(x_i; \theta) = \frac{e^{c_i \theta}}{\Gamma(c_i)} (c_i \theta)^{c_i-1}$$

$$\text{know } T = \sum_{i=1}^n X_i \sim \text{Gamma}(n, \frac{1}{\theta})$$

$$(a) \int \eta(\theta) = \int L(x; \theta) + \dots + \int L(x_n; \theta)$$

$$T \text{ is } E\left(\frac{1}{T}\right) = \int_0^\infty t^{-1} \frac{1}{\Gamma(n) (\frac{1}{\theta})^n} t^{n-1} \exp(-\theta t) dt$$

$$\text{which } \int x_i(\theta) = E\left(\left(\frac{\partial}{\partial \theta} \log f(x_i; \theta)\right)^2\right)$$

$$= \int_0^\infty \frac{1}{\Gamma(n-1) (\frac{1}{\theta})^{n-1}} t^{n-2} \exp(-\theta t) dt \cdot \frac{\Gamma(n-1)}{\Gamma(n)} \theta$$

$$= E\left(\left(\frac{\partial}{\partial \theta} (-c_i \theta + X_i \log c_i \theta - \log X_i!)\right)^2\right)$$

$$= \underbrace{\text{pdf of Gamma}(n-1, \frac{1}{\theta})}_{\text{pdf of Gamma}(n-1, \frac{1}{\theta})}$$

$$= E\left(\left(-c_i + c_i \frac{X_i}{c_i \theta}\right)^2\right)$$

$$= \frac{\theta}{n-1}$$

$$= \frac{1}{\theta} E\left((X - c_i \theta)^2\right)$$

$$\therefore E\left(\frac{n-1}{T}\right) = \theta$$

$$= \frac{1}{\theta} \text{Var}(X)$$

$$\Rightarrow \frac{n-1}{T} = \frac{n-1}{\sum X_i} \text{ is UMVUE for } \theta$$

$$= \frac{c_i}{\theta}$$

Goal: Check CRLB for θ is attained

$$\therefore \text{CRLB} = \frac{(\frac{\partial}{\partial \theta} \log f(\theta))^2}{L(\theta)} = \frac{1^2}{(\frac{\theta}{n} + \dots + \frac{\theta}{n})} = \frac{\theta}{\sum c_i} \checkmark$$

$$L(\theta) = E\left(\left(\frac{\partial}{\partial \theta} \log f(x; \theta)\right)^2\right)$$

$$(b) L(\theta|x) = f(x; \theta)$$

$$= E\left(\left(\frac{\partial}{\partial \theta} \log \theta - \theta x\right)^2\right)$$

$$= \prod_{i=1}^n e^{-c_i \theta} \frac{(c_i \theta)^{c_i-1}}{c_i!}$$

$$= E\left(\left(\frac{1}{\theta} - x\right)^2\right)$$

$$= e^{-\theta \sum c_i} \prod_{i=1}^n (c_i \theta)^{c_i-1} \prod_{i=1}^n \frac{1}{c_i!}$$

$$= \text{Var } X = \frac{1}{\theta^2}$$

$$\log L(\theta|x) = -\theta \sum c_i + \sum_{i=1}^n X_i \ln c_i \theta + \sum_{i=1}^n (-\ln X_i!)$$

$$\text{CRLB} = \frac{(\frac{\partial}{\partial \theta} \log f(\theta))^2}{L(\theta)} = \frac{\theta^2}{n}$$

$$\frac{\partial}{\partial \theta} \log L(\theta|x) = -\sum c_i + \sum_{i=1}^n \frac{X_i}{c_i \theta} c_i$$

$$\text{Var}\left(\frac{n-1}{T}\right) = (n-1)^2 \text{Var}\left(\frac{1}{T}\right)$$

$$= -\sum c_i + \sum_{i=1}^n \frac{X_i}{\theta}$$

$$\text{where } \text{Var}\left(\frac{1}{T}\right) = \left(E\left(\frac{1}{T}\right) - \left(E\left(\frac{1}{T}\right)\right)^2\right) = \left(\frac{\theta}{n-1}\right)^2$$

$$= -\sum c_i + \frac{1}{\theta} \sum_{i=1}^n X_i > 0 \text{ if } \frac{1}{\theta} \sum_{i=1}^n X_i > \sum c_i \Leftrightarrow \theta < \frac{\sum X_i}{\sum c_i}$$

$$\text{where } E\left(\frac{1}{T}\right)^2 = \int_0^\infty t^{-2} \frac{1}{\Gamma(n) (\frac{1}{\theta})^n} t^{n-1} e^{-\theta t} dt$$

$$= 0 \text{ if } \theta = \frac{\sum X_i}{\sum c_i}$$

$$= \frac{\Gamma(n-2)}{\Gamma(n)} \theta^{-2} \int_0^\infty \frac{1}{\Gamma(n-2) (\frac{1}{\theta})^{n-2}} t^{n-3} e^{-\theta t} dt$$

$$< 0 \text{ if } \theta > \frac{\sum X_i}{\sum c_i}$$

$$= \frac{\theta^2}{(n-1)(n-2)}$$

$$\therefore \hat{\theta}_{MLE} = \frac{\sum X_i}{\sum c_i} \checkmark$$

$$= (n-1)^2 \left(\frac{\theta^2}{(n-1)(n-2)} - \left(\frac{\theta^2}{(n-1)^2}\right) \right)$$

$$E\left(\frac{\sum X_i}{\sum c_i}\right) = \frac{1}{\sum c_i} \sum E(X_i) = \frac{1}{\sum c_i} \sum c_i \theta = \theta \frac{\sum c_i}{\sum c_i} = \theta$$

$$= \theta^2 \left(\frac{n-1}{n-2} - 1 \right) = \theta^2 \left(\frac{1}{n-2} \right) > \frac{\theta^2}{n}$$

$$\therefore \frac{\sum X_i}{\sum c_i} \text{ is unbiased est.}$$

Continue b(b)

$$\text{Var}\left(\frac{\sum X_i}{\sum c_i}\right) = \frac{1}{(\sum c_i)^2} \sum_{i=1}^n \text{Var} X_i$$

$$= \frac{1}{(\sum c_i)^2} \sum c_i^2 \theta$$

$$= \frac{\theta}{\sum c_i} = \text{CRLB}$$

 $\Rightarrow \frac{\sum X_i}{\sum c_i}$ is UMVUE for θ

and by 5.

we know [if $X_i \sim \text{NegE}(\theta)$, then $\frac{n-1}{\sum X_i}$ is UMVUE for θ]

so, 在這題裡

$$\frac{n-1}{\sum X_i} = \frac{n-1}{\sum (-\ln X_i)} = -\frac{n-1}{\ln(X_1 \cdots X_n)} = -\frac{n-1}{T}$$

is UMVUE for θ

$$1. X_1, \dots, X_n \stackrel{\text{iid}}{\sim} f(x; \theta) = \theta x^{\theta-1} x \in (0,1), \theta \in (0, \infty)$$

$$4. X_1, \dots, X_n \stackrel{\text{iid}}{\sim} B(1, p) \quad p \in [0,1] \quad n \geq 6$$

$$f(x; \theta) = \prod_{i=1}^n \theta x_i^{\theta-1}$$

$$f(x; p) = \prod p^{x_i} (1-p)^{1-x_i}$$

$$= \theta^n (x_1 \cdots x_n)^{\theta-1}$$

$$= p^{\sum x_i} (1-p)^{n-\sum x_i}$$

$$= \theta^n \exp\left((\theta-1) \ln(x_1 \cdots x_n)\right) \cdot 1$$

$$= (1-p)^n \exp\left(\frac{p}{1-p} \sum x_i\right) \cdot 1$$

$$= (1-\theta) \exp\left(\theta \ln\left(\frac{1}{T}\right)\right) \cdot h(x)$$

$$= (1-\theta) \exp\left(\theta \ln\left(\frac{1}{T}\right)\right) \cdot h(x)$$

$$w/ \mathcal{X} = \{x \mid x_i \in (0,1)\}$$

$$w/ \mathcal{X} = \{x \mid x_i = 0,1\}$$

 $\Rightarrow f(x; \theta) \in 1\text{-par. exp. family}$
 $\Rightarrow f(x; p) \in 1\text{-par. exp. family}$

$$B = \{Q(\theta) \mid \theta > 0\}$$

$$B = \{Q(\theta) \mid \theta \in (0,1)\}$$

$$= (-1, \infty)$$

$$= \left\{ \frac{p}{1-p} \mid p \in (0,1) \right\}$$

$$\Rightarrow B^0 \neq \emptyset$$

$$= (0, \infty)$$

$$\Rightarrow T = \ln(X_1 \cdots X_n) = \sum \ln X_i$$

$$\Rightarrow B^0 \neq \emptyset$$

is suff. stat. and complete.

 $\Rightarrow T = \sum X_i$ is suff. stat. and complete

$$\text{Goal: find } g(T) \text{ s.t. } E(g(T)) = \theta$$

$$\text{Try } E(T) = E(\sum X_i) = np$$

$$Y_i = 1, 2, \dots, n$$

$$\text{know } T \sim B(n, p)$$

$$\text{Let } Y_i = -\ln X_i \quad \Rightarrow e^{-Y_i} = X_i \quad w/ \quad X \in (0,1)$$

$$\text{Try } E(T(T-1)(n-T)(n-T-1)(n-T-2))$$

$$f_{Y_i}(y) = f_{X_i}(e^y) \left| \frac{dX(y)}{dy} \right|$$

$$= \sum_{t=0}^n \frac{n!}{t!(n-t)!} (n-t-1)(n-t-2) \binom{n}{t} p^t (1-p)^{n-t}$$

$$= \theta(e^y)^{\theta-1} \left| \frac{d\theta(y)}{dy} \right|$$

$$\stackrel{\because n \geq 5}{=} \sum_{t=2}^n \frac{n!}{t!(n-t)!} (n-t-1)(n-t-2) \frac{n!}{t!(n-t)!} p^{t-2} (1-p)^{n-t-2}$$

$$= \theta e^{-Y(\theta-1)} | -e^{-Y} |$$

$$(t+2)!(n-t-3)!$$

$$= \theta e^{-Y\theta}, \quad Y \in (0, \infty)$$

$$= \left[\sum_{t=2}^{n-5} \frac{(n-5)!}{(t+2)(n-t-3)!} p^{t-2} (1-p)^{n-t-3} \right] \left[\frac{n!}{(n-5)!} \cdot p^2 (1-p)^3 \right]$$

$$\text{i.e. } Y_i \sim \text{NegE}(\theta)$$

$$\text{pdf of } B(n-5, p)$$

$$\therefore X_i \stackrel{\text{iid}}{\sim}$$

$$= 1 \cdot \frac{n!}{(n-5)!} p^2 (1-p)^3$$

$$\Rightarrow Y_i \stackrel{\text{iid}}{\sim}$$

$$\frac{(n-5)!}{n!} T(T-1)(n-T)(n-T-1)(n-T-2)$$

$$\text{is UMVUE for } p^2(1-p)^3$$

$$2. X_1, \dots, X_n \sim N(\theta, 1) \Rightarrow X_i - \theta \sim N(0, 1)$$

$$Y \stackrel{\text{def}}{=} \sum_{i=1}^n X_i, X_n > 0$$

$$\text{s.t. } Y \sim \text{Bin}(n, p) \text{ w/ } p = 1 - \bar{\Phi}(0 - \theta) \text{ w/ } \bar{\Phi} \text{ is cdf. of } N(0, 1)$$

$$= \bar{\Phi}(\theta)$$

Goal: Find MLE of θ

$$L(\theta | Y) = f_Y(Y; \theta)$$

$$= \binom{n}{Y} (\bar{\Phi}(\theta))^Y (1 - \bar{\Phi}(\theta))^{n-Y}$$

$$= \bar{\Phi}(\theta)$$

$$\Rightarrow \frac{\partial}{\partial \theta} \log L(\theta | Y) = \frac{\partial}{\partial \theta} \left(Y \log(\bar{\Phi}(\theta)) + (n-Y) \log(1 - \bar{\Phi}(\theta)) \right)$$

$$= Y \frac{1}{\bar{\Phi}(\theta)} \phi(\theta) + (n-Y) \frac{-1}{1 - \bar{\Phi}(\theta)} (-\phi(\theta)) \quad \text{w/ } \phi \text{ is p.d.f. of } N(0, 1)$$

$$= \phi(\theta) \left[\frac{Y}{\bar{\Phi}(\theta)} + \frac{Y-n}{1 - \bar{\Phi}(\theta)} \right]$$

$$= \phi(\theta) \left[\frac{Y - n \bar{\Phi}(\theta)}{\bar{\Phi}(\theta) (1 - \bar{\Phi}(\theta))} \right]$$

$$= \frac{\phi(\theta)}{\bar{\Phi}(\theta) (1 - \bar{\Phi}(\theta))} (Y - n \bar{\Phi}(\theta)) > 0 \text{ if } Y > n \bar{\Phi}(\theta) \Leftrightarrow \bar{\Phi}^{-1}\left(\frac{Y}{n}\right) > \theta$$

$$> 0, \forall \theta$$

$$= 0 \text{ if } \theta = \bar{\Phi}^{-1}\left(\frac{Y}{n}\right)$$

$$< 0 \text{ if } \theta > \bar{\Phi}^{-1}\left(\frac{Y}{n}\right)$$

$$\therefore \hat{\theta}_{MLE} = \bar{\Phi}^{-1}\left(\frac{Y}{n}\right)$$