

HW#2 due March 10

$$f(y; \mu, \sigma^2) = \frac{1}{\pi \{1 + (y - \mu)^2\}} \cdot \frac{1}{\sigma} \quad -\infty < y < \infty$$

$$\sigma \equiv 1$$

$$L(\mu; y) = \prod_{i=1}^n \frac{1}{\pi \{1 + (y_i - \mu)^2\}}$$

$$l(\mu) = \sum_{i=1}^n \frac{-\log \{1 + (y_i - \mu)^2\} - n \log \pi}{}$$

$$\begin{aligned} E\{-l''(\mu)\} &= i(\mu) && \text{expected Fisher info.} \\ &= ni_1(\mu) && i_1 = E[-\log \{1 + (Y - \mu)^2\}]'' \end{aligned}$$

$$E(\sum X_i) = nEX_1 \quad \text{if } X_1, \dots, X_n \text{ iid}$$

$$l' = \frac{2(y - \mu)}{\{1 + (y - \mu)^2\}} \quad l'' = \dots$$

$$\int_{-\infty}^{\infty} \frac{1 - (y - \mu)^2}{\{1 + (y - \mu)^2\}^2} \cdot \frac{1}{\pi \{1 + (y - \mu)^2\}} dy \quad z = y - \mu$$

$$l(\mu, \sigma; y_1, \dots, y_n) = -n \log \sigma - \sum \log \left\{ 1 + \left(\frac{y_i - \mu}{\sigma} \right)^2 \right\}$$

$$\frac{\partial^2 l}{\partial \mu^2}, \quad \frac{\partial^2 l}{\partial \mu \partial \sigma}, \quad \frac{\partial^2 l}{\partial \sigma^2}$$

$$E\left(-\frac{\partial^2 l}{\partial \mu \partial \sigma}\right) = \dots = \frac{1}{\sigma^2} \cdot \int \frac{1-z^2}{(1+z^2)^3} dz$$

??
??

m.l.e. $\hat{\mu}$ solves $l'(\hat{\mu}) = 0$

$$\sum_{i=1}^n \frac{(y_i - \hat{\mu})}{\{1 + (y_i - \hat{\mu})^2\}} = 0 \quad \text{defines } \hat{\mu} \quad (\sigma=1)$$

$$\text{a. var}(\hat{\mu}) = \left[E\{-l''(\hat{\mu})\} \right]^{-1} \quad (\text{likelihood theory})$$

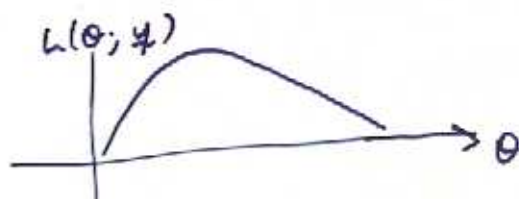
$$\text{a. var}(\tilde{\mu}_{\text{med}}) = ? \frac{1}{n^4 f^2(0)} ? \quad \text{for } \tilde{\mu} = \text{med}(y_i)$$

§ 2.2 General lik. theory, vector param. θ

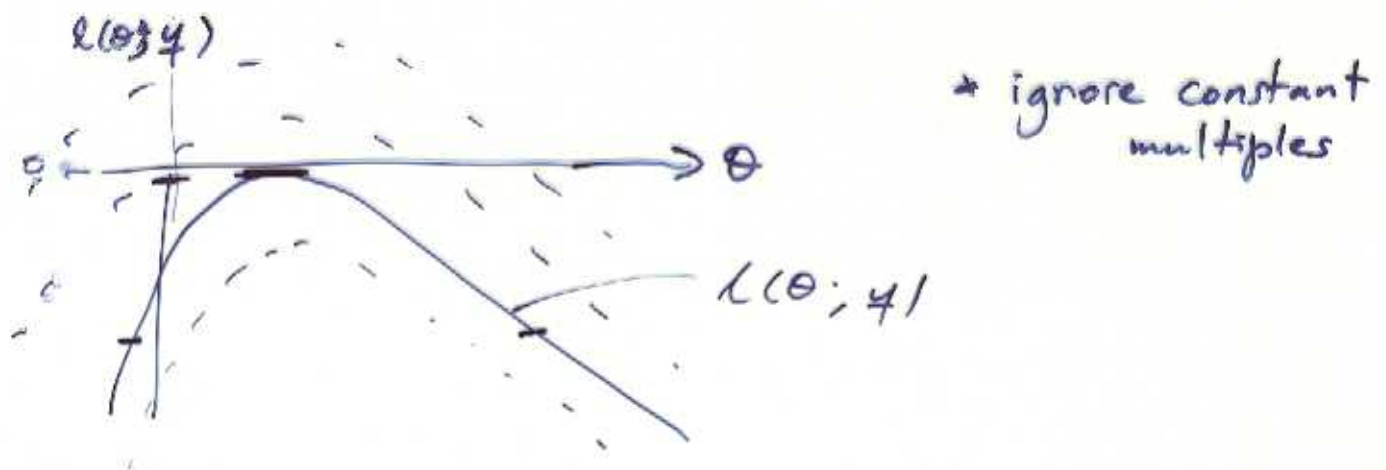
$$\text{Lik } f = L(\theta) = \prod_{i=1}^n f(y_i; \theta) \propto \prod_{i=1}^n f(y_i; \theta) \in \mathbb{R}^*$$

Assume Y_1, \dots, Y_n ind't, id. dist'd, w density $f(y; \theta)$

Function of θ



$$l(\theta) = \log\text{-likelihood} = \sum_{i=1}^n \log f(y_i; \theta) + a(y)$$



relative values of $l(\theta)$ are important
but absolute values are not

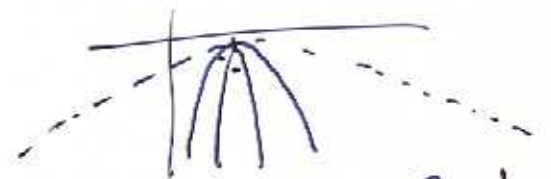
3 quantities used to summarize this plot

- $\hat{\theta}$ max. lik. est. $\sup_{\theta} l(\theta) = l(\hat{\theta})$

- $-l''(\hat{\theta})$ curvature at maximum $(j(\hat{\theta}))$

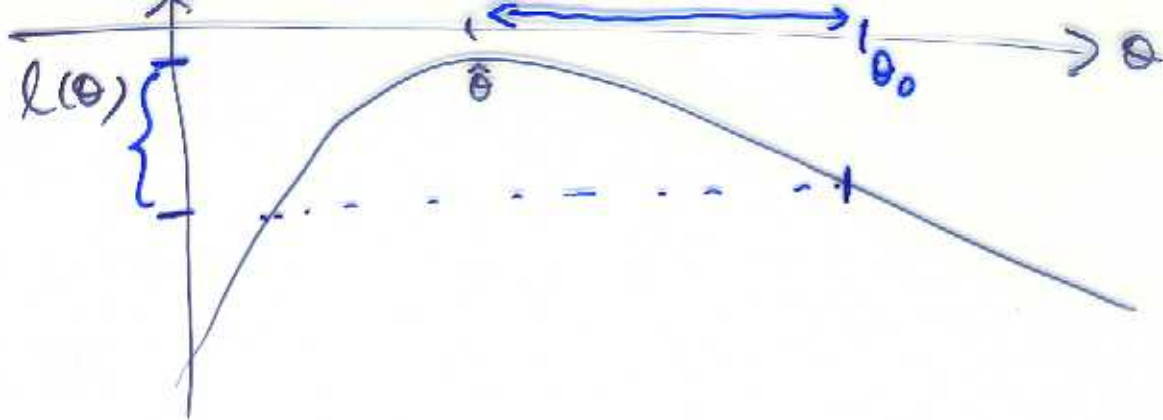
$\hat{\theta}$ asymptotically $N(\theta, i^{-1}(\theta))$ (*) or ...
 $n \rightarrow \infty$

where $i(\theta) = E_{\theta} j(\theta)$



- $l'(\theta)$ score statistic because we find $\hat{\theta}$ by

$$l'(\hat{\theta}) = 0$$



$$\hat{\theta} \sim N(\theta, i^{-1}(\theta)) \quad (\theta, \sigma^2) \quad \checkmark$$

$$\Rightarrow \hat{\theta} \pm z_{\alpha/2} \cdot \underset{\substack{\uparrow \\ 1/\hat{\sigma}}}{j(\hat{\theta})^{1/2}} \quad \text{is approx } 1-\alpha \text{ CI for } \theta$$

$$\checkmark \Rightarrow P\left(\frac{|\hat{\theta} - \theta|}{\{i^{-1}(\theta)\}^{1/2}} \leq x\right) \simeq P_z(|N(0,1)| \leq x)$$

$$P(|\hat{\theta} - \theta| \leq z_{\alpha/2} i^{-1/2}(\theta)) \simeq 1 - \alpha \quad \text{if } x = z_{\alpha/2}$$

$$\hat{\theta} \pm z_{\alpha/2} \underset{\substack{\downarrow \\ j^{-1/2}(\hat{\theta})}}{i^{-1/2}(\theta)} \quad \text{is a } 1-\alpha \text{ CI}$$

\uparrow uses a measure of $\hat{\theta} - \theta_0$

(any fixed θ_0 -value)

Alternative $2\{l(\hat{\theta}) - l(\theta)\} \simeq \chi^2_1$ if θ is 'true' value

How to compute mle's ??

1) scalar $l'(\hat{\theta}) = 0 = l'(\theta_0) + (\hat{\theta} - \theta_0) l''(\theta_0) + \dots$

Taylor series

$$\hat{\theta} - \theta_0 \approx -\frac{l'(\theta_0)}{l''(\theta_0)}$$

] Newton-Raphson

$\hat{\theta}^{(0)}$ guess

$$\hat{\theta}^{(t+1)} = \hat{\theta}^{(t)} - \frac{l'(\hat{\theta}^{(t)})}{l''(\hat{\theta}^{(t)})}$$

$$\left| \hat{\theta}^{(t+1)} - \hat{\theta}^{(t)} \right| < \epsilon$$

2)

$$\frac{\hat{\theta}^{(t+1)}}{p \times 1} = \frac{\hat{\theta}^{(t)}}{p \times 1} - \left[\frac{l''(\hat{\theta}^{(t)})}{p \times p} \right]^{-1} \frac{l'(\hat{\theta}^{(t)})}{p \times 1}$$

Poisson with no zeros

$$f(y_i; \lambda) = \frac{\lambda^{y_i} e^{-\lambda}}{y_i! (1 - e^{-\lambda})}, \quad y_i = 1, 2, \dots$$

(he called $y_i; n_i$)

Vector $\underline{\theta}$ (H0) $\underline{\theta}_{p \times 1}$

$$1) \quad \hat{\underline{\theta}} : \sup_{\underline{\theta}} l(\underline{\theta}) = \sup_{\underline{\theta}} l(\underline{\theta}; \underline{y})$$

\uparrow
 y_1, \dots, y_n

$\hat{\underline{\theta}}(y)$

$$2) \quad -l''(\underline{\theta}) \quad p \times p \quad \text{matrix} \quad \left[-\frac{\partial^2 l(\underline{\theta})}{\partial \theta_i \partial \theta_j} \right]_{\substack{i=1, \dots, p \\ j=1, \dots, p}}$$

$$\{-l''(\hat{\underline{\theta}})\}^{1/2} (\hat{\underline{\theta}} - \underline{\theta}) \sim N_p(0, \mathbf{I})$$

$$3) \quad l'(\hat{\underline{\theta}}) = \underline{0} \quad \leftarrow p \text{ eq}^n \text{ s in } p \text{ unknowns}$$

$$4) \quad 2 \{ l(\hat{\underline{\theta}}) - l(\underline{\theta}_0) \} \sim \chi_p^2 \quad \text{where } \underline{\theta}_0 \text{ is fixed}$$

$\& y_1, \dots, y_n \sim f(y; \underline{\theta}_0)$

Fancier (Hw 2 Qu #1)

$$l(\hat{\underline{\beta}}, \hat{\underline{\mu}}) \quad \beta_1, \dots, \beta_{10}, \mu_1, \dots, \mu_{10} \text{ separate gammas}$$

$$\rightarrow l(\hat{\underline{\beta}}, \hat{\underline{\mu}}) \quad \beta; \mu_1, \dots, \mu_{10}$$

$$4') \quad 2 \{ l(\hat{\underline{\beta}}, \hat{\underline{\mu}}) - l(\hat{\underline{\beta}}, \hat{\underline{\mu}}) \} \rightarrow \chi_{g=20-11}^2$$

$$\rightarrow l(1, \hat{\underline{\mu}}) \quad 2 \{ l(\hat{\underline{\beta}}; \hat{\underline{\mu}}) - l(1; \hat{\underline{\mu}}) \} \rightarrow \chi_1^2$$

$$L(\lambda) = \prod_{i=1}^n \frac{\lambda^{y_i} e^{-\lambda}}{y_i! (1-e^{-\lambda})}$$

$$\begin{aligned} \ell(\lambda) &= \sum_{i=1}^n y_i \log \lambda - \underline{n} \lambda - \underline{n} \log(1-e^{-\lambda}) \\ &= \underline{y+} \log \lambda - \underline{n} \lambda - n \log(1-e^{-\lambda}) \\ &\quad \text{sum}(n) \quad \uparrow \quad \uparrow \quad \text{length}(n) \{ \lambda + \log(1-e^{-\lambda}) \} \end{aligned}$$

nzp. log. likelihood

$$n \leftarrow c(1, 2, 3, 2, 2)$$

$$n \leftarrow c(1, 2, 3, 2, 2, 5, 7, 1, 6, \dots)$$