

Lec. 20

Announcement

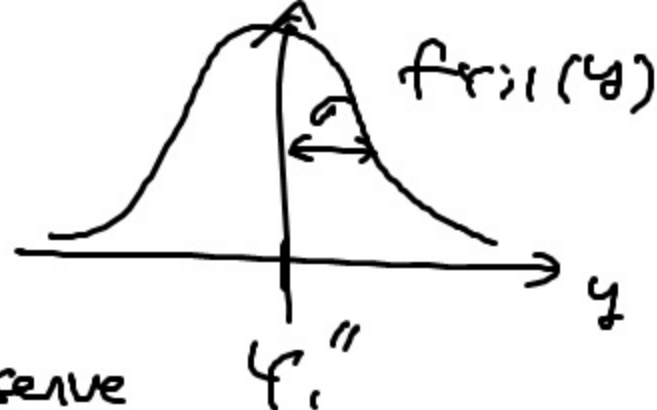
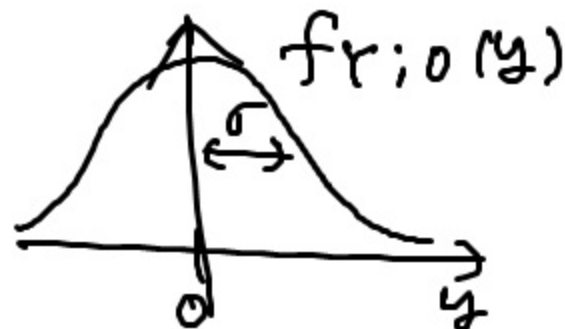
Midterm Exam #2 on 1/15.

Review

- A family of distribution functions
 $\{f_Y; \theta(Y); \theta \in A\}$
- a statistic $\underline{Y} \rightarrow \boxed{T(\underline{Y})} \rightarrow T(\underline{Y})$
- a sufficient statistic: $T(\underline{Y})$ is a s.s. $\Leftrightarrow$ 

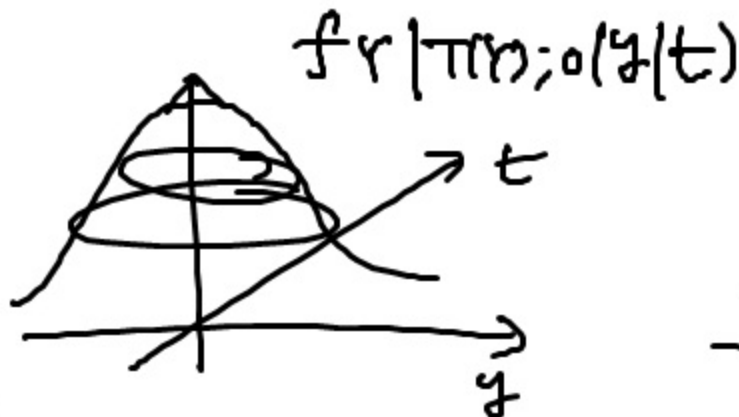
p.2

tex1



"We don't need to observe Y_1 ."

tex1



"We don't need to observe Y given $\{T(Y)=t\}$."

" $f_{r|T(r);0}(y|t)$ not a function of $\underline{0}, \underline{t_0}, \underline{t_1}$ "

p.3 "a sufficient statistic for θ "
for a family of
distribution functions $\{f_X; \theta(x); \theta \in A\}$
parameterized by θ .

The trouble w/ definition:

$f_X(I(X); \theta(x|t))$ not a function of θ .

1) $I(X)$ must be given.

2) Need to compute $f_X(I(X); \theta(x|t))$

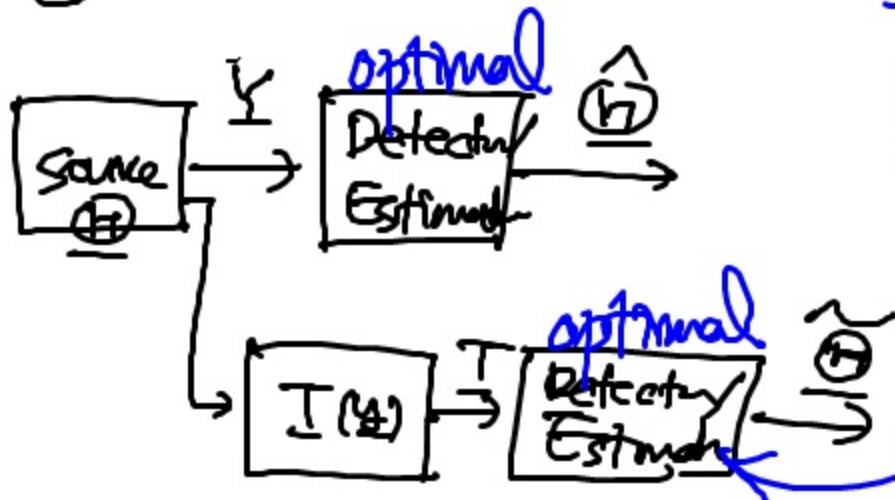
3) Need to check whether it is a ft of θ or not



p.4

 $\begin{cases} \underline{0} \\ \underline{\theta} \end{cases}$
 $f_Y(I(Y); \underline{0}(\underline{y}(\underline{t})) \dots$
 $\underline{f_Y(I(Y), \underline{\theta})(\underline{y}(\underline{t}, \underline{0}))}$
 $f_{Y|\underline{\theta}}(\underline{y}|\underline{\theta})$

Theorem.

If $\underline{\theta}$ is random, then
 $f_{\underline{\theta}}(Y|\underline{0}(\underline{y}))$ is
a function of $\underline{y}$
& $\underline{0}$ only through
 $I(\underline{y})$.

 $\{f_{I|\underline{\theta}}(\underline{t}|\underline{\theta})\}$

p5 Given $P_Y(I(Y), \underline{\Theta}) (Y|t, \underline{\Theta})$ not a function of $\underline{\Theta}$,
 Show, $P_{\underline{\Theta}}(Y | \underline{\Theta} | Y)$ is a function of $\underline{\Theta}$ & $I(Y)$.

sol)

$$\begin{aligned}
 P_Y(I(Y), \underline{\Theta}) (Y|t, \underline{\Theta}) &= P(Y=Y | I(Y)=t, \underline{\Theta}=\underline{\Theta}) \\
 &= P(\underline{\Theta}=\underline{\Theta} | \boxed{Y=Y} | \cancel{I(Y)=t}) \cdot \frac{P(Y=Y | I(Y)=t)}{P(\underline{\Theta}=\underline{\Theta} | I(Y)=t)} \\
 &= \text{not a function of } \underline{\Theta} \quad \Leftrightarrow
 \end{aligned}$$

p.6

We start from " $P_{Y|I(Y), \Theta}(y(t), \Theta)$ is not a function of Θ ".

$$Pr(Y=y | I(Y)=t, \Theta=\Theta) = \begin{cases} 0, & \text{if } t \neq I(y) \end{cases}$$

$$\left(\begin{aligned} &Pr(Y=y | I(Y)=I(y), \Theta=\Theta) \\ &\text{, otherwise } (t = I(y)) \end{aligned} \right)$$

$$Pr(Y=y | I(Y)=I(y), \Theta=\Theta) = \dots$$

p7

Example 2.

$$\underline{Y} = A\underline{1} + \underline{N}$$

" $f_{A|Y}(a|\underline{y})$ is a function of a & $\underline{1}^T \underline{y}$."

$$\begin{aligned}\mu &\triangleq E[A|Y=\underline{y}] = E[A] + \text{Cov}(A, Y) \text{Cov}(Y)^{-1} (\underline{y} - E[Y]) \\ &= 0 + \sigma_a^2 (\sigma_a^2 \underline{1}\underline{1}^T + \sigma_n^2 I)^{-1} \underline{y}\end{aligned}$$

$$\begin{aligned}\sigma^2 &\triangleq \text{Var}[A|Y=\underline{y}] = \text{Var}[A] - \text{Cov}(A, Y) \text{Cov}(Y)^{-1} \text{Cov}(Y, A) \\ &= \sigma_a^2 - \sigma_a^2 (\sigma_a^2 \underline{1}\underline{1}^T + \sigma_n^2 I)^{-1} \sigma_a^2.\end{aligned}$$

$$\rightarrow \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(a-\mu)^2}{2\sigma^2}\right)$$

p.8

Matrix Inversion Lemma.

$$(A + BC)^T = \dots \text{ in terms of } A^{-1} \dots$$

$$(A + PB^T)^T = \dots \text{ " " " } A^{-1} \dots$$

$$(A + PP^T)^T = \dots \text{ " " " } A^{-1} \dots$$

⇒ The Sherman-Morrison Formula.

$$A^{-1} \rightarrow \frac{A^{-1}PB^TA^{-1}}{1 + B^TA^{-1}P}$$

p.9

$$E[A|Y=y] = \underbrace{\sigma_a^2 \mathbf{1}}_P (\underbrace{\sigma_a^2 \mathbf{1} \mathbf{1}^T}_R + \sigma_n^2 I)^T \underline{y}$$

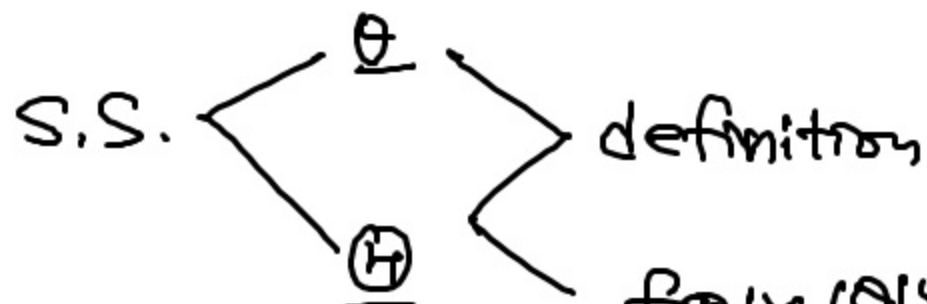
$$= \sigma_a^2 \mathbf{1} \left(\frac{I}{\sigma_n^2} - \frac{\frac{I}{\sigma_n^2} \underbrace{\sigma_a^2 \mathbf{1} \mathbf{1}^T}_R \frac{I}{\sigma_n^2}}{1 + \underbrace{\mathbf{1}^T \frac{I}{\sigma_n^2} \mathbf{1}}_R \underbrace{\sigma_a^2 \mathbf{1}}_P} \right) \underline{y}$$

$$= \sigma_a^2 \mathbf{1} \left(\frac{\underline{y}}{\sigma_n^2} - \frac{\frac{\sigma_a^2 \mathbf{1}}{\sigma_n^2} \frac{\mathbf{1}^T \underline{y}}{\sigma_n^2}}{1 + \frac{N}{\sigma_n^2} \sigma_a^2} \right)$$

$$= \frac{\sigma_a^2}{\sigma_n^2} \underbrace{\mathbf{1}^T \underline{y}}_R - \frac{\frac{\sigma_a^4}{\sigma_n^4} N \underbrace{\mathbf{1}^T \underline{y}}_R}{\sigma_n^2} = \dots =$$

$$= \left(\frac{\sigma_a^2}{\sigma_n^2} - \frac{\frac{\sigma_a^4}{\sigma_n^4} N}{1 + \frac{\sigma_a^2}{\sigma_n^2} N} \right) \underbrace{\mathbf{1}^T \underline{y}}_R$$

p10



$f_{\eta}(x | \theta(y))$ is a function only of θ & $\eta(y)$.

Must have a candidate $\eta(y)$.



Neyman-Fisher Factorization theorem.

$\eta(y)$ is sufficient $\Leftrightarrow f_{Y;\theta}(y) = \dots \forall \theta.$

P.11

$$f_{Y; \underline{0}}(\underline{y}) = \underbrace{\text{a ft only of } \underline{y}} \times \underbrace{\text{a ft of } \underline{I(\underline{y})} \& \underline{0}} \quad \forall \underline{y}, \forall \underline{0}$$

$\Leftrightarrow \underline{I(\underline{y})}$ is sufficient. ~~$\times \circ$~~ $\nearrow$ a function of $\underline{I(\underline{y})}$.

$\Leftrightarrow \exists \underline{I(\underline{y})}, \underbrace{h(\underline{y})}_{\text{co-factor.}}$ $\nwarrow$ a ft of $\underline{0}$.

$$g(\underline{z}, \underline{0})$$

s.t. $f_{Y; \underline{0}}(\underline{y}) = h(\underline{y}) g(\underline{I(\underline{y})}, \underline{0}),$

$\forall \underline{y}, \forall \underline{0}.$

p. 12.

Example 1. $\underline{Y} = \begin{bmatrix} Y_1 \\ Y_2 \\ \vdots \\ Y_N \end{bmatrix}$

Y_i 's i.i.d.

$$Pr(Y=1) = \theta = 1 - Pr(Y=0)$$

$$\begin{aligned} P(\underline{Y}; \theta | \underline{y}) &= Pr(\underline{Y} = \underline{y}; \theta) = \prod_{i=1}^N Pr(Y_i = y_i; \theta) \\ &= \prod_{i=1}^N \theta^{y_i} (1-\theta)^{1-y_i} \\ &= \theta^{\sum y_i} (1-\theta)^{N - \sum y_i} \end{aligned}$$

To apply the N-F factorization theorem, we search for $T(\underline{y})$, $h(\underline{y}) = 1$, $g(t, \theta)$.

$$\sum y_i = \underline{1}^T \underline{y} \quad \theta^t (1-\theta)^{N-t}$$