

STAT 5430

Lec 39, F , May 2

- Homework 9 is assigned & due Sunday, May 4
but you can submit this on Monday, May 5
- Lecture on M, May 5
- Lecture/review on W, May 7
- No class on F, May 9
- Final Exam on Tuesday, May 13, 7:30-9:30 PM

see
Canvas

- Comprehensive - but focus on material since Exam 2 (interval estimation)
- Formula sheet for new material/interval & 2 formula sheets previous material
- (3 sheets (ready, front/back) total)
- Practice Exams

Interval Estimation II

Bayes Intervals

Definition: A highest posterior density (HPD) credible set of level $(1 - \alpha)$ is a set of the form

$$C_{\tilde{x}} = \{\theta \in \Theta : f_{\theta|\tilde{x}}(\theta) \geq C\}, \text{ for some } C > 0$$

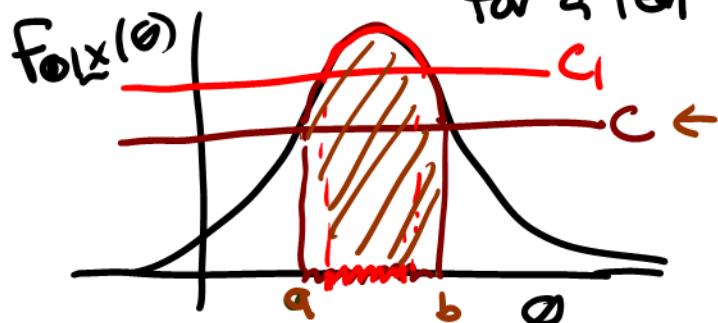
such that $P(\theta \in C_{\tilde{x}} | X = \tilde{x}) = 1 - \alpha, \forall \tilde{x}$.

posterior prob

*has right post. prob
to be a $(1-\alpha)$ credibility set*

Discussion: Why do this?

Consider posterior density $f_{\theta|\tilde{x}}(\theta)$
for a real-valued θ , which is unimodal

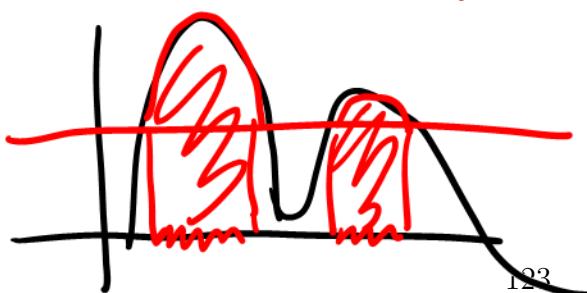


Pick C so that $\int_a^b f_{\theta|\tilde{x}}(\theta) d\theta = 1 - \alpha \Rightarrow$ gives the HPD
credible set as $C_x = [a, b]$

Note: When $f_{\theta|\tilde{x}}(\theta)$ is "high", want to "pack in"
an area of $1 - \alpha$ over a short region of θ .

So, HPD credible sets achieve $(1 - \alpha)$ posterior coverage
but tend to be small/informative sets for θ

(guesses for θ)



Interval Estimation II

Bayes HPD Intervals: Illustration

Example: Let X_1, \dots, X_n be iid $N(\theta, \sigma^2)$ with $\theta \in \mathbb{R}$ and known $\sigma^2 > 0$. Suppose a prior distribution for θ is $N(\mu, \tau^2)$ for some known $\mu \in \mathbb{R}, \tau^2 > 0$.

\Rightarrow posterior distribution of θ given \underline{x} is

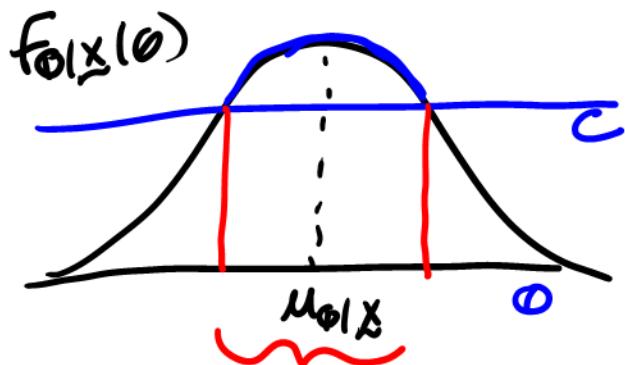
$$\theta | \underline{x} \sim \text{Normal}(\mu_{\theta|\underline{x}}, \sigma_{\theta|\underline{x}}^2)$$

unimodal

depend on $\underline{x}, n, \mu, \tau^2$
(done this before)

Find $\alpha(1-\alpha)$ HPD credible set for θ :

$$C_{\underline{x}} = \{\theta : f_{\theta|\underline{x}}(\theta) \geq C\} = \{\theta : \frac{1}{\sqrt{2\pi}\sigma_{\theta|\underline{x}}} e^{-\frac{(\theta-\mu_{\theta|\underline{x}})^2}{2\sigma_{\theta|\underline{x}}^2}} \geq C\} \\ = \{\theta : \left| \frac{\theta - \mu_{\theta|\underline{x}}}{\sigma_{\theta|\underline{x}}} \right| \leq C_1\}$$

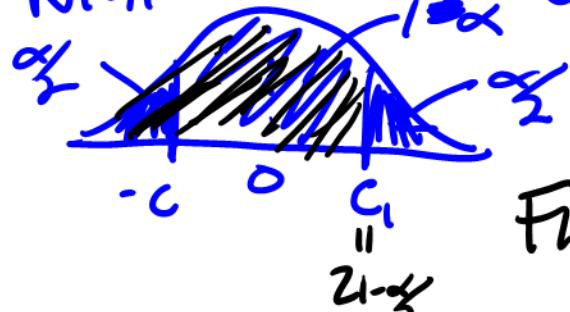


Need to pick C_1 (or C) so

that

$$1-\alpha = P(\theta \in C_{\underline{x}} | \underline{x}) = P\left(\left| \frac{\theta - \mu_{\theta|\underline{x}}}{\sigma_{\theta|\underline{x}}} \right| \leq C_1 | \underline{x}\right)$$

$$\text{Note: } \frac{\theta - \mu_{\theta|\underline{x}}}{\sigma_{\theta|\underline{x}}} | \underline{x} \sim N(0, 1) \Rightarrow P(|Z| \leq C_1) = P(|Z| \leq z_{1-\alpha/2})$$



pick $C_1 = z_{1-\alpha/2}$

Finally, HPD credible set for θ is

$$124 \quad \{\theta : \left| \frac{\theta - \mu_{\theta|\underline{x}}}{\sigma_{\theta|\underline{x}}} \right| \leq z_{1-\alpha/2}\}$$

Interval Estimation II

Evaluating Interval Estimators

Remark 1: For two interval estimators $I_C = [L_C(X), U_C(X)]$ and $I_D = [L_D(X), U_D(X)]$ with the same C.C. $1 - \alpha$, then I_D is **preferred** to I_C if

most important

$$E_\theta[\text{length } I_D] \leq E_\theta[\text{length } I_C], \quad \forall \theta \in \Theta$$

*better interval estimator (for a given C.C. $1 - \alpha$)
has shorter expected length (more informative)*

Remark 2: For confidence regions C_X and D_X for $\theta \in \Theta \subset \mathbb{R}^p$ with the same C.C. $1 - \alpha$, D_X is preferred to C_X if

$$E_\theta[\text{volume } D_X] \leq E_\theta[\text{volume } C_X], \quad \forall \theta \in \Theta.$$

↙ For finding CIs with short length

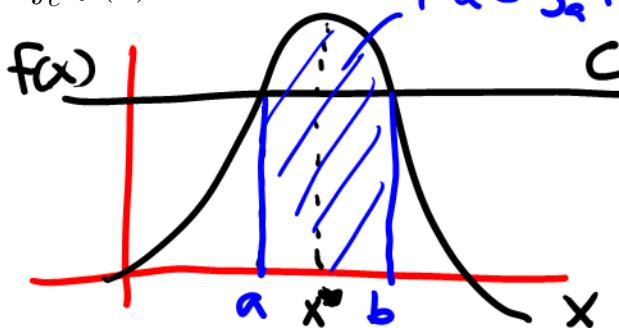
Theorem on Interval Lengths (Theorem 9.3.1 [CB]): Let $f(x)$, $x \in \mathbb{R}$, be a unimodal pdf. If an interval $[a, b]$ satisfies

1. $\int_a^b f(x)dx = 1 - \alpha$
2. $f(a) = f(b) > 0$
3. $a \leq x^* \leq b$, where x^* is the mode/peak of $f(x)$,

then $[a, b]$ has smallest (shortest) length out of all possible intervals $[c, d]$ satisfying

$$\int_c^d f(x)dx = 1 - \alpha.$$

$$C = f(a) = f(b)$$



*(Same principle
as HPD Bayes
intervals)*

Interval Estimation II

Evaluating Interval Estimators: Illustration

Example: Let X_1, \dots, X_n iid $N(\mu, \sigma^2)$, $\sigma^2 > 0$, $\mu \in \mathbb{R}$ both unknown. Find a CI to estimate μ of the form

$$I_{a,b} = \left[\bar{X}_n - \frac{bS}{\sqrt{n}}, \bar{X}_n - \frac{aS}{\sqrt{n}} \right]$$

\leftarrow sample mean
 $S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2$
 \uparrow sample variance
 $S = \sqrt{S^2}$

(where $-\infty < a < b < \infty$ and S^2 is the sample variance) such that $P_{\mu, \sigma^2}(\mu \in I_{a,b}) = 1 - \alpha$, $\forall \mu, \sigma^2$ and such that the expected length $E_{\mu, \sigma^2}(I_{a,b})$ is as short as possible, $\forall \mu, \sigma^2$.

Solution: $1 - \alpha = P_{\mu, \sigma^2}(\mu \in I_{a,b})$

$$= P_{\mu, \sigma^2}\left(\frac{\bar{X}_n - bS}{\sqrt{n}} \leq \mu \leq \frac{\bar{X}_n - aS}{\sqrt{n}}\right)$$

$$= P_{\mu, \sigma^2}\left(a \leq \frac{\bar{X}_n - \mu}{S/\sqrt{n}} \leq b\right)$$

$\underbrace{\quad}_{T_{n-1}}$

$$= \int_a^b f_{T_{n-1}}(x) dx$$

①

unif. dist.

Note: expected length $E_{\mu, \sigma^2}(I_{a,b}) = \frac{b-a}{\sqrt{n}} E_{\mu, \sigma^2}(S)$ ②

Hence, of all intervals $I_{a,b}$ satisfying ①, I_{a^*, b^*} minimizes ② where a^* & b^* satisfy

$$\textcircled{1} \quad \int_{a^*}^{b^*} f_{T_{n-1}}(x) dx = 1 - \alpha, \quad \textcircled{2} \quad f_{T_{n-1}}(a^*) = f_{T_{n-1}}(b^*)$$

+ ③ $a^* \leq 0 \leq b^*$.

So, $a^* = -b^*$, $b^* = t_{1-\alpha/2, n-1}$

Interval Estimation II

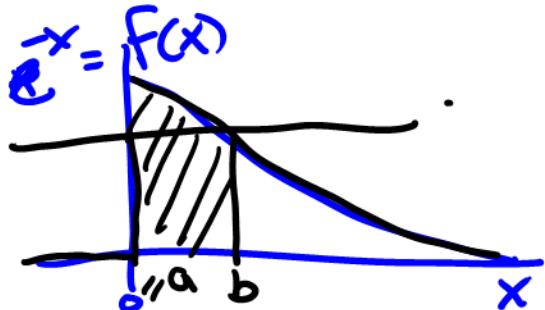
Evaluating Interval Estimators

Remark 3: The main idea of the theorem on interval lengths (Theorem 9.3.1) above is to pack in “area” under a density curve and use this to get short CIs. This concept can apply if Theorem 9.3.1 doesn’t.

Example: Let $X \sim \text{Exponential}(\frac{1}{\beta})$, $\beta > 0$. Find a CI for β of the form $[\frac{a}{X}, \frac{b}{X}]$, $b > a \geq 0$ such that the C.C. is $1 - \alpha$ and the length $(b - a)$ is as short as possible.

Solution Note: $\frac{X}{\beta} = \beta X \sim \text{Exponential}(1)$

$$\text{So, } 1 - \alpha = P_{\beta}(\beta \in [\frac{a}{X}, \frac{b}{X}]) = P_{\beta}(a \leq \underbrace{\beta X}_{\text{Exp}(1)} \leq b)$$



$$\begin{aligned} \text{So, } 1 - \alpha &= \int_a^b e^{-x} dx \\ &= \int_a^b e^{-x} dx \end{aligned}$$

So, pick $a = 0$ $1 - \alpha = e^{-a} - e^{-b} = 1 - e^{-b}$
 $\Rightarrow b = -\log(\alpha)$