

STAT 5430

Lec 32, M, Apr 14

- No homework this week!

- Exam 2 is

on W, April 16, 6:15-8:15 PM, 3rd floor seminar room

- No class on that W.

- I'll post: study guide (sufficiency/completeness/tests)

- practice exams

- bring new 1 page (front/back)

Formula sheet on exam 2 material

(I'll post one to use if you'd like)

- can bring calculator & previous formula sheet for exam 1

- I'll provide table of distributions /
STAT 542 facts on test as before

No Bayes tests on exam

Interval Estimation I

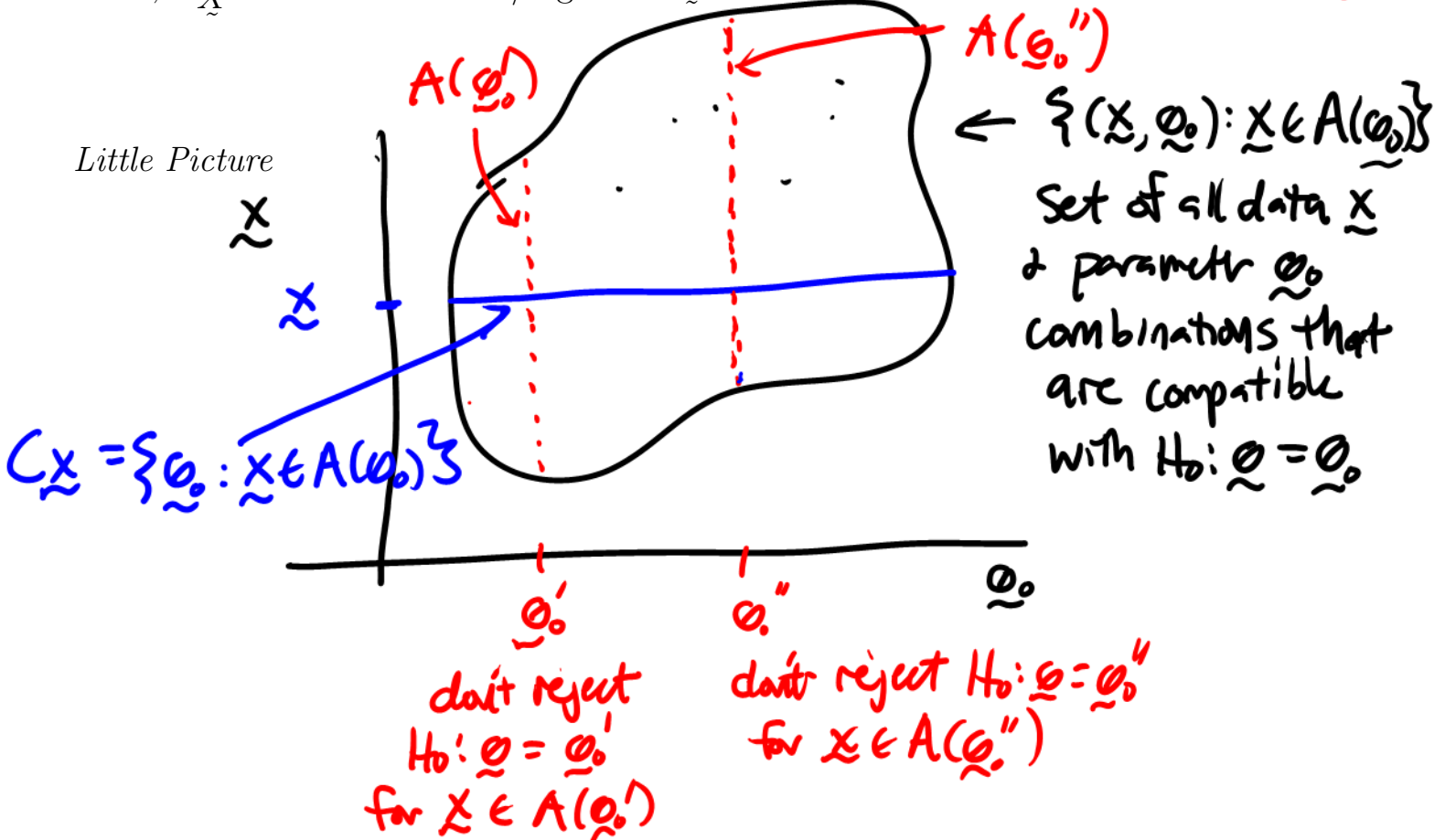
Inverting a Test

Theorem: Let X_1, \dots, X_n have joint pdf/pmf $f(x|\theta)$, $\theta \in \Theta \subset \mathbb{R}^p$ and let $A(\theta_0)$ denote the acceptance region of a test is 0 or 1 simple test of size α for testing $H_0 : \theta = \theta_0$ vs. $H_1 : \theta \neq \theta_0$ (for $p = 1$, $H_1 : \theta < \theta_0$ or $\theta > \theta_0$ is allowed too). Define sets $C_{\underline{x}} \subset \Theta$, $\underline{x} \in \mathbb{R}^n$ as

$$C_{\underline{x}} = \{\theta_0 : \underline{x} \in A(\theta_0)\}$$

Then, $C_{\underline{x}}$ is a confidence set/region for θ with confidence coefficient $1 - \alpha$.

Little Picture



Interval Estimation I

Inverting a Test, cont'd

Proof of Theorem: Note that

1.

$$\begin{aligned}\min_{\theta_0 \in \Theta} P_{\theta_0}(X \in A(\theta_0)) &= \min_{\theta_0 \in \Theta} P_{\theta_0}(\text{“do not reject } H_0 : \theta = \theta_0\text{”}) \\ &= \min_{\theta_0 \in \Theta} [1 - P_{\theta_0}(\text{“reject } H_0 : \theta = \theta_0\text{”})] \\ &= \min_{\theta_0 \in \Theta} [1 - \alpha] \\ &= 1 - \alpha,\end{aligned}$$

and

2. for any $\theta_0 \in \Theta$, any $x \in \mathbb{R}^n$, it holds that

$$x \in A(\theta_0) \Leftrightarrow \theta_0 \in C_x.$$

Hence,

$$\begin{aligned}\min_{\theta_0 \in \Theta} P_{\theta_0}(\theta_0 \in C_X) &= \min_{\theta_0 \in \Theta} P_{\theta_0}(X \in A(\theta_0)) \\ &= 1 - \alpha.\end{aligned}$$

Interval Estimation I

Inverting a Test: Illustration

Example: Let X_1, \dots, X_n be iid $N(\mu, \sigma^2)$, $-\infty < \mu < \infty$, $\sigma > 0$. Find

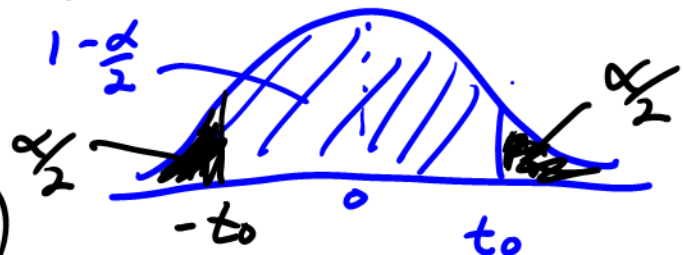
1. a C.I. (confidence interval) for μ with C.C. $1 - \alpha$ (two-sided)
2. a 1-sided lower confidence bound for μ with C.C. $1 - \alpha$, i.e., $(L(X), \infty)$

Solution for 1. Consider a test function

$$\phi_{\mu_0}(x) = \begin{cases} 1 & \text{if } \frac{|\bar{X}_n - \mu_0|}{s/\sqrt{n}} > t_\alpha \\ 0 & \text{o.w.} \end{cases}$$

for testing $H_0: \mu = \mu_0$ vs $H_1: \mu \neq \mu_0$ (given some $\mu_0 \in \mathbb{R}$)
where $t_\alpha \equiv (1 - \frac{\alpha}{2})$ percentile of T_{n-1} distribution.

$$\begin{aligned} E_{\mu_0} \phi_{\mu_0}(x) &= P_{\mu_0} (|\bar{X}_n - \mu_0| / (s/\sqrt{n}) > t_\alpha) \\ &= \alpha, \text{ i.e. } \phi_{\mu_0}(x) \text{ is a simple test (0 or 1) of size } \alpha \\ &\quad \text{for } H_0: \mu = \mu_0 \text{ vs } H_1: \mu \neq \mu_0 \end{aligned}$$



Note: acceptance region of $\phi_{\mu_0}(x)$ is $A(\mu_0) = \{x: \frac{|\bar{X}_n - \mu_0|}{s/\sqrt{n}} \leq t_\alpha\}$

Hence, $C_x = \{ \mu_0: x \in A(\mu_0) \} = \{x: \phi_{\mu_0}(x) = 0\}$

$$= \{ \mu_0: \frac{|\bar{X}_n - \mu_0|}{s/\sqrt{n}} \leq t_\alpha \}$$

$$= \{ \mu_0: -t_\alpha \leq \bar{X}_n - \mu_0 \leq t_\alpha \}$$

$$= \{ \mu_0: -t_\alpha \frac{s}{\sqrt{n}} \leq \bar{X}_n - \mu_0 \leq t_\alpha \frac{s}{\sqrt{n}} \}$$

$$= \{ \mu_0: \bar{X}_n - t_\alpha \frac{s}{\sqrt{n}} \leq \mu_0 \leq \bar{X}_n + t_\alpha \frac{s}{\sqrt{n}} \}$$

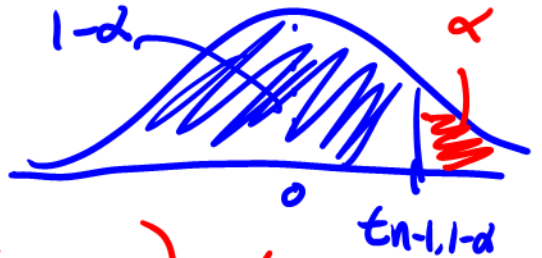
$$= [\bar{X}_n - t_\alpha \frac{s}{\sqrt{n}}, \bar{X}_n + t_\alpha \frac{s}{\sqrt{n}}]$$

depends on x
and
 C_x is a C.I.
with C.C. $1 - \alpha$

Solution to 2. Consider the following test for $H_0: \mu = \mu_0$ vs $H_1: \mu > \mu_0$ given by

$$\phi_{\mu_0}(\underline{x}) = \begin{cases} 1 & (\bar{X}_n - \mu_0) / (S/\sqrt{n}) > t_{n-1, 1-\alpha} \\ 0 & \text{o.w.} \end{cases}$$

where $t_{n-1, 1-\alpha} \equiv 1-\alpha$ percentile



Note: $E_{\mu_0} \phi(\underline{X}) = P_{\mu_0} \left(\frac{\bar{X}_n - \mu_0}{S/\sqrt{n}} > t_{n-1, 1-\alpha} \right) = \alpha$
 \uparrow size

Acceptance region for $\phi_{\mu_0}(\underline{x})$ is

$$A(\mu_0) \equiv \{ \underline{x} : \frac{\bar{X}_n - \mu_0}{S/\sqrt{n}} \leq t_{n-1, 1-\alpha} \}$$

For given \underline{x} ,

$$\begin{aligned} C_{\underline{x}} &= \{ \mu_0 : \underline{x} \in A(\mu_0) \} \\ &= \{ \mu_0 : \frac{\bar{X}_n - \mu_0}{S/\sqrt{n}} \leq t_{n-1, 1-\alpha} \} \\ &= \{ \mu_0 : \bar{X}_n - \mu_0 \leq \frac{S}{\sqrt{n}} t_{n-1, 1-\alpha} \} \\ &= \{ \mu_0 : \bar{X}_n - \frac{S}{\sqrt{n}} t_{n-1, 1-\alpha} \leq \mu_0 \} \\ &= [\bar{X}_n - \frac{S}{\sqrt{n}} t_{n-1, 1-\alpha}, \infty) \end{aligned}$$

\uparrow lower confidence interval for μ

Interval Estimation I

Pivotal Quantities

↓ vector of parameters $\underline{\theta}$

Definition: Let X_1, \dots, X_n be joint pdf/pmf $f(x|\underline{\theta})$, $\underline{\theta} \in \Theta \subset \mathbb{R}^p$. Then a random variable $Q(\underline{X}, \underline{\theta})$ is called a **pivot** or **pivotal quantity** if the distribution of $Q(\underline{X}, \underline{\theta})$ under $\underline{\theta}$ does not depend on $\underline{\theta}$.

Note: $Q(\underline{X}, \underline{\theta})$ is NOT a statistic (because can depend on $\underline{\theta}$)
 $P_{\underline{\theta}}(Q(\underline{X}, \underline{\theta}) \in A) = P(Q(\underline{X}, \underline{\theta}) \in A)$

Some examples: (pivots, unlike statistics, can be functions of parameters $\underline{\theta}$)

$\frac{X_1 - X_2}{X_3 - X_4}$ ancillary statistic

1. Let $X_1 \dots X_n$ be iid $N(\mu, \sigma^2)$ random variables.

$$\bar{X}_n \sim N(\mu, \frac{\sigma^2}{n})$$

pivot $\left\{ \frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} \sim N(0,1) \right\}$ pivot $\left\{ \frac{\bar{X}_n - \mu}{s/\sqrt{n}} \sim T_{n-1} \text{ distribution} \right\}$

2. Let f_0 be a pdf on \mathbb{R} . Let $X_1 \dots X_n$ be iid with random variables common pdf $f(x|\underline{\theta})$ where

$$\underline{\theta} = (\theta_1, \theta_2) \quad f(x|\underline{\theta}) = \frac{1}{\theta_2} f_0\left(\frac{x - \theta_1}{\theta_2}\right), \quad x \in \mathbb{R},$$

for $\underline{\theta} = (\theta_1, \theta_2)$, $\theta_1 \in \mathbb{R}$ (location parameter) and $\theta_2 > 0$ (scale parameter).

Then, $Q(\underline{X}, \underline{\theta}) = \frac{\bar{X}_n - \theta_1}{\theta_2} \leftarrow \text{pivot}$

Why? Note: $Y_i \equiv \frac{X_i - \theta_1}{\theta_2} \sim f_0(y)$ (i.e. Y_1, \dots, Y_n iid $f_0(y)$)

and $Q(\underline{X}, \underline{\theta}) = \frac{1}{n} \sum_{i=1}^n Y_i \leftarrow \text{distribution doesn't depend on } \underline{\theta}$

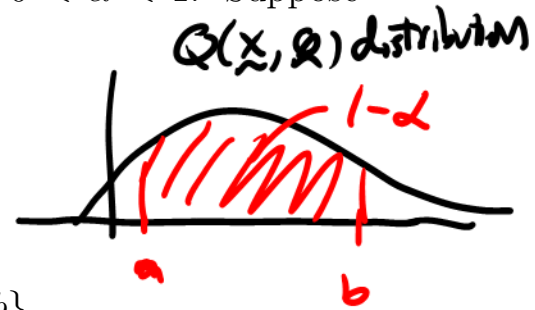
Interval Estimation I

Interval Estimation via Pivotal Quantities

Remarks:

1. Let $Q(\underline{X}, \underline{\theta})$ be a pivotal quantity ($\underline{\theta} \in \Theta \subset \mathbb{R}^p$) and $0 < \alpha < 1$. Suppose $-\infty \leq a \leq b \leq \infty$ are such that

$$P(a \leq Q(\underline{X}, \underline{\theta}) \leq b) = P_{\underline{\theta}}(a \leq Q(\underline{X}, \underline{\theta}) \leq b) = 1 - \alpha.$$



Then,

$$C_X = \{\underline{\theta} : \underline{\theta} \in \Theta, a \leq Q(\underline{X}, \underline{\theta}) \leq b\}$$

is a **confidence region** for $\underline{\theta}$ with CC $(1 - \alpha)$

$$\text{That is, } \min_{\underline{\theta} \in \Theta} P_{\underline{\theta}}(\underline{\theta} \in C_X) = \min_{\underline{\theta} \in \Theta} P_{\underline{\theta}}(a \leq Q(\underline{X}, \underline{\theta}) \leq b) = 1 - \alpha.$$

$$P(a \leq Q(\underline{X}, \underline{\theta}) \leq b)$$

2. If $\Theta \subset \mathbb{R}$ and $Q(\underline{X}, \underline{\theta})$ is monotone in $\underline{\theta} \in \mathbb{R}$, then the region C_X will be an interval.

Interval Estimation I

Interval Estimation via Pivotal Quantities

Example: Let $X_1 \dots X_n$ be iid $\text{Gamma}(\delta_0, \theta)$ where $\theta > 0$ (δ_0 fixed/known). Using a pivotal quantity based on $\sum_{i=1}^n X_i$, find a CI for θ with C.C. $1 - \alpha$.

Shape is known
Scale (unknown)

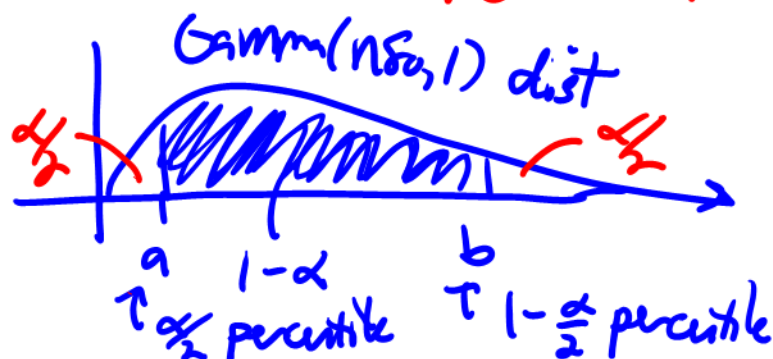
Solution: $\sum_{i=1}^n X_i \sim \text{Gamma}(n\delta_0, \theta)$

$Q(X, \theta) = \frac{\sum_{i=1}^n X_i}{\theta} \sim \text{Gamma}(n\delta_0, 1)$

pivot

dist. doesn't depend on θ

Find a & b such that



Confidence interval for $\theta > 0$

is $\{\theta : a \leq Q(\underline{X}, \theta) \leq b\} = \{\theta : a \leq \frac{\sum X_i}{\theta} \leq b\}$

$= \{\theta > 0 : \frac{a}{\sum X_i} \leq \theta \leq \frac{b}{\sum X_i}\}$

$= \left[\frac{a}{\sum X_i}, \frac{b}{\sum X_i} \right]$