

STAT 5430

Lec 20, M, Mar 10

sufficiency
& completeness
→

Homework 4 posted, due M, Mar 10

Bas's theorem
& Intro
to testing
→

Homework 5 posted, due M, Mar 24
(after break)

STAT 5430: Summary to date

Where we have been & where we are headed

- Completed
 - Introduction to Statistical Inference
 - Point Estimation
 - * MME/MLE
 - Criteria for Evaluating Point Estimators
 - * bias, variance, UMVUE, MSE
 - Elements of Decision Theory
 - * Minimax, finding Bayes estimators
 - Sufficiency and Point Estimation
 - * Factorization Theorem, Rao-Blackwell Theorem, Completeness, Lehman-Scheffe Theorem/UMVUE
- Next: Hypothesis Testing I
 - General Concepts: hypotheses, size, power
 - Most Powerful Tests/Neyman-Pearson Lemma
 - Uniformly Most Powerful Tests/Monotone Likelihood Ratios

Hypothesis Testing I

Terminology

Definitions

1. A (statistical) hypothesis is a statement about a population parameter. \mathcal{H}
2. The two complementary hypotheses in a testing problem are called the **null hypothesis** (denoted H_0) and the **alternative hypothesis** (denoted H_1).

↑ "starting claim"

H_0

↑ "counter-claim"

(what you want to show)

H_1 or H_a

Example: $\theta \equiv (\theta_1, \theta_2) \rightarrow$ average blood sugar level of a group of patients before and after taking a new drug

before \rightarrow after

H_0 : no effect $\leftrightarrow H_0: \theta_1 = \theta_2$ (no drug effect)

H_1 : effective $\leftrightarrow H_a: \theta_1 \neq \theta_2$ or $\theta_1 > \theta_2$ (drug has effect)

↓ H_0 or H_1

Definition: If a statistical hypothesis H completely specifies the distribution of (X_1, \dots, X_n) , then it is **simple**; otherwise H is called **composite**.

eg X_1, \dots, X_n iid $N(\mu, 1)$, $\mu \in \mathbb{R}$

$H_0: \mu = 0$ (simple hypothesis, we know $X_i \sim N(0, 1)$)

$H_0: \mu = 1$ (simple hypothesis)

$H_0: \mu \leq 0$ (composite hypothesis)

$H_0: \mu \neq 0$ (composite hypothesis)

$H_1: \mu > 1$ (composite hypothesis)

$H_1: \mu = 2$ (simple hypothesis)

eg X_1, \dots, X_n
iid $N(\mu, \sigma^2)$

$H_0: \mu = 0$
(composite)

Hypothesis Testing I

Terminology: Test Functions

Definition: Let \mathcal{X} be the set of all possible values of $\underline{X} = (X_1, \dots, X_n)$. Then, a **test function** or a **test rule** $\phi(X_1, \dots, X_n)$ is a function from \mathcal{X} into $[0, 1]$ with the interpretation that if $\underline{X} = (X_1, \dots, X_n)$ is observed then H_0 is rejected with probability $\phi(\underline{X}) \equiv \phi(x_1, \dots, x_n)$.

data $\underline{X} \rightarrow \phi(\underline{X}) \in [0, 1] \Rightarrow$ "Coin Flip" $Y \sim \text{Bernoulli}(\phi(\underline{X}))$
 \Rightarrow if $\begin{cases} Y = 1 & (\text{w. prob } \phi(\underline{X})) \\ Y = 0 & (\text{w. prob } 1 - \phi(\underline{X})) \end{cases}$, reject H_0
 don't reject H_0 (accept H_0)

e.g. $\underline{X} \rightarrow \phi(\underline{X}) = 3/7$ reject H_0 w. prob $3/7$
 don't reject H_0 w. prob $4/7$

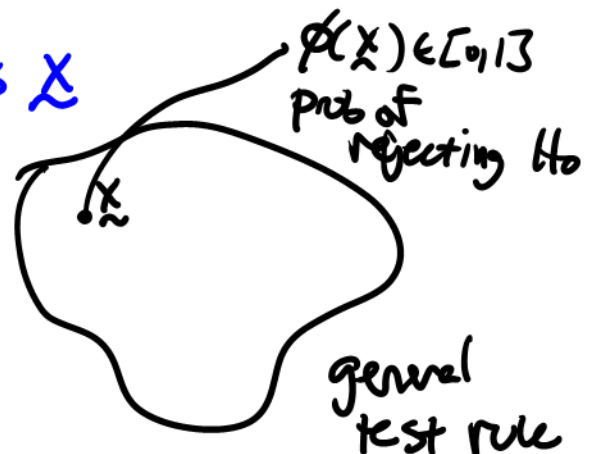
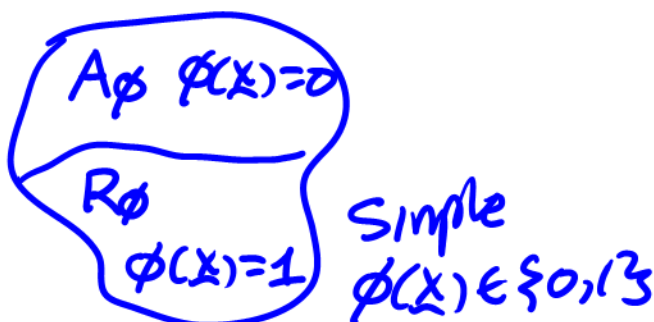
Note: $\phi(\underline{X}) = 1 \Rightarrow$ reject H_0
 $\phi(\underline{X}) = 0 \Rightarrow$ don't reject H_0

Definition: If $\phi(\underline{X}) \in \{0, 1\}$, $\forall \underline{X} \in \mathcal{X}$, then

1. $\phi(\underline{X})$ is called a **simple test function** (rule).
2. $R_\phi = \{\underline{X} : \phi(\underline{X}) = 1\}$ is called the **rejection region** of $\phi(\underline{X})$.
3. $A_\phi = \{\underline{X} : \phi(\underline{X}) = 0\}$ is called the **acceptance region** of $\phi(\underline{X})$.

← two possibilities
 reject H_0 ($\phi(\underline{X}) = 1$)
 or don't reject H_0 ($\phi(\underline{X}) = 0$)

Little Picture: all possible data outcomes \underline{X}



Hypothesis Testing I

Terminology: Error Probabilities

"accept" = don't reject H_0

action based on $\phi(\cdot)$

		fail to reject H_0	reject H_0
the state of nature	H_0 is true	OK	Type I error
	H_0 is false	Type II error	OK

Note: Suppose $\phi(\underline{X}) \in \{0, 1\}$ is a simple test rule for testing

$H_0 : \theta \in \Theta_0$ vs. $H_1 : \theta \notin \Theta_0$ for some subset $\Theta_0 \subset \Theta$. Then,

(Some parameter θ must generate data \underline{X})

1. for any $\theta \in \Theta_0$, the probability of type I error at θ

H_0 is true

$$P_{\theta}(\text{reject } H_0) = P_{\theta}(\underline{X} = (X_1, \dots, X_n) \in R_{\phi}) = E_{\theta} \phi(\underline{X}),$$

key step

$$\text{since } \phi(\underline{X}) = \begin{cases} 1 & \text{if } \underline{X} \in R_{\phi} \\ 0 & \text{if } \underline{X} \notin R_{\phi} \end{cases}$$

$$\Rightarrow E_{\theta} \phi(\underline{X}) = 1 \cdot P_{\theta}(\underline{X} \in R_{\phi}) + 0 \cdot P_{\theta}(\underline{X} \in A_{\phi}) = P_{\theta}(\underline{X} \in R_{\phi}) = P_{\theta}(\text{reject } H_0)$$

2. and for any $\theta \notin \Theta_0$, the probability of type II error at θ is

for $\theta \in \Theta_0^c$

$$P_{\theta}(\text{fail to reject } H_0) = P_{\theta}(\underline{X} \in A_{\phi}) = 1 - P_{\theta}(\underline{X} \in R_{\phi}) = 1 - E_{\theta} \phi(\underline{X}) = 1 - P_{\theta}(\text{reject } H_0)$$

$$\theta \xrightarrow{\text{generates}} \underline{X} \xrightarrow{\text{rule}} \phi(\underline{X}) \quad E_{\theta} \phi(\underline{X}) = P_{\theta}(\text{reject } H_0)$$

Want $E_{\theta} \phi(\underline{X})$ to be small whenever θ satisfies H_0
 $E_{\theta} \phi(\underline{X})$ to be large whenever θ doesn't satisfy H_0

Hypothesis Testing I

Terminology: Error Probabilities

$$\rightarrow \phi(X) \in [0, 1]$$

Remark: For any **general** test function, the same holds true: $\phi(\cdot)$,

1. Prob. of a type I error at θ ($\theta \in \Theta_0$) = $P_\theta(\text{reject } H_0) = E_\theta \phi(X)$

2. Prob. of a type II error at θ ($\theta \notin \Theta_0$) = $P_\theta(\text{fail to reject } H_0) = 1 - E_\theta \phi(X)$.

Same as for simple test rules ($\phi(X) = 0$ or 1)

e.g. $X \sim \text{Binomial}(2, \theta)$, $0 < \theta < 1$

general
test
rule

$$\phi(x) = \begin{cases} 1 & \text{if } x=0 \\ \frac{1}{2} & \text{if } x=1 \\ 0 & \text{if } x=2 \end{cases}$$

$$\Rightarrow E_\theta \phi(X) = 1 \cdot P_\theta(X=0) + \frac{1}{2} P_\theta(X=1) + 0 \cdot P_\theta(X=2) = P_\theta(\text{reject } H_0)$$

Definition: Let $\phi(X)$ be a test rule for testing $H_0 : \theta \in \Theta_0$ vs $H_1 : \theta \notin \Theta_0$,

1. $\max_{\theta \in \Theta_0} E_\theta \phi(X)$ is called **the size** or **the level** of $\phi(X)$

2. $\Pi_\phi(\theta) = E_\theta \phi(X)$ is called the **power function** of $\phi(X)$.

Note: For $\theta \in \Theta_0$, $\Pi_\phi(\theta) = E_\theta \phi(X)$ = probability of type I error

For $\theta \notin \Theta_0$, probability of type II error = $1 - \Pi_\phi(\theta) = 1 - E_\theta \phi(X)$