

STAT 5430

Lec 20, M , Mar 10

sufficiency
& completeness Homework 4 posted, due M, Mar 10

Basv's theorem
Intro to testing → Homework 5 posted, due M, Mar 29
(aft'r 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.
"claim"
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

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

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

$\downarrow H_0 \text{ or } H_1$

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**.

e.g. 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)

e.g. 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 $\phi(\underline{X}_1, \dots, \underline{X}_n)$ 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 $\phi(\underline{X}) = \phi(X_1, \dots, X_n)$.

data $\underline{X} \rightarrow \phi(\underline{X}) \in [0, 1] \Rightarrow$ "coin flip" $Y \sim \text{Bernoulli}(\phi(\underline{X}))$

\Rightarrow if $\begin{cases} Y=1 & (\text{w. prob } \phi(\underline{X})) \text{, reject } H_0 \\ Y=0 & (\text{w. prob } 1-\phi(\underline{X})) \text{, don't reject } H_0 \end{cases}$ (accept H_0)

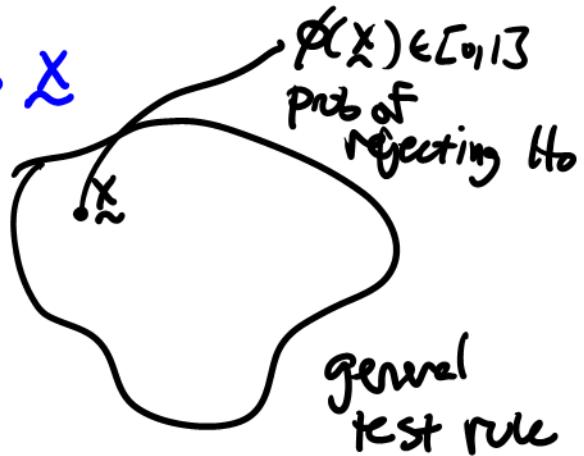
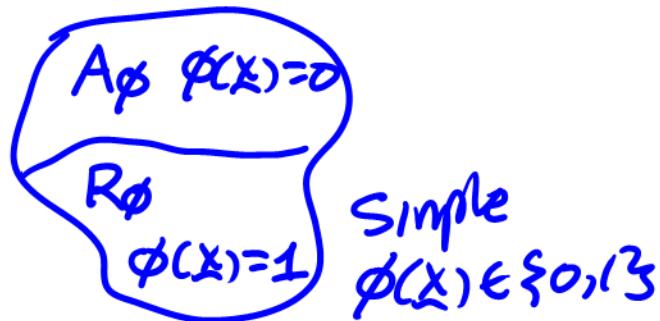
e.g. $\underline{X} \rightarrow \phi(\underline{X}) = \frac{3}{7}$ reject H_0 w. prob $\frac{3}{7}$
don't reject H_0 w. prob $\frac{4}{7}$

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

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



Hypothesis Testing I

Terminology: Error Probabilities

		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 $\phi(X) \in \{0, 1\}$ 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 X)

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

H_0 is true $P_{\theta}(X = (X_1, \dots, X_n) \in R_{\phi}) = E_{\theta}\phi(X) = \text{key step}$

since $\phi(X) = \begin{cases} 1 & \text{if } X \in R_{\phi} \\ 0 & \text{if } X \notin R_{\phi} \end{cases} \Rightarrow E_{\theta}\phi(X) = P_{\theta}(X \in R_{\phi}) + 0 \cdot P_{\theta}(X \in A_{\phi}) = P_{\theta}(X \in R_{\phi}) = P_{\theta}(\text{reject } H_0)$

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

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

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

Want $E_{\theta}\phi(X)$ to be small whenever θ satisfies H_0
 $E_{\theta}\phi(X)$ 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. $X \sim \text{Binomial}(2, \theta)$, $\theta < 1$

general test rule $\phi(x) = \begin{cases} 1 & \text{if } x=0 \\ \frac{1}{2} & \text{if } x=1 \\ 0 & \text{if } x=2 \end{cases} \Rightarrow E_{\theta}\phi(X) = 1 \cdot P_{\theta}(X=0) + \frac{1}{2} P_{\theta}(X=1) + 0 P_{\theta}(X=2) = P_{\theta}(\text{reject } H_0)$

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 = $1 - \Pi_{\phi}(\theta) = 1 - E_{\theta}\phi(X)$