

Hypothesis Testing

Stat 523

4-9-19

There will be 2 competing hypotheses, $H_0 \neq H_1$,
that make contradictory claims about a parameter. ①

Simple example: $H_0: \theta = 2$

$H_1: \theta = 3$

H_0 is the null hypothesis

H_1 is the alternative hypothesis

There will be a test statistic Λ

And a decision rule based on Λ

That is, \mathbb{R} will be partitioned into 2 sets

R_1 and R_2 , so that

if $\Lambda \in R_1$, H_0 is selected

if $\Lambda \in R_2$, H_1 is selected.

		Decision	
		H_0	H_1
Actually true	H_0	✓	Type I
	H_1	Type II	✓

(3)

$$\text{Let } \alpha = P(\text{rejecting } H_0 \mid H_0 \text{ was actually true}) \\ = P(\text{Type I error})$$

Select $H_1 \Rightarrow$ "Reject H_0 "

Select $H_0 \Rightarrow$ "Fail to reject H_0 "

Likelihood Ratio Test (LRT)

$H_0: \theta \in \Omega_0$ where $\Omega_0 \subset \Omega$ (parameter space)

$H_1: \theta \notin \Omega_0$

$$\text{Let } \Lambda = \frac{\sup_{H_0} L(\theta)}{\sup_{\Omega} L(\theta)}$$

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Decision rule: Reject H_0 when $\Lambda \leq c$,

where c is selected so that $P_{H_0}(\Lambda \leq c) = \alpha$.

Example: $X_1, \dots, X_n \sim \text{iid Exp}(\lambda)$ $f(x) = \lambda e^{-\lambda x}$ $x > 0$

$H_0: \lambda = \lambda_0$

$\Omega = \{\lambda \mid \lambda > 0\}$

$H_1: \lambda \neq \lambda_0$

$\Omega_0 = \{\lambda_0\}$

$$L(\lambda) = \prod_{i=1}^n \lambda e^{-\lambda x_i} = \lambda^n e^{-\lambda \sum x_i}$$

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Denominator of Λ : $\hat{\lambda}_{MLE} = \frac{1}{\bar{x}}$

$$L(\hat{\lambda}_{MLE}) = \frac{1}{\bar{x}^n} e^{-\frac{1}{\bar{x}} \sum x_i} = \frac{e^{-n}}{\bar{x}^n}$$

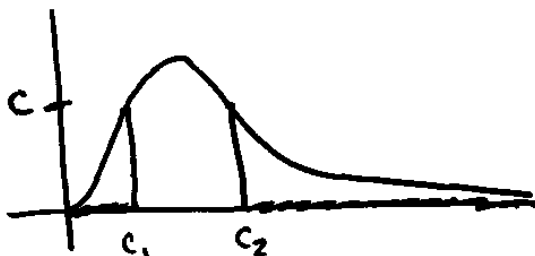
Numerator of Λ : $L(\lambda_0) = \lambda_0^n e^{-\lambda_0 \sum x_i}$

$$\begin{aligned} \Lambda &= \frac{\lambda_0^n e^{-\lambda_0 \sum x_i}}{e^{-n}/\bar{x}^n} = e^n \lambda_0^n \bar{x}^n e^{-n\lambda_0 \bar{x}} \\ &= e^n (\lambda_0 \bar{x})^n e^{-n\lambda_0 \bar{x}} \end{aligned}$$

Decision rule: reject H_0 when $\Lambda \leq c$.

(6)

What does $g(t) = t^n e^{-nt}$ look like?



So $\Lambda \leq c$ is equivalent to $\lambda_0 \bar{x} \geq c_2$
or $\lambda_0 \bar{x} \leq c_1$

OR

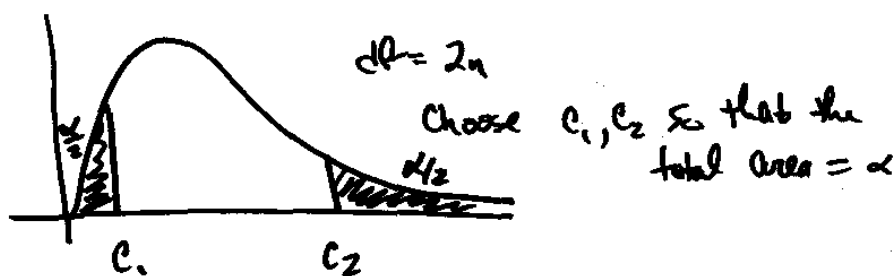
$$\begin{aligned} \sum x_i &\geq c_2' \\ \sum x_i &\leq c_1' \end{aligned}$$

(7)

Under H_0 , $X_1, \dots, X_n \sim \text{Exp}(\lambda_0)$

$$\text{So } \sum X_i \sim \text{Gamma}(\alpha = n, \beta = \frac{1}{\lambda_0})$$

$$\text{Thus } 2\lambda_0 \sum X_i \sim \text{Gamma}(\alpha = n, \beta = 2) \\ \sim \chi^2_{2n}$$



Usually, we split the area into 2 equal parts

(8)

Final decision rule:

Reject H_0 if $2\lambda_0 \sum X_i \geq c_2$ or

$2\lambda_0 \sum X_i \leq c_1$, where

c_2 cuts off the upper $\alpha/2$ area

and c_1 " " " lower $\alpha/2$ area in

the χ^2 distr. with $2n$ df.

Wording of your conclusion:

Reject H_0 : found sufficient evidence favoring H_1 .

Fail to reject H_0 : failed to find sufficient evidence favoring H_1 .

Example: $X_1, \dots, X_n \sim \text{iid } N(\mu, \sigma^2)$
 \uparrow
 known

(9)

$$H_0: \mu \leq \mu_0$$

$$H_1: \mu > \mu_0$$

$$L(\mu) = \prod_{i=1}^n \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2\sigma^2}(x_i - \mu)^2}$$

$$= \sigma^{-n} (2\pi)^{-\frac{n}{2}} e^{-\frac{1}{2\sigma^2} \sum (x_i - \mu)^2}$$

$$\text{know } \hat{\mu}_{MLE} = \bar{x}$$

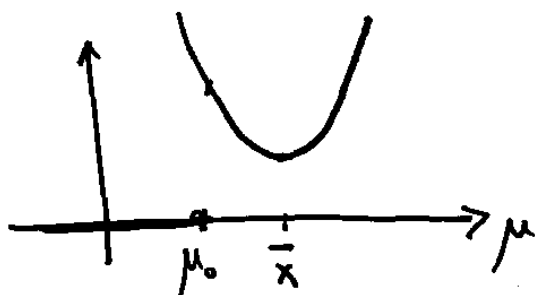
$$L(\hat{\mu}_{MLE}) = \sigma^{-n} (2\pi)^{-\frac{n}{2}} e^{-\frac{1}{2\sigma^2} \sum (x_i - \bar{x})^2}$$

$$= \sigma^{-n} (2\pi)^{-\frac{n}{2}} e^{-\frac{(n-1)s^2}{2\sigma^2}}$$

(10)

For the numerator, we need to maximize $L(\mu)$,
 subject to $\mu \leq \mu_0$

$$l(\mu) = \ln L(\mu) = -n \ln \sigma - \frac{n}{2} \ln(2\pi) - \frac{1}{2\sigma^2} \sum (x_i - \mu)^2$$



$$\sum x_i^2 - 2\mu \sum x_i + n\mu^2$$

$$n \left(\frac{1}{n} \sum x_i^2 - 2\mu \bar{x} + \mu^2 \right)$$

Subject to H_0 , $l(\mu)$

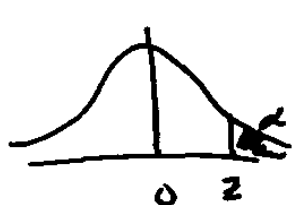
is maximized at
 $\mu = \mu_0$

$$\begin{aligned}
 \text{Now } \Lambda &= \frac{\sigma^n (2\pi)^{\frac{n}{2}} e^{-\frac{1}{2\sigma^2} \sum (x_i - \mu_0)^2}}{\sigma^n (2\pi)^{\frac{n}{2}} e^{-\frac{1}{2\sigma^2} \sum (x_i - \bar{x})^2}} \quad (11) \\
 &= e^{-\frac{1}{2\sigma^2} \left(\sum x_i^2 - 2\mu_0 \sum x_i + n\mu_0^2 - \left(\sum x_i^2 - 2\bar{x} \sum x_i + n\bar{x}^2 \right) \right)} \\
 &= e^{-\frac{n}{2\sigma^2} (-2\mu_0 \bar{x} + \mu_0^2 + 2\bar{x}^2 - \bar{x}^2)} \\
 &= e^{-\frac{n}{2\sigma^2} (\bar{x} - \mu_0)^2}
 \end{aligned}$$

Rule: Reject H_0 when $\Lambda \leq c$,
 i.e. when $\frac{(\bar{x} - \mu_0)^2}{\sigma^2/n} \geq c'$

This is equivalent to rejecting H_0 when

$$\frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \geq \sqrt{c_1} \quad \text{or} \quad \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \leq \sqrt{c_2}$$



Not feasible