STAT 135 Lecture 9

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Remark 0.1

Simple vs Simple Hypothesis Testing

Two possible values, $H_0: \theta = \theta_0 \text{ vs } H_1: \theta = \theta_1$

Simple vs simple hypothesis test takes the form

 $X_1,\ldots,X_n\sim f$

 $H_0: f = f_0$

 $H_1: f = f_1$

Lemma 0.2 (Neyman-Pearson Lemma)

Likelihood ratio tests are good in simple vs simple hypothesis testing.

Method 0.3 (Likelihood Ratio Test playbook)

- 1. Write down likelihood ratios (LR, Λ)
- 2. Find c such that $P(LR < c|H_0) = \alpha$

Example 0.4

Normal with known variance $X_1, X_2, \ldots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2), \sigma^2$ is known

 $H_0: \mu = \mu_0$

 $H_1: \mu = \mu_1$

Solution.

$$LR = \frac{\exp\left[\frac{1}{-2\sigma^2} \sum_{i=1}^n (X_i - \mu_0)^2\right]}{\exp\left[\frac{1}{-2\sigma^2} \sum_{i=1}^n (X_i - \mu_1)^2\right]}$$
 Take log of LR

$$\log(LR) = \sum_{i=1}^{n} (X_i - \mu_0)^2 - \sum_{i=1}^{n} (X_i - \mu_1)^2 = 2n\overline{X}(\mu_0 - \mu_1) + n\mu_1^2 - n\mu_0^2$$

Reject \overline{X} large, find a c such that $P(\overline{X} > c|H_0) = \alpha$

Example 0.5

Laplace example

Have 1 observation $X \sim f(x|\sigma) = \frac{1}{2\sigma}e^{\frac{-|x|}{\sigma}}$

 $H_0: \sigma = 1$ $H_1: \sigma = 2$

Solution.

$$\Lambda = \frac{\frac{1}{2}e^{-|x|}}{\frac{1}{4}e^{\frac{-|x|}{2}}}$$

Reject for small values of Λ ie $-|x|+\frac{|x|}{2}$, small values $-\frac{|x|}{2}$, ie large values of |x| Need to find a critical value, ie c such that $P(|X|>c|H_0)=\alpha$ Under H_0 $X\sim f(x|1)=\frac{1}{2}e^{-|x|}$

$$2\int_c^\infty \frac{1}{2}e^{-x}dx = \alpha, \qquad \alpha = 0.05 \text{normal value of } \alpha$$

$$-e^{-x}\big|_c^\infty = 0.05 \to c = -\log 0.05 \approx 3$$

Example 0.6

p-values

 $X_1, X_2, \dots, X_{25} \stackrel{iid}{\sim} \mathcal{N}(\mu, 1)$ $H_0: \mu = 1 \ H_1: \mu = 2$

Solution. Reject for large values of \overline{X} . Under the null hypothesis $\overline{X} \sim N(1, \frac{1}{25})$

$$\begin{array}{c|c} \alpha & \text{Reject for } \overline{X} > c \\ \hline 0.10 & \overline{X} \geq 1.26 \\ 0.05 & \overline{X} \geq 1.33 \\ 0.03 & \overline{X} \geq 1.37 \\ \end{array}$$

Suppose we have that $\overline{X} = 1.37$, then we reject the null if we have $\alpha = 0.05$ To find the minimal α value where we would reject \overline{X} , we just take $1 - cdf(\overline{X})$

Definition 0.7 (p-value)

Without setting α in advance, can look at test statistic and ask what is the smallest α for which we reject H_0

Reject if p-value $\leq \alpha$

Ways to interpret the p-value

- 1. Smallest α for which we would reject H_0
- 2. P(Observing more extreme test statistics than we observed)
- 3. "Strength of evidence" against H_0

Method 0.8 (Composite Tests)

 $\theta \in \Omega$

 $H_0: \theta \in \omega_0$

 $H_1:\theta\in\omega_1$

 $\omega_0 \cup \omega_1 = \Omega$

 $\omega_0 \cap \omega_1 = \emptyset$

 $X_1, X_2, \dots, X_n \sim f(x|\theta)$

Method 0.9 (Generalized Likelihood Ratio Test)

$$\Lambda^* = \frac{\max_{\theta \in \omega_0} f(x_1, \dots, x_n | \theta)}{\max_{\theta \in \omega_1} f(x_1, \dots, x_n | \theta)} = \frac{\text{Strongest case for } H_0}{\text{Strongest case for } H_1}$$

$$\Lambda = \frac{\max_{\theta \in \omega_0} f(x_1, \dots, x_n | \theta)}{\max_{\theta \in \Omega} f(x_1, \dots, x_n | \theta)} = \min(\Lambda^*, 1)$$

Example 0.10

 $X_1, \ldots, X_n \stackrel{iid}{\sim} \mathcal{N}(\mu, 1)$

 $H_0: \mu = 5$

 $H_1: \mu \neq 5$

Solution.

$$\omega_0 = \{5\}, \omega_1 = \mathbb{R} - \{5\}$$

Generalized LR Stat :
$$\Lambda = \frac{\exp\left[-\frac{1}{2}\sum_{i=1}^{n}(x_i - 5)^2\right]}{\exp\left[-\frac{1}{2}\sum_{i=1}^{n}(x_i - \overline{X})^2\right]}$$

Reject for small values of Λ

Take logs etc., simplifies to

"reject for large values of $n(\overline{x}-5)^2$ ", $\sqrt{n}(\overline{X}-5) \sim \mathcal{N}(0,1)$, $n(\overline{X}-5)^2 \sim \chi_1^2$ Find k such that $P(n(\overline{X}-5)^2 > k|H_0) = \alpha$, reject for large values of $n(\overline{X}-5)^2$, same as reject large values of $|\overline{X} - 5|$

Theorem 0.11 (Wilk's Theorem)

Why am I mentioning χ^2 , more generally,

$$-2\log\Lambda \sim \chi_k^2, k = \dim\omega_1 - \dim\omega_0$$

Use the known distribution to get critical values

Reject for large values of $-2 \log \Lambda$ (quantiles of χ_k^2)