

Chapter 5 Hypothesis Testing

BY YUEJIAN MO

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Here are two important areas of **statistical inference**. The first one is the estimation of **parameters**, and the second is the **testing of hypotheses**. Based on observations from a random sample, statisticians follow a formal process to determine whether or not to reject a null hypothesis(H_0). This process is called **hypothesis testing**.

Here are four steps of hypothesis testing:

- *State the hypothesis*. This involves stating the null and alternative by hypotheses.
- *formulate an analysis plan*. The analysis plan describe how to use sample data to evaluate the null hypothesis.
- *analyze sample data*.
- *Interpret results*. Apply the decision rule decribed in the anylysis plan. If the value of the test statistic is unlikely, based on the null hypothesis, reject the null hypothesis.

A *statistical hypothesis* is an assumption about a population parameter. This assumption may or may not be true. A research can conduct a statistical experiment to test the validity of this hypothesis. If the statistical hypothesis specifies the population distribution, it is called *simple hypothesis*. Otherwise, it is called a *composite hypothesis*.

The first, the hypothesis being testing, is called the *null hypothesis*, denoted by H_0 . The second is called the *alternative hypothesis*, denoted by H_1 or H_a .

We

1 Type I error and Type II error

Being adjudged to be	The man is crimeless	The man is crime
Guilty(yes)	Type I Error(弃真)	
Guiltess(no)		Type II Error(取伪)

Table 1.

Rejeciton of the null hypothesis H_0 when it is ture is called *Type I error*. The probalibity of making a Type I error is denoted by

$$\begin{aligned}\alpha(\theta) = \Pr(\text{Type I error}) &= \Pr(\text{rejecting } H_0 | H_0 \text{ is true}) \\ &= \Pr(\mathbf{x} \in C | \theta \in \Theta_0)\end{aligned}$$

which is a function of θ defined in Θ_0 . $\alpha(\theta)$ is called the *Type I error function*.

Typically, the Type I error is more serious thant the Tyoe] II error. But we can adjust to H_0 and H_1 .

The values of the power function are the probabilities of rejecting the null hypothesis H_0 for various values of the parameter θ . The power function plays the same role in hypothesis testing as that (MSE) played in estimation. The power function is golden standard in assessing the goodness of a test T or in comparing two competing tests T_1 and T_2 .

$$\begin{aligned} p(\theta) &= \Pr \theta \in \Theta_0 \\ &= \Pr \theta \in \Theta_1 \\ &\quad \theta \notin \Theta_0 \cup \Theta_1 \end{aligned}$$

Fix the probability of Type I error at preassigned (small) level α^* , then minimize the probability of Type II error. That is, consider the test with

$$\sup_{\theta \in \Theta_0} p(\theta) = \sup_{\theta \in \Theta_0} \alpha(\theta) \leq \alpha^*$$

and choose the one with the probability of Type II error $\beta(\theta)$ being minimized. So if $\alpha_{T_1}(\theta)$

2 The Neyman-Pearson Lemma

the most powerful test(MPT)

the uniformly most powerful test(UMPT) 一致最优检验