Packet 6: Hypothesis Testing

Learning Objects:

- Understand basic concepts and formulating the hypothesis.
- Learn how to conduct and interpret classical hypothesis testing.

Hypothesis Testing: Simply put, to answer yes/no questions; E.g., whether people eating saturated fat are more likely to develop heart disease. θ : propertion of people with heart disease.

Statistical hypothesis (or hypothesis) is a statement about parameter(s) θ of a population.

Null v.s. Alternative: The claim or the research hypothesis we want to establish is called alternative hypothesis, H_1 , opposite of which is called the null hypothesis, H_0 .

Rodolfo Gonzalez's hypothesis:

 H_0 : human intelligence genes are not only carried by X chromesome.

 H_1 : human intelligence genes are only carried by X chromesome.



Gonzalez' hypothesis has some interesting implications. If a male has some outstanding intellectual ability (associated with the X-chromosome) he is likely to be disappointed in the abilities of his sons because that ability can only be passed on to his daughters.

Decision:

Reject H_0 and conclude that H_1 is substantiated

Not reject
$$H_0$$
 We never say accept H_0'' .

Type I and Type II Errors: Errors occur when decision is wrong

Type I error, H_0 is rejected when H_0 is true.

Type II error, H_0 is accepted when H_0 is false. not rejected

Court example: A suspect is not guilty v.s. A suspect is guilty.

 H_0 v.S. H,

Type I error: an innocent person is found guilty. (false rejection).

Type II error: a guilty person is found innocent. (false acceptance).

In the above criminal trail example, reject Ho presumption of innocence

Presumption of Innocence One is considered innocent until proven guilty. v.s. No person shall be found guilty without being judged as such by a court.

Evidence is needed to reject H_0 . quantitative evidence -> data

In statistics, we denote

If there is no evidence to prove guilty, we do not reject the, but it does not mean

 $\alpha = P(\text{type I error}) = P(\text{reject } H_0 \mid H_0 \text{ is true}).$ We have proved to is true

 $\beta = P(\text{type II error}) = P(\text{fail to reject } H_0 \mid H_0 \text{ is false}). \text{ Power } = 1 - \beta.$

Fire alarm example: VS. no fire

truth

legal right

of accused

decision alarm v.s. no alarm

Ho notire us. H. fire

which is Ho? yes: Ho five v.s. H, no five

mo: Ho no five v.s. H, five

smoke detector will need a

certain level of GO concentration to reject Ho no fire

Increase sensitivity of the detector: more likely to ring the alarm

 $d = P(alarm \mid nor fire)$ false alarm $\int never ring alarm <math>d = 0$ $\beta = 1$ $\beta = P(nor alarm \mid fire)$ λ always alarm d = 1 $\beta = 0$

There is a trade off between α and β . Given a hypothesis testing problem, we need to design a test such that α and β are balanced.

When type I error is more serious (like the criminal example), we design a test such that a preferred value of α is obtained (e.g., $\alpha = 0.05$). A good decision rule gives a small β .

Uniformly most powerful test (UMPT): a uniformly most powerful (UMP) test is a hypothesis test which has the greatest power, $1-\beta$, among all possible tests of a given size α .