

# Introduction and Basics in TOH

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# Introduction

- The objective of statistics is to make inference about a population based on information contained in the sample.
- **Population** denotes all items the researcher is interested in.
- **Sample** A subset of the population
- There are two major areas of statistical inference namely:
  - i. Estimation of parameters
  - ii. Testing of hypothesis
- We will develop general methods for testing hypothesis and apply to some common problems.

- We will consider a random variable  $X$  with pdf  $f(x; \theta)$  where  $\theta$  is an unknown parameter.
- **Statistical hypothesis** is a statement or assertion about the distribution of one or more random variables.
- **Hypothesis testing** is a rule for deciding whether to reject or accept the statistical hypothesis.
- **Null Hypothesis** This represents a theory that has been put forward either because it is believed to be true or because it is used as a basis for argument but has not been proved. Denoted by  $H_0$
- **Alternative Hypothesis** denoted by  $H_1$  means an opposing theory to the null hypothesis.

# Basics Cont'd

- The final conclusion once the test have been carried out is always given in terms of the null hypothesis.
- The two possible conclusions are:
  - i. Reject  $H_0$
  - ii. Fail to reject  $H_0$

# Simple and Composite Hypothesis

- A hypothesis is said to be **simple** if it specifies the population completely *i.e it specifies the probability distribution uniquely* whereas a **composite** hypothesis leads to two or more possibilities.
- Examples:
  - i. Suppose  $X \sim N(\mu, 1)$

$$H_0 : \mu = 0 \rightarrow X \sim N(0, 1)$$

$$H_a : \mu = 2 \rightarrow X \sim N(0, 2)$$

Both  $H_0$  and  $H_1$  are simple hypothesis.

ii. Suppose  $X \sim N(\mu, \sigma^2)$

$$H_0 : \mu = 5, \sigma^2 = 25 \rightarrow X \sim N(5, 25)$$

$$H_1 : \mu \neq 5, \sigma^2 = 6$$

$H_0$  is simple while  $H_1$  is composite.

iii. Suppose  $X \sim N(\mu, \sigma^2)$

$$H_0 : \mu \leq 25, \sigma^2 \leq 9$$

$$H_a : \mu > 25, \sigma^2 > 9$$

Both  $H_0$  and  $H_1$  are composite since we do not know the exact values of  $\mu$  and  $\sigma^2$

# Rejection and Acceptance Regions

- Let  $X_1, X_2, \dots, X_n$  be a random sample of size  $n$  from  $X$ . These observations generate a  $n$ -dimensional space which we denote by  $X$ , a sample space in  $\mathbb{R}^n$
- A statistical test partitions the sample space  $X$  into two disjoint subsets  $\omega$  and  $\bar{\omega}$  that if
  - a.  $x \in \omega$  we reject  $H_0$
  - b.  $x \in \bar{\omega}$  we accept  $H_0$  where  $X = (x_1, x_2, \dots, x_n)$  is a particular value of the observed vector.
- We call  $\omega$  the rejection or critical region and  $\bar{\omega}$  is called the acceptance region.
- Acceptance region is the set of values for which the null hypothesis is accepted while rejection(critical) region is the set of values for the test statistic for which  $H_0$  is rejected.

# Errors in decisions

- There are 2 possible errors:
  - i. Type I error: Reject  $H_0$  if it is true
  - ii. Type II error: Accepting  $H_0$  if it is false

	$H_0$ True	$H_0$ False
Reject $H_0$	Type I Error	Correct Rejection
Fail to Reject $H_0$	Correct Decision	Type II Error

- $\alpha = P(\text{Type I error})$
- $\beta = P(\text{Type II error})$
- $1 - \beta = \text{Power of test}$



# Significance level of a test

- The significance level of a statistical hypothesis test is a fixed probability of wrongly rejecting the null hypothesis  $H_0$  if in fact  $H_0$  is true.
- It is the probability of type I error and is set by the investigator in relation to the consequences of such an error.
- The significance level is usually denoted by  $\alpha$  thus  $\alpha = P(\text{Type I error})$
- Usually the level of significance is chosen to be  $0.001 = 0.1\%$ ,  $0.01 = 1\%$ ,  $0.05 = 5\%$ ,  $0.10 = 10\%$

# P-Value

- The probability value, p-value of a statistical hypothesis is the probability of getting a value of the test statistic at least as extreme as that observed by chance alone if the null hypothesis is true.
- It is the probability of wrongly rejecting null hypothesis if it true.
- We can make a decision on whether to reject or fail to reject  $H_0$  based on the p-value.
- If the p-value is less than 0.05  $\alpha$  level of significance we reject  $H_0$ .

# Example 1

Suppose a researcher is to test the hypothesis  $\theta = 0.90$  against the alternative  $\theta = 0.60$ . Let this test statistic be  $X$ , the observed number of successes in  $n = 20$  trials. If he will accept the hypothesis if  $X > 15$  or otherwise conclude  $\theta = 0.60$ . Obtain the significance level of the test and the probability of obtaining type II error.

# Solution

- We have the hypothesis

$$H_0 : \theta = 0.90$$

vs

$$H_a = 0.60$$

- The critical region is

$$\omega = \{0 \leq X \leq 14\}$$

- The acceptance region is

$$\bar{\omega} = \{15 \leq X \leq 20\}$$

- The distribution is:

$$f(x, \theta) = \binom{20}{x} \theta^x (1 - \theta)^{20-x}, \quad x = 0, 1, 2 \dots 20 \quad (1)$$

$$\begin{aligned} \alpha &= P(\text{Type I error}) \\ &= P(X \in \omega | H_0 \text{ is true}) \\ &= P(X \leq 15 | \theta = 0.90) \end{aligned} \quad (2)$$

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$$\begin{aligned} f(x, \theta) &= \binom{20}{x} (0.9)^x (0.1)^{20-x} \\ &= \sum_{x=0}^{14} \binom{20}{x} (0.9)^x (0.1)^{20-x} = 0.0114 \end{aligned} \quad (3)$$



$$\begin{aligned}\beta &= P(X \in \bar{\omega} | H_0 \text{ is false}) \\ &= P(x \in \bar{\omega} | H_1 \text{ is true}) \\ &= P(X \in \bar{\omega} | \theta = 0.60)\end{aligned}\tag{4}$$

$$\begin{aligned}f(x, \theta) &= \binom{20}{x} (0.6)^x (0.4)^{20-x} \\ &= \sum_{x=15}^{20} \binom{20}{x} (0.9)^x (0.1)^{20-x} = 0.1256\end{aligned}\tag{5}$$

## Example 2

If  $X \geq 1$  is the critical region for testing  $\theta = 0.2$  against  $\theta = 1$  on the basis of a single observation from the population whose pdf is

$$f(x, \theta) = \begin{cases} \theta e^{-\theta x} & \text{if } 0 \leq X < \theta \\ 0 & \text{otherwise} \end{cases} \quad (6)$$

Obtain the values of Type I and Type II errors.

# Solution



$$H_0 : \theta = 2 \text{ vs } H_1 : \theta = 1$$



$$\omega = \{X : X \geq 1\} \quad \bar{\omega} = \{X : X < 1\}$$



$$\begin{aligned} \alpha &= P(\text{Reject } H_0 | H_0 \text{ is true}) \\ &= P(x \in \omega | \omega = 2) \\ &= P(1 \leq X < 2 | \theta = 2) \end{aligned} \tag{7}$$





$$\begin{aligned}f(x, \theta = 2) &= 2e^{-2x} \\&= \int_1^{\infty} 2e^{-2x} dx \\&= -e^{-2x}]_1^{\infty} \\&= 0 + e^{-2} = 0.1353\end{aligned}\tag{8}$$



$$\begin{aligned}\beta &= P(\text{accept } H_0 | H_1 \text{ is True}) \\&= P(x \in \bar{\omega} | \theta = 1) \\&= P(X < 1 | \theta = 1) \\&= f(x, \theta = 1) = e^{-x}, x \geq 0 \\&= \int_0^1 e^{-x} dx = -e^{-x}]_0^1 = 0.6321\end{aligned}\tag{9}$$

# Power of a test

- The power of a test is the probability of rejecting the null hypothesis when it is false.
- It is given by  $1 - \beta$  where  $\beta$  is the probability of type II error.
- It is therefore the probability of not committing type II error.
- The power function of a statistical test of hypothesis  $H_0$  against an alternative  $H_1$  is given by:

$$\pi(\theta) = \alpha(\theta)$$

for values of  $\theta$  under  $H_0$

$$\pi(\theta) = 1 - \beta(\theta)$$

for values of  $\theta$  under  $H_1$

# Example

Let  $p$  be the probability that a coin will fall as a head in a single toss in order to test

$$H_0 : p = \frac{1}{2} \text{ vs } H_1 : p = \frac{3}{4}$$

The coin is tossed 5 times and  $H_0$  is rejected if more than 3 heads are obtained. Find the power of the test.

# Solution

- $X$  forms a binomial distribution with parameters  $n = 5$  and  $p = 0.5$  thus

$$P(X = x) = \binom{n}{x} p^x (1 - p)^{5-x}, x = 0, 1, 2, 3, 4, 5$$

- The Rejection region  $\omega = \{x : x > 3\}$  and  $\bar{\omega} = \{x : x \leq 3\}$
- $\beta = P(x \in \bar{\omega} | H_1 \text{ is true}) = P(X \leq 3 | p = \frac{3}{4})$

- $$\sum_{x=0}^3 \binom{5}{x} \left(\frac{3}{4}\right)^x \left(\frac{1}{4}\right)^{5-x} = 0.3671$$

- The power of the test is  $1 - \beta = 1 - 0.3671 = 0.6329$

Thank You!