

101學年 統計學(I) 授課老師：蔡碧紋

Statistical Inference : Tests of Hypotheses

# Scientific questions

1. Whether the true average lifetime of a certain brand of tire is at least 22,000 kilometers.
2. whether fertilizer A produces a higher yield of soybeans than fertilizer B.
3. Pharmaceutical company: decide on the basis of samples whether at least 90% of all patients given a new medication will recover from a certain disease.
4. 廠商宣稱每杯citi coffee 重量至少 300 g.

# Statistical Hypotheses testing

1. if the lifetime of the tire has pdf  $f(x) = \lambda e^{-\lambda x}$ ,  $x > 0$ , then the expected lifetime,  $\frac{1}{\lambda}$ , is at least greater than 22000.
2. decide whether  $\mu_A > \mu_B$ , where  $\mu_A$ , and  $\mu_B$  are the means of the two populations
3. whether  $p$ , the parameter of a binomial distribution is greater or equal to 0.9.
4. whether  $\mu > 300$ .

In each case, it is assumed that the stated distribution correctly describes the experimental conditions, and the hypothesis concerns the parameter(s) of the distribution.

# Hypotheses Testing

假設 (hypothesis) 就是我們對於母體參數的宣稱(claim, statement)

The Scientist formulates a statement concerning the value of a parameter.

A **test** of a statistical hypothesis is a procedure for deciding whether to “accept” or “reject” the hypothesis.

假設檢定 (hypothesis testing) 的目的就是要對這些宣稱提供統計上的檢驗, 以統計的檢定方法來推論假設的"真偽".

# Terms you should know about hypotheses testing

1. Null and alternative hypotheses ( $H_0$  vs  $H_1$ )
2. Test statistic,  $T(\mathbf{X})$ , and  $T(\mathbf{x})$  (the distribution of  $T(\mathbf{X})$ )
3. Significance level of the test  $\alpha$
4. Rejection region (critical region) (RR) and acceptance region
5. Type I and Type II error probabilities.
6.  $p$ -value
7. power

# Null and Alternative Hypotheses

Assume that the form of the distribution for the population is known ,  $X \sim F(x; \theta)$  where  $\theta \in \Omega$ , where  $\Omega$  is the set of all possible values of  $\theta$  can take, and is called the parameter space.

The statistical hypothesis is a statement about the value of the parameter(s) of the distribution, such as

$$“\theta \in \omega”$$

$\omega$  be a subset of  $\Omega$ .

This is a statistical hypothesis and is denoted by  $H$  ( $H_0$ ), called **Null Hypothesis**

# Null and Alternative Hypotheses

On the other hand, the statement “ $\theta \in \bar{\omega}$ ” (where  $\bar{\omega}$  is the complement of  $\omega$  w.r.t  $\Omega$ ) is called the **alternative to  $H_0$**  and is denoted by  $H_a$ (or  $H_1$ ).

we write

$$H_0 : \theta \in \omega \quad \text{and} \quad H_a : \theta \in \bar{\omega} \quad (\text{or } \theta \notin \omega)$$

## $H_0, H_1$

1. In some case, we want to know the mean of something is as what people stated (or represents the *status quo*) we put it in the null hypothesis  $H_0$   
主計處調查國民平均月所得為20000元

$H_0$  : \_\_\_\_\_ vs  $H_1$  : \_\_\_\_\_

合歡山一月平均降雪量為20 公分。

$H_0$  : \_\_\_\_\_ vs  $H_1$  : \_\_\_\_\_

2. Often hypothesis arise in the form that we want to know if a new product, technique, teaching method, etc., is better than the existing one. In this context,  $H_0$  is a statement that **nullifies** the theory and is sometimes called a null hypothesis. In this case, 我們把想要檢定的假設定為  $H_1$  ,  $H_0$  則為其相反之假設。



## $H_0, H_1$

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# Hypotheses Testing

- ▶ 假設檢定係指在尚未蒐集樣本資料、進行推論之前，就事先對母體的某種特徵性質作一合理的假設敘述，再利用隨機抽出的樣本及抽樣分配，配合機率原理，以判斷此項假設是否為真。

以統計方法進行決策的過程中，會提出兩個假設：

$H_0$ : null hypothesis (虛無假設)。

$H_1$ : alternative or research hypothesis(對立假設、研究假設)。

可能的結論：

1. 有足夠的統計證據可推論  $H_1$  為真 (reject  $H_0$  and accept  $H_1$ )。
2. 沒有足夠的統計證據可推論  $H_1$  為真 (do not reject (Fail to reject, retain)  $H_0$ . The data doesn't provide

# Hypotheses Testing

- ▶ 假設檢定的主要精神在於尋找證據來拒絕 $H_0$ 而接受 $H_1$ ，我們無法證明 $H_0$ 為絕對正確，只有不能拒斥它。

**證據的角色:** 假設  $H_0$  為真的情況下，嘗試在其間找出矛盾，然後進行推論。

假設  $H_0$  為真，收集到此資料的可能性，如果是異常稀少事件（顯著的異常），則判定 $H_0$ 的假設是錯誤，所以拒絕 $H_0$ 。

因此假設檢定又稱為『顯著性檢定』（significant test）

# simple and composite hypothesis

$H_0 : \mu = 166$  vs  $H_1 : \mu > 166$ ,

$H_1 : \mu < 166$  or

$H_1 : \mu \neq 166$

If  $\omega$  contains only **one point**, i.e., if  $\omega = \{\theta : \theta = \theta_0\}$  then  $H_0$  is called a **simple** hypothesis which completely specifies the null distribution. We write it as  $H_0 : \theta = \theta_0$ .

Otherwise, if it does not completely specify the distribution. It is called **composite** hypothesis.

# Test Statistics

**Test Statistics:**  $T(\mathbf{X})$ ,

a function of a set of i.i.d. random variables  $X_1, \dots, X_n$   
which follow some distribution  $F(x, \theta)$ .

Such as  $\bar{X}$  or  $S^2$  etc.

Or a pivotal quantity for the test statistic

# Significance level of the test

We defined an event with small probability.

When  $H_0$  is true, the probability of an extreme event such as that  $\{\bar{X} > c\}$  is very small. If  $\{\bar{X} > c\}$ , we will reject the null hypothesis.

This small probability is called **the significance level of the test**, denoted by  $\alpha$ .

$$\mathbf{P}(\bar{X} > c | H_0 : \mu = 160) = \alpha$$

Thus the hypothesis testing is also called the test of significance.

# Rejection Region

If  $\bar{X} > c$ , then we will reject  $H_0$ . T

$\{\bar{X} > c\}$  is called the **rejection region, RR** (or **critical region**) denoted by  $R$ .

Note that critical value is defined before (ex-ante) we collect the data.

We will make the decision by comparing  $\bar{x}$  with  $c$

# Decision rules

we use data (random sample) to test if the data provides significant evidence to reject the null hypothesis.

If  $\bar{X} > c$  reject  $H_0$

A **test of hypotheses** is a rule, or decision, based on a sample from a given distribution to show whether the data support our hypothesis.



# Decision

Data:

If  $\bar{x} > c$  reject  $H_0$

Normally, the conclusion is either

1. Reject  $H_0$  and conclude that  $H_1$  is true, or
2. Do not reject  $H_0$

## Remark:

Note that, we don't say we "accept"  $H_0$ , because it implies stronger action than is really warranted. We can't find enough evidence to reject  $H_0$  but it does not mean that  $H_0$  is absolutely true.

## Example: 7.2-1

Test if  $X$ , the breaking strength of a steel bar.

$X \sim \mathcal{N}(50, 36)$  or  $X \sim \mathcal{N}(55, 36)$ . We want to know if new method increases the mean of the strength to 55.

Draw sample and test if they are from normally distributed population with mean 55.

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Draw sample and test if they are from normally distributed population with mean 55.

1. two hypotheses:  $H_0 : \mu = 50$  vs  $H_a : \mu = 55$
2. test statistic:  $\bar{X}$
3. Decision rule: If  $\bar{X} > 53$ , we will reject  $H_0$ .  
Rejection Region  $R = \{\bar{x} > 53\}$

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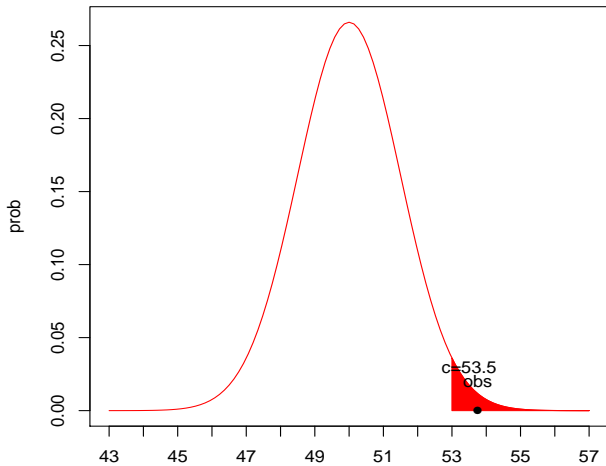
Draw sample and test if they are from normally distributed population with mean 55.

1. two hypotheses:  $H_0 : \mu = 50$  vs  $H_a : \mu = 55$
2. test statistic:  $\bar{X}$
3. Decision rule: If  $\bar{X} > 53$ , we will reject  $H_0$ .  
Rejection Region  $R = \{\bar{x} > 53\}$

**Data:** 16 random samples are drawn.

We observed  $\bar{x} = 53.75$ , then we will reject  $H_0$ .

### Distribution for $\bar{x}$ under $H_0$



## Example: The significance level of the test

The probability of a rare event when  $H_0$  is true.

The significance level of the test is

$$\mathbf{P}(\bar{X} > 53 | H_0 : X_i \sim \mathcal{N}(50, 36))$$

$$\mathbf{P}\left(\frac{\bar{X} - 50}{6/\sqrt{16}} > \frac{53 - 50}{6/\sqrt{16}}\right) = \mathbf{P}(Z > 12/3) = 0.023$$

The significance level of the test given  $R = \{\bar{x} > 53\}$  is  $\alpha = 0.02$ .

# Type I and Type II error probabilities

Probability of making a wrong decision:

There are two types of errors that can occur.

		Our decision	
		Reject $H_0$	Accept $H_0$
Actual situation	$H_0$ is true	Type I error	good
	$H_0$ is false	good	Type II error

Probabilities associated with the two incorrect decisions are denoted by type I and type II error probabilities.

# Type I and Type II error probabilities

1. Reject  $H_0$  when it is true.

$$\alpha = \mathbf{P}(\text{Type I error}) = \mathbf{P}(\text{reject a true } H_0) = \mathbf{P}(\text{reject } H_0 | H_0 \text{ is true})$$

2. Fail to reject  $H_0$  when it is false (Fail to accept  $H_1$  when  $H_1$  is true)

$$\beta = \mathbf{P}(\text{Type II error}) = \mathbf{P}(\text{Retain a false } H_0) = \mathbf{P}(\text{retain } H_0 | H_1 \text{ is true})$$



# Test statistics I

Often we work with the distribution of  $T(\mathbf{X})$ , pivotal quantity for the test statistic, such as a standard normal,  $t$ ,  $\chi^2$ , or  $F$ .

Pivotal quantity: a function of (data) observations and unknown parameters whose distribution does not depend on the unknown parameters

## Test statistics II

1. If  $X_1, \dots, X_n \sim \mathcal{N}(\mu, \sigma^2)$  with  $\mu$  is the unknown parameter and  $\sigma^2$  is some known constant, we have  $\bar{X} \sim \mathcal{N}(\mu, \sigma^2/\sqrt{n})$ ,

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim \mathcal{N}(0, 1)$$

2. If  $X_1, \dots, X_n \sim \mathcal{N}(\mu, \sigma^2)$  where  $\mu$  and  $\sigma^2$  are unknown parameters, we have

$$\frac{\bar{X} - \mu}{S/\sqrt{n}} \sim t(n - 1)$$

## Test statistics III

3.  $X_1, \dots, X_n \sim \text{Bernoulli}(p)$  and  $Y = X_1 + \dots + X_n$  we have

$$\frac{(Y/n) - p}{\sqrt{\frac{p(1-p)}{n}}} \rightarrow \mathcal{N}(0, 1)$$

4.  $X_1, \dots, X_n \sim \mathcal{N}(\mu, \sigma^2)$  where  $\mu$  and  $\sigma^2$  are unknown parameters and let  $S^2 = \sum(X_i - \bar{X})^2 / (n - 1)$ .

$$W = \frac{(n-1)S^2}{\sigma^2} \sim \chi^2(n-1)$$

## Test statistics IV

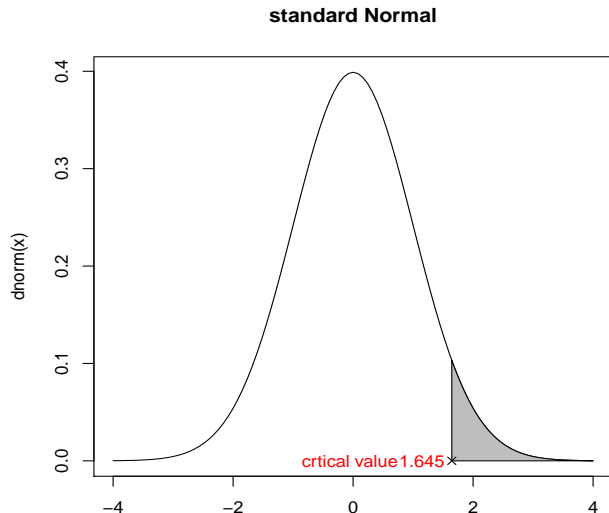
Based on the distribution of the test statistic we define the **Critical value  $t^*$** , such as

$z_\alpha$ ,  $t_\alpha$ ,  $\chi_\alpha^2$ , or  $F_\alpha$  and

construct the decision rule (rejection region) for a given the significance level of the test  $\alpha$ .

# Critical value for $\alpha=0.05$

EX:  $\mathbf{P}(Z \geq 1.645) = 0.05$ ,  $z_{0.05} = 1.645$



# The procedure

1. Specify the null and alternative hypotheses.
2. Specify the significant level of the test  $\alpha$ . (control the Type I error)
3. Define a test statistic  $T(\mathbf{X})$  and its distribution under  $H_0$
4. Decision rule: obtain the rejection region  
$$R = \{\mathbf{x} : T(\mathbf{x}) \in R(\theta_0)\}.$$
5. Obtain the data and calculate the value of the test statistic  $T(\mathbf{x})$  ( $T_{\text{obs}}$ )
6. Conclusion: If the test statistic  $T(\mathbf{x}) \in R(\theta_0)$  reject  $H_0$  and conclude that there is strong evidence to reject the null hypothesis at the significant level  $\alpha$

## Example ( $n=16$ , $\bar{x} = 53.75$ )

1. Hypotheses:  $H_0 : \mu = 50$  vs  $H_a : \mu = 55$
2. Given  $\alpha = 0.05$  (Significance level of the test)
3. Test statistic:

$$Z = \frac{\bar{X} - 50}{6/\sqrt{16}} \sim^{H_0} N(0, 1)$$

4. Decision rule: if  $Z_{\text{obs}} > Z_{0.05} = 1.645$  (critical value)  
Reject  $H_0$
5. Data  $\bar{x} = 53.75$  we have

$$z_{\text{obs}} = \frac{53.75 - 50}{6/4} = 2.5$$

6. Because  $z_{\text{obs}} = 2.5 > 1.645$   
Reject  $H_0 : \mu = 50$  at  $\alpha = 0.05$ .

## Probability value (p-value) of the test

When  $H_0$  is true, the probability that the test statistic is equal to or exceeds the actually observed value toward the direction of the alternative hypothesis.

Tail-end probability under  $H_0$  toward  $H_1$ .

$$\mathbf{P}(\bar{X} \geq 53.75 | H_0) = \mathbf{P}(Z \geq \frac{3.75}{6/4}) = \phi(2.5) = 0.006$$

$$\bar{X} \sim^{H_0} \mathcal{N}(50, 36/16)$$

The **p-value of the test** is 0.006.

**Small p-value** provides evidence to **reject the null hypothesis  $H_0$**  given the data.



## Decision rule

If the  $p$ -value of a test is as small or smaller than the significance level of a test,  $\alpha$ , we say the data are statistically significant at an  $\alpha$  significant level.

The  $p$ -values : if  $p\text{-value} < \alpha$  reject  $H_0$  at significance level  $\alpha$ .

(We don't need to find different  $t^*$  for different significance level  $\alpha$ .)

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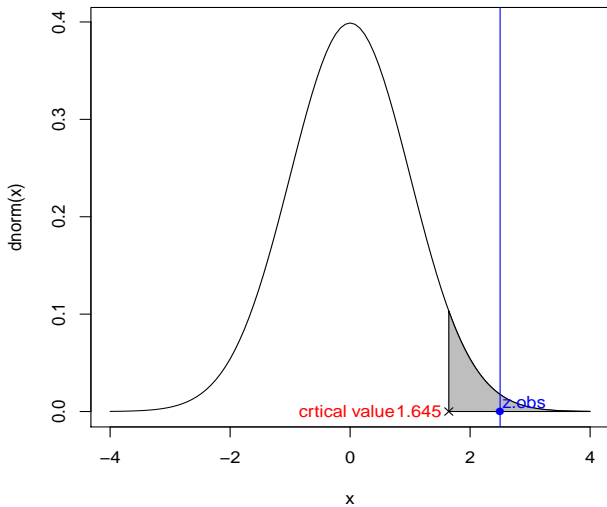
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(We don't need to find different  $t^*$  for different significance level  $\alpha$ .)

Recall: Decision rule by critical value

if  $T(\mathbf{x}) \in R(\theta_0)$  reject  $H_0$  at significance level  $\alpha$ .

### standard Normal



# Power

Power= $\mathbf{P}$ (Accept  $H_1$  when  $H_1$  is true)

Power= $1 - \beta$ ,  $1 - \beta$  is defined as the **power** of the test.

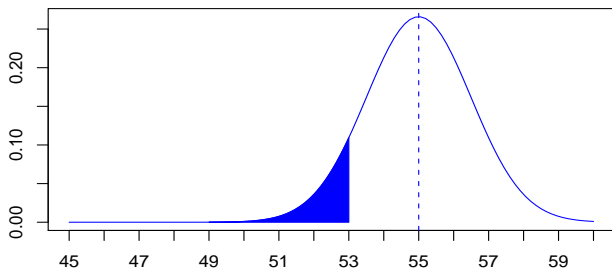
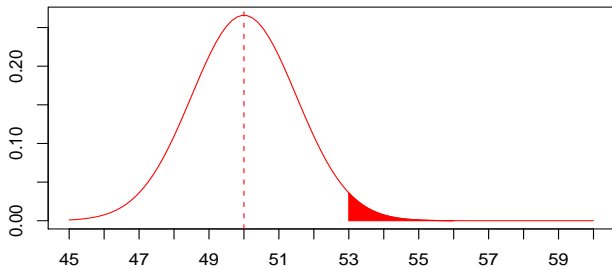
## $\alpha$ , $\beta$ and Power

### Example (Ex 7.2-1)

Let  $X$  be the breaking strength of a steel bar.  $H_0 : \mu = 50$   
vs  $H_a : \mu = 55$  Given  $C = \{(x_1, \dots, x_n) : \bar{x} \geq 53\}$  or  
 $C = \{\bar{x} : \bar{x} \geq 53\}$  Data:  $n=16$ , what are the type I and  
type II error probabilities?

$$\bar{X} \sim N(50, 36/16) \text{ under } H_0 : \mu = 50$$

$$\bar{X} \sim N(55, 36/16) \text{ under } H_1 : \mu = 55$$



## $\alpha$ , $\beta$ and Power

Type I error rate = 0.0228

Type II error rate = 0.0912

The significance level of the test  $\alpha = 0.0228$ .

The power of the test is  $1 - \beta = 0.9088$ .

# The relationship between $\alpha$ and $\beta$

