

## Testing of Hypothesis

A statistical hypothesis is an assertion about the probability distribution of a population.

Suppose we consider marks of students in two tests that have been conducted

Let  $f(x)$  be the dist<sup>n</sup>. Then we want to test hypothesis

Null hypothesis  $H_0$ :  $f(x)$  is  $N(\mu, \sigma^2)$ .

Alternative hyp  $H_1$ :  $f(x)$  is not  $N(\mu, \sigma^2)$

Another example: The average marks of students in PS class.

Consider two groups. Group I: the

students who regularly follow the class . Group I : the students who are irregular .

$\mu_1 \rightarrow$  average marks of the first group .

$\mu_2 \rightarrow$  average marks of the second group

Then we would like to test

$$H_0: \mu_1 = \mu_2 \rightarrow \text{Null Hyp}$$

$$H_1: \mu_1 > \mu_2 \rightarrow \text{Alt. Hyp.}$$

Consider a drug being used for a certain disease. Let  $p$  be the proportion of patients who recover using the drug. Suppose it is known that  $\boxed{p = 0.8} \rightarrow$

A new drug is discovered and let  $P$  is the proportion of recovery using the new drug. Then we want

to test  $\neq$   $H_0: P \leq 0.8$

$\leftarrow H_1: P > 0.8$

null hyp

alt. hyp

$$\begin{cases} H_0: P > 0.8 \\ H_1: P \leq 0.8 \end{cases}$$

Simple Hypothesis  $\rightarrow$  The prob. dist<sup>n</sup> is completely specified.

Composite Hypothesis → The prob.  
dist<sup>n</sup> is not completely specified.

Test of Statistical Hypothesis → is  
a procedure based on a random  
sample from the given population to  
check if the null or alternative  
are more likely under the sample.

So we reject  $H_0$  on the basis of  
sample (accept  $H_1$ )

or we do not reject  $H_0$  (accept  $H_0$ )

## Acceptance and Rejection Regions.

Let  $\Sigma$  be the sample space of  
the random sample.

A rejection region (critical region)

of the test is that part of the sample space which corresponds to the rejection of the null hypothesis

$$x \in R$$

$$R \subset S$$

$$x \in R^c \rightarrow \text{acceptance region} \quad R^c \subset S$$

Example: A coin is tossed thrice independently and  $p = P(\text{Head})$

$$H_0: p = \frac{1}{4}$$

$$H_1: p = \frac{3}{4}$$

$$S = \{ HHH, HHT, HTH, THH, HTT, THT, TTH, TTT \}$$

We may choose

$$A = R^c = \{ HTT, THT, TTH, TTT \}$$

$$R = \{ HHH, HHT, HTH, THH \}$$

That is , if one or none heads are observed , we feel  $H_0$  is true and if two or three heads are observed then we feel  $H_1$  is true .

Two Types of Errors : We are likely to make two types of errors when we conduct a test of hypothesis :

Type I Error : Rejecting  $H_0$  when  
it is true

→ Error of the first kind

Type II Error : Accepting  $H_0$  when it  
is false

→ Error of the second kind

$$\alpha = P(\text{Type I error})$$

$$\beta = P(\text{Type II error})$$

In the above example , we can find

$$\alpha = P(X \in R \mid H_0 \text{ is true})$$

$X \rightarrow$  number of heads

$$X \sim \text{Bin}(3, p)$$

$$\begin{aligned} \alpha &= P(X \geq 2) = \binom{3}{2} \left(\frac{1}{4}\right)^2 \frac{3}{4} + \binom{3}{3} \left(\frac{1}{4}\right)^3 \\ &\stackrel{p=1/4}{=} \frac{10}{64} = \frac{5}{32} \end{aligned}$$

$$\beta = P(X \leq 1) = \binom{3}{0} \left(\frac{1}{4}\right)^3 + \binom{3}{1} \frac{3}{4} \left(\frac{1}{4}\right)^2$$

$\downarrow p=3/4$

$$= \frac{1}{64} = \frac{1}{512}.$$

$1 - \beta = r =$  Power of the test  
 $= P(\text{Accepting } H_0 \text{ when it is false})$

Most Powerful Test : It is not

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possible to minimize both  $\alpha$  &  $\beta$

simultaneously. So generally we fix  $\alpha$  ( up to a certain level ) and then consider test procedure which minimizes  $\beta$  ( or maximizes  $\gamma = 1 - \beta$  ). This is called most powerful test ( MP test ).

Neyman & Pearson proved that

for simple null hyp. against simple alternative hypothesis the MP test exists and can be found.

## Testing for Parameters of a Normal

Population: Let  $x_1, \dots, x_n$  be a random sample from  $N(\mu, \sigma^2)$  population.

Testing for mean  $\mu$ .

Case I :  $\sigma^2$  is known

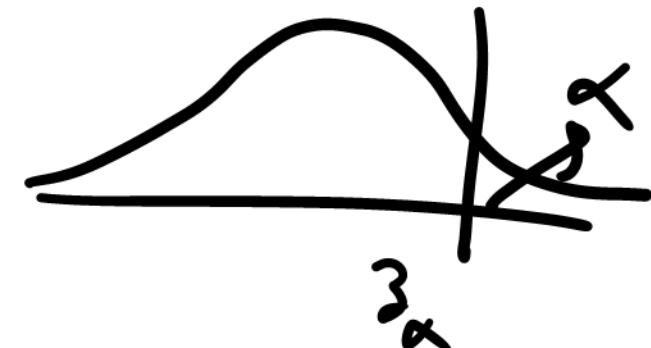
$$Z = \frac{\sqrt{n}(\bar{X} - \mu_0)}{\sigma} \quad \left\{ \begin{array}{l} \text{Test} \\ \text{statistic} \end{array} \right.$$

(P<sub>1</sub>)  $\left\{ \begin{array}{l} H_0: \mu \leq \mu_0 \text{ (fixed number)} \\ H_1: \mu > \mu_0 \end{array} \right.$

We reject  $H_0$  if  $Z > z_\alpha$

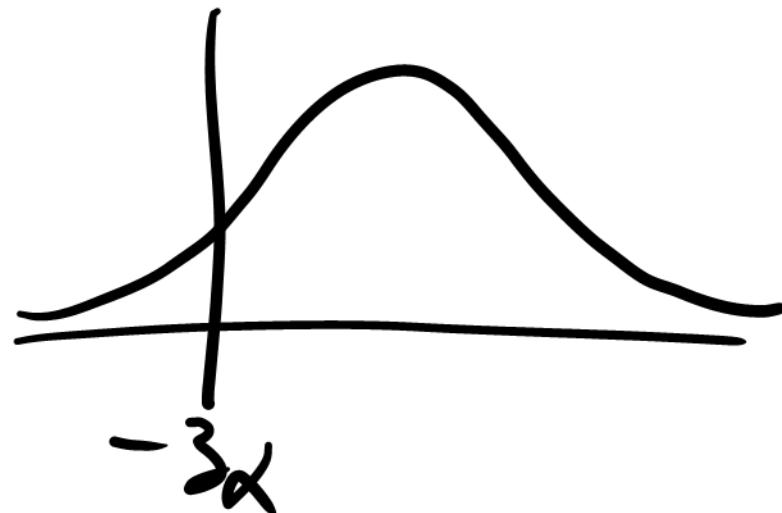
$z_\alpha \rightarrow$  Upper  $100(1-\alpha)\%$  point

of standard normal distn



Accept  $H_0$  otherwise.

(P2)  $\left. \begin{array}{l} H_0: \mu \geq \mu_0 \\ H_1: \mu < \mu_0 \end{array} \right\}$



Reject  $H_0$  if  $Z < -z_\alpha$ .

$$P_3 \left\{ \begin{array}{l} H_0: \mu = \mu_0 \\ H_1: \mu \neq \mu_0 \end{array} \right.$$



Rejecting  $H_0$  if  $|Z| \geq 3\sigma\gamma_2$

Case II:  $\sigma^2$  unknown

$$T = \frac{\sqrt{n}(\bar{X} - \mu_0)}{S}$$

$$S^2 = \frac{1}{n-1} \sum (X_i - \bar{X})^2$$

For  $\textcircled{P1} \rightarrow \text{Reject } H_0 \text{ if } T > t_{\alpha, n-1}$

$100(1-\alpha)$ . point  
of  $t_{n-1}$  dist<sup>n</sup>.



For  $\textcircled{P2} \rightarrow \text{Rejecting } H_0 \text{ if } T < -t_{\alpha, n-1}$

For  $\textcircled{P3} \rightarrow \text{Reject } H_0 \text{ if } |T| \geq t_{\alpha/2, n-1}$

Example : Suppose 12 electronic items

are tested and their lifetimes are (in months) recorded as (36.1, 40.2, 33.8, 38.5, 42, 35.8, 37, 41, 36.8, 37.2, 33, 36)

$$H_0: \mu \geq 40 \text{ vs } H_1: \mu < 40$$

$$\bar{x} = 37.2833, \quad \delta = 2.7319$$

$$T = \frac{\sqrt{n}(\bar{x}-40)}{\delta} = \frac{\sqrt{12}(37.2833-40)}{2.7319}$$

$$= -3.44, \quad t_{0.05, 11} = 1.796$$

We want to see if  $T \leq -t_{\alpha, n-1}$

$$-3.44 < -1.796$$

So  $H_0$  is rejected.  
is true

## Testing for Variance

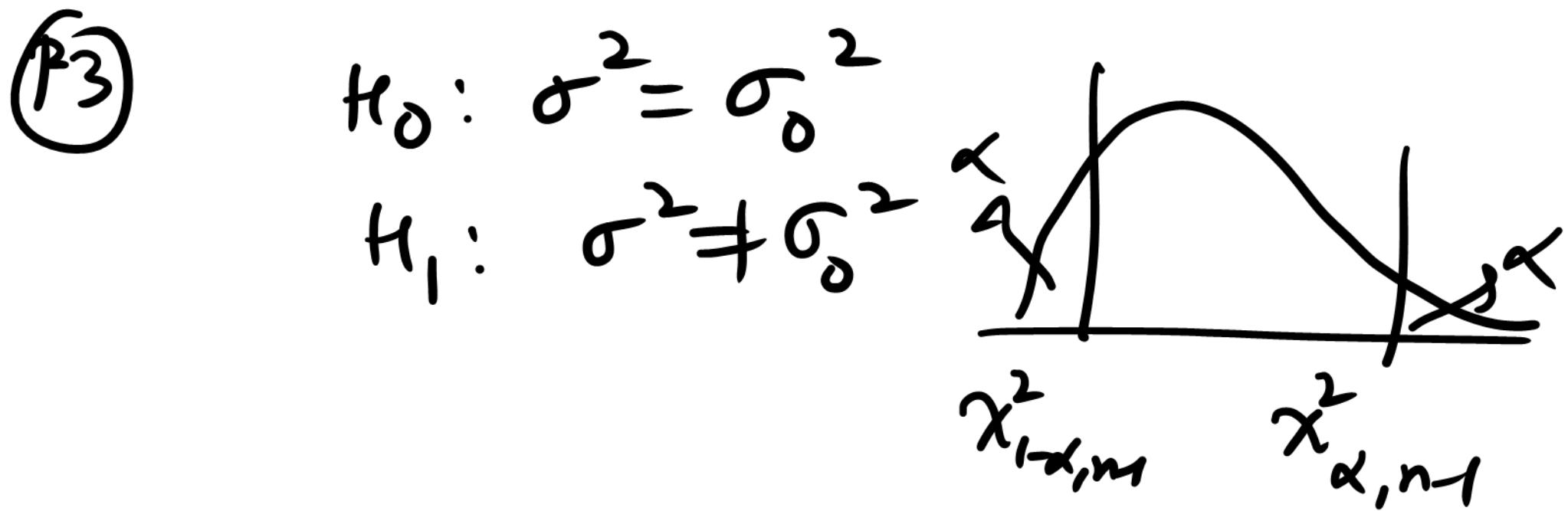
$$W = \frac{(n-1) S^2}{\sigma_0^2}$$

$$(P1) \quad H_0: \sigma^2 \leq \sigma_0^2$$

$$H_1: \sigma^2 > \sigma_1^2$$

$$(P2) \quad H_0: \sigma^2 \geq \sigma_0^2$$

$$H_1: \sigma^2 < \sigma_1^2$$



For (P1) : Reject  $H_0 \nabla W > \chi^2_{\alpha, n-1}$

For P2 : Reject  $H_0$  if  $W < \chi^2_{1-\alpha, n-1}$

For P3 : Reject  $H_0$  if  $W < \chi^2_{1-\frac{\alpha}{2}, n-1}$   
or  $W > \chi^2_{\alpha/2, n-1}$ .

Example: A random sample of size 25

has  $S^2 = 2$ . We want to test

$$H_0: \sigma^2 \leq 1$$

$$H_1: \sigma^2 > 1$$

$$W = \frac{(n-1)S^2}{\sigma_0^2} = \frac{24 \times 2}{1} = 48$$

$$\chi^2_{0.05, 24} = 37 \dots$$

$W > \chi^2_{0.05, 24}$ . So  $H_0$  is rejected

Testing for Parameters of Two Normal Populations :

$x_1, \dots, x_m \sim N(\mu_1, \sigma_1^2)$

indep

$y_1, \dots, y_n \sim N(\mu_2, \sigma_2^2)$

For comparing  $\mu_1 \Sigma \mu_2$

Case I:  $\sigma_1^2$  and  $\sigma_2^2$  are known

(P1)

$H_0: \mu_1 \leq \mu_2$

$H_1: \mu_1 > \mu_2$

(P2)

$H_0: \mu_1 \geq \mu_2$

$H_1: \mu_1 < \mu_2$

(P3) :  $H_0: \mu_1 = \mu_2$   
 $H_1: \mu_1 \neq \mu_2.$

$$Z = \frac{\bar{X} - \bar{Y}}{\sqrt{\frac{\sigma_1^2}{m} + \frac{\sigma_2^2}{n}}}$$

For (P1) : Reject  $H_0$  if  $Z > 3\alpha$

For (P2) : Reject  $H_0$  if  $Z < -3\alpha$

For P3 : Reject  $H_0$  if  $|z| \geq 3\alpha_2$ .

Case II :  $\sigma_1^2 = \sigma_2^2 = \sigma^2$  unknown

$$T = \sqrt{\frac{mn}{m+n}} \frac{(\bar{x} - \bar{y})}{S_p}$$

$$S_p^2 = \frac{(m-1)S_1^2 + (n-1)S_2^2}{(m+n-2)}$$

For P1 Reject  $H_0$  if  $T > t_{\alpha, m+n-2}$

For P2 Reject  $H_0$  if  $T < -t_{\alpha, m+n-2}$

For P3 Reject  $H_0$  if  $|T| \geq t_{\frac{\alpha}{2}, m+n-2}$

Case III :  $\sigma_1^2$  &  $\sigma_2^2$  are completely unknown

Welch - Smith - Satterwaite procedure

$$T^* = \frac{\bar{x} - \bar{y}}{\sqrt{\frac{s_1^2}{m} + \frac{s_2^2}{n}}}$$

$$\nu = \frac{\left( \frac{s_1^2}{m} + \frac{s_2^2}{n} \right)^2}{\left[ \frac{s_1^4}{m^2(m-1)} + \frac{s_2^4}{n^2(n-1)} \right]}$$

we take  $\nu$  rounded to nearest integer.

- (P1) Reject  $H_0$  if  $T^* > t_{\alpha/2}$
- (P2) Reject  $H_0$  if  $T^* < -t_{\alpha/2}$
- (P3) Reject  $H_0$  if  $|T^*| \geq t_{\frac{\alpha}{2}}$ .

Paired t-test: Here  $X_i$  &  $Y_i$  are not independent but they are paired.

$$(X_i, Y_i) \sim \text{BUN}(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2, \rho)$$

$$d_i = X_i - Y_i, \quad d_i \sim N(\mu_1 - \mu_2, \sigma_d^2)$$

$$\bar{d} = \frac{1}{n} \sum d_i, \quad S_d^2 = \frac{1}{n-1} \sum (d_i - \bar{d})^2$$

$$T_1 = \frac{\sqrt{n} \bar{d}}{S_d}$$

For ① Reject  $H_0$  if  $T_1 > t_{\alpha/2, n-1}$

for (P2) Reject  $H_0$  if  $T_1 < -t_{\alpha, n-1}$

for (P3) Reject  $H_0$  if  $|T_1| \geq t_{\alpha/2, n-1}$

## Testing for Variances

(P1)  $H_0: \sigma_1^2 \leq \sigma_2^2$

$$H_1: \sigma_1^2 > \sigma_2^2$$

(P2)  $H_0: \sigma_1^2 \geq \sigma_2^2$

$$H_1: \sigma_1^2 < \sigma_2^2$$

(P3)  $H_0: \sigma_1^2 = \sigma_2^2$  vs  $H_1: \sigma_1^2 \neq \sigma_2^2$

$$F = \frac{S_1^2}{S_2^2}$$

- For (P1) Reject  $H_0$  if  $F > f_{\alpha, m-1, n-1}$
- (P2) Reject  $H_0$  if  $F < f_{1-\alpha, m-1, n-1}$
- (P3) Reject  $H_0$  if  $F < f_{1-\frac{\alpha}{2}, m-1, n-1}$   
or  $F > f_{\alpha/2, m-1, n-1}$

Examples: Comparing average yields  
of a crop in two different states

State 1:

$$m = 10$$

$$\bar{x} = 825 \text{ mt}$$

$$\sigma_1^2 = 100$$

State 2:

$$n = 10$$

$$\bar{y} = 815 \cdot \text{mt}$$

$$\sigma_2^2 = 60$$

$$Z = \frac{\bar{x} - \bar{y}}{\sqrt{\frac{\sigma_1^2}{m} + \frac{\sigma_2^2}{n}}} = 2.5$$

$$z_{0.05} = 1.645$$

$$Z > z_{0.05}.$$

So for P1 :  $H_0$  is rejected

2. weights of athletes in two teams

Team 1 :  $m=8, \bar{x}=90 \text{ (kg)}$

$$\sigma_1^2 = 3.9$$

Team 2 :  $n=8, \bar{y}=95 \text{ (kg)}, \sigma_2^2 = 4.$

For testing  $H_0: \sigma_1^2 = \sigma_2^2$   
 $H_1: \sigma_1^2 \neq \sigma_2^2$

$$F = \frac{s_2^2}{s_1^2} = 1.026, \quad f_{0.05, 7, 7} = 3.78$$
$$f_{0.95, 7, 7} = 0.3146$$

So  $H_0$  cannot be rejected  
if the two variances are equal.

So for testing for equality of means

we can use pooled sample variance

procedur.

$$T^* = \sqrt{\frac{mn}{m+n}} \quad \frac{\bar{x} - \bar{y}}{s_p} = \sqrt{\frac{64}{16}} \quad \frac{(-5)}{\sqrt{3.95}}$$

$$s_p^2 = \frac{7 \times 3.9 + 7 \times 4}{14} = 3.95$$

$$\frac{\bar{x} - \bar{y}}{s_p} \approx -5.032,$$

$$t_{0.025, 14} = 2.145, \quad t_{0.05, 14} = 1.761$$

For P<sub>2</sub>  $T^* \leftarrow t_{0.05, 14}$

So H<sub>0</sub> is rejected

i.e average weight of athletes in Team 1  
are less than average weight of  
athletes in Team 2.

Example : Lives of some electronic  
items. (in hours)

Brand 1 :  $m = 25, \bar{x} = 380, s_1^2 = 100$

Brand 2 :  $n = 16, \bar{y} = 370, s_2^2 = 400$

$$H_0: \sigma_1^2 \leq \sigma_2^2 \quad F \doteq \frac{s_2^2}{s_1^2} = 4$$

$$H_1: \sigma_1^2 > \sigma_2^2$$

$$f_{0.05, 15, 24} = 2.1077$$

So  $H_0$  is rejected.

We cannot pool sample variances

here as variances are not the same.

$$T_1 = \frac{\bar{X} - \bar{Y}}{\sqrt{\frac{s_1^2}{m} + \frac{s_2^2}{n}}} \approx 1.857$$

$$\nu = \frac{\left( \frac{s_1^2}{m} + \frac{s_2^2}{n} \right)^2}{\left[ \frac{s_1^4}{m^2 n^2} + \frac{s_2^4}{n^2 m^2} \right]} = 19.86$$

$$t_{0.05, 19} = 1.729, \quad t_{0.025, 19} = 2.093$$

$\equiv$

For  $P_3$ ,  $H_0$  cannot be rejected  
 at 5% level of significance  
 So the lives of electronic items  
 in two firms are similar.

Example (Paired-t-test)

15 persons underwent a diet program for 3 months. and the weights  $x_i$  ( before ) ,  $y_i$  ( after ) are as follows.

Person	1	2	3	4	5	6	7
$x_i$	70	80	72	76	76	76	72
$y_i$	68	72	62	70	58	66	68
$d_i = x_i - y_i$	2	8	10	6	18	10	4

Person	8	9	10	11	12	13	14	15
$x_i$	78	82	64	74	92	74	68	84
$y_i$	52	64	72	74	60	74	72	74
$d_i$	26	18	-8	0	32	0	-4	10

$$\bar{d} = 8.8, \quad s_d = 10.98, \quad n=15$$

$$\frac{\sqrt{n} \bar{d}}{s_d} = 3.1, \quad t_{0.025, 14} = 2.145$$

$$S_0 \text{ at } \alpha = 0.025, \quad H_0: \mu_1 \leq \mu_2 \\ H_1: \mu_1 > \mu_2$$

$H_0$  is rejected. So the training program is effective.

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## Law of Large Numbers (Weak)

Let  $X_1, X_2, \dots$  be independent and identically distributed random variables with mean  $\mu$  and

variance  $\sigma^2$ .

Let  $\bar{Y}_n = \frac{1}{n} \sum_{i=1}^n X_i$

Then for any  $k > 0$ ,

$$P(|\bar{Y}_n - \mu| > k) \rightarrow 0$$

as  $n \rightarrow \infty$

or  $P(|\bar{Y}_n - \mu| \leq k) \rightarrow 1$  as  $n \rightarrow \infty$

Or  $y_n \xrightarrow{P} \mu$

Example: Let  $x_1, \dots, x_n, \dots \stackrel{i.i.d.}{\sim} \mathcal{P}(\lambda)$

If  $\lambda = 3$

8.  $y_n = \frac{1}{n} \sum_{i=1}^n x_i \xrightarrow{P} 3$

2. Let  $x_1, x_2, \dots \sim N(2, 4)$ .

$$E(x_i^2) = \mu^2 + \sigma^2 = 4 + 4 = 8$$

$$\frac{\sum x_i^2}{n} \xrightarrow{P} E(x_i^2) = 8$$

as  $n \rightarrow \infty$