

(ii) Similarly for the multivariate case

we have $\underline{x} \sim N_p(\underline{\mu}, \Sigma)$ then the
sample mean $\bar{\underline{x}} = \hat{\underline{\mu}}$

$$\bar{\underline{x}} = \hat{\underline{\mu}} = \frac{1}{n} \sum_{i=1}^n X_{ia} \sim N_p\left(\underline{\mu}, \frac{1}{n} \Sigma\right), \forall i=1, 2, \dots, p$$

To test.

one-sample

$$H_0: \underline{\mu} = \underline{\mu}_0$$

$$H_1: \underline{\mu} \neq \underline{\mu}_0$$

((Σ is unknown
(one sample.)))

$$\therefore \omega = \{ \underline{\mu} = \underline{\mu}_0, \Sigma_{\omega} \text{ is unknown} \} \text{ under } H_0$$

$$\therefore \hat{\omega} = \left\{ \underline{\mu} = \underline{\mu}_0, \hat{\Sigma}_{\omega} = \frac{\sum_{i=1}^n (X_{ia} - \underline{\mu}_0)(X_{ia} - \underline{\mu}_0)'}{n} \right\}$$

$$\Rightarrow L(\hat{\omega}) = \frac{1}{(2\pi)^{\frac{np}{2}}} |\hat{\Sigma}_{\omega}|^{-\frac{n}{2}} e^{-\frac{1}{2} \sum_{i=1}^n (X_{ia} - \underline{\mu}_0)' \hat{\Sigma}_{\omega}^{-1} (X_{ia} - \underline{\mu}_0)}$$

But $\hat{\Sigma}$ unbiased

$$= \sum_{i=1}^n (X_{ia} - \underline{\mu}_0)' \hat{\Sigma}^{-1} (X_{ia} - \underline{\mu}_0)$$

$$= \text{tr} \sum_{i=1}^n \hat{\Sigma}^{-1} (X_{ia} - \underline{\mu}_0)' (X_{ia} - \underline{\mu}_0)'$$

$$= \text{tr} n \left[\sum_{i=1}^n (X_{ia} - \underline{\mu}_0)' (X_{ia} - \underline{\mu}_0)' \right]^{-1} \left[\sum_{i=1}^n (X_{ia} - \underline{\mu}_0)' (X_{ia} - \underline{\mu}_0) \right]$$

$$= n \text{tr } I_{p \times p}$$

$$= \boxed{np}$$

Then

$$L(\hat{\omega}) = \frac{1}{(2\pi)^{\frac{np}{2}}} |\hat{\Sigma}_{\omega}|^{-\frac{n}{2}} e^{-\frac{np}{2}} \dots \dots \textcircled{1}$$

$$\lambda = \left[\frac{|\hat{\Sigma}_w|}{|\hat{\Sigma}_a|} \right]^{-n/2}$$

تقل
البيانات في المقام

$$\lambda = \left[\frac{|\hat{\Sigma}_a|}{|\hat{\Sigma}_w|} \right]^{n/2}$$

ترفع للمقام
($\frac{\Sigma_a}{\Sigma_w}$)

$$\lambda^{n/2} = \frac{|\hat{\Sigma}_a|}{|\hat{\Sigma}_w|}$$

$$\lambda^{n/2} = \frac{|\sum_{i=1}^n (X_i - \bar{X})(X_i - \bar{X})'|}{|\sum_{i=1}^n (X_i - \bar{X}_0)(X_i - \bar{X}_0)'|}$$

بعد اختصار \bar{X}

$$= \frac{|\sum_{i=1}^n (X_i - \bar{X})(X_i - \bar{X})'|}{|\sum_{i=1}^n (X_i - \bar{X})(X_i - \bar{X})' + n(\bar{X} - \bar{X}_0)(\bar{X} - \bar{X}_0)'|}$$

$$= \frac{|(n-1)S|}{|(n-1)S + n(\bar{X} - \bar{X}_0)(\bar{X} - \bar{X}_0)'|}$$

$$\lambda^{n/2} = \frac{1}{1 + \frac{n(\bar{X} - \bar{X}_0)(\bar{X} - \bar{X}_0)'}{(n-1)S}}$$

$$\text{since } \frac{|a|}{|a+b|} = \frac{1}{1 + \frac{b}{a}}$$

$$\lambda^{n/2} = \frac{1}{1 + \frac{n(\bar{X} - \bar{X}_0)S^{-1}(\bar{X} - \bar{X}_0)'}{(n-1)}}$$

$$\text{But } n(\bar{X} - \bar{X}_0)S^{-1}(\bar{X} - \bar{X}_0)' = T^2$$

$$\therefore \lambda^{\frac{2}{n}} = \frac{1}{1 + \frac{T^2}{n-1}}$$

where T^2 is Hotelling T^2 test.

* Know must find the critical region إيجاد المنطقة
الرجعية في
صالة
عينة واحدة
to test $H_0: \underline{\mu} = \underline{\mu}_0$ (one sample)

is $\lambda \leq \lambda_0$
 $\lambda^{\frac{2}{n}} \leq \lambda_0^{\frac{2}{n}}$

$$\frac{1}{1 + \frac{T^2}{n-1}} \leq \lambda_0^{\frac{2}{n}}$$

نقلب الطرفين $1 + \frac{T^2}{n-1} \leq \frac{1}{\lambda_0^{\frac{2}{n}}}$

$$1 + \frac{T^2}{n-1} \leq \lambda_0^{-\frac{2}{n}}$$

$$\frac{T^2}{n-1} \leq \lambda_0^{-\frac{2}{n}} - 1$$

$$T^2 \leq (n-1) (\lambda_0^{-\frac{2}{n}} - 1)$$

* $T_{col}^2 \geq T_{table}^2$ (The critical region one-sample)

where $T_{table}^2 = \frac{(n-1)P}{(n-P)} F_{P, n-P, \alpha}$

$$T_{col}^2 = n(\bar{X} - \underline{\mu}_0)' S^{-1} (\bar{X} - \underline{\mu}_0)$$

(iii) To test the null hypothesis
 That the mean of one normal population equal
 to the mean of another normal population
 with Common unknown Dispersion matrix.

2-Sample

i.e

$$X_{\alpha}^{(i)} \sim N_p(\mu^{(i)}, \Sigma), i=1,2$$

where Σ is unknown
 (2-Sample)

To test $H_0: \mu^{(1)} = \mu^{(2)}$

$$\bar{X}^{(1)} \sim N_p(\mu^{(1)}, \frac{1}{n_1} \Sigma)$$

$$\bar{X}^{(2)} \sim N_p(\mu^{(2)}, \frac{1}{n_2} \Sigma)$$

$$(\bar{X}^{(1)} - \bar{X}^{(2)}) \sim N_p[(\mu^{(1)} - \mu^{(2)}), (\frac{1}{n_1} + \frac{1}{n_2}) \Sigma]$$

$$\therefore \sqrt{\frac{n_1+n_2}{n_1 n_2}} (\bar{X}^{(1)} - \bar{X}^{(2)}) \sim N_p[\mu^{(1)} - \mu^{(2)}, \Sigma]$$

$$\therefore \sqrt{\frac{n_1+n_2}{n_1 n_2}} (\bar{X}^{(1)} - \bar{X}^{(2)}) \sim N_p[0, \Sigma]$$

where

$$S = \frac{\sum_{\alpha=1}^{n_1} (X_{\alpha}^{(1)} - \bar{X}^{(1)})(X_{\alpha}^{(1)} - \bar{X}^{(1)})' + \sum_{\alpha=1}^{n_2} (X_{\alpha}^{(2)} - \bar{X}^{(2)})(X_{\alpha}^{(2)} - \bar{X}^{(2)})'}{n_1 + n_2 - 2}$$

Then for

$$T^2 = \frac{n_1 n_2}{n_1 + n_2} (\bar{X}^{(1)} - \bar{X}^{(2)})' S^{-1} (\bar{X}^{(1)} - \bar{X}^{(2)})$$

The critical region to test

$$H_0: \underline{\mu}^{(1)} = \underline{\mu}^{(2)} \quad (\Sigma \text{ unknown}) \text{ is:}$$

$$T^2 \geq T_0^2 \quad (2\text{-Sample})$$

Since

$$T_0^2 = \frac{(n_1+n_2-2)P}{n_1+n_2-P-1} F_{P, n_1+n_2-P-1}$$

$$T_{Col}^2 = \frac{n_1 n_2}{n_1+n_2} (\underline{\bar{X}}^{(1)} - \underline{\bar{X}}^{(2)})' \underline{S}^{-1} (\underline{\bar{X}}^{(1)} - \underline{\bar{X}}^{(2)})$$

1. let $\underline{X} \sim N_2(\underline{\mu}, \Sigma)$, The Sample mean vector
 $\bar{\underline{X}} = \begin{pmatrix} 55.24 \\ 34.97 \end{pmatrix}$, The Sample Covariance matrix is
 $\underline{S} = \begin{pmatrix} 210.54 & 126.97 \\ & 119.68 \end{pmatrix}$, The Sample size

$N=101$, Test the hypothesis at $\alpha=0.01$

$$H_0: \underline{\mu} = \underline{\mu}_0 = \begin{bmatrix} 60 \\ 50 \end{bmatrix}$$

Sol we have Σ know, $N=101$ (one sample)

By using Hotelling T^2 :

$$T^2 = n (\bar{\underline{X}} - \underline{\mu}_0)' \underline{S}^{-1} (\bar{\underline{X}} - \underline{\mu}_0)$$

$$= (101) \begin{pmatrix} 55.24 & 34.97 \end{pmatrix} \begin{pmatrix} 210.54 & 126.97 \\ & 119.68 \end{pmatrix}^{-1} \begin{pmatrix} 55.24 \\ 34.97 \end{pmatrix}$$

$$T_{\text{Calc}}^2 = 357.43$$

$$F(2, 99, 0.01) = 4.83 \text{ From } F \text{ table}$$

$$\therefore T_0^2 = \frac{(100)(2)}{99} (4.83) = 9.76$$

$$T_{\text{Calc}}^2 = 357.43 > T_0^2 = 9.76$$

\therefore we reject H_0

i.e. $\underline{\mu} \neq \underline{\mu}_0$

example 2: test if normal distribution? test if normal distribution?
 where $n_1 = 10$, $n_2 = 10$, the sample means

$$\bar{x} = \begin{pmatrix} 2.01 \\ 2.21 \\ 2.31 \\ 2.41 \end{pmatrix}, \bar{y} = \begin{pmatrix} 2.01 \\ 2.11 \\ 2.21 \\ 2.31 \end{pmatrix}$$

and the v.c matrix

$$S = \begin{pmatrix} 0.2502 & 0.0011 & 0.0001 & 0.001 \\ 0.0011 & 0.0010 & 0.001 & 0.001 \\ 0.0001 & 0.001 & 0.001 & 0.001 \\ 0.001 & 0.001 & 0.001 & 0.001 \end{pmatrix}$$

Test the hypothesis: $H_0: \mu_1 = \mu_2$

vs $H_1: \mu_1 \neq \mu_2$

2-sample

Sol (2-sample) ($n_1 = n_2 = 10$) then:

$$T^2 = \frac{n_1 n_2}{n_1 + n_2} \left(\bar{x} - \bar{y} \right)' S^{-1} \left(\bar{x} - \bar{y} \right)$$

$$= \frac{10 \cdot 10}{10 + 10} \left[\begin{matrix} 2.01 & 2.21 & 2.31 & 2.41 \end{matrix} \right] \begin{matrix} \downarrow \\ \downarrow \\ \downarrow \\ \downarrow \end{matrix}$$

$$\begin{pmatrix} 0.0011 & 0.0011 & 0.0011 & 0.0011 \\ 0.0011 & 0.0010 & 0.0010 & 0.0010 \\ 0.0011 & 0.0010 & 0.0010 & 0.0010 \\ 0.0011 & 0.0010 & 0.0010 & 0.0010 \end{pmatrix}$$

$$T_{10}^2 = 22.05$$

$$T_{table} = \frac{n_1 n_2 + 2}{n_1 + n_2 + 2} \sqrt{\frac{(n_1 + n_2 - 2) F_{\alpha/2, n_1 + n_2 - 2, 1}}{n_1 + n_2}}$$

$$= \frac{10 \cdot 10 + 2}{10 + 10 + 2} \sqrt{\frac{(10 + 10 - 2) F_{0.05, 10 + 10 - 2, 1}}{10 + 10}} = \frac{22}{22} \sqrt{5.16} = 5.51$$

$$\therefore T_{10}^2 = 22.05 > T_{table} = 5.51$$

\therefore we reject H_0

Then $\mu_1 \neq \mu_2$

2/2/2017

Note

see page

5' x 12' x 12' (12' x 12' x 12')

5' x 12' x 12'

see page 12/12/12

5' x 12' x 12'

Q11/11/2017

Q11/11/2017

Assume $H_0: \mu = 0$ vs $H_1: \mu > 0$
 Test $Z = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$
 To test $H_0: \mu = 0$ vs $H_1: \mu > 0$
 $Z = \frac{\bar{X} - 0}{\sigma/\sqrt{n}}$

(1) If $\mu = 0$ (i.e. H_0 is true)

$Z \sim N(0, 1)$

$Z \sim N(0, 1)$

Let $\alpha = P(Z > z_\alpha)$

and $\beta = P(Z < z_\alpha)$

$Z = Z_0 + Z_1$

Therefore will be

$H_0: \mu = 0$

vs $H_1: \mu > 0$

When Z is known.

$Z \sim N(0, 1)$

" we reject H_0 if $Z > z_\alpha$

Let $\alpha = P(Z > z_\alpha)$ and $\beta = P(Z < z_\alpha)$

(ii) if $n_1 = n_2 = n$ (Σ_1 and Σ_2 unknown)

$$\bar{X}^{(1)} \sim N_p(\mu^{(1)}, \frac{1}{n_1} \Sigma_1), \quad \bar{X}^{(2)} \sim N_p(\mu^{(2)}, \frac{1}{n_2} \Sigma_2)$$

$$\text{Let } Y_a = X_a^{(1)} - X_a^{(2)} \sim (v, \Sigma)$$

$$\text{where } v = \bar{X}^{(1)} - \bar{X}^{(2)}$$

$$\Sigma = \Sigma_1 + \Sigma_2$$

To test $H_0: v = 0$ (Σ is unknown)

v.s. $H_1: v \neq 0$

$$\therefore T^2 = n \bar{Y}' S^{-1} \bar{Y} \sim T_{(n-1)}^2$$

$$n (\bar{X}^{(1)} - \bar{X}^{(2)})' S^{-1} (\bar{X}^{(1)} - \bar{X}^{(2)}) \sim T_{(n-1)}^2$$

$$\text{Since } S = \frac{\sum (Y_a - \bar{Y}_a)(Y_a - \bar{Y}_a)'}{n-1}$$

and

$$S = \frac{\sum [(X_a^{(1)} - X_a^{(2)}) - (\bar{X}^{(1)} - \bar{X}^{(2)})] [(X_a^{(1)} - X_a^{(2)}) - (\bar{X}^{(1)} - \bar{X}^{(2)})]'}{n-1}$$

notes

(1) if we found that $\bar{X}_1 = \bar{X}_2$ then we will have the case (ii) with T' not with $d.f = 2n - 2$ (in case (ii))

(2) if $n_1 < n_2$ then $n_1 T' \leq T$ for T' For testing $H_0: \mu_1^{(1)} - \mu_1^{(2)} = 0$

where

$$S = \frac{\sum_{i=1}^{n_2} \left[X_{2i}^{(1)} - \bar{X}_2^{(1)} - \sqrt{\frac{n_2}{n_1}} \left(X_{2i}^{(2)} - \frac{1}{n_2} \sum_{j=1}^{n_2} X_{2j}^{(2)} \right) \right]^2}{(n_2 - 1)}$$

where $\bar{X} = \bar{X}^{(1)} - \bar{X}^{(2)}$

3. Comparing Two Mean vectors

Assume that we have a random sample consisting of $X_i \sim N_p(\mu_1, \Sigma)$ of size n_1 and $Y_i \sim N_p(\mu_2, \Sigma)$ of size n_2 . We assume that the two samples are independent and that $\Sigma_1 = \Sigma_2 = \Sigma$, say with Σ unknown.

- The test of the equality of means μ_1 and μ_2 can be formally written as follows:

$$H_0: \mu_1 = \mu_2 \text{ vs } H_1: \mu_1 \neq \mu_2$$

or

$$H_0: \mu_1 - \mu_2 = 0 \text{ vs } H_1: \mu_1 - \mu_2 \neq 0$$

Both samples provide the statistics

- $\bar{X}_1, \bar{X}_2, S_1$ and S_2 .

Let $\delta = \mu_1 - \mu_2$ and $n = n_1 + n_2$.

We have,

$$(\bar{X}_1 - \bar{X}_2) \sim N_p\left(\delta, \frac{n}{n_1 n_2} \Sigma\right)$$

and $(n_1 - 1)S_1 + (n_2 - 1)S_2 \sim W_p(n_1 + n_2 - 2, \Sigma)$

Let $S_p = \frac{1}{n_1+n_2-2} [(n_1-1)S_1 + (n_2-1)S_2]$ is the pool sample covariance matrix.

This lead to a test statistic with Hotelling - T^2 distribution.

$$\frac{n_1 n_2}{n_1+n_2} \left[\{(\bar{X}_1 - \bar{X}_2) - \delta\}' S_p^{-1} \{(\bar{X}_1 - \bar{X}_2) - \delta\} \right] \sim T_{(p, n_1+n_2-2)}^2$$

This result can be used to test the null hypothesis

$$H_0: \delta = \underline{0} \quad \text{or} \quad H_0: \underline{M}_1 - \underline{M}_2 = \underline{0}$$

That is:

$$T^2 = \frac{n_1 n_2}{n_1+n_2} (\bar{X}_1 - \bar{X}_2)' S_p^{-1} (\bar{X}_1 - \bar{X}_2)$$

which is known Hotelling - T^2 statistic.

Then we reject $H_0: \underline{M}_1 = \underline{M}_2$ if:

$$T^2 \geq T_{(\alpha, p, n_1+n_2-2)}^2$$

If H_0 is true, then

$$\frac{(n_1+n_2-p-1)}{p(n_1+n_2-2)} T^2 \sim F_{(p, n_1+n_2-p-1)}$$

The rejection region of the test is:

$$\frac{(n_1+n_2-p-1)}{p(n_1+n_2-2)} T^2 \geq F_{(k,p,n_1+n_2-2)}$$

EX: Four psychological test were given
(32) men and (32) woman.
If you have the following data:

$$\bar{X}_1 = \begin{pmatrix} 18.97 \\ 19.91 \\ 27.19 \\ 22.75 \end{pmatrix}, \bar{X}_2 = \begin{pmatrix} 12.34 \\ 13.91 \\ 16.66 \\ 21.94 \end{pmatrix}$$

$$S_1 = \begin{pmatrix} 5.192 & 4.549 & 6.622 & 5.250 \\ & 13.18 & 6.760 & 6.266 \\ & & 28.67 & 14.47 \\ & & & 16.69 \end{pmatrix}$$

$$S_2 = \begin{pmatrix} 9.136 & 7.549 & 4.864 & 4.151 \\ & 18.60 & 10.22 & 5.446 \\ & & 30.04 & 13.49 \\ & & & 28.00 \end{pmatrix}$$

Test the hypothesis $H_0: \underline{M}_1 = \underline{M}_2$ vs.

$H_1: \underline{M}_1 \neq \underline{M}_2$ at $\alpha = 0.05$

Sol

$$S_p = \frac{1}{n_1 + n_2 - 2} [(n_1 - 1)S_1 + (n_2 - 1)S_2]$$

$$= \frac{1}{32+32-2} [(32-1) \begin{pmatrix} 5.192 & 4.545 & 6.522 & 5.260 \\ & 13.18 & 6.760 & 6.268 \\ & & 28.67 & 14.47 \\ & & & 8.64 \end{pmatrix}$$

$$+ (32-1) \begin{pmatrix} 9.136 & 7.549 & 4.864 & 4.151 \\ & 18.60 & 10.22 & 5.446 \\ & & 30.04 & 13.49 \\ & & & 28.00 \end{pmatrix}]$$

$$\therefore S_p = \begin{pmatrix} 7.164 & 6.047 & 5.693 & 4.701 \\ & 15.89 & 8.492 & 5.856 \\ & & 29.36 & 13.98 \\ & & & 22.32 \end{pmatrix}$$

$$\therefore T = \frac{n_1 n_2}{n_1 + n_2} (\bar{\underline{X}}_1 - \bar{\underline{X}}_2)' S_p^{-1} (\bar{\underline{X}}_1 - \bar{\underline{X}}_2)$$

$$= \frac{(32)(32)}{32+32} \begin{pmatrix} 19.97 - 12.34 \\ 19.91 - 13.91 \\ 27.19 - 16.66 \\ 22.75 - 21.94 \end{pmatrix}$$

$$\begin{pmatrix} 7.164 & 6.047 & 5.693 & 4.701 \\ & 19.89 & 8.492 & 5.856 \\ & & 29.36 & 13.98 \\ & & & 22.32 \end{pmatrix}^{-1} \begin{pmatrix} 19.97 - 12.34 \\ 19.91 - 13.91 \\ 27.19 - 16.66 \\ 22.75 - 21.94 \end{pmatrix}$$

$$\therefore T^2 = 97.601$$

$$F_{cal} = \frac{(n_1+n_2-p-1)}{p(n_1+n_2-2)} T^2 = \frac{(32+32-4-1)}{4(32+32-2)} (97.601) = 23.2$$

$$F_{(1,p,n_1+n_2-p-1)} = F_{(0.05, 4, 59)} = 2.335$$

$$\text{Since } F_{cal} = 23.219 > F_{tab} = 2.335$$

Therefore, we reject $H_0: \underline{\mu}_1 = \underline{\mu}_2$ and
accept $H_1: \underline{\mu}_1 \neq \underline{\mu}_2$

???

(87)

properties of T^2

Case A

a.r.s. X_1, \dots, X_n is drawn from $N_p(\mu, \Sigma)$,
(Σ is unknown), to test

$$H_0: \mu = \mu_0$$

$$\text{v.s. } H_1: \mu \neq \mu_0$$

The tests based on the following facts:

$$(1) T^2 = n(\bar{X} - \mu_0)' S^{-1} (\bar{X} - \mu_0) \sim T^2$$

where

$$S = \sum_{i=1}^n \frac{(X_i - \bar{X})(X_i - \bar{X})'}{n-1}$$

where S is unbiased estimate of Σ .

(2) We can compute T^2 as:

$$T^2 = \frac{|S + n(\bar{X} - \mu_0)(\bar{X} - \mu_0)'|}{|S|} - 1$$

$$(3) \text{ The dist. of } U = \frac{n-p}{(n-1)(n-p)} T^2 \sim F'_{(p-1, n-p, \lambda)}$$

where F' is the non central F with non centrality
 $n(\mu - \mu_0)' \Sigma^{-1} (\mu - \mu_0)$ if H_0 is true we have
central F

(4) The GLR test for $H_0: \mu = \mu_0$
v.s. $H_1: \mu \neq \mu_0$

is $T^2 \geq T_0^c$ where T_0^c in term of F .

Case B let $X_{a_i}^{(i)}$, $a=1, 2, \dots, n_i$ is a set from $N(\mu^{(i)}, \Sigma)$ for

to test $H_0: \mu = \mu_0$

v.s. $H_1: \mu \neq \mu_0$

The test based on the following facts.

$$(1) \bar{X}_i \sim N_p(\mu^{(i)}, \frac{1}{n_i} \Sigma), i=1, 2$$

$$\therefore \sqrt{\frac{n_1 n_2}{n_1 + n_2}} (\bar{X}^{(i)} - \mu^{(i)}) \sim N_p(0, \Sigma)$$

$$\text{where } S = \frac{\sum_{a=1}^n (X_a^{(i)} - \bar{X})(X_a^{(i)} - \bar{X})' + \sum_{a=1}^n (X_a^{(j)} - \bar{X})(X_a^{(j)} - \bar{X})'}{n_1 + n_2 - 2}$$

Therefore, when H_0 is true then

$$\frac{n_1 n_2}{n_1 + n_2} (\bar{X}^{(i)} - \bar{X}^{(j)})' S^{-1} (\bar{X}^{(i)} - \bar{X}^{(j)}) \sim T_{n_1 + n_2 - 2}^2$$

$$(2) \text{ The dist. of } T^2 = \frac{n_1 n_2 - p + 1}{(p-1)(n_1 + n_2 - 2)} T^2 \sim F_{p-1, n_1 + n_2 - 2}$$

(3) The GLR test for $H_0: \mu^{(i)} = \mu^{(j)}$

$$\text{is } T^2 \geq T_0^2$$

where T_0^2 in term of F .

Case C

بِسْمِ اللَّهِ الرَّحْمَنِ الرَّحِيمِ

بِسْمِ اللَّهِ الرَّحْمَنِ الرَّحِيمِ

بِسْمِ اللَّهِ الرَّحْمَنِ الرَّحِيمِ
والصلاة والسلام على محمد وآله

Case C: let $X_a^{(i)}$, $a=1,2,\dots,n$ is a r.s. from $N_p(\underline{\mu}^{(i)}, \Sigma_i)$, $i=1,2$
to test:

$H_0: \underline{\mu}^{(1)} = \underline{\mu}^{(2)}$, v.s. $H_1: \underline{\mu}^{(1)} \neq \underline{\mu}^{(2)}$

The test based on the following facts:

① $\bar{x}^{(i)} \sim N_p(\underline{\mu}^{(i)}, \frac{1}{n} \Sigma_i)$, $i=1,2$

$X^{(1)} \sim N_p(\underline{\mu}^{(1)}, \Sigma_1)$, $X^{(2)} \sim N_p(\underline{\mu}^{(2)}, \Sigma_2)$

② To test $H_0: \underline{\mu}^{(1)} = \underline{\mu}^{(2)}$ is equivalent to ~~$\underline{\mu}^{(1)} - \underline{\mu}^{(2)} = 0$~~

$H_0: \underline{\mu}^{(1)} - \underline{\mu}^{(2)} = 0$

let $Y_a = X_a^{(1)} - X_a^{(2)}$, $\underline{\mu}_Y = \underline{\mu}^{(1)} - \underline{\mu}^{(2)}$

$\Sigma_Y = \Sigma_1 - \Sigma_2$

$\therefore \bar{Y} \sim N_p(0, \frac{1}{n} \Sigma_Y)$ if H_0 is true.

③ To test $H_0: \underline{\mu}^{(1)} - \underline{\mu}^{(2)} = 0$
v.s. $H_1: \underline{\mu}^{(1)} - \underline{\mu}^{(2)} \neq 0$

$\therefore n \bar{Y}' S^{-1} \bar{Y} \sim T_{(n-1)}^2$

where $S = \frac{\sum (Y_a - \bar{Y})(Y_a - \bar{Y})'}{n-1}$

Simultaneous inferences for means

Case (1) (one-sample)

Let a single sample of size N has been drawn from $MVN(\underline{\mu}, \Sigma)$, ^{then} ~~then~~ for any vector

$\underline{a}' = [a_1, a_2, \dots, a_p]$ to test the hypothesis

$$H_0: \underline{a}'\underline{\mu} = \underline{a}'\underline{\mu}_0$$

$$t_{(a)} = \frac{\underline{a}'(\bar{X} - \underline{\mu})\sqrt{N}}{\sqrt{\underline{a}'S\underline{a}}} \sim T_{\alpha, p, N}^2$$

$$\Rightarrow t_{(a)}^2 = \frac{N(\underline{a}'(\bar{X} - \underline{\mu}))^2}{\underline{a}'S\underline{a}} \leq N(\bar{X} - \underline{\mu})S^{-1}(\bar{X} - \underline{\mu})$$

and from the dist. of T^2 .

$$P[t_{(a)}^2 \leq T_{\alpha, p, N-p}^2] = 1 - \alpha$$

$$P\left(-T_{\alpha, p, N-p} < \frac{\underline{a}'\bar{X} - \underline{a}'\underline{\mu}}{\sqrt{\frac{\underline{a}'S\underline{a}}{N}}} < T_{\alpha, p, N-p}\right) = 1 - \alpha$$

$$P\left\{ \underline{d}\bar{X} - \sqrt{\frac{d^2 S^2}{N}} T_{\alpha, p, N-p} \leq \underline{d}\mu < \underline{d}\bar{X} + \sqrt{\frac{d^2 S^2}{N}} T_{1-\alpha, p, N-p} \right\}$$

Since $T_{\alpha, p, N-p} = \sqrt{\frac{(n-1)F}{(n-p)}} T_{\alpha, p, N-p}$

This is the family of the simultaneous confidence intervals of Roy & Bose (1953) with coefficient $(1-\alpha)$.
 So for a given choice of α we may construct intervals for any and all linear compounds of the means with any hypothesis:

$$H_0: \underline{d}\mu = \underline{d}\mu_0$$

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ex.) A sample of size $n=101$ with $\bar{x} = \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = \begin{pmatrix} 55.24 \\ 24.77 \end{pmatrix}$

$$\text{and } S = \begin{pmatrix} 210.54 & 126.77 \\ & 119.68 \end{pmatrix}$$

Test the hypothesis.

① $H_0: \mu_1 = 60$ with $\alpha = 0.01$

sol

$$H_0: \underline{a}'\underline{\mu} = \mu_1 = 60$$

$$\underline{a}'\underline{\mu} = (1 \ 0) \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} = (1 \ 0) \begin{pmatrix} 60 \\ 50 \end{pmatrix} = 60$$

we find

$$\sqrt{\frac{\underline{a}'S\underline{a}}{n}} = \left[\frac{1}{101} (1 \ 0) \begin{pmatrix} 210.54 & 126.77 \\ & 119.68 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right]^{1/2} = 1.4439$$

$$T_{0.01, 2, 99} = \sqrt{\frac{(101-1)(2)}{(101-2)}} F_{0.01, 2, 99} = \sqrt{\frac{200}{99}} (1.779) = 3.11$$

Then the C.I.

$$Pr(55.24 - 1.4439(3.11) \leq \mu_1 \leq 55.24 + 1.4439(3.11)) = 0.99$$

$$Pr(50.73 \leq \mu_1 \leq 59.75) = 0.99$$

\therefore we reject H_0

$$H_0: \mu_1 = 60$$

H.w:

Test

① $H_0: \mu_2 = 50$

② $H_0: \mu_1 + \mu_2 = 110$

③ $H_0: \mu_1 - \mu_2 = 10$

from the example above
