

# Comparing mean vectors from 2 MVN sample

## Paired test

requires each test  
object split in 2 parts



example: before and  
after

pairwise observation

$$(x_{1j}, x_{2j}), j=1, \dots, n$$

$(px_j) \quad (px_j)$

$$x_{1j} \sim Np(\mu_1, \Sigma_1) \quad \text{Model}$$
$$x_{2j} \sim Np(\mu_2, \Sigma_2) \quad \text{Check}$$

# Difference model

$D_j \stackrel{\text{def}}{=} X_{1j} - X_{2j}$  ← the difference between the 2 vectors  
 $\sim N_p(\mu_1 - \mu_2, \Sigma_d)$

test if the difference is a value

$H_0: \delta \stackrel{\text{def}}{=} \mu_1 - \mu_2 = \delta_0$  ?  
(px1) often set to  $\delta_0 = 0$

$H_1: \delta \neq \delta_0$  ← not plausible

Estimates  $\hat{\mu}_1 - \hat{\mu}_2 = \bar{D} = \frac{1}{n} \sum_{j=1}^n D_j \sim$   
 $\sim N_p(\mu_1 - \mu_2, \frac{\Sigma_d}{n})$

$$\hat{\Sigma}_d = S_d = \frac{1}{n-1} \sum_{j=1}^n (D_j - \bar{D})(D_j - \bar{D})^T$$

## Hotelling $T^2$ test Stat

$$T^2 = \bar{D} - (\mu_1 - \mu_2) \left( \frac{S_d}{n} \right)^{-1} (\bar{D} - (\mu_1 - \mu_2))$$

true mean

$$\sim \frac{P(n-1)}{n-p} F_{p, n-p}$$

### Test

$$T_0^2 = (\bar{D} - \delta_0)^T \left( \frac{S_d}{n} \right)^{-1} (\bar{D} - \delta_0) \sim \frac{P(n-1)}{n-p} F_{p, n-p}$$

$$T_0^2 \text{ data} = t_0^2 \quad \text{choose } \alpha: \text{reject } H_0 \text{ IFF}$$

$$t_0^2 > \frac{P(n-1)}{n-p} F_{p, n-p, \alpha}$$

$$p\text{-value} = P\left(\frac{P(n-1)}{n-p} F_{p, n-p} > t_0^2\right)$$

$f_{cdf}$

$100(1-\alpha)\%$  CR for  $\mu_1 - \mu_2$

$$\forall (\mu_1 - \mu_2) \in \mathbb{R}: T^2 \leq \frac{\chi_{n-p}^{2-\alpha}}{n-p} F_{p, n-p, \alpha}$$

Hyperellipsoid in  $\mathbb{R}^p$

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Confidence intervals

Bonferroni CI for each

variable

Student-t dist

$$[\bar{D}_i - t_{n-1, \frac{\alpha}{2p}} \cdot \sqrt{\frac{s_{D,ii}}{n}}, \bar{D}_i + t_{n-1, \frac{\alpha}{2p}} \cdot \sqrt{\frac{s_{D,ii}}{n}}]$$

variable

$i=1, p$

Non-paired  
always possible

Model  $X_{1,j} \sim N_p(\mu_1, \Sigma_1), j=1, \dots, n_1$   
 $X_{2,j} \sim N_p(\mu_2, \Sigma_2), j=1, \dots, n_2$

model check

$$H_0: f^{\text{obs}} = \mu_1 - \mu_2 \stackrel{?}{=} f_0 \quad \text{often } S_0 = 0$$

$$H_1: f \neq f_0$$

Estimates  $\hat{\mu}_1 = \bar{x}_1 \sim N_p\left(\mu_1, \frac{\Sigma_1}{n_1}\right)$   
 $\hat{\mu}_2 = -11-$   
 $(P_{X_1})$

Case 1

Homoscedacity

$$\underline{\Sigma_1 = \Sigma_2 = \Sigma} \text{ (test with Bartlett test)}$$

$$S_1 = \frac{1}{n_1 - 1} \sum_{j=1}^{n_1} (X_{1j} - \bar{X}_1)(X_{2j} - \bar{X}_2)^T$$

$$S_2 = \frac{1}{n_2 - 1} \sum_{j=1}^{n_2} (X_{2j} - \bar{X}_2)(X_{2j} - \bar{X}_2)^T$$

if  $n_1 = n_2 \rightarrow$  use average

else:

pooled estimate  $\rightarrow$  weighted avg

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

↑

doable only because we assume

$\Sigma_1 = \Sigma_2 \rightarrow$  tested with Bartlett

Hotelling  $T^2$  TS

$$T^2 = (\bar{\mathbf{x}}_1 - \bar{\mathbf{x}}_2) - (\mu_1 - \mu_2) \left[ S_p \left( \frac{1}{n_1} + \frac{1}{n_2} \right) \right]^{-1} \rightarrow$$

$$\rightarrow (\bar{\mathbf{x}}_1 - \bar{\mathbf{x}}_2) - (\mu_1 - \mu_2) \sim \frac{P(n_1 + n_2 - 2)}{n_1 + n_2 - p - 1} F_{p, n_1 + n_2 - p - 1}$$

↑  
test Statistics      ↓  
distribution

test  $T^2 \stackrel{\mu_1 - \mu_2 = \delta_0}{=} T_0^2 = t^2$

insert data  $\delta_0$   
set number  $t^2$

Reject  $H_0 \Rightarrow t^2 > \frac{P(n_1 + n_2 - 2)}{n_1 + n_2 - p - 1} F_{p, n_1 + n_2 - p - 1, \alpha}$



$f_{\text{inv}} \rightarrow F_\alpha$

CR for  $\mu_1 - \mu_2$

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

Bonferroni CI's

$$\left( \bar{X}_{1j} - \bar{X}_{2j} \right) \pm t_{n-1, \alpha/2P} \cdot \sqrt{S_{Pii} \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}$$

$\uparrow$   
covar matrix

## Non-paired case 2

$$\sum_1 \neq \sum_2$$

Large Sample  
approximation

$$\sum_1 = S_1, \sum_2 = S_2 \quad n_1, n_2 \rightarrow \infty$$

$$T^2 = ((\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)) \left( \frac{S_1}{n_1} + \frac{S_2}{n_2} \right)^{-1} ((\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2))$$

$T^2$  test  
approx  $\chi^2_p \rightarrow CR$   
 $\downarrow CI's$

Case 1 or Case 2?

Bartlett test (Box's M-test)

test for homoscedacity

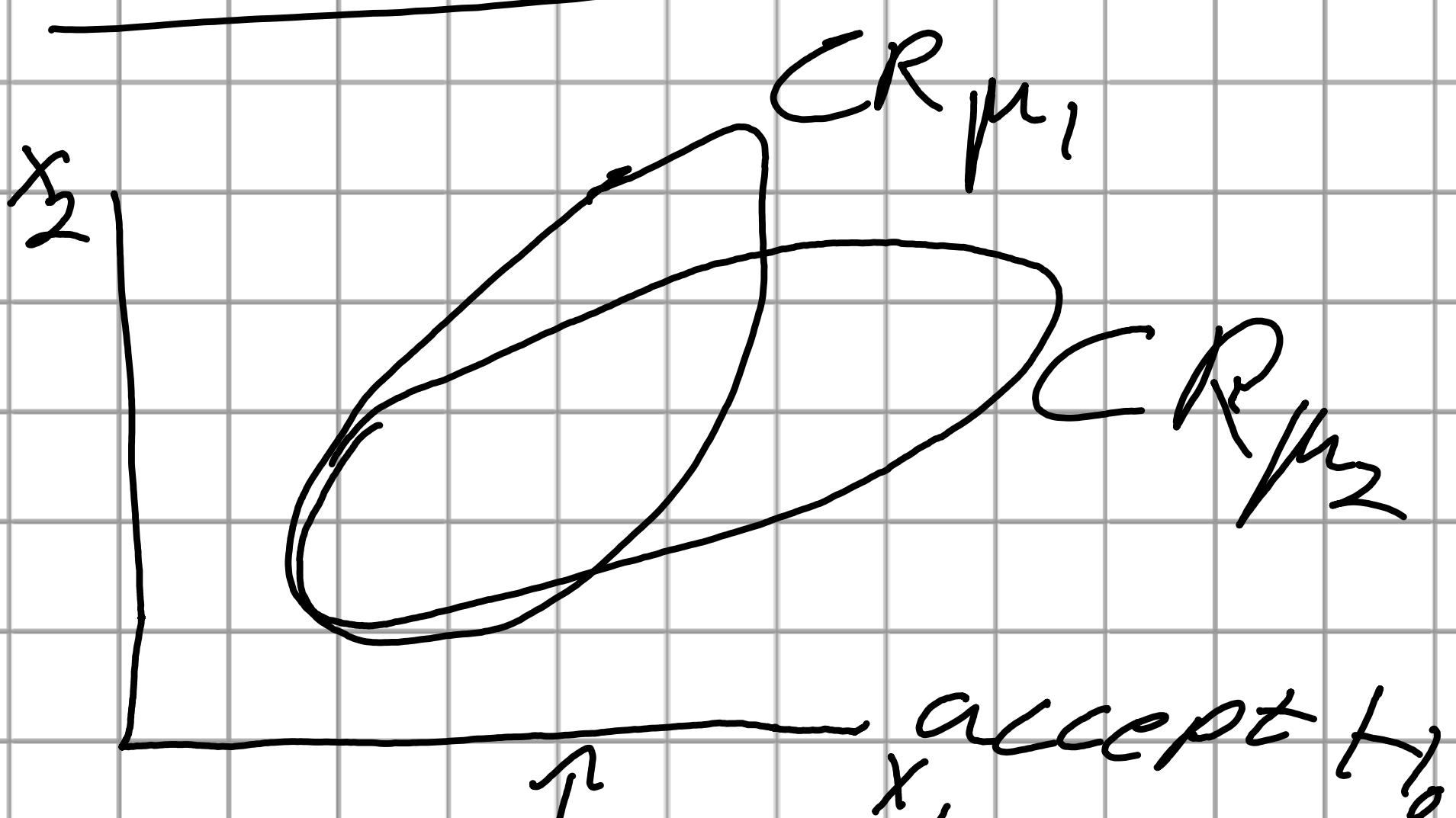
Model  $X_{ij} \sim N(\mu_i, \Sigma_i)$

$$X_{2j} \sim N(\mu_2, \Sigma_2)$$

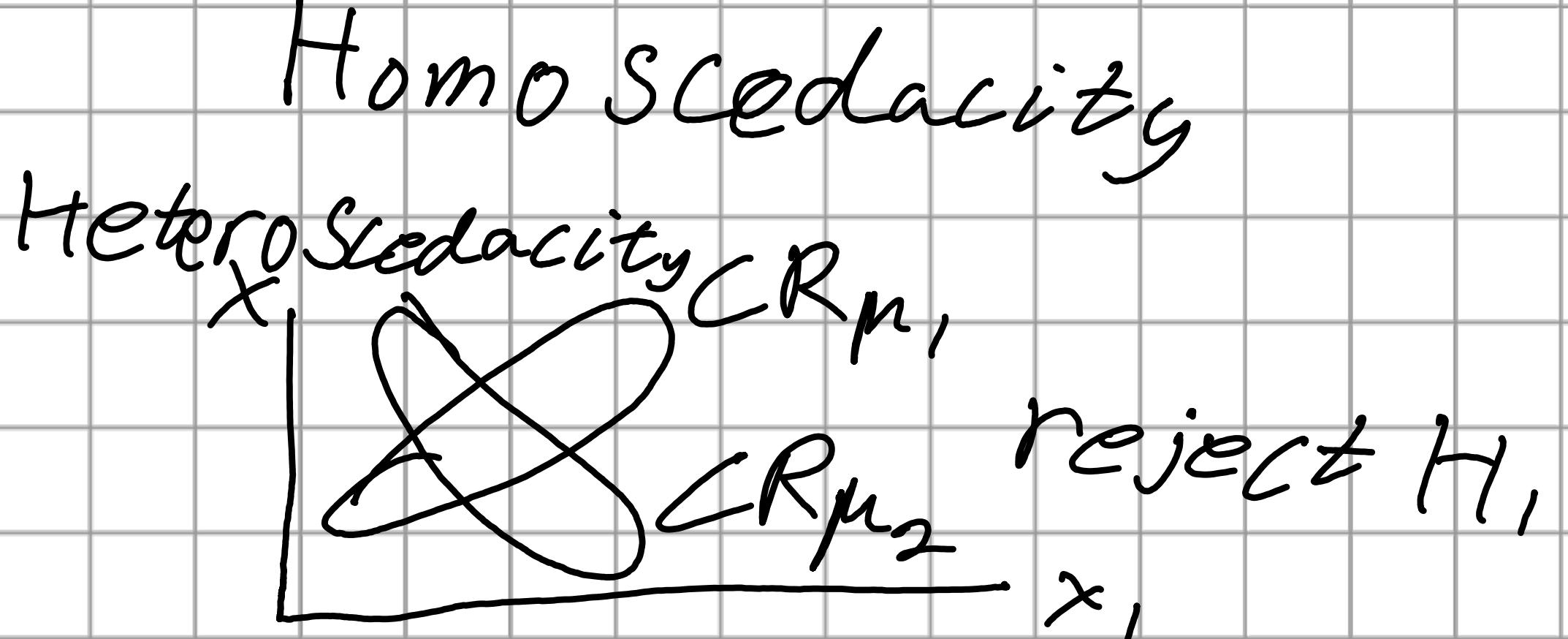
$$H_0: \Sigma_1 = \Sigma_2$$

$\rho \times \rho$        $\rho \times \rho$

2D visual



$$H_1: \Sigma_1 \neq \Sigma_2$$



Full model, 2 different  $\Sigma$ 's:  
Likelihood function  
 $\Sigma_1 \neq \Sigma_2$

Constrained model

Only 1 common  $\Sigma$ :  $\Sigma_1 = \Sigma_2$

Likelihood function  $L_{H_0}(\Sigma)$

Likelihood ratio  $\lambda \stackrel{\text{def}}{=} \frac{\max_{\Sigma} L_{H_0}(\Sigma)}{\max_{\Sigma_1, \Sigma_2} L(\Sigma_1, \Sigma_2)}$

$$0 \leq \lambda \leq 1$$

$$-2\log(\lambda) = (n_1 + n_2 - 2) \log |S_p| - (n_1 - 1) \log |S_1|$$

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