

Multivariate Inference - II

Two multivariate sample tests

STAT 32950-24620

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Groups of random vectors

$$X = \begin{bmatrix} X_1 \\ \dots \\ X_2 \\ \dots \\ \vdots \\ \dots \\ X_g \end{bmatrix},$$

Each random vector

$$X_k = [X_{k1} \cdots X_{ki} \cdots X_{kp}]' = (X_{k1}, \cdots, X_{ki}, \cdots, X_{kp})$$

may have several observations

$$\begin{aligned} &(X_{k,11}, \cdots, X_{k,1i}, \cdots, X_{k,1p}) \\ &\vdots \\ &(X_{k,n_k1}, \cdots, X_{k,n_ki}, \cdots, X_{k,n_kp}) \end{aligned}$$

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Observed data

$$X = \begin{bmatrix} X_1 \\ \dots \\ X_2 \\ \dots \\ \vdots \\ \dots \\ X_g \end{bmatrix} = \begin{bmatrix} x_{1,11} & x_{1,12} & \cdots & x_{1,1p} \\ \vdots & \vdots & \vdots & \vdots \\ x_{1,n_11} & x_{1,n_12} & \cdots & x_{1,n_1p} \\ x_{2,11} & x_{2,12} & \cdots & x_{2,1p} \\ \vdots & \vdots & \vdots & \vdots \\ x_{2,n_21} & x_{2,n_22} & \cdots & x_{2,n_2p} \\ \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots \\ x_{g,11} & x_{g,12} & \cdots & x_{g,1p} \\ \vdots & \vdots & \vdots & \vdots \\ x_{g,n_g1} & x_{g,n_g2} & \cdots & x_{g,n_gp} \end{bmatrix}$$

(notice different usages of notations)

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Example

Data: Turtle shell measurements (n = 48 obs., each of 4 variables)

```
load(file="turtles.rda")
attach(turtles) # from package Flurry
#str(turtles)   #original data n=48, var=4: Sex,L,W,H
#summary(turtles)
```

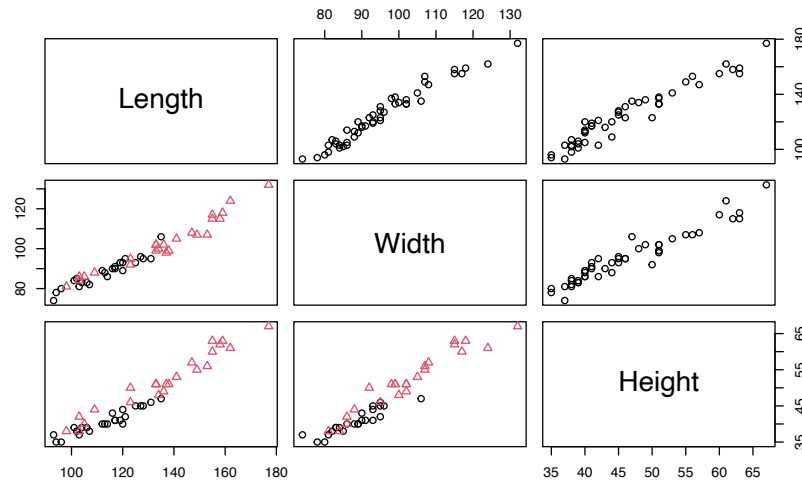
Plot

```
pairs(turtles[,-1],
      lower.panel = function(x, y){
        points(x, y, pch = unclass(turtles[,1]),
              col = as.numeric(turtles[,1]))},
      main =
        'Data "turtles": male (circle), female (triangle)')
```

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Data pairwise plots

Data "turtles": male (circle), female (triangle)



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Subset observations into two samples

Subset observations by Gender ($g = 2$ "treatment" groups)

```
male=subset(turtles[,2:4],Gender=="Male")
```

```
female=subset(turtles[,2:4],Gender=="Female")
```

Random vectors

$$X = \begin{bmatrix} X_{male} \\ X_{female} \end{bmatrix}$$

Observation data by groups

$$X = \begin{bmatrix} X_{male} \\ X_{female} \end{bmatrix} = \begin{bmatrix} x_{m,11} & x_{m,12} & x_{m,13} \\ \vdots & \vdots & \vdots \\ x_{m,n_11} & x_{m,n_12} & x_{m,n_13} \\ x_{f,11} & x_{f,12} & x_{f,13} \\ \vdots & \vdots & \vdots \\ x_{f,n_21} & x_{f,n_22} & x_{f,n_23} \end{bmatrix}$$

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Two independent samples with common covariance

Assuming two samples from X_1 and X_2 are **independent**.

$$X_1 \sim N_p(\mu_1, \Sigma_1), \quad X_2 \sim N_p(\mu_2, \Sigma_2)$$

Consider the equal-covariance case

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

To estimate Σ , use the **pooled sample covariance matrix**

$$S_{pool} = \frac{(n_1 - 1)S_1 + (n_2 - 1)S_2}{n_1 + n_2 - 2}$$

S_{pool} is an unbiased estimator of the population covariance matrix.

$$E(S_{pool}) = \Sigma$$

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Two independent samples: mean vectors $\mu_1 = \mu_2$?

To test

$$H_o : \mu_1 = \mu_2$$

$$H_a : \mu_1 \neq \mu_2$$

Apply Hotelling's T^2 to $\bar{X}_1 - \bar{X}_2$.

Use

$$\widehat{Cov}(\bar{X}_1 - \bar{X}_2) = \left(\frac{1}{n_1} + \frac{1}{n_2} \right) S_{pool}$$

(compared with S/n for one sample)

$$E \left[\left(\frac{1}{n_1} + \frac{1}{n_2} \right) S_{pool} \right] = \left(\frac{1}{n_1} + \frac{1}{n_2} \right) \Sigma = Cov(\bar{X}_1 - \bar{X}_2)$$

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Hotelling's T^2 for two samples

Under $H_0 : \mu_1 - \mu_2 = 0$,

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

where

$$T^2 = [(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)]' \left[\left(\frac{1}{n_1} + \frac{1}{n_2} \right) S_{pool} \right]^{-1} [(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)]$$

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Apply Hotelling's T^2 on mean difference

Example (by steps)

Comparing two p -variate means ($p = 3, n_1 = 24, n_2 = 24$)

$$H_0 : \mu_{male} = \mu_{female}$$

$$H_a : \mu_{male} \neq \mu_{female}$$

```
male=subset(turtles[,2:4],Gender=="Male")
female=subset(turtles[,2:4],Gender=="Female")
mbar=colMeans(male); fbar=colMeans(female);
diffmean = mbar - fbar
n1=24; n2=24; p=3
Spool=
(n1-1)/(n1+n2-2)*cov(male)+(n2-1)/(n1+n2-2)*cov(female)
T2=t(diffmean)%*%solve((1/n1+1/n2)*Spool)%*%diffmean
```

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Assumption $\Sigma_1 = \Sigma_2$

$g = 2, p = 3, n_1 = 24, n_2 = 24$.

```
cov(male)
```

```
##           Length Width Height
## Length 138.77 79.15  37.38
## Width   79.15 50.04  21.65
## Height  37.38 21.65  11.26
```

```
cov(female)
```

```
##           Length Width Height
## Length 451.4 271.2 168.70
## Width  271.2 171.7 103.29
## Height 168.7 103.3  66.65
```

Assuming $\Sigma_1 = \Sigma_2$, for now.

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Check descriptive sample statistics

```
diffmean; Spool
```

```
## Length Width Height
## -22.62 -14.29 -11.25

##           Length Width Height
## Length 295.1 175.16 103.04
## Width   175.2 110.89  62.47
## Height  103.0  62.47  38.95
T2      # 66.8
```

```
##           [,1]
## [1,] 66.76
```

```
# F test p-value
```

```
1-pf((n1+n2-1-p)*T2/(p*(n1+n2-2)),df1=p,df2=n1+n2-1-p)
```

```
##           [,1]
## [1,] 1.141e-08
```

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Hottelling's T^2 using "manova" in R (Turtle data, $g = 2, p = 3$)

```
data=cbind(Length, Width, Height)
summary(manova(data~Gender),test="Hotelling")
```

```
##           Df Hotelling-Lawley approx F num Df den Df  P1
## Gender      1           1.45      21.3      3    44 1.1
## Residuals 46
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1
```

Remark: Given F critical, T^2 value can be recovered by

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

```
# F*(p*(n1+n2-2))/(n1+n2-1-p), F(3,44)=21.284
21.3*(3*(24+24-2))/(24+24-1-3)
```

```
## [1] 66.8
```

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Univariate marginal test of mean diff. on variable Length

Test each pair of component means by univariate t-test. On Length:

$$H_o : \mu_{male, Length} = \mu_{female, Length}$$

$$H_a : \mu_{male, Length} \neq \mu_{female, Length}$$

```
t.test(Length~Gender) # Marginal t=test
```

```
##
## Welch Two Sample t-test
##
## data: Length by Gender
## t = -4.6, df = 36, p-value = 6e-05
## alternative hypothesis: true difference in means between
## 95 percent confidence interval:
## -32.68 -12.57
## sample estimates:
## mean in group Male mean in group Female
## 113.4 136.0
```

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Marginal test on variable Width

$$H_o : \mu_{male, Width} = \mu_{female, Width}$$

$$H_a : \mu_{male, Width} \neq \mu_{female, Width}$$

```
t.test(Width~Gender)
```

```
##
## Welch Two Sample t-test
##
## data: Width by Gender
## t = -4.7, df = 35, p-value = 4e-05
## alternative hypothesis: true difference in means between
## 95 percent confidence interval:
## -20.461 -8.123
## sample estimates:
## mean in group Male mean in group Female
## 88.29 102.58
```

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Marginal test on variable Height

$$H_o : \mu_{male, Height} = \mu_{female, Height}$$

$$H_a : \mu_{male, Height} \neq \mu_{female, Height}$$

```
t.test(Height~Gender)
```

```
##
## Welch Two Sample t-test
##
## data: Height by Gender
## t = -6.2, df = 31, p-value = 7e-07
## alternative hypothesis: true difference in means between
## 95 percent confidence interval:
## -14.927 -7.573
## sample estimates:
## mean in group Male mean in group Female
## 40.71 51.96
```

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