## STAT 3690 Lecture 11

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### Testing on $A\mu$ (J&W pp. 279)

- **A** is of  $q \times p$  and  $\operatorname{rk}(\mathbf{A}) = q$ , i.e.,  $\mathbf{A} \mathbf{\Sigma} \mathbf{A}^{\top} > 0$
- Model: iid  $\mathbf{A}\mathbf{X}_i \sim MVN_q(\mathbf{A}\boldsymbol{\mu}, \mathbf{A}\boldsymbol{\Sigma}\mathbf{A}^{\top})$ .
- LRT for  $H_0: \mathbf{A}\boldsymbol{\mu} = \boldsymbol{\nu}_0$  v.s.  $H_1: \mathbf{A}\boldsymbol{\mu} \neq \boldsymbol{\nu}_0$ 
  - Test statistic:  $n(\mathbf{A}\bar{\mathbf{X}} \boldsymbol{\nu}_0)^{\top}(\mathbf{A}\mathbf{S}\mathbf{A}^{\top})^{-1}(\mathbf{A}\bar{\mathbf{X}} \boldsymbol{\nu}_0) \sim T^2(q, n-1) = \frac{(n-1)q}{n-q}F(q, n-q)$  under  $H_0$
  - Rejction region at level  $\alpha$ :  $R = \{\boldsymbol{x}_1, \dots, \boldsymbol{x}_n : \frac{n(n-q)}{q(n-1)} (\mathbf{A}\bar{\boldsymbol{x}} \boldsymbol{\nu}_0)^\top (\mathbf{A}\mathbf{S}\mathbf{A}^\top)^{-1} (\mathbf{A}\bar{\boldsymbol{x}} \boldsymbol{\nu}_0) \geq F_{1-\alpha,q,n-q} \}$
  - p-value:  $p(\mathbf{x}_1, ..., \mathbf{x}_n) = 1 F_{F(q, n-q)} \{ \frac{n(n-q)}{q(n-1)} (\mathbf{A}\bar{\mathbf{x}} \boldsymbol{\nu}_0)^{\top} (\mathbf{A}\mathbf{S}\mathbf{A}^{\top})^{-1} (\mathbf{A}\bar{\mathbf{x}} \boldsymbol{\nu}_0) \}$
- Multiple comparison
  - Interested in  $H_0: \mu_1 = \cdots = \mu_p$  v.s.  $H_1:$  Not all entries of  $\mu$  are equal. \*  $\mu_k$ : the kth entry of  $\mu$
  - Take

$$\boldsymbol{\nu}_0 = \mathbf{0}_{(p-1)\times 1}, \quad \mathbf{A} = \begin{bmatrix} 1 & -1 & 0 & \cdots & 0 \\ 1 & 0 & -1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 0 & 0 & \cdots & -1 \end{bmatrix}_{(p-1)\times p}.$$

$$- p = 2$$
 (i.e.,  $\mathbf{A} = [1, -1]$ ): A/B testing

# $(1-\alpha) \times 100\%$ confidence region (CR) for $\mu$ (J&W Sec. 5.4)

- $Pr((1-\alpha) \times 100\%CR \text{ covers } \boldsymbol{\mu}) = 1-\alpha$ 
  - CR is a set made of observations and is hence random
  - $-\mu$  is fixed
  - $(1-\alpha) \times 100\%$  CR covers  $\mu$  with probability  $(1-\alpha) \times 100\%$
- Dual problem of testing  $H_0: \mu = \mu_0$  v.s.  $H_1: \mu \neq \mu_0$  at the  $\alpha$  level
  - Translated from rejection region. Steps:
    - 1. Take R as a function of  $\mu_0$ ;
    - 2. Replace  $\mu_0$  with  $\mu$ ;
    - 3. Take the complement.
  - $(1 \alpha) \times 100\% \text{ CR} = \{ \boldsymbol{\mu} : n(\bar{\boldsymbol{x}} \boldsymbol{\mu})^{\top} \boldsymbol{\Sigma}^{-1} (\bar{\boldsymbol{x}} \boldsymbol{\mu}) < \chi^2_{1-\alpha,p} \} \text{ if } \boldsymbol{\Sigma} \text{ is known}$
  - $-(1-\alpha) \times 100\% \text{ CR} = \{ \boldsymbol{\mu} : \frac{n(n-p)}{p(n-1)} (\bar{\boldsymbol{x}} \boldsymbol{\mu})^{\top} \mathbf{S}^{-1} (\bar{\boldsymbol{x}} \boldsymbol{\mu}) < F_{1-\alpha,p,n-p} \} \text{ if } \boldsymbol{\Sigma} \text{ is not known}$

 $(1-\alpha) \times 100\%$  CR for  $\nu = A\mu$ 

• 
$$\mathbf{X}_1, \dots, \mathbf{X}_n \stackrel{\mathrm{iid}}{\sim} MVN_p(\boldsymbol{\mu}, \boldsymbol{\Sigma})$$

– Unknown 
$$\Sigma$$

$$-n>p$$

- **A** is of  $q \times p$  and  $\operatorname{rk}(\mathbf{A}) = q$ , i.e.,  $\mathbf{A} \mathbf{\Sigma} \mathbf{A}^{\top} > 0$
- Then iid  $\mathbf{A}\mathbf{X}_i \sim MVN_q(\boldsymbol{\nu}, \mathbf{A}\boldsymbol{\Sigma}\mathbf{A}^\top)$
- $(1-\alpha) \times 100\%$  CR for  $\nu$  is  $\{\nu : \frac{n(n-q)}{q(n-1)} (\mathbf{A}\bar{x} \nu)^{\top} (\mathbf{A}\mathbf{S}\mathbf{A}^{\top})^{-1} (\mathbf{A}\bar{x} \nu) < F_{1-\alpha,q,n-q} \}$
- Special case:  $\mathbf{A} = \boldsymbol{a} \in \mathbb{R}^p$ 
  - $-(1-\alpha)\times 100\%$  confidence interval (CI) for scalar  $\nu=\boldsymbol{a}^{\top}\boldsymbol{\mu}$  is

$$\{\nu: n(\boldsymbol{a}^{\top}\bar{\boldsymbol{x}}-\nu)^{2}(\boldsymbol{a}^{\top}\mathbf{S}\boldsymbol{a})^{-1} < F_{1-\alpha,1,n-1}\} = \left(\boldsymbol{a}^{\top}\bar{\boldsymbol{x}}-t_{1-\alpha/2,n-1}\sqrt{\boldsymbol{a}^{\top}\mathbf{S}\boldsymbol{a}/n}, \boldsymbol{a}^{\top}\bar{\boldsymbol{x}}+t_{1-\alpha/2,n-1}\sqrt{\boldsymbol{a}^{\top}\mathbf{S}\boldsymbol{a}/n}\right)$$

#### Simultaneous confidence intervals

- Construct  $(1 \alpha_k)$  CI for scalars  $\boldsymbol{a}_k^{\top} \boldsymbol{\mu}$ , say  $\text{CI}_k$ ,  $k = 1, \dots, m$ , simultaneously
- Make sure  $\Pr(\bigcap_{k} \{ \boldsymbol{a}_{k}^{\top} \boldsymbol{\mu} \in \operatorname{CI}_{k} \}) \geq 1 \alpha$
- Bonferroni correction
  - Bonferroni inequality:

$$\Pr(\bigcap_{k=1}^{m} \{\boldsymbol{a}_{k}^{\top} \boldsymbol{\mu} \in \operatorname{CI}_{k}\}) = 1 - \Pr(\bigcup_{k=1}^{m} \{\boldsymbol{a}_{k}^{\top} \boldsymbol{\mu} \notin \operatorname{CI}_{k}\}) \ge 1 - \sum_{k=1}^{m} \Pr(\boldsymbol{a}_{k}^{\top} \boldsymbol{\mu} \notin \operatorname{CI}_{k}) = 1 - \sum_{k=1}^{m} \alpha_{k}$$

– Taking  $\alpha_k$  such that  $\alpha = \sum_{k=1}^m \alpha_k$ , e.g.,  $\alpha_k = \alpha/m$ , i.e.,

$$(\boldsymbol{a}_k^{\top}\bar{\boldsymbol{x}} - t_{1-\alpha/(2m),n-1}\sqrt{\boldsymbol{a}_k^{\top}\mathbf{S}\boldsymbol{a}_k/n},\boldsymbol{a}_k^{\top}\bar{\boldsymbol{x}} + t_{1-\alpha/(2m),n-1}\sqrt{\boldsymbol{a}_k^{\top}\mathbf{S}\boldsymbol{a}_k/n})$$

- Working for small m
- · Scheffé's method

– Let 
$$CI_{\boldsymbol{w}} = (\boldsymbol{w}^{\top} \bar{\boldsymbol{x}} - c\sqrt{\boldsymbol{w}^{\top} \mathbf{S} \boldsymbol{w}/n}, \boldsymbol{w}^{\top} \bar{\boldsymbol{x}} + c\sqrt{\boldsymbol{w}^{\top} \mathbf{S} \boldsymbol{w}/n})$$
 for all  $\boldsymbol{w} \in \mathbb{R}^p$ 

– Derive that  $c = \sqrt{p(n-1)(n-p)^{-1}F_{1-\alpha,p,n-p}}$ 

\* By Cauchy–Schwarz: 
$$\{\boldsymbol{w}^{\top}(\bar{\boldsymbol{x}}-\boldsymbol{\mu})\}^2 = [(\mathbf{S}^{1/2}\boldsymbol{w})^{\top}\{\mathbf{S}^{-1/2}(\bar{\boldsymbol{x}}-\boldsymbol{\mu})\}]^2 \le \{(\boldsymbol{w}^{\top}\mathbf{S}\boldsymbol{w})^{\top}/n\}\{n(\bar{\boldsymbol{x}}-\boldsymbol{\mu})^{\top}\mathbf{S}^{-1}(\bar{\boldsymbol{x}}-\boldsymbol{\mu})\} \Rightarrow$$

$$\Pr(\bigcap_{k=1}^{m} \{\boldsymbol{a}_{k}^{\top} \boldsymbol{\mu} \in \operatorname{CI}_{k}\}) \ge \Pr(\bigcap_{\boldsymbol{w} \in \mathbb{R}^{p}} \{\boldsymbol{w}^{\top} \boldsymbol{\mu} \in \operatorname{CI}_{\boldsymbol{w}}\}) = 1 - \Pr(\bigcup_{\boldsymbol{w} \in \mathbb{R}^{p}} \{\boldsymbol{w}^{\top} \boldsymbol{\mu} \notin \operatorname{CI}_{\boldsymbol{w}}\})$$

$$= 1 - \Pr(\bigcup_{\boldsymbol{w} \in \mathbb{R}^{p}} [\{\boldsymbol{w}^{\top} (\bar{\mathbf{X}} - \boldsymbol{\mu})\}^{2} / \{(\boldsymbol{w}^{\top} \mathbf{S} \boldsymbol{w})^{\top} / n\} > c^{2}])$$

$$\ge 1 - \Pr(\{n(\bar{\mathbf{X}} - \boldsymbol{\mu})^{\top} \mathbf{S}^{-1} (\bar{\mathbf{X}} - \boldsymbol{\mu}) > c^{2}\})$$

\* 
$$\Pr(\{n(\bar{\mathbf{X}} - \boldsymbol{\mu})^{\top} \mathbf{S}^{-1}(\bar{\mathbf{X}} - \boldsymbol{\mu}) > c^2\}) = \alpha \Rightarrow c = \sqrt{p(n-1)(n-p)^{-1} F_{1-\alpha,p,n-p}}$$

- Working for large even infinite m

## Comparing two multivariate means (J&W Sec. 6.3)

• Two independent samples of (potentially) different sizes from two distributions with equal covariance

$$-\mathbf{X}_{11}, \dots, \mathbf{X}_{1n_1} \overset{\text{iid}}{\sim} MVN_p(\boldsymbol{\mu}_1, \boldsymbol{\Sigma}) \\ -\mathbf{X}_{21}, \dots, \mathbf{X}_{2n_2} \overset{\text{iid}}{\sim} MVN_p(\boldsymbol{\mu}_2, \boldsymbol{\Sigma})$$

- Let  $\bar{\mathbf{X}}_i$  and  $\mathbf{S}_i$  be the sample mean and sample covariance for the *i*th sample
- Hypotheses  $H_0: \boldsymbol{\mu}_1 = \boldsymbol{\mu}_2$  v.s.  $H_1: \boldsymbol{\mu}_1 \neq \boldsymbol{\mu}_2$
- Test statistic following LRT

$$(\bar{\mathbf{X}}_1 - \bar{\mathbf{X}}_2)^{\top} \{ (n_1^{-1} + n_2^{-1}) \mathbf{S}_{\text{pool}} \}^{-1} (\bar{\mathbf{X}}_1 - \bar{\mathbf{X}}_2) \sim \frac{(n_1 + n_2 - 2)p}{n_1 + n_2 - p - 1} F(p, n_1 + n_2 - p - 1)$$

• Rejection region at level  $\alpha$ 

$$\left\{ x_{11}, \dots, x_{1n_1}, x_{21}, \dots, x_{2n_2} : (\bar{\boldsymbol{x}}_1 - \bar{\boldsymbol{x}}_2)^\top \{ (n_1^{-1} + n_2^{-1}) \mathbf{S}_{\text{pool}} \}^{-1} (\bar{\boldsymbol{x}}_1 - \bar{\boldsymbol{x}}_2) \ge \frac{p(n_1 + n_2 - 2)}{n_1 + n_2 - p - 1} F_{1-\alpha, p, n_1 + n_2 - p - 1} \right\} - \mathbf{S}_{\text{pool}} = \frac{(n_1 - 1) \mathbf{S}_1 + (n_2 - 1) \mathbf{S}_2}{n_1 + n_2 - 2}$$

• *p*-value

$$1 - F_{F_{1-\alpha,p,n_1+n_2-p-1}} \left[ \frac{n_1 + n_2 - p - 1}{p(n_1 + n_2 - 2)} (\bar{\boldsymbol{x}}_1 - \bar{\boldsymbol{x}}_2)^\top \{ (n_1^{-1} + n_2^{-1}) \mathbf{S}_{\text{pool}} \}^{-1} (\bar{\boldsymbol{x}}_1 - \bar{\boldsymbol{x}}_2) \right]$$