

- 1 a In a standard 1-way ANOVA with  $k$  groups and  $n$  observations per group with the usual notation, show that the statistic

$$T_{k,n} = \max_{1 \leq i < j \leq k} n^{1/2} |\bar{Y}_i - \bar{Y}_j - (\mu_i - \mu_j)| / \hat{\sigma}$$

is distributed as the studentized range distribution and identify its degrees of freedom.

a. in one-way ANOVA, the  $n$  observations in the  $i$ th group follows:  $Y_{i1} \dots Y_{in} \stackrel{iid}{\sim} N(\mu_i, \sigma^2)$

then:  $\bar{Y}_i \sim N(\mu_i, \frac{\sigma^2}{n})$ ,  $i = 1, 2, \dots, k$

$$\Rightarrow \bar{Y}_i - \mu_i \stackrel{iid}{\sim} N(0, \frac{\sigma^2}{n}), i = 1, 2, \dots, k$$

$$\begin{aligned} \Rightarrow q &= \max_{1 \leq i < j \leq k} \frac{|(\bar{Y}_i - \mu_i) - (\bar{Y}_j - \mu_j)|}{\sqrt{s^2/n}} \\ &= \max_{1 \leq i < j \leq k} \frac{|(\bar{Y}_i - \bar{Y}_j) - (\mu_i - \mu_j)|}{\hat{\sigma}} \cdot \sqrt{n} \end{aligned}$$

$\sim$  studentized range distribution with  $df = (a, b)$

where  $b = (nk - k)$ , as  $s^2/\sigma^2 \sim \chi^2(nk - k)$ ;  $a = k$ , as  $(\bar{Y}_i - \mu_i) \stackrel{iid}{\sim} N(0, \frac{\sigma^2}{n})$ ,  $i = 1, \dots, k$

$\Rightarrow T_{k,n}$  follows studentized range distribution with  $df = (k, nk - k)$   
(notated in the setting)

- b Find the asymptotic distribution of  $T_{k,n}$  as  $n$  tends to infinity.

b. under setting of standard 1-way ANOVA:

denote  $e_i = \bar{Y}_i - \mu_i$ , then  $e_i \stackrel{iid}{\sim} N(0, \frac{\sigma^2}{n})$ ,  $i = 1, 2, \dots, k$

$$\frac{e_i}{\sigma/\sqrt{n}} \stackrel{iid}{\sim} N(0, 1), i = 1, 2, \dots, k$$

$$\begin{aligned} T_{k,n} &= \max_{1 \leq i < j \leq k} \frac{|e_i - e_j| / (\sigma/\sqrt{n})}{\hat{\sigma} / \sigma} \\ &= \left(\frac{\hat{\sigma}}{\sigma}\right) \cdot \max_{1 \leq i < j \leq k} |e_i - e_j| / \sigma/\sqrt{n} \\ &\triangleq \left(\frac{\hat{\sigma}}{\sigma}\right) \cdot Q_{k,n} \end{aligned}$$

as  $\lim_{n \rightarrow \infty} \left(\frac{\hat{\sigma}}{\sigma}\right) = 1$ , and by definition,  $Q_{k,n} \sim$  range distribution  $q_{k,n}$

by Slutsky's theorem:  $T_{k,n} \xrightarrow{d} q_{k,n}$

Also, the result remains valid when normality assumption is violated as  $e_i/\sigma/\sqrt{n} \xrightarrow{d} N(0, 1)$  by CLT.

- 4 Let  $U = (U_1, U_2, \dots, U_p)'$  be a vector of principal components of  $X$ . Then  $U_i = a_i' X$  for some vector  $a_i$  of length 1,  $i=1, 2, \dots, p$ . Show that
- a  $\text{var } a'X \leq \text{var } U_1$

a.  $U = \begin{pmatrix} U_1 \\ \vdots \\ U_p \end{pmatrix} = A'X = \begin{pmatrix} a_1' \\ \vdots \\ a_p' \end{pmatrix} X$ ,  $\text{cov}(X) = \Sigma \succ 0$

by definition of PCA, the  $A$  is an orthogonal matrix s.t.  $A \Sigma A' = \Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$   
 $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$

therefore,  $\text{cov}(A'X) = \text{cov}(U) = A' \text{cov}(X) A$   
 $= A' \Sigma A$   
 $= \Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$

then:  $\text{var}(U_1) = \lambda_1$

let  $a$  be any  $n \times 1$  vector s.t.  $a'a = 1$

$$\begin{aligned} \text{var}(a'X) &= a' \Sigma a \\ &= a' (A \Lambda A') a \\ &= (A'a)' \Lambda (A'a) \\ &= \begin{pmatrix} a_1'a \\ \vdots \\ a_n'a \end{pmatrix}' \text{diag}(\lambda_1, \dots, \lambda_n) \begin{pmatrix} a_1'a \\ \vdots \\ a_n'a \end{pmatrix} \\ &= \sum_{i=1}^n a_i' a \lambda_i a_i' a \\ &= a' \left( \sum_{i=1}^n a_i \lambda_i a_i' \right) a \\ &\leq a' \left( \sum_{i=1}^n \lambda_1 \cdot a_i a_i' \right) a \end{aligned}$$

since  $AA' = I$ ,  $\sum_{i=1}^n a_i a_i' = AA' = I$

$\text{var}(a'X) \leq \lambda_1 \cdot a'a = \lambda_1 = \text{var}(U_1)$ , for any  $a$  s.t.  $\|a\| = 1$   $\square$

b if  $a'X$  is uncorrelated with  $U_1, U_2, \dots, U_{i-1}$ , then  $\text{var } a'X \leq \text{var } U_i$ .

b.  $a'X \perp U_1, \dots, U_{i-1} \Leftrightarrow \text{cov}(a'X, a_k'X) = 0$ ,  $k=1, 2, \dots, i-1$

$\text{cov}(a'X, a_k'X) = a' \Sigma a_k = a' A \Lambda A' a_k = a' \left( \sum_{i=1}^n a_i \lambda_i a_i' \right) a_k$

for  $AA' = I$ :  $= a' a_k \lambda_k a_k' a_k = \lambda_k \cdot a'a_k \geq 0 \Rightarrow a'a_k = 0, k=1, \dots, i-1$

then  $\text{var}(a'X) = a' \Sigma a = a' A \Lambda A' a$

$$= \sum_{i=1}^n a_i' a_i \lambda_i a_i' a$$

$$\text{for } a_k' a = 0, k < i: \quad = \sum_{k=i}^n a_k' a_k \lambda_k a_k' a$$

$$\leq \lambda_i \cdot a_i' \sum_{k=i}^n a_k a_k' \cdot a$$

$$= \lambda_i \cdot a_i' \cdot \begin{bmatrix} 0 \\ I_{n-i+1} \end{bmatrix} a$$

$$\text{for } a_i' a = 1: \quad \leq \lambda_i \cdot 1 = \lambda_i$$

therefore, for any  $a'x$  s.t.  $a'x \perp u_1 \dots u_{i-1}$ ,  $\text{var}(a'x) \leq \lambda_i$ .  $\square$