

(a) i. $\Lambda = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ -1 & 0 \\ 0 & -1 \\ 0 & 1 \\ \parallel & \parallel \\ \lambda_1 & \lambda_2 \end{bmatrix} \Rightarrow \text{rank}(\Lambda) = 2$ (columns are linearly indep) ... (A)

$$X = \begin{bmatrix} 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Basis vectors for $e(X^T)$:

$$\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

" " " "

u_1 u_2 u_3 u_4

$\Rightarrow \text{rank}(X) = 4$

(B) $\begin{cases} \lambda_1 = u_1 - u_2 \Rightarrow \lambda_1 \in e(X^T) \Rightarrow \lambda_1^T \beta \text{ estimable} \\ \lambda_2 = u_3 - u_4 \Rightarrow \lambda_2 \in e(X^T) \Rightarrow \lambda_2^T \beta \text{ estimable} \end{cases}$

By (A) & (B) $\Lambda^T \beta = 0$ is testable.

(b) From lecture, BLUE of $\Lambda^T \beta$ is $\Lambda^T \hat{\beta}$ where $\hat{\beta}$: LSE

$$\Lambda^T \hat{\beta} = \Lambda^T (X^T X)^{-1} X^T y \sim N_0 \left(\underset{\substack{\uparrow \\ \text{unbiased}}}{\Lambda^T \beta}, \sigma^2 \Lambda^T (X^T X)^{-1} \Lambda \right)$$

y follows a n -dim normal ($n=8$)
 \Rightarrow linear combinations of y follow a normal

Given in the question

$$\hat{\beta} = (X^T X)^{-1} X^T y = \begin{bmatrix} \bar{y}_{..} \\ \bar{y}_{1.} - \bar{y}_{..} \\ \bar{y}_{2.} - \bar{y}_{..} \\ \bar{y}_{3.} - \bar{y}_{..} \\ \bar{y}_{4.} - \bar{y}_{..} \end{bmatrix} \Rightarrow \Lambda^T \hat{\beta} = \begin{bmatrix} \bar{y}_{1.} - \bar{y}_{2.} \\ \bar{y}_{3.} - \bar{y}_{4.} \end{bmatrix}$$

$$\sigma^2 \Lambda^T (X^T X)^{-1} \Lambda = \sigma^2 \begin{bmatrix} 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{2} & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{2} & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 & 0 \\ -1 & 0 \\ 0 & 0 \\ 0 & -1 \\ 0 & 1 \end{bmatrix}$$

$$= \sigma^2 \begin{bmatrix} 0 & \frac{1}{2} & -\frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ -1 & 0 \\ 0 & 1 \\ 0 & -1 \end{bmatrix} = \sigma^2 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\Rightarrow \Lambda^T \hat{\beta} = \begin{bmatrix} \bar{y}_1 & -\bar{y}_2 \\ \bar{y}_3 & -\bar{y}_4 \end{bmatrix} \sim N_2 \left(\begin{bmatrix} \alpha_1 - \alpha_2 \\ \alpha_3 - \alpha_4 \end{bmatrix}, \sigma^2 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right)$$

(c) (i) $\text{rank}(X) = 4$

(iii) $\frac{\text{SSE}}{\sigma^2} = \frac{y^T (I-P)y}{\sigma^2} \sim \chi^2(8-4)$

TRUE

$$\Lambda^T \hat{\beta} = \Lambda^T (X^T X)^{-1} X^T y = (X^T \Lambda)^T (X^T X)^{-1} X^T y = A^T X (X^T X)^{-1} X^T y = A^T P y$$

since $\lambda_1, \lambda_2 \in e(X^T)$

$$\exists a_1, a_2 \text{ s.t. } \lambda_1 = X^T a_1 \text{ \& } \lambda_2 = X^T a_2$$

$$\text{Let } A = [a_1 \mid a_2] \Rightarrow X^T A = \Lambda$$

$$\begin{aligned} \text{cov}((I-P)y, A^T P y) &= (I-P) \text{cov}(y, y) P A \\ &= \sigma^2 (I-P) P A \\ &= 0 \end{aligned}$$

\Rightarrow since $(I-P)y$ and $A^T P y$ are normal,

$$\text{cov}((I-P)y, A^T P y) = 0 \Leftrightarrow (I-P)y \text{ \& } A^T P y \text{ are indep.}$$

$$\left. \begin{array}{l} \frac{\text{SSE}}{\sigma^2} \text{ is a function of } (I-P)y \\ \Lambda^T \hat{\beta} \text{ is a function of } A^T P y \end{array} \right\} \Rightarrow \frac{\text{SSE}}{\sigma^2} \text{ \& } \Lambda^T \hat{\beta} \text{ are indep.}$$

(d) $\Lambda^T \hat{\beta} \sim N_2(\Lambda^T \beta, \sigma^2 I_2)$

$$\Rightarrow \left[\begin{array}{l} \frac{1}{\sigma^2} (\Lambda^T \hat{\beta})^T (\Lambda^T \hat{\beta}) \sim \chi^2(2, \frac{(\Lambda^T \beta)^T (\Lambda^T \beta)}{2}) \\ \frac{SSE}{\sigma^2} \sim \chi^2(4) \end{array} \right] \text{ indep.}$$

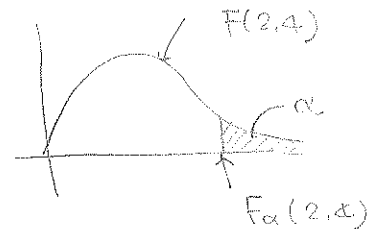
$$F = \frac{(\Lambda^T \hat{\beta})^T (\Lambda^T \hat{\beta}) / 2}{SSE / 4} \sim F(2, 4, \frac{(\Lambda^T \beta)^T (\Lambda^T \beta)}{2})$$

$$\Rightarrow F = \frac{\left\{ (\bar{y}_1 - \bar{y}_2)^2 + (\bar{y}_3 - \bar{y}_4)^2 \right\} / 2}{\sum_{i,j=1}^4 (\hat{y}_{ij} - \bar{y}_i)^2 / 4} \sim F(2, 4, \frac{(d_1 - d_2)^2 + (d_3 - d_4)^2}{2})$$

= ϕ
= 0 under H_0

The test procedure is

Reject $H_0: d_1 - d_2 = 0$ & $d_3 - d_4 = 0$ if $F > F_\alpha(2, 4)$



v. As $|d_1 - d_2|$ becomes bigger OR $|d_3 - d_4|$ becomes bigger

$(d_1 - d_2)^2$ and/or $(d_3 - d_4)^2$ increases. \Rightarrow the noncentrality

parameter is increasing.

Since the noncentral F distribution is stochastically increasing in ϕ ,

the power of the developed test is increasing.

$$\text{power} = P(F > F_\alpha(2, 4) | H_0 \text{ is not true})$$

2 (a)

$$X_0 = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \end{bmatrix}$$

$$\beta_0 = \begin{bmatrix} \mu \\ \alpha_1 \\ \alpha_2 \end{bmatrix}$$

$$\text{rank}(X_0) = 2.$$

(b) $P_0 = X_0 (X_0^T X_0)^{-1} X_0^T =$

$$X_0^T X_0 = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 8 & 4 & 4 \\ 4 & 4 & 0 \\ 4 & 0 & 4 \end{bmatrix}$$

$$\Rightarrow (X_0^T X_0)^{-1} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & \frac{1}{4} & 0 \\ 0 & 0 & \frac{1}{4} \end{bmatrix}$$

$$(X_0^T X_0)^{-1} X_0^T = \begin{bmatrix} 0 & 0 & 0 \\ 0 & \frac{1}{4} & 0 \\ 0 & 0 & \frac{1}{4} \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \end{bmatrix}$$

$$P_0 = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \end{bmatrix} = \begin{bmatrix} \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & 0 & 0 & 0 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & 0 & 0 & 0 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & 0 & 0 & 0 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ 0 & 0 & 0 & 0 & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ 0 & 0 & 0 & 0 & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ 0 & 0 & 0 & 0 & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \end{bmatrix}$$

$$Py = \begin{bmatrix} \bar{y}_{1.} \\ \bar{y}_{1.} \\ \bar{y}_{2.} \\ \bar{y}_{2.} \\ \bar{y}_{3.} \\ \bar{y}_{3.} \\ \bar{y}_{4.} \\ \bar{y}_{4.} \end{bmatrix}$$

$$P_0 y = \begin{bmatrix} \frac{\bar{y}_{1.} + \bar{y}_{2.}}{2} \\ \frac{\bar{y}_{1.} + \bar{y}_{2.}}{2} \\ \frac{\bar{y}_{3.} + \bar{y}_{4.}}{2} \\ \frac{\bar{y}_{3.} + \bar{y}_{4.}}{2} \\ \frac{\bar{y}_{3.} + \bar{y}_{4.}}{2} \\ \frac{\bar{y}_{3.} + \bar{y}_{4.}}{2} \\ \frac{\bar{y}_{3.} + \bar{y}_{4.}}{2} \\ \frac{\bar{y}_{3.} + \bar{y}_{4.}}{2} \end{bmatrix}$$

$$\Rightarrow (P - P_0)y = \begin{bmatrix} \frac{\bar{y}_{1.} - \bar{y}_{2.}}{2} \\ \frac{\bar{y}_{1.} - \bar{y}_{2.}}{2} \\ \frac{\bar{y}_{2.} - \bar{y}_{1.}}{2} \\ \frac{\bar{y}_{2.} - \bar{y}_{1.}}{2} \\ \frac{\bar{y}_{3.} - \bar{y}_{4.}}{2} \\ \frac{\bar{y}_{3.} - \bar{y}_{4.}}{2} \\ \frac{\bar{y}_{4.} - \bar{y}_{3.}}{2} \\ \frac{\bar{y}_{4.} - \bar{y}_{3.}}{2} \end{bmatrix}$$

$$\Rightarrow \frac{y^T (P - P_0) y}{\sigma^2} = \frac{4(\bar{y}_{1.} - \bar{y}_{2.})^2 + 4(\bar{y}_{3.} - \bar{y}_{4.})^2}{4\sigma^2} = \frac{(\bar{y}_{1.} - \bar{y}_{2.})^2 + (\bar{y}_{3.} - \bar{y}_{4.})^2}{\sigma^2}$$

$$\frac{y}{\sigma} \sim N_n(X\beta, I_n) \quad \& \quad (P - P_0) \text{ is symmetric \& idempotent}$$

by the results in lecture

$$\Rightarrow \frac{y^T (P - P_0) y}{\sigma^2} \sim \chi^2 \left(\begin{array}{c} r(P - P_0), \\ \parallel \\ 4 - 2 \\ = 2 \end{array}, \frac{(X\beta)^T (P - P_0) X\beta}{2\sigma^2} \right)$$

(c) TRUE $\text{Cov}((P - P_0)y, (I - P)y) = \sigma^2 (P - P_0)(I - P)$

$$= \sigma^2 (P - P^2 - P_0 - P_0 P)$$

$$= \sigma^2 (P - P - P_0 - P_0)$$

$$= 0$$

since P is idempotent and $E(x_0) \subset C(X)$

\Rightarrow Since $(P - P_0)y$ and $(I - P)y$ are normal,
 $\text{Cov} = 0 \Leftrightarrow (P - P_0)y$ and $(I - P)y$ are indep.

(d) Test statistic since $\frac{y^T(P-P_0)y}{\sigma^2}$ and $\frac{SSE}{\sigma^2}$ are indep,

$$F = \frac{\frac{y^T(P-P_0)y}{\sigma^2} / 2}{\frac{SSE}{\sigma^2} / 4} = \frac{2 \cdot \left((\bar{y}_{1\cdot} - \bar{y}_{2\cdot})^2 + (\bar{y}_{3\cdot} - \bar{y}_{4\cdot})^2 \right)}{\sum_i \sum_j (y_{ij} - \bar{y}_{i\cdot})^2}$$

$$\sim F(2, 4, \underbrace{\frac{(XP)^T(P-P_0)(XP)}{2\sigma^2}}_{\phi})$$

since $E(X) \subset E(X)$, $PX_0\beta_0 = P_0X_0\beta_0$

under H_0 , $E((P-P_0)y) = (P-P_0)X_0\beta_0 = 0 \Rightarrow \phi = 0$

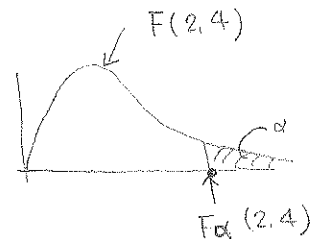
under H_0 not true, $E((P-P_0)y) = (P-P_0)X\beta = PX\beta - P_0X\beta = X\beta - P_0X\beta$

$$= \begin{bmatrix} \mu + \alpha_1 \\ \mu + \alpha_1 \\ \vdots \\ \mu + \alpha_4 \\ \mu + \alpha_4 \end{bmatrix} - \begin{bmatrix} \mu + \frac{\alpha_1 + \alpha_2}{2} \\ \mu + \frac{\alpha_1 + \alpha_2}{2} \\ \vdots \\ \mu + \frac{\alpha_3 + \alpha_4}{2} \end{bmatrix} = \begin{bmatrix} \frac{\alpha_1 - \alpha_2}{2} \\ \frac{\alpha_1 - \alpha_2}{2} \\ \frac{\alpha_2 - \alpha_1}{2} \\ \frac{\alpha_2 - \alpha_1}{2} \\ \frac{\alpha_3 - \alpha_4}{2} \\ \frac{\alpha_3 - \alpha_4}{2} \\ \frac{\alpha_4 - \alpha_3}{2} \\ \frac{\alpha_4 - \alpha_3}{2} \end{bmatrix}$$

$$\phi = \frac{(XP)^T(P-P_0)X\beta}{2\sigma^2} = \frac{(\alpha_1 - \alpha_2)^2 + (\alpha_3 - \alpha_4)^2}{2\sigma^2}$$

Test procedure,

Reject H_0 if $F > F_{0.05}(2, 4)$



As I showed the above, we see the test statistic is the same

$$F = \frac{2 \cdot \left((\bar{y}_{1\cdot} - \bar{y}_{2\cdot})^2 + (\bar{y}_{3\cdot} - \bar{y}_{4\cdot})^2 \right)}{\sum_i \sum_j (y_{ij} - \bar{y}_{i\cdot})^2} \text{ and the distribution of } F \text{ is}$$

the same, $F \sim F(2, 4, \phi = \frac{(\alpha_1 - \alpha_2)^2 + (\alpha_3 - \alpha_4)^2}{2\sigma^2})$