

Solution TMA 4267, May 25 2013

Problem 1

a)  $\rho_{x_3, x_1} = \frac{1}{\sqrt{3} \cdot \sqrt{1}} = \frac{\sqrt{3}}{3}$

$\rho_{x_3, x_2} = \frac{-1}{\sqrt{2} \sqrt{3}} = -\frac{\sqrt{6}}{6}$ , since  $|\frac{\sqrt{3}}{3}| > |-\frac{\sqrt{6}}{6}|$

it follows that  $x_1$  have the strongest correlation with  $x_3$  in absolute value

let  $\underline{z} = \begin{bmatrix} -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \underline{x}$ , i. e. a linear transformation of  $\underline{x}$  which is trivariate normal

and thereby bivariate normal distributed

since  $\underline{A}$  is a  $2 \times 3$  matrix

$E[\underline{z}] = \underline{A} \underline{\mu} = \begin{bmatrix} -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \end{bmatrix} = \begin{bmatrix} \mu_2 - \mu_1 \\ \mu_3 - \mu_1 \end{bmatrix} = \begin{bmatrix} 6 - 2 \\ 4 - 2 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$

$\Sigma_z = \underline{A} \underline{\Sigma} \underline{A}^d = \begin{bmatrix} -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & -1 \\ 1 & -1 & 3 \end{bmatrix} \begin{bmatrix} -1 & -1 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} -1 & 0 \\ 2 & -1 \\ -2 & 2 \end{bmatrix} = \begin{bmatrix} 3 & -1 \\ -1 & 2 \end{bmatrix}$

b)  $\underline{y} = \begin{bmatrix} \underline{e}_1^+ \\ \underline{e}_2^+ \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \underline{A} \underline{x}$  and is thereby a linear transformation

of  $\text{Var}(x_i) = \lambda_i$ .  $\underline{A}$  is a  $2 \times 3$  matrix

$\text{Cov}(y_1, y_2) = \text{Cov}(\underline{e}_1^+ \underline{x}, \underline{e}_2^+ \underline{x}) = E[\underline{e}_1^+ (\underline{x} - \underline{\mu})(\underline{x} - \underline{\mu})^+ \underline{e}_2^+]$   
 $= \underline{e}_1^+ \underline{\Sigma} \underline{e}_2 = \underline{e}_1^+ \lambda \underline{e}_2 = \lambda \underline{e}_1^+ \underline{e}_2 = \underline{0}$

Since  $y_1$  and  $y_2$  are univariate normal and uncorrelated, they are independent.

$\frac{\lambda_1 + \lambda_2}{\lambda_1 + \lambda_2 + \lambda_3} = \frac{3.88 + 1.65}{6} = 0.921$ , i. e. about 92.1%

of the total variation in  $\underline{x}$  can be explained by  $\underline{y}$ .

c) Since  $\underline{Y}$  is bivariate normal distributed we know that  $Q$  is  $\chi^2$ -distributed with two degrees of freedom. If  $Q \geq \chi_{\alpha, 2}^2$  for some  $\alpha$  the product would not pass the control.

Suppose  $\underline{U} = \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}$  is trivariate normal distributed with mean vector  $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$  and covariance matrix  $\begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} = \underline{\Lambda}$

Let  $\underline{X} = \underline{P}\underline{U} = [\underline{e}_1, \underline{e}_2, \underline{e}_3]\underline{U}$ . Then  $\text{Cov}(\underline{X}) = \underline{P}\underline{\Lambda}\underline{P}^t = \underline{\Sigma}$

Hence simulate  $n$  observation vectors from  $\underline{U}$  and let

$$\underline{X} = \underline{P}\underline{U} + \underline{\mu}$$

### Problem 2

$H_0: \alpha_1 = \alpha_2 = \alpha_3 = 0$        $H_1: \alpha_i \neq 0$  for *at least one*  $i, i=1, 2, 3$

$F_{obs} = 15.28$ .       $P(F_{2,6} \geq 15.28) = 0.004 < 0.05$ . We conclude

that ~~the~~ at least one  $\alpha_i \neq 0$ , i.e. the declination has changed.

There are three pairwise comparisons  $\mu_2 - \mu_1$ ,  $\mu_3 - \mu_1$ ,  $\mu_3 - \mu_2$

Only the confidence interval  $\mu_3 - \mu_1$  does not cover 0 using the exact Tukey test. Hence we can only conclude that  $\mu_3 \neq \mu_1$

The corresponding time span is  $1865 - 1669 = 196$  year.

The two others time span are: 111 and 85 respectively, but according to the Tukey test we cannot conclude that 100 year is sufficient to conclude there has been a change though the comparison for  $\mu_3 - \mu_2$  is pretty close.

$$b) \quad \beta_1 = \mu + d_1, \quad \beta_2 = \mu + d_2, \quad \beta_3 = \mu + d_3$$

$$\text{cov}(\hat{\beta}) = \sigma^2 (X'X)^{-1} = \sigma^2 \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}^{-1} = \sigma^2 \begin{bmatrix} \frac{1}{3} & 0 & 0 \\ 0 & \frac{1}{3} & 0 \\ 0 & 0 & \frac{1}{3} \end{bmatrix} = \sigma^2 \frac{1}{3} I$$

Hence  $\hat{\beta}_1$ , ~~and~~  $\hat{\beta}_2$  and  $\hat{\beta}_3$  are uncorrelated. Since the  $Y$ 's are normal distributed they are also independent.

c)

$$\underline{H} = \underline{X} \cdot \frac{1}{3} \underline{I} \cdot \underline{X}' = \frac{1}{3} \underline{X} \underline{X}' = \frac{1}{3} \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & \sqrt{3} & 0 \\ 0 & 0 & \sqrt{3} \end{bmatrix}$$

Since  $\underline{H} = \underline{J}^{\circ}$ , ~~it~~ it follows that expression 1 and 2 are equal.

All the matrices defined are shown in to be projection matrices in the course. Hence

$$\text{rank}(\underline{I} - \frac{1}{9} \underline{J}) = \text{tr}(\underline{I}) - \frac{1}{9} \text{tr}(\underline{J}) = 9 - \frac{1}{9} \cdot 9 = 8$$

$$\text{rank}(\underline{I} - \underline{J}^{\circ}) = \text{rank}(\underline{I} - \underline{H}) = \text{tr}(\underline{I}) - \text{tr}(\underline{H}) = 9 - \frac{1}{3} \cdot 9 = 6$$

$$\text{rank}(\underline{J}^{\circ} - \frac{1}{9} \underline{J}) = \text{rank}(\underline{H} - \frac{1}{9} \underline{J}) = \text{tr}(\underline{H}) - \frac{1}{9} \text{tr}(\underline{J}) = 3 - 1 = 2$$

$$\text{rank}(\underline{H}) = \text{tr}(\underline{H}) = 3$$

d) According to our theoretical results since  $(\underline{I} - \underline{H})(\underline{Y} - \underline{X}\beta)$  and  $Y \sim N(\underline{X}\beta, \sigma^2 \underline{I})$   
 $= (\underline{I} - \underline{H})\underline{Y}$ , the two quadratic forms are independent

$$\text{i.f. } (\underline{I} - \underline{H})\underline{H} = \underline{0}. \quad (\underline{I} - \underline{H})\underline{H} = \underline{H} - \underline{H}^2 = \underline{H} - \underline{H} = \underline{0}.$$

Hence they are independent.

$$\text{Also: } \frac{(\underline{Y} - \underline{X}\beta)'(\underline{I} - \underline{H})(\underline{Y} - \underline{X}\beta)}{\sigma^2} = \frac{Y'(\underline{I} - \underline{H})Y}{\sigma^2} \text{ is } \chi^2(\text{rank}(\underline{I} - \underline{H})), \text{ i.e. } \chi^2(6)$$

If  $\beta_1 = \beta_2 = \beta_3 = 0$ .  $E(\underline{y}) = 0$ . The  $y$ 's are independent and  $\text{Var}(y_{i,j}) = \sigma^2$ ,  $\forall i, j$ . Hence  $\frac{\underline{y}^t H \underline{y}}{\sigma^2} \sim \chi^2(\text{rank}(H)) \sim \chi^2(3)$

e)

$$H_0^1: \mu_1 = \mu_2 = \mu_3$$

$$H_1^1: \mu_i \neq \mu_j \text{ for at least one } i, j$$

This is the normal  $F$ -test in a one-way anova.

$$F = \frac{SSA/2}{SSE/6} = \frac{\underline{y}^t (\underline{J} - \frac{1}{9} \underline{J} \underline{J}) \underline{y} / 2}{\underline{y}^t (\underline{I} - \underline{J} \underline{J}) \underline{y} / 6} \sim F_{2,6}$$

From theory we know that  $\frac{SSA}{\sigma^2}$  and  $\frac{SSE}{\sigma^2}$  are independent and  $\chi^2(\text{rank}(\underline{J} - \frac{1}{9} \underline{J} \underline{J}))$  and  $\chi^2(\text{rank}(\underline{I} - \underline{J} \underline{J}))$  respectively.

Therefore the distribution. From 2a we know that  $H_0^1$  will be rejected.

$$H_0^2: \mu_1 = \mu_2 = \mu_3 = 0 \Leftrightarrow \beta_1 = \beta_2 = \beta_3 = 0, \quad H_1^2: \text{at least one } \mu_i \neq 0$$

From the discussion in 2c and 2d we know that

$$\text{under } H_0^2: \mu_1 = \mu_2 = \mu_3 = 0$$

$$\frac{\underline{y}^t H \underline{y} / 3}{\underline{y}^t (\underline{I} - H) \underline{y} / 6} \sim F_{3,6}$$

From the output  $P(F_{3,6} \geq F_{obs})$

$$= 5.482 \cdot 10^{-10}, \text{ Hence } H_0^2 \text{ is rejected.}$$

$$SS_R = \underline{y}^t (H - \frac{1}{9} \underline{J} \underline{J}) \underline{y} = \underline{y}^t (\underline{J} - \frac{1}{9} \underline{J} \underline{J}) \underline{y} = 90.03 \text{ from R-output before 2a}$$

$$SST = SS_R + SSE = 90.03 + 17.67 = 107.70,$$

$$R^2 = \frac{90.03}{107.70} \approx \underline{0.84}$$