

Canonical Correlation

Stat 571

11/12/13

①

Start with 2 sets of variables

$$X_1, \dots, X_p \text{ and } Y_1, \dots, Y_q$$

$$\text{let } W = a_1 X_1 + \dots + a_p X_p = \vec{a}' \vec{X}$$

$$\text{and } Z = b_1 Y_1 + \dots + b_q Y_q = \vec{b}' \vec{Y}$$

Goal: find \vec{a} and \vec{b} that maximize the correlation between W and Z .

$$\text{Let } \begin{bmatrix} \vec{X} \\ \vec{Y} \end{bmatrix} = \begin{bmatrix} X_1 \\ \vdots \\ X_p \\ Y_1 \\ \vdots \\ Y_q \end{bmatrix}_{(p+q) \times 1}$$

$$\text{Let } \Sigma = \begin{bmatrix} \Sigma_{xx} & \Sigma_{xy} \\ \Sigma_{yx} & \Sigma_{yy} \end{bmatrix}$$

(2) $\begin{matrix} p \times p & p \times q \\ \uparrow & \uparrow \\ \Sigma_{xx} & \Sigma_{xy} \\ \uparrow & \uparrow \\ \Sigma_{yx} & \Sigma_{yy} \end{matrix}$

$$\text{Corr}(W, Z) = \frac{\text{Cov}(W, Z)}{\sqrt{V(W) V(Z)}}$$

$$\begin{aligned} \text{Cov}(W, Z) &= \text{Cov}(\vec{a}' \vec{X}, \vec{b}' \vec{Y}) = \vec{a}' \text{Cov}(\vec{X}, \vec{Y}) \vec{b} \\ &= \vec{a}' \Sigma_{xy} \vec{b} \end{aligned}$$

$$V(W) = V(\vec{a}' \vec{X}) = \vec{a}' \text{Cov}(\vec{X}) \vec{a} \\ = \vec{a}' \Sigma_{xx} \vec{a}$$

$$V(Z) = V(\vec{b}' \vec{Y}) = \vec{b}' \text{Cov}(\vec{Y}) \vec{b} \\ = \vec{b}' \Sigma_{yy} \vec{b}$$

$$\text{Corr}(W, Z) = \frac{\vec{a}' \Sigma_{xy} \vec{b}}{\sqrt{(\vec{a}' \Sigma_{xx} \vec{a})(\vec{b}' \Sigma_{yy} \vec{b})}}$$

Without loss of generality, we can choose \vec{a} and \vec{b} such that $\vec{a}' \Sigma_{xx} \vec{a} = \vec{b}' \Sigma_{yy} \vec{b} = 1$

Restatement of our goal:

Maximize $r = \vec{a}' \Sigma_{xy} \vec{b}$, subject to the constraints $\vec{a}' \Sigma_{xx} \vec{a} = \vec{b}' \Sigma_{yy} \vec{b} = 1$

Use Lagrange multipliers:

$$\text{Let } F(\vec{a}, \vec{b}, \delta_1, \delta_2)$$

$$= \vec{a}' \Sigma_{xy} \vec{b} - \delta_1 (\vec{a}' \Sigma_{xx} \vec{a} - 1) - \delta_2 (\vec{b}' \Sigma_{yy} \vec{b} - 1)$$

$$\textcircled{1} \quad \frac{\partial F}{\partial \vec{a}} = \vec{a} \cdot \vec{b} - \delta_1 2 \vec{a} \cdot \vec{a} \stackrel{\text{set}}{=} \vec{0}$$

⑤

$$\textcircled{2} \quad \frac{\partial F}{\partial \vec{b}} = \vec{a} \cdot \vec{b} - \delta_2 2 \vec{b} \cdot \vec{b} \stackrel{\text{set}}{=} \vec{0}$$

$$\textcircled{3} \quad \frac{\partial F}{\partial \delta_1} = -(\vec{a}' \cdot \vec{a} - 1) \stackrel{\text{set}}{=} 0$$

$$\textcircled{4} \quad \frac{\partial F}{\partial \delta_2} = -(\vec{b}' \cdot \vec{b} - 1) \stackrel{\text{set}}{=} 0$$

$$\textcircled{1} \quad \vec{a} \cdot \vec{b} = 2\delta_1 \vec{a} \cdot \vec{a}$$

⑥

$$\vec{a}' \cdot \vec{b} = 2\delta_1 \vec{a}' \cdot \vec{a} = 2\delta_1$$

$$\textcircled{2} \quad \vec{a} \cdot \vec{b} = 2\delta_2 \vec{b} \cdot \vec{b}$$

$$\vec{a}' \cdot \vec{b} = 2\delta_2 \vec{b}' \cdot \vec{b}$$

$$\vec{a}' \cdot \vec{b} = 2\delta_2 \vec{b}' \cdot \vec{b} = 2\delta_2$$

$$\therefore \delta_1 = \delta_2 = \frac{1}{2} \vec{a}' \cdot \vec{b}$$

(7)

Back to ①:

$$\mathbf{K}_{xy} \vec{b} = 2\delta_1 \mathbf{K}_{xx} \vec{a}$$

$$\mathbf{K}_{xx}^{-1} \mathbf{K}_{xy} \vec{b} = 2\delta_1 \underbrace{\mathbf{K}_{xx}^{-1} \mathbf{K}_{xx}}_{\mathbf{I}} \vec{a}$$

$$\vec{a} = \frac{1}{2\delta_1} \mathbf{K}_{xx}^{-1} \mathbf{K}_{xy} \vec{b}$$

Substitute into ②:

$$\mathbf{K}_{yx} \left(\frac{1}{2\delta_1} \mathbf{K}_{xx}^{-1} \mathbf{K}_{xy} \vec{b} \right) = 2\delta_2 \mathbf{K}_{yy} \vec{b}$$

$$\mathbf{K}_{yy}^{-1} \mathbf{K}_{yx} \mathbf{K}_{xx}^{-1} \mathbf{K}_{xy} \vec{b} = \underbrace{4\delta_1 \delta_2}_{4\delta^2} \underbrace{\mathbf{K}_{yy}^{-1} \mathbf{K}_{yy}}_{\mathbf{I}} \vec{b} \quad (8)$$

$$\underbrace{(\mathbf{K}_{yy}^{-1} \mathbf{K}_{yx} \mathbf{K}_{xx}^{-1} \mathbf{K}_{xy})}_M - \underbrace{4\delta^2}_{\lambda} \mathbf{I} \vec{b} = \vec{0}$$

$$(M - \lambda \mathbf{I}) \vec{b} = \vec{0}$$

That is, \vec{b} is the eigenvector of M , corresponding to the eigenvalue λ .

$$\begin{aligned}\text{But } \lambda &= 4\delta_1^2 = 4\left(\frac{1}{2}\vec{a}'\Sigma_{xy}\vec{b}\right)^2 \\ &= (\vec{a}'\Sigma_{xy}\vec{b})^2 \\ &= r^2\end{aligned}$$

⑨

Since our goal was to maximize r ,
take λ to be λ_1 , the largest eigenvalue
of M .

Continuation of the process: Once w_1 and z_1 are
found, find w_2 & z_2 that have the highest possible
correlation, subject to being uncorrelated with w_1 & z_1 .

That is, maximize

$$r = \frac{\vec{a}_2' \Sigma_{xy} \vec{b}_2}{\sqrt{(\vec{a}_2' \Sigma_{xx} \vec{a}_2)(\vec{b}_2' \Sigma_{yy} \vec{b}_2)}}$$

⑩

Subject to: $\vec{a}_2' \Sigma_{xx} \vec{a}_2 = \vec{b}_2' \Sigma_{yy} \vec{b}_2 = 1$ and

$$\vec{a}_2' \Sigma_{xx} \vec{a}_1 = \vec{a}_2' \Sigma_{xy} \vec{b}_1 = \vec{b}_2' \Sigma_{xy} \vec{b}_1 = \vec{b}_2' \Sigma_{yx} \vec{a}_1 = 0$$

Solution is again found by Lagrange multipliers,

and it says to select λ_2 , the 2nd-largest eigenvalue
of M , and its eigenvector \vec{b}_2 .

$$H_0: \lambda_k = \lambda_{k+1} = \dots = \lambda_s = 0 \quad (s = \min(p, q)) \quad (11)$$

Bartlett's Test let $\Lambda_k = \prod_{i=k}^s (1 - \lambda_i)$

and $V_k = -[n-1 - \frac{1}{2}(p+q+1)] \ln \Lambda_k$

Then $V_k \sim \chi^2$ df = $(p-k+1)(q-k+1)$

Rao's Test let $F = \frac{v_2}{v_1} \left[\frac{1 - \Lambda_k^{v_b}}{\Lambda_k^{v_b}} \right],$

where $a = n-1 - \frac{1}{2}(p+q+1),$

$$b = \sqrt{\frac{(p-k+1)^2 (q-k+1)^2 - 4}{(p-k+1)^2 + (q-k+1)^2 - 5}} \quad (12)$$

$v_1 = (p-k+1)(q-k+1), \quad v_2 = ab - \frac{1}{2}v_1 + 1$

$$F \sim F_{v_1, v_2}$$

SAS: PROC CANCORR (PSP)

Minitab: non existent

R : CANCORR?
(CA)

SPSS: CANCORR
Statement
(Syntax file)

HW #5

p. 573 #10.18

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(13)