

Canonical Correlation Analysis (CCA)

- Recall: Regression analysis --- concerned with the relationship between a *single* response and a set of predictors

$$\mathbf{X}_{(n \times p)} = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1p} \\ x_{21} & x_{22} & \cdots & x_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ x_{n1} & x_{n2} & \cdots & x_{np} \end{bmatrix}$$

$$\triangleright Y = \beta_0 + \beta_1 X_1 + \cdots + \beta_k X_k + \epsilon$$

$\triangleright$ connection between regression model and conditional normal distribution

$$\triangleright R^2 = \text{cor}(Y, \hat{Y})^2$$

- **Q**: What if we have *more than one* response?

$$\mathbf{X}_{(n \times p)} = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1p} \\ x_{21} & x_{22} & \cdots & x_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ x_{n1} & x_{n2} & \cdots & x_{np} \end{bmatrix}$$

$\triangleright$ Examples

- variables related to arithmetic power & variables related to reading power
- variables related to governmental policy & variables related to economic goal
- variables related to college performance & variables related to precollege achievement
- $\triangleright$ CCA seeks to identify and quantify the linear associations between *two sets* of variables

NTHU STAT 5191, 2010, Lecture Notes
made by S.-W. Cheng (NTHU, Taiwan)

- Population Canonical Variables and Canonical Correlations

$\triangleright$ Data: two groups of variables

The first group, of p variables, is represented by the $(p \times 1)$ random vector $\mathbf{X}^{(1)}$. The second group, of q variables, is represented by the $(q \times 1)$ random vector $\mathbf{X}^{(2)}$. We assume, in the theoretical development, that $\mathbf{X}^{(1)}$ represents the *smaller* set, so that $p \leq q$.

$\triangleright$ Some notations

$$\mathbf{X}_{((p+q) \times 1)} = \begin{bmatrix} \mathbf{X}^{(1)} \\ \mathbf{X}^{(2)} \end{bmatrix} = \begin{bmatrix} X_1^{(1)} \\ X_2^{(1)} \\ \vdots \\ X_p^{(1)} \\ X_1^{(2)} \\ X_2^{(2)} \\ \vdots \\ X_q^{(2)} \end{bmatrix}$$

$$\mathbf{\mu}_{((p+q) \times 1)} = E(\mathbf{X}) = \begin{bmatrix} E(\mathbf{X}^{(1)}) \\ E(\mathbf{X}^{(2)}) \end{bmatrix} = \begin{bmatrix} \mathbf{\mu}^{(1)} \\ \mathbf{\mu}^{(2)} \end{bmatrix}$$

$$\mathbf{\Sigma}_{(p+q) \times (p+q)} = E(\mathbf{X} - \mathbf{\mu})(\mathbf{X} - \mathbf{\mu})' = \begin{bmatrix} \mathbf{\Sigma}_{11} & \mathbf{\Sigma}_{12} \\ \mathbf{\Sigma}_{21} & \mathbf{\Sigma}_{22} \end{bmatrix}$$

$$\begin{matrix} \mathbf{\Sigma}_{11} & \mathbf{\Sigma}_{12} \\ (p \times p) & (p \times q) \\ \mathbf{\Sigma}_{21} & \mathbf{\Sigma}_{22} \\ (q \times p) & (q \times q) \end{matrix}$$

$\triangleright$ the pq elements of $\mathbf{\Sigma}_{12}$ measure the association between the two sets.

- p and q are relatively large, hopeless to interpret elements of $\mathbf{\Sigma}_{12}$ collectively
- moreover, it's often linear combinations of variables that are interesting and useful for predictive or comparative purposes

➤ CCA finds linear functions of one set of variables that maximally correlated with linear function of the other set of variables

▪ Set $U = \mathbf{a}'\mathbf{X}^{(1)}$ and $V = \mathbf{b}'\mathbf{X}^{(2)}$ for some pair of coefficient vectors $\mathbf{a}$ and $\mathbf{b}$.

▪ $\text{Var}(U) = \mathbf{a}' \text{Cov}(\mathbf{X}^{(1)}) \mathbf{a} = \mathbf{a}' \boldsymbol{\Sigma}_{11} \mathbf{a}$

$\text{Var}(V) = \mathbf{b}' \text{Cov}(\mathbf{X}^{(2)}) \mathbf{b} = \mathbf{b}' \boldsymbol{\Sigma}_{22} \mathbf{b}$

$\text{Cov}(U, V) = \mathbf{a}' \text{Cov}(\mathbf{X}^{(1)}, \mathbf{X}^{(2)}) \mathbf{b} = \mathbf{a}' \boldsymbol{\Sigma}_{12} \mathbf{b}$

▪ We shall seek coefficient vectors $\mathbf{a}$ and $\mathbf{b}$ such that

$$\text{Corr}(U, V) = \frac{\mathbf{a}' \boldsymbol{\Sigma}_{12} \mathbf{b}}{\sqrt{\mathbf{a}' \boldsymbol{\Sigma}_{11} \mathbf{a}} \sqrt{\mathbf{b}' \boldsymbol{\Sigma}_{22} \mathbf{b}}} \quad (\diamond)$$

is as large as possible.

▪ The *first pair of canonical variables*, or *first canonical variate pair*, is the pair of linear combinations U_1, V_1 having unit variances, which maximize the correlation ($\diamond$)

▪ The *second pair of canonical variables*, or *second canonical variate pair*, is the pair of linear combinations U_2, V_2 having unit variances, which maximize the correlation ($\diamond$) among all choices that are uncorrelated with the first pair of canonical variables.

▪ The *kth pair of canonical variables*, or *kth canonical variate pair*, is the pair of linear combinations U_k, V_k having unit variances, which maximize the correlation ($\diamond$) among all choices uncorrelated with the previous $k - 1$ canonical variable pairs.

NTHU STAT 5191, 2010, Lecture Notes
made by S.-W. Cheng (NTHU, Taiwan)

▪ **Result 10.1.** Suppose $p \leq q$ and let the random vectors $\mathbf{X}^{(1)}_{(p \times 1)}$ and $\mathbf{X}^{(2)}_{(q \times 1)}$ have $\text{Cov}(\mathbf{X}^{(1)}) = \boldsymbol{\Sigma}_{11}_{(p \times p)}$, $\text{Cov}(\mathbf{X}^{(2)}) = \boldsymbol{\Sigma}_{22}_{(q \times q)}$ and $\text{Cov}(\mathbf{X}^{(1)}, \mathbf{X}^{(2)}) = \boldsymbol{\Sigma}_{12}_{(p \times q)}$, where $\boldsymbol{\Sigma}$ has full rank. For coefficient vectors $\mathbf{a}_{(p \times 1)}$ and $\mathbf{b}_{(q \times 1)}$, form the linear combinations $U = \mathbf{a}'\mathbf{X}^{(1)}$ and $V = \mathbf{b}'\mathbf{X}^{(2)}$. Then

$$\max_{\mathbf{a}, \mathbf{b}} \text{Corr}(U, V) = \rho_1^*$$

attained by the linear combinations (first canonical variate pair)

$$U_1 = \underbrace{\mathbf{e}_1' \boldsymbol{\Sigma}_{11}^{-1/2}}_{\mathbf{a}_1'} \mathbf{X}^{(1)} \quad \text{and} \quad V_1 = \underbrace{\mathbf{f}_1' \boldsymbol{\Sigma}_{22}^{-1/2}}_{\mathbf{b}_1'} \mathbf{X}^{(2)}$$

The *kth* pair of canonical variates, $k = 2, 3, \dots, p$,

$$U_k = \mathbf{e}_k' \boldsymbol{\Sigma}_{11}^{-1/2} \mathbf{X}^{(1)} \quad V_k = \mathbf{f}_k' \boldsymbol{\Sigma}_{22}^{-1/2} \mathbf{X}^{(2)}$$

maximizes

$$\text{Corr}(U_k, V_k) = \rho_k^*$$

among those linear combinations uncorrelated with the preceding $1, 2, \dots, k - 1$ canonical variables.

Here $\rho_1^{*2} \geq \rho_2^{*2} \geq \dots \geq \rho_p^{*2}$ are the eigenvalues of $\boldsymbol{\Sigma}_{11}^{-1/2} \boldsymbol{\Sigma}_{12} \boldsymbol{\Sigma}_{22}^{-1} \boldsymbol{\Sigma}_{21} \boldsymbol{\Sigma}_{11}^{-1/2}$, and $\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_p$ are the associated $(p \times 1)$ eigenvectors. [The quantities $\rho_1^{*2}, \rho_2^{*2}, \dots, \rho_p^{*2}$ are also the p largest eigenvalues of the matrix $\boldsymbol{\Sigma}_{22}^{-1/2} \boldsymbol{\Sigma}_{21} \boldsymbol{\Sigma}_{11}^{-1} \boldsymbol{\Sigma}_{12} \boldsymbol{\Sigma}_{22}^{-1/2}$ with corresponding $(q \times 1)$ eigenvectors $\mathbf{f}_1, \mathbf{f}_2, \dots, \mathbf{f}_p$. Each $\mathbf{f}_i$ is proportional to $\boldsymbol{\Sigma}_{22}^{-1/2} \boldsymbol{\Sigma}_{21} \boldsymbol{\Sigma}_{11}^{-1/2} \mathbf{e}_i$.]

- **Q**: why $\Sigma_{11}^{-1/2} \Sigma_{12} \Sigma_{22}^{-1} \Sigma_{21} \Sigma_{11}^{-1/2}$ (or $\Sigma_{22}^{-1/2} \Sigma_{21} \Sigma_{11}^{-1} \Sigma_{12} \Sigma_{22}^{-1/2}$)?

- geometrical interpretation

Let $\mathbf{A} = [\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_p]'$ and $\mathbf{B} = [\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_q]'$, so that the vectors of canonical variables are

$$\underset{(p \times 1)}{\mathbf{U}} = \underset{(p \times p)}{\mathbf{A}} \underset{(q \times 1)}{\mathbf{X}}^{(1)} \quad \underset{(q \times 1)}{\mathbf{V}} = \underset{(q \times q)}{\mathbf{B}} \underset{(q \times 1)}{\mathbf{X}}^{(2)}$$

$\mathbf{A} = \mathbf{E}' \Sigma_{11}^{-1/2} = \mathbf{E}' \mathbf{P}_1 \Lambda_1^{-1/2} \mathbf{P}_1'$ where $\mathbf{E}'$ is an orthogonal matrix with row $\mathbf{e}_i'$

NTHU STAT 5191, 2010, Lecture Notes
made by S.-W. Cheng (NTHU, Taiwan)

- The canonical variates have the properties

$$\text{Var}(U_k) = \text{Var}(V_k) = 1$$

$$\text{Cov}(U_k, U_\ell) = \text{Corr}(U_k, U_\ell) = 0 \quad k \neq \ell$$

$$\text{Cov}(V_k, V_\ell) = \text{Corr}(V_k, V_\ell) = 0 \quad k \neq \ell$$

$$\text{Cov}(U_k, V_\ell) = \text{Corr}(U_k, V_\ell) = 0 \quad k \neq \ell$$

for $k, \ell = 1, 2, \dots, p$.

- $\text{Cov}(\mathbf{U}, \mathbf{X}^{(1)}) = \text{Cov}(\mathbf{A}\mathbf{X}^{(1)}, \mathbf{X}^{(1)}) = \mathbf{A}\Sigma_{11}$

$$\begin{aligned} \underset{(p \times p)}{\boldsymbol{\rho}_{\mathbf{U}, \mathbf{X}^{(1)}}} &= \text{Corr}(\mathbf{U}, \mathbf{X}^{(1)}) = \text{Cov}(\mathbf{U}, \mathbf{V}_{11}^{-1/2} \mathbf{X}^{(1)}) = \text{Cov}(\mathbf{A}\mathbf{X}^{(1)}, \mathbf{V}_{11}^{-1/2} \mathbf{X}^{(1)}) \\ &= \mathbf{A}\Sigma_{11} \mathbf{V}_{11}^{-1/2} \end{aligned}$$

Similar calculations for the pairs $(\mathbf{U}, \mathbf{X}^{(2)})$, $(\mathbf{V}, \mathbf{X}^{(2)})$ and $(\mathbf{V}, \mathbf{X}^{(1)})$ yield

$$\begin{aligned} \underset{(p \times p)}{\boldsymbol{\rho}_{\mathbf{U}, \mathbf{X}^{(1)}}} &= \mathbf{A}\Sigma_{11} \mathbf{V}_{11}^{-1/2} & \underset{(q \times q)}{\boldsymbol{\rho}_{\mathbf{V}, \mathbf{X}^{(2)}}} &= \mathbf{B}\Sigma_{22} \mathbf{V}_{22}^{-1/2} \\ \underset{(p \times q)}{\boldsymbol{\rho}_{\mathbf{U}, \mathbf{X}^{(2)}}} &= \mathbf{A}\Sigma_{12} \mathbf{V}_{22}^{-1/2} & \underset{(q \times p)}{\boldsymbol{\rho}_{\mathbf{V}, \mathbf{X}^{(1)}}} &= \mathbf{B}\Sigma_{21} \mathbf{V}_{11}^{-1/2} \end{aligned}$$

- To ease the computation burden, many people prefer to get the canonical correlations from

$$|\Sigma_{11}^{-1}\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21} - \rho^{*2}\mathbf{I}| = 0$$

The coefficient vectors **a** and **b** follow directly from the eigenvector equations

$$\Sigma_{11}^{-1}\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}\mathbf{a} = \rho^{*2}\mathbf{a}$$

$$\Sigma_{22}^{-1}\Sigma_{21}\Sigma_{11}^{-1}\Sigma_{12}\mathbf{b} = \rho^{*2}\mathbf{b}$$

➤ Interpreting the population canonical variables

- identify the canonical variables

- ◆ the linear combinations can be interpreted much as in principal components
- ◆ it is helpful to compute the correlation between the canonical variables and the original variables as way to determine the relative important of original variables to a particular canonical variable

- canonical correlation

- ◆ canonical correlation generalizes the correlation between 2 variables

$$|\text{Corr}(\mathbf{X}_i^{(1)}, \mathbf{X}_k^{(2)})| = |\text{Corr}(\mathbf{a}'\mathbf{X}^{(1)}, \mathbf{b}'\mathbf{X}^{(2)})| \leq \max_{\mathbf{a}, \mathbf{b}} \text{Corr}(\mathbf{a}'\mathbf{X}^{(1)}, \mathbf{b}'\mathbf{X}^{(2)}) = \rho_1^*$$

- ◆ An R^2 type statistic can be obtained to explain the total variance explained by a given set of canonical variables:

Because of its multiple correlation coefficient interpretation, the k th *squared* canonical correlation ρ_k^{*2} is the proportion of the variance of canonical variate U_k “explained” by the set $\mathbf{X}^{(2)}$. It is also the proportion of the variance of canonical variate V_k “explained” by the set $\mathbf{X}^{(1)}$.

NTHU STAT 5191, 2010, Lecture Notes
made by S.-W. Cheng (NTHU, Taiwan)

- first r canonical variables as a summary of variability

- ◆ remember that the linear combinations are chosen to maximize correlation between the canonical variables
- ◆ these linear combinations may be quite different from those obtained from principal component within a particular set of variables
- ◆ the canonical may not necessarily explain the total variation within a set of original variables

➤ CCA for standardized data

- If the original variables are standardized with $\mathbf{Z}^{(1)} = [Z_1^{(1)}, Z_2^{(1)}, \dots, Z_p^{(1)}]'$ and $\mathbf{Z}^{(2)} = [Z_1^{(2)}, Z_2^{(2)}, \dots, Z_q^{(2)}]'$, from first principles, the canonical variates are of the form

$$U_k = \mathbf{a}_k' \mathbf{Z}^{(1)} = \mathbf{e}_k' \boldsymbol{\rho}_{11}^{-1/2} \mathbf{Z}^{(1)}$$

$$V_k = \mathbf{b}_k' \mathbf{Z}^{(2)} = \mathbf{f}_k' \boldsymbol{\rho}_{22}^{-1/2} \mathbf{Z}^{(2)}$$

Here, $\text{Cov}(\mathbf{Z}^{(1)}) = \boldsymbol{\rho}_{11}$, $\text{Cov}(\mathbf{Z}^{(2)}) = \boldsymbol{\rho}_{22}$, $\text{Cov}(\mathbf{Z}^{(1)}, \mathbf{Z}^{(2)}) = \boldsymbol{\rho}_{12} = \boldsymbol{\rho}_{21}'$, and $\mathbf{e}_k$ and $\mathbf{f}_k$ are the eigenvectors of $\boldsymbol{\rho}_{11}^{-1/2} \boldsymbol{\rho}_{12} \boldsymbol{\rho}_{22}^{-1} \boldsymbol{\rho}_{21} \boldsymbol{\rho}_{11}^{-1/2}$ and $\boldsymbol{\rho}_{22}^{-1/2} \boldsymbol{\rho}_{21} \boldsymbol{\rho}_{11}^{-1} \boldsymbol{\rho}_{12} \boldsymbol{\rho}_{22}^{-1/2}$, respectively. The canonical correlations, ρ_k^* , satisfy

$$\text{Corr}(U_k, V_k) = \rho_k^*, \quad k = 1, 2, \dots, p$$

where $\rho_1^{*2} \geq \rho_2^{*2} \geq \dots \geq \rho_p^{*2}$ are the nonzero eigenvalues of the matrix $\boldsymbol{\rho}_{11}^{-1/2} \boldsymbol{\rho}_{12} \boldsymbol{\rho}_{22}^{-1} \boldsymbol{\rho}_{21} \boldsymbol{\rho}_{11}^{-1/2}$ (or, equivalently, the largest eigenvalues of $\boldsymbol{\rho}_{22}^{-1/2} \boldsymbol{\rho}_{21} \boldsymbol{\rho}_{11}^{-1} \boldsymbol{\rho}_{12} \boldsymbol{\rho}_{22}^{-1/2}$).

- the canonical correlations are unchanged by the standardization (cf. PCA)