

# STAT 3690 Lecture 29

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## CCA (Con'd)

- Sample version
  - $(\mathbf{Y}_1, \mathbf{X}_1), \dots, (\mathbf{Y}_n, \mathbf{X}_n) \stackrel{\text{iid}}{\sim} (\mathbf{Y}, \mathbf{X})$ 
    - \*  $\mathbf{Y}_i$  and  $\mathbf{X}_i$  jointly sampled
    - \*  $p \leq q < n$
  - $n \times p$  matrix  $\mathbb{Y} = [\mathbf{Y}_1, \dots, \mathbf{Y}_n]^\top$  and  $n \times q$  matrix  $\mathbb{X} = [\mathbf{X}_1, \dots, \mathbf{X}_n]^\top$
  - Sample covariance matrices
    - \*  $\mathbf{S}_\mathbf{Y} = (n-1)^{-1} \sum_i (\mathbf{Y}_i - \bar{\mathbf{Y}})(\mathbf{Y}_i - \bar{\mathbf{Y}})^\top$
    - \*  $\mathbf{S}_\mathbf{X} = (n-1)^{-1} \sum_i (\mathbf{X}_i - \bar{\mathbf{X}})(\mathbf{X}_i - \bar{\mathbf{X}})^\top$
    - \*  $\mathbf{S}_{\mathbf{YX}} = \mathbf{S}_{\mathbf{XY}}^\top = (n-1)^{-1} \sum_i (\mathbf{Y}_i - \bar{\mathbf{Y}})(\mathbf{X}_i - \bar{\mathbf{X}})^\top$
  - Vocabulary
    - \* (The  $k$ th pair of) sample canonical directions:  $(\hat{\mathbf{a}}_k \in \mathbb{R}^p, \hat{\mathbf{b}}_k \in \mathbb{R}^q)$
    - \* (The  $k$ th pair of) sample canonical variates:  $(\mathbb{Y}_C \hat{\mathbf{a}}_k, \mathbb{X}_C \hat{\mathbf{b}}_k)$
    - \* (The  $k$ th) canonical correlation:  $\hat{\rho}_k$
  - Goal: find  $\hat{\mathbf{a}}_k$  and  $\hat{\mathbf{b}}_k$ ,  $k = 1, \dots, r \leq p$ , to maximize

$$\hat{\rho}_k = \frac{\hat{\mathbf{a}}_k^\top \mathbf{S}_{\mathbf{YX}} \hat{\mathbf{b}}_k}{\sqrt{\hat{\mathbf{a}}_k^\top \mathbf{S}_\mathbf{Y} \hat{\mathbf{a}}_k} \sqrt{\hat{\mathbf{b}}_k^\top \mathbf{S}_\mathbf{X} \hat{\mathbf{b}}_k}}$$

subject to

- \*  $\hat{\mathbf{a}}_k^\top \mathbf{S}_\mathbf{Y} \hat{\mathbf{a}}_k = 1$
- \*  $\hat{\mathbf{b}}_k^\top \mathbf{S}_\mathbf{X} \hat{\mathbf{b}}_k = 1$
- \*  $\hat{\mathbf{a}}_k^\top \mathbf{S}_\mathbf{Y} \hat{\mathbf{a}}_\ell = 0, \ell = 1, \dots, k-1$
- \*  $\hat{\mathbf{a}}_k^\top \mathbf{S}_{\mathbf{YX}} \hat{\mathbf{b}}_\ell = 0, \ell = 1, \dots, k-1$
- \*  $\hat{\mathbf{b}}_k^\top \mathbf{S}_\mathbf{X} \hat{\mathbf{b}}_\ell = 0, \ell = 1, \dots, k-1$
- \*  $\hat{\mathbf{b}}_k^\top \mathbf{S}_{\mathbf{XY}} \hat{\mathbf{a}}_\ell = 0, \ell = 1, \dots, k-1$
- Solution: Let  $\widehat{\mathbf{M}} = \mathbf{S}_\mathbf{Y}^{-1/2} \mathbf{S}_{\mathbf{YX}} \mathbf{S}_\mathbf{X}^{-1/2}$ 
  - \*  $\hat{\rho}_k = \sqrt{\hat{\lambda}_k}$  is the  $k$ th largest singular value of  $\widehat{\mathbf{M}}$ 
    - $\hat{\lambda}_k$ : the  $k$ th largest eigenvalue of  $\widehat{\mathbf{M}} \widehat{\mathbf{M}}^\top$  (or  $\widehat{\mathbf{M}}^\top \widehat{\mathbf{M}}$ )
  - \*  $\hat{\mathbf{a}}_k = \mathbf{S}_\mathbf{Y}^{-1/2} \hat{\mathbf{e}}_k$ 
    - $\hat{\mathbf{e}}_k$ : the left-singular vector corresponding to the  $k$ th largest singular value of  $\widehat{\mathbf{M}}$ , i.e., the eigenvector corresponding to the  $k$ th largest eigenvalue of  $\widehat{\mathbf{M}} \widehat{\mathbf{M}}^\top$
  - \*  $\hat{\mathbf{b}}_k = \mathbf{S}_\mathbf{X}^{-1/2} \hat{\mathbf{f}}_k$ 
    - $\hat{\mathbf{f}}_k$ : the right-singular vector corresponding to the  $k$ th largest singular value of  $\widehat{\mathbf{M}}$ , i.e., the eigenvector corresponding to the  $k$ th largest eigenvalue of  $\widehat{\mathbf{M}}^\top \widehat{\mathbf{M}}$

- Example: olive oil data
    - 572 olive oils
    - 10 features
      - \* **region** indicates the general region (in Italy) of origin.
      - \* **area** details the area of Italy.
      - \* Remaining variables are continuous valued and measure the percentage composition of 8 different fatty acids
    - Interested in the correlations between the region of origin and the fatty acid measurements
      - \*  $\mathbf{Y} \in \mathbb{R}^{572 \times 3}$  an indicator matrix, i.e., each row of  $\mathbf{Y}$  indicates the region with a 1 and otherwise has 0
      - \*  $\mathbf{X} \in \mathbb{R}^{572 \times 8}$  contains the 8 fatty acid measurements
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- Proportion of explained correlation
    - Determining  $r$ , the number of pairs of canonical variates to retain
    - $p \times r$  matrix  $\text{corr}(\mathbf{Y}, \mathbf{A}_r^\top \mathbf{Y})$  and  $q \times r$  matrix  $\text{corr}(\mathbf{X}, \mathbf{B}_r^\top \mathbf{X})$ 
      - \* The correlation matrix between  $\mathbf{Y}$  (or  $\mathbf{X}$ ) and canonical variates
      - \*  $\mathbf{A}_r = [\mathbf{a}_1, \dots, \mathbf{a}_r]$  and  $\mathbf{B}_r = [\mathbf{b}_1, \dots, \mathbf{b}_r]$
    - $\|\text{corr}(\mathbf{Y}, \mathbf{A}_r^\top \mathbf{Y})\|_F^2/p$  and  $\|\text{corr}(\mathbf{X}, \mathbf{B}_r^\top \mathbf{X})\|_F^2/q$ 
      - \* Proportion of explained correlation of  $\mathbf{Y}$  (or  $\mathbf{X}$ )
      - \*  $\|\cdot\|_F^2$ : squared Frobenius norm, i.e., sum of squared entries
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- Interpreting canonical variates
  - $\text{corr}(\mathbf{Y}, \mathbf{A}_r^\top \mathbf{Y})$
  - $\text{corr}(\mathbf{X}, \mathbf{B}_r^\top \mathbf{X})$
  - $\text{corr}(\mathbf{Y}, \mathbf{B}_r^\top \mathbf{X})$
  - $\text{corr}(\mathbf{X}, \mathbf{A}_r^\top \mathbf{Y})$

## Testing the uncorrelatedness of canonical variates

- LRT for  $H_0 : \Sigma_{\mathbf{YX}} = 0$  vs.  $H_1$  : otherwise
    - LRT statistic  $\lambda = \prod_{k=1}^p (1 - \hat{\rho}_k^2)^{n/2}$ 
      - \*  $\hat{\rho}_k$ : the  $k$ th sample canonical correlation
      - \* Under  $H_0$ ,  $-2 \ln \lambda = -n \sum_{k=1}^p \ln(1 - \hat{\rho}_k^2) \approx \chi^2(pq)$
  - Sequential inference
    - Determining  $r$ , the number of pairs of canonical variates to retain
    - Note that  $\Sigma_{\mathbf{YX}} = 0 \Leftrightarrow \rho_1 = \dots = \rho_p = 0 \Leftrightarrow \rho_1 = 0$ 
      - \* Since  $\rho_1 \geq \dots \geq \rho_p$
    - Consider a sequence of  $p$  pairs of hypotheses:  $H_{0,k} : \rho_{k-1} > 0, \rho_k = 0$  vs.  $H_{1,k} : \rho_k > 0$ 
      - \* LRT statistic  $\lambda_k = \prod_{\ell=k}^p (1 - \hat{\rho}_\ell^2)^{n/2}$ 
        - Under  $H_{0,k}$ ,  $-2 \ln \lambda_k = -n \sum_{\ell=k}^p \ln(1 - \hat{\rho}_\ell^2) \approx \chi^2((p-k+1)(q-k+1))$
    - Stopping rules
      - \*  $p_k$ : the  $p$ -value associated with the testing on  $H_{0,k}$  vs.  $H_{1,k}$
      - \*  $p_{(k)}$ : the  $k$ th smallest value among  $\{p_1, \dots, p_p\}$
      - \* Holm-Bonferroni procedure (Holm (1979), Scandinavian Journal of Statistics, 6, 65–70): if  $p_{(k)} < \alpha/(p+1-k)$ , reject  $H_{0,(k)}$  and proceed to larger  $p$ -values; otherwise EXIT.
      - \* B-H procedure (Benjamini & Hochberg (1995), Journal of the Royal Statistical Society, Series B., 57, 289–300):
        1. For a given level  $\alpha$ , find  $k^* = \max\{k \in \{1, \dots, p\} \mid p_{(k)} \leq k\alpha/p\}$
        2. Reject  $H_{0,(k)}$  for  $k = 1, \dots, k^*$
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