

# STAT 3690 Lecture 30

zhiyanggeezhou.github.io

Zhiyang Zhou (zhiyang.zhou@umanitoba.ca)

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## Testing the uncorrelatedness of canonical variates

- LRT for  $H_0 : \Sigma_{\mathbf{Y}\mathbf{X}} = 0$  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  $H_0$ ,  $-2 \ln \lambda = -n \sum_{k=1}^p \ln(1 - \hat{\rho}_k^2) \approx \chi^2(pq)$

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- Sequential inference
  - Determining  $r$ , the number of pairs of canonical variates to retain
  - Note that  $\Sigma_{\mathbf{Y}\mathbf{X}} = 0 \Leftrightarrow \rho_1 = \dots = \rho_p = 0 \Leftrightarrow \rho_1 = 0$ 
    - \* 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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## Summary of CCA

- Dimension reduction method
  - Maximize correlation
  - Treat  $\mathbf{Y}$  and  $\mathbf{X}$  equally/reduce the dimension of both  $\mathbf{Y}$  and  $\mathbf{X}$  simultaneously
- Limitation: in need of invertible  $\Sigma_{\mathbf{Y}}$  and  $\Sigma_{\mathbf{X}}$