1 Pillai's Trace as an M-estimator

Given two sets of residuals $\mathbf{R}^x = \{R_1^x, \dots, R_p^x\}$ and $\mathbf{R}^y = \{R_1^y, \dots, R_q^y\}$.

$$\theta = (\theta^{x}, \theta^{y}, \theta^{xy})$$

$$\theta^{x} = (\theta_{1}^{x}, \dots, \theta_{p}^{x})$$

$$\theta^{y} = (\theta_{1}^{y}, \dots, \theta_{q}^{y})$$

$$\theta^{xy} = (\theta_{11}^{xy}, \dots, \theta_{pq}^{xy})$$
(1)

$$\Psi = \begin{cases} R_{i}^{x} - \theta_{i}^{x} & \forall i \in \{1, \dots, p\} \\ R_{i}^{y} - \theta_{i}^{y} & \forall i \in \{1, \dots, q\} \\ R_{i}^{x} R_{j}^{x} - \theta_{ij}^{xx} & \forall i, j \in \{1, \dots, p\} \\ R_{i}^{y} R_{j}^{y} - \theta_{ij}^{yy} & \forall i, j \in \{1, \dots, q\} \\ R_{i}^{x} R_{j}^{y} - \theta_{ij}^{xy} & \forall i \in \{1, \dots, p\}, j \in \{1, \dots, q\} \end{cases}$$
(2)

$$\hat{\theta}_{i}^{x} = \mathbb{E}[R_{i}^{x}]
\hat{\theta}_{i}^{y} = \mathbb{E}[R_{i}^{y}]
\hat{\theta}_{ij}^{xx} = \mathbb{E}[R_{i}^{x}R_{i}^{x}]
\hat{\theta}_{ij}^{yy} = \mathbb{E}[R_{i}^{y}R_{i}^{y}]
\hat{\theta}_{ij}^{xy} = \mathbb{E}[R_{i}^{x}R_{i}^{y}]$$
(3)

$$\begin{split} &\Sigma_{ij}^{x} = \mathbb{E}[R_{i}^{x}R_{j}^{x}] - \mathbb{E}[R_{i}^{x}]\mathbb{E}[R_{j}^{x}] = \mathbb{E}[R_{i}^{x}R_{j}^{x}] \quad (under \ null) \\ &\Sigma_{ij}^{y} = \mathbb{E}[R_{i}^{y}R_{j}^{y}] - \mathbb{E}[R_{i}^{y}]\mathbb{E}[R_{j}^{y}] = \mathbb{E}[R_{i}^{y}R_{j}^{y}] \quad (under \ null) \\ &\Sigma_{ij}^{xy} = \mathbb{E}[R_{i}^{x}R_{j}^{y}] - \mathbb{E}[R_{i}^{x}]\mathbb{E}[R_{j}^{y}] = 0 \quad (under \ null \ as \ no \ cross \ association) \\ &\Lambda = (\Sigma^{x})^{-\frac{1}{2}} \Sigma^{xy} (\Sigma^{y})^{-\frac{1}{2}} \\ &g(\boldsymbol{\theta}) = \sum_{i=\{1,\cdots,p\}, j=\{1,\cdots,q\}} \Lambda_{i,j}^{2} \quad elementwise\text{-square} \end{split}$$

$$\kappa^{x} = (\Sigma^{x})^{-\frac{1}{2}}$$

$$\kappa^{y} = (\Sigma^{y})^{-\frac{1}{2}}$$

$$\Lambda_{i,j} = \sum_{k=1}^{q} \left[\left(\sum_{l=1}^{p} \kappa_{il}^{x} \Sigma_{lk}^{xy} \right) \kappa_{kj}^{y} \right]$$
(5)

Under null all residual expectations are 0.

$$A(\theta) = \mathbb{E}\left[-\frac{\partial}{\partial \theta} \Psi_i(\theta)\right] = \mathbb{I}_{2(p+q)(pq)}$$

$$B(\theta) = \mathbb{E}\left[\Psi_i(\theta)\Psi_i(\theta)'\right]$$

$$V(\theta) = A(\theta)B(\theta)A(\theta)' = B(\theta)$$
(6)

$$\sqrt{n}(g(\hat{\theta}) - g(\theta)) \to \mathcal{N}(0, \sigma^2)$$

$$\sqrt{n}(g(\hat{\theta})) \to \mathcal{N}(0, \sigma^2)$$

$$\sigma^2 = B \frac{\partial}{\partial \theta} g(\theta) B'$$
(7)

2 Q1 as an M-Estimator

Given two sets of residuals, (X, Y), trying to write:

• Canonical correlation: $\frac{\mathbb{E}[XY]}{\mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]}$

• Q1: $\frac{\mathbb{E}[XY]}{\mathbb{E}[(XY)^2] - \mathbb{E}[XY]^2}$