## **Canonical Correlations**

General goal: explore correlation structure between two sets of variables  $X_1$  and  $X_2$ .

Begin with population definitions.

Assume

$$\mathbf{X} = \left[ \begin{array}{c} \mathbf{X}_1 \\ \mathbf{X}_2 \end{array} \right]$$

where  $\mathbf{X_1}$  has  $p_1$  components and  $\mathbf{X_2}$  has  $p_2$  components.

Partition variance covariance matrix of  ${f X}$  as usual into  $\Sigma_{ij}.$ 

Which linear combination of  $\mathbf{X}_1$  entries is most correlated with which linear combination of  $\mathbf{X}_2$  entries?

Consider vectors a, b.

$$Corr(\mathbf{a}^T \mathbf{X}_1, \mathbf{b}^T \mathbf{X}_2) = \frac{\mathbf{a}^T \mathbf{\Sigma}_{12} \mathbf{b}}{\sqrt{\mathbf{a}^T \mathbf{\Sigma}_{11} \mathbf{a} \mathbf{b}^T \mathbf{\Sigma}_{22} \mathbf{b}}}$$

Scale invariant as function of either a or b.

So: maximize  $\mathbf{a}^T \mathbf{\Sigma}_{12} \mathbf{b}$  subject to two conditions:

$$\mathbf{a}^T \mathbf{\Sigma}_{11i} \mathbf{a} = \mathbf{1} = \mathbf{b}^T \mathbf{\Sigma}_{22} \mathbf{b}.$$

Two Lagrange multipliers gives equations:

$$\Sigma_{12}b = \lambda_1 \Sigma_{11}a$$
$$\Sigma_{21}a = \lambda_2 \Sigma_{22}b$$

Manipulate to get

$$\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}\mathbf{a} = \lambda_1\lambda_2\Sigma_{11}\mathbf{a}$$

Homework problem: maximize  $\mathbf{a}^T \mathbf{x}$  subject to  $\mathbf{x}^T \mathbf{Q} \mathbf{x} = \mathbf{1}$  over  $\mathbf{x}$ .

Equivalent to maximizing

$$\frac{\mathbf{a}^T \mathbf{x}}{\sqrt{\mathbf{x}^T \mathbf{Q} \mathbf{x}}}$$

Solution is

$$x = Q^{-1}a$$

Maximum value is

$$\sqrt{\mathbf{a}^T \mathbf{Q}^{-1} \mathbf{a}}$$

Apply to our problem to maximize over  ${\bf b}$  with a fixed. Get

$$\sqrt{\frac{\mathbf{a}^T \boldsymbol{\Sigma}_{12} \boldsymbol{\Sigma}_{22}^{-1} \boldsymbol{\Sigma}_{21} \mathbf{a}}{\mathbf{a}^T \boldsymbol{\Sigma}_{11} \mathbf{a}}}$$

Generalized eigenvalue problems: suppose  ${\bf A}$ ,  ${\bf B}$  symmetric and suppose  ${\bf B}$  not singular. Find solutions of

$$(\mathbf{A} - \lambda \mathbf{B})\mathbf{v} = 0$$

for non-zero  $\mathbf{v}$ . (Notice solution set is a vector space.)

Write  $B = B^{1/2}B^{1/2}$  for symmetric  $B^{1/2}$ .

Define  $\mathbf{w}=\mathbf{B}^{1/2}\mathbf{v},$  multiply basic equation by  $\mathbf{B}^{-1/2}$  to get

$$B^{-1/2}AB^{-1/2}w = \lambda w$$

Thus solutions are of form  $(\lambda, \mathbf{B}^{-1/2}\mathbf{w})$  where  $(\lambda, \mathbf{w})$  is an eigenvalue-eigenvector pair for

$$B^{-1/2}AB^{-1/2}w$$
.

Equivalently: find eigenvalues of  ${\bf B}^{-1}{\bf A}$  or of  ${\bf AB}^{-1}.$ 

Corresponding maximization problem: maximize

$$\frac{\mathbf{v}^t \mathbf{A} \mathbf{v}}{\mathbf{v}^t \mathbf{B} \mathbf{v}}$$

Maximum value is largest  $\lambda$ .

Application to our problem:

Having maximized over b now must maximize correlation squared by maximizing

$$\frac{\mathbf{a}^T \boldsymbol{\Sigma}_{12} \boldsymbol{\Sigma}_{22}^{-1} \boldsymbol{\Sigma}_{21} \mathbf{a}}{\mathbf{a}^T \boldsymbol{\Sigma}_{11} \mathbf{a}}$$

so 
$$B = \Sigma_{11}$$
 and  $A = \Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}$ .

Maximal squared correlation is largest eigenvalue of

$$\Sigma_{11}^{-1}\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}.$$

First canonical correlates: find largest eigenvalue of

$$\Sigma_{11}^{-1}\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}$$

and  $\mathbf{a}_1$  the corresponding eigenvector. Then

$$\mathbf{b}_1 = \Sigma_{22}^{-1} \Sigma_{21} \mathbf{a}_1$$

Maximal value of squared correlation is largest eigenvalue

Corresponding  $a_1$ ,  $b_1$  are eigenvectors of

$$\Sigma_{11}^{-1}\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}$$

and

$$\Sigma_{22}^{-1}\Sigma_{21}\Sigma_{11}^{-1}\Sigma_{12}$$

Usually normalized.

Second canonical correlates: repeat maximization but require:

Independence of  $\mathbf{a}_2^T\mathbf{X}_1$  and  $\mathbf{a}_1^T\mathbf{X}_1$  and of  $\mathbf{b}_2^T\mathbf{X}_2$  and  $\mathbf{b}_1^T\mathbf{X}_2$ .

And so on to get  $min\{p_1, p_2\}$  triples:

$$\lambda_i, \mathbf{a}_i, \mathbf{b}_i$$

with  $\lambda_i$  in decreasing order.

With data: just replace  $\Sigma$  by S.

## **Canonical Correlations example**

Data: Table 9.12 in Johnson and Wichern.

Sales data: 3 measurements on the sales performance of 50 salespeople for a large firm and 4 test scores.

Use SAS to do canonical correlation analysis between the first 3 variables and the remaining 4.

Here is the SAS code.

And here is the output.

Canonical Correlation Analysis

Adjusted Approx Squared

Canonical Canonical Standard Canonical

Correln Correln Error Correln

1 0.994483 0.994021 0.001572 0.988996

2 0.878107 0.872097 0.032704 0.771071

3 0.383606 0.366795 0.121835 0.147153

Eigenvalues of INV(E)\*H = CanRsq/(1-CanRsq)

E	igenvalue	Difference	Proportion	Cumul
1	89.8745	86.5063	0.9621	0.9621
2	3.3682	3.1956	0.0361	0.9982
3	0.1725	•	0.0018	1.0000

Notice the first canonical correlates account for most of the correlation between the two sets of variables. Canonical Correlation Analysis

Test of HO: The canonical correlations in current row and all that follow are zero
Likelihood

Ratio Approx F Num DF Den DF Pr > F 1 0.00214847 87.3915 12 114.06 0.0001 2 0.19524127 18.5263 6 88 0.0001 3 0.85284669 3.8822 2 45 0.0278 Multivariate Statistics and F Approximations S=3 M=0 N=20.5Statistic Value F NumDF DenDF Pr > F Wilks' 0.0022 87.39 12 114.1 0.0001 Pillai's 1.9072 19.63 12 135 0.0001 Hotel'g-Ly 93.4152 324.35 12 125 0.0001 89.8745 1011.08 4 45 0.0001 Roy's NOTE: F Statistic for Roy's Greatest Root is an upper bound.

Clearly all 3 of the canonical correlations are non-zero. The multivariate tests are of  $H_o$ :  $\Sigma_{12}=0$ .

```
Canonical Correlation Analysis
Raw Canonical Coefficients
       for the 'VAR' Variables
            V1
                    V2
GROWTH 0.06238 -0.17407 -0.37715
PROFIT 0.02093 0.24216 0.10352
NEW
        0.07826 - 0.23829
                           0.38342
Raw Canonical Coefficients
       for the 'WITH' Variables
          W1
                   W2
                            W3
CREATE 0.06975 -0.19239 0.24656
      0.03074 \quad 0.20157 \quad -0.14190
MECH
      0.08956 -0.49576 -0.28022
ABST
      0.06283 0.06832 0.01133
MATH
Canonical Correlation Analysis
Standardized Canonical Coefficients
     for the 'VAR' Variables
          V1
                 V2
                         V3
GROWTH 0.4577 -1.2772 -2.7673
PROFIT 0.2119 2.4517 1.0480
      0.3688 -1.1229 1.8067
NEW
Standardized Canonical Coefficients
     for the 'WITH' Variables
         W1
                 W2
                         W3
CREATE 0.2755 -0.7600 0.9739
MECH 0.1040 0.6823 -0.4803
ABST 0.1916 -1.0607 -0.5996
MATH 0.6621 0.7199 0.1194
```

The standardized coefficients are easiest to interpret.

For instance a quantity which is not too different from the average of the 3 sales indices is strongly correlated with a weighted average of the psychological test scores which puts most of the weight on Math.

The second set of correlates focus on the relation between profitability minus the average of the other two sales indices correlated with (Math + Mech) - (Abstract + Creativity).

The last one is not particularly meaningful to me but it is a very small part of the correlation structure between the two sets of variables. Canonical Structure

Correlations Between the 'VAR' Variables and Their Canonical Variables

V1 V2 V3

GROWTH 0.9799 0.0006 -0.1996

PROFIT 0.9464 0.3229 0.0075

NEW 0.9519 -0.1863 0.2434

Correlations Between the 'WITH' Variables and Their Canonical Variables

W1 W2 W3

CREATE 0.6383 -0.2157 0.6514

MECH 0.7212 0.2376 -0.0677

ABST 0.6472 -0.5013 -0.5742

MATH 0.9441 0.1975 -0.0942

Canonical Structure

Correlations Between the 'VAR' Variables and the Canonical Variables of the 'WITH' Variables

W1 W2 W3

GROWTH 0.9745 0.0006 -0.0766

PROFIT 0.9412 0.2835 0.0029

NEW 0.9466 -0.1636 0.0934

Correlations Between the 'WITH' Variables and the Canonical Variables of the 'VAR' Variables

V1 V2 V3

CREATE 0.6348 -0.1894 0.2499

MECH 0.7172 0.2086 -0.0260

ABST 0.6437 -0.4402 -0.2203

MATH 0.9389 0.1735 -0.0361