

NFL Combine Example – Canonical Correlation Analysis

$\mathbf{X}^{(1)} \equiv$ Body Dimension Variables: Height, Arm Length, Weight, Hand Length

$\mathbf{X}^{(2)} \equiv$ Physical Performance Variables: 40 Yard Time, 225 lb Bench Press Reps, Vertical Jump, Broad Jump

Correlation Matrix: $\mathbf{R} = \begin{bmatrix} \mathbf{R}_{11} & \mathbf{R}_{12} \\ \mathbf{R}_{21} & \mathbf{R}_{22} \end{bmatrix}$

	height	armlen	wt	handlen	time40	bench	vjump	bjump
height	1.0000	0.7644	0.7208	0.4064	0.5895	0.3871	-0.4425	-0.4218
armlen	0.7644	1.0000	0.6347	0.4766	0.4812	0.3333	-0.3828	-0.3514
wt	0.7208	0.6347	1.0000	0.4619	0.8910	0.7123	-0.7103	-0.7512
handlen	0.4064	0.4766	0.4619	1.0000	0.3874	0.3042	-0.2539	-0.2726
time40	0.5895	0.4812	0.8910	0.3874	1.0000	0.5671	-0.7691	-0.8456
bench	0.3871	0.3333	0.7123	0.3042	0.5671	1.0000	-0.4098	-0.4427
vjump	-0.4425	-0.3828	-0.7103	-0.2539	-0.7691	-0.4098	1.0000	0.8432
bjump	-0.4218	-0.3514	-0.7512	-0.2726	-0.8456	-0.4427	0.8432	1.0000

Matrices used to obtain Canonical Variates and (Squared) Correlations:

$\mathbf{R}_{11}^{-1/2} \mathbf{R}_{12} \mathbf{R}_{22}^{-1} \mathbf{R}_{21} \mathbf{R}_{11}^{-1/2}$ and $\mathbf{R}_{22}^{-1/2} \mathbf{R}_{21} \mathbf{R}_{11}^{-1} \mathbf{R}_{12} \mathbf{R}_{22}^{-1/2}$

> CC.mat1

	[,1]	[,2]	[,3]	[,4]
[1,]	0.10717575	0.05303344	0.1853638	0.05396137
[2,]	0.05303344	0.03296692	0.1166684	0.02490461
[3,]	0.18536382	0.11666837	0.7876753	0.14152964
[4,]	0.05396137	0.02490461	0.1415296	0.03989531

> CC.mat2

	[,1]	[,2]	[,3]	[,4]
[1,]	0.4191117	0.2893379	-0.2097069	-0.2010653
[2,]	0.2893379	0.2554468	-0.1607914	-0.2011895
[3,]	-0.2097069	-0.1607914	0.1225738	0.1211551
[4,]	-0.2010653	-0.2011895	0.1211551	0.1705809

Eigenvalues and Eigenvectors of $\mathbf{R}_{11}^{-1/2} \mathbf{R}_{12} \mathbf{R}_{22}^{-1} \mathbf{R}_{21} \mathbf{R}_{11}^{-1/2} \left(\hat{\rho}_k^2, \hat{\mathbf{e}}_k \right)$ and Eigenvectors of $\mathbf{R}_{22}^{-1/2} \mathbf{R}_{21} \mathbf{R}_{11}^{-1} \mathbf{R}_{12} \mathbf{R}_{22}^{-1/2} \left(\hat{\mathbf{f}}_k \right)$:

> eigen(CC.mat1)\$val

[1] 0.881941333 0.073108516 0.011072002 0.001591423

> eigen(CC.mat1)\$vec

	[,1]	[,2]	[,3]	[,4]
[1,]	-0.2476794	0.8492400	-0.006944013	-0.4662597
[2,]	-0.1499372	0.3342368	-0.615766637	0.6975930
[3,]	-0.9404034	-0.3241813	-0.049144939	-0.0901814
[4,]	-0.1783682	0.2489651	0.786363786	0.5365009

> eigen(CC.mat2)\$vec

	[,1]	[,2]	[,3]	[,4]
[1,]	0.6626096	0.65062063	0.277057585	-0.2467397
[2,]	0.5251578	-0.38070053	0.271325808	0.7110968
[3,]	-0.3574770	-0.07740005	0.921708783	-0.1291213
[4,]	-0.3966964	0.65250988	-0.008620857	0.6455916

$$\hat{\rho}_1^* = \sqrt{.88194} = .93912 \quad \hat{\rho}_2^* = \sqrt{.07311} = .27039 \quad \hat{\rho}_3^* = \sqrt{.01107} = .10522 \quad \hat{\rho}_4^* = \sqrt{.00159} = .03989$$

$$\hat{\mathbf{a}}_k = \mathbf{R}_{11}^{-1/2} \hat{\mathbf{e}}_k \quad \hat{\mathbf{b}}_k = \mathbf{R}_{22}^{-1/2} \hat{\mathbf{f}}_k \quad k=1,2,3,4$$

```
> round(AB.out,3)
      a1      b1      a2      b2      a3      b3      a4      b4
[1,]  0.115  0.654  1.321  1.809  0.322  0.744 -1.115 -0.279
[2,]  0.134  0.358  0.050 -0.651 -1.069  0.265  1.220  0.929
[3,] -1.156 -0.076 -0.953 -0.392 -0.084  1.701 -0.152 -0.738
[4,]  0.004 -0.047  0.244  1.847  1.027 -0.467  0.507  1.244
```

Canonical Variates:

$$\hat{U}_k = \hat{\mathbf{a}}_k' \mathbf{z}^{(1)} \quad \hat{V}_k = \hat{\mathbf{b}}_k' \mathbf{z}^{(2)} \quad k=1,2,3,4 \quad V\{\hat{U}_k\} = V\{\hat{V}_k\} = 1 \quad \text{COV}\{\hat{U}_k, \hat{U}_l\} = \text{COV}\{\hat{V}_k, \hat{V}_l\} = \text{COV}\{\hat{U}_k, \hat{V}_l\} = 0$$

$$\hat{\mathbf{A}} = \begin{bmatrix} \hat{\mathbf{a}}_1' \\ \vdots \\ \hat{\mathbf{a}}_p' \end{bmatrix} \quad \hat{\mathbf{B}} = \begin{bmatrix} \hat{\mathbf{b}}_1' \\ \vdots \\ \hat{\mathbf{b}}_q' \end{bmatrix} \quad \hat{\mathbf{U}} = \hat{\mathbf{A}} \mathbf{z}^{(1)} \quad \hat{\mathbf{V}} = \hat{\mathbf{B}} \mathbf{z}^{(2)} \quad \hat{\mathbf{A}}^{-1} = \begin{bmatrix} \hat{\mathbf{a}}^{(1)} & \dots & \hat{\mathbf{a}}^{(p)} \end{bmatrix} \quad \hat{\mathbf{B}}^{-1} = \begin{bmatrix} \hat{\mathbf{b}}^{(1)} & \dots & \hat{\mathbf{b}}^{(q)} \end{bmatrix}$$

$$\text{COV}\{\mathbf{z}^{(1)}, \hat{\mathbf{U}}\} = \text{COV}\{\hat{\mathbf{A}}^{-1} \hat{\mathbf{U}}, \hat{\mathbf{U}}\} = \hat{\mathbf{A}}^{-1} \quad \text{COV}\{\mathbf{z}^{(2)}, \hat{\mathbf{V}}\} = \text{COV}\{\hat{\mathbf{B}}^{-1} \hat{\mathbf{V}}, \hat{\mathbf{V}}\} = \hat{\mathbf{B}}^{-1}$$

```
> round(AB.inv.out,4)
      INV(A)
[1,] -0.6147  0.7717 -0.1389 -0.0862  0.9546  0.1792 -0.0192 -0.2372
[2,] -0.5105  0.5715 -0.3873  0.5126  0.7807 -0.2822  0.1961  0.5219
[3,] -0.9866  0.1439 -0.0566  0.0526 -0.7646  0.0410  0.6265  0.1453
[4,] -0.4198  0.3650  0.6094  0.5650 -0.8220  0.2751  0.2210  0.4471
      INV(B)
```

Proportions of Sample Variances of $\mathbf{Z}^{(1)}$ and $\mathbf{Z}^{(2)}$ explained by first r canonical variates.

$$R^2_{z^{(1)} \bullet \hat{U}_1 \dots \hat{U}_r} = \sum_{i=1}^r \left\{ \left(\hat{\mathbf{A}}^{-1} \right)' \hat{\mathbf{A}}^{-1} \right\}_{i,i} \quad R^2_{z^{(2)} \bullet \hat{V}_1 \dots \hat{V}_r} = \sum_{i=1}^r \left\{ \left(\hat{\mathbf{B}}^{-1} \right)' \hat{\mathbf{B}}^{-1} \right\}_{i,i}$$

```
> round(AiAiBiBi.out,4)
      INV(A)'INV(A)
[1,]  1.7879 -1.0613  0.0831 -0.4977
[2,] -1.0613  1.0761 -0.1142  0.4403
[3,]  0.0831 -0.1142  0.5438  0.1548
[4,] -0.4977  0.4403  0.1548  0.5922
      INV(B)'INV(B)
[1,]  2.7810 -0.3068 -0.5259 -0.2975
[2,] -0.3068  0.1891  0.0277 -0.0608
[3,] -0.5259  0.0277  0.4802  0.2967
[4,] -0.2975 -0.0608  0.2967  0.5496
```

```
> round(prop.out,4)
      z1,r=1  r=2  r=3  r=4  z2,r=1  r=2  r=3  r=4
[1,]  0.447  0.716  0.8519  1  0.6953  0.7425  0.8626  1
```

$$\text{Test whether } \Sigma_{12} = \mathbf{0} \quad TS: -\left(n-1 - \frac{p+q+1}{2} \right) \ln \left(\prod_{i=1}^p \hat{\rho}_i^* \right)$$

```
> round(test.out,4)
      Test Stat DF X2(.05) Pr>TS
[1,] 432.8051 16 26.2962 0
```

Testing whether subsequent (smaller) correlations are 0. $H_0 : \rho_{k+1} = \dots = \rho_p = 0 \quad k+1=1, \dots, p$

```
> round(test.out,4)
      k+1 Test Stat DF X2(.05) Pr>TS
[1,]   1  432.8051 16 26.2962 0.0000
[2,]   2   17.2415  9 16.9190 0.0451
[3,]   3    2.4753  4  9.4877 0.6491
[4,]   4    0.3098  1  3.8415 0.5778
```