

## R Solution Exercise Sheet 5: Canonical Correlation Analysis

### Computer Problems:

1. (a) 

```
> dim(mmreg)
[1] 600 9
```
- (b) 

```
psych <- mmreg[, 2:4]
acad <- mmreg[, 5:8]
```
- (c) 

```
require(ggplot2)
require(GGally)
require(CCA)
colnames(mmreg) <- c("No", "Control", "Concept", "Motivation", "Read", "Write", "Math",
+ "Science", "Sex")
> summary(mmreg)
```

No		Control		Concept		Motivation		Read	
Min.	: 1.0	Min.	:-2.23000	Min.	:-2.620000	Min.	:0.0000	Min.	:28.3
1st Qu.	:150.8	1st Qu.	:-0.37250	1st Qu.	:-0.300000	1st Qu.	:0.3300	1st Qu.	:44.2
Median	:300.5	Median	: 0.21000	Median	: 0.030000	Median	:0.6700	Median	:52.1
Mean	:300.5	Mean	: 0.09653	Mean	: 0.004917	Mean	:0.6608	Mean	:51.9
3rd Qu.	:450.2	3rd Qu.	: 0.51000	3rd Qu.	: 0.440000	3rd Qu.	:1.0000	3rd Qu.	:60.1
Max.	:600.0	Max.	: 1.36000	Max.	: 1.190000	Max.	:1.0000	Max.	:76.0

  

Write		Math		Science		Sex	
Min.	:25.50	Min.	:31.80	Min.	:26.00	Min.	:0.000
1st Qu.	:44.30	1st Qu.	:44.50	1st Qu.	:44.40	1st Qu.	:0.000
Median	:54.10	Median	:51.30	Median	:52.60	Median	:1.000
Mean	:52.38	Mean	:51.85	Mean	:51.76	Mean	:0.545
3rd Qu.	:59.90	3rd Qu.	:58.38	3rd Qu.	:58.65	3rd Qu.	:1.000
Max.	:67.10	Max.	:75.50	Max.	:74.20	Max.	:1.000
- (d) 

```
> cc1 <- cc(psych, acad)
```
- (e) 

```
> # display the canonical correlations
> cc1$cor
[1] 0.4640861 0.1675091 0.1039911
```
- (f) 

```
> # raw canonical coefficients
> cc1[3:4]
$xccoef
          [,1]          [,2]          [,3]
Control -1.2538339 -0.6214775 -0.6616896
Concept  0.3513499 -1.1876867  0.8267209
Motivation -1.2624203  2.0272641  2.0002284

$ycoef
          [,1]          [,2]          [,3]
Read -0.044620596 -0.004910018  0.02138056
Write -0.035877112  0.042071471  0.09130733
Math -0.023417185  0.004229472  0.00939821
Science -0.005025157 -0.085162175 -0.10983502
```

```
Sex      -0.632119239  1.084642482 -1.79464692
```

- (g) The raw canonical coefficients are interpreted in a manner analogous to interpreting regression coefficients i.e., for the variable read, a one unit increase in reading leads to a .0446 decrease in the first canonical variate of set 2 when all of the other variables are held constant. Here is another example: being female leads to a .6321 decrease in the dimension 1 for the academic set with the other predictors held constant.

```
(h) > # compute canonical loadings
> cc2 <- comput(psych, acad, cc1)
>
> # display canonical loadings
> cc2[3:6]
$corr.X.xscores
      [,1]      [,2]      [,3]
Control -0.90404632 -0.3896883 -0.1756227
Concept -0.02084327 -0.7087386  0.7051632
Motivation -0.56715105  0.3508882  0.7451290

$corr.Y.xscores
      [,1]      [,2]      [,3]
Read    -0.3900402 -0.06010654  0.01407660
Write   -0.4067914  0.01086074  0.02647208
Math    -0.3545378 -0.04990916  0.01536586
Science -0.3055607 -0.11336979 -0.02395489
Sex     -0.1689796  0.12645737 -0.05650916

$corr.X.yscores
      [,1]      [,2]      [,3]
Control -0.419555308 -0.06527635 -0.01826320
Concept -0.009673071 -0.11872021  0.07333073
Motivation -0.263206905  0.05877698  0.07748682

$corr.Y.yscores
      [,1]      [,2]      [,3]
Read    -0.8404480 -0.35882539  0.1353635
Write   -0.8765429  0.06483672  0.2545609
Math    -0.7639483 -0.29794886  0.1477612
Science -0.6584139 -0.67679758 -0.2303551
Sex     -0.3641127  0.75492816 -0.5434035
```

These loadings are correlations between variables and the canonical variates.

The above correlations are between observed variables and canonical variables which are known as the canonical loadings. These canonical variates are actually a type of latent variable.

```
(i) > # tests of canonical dimensions
> ev <- (1 - cc1$cor^2)
>
> n <- dim(psych)[1]
> p <- length(psych)
> q <- length(acad)
> k <- min(p, q)
> m <- n - 3/2 - (p + q)/2
>
> w <- rev(cumprod(rev(ev)))
>
> # initialize
> d1 <- d2 <- f <- vector("numeric", k)
>
> for (i in 1:k) {
+   s <- sqrt((p^2 * q^2 - 4)/(p^2 + q^2 - 5))
+   si <- 1/s
+   d1[i] <- p * q
+   d2[i] <- m * s - p * q/2 + 1
+   r <- (1 - w[i]^si)/w[i]^si
+   f[i] <- r * d2[i]/d1[i]
+   p <- p - 1
+   q <- q - 1
+ }
>
> pv <- pf(f, d1, d2, lower.tail = FALSE)
> (dmat <- cbind(WilksL = w, F = f, df1 = d1, df2 = d2, p = pv))
      WilksL      F df1      df2      p
[1,] 0.7543611 11.715733 15 1634.653 7.497602e-28
[2,] 0.9614300  2.944459  8 1186.000 2.905057e-03
[3,] 0.9891858  2.164612  3  594.000 9.109217e-02
```

As shown in the table above, the first test of the canonical dimensions tests whether all three dimensions are significant (they are,  $F = 11.72$ ), the next test tests whether dimensions 2 and 3 combi-

ned are significant (they are,  $F = 2.94$ ). Finally, the last test tests whether dimension 3, by itself, is significant (it is not). Therefore dimensions 1 and 2 must each be significant while dimension three is not.