

Multivariate Analysis for the Behavioral Sciences,
Second Edition (Chapman and Hall/CRC, 2019)
Solutions to Exercises of Chapter 18:
Grouped Multivariate Data

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Solutions

Exercise 18.4

```
pottery <- read.csv("data/pottery.csv")
str(pottery)

## 'data.frame':    45 obs. of  10 variables:
## $ kiln : int  1 1 1 1 1 1 1 1 1 1 ...
## $ AL203: num  18.8 16.9 18.2 16.9 17.8 18.8 16.5 18 15.8 14.6 ...
## $ FE203: num   9.52 7.33 7.64 7.29 7.24 7.45 7.05 7.42 7.15 6.87 ...
## $ MGO  : num   2 1.65 1.82 1.56 1.83 2.06 1.81 2.06 1.62 1.67 ...
## $ CAO  : num   0.79 0.84 0.77 0.76 0.92 0.87 1.73 1 0.71 0.76 ...
## $ NA20 : num   0.4 0.4 0.4 0.4 0.43 0.25 0.33 0.28 0.38 0.33 ...
## $ K20  : num   3.2 3.05 3.07 3.05 3.12 3.26 3.2 3.37 3.25 3.06 ...
## $ TIO2 : num   1.01 0.99 0.98 1 0.93 0.98 0.95 0.96 0.93 0.91 ...
## $ MNO  : num   0.077 0.067 0.087 0.063 0.061 0.072 0.066 0.072 0.062 0.055 ...
## $ BAO  : num   0.015 0.018 0.014 0.019 0.019 0.017 0.019 0.017 0.017 0.012 ...

head(pottery)

##   kiln AL203 FE203  MGO  CAO NA20  K20 TIO2  MNO  BAO
## 1    1  18.8  9.52 2.00 0.79 0.40 3.20 1.01 0.077 0.015
## 2    1  16.9  7.33 1.65 0.84 0.40 3.05 0.99 0.067 0.018
## 3    1  18.2  7.64 1.82 0.77 0.40 3.07 0.98 0.087 0.014
## 4    1  16.9  7.29 1.56 0.76 0.40 3.05 1.00 0.063 0.019
## 5    1  17.8  7.24 1.83 0.92 0.43 3.12 0.93 0.061 0.019
## 6    1  18.8  7.45 2.06 0.87 0.25 3.26 0.98 0.072 0.017

pottery_manova <- manova(cbind(AL203, FE203, MGO, CAO, NA20, K20, TIO2, MNO) ~ kiln,
                        data = pottery)

summary(pottery_manova, test = "Pillai")

##           Df  Pillai approx F num Df den Df    Pr(>F)
## kiln         1 0.92986   59.657     8    36 < 2.2e-16 ***
## Residuals 43
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(pottery_manova, test = "Wilks")

##           Df   Wilks approx F num Df den Df    Pr(>F)
## kiln       1 0.07014   59.657      8    36 < 2.2e-16 ***
## Residuals 43
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

summary(pottery_manova, test = "Hotelling")

##           Df Hotelling-Lawley approx F num Df den Df    Pr(>F)
## kiln       1      13.257   59.657      8    36 < 2.2e-16 ***
## Residuals 43
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

summary(pottery_manova, test = "Roy")

##           Df      Roy approx F num Df den Df    Pr(>F)
## kiln       1 13.257   59.657      8    36 < 2.2e-16 ***
## Residuals 43
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# all highly significant

region <- rep(1, length(pottery$kiln))
region[pottery$kiln == 2 | pottery$kiln == 3] <- 2
region[pottery$kiln == 4 | pottery$kiln == 5] <- 3

# calculate means of each region
m1 <- apply(pottery[region == 1, -1], 2, mean)
m2 <- apply(pottery[region == 2, -1], 2, mean)
m3 <- apply(pottery[region == 3, -1], 2, mean)

# find numbers in each class
n1 <- length(pottery[region==1, 1])
n2 <- length(pottery[region==2, 1])
n3 <- length(pottery[region==3, 1])

# find pooled covariance matrix
S123 <- ((n1-1)*var(pottery[region==1, -1]) +
          (n2-1)*var(pottery[region==2, -1]) +
          (n3-1)*var(pottery[region==3, -1])) / ((n1-1)+(n2-1)+(n3-1))

# find coefficients for each classification class
invS <- solve(S123)
a1 <- invS %*% (m1-m2)
a2 <- invS %*% (m1-m3)
a3 <- invS %*% (m2-m3)

# find thresholds
z12 <- (m1 %*% a1 + m2 %*% a1) / 2
z13 <- (m1 %*% a2 + m3 %*% a2) / 2
z23 <- (m2 %*% a3 + m3 %*% a3) / 2
```

```

# new pot with the following chemical composition:
newvalues<-c(15.5, 5.71, 2.07, 0.98, 0.65, 3.01, 0.76, 0.09, 0.012)

(newvalues - as.vector(z12)) %*% a1

##          [,1]
## [1,] -2276.902

(newvalues - as.vector(z13)) %*% a2

##          [,1]
## [1,] -10225.13

(newvalues - as.vector(z23)) %*% a3

##          [,1]
## [1,] 29623.82

# conclusion: allocate to region 2 (plot a graph!)

```