Lab 7

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 \mathbf{a}

```
library(ISLR)
library(dplyr)

##

## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':

##

## filter, lag

## The following objects are masked from 'package:base':

##

intersect, setdiff, setequal, union

load("../data/Auto-3.rda")
```

b

```
Auto <- Auto %>%
mutate(Economy = case_when(
   mpg <= 17 ~ "Heavy",
   mpg <= 22.75 ~ "OK",
   mpg <= 29 ~ "Eco",
   mpg > 29 ~ "Excellent"
)) %>%
mutate(Economy = as.factor(Economy)) %>%
mutate(origin = as.factor(origin))
```

 \mathbf{c}

```
library(MASS)

##
## Attaching package: 'MASS'

## The following object is masked from 'package:dplyr':
##
## select
```

```
table(Auto$Economy)
##
##
                                        OK
         Eco Excellent
                           Heavy
##
         101
                    95
                              99
                                        97
lda_result <- lda(Economy ~ mpg + cylinders + displacement + weight + acceleration + year + origin, dat
lda_result
## Call:
## lda(Economy ~ mpg + cylinders + displacement + weight + acceleration +
      year + origin, data = Auto)
##
##
## Prior probabilities of groups:
        Eco Excellent
                          Heavy
                                        OK
## 0.2576531 0.2423469 0.2525510 0.2474490
##
## Group means:
##
                 mpg cylinders displacement
                                             weight acceleration
                                                                       vear
            25.88911 4.277228
                                    128.7178 2501.069
                                                         16.35248 76.03960
## Eco
## Excellent 34.30842 4.073684
                                    101.7895 2157.958
                                                         16.65053 79.17895
                                    335.7879 4114.384 13.37071 73.38384
## Heavy
            14.26869 7.717172
## OK
            19.62990 5.793814
                                    209.2371 3116.237
                                                         15.82577 75.43299
##
               origin2
                          origin3
## Eco
            0.31683168 0.20792079
## Excellent 0.23157895 0.51578947
            0.03030303 0.00000000
## Heavy
## OK
            0.11340206 0.09278351
##
## Coefficients of linear discriminants:
##
                         I.D1
                                      LD2
                                                    LD3
## mpg
               -0.3389905364 -0.264405943 -0.103651807
                0.2317355612 -0.422308212 1.163393044
## cylinders
## displacement 0.0018855775 -0.005475355 -0.004392613
## weight
                0.0003569182 -0.001252238 -0.001962502
## acceleration 0.0300076370 0.061115872 0.180788046
            -0.0378684709 0.085925497 0.194580341
## year
## origin2
                0.1702613411 -0.399663658 -0.729557654
               -0.1481699449 -1.137048668 0.234903830
## origin3
## Proportion of trace:
##
     LD1
            LD2
                   LD3
```

Heavy economy cars tend to have more cylinders, more displacement, higher weight, and lower mpg. Excellent and Eco economy cars tend to be lighter and more fuel efficient.

d

0.9406 0.0525 0.0069

```
library(MASS)

lda_cv <- lda(Economy ~ mpg + cylinders + displacement + weight + acceleration + year + origin, data = .

conf_matrix <- table(True = Auto$Economy, Predicted = lda_cv$class)</pre>
```

```
print(conf_matrix)
##
              Predicted
## True
               Eco Excellent Heavy OK
##
     Eco
                95
                            0
                                  0 6
##
     Excellent 13
                           82
                                  0 0
##
     Heavy
                 0
                            0
                                 85 14
                                  8 78
                            0
##
                 11
accuracy <- mean(Auto$Economy == lda_cv$class)</pre>
print(paste("Classification accuracy (CV):", round(accuracy * 100, 2), "%"))
## [1] "Classification accuracy (CV): 86.73 %"
\mathbf{e}
lda_cv_prior <- lda(Economy ~ mpg + cylinders + displacement + weight + acceleration + year + origin,</pre>
                     data = Auto,
                     prior = c(0.25, 0.25, 0.25, 0.25),
                     CV = TRUE
conf_matrix_prior <- table(True = Auto$Economy, Predicted = lda_cv_prior$class)</pre>
print(conf_matrix_prior)
              Predicted
##
## True
               Eco Excellent Heavy OK
##
     Eco
                94
                            0
                                  0 7
     Excellent 12
                           83
                                  0 0
##
##
     Heavy
                 0
                            0
                                 85 14
                                  7 79
##
     OK
                 11
                            0
accuracy_prior <- mean(Auto$Economy == lda_cv_prior$class)</pre>
print(paste("Classification accuracy with custom priors:", round(accuracy_prior * 100, 2), "%"))
## [1] "Classification accuracy with custom priors: 86.99 %"
the classification accuracy slightly decreased.
f
lda_cv_prior2 <- lda(Economy ~ mpg + cylinders + displacement + weight + acceleration + year + origin,</pre>
                     data = Auto,
                     prior = c(0.4,0.3,0.2,0.1),
                     CV = TRUE)
conf_matrix_prior2 <- table(True = Auto$Economy, Predicted = lda_cv_prior2$class)</pre>
print(conf_matrix_prior2)
##
              Predicted
               Eco Excellent Heavy OK
## True
##
     Eco
                98
                            0
                                  0 3
                           81
                                  0 0
##
     Excellent 14
##
     Heavy
                 0
                            0
                                 88 11
##
                17
                            0
                                  9 71
     OK
```

```
accuracy_prior2 <- mean(Auto$Economy == lda_cv_prior2$class)
print(paste("Classification accuracy with custom priors:", round(accuracy_prior2 * 100, 2), "%"))
## [1] "Classification accuracy with custom priors: 86.22 %"
the classification accuracy slightly decreased again.</pre>
```