```
#use data set Cracker from Ecdat package
install.packages(Ecdat)
library(Ecdat)
data("Cracker")
#remove the Nabisco prices that equal zero
cracker <- Cracker
for (t in 1:3289) {
  if(cracker$price.nabisco[t] == 0){
    cracker <- cracker[-t,]</pre>
  }
}
#split the data in a train and test data set
set.seed(218)
sample <- sample.split(cracker, SplitRatio = 0.75)</pre>
train <- subset(cracker, sample == TRUE)</pre>
test <- subset(cracker, sample == FALSE)</pre>
#create matrices for train and test data with the marketing mix
variables (display, feature and log of price)
train_matrix <- sapply(train[,2:13], as.numeric)</pre>
test_matrix <- sapply(test[,2:13], as.numeric)</pre>
x_train <- as.matrix(train_matrix)</pre>
x_test <- as.matrix(test_matrix)</pre>
x_train_good <- cbind(x_train[,1:8],log(x_train[,9:12]))</pre>
x_{test_good} \leftarrow cbind(x_{test_good}, 1:8], log(x_{test_good}, 9:12]))
#create binary vectors containing the choice of the individuals for
training and test data
choice_train_sunshine <- ifelse(train$choice=="sunshine",1,0)</pre>
choice_train_kleebler <- ifelse(train$choice=="kleebler",1,0)</pre>
choice train nabisco <- ifelse(train$choice=="nabisco",1,0)</pre>
choice_train_private <- ifelse(train$choice=="private",1,0)</pre>
choice_sunshine_test <- ifelse(test$choice=="sunshine",1,0)</pre>
choice kleebler test <- ifelse(test$choice=="kleebler",1,0)</pre>
choice_nabisco_test <- ifelse(test$choice=="nabisco",1,0)</pre>
choice_private_test <- ifelse(test$choice=="private",1,0)</pre>
choice all train <- cbind(choice train sunshine,
choice train kleebler, choice train nabisco, choice train private)
choice_all_test <- cbind(choice_sunshine_test, choice_kleebler_test,</pre>
choice_nabisco_test, choice_private_test)
#make one matrix with all the variables
matrix all <- cbind(x train good, choice all train)</pre>
matrix_all_test <- cbind(x_test_good,choice_all_test)</pre>
#create the parameters
parm < - rep(1,6)
#create log likelihood function for Multinomial Logit model
LL <- function(parm,matrix_all){</pre>
```

```
# Computes the log likelihood for the Multinomial Logit model.
  #
  # Args:
      parm: A vector containing the parameters of the log likelihood
function (beta and gamma)
      matrix_all: A matrix containing all variables in the training
data set used for the log likelihood function (display,
      feature and log of price for the brands Sunshine, Kleebler,
Nabisco and Private for every observation)
  # Returns:
  #
      The log likelihood value for the Multinomial Logit model
  beta <- parm[1:3]
  gamma <- parm[4:6]</pre>
  x_train_good <- matrix_all[,1:12]</pre>
  choice_all_train <- matrix_all[,13:16]</pre>
  x_sunshine <- cbind(x_train_good[,1],x_train_good[,</pre>
5],x_train_good[,9])
  x_kleebler <- cbind(x_train_good[,2],x_train_good[,</pre>
6],x train good[,10])
  x_nabisco <- cbind(x_train_good[,3],x_train_good[,</pre>
7],x_train_good[,11])
  x_private <- cbind(x_train_good[,4],x_train_good[,</pre>
8],x_train_good[,12])
  n <- nrow(x_sunshine)</pre>
  output <- rep(1,n)</pre>
  #Log likelihood function
  for (i in 1:n) {
    q <- choice_all_train[i,1]%**(beta[1]+gamma%*%x_sunshine[i,1:3]-</pre>
log(exp(beta[1]+gamma%*%x_sunshine[i,1:3])
+exp(beta[2]+gamma%*%x kleebler[i,1:3])
+exp(beta[3]+gamma%*%x_nabisco[i,1:3])+exp(gamma%*%x_private[i,
1:3])))+choice_all_train[i,2]%*%(beta[2]+gamma%*%x_kleebler[i,1:3]-
log(exp(beta[1]+gamma%*%x sunshine[i,1:3])
+exp(beta[2]+gamma%*%x kleebler[i,1:3])
+exp(beta[3]+gamma%*%x_nabisco[i,1:3])+exp(gamma%*%x_private[i,
1:3])))+choice all train[i,3]%*%(beta[3]+gamma%*%x nabisco[i,1:3]-
log(exp(beta[1]+gamma%*%x sunshine[i,1:3])
+exp(beta[2]+gamma%*%x kleebler[i,1:3])
+exp(beta[3]+gamma%*%x_nabisco[i,1:3])+exp(gamma%*%x_private[i,
1:3])))+choice all train[i,4]%*%(gamma%*%x private[i,1:3]-
log(exp(beta[1]+gamma%*%x_sunshine[i,1:3])
+exp(beta[2]+gamma%*%x_kleebler[i,1:3])
+exp(beta[3]+gamma%*%x nabisco[i,1:3])+exp(gamma%*%x private[i,
1:31)))
    output[i] <- q
  loglikelihood <- -sum(output)</pre>
}
```

```
#optimise the log likelihood of MNL using "BFGS"
MNL <- optim(par = parm, fn = LL, matrix_all = matrix_all, method = "BFGS")</pre>
```