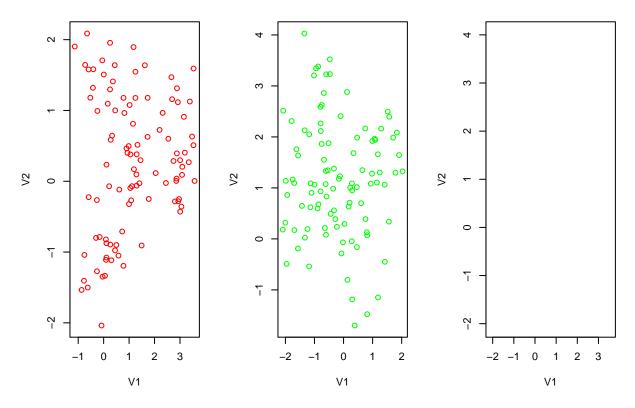
Example1(sh)

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2020 10 25

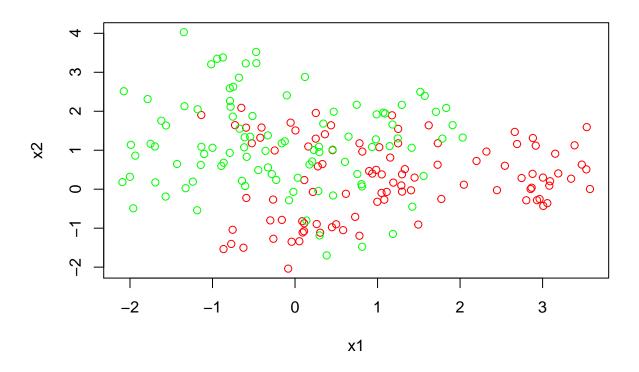
```
setwd("C:/Users/ghkstbd/Desktop/2020 고급데이터마이닝/R Example 1")
# red points form training daaset
redpoints=read.table("trainred.txt",sep="\t",header=FALSE)
# green points from training dataset
greenpoints=read.table("traingreen.txt",sep="\t",header=FALSE)
# red points from test dataset
redtestpoints=read.table("testred.txt",sep="\t",header=FALSE)
# green points from test dataset
greentestpoints=read.table("testgreen.txt",sep="\t",header=FALSE)
# Combine red and green points.
dim(redpoints) # 100 by 2
## [1] 100
dim(greenpoints) # 100 by 2
## [1] 100
X=rbind(redpoints, greenpoints) # 1~100:red, 101~200:green
dim(X) # 200 by 2
## [1] 200
y=c(rep(1,100),rep(0,100))
y # 1~100:red(1), 101~200:green(0)
   ## [186] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
par(mfrow=c(1,3))
plot(redpoints,col="red")
plot(greenpoints,col="green")
```

plot(X,type="n")



```
par(mfrow=c(1,1))
plot(X,type="n",main='Red & Green',xlab='x1',ylab='x2') # type="n":nothing
points(redpoints,col="red")
points(greenpoints,col="green")
```

Red & Green

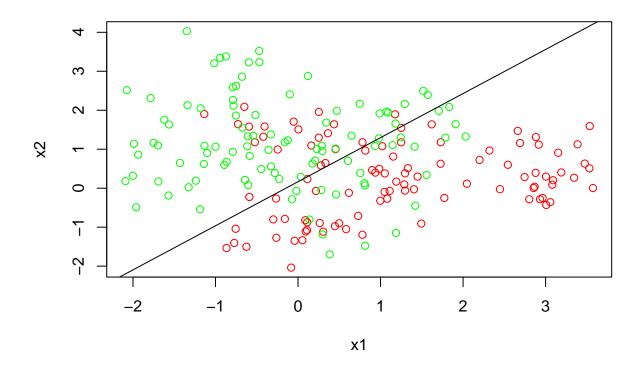


```
############# Q3. ###############
is.matrix(X)
## [1] FALSE
is.data.frame(X)
## [1] TRUE
X=as.matrix(X)
is.matrix(X)
## [1] TRUE
# Regression 1
X1=cbind(rep(1,nrow(X)),X) # design matrix : 200 by 3
dim(X1)
## [1] 200
beta.hat=solve(t(X1)%*%X1)%*%t(X1)%*%y
beta.hat
            [,1]
##
       0.5235718
##
## V1 0.1601129
## V2 -0.1415877
```

```
y.hat=X1%*%beta.hat # 200 by 1
result <- ifelse(y.hat>0.5,1,0)
result[,1]
    [38] 1 1 1 1 1 1 1 1 1 0 1 1 1 1 1 1 0 0 0 1 1 1 1 1 0 1 1 1 1 1 1 1 1 1 0 1 1 1 1
## [75] 1 0 1 1 1 0 1 1 1 0 1 1 1 1 0 0 1 1 1 1 1 0 0 1 1 1 1 1 0 0 0 1 0 0 0 0 0 0 1 0 1
## [112] 0 0 1 0 0 0 0 0 0 0 1 1 0 0 0 0 0 0 1 1 0 0 0 0 0 1 0 0 0 0 1 1 0 0 0 0 0 1 1 0 0 0
## [186] 0 0 0 0 0 0 1 0 0 1 0 0 1 0
sum(result[,1]==y)
## [1] 155
te=1-sum(result[,1]==y)/length(y); te
## [1] 0.225
# Regression 2
fit=lm(y~X[,1]+X[,2])
summary(fit)
##
## Call:
## lm(formula = y \sim X[, 1] + X[, 2])
##
## Residuals:
##
                1Q
                   Median
## -0.87609 -0.33002 0.00249 0.30485 0.92773
##
## Coefficients:
             Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 0.52357
                        0.03653 14.332 < 2e-16 ***
             0.16011
                        0.02124
                                7.539 1.69e-12 ***
## X[, 1]
## X[, 2]
             -0.14159
                        0.02545 -5.563 8.59e-08 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.408 on 197 degrees of freedom
## Multiple R-squared: 0.344, Adjusted R-squared: 0.3374
## F-statistic: 51.66 on 2 and 197 DF, p-value: < 2.2e-16
summary(fit)$coef
##
               Estimate Std. Error
                                   t value
                                              Pr(>|t|)
## (Intercept) 0.5235718 0.03653068 14.332387 2.199775e-32
## X[, 1]
              0.1601129 0.02123825 7.538895 1.692278e-12
## X[, 2]
             -0.1415877 0.02545322 -5.562664 8.587034e-08
summary(fit)$coef[,1]
## (Intercept)
                  X[, 1]
                             X[, 2]
    0.5235718
               0.1601129 -0.1415877
beta.hat2 = fit$coefficients
y.hat2 = fit$fitted.values # y.hat2 = fitted.values(fit)
```

```
g.hat=as.numeric(y.hat>0.5) # T/F로 반환해서 numeric으로 1/0
g.hat
    ##
## [38] 1 1 1 1 1 1 1 1 1 0 1 1 1 1 1 0 0 0 1 1 1 1 1 0 1 1 1 1 1 1 1 1 1 0 1 1 1 1
## [75] 1 0 1 1 1 0 1 1 1 0 1 1 1 1 0 0 1 1 1 1 1 0 0 1 1 1 1 1 0 0 0 1 0 0 0 0 1 0 1
## [112] 0 0 1 0 0 0 0 0 0 0 1 1 0 0 0 0 0 0 1 1 0 0 0 0 0 1 1 0 0 0 0 0 1 1 0 0 0 0 0 1 1 0 0 0
## [149] 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 1 1 0 0 0 1 0 1 0 1 0 0 0 0 1 1 0 1 0 0 0 0 0 0 1 0 1 0 0 0
## [186] 0 0 0 0 0 0 1 0 0 1 0 0 0 1 0
z=sum(g.hat==y)
training.error=1-z/200 # 오분류율을 찾는 거니까!
training.error # 0.225
## [1] 0.225
############## Q4. ###############
plot(X,type="n",main='Red & Green',xlab='x1',ylab='x2') # type="n":nothing
points(redpoints,col="red")
points(greenpoints,col="green")
beta.hat
##
           [,1]
      0.5235718
##
## V1 0.1601129
## V2 -0.1415877
abline((.5-beta.hat[1])/beta.hat[3],-beta.hat[2]/beta.hat[3])
```

Red & Green



#abline((.5-beta.hat[1,1])/beta.hat[3,1],-beta.hat[2,1]/beta.hat[3,1])

```
############# Q5. ################
X0=rbind(redtestpoints, greentestpoints) # testset은 각 1000 by 2
X0=as.matrix(X0) # X0 \( \subseteq \) 2000 by 2
y0=c(rep(1,1000),rep(0,1000)) # 마찬가지로 y를 1~1000(red)=1, 1001~2000(green)=0
X01=cbind(rep(1,2000),X0) # design matrix(2000 by 3)
# 추정한 모수들을 이용하여 y=xb, 즉 testset에서의 예측된 색(y)을 구함, testerror
y0.hat=X01%*%beta.hat
g0.hat=as.numeric(y0.hat>0.5)
test.error=1-sum(g0.hat==y0)/2000
test.error # 0.245 > 0.225(training\ error)
## [1] 0.245
# testerror is always larger than training error !!!
############# Q6. ################
# Training observations 사이의 거리를 계산해야 함.
# Euclidean distance matrix
D=matrix(0,200,200) # 200 by 200 matrix all values are zero.
```

```
dim(D) # -> Nrow of training set is 200
## [1] 200 200
# Training set X is 200 by 2
for (i in 1:200) for (j in 1:200) D[i,j]=sqrt(sum((X[i,]-X[j,])^2))
# => D 200 by 200 계산으로 쭉 채움
# To find k-closest observation, Example
s=c(1,4,5,3)
sort(s) # 1 3 4 5 (ascending dataset)
## [1] 1 3 4 5
order(s) # ordered position of dataset, it return the index value
## [1] 1 4 2 3
# 1 4 2 3
# 첫 번째로 작은 값은 첫 번째 인덱스(1)에 있다.
# 두 번째로 작은 값은 네 번째 인덱스(3)에 있다.
# 세 번째로 작은 값은 두 번째 인덱스(4)에 있다.
# 네 번째로 작은 값은 세 번째 인덱스(5)에 있다.
# ex) (x1,x1)~(x1,x200)중 closest training obs 찾기
# D[1,] : (x1,x1) ~ (x1,x200)의 거리를 나타냄
order(D[1,])
          23 32 74 81 114 12 49 66 130 100 35 159 31 26 152 171
   [1]
        1
##
   [19] 95
         58
             2 104
                   67 17
                         77
                            36 192 111 44 183 51 99 160 145
                                                          56
  [37] 164 63
            10
                 3
                   28
                       9 92
                            70 97
                                 72 90 181 194 22 14 153
                                                       79
                                                          61
              7 170
                   84 187 109
                            29 185
                                 43 166 125 197 173 110 149
                                                          75
 [55] 13 15
## [73] 131 47 60 46 200 188 27 140 195
                                  37 54 169 175 89 158 126 144
  [91] 128 134 174 121
                   87
                      18 117 132
                              71 101 147 83 184 127 163 123
                                                          73
## [109] 96 21 50 85
                   42
                      39 33 98
                              19
                                  30 167 118 129 142 52 108 137 146
## [127] 190
           6 135
                45 11
                      48 189 112
                              91
                                  93 107 199 69 80 16
                           20 161 113 82 103 78 179 191 139 24
## [145] 150 119 41 68 151 122 53
                                                         88
## [163] 156 105 115 57 138 25 136 94 65
                                 86
                                      8 172 141 76 116 198 154
## [181] 176 120 143 165 177 180 178 155 186 62 157 168 182 162 196 193 106 133
## [199] 148 124
order(D[1,])[1:7] # 7-closest training obs
     1 23 32 74 81 114 12
## [1]
y # y=c(rep(1,100),rep(0,100))
   ## [186] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
y[order(D[1,])[1:7]]
## [1] 1 1 1 1 0 1
```

```
mean(y[order(D[1,])[1:7]]) # y_hat value
## [1] 0.8571429
g.hat=(mean(y[order(D[1,])[1:7]])>0.5)
g.hat
## [1] TRUE
# Q6.
k=7
g.hat7 = rep(0,200)
for (i in 1:200) g.hat7[i]=(mean(y[order(D[i,])[1:k]])>0.5)
g.hat7
##
   ## [38] 1 1 0 1 1 1 0 1 0 0 1 1 1 1 1 1 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 0 1 1 1 0 1 1 1 1 1
## [186] 0 0 0 0 0 0 1 0 0 1 0 0 1 0
training.error=1-sum(g.hat7==y)/200
training.error
## [1] 0.2
k=3
g.hat3=rep(0,200)
for (i in 1:200) g.hat3[i]=(mean(y[order(D[i,])[1:k]])>0.5)
training.error=1-sum(g.hat3==y)/200
training.error
## [1] 0.17
k=1
g.hat1=rep(0,200)
for (i in 1:200) g.hat1[i]=(mean(y[order(D[i,])[1:k]])>0.5)
training.error=1-sum(g.hat1==y)/200
training.error
## [1] 0
g <- NULL
for(k in c(1,3,7)){
     for(i in 1:200){
            g[i] <- mean(y[order(D[i,])[1:k]])>0.5
     print(1-sum(g==y)/200)
}
## [1] 0
## [1] 0.17
## [1] 0.2
############## Q7. ###############
# Test observations 사이의 거리를 계산해야 함.
# Euclidean distance matrix
```

```
DO=matrix(0,2000,200) # row = testset, col = trainingset
dim(DO)
## [1] 2000 200
# (x1,x1) \sim (x1,x200) \sim \dots \sim (x2000,x1) \sim (x2000,x200)
for (i in 1:2000) for (j in 1:200) DO[i,j]=sqrt(sum((X0[i,]-X[j,])^2))
k=7
g0.hat=rep(0,2000)
for (i in 1:2000) g0.hat[i]=(mean(y[order(D0[i,])[1:k]])>0.5)
head(g0.hat,100)
    [1] 1 0 1 1 0 1 1 1 0 1 1 1 0 0 1 1 1 1 0 0 1 1 1 1 0 1 1 1 0 1 0 1 0 0 0 1 1 1 0 1 0 1 0 1 0 1
##
   [38] 1 0 0 0 1 1 1 0 1 1 1 1 1 0 1 1 1 1 0 1 1 0 0 0 1 0 1 1 1 1 0 1 1 1 1 0 0 1 1
test.error=1-sum(g0.hat==y0)/2000
test.error
## [1] 0.2965
k=3
g0.hat=rep(0,2000)
for (i in 1:2000) g0.hat[i]=(mean(y[order(D0[i,])[1:k]])>0.5)
head(g0.hat,100)
##
    [38] 1 1 0 0 1 0 1 0 1 1 1 1 1 1 1 1 1 1 0 0 1 0 0 0 1 0 1 0 1 0 1 1 1 1 1 1 0 1 1 0
## [75] 1 1 1 1 0 1 0 0 1 1 0 1 1 0 0 0 1 0 0 0 0 1 1 1 1 1 0
test.error=1-sum(g0.hat==y0)/2000
test.error
## [1] 0.328
k=1
g0.hat=rep(0,2000)
for (i in 1:2000) g0.hat[i]=(mean(y[order(D0[i,])[1:k]])>0.5)
head(g0.hat, 100)
##
    [1] 1 1 1 1 0 1 0 1 0 1 1 1 1 0 0 1 0 1 1 1 0 0 1 0 1 1 1 1 0 1 0 0 0 0 1 1 1 1 1 0 0 1 0 1
## [38] 1 0 0 0 1 1 0 0 1 1 1 1 1 1 1 1 1 0 1 0 1 1 1 0 0 0 1 0 1 0 0 0 1 1 1 1 1 1 1 1 1
test.error=1-sum(g0.hat==y0)/2000
test.error
## [1] 0.3225
g2 <- NULL
for(k in c(1,3,7)){
       for(i in 1:2000){
               g2[i] <- mean(y[order(D0[i,])[1:k]])>0.5
       print(1-sum(g2==y0)/2000)
}
```

[1] 0.3225

[1] 0.328 ## [1] 0.2965