# Homework 4

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## HW4 Part1: SVM classification for Simulated Data

## Question 1 : Generate a Data Set by Simulations

We seek to generate 5000 cases  $x1 \dots x5000$  in R4 each case  $x = [x1 \ x2 \ x3 \ x4]$  has 4 numerical features

#### Step 1

Using random sampling of uniform distribution over the interval [-2, +2] to create:

- Aij 4x4 matrix
- Bi 1x4 matrix
- C

And define the polynomial of degree 2 in the 4 variables x1 x2 x3 x4 as follows

$$Pol(x) = \sum_{i} \sum_{j} A_{ij} x_i x_j + \sum_{i} B_i x_i + c/20$$

```
## [1] "Aij 4x4 matrix:"
##
           [,1]
                  [,2]
                          [,3]
                                  [,4]
## [1,] -1.0243 0.933 0.224 1.6710
## [2,] 0.0518 0.651 -0.704 -0.0622
## [3,] -1.8455 1.913 1.792 -1.2401
## [4,] -1.3353 -0.468 -0.373 0.1972
## [1] "Bi 4x1 matrix:"
##
          [,1]
## [1,] -1.393
## [2,] -1.710
## [3,]
        1.317
## [4,] 0.916
## [1] "C = 1.852"
## [1] "Test the polynomial function"
## [1] "If x = "
## [1] -0.381 0.885 -0.766 0.924
## [1] ", then Pol(x) = -0.5049"
If x = , then Pol(x) = 3.5853095
```

#### Step 2

Using random sampling of uniform distribution over the interval [-2, 2] select 10000 vectors x1 ... x10000 in R4 each such vector xn has 4 randomly chosen coordinates with values in [-2, 2]] for each selected xn compute U(n) = Pol(xn) and y(n) = sign[U(n)]

define two classes by CL(1) = class1 = set of all xn such that y(n) = +1 CL(-1) = class1 = set of all xn such that y(n) = -1 keep only 2500 cases in CL(1) and 2500 cases in CL(-1), Center and Rescale this data set of size 5000 so that the standardized data set will have mean = 0 and dispersion =1 Then Split each class into a training set and a test set , using the proportions 80% and 20% this defines a training set TRAIN and a test set TEST of resp. sizes 4000 and 1000

Table 1: Head of TRAIN set

|      | X1     | X2     | Х3     | X4     | Yn |
|------|--------|--------|--------|--------|----|
| 4379 | 0.693  | 0.119  | 0.234  | 1.284  | -1 |
| 4725 | 1.469  | 0.542  | -0.766 | -1.065 | -1 |
| 46   | -0.416 | -0.214 | 1.630  | -0.139 | 1  |
| 913  | 0.058  | 1.492  | 1.134  | 0.649  | 1  |
| 2369 | -1.590 | -1.647 | -1.484 | -0.238 | 1  |
| 4997 | -1.308 | 1.493  | -0.930 | 1.666  | -1 |
|      |        |        |        |        |    |

Table 2: Head of TEST set

|      | X1     | X2     | Х3     | X4     | Yn |
|------|--------|--------|--------|--------|----|
| 1444 | -1.309 | 0.190  | 0.492  | -1.592 | 1  |
| 2827 | -0.618 | 0.614  | -0.680 | -0.086 | -1 |
| 596  | 0.805  | 1.377  | 1.357  | -0.695 | 1  |
| 2628 | 0.483  | -0.014 | -0.394 | -0.397 | -1 |
| 4519 | 0.456  | -0.742 | -0.903 | -0.842 | -1 |
| 277  | -0.959 | -0.850 | 0.280  | 1.229  | 1  |

### Question 2: SVM classification by linear kernel

- Fix arbitrarily the "cost" parameter in the sym() function, for instance cost = 5
- Select the kernel parameter kernel = "linear"
- Run the svm() function on the set TRAIN
- Compute the number S of support vectors and the ratio s = S/4000
- Compute the percentages of correct prediction PredTrain and PredTest on the sets TRAIN and TEST
- Compute two confusion matrices (one for the set TRAIN and one for the test set. Confusion matrices must be converted in terms of frequencies of correct predictions within each class
- Compute the errors of estimation on PredTRAIN, PredTEST, and on the terms of the confusion matrices interpret your results

```
############ A function to calculate the error of estimation
std_p = function(acc,n) {sqrt(acc*(1-acc)/n)*100}

############ A function to generate a nice confusion matrix
generate_cfm=function(predi,truec,cap){
    cfm=table(truec, predi)
    #colnames(cfm)=c('Pred_CL-1','Pred_CL1')
    #rownames(cfm)=c('True_CL-1','True_CL1')
    #print(cfm,caption = cap)
    nocc=rowSums(cfm)
    pocp=cfm/nocc
    kable(pocp,caption = paste('Confustion matrix for the set ',cap),digits = 3)
    acc=(cfm[1,1]+cfm[2,2])/length(truec)
    std=std_p(acc,length(truec))
```

```
std_1=std_p(pocp[1,1],nocc[1])
     std_2=std_p(pocp[2,2],nocc[2])
     ac=c(acc,pocp[1,1],pocp[2,2])
     st=c(std,std_1,std_2)
     pred_table=data.frame(ac,st)
     rownames(pred_table) = c('Global accuracy','CL-1','CL1')
     colnames(pred_table) = c('Accuracy','Standard deviation')
     kable(pred table,caption=paste('Prediction table for the set ',cap),digits = 3 )
}
N=5000
\#x=TRAIN[,1:4]
#y=TRAIN$Yn
model = svm(Yn~., cost=5, kernel='linear', type='C-classification', scale=FALSE, data=TRAIN)
#Alternative way to use sum without specify the type is to conver Y to factor
\#dat = data.frame(x=x, y=as.factor(y)) \# must convert to factor to avoid the sum regression
\#model = svm(y^{-}, cost = 5, kernel = 'linear', scale = FALSE, data = dat)
summary(model)
Call: svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data = TRAIN, cost = 5, kernel = "linear", type = "C-classification", scale = 1, svm(formula = Yn \sim ., data 
FALSE)
Parameters: SVM-Type: C-classification SVM-Kernel: linear cost: 5
Number of Support Vectors: 2498
(1250 1248)
Number of Classes: 2
Levels: -1 1
S=sum(model$nSV)
print(paste('Number S of support vectors is ',S))
[1] "Number S of support vectors is 2498"
print(paste('And the ratio s = ',s))
[1] "And the ratio s = 0.4996"
PredTRAIN =predict(model,TRAIN[,1:4])
PredTEST =predict(model,TEST[,1:4])
# Check accuracy:
generate_cfm(PredTRAIN, TRAIN$Yn,'Confusion matrix for the set TRAIN')
```

Table 3: Prediction table for the set Confusion matrix for the set TRAIN

|                 | Accuracy | Standard deviation |
|-----------------|----------|--------------------|
| Global accuracy | 0.731    | 0.701              |
| CL-1            | 0.756    | 0.960              |
| CL1             | 0.705    | 1.020              |

generate\_cfm(PredTEST, TEST\$Yn,'Confusion matrix for the set TEST')

Table 4: Prediction table for the set Confusion matrix for the set  $\operatorname{TEST}$ 

|                 | Accuracy | Standard deviation |
|-----------------|----------|--------------------|
| Global accuracy | 0.752    | 1.366              |
| CL-1            | 0.774    | 1.870              |
| CL1             | 0.730    | 1.985              |