

1. Objective: Build SVM classification model to predict if the customer is likely to accept the personal loan offered by the bank.
2. Another library kernlab for kernel SVM
3. Grid search

Dataset Details

Attribute	Description
ID	Customer ID
Age	Customer's age in completed years
Experience	#years of professional experience
Income	Annual income of the customer (\$000)
ZIPCode	Home Address ZIP code.
Family	Family size of the customer
CCAvg	Avg. spending on credit cards per month (\$000)
Education	Education Level. 1: Undergrad; 2: Graduate; 3: Advanced/Professional
Mortgage	Value of house mortgage if any. (\$000)
Personal Loan	Did this customer accept the personal loan offered in the last campaign? (Target attribute)
Securities Account	Does the customer have a securities account with the bank?
CD Account	Does the customer have a certificate of deposit (CD) account with the bank?
Online	Does the customer use internet banking facilities?
CreditCard	Does the customer use a credit card issued by UniversalBank?

Classification using e1071

1. Load Data into R
2. Data preparation
 - a. Remove the columns ID & ZIP
 - b. Variable "Education" has 3 categories , so create dummy variables
 - c. Standardization of data – use range method
 - d. Split the data into train and test datasets

3. Model Building

```
#install.packages("e1071")
```

```
library(e1071)
```

Store the independent variables and target variable separately (for easy use)

- `x = subset (train_bankdata, select = -Personal.Loan) #remove response variable`
- `y = as.factor (train_bankdata$Personal.Loan)`

#Build the model on train data

```
model = svm(x,y, method = "C-classification", kernel = "linear", cost = 10, gamma = 0.1)
summary(model)
#The "cost" parameter balances the trade-off between having a large margin and
#classifying all points correctly. It is important to choose it well to have
#good generalization.
```

4. Predict on train & test data**5. Build the confusion matrix****6. Compute the error metrics**

Note: Build SVM model by changing the kernel function to “radial” and check if the accuracies are better.

Classification using K SVM

```
#install.packages("kernlab")
library(kernlab)
names(train_bankdata)
#Build model using ksvm with "rbf" kernel
kern_rbf <- ksvm(as.matrix(train_bankdata[,-7]),train_bankdata[,7],
               type='C-svc',kernel="rbf",kpar=list(sigma=(0:1)),
               C=10, cross=5)
kern_rbf

kern_rbf <- ksvm(as.matrix(train_bankdata[,-7]),train_bankdata[,7],
               type='C-svc',kernel="rbf",kpar="automatic",
               C=10, cross=5)

#Build model using ksvm with "vanilladot" kernel
kern_vanilla <- ksvm(as.matrix(train_bankdata[,-7]),train_bankdata[,7],
                  type='C-svc',kernel="vanilladot", C = 10)
kern_vanilla

#Predict model "kern_rbf" (on test data)
kpred_rbf<- predict(kern_rbf,test_bankdata[-7])
confMatrix <- table(test_bankdata$Personal.Loan, kpred_rbf)
acc_rbf = sum(diag(confMatrix))/sum(confMatrix);acc_rbf
rec_rbf = (confMatrix[2,2]/(confMatrix[2,2]+confMatrix[2,1]));rec_rbf

#Predict model "kern_vanilla" (on test data)
kpred_vanilla<- predict(kern_vanilla,test_bankdata[-7])
confMatrix <- table(test_bankdata$Personal.Loan, kpred_vanilla)
acc_vanilla = sum(diag(confMatrix))/sum(confMatrix);acc_vanilla
rec_vanilla = (confMatrix[2,2]/(confMatrix[2,2]+confMatrix[2,1]));rec_vanilla
```

#Perform a grid search

```
tuneResult <- tune(svm, train.x = x, train.y = y, ranges = list(gamma = 10^(-6:-1), cost =  
2^(2:3)))  
print(tuneResult)
```

#Predict model and calculate errors

```
tunedModel <- tuneResult$best.model  
tunedModelY <- predict(tunedModel, as.matrix(x))  
Conf <- table(y, tunedModelY)  
# you can now compute the metrics.
```

Assignment: SVM model building for regression

- Given the data BostonHousing.csv, we need to predict the variable 'medv', which is the median value of owner-occupied homes in USD in 1000's.
- Perform required preprocessing steps.
- Split the data into test and train.
- Run a regression using svm. Read the help function to understand how to perform a regression.
- Perform tuning to obtain the best metrics on test data