

# Machine Learning - Block02 Assignment 01

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## 1. Ensemble Methods

The file `spambase.csv` contains information about the frequency of various words, characters, etc. for a total of 4601 e-mails. Furthermore, these e-mails have been classified as spams (`spam = 1`) or regular e-mails (`spam = 0`). You can find more information about these data at <https://archive.ics.uci.edu/ml/datasets/Spambase>.

Your task is to evaluate the performance of Adaboost classification trees and random forests on the spam data. Specifically, provide a plot showing the error rates when the number of trees considered are 10, 20, . . . , 100. To estimate the error rates, use 2/3 of the data for training and 1/3 as hold-out test data.

To learn Adaboost classification trees, use the function `blackboost()` of the R package `mboost`. Specify the loss function corresponding to Adaboost with the parameter `family`. To learn random forests, use the function `randomForest` of the R package `randomForest`. To load the data, you may want to use the following code:

```
sp <- read.csv2("data/spambase.csv")
sp$Spam <- as.factor(sp$Spam)
```

## Solution

For trees trained using adaboost we have:

```
## Splitting data
set.seed(42)
n <- dim(sp)[1]
idxs <- sample(1:n, floor(2*n/3))
train <- sp[idxs,]
test <- sp[-idxs,]

get_missed <- function (true, predicted) {
  confusion <- table(true, predicted)
  #tn <- confusion[1,1]
  #tp <- confusion[2,2]
  fn <- confusion[1,2]
  fp <- confusion[2,1]
  total <- sum(confusion)
  #success <- (tp + tn) / total * 100
  miss <- (fp + fn) / total * 100
  return(miss)
}

# Training
nums <- seq(10,100,10)
formula <- Spam ~ .
error_rates_train_bb <- c()
error_rates_test_bb <- c()
error_rates_train_rf <- c()
error_rates_test_rf <- c()
depths <- c()
for (i in nums) {
  bb <- blackboost (
```

```

    Spam ~ .,
    data = train,
    family = AdaExp(),
    control = boost_control(mstop = i)
  )

rf <- randomForest(
  Spam ~ .,
  data = train,
  ntree = i
)

predicted <- predict(bb, train, type = "class")
miss <- get_missed(train$Spam, predicted)
error_rates_train_bb <- append(error_rates_train_bb, miss)

predicted <- predict(bb, test, type = "class")
miss <- get_missed(test$Spam, predicted)
error_rates_test_bb <- append(error_rates_test_bb, miss)

predicted <- predict(rf, train, type = "class")
miss <- get_missed(train$Spam, predicted)
error_rates_train_rf <- append(error_rates_train_rf, miss)

predicted <- predict(rf, test, type = "class")
miss <- get_missed(test$Spam, predicted)
error_rates_test_rf <- append(error_rates_test_rf, miss)

depths <- append(depths, i)
}

p <- ggplot()
p <- p + geom_line(aes(x = depths, y = error_rates_train_bb), color = "red")
p <- p + geom_line(aes(x = depths, y = error_rates_test_bb), color = "blue")
p <- p + geom_line(aes(x = depths, y = error_rates_train_rf), color = "black")
p <- p + geom_line(aes(x = depths, y = error_rates_test_rf), color = "green")
p <- p + geom_point(aes(x = depths, y = error_rates_train_bb), color = "red")
p <- p + geom_point(aes(x = depths, y = error_rates_test_bb), color = "blue")
p <- p + geom_point(aes(x = depths, y = error_rates_train_rf), color = "black")
p <- p + geom_point(aes(x = depths, y = error_rates_test_rf), color = "green")
p

```

