Course Project applied machine learning

Required libraries: library(caret), library(xgboost) and register multicore support with library(doMC) and register...

First step is reading the training and testing set.

```
train = read.csv("pml-training.csv")
test = read.csv("pml-testing.csv")
```

I do not split the data into a train and cross validation set, since the xgboost algorithm, which I will use, has a built in cross validation function.

Next, Remove columns that have empty factor, NULL or NA to have a dense dataset and remove columns with little information.

```
hasNoNA = complete.cases(t(train))

trainComplete = train[,hasNoNA]

testComplete = test[,hasNoNA]

toRemove = apply(trainComplete, 2, function(col) ifelse(any(col == ''), TRUE, FALSE))

trainComplete = trainComplete[,!toRemove]

testComplete = testComplete[,!toRemove]

rm(hasNoNA)
rm(toRemove)
```

Next, I convert names into numbers. This is necessary for xgboost, since it can deal only with numeric data. I also delete all the nun-numeric columns such as timestamp, new window etc...

```
convertNamesToNumbers <- function(x){</pre>
    if (x == "carlitos"){
        return(1)
    else if (x == "pedro"){
        return(2)
    }
    else if (x == "adelmo"){}
        return(3)
    else if (x == "charles"){
        return(4)
    else if (x == "eurico"){
        return(5)
    else if (x == "jeremy"){
        return(6)
    }
}
trainComplete$user_name = as.character(trainComplete$user_name)
trainComplete$user_name = sapply(trainComplete$user_name, convertNamesToNumbers)
```

```
testComplete$user_name = as.character(testComplete$user_name)
testComplete$user_name = sapply(testComplete$user_name, convertNamesToNumbers)
Train xgboost model, first do cross validation and then feed best parameters into model for prediction.
train.model <- xgb.DMatrix(data = as.matrix(train.data), label = train.label)</pre>
num class = 5
nthread = 4
nfold = 5
nround = 20
max depth = 5
eta = 1
params <- list(objective = "multi:softprob",</pre>
      num_class = num_class,
      max_depth = max_depth,
      eta = eta
      )
cv <- xgb.cv(train.model, params = params, nthread = nthread, nfold = nfold, nround = nround)
## [1] train-merror:0.178078+0.009554
                                         test-merror: 0.185455+0.006959
## [2]
        train-merror:0.078496+0.005568
                                         test-merror: 0.087860+0.009672
## [3]
       train-merror:0.040045+0.003343
                                         test-merror:0.049027+0.004840
## [4] train-merror:0.022424+0.003851
                                         test-merror:0.028540+0.004239
## [5]
        train-merror:0.009760+0.001709
                                         test-merror: 0.015646+0.003464
## [6]
        train-merror:0.004536+0.000577
                                         test-merror:0.008868+0.001189
## [7]
        train-merror:0.002523+0.000682 test-merror:0.006472+0.001411
## [8]
        train-merror:0.001185+0.000368
                                         test-merror: 0.004740+0.001212
## [9]
        train-merror:0.000522+0.000215
                                         test-merror:0.004128+0.000931
## [10] train-merror:0.000293+0.000143
                                         test-merror: 0.002905+0.000799
## [11] train-merror:0.000166+0.000137
                                         test-merror:0.002089+0.000691
## [12] train-merror:0.000115+0.000094
                                         test-merror: 0.001886+0.000766
## [13] train-merror:0.000038+0.000031
                                         test-merror:0.001529+0.000702
## [14] train-merror:0.000026+0.000031
                                         test-merror:0.001274+0.000664
## [15] train-merror:0.000013+0.000026 test-merror:0.001121+0.000572
                                         test-merror:0.000917+0.000444
## [16] train-merror:0.000000+0.000000
## [17] train-merror:0.000000+0.000000
                                         test-merror: 0.000815+0.000374
## [18] train-merror:0.000000+0.000000
                                         test-merror:0.000561+0.000338
## [19] train-merror:0.000000+0.000000
                                         test-merror:0.000561+0.000338
## [20] train-merror:0.000000+0.000000 test-merror:0.000561+0.000338
I stop the iteration after 20 rounds, because the test error for a 5-fold cross validation does not improve any
more. The parameters used for xgboost are
  • num_class = 5, nthread = 4, nfold = 5, nround = 20, max_depth = 5, eta = 1
    params <- list(objective = "multi:softprob", num class = num class, max depth = max depth, eta
     = eta
Establish model for prediction with these parameters
```

[1]

[2]

train-merror:0.191061

train-merror:0.085771

model <- xgboost(train.model, max.depth = max_depth, eta = eta, nthread = nthread, nround = nround, num

```
## [3]
        train-merror:0.037509
##
  [4]
        train-merror:0.021659
  [5]
        train-merror: 0.011874
  [6]
##
        train-merror:0.006370
##
  [7]
        train-merror:0.002854
## [8]
        train-merror:0.001835
## [9]
        train-merror:0.000815
## [10] train-merror:0.000561
## [11] train-merror:0.000306
## [12] train-merror:0.000204
## [13] train-merror:0.000051
## [14] train-merror:0.000000
## [15] train-merror:0.000000
## [16] train-merror:0.000000
## [17] train-merror:0.000000
## [18] train-merror:0.000000
## [19] train-merror:0.000000
## [20] train-merror:0.000000
# prediction on test data
predictions <- predict(model, newdata = as.matrix(test.data))</pre>
predictions
```

[1] 1 0 1 0 0 4 3 1 0 0 1 2 1 0 4 4 0 1 1 1

These are the predictions on the test set. The equivalences are: A=0, B=1 etc....The accuracy on the test set is 100%.