## 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:0.181302+0.007732
                                         test-merror:0.189073+0.009778
## [2]
        train-merror:0.077502+0.006340
                                         test-merror: 0.087861+0.008692
## [3]
       train-merror:0.034923+0.005706
                                         test-merror: 0.043420+0.004676
## [4] train-merror:0.016028+0.002247
                                         test-merror: 0.022475+0.002163
## [5]
        train-merror:0.009020+0.001945 test-merror:0.014932+0.001584
## [6]
        train-merror:0.004039+0.000598
                                         test-merror:0.008511+0.000862
## [7]
        train-merror:0.001860+0.000338 test-merror:0.005504+0.001835
## [8]
        train-merror:0.001032+0.000130
                                         test-merror:0.003873+0.000544
## [9]
        train-merror:0.000739+0.000200
                                         test-merror: 0.003160+0.000524
## [10] train-merror:0.000319+0.000107
                                         test-merror: 0.002599+0.000692
## [11] train-merror:0.000153+0.000065
                                         test-merror:0.001733+0.000710
## [12] train-merror:0.000089+0.000051
                                         test-merror: 0.001631+0.000731
## [13] train-merror:0.000064+0.000040
                                         test-merror: 0.001172+0.000695
## [14] train-merror:0.000013+0.000026
                                         test-merror:0.000918+0.000616
## [15] train-merror:0.000000+0.000000
                                         test-merror: 0.000918+0.000594
                                         test-merror:0.000917+0.000657
## [16] train-merror:0.000000+0.000000
## [17] train-merror:0.000000+0.000000
                                         test-merror: 0.000612+0.000525
## [18] train-merror:0.000000+0.000000
                                         test-merror:0.000561+0.000544
## [19] train-merror:0.000000+0.000000
                                         test-merror: 0.000561+0.000544
## [20] train-merror:0.000000+0.000000 test-merror:0.000510+0.000581
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
```

```
2
```

## [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%.