

# The Course Project in Practical Machine Learning

*Atsuko Yamamoto*

*April 26, 2015*

## Executive Summary

This is the Course Project for the class “Practical Machine Learning”. In this report, I predict the manner using the exercise data from accelerometers on the belt, forearm, arm and dumbbell of 6 participants. I build the prediction model to predict the objective variable “classe”.

## Result

I use caret with random forest as my model with 5 fold cross validation. The model is mtry : 27, accuracy : 0.99, OOB estimate of error rate is less than 1%.

Predictions of testing: B A B A A E D B A A B C B A E E A B B B

## Data preprocessing

```
# loading data
temp <- tempfile()
urltrain="https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv"
urltest="https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv"
download.file(urltrain,temp,"curl")
pmltrain <- read.csv(temp)
download.file(urltest,temp,"curl")
pmltest <- read.csv(temp)
dim(pmltrain);dim(pmltest)
```

```
## [1] 19622 160
```

```
## [1] 20 160
```

```
# summary(pmltrain)
```

There are a lot of missing values in csvfiles. I get rid of the columns that are independent of exercises and mostly NA.

```
# Cleaning the data
pmldata <- pmltrain[,-1:-7]
naCnt <- apply(pmldata,2,function(x) {sum(is.na(x)|x=="")})
pmldata <- pmldata[,which(naCnt < 19216)]
dim(pmldata)
```

```
## [1] 19622 53
```

## Data splitting

Devide the pmldata between training and testing.

```
library(caret)
```

```
## Loading required package: lattice
## Loading required package: ggplot2
```

```
set.seed(1000)
inTrain <- createDataPartition(y=pmldata$classe, p=0.6, list=FALSE)
training <- pmldata[inTrain,]
testing <- pmldata[-inTrain,]
dim(training); dim(testing)
```

```
## [1] 11776    53
```

```
## [1] 7846    53
```

## Training

I use caret with random forest as my model with 5 fold cross validation. Because random forests are usually one of the top performing algorithms along with boosting in any prediction contests.

```
rfmodel <- train(classe ~ .,method="rf", data=training,
                 trControl=trainControl(method="cv", number=5),
                 prox=TRUE, allowParallel=TRUE)
```

```
## Loading required package: randomForest
## randomForest 4.6-10
## Type rfNews() to see new features/changes/bug fixes.
```

```
rfmodel
```

```
## Random Forest
##
## 11776 samples
##    52 predictor
##    5 classes: 'A', 'B', 'C', 'D', 'E'
##
## No pre-processing
## Resampling: Cross-Validated (5 fold)
##
## Summary of sample sizes: 9420, 9422, 9420, 9421, 9421
##
## Resampling results across tuning parameters:
##
##   mtry  Accuracy  Kappa      Accuracy SD  Kappa SD
##    2    0.9883664  0.9852807  0.004304795  0.005450606
##   27    0.9886217  0.9856048  0.003928684  0.004973259
```

```
## 52 0.9857344 0.9819537 0.004385470 0.005550943
```

```
##
```

```
## Accuracy was used to select the optimal model using the largest value.
```

```
## The final value used for the model was mtry = 27.
```

```
rfmodel$finalModel
```

```
##
```

```
## Call:
```

```
## randomForest(x = x, y = y, mtry = param$mtry, proximity = TRUE, allowParallel = TRUE)
```

```
## Type of random forest: classification
```

```
## Number of trees: 500
```

```
## No. of variables tried at each split: 27
```

```
##
```

```
## OOB estimate of error rate: 0.84%
```

```
## Confusion matrix:
```

```
## A B C D E class.error
```

```
## A 3341 4 1 0 2 0.002090800
```

```
## B 17 2254 8 0 0 0.010969724
```

```
## C 0 13 2033 8 0 0.010223953
```

```
## D 0 3 27 1898 2 0.016580311
```

```
## E 0 2 5 7 2151 0.006466513
```

```
pred <- predict(rfmodel, testing)
```

```
table(pred, testing$classe)
```

```
##
```

```
## pred A B C D E
```

```
## A 2230 23 0 0 0
```

```
## B 2 1493 11 1 2
```

```
## C 0 2 1352 24 2
```

```
## D 0 0 5 1261 5
```

```
## E 0 0 0 0 1433
```

The model is  $mtry = 27$ , and accuracy is 0.99. OOB estimate of error rate is less than 1%. That is a good model.

## Predictions of testing

I use my prediction model to predict 20 different test cases.

```
answers <- predict(rfmodel, pmltest)
```

```
answers
```

```
## [1] B A B A A E D B A A B C B A E E A B B B
```

```
## Levels: A B C D E
```