Practical Machine Learning (Project)

TD

Monday, February 16, 2015

Introduction

In this project, we will use the data from accelerometers on the belt, forearm, arm, and dumbell of 6 participants to find patterns in their behavior. They were asked to perform barbell lifts correctly and incorrectly in 5 different ways. More information is available from the website here: (http://groupware.les.inf.puc-rio.br/har) (see the section on the Weight Lifting Exercise Dataset).

Synopsis

In this project, two methods were used to fit the training data set. Cross validation were applied. Based on the model evaluation, it reveals that model using the method of random forest is better than the model using the method of 'rpart' in terms of accuracy of prediction. Finally, the random forest model were applied to predict the new test data set (Test_final).

Load data sets

```
#setwd("E:\\Coursera\\Practical Machine Learning\\project")
#UrlTrain <- "https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv"
#UrlTest <- "https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv"
#download.file (UrlTrain, destfile = "training.csv")
#download.file (UrlTest, destfile = "testing.csv")

setwd("E:\\Coursera\\Practical Machine Learning\\project")
setTrain<-read.table(file="training.csv", header = TRUE, sep = ",")
setTest<-read.table(file="testing.csv", header = TRUE, sep = ",")</pre>
```

Preprocess data sets

Keep variables that satisfying the following criteria:

- 1. variables that indicate measures from accelerometers on the belt, forearm, arm, and dumbell
- 2. no missing values
- 3. no near zero variables

```
dim(setTrain);dim(setTest)
```

```
## [1] 19622 160
## [1] 20 160
```

```
#select the variables that relates to
#the data from accelerometers on the belt, forearm, arm, and dumbell
library(dplyr)
Train<-select(setTrain, contains("accel"),raw_timestamp_part_1,raw_timestamp_part_2, classe)</pre>
names(Train)
## [1] "total_accel_belt"
                               "var_total_accel_belt" "accel_belt_x"
## [4] "accel_belt_y"
                               "accel_belt_z"
                                                      "total_accel_arm"
## [7] "var_accel_arm"
                               "accel arm x"
                                                      "accel_arm_y"
## [10] "accel_arm_z"
                               "total_accel_dumbbell" "var_accel_dumbbell"
## [13] "accel_dumbbell_x"
                               "accel_dumbbell_y"
                                                    "accel_dumbbell_z"
## [16] "total accel forearm"
                               "var accel forearm"
                                                      "accel forearm x"
                                                     "raw_timestamp_part_1"
## [19] "accel_forearm_y"
                               "accel_forearm_z"
## [22] "raw timestamp part 2" "classe"
Test<-select(setTest, contains("accel"),raw_timestamp_part_1,raw_timestamp_part_2)
names(Test)
## [1] "total_accel_belt"
                               "var_total_accel_belt" "accel_belt_x"
                               "accel_belt_z"
## [4] "accel_belt_y"
                                                      "total_accel_arm"
## [7] "var_accel_arm"
                               "accel_arm_x"
                                                      "accel_arm_y"
                               "total_accel_dumbbell" "var_accel_dumbbell"
## [10] "accel_arm_z"
## [13] "accel_dumbbell_x"
                               "accel_dumbbell_y"
                                                      "accel_dumbbell_z"
## [16] "total_accel_forearm"
                               "var_accel_forearm"
                                                      "accel_forearm_x"
## [19] "accel forearm y"
                               "accel forearm z"
                                                      "raw_timestamp_part_1"
## [22] "raw timestamp part 2"
#check missing data and keep complete columns
VarMtest<-which(colSums(is.na(Test))>0)
Test final<-Test[,-c(VarMtest)]</pre>
include<- names(Train) %in% c("classe", names(Test[,-c(VarMtest)]))</pre>
Train final <- Train[include]</pre>
#check near zero variables
library(caret)
nsv <- nearZeroVar(Train_final, saveMetrics=TRUE)</pre>
##
                        freqRatio percentUnique zeroVar
## total_accel_belt
                        1.063160
                                      0.1477933 FALSE FALSE
                                      0.8357966 FALSE FALSE
## accel_belt_x
                        1.055412
## accel_belt_y
                                     0.7287738 FALSE FALSE
                        1.113725
## accel_belt_z
                        1.078767
                                     1.5237998 FALSE FALSE
                                     0.3363572 FALSE FALSE
## total_accel_arm
                        1.024526
                                     3.9598410 FALSE FALSE
## accel_arm_x
                        1.017341
## accel_arm_y
                                     2.7367241 FALSE FALSE
                        1.140187
## accel arm z
                                    4.0362858 FALSE FALSE
                        1.128000
## total_accel_dumbbell 1.072634
                                     0.2191418 FALSE FALSE
## accel_dumbbell_x
                        1.018018
                                     2.1659362 FALSE FALSE
## accel_dumbbell_y
                        1.053061
                                     2.3748853 FALSE FALSE
## accel_dumbbell_z
                        1.133333
                                     2.0894914 FALSE FALSE
                                    0.3567424 FALSE FALSE
## total accel forearm 1.128928
```

```
FALSE FALSE
## accel_forearm_x
                         1.126437
                                       4.0464784
## accel_forearm_y
                         1.059406
                                       5.1116094 FALSE FALSE
## accel forearm z
                          1.006250
                                       2.9558659 FALSE FALSE
## raw_timestamp_part_1
                         1.000000
                                       4.2656202
                                                   FALSE FALSE
## raw_timestamp_part_2 1.000000
                                      85.5315462
                                                   FALSE FALSE
## classe
                          1.469581
                                       0.0254816
                                                   FALSE FALSE
#double check
dim(Train_final);dim(Test_final)
## [1] 19622
## [1] 20 18
sapply(Train_final, function(x) sum(is.na(x) | x == ""))
##
       total_accel_belt
                                 accel_belt_x
                                                       accel_belt_y
##
##
           accel_belt_z
                              total_accel_arm
                                                        accel_arm_x
##
                       0
##
            accel arm y
                                  accel arm z total accel dumbbell
##
                      0
                                            0
##
       accel_dumbbell_x
                             accel_dumbbell_y
                                                   accel dumbbell z
##
                                                                  0
##
    total_accel_forearm
                              accel_forearm_x
                                                    accel_forearm_y
##
                                            0
##
        accel_forearm_z raw_timestamp_part_1 raw_timestamp_part_2
                      0
                                            0
##
##
                 classe
##
                      0
sum(!complete.cases(Train_final))
## [1] 0
sapply(Test_final, function(x) sum(is.na(x) | x == ""))
##
       total_accel_belt
                                 accel_belt_x
                                                       accel_belt_y
##
##
           accel_belt_z
                              total_accel_arm
                                                        accel_arm_x
##
##
                                  accel_arm_z total_accel_dumbbell
            accel_arm_y
##
       accel_dumbbell_x
##
                             accel_dumbbell_y
                                                   {\tt accel\_dumbbell\_z}
##
##
    total_accel_forearm
                              accel_forearm_x
                                                    accel_forearm_y
##
##
        accel_forearm_z raw_timestamp_part_1 raw_timestamp_part_2
##
```

```
sum(!complete.cases(Test_final))
## [1] 0
set.seed(1968)
library(caret)
inTrain = createDataPartition(y=Train_final$classe, p=0.6, list=FALSE)
training = Train_final[inTrain,]
testing = Train_final[-inTrain,]
dim(training);dim(testing)
Split the 'Train_final' set into 'training' \operatorname{set}(60\%) and 'testing' \operatorname{set}(40\%)
## [1] 11776
                 19
## [1] 7846 19
summary(training$classe)
                            Ε
      Α
           В
                 C
                       D
## 3348 2279 2054 1930 2165
```

Fit Models

```
library(rattle)
library(rpart.plot)

model<-train(classe~., method="rpart", data=training)
#print(model, digits=3)
#print(model$finalModel, digits=3)</pre>
```

Model1: Predicting with trees using 'rpart'

```
library(randomForest)
model2<- randomForest(classe~., data=training,proximity=TRUE)
#print(model2, digits=3)</pre>
```

Model2: Predicting with random forest

Model evaluation

```
predictions<-predict(model, newdata=testing)</pre>
confusionMatrix(predictions, testing$classe)
Model1 (method="rpart")
## Confusion Matrix and Statistics
##
             Reference
## Prediction
                Α
                      В
                           C
                                D
                                     Ε
##
           A 1125
                   150
                         175
                               59
                                    83
                              163
##
           B 619
                   883
                         561
                                  428
            С
                         208
##
                0
                      0
                                0
                                     0
##
           D
              485
                   485
                         424 1064
                                  307
            Ε
##
                 3
                      0
                           0
                                   624
##
## Overall Statistics
##
##
                  Accuracy : 0.4976
##
                    95% CI: (0.4865, 0.5087)
##
      No Information Rate: 0.2845
##
      P-Value [Acc > NIR] : < 2.2e-16
##
##
                     Kappa: 0.3718
##
  Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##
                        Class: A Class: B Class: C Class: D Class: E
## Sensitivity
                          0.5040
                                   0.5817 0.15205
                                                     0.8274 0.43273
## Specificity
                          0.9168
                                   0.7201 1.00000
                                                     0.7407
                                                             0.99953
## Pos Pred Value
                          0.7067
                                   0.3327
                                           1.00000
                                                     0.3848
                                                             0.99522
## Neg Pred Value
                          0.8230
                                  0.8777 0.84813
                                                     0.9563
                                                             0.88669
## Prevalence
                          0.2845
                                  0.1935 0.17436
                                                     0.1639
                                                             0.18379
## Detection Rate
                                   0.1125 0.02651
                          0.1434
                                                     0.1356
                                                             0.07953
## Detection Prevalence
                          0.2029
                                   0.3383
                                           0.02651
                                                     0.3524
                                                             0.07991
## Balanced Accuracy
                          0.7104
                                  0.6509 0.57602
                                                     0.7840 0.71613
predictions2<-predict(model2, newdata=testing, type="class")</pre>
confusionMatrix(predictions2, testing$classe)
Model2 (method="randomForest")
## Confusion Matrix and Statistics
##
##
             Reference
## Prediction
                Α
                           С
                                D
                                     Ε
##
            A 2231
                      3
                           0
                                0
                                     0
```

##

B 1 1513

```
С
##
                      2 1364
                                 3
##
            D
                 0
                      0
                           4 1283
                                      2
            Ε
##
                           0
                                 0 1440
##
## Overall Statistics
##
##
                  Accuracy: 0.9981
                    95% CI : (0.9968, 0.9989)
##
##
       No Information Rate: 0.2845
##
       P-Value [Acc > NIR] : < 2.2e-16
##
##
                     Kappa: 0.9976
    Mcnemar's Test P-Value : NA
##
##
## Statistics by Class:
##
##
                        Class: A Class: B Class: C Class: D Class: E
## Sensitivity
                          0.9996
                                    0.9967
                                             0.9971
                                                      0.9977
                                                                0.9986
## Specificity
                          0.9995
                                    0.9998
                                             0.9992
                                                      0.9991
                                                                1.0000
## Pos Pred Value
                          0.9987
                                    0.9993
                                             0.9963
                                                      0.9953
                                                                1.0000
## Neg Pred Value
                          0.9998
                                   0.9992
                                             0.9994
                                                      0.9995
                                                                0.9997
## Prevalence
                          0.2845
                                    0.1935
                                             0.1744
                                                      0.1639
                                                                0.1838
## Detection Rate
                          0.2843
                                    0.1928
                                                      0.1635
                                                                0.1835
                                             0.1738
## Detection Prevalence
                          0.2847
                                    0.1930
                                             0.1745
                                                      0.1643
                                                                0.1835
## Balanced Accuracy
                          0.9995
                                   0.9983
                                             0.9982
                                                      0.9984
                                                                0.9993
```

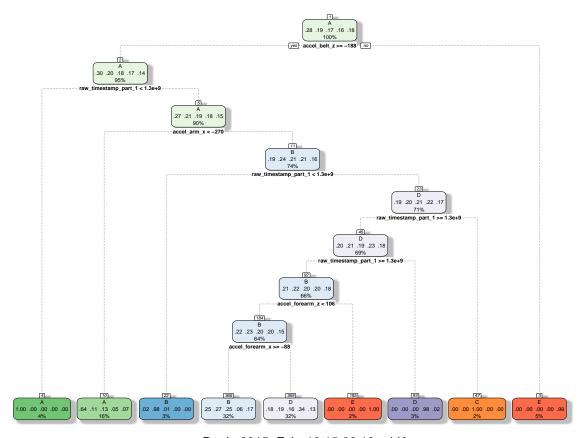
Prediction on the 'Test_final' set

```
predict_new<-predict(model2, newdata=Test_final, type="class")
predict_new

## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## 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</pre>
```

Appendix: Tree plot of the final model that using the method of 'rpart'

```
fancyRpartPlot(model$finalModel)
```



Rattle 2015-Feb-16 15:09:18 gridA