## Machine Learning\_Project

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December 26, 2015

#### BackGround

Using devices such as Jawbone Up, Nike FuelBand, and Fitbit it is now possible to collect a large amount of data about personal activity relatively inexpensively. These type of devices are part of the quantified self movement - a group of enthusiasts who take measurements about themselves regularly to improve their health, to find patterns in their behavior, or because they are tech geeks. One thing that people regularly do is quantify how much of a particular activity they do, but they rarely quantify how well they do it. In this project, your goal will be to use data from accelerometers on the belt, forearm, arm, and dumbell of 6 participants. They were asked to perform barbell lifts correctly and incorrectly in 5 different ways. More information is available from the website here: <a href="http://groupware.les.inf.puc-rio.br/har">http://groupware.les.inf.puc-rio.br/har</a> (see the section on the Weight Lifting Exercise Dataset).

## **Installing Packages**

```
library(caret)

## Warning: package 'caret' was built under R version 3.2.3

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

library(ggplot2)
library(lattice)
library(kernlab)

## Warning: package 'kernlab' was built under R version 3.2.3

library(randomForest)

## TandomForest 4.6-12
## Type rfNews() to see new features/changes/bug fixes.

library(rpart)

## Warning: package 'rpart' was built under R version 3.2.3

library(rattle)

## Warning: package 'rattle' was built under R version 3.2.3
```

```
## Rattle: A free graphical interface for data mining with R.
## Version 4.0.5 Copyright (c) 2006-2015 Togaware Pty Ltd.
## Type 'rattle()' to shake, rattle, and roll your data.

library(rpart)
library(rpart.plot)

## Warning: package 'rpart.plot' was built under R version 3.2.3
```

## Downloading datasets

```
traindata <- read.csv('R/pml-training.csv')
validdata <- read.csv('R/pml-testing.csv')</pre>
```

## **Preprocessing Operations**

### Clearing zero variance

```
set.seed(32768)
nzv <- nearZeroVar(traindata)
traindata <- traindata[-nzv]
validdata <-validdata[-nzv]</pre>
```

#### Clearing irrelevant columns

```
Cl <- grep("name|timestamp|window|X|^max|^min|^std|^amplitude", colnames(traindata), value=F)
traindata <- traindata[,-Cl]
validdata <-validdata[,-Cl]</pre>
```

#### Clearing >95% NAs

```
traindata[traindata==""] <- NA
NAs <- apply(traindata, 2, function(x) sum(is.na(x)))/nrow(traindata)
traindata <- traindata[!(NAs>0.95)]

validdata[validdata==""] <- NA
NAs <- apply(validdata, 2, function(x) sum(is.na(x)))/nrow(validdata)
validdata <- validdata[!(NAs>0.95)]
```

#### **Cross Validation**

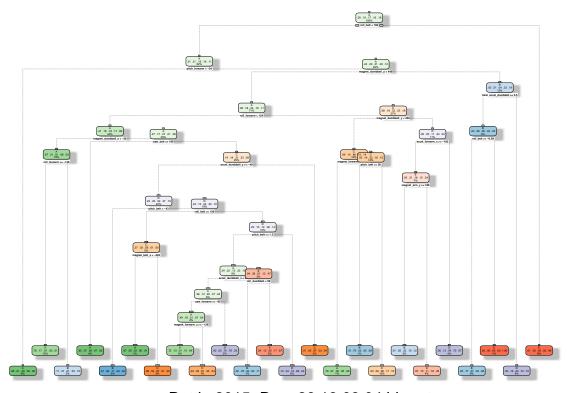
```
trainidx <- createDataPartition(traindata$classe,p=.9,list=FALSE)
traindata = traindata[trainidx,]
testdata = traindata[-trainidx,]</pre>
```

## Model Application

#### **CART**

```
model1 <- rpart(classe ~ ., data=traindata, method="class")
fancyRpartPlot(model1)</pre>
```

## Warning: labs do not fit even at cex 0.15, there may be some overplotting



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```
#rpart.plot(model1, main="Classification Tree", extra=102, under=TRUE, faclen=0)

# Predicting:
prediction1 <- predict(model1, testdata, type = "class")

# Testing:
confusionMatrix(prediction1, testdata$classe)</pre>
```

```
## Confusion Matrix and Statistics
##
##
            Reference
## Prediction A B
                              Ε
                      С
                          D
##
           A 463 45
                      5 22
                              3
##
           B 10 229 33 24 28
##
              9 31 233 44 48
           D 16 24 17 190 21
##
##
           Ε
             6 20
                      8 18 231
##
## Overall Statistics
##
                 Accuracy: 0.757
##
##
                  95% CI: (0.7364, 0.7768)
##
      No Information Rate: 0.2835
      P-Value [Acc > NIR] : < 2.2e-16
##
##
##
                   Kappa: 0.6922
## Mcnemar's Test P-Value : 1.337e-10
##
## Statistics by Class:
##
                      Class: A Class: B Class: C Class: D Class: E
##
## Sensitivity
                        0.9187 0.6562
                                        0.7872
                                                 0.6376
                                                          0.6979
## Specificity
                        0.9411 0.9335
                                        0.9109
                                                0.9473
                                                          0.9641
## Pos Pred Value
                        0.8606 0.7068
                                        0.6384
                                                 0.7090
                                                         0.8163
## Neg Pred Value
                        0.9669 0.9175
                                        0.9554
                                                 0.9285
                                                          0.9331
## Prevalence
                        0.2835 0.1963
                                        0.1665
                                                 0.1676
                                                          0.1862
## Detection Rate
                                                0.1069
                        0.2604 0.1288
                                        0.1310
                                                          0.1299
## Detection Prevalence 0.3026 0.1822
                                        0.2053
                                                 0.1507
                                                          0.1592
                        0.9299 0.7948
                                                0.7924
## Balanced Accuracy
                                        0.8490
                                                          0.8310
```

#### Random Forest

```
model2 <- randomForest(classe ~. , data=traindata, method="class")

# Predicting:
prediction2 <- predict(model2, testdata, type = "class")

# Test results on subTesting data set:
confusionMatrix(prediction2, testdata$classe)</pre>
```

```
## Confusion Matrix and Statistics
##
##
             Reference
## Prediction
                    В
                               Ε
               Α
                        C
                            D
##
            A 504
                    0
                        0
                                0
##
            В
                0 349
                        0
                            0
                                0
##
            С
                0
                    0 296
                            0
                                0
##
            D
                    0
                        0 298
                                0
                0
##
            Ε
                            0 331
##
```

```
## Overall Statistics
##
##
                  Accuracy: 1
                    95% CI : (0.9979, 1)
##
       No Information Rate: 0.2835
##
##
       P-Value [Acc > NIR] : < 2.2e-16
##
##
                     Kappa: 1
## Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##
                        Class: A Class: B Class: C Class: D Class: E
## Sensitivity
                                  1.0000
                                           1.0000
                                                              1.0000
                          1.0000
                                                     1.0000
## Specificity
                          1.0000
                                   1.0000
                                            1.0000
                                                     1.0000
                                                              1.0000
## Pos Pred Value
                          1.0000
                                   1.0000
                                            1.0000
                                                     1.0000
                                                              1.0000
## Neg Pred Value
                          1.0000
                                  1.0000
                                            1.0000
                                                     1.0000
                                                              1.0000
## Prevalence
                          0.2835
                                  0.1963
                                            0.1665
                                                     0.1676
                                                              0.1862
## Detection Rate
                          0.2835
                                  0.1963
                                            0.1665
                                                     0.1676
                                                              0.1862
## Detection Prevalence
                          0.2835
                                   0.1963
                                            0.1665
                                                     0.1676
                                                              0.1862
## Balanced Accuracy
                          1.0000
                                   1.0000
                                            1.0000
                                                     1.0000
                                                              1.0000
```

# #Predictor importance importance(model2)

##		MeanDecreaseGini
##	roll belt	1110.50327
##	pitch_belt	638.51628
	yaw_belt	820.50441
##	total_accel_belt	205.12620
##		88.89922
	gyros_belt_x	107.75418
	gyros_belt_y	
##	gyros_belt_z	277.71754
##	accel_belt_x	106.01851
##	accel_belt_y	110.26759
	accel_belt_z	400.36822
##	magnet_belt_x	229.91302
##	magnet_belt_y	320.78291
##	magnet_belt_z	350.11253
##	roll_arm	285.84205
##	pitch_arm	157.76993
##	yaw_arm	221.17800
##	total_accel_arm	89.83605
##	gyros_arm_x	115.12857
##	gyros_arm_y	128.18081
##	gyros_arm_z	51.82840
##	accel_arm_x	206.81301
##	accel_arm_y	135.33070
##	accel_arm_z	122.28596
##	magnet_arm_x	250.80318
##	magnet_arm_y	197.60666
##	magnet_arm_z	164.44469
##	roll_dumbbell	363.36475
##	pitch_dumbbell	155.10177

```
## yaw_dumbbell
                               224.88719
## total_accel_dumbbell
                               251.09667
## gyros dumbbell x
                               111.96948
## gyros_dumbbell_y
                               223.33236
## gyros_dumbbell_z
                               73.77542
## accel dumbbell x
                              232.33825
## accel dumbbell y
                               353.47274
## accel_dumbbell_z
                               296.68872
## magnet_dumbbell_x
                               449.99339
## magnet_dumbbell_y
                               596.63281
## magnet_dumbbell_z
                               681.26192
## roll_forearm
                               563.72357
## pitch_forearm
                               662,29875
## yaw_forearm
                              160.38070
## total_accel_forearm
                               98.29794
## gyros_forearm_x
                               68.43074
## gyros_forearm_y
                             115.76433
## gyros forearm z
                              72.76006
## accel_forearm_x
                              292.74951
## accel_forearm_y
                              129.26252
## accel_forearm_z
                              225.34973
## magnet_forearm_x
                             199.81598
## magnet_forearm_y
                              201.39052
## magnet_forearm_z
                               265.29076
```

## Predicting on Validation Data Set - FINAL

```
final <- predict(model2, validdata, type="class")
final

## 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>
```

#### Write files for submission

```
pml_write_files = function(x){
    n = length(x)
    for(i in 1:n){
        filename = paste0("problem_id_",i,".txt")
        write.table(x[i],file=filename,quote=FALSE,row.names=FALSE,col.names=FALSE)
    }
}
pml_write_files(final)
```