

# Prediction Assignment Writeup

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## Overview

This document summarizes the work done for the *Prediction Assignment Writeup* project for the *Coursera Practical Machine Learning* course. It's created using the functionalities of the *knitr* package in *RStudio* using the actual analysis code. The repository for this work can be found at <https://github.com/amete/PracticalMachineLearningAssignment>.

## 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 dumbbell 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: <http://groupware.les.inf.puc-rio.br/har> (see the section on the Weight Lifting Exercise Dataset).

## Analysis

```
require(data.table)
require(dplyr)
require(caret)

training <- tbl_df(fread("training.csv", na.strings=c('#DIV/0!', '', 'NA')))
testing  <- tbl_df(fread("testing.csv", na.strings=c('#DIV/0!', '', 'NA')))

set.seed(1234)
trainingDS <- createDataPartition( y = training$classe,
                                   p = 0.7,
                                   list = FALSE)
actual.training <- training[trainingDS,] actual.validation <-
  training[-trainingDS,]

nzv <- nearZeroVar(actual.training)
actual.training <- actual.training[, -nzv]
actual.validation <- actual.validation[, -nzv]

mostlyNA <- sapply(actual.training, function(x) mean(is.na(x))) > 0.95
actual.training <- actual.training[, mostlyNA==FALSE]
actual.validation <- actual.validation[, mostlyNA==FALSE]

actual.training <- actual.training[, -(1:5)]
actual.validation <- actual.validation[, -(1:5)]

set.seed(1234)
modelRF <- train( classe ~.,
  data = actual.training,
  method = "rf",
  trControl = trainControl(method="cv", number=3) )

set.seed(1234)
modelBM <- train( classe ~.,
  data = actual.training,
  method = "gbm",
  trControl = trainControl(method="repeatedcv", number = 5, repeats = 1),
  verbose = FALSE)
```

```

prediction.validation.rf <- predict(modelRF,actual.validation)
conf.matrix.rf <- confusionMatrix(prediction.validation.rf,actual.validation$classe)
print(conf.matrix.rf)

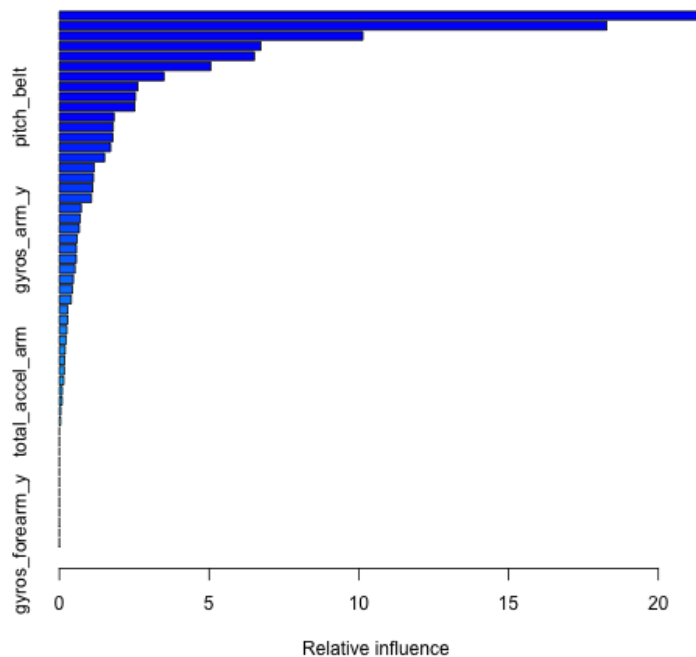
## Confusion Matrix and Statistics
##
##           Reference
## Prediction   A    B    C    D    E
##           A 1674    4    0    0    0
##           B    0 1134    4    0    0
##           C    0    1 1022    1    0
##           D    0    0    0  963    2
##           E    0    0    0    0 1080
##
##
## Overall Statistics
##
##           Accuracy : 0.9
##           95% CI : (0.9964, 0.998
##           No Information Rate : 0.28
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9974
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      1.0000  0.9956  0.9961  0.9990  0.9982
## Specificity      0.9991  0.9992  0.9996  0.9996  1.0000
## Pos Pred Value   0.9976  0.9965  0.9980  0.9979  1.0000
## Neg Pred Value   1.0000  0.9989  0.9992  0.9998  0.9996
## Prevalence       0.2845  0.1935  0.1743  0.1638  0.1839
## Detection Rate   0.2845  0.1927  0.1737  0.1636  0.1835
## Detection Prevalence 0.2851  0.1934  0.1740  0.1640  0.1835
## Balanced Accuracy 0.9995  0.9974  0.9978  0.9993  0.9991

prediction.validation.bm <- predict(modelBM,actual.validation)
conf.matrix.bm <- confusionMatrix(prediction.validation.bm,actual.validation$classe)
print(conf.matrix.bm)

## Confusion Matrix and Statistics
##
##           Reference
## Prediction   A    B    C    D    E
##           A 1673   14    0    0    0
##           B    1 1107    9    7    1
##           C    0   15 1017   11    3
##           D    0    3    0  946    9
##           E    0    0    0    0 1069
##
## Overall Statistics
##
##           Accuracy : 0.9876
##           95% CI : (0.9844, 0.9903)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9843
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9994  0.9719  0.9912  0.9813  0.9880
## Specificity      0.9967  0.9962  0.9940  0.9976  1.0000
## Pos Pred Value   0.9917  0.9840  0.9723  0.9875  1.0000
## Neg Pred Value   0.9998  0.9933  0.9981  0.9963  0.9973
## Prevalence       0.2845  0.1935  0.1743  0.1638  0.1839
## Detection Rate   0.2843  0.1881  0.1728  0.1607  0.1816
## Detection Prevalence 0.2867  0.1912  0.1777  0.1628  0.1816
## Balanced Accuracy 0.9980  0.9841  0.9926  0.9894  0.9940

```

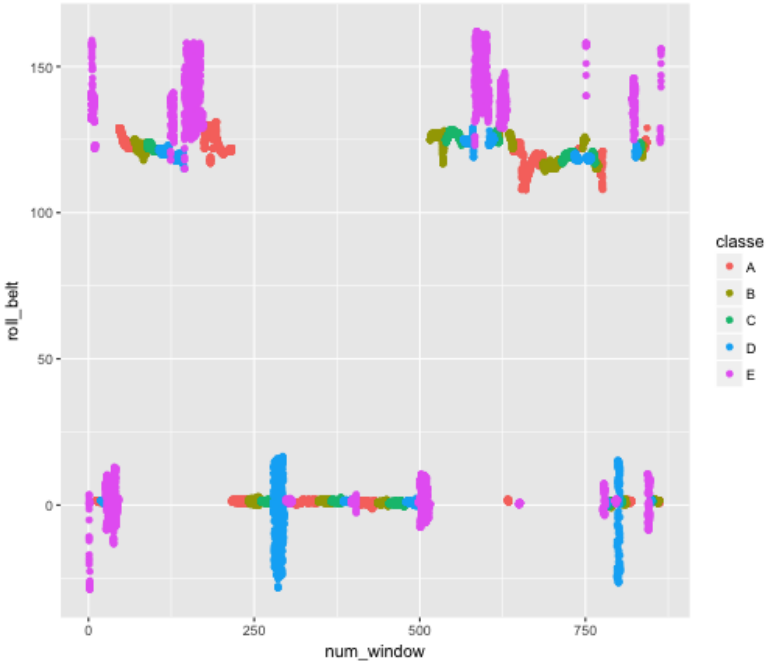
```
print(summary(modelBM))
```



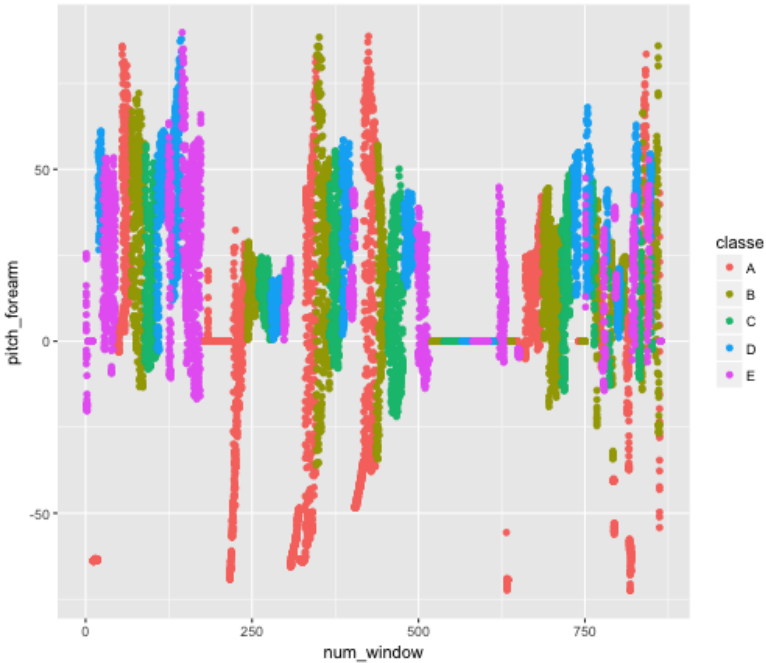
```
##               var      rel.inf
## num_window      num_window 21.35545512
## roll_belt        roll_belt 18.27724938
## pitch_forearm    pitch_forearm 10.13583112
## magnet_dumbbell_z magnet_dumbbell_z 6.72541291
## yaw_belt         yaw_belt 6.51200291
## magnet_dumbbell_y magnet_dumbbell_y 5.05649194
## roll_forearm     roll_forearm 3.49549946
## accel_forearm_x  accel_forearm_x 2.61845321
## magnet_belt_z    magnet_belt_z 2.53921239
## pitch_belt       pitch_belt 2.51716927
## gyros_belt_z     gyros_belt_z 1.82833065
## gyros_dumbbell_y gyros_dumbbell_y 1.79600229
## accel_dumbbell_z accel_dumbbell_z 1.78978003
## roll_dumbbell    roll_dumbbell 1.71291149
## accel_dumbbell_y accel_dumbbell_y 1.51172485
## yaw_arm          yaw_arm 1.16790745
## magnet_forearm_z magnet_forearm_z 1.13838373
## accel_forearm_z  accel_forearm_z 1.11085165
## accel_dumbbell_x accel_dumbbell_x 1.07112326
## magnet_belt_y    magnet_belt_y 0.73220505
## roll_arm         roll_arm 0.69769807
## magnet_arm_z     magnet_arm_z 0.66713578
## gyros_arm_y      gyros_arm_y 0.59235410
## magnet_belt_x    magnet_belt_x 0.56962809
## accel_belt_z     accel_belt_z 0.56689770
## gyros_belt_y     gyros_belt_y 0.52256127
## magnet_arm_x     magnet_arm_x 0.46724034
## magnet_dumbbell_x magnet_dumbbell_x 0.43441017
## total_accel_dumbbell total_accel_dumbbell 0.39610082
## magnet_forearm_x magnet_forearm_x 0.28197085
## total_accel_forearm total_accel_forearm 0.27783955
## gyros_dumbbell_x gyros_dumbbell_x 0.25152542
## magnet_arm_y     magnet_arm_y 0.22235172
## pitch_dumbbell   pitch_dumbbell 0.19453168
## accel_arm_z      accel_arm_z 0.17252396
## accel_forearm_y  accel_forearm_y 0.16660947
## gyros_belt_x     gyros_belt_x 0.14726266
## total_accel_arm  total_accel_arm 0.09844732
```

```
## gyros_forearm_z      gyros_forearm_z  0.09281553
## gyros_dumbbell_z     gyros_dumbbell_z  0.04731855
## magnet_forearm_y     magnet_forearm_y  0.04077878
## total_accel_belt     total_accel_belt  0.00000000
## accel_belt_x         accel_belt_x     0.00000000
## accel_belt_y         accel_belt_y     0.00000000
## pitch_arm            pitch_arm        0.00000000
## gyros_arm_x          gyros_arm_x      0.00000000
## gyros_arm_z          gyros_arm_z      0.00000000
## accel_arm_x          accel_arm_x      0.00000000
## accel_arm_y          accel_arm_y      0.00000000
## yaw_dumbbell         yaw_dumbbell     0.00000000
## yaw_forearm          yaw_forearm      0.00000000
## gyros_forearm_x      gyros_forearm_x  0.00000000
## gyros_forearm_y      gyros_forearm_y  0.00000000
```

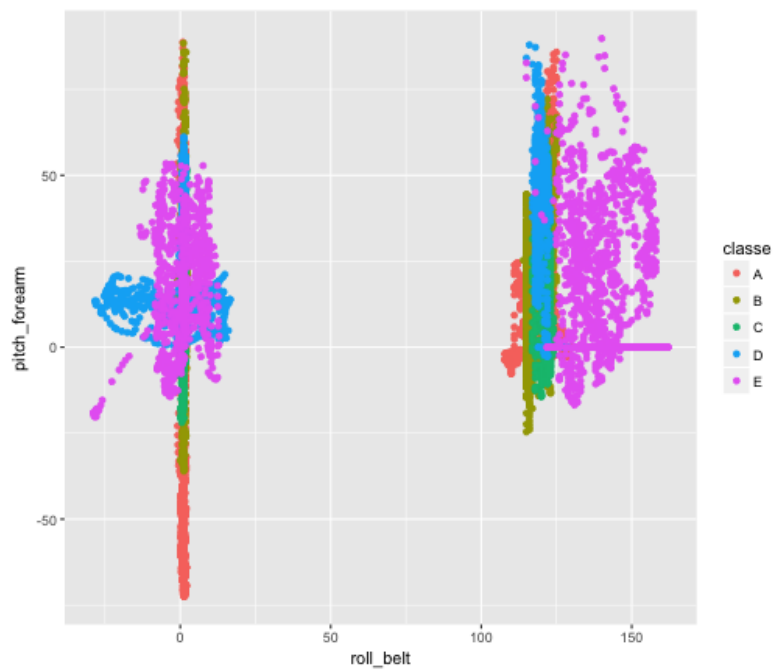
```
qplot(num_window, roll_belt , data = actual.training, col = classe)
```



```
qplot(num_window, pitch_forearm, data = actual.training, col = classe)
```



```
qplot(roll_belt , pitch_forearm, data = actual.training, col = classe)
```



```
prediction.testing.rf <- predict(modelRF,testing)
print(prediction.testing.rf)
```

```
## [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
```