

# Predicting the manner and quality of Barbell Lifts using data from Fitness Devices

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## Introduction

The aim of this project is to build a predictive model to know how a particular user is lifting weights based on data obtained from an accelerometer.

The dataset consists of 5 classes:

Class A => The subject is lifting weights exactly according to the specifications.

Class B => Throwing the elbow to the front.

Class C => Lifting the dumbbell only halfway.

Class D => Lowering the dumbbell only halfway.

Class E => Throwing the hips to the front.

Further details about the dataset can be found using the following URL: <http://groupware.les.inf.puc-rio.br/har>

## Getting The Data

The file “pml-training” will be used as the training set. The file “pml-testing” is a data set without the classes i.e. the classes will be predicted using the model.

```
if(!file.exists("pml-training.csv"))
{
  download.file("https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv", "pml-training.csv")
}
dataset <- read.csv("pml-training.csv", na.strings = c("NA", ""))
if(!file.exists("pml-testing.csv"))
{
  download.file("https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv", "pml-testing.csv")
}
validation <- read.csv("pml-testing.csv")
```

## Data Preprocessing

The necessary packages are being imported below.

```
library(caret)
```

```
## Warning: package 'caret' was built under R version 3.6.3
```

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

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

```
## Warning: package 'ggplot2' was built under R version 3.6.3
```

```
library(randomForest)
```

```
## Warning: package 'randomForest' was built under R version 3.6.3
```

```
## randomForest 4.6-14
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
##
```

```
## Attaching package: 'randomForest'
```

```
## The following object is masked from 'package:ggplot2':
```

```
##
```

```
##     margin
```

A suitable seed is being set below to ensure reproducibility.

```
set.seed(17)
```

Once the data is partitioned, 70% of it will go to the training set and the rest will be the test set.

```
inTrain = createDataPartition(y=dataset$classe, p=0.7, list=FALSE)
training = dataset[inTrain,]
testing = dataset[-inTrain,]
```

The NA entries are being eliminated below.

```
naColumns = sapply(training, function(x) {sum(is.na(x))}) #Make a vector of all the columns and the number of NA entries
naColumns
```

```
##           X           user_name      raw_timestamp_part_1
##           0              0              0
## raw_timestamp_part_2      cvtd_timestamp      new_window
##           0              0              0
##      num_window      roll_belt      pitch_belt
##           0              0              0
##      yaw_belt      total_accel_belt      kurtosis_roll_belt
##           0              0              13443
##      kurtosis_pitch_belt      kurtosis_yaw_belt      skewness_roll_belt
##           13443              13443              13443
##      skewness_roll_belt.1      skewness_yaw_belt      max_roll_belt
##           13443              13443              13443
##      max_pitch_belt      max_yaw_belt      min_roll_belt
##           13443              13443              13443
##      min_pitch_belt      min_yaw_belt      amplitude_roll_belt
##           13443              13443              13443
```

##	amplitude_pitch_belt	amplitude_yaw_belt	var_total_accel_belt
##	13443	13443	13443
##	avg_roll_belt	stddev_roll_belt	var_roll_belt
##	13443	13443	13443
##	avg_pitch_belt	stddev_pitch_belt	var_pitch_belt
##	13443	13443	13443
##	avg_yaw_belt	stddev_yaw_belt	var_yaw_belt
##	13443	13443	13443
##	gyros_belt_x	gyros_belt_y	gyros_belt_z
##	0	0	0
##	accel_belt_x	accel_belt_y	accel_belt_z
##	0	0	0
##	magnet_belt_x	magnet_belt_y	magnet_belt_z
##	0	0	0
##	roll_arm	pitch_arm	yaw_arm
##	0	0	0
##	total_accel_arm	var_accel_arm	avg_roll_arm
##	0	13443	13443
##	stddev_roll_arm	var_roll_arm	avg_pitch_arm
##	13443	13443	13443
##	stddev_pitch_arm	var_pitch_arm	avg_yaw_arm
##	13443	13443	13443
##	stddev_yaw_arm	var_yaw_arm	gyros_arm_x
##	13443	13443	0
##	gyros_arm_y	gyros_arm_z	accel_arm_x
##	0	0	0
##	accel_arm_y	accel_arm_z	magnet_arm_x
##	0	0	0
##	magnet_arm_y	magnet_arm_z	kurtosis_roll_arm
##	0	0	13443
##	kurtosis_pitch_arm	kurtosis_yaw_arm	skewness_roll_arm
##	13443	13443	13443
##	skewness_pitch_arm	skewness_yaw_arm	max_roll_arm
##	13443	13443	13443
##	max_pitch_arm	max_yaw_arm	min_roll_arm
##	13443	13443	13443
##	min_pitch_arm	min_yaw_arm	amplitude_roll_arm
##	13443	13443	13443
##	amplitude_pitch_arm	amplitude_yaw_arm	roll_dumbbell
##	13443	13443	0
##	pitch_dumbbell	yaw_dumbbell	kurtosis_roll_dumbbell
##	0	0	13443
##	kurtosis_pitch_dumbbell	kurtosis_yaw_dumbbell	skewness_roll_dumbbell
##	13443	13443	13443
##	skewness_pitch_dumbbell	skewness_yaw_dumbbell	max_roll_dumbbell
##	13443	13443	13443
##	max_pitch_dumbbell	max_yaw_dumbbell	min_roll_dumbbell
##	13443	13443	13443
##	min_pitch_dumbbell	min_yaw_dumbbell	amplitude_roll_dumbbell
##	13443	13443	13443
##	amplitude_pitch_dumbbell	amplitude_yaw_dumbbell	total_accel_dumbbell
##	13443	13443	0
##	var_accel_dumbbell	avg_roll_dumbbell	stddev_roll_dumbbell
##	13443	13443	13443

```
##      var_roll_dumbbell      avg_pitch_dumbbell      stddev_pitch_dumbbell
##      13443                13443                13443
##      var_pitch_dumbbell      avg_yaw_dumbbell      stddev_yaw_dumbbell
##      13443                13443                13443
##      var_yaw_dumbbell      gyros_dumbbell_x      gyros_dumbbell_y
##      13443                0                    0
##      gyros_dumbbell_z      accel_dumbbell_x      accel_dumbbell_y
##      0                    0                    0
##      accel_dumbbell_z      magnet_dumbbell_x      magnet_dumbbell_y
##      0                    0                    0
##      magnet_dumbbell_z      roll_forearm      pitch_forearm
##      0                    0                    0
##      yaw_forearm      kurtosis_roll_forearm      kurtosis_pitch_forearm
##      0                13443                13443
##      kurtosis_yaw_forearm      skewness_roll_forearm      skewness_pitch_forearm
##      13443                13443                13443
##      skewness_yaw_forearm      max_roll_forearm      max_pitch_forearm
##      13443                13443                13443
##      max_yaw_forearm      min_roll_forearm      min_pitch_forearm
##      13443                13443                13443
##      min_yaw_forearm      amplitude_roll_forearm      amplitude_pitch_forearm
##      13443                13443                13443
##      amplitude_yaw_forearm      total_accel_forearm      var_accel_forearm
##      13443                0                    13443
##      avg_roll_forearm      stddev_roll_forearm      var_roll_forearm
##      13443                13443                13443
##      avg_pitch_forearm      stddev_pitch_forearm      var_pitch_forearm
##      13443                13443                13443
##      avg_yaw_forearm      stddev_yaw_forearm      var_yaw_forearm
##      13443                13443                13443
##      gyros_forearm_x      gyros_forearm_y      gyros_forearm_z
##      0                    0                    0
##      accel_forearm_x      accel_forearm_y      accel_forearm_z
##      0                    0                    0
##      magnet_forearm_x      magnet_forearm_y      magnet_forearm_z
##      0                    0                    0
##      classe
##      0
```

```
columnsWithNA = names(naColumns[naColumns > 0]) #Vector with all the columns that has NA values
training = training[, !names(training) %in% columnsWithNA] #Remove those columns from the training set
names(training)
```

```
## [1] "X" "user_name" "raw_timestamp_part_1"
## [4] "raw_timestamp_part_2" "cvt_d_timestamp" "new_window"
## [7] "num_window" "roll_belt" "pitch_belt"
## [10] "yaw_belt" "total_accel_belt" "gyros_belt_x"
## [13] "gyros_belt_y" "gyros_belt_z" "accel_belt_x"
## [16] "accel_belt_y" "accel_belt_z" "magnet_belt_x"
## [19] "magnet_belt_y" "magnet_belt_z" "roll_arm"
## [22] "pitch_arm" "yaw_arm" "total_accel_arm"
## [25] "gyros_arm_x" "gyros_arm_y" "gyros_arm_z"
## [28] "accel_arm_x" "accel_arm_y" "accel_arm_z"
## [31] "magnet_arm_x" "magnet_arm_y" "magnet_arm_z"
```

```
## [34] "roll_dumbbell"      "pitch_dumbbell"      "yaw_dumbbell"
## [37] "total_accel_dumbbell" "gyros_dumbbell_x"    "gyros_dumbbell_y"
## [40] "gyros_dumbbell_z"    "accel_dumbbell_x"    "accel_dumbbell_y"
## [43] "accel_dumbbell_z"    "magnet_dumbbell_x"   "magnet_dumbbell_y"
## [46] "magnet_dumbbell_z"   "roll_forearm"        "pitch_forearm"
## [49] "yaw_forearm"         "total_accel_forearm" "gyros_forearm_x"
## [52] "gyros_forearm_y"     "gyros_forearm_z"     "accel_forearm_x"
## [55] "accel_forearm_y"     "accel_forearm_z"     "magnet_forearm_x"
## [58] "magnet_forearm_y"    "magnet_forearm_z"    "classe"
```

*#Remove unnecessary columns (the first 7 columns)*

```
training <- training[, !names(training) %in% c("X", "user_name", "raw_timestamp_part_1", "raw_timestamp_part_2")]
```

The same procedure is followed for the validation set.

```
naColumns = sapply(validation, function(x) {sum(is.na(x))}) #Make a vector of all the columns and the number of NA values
columnsWithNA = names(naColumns[naColumns > 0]) #Vector with all the columns that has NA values
validation = validation[, !names(validation) %in% columnsWithNA] #Remove those columns from the training set.
validation <- validation[, !names(validation) %in% c("X", "user_name", "raw_timestamp_part_1", "raw_timestamp_part_2")]
```

The same procedure is followed for the testing set.

```
naColumns = sapply(testing, function(x) {sum(is.na(x))}) #Make a vector of all the columns and the number of NA values
columnsWithNA = names(naColumns[naColumns > 0]) #Vector with all the columns that has NA values
testing = testing[, !names(testing) %in% columnsWithNA] #Remove those columns from the training set.
testing <- testing[, !names(testing) %in% c("X", "user_name", "raw_timestamp_part_1", "raw_timestamp_part_2")]
```

Next, the predictive model is built using Random Forest.

```
model <- randomForest(classe ~ ., data=training, ntree = 50)
predictions <- predict(model, testing)
confusionMatrix(predictions, testing$classe)
```

## Confusion Matrix and Statistics

```
##
##           Reference
## Prediction    A    B    C    D    E
##           A 1671   10    0    0    0
##           B    3 1124    2    0    0
##           C    0    5 1024   12    0
##           D    0    0    0  950    2
##           E    0    0    0    2 1080
##
```

## Overall Statistics

```
##
##           Accuracy : 0.9939
##           95% CI : (0.9915, 0.9957)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9923
##
```

```
## McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9982  0.9868  0.9981  0.9855  0.9982
## Specificity      0.9976  0.9989  0.9965  0.9996  0.9996
## Pos Pred Value   0.9941  0.9956  0.9837  0.9979  0.9982
## Neg Pred Value   0.9993  0.9968  0.9996  0.9972  0.9996
## Prevalence       0.2845  0.1935  0.1743  0.1638  0.1839
## Detection Rate   0.2839  0.1910  0.1740  0.1614  0.1835
## Detection Prevalence 0.2856  0.1918  0.1769  0.1618  0.1839
## Balanced Accuracy 0.9979  0.9929  0.9973  0.9925  0.9989
```

```
modelAcc <- confusionMatrix(predictions, testing$classe)$overall[[1]]
```

The model is 0.9938828 accurate.

Now, the unknown classes of the validation set are predicted.

```
predictions <- predict(model, validation)
predictions
```

```
##  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
```