

Practical Machine Learning Project

Angelika

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Problem Description (from Instructor)

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 types 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, the goal is 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).

The goal of your project is to predict the manner in which they did the exercise. This is the “classe” variable in the training set.

Loading the data

The code below creates a “coursera_practical_machine_learning” directory within user’s current working directory, downloads both the train and test datafiles directly from the source, and reads them in using the data.table package. The first few entries of the dataset are presented in the output below.

```
library("data.table")

if (!file.exists("coursera_practical_machine_learning")) {dir.create("coursera
_practical_machine_learning")}

fileUrl <- "https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.
csv"

download.file(fileUrl, destfile = "./coursera_practical_machine_learning/pml-
training.csv", method = "curl")

train<-read.csv("./coursera_practical_machine_learning/pml-training.csv", hea
der=TRUE)

fileUrl <- "https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.c
sv"

download.file(fileUrl, destfile = "./coursera_practical_machine_learning/pml-
testing.csv", method = "curl")
```

```
test<-read.csv("../coursera_practical_machine_learning/pml-testing.csv", header=TRUE)
```

```
head(train)
```

```
##   X user_name raw_timestamp_part_1 raw_timestamp_part_2   cvtd_timestamp
## 1 1  carlitos          1323084231          788290 05/12/2011 11:23
## 2 2  carlitos          1323084231          808298 05/12/2011 11:23
## 3 3  carlitos          1323084231          820366 05/12/2011 11:23
## 4 4  carlitos          1323084232          120339 05/12/2011 11:23
## 5 5  carlitos          1323084232          196328 05/12/2011 11:23
## 6 6  carlitos          1323084232          304277 05/12/2011 11:23
##   new_window num_window roll_belt pitch_belt yaw_belt total_accel_belt
## 1         no          11      1.41      8.07    -94.4                3
## 2         no          11      1.41      8.07    -94.4                3
## 3         no          11      1.42      8.07    -94.4                3
## 4         no          12      1.48      8.05    -94.4                3
## 5         no          12      1.48      8.07    -94.4                3
## 6         no          12      1.45      8.06    -94.4                3
##   kurtosis_roll_belt kurtosis_picth_belt kurtosis_yaw_belt
## 1
## 2
## 3
## 4
## 5
## 6
##   skewness_roll_belt skewness_roll_belt.1 skewness_yaw_belt max_roll_belt
## 1
## 2
## 3
## 4
## 5
## 6
##   max_picth_belt max_yaw_belt min_roll_belt min_pitch_belt min_yaw_belt
## 1              NA              NA              NA
```

## 2	NA	NA	NA		
## 3	NA	NA	NA		
## 4	NA	NA	NA		
## 5	NA	NA	NA		
## 6	NA	NA	NA		
##	amplitude_roll_belt	amplitude_pitch_belt	amplitude_yaw_belt		
## 1	NA	NA			
## 2	NA	NA			
## 3	NA	NA			
## 4	NA	NA			
## 5	NA	NA			
## 6	NA	NA			
##	var_total_accel_belt	avg_roll_belt	stddev_roll_belt	var_roll_belt	
## 1	NA	NA	NA	NA	
## 2	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	
## 6	NA	NA	NA	NA	
##	avg_pitch_belt	stddev_pitch_belt	var_pitch_belt	avg_yaw_belt	
## 1	NA	NA	NA	NA	
## 2	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	
## 6	NA	NA	NA	NA	
##	stddev_yaw_belt	var_yaw_belt	gyros_belt_x	gyros_belt_y	gyros_belt_z
## 1	NA	NA	0.00	0.00	-0.02
## 2	NA	NA	0.02	0.00	-0.02
## 3	NA	NA	0.00	0.00	-0.02
## 4	NA	NA	0.02	0.00	-0.03
## 5	NA	NA	0.02	0.02	-0.02
## 6	NA	NA	0.02	0.00	-0.02
##	accel_belt_x	accel_belt_y	accel_belt_z	magnet_belt_x	magnet_belt_y

## 1	-21	4	22	-3	599	
## 2	-22	4	22	-7	608	
## 3	-20	5	23	-2	600	
## 4	-22	3	21	-6	604	
## 5	-21	2	24	-6	600	
## 6	-21	4	21	0	603	
##	magnet_belt_z	roll_arm	pitch_arm	yaw_arm	total_accel_arm	var_accel_arm
## 1	-313	-128	22.5	-161	34	NA
## 2	-311	-128	22.5	-161	34	NA
## 3	-305	-128	22.5	-161	34	NA
## 4	-310	-128	22.1	-161	34	NA
## 5	-302	-128	22.1	-161	34	NA
## 6	-312	-128	22.0	-161	34	NA
##	avg_roll_arm	stddev_roll_arm	var_roll_arm	avg_pitch_arm	stddev_pitch_arm	
## 1	NA	NA	NA	NA	NA	NA
## 2	NA	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA	NA
##	var_pitch_arm	avg_yaw_arm	stddev_yaw_arm	var_yaw_arm	gyros_arm_x	
## 1	NA	NA	NA	NA	0.00	
## 2	NA	NA	NA	NA	0.02	
## 3	NA	NA	NA	NA	0.02	
## 4	NA	NA	NA	NA	0.02	
## 5	NA	NA	NA	NA	0.00	
## 6	NA	NA	NA	NA	0.02	
##	gyros_arm_y	gyros_arm_z	accel_arm_x	accel_arm_y	accel_arm_z	magnet_arm_x
## 1	0.00	-0.02	-288	109	-123	-368
## 2	-0.02	-0.02	-290	110	-125	-369
## 3	-0.02	-0.02	-289	110	-126	-368
## 4	-0.03	0.02	-289	111	-123	-372
## 5	-0.03	0.00	-289	111	-123	-374
## 6	-0.03	0.00	-289	111	-122	-369

	magnet_arm_y	magnet_arm_z	kurtosis_roll_arm	kurtosis_pitch_arm	
## 1	337	516			
## 2	337	513			
## 3	344	513			
## 4	344	512			
## 5	337	506			
## 6	342	513			

	kurtosis_yaw_arm	skewness_roll_arm	skewness_pitch_arm	skewness_yaw_arm	
## 1					
## 2					
## 3					
## 4					
## 5					
## 6					

	max_roll_arm	max_pitch_arm	max_yaw_arm	min_roll_arm	min_pitch_arm
## 1	NA	NA	NA	NA	NA
## 2	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA

	min_yaw_arm	amplitude_roll_arm	amplitude_pitch_arm	amplitude_yaw_arm	
## 1	NA	NA	NA	NA	
## 2	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	
## 6	NA	NA	NA	NA	

	roll_dumbbell	pitch_dumbbell	yaw_dumbbell	kurtosis_roll_dumbbell
## 1	13.05217	-70.49400	-84.87394	
## 2	13.13074	-70.63751	-84.71065	
## 3	12.85075	-70.27812	-85.14078	
## 4	13.43120	-70.39379	-84.87363	
## 5	13.37872	-70.42856	-84.85306	

## 6	13.38246	-70.81759	-84.46500
##	kurtosis_picth_dumbbell	kurtosis_yaw_dumbbell	skewness_roll_dumbbell
## 1			
## 2			
## 3			
## 4			
## 5			
## 6			
##	skewness_pitch_dumbbell	skewness_yaw_dumbbell	max_roll_dumbbell
## 1			NA
## 2			NA
## 3			NA
## 4			NA
## 5			NA
## 6			NA
##	max_picth_dumbbell	max_yaw_dumbbell	min_roll_dumbbell
## 1	NA		NA
## 2	NA		NA
## 3	NA		NA
## 4	NA		NA
## 5	NA		NA
## 6	NA		NA
##	min_yaw_dumbbell	amplitude_roll_dumbbell	amplitude_pitch_dumbbell
## 1		NA	NA
## 2		NA	NA
## 3		NA	NA
## 4		NA	NA
## 5		NA	NA
## 6		NA	NA
##	amplitude_yaw_dumbbell	total_accel_dumbbell	var_accel_dumbbell
## 1		37	NA
## 2		37	NA
## 3		37	NA
## 4		37	NA

## 5		37	NA	
## 6		37	NA	
##	avg_roll_dumbbell	stddev_roll_dumbbell	var_roll_dumbbell	
## 1	NA	NA	NA	
## 2	NA	NA	NA	
## 3	NA	NA	NA	
## 4	NA	NA	NA	
## 5	NA	NA	NA	
## 6	NA	NA	NA	
##	avg_pitch_dumbbell	stddev_pitch_dumbbell	var_pitch_dumbbell	
## 1	NA	NA	NA	
## 2	NA	NA	NA	
## 3	NA	NA	NA	
## 4	NA	NA	NA	
## 5	NA	NA	NA	
## 6	NA	NA	NA	
##	avg_yaw_dumbbell	stddev_yaw_dumbbell	var_yaw_dumbbell	gyros_dumbbell_x
## 1	NA	NA	NA	0
## 2	NA	NA	NA	0
## 3	NA	NA	NA	0
## 4	NA	NA	NA	0
## 5	NA	NA	NA	0
## 6	NA	NA	NA	0
##	gyros_dumbbell_y	gyros_dumbbell_z	accel_dumbbell_x	accel_dumbbell_y
## 1	-0.02	0.00	-234	47
## 2	-0.02	0.00	-233	47
## 3	-0.02	0.00	-232	46
## 4	-0.02	-0.02	-232	48
## 5	-0.02	0.00	-233	48
## 6	-0.02	0.00	-234	48
##	accel_dumbbell_z	magnet_dumbbell_x	magnet_dumbbell_y	magnet_dumbbell_z
## 1	-271	-559	293	-65
## 2	-269	-555	296	-64
## 3	-270	-561	298	-63

## 4	-269	-552	303	-60
## 5	-270	-554	292	-68
## 6	-269	-558	294	-66
##	roll_forearm	pitch_forearm	yaw_forearm	kurtosis_roll_forearm
## 1	28.4	-63.9	-153	
## 2	28.3	-63.9	-153	
## 3	28.3	-63.9	-152	
## 4	28.1	-63.9	-152	
## 5	28.0	-63.9	-152	
## 6	27.9	-63.9	-152	
##	kurtosis_picth_forearm	kurtosis_yaw_forearm	skewness_roll_forearm	
## 1				
## 2				
## 3				
## 4				
## 5				
## 6				
##	skewness_pitch_forearm	skewness_yaw_forearm	max_roll_forearm	
## 1			NA	
## 2			NA	
## 3			NA	
## 4			NA	
## 5			NA	
## 6			NA	
##	max_picth_forearm	max_yaw_forearm	min_roll_forearm	min_pitch_forearm
## 1	NA		NA	NA
## 2	NA		NA	NA
## 3	NA		NA	NA
## 4	NA		NA	NA
## 5	NA		NA	NA
## 6	NA		NA	NA
##	min_yaw_forearm	amplitude_roll_forearm	amplitude_pitch_forearm	
## 1		NA	NA	
## 2		NA	NA	

##	3		NA		NA
##	4		NA		NA
##	5		NA		NA
##	6		NA		NA
##		amplitude_yaw_forearm	total_accel_forearm	var_accel_forearm	
##	1		36		NA
##	2		36		NA
##	3		36		NA
##	4		36		NA
##	5		36		NA
##	6		36		NA
##		avg_roll_forearm	stddev_roll_forearm	var_roll_forearm	avg_pitch_forearm
##	1	NA	NA	NA	NA
##	2	NA	NA	NA	NA
##	3	NA	NA	NA	NA
##	4	NA	NA	NA	NA
##	5	NA	NA	NA	NA
##	6	NA	NA	NA	NA
##		stddev_pitch_forearm	var_pitch_forearm	avg_yaw_forearm	
##	1	NA	NA	NA	
##	2	NA	NA	NA	
##	3	NA	NA	NA	
##	4	NA	NA	NA	
##	5	NA	NA	NA	
##	6	NA	NA	NA	
##		stddev_yaw_forearm	var_yaw_forearm	gyros_forearm_x	gyros_forearm_y
##	1	NA	NA	0.03	0.00
##	2	NA	NA	0.02	0.00
##	3	NA	NA	0.03	-0.02
##	4	NA	NA	0.02	-0.02
##	5	NA	NA	0.02	0.00
##	6	NA	NA	0.02	-0.02
##		gyros_forearm_z	accel_forearm_x	accel_forearm_y	accel_forearm_z
##	1	-0.02	192	203	-215

```
## 2      -0.02      192      203      -216
## 3       0.00      196      204      -213
## 4       0.00      189      206      -214
## 5      -0.02      189      206      -214
## 6      -0.03      193      203      -215

##   magnet_forearm_x magnet_forearm_y magnet_forearm_z classe
## 1          -17          654          476      A
## 2          -18          661          473      A
## 3          -18          658          469      A
## 4          -16          658          469      A
## 5          -17          655          473      A
## 6           -9          660          478      A
```

Date Cleaning and Pre-Processing

There appear to be some columns that have a lot of null values, perhaps even the entire column is null. We remove such columns for both the test and train datasets.

```
library(dplyr)

## Warning: package 'dplyr' was built under R version 3.5.3
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:data.table':
##
##   between, first, last
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

dim(train)
## [1] 19622  160

not_any_na <- function(x) all(!is.na(x))
train<-train %>% select_if(not_any_na)
test<-test %>% select_if(not_any_na)
```

```
head(train)
```

```
##      X user_name raw_timestamp_part_1 raw_timestamp_part_2   cvtd_timestamp
## 1 1  carlitos      1323084231              788290 05/12/2011 11:23
## 2 2  carlitos      1323084231              808298 05/12/2011 11:23
## 3 3  carlitos      1323084231              820366 05/12/2011 11:23
## 4 4  carlitos      1323084232              120339 05/12/2011 11:23
## 5 5  carlitos      1323084232              196328 05/12/2011 11:23
## 6 6  carlitos      1323084232              304277 05/12/2011 11:23
```

```
##      new_window num_window roll_belt pitch_belt yaw_belt total_accel_belt
## 1           no          11      1.41      8.07     -94.4              3
## 2           no          11      1.41      8.07     -94.4              3
## 3           no          11      1.42      8.07     -94.4              3
## 4           no          12      1.48      8.05     -94.4              3
## 5           no          12      1.48      8.07     -94.4              3
## 6           no          12      1.45      8.06     -94.4              3
```

```
##      kurtosis_roll_belt kurtosis_pitch_belt kurtosis_yaw_belt
```

```
## 1
```

```
## 2
```

```
## 3
```

```
## 4
```

```
## 5
```

```
## 6
```

```
##      skewness_roll_belt skewness_roll_belt.1 skewness_yaw_belt max_yaw_belt
```

```
## 1
```

```
## 2
```

```
## 3
```

```
## 4
```

```
## 5
```

```
## 6
```

```
##      min_yaw_belt amplitude_yaw_belt gyros_belt_x gyros_belt_y gyros_belt_z
```

```
## 1              0.00              0.00          -0.02
```

```
## 2              0.02              0.00          -0.02
```

```
## 3              0.00              0.00          -0.02
```

```
## 4              0.02              0.00          -0.03
```

## 5			0.02	0.02	-0.02	
## 6			0.02	0.00	-0.02	
##	accel_belt_x	accel_belt_y	accel_belt_z	magnet_belt_x	magnet_belt_y	
## 1	-21	4	22	-3	599	
## 2	-22	4	22	-7	608	
## 3	-20	5	23	-2	600	
## 4	-22	3	21	-6	604	
## 5	-21	2	24	-6	600	
## 6	-21	4	21	0	603	
##	magnet_belt_z	roll_arm	pitch_arm	yaw_arm	total_accel_arm	gyros_arm_x
## 1	-313	-128	22.5	-161	34	0.00
## 2	-311	-128	22.5	-161	34	0.02
## 3	-305	-128	22.5	-161	34	0.02
## 4	-310	-128	22.1	-161	34	0.02
## 5	-302	-128	22.1	-161	34	0.00
## 6	-312	-128	22.0	-161	34	0.02
##	gyros_arm_y	gyros_arm_z	accel_arm_x	accel_arm_y	accel_arm_z	magnet_arm_x
## 1	0.00	-0.02	-288	109	-123	-368
## 2	-0.02	-0.02	-290	110	-125	-369
## 3	-0.02	-0.02	-289	110	-126	-368
## 4	-0.03	0.02	-289	111	-123	-372
## 5	-0.03	0.00	-289	111	-123	-374
## 6	-0.03	0.00	-289	111	-122	-369
##	magnet_arm_y	magnet_arm_z	kurtosis_roll_arm	kurtosis_pitch_arm		
## 1	337	516				
## 2	337	513				
## 3	344	513				
## 4	344	512				
## 5	337	506				
## 6	342	513				
##	kurtosis_yaw_arm	skewness_roll_arm	skewness_pitch_arm	skewness_yaw_arm		
## 1						
## 2						
## 3						

```

## 4
## 5
## 6
## roll_dumbbell pitch_dumbbell yaw_dumbbell kurtosis_roll_dumbbell
## 1      13.05217      -70.49400      -84.87394
## 2      13.13074      -70.63751      -84.71065
## 3      12.85075      -70.27812      -85.14078
## 4      13.43120      -70.39379      -84.87363
## 5      13.37872      -70.42856      -84.85306
## 6      13.38246      -70.81759      -84.46500
## kurtosis_pitch_dumbbell kurtosis_yaw_dumbbell skewness_roll_dumbbell
## 1
## 2
## 3
## 4
## 5
## 6
## skewness_pitch_dumbbell skewness_yaw_dumbbell max_yaw_dumbbell
## 1
## 2
## 3
## 4
## 5
## 6
## min_yaw_dumbbell amplitude_yaw_dumbbell total_accel_dumbbell
## 1                                     37
## 2                                     37
## 3                                     37
## 4                                     37
## 5                                     37
## 6                                     37
## gyros_dumbbell_x gyros_dumbbell_y gyros_dumbbell_z accel_dumbbell_x
## 1              0          -0.02          0.00          -234
## 2              0          -0.02          0.00          -233

```

## 3	0	-0.02	0.00	-232
## 4	0	-0.02	-0.02	-232
## 5	0	-0.02	0.00	-233
## 6	0	-0.02	0.00	-234
##	accel_dumbbell_y	accel_dumbbell_z	magnet_dumbbell_x	magnet_dumbbell_y
## 1	47	-271	-559	293
## 2	47	-269	-555	296
## 3	46	-270	-561	298
## 4	48	-269	-552	303
## 5	48	-270	-554	292
## 6	48	-269	-558	294
##	magnet_dumbbell_z	roll_forearm	pitch_forearm	yaw_forearm
## 1	-65	28.4	-63.9	-153
## 2	-64	28.3	-63.9	-153
## 3	-63	28.3	-63.9	-152
## 4	-60	28.1	-63.9	-152
## 5	-68	28.0	-63.9	-152
## 6	-66	27.9	-63.9	-152
##	kurtosis_roll_forearm	kurtosis_pitch_forearm	kurtosis_yaw_forearm	
## 1				
## 2				
## 3				
## 4				
## 5				
## 6				
##	skewness_roll_forearm	skewness_pitch_forearm	skewness_yaw_forearm	
## 1				
## 2				
## 3				
## 4				
## 5				
## 6				
##	max_yaw_forearm	min_yaw_forearm	amplitude_yaw_forearm	
## 1				

```
## 2
## 3
## 4
## 5
## 6
## total_accel_forearm gyros_forearm_x gyros_forearm_y gyros_forearm_z
## 1          36          0.03          0.00          -0.02
## 2          36          0.02          0.00          -0.02
## 3          36          0.03         -0.02          0.00
## 4          36          0.02         -0.02          0.00
## 5          36          0.02          0.00          -0.02
## 6          36          0.02         -0.02         -0.03
## accel_forearm_x accel_forearm_y accel_forearm_z magnet_forearm_x
## 1          192          203          -215          -17
## 2          192          203          -216          -18
## 3          196          204          -213          -18
## 4          189          206          -214          -16
## 5          189          206          -214          -17
## 6          193          203          -215          -9
## magnet_forearm_y magnet_forearm_z classe
## 1          654          476          A
## 2          661          473          A
## 3          658          469          A
## 4          658          469          A
## 5          655          473          A
## 6          660          478          A
```

We started out with 160 variables and now have 93. Additionally, the first few columns do not contain information regarding how well an activity is performed and mostly serve the role of primary keys for these data. For the purposes of the analysis, they can be ignored and, hence, removed from the dataset.

```
train<- train[,-c(1:7)]
test<- test[,-c(1:7)]
dim(train)
## [1] 19622    86
dim(test)
```

```
## [1] 20 53
```

We are not left with 86 variables in the train set and 52 variables in the test set. Since we need to have the same variables in both the training and testing steps, we remove the columns from the train set that are not part of the test set.

```
names<-names(test)
names<-append(names, "classe")
train<-train[,names(train) %in% names]
dim(train)
## [1] 19622 53
```

We now have comparable datasets and now we can move onto modeling stage. We note that the train data includes the variable “classe”, which we are attempting to predict.

Splitting Data into Train and Validate Sets

We split training data into the train and validation sets in the traditional 70/30 manner. The validation set will be used to estimate an out-of-sample error.

```
library(caret)
## Warning: package 'caret' was built under R version 3.5.3
## Loading required package: lattice
## Loading required package: ggplot2
inTrain = createDataPartition(train$classe, p = 0.70)[[1]]
training= train[ inTrain,]
validation = train[-inTrain,]
```

Modeling Using Classification Tree

We first fit a classification tree model to these data.

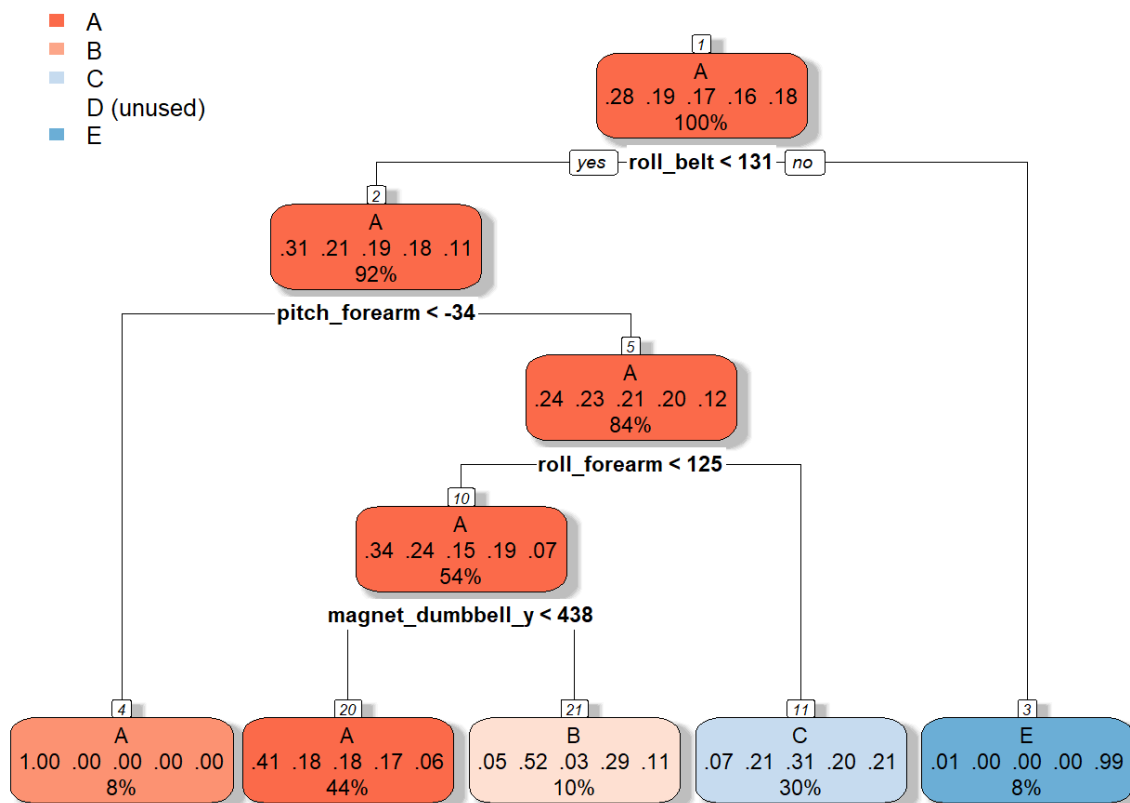
```
library(rpart)
## Warning: package 'rpart' was built under R version 3.5.3
library(rpart.plot)
## Warning: package 'rpart.plot' was built under R version 3.5.3
tree_model<- train(classe~., data=training, method="rpart")
tree_model$finalModel
## n= 13737
##
```



```
## node), split, n, loss, yval, (yprob)
##      * denotes terminal node
##
## 1) root 13737 9831 A (0.28 0.19 0.17 0.16 0.18)
##    2) roll_belt< 130.5 12599 8703 A (0.31 0.21 0.19 0.18 0.11)
##      4) pitch_forearm< -33.95 1083    3 A (1 0.0028 0 0 0) *
##      5) pitch_forearm>=-33.95 11516 8700 A (0.24 0.23 0.21 0.2 0.12)
##      10) roll_forearm< 124.5 7375 4836 A (0.34 0.24 0.15 0.19 0.07)
##        20) magnet_dumbbell_y< 437.5 6068 3590 A (0.41 0.18 0.18 0.17 0.06
## 2) *
##        21) magnet_dumbbell_y>=437.5 1307  622 B (0.047 0.52 0.031 0.29 0.
## 11) *
##      11) roll_forearm>=124.5 4141 2875 C (0.067 0.21 0.31 0.2 0.21) *
##    3) roll_belt>=130.5 1138    10 E (0.0088 0 0 0 0.99) *
```

Next, we visualize the tree model via a graphic.

```
rpart.plot(tree_model$finalModel, box.palette="RdBu", shadow.col="gray", nn=T
RUE)
```



Note that rpart automatically performs 10-fold cross-validation when growing a tree. The variables used in the final tree are magnet_dumbbell_y, pitch_forearm, roll_bel, roll_forearm with the relative error of 71 %.

We can estimate the out of sample error by applying the trained model on the validation set.

```
validate_tree <- predict(tree_model$finalModel, validation, type="class")
confusionMatrix(validation$classe, validate_tree)
```

Confusion Matrix and Statistics

##

Reference

Prediction A B C D E

A 1520 25 125 0 4

B 480 277 382 0 0

C 503 13 510 0 0

D 425 167 372 0 0

E 146 73 360 0 503

##

```
## Overall Statistics
##
##           Accuracy : 0.4775
##           95% CI : (0.4647, 0.4903)
##       No Information Rate : 0.5223
##       P-Value [Acc > NIR] : 1
##
##           Kappa : 0.3174
##
##  McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.4945  0.49910  0.29160      NA  0.99211
## Specificity      0.9452  0.83827  0.87524  0.8362  0.89234
## Pos Pred Value   0.9080  0.24320  0.49708      NA  0.46488
## Neg Pred Value   0.6310  0.94142  0.74501      NA  0.99917
## Prevalence       0.5223  0.09431  0.29720  0.0000  0.08615
## Detection Rate   0.2583  0.04707  0.08666  0.0000  0.08547
## Detection Prevalence 0.2845  0.19354  0.17434  0.1638  0.18386
## Balanced Accuracy 0.7198  0.66869  0.58342      NA  0.94222
```

The accuracy does not look too good but overall, the model is statistically significant, so we use it for the final prediction.

```
test_tree <- predict(tree_model$finalModel,test, type="class")
test_tree
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20
##  C  A  C  A  A  C  C  A  A  A  C  C  C  A  C  A  A  A  A  C
## Levels: A B C D E
```

Unfortunately, after submitting these results to the quiz engine, I did not pass the assignment.

Modeling with Random Forest

So, I try a random forest next because it uses an ensemble of trees so maybe it will work better.
First, I train the model

```
random_forest <- train(classe ~ ., data = train, method = "rf")
print(random_forest)

## Random Forest
##
## 19622 samples
##    52 predictor
##    5 classes: 'A', 'B', 'C', 'D', 'E'
##
## No pre-processing
## Resampling: Bootstrapped (25 reps)
## Summary of sample sizes: 19622, 19622, 19622, 19622, 19622, 19622, ...
## Resampling results across tuning parameters:
##
##    mtry  Accuracy   Kappa
##    2     0.9930247  0.9911747
##   27     0.9926283  0.9906736
##   52     0.9846718  0.9806069
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was mtry = 2.
```

Then, estimate an out-of-sample error on the validation set and display the confusion matrix.

```
validate <- predict(random_forest, validation)
# Show prediction result
confusionMatrix(validation$classe, validate)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction    A    B    C    D    E
##      A 1674     0     0     0     0
##      B     0 1139     0     0     0
##      C     0     0 1026     0     0
```

```
##           D      0      0      0  964      0
##           E      0      0      0      0 1082
##
## Overall Statistics
##
##           Accuracy : 1
##           95% CI : (0.9994, 1)
##           No Information Rate : 0.2845
##           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            0.2845   0.1935   0.1743   0.1638   0.1839
## Detection Rate        0.2845   0.1935   0.1743   0.1638   0.1839
## Detection Prevalence  0.2845   0.1935   0.1743   0.1638   0.1839
## Balanced Accuracy      1.0000   1.0000   1.0000   1.0000   1.0000
```

Finally, pass the quiz with the following predictions

```
predict(random_forest, test)
## [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
```

Conclusions

In this project we have examined exercise data and were able to characterize the quality of the exercise using only a few variables. We have leveraged a random forest model to achieve a high classification on both in-sample and out-of-sample data. We conclude that ensemble models rock

and that the government can recruit the future generation of secret service agents by hacking into the fitness watches of the daily users of such devices!

References

In preparing this report, I have consulted the following web-sites

<https://stackoverflow.com/questions/2643939/remove-columns-from-dataframe-where-all-values-are-na>

<https://stackoverflow.com/questions/33282174/r-caret-package-error-createdataparition-no-observation>

<https://blog.exploratory.io/visualizing-a-decision-tree-using-r-packages-in-exploratory-b26d4cb5e71f>

<https://www.gormananalysis.com/blog/decision-trees-in-r-using-rpart/>

<https://stackoverflow.com/questions/39620287/how-to-create-a-confusion-matrix-for-a-decision-tree-model>