

Assesment

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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://web.archive.org/web/20161224072740/http://groupware.les.inf.puc-rio.br/har> (see the section on the Weight Lifting Exercise Dataset).

```
library(caret)
library(ggplot2)
library(rpart)
library(randomForest)
library(dplyr)
```

Loading the data set and exploring it!

```
#along with importing data we are adding NA to blank cells of csv files
raw_train <- read.csv("data/pml-training.csv", na.strings=c("", "NA"))
raw_test <- read.csv("data/pml-testing.csv", na.strings=c("", "NA"))

head(raw_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_pichth_belt kurtosis_yaw_belt skewness_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>		
##	skewness_roll_belt.1	skewness_yaw_belt	max_roll_belt	max_pitch_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		
##	max_yaw_belt	min_roll_belt	min_pitch_belt	min_yaw_belt	amplitude_roll_belt	
## 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	
##	amplitude_pitch_belt	amplitude_yaw_belt	var_total_accel_belt	avg_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		
##	stddev_roll_belt	var_roll_belt	avg_pitch_belt	stddev_pitch_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		
##	var_pitch_belt	avg_yaw_belt	stddev_yaw_belt	var_yaw_belt	gyros_belt_x	
## 1	NA	NA	NA	NA	0.00	
## 2	NA	NA	NA	NA	0.02	
## 3	NA	NA	NA	NA	0.00	
## 4	NA	NA	NA	NA	0.02	
## 5	NA	NA	NA	NA	0.02	
## 6	NA	NA	NA	NA	0.02	
##	gyros_belt_y	gyros_belt_z	accel_belt_x	accel_belt_y	accel_belt_z	
## 1	0.00	-0.02	-21	4	22	
## 2	0.00	-0.02	-22	4	22	
## 3	0.00	-0.02	-20	5	23	
## 4	0.00	-0.03	-22	3	21	
## 5	0.02	-0.02	-21	2	24	
## 6	0.00	-0.02	-21	4	21	
##	magnet_belt_x	magnet_belt_y	magnet_belt_z	roll_arm	pitch_arm	yaw_arm
## 1	-3	599	-313	-128	22.5	-161
## 2	-7	608	-311	-128	22.5	-161
## 3	-2	600	-305	-128	22.5	-161
## 4	-6	604	-310	-128	22.1	-161
## 5	-6	600	-302	-128	22.1	-161
## 6	0	603	-312	-128	22.0	-161
##	total_accel_arm	var_accel_arm	avg_roll_arm	stddev_roll_arm	var_roll_arm	

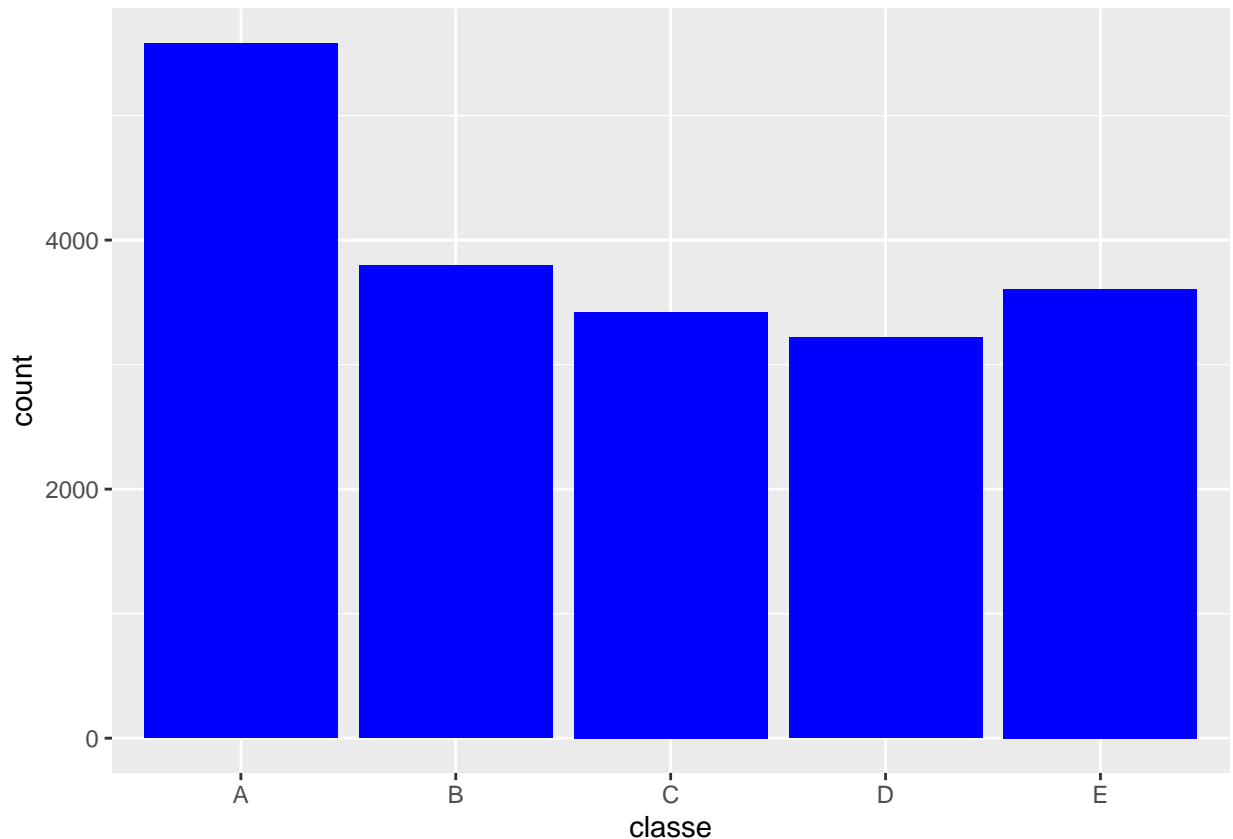
## 1	34	NA	NA	NA	NA	
## 2	34	NA	NA	NA	NA	
## 3	34	NA	NA	NA	NA	
## 4	34	NA	NA	NA	NA	
## 5	34	NA	NA	NA	NA	
## 6	34	NA	NA	NA	NA	
##	avg_pitch_arm	stddev_pitch_arm	var_pitch_arm	avg_yaw_arm	stddev_yaw_arm	
## 1	NA	NA	NA	NA	NA	
## 2	NA	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	NA	
## 6	NA	NA	NA	NA	NA	
##	var_yaw_arm	gyros_arm_x	gyros_arm_y	gyros_arm_z	accel_arm_x	accel_arm_y
## 1	NA	0.00	0.00	-0.02	-288	109
## 2	NA	0.02	-0.02	-0.02	-290	110
## 3	NA	0.02	-0.02	-0.02	-289	110
## 4	NA	0.02	-0.03	0.02	-289	111
## 5	NA	0.00	-0.03	0.00	-289	111
## 6	NA	0.02	-0.03	0.00	-289	111
##	accel_arm_z	magnet_arm_x	magnet_arm_y	magnet_arm_z	kurtosis_roll_arm	
## 1	-123	-368	337	516	<NA>	
## 2	-125	-369	337	513	<NA>	
## 3	-126	-368	344	513	<NA>	
## 4	-123	-372	344	512	<NA>	
## 5	-123	-374	337	506	<NA>	
## 6	-122	-369	342	513	<NA>	
##	kurtosis_pitch_arm	kurtosis_yaw_arm	skewness_roll_arm	skewness_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>	
##	skewness_yaw_arm	max_roll_arm	max_pitch_arm	max_yaw_arm	min_roll_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_pitch_arm	min_yaw_arm	amplitude_roll_arm	amplitude_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	
##	amplitude_yaw_arm	roll_dumbbell	pitch_dumbbell	yaw_dumbbell		
## 1	NA	13.05217	-70.49400	-84.87394		
## 2	NA	13.13074	-70.63751	-84.71065		
## 3	NA	12.85075	-70.27812	-85.14078		
## 4	NA	13.43120	-70.39379	-84.87363		
## 5	NA	13.37872	-70.42856	-84.85306		

## 6	NA	13.38246	-70.81759	-84.46500
##	kurtosis_roll_dumbbell	kurtosis_picth_dumbbell	kurtosis_yaw_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>
##	skewness_roll_dumbbell	skewness_pitch_dumbbell	skewness_yaw_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>
##	max_roll_dumbbell	max_picth_dumbbell	max_yaw_dumbbell	min_roll_dumbbell
## 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
##	min_pitch_dumbbell	min_yaw_dumbbell	amplitude_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	
##	amplitude_pitch_dumbbell	amplitude_yaw_dumbbell	total_accel_dumbbell	
## 1	NA	<NA>	37	
## 2	NA	<NA>	37	
## 3	NA	<NA>	37	
## 4	NA	<NA>	37	
## 5	NA	<NA>	37	
## 6	NA	<NA>	37	
##	var_accel_dumbbell	avg_roll_dumbbell	stddev_roll_dumbbell	var_roll_dumbbell
## 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_dumbbell	stddev_pitch_dumbbell	var_pitch_dumbbell	avg_yaw_dumbbell
## 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_dumbbell	var_yaw_dumbbell	gyros_dumbbell_x	gyros_dumbbell_y
## 1	NA	NA	0	-0.02
## 2	NA	NA	0	-0.02
## 3	NA	NA	0	-0.02

## 4	NA	NA	0	-0.02
## 5	NA	NA	0	-0.02
## 6	NA	NA	0	-0.02
##	gyros_dumbbell_z	accel_dumbbell_x	accel_dumbbell_y	accel_dumbbell_z
## 1	0.00	-234	47	-271
## 2	0.00	-233	47	-269
## 3	0.00	-232	46	-270
## 4	-0.02	-232	48	-269
## 5	0.00	-233	48	-270
## 6	0.00	-234	48	-269
##	magnet_dumbbell_x	magnet_dumbbell_y	magnet_dumbbell_z	roll_forearm
## 1	-559	293	-65	28.4
## 2	-555	296	-64	28.3
## 3	-561	298	-63	28.3
## 4	-552	303	-60	28.1
## 5	-554	292	-68	28.0
## 6	-558	294	-66	27.9
##	pitch_forearm	yaw_forearm	kurtosis_roll_forearm	kurtosis_pitch_forearm
## 1	-63.9	-153	<NA>	<NA>
## 2	-63.9	-153	<NA>	<NA>
## 3	-63.9	-152	<NA>	<NA>
## 4	-63.9	-152	<NA>	<NA>
## 5	-63.9	-152	<NA>	<NA>
## 6	-63.9	-152	<NA>	<NA>
##	kurtosis_yaw_forearm	skewness_roll_forearm	skewness_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>	
##	skewness_yaw_forearm	max_roll_forearm	max_pitch_forearm	max_yaw_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>
##	min_roll_forearm	min_pitch_forearm	min_yaw_forearm	amplitude_roll_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
##	amplitude_pitch_forearm	amplitude_yaw_forearm	total_accel_forearm	
## 1	NA	<NA>	36	
## 2	NA	<NA>	36	
## 3	NA	<NA>	36	
## 4	NA	<NA>	36	
## 5	NA	<NA>	36	
## 6	NA	<NA>	36	
##	var_accel_forearm	avg_roll_forearm	stddev_roll_forearm	var_roll_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
##	avg_pitch_forearm	stddev_pitch_forearm	var_pitch_forearm	avg_yaw_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_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

```
raw_train %>% ggplot(aes(classe)) + geom_bar(stat = "count", fill = "blue")
```



As there are null values and many columns in the columns we need to do some data wrangling!

```
# removing columns with NA greater than 50%
training <- raw_train[, which(colMeans(!is.na(raw_train)) > 0.5)]
testing <- raw_test[,which(colMeans(!is.na(raw_train))>0.5)]

#First seven columns of the dataset are not so useful we can remove them
sam <- training[,-c(1:7)]
testing <- testing[,-c(1:7)]

#creating data portion of training sample for modeling
intrain <- createDataPartition(y= sam$classe, p=0.7, list = FALSE)

sam_train <- sam[intrain,]
sam_test <- sam[-intrain,]
```

#Decision Tree

```
model_dt <- rpart(classe ~ ., data = sam_train, method = "class")
pred_dt <- predict(model_dt,sam_test, type = "class")

confusionMatrix(table(pred_dt,sam_test$class))
```

```
## Confusion Matrix and Statistics
##
```

```
##
## pred_dt      A      B      C      D      E
##      A 1515  250   18  115   44
##      B   37  610   52   23   76
##      C   39  119  822  137  118
##      D   55   70   55  609   49
##      E   28   90   79   80  795
##
## Overall Statistics
##
##              Accuracy : 0.7393
##              95% CI : (0.7279, 0.7505)
##      No Information Rate : 0.2845
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.6686
##
## McNemar's Test P-Value : < 2.2e-16
##
## Statistics by Class:
##
##              Class: A Class: B Class: C Class: D Class: E
## Sensitivity          0.9050   0.5356   0.8012   0.6317   0.7348
## Specificity          0.8986   0.9604   0.9150   0.9535   0.9423
## Pos Pred Value       0.7801   0.7644   0.6656   0.7267   0.7416
## Neg Pred Value       0.9597   0.8960   0.9561   0.9297   0.9404
## Prevalence           0.2845   0.1935   0.1743   0.1638   0.1839
## Detection Rate       0.2574   0.1037   0.1397   0.1035   0.1351
## Detection Prevalence 0.3300   0.1356   0.2099   0.1424   0.1822
## Balanced Accuracy    0.9018   0.7480   0.8581   0.7926   0.8385
```

#Random Forest

```
library(parallel)
library(doParallel)
```

```
## Loading required package: foreach
```

```
## Loading required package: iterators
```

```
cluster <- makeCluster(detectCores() - 1) # convention to leave 1 core for OS
registerDoParallel(cluster)

fitControl <- trainControl(method = "cv",
                           number = 5,
                           allowParallel = TRUE)

model_rf <- train(classe ~ ., method = "rf", data = sam_train, trControl = fitControl )

stopCluster(cluster)
registerDoSEQ()
```



```
pred_rf <- predict(model_rf,sam_test)

confusionMatrix(table(pred_rf,sam_test$class))
```

```
## Confusion Matrix and Statistics
##
##
## pred_rf      A      B      C      D      E
##      A 1672     11      0      0      0
##      B      0 1124      4      1      0
##      C      2      4 1018     10      1
##      D      0      0      4  952      5
##      E      0      0      0      1 1076
##
## Overall Statistics
##
##              Accuracy : 0.9927
##              95% CI : (0.9902, 0.9947)
##      No Information Rate : 0.2845
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.9908
##
##      McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##              Class: A Class: B Class: C Class: D Class: E
## Sensitivity          0.9988   0.9868   0.9922   0.9876   0.9945
## Specificity          0.9974   0.9989   0.9965   0.9982   0.9998
## Pos Pred Value       0.9935   0.9956   0.9836   0.9906   0.9991
## Neg Pred Value       0.9995   0.9968   0.9984   0.9976   0.9988
## Prevalence           0.2845   0.1935   0.1743   0.1638   0.1839
## Detection Rate       0.2841   0.1910   0.1730   0.1618   0.1828
## Detection Prevalence 0.2860   0.1918   0.1759   0.1633   0.1830
## Balanced Accuracy    0.9981   0.9929   0.9944   0.9929   0.9971
```

Final prediction using testing data set on random forest model as we are getting 99.25 accuracy

```
# This will give us the prediction of the 20 variables in testing dataset
list(predict(model_rf, testing))
```

```
## [[1]]
##      [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
```