

Machine learning project"

Introduction

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, 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

Downloading data for hte project

```
download.file("http://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv",
              "train_set.csv",mode="wb")
download.file("http://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv",
              "test_set.csv", mode="wb")

train_set <- read.csv("train_set.csv",na.strings=c("NA","#DIV/0!", ""), row.names = 1)
test_set <- read.csv("test_set.csv",na.strings=c("NA","#DIV/0!", ""), row.names = 1)

dim(train_set)
```

```
## [1] 19622 159
```

```
dim(test_set)
```

```
## [1] 20 159
```

```
table(train_set$classe)
```

```
##
##      A      B      C      D      E
## 5580 3797 3422 3216 3607
```

Data Processing

First round of data processing will include cleaning data.

```
##check for NAs
nasPerColumn <- apply(train_set, 2, function(x) length(which(is.na(x))))
nasPerColumn
```

```
##           user_name  raw_timestamp_part_1  raw_timestamp_part_2
##                0                0                0
##      cvtd_timestamp      new_window      num_window
##                0                0                0
```

##	roll_belt	pitch_belt	yaw_belt
##	0	0	0
##	total_accel_belt	kurtosis_roll_belt	kurtosis_picth_belt
##	0	19226	19248
##	kurtosis_yaw_belt	skewness_roll_belt	skewness_roll_belt.1
##	19622	19225	19248
##	skewness_yaw_belt	max_roll_belt	max_picth_belt
##	19622	19216	19216
##	max_yaw_belt	min_roll_belt	min_pitch_belt
##	19226	19216	19216
##	min_yaw_belt	amplitude_roll_belt	amplitude_pitch_belt
##	19226	19216	19216
##	amplitude_yaw_belt	var_total_accel_belt	avg_roll_belt
##	19226	19216	19216
##	stddev_roll_belt	var_roll_belt	avg_pitch_belt
##	19216	19216	19216
##	stddev_pitch_belt	var_pitch_belt	avg_yaw_belt
##	19216	19216	19216
##	stddev_yaw_belt	var_yaw_belt	gyros_belt_x
##	19216	19216	0
##	gyros_belt_y	gyros_belt_z	accel_belt_x
##	0	0	0
##	accel_belt_y	accel_belt_z	magnet_belt_x
##	0	0	0
##	magnet_belt_y	magnet_belt_z	roll_arm
##	0	0	0
##	pitch_arm	yaw_arm	total_accel_arm
##	0	0	0
##	var_accel_arm	avg_roll_arm	stddev_roll_arm
##	19216	19216	19216
##	var_roll_arm	avg_pitch_arm	stddev_pitch_arm
##	19216	19216	19216
##	var_pitch_arm	avg_yaw_arm	stddev_yaw_arm
##	19216	19216	19216
##	var_yaw_arm	gyros_arm_x	gyros_arm_y
##	19216	0	0
##	gyros_arm_z	accel_arm_x	accel_arm_y
##	0	0	0
##	accel_arm_z	magnet_arm_x	magnet_arm_y
##	0	0	0
##	magnet_arm_z	kurtosis_roll_arm	kurtosis_picth_arm
##	0	19294	19296
##	kurtosis_yaw_arm	skewness_roll_arm	skewness_pitch_arm
##	19227	19293	19296
##	skewness_yaw_arm	max_roll_arm	max_picth_arm
##	19227	19216	19216
##	max_yaw_arm	min_roll_arm	min_pitch_arm
##	19216	19216	19216
##	min_yaw_arm	amplitude_roll_arm	amplitude_pitch_arm
##	19216	19216	19216
##	amplitude_yaw_arm	roll_dumbbell	pitch_dumbbell
##	19216	0	0
##	yaw_dumbbell	kurtosis_roll_dumbbell	kurtosis_picth_dumbbell
##	0	19221	19218

```

##      kurtosis_yaw_dumbbell      skewness_roll_dumbbell      skewness_pitch_dumbbell
##      19622                      19220                      19217
##      skewness_yaw_dumbbell      max_roll_dumbbell      max_pitch_dumbbell
##      19622                      19216                      19216
##      max_yaw_dumbbell      min_roll_dumbbell      min_pitch_dumbbell
##      19221                      19216                      19216
##      min_yaw_dumbbell      amplitude_roll_dumbbell      amplitude_pitch_dumbbell
##      19221                      19216                      19216
##      amplitude_yaw_dumbbell      total_accel_dumbbell      var_accel_dumbbell
##      19221                      0                      19216
##      avg_roll_dumbbell      stddev_roll_dumbbell      var_roll_dumbbell
##      19216                      19216                      19216
##      avg_pitch_dumbbell      stddev_pitch_dumbbell      var_pitch_dumbbell
##      19216                      19216                      19216
##      avg_yaw_dumbbell      stddev_yaw_dumbbell      var_yaw_dumbbell
##      19216                      19216                      19216
##      gyros_dumbbell_x      gyros_dumbbell_y      gyros_dumbbell_z
##      0                      0                      0
##      accel_dumbbell_x      accel_dumbbell_y      accel_dumbbell_z
##      0                      0                      0
##      magnet_dumbbell_x      magnet_dumbbell_y      magnet_dumbbell_z
##      0                      0                      0
##      roll_forearm      pitch_forearm      yaw_forearm
##      0                      0                      0
##      kurtosis_roll_forearm      kurtosis_pitch_forearm      kurtosis_yaw_forearm
##      19300                      19301                      19622
##      skewness_roll_forearm      skewness_pitch_forearm      skewness_yaw_forearm
##      19299                      19301                      19622
##      max_roll_forearm      max_pitch_forearm      max_yaw_forearm
##      19216                      19216                      19300
##      min_roll_forearm      min_pitch_forearm      min_yaw_forearm
##      19216                      19216                      19300
##      amplitude_roll_forearm      amplitude_pitch_forearm      amplitude_yaw_forearm
##      19216                      19216                      19300
##      total_accel_forearm      var_accel_forearm      avg_roll_forearm
##      0                      19216                      19216
##      stddev_roll_forearm      var_roll_forearm      avg_pitch_forearm
##      19216                      19216                      19216
##      stddev_pitch_forearm      var_pitch_forearm      avg_yaw_forearm
##      19216                      19216                      19216
##      stddev_yaw_forearm      var_yaw_forearm      gyros_forearm_x
##      19216                      19216                      0
##      gyros_forearm_y      gyros_forearm_z      accel_forearm_x
##      0                      0                      0
##      accel_forearm_y      accel_forearm_z      magnet_forearm_x
##      0                      0                      0
##      magnet_forearm_y      magnet_forearm_z      classe
##      0                      0                      0

```

```

##remove columns that contain mostly NA values (90%)
train_set <- train_set[,which(nasPerColumn < nrow(train_set)*0.9)]

library(caret)

```

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

##remove near zero values
nearZeroColumns <- nearZeroVar(train_set, saveMetrics = TRUE)
train_set <- train_set[, nearZeroColumns$nzv==FALSE]

##subset the taining set to only predictor columns
train_set <-train_set[, -c(1:7)]
test_set <-test_set[, -c(1:7)]

##Classe as factor
train_set$classe <- factor(train_set$classe)
```

Now let's do preprocessing for the model training

```
library(caret)
set.seed(23232)
inTrain <- createDataPartition(train_set$classe, p=0.75, list = FALSE)
train <- train_set[inTrain,]
test <- train_set[-inTrain,]
```

Model Training and validation

First, we will build descision tree model.

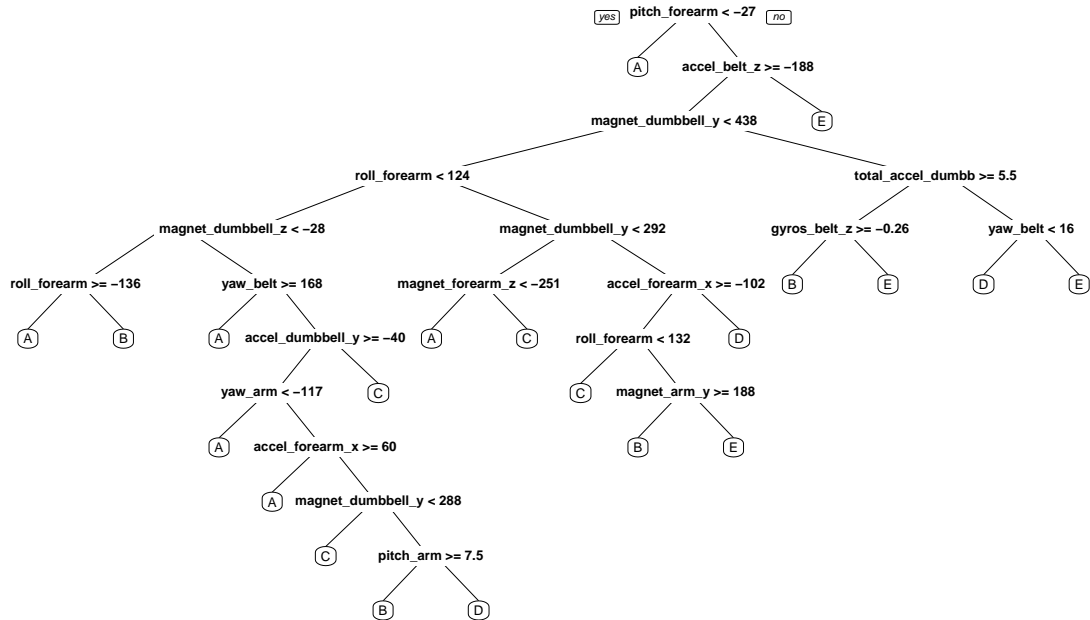
```
library(randomForest)

## randomForest 4.6-10
## Type rfNews() to see new features/changes/bug fixes.

library(rpart)
library(rpart.plot)

firstmodel <- rpart(classe ~ ., data=train, method="class")
prediction <- predict(firstmodel, test, type = "class")
rpart.plot(firstmodel, main="Classification Tree")
```

Classification Tree



```
##review confusin matrix
confusionMatrix(prediction, test$classe)
```

Confusion Matrix and Statistics

```
##
##           Reference
## Prediction   A    B    C    D    E
##           A 1256  179   36   62   64
##           B   37  423   66   88  133
##           C   38  211  732  209  219
##           D   49  104   14  407   61
##           E   15   32    7   38  424
```

Overall Statistics

```
##
##           Accuracy : 0.6611
##           95% CI : (0.6477, 0.6743)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
```

```
##
##           Kappa : 0.5695
##           McNemar's Test P-Value : < 2.2e-16
```

```
##
## Statistics by Class:
```

```
##
##           Class: A Class: B Class: C Class: D Class: E
```

## Sensitivity	0.9004	0.44573	0.8561	0.50622	0.47059
## Specificity	0.9028	0.91808	0.8328	0.94439	0.97702
## Pos Pred Value	0.7865	0.56627	0.5195	0.64094	0.82171
## Neg Pred Value	0.9580	0.87347	0.9648	0.90700	0.89129
## Prevalence	0.2845	0.19352	0.1743	0.16395	0.18373
## Detection Rate	0.2561	0.08626	0.1493	0.08299	0.08646
## Detection Prevalence	0.3257	0.15232	0.2873	0.12949	0.10522
## Balanced Accuracy	0.9016	0.68191	0.8445	0.72530	0.72380

Now we will build random forest model.

```
scondmodel <- randomForest(classe ~ ., data=train, method="class")
prediction <- predict(scondmodel, test, type = "class")

##review confusin matrix
confusionMatrix(prediction, test$classe)
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    A    B    C    D    E
##           A 1392    7    0    0    0
##           B    2   939    5    0    0
##           C    0    3   849   18    0
##           D    0    0    1   786    3
##           E    1    0    0    0   898
##
## Overall Statistics
##
##           Accuracy : 0.9918
##           95% CI : (0.9889, 0.9942)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9897
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9978   0.9895   0.9930   0.9776   0.9967
## Specificity      0.9980   0.9982   0.9948   0.9990   0.9998
## Pos Pred Value   0.9950   0.9926   0.9759   0.9949   0.9989
## Neg Pred Value   0.9991   0.9975   0.9985   0.9956   0.9993
## Prevalence       0.2845   0.1935   0.1743   0.1639   0.1837
## Detection Rate   0.2838   0.1915   0.1731   0.1603   0.1831
## Detection Prevalence 0.2853   0.1929   0.1774   0.1611   0.1833
## Balanced Accuracy 0.9979   0.9938   0.9939   0.9883   0.9982
```

We see that random forest model accuracy is better than the decision tree model accuracy, so we should be using that model to predict results for the test_set file.

Test Set predictions

```
test_predictions <- predict(scondmodel, test_set, type="class")
```

```
##the list of answers
```

```
test_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
```

Creating submission files.

```
# 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(test_predictions)
```