

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).

What you should submit

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. You may use any of the other variables to predict with. You should create a report describing how you built your model, how you used cross validation, what you think the expected out of sample error is, and why you made the choices you did. You will also use your prediction model to predict 20 different test cases.

Data Preperation

Note: Please see appendix for data exploration

```
#Set Seed
#set.seed(12345)
#Load packages
library(caret)
```

```
## Warning: package 'caret' was built under R version 3.2.5
```

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

```
## Warning: package 'lattice' was built under R version 3.2.3
```

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

```
## Warning: package 'ggplot2' was built under R version 3.2.3
```

```
library(randomForest)
```

```
## Warning: package 'randomForest' was built under R version 3.2.5
```

```
## randomForest 4.6-12
```

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

```
##
```

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

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

```
#load data
TrainingData <- read.csv('pml-training.csv', stringsAsFactors = F, na.strings = "")
TestData <- read.csv('pml-testing.csv', stringsAsFactors = F, na.strings = "")

#clean data

#Remove first seven columns of data as they are not predictors of #form, these fields include record ID
TrainingData <- TrainingData[, -(1:7)]
TestData <- TestData[, -(1:7)]

#set classe as factor data type
TrainingData$classe <- as.factor(TrainingData$classe)

#remove fields with near zero variance as they will not be good predictors
nzv <- nearZeroVar(TrainingData)
TrainingData <- TrainingData[-nzv]
nzv <- nearZeroVar(TestData)
TestData <- TestData[-nzv]

#remove fields with missing data
TrainingData <- TrainingData[colSums(is.na(TrainingData)) == 0]
TestData <- TestData[colSums(is.na(TestData)) == 0]

#Split training data into training and validation data subsets for cross validation
trainIndex <- createDataPartition(TrainingData$classe, list = F, p = 0.7)
TrainSubset <- TrainingData[trainIndex, ]
ValidationSubset <- TrainingData[- trainIndex, ]
```

Modelselection - training data

Begin with random forest model with the training data

```
# Train random forest model
RFModel1 <- randomForest(classe ~ ., data = TrainSubset, mtry = 25, ntree = 1000, proximity = F)
print(RFModel1)
```

```
##
## Call:
## randomForest(formula = classe ~ ., data = TrainSubset, mtry = 25,          ntree = 1000, proximity = F)
##           Type of random forest: classification
##           Number of trees: 1000
## No. of variables tried at each split: 25
##
##           OOB estimate of  error rate: 0.75%
## Confusion matrix:
##      A    B    C    D    E class.error
## A 3901     3     0     0     2 0.001280082
## B   18 2630    10     0     0 0.010534236
## C     0   17 2371     7     1 0.010434057
## D     0    1  31 2217     3 0.015541741
```

```
## E    0    0    3    7 2515 0.003960396
```

confusion matrix

Use validation data set to estimate out of sample error

```
RFPrediction1 <- predict(RFModel1, ValidationSubset)
ConMat <- confusionMatrix(ValidationSubset$classe, RFPrediction1)
print(ConMat)
```

Confusion Matrix and Statistics

```
##
```

```
##           Reference
```

```
## Prediction    A    B    C    D    E
##           A 1673    1    0    0    0
##           B    6 1130    3    0    0
##           C    0    4 1018    4    0
##           D    0    0   12  952    0
##           E    0    1    4    2 1075
```

```
##
```

Overall Statistics

```
##
```

```
##           Accuracy : 0.9937
##           95% CI : (0.9913, 0.9956)
##    No Information Rate : 0.2853
##    P-Value [Acc > NIR] : < 2.2e-16
```

```
##
```

```
##           Kappa : 0.992
```

```
## Mcnemar's Test P-Value : NA
```

```
##
```

Statistics by Class:

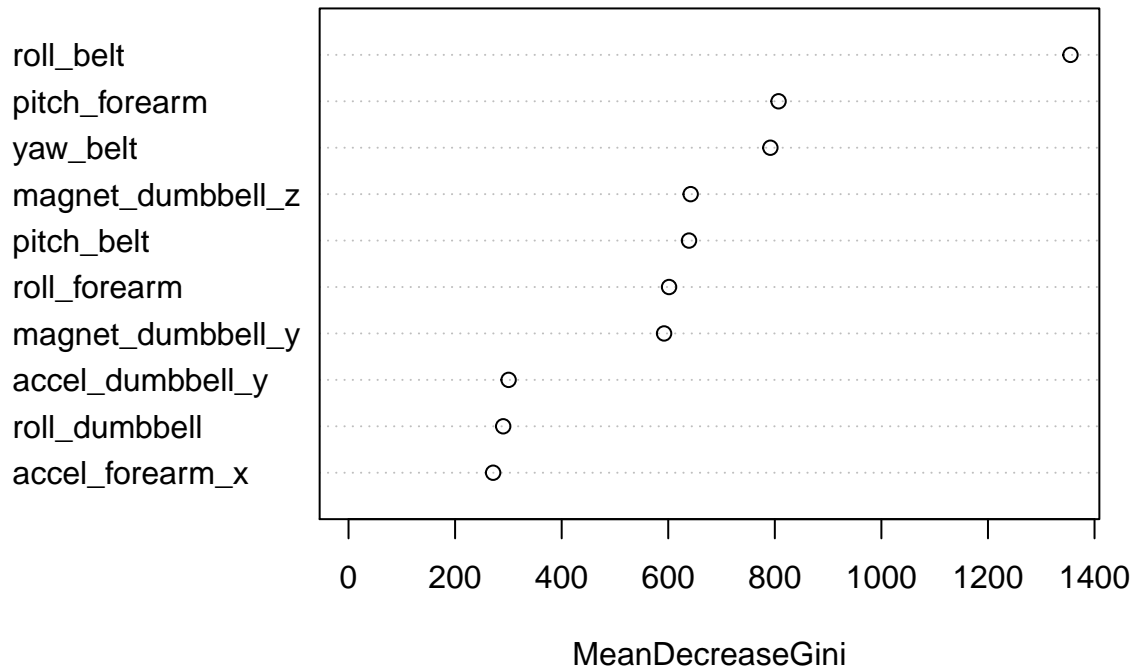
```
##
```

```
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9964  0.9947  0.9817  0.9937  1.0000
## Specificity      0.9998  0.9981  0.9983  0.9976  0.9985
## Pos Pred Value   0.9994  0.9921  0.9922  0.9876  0.9935
## Neg Pred Value   0.9986  0.9987  0.9961  0.9988  1.0000
## Prevalence       0.2853  0.1930  0.1762  0.1628  0.1827
## Detection Rate   0.2843  0.1920  0.1730  0.1618  0.1827
## Detection Prevalence 0.2845  0.1935  0.1743  0.1638  0.1839
## Balanced Accuracy 0.9981  0.9964  0.9900  0.9957  0.9993
```

Plot variables by importance

```
# Top Ten most important predictors
varImpPlot(RFModel1, n.var = 10, main = "Top 10 Predictors")
```

Top 10 Predictors



Appendix

Data Exploration

```
#Review file variables
str(TrainingData)
```

```
## 'data.frame': 19622 obs. of 53 variables:
## $ roll_belt : num 1.41 1.41 1.42 1.48 1.48 1.45 1.42 1.42 1.43 1.45 ...
## $ pitch_belt : num 8.07 8.07 8.07 8.05 8.07 8.06 8.09 8.13 8.16 8.17 ...
## $ yaw_belt : num -94.4 -94.4 -94.4 -94.4 -94.4 -94.4 -94.4 -94.4 -94.4 -94.4 ...
## $ total_accel_belt : int 3 3 3 3 3 3 3 3 3 3 ...
## $ gyros_belt_x : num 0 0.02 0 0.02 0.02 0.02 0.02 0.02 0.02 0.03 ...
## $ gyros_belt_y : num 0 0 0 0 0.02 0 0 0 0 0 ...
## $ gyros_belt_z : num -0.02 -0.02 -0.02 -0.03 -0.02 -0.02 -0.02 -0.02 -0.02 -0.02 0 ...
## $ accel_belt_x : int -21 -22 -20 -22 -21 -21 -22 -22 -20 -21 ...
## $ accel_belt_y : int 4 4 5 3 2 4 3 4 2 4 ...
## $ accel_belt_z : int 22 22 23 21 24 21 21 21 24 22 ...
## $ magnet_belt_x : int -3 -7 -2 -6 -6 0 -4 -2 1 -3 ...
## $ magnet_belt_y : int 599 608 600 604 600 603 599 603 602 609 ...
## $ magnet_belt_z : int -313 -311 -305 -310 -302 -312 -311 -313 -312 -308 ...
## $ roll_arm : num -128 -128 -128 -128 -128 -128 -128 -128 -128 -128 ...
## $ pitch_arm : num 22.5 22.5 22.5 22.1 22.1 22 21.9 21.8 21.7 21.6 ...
## $ yaw_arm : num -161 -161 -161 -161 -161 -161 -161 -161 -161 -161 ...
## $ total_accel_arm : int 34 34 34 34 34 34 34 34 34 34 ...
## $ gyros_arm_x : num 0 0.02 0.02 0.02 0 0.02 0 0.02 0.02 0.02 ...
```

```
## $ gyros_arm_y      : num  0 -0.02 -0.02 -0.03 -0.03 -0.03 -0.03 -0.02 -0.03 -0.03 ...
## $ gyros_arm_z      : num -0.02 -0.02 -0.02  0.02  0  0  0  0 -0.02 -0.02 ...
## $ accel_arm_x      : int  -288 -290 -289 -289 -289 -289 -289 -289 -288 -288 ...
## $ accel_arm_y      : int  109 110 110 111 111 111 111 111 109 110 ...
## $ accel_arm_z      : int  -123 -125 -126 -123 -123 -122 -125 -124 -122 -124 ...
## $ magnet_arm_x     : int  -368 -369 -368 -372 -374 -369 -373 -372 -369 -376 ...
## $ magnet_arm_y     : int   337 337 344 344 337 342 336 338 341 334 ...
## $ magnet_arm_z     : int   516 513 513 512 506 513 509 510 518 516 ...
## $ roll_dumbbell    : num  13.1 13.1 12.9 13.4 13.4 ...
## $ pitch_dumbbell   : num  -70.5 -70.6 -70.3 -70.4 -70.4 ...
## $ yaw_dumbbell     : num  -84.9 -84.7 -85.1 -84.9 -84.9 ...
## $ total_accel_dumbbell: int  37 37 37 37 37 37 37 37 37 37 ...
## $ gyros_dumbbell_x  : num   0  0  0  0  0  0  0  0  0  0 ...
## $ gyros_dumbbell_y  : num  -0.02 -0.02 -0.02 -0.02 -0.02 -0.02 -0.02 -0.02 -0.02 -0.02 ...
## $ gyros_dumbbell_z  : num   0  0  0 -0.02  0  0  0  0  0  0 ...
## $ accel_dumbbell_x  : int  -234 -233 -232 -232 -233 -234 -232 -234 -232 -235 ...
## $ accel_dumbbell_y  : int   47 47 46 48 48 48 47 46 47 48 ...
## $ accel_dumbbell_z  : int  -271 -269 -270 -269 -270 -269 -270 -272 -269 -270 ...
## $ magnet_dumbbell_x : int  -559 -555 -561 -552 -554 -558 -551 -555 -549 -558 ...
## $ magnet_dumbbell_y : int   293 296 298 303 292 294 295 300 292 291 ...
## $ magnet_dumbbell_z : num  -65 -64 -63 -60 -68 -66 -70 -74 -65 -69 ...
## $ roll_forearm     : num  28.4 28.3 28.3 28.1 28 27.9 27.9 27.8 27.7 27.7 ...
## $ pitch_forearm    : num  -63.9 -63.9 -63.9 -63.9 -63.9 -63.9 -63.9 -63.8 -63.8 -63.8 ...
## $ yaw_forearm      : num  -153 -153 -152 -152 -152 -152 -152 -152 -152 -152 ...
## $ total_accel_forearm: int  36 36 36 36 36 36 36 36 36 36 ...
## $ gyros_forearm_x   : num   0.03 0.02 0.03 0.02 0.02 0.02 0.02 0.02 0.03 0.02 ...
## $ gyros_forearm_y   : num   0  0 -0.02 -0.02  0 -0.02  0 -0.02  0  0 ...
## $ gyros_forearm_z   : num  -0.02 -0.02  0  0 -0.02 -0.03 -0.02  0 -0.02 -0.02 ...
## $ accel_forearm_x   : int  192 192 196 189 189 193 195 193 193 190 ...
## $ accel_forearm_y   : int  203 203 204 206 206 203 205 205 204 205 ...
## $ accel_forearm_z   : int  -215 -216 -213 -214 -214 -215 -215 -213 -214 -215 ...
## $ magnet_forearm_x  : int  -17 -18 -18 -16 -17 -9 -18 -9 -16 -22 ...
## $ magnet_forearm_y  : num  654 661 658 658 655 660 659 660 653 656 ...
## $ magnet_forearm_z  : num  476 473 469 469 473 478 470 474 476 473 ...
## $ classe            : Factor w/ 5 levels "A","B","C","D",...: 1 1 1 1 1 1 1 1 1 1 ...
```

```
summary(TrainingData)
```

```
##      roll_belt      pitch_belt      yaw_belt      total_accel_belt
## Min.      :-28.90    Min.      :-55.8000    Min.      :-180.00    Min.      : 0.00
## 1st Qu.:   1.10     1st Qu.:   1.7600     1st Qu.:  -88.30     1st Qu.:   3.00
## Median :  113.00     Median :   5.2800     Median :  -13.00     Median : 17.00
## Mean      :  64.41     Mean      :  0.3053     Mean      : -11.21     Mean      :11.31
## 3rd Qu.:  123.00     3rd Qu.:  14.9000     3rd Qu.:   12.90     3rd Qu.: 18.00
## Max.      : 162.00     Max.      : 60.3000     Max.      : 179.00     Max.      :29.00
##      gyros_belt_x      gyros_belt_y      gyros_belt_z
## Min.      :-1.040000    Min.      :-0.64000    Min.      :-1.4600
## 1st Qu.: -0.030000     1st Qu.:  0.00000     1st Qu.: -0.2000
## Median :  0.030000     Median :  0.02000     Median : -0.1000
## Mean      : -0.005592    Mean      :  0.03959     Mean      : -0.1305
## 3rd Qu.:  0.110000     3rd Qu.:  0.11000     3rd Qu.: -0.0200
## Max.      :  2.220000     Max.      :  0.64000     Max.      :  1.6200
##      accel_belt_x      accel_belt_y      accel_belt_z      magnet_belt_x
## Min.      :-120.000    Min.      :-69.00     Min.      :-275.00    Min.      :-52.0
```

```

## 1st Qu.: -21.000 1st Qu.: 3.00 1st Qu.: -162.00 1st Qu.: 9.0
## Median : -15.000 Median : 35.00 Median : -152.00 Median : 35.0
## Mean : -5.595 Mean : 30.15 Mean : -72.59 Mean : 55.6
## 3rd Qu.: -5.000 3rd Qu.: 61.00 3rd Qu.: 27.00 3rd Qu.: 59.0
## Max. : 85.000 Max. : 164.00 Max. : 105.00 Max. : 485.0
## magnet_belt_y magnet_belt_z roll_arm pitch_arm
## Min. :354.0 Min. : -623.0 Min. : -180.00 Min. : -88.800
## 1st Qu.:581.0 1st Qu.: -375.0 1st Qu.: -31.77 1st Qu.: -25.900
## Median :601.0 Median : -320.0 Median : 0.00 Median : 0.000
## Mean :593.7 Mean : -345.5 Mean : 17.83 Mean : -4.612
## 3rd Qu.:610.0 3rd Qu.: -306.0 3rd Qu.: 77.30 3rd Qu.: 11.200
## Max. :673.0 Max. : 293.0 Max. : 180.00 Max. : 88.500
## yaw_arm total_accel_arm gyros_arm_x gyros_arm_y
## Min. : -180.0000 Min. : 1.00 Min. : -6.37000 Min. : -3.4400
## 1st Qu.: -43.1000 1st Qu.:17.00 1st Qu.: -1.33000 1st Qu.: -0.8000
## Median : 0.0000 Median :27.00 Median : 0.08000 Median : -0.2400
## Mean : -0.6188 Mean :25.51 Mean : 0.04277 Mean : -0.2571
## 3rd Qu.: 45.8750 3rd Qu.:33.00 3rd Qu.: 1.57000 3rd Qu.: 0.1400
## Max. : 180.0000 Max. :66.00 Max. : 4.87000 Max. : 2.8400
## gyros_arm_z accel_arm_x accel_arm_y accel_arm_z
## Min. : -2.3300 Min. : -404.00 Min. : -318.0 Min. : -636.00
## 1st Qu.: -0.0700 1st Qu.: -242.00 1st Qu.: -54.0 1st Qu.: -143.00
## Median : 0.2300 Median : -44.00 Median : 14.0 Median : -47.00
## Mean : 0.2695 Mean : -60.24 Mean : 32.6 Mean : -71.25
## 3rd Qu.: 0.7200 3rd Qu.: 84.00 3rd Qu.: 139.0 3rd Qu.: 23.00
## Max. : 3.0200 Max. : 437.00 Max. : 308.0 Max. : 292.00
## magnet_arm_x magnet_arm_y magnet_arm_z roll_dumbbell
## Min. : -584.0 Min. : -392.0 Min. : -597.0 Min. : -153.71
## 1st Qu.: -300.0 1st Qu.: -9.0 1st Qu.: 131.2 1st Qu.: -18.49
## Median : 289.0 Median : 202.0 Median : 444.0 Median : 48.17
## Mean : 191.7 Mean : 156.6 Mean : 306.5 Mean : 23.84
## 3rd Qu.: 637.0 3rd Qu.: 323.0 3rd Qu.: 545.0 3rd Qu.: 67.61
## Max. : 782.0 Max. : 583.0 Max. : 694.0 Max. : 153.55
## pitch_dumbbell yaw_dumbbell total_accel_dumbbell
## Min. : -149.59 Min. : -150.871 Min. : 0.00
## 1st Qu.: -40.89 1st Qu.: -77.644 1st Qu.: 4.00
## Median : -20.96 Median : -3.324 Median :10.00
## Mean : -10.78 Mean : 1.674 Mean :13.72
## 3rd Qu.: 17.50 3rd Qu.: 79.643 3rd Qu.:19.00
## Max. : 149.40 Max. : 154.952 Max. :58.00
## gyros_dumbbell_x gyros_dumbbell_y gyros_dumbbell_z
## Min. : -204.0000 Min. : -2.10000 Min. : -2.380
## 1st Qu.: -0.0300 1st Qu.: -0.14000 1st Qu.: -0.310
## Median : 0.1300 Median : 0.03000 Median : -0.130
## Mean : 0.1611 Mean : 0.04606 Mean : -0.129
## 3rd Qu.: 0.3500 3rd Qu.: 0.21000 3rd Qu.: 0.030
## Max. : 2.2200 Max. :52.00000 Max. :317.000
## accel_dumbbell_x accel_dumbbell_y accel_dumbbell_z magnet_dumbbell_x
## Min. : -419.00 Min. : -189.00 Min. : -334.00 Min. : -643.0
## 1st Qu.: -50.00 1st Qu.: -8.00 1st Qu.: -142.00 1st Qu.: -535.0
## Median : -8.00 Median : 41.50 Median : -1.00 Median : -479.0
## Mean : -28.62 Mean : 52.63 Mean : -38.32 Mean : -328.5
## 3rd Qu.: 11.00 3rd Qu.: 111.00 3rd Qu.: 38.00 3rd Qu.: -304.0
## Max. : 235.00 Max. : 315.00 Max. : 318.00 Max. : 592.0

```

```

## magnet_dumbbell_y magnet_dumbbell_z roll_forearm pitch_forearm
## Min. : -3600 Min. : -262.00 Min. : -180.0000 Min. : -72.50
## 1st Qu.: 231 1st Qu.: -45.00 1st Qu.: -0.7375 1st Qu.: 0.00
## Median : 311 Median : 13.00 Median : 21.7000 Median : 9.24
## Mean : 221 Mean : 46.05 Mean : 33.8265 Mean : 10.71
## 3rd Qu.: 390 3rd Qu.: 95.00 3rd Qu.: 140.0000 3rd Qu.: 28.40
## Max. : 633 Max. : 452.00 Max. : 180.0000 Max. : 89.80
## yaw_forearm total_accel_forearm gyros_forearm_x
## Min. : -180.00 Min. : 0.00 Min. : -22.000
## 1st Qu.: -68.60 1st Qu.: 29.00 1st Qu.: -0.220
## Median : 0.00 Median : 36.00 Median : 0.050
## Mean : 19.21 Mean : 34.72 Mean : 0.158
## 3rd Qu.: 110.00 3rd Qu.: 41.00 3rd Qu.: 0.560
## Max. : 180.00 Max. : 108.00 Max. : 3.970
## gyros_forearm_y gyros_forearm_z accel_forearm_x accel_forearm_y
## Min. : -7.02000 Min. : -8.0900 Min. : -498.00 Min. : -632.0
## 1st Qu.: -1.46000 1st Qu.: -0.1800 1st Qu.: -178.00 1st Qu.: 57.0
## Median : 0.03000 Median : 0.0800 Median : -57.00 Median : 201.0
## Mean : 0.07517 Mean : 0.1512 Mean : -61.65 Mean : 163.7
## 3rd Qu.: 1.62000 3rd Qu.: 0.4900 3rd Qu.: 76.00 3rd Qu.: 312.0
## Max. : 311.00000 Max. : 231.0000 Max. : 477.00 Max. : 923.0
## accel_forearm_z magnet_forearm_x magnet_forearm_y magnet_forearm_z
## Min. : -446.00 Min. : -1280.0 Min. : -896.0 Min. : -973.0
## 1st Qu.: -182.00 1st Qu.: -616.0 1st Qu.: 2.0 1st Qu.: 191.0
## Median : -39.00 Median : -378.0 Median : 591.0 Median : 511.0
## Mean : -55.29 Mean : -312.6 Mean : 380.1 Mean : 393.6
## 3rd Qu.: 26.00 3rd Qu.: -73.0 3rd Qu.: 737.0 3rd Qu.: 653.0
## Max. : 291.00 Max. : 672.0 Max. : 1480.0 Max. : 1090.0
## classe
## A:5580
## B:3797
## C:3422
## D:3216
## E:3607
##

```