

Practical Machine Learning - Write-Up Project

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The goal of this project is to predict the manner in which people did the exercise. This is the “classe” variable in the training set. I created this report to describe how I built the model, how I used cross validation, what I think the expected out of sample error is. There is also the script for the 20 predictions.

Data Processing and Analysis

The training and testing datasets are available here:

Training dataset: <https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv>

Testing dataset: <https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv>

Load all the librarys necessary to run the project.

```
library(caret)
```

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

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

```
library(ggplot2)
```

```
library(corrplot)
```

```
library(randomForest)
```

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

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

assing the data from pml-training.csv and pml-testing.csv files to the variables.

```
setwd("C:/Users/220194/Documents/Data Science Specialization/08 Practical Machine Learning/PracticalMac  
training <- read.csv("pml-training.csv",na.strings = c("NA", ""))  
test <- read.csv("pml-testing.csv",na.strings = c("NA", ""))
```

We search in the variables those columns that dont contibute to the analysis because they are character or non-numerical.

```
sapply(test,class)
```

##	X	user_name	raw_timestamp_part_1
##	"integer"	"factor"	"integer"
##	raw_timestamp_part_2	cvtd_timestamp	new_window
##	"integer"	"factor"	"factor"
##	num_window	roll_belt	pitch_belt
##	"integer"	"numeric"	"numeric"
##	yaw_belt	total_accel_belt	kurtosis_roll_belt
##	"numeric"	"integer"	"logical"
##	kurtosis_picth_belt	kurtosis_yaw_belt	skewness_roll_belt
##	"logical"	"logical"	"logical"

## skewness_roll_belt.1	skewness_yaw_belt	max_roll_belt
## "logical"	"logical"	"logical"
## max_picth_belt	max_yaw_belt	min_roll_belt
## "logical"	"logical"	"logical"
## min_pitch_belt	min_yaw_belt	amplitude_roll_belt
## "logical"	"logical"	"logical"
## amplitude_pitch_belt	amplitude_yaw_belt	var_total_accel_belt
## "logical"	"logical"	"logical"
## avg_roll_belt	stddev_roll_belt	var_roll_belt
## "logical"	"logical"	"logical"
## avg_pitch_belt	stddev_pitch_belt	var_pitch_belt
## "logical"	"logical"	"logical"
## avg_yaw_belt	stddev_yaw_belt	var_yaw_belt
## "logical"	"logical"	"logical"
## gyros_belt_x	gyros_belt_y	gyros_belt_z
## "numeric"	"numeric"	"numeric"
## accel_belt_x	accel_belt_y	accel_belt_z
## "integer"	"integer"	"integer"
## magnet_belt_x	magnet_belt_y	magnet_belt_z
## "integer"	"integer"	"integer"
## roll_arm	pitch_arm	yaw_arm
## "numeric"	"numeric"	"numeric"
## total_accel_arm	var_accel_arm	avg_roll_arm
## "integer"	"logical"	"logical"
## stddev_roll_arm	var_roll_arm	avg_pitch_arm
## "logical"	"logical"	"logical"
## stddev_pitch_arm	var_pitch_arm	avg_yaw_arm
## "logical"	"logical"	"logical"
## stddev_yaw_arm	var_yaw_arm	gyros_arm_x
## "logical"	"logical"	"numeric"
## gyros_arm_y	gyros_arm_z	accel_arm_x
## "numeric"	"numeric"	"integer"
## accel_arm_y	accel_arm_z	magnet_arm_x
## "integer"	"integer"	"integer"
## magnet_arm_y	magnet_arm_z	kurtosis_roll_arm
## "integer"	"integer"	"logical"
## kurtosis_picth_arm	kurtosis_yaw_arm	skewness_roll_arm
## "logical"	"logical"	"logical"
## skewness_pitch_arm	skewness_yaw_arm	max_roll_arm
## "logical"	"logical"	"logical"
## max_picth_arm	max_yaw_arm	min_roll_arm
## "logical"	"logical"	"logical"
## min_pitch_arm	min_yaw_arm	amplitude_roll_arm
## "logical"	"logical"	"logical"
## amplitude_pitch_arm	amplitude_yaw_arm	roll_dumbbell
## "logical"	"logical"	"numeric"
## pitch_dumbbell	yaw_dumbbell	kurtosis_roll_dumbbell
## "numeric"	"numeric"	"logical"
## kurtosis_picth_dumbbell	kurtosis_yaw_dumbbell	skewness_roll_dumbbell
## "logical"	"logical"	"logical"
## skewness_pitch_dumbbell	skewness_yaw_dumbbell	max_roll_dumbbell
## "logical"	"logical"	"logical"
## max_picth_dumbbell	max_yaw_dumbbell	min_roll_dumbbell
## "logical"	"logical"	"logical"

```

##      min_pitch_dumbbell      min_yaw_dumbbell      amplitude_roll_dumbbell
##      "logical"              "logical"              "logical"
##      amplitude_pitch_dumbbell      amplitude_yaw_dumbbell      total_accel_dumbbell
##      "logical"              "logical"              "integer"
##      var_accel_dumbbell      avg_roll_dumbbell      stddev_roll_dumbbell
##      "logical"              "logical"              "logical"
##      var_roll_dumbbell      avg_pitch_dumbbell      stddev_pitch_dumbbell
##      "logical"              "logical"              "logical"
##      var_pitch_dumbbell      avg_yaw_dumbbell      stddev_yaw_dumbbell
##      "logical"              "logical"              "logical"
##      var_yaw_dumbbell      gyros_dumbbell_x      gyros_dumbbell_y
##      "logical"              "numeric"              "numeric"
##      gyros_dumbbell_z      accel_dumbbell_x      accel_dumbbell_y
##      "numeric"              "integer"              "integer"
##      accel_dumbbell_z      magnet_dumbbell_x      magnet_dumbbell_y
##      "integer"              "integer"              "integer"
##      magnet_dumbbell_z      roll_forearm      pitch_forearm
##      "integer"              "numeric"              "numeric"
##      yaw_forearm      kurtosis_roll_forearm      kurtosis_pitch_forearm
##      "numeric"              "logical"              "logical"
##      kurtosis_yaw_forearm      skewness_roll_forearm      skewness_pitch_forearm
##      "logical"              "logical"              "logical"
##      skewness_yaw_forearm      max_roll_forearm      max_pitch_forearm
##      "logical"              "logical"              "logical"
##      max_yaw_forearm      min_roll_forearm      min_pitch_forearm
##      "logical"              "logical"              "logical"
##      min_yaw_forearm      amplitude_roll_forearm      amplitude_pitch_forearm
##      "logical"              "logical"              "logical"
##      amplitude_yaw_forearm      total_accel_forearm      var_accel_forearm
##      "logical"              "integer"              "logical"
##      avg_roll_forearm      stddev_roll_forearm      var_roll_forearm
##      "logical"              "logical"              "logical"
##      avg_pitch_forearm      stddev_pitch_forearm      var_pitch_forearm
##      "logical"              "logical"              "logical"
##      avg_yaw_forearm      stddev_yaw_forearm      var_yaw_forearm
##      "logical"              "logical"              "logical"
##      gyros_forearm_x      gyros_forearm_y      gyros_forearm_z
##      "numeric"              "numeric"              "numeric"
##      accel_forearm_x      accel_forearm_y      accel_forearm_z
##      "integer"              "integer"              "integer"
##      magnet_forearm_x      magnet_forearm_y      magnet_forearm_z
##      "integer"              "integer"              "integer"
##      problem_id
##      "integer"

```

Using What we saw in the previous part we should remove some of the columns that not make sense to have with the grep function, because they are not numeric and create a new vector from this operation.

```

training_aux <- training[, -(grep("timestamp|X|user_name|num_window|new_window", names(training)))]
test_aux <- test[, -(grep("timestamp|X|user_name|num_window|new_window", names(test)))]

```

Many of the columns we load have NAs values. We take out from the variables those data with NA's values The NAs records make our machine learning algorithm less precise. Finally we update the vector with out any NA's values.

```
NAs <- apply(training_aux, 2, function(x) {
  sum(is.na(x))})

training_aux <- training_aux[, which(NAs == 0)]
test_aux <- test_aux[, which(NAs == 0)]
```

We split the data into two variables. First the training data with the 60 % and the test data with 40%.

```
training.idx <- training_aux[createDataPartition(y = training_aux$classe, p = 0.6, list = FALSE), ]
test.idx <- training_aux[-createDataPartition(y = training_aux$classe, p = 0.6, list = FALSE), ]
```

We take a look how our variables shrink, with out the NA's values.

```
dim(training.idx)
```

```
## [1] 11776    53
```

```
head(training.idx)
```

```
##   roll_belt pitch_belt yaw_belt total_accel_belt gyros_belt_x gyros_belt_y
## 1      1.41      8.07   -94.4              3         0.00         0.00
## 3      1.42      8.07   -94.4              3         0.00         0.00
## 5      1.48      8.07   -94.4              3         0.02         0.02
## 6      1.45      8.06   -94.4              3         0.02         0.00
## 7      1.42      8.09   -94.4              3         0.02         0.00
## 9      1.43      8.16   -94.4              3         0.02         0.00
##   gyros_belt_z accel_belt_x accel_belt_y accel_belt_z magnet_belt_x
## 1      -0.02         -21          4          22          -3
## 3      -0.02         -20          5          23          -2
## 5      -0.02         -21          2          24          -6
## 6      -0.02         -21          4          21           0
## 7      -0.02         -22          3          21          -4
## 9      -0.02         -20          2          24           1
##   magnet_belt_y magnet_belt_z roll_arm pitch_arm yaw_arm total_accel_arm
## 1           599         -313    -128     22.5    -161           34
## 3           600         -305    -128     22.5    -161           34
## 5           600         -302    -128     22.1    -161           34
## 6           603         -312    -128     22.0    -161           34
## 7           599         -311    -128     21.9    -161           34
## 9           602         -312    -128     21.7    -161           34
##   gyros_arm_x gyros_arm_y gyros_arm_z accel_arm_x accel_arm_y accel_arm_z
## 1           0.00         0.00     -0.02     -288        109     -123
## 3           0.02        -0.02     -0.02     -289        110     -126
## 5           0.00        -0.03         0.00     -289        111     -123
## 6           0.02        -0.03         0.00     -289        111     -122
## 7           0.00        -0.03         0.00     -289        111     -125
## 9           0.02        -0.03     -0.02     -288        109     -122
##   magnet_arm_x magnet_arm_y magnet_arm_z roll_dumbbell pitch_dumbbell
## 1          -368         337         516    13.05217    -70.49400
## 3          -368         344         513    12.85075    -70.27812
## 5          -374         337         506    13.37872    -70.42856
```

```

## 6      -369      342      513      13.38246      -70.81759
## 7      -373      336      509      13.12695      -70.24757
## 9      -369      341      518      13.15463      -70.42520
##  yaw_dumbbell total_accel_dumbbell gyros_dumbbell_x gyros_dumbbell_y
## 1     -84.87394      37      0      -0.02
## 3     -85.14078      37      0      -0.02
## 5     -84.85306      37      0      -0.02
## 6     -84.46500      37      0      -0.02
## 7     -85.09961      37      0      -0.02
## 9     -84.91563      37      0      -0.02
##  gyros_dumbbell_z accel_dumbbell_x accel_dumbbell_y accel_dumbbell_z
## 1      0      -234      47      -271
## 3      0      -232      46      -270
## 5      0      -233      48      -270
## 6      0      -234      48      -269
## 7      0      -232      47      -270
## 9      0      -232      47      -269
##  magnet_dumbbell_x magnet_dumbbell_y magnet_dumbbell_z roll_forearm
## 1     -559      293      -65      28.4
## 3     -561      298      -63      28.3
## 5     -554      292      -68      28.0
## 6     -558      294      -66      27.9
## 7     -551      295      -70      27.9
## 9     -549      292      -65      27.7
##  pitch_forearm yaw_forearm total_accel_forearm gyros_forearm_x
## 1     -63.9      -153      36      0.03
## 3     -63.9      -152      36      0.03
## 5     -63.9      -152      36      0.02
## 6     -63.9      -152      36      0.02
## 7     -63.9      -152      36      0.02
## 9     -63.8      -152      36      0.03
##  gyros_forearm_y gyros_forearm_z accel_forearm_x accel_forearm_y
## 1      0.00      -0.02      192      203
## 3     -0.02      0.00      196      204
## 5      0.00      -0.02      189      206
## 6     -0.02      -0.03      193      203
## 7      0.00      -0.02      195      205
## 9      0.00      -0.02      193      204
##  accel_forearm_z magnet_forearm_x magnet_forearm_y magnet_forearm_z
## 1     -215      -17      654      476
## 3     -213      -18      658      469
## 5     -214      -17      655      473
## 6     -215      -9      660      478
## 7     -215      -18      659      470
## 9     -214      -16      653      476
##  classe
## 1      A
## 3      A
## 5      A
## 6      A
## 7      A
## 9      A

```

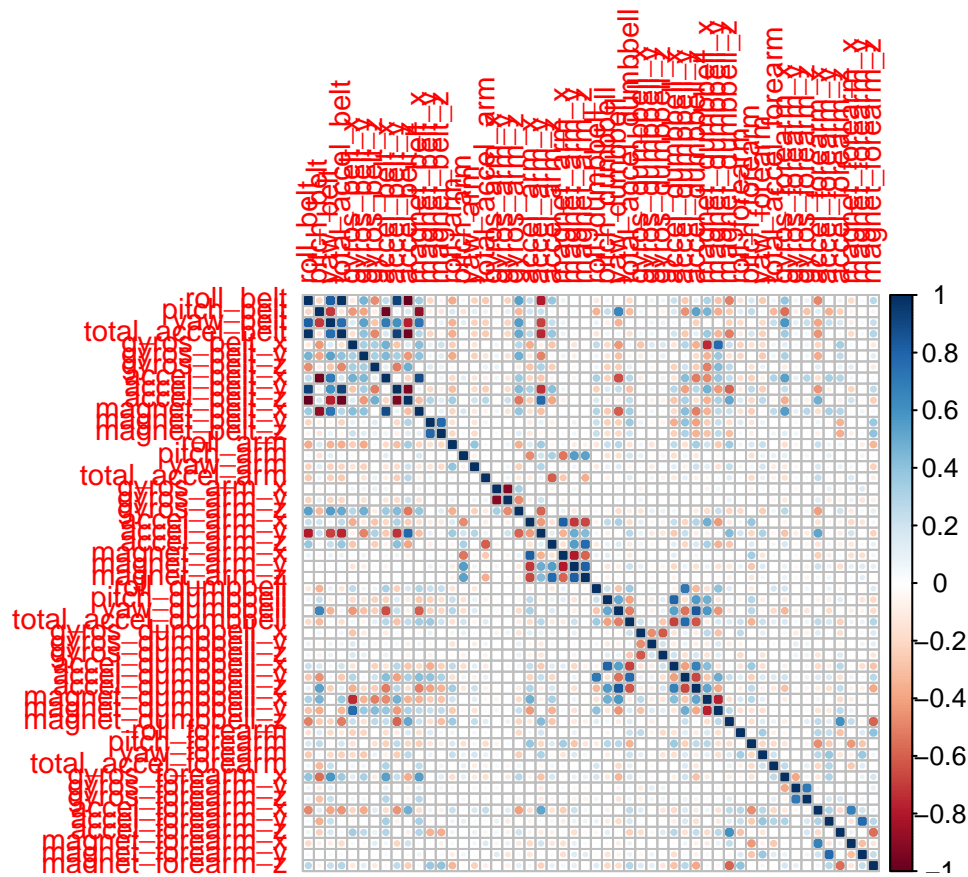
this graph is usefull to look at the correlation with some variables to each others. Note that the dark blue

indicated strong correlation and red negative correlation. Our dataset is 11776 X 53 and one of those is Classe, the variable we want to predict. This graph shows us nice information to implement a Random Forest Study.

```
Graph <- training.idx

NAs <- apply(training.idx, 2, function(x) {
  sum(is.na(x))
})
Graph<- training.idx[, which(NAs == 0)]

CorrePlot = cor( Graph[, -c(grep("timestamp|X|user_name|num_window|new_window", names(Graph))), length(Graph)
corrplot(CorrePlot, method="circle", tl.cex=0.9)
```



At this point we create our model using Random Forest and the set of data originally given but reduced. We train our model over the dataset using the “boot” method inside trainControl.

```
set.seed(10)

MyModel.Forest<- train(training.idx$classe ~ ., data = training.idx, method = "rf",
  prof = TRUE, trControl = trainControl(method = "boot", number = 5, allowParallel = TRUE))

summary(MyModel.Forest)
```

```
##           Length Class      Mode
## call           5 -none-    call
## type           1 -none-   character
## predicted     11776 factor    numeric
## err.rate       3000 -none-    numeric
## confusion       30 -none-    numeric
## votes         58880 matrix   numeric
## oob.times      11776 -none-    numeric
## classes        5 -none-   character
## importance      52 -none-    numeric
## importanceSD     0 -none-    NULL
## localImportance  0 -none-    NULL
## proximity       0 -none-    NULL
## ntree          1 -none-    numeric
## mtry           1 -none-    numeric
## forest         14 -none-    list
## y             11776 factor    numeric
## test           0 -none-    NULL
## inbag           0 -none-    NULL
## xNames         52 -none-   character
## problemType     1 -none-   character
## tuneValue       1 data.frame list
## obsLevels       5 -none-   character
```

```
MyModel.Forest
```

```
## Random Forest
##
## 11776 samples
##   52 predictor
##   5 classes: 'A', 'B', 'C', 'D', 'E'
##
## No pre-processing
## Resampling: Bootstrapped (5 reps)
## Summary of sample sizes: 11776, 11776, 11776, 11776, 11776
## Resampling results across tuning parameters:
##
##   mtry Accuracy   Kappa      Accuracy SD   Kappa SD
##    2   0.9863446  0.9827060  0.001599790  0.002023190
##   27   0.9862003  0.9825210  0.002381382  0.003031075
##   52   0.9788633  0.9732262  0.003715593  0.004725817
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was mtry = 2.
```

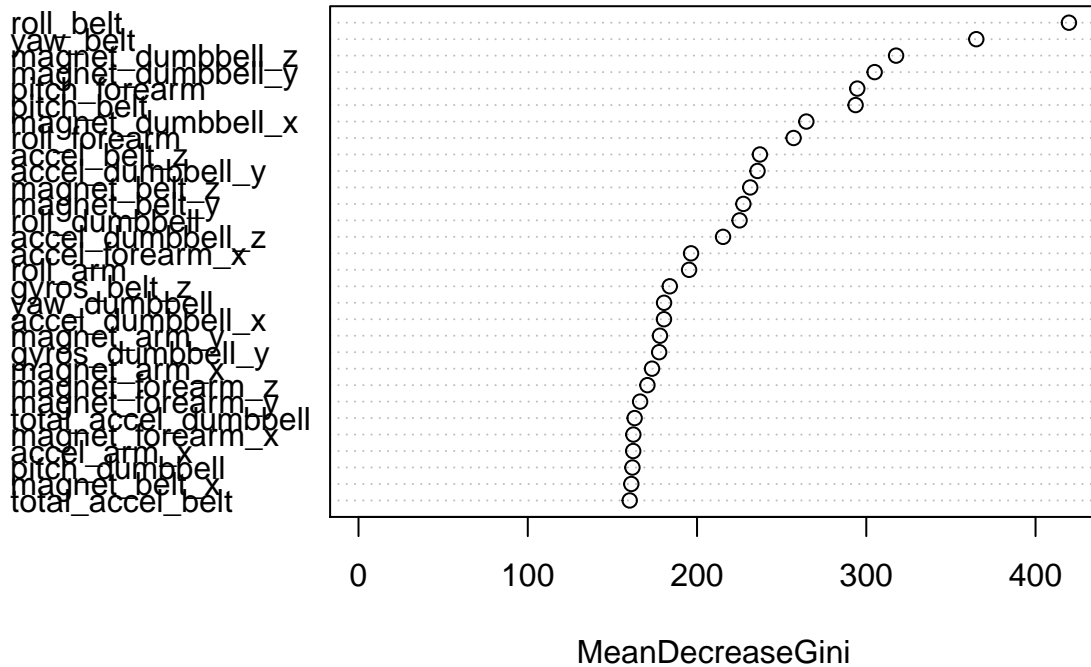
```
MyModel.Forest.Final<- MyModel.Forest$results
round(max(MyModel.Forest.Final$Accuracy), 3) * 100
```

```
## [1] 98.6
```

We got a very nice result of the accuracy at the first attempt: 98.6 %. That's ok.

```
varImpPlot(MyModel.Forest$finalModel,
  main = "Principal Components with high importance")
```

Principal Components with high importance



The cross validation study of our model.

```
test.idx$predRight <- (predict(MyModel.Forest, test.idx)) == test.idx$classe
table(predict(MyModel.Forest, test.idx), test.idx$classe)
```

```
##
##      A      B      C      D      E
## A 2231      9      0      0      0
## B      1 1505      6      0      0
## C      0      4 1361     16      2
## D      0      0      1 1270      4
## E      0      0      0      0 1436
```

```
CrossValidated<- postResample((predict(MyModel.Forest, test.idx)), test.idx$classe)
CrossValidated
```

```
## Accuracy      Kappa
## 0.9945195 0.9930670
```

We got a nice high level of accuracy.

We try out with the ConfussionMatrix

```
set.seed(10)
CrossValidatedError <- confusionMatrix((predict(MyModel.Forest, test.idx)), test.idx$classe)
CrossValidatedError
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    A    B    C    D    E
##           A 2231    9    0    0    0
##           B    1 1505    6    0    0
##           C    0    4 1361   16    2
##           D    0    0    1 1270    4
##           E    0    0    0    0 1436
##
## Overall Statistics
##
##           Accuracy : 0.9945
##           95% CI : (0.9926, 0.996)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9931
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9996  0.9914  0.9949  0.9876  0.9958
## Specificity      0.9984  0.9989  0.9966  0.9992  1.0000
## Pos Pred Value   0.9960  0.9954  0.9841  0.9961  1.0000
## Neg Pred Value   0.9998  0.9979  0.9989  0.9976  0.9991
## Prevalence       0.2845  0.1935  0.1744  0.1639  0.1838
## Detection Rate   0.2843  0.1918  0.1735  0.1619  0.1830
## Detection Prevalence 0.2855  0.1927  0.1763  0.1625  0.1830
## Balanced Accuracy 0.9990  0.9952  0.9957  0.9934  0.9979
```

```
postResample((predict(MyModel.Forest, test.idx)), test.idx$classe)[[1]]
```

```
## [1] 0.9945195
```

```
1- postResample((predict(MyModel.Forest, test.idx)), test.idx$classe)[[1]]
```

```
## [1] 0.0054805
```

we can see that our calculus were very close with the result of the matrix. The accuracy is 99.6%

The 20 cases to predict

Prepare the function to write the files that we are going to submit to Coursera

```
pml_write_files = function(x, directory="solutionfiles"){  
  dir.create (directory)  
  
  n = length(x)  
  for(i in 1:n){  
    filename = paste0("problem_id_",i,".txt")  
    filename=file.path(directory, filename)  
    write.table(x[i],file=filename,quote=FALSE,row.names=FALSE,col.names=FALSE)  
  }  
}
```

```
Model.Prediction <- predict(MyModel.Forest, test)  
Model.Prediction
```

```
## [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
```

```
pml_write_files(Model.Prediction)
```

```
## Warning in dir.create(directory): 'solutionfiles' already exists
```