

# Prediction Assignment Writeup

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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, my goal will be to use data from accelerometers on the belt, forearm, arm, and dumbbell of 6 participants.

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## What I 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.

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## Receiving data

All data in two tables pml-training.csv and pml-testing.csv. I will use pml-training data for model creating and testing. In code below I loading needable libraries and data.

```
#### Libraries
library(caret)

## Loading required package: lattice

## Loading required package: ggplot2

library(rpart)
library(rpart.plot)
library(randomForest)

## randomForest 4.6-14

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

##
## Attaching package: 'randomForest'

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

### Data loading
urlTrain<-"https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv"
url_final_test<-"https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv"
my_data<-read.csv(urlTrain,na.strings = c("NA","#DIV/0!", ""))
```

```

final_test<-read.csv(url_final_test,na.strings = c("NA","#DIV/0!",""))

###loaded data
str(my_data)

## 'data.frame':    19622 obs. of  160 variables:
## $ X                : int  1 2 3 4 5 6 7 8 9 10 ...
## $ user_name         : Factor w/ 6 levels "adelmo","carlitos",...: 2 2 2 2 2 2 2 2 2 2 ...
## $ raw_timestamp_part_1 : int  1323084231 1323084231 1323084231 1323084232 1323084232 1323084232 1323084232 ...
## $ raw_timestamp_part_2 : int  788290 808298 820366 120339 196328 304277 368296 440390 484323 484434 ...
## $ cvtd_timestamp      : Factor w/ 20 levels "02/12/2011 13:32",...: 9 9 9 9 9 9 9 9 9 9 ...
## $ new_window          : Factor w/ 2 levels "no","yes": 1 1 1 1 1 1 1 1 1 1 ...
## $ num_window          : int  11 11 11 12 12 12 12 12 12 12 ...
## $ 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 ...
## $ kurtosis_roll_belt  : num  NA NA NA NA NA NA NA NA NA NA ...
## $ kurtosis_pitch_belt : num  NA NA NA NA NA NA NA NA NA NA ...
## $ kurtosis_yaw_belt   : logi  NA NA NA NA NA NA NA NA NA NA ...
## $ skewness_roll_belt  : num  NA NA NA NA NA NA NA NA NA NA ...
## $ skewness_roll_belt.1 : num  NA NA NA NA NA NA NA NA NA NA ...
## $ skewness_yaw_belt   : logi  NA NA NA NA NA NA NA NA NA NA ...
## $ max_roll_belt       : num  NA NA NA NA NA NA NA NA NA NA ...
## $ max_pitch_belt      : int  NA NA NA NA NA NA NA NA NA NA ...
## $ max_yaw_belt        : num  NA NA NA NA NA NA NA NA NA NA ...
## $ min_roll_belt       : num  NA NA NA NA NA NA NA NA NA NA ...
## $ min_pitch_belt      : int  NA NA NA NA NA NA NA NA NA NA ...
## $ min_yaw_belt        : num  NA NA NA NA NA NA NA NA NA NA ...
## $ amplitude_roll_belt : num  NA NA NA NA NA NA NA NA NA NA ...
## $ amplitude_pitch_belt : int  NA NA NA NA NA NA NA NA NA NA ...
## $ amplitude_yaw_belt  : num  NA NA NA NA NA NA NA NA NA NA ...
## $ var_total_accel_belt : num  NA NA NA NA NA NA NA NA NA NA ...
## $ avg_roll_belt       : num  NA NA NA NA NA NA NA NA NA NA ...
## $ stddev_roll_belt    : num  NA NA NA NA NA NA NA NA NA NA ...
## $ var_roll_belt       : num  NA NA NA NA NA NA NA NA NA NA ...
## $ avg_pitch_belt      : num  NA NA NA NA NA NA NA NA NA NA ...
## $ stddev_pitch_belt   : num  NA NA NA NA NA NA NA NA NA NA ...
## $ var_pitch_belt      : num  NA NA NA NA NA NA NA NA NA NA ...
## $ avg_yaw_belt        : num  NA NA NA NA NA NA NA NA NA NA ...
## $ stddev_yaw_belt     : num  NA NA NA NA NA NA NA NA NA NA ...
## $ var_yaw_belt        : num  NA NA NA NA NA NA NA NA NA NA ...
## $ 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 ...
## $ 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 ...
## $ var_accel_arm      : num   NA NA NA NA NA NA NA NA NA NA ...
## $ avg_roll_arm       : num   NA NA NA NA NA NA NA NA NA NA ...
## $ stddev_roll_arm    : num   NA NA NA NA NA NA NA NA NA NA ...
## $ var_roll_arm       : num   NA NA NA NA NA NA NA NA NA NA ...
## $ avg_pitch_arm      : num   NA NA NA NA NA NA NA NA NA NA ...
## $ stddev_pitch_arm   : num   NA NA NA NA NA NA NA NA NA NA ...
## $ var_pitch_arm      : num   NA NA NA NA NA NA NA NA NA NA ...
## $ avg_yaw_arm        : num   NA NA NA NA NA NA NA NA NA NA ...
## $ stddev_yaw_arm     : num   NA NA NA NA NA NA NA NA NA NA ...
## $ var_yaw_arm        : num   NA NA NA NA NA NA NA NA NA NA ...
## $ 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 ...
## $ kurtosis_roll_arm  : num   NA NA NA NA NA NA NA NA NA NA ...
## $ kurtosis_pitch_arm : num   NA NA NA NA NA NA NA NA NA NA ...
## $ kurtosis_yaw_arm   : num   NA NA NA NA NA NA NA NA NA NA ...
## $ skewness_roll_arm  : num   NA NA NA NA NA NA NA NA NA NA ...
## $ skewness_pitch_arm : num   NA NA NA NA NA NA NA NA NA NA ...
## $ skewness_yaw_arm   : num   NA NA NA NA NA NA NA NA NA NA ...
## $ max_roll_arm       : num   NA NA NA NA NA NA NA NA NA NA ...
## $ max_pitch_arm      : num   NA NA NA NA NA NA NA NA NA NA ...
## $ max_yaw_arm        : int   NA NA NA NA NA NA NA NA NA NA ...
## $ min_roll_arm       : num   NA NA NA NA NA NA NA NA NA NA ...
## $ min_pitch_arm      : num   NA NA NA NA NA NA NA NA NA NA ...
## $ min_yaw_arm        : int   NA NA NA NA NA NA NA NA NA NA ...
## $ amplitude_roll_arm : num   NA NA NA NA NA NA NA NA NA NA ...
## $ amplitude_pitch_arm : num   NA NA NA NA NA NA NA NA NA NA ...
## $ amplitude_yaw_arm  : int   NA NA NA NA NA NA NA NA NA NA ...
## $ 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 ...
## $ kurtosis_roll_dumbbell : num  NA NA NA NA NA NA NA NA NA NA ...
## $ kurtosis_pitch_dumbbell : num  NA NA NA NA NA NA NA NA NA NA ...
## $ kurtosis_yaw_dumbbell : logi  NA NA NA NA NA NA NA ...

```

```
## $ skewness_roll_dumbbell : num NA NA NA NA NA NA NA NA NA NA ...
## $ skewness_pitch_dumbbell : num NA NA NA NA NA NA NA NA NA NA ...
## $ skewness_yaw_dumbbell : logi NA NA NA NA NA NA NA ...
## $ max_roll_dumbbell : num NA NA NA NA NA NA NA NA NA NA NA ...
## $ max_pitch_dumbbell : num NA NA NA NA NA NA NA NA NA NA NA ...
## $ max_yaw_dumbbell : num NA NA NA NA NA NA NA NA NA NA NA ...
## $ min_roll_dumbbell : num NA NA NA NA NA NA NA NA NA NA NA ...
## $ min_pitch_dumbbell : num NA NA NA NA NA NA NA NA NA NA NA ...
## $ min_yaw_dumbbell : num NA NA NA NA NA NA NA NA NA NA NA ...
## $ amplitude_roll_dumbbell : num NA NA NA NA NA NA NA NA NA NA ...
## [list output truncated]
```

## Preprocessing the Data

As we can see, in tables with data many fields with “NA” variables. I must clean data for analysis. In code below I will do it. Some variables have no sense, like “X”, “raw\_timestamp\_part\_1”, “raw\_timestamp\_part\_2”, “cvtd\_timestamp”, I remove these variables too. For training and testing I will use 75 and 25 percents of data “pml-training.csv”.

```
cut_off <- apply(my_data, 2, function(x) sum(is.na(x)))/nrow(my_data)
```

```
my_data <- my_data[!(cut_off > .95)]
final_test <- final_test[!(cut_off > .95)]
```

```
cut_off <- nearZeroVar(my_data)
```

```
my_data <- my_data[, -cut_off]
final_test <- final_test[, -cut_off]
```

```
my_data <- my_data[, -c(1,3,4,5)]
final_test <- final_test[, -c(1,3,4,5)]
```

```
str(my_data)
```

```
## 'data.frame':    19622 obs. of  55 variables:
## $ user_name      : Factor w/ 6 levels "adelmo","carlitos",...: 2 2 2 2 2 2 2 2 2 2 ...
## $ num_window     : int  11 11 11 12 12 12 12 12 12 12 ...
## $ 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 ...
## $ 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 -63.9 -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 0.02 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
...
```

```
my_train<-createDataPartition(my_data$classe,p=.75,list = FALSE)
```

```
train_data<-my_data[my_train,]
test_data<-my_data[-my_train,]
```

## Processing the Data

### Rpart model

```
set.seed(1)
fit_rpart<-rpart(data = train_data,classe ~ ., method = "class")
pred_rpart<-predict(fit_rpart,test_data,type = "class")

rpart_prediction<-confusionMatrix(pred_rpart,test_data$classe)
print(rpart_prediction)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction    A    B    C    D    E
##      A 1248  200   41   74   46
##      B   38  511   27   19   25
##      C    7   55  695  110   60
##      D   81  136   48  536   94
##      E   21   47   44   65  676
##
## Overall Statistics
##
##              Accuracy : 0.7476
##              95% CI : (0.7351, 0.7597)
##      No Information Rate : 0.2845
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.6794
##  Mcnemar's Test P-Value : < 2.2e-16
##
## Statistics by Class:
##
##              Class: A Class: B Class: C Class: D Class: E
## Sensitivity          0.8946   0.5385   0.8129   0.6667   0.7503
## Specificity          0.8971   0.9724   0.9427   0.9124   0.9558
## Pos Pred Value       0.7756   0.8242   0.7497   0.5989   0.7925
## Neg Pred Value       0.9554   0.8978   0.9598   0.9332   0.9445
## Prevalence           0.2845   0.1935   0.1743   0.1639   0.1837
## Detection Rate       0.2545   0.1042   0.1417   0.1093   0.1378
## Detection Prevalence 0.3281   0.1264   0.1890   0.1825   0.1739
## Balanced Accuracy     0.8959   0.7555   0.8778   0.7896   0.8530

varImp(fit_rpart)

##              Overall
## accel_arm_x      420.84237
## accel_belt_x     230.13815
## accel_belt_y     222.50004
## accel_belt_z     662.20412
## accel_dumbbell_x  86.17580
## accel_dumbbell_y 1350.86993
## accel_dumbbell_z  271.50459
## accel_forearm_x  296.55551
## accel_forearm_z  246.33351
## gyros_belt_z     68.50919
```

```
## gyros_dumbbell_x      72.32089
## magnet_arm_x          437.21928
## magnet_arm_y          214.11438
## magnet_belt_y         616.82535
## magnet_belt_z         333.23569
## magnet_dumbbell_x     586.90553
## magnet_dumbbell_y     1450.13062
## magnet_dumbbell_z     1807.25191
## magnet_forearm_y      266.21022
## magnet_forearm_z      471.33506
## num_window            2563.87184
## pitch_belt            1343.02014
## pitch_dumbbell        92.59137
## pitch_forearm         1728.85460
## roll_arm              164.99628
## roll_belt             1450.77656
## roll_dumbbell         767.45005
## roll_forearm          1987.99319
## total_accel_belt      542.94590
## total_accel_dumbbell  321.14721
## user_name             410.66142
## yaw_belt              1675.93996
## yaw_dumbbell          125.92478
## gyros_belt_x           0.00000
## gyros_belt_y           0.00000
## magnet_belt_x          0.00000
## pitch_arm              0.00000
## yaw_arm                0.00000
## total_accel_arm       0.00000
## gyros_arm_x            0.00000
## gyros_arm_y            0.00000
## gyros_arm_z            0.00000
## accel_arm_y            0.00000
## accel_arm_z            0.00000
## magnet_arm_z           0.00000
## gyros_dumbbell_y       0.00000
## gyros_dumbbell_z       0.00000
## yaw_forearm            0.00000
## total_accel_forearm    0.00000
## gyros_forearm_x        0.00000
## gyros_forearm_y        0.00000
## gyros_forearm_z        0.00000
## accel_forearm_y        0.00000
## magnet_forearm_x       0.00000
```

We can see not all variables take part in model.

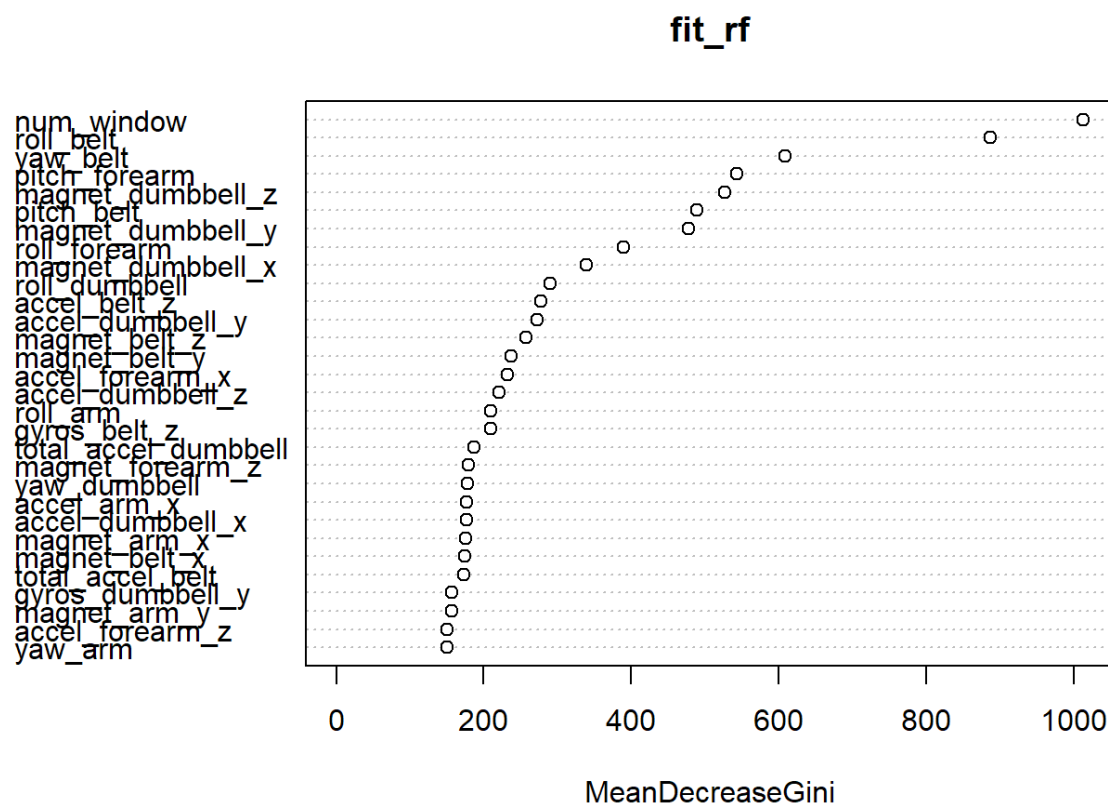
---

## Random Forest model

```
set.seed(1)
fit_rf<-randomForest(data = train_data, classe ~ ., method = "class")
pred_rf<-predict(fit_rf,test_data, type = "class")
rf_prediction<-confusionMatrix(pred_rf,test_data$classe)
print(rf_prediction)
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction   A    B    C    D    E
##           A 1395    2    0    0    0
##           B    0  947    0    0    0
##           C    0    0  855    9    0
##           D    0    0    0  793    0
##           E    0    0    0    2  901
##
## Overall Statistics
##
##           Accuracy : 0.9973
##           95% CI : (0.9955, 0.9986)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9966
##           Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      1.0000  0.9979  1.0000  0.9863  1.0000
## Specificity      0.9994  1.0000  0.9978  1.0000  0.9995
## Pos Pred Value   0.9986  1.0000  0.9896  1.0000  0.9978
## Neg Pred Value   1.0000  0.9995  1.0000  0.9973  1.0000
## Prevalence       0.2845  0.1935  0.1743  0.1639  0.1837
## Detection Rate   0.2845  0.1931  0.1743  0.1617  0.1837
## Detection Prevalence 0.2849  0.1931  0.1762  0.1617  0.1841
## Balanced Accuracy 0.9997  0.9989  0.9989  0.9932  0.9998
varImpPlot(fit_rf)
```





This model more slower, but more accuracy, in my mind all variables take part in model creation.

## Final result

For final result I use random forest model, because it more accuracy.

```
final_predict<-predict(fit_rf,final_test, method = "class")
print(final_predict)
```

```
##  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
```

## RESULT

In this work I use 2 models random forest and rpart. Rpart more faster but random forest have more accuracy.