

# Coursera Peer Reviewed Project - Practical Machine Learning Week 4 - Using accelerometers to predict if exercise is being performed correctly

*Karl Eiholzer*

*31 December 2016*

## Table of Contents

- Executive Summary
- Preliminary Steps
- Exploratory Data Analysis
- Data Cleaning
- Fitting Multiple Prediction Models
- Simplifying the Random Forest Model
- Conclusions
- Appendix 1: Timestamps and Data
- Appendix 2: Random Forest Data
- Appendix 3: Simplified Random Forest Data
- System and Version Information

## Executive Summary

[Skip to Top or next section](#)

The goal of the project is to predict if a person is exercising correctly based on accelerometer measurements:

- Lorem ipsum dolor sit amet, consectetur adipiscing elit.
- Lorem ipsum dolor sit amet, consectetur adipiscing elit.

## Preliminaries: load libraries we will be using

[Skip to prior section or next section](#)

Loading ggplot2 and caret packages, setting seed and loading in the data files.

## Exploratory Data Analysis

[Skip to prior section or next section](#)

The data comes from six test subjects (adelmo, carlitos, charles, eurico, jeremy, and pedro) engaged in five activities (labelled A thru E). As described by the original authors: “Class A corresponds to the specified execution of the exercise, while the other 4 classes correspond to common mistakes. Participants were

supervised by an experienced weight lifter to make sure the execution complied to the manner they were supposed to simulate. The exercises were performed by six male participants aged between 20-28 years, with little weight lifting experience. We made sure that all participants could easily simulate the mistakes in a safe and controlled manner by using a relatively light dumbbell (1.25kg).” Accelerometer measurements were captured as the participants exercised. [Read more: <http://groupware.les.inf.puc-rio.br/har#ixzz4TZMpuEj6>]

We see a well distributed number of observations for each person for each activity, from 469 to 1,177:

```
with(raw.training, table(classe, user_name))
```

```
##          user_name
## classe adelmo carlitos charles eurico jeremy pedro
##      A    1165      834      899      865    1177    640
##      B     776      690      745      592     489    505
##      C     750      493      539      489     652    499
##      D     515      486      642      582     522    469
##      E     686      609      711      542     562    497
```

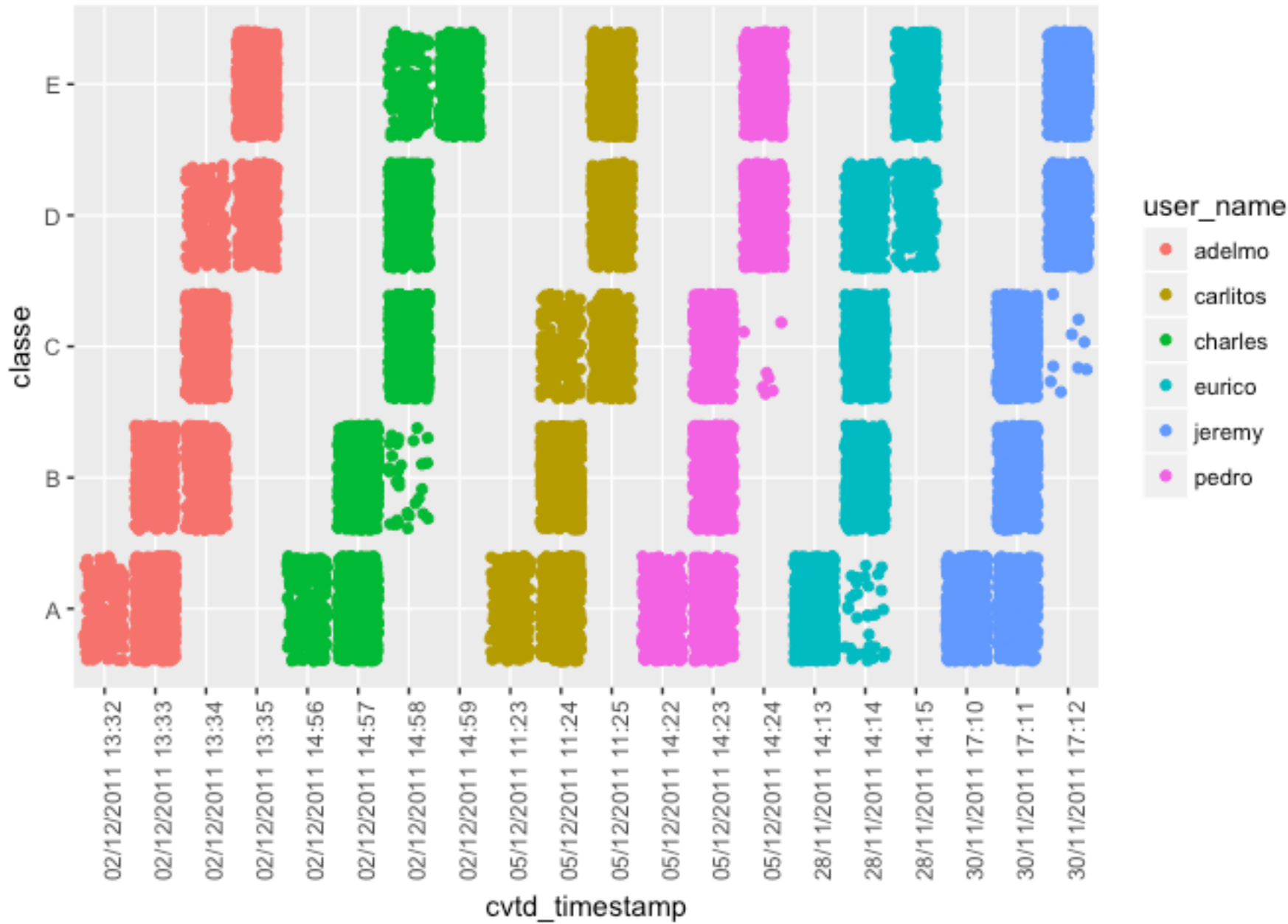
Many of the columns appear to be mostly comprised of NA values. The distribution of columns with high rates of NA values suggests on 59 columns are useful as predictors:

```
NA.rates <- cbind(Column = names(raw.training), Percent.NA = 0, High.NAs = FALSE)
y <- nrow(raw.training)
for(i in 1:length(raw.training) )
  { x <- round( sum( is.na( raw.training[, i] ) ) / y * 100 , digits = 0 )
    NA.rates[i , 2 ] <- x
    if (x > 95)
      { NA.rates[i , 3 ] <- TRUE
      }
  }
table(NA.rates[,2])
```

```
##
##      0 100  98
##     60   6  94
```

Of the 60 columns contain zero NA values, while 100 columns contain 98 or 100 percent NA values.

Also we see that the training data is organized by timestamp. The individuals worked through the various activities from A to E in order each time. Given that we would not normally have access to that information when predicting the activity, we want to build a model that is not dependent on timestamps. Finally we also see that the data is ordered by row number (“X”), which we would also not expect to occur in a normal environment:



See Appendix 1 for more data on this topic.

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

## Data Cleaning

[Skip to prior section](#) or [next section](#)

I'll remove the columns that contain mostly NA values and also any columns with near zero variance:

```

sig.training <- raw.training
for(i in 1:length(raw.training))
{ if( sum( is.na( raw.training[, i] ) ) /nrow(raw.training) >= .95)
  { for(j in 1:length(sig.training))
    { if( length( grep(names(raw.training[i]), names(sig.training)[j]) ) == 1)
      { sig.training <- sig.training[ , -j]
    }
  }
}
}

col.no <- which(names(sig.training)=="classe")
nzv <- nearZeroVar(sig.training[, -col.no],
                  freqCut = 99/1,
                  uniqueCut = 5,
                  saveMetrics=TRUE)
nzv.training <- sig.training[,nzv$nzv==FALSE]

dim(nzv.training)

```

```
## [1] 19622    60
```

The near zero variance test does not remove any columns, indicating that any of the remaining columns may be useful for prediction.

Lastly we will remove the user names and timestamp columns, as I believe they will not be useful for prediction. I also shuffle the rows randomly to assist with sampling techniques when fitting and cross validating.

```

col.no <- which(names(nzv.training)%in%c("user_name", "new_window", "X", "raw_timestamp_
part_1", "raw_timestamp_part_2", "cvtd_timestamp"))
clean.training <- nzv.training[, -col.no]
clean.training <- clean.training[sample(nrow(clean.training)),]
# Clean up the objects
rm(nzv.training)
rm(sig.training)

```

## Fitting a Random Forest Prediction Models

[Skip to prior section](#) or [next section](#)

After testing kNN, gbm and random forest models, I found the highest accuracy using random forest. First I train a model using all available predictors.

I start by creating separate training and validation stes:

```
set.seed(102767)
TrainIndex <- createDataPartition(clean.training$classe,
                                  p = .7,
                                  list = FALSE,
                                  times = 2)

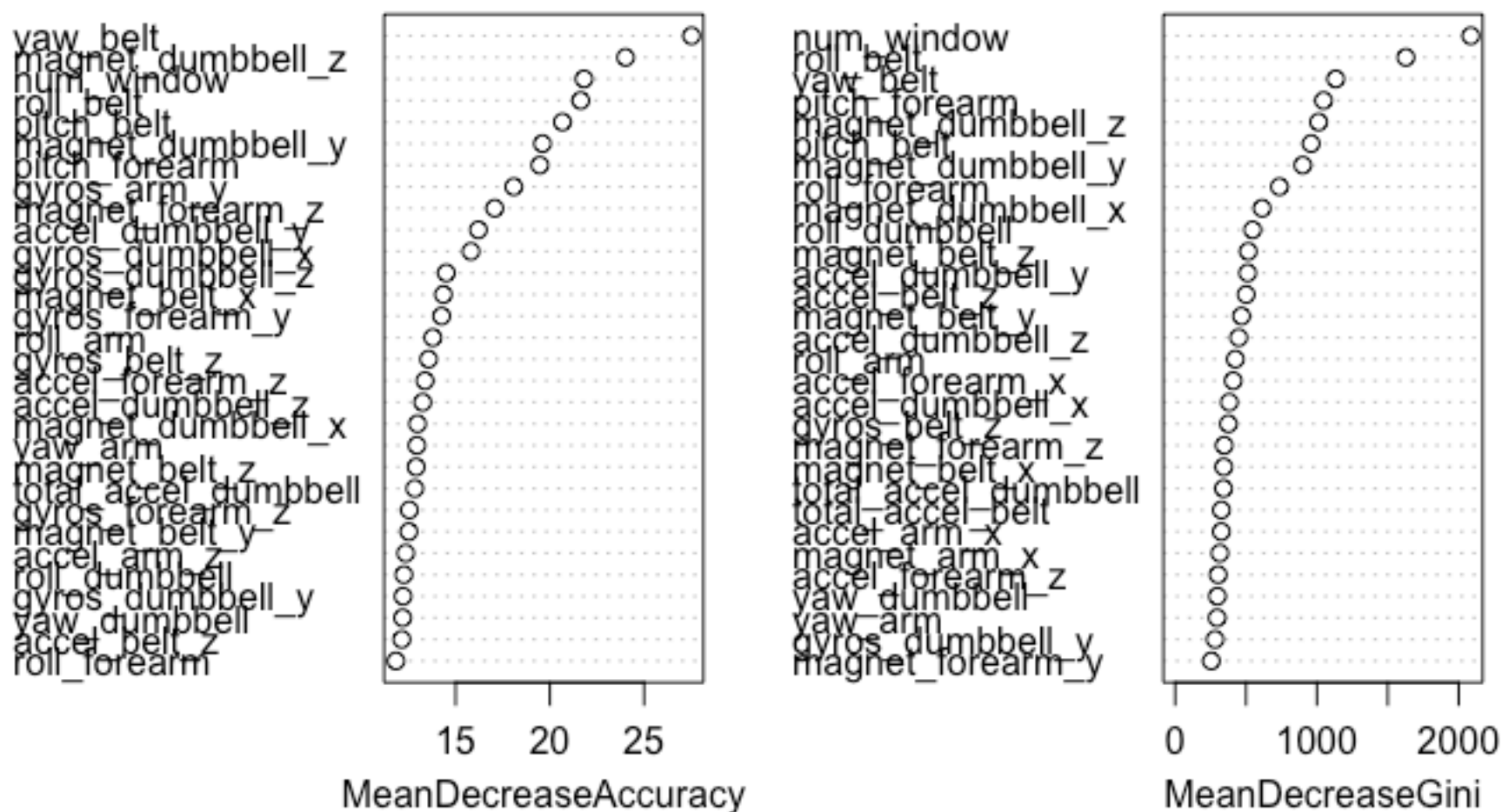
Training <- clean.training[TrainIndex,]
Validation <- clean.training[-TrainIndex,]
```

Due to mimitation of the machine I am using, I will run 100 trees, instead of the recomended 400 or more.

Full informatiomn about the random forest fit may be found in Appendix 2 :

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

rf.fit



This analysis identifies the most important variables for predicting which class of acticiy the exerciser was engaged in:

```
rownames(importance(rf.fit)[1:25,0])
```

```
## [1] "num_window"      "roll_belt"      "pitch_belt"
## [4] "yaw_belt"        "total_accel_belt" "gyros_belt_x"
## [7] "gyros_belt_y"    "gyros_belt_z"   "accel_belt_x"
## [10] "accel_belt_y"    "accel_belt_z"   "magnet_belt_x"
## [13] "magnet_belt_y"   "magnet_belt_z"  "roll_arm"
## [16] "pitch_arm"       "yaw_arm"        "total_accel_arm"
## [19] "gyros_arm_x"     "gyros_arm_y"    "gyros_arm_z"
## [22] "accel_arm_x"     "accel_arm_y"    "accel_arm_z"
## [25] "magnet_arm_x"
```

My next goal will be to create a simpler model with just those predictors.

## Simplifying the Random Forest Model

[Skip to prior section or next section](#)

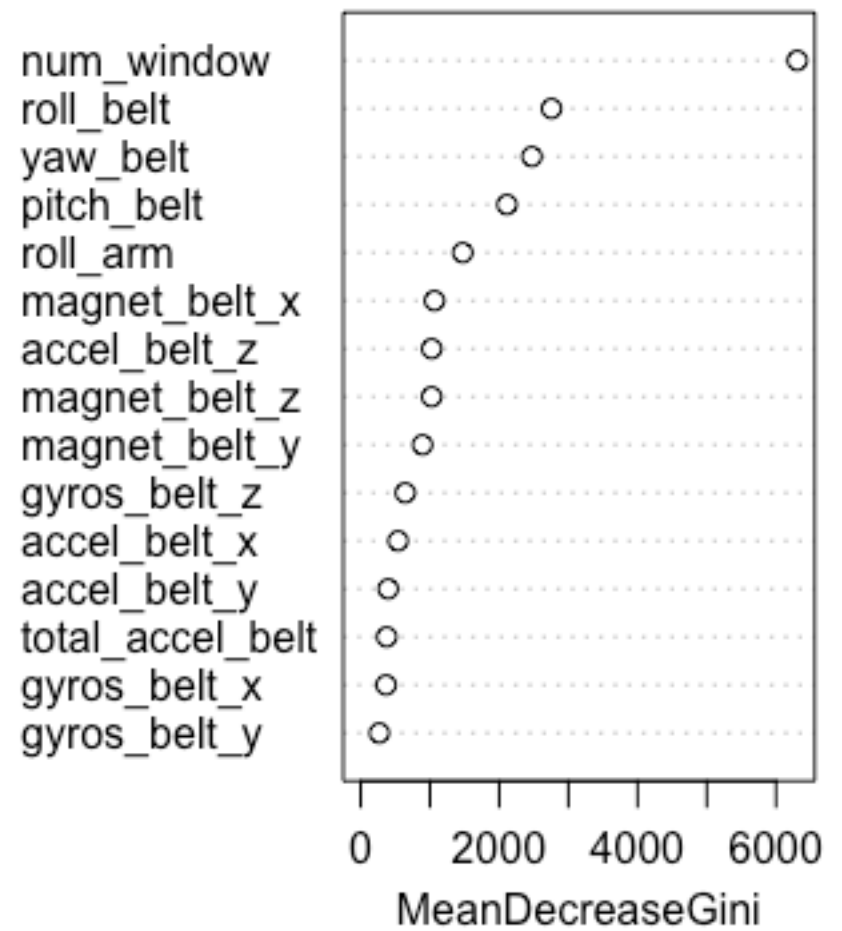
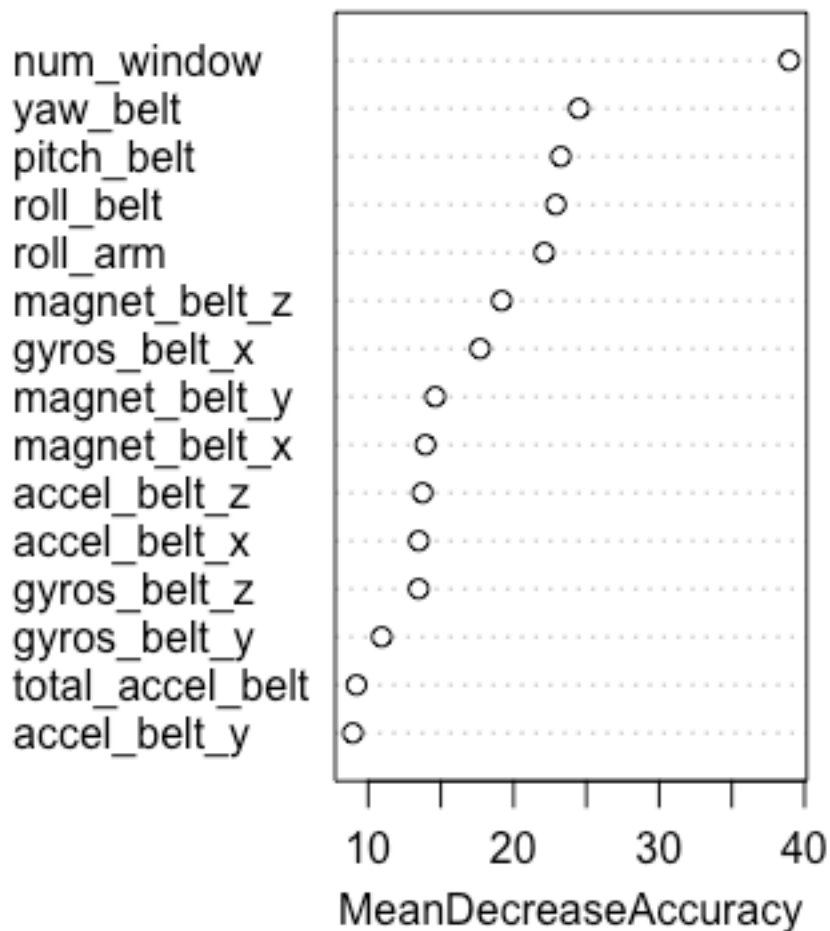
After testing a few smaller sets of predictors, I found that 15 variables would still achieve over 99% accuracy. I also found that accuracy improvements were not found after more than 60 trees. This allows me to run a more simplified version of the model:

```
set.seed(102767)
key.var <- rownames(importance(rf.fit)[1:15,0])
rf.fit2 <- randomForest(classe ~ .,
                        data = Training[,c("classe",key.var)],
                        ntree=60,
                        importance = TRUE,
                        proximity = TRUE )
rf.result2 <- predict(rf.fit2, Validation[,c("classe",key.var)])
```

Full information about the simplified model may be found in Appendix 3 :

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

rf.fit2



## Conclusions

[Skip to prior section](#) or [next section](#)

To confirm that the simplified model is as effective as the original full model, we can test both models against the testing data. :

```
# make columns identical between training and testing sets
col.no <- which(names(testing)%in%as.vector(colnames(Training)))
final.testing1 <- testing[,col.no]
# make columns identical between training and testing sets
col.no <- which(names(testing)%in%as.vector(c("classe",key.var)))
final.testing2 <- testing[,col.no]
# predict results
final.result1 <- predict(rf.fit, newdata = final.testing1)
final.result2 <- predict(rf.fit2, newdata = final.testing2)
```

A confusion matrix shows that both models return the same results:

```
confusionMatrix(final.result1, final.result2)$table
```

```
##           Reference
## Prediction A B C D E
##           A 7 0 0 0 0
##           B 0 8 0 0 0
##           C 0 0 1 0 0
##           D 0 0 0 1 0
##           E 0 0 0 0 3
```

Using the simplified model, the final predictions for the test set are as follows:

```
final.result2
```

```
##  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
```

## Appendix 1: Timestamps and Data

```
con <- file("Documentation/Exercise_Timestamp_data.txt", "r", blocking = FALSE)
readLines("Documentation/Exercise_Timestamp_data.txt")
```

```
## [1] "##### "
## [2] "# This is a Log Decsribing Relationships between timestamps and users in th
e data"
## [3] "# Created: Mon Jan 2 08:02:03 2017 "
## [4] "##### "
## [5] " "
## [6] " "
## [7] ""
## [8] ", , user_name = adelmo"
## [9] ""
## [10] "         classe"
## [11] "cvtd_timestamp      A      B      C      D      E"
## [12] " 02/12/2011 13:32 177      0      0      0      0"
## [13] " 02/12/2011 13:33 988 333      0      0      0"
## [14] " 02/12/2011 13:34      0 443 750 182      0"
## [15] " 02/12/2011 13:35      0      0      0 333 686"
## [16] " 02/12/2011 14:56      0      0      0      0      0"
## [17] " 02/12/2011 14:57      0      0      0      0      0"
## [18] " 02/12/2011 14:58      0      0      0      0      0"
## [19] " 02/12/2011 14:59      0      0      0      0      0"
## [20] " 05/12/2011 11:23      0      0      0      0      0"
## [21] " 05/12/2011 11:24      0      0      0      0      0"
## [22] " 05/12/2011 11:25      0      0      0      0      0"
## [23] " 05/12/2011 14:22      0      0      0      0      0"
## [24] " 05/12/2011 14:23      0      0      0      0      0"
```



```

## [25] " 05/12/2011 14:24 0 0 0 0 0"
## [26] " 28/11/2011 14:13 0 0 0 0 0"
## [27] " 28/11/2011 14:14 0 0 0 0 0"
## [28] " 28/11/2011 14:15 0 0 0 0 0"
## [29] " 30/11/2011 17:10 0 0 0 0 0"
## [30] " 30/11/2011 17:11 0 0 0 0 0"
## [31] " 30/11/2011 17:12 0 0 0 0 0"
## [32] ""
## [33] ", , user_name = carlitos"
## [34] ""
## [35] "               classe"
## [36] "cvtd_timestamp      A   B   C   D   E"
## [37] " 02/12/2011 13:32 0 0 0 0 0"
## [38] " 02/12/2011 13:33 0 0 0 0 0"
## [39] " 02/12/2011 13:34 0 0 0 0 0"
## [40] " 02/12/2011 13:35 0 0 0 0 0"
## [41] " 02/12/2011 14:56 0 0 0 0 0"
## [42] " 02/12/2011 14:57 0 0 0 0 0"
## [43] " 02/12/2011 14:58 0 0 0 0 0"
## [44] " 02/12/2011 14:59 0 0 0 0 0"
## [45] " 05/12/2011 11:23 190 0 0 0 0"
## [46] " 05/12/2011 11:24 644 690 163 0 0"
## [47] " 05/12/2011 11:25 0 0 330 486 609"
## [48] " 05/12/2011 14:22 0 0 0 0 0"
## [49] " 05/12/2011 14:23 0 0 0 0 0"
## [50] " 05/12/2011 14:24 0 0 0 0 0"
## [51] " 28/11/2011 14:13 0 0 0 0 0"
## [52] " 28/11/2011 14:14 0 0 0 0 0"
## [53] " 28/11/2011 14:15 0 0 0 0 0"
## [54] " 30/11/2011 17:10 0 0 0 0 0"
## [55] " 30/11/2011 17:11 0 0 0 0 0"
## [56] " 30/11/2011 17:12 0 0 0 0 0"
## [57] ""
## [58] ", , user_name = charles"
## [59] ""
## [60] "               classe"
## [61] "cvtd_timestamp      A   B   C   D   E"
## [62] " 02/12/2011 13:32 0 0 0 0 0"
## [63] " 02/12/2011 13:33 0 0 0 0 0"
## [64] " 02/12/2011 13:34 0 0 0 0 0"
## [65] " 02/12/2011 13:35 0 0 0 0 0"
## [66] " 02/12/2011 14:56 235 0 0 0 0"
## [67] " 02/12/2011 14:57 664 716 0 0 0"
## [68] " 02/12/2011 14:58 0 29 539 642 154"
## [69] " 02/12/2011 14:59 0 0 0 0 557"
## [70] " 05/12/2011 11:23 0 0 0 0 0"
## [71] " 05/12/2011 11:24 0 0 0 0 0"
## [72] " 05/12/2011 11:25 0 0 0 0 0"
## [73] " 05/12/2011 14:22 0 0 0 0 0"
## [74] " 05/12/2011 14:23 0 0 0 0 0"

```

```

## [75] " 05/12/2011 14:24 0 0 0 0 0"
## [76] " 28/11/2011 14:13 0 0 0 0 0"
## [77] " 28/11/2011 14:14 0 0 0 0 0"
## [78] " 28/11/2011 14:15 0 0 0 0 0"
## [79] " 30/11/2011 17:10 0 0 0 0 0"
## [80] " 30/11/2011 17:11 0 0 0 0 0"
## [81] " 30/11/2011 17:12 0 0 0 0 0"
## [82] ""
## [83] ", , user_name = eurico"
## [84] ""
## [85] "               classe"
## [86] "cvtd_timestamp      A   B   C   D   E"
## [87] " 02/12/2011 13:32 0 0 0 0 0"
## [88] " 02/12/2011 13:33 0 0 0 0 0"
## [89] " 02/12/2011 13:34 0 0 0 0 0"
## [90] " 02/12/2011 13:35 0 0 0 0 0"
## [91] " 02/12/2011 14:56 0 0 0 0 0"
## [92] " 02/12/2011 14:57 0 0 0 0 0"
## [93] " 02/12/2011 14:58 0 0 0 0 0"
## [94] " 02/12/2011 14:59 0 0 0 0 0"
## [95] " 05/12/2011 11:23 0 0 0 0 0"
## [96] " 05/12/2011 11:24 0 0 0 0 0"
## [97] " 05/12/2011 11:25 0 0 0 0 0"
## [98] " 05/12/2011 14:22 0 0 0 0 0"
## [99] " 05/12/2011 14:23 0 0 0 0 0"
## [100] " 05/12/2011 14:24 0 0 0 0 0"
## [101] " 28/11/2011 14:13 833 0 0 0 0"
## [102] " 28/11/2011 14:14 32 592 489 385 0"
## [103] " 28/11/2011 14:15 0 0 0 197 542"
## [104] " 30/11/2011 17:10 0 0 0 0 0"
## [105] " 30/11/2011 17:11 0 0 0 0 0"
## [106] " 30/11/2011 17:12 0 0 0 0 0"
## [107] ""
## [108] ", , user_name = jeremy"
## [109] ""
## [110] "               classe"
## [111] "cvtd_timestamp      A   B   C   D   E"
## [112] " 02/12/2011 13:32 0 0 0 0 0"
## [113] " 02/12/2011 13:33 0 0 0 0 0"
## [114] " 02/12/2011 13:34 0 0 0 0 0"
## [115] " 02/12/2011 13:35 0 0 0 0 0"
## [116] " 02/12/2011 14:56 0 0 0 0 0"
## [117] " 02/12/2011 14:57 0 0 0 0 0"
## [118] " 02/12/2011 14:58 0 0 0 0 0"
## [119] " 02/12/2011 14:59 0 0 0 0 0"
## [120] " 05/12/2011 11:23 0 0 0 0 0"
## [121] " 05/12/2011 11:24 0 0 0 0 0"
## [122] " 05/12/2011 11:25 0 0 0 0 0"
## [123] " 05/12/2011 14:22 0 0 0 0 0"
## [124] " 05/12/2011 14:23 0 0 0 0 0"

```

```
## [125] " 05/12/2011 14:24 0 0 0 0 0"
## [126] " 28/11/2011 14:13 0 0 0 0 0"
## [127] " 28/11/2011 14:14 0 0 0 0 0"
## [128] " 28/11/2011 14:15 0 0 0 0 0"
## [129] " 30/11/2011 17:10 869 0 0 0 0"
## [130] " 30/11/2011 17:11 308 489 643 0 0"
## [131] " 30/11/2011 17:12 0 0 9 522 562"
## [132] ""
## [133] ", , user_name = pedro"
## [134] ""
## [135] " classe"
## [136] "cvtd_timestamp A B C D E"
## [137] " 02/12/2011 13:32 0 0 0 0 0"
## [138] " 02/12/2011 13:33 0 0 0 0 0"
## [139] " 02/12/2011 13:34 0 0 0 0 0"
## [140] " 02/12/2011 13:35 0 0 0 0 0"
## [141] " 02/12/2011 14:56 0 0 0 0 0"
## [142] " 02/12/2011 14:57 0 0 0 0 0"
## [143] " 02/12/2011 14:58 0 0 0 0 0"
## [144] " 02/12/2011 14:59 0 0 0 0 0"
## [145] " 05/12/2011 11:23 0 0 0 0 0"
## [146] " 05/12/2011 11:24 0 0 0 0 0"
## [147] " 05/12/2011 11:25 0 0 0 0 0"
## [148] " 05/12/2011 14:22 267 0 0 0 0"
## [149] " 05/12/2011 14:23 373 505 492 0 0"
## [150] " 05/12/2011 14:24 0 0 7 469 497"
## [151] " 28/11/2011 14:13 0 0 0 0 0"
## [152] " 28/11/2011 14:14 0 0 0 0 0"
## [153] " 28/11/2011 14:15 0 0 0 0 0"
## [154] " 30/11/2011 17:10 0 0 0 0 0"
## [155] " 30/11/2011 17:11 0 0 0 0 0"
## [156] " 30/11/2011 17:12 0 0 0 0 0"
## [157] ""
## [158] " "
## [159] ""
## [160] ""
```

```
close(con)
```

## Appendix 2: Log of Random Forest Model Data

```
con <- file("Documentation/Random_Forest_Model.txt", "r", blocking = FALSE)
readLines("Documentation/Random_Forest_Model.txt")
```

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

```
## [2] "# This is a Log Capturing the Random Forest Tree Results"
## [3] "# Created: Mon Jan 2 08:17:40 2017 "
## [4] "##### "
## [5] " "
## [6] " "
## [7] ""
## [8] "
           A           B           C           D           E"
## [9] "num_window      16.024718 20.763105 22.884317 21.042931 18.135973"
## [10] "roll_belt       15.061804 20.069067 18.121956 21.421957 17.159104"
## [11] "pitch_belt      11.521232 20.271608 16.186241 18.682291 16.258959"
## [12] "yaw_belt        17.761668 21.710618 21.203173 21.762556 14.281758"
## [13] "total_accel_belt 7.985035 10.013000 8.026394 8.324065 7.294517"
## [14] "gyros_belt_x    6.489707 9.222908 8.962467 6.754466 11.958591"
## [15] "gyros_belt_y    5.062034 7.525845 6.663050 6.655441 9.447508"
## [16] "gyros_belt_z    9.860442 12.899770 10.312663 10.699562 12.141820"
## [17] "accel_belt_x    6.723981 9.113073 8.819301 7.318663 7.538151"
## [18] "accel_belt_y    6.580888 6.292784 7.093961 7.721691 5.768616"
## [19] "accel_belt_z    8.929455 9.790304 10.851120 10.403680 9.073536"
## [20] "magnet_belt_x   7.643335 15.024559 10.940130 11.507913 11.324437"
## [21] "magnet_belt_y   10.164274 11.661418 11.841930 10.460043 9.930806"
## [22] "magnet_belt_z   8.610183 11.757244 10.537555 13.671386 11.347376"
## [23] "roll_arm        9.633819 11.718844 11.389241 12.418387 9.689523"
## [24] "pitch_arm       7.073879 10.688642 9.337041 8.702393 8.319898"
## [25] "yaw_arm         9.173545 11.939660 10.645555 11.988869 9.033401"
## [26] "total_accel_arm 4.065773 12.255909 9.802004 8.862552 9.268660"
## [27] "gyros_arm_x     6.837997 11.437193 10.861925 10.785481 9.099866"
## [28] "gyros_arm_y     7.971706 14.940366 14.673269 12.732876 14.524770"
## [29] "gyros_arm_z     5.875256 8.059275 8.957054 6.546702 5.483843"
## [30] "accel_arm_x     8.073746 9.365019 9.578887 11.111666 9.078450"
## [31] "accel_arm_y     8.930798 9.947869 8.872118 8.288839 8.252510"
## [32] "accel_arm_z     5.751341 11.962215 10.149486 11.749863 9.531502"
## [33] "magnet_arm_x    6.816815 6.693087 7.484870 7.446436 6.181297"
## [34] "magnet_arm_y    5.344110 7.884725 8.305119 9.261945 4.847302"
## [35] "magnet_arm_z    8.965879 9.517644 10.359628 8.870976 7.686243"
## [36] "roll_dumbbell   9.505241 11.000897 11.837891 11.557742 11.799073"
## [37] "pitch_dumbbell  5.788875 10.194994 6.994982 6.316711 6.727654"
## [38] "yaw_dumbbell    7.643916 10.647247 11.281808 9.802804 9.502272"
## [39] "total_accel_dumbbell 10.199545 10.755427 10.441957 10.456604 11.596779"
## [40] "gyros_dumbbell_x 6.489330 12.204698 10.340114 10.152866 10.660845"
## [41] "gyros_dumbbell_y 9.501006 11.032729 12.581875 9.683094 8.360437"
## [42] "gyros_dumbbell_z 9.068930 10.599558 10.667101 9.234897 8.032108"
## [43] "accel_dumbbell_x 6.808127 10.951642 8.703480 8.802517 12.044264"
## [44] "accel_dumbbell_y 13.243606 15.250873 12.722820 13.034037 12.410119"
## [45] "accel_dumbbell_z 9.749849 10.720788 10.541446 10.879988 13.819947"
## [46] "magnet_dumbbell_x 11.123105 11.951287 12.775915 12.074963 11.065135"
## [47] "magnet_dumbbell_y 16.584411 17.133598 18.400051 15.858149 15.173647"
## [48] "magnet_dumbbell_z 19.610103 19.604542 20.571914 18.842012 17.182463"
## [49] "roll_forearm    12.012509 10.960342 12.472810 10.011947 9.625897"
## [50] "pitch_forearm   15.495576 16.432206 18.601045 14.862421 14.638193"
## [51] "yaw_forearm     9.091583 8.621935 9.578849 8.638759 8.376614"
```

##	[52]	"total_accel_forearm"	7.476333	9.409764	9.291175	9.505572	9.274844"	
##	[53]	"gyros_forearm_x"	6.171802	9.217549	8.574533	7.464346	7.878066"	
##	[54]	"gyros_forearm_y"	6.029547	12.686295	12.245724	11.465795	8.036672"	
##	[55]	"gyros_forearm_z"	7.281156	10.690183	9.885007	11.040826	9.143009"	
##	[56]	"accel_forearm_x"	7.013529	12.469541	11.209159	13.021706	10.339376"	
##	[57]	"accel_forearm_y"	6.755595	10.711374	10.587576	9.340770	10.125554"	
##	[58]	"accel_forearm_z"	9.568225	10.744983	10.790727	10.122767	11.917588"	
##	[59]	"magnet_forearm_x"	6.253818	9.352312	8.039123	8.192869	8.275685"	
##	[60]	"magnet_forearm_y"	8.626153	9.361469	8.495428	7.425499	7.871474"	
##	[61]	"magnet_forearm_z"	8.811970	14.265156	13.322005	14.713244	11.355168"	
##	[62]	"	MeanDecreaseAccuracy					MeanDecreaseGini"
##	[63]	"num_window"		21.812618		2086.02468		
##	[64]	"roll_belt"		21.637675		1627.72503		
##	[65]	"pitch_belt"		20.668116		959.11980		
##	[66]	"yaw_belt"		27.520206		1133.04870		
##	[67]	"total_accel_belt"		10.257521		325.90343		
##	[68]	"gyros_belt_x"		10.374099		122.18924		
##	[69]	"gyros_belt_y"		7.983096		158.39258		
##	[70]	"gyros_belt_z"		13.564700		375.72891		
##	[71]	"accel_belt_x"		11.340411		157.84164		
##	[72]	"accel_belt_y"		8.568848		169.02132		
##	[73]	"accel_belt_z"		12.143365		501.03548		
##	[74]	"magnet_belt_x"		14.350032		342.96042		
##	[75]	"magnet_belt_y"		12.517295		468.36757		
##	[76]	"magnet_belt_z"		12.906328		515.44606		
##	[77]	"roll_arm"		13.790242		424.43246		
##	[78]	"pitch_arm"		10.859147		230.57481		
##	[79]	"yaw_arm"		12.943445		295.89065		
##	[80]	"total_accel_arm"		10.170574		117.55136		
##	[81]	"gyros_arm_x"		11.000457		163.76744		
##	[82]	"gyros_arm_y"		18.088131		152.42418		
##	[83]	"gyros_arm_z"		11.060543		63.26546		
##	[84]	"accel_arm_x"		10.295814		325.42604		
##	[85]	"accel_arm_y"		11.223355		180.45479		
##	[86]	"accel_arm_z"		12.350177		136.62487		
##	[87]	"magnet_arm_x"		7.222631		315.26480		
##	[88]	"magnet_arm_y"		7.480713		249.11900		
##	[89]	"magnet_arm_z"		10.912352		194.22427		
##	[90]	"roll_dumbbell"		12.252131		546.27555		
##	[91]	"pitch_dumbbell"		7.869416		231.89803		
##	[92]	"yaw_dumbbell"		12.189236		297.90259		
##	[93]	"total_accel_dumbbell"		12.833809		341.31610		
##	[94]	"gyros_dumbbell_x"		15.813607		139.85569		
##	[95]	"gyros_dumbbell_y"		12.207088		281.77080		
##	[96]	"gyros_dumbbell_z"		14.504690		90.19071		
##	[97]	"accel_dumbbell_x"		10.305355		381.66965		
##	[98]	"accel_dumbbell_y"		16.207887		511.56615		
##	[99]	"accel_dumbbell_z"		13.250517		448.37199		
##	[100]	"magnet_dumbbell_x"		12.976302		613.45545		
##	[101]	"magnet_dumbbell_y"		19.609452		900.27515		

```
## [102] "magnet_dumbbell_z" 24.019058 1012.00935"
## [103] "roll_forearm" 11.830875 735.83625"
## [104] "pitch_forearm" 19.469173 1046.87957"
## [105] "yaw_forearm" 10.995771 197.48178"
## [106] "total_accel_forearm" 11.256279 122.83480"
## [107] "gyros_forearm_x" 11.727381 85.62149"
## [108] "gyros_forearm_y" 14.257345 130.06487"
## [109] "gyros_forearm_z" 12.544092 100.66151"
## [110] "accel_forearm_x" 11.552417 409.22500"
## [111] "accel_forearm_y" 11.502167 148.49568"
## [112] "accel_forearm_z" 13.374053 302.88610"
## [113] "magnet_forearm_x" 8.652204 249.72640"
## [114] "magnet_forearm_y" 9.600141 254.96171"
## [115] "magnet_forearm_z" 17.073383 345.73828"
## [116] " "
## [117] ""
## [118] " "
## [119] ""
## [120] ""
## [121] " "
## [122] ""
## [123] ""
```

```
close(con)
```

## Appendix 3: Log of Simplified Random Forest Model Data

```
con <- file("Documentation/Random_Forest_Model2.txt", "r", blocking = FALSE)
readLines("Documentation/Random_Forest_Model2.txt")
```

```
## [1] "##### "
## [2] "# This is a Log Capturing the Simplified Random Forest Tree Results"
## [3] "# Created: Mon Jan 2 08:25:33 2017 "
## [4] "##### "
## [5] " "
## [6] " "
## [7] ""
## [8] "Confusion Matrix and Statistics"
## [9] ""
## [10] "          Reference"
## [11] "Prediction    A    B    C    D    E"
## [12] "          A 499    0    0    0    0"
## [13] "          B    0 347    1    0    0"
## [14] "          C    0    0 316    1    0"
## [15] "          D    0    0    0 299    0"
## [16] "          E    0    0    0    0 344"
## [17] ""
```

```

## [18] "Overall Statistics"
## [19] "
## [20] "          Accuracy : 0.9989          "
## [21] "          95% CI : (0.996, 0.9999)"
## [22] "    No Information Rate : 0.2761      "
## [23] "    P-Value [Acc > NIR] : < 2.2e-16  "
## [24] "
## [25] "          Kappa : 0.9986              "
## [26] " McNemar's Test P-Value : NA          "
## [27] ""
## [28] "Statistics by Class:"
## [29] ""
## [30] "          Class: A Class: B Class: C Class: D Class: E"
## [31] "Sensitivity          1.0000    1.0000    0.9968    0.9967    1.0000"
## [32] "Specificity          1.0000    0.9993    0.9993    1.0000    1.0000"
## [33] "Pos Pred Value       1.0000    0.9971    0.9968    1.0000    1.0000"
## [34] "Neg Pred Value       1.0000    1.0000    0.9993    0.9993    1.0000"
## [35] "Prevalence           0.2761    0.1920    0.1754    0.1660    0.1904"
## [36] "Detection Rate       0.2761    0.1920    0.1749    0.1655    0.1904"
## [37] "Detection Prevalence 0.2761    0.1926    0.1754    0.1655    0.1904"
## [38] "Balanced Accuracy     1.0000    0.9997    0.9981    0.9983    1.0000"
## [39] " "
## [40] ""
## [41] " "
## [42] ""
## [43] "          A          B          C          D          E"
## [44] "num_window    31.361124 37.561341 44.682921 32.721379 21.793081"
## [45] "roll_belt     17.219486 18.678343 21.290510 19.924290 17.445971"
## [46] "pitch_belt    20.887441 20.632304 18.189963 17.307170 16.336910"
## [47] "yaw_belt      16.281578 21.135192 17.641623 27.270936 15.397867"
## [48] "total_accel_belt 8.137787  7.815665  6.274316  8.234807  7.489202"
## [49] "gyros_belt_x  12.311223 10.787975 14.091696 10.619257 13.924750"
## [50] "gyros_belt_y   6.860335  8.282067  8.535046  8.610661  9.379573"
## [51] "gyros_belt_z  12.242936 10.992555 12.326802 10.812217 10.498117"
## [52] "accel_belt_x  10.673199 10.487307 12.052757  8.749628 12.620318"
## [53] "accel_belt_y   7.818817  7.701176  7.169086  8.541795  6.048712"
## [54] "accel_belt_z  11.259712  9.816876 13.752961 12.496770 11.501801"
## [55] "magnet_belt_x  11.070942 12.467056 14.485769 12.519904 14.898076"
## [56] "magnet_belt_y  12.102507 10.831155 12.788134 15.003102 13.038608"
## [57] "magnet_belt_z  13.911470 15.311670 16.837849 18.751685 12.188533"
## [58] "roll_arm      18.463348 15.898673 15.308792 19.813159 15.911326"
## [59] "          MeanDecreaseAccuracy MeanDecreaseGini"
## [60] "num_window          38.953899          6308.4931"
## [61] "roll_belt           22.907072          2756.5043"
## [62] "pitch_belt          23.240817          2112.7697"
## [63] "yaw_belt            24.488686          2473.3230"
## [64] "total_accel_belt     9.201700          375.1985"
## [65] "gyros_belt_x         17.694883          365.8504"
## [66] "gyros_belt_y         10.931989          267.9115"
## [67] "gyros_belt_z         13.472726          643.4181"

```

```
## [68] "accel_belt_x" 13.483021 539.0120"
## [69] "accel_belt_y" 8.942374 396.6869"
## [70] "accel_belt_z" 13.737058 1025.0580"
## [71] "magnet_belt_x" 13.941318 1065.3088"
## [72] "magnet_belt_y" 14.615295 892.6668"
## [73] "magnet_belt_z" 19.179077 1024.3218"
## [74] "roll_arm" 22.106129 1474.0091"
## [75] " "
## [76] ""
## [77] " "
## [78] ""
## [79] ""
## [80] " "
## [81] ""
## [82] ""
```

```
close(con)
```

## System and Version Infomation

Code Origininally Executed on:	Value
R Version	R version 3.3.2 (2016-10-31)
Operating System	darwin13.4.0
Architecture	x86_64

Return to top.

File created *Mon Jan 2 08:25:35 2017*