

August 30, 2020

The results below are generated from an R script.

```
## The required packages

library(lattice)
library(caret)
library(ggplot2)
library(randomForest)
library(rpart)
library(rattle)

set.seed(1234)

## Downloading the files

trainingUrl <- "https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv"
testingUrl <- "https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv"

if (!file.exists('./pml-testing.csv') & !file.exists('./pml-training.csv')){
  download.file(testingUrl, './pml-testing.csv', mode = 'wb')
  download.file(trainingUrl, './pml-training.csv', mode = 'wb')
}

## Loading the files

train <- read.csv("pml-training.csv", na.strings = c("NA", "#DIV/0!", ""))
test <- read.csv("pml-testing.csv", na.strings = c("NA", "#DIV/0!", ""))

## Deleting columns with missing values only

train<-train[,colSums(is.na(train)) == 0]
test<-test[,colSums(is.na(test)) == 0]

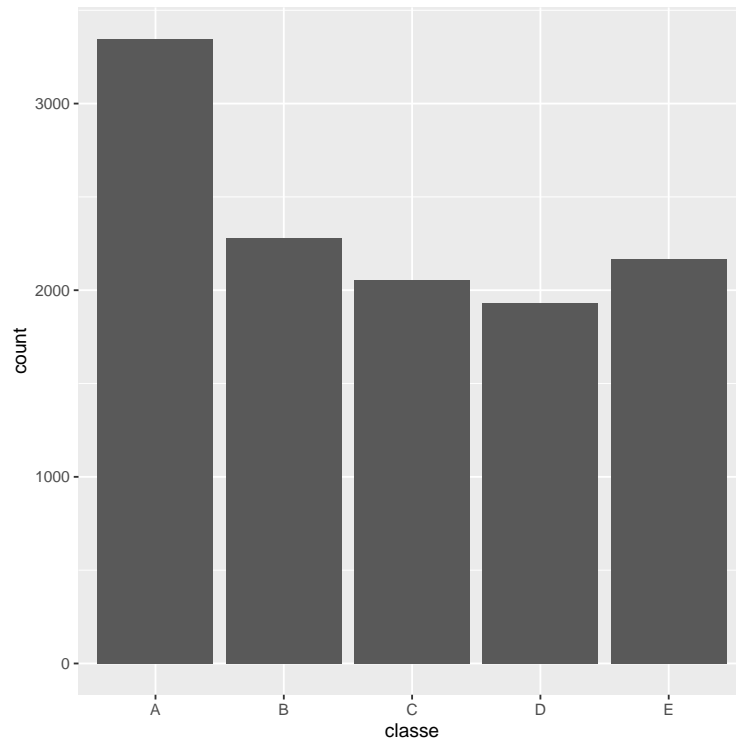
## Removing non-relevant variables

train <- train[, -c(1:7)]
test <- test[, -c(1:7)]

## Partitioning the training set

inTrain <- createDataPartition(y=train$classe, p = 0.6, list = FALSE)
trainTrain <- train [inTrain, ]
trainTest <- train[-inTrain, ]
```

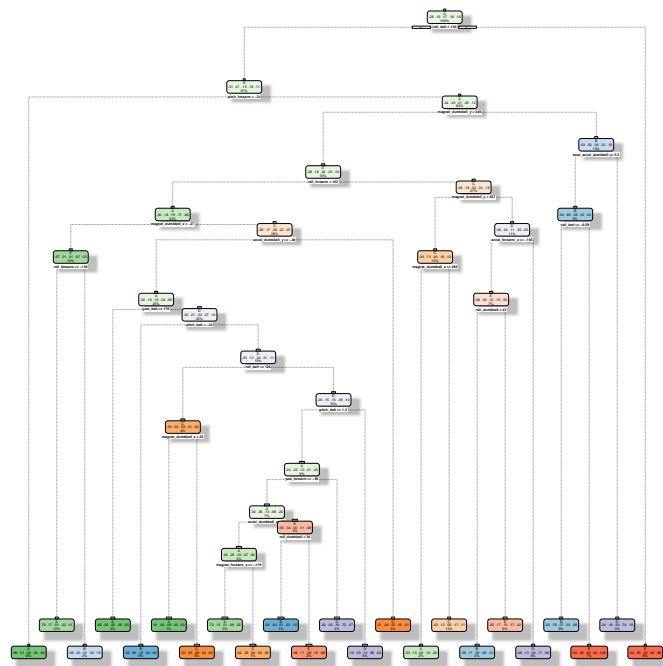
```
## Visualizing the training set
classe <- trainTrain$classe
ggplot(data.frame(classe), aes (x = classe)) + geom_bar()
```



```
## Decision Tree model with its plot and prediction

modFit1 <- rpart(classe ~ ., data=trainTrain, method = "class")
fancyRpartPlot(modFit1)

## Warning: labs do not fit even at cex 0.15, there may be some overplotting
```



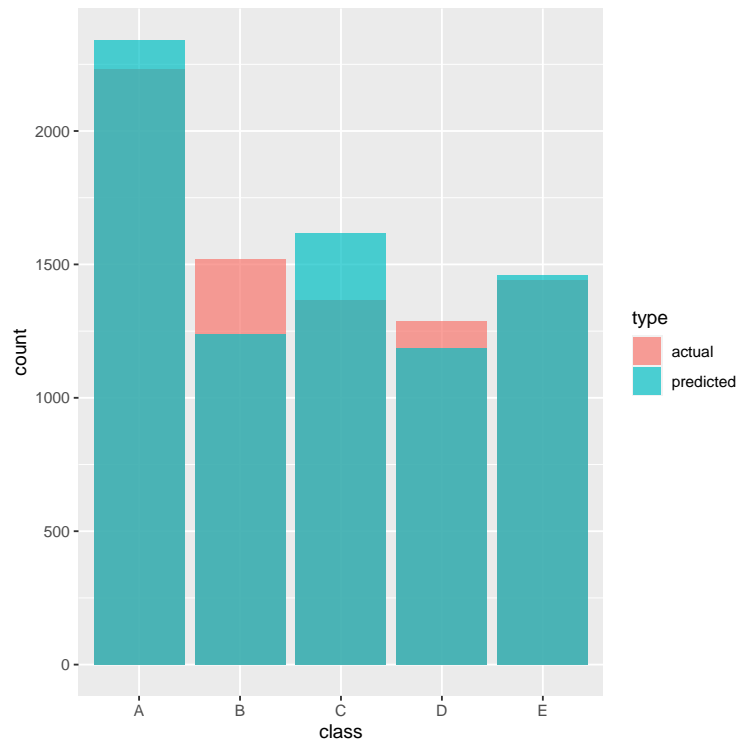
Rattle 2020-8-30 22:25:55 TA

```
prediction1 <- predict(modFit1, trainTest, type = "class")

## Plot the prediction in comparison to the actual value

p1 <- data.frame(class = prediction1)
cla <- data.frame(class = trainTest$classe)
p1$type <- 'predicted'
cla$type <- 'actual'
v <- rbind(p1, cla)
ggplot(v, aes(class, fill = type))+ geom_histogram(alpha = 0.7,
                                                    stat = "count", position = 'identity')

## Warning: Ignoring unknown parameters: binwidth, bins, pad
```



```
## Checking if the predictions and classe variables are same level factors

str(prediction1)

## Factor w/ 5 levels "A","B","C","D",...: 1 1 1 1 1 1 1 1 1 1 ...
## - attr(*, "names")= chr [1:7846] "1" "2" "6" "8" ...

str(trainTest$classe)

## chr [1:7846] "A" "A" "A" "A" "A" "A" "A" "A" "A" "A" "A" "A" "A" "A" "A" "A" "A" "A" ...

## Turning the classe variable into appropriate factor
factoredClasse <- factor(trainTest$classe, levels = c("A", "B", "C", "D", "E"))

## Testing the prediction
confusionMatrix(prediction1, factoredClasse)

## Confusion Matrix and Statistics
##
##           Reference
## Prediction   A    B    C    D    E
##      A 1930  231   47   81   53
##      B   82  866   67  103  121
##      C   59  189 1062  173  134
##      D   94  118   85  821   70
##      E   67  114  107  108 1064
##
## Overall Statistics
##
##               Accuracy : 0.732
##               95% CI : (0.722, 0.7417)
```

```

##      No Information Rate : 0.2845
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.6605
##
##      McNemar's Test P-Value : < 2.2e-16
##
## Statistics by Class:
##
##              Class: A Class: B Class: C Class: D Class: E
## Sensitivity          0.8647   0.5705   0.7763   0.6384   0.7379
## Specificity          0.9266   0.9411   0.9143   0.9441   0.9382
## Pos Pred Value       0.8241   0.6990   0.6568   0.6911   0.7288
## Neg Pred Value       0.9451   0.9013   0.9509   0.9302   0.9408
## Prevalence           0.2845   0.1935   0.1744   0.1639   0.1838
## Detection Rate       0.2460   0.1104   0.1354   0.1046   0.1356
## Detection Prevalence 0.2985   0.1579   0.2061   0.1514   0.1861
## Balanced Accuracy     0.8957   0.7558   0.8453   0.7912   0.8380

## Random forest model with its prediction

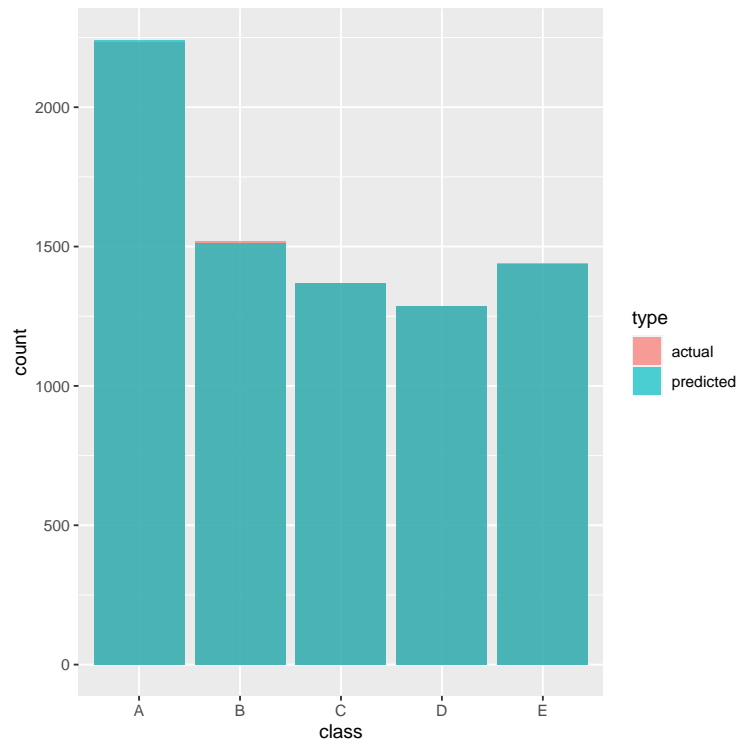
modFit2 <- randomForest(as.factor(classe) ~. , data = trainTrain)
prediction2 <- predict(modFit2, trainTest, type = "class")

## Plot the prediction in comparison to the actual value

p2 <- data.frame(class = prediction2)
p2$type <- 'predicted'
v2 <- rbind(p2, cla)
ggplot(v2, aes(class, fill = type))+ geom_histogram(alpha = 0.7,
                                                    stat = "count", position = 'identity')

## Warning: Ignoring unknown parameters: binwidth, bins, pad

```



Testing the result

`confusionMatrix(prediction2, factoredClasse)`

Confusion Matrix and Statistics

##

Reference

Prediction	A	B	C	D	E
A	2230	12	0	0	0
B	1	1506	5	0	0
C	0	0	1357	10	3
D	1	0	6	1274	5
E	0	0	0	2	1434

##

Overall Statistics

##

Accuracy : 0.9943

95% CI : (0.9923, 0.9958)

No Information Rate : 0.2845

P-Value [Acc > NIR] : < 2.2e-16

##

Kappa : 0.9927

##

McNemar's Test P-Value : NA

##

Statistics by Class:

##

	Class: A	Class: B	Class: C	Class: D	Class: E
Sensitivity	0.9991	0.9921	0.9920	0.9907	0.9945
Specificity	0.9979	0.9991	0.9980	0.9982	0.9997

## Pos Pred Value	0.9946	0.9960	0.9905	0.9907	0.9986
## Neg Pred Value	0.9996	0.9981	0.9983	0.9982	0.9988
## Prevalence	0.2845	0.1935	0.1744	0.1639	0.1838
## Detection Rate	0.2842	0.1919	0.1730	0.1624	0.1828
## Detection Prevalence	0.2858	0.1927	0.1746	0.1639	0.1830
## Balanced Accuracy	0.9985	0.9956	0.9950	0.9944	0.9971

The R session information (including the OS info, R version and all packages used):

```
sessionInfo()

## R version 4.0.2 (2020-06-22)
## Platform: x86_64-w64-mingw32/x64 (64-bit)
## Running under: Windows 10 x64 (build 19041)
##
## Matrix products: default
##
## locale:
## [1] LC_COLLATE=Korean_Korea.949 LC_CTYPE=Korean_Korea.949 LC_MONETARY=Korean_Korea.949
## [4] LC_NUMERIC=C LC_TIME=Korean_Korea.949
##
## attached base packages:
## [1] stats graphics grDevices utils datasets methods base
##
## other attached packages:
## [1] Hmisc_4.4-1 Formula_1.2-3 survival_3.2-3 RColorBrewer_1.1-2
## [5] rpart.plot_3.0.8 rattle_5.4.0 bitops_1.0-6 tibble_3.0.1
## [9] rpart_4.1-15 randomForest_4.6-14 caret_6.0-86 ggplot2_3.3.2
## [13] lattice_0.20-41
##
## loaded via a namespace (and not attached):
## [1] Rcpp_1.0.4.6 lubridate_1.7.9 png_0.1-7 class_7.3-17
## [5] packrat_0.5.0 digest_0.6.25 ipred_0.9-9 foreach_1.5.0
## [9] R6_2.4.1 plyr_1.8.6 backports_1.1.7 stats4_4.0.2
## [13] evaluate_0.14 e1071_1.7-3 highr_0.8 pillar_1.4.4
## [17] rlang_0.4.6 rstudioapi_0.11 data.table_1.13.0 Matrix_1.2-18
## [21] checkmate_2.0.0 rmarkdown_2.3 labeling_0.3 splines_4.0.2
## [25] gower_0.2.2 stringr_1.4.0 foreign_0.8-80 htmlwidgets_1.5.1
## [29] tinytex_0.24 munsell_0.5.0 xfun_0.15 compiler_4.0.2
## [33] pkgconfig_2.0.3 base64enc_0.1-3 htmltools_0.5.0 nnet_7.3-14
## [37] tidyselect_1.1.0 gridExtra_2.3 htmlTable_2.0.1 prodlim_2019.11.13
## [41] codetools_0.2-16 crayon_1.3.4 dplyr_1.0.0 withr_2.2.0
## [45] MASS_7.3-51.6 recipes_0.1.13 ModelMetrics_1.2.2.2 grid_4.0.2
## [49] nlme_3.1-148 gtable_0.3.0 lifecycle_0.2.0 magrittr_1.5
## [53] pROC_1.16.2 scales_1.1.1 stringi_1.4.6 farver_2.0.3
## [57] reshape2_1.4.4 latticeExtra_0.6-29 timeDate_3043.102 ellipsis_0.3.1
## [61] generics_0.0.2 vctrs_0.3.1 lava_1.6.7 iterators_1.0.12
## [65] tools_4.0.2 glue_1.4.1 purrr_0.3.4 jpeg_0.1-8.1
## [69] rsconnect_0.8.16 yaml_2.2.1 colorspace_1.4-1 cluster_2.1.0
## [73] knitr_1.29

Sys.time()

## [1] "2020-08-30 22:26:42 KST"
```