

# Coursera ML Week4: Practical Machine Learning

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## Overview:

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.

## Links For The Raw Data:

The data for this project are available here:

- Link to download training data: Training Data CSV  
(%22https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv%22)
- Link to download test data: Test Data CSV  
(%22https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv%22)
- Link to main source: Main Source (%22http://groupware.les.inf.puc-rio.br/har%22)

```
# To reproduce the result please change the working directory
workingDirectory = "/Projects/Rworkspace/coursera/courseraMLWeek4"
setwd(workingDirectory)
```

## Loading data & performing some basic exploratory data analysis:

```
# Packages
library(caret)
```

```
# checking if data directory exists if not creating it
if(!file.exists("./data")){
  dir.create("data")

  # Downloading the file
  trainDataUrlPath <- "https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv"
  download.file(trainDataUrlPath, "./data/trainData.csv", method="curl")

  # Downloading the file
  testDataUrlPath <- "https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv"
  download.file(testDataUrlPath, "./data/testData.csv", method="curl")
}
```

Taking the overview of the training data before processing.

```

# Loading the dataset
trainDataset <- read.csv("./data/trainData.csv", header = TRUE, sep = ",", quote =
"\\"", na.strings=c("NA", "#DIV/0!", ""))
validationDataset <- read.csv("./data/testData.csv", header = TRUE, sep = ",", quote
= "\\"", na.strings=c("NA", "#DIV/0!", ""))

# Cleaning the data
# removing near zero values
nearZeroVar <- nearZeroVar(trainDataset)
trainDataset <- trainDataset[, -nearZeroVar]

# removing na's
na1Val <- sapply(trainDataset, function(x) mean(is.na(x))) > 0.95
trainDataset <- trainDataset[, na1Val==FALSE]

# removing the column 1:5
trainDataset <- trainDataset[, -c(1:5)]

# Partitioning the data
inTrain <- createDataPartition(trainDataset$classe, p=0.7, list=FALSE)
trainingDataset <- trainDataset[inTrain,]
testingDataset <- trainDataset[-inTrain,]

# Checking the dimension of the training data
dim(trainingDataset)

```

```
## [1] 13737    54
```

```

# Checking the dimension of the validation data
dim(testingDataset)

```

```
## [1] 5885    54
```

## Model Building

### Using Decision Tree - Classification Model:

```

# setting seed
set.seed(987)

# Fitting model using caret package - (method = rpart, i.e: decision tree)
modelFitDecionTree <- train(classe ~ ., data=trainingDataset, method="rpart")
modelFitDecionTree

```

```
## CART
##
## 13737 samples
##    53 predictor
##    5 classes: 'A', 'B', 'C', 'D', 'E'
##
## No pre-processing
## Resampling: Bootstrapped (25 reps)
## Summary of sample sizes: 13737, 13737, 13737, 13737, 13737, 13737, ...
## Resampling results across tuning parameters:
##
##    cp          Accuracy    Kappa
##    0.03880582  0.5546425    0.42955823
##    0.05538602  0.4099944    0.19504182
##    0.11351846  0.3214589    0.05315859
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was cp = 0.03880582.
```

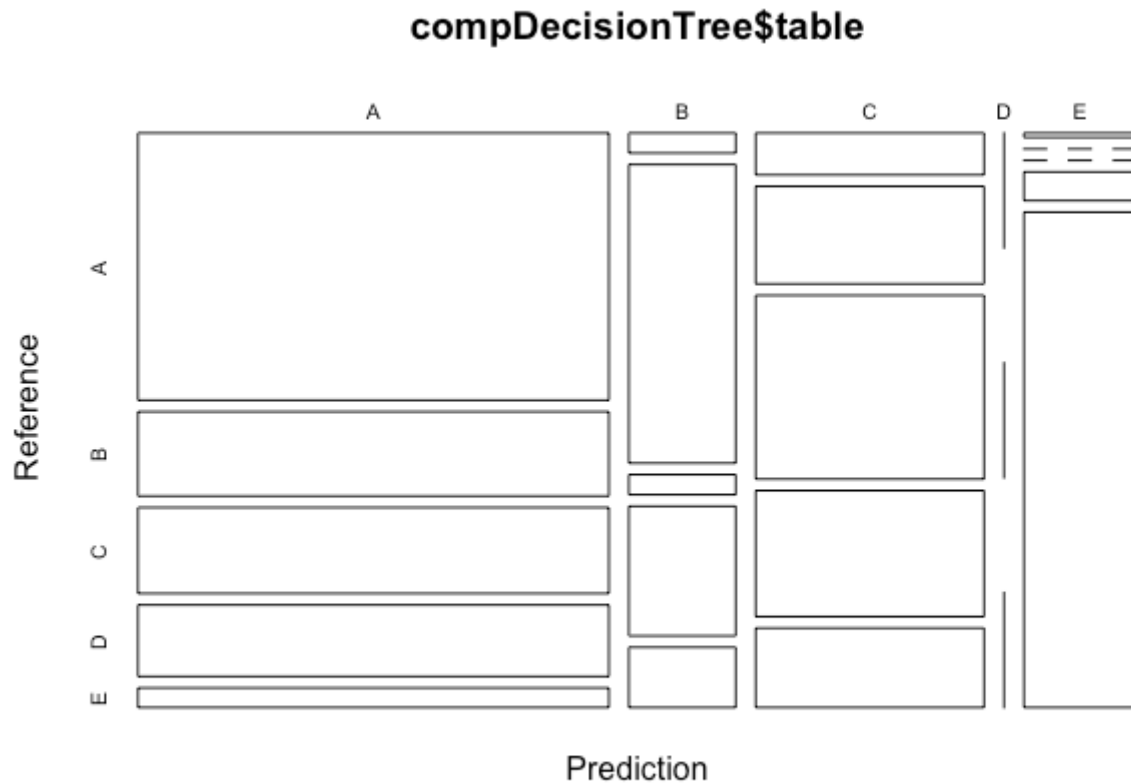
```
# Predicting on the test set
modelPredictDecionTree <- predict(modelFitDecionTree, newdata=testingDataset)
compDecisionTree <- confusionMatrix(modelPredictDecionTree, testingDataset$classe)
compDecisionTree
```

```
## Confusion Matrix and Statistics
##
##              Reference
## Prediction    A    B    C    D    E
##      A 1526  482  491  408  112
##      B   26  387   26  168   78
##      C  116  270  509  349  220
##      D    0    0    0    0    0
##      E    6    0    0   39  672
##
## Overall Statistics
##
##              Accuracy : 0.5257
##              95% CI : (0.5129, 0.5386)
##      No Information Rate : 0.2845
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.3807
##      McNemar's Test P-Value : < 2.2e-16
##
## Statistics by Class:
##
##              Class: A Class: B Class: C Class: D Class: E
## Sensitivity          0.9116  0.33977  0.49610  0.0000  0.6211
## Specificity          0.6455  0.93721  0.80346  1.0000  0.9906
## Pos Pred Value       0.5055  0.56496  0.34768   NaN  0.9372
## Neg Pred Value       0.9484  0.85538  0.88306  0.8362  0.9207
## Prevalence           0.2845  0.19354  0.17434  0.1638  0.1839
## Detection Rate       0.2593  0.06576  0.08649  0.0000  0.1142
## Detection Prevalence 0.5130  0.11640  0.24877  0.0000  0.1218
## Balanced Accuracy     0.7785  0.63849  0.64978  0.5000  0.8059
```

The following are the results obtained by Decision Tree model

- Decision Tree Accuracy: 0.5257434
- The out-of-sample error: 0.4742566

```
# graphical overview
plot(compDecisionTree$table, col = compDecisionTree$byClass)
```



## Using Random Forest:

```
# setting seed
set.seed(657)

# defining cross validation parameter for the model
crossValidationParamRF <- trainControl(
  method="cv",
  number=5,
  savePredictions = TRUE,
  classProbs = TRUE,
  verboseIter=FALSE
)

# Fitting model using caret package - (method = rf, i.e: random forest)
modelFitRandomForest <- train(classe ~ ., data=trainingDataset, method="rf",
                              trControl=crossValidationParamRF)

modelFitRandomForest
```

```

## Random Forest
##
## 13737 samples
##    53 predictor
##    5 classes: 'A', 'B', 'C', 'D', 'E'
##
## No pre-processing
## Resampling: Cross-Validated (5 fold)
## Summary of sample sizes: 10990, 10990, 10990, 10988, 10990
## Resampling results across tuning parameters:
##
##    mtry  Accuracy   Kappa
##    2     0.9938852  0.9922644
##    27     0.9965055  0.9955798
##    53     0.9936665  0.9919882
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was mtry = 27.

```

```

# Predicting on the test set
modelPredictRandomForest <- predict(modelFitRandomForest, newdata=testingDataset)

compRandomForest <- confusionMatrix(modelPredictRandomForest, testingDataset$classe)
compRandomForest

```

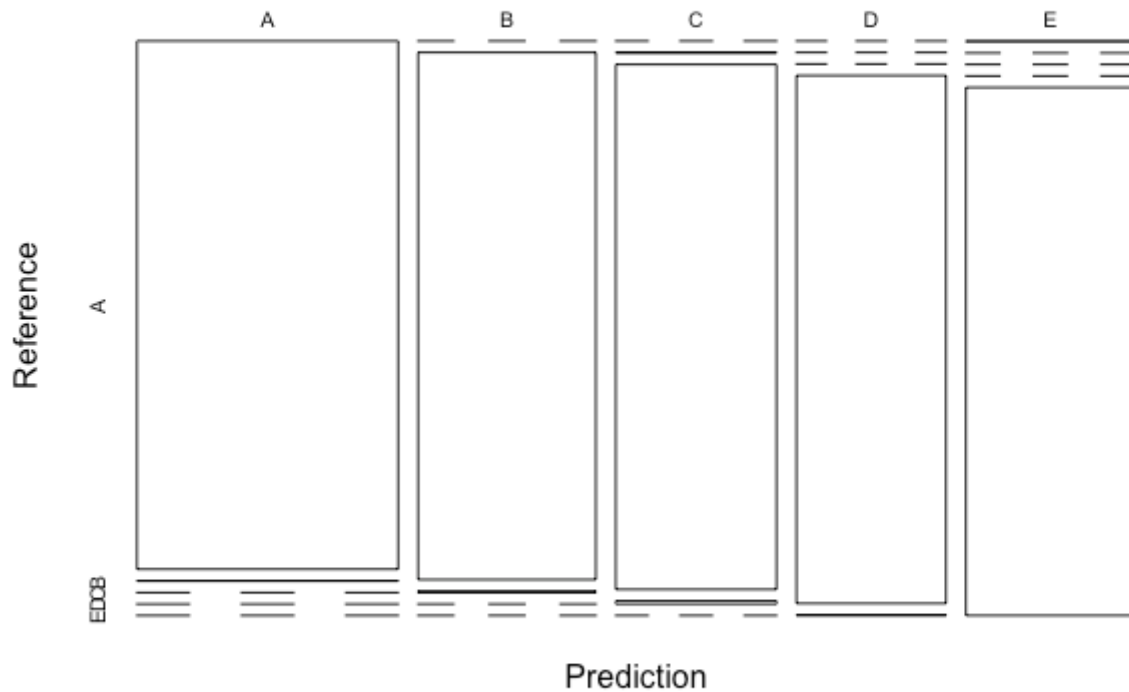
```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction   A      B      C      D      E
##           A 1673      1      0      0      0
##           B      0 1137      3      0      0
##           C      0      1 1023      6      0
##           D      0      0      0 958      1
##           E      1      0      0      0 1081
##
## Overall Statistics
##
##           Accuracy : 0.9978
##           95% CI : (0.9962, 0.9988)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9972
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9994  0.9982  0.9971  0.9938  0.9991
## Specificity      0.9998  0.9994  0.9986  0.9998  0.9998
## Pos Pred Value    0.9994  0.9974  0.9932  0.9990  0.9991
## Neg Pred Value    0.9998  0.9996  0.9994  0.9988  0.9998
## Prevalence        0.2845  0.1935  0.1743  0.1638  0.1839
## Detection Rate    0.2843  0.1932  0.1738  0.1628  0.1837
## Detection Prevalence 0.2845  0.1937  0.1750  0.1630  0.1839
## Balanced Accuracy 0.9996  0.9988  0.9978  0.9968  0.9994
```

*The following are the results obtained by Random Forest*

- Random Forest Accuracy: 0.997791
- The out-of-sample error: 0.002209

```
# graphical overview
plot(compRandomForest$table, col = compRandomForest$byClass)
```

## compRandomForest\$stable



### Using Generalized Boosted Model:

```
# setting seed
set.seed(345)

# defining cross validation parameter for the Generalized Boosted Regression Model
crossValidationParamGB <- trainControl(
  method="repeatedcv",
  number=5,
  repeats = 1,
  savePredictions = TRUE,
  classProbs = TRUE
)

# Fitting model using caret package - (method = gbm, i.e: generalized boosted regression)
modelFitGenBoost <- train(classe ~ ., data=trainingDataset, method="gbm",
  trControl=crossValidationParamGB)
```

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1282
##	2	1.5243	nan	0.1000	0.0833
##	3	1.4670	nan	0.1000	0.0667
##	4	1.4233	nan	0.1000	0.0542
##	5	1.3880	nan	0.1000	0.0490
##	6	1.3559	nan	0.1000	0.0447
##	7	1.3272	nan	0.1000	0.0357
##	8	1.3042	nan	0.1000	0.0401
##	9	1.2779	nan	0.1000	0.0319
##	10	1.2556	nan	0.1000	0.0281
##	20	1.0965	nan	0.1000	0.0212
##	40	0.9130	nan	0.1000	0.0092
##	60	0.7993	nan	0.1000	0.0063
##	80	0.7162	nan	0.1000	0.0048
##	100	0.6509	nan	0.1000	0.0036
##	120	0.5975	nan	0.1000	0.0044
##	140	0.5503	nan	0.1000	0.0024
##	150	0.5287	nan	0.1000	0.0024

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1891
##	2	1.4866	nan	0.1000	0.1319
##	3	1.4019	nan	0.1000	0.1040
##	4	1.3346	nan	0.1000	0.0895
##	5	1.2776	nan	0.1000	0.0727
##	6	1.2307	nan	0.1000	0.0680
##	7	1.1870	nan	0.1000	0.0627
##	8	1.1479	nan	0.1000	0.0543
##	9	1.1130	nan	0.1000	0.0538
##	10	1.0788	nan	0.1000	0.0423
##	20	0.8606	nan	0.1000	0.0238
##	40	0.6314	nan	0.1000	0.0086
##	60	0.5014	nan	0.1000	0.0077
##	80	0.4046	nan	0.1000	0.0065
##	100	0.3330	nan	0.1000	0.0030
##	120	0.2753	nan	0.1000	0.0027
##	140	0.2339	nan	0.1000	0.0027
##	150	0.2152	nan	0.1000	0.0023

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.2419
##	2	1.4552	nan	0.1000	0.1593
##	3	1.3512	nan	0.1000	0.1198
##	4	1.2759	nan	0.1000	0.1048
##	5	1.2092	nan	0.1000	0.1033
##	6	1.1461	nan	0.1000	0.0712
##	7	1.1000	nan	0.1000	0.0856
##	8	1.0466	nan	0.1000	0.0644
##	9	1.0067	nan	0.1000	0.0597
##	10	0.9693	nan	0.1000	0.0620
##	20	0.6970	nan	0.1000	0.0310
##	40	0.4497	nan	0.1000	0.0120
##	60	0.3317	nan	0.1000	0.0054
##	80	0.2507	nan	0.1000	0.0034
##	100	0.1990	nan	0.1000	0.0032
##	120	0.1570	nan	0.1000	0.0018



```

##      140      0.1273      nan      0.1000      0.0009
##      150      0.1160      nan      0.1000      0.0021
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.6094      nan      0.1000      0.1293
##      2      1.5254      nan      0.1000      0.0860
##      3      1.4686      nan      0.1000      0.0659
##      4      1.4250      nan      0.1000      0.0499
##      5      1.3916      nan      0.1000      0.0482
##      6      1.3592      nan      0.1000      0.0398
##      7      1.3333      nan      0.1000      0.0404
##      8      1.3073      nan      0.1000      0.0407
##      9      1.2799      nan      0.1000      0.0347
##     10      1.2580      nan      0.1000      0.0289
##     20      1.0938      nan      0.1000      0.0172
##     40      0.9124      nan      0.1000      0.0090
##     60      0.7999      nan      0.1000      0.0058
##     80      0.7176      nan      0.1000      0.0050
##    100      0.6543      nan      0.1000      0.0044
##    120      0.5977      nan      0.1000      0.0039
##    140      0.5495      nan      0.1000      0.0026
##    150      0.5284      nan      0.1000      0.0022
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.6094      nan      0.1000      0.1952
##      2      1.4856      nan      0.1000      0.1343
##      3      1.4008      nan      0.1000      0.0984
##      4      1.3357      nan      0.1000      0.0917
##      5      1.2780      nan      0.1000      0.0727
##      6      1.2312      nan      0.1000      0.0726
##      7      1.1851      nan      0.1000      0.0645
##      8      1.1451      nan      0.1000      0.0557
##      9      1.1105      nan      0.1000      0.0560
##     10      1.0752      nan      0.1000      0.0406
##     20      0.8491      nan      0.1000      0.0303
##     40      0.6115      nan      0.1000      0.0116
##     60      0.4834      nan      0.1000      0.0086
##     80      0.3948      nan      0.1000      0.0055
##    100      0.3257      nan      0.1000      0.0061
##    120      0.2732      nan      0.1000      0.0032
##    140      0.2285      nan      0.1000      0.0011
##    150      0.2114      nan      0.1000      0.0027
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.6094      nan      0.1000      0.2357
##      2      1.4571      nan      0.1000      0.1608
##      3      1.3563      nan      0.1000      0.1257
##      4      1.2759      nan      0.1000      0.1108
##      5      1.2073      nan      0.1000      0.1044
##      6      1.1415      nan      0.1000      0.0778
##      7      1.0932      nan      0.1000      0.0608
##      8      1.0535      nan      0.1000      0.0676
##      9      1.0097      nan      0.1000      0.0709
##     10      0.9675      nan      0.1000      0.0593
##     20      0.6963      nan      0.1000      0.0257
##     40      0.4514      nan      0.1000      0.0168
##     60      0.3267      nan      0.1000      0.0068
##     80      0.2460      nan      0.1000      0.0036

```

```

##      100      0.1950      nan      0.1000      0.0057
##      120      0.1544      nan      0.1000      0.0027
##      140      0.1233      nan      0.1000      0.0019
##      150      0.1120      nan      0.1000      0.0021
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.6094      nan      0.1000      0.1290
##      2      1.5238      nan      0.1000      0.0854
##      3      1.4674      nan      0.1000      0.0667
##      4      1.4235      nan      0.1000      0.0527
##      5      1.3895      nan      0.1000      0.0483
##      6      1.3574      nan      0.1000      0.0408
##      7      1.3308      nan      0.1000      0.0393
##      8      1.3059      nan      0.1000      0.0393
##      9      1.2795      nan      0.1000      0.0328
##     10      1.2579      nan      0.1000      0.0312
##     20      1.0970      nan      0.1000      0.0176
##     40      0.9112      nan      0.1000      0.0114
##     60      0.8014      nan      0.1000      0.0055
##     80      0.7210      nan      0.1000      0.0055
##    100      0.6556      nan      0.1000      0.0035
##    120      0.5998      nan      0.1000      0.0030
##    140      0.5533      nan      0.1000      0.0027
##    150      0.5327      nan      0.1000      0.0026
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.6094      nan      0.1000      0.1860
##      2      1.4863      nan      0.1000      0.1261
##      3      1.4043      nan      0.1000      0.1095
##      4      1.3337      nan      0.1000      0.0853
##      5      1.2788      nan      0.1000      0.0829
##      6      1.2270      nan      0.1000      0.0696
##      7      1.1827      nan      0.1000      0.0604
##      8      1.1450      nan      0.1000      0.0504
##      9      1.1133      nan      0.1000      0.0409
##     10      1.0863      nan      0.1000      0.0489
##     20      0.8549      nan      0.1000      0.0285
##     40      0.6276      nan      0.1000      0.0121
##     60      0.4923      nan      0.1000      0.0084
##     80      0.4004      nan      0.1000      0.0032
##    100      0.3308      nan      0.1000      0.0065
##    120      0.2782      nan      0.1000      0.0032
##    140      0.2387      nan      0.1000      0.0040
##    150      0.2200      nan      0.1000      0.0018
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.6094      nan      0.1000      0.2436
##      2      1.4554      nan      0.1000      0.1572
##      3      1.3540      nan      0.1000      0.1322
##      4      1.2702      nan      0.1000      0.1064
##      5      1.2018      nan      0.1000      0.0859
##      6      1.1461      nan      0.1000      0.0845
##      7      1.0924      nan      0.1000      0.0775
##      8      1.0449      nan      0.1000      0.0709
##      9      1.0014      nan      0.1000      0.0625
##     10      0.9625      nan      0.1000      0.0497
##     20      0.6913      nan      0.1000      0.0272
##     40      0.4472      nan      0.1000      0.0129

```

##	60	0.3269	nan	0.1000	0.0087
##	80	0.2500	nan	0.1000	0.0062
##	100	0.1925	nan	0.1000	0.0023
##	120	0.1552	nan	0.1000	0.0025
##	140	0.1252	nan	0.1000	0.0023
##	150	0.1117	nan	0.1000	0.0014
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1246
##	2	1.5254	nan	0.1000	0.0899
##	3	1.4678	nan	0.1000	0.0667
##	4	1.4238	nan	0.1000	0.0510
##	5	1.3888	nan	0.1000	0.0515
##	6	1.3557	nan	0.1000	0.0426
##	7	1.3287	nan	0.1000	0.0353
##	8	1.3052	nan	0.1000	0.0375
##	9	1.2791	nan	0.1000	0.0332
##	10	1.2577	nan	0.1000	0.0311
##	20	1.0947	nan	0.1000	0.0171
##	40	0.9152	nan	0.1000	0.0097
##	60	0.8000	nan	0.1000	0.0072
##	80	0.7190	nan	0.1000	0.0032
##	100	0.6536	nan	0.1000	0.0040
##	120	0.5986	nan	0.1000	0.0038
##	140	0.5524	nan	0.1000	0.0022
##	150	0.5304	nan	0.1000	0.0017
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1817
##	2	1.4895	nan	0.1000	0.1326
##	3	1.4047	nan	0.1000	0.1023
##	4	1.3385	nan	0.1000	0.0897
##	5	1.2816	nan	0.1000	0.0728
##	6	1.2350	nan	0.1000	0.0696
##	7	1.1921	nan	0.1000	0.0657
##	8	1.1501	nan	0.1000	0.0541
##	9	1.1154	nan	0.1000	0.0538
##	10	1.0821	nan	0.1000	0.0438
##	20	0.8598	nan	0.1000	0.0207
##	40	0.6355	nan	0.1000	0.0193
##	60	0.4920	nan	0.1000	0.0084
##	80	0.3987	nan	0.1000	0.0039
##	100	0.3279	nan	0.1000	0.0054
##	120	0.2745	nan	0.1000	0.0019
##	140	0.2304	nan	0.1000	0.0022
##	150	0.2109	nan	0.1000	0.0024
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.2348
##	2	1.4602	nan	0.1000	0.1654
##	3	1.3558	nan	0.1000	0.1262
##	4	1.2768	nan	0.1000	0.1060
##	5	1.2086	nan	0.1000	0.0900
##	6	1.1530	nan	0.1000	0.0857
##	7	1.0980	nan	0.1000	0.0758
##	8	1.0501	nan	0.1000	0.0751
##	9	1.0030	nan	0.1000	0.0625
##	10	0.9639	nan	0.1000	0.0492

##	20	0.7105	nan	0.1000	0.0239
##	40	0.4575	nan	0.1000	0.0087
##	60	0.3347	nan	0.1000	0.0071
##	80	0.2538	nan	0.1000	0.0046
##	100	0.1978	nan	0.1000	0.0049
##	120	0.1572	nan	0.1000	0.0034
##	140	0.1265	nan	0.1000	0.0022
##	150	0.1140	nan	0.1000	0.0013

##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1274
##	2	1.5236	nan	0.1000	0.0840
##	3	1.4671	nan	0.1000	0.0678
##	4	1.4231	nan	0.1000	0.0539
##	5	1.3875	nan	0.1000	0.0434
##	6	1.3588	nan	0.1000	0.0430
##	7	1.3300	nan	0.1000	0.0413
##	8	1.3035	nan	0.1000	0.0366
##	9	1.2795	nan	0.1000	0.0328
##	10	1.2566	nan	0.1000	0.0316
##	20	1.0940	nan	0.1000	0.0188
##	40	0.9152	nan	0.1000	0.0105
##	60	0.8029	nan	0.1000	0.0049
##	80	0.7181	nan	0.1000	0.0032
##	100	0.6559	nan	0.1000	0.0039
##	120	0.6005	nan	0.1000	0.0030
##	140	0.5519	nan	0.1000	0.0025
##	150	0.5307	nan	0.1000	0.0022

##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1926
##	2	1.4848	nan	0.1000	0.1298
##	3	1.4009	nan	0.1000	0.1037
##	4	1.3350	nan	0.1000	0.0805
##	5	1.2813	nan	0.1000	0.0781
##	6	1.2323	nan	0.1000	0.0699
##	7	1.1891	nan	0.1000	0.0645
##	8	1.1482	nan	0.1000	0.0568
##	9	1.1120	nan	0.1000	0.0490
##	10	1.0808	nan	0.1000	0.0429
##	20	0.8568	nan	0.1000	0.0282
##	40	0.6166	nan	0.1000	0.0153
##	60	0.4879	nan	0.1000	0.0120
##	80	0.3923	nan	0.1000	0.0058
##	100	0.3254	nan	0.1000	0.0054
##	120	0.2741	nan	0.1000	0.0027
##	140	0.2330	nan	0.1000	0.0037
##	150	0.2170	nan	0.1000	0.0029

##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.2411
##	2	1.4583	nan	0.1000	0.1595
##	3	1.3554	nan	0.1000	0.1321
##	4	1.2732	nan	0.1000	0.1031
##	5	1.2076	nan	0.1000	0.0926
##	6	1.1495	nan	0.1000	0.0834
##	7	1.0967	nan	0.1000	0.0861
##	8	1.0421	nan	0.1000	0.0671

```
##      9      1.0003      nan      0.1000      0.0640
##     10      0.9605      nan      0.1000      0.0659
##     20      0.7023      nan      0.1000      0.0295
##     40      0.4488      nan      0.1000      0.0122
##     60      0.3310      nan      0.1000      0.0052
##     80      0.2494      nan      0.1000      0.0060
##    100      0.1934      nan      0.1000      0.0028
##    120      0.1538      nan      0.1000      0.0023
##    140      0.1249      nan      0.1000      0.0022
##    150      0.1140      nan      0.1000      0.0021
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1         1.6094             nan      0.1000      0.2361
##      2         1.4598             nan      0.1000      0.1640
##      3         1.3578             nan      0.1000      0.1260
##      4         1.2775             nan      0.1000      0.1015
##      5         1.2133             nan      0.1000      0.0919
##      6         1.1559             nan      0.1000      0.0869
##      7         1.1016             nan      0.1000      0.0746
##      8         1.0542             nan      0.1000      0.0793
##      9         1.0056             nan      0.1000      0.0508
##     10         0.9739             nan      0.1000      0.0512
##     20         0.6973             nan      0.1000      0.0306
##     40         0.4472             nan      0.1000      0.0146
##     60         0.3280             nan      0.1000      0.0046
##     80         0.2522             nan      0.1000      0.0046
##    100         0.1995             nan      0.1000      0.0029
##    120         0.1589             nan      0.1000      0.0020
##    140         0.1255             nan      0.1000      0.0021
##    150         0.1142             nan      0.1000      0.0008
```

```
# because it has huge output, i have commented it
# modelFitGenBoost

# Predicting on the test set
modelPredictGenBoost <- predict(modelFitGenBoost, newdata=testingDataset)

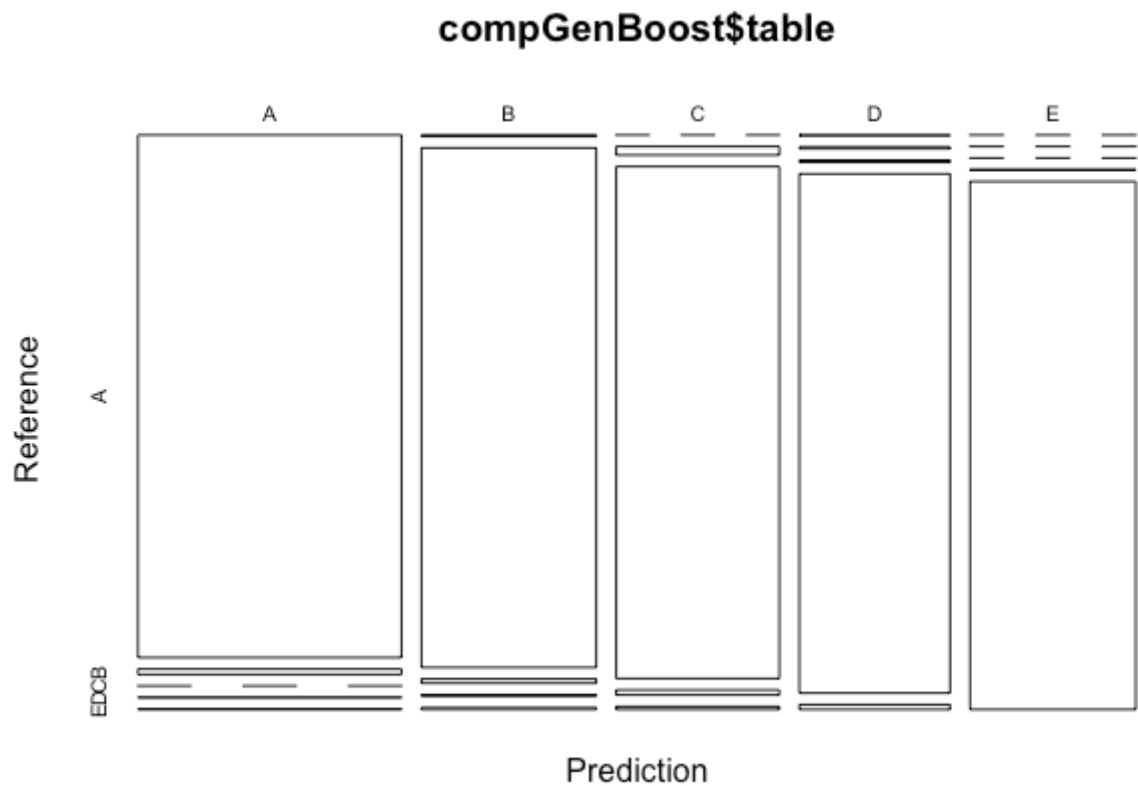
compGenBoost <- confusionMatrix(modelPredictGenBoost, testingDataset$classe)
compGenBoost
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    A    B    C    D    E
##           A 1670    17    0    1    1
##           B    3 1102    9    3    4
##           C    0    17 1013   10    6
##           D    1    3    4  949    9
##           E    0    0    0    1 1062
##
## Overall Statistics
##
##           Accuracy : 0.9849
##           95% CI : (0.9814, 0.9878)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9809
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9976  0.9675  0.9873  0.9844  0.9815
## Specificity      0.9955  0.9960  0.9932  0.9965  0.9998
## Pos Pred Value   0.9888  0.9831  0.9685  0.9824  0.9991
## Neg Pred Value   0.9990  0.9922  0.9973  0.9970  0.9959
## Prevalence       0.2845  0.1935  0.1743  0.1638  0.1839
## Detection Rate   0.2838  0.1873  0.1721  0.1613  0.1805
## Detection Prevalence 0.2870  0.1905  0.1777  0.1641  0.1806
## Balanced Accuracy 0.9965  0.9818  0.9903  0.9905  0.9907
```

*The following are the results obtained by Generalized boosted model*

- Generalized boosted model Accuracy: 0.9848768
- The out-of-sample error: 0.0151232

```
# graphical overview
plot(compGenBoost$table, col = compGenBoost$byClass)
```



## Model Selection/Conclusion:

So based on the accuracy, the results of random forest model are more accurate than the other models. so applying the results to the validation dataset.

```
# Predicting on the validation Dataset
finalPrediction <- predict(modelFitRandomForest, newdata=validationDataset)
finalPrediction
```

```
## [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
```