

Practical Machine Learning Final Report: Exercise Prediction

I have downloaded the file, reading it then creating partition.

Reading and creating partition in the File

```
library(caret)
```

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

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

```
library(randomForest)
```

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

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

```
##  
## Attaching package: 'randomForest'
```

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

```
library(rpart)  
train_in <- read.csv("pml-training.csv")  
validation <- read.csv('pml-testing.csv', header=T)  
set.seed(127)  
training_sample <- createDataPartition(y=train_in$classe, p=0.7, list=FALSE)  
training <- train_in[training_sample, ]  
testing <- train_in[-training_sample, ]  
dim(training)
```

```
## [1] 13737 160
```

```
dim(testing)
```

```
## [1] 5885 160
```

Removing near 0 variables and then removing variable

```
NZV <- nearZeroVar(training)  
training <- training[, -NZV]  
testing <- testing[, -NZV]  
dim(training)
```

```
## [1] 13737 109
```

```
dim(testing)
```

```
## [1] 5885 109
```

```
#Now removing variables that are mostly NA  
AllNA <- sapply(training, function(x) mean(is.na(x))) > 0.95  
training <- training[, AllNA==FALSE]  
testing <- testing[, AllNA==FALSE]  
dim(training)
```

```
## [1] 13737    59
```

```
dim(testing)
```

```
## [1] 5885    59
```

Model building

The three model types I'm going to test are:

1.Random forest decision trees (rf) 2.Decision trees with CART (rpart) 3.Method: Generalized Boosted Model

1. Random forest decision trees (rf)

```
controlRF <- trainControl(method="cv", number=3, verboseIter=FALSE)
modFitRandForest <- train(classe ~ ., data=training, method="rf",trControl=controlRF)
predictRandForest <- predict(modFitRandForest, newdata=testing)
confMatRandForest <- confusionMatrix(predictRandForest, testing$classe)
confMatRandForest
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    A    B    C    D    E
##           A 1674    0    0    0    0
##           B    0 1139    0    0    0
##           C    0    0 1026    0    0
##           D    0    0    0  964    0
##           E    0    0    0    0 1082
##
## Overall Statistics
##
##           Accuracy : 1
##           95% CI : (0.9994, 1)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 1
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity          1.0000   1.0000   1.0000   1.0000   1.0000
## Specificity          1.0000   1.0000   1.0000   1.0000   1.0000
## Pos Pred Value       1.0000   1.0000   1.0000   1.0000   1.0000
## Neg Pred Value       1.0000   1.0000   1.0000   1.0000   1.0000
## Prevalence           0.2845   0.1935   0.1743   0.1638   0.1839
## Detection Rate       0.2845   0.1935   0.1743   0.1638   0.1839
## Detection Prevalence 0.2845   0.1935   0.1743   0.1638   0.1839
## Balanced Accuracy     1.0000   1.0000   1.0000   1.0000   1.0000
```

2. Decision trees

```
modFitDecTree <- rpart(classe ~ ., data=training, method="class")
predictDecTree <- predict(modFitDecTree, newdata=testing, type="class")
confMatDecTree <- confusionMatrix(predictDecTree, testing$classe)
confMatDecTree
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    A    B    C    D    E
##           A 1673    0    0    0    0
##           B    1 1138    0    0    0
##           C    0    1 1025    0    0
##           D    0    0    1   964    0
##           E    0    0    0    0 1082
##
## Overall Statistics
##
##           Accuracy : 0.9995
##           95% CI : (0.9985, 0.9999)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9994
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9994  0.9991  0.9990  1.0000  1.0000
## Specificity      1.0000  0.9998  0.9998  0.9998  1.0000
## Pos Pred Value   1.0000  0.9991  0.9990  0.9990  1.0000
## Neg Pred Value   0.9998  0.9998  0.9998  1.0000  1.0000
## Prevalence       0.2845  0.1935  0.1743  0.1638  0.1839
## Detection Rate   0.2843  0.1934  0.1742  0.1638  0.1839
## Detection Prevalence 0.2843  0.1935  0.1743  0.1640  0.1839
## Balanced Accuracy 0.9997  0.9995  0.9994  0.9999  1.0000
```

3. Method: Generalized Boosted Model

```
controlGBM <- trainControl(method = "repeatedcv", number = 5, repeats = 1)
modFitGBM  <- train(classe ~ ., data=training, method = "gbm",trControl = controlGBM, verbose = FALSE)
```

```
## Loading required package: gbm
```

```
## Loading required package: survival
```

```
##
## Attaching package: 'survival'
```

```
## The following object is masked from 'package:caret':
##
##      cluster
```

```
## Loading required package: splines
```

```
## Loading required package: parallel
```

```
## Loaded gbm 2.1.1
```

```
## Loading required package: plyr
```

```
modFitGBM$finalModel
```

```
## A gradient boosted model with multinomial loss function.
## 50 iterations were performed.
## There were 80 predictors of which 1 had non-zero influence.
```

```
predictGBM <- predict(modFitGBM, newdata=testing)
confMatGBM <- confusionMatrix(predictGBM, testing$classe)
confMatGBM
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    A     B     C     D     E
##           A 1673     0     0     0     0
##           B     1 1138     0     0     0
##           C     0     1 1025     0     0
##           D     0     0     1   964     0
##           E     0     0     0     0 1082
##
## Overall Statistics
##
##           Accuracy : 0.9995
##           95% CI   : (0.9985, 0.9999)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9994
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9994  0.9991  0.9990  1.0000  1.0000
## Specificity      1.0000  0.9998  0.9998  0.9998  1.0000
## Pos Pred Value   1.0000  0.9991  0.9990  0.9990  1.0000
## Neg Pred Value   0.9998  0.9998  0.9998  1.0000  1.0000
## Prevalence       0.2845  0.1935  0.1743  0.1638  0.1839
## Detection Rate   0.2843  0.1934  0.1742  0.1638  0.1839
## Detection Prevalence 0.2843  0.1935  0.1743  0.1640  0.1839
## Balanced Accuracy 0.9997  0.9995  0.9994  0.9999  1.0000
```

Model Assessment (Out of sample error)

The accuracy of the 3 regression modeling methods above are:

Random Forest : 1 Decision Tree : 0.9995 GBM : 0.9995 In that case, the Random Forest model will be applied to predict the 20 quiz results (testing dataset) as shown below.

```
predictTEST <- predict(modFitRandForest, newdata=testing)
predictTEST
```

```
##      [1] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##     [35] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##     [69] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [103] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [137] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [171] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [205] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [239] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [273] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [307] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [341] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [375] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [409] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [443] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [477] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [511] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [545] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [579] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [613] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [647] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [681] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [715] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [749] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [783] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [817] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [851] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [885] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [919] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [953] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##    [987] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
##   [1021] A A A A A A A A A A A A A A A A A A A A A A A A A A A A A
```

[illegible]

[illegible]