

Weight_Exercise_prediction

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Executive Summary

Based on a dataset provide by HAR <http://groupware.les.inf.puc-rio.br/har> (<http://groupware.les.inf.puc-rio.br/har>) we will try to train a predictive model to predict what exercise was performed using a dataset with 159 features

We'll take the following steps:

- Process the data, for use of this project
- Explore the data, especially focussing on the two paramaters we are interested in
- Model selection, where we try different models to help us answer our questions
- Model examination, to see wether our best model holds up to our standards
- A Conclusion where we answer the questions based on the data
- Predicting the classification of the model on test set

Processing

```
set.seed(111)
training_data = read.csv("pml-training.csv")
testing_data = read.csv("pml-testing.csv")
```

Exploratory data analyses

```
#Remove columns with more than 20% missing values
maxNAallowed = ceiling(nrow(training_data)/100 * 20)
removeColumns = which(colSums(is.na(training_data)| training_data=="")>maxNAallowed)
training_data_clean = training_data[, -c(1:7,removeColumns)]
testing_data_clean = testing_data[, -c(1:7,removeColumns)]

#remove time related columns
remove_time = grep("timestamp",names(training_data_clean))
training_without_time = training_data_clean[, -c(1,remove_time)]
testing_without_time = testing_data_clean[, -c(1,remove_time)]

#final data
train_data = training_without_time
testing_data = testing_without_time
```

Model selection

```
#split train data into test and train
partition <- createDataPartition(y=train_data$classe, p=0.8, list=FALSE)
train_sub_Train <- train_data[partition, ]
train_sub_Test <- train_data[-partition, ]

#Decision Tree
system.time(
  modelDT <- rpart(classe ~ ., method = "class", data = train_sub_Train)
)
```

```
##      user  system elapsed
##    5.34    0.00    5.48
```

```
predictDT <- predict(modelDT, train_sub_Test, type = "class")
cM <- confusionMatrix(predictDT, train_sub_Test$classe)
cM
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction  A   B   C   D   E
##           A 990 117  10  23  29
##           B  30 388  66  46 107
##           C  22  99 479  80 118
##           D  58 127 109 459  95
##           E  16  28  20  35 372
##
## Overall Statistics
##
##           Accuracy : 0.6852
##           95% CI : (0.6704, 0.6997)
## No Information Rate : 0.2845
## P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.6019
##
## Mcnemar's Test P-Value : < 2.2e-16
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.8871  0.5112  0.7003  0.7138  0.51595
## Specificity      0.9362  0.9213  0.9015  0.8814  0.96908
## Pos Pred Value   0.8469  0.6091  0.6003  0.5413  0.78981
## Neg Pred Value   0.9542  0.8871  0.9344  0.9402  0.89890
## Prevalence       0.2845  0.1935  0.1744  0.1639  0.18379
## Detection Rate   0.2524  0.0989  0.1221  0.1170  0.09483
## Detection Prevalence 0.2980  0.1624  0.2034  0.2162  0.12006
## Balanced Accuracy 0.9117  0.7163  0.8009  0.7976  0.74252
```

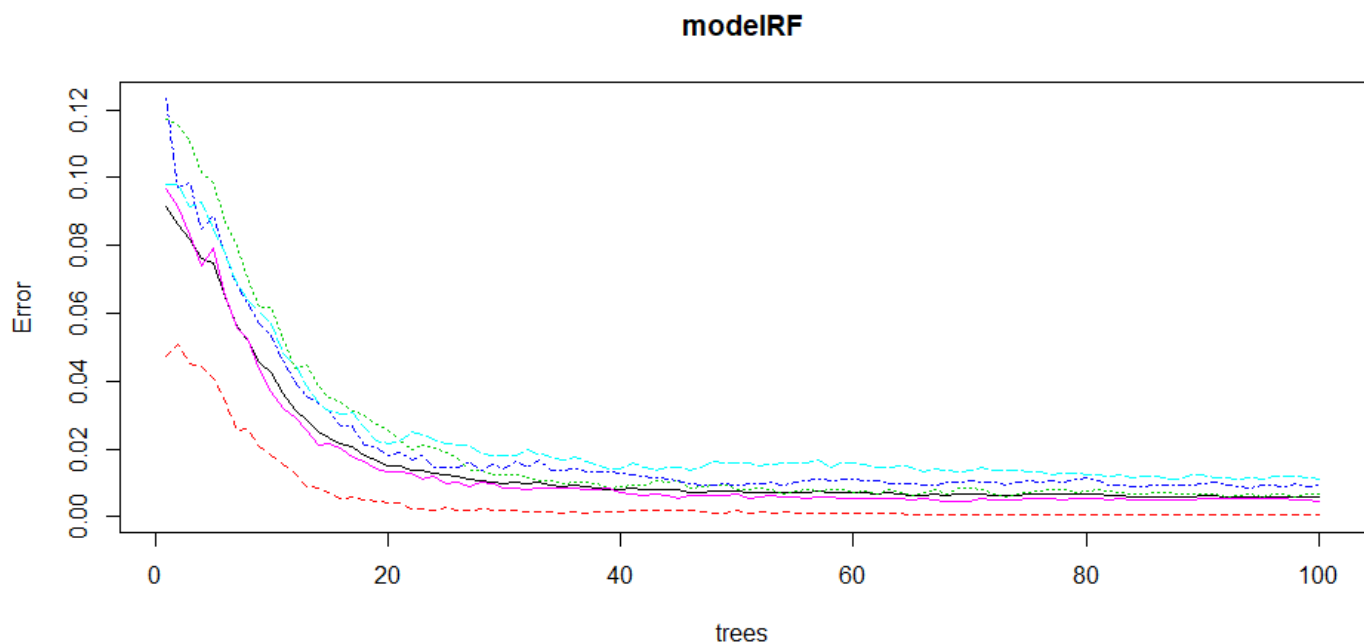
```
round(cM$overall["Accuracy"][[1]], 4) * 100
```

```
## [1] 68.52
```

```
#Random Forest
system.time(
  modelRF <- randomForest(classe ~ ., data = train_sub_Train, ntree = 100)
)
```

```
##      user  system elapsed
##    22.05    0.15    22.68
```

```
plot(modelRF)
```



```
predictRF <- predict(modelRF, train_sub_Test, type = "class")
cM <- confusionMatrix(predictRF, train_sub_Test$classe)
cM
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction   A    B    C    D    E
##           A 1116    5    0    0    0
##           B    0  754    3    0    0
##           C    0    0  681   10    1
##           D    0    0    0  633    1
##           E    0    0    0    0  719
##
## Overall Statistics
##
##           Accuracy : 0.9949
##           95% CI : (0.9921, 0.9969)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9936
##
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity           1.0000   0.9934   0.9956   0.9844   0.9972
## Specificity           0.9982   0.9991   0.9966   0.9997   1.0000
## Pos Pred Value        0.9955   0.9960   0.9841   0.9984   1.0000
## Neg Pred Value        1.0000   0.9984   0.9991   0.9970   0.9994
## Prevalence            0.2845   0.1935   0.1744   0.1639   0.1838
## Detection Rate        0.2845   0.1922   0.1736   0.1614   0.1833
## Detection Prevalence  0.2858   0.1930   0.1764   0.1616   0.1833
## Balanced Accuracy      0.9991   0.9962   0.9961   0.9921   0.9986
```

```
round(cM$overall["Accuracy"][[1]], 4) * 100
```

```
## [1] 99.49
```

```
#GBM
system.time(
  modelGBM <- train(classe ~ ., method = "gbm", data = train_sub_Train,
                    trControl = trainControl(method = "repeatedcv", number = 5, repeats = 1))
)
```

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1155
##	2	1.5371	nan	0.1000	0.0767
##	3	1.4874	nan	0.1000	0.0608
##	4	1.4493	nan	0.1000	0.0518
##	5	1.4168	nan	0.1000	0.0389
##	6	1.3923	nan	0.1000	0.0425
##	7	1.3655	nan	0.1000	0.0369
##	8	1.3410	nan	0.1000	0.0341
##	9	1.3182	nan	0.1000	0.0326
##	10	1.2979	nan	0.1000	0.0263
##	20	1.1508	nan	0.1000	0.0169
##	40	0.9747	nan	0.1000	0.0105
##	60	0.8637	nan	0.1000	0.0065
##	80	0.7798	nan	0.1000	0.0038
##	100	0.7133	nan	0.1000	0.0037
##	120	0.6593	nan	0.1000	0.0036
##	140	0.6135	nan	0.1000	0.0023
##	150	0.5937	nan	0.1000	0.0024

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1720
##	2	1.5046	nan	0.1000	0.1241
##	3	1.4266	nan	0.1000	0.0953
##	4	1.3678	nan	0.1000	0.0732
##	5	1.3206	nan	0.1000	0.0675
##	6	1.2771	nan	0.1000	0.0730
##	7	1.2315	nan	0.1000	0.0582
##	8	1.1950	nan	0.1000	0.0523
##	9	1.1626	nan	0.1000	0.0463
##	10	1.1341	nan	0.1000	0.0381
##	20	0.9434	nan	0.1000	0.0195
##	40	0.7059	nan	0.1000	0.0155
##	60	0.5651	nan	0.1000	0.0078
##	80	0.4783	nan	0.1000	0.0055
##	100	0.4144	nan	0.1000	0.0047
##	120	0.3568	nan	0.1000	0.0030
##	140	0.3157	nan	0.1000	0.0019
##	150	0.2981	nan	0.1000	0.0018

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.2166
##	2	1.4724	nan	0.1000	0.1527
##	3	1.3790	nan	0.1000	0.1165
##	4	1.3063	nan	0.1000	0.0937
##	5	1.2467	nan	0.1000	0.0891
##	6	1.1915	nan	0.1000	0.0718
##	7	1.1461	nan	0.1000	0.0777
##	8	1.0994	nan	0.1000	0.0581
##	9	1.0633	nan	0.1000	0.0568
##	10	1.0277	nan	0.1000	0.0519
##	20	0.8029	nan	0.1000	0.0242
##	40	0.5531	nan	0.1000	0.0165

##	60	0.4151	nan	0.1000	0.0075
##	80	0.3306	nan	0.1000	0.0051
##	100	0.2742	nan	0.1000	0.0038
##	120	0.2327	nan	0.1000	0.0024
##	140	0.2001	nan	0.1000	0.0019
##	150	0.1857	nan	0.1000	0.0011

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1163
##	2	1.5376	nan	0.1000	0.0748
##	3	1.4903	nan	0.1000	0.0609
##	4	1.4521	nan	0.1000	0.0467
##	5	1.4230	nan	0.1000	0.0477
##	6	1.3933	nan	0.1000	0.0380
##	7	1.3685	nan	0.1000	0.0392
##	8	1.3447	nan	0.1000	0.0297
##	9	1.3247	nan	0.1000	0.0298
##	10	1.3058	nan	0.1000	0.0296
##	20	1.1567	nan	0.1000	0.0161
##	40	0.9821	nan	0.1000	0.0092
##	60	0.8668	nan	0.1000	0.0057
##	80	0.7810	nan	0.1000	0.0046
##	100	0.7148	nan	0.1000	0.0035
##	120	0.6614	nan	0.1000	0.0025
##	140	0.6160	nan	0.1000	0.0017
##	150	0.5959	nan	0.1000	0.0024

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1679
##	2	1.5038	nan	0.1000	0.1129
##	3	1.4337	nan	0.1000	0.0980
##	4	1.3734	nan	0.1000	0.0830
##	5	1.3218	nan	0.1000	0.0650
##	6	1.2812	nan	0.1000	0.0654
##	7	1.2409	nan	0.1000	0.0592
##	8	1.2042	nan	0.1000	0.0523
##	9	1.1716	nan	0.1000	0.0468
##	10	1.1420	nan	0.1000	0.0408
##	20	0.9354	nan	0.1000	0.0194
##	40	0.7083	nan	0.1000	0.0086
##	60	0.5733	nan	0.1000	0.0076
##	80	0.4847	nan	0.1000	0.0043
##	100	0.4151	nan	0.1000	0.0041
##	120	0.3605	nan	0.1000	0.0033
##	140	0.3171	nan	0.1000	0.0038
##	150	0.2985	nan	0.1000	0.0016

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.2175
##	2	1.4747	nan	0.1000	0.1441
##	3	1.3853	nan	0.1000	0.1203
##	4	1.3104	nan	0.1000	0.1025
##	5	1.2463	nan	0.1000	0.0818
##	6	1.1957	nan	0.1000	0.0740

##	7	1.1504	nan	0.1000	0.0698
##	8	1.1072	nan	0.1000	0.0686
##	9	1.0658	nan	0.1000	0.0435
##	10	1.0374	nan	0.1000	0.0538
##	20	0.7928	nan	0.1000	0.0317
##	40	0.5521	nan	0.1000	0.0114
##	60	0.4127	nan	0.1000	0.0063
##	80	0.3288	nan	0.1000	0.0034
##	100	0.2723	nan	0.1000	0.0033
##	120	0.2277	nan	0.1000	0.0024
##	140	0.1929	nan	0.1000	0.0016
##	150	0.1790	nan	0.1000	0.0020

##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1176
##	2	1.5363	nan	0.1000	0.0791
##	3	1.4869	nan	0.1000	0.0627
##	4	1.4477	nan	0.1000	0.0501
##	5	1.4163	nan	0.1000	0.0447
##	6	1.3875	nan	0.1000	0.0403
##	7	1.3613	nan	0.1000	0.0371
##	8	1.3377	nan	0.1000	0.0320
##	9	1.3164	nan	0.1000	0.0333
##	10	1.2947	nan	0.1000	0.0332
##	20	1.1464	nan	0.1000	0.0151
##	40	0.9719	nan	0.1000	0.0101
##	60	0.8583	nan	0.1000	0.0072
##	80	0.7745	nan	0.1000	0.0043
##	100	0.7080	nan	0.1000	0.0037
##	120	0.6533	nan	0.1000	0.0027
##	140	0.6074	nan	0.1000	0.0025
##	150	0.5877	nan	0.1000	0.0023

##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1754
##	2	1.5015	nan	0.1000	0.1142
##	3	1.4282	nan	0.1000	0.0995
##	4	1.3663	nan	0.1000	0.0803
##	5	1.3166	nan	0.1000	0.0635
##	6	1.2765	nan	0.1000	0.0727
##	7	1.2314	nan	0.1000	0.0621
##	8	1.1924	nan	0.1000	0.0503
##	9	1.1606	nan	0.1000	0.0469
##	10	1.1306	nan	0.1000	0.0392
##	20	0.9317	nan	0.1000	0.0258
##	40	0.7113	nan	0.1000	0.0123
##	60	0.5690	nan	0.1000	0.0073
##	80	0.4789	nan	0.1000	0.0039
##	100	0.4103	nan	0.1000	0.0048
##	120	0.3581	nan	0.1000	0.0023
##	140	0.3159	nan	0.1000	0.0017
##	150	0.2957	nan	0.1000	0.0025

##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve

##	1	1.6094	nan	0.1000	0.2201
##	2	1.4721	nan	0.1000	0.1502
##	3	1.3797	nan	0.1000	0.1283
##	4	1.3006	nan	0.1000	0.1016
##	5	1.2372	nan	0.1000	0.0840
##	6	1.1851	nan	0.1000	0.0752
##	7	1.1399	nan	0.1000	0.0751
##	8	1.0941	nan	0.1000	0.0600
##	9	1.0573	nan	0.1000	0.0492
##	10	1.0269	nan	0.1000	0.0459
##	20	0.7941	nan	0.1000	0.0228
##	40	0.5473	nan	0.1000	0.0153
##	60	0.4179	nan	0.1000	0.0066
##	80	0.3338	nan	0.1000	0.0062
##	100	0.2772	nan	0.1000	0.0026
##	120	0.2338	nan	0.1000	0.0019
##	140	0.2002	nan	0.1000	0.0024
##	150	0.1854	nan	0.1000	0.0006

##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1141
##	2	1.5368	nan	0.1000	0.0767
##	3	1.4869	nan	0.1000	0.0587
##	4	1.4493	nan	0.1000	0.0475
##	5	1.4189	nan	0.1000	0.0436
##	6	1.3908	nan	0.1000	0.0418
##	7	1.3644	nan	0.1000	0.0411
##	8	1.3392	nan	0.1000	0.0321
##	9	1.3188	nan	0.1000	0.0287
##	10	1.2994	nan	0.1000	0.0307
##	20	1.1530	nan	0.1000	0.0171
##	40	0.9770	nan	0.1000	0.0108
##	60	0.8650	nan	0.1000	0.0077
##	80	0.7759	nan	0.1000	0.0034
##	100	0.7138	nan	0.1000	0.0038
##	120	0.6589	nan	0.1000	0.0033
##	140	0.6152	nan	0.1000	0.0044
##	150	0.5946	nan	0.1000	0.0020

##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1728
##	2	1.5032	nan	0.1000	0.1185
##	3	1.4280	nan	0.1000	0.0991
##	4	1.3658	nan	0.1000	0.0798
##	5	1.3170	nan	0.1000	0.0656
##	6	1.2760	nan	0.1000	0.0595
##	7	1.2380	nan	0.1000	0.0573
##	8	1.2029	nan	0.1000	0.0496
##	9	1.1720	nan	0.1000	0.0555
##	10	1.1384	nan	0.1000	0.0431
##	20	0.9315	nan	0.1000	0.0219
##	40	0.7029	nan	0.1000	0.0110
##	60	0.5763	nan	0.1000	0.0080
##	80	0.4824	nan	0.1000	0.0057

##	100	0.4096	nan	0.1000	0.0028
##	120	0.3581	nan	0.1000	0.0029
##	140	0.3177	nan	0.1000	0.0028
##	150	0.2988	nan	0.1000	0.0017
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.2167
##	2	1.4744	nan	0.1000	0.1489
##	3	1.3806	nan	0.1000	0.1199
##	4	1.3060	nan	0.1000	0.1002
##	5	1.2448	nan	0.1000	0.0824
##	6	1.1927	nan	0.1000	0.0802
##	7	1.1437	nan	0.1000	0.0643
##	8	1.1030	nan	0.1000	0.0546
##	9	1.0678	nan	0.1000	0.0489
##	10	1.0365	nan	0.1000	0.0496
##	20	0.7997	nan	0.1000	0.0211
##	40	0.5520	nan	0.1000	0.0101
##	60	0.4207	nan	0.1000	0.0052
##	80	0.3351	nan	0.1000	0.0044
##	100	0.2748	nan	0.1000	0.0032
##	120	0.2325	nan	0.1000	0.0033
##	140	0.1994	nan	0.1000	0.0015
##	150	0.1859	nan	0.1000	0.0020
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1146
##	2	1.5364	nan	0.1000	0.0790
##	3	1.4867	nan	0.1000	0.0628
##	4	1.4472	nan	0.1000	0.0496
##	5	1.4158	nan	0.1000	0.0432
##	6	1.3895	nan	0.1000	0.0418
##	7	1.3631	nan	0.1000	0.0394
##	8	1.3385	nan	0.1000	0.0310
##	9	1.3189	nan	0.1000	0.0314
##	10	1.2987	nan	0.1000	0.0304
##	20	1.1504	nan	0.1000	0.0205
##	40	0.9755	nan	0.1000	0.0092
##	60	0.8646	nan	0.1000	0.0067
##	80	0.7784	nan	0.1000	0.0058
##	100	0.7127	nan	0.1000	0.0040
##	120	0.6579	nan	0.1000	0.0032
##	140	0.6134	nan	0.1000	0.0033
##	150	0.5935	nan	0.1000	0.0030
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.6094	nan	0.1000	0.1694
##	2	1.5031	nan	0.1000	0.1197
##	3	1.4272	nan	0.1000	0.0958
##	4	1.3665	nan	0.1000	0.0822
##	5	1.3157	nan	0.1000	0.0692
##	6	1.2732	nan	0.1000	0.0595
##	7	1.2361	nan	0.1000	0.0488
##	8	1.2052	nan	0.1000	0.0453

```
##      9      1.1755      nan    0.1000    0.0475
##     10      1.1448      nan    0.1000    0.0428
##     20      0.9415      nan    0.1000    0.0223
##     40      0.7049      nan    0.1000    0.0081
##     60      0.5687      nan    0.1000    0.0067
##     80      0.4783      nan    0.1000    0.0052
##    100      0.4100      nan    0.1000    0.0034
##    120      0.3584      nan    0.1000    0.0037
##    140      0.3149      nan    0.1000    0.0015
##    150      0.2964      nan    0.1000    0.0020
```

```
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.6094      nan    0.1000    0.2147
##      2      1.4747      nan    0.1000    0.1493
##      3      1.3818      nan    0.1000    0.1304
##      4      1.3031      nan    0.1000    0.0950
##      5      1.2426      nan    0.1000    0.0897
##      6      1.1873      nan    0.1000    0.0813
##      7      1.1387      nan    0.1000    0.0611
##      8      1.1002      nan    0.1000    0.0702
##      9      1.0583      nan    0.1000    0.0528
##     10      1.0250      nan    0.1000    0.0531
##     20      0.7882      nan    0.1000    0.0315
##     40      0.5489      nan    0.1000    0.0097
##     60      0.4177      nan    0.1000    0.0083
##     80      0.3346      nan    0.1000    0.0043
##    100      0.2757      nan    0.1000    0.0030
##    120      0.2306      nan    0.1000    0.0031
##    140      0.1975      nan    0.1000    0.0016
##    150      0.1828      nan    0.1000    0.0014
```

```
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.6094      nan    0.1000    0.2132
##      2      1.4775      nan    0.1000    0.1544
##      3      1.3828      nan    0.1000    0.1090
##      4      1.3149      nan    0.1000    0.1104
##      5      1.2484      nan    0.1000    0.0759
##      6      1.2001      nan    0.1000    0.0807
##      7      1.1510      nan    0.1000    0.0789
##      8      1.1045      nan    0.1000    0.0582
##      9      1.0677      nan    0.1000    0.0495
##     10      1.0366      nan    0.1000    0.0479
##     20      0.7962      nan    0.1000    0.0264
##     40      0.5474      nan    0.1000    0.0101
##     60      0.4202      nan    0.1000    0.0070
##     80      0.3363      nan    0.1000    0.0056
##    100      0.2783      nan    0.1000    0.0032
##    120      0.2343      nan    0.1000    0.0025
##    140      0.2012      nan    0.1000    0.0009
##    150      0.1878      nan    0.1000    0.0021
```

```
##   user  system elapsed
## 730.75    2.37  746.42
```

```

predictGBM <- predict(modelGBM, train_sub_Test)
cM <- confusionMatrix(predictGBM, train_sub_Test$classe)
cM

```

``` ## Confusion Matrix and Statistics ```

```

##
##           Reference
## Prediction    A    B    C    D    E
##           A 1099   28    0    2    2
##           B   14  701   22    5   10
##           C    1   30  653   27    8
##           D    1    0    8  601    6
##           E    1    0    1    8  695
##

```

``` ## Overall Statistics ```

```

##
##           Accuracy : 0.9556
##           95% CI : (0.9487, 0.9619)
##    No Information Rate : 0.2845
##    P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9439
##
##    McNemar's Test P-Value : 2.972e-05
##

```

``` ## Statistics by Class: ```

```

##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.9848  0.9236  0.9547  0.9347  0.9639
## Specificity      0.9886  0.9839  0.9796  0.9954  0.9969
## Pos Pred Value   0.9717  0.9322  0.9082  0.9756  0.9858
## Neg Pred Value   0.9939  0.9817  0.9903  0.9873  0.9919
## Prevalence       0.2845  0.1935  0.1744  0.1639  0.1838
## Detection Rate   0.2801  0.1787  0.1665  0.1532  0.1772
## Detection Prevalence 0.2883  0.1917  0.1833  0.1570  0.1797
## Balanced Accuracy 0.9867  0.9537  0.9672  0.9651  0.9804

```

```

round(cM$overall["Accuracy"][[1]], 4) * 100

```

```

## [1] 95.56

```

``` #LDA ```

```

system.time(
  modelLDA <- train(classe ~ ., method = "lda", data = train_sub_Train,
    trControl = trainControl(method = "repeatedcv", number = 5, repeats = 1))
)

```

```

##    user  system elapsed
##   5.02    0.19    5.23

```

```

predictLDA <- predict(modelLDA, train_sub_Test)
cM <- confusionMatrix(predictLDA, train_sub_Test$classe)
cM

```

```

## Confusion Matrix and Statistics
##
##           Reference
## Prediction  A   B   C   D   E
##           A 908 112  72  32  25
##           B  28 491  51  31 133
##           C 101  92 461  87  74
##           D  77  23  84 447  91
##           E   2  41  16  46 398
##
## Overall Statistics
##
##           Accuracy : 0.6895
##           95% CI : (0.6748, 0.704)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.6073
##
##           Mcnemar's Test P-Value : < 2.2e-16
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity           0.8136  0.6469  0.6740  0.6952  0.5520
## Specificity           0.9141  0.9232  0.8907  0.9162  0.9672
## Pos Pred Value        0.7903  0.6689  0.5656  0.6191  0.7913
## Neg Pred Value        0.9250  0.9160  0.9282  0.9388  0.9056
## Prevalence            0.2845  0.1935  0.1744  0.1639  0.1838
## Detection Rate        0.2315  0.1252  0.1175  0.1139  0.1015
## Detection Prevalence  0.2929  0.1871  0.2077  0.1840  0.1282
## Balanced Accuracy      0.8639  0.7851  0.7823  0.8057  0.7596

```

```

round(cM$overall["Accuracy"][[1]], 4) * 100

```

```

## [1] 68.95

```

```

#Neural Networks
system.time(
  modelDL <- nnet(classe ~ ., train_sub_Train, size = 5, rang = .1, decay = 5e-4, maxit = 100,
    trControl = trainControl(method = "repeatedcv", number = 5, repeats = 1))
)

```

```
## # weights: 290
## initial value 25344.736575
## iter 10 value 23765.751311
## iter 20 value 23495.771752
## iter 30 value 22861.470413
## iter 40 value 22336.941867
## iter 50 value 22025.974366
## iter 60 value 21671.166621
## iter 70 value 21225.465381
## iter 80 value 21139.356477
## iter 90 value 20913.854464
## iter 100 value 20649.707306
## final value 20649.707306
## stopped after 100 iterations
```

```
## user system elapsed
## 22.28 0.00 22.59
```

```
predictDL <- predict(modelDL, train_sub_Test, type = "class")
cM <- confusionMatrix(as.factor(predictDL), train_sub_Test$classe)
cM
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction  A   B   C   D   E
##           A 710  66  76  23  32
##           B  23 116  30  11  79
##           C 206 216 514 216 244
##           D 177 330  56 393 318
##           E   0  31   8   0  48
##
## Overall Statistics
##
##           Accuracy : 0.454
##           95% CI : (0.4383, 0.4697)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.3193
##
##           Mcnemar's Test P-Value : < 2.2e-16
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class: E
## Sensitivity      0.6362 0.15283 0.7515 0.6112 0.06657
## Specificity      0.9298 0.95480 0.7277 0.7314 0.98782
## Pos Pred Value   0.7828 0.44788 0.3682 0.3085 0.55172
## Neg Pred Value   0.8654 0.82451 0.9327 0.9056 0.82456
## Prevalence       0.2845 0.19347 0.1744 0.1639 0.18379
## Detection Rate   0.1810 0.02957 0.1310 0.1002 0.01224
## Detection Prevalence 0.2312 0.06602 0.3559 0.3248 0.02218
## Balanced Accuracy 0.7830 0.55382 0.7396 0.6713 0.52720
```

```
round(cM$overall["Accuracy"][[1]], 4) * 100
```

```
## [1] 45.4
```

#using Random Forest because of best accuracy

```
predict(modelRF, testing_data, type = "class")
```

```
##  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
```