

Practical Machine Learning CourseProject

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Step1. Download & assign data

The 1st step is to load the data into R dataset. And assign the training & testing data into variables `final_training`, `final_testing` respectively.

Then, for the `final_training` dataset, we separate it `pre_training` & `pre_testing` (70% & 30%) for modeling.

```
final_training <- read.csv("D:\\Coursera\\Material\\08. Practical Machine Learning\\CourseProject\\pml-
final_testing <- read.csv("D:\\Coursera\\Material\\08. Practical Machine Learning\\CourseProject\\pml-t

library(lattice);library(ggplot2);library(caret);
set.seed(33833)
inTrain <- createDataPartition(y=final_training$classe,p=0.7,list=FALSE)
pre_training <- final_training[inTrain,]
pre_testing <- final_testing[-inTrain,]
```

Step2. Data preprocess & Variable selection

Let's do the briefly data explorer. As you can see, there are too many NA or Null variables in the dataset. We remove it from our modeling.

```
head(pre_training)
```

```
##      X user_name raw_timestamp_part_1 raw_timestamp_part_2  cvtd_timestamp
## 1 1  carlitos      1323084231      788290 05/12/2011 11:23
## 3 3  carlitos      1323084231      820366 05/12/2011 11:23
## 4 4  carlitos      1323084232      120339 05/12/2011 11:23
## 5 5  carlitos      1323084232      196328 05/12/2011 11:23
## 7 7  carlitos      1323084232      368296 05/12/2011 11:23
## 8 8  carlitos      1323084232      440390 05/12/2011 11:23
##      new_window num_window roll_belt pitch_belt yaw_belt total_accel_belt
## 1          no         11      1.41      8.07    -94.4              3
## 3          no         11      1.42      8.07    -94.4              3
## 4          no         12      1.48      8.05    -94.4              3
## 5          no         12      1.48      8.07    -94.4              3
## 7          no         12      1.42      8.09    -94.4              3
## 8          no         12      1.42      8.13    -94.4              3
##      kurtosis_roll_belt kurtosis_picth_belt kurtosis_yaw_belt
## 1
## 3
## 4
## 5
## 7
## 8
##      skewness_roll_belt skewness_roll_belt.1 skewness_yaw_belt max_roll_belt
## 1                                                    NA
```

## 3					NA	
## 4					NA	
## 5					NA	
## 7					NA	
## 8					NA	
##	max_picth_belt	max_yaw_belt	min_roll_belt	min_pitch_belt	min_yaw_belt	
## 1	NA		NA	NA		
## 3	NA		NA	NA		
## 4	NA		NA	NA		
## 5	NA		NA	NA		
## 7	NA		NA	NA		
## 8	NA		NA	NA		
##	amplitude_roll_belt	amplitude_pitch_belt	amplitude_yaw_belt			
## 1	NA		NA			
## 3	NA		NA			
## 4	NA		NA			
## 5	NA		NA			
## 7	NA		NA			
## 8	NA		NA			
##	var_total_accel_belt	avg_roll_belt	stddev_roll_belt	var_roll_belt		
## 1	NA	NA		NA	NA	
## 3	NA	NA		NA	NA	
## 4	NA	NA		NA	NA	
## 5	NA	NA		NA	NA	
## 7	NA	NA		NA	NA	
## 8	NA	NA		NA	NA	
##	avg_pitch_belt	stddev_pitch_belt	var_pitch_belt	avg_yaw_belt		
## 1	NA	NA	NA	NA		
## 3	NA	NA	NA	NA		
## 4	NA	NA	NA	NA		
## 5	NA	NA	NA	NA		
## 7	NA	NA	NA	NA		
## 8	NA	NA	NA	NA		
##	stddev_yaw_belt	var_yaw_belt	gyros_belt_x	gyros_belt_y	gyros_belt_z	
## 1	NA	NA	0.00	0.00	-0.02	
## 3	NA	NA	0.00	0.00	-0.02	
## 4	NA	NA	0.02	0.00	-0.03	
## 5	NA	NA	0.02	0.02	-0.02	
## 7	NA	NA	0.02	0.00	-0.02	
## 8	NA	NA	0.02	0.00	-0.02	
##	accel_belt_x	accel_belt_y	accel_belt_z	magnet_belt_x	magnet_belt_y	
## 1	-21	4	22	-3	599	
## 3	-20	5	23	-2	600	
## 4	-22	3	21	-6	604	
## 5	-21	2	24	-6	600	
## 7	-22	3	21	-4	599	
## 8	-22	4	21	-2	603	
##	magnet_belt_z	roll_arm	pitch_arm	yaw_arm	total_accel_arm	var_accel_arm
## 1	-313	-128	22.5	-161	34	NA
## 3	-305	-128	22.5	-161	34	NA
## 4	-310	-128	22.1	-161	34	NA
## 5	-302	-128	22.1	-161	34	NA
## 7	-311	-128	21.9	-161	34	NA
## 8	-313	-128	21.8	-161	34	NA

##	avg_roll_arm	stddev_roll_arm	var_roll_arm	avg_pitch_arm	stddev_pitch_arm	
## 1	NA	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	NA	
## 7	NA	NA	NA	NA	NA	
## 8	NA	NA	NA	NA	NA	
##	var_pitch_arm	avg_yaw_arm	stddev_yaw_arm	var_yaw_arm	gyros_arm_x	
## 1	NA	NA	NA	NA	0.00	
## 3	NA	NA	NA	NA	0.02	
## 4	NA	NA	NA	NA	0.02	
## 5	NA	NA	NA	NA	0.00	
## 7	NA	NA	NA	NA	0.00	
## 8	NA	NA	NA	NA	0.02	
##	gyros_arm_y	gyros_arm_z	accel_arm_x	accel_arm_y	accel_arm_z	magnet_arm_x
## 1	0.00	-0.02	-288	109	-123	-368
## 3	-0.02	-0.02	-289	110	-126	-368
## 4	-0.03	0.02	-289	111	-123	-372
## 5	-0.03	0.00	-289	111	-123	-374
## 7	-0.03	0.00	-289	111	-125	-373
## 8	-0.02	0.00	-289	111	-124	-372
##	magnet_arm_y	magnet_arm_z	kurtosis_roll_arm	kurtosis_pitch_arm		
## 1	337	516				
## 3	344	513				
## 4	344	512				
## 5	337	506				
## 7	336	509				
## 8	338	510				
##	kurtosis_yaw_arm	skewness_roll_arm	skewness_pitch_arm	skewness_yaw_arm		
## 1						
## 3						
## 4						
## 5						
## 7						
## 8						
##	max_roll_arm	max_pitch_arm	max_yaw_arm	min_roll_arm	min_pitch_arm	
## 1	NA	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	NA	
## 7	NA	NA	NA	NA	NA	
## 8	NA	NA	NA	NA	NA	
##	min_yaw_arm	amplitude_roll_arm	amplitude_pitch_arm	amplitude_yaw_arm		
## 1	NA	NA	NA	NA		
## 3	NA	NA	NA	NA		
## 4	NA	NA	NA	NA		
## 5	NA	NA	NA	NA		
## 7	NA	NA	NA	NA		
## 8	NA	NA	NA	NA		
##	roll_dumbbell	pitch_dumbbell	yaw_dumbbell	kurtosis_roll_dumbbell		
## 1	13.05217	-70.49400	-84.87394			
## 3	12.85075	-70.27812	-85.14078			
## 4	13.43120	-70.39379	-84.87363			
## 5	13.37872	-70.42856	-84.85306			

```

## 7      13.12695      -70.24757      -85.09961
## 8      12.75083      -70.34768      -85.09708
## kurtosis_picth_dumbbell kurtosis_yaw_dumbbell skewness_roll_dumbbell
## 1
## 3
## 4
## 5
## 7
## 8
## skewness_pitch_dumbbell skewness_yaw_dumbbell max_roll_dumbbell
## 1                                     NA
## 3                                     NA
## 4                                     NA
## 5                                     NA
## 7                                     NA
## 8                                     NA
## max_picth_dumbbell max_yaw_dumbbell min_roll_dumbbell min_pitch_dumbbell
## 1      NA                                     NA      NA
## 3      NA                                     NA      NA
## 4      NA                                     NA      NA
## 5      NA                                     NA      NA
## 7      NA                                     NA      NA
## 8      NA                                     NA      NA
## min_yaw_dumbbell amplitude_roll_dumbbell amplitude_pitch_dumbbell
## 1      NA                                     NA
## 3      NA                                     NA
## 4      NA                                     NA
## 5      NA                                     NA
## 7      NA                                     NA
## 8      NA                                     NA
## amplitude_yaw_dumbbell total_accel_dumbbell var_accel_dumbbell
## 1      37      NA
## 3      37      NA
## 4      37      NA
## 5      37      NA
## 7      37      NA
## 8      37      NA
## avg_roll_dumbbell stddev_roll_dumbbell var_roll_dumbbell
## 1      NA      NA      NA
## 3      NA      NA      NA
## 4      NA      NA      NA
## 5      NA      NA      NA
## 7      NA      NA      NA
## 8      NA      NA      NA
## avg_pitch_dumbbell stddev_pitch_dumbbell var_pitch_dumbbell
## 1      NA      NA      NA
## 3      NA      NA      NA
## 4      NA      NA      NA
## 5      NA      NA      NA
## 7      NA      NA      NA
## 8      NA      NA      NA
## avg_yaw_dumbbell stddev_yaw_dumbbell var_yaw_dumbbell gyros_dumbbell_x
## 1      NA      NA      NA      0
## 3      NA      NA      NA      0

```

## 4	NA	NA	NA	0
## 5	NA	NA	NA	0
## 7	NA	NA	NA	0
## 8	NA	NA	NA	0
##	gyros_dumbbell_y	gyros_dumbbell_z	accel_dumbbell_x	accel_dumbbell_y
## 1	-0.02	0.00	-234	47
## 3	-0.02	0.00	-232	46
## 4	-0.02	-0.02	-232	48
## 5	-0.02	0.00	-233	48
## 7	-0.02	0.00	-232	47
## 8	-0.02	0.00	-234	46
##	accel_dumbbell_z	magnet_dumbbell_x	magnet_dumbbell_y	magnet_dumbbell_z
## 1	-271	-559	293	-65
## 3	-270	-561	298	-63
## 4	-269	-552	303	-60
## 5	-270	-554	292	-68
## 7	-270	-551	295	-70
## 8	-272	-555	300	-74
##	roll_forearm	pitch_forearm	yaw_forearm	kurtosis_roll_forearm
## 1	28.4	-63.9	-153	
## 3	28.3	-63.9	-152	
## 4	28.1	-63.9	-152	
## 5	28.0	-63.9	-152	
## 7	27.9	-63.9	-152	
## 8	27.8	-63.8	-152	
##	kurtosis_pitch_forearm	kurtosis_yaw_forearm	skewness_roll_forearm	
## 1				
## 3				
## 4				
## 5				
## 7				
## 8				
##	skewness_pitch_forearm	skewness_yaw_forearm	max_roll_forearm	
## 1			NA	
## 3			NA	
## 4			NA	
## 5			NA	
## 7			NA	
## 8			NA	
##	max_pitch_forearm	max_yaw_forearm	min_roll_forearm	min_pitch_forearm
## 1	NA		NA	NA
## 3	NA		NA	NA
## 4	NA		NA	NA
## 5	NA		NA	NA
## 7	NA		NA	NA
## 8	NA		NA	NA
##	min_yaw_forearm	amplitude_roll_forearm	amplitude_pitch_forearm	
## 1		NA	NA	
## 3		NA	NA	
## 4		NA	NA	
## 5		NA	NA	
## 7		NA	NA	
## 8		NA	NA	
##	amplitude_yaw_forearm	total_accel_forearm	var_accel_forearm	

```

## 1 36 NA
## 3 36 NA
## 4 36 NA
## 5 36 NA
## 7 36 NA
## 8 36 NA
## avg_roll_forearm stddev_roll_forearm var_roll_forearm avg_pitch_forearm
## 1 NA NA NA NA
## 3 NA NA NA NA
## 4 NA NA NA NA
## 5 NA NA NA NA
## 7 NA NA NA NA
## 8 NA NA NA NA
## stddev_pitch_forearm var_pitch_forearm avg_yaw_forearm
## 1 NA NA NA
## 3 NA NA NA
## 4 NA NA NA
## 5 NA NA NA
## 7 NA NA NA
## 8 NA NA NA
## stddev_yaw_forearm var_yaw_forearm gyros_forearm_x gyros_forearm_y
## 1 NA NA 0.03 0.00
## 3 NA NA 0.03 -0.02
## 4 NA NA 0.02 -0.02
## 5 NA NA 0.02 0.00
## 7 NA NA 0.02 0.00
## 8 NA NA 0.02 -0.02
## gyros_forearm_z accel_forearm_x accel_forearm_y accel_forearm_z
## 1 -0.02 192 203 -215
## 3 0.00 196 204 -213
## 4 0.00 189 206 -214
## 5 -0.02 189 206 -214
## 7 -0.02 195 205 -215
## 8 0.00 193 205 -213
## magnet_forearm_x magnet_forearm_y magnet_forearm_z classe
## 1 -17 654 476 A
## 3 -18 658 469 A
## 4 -16 658 469 A
## 5 -17 655 473 A
## 7 -18 659 470 A
## 8 -9 660 474 A

```

```

colIdx <- c(7:11,37:49,60:68,84:86,102,113:124,140,151:159,160)
training <- final_training[inTrain,colIdx]
testing <- final_training[-inTrain,colIdx]

```

Step3. Start modeling

To start the modeling procedure. Here, I choose two model “rpart” & “lda” due to performance consideration. To compare these to model, I will evaluate the out of sample error estimation (accuracy) to choose the better one as the final model!

```
library(rpart);library(MASS);library(randomForest);library(ggplot2);
```

```
## 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
```

```
memory.limit(60000)
```

```
## [1] 60000
```

```
set.seed(33833)
```

```
rpart <- train(classe~., data=training[,-1],method="rpart")
```

```
lda <- train(classe~., data=training[,-1],method="lda")
```

```
rf <- train(classe~., data=training[,-1],method="rf", ntree = 150)
```

```
rpart
```

```
## CART
```

```
##
```

```
## 13737 samples
```

```
##      52 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
```

```
## 0.03722917 0.5017537 0.35068186
```

```
## 0.06133659 0.4229226 0.22020767
```

```
## 0.11484081 0.3323672 0.07387813
```

```
##
```

```
## Accuracy was used to select the optimal model using the largest value.
```

```
## The final value used for the model was cp = 0.03722917.
```

```
lda
```

```
## Linear Discriminant Analysis
```

```
##
```

```
## 13737 samples
```

```
##      52 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:
##
##   Accuracy   Kappa
##   0.6998909  0.6204494
##
##
```

```
rf
```

```
## Random Forest
##
## 13737 samples
##   52 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:
##
##   mtry Accuracy   Kappa
##   2    0.9880023  0.9848167
##   27    0.9890561  0.9861515
##   52    0.9779916  0.9721519
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was mtry = 27.
```

Step4. Out of Sample Error Estimation (Comparing model by it)

Select to better model by accuracy. Here, the rpart get 49% score & lda get 70% score & random forest get 99% score. Hence, I choose the random forest as my final model.

```
pred.rpart <- predict(rpart,testing)
pred.lda <- predict(lda,testing)
pred.rf <- predict(rf,testing)
sum(pred.rpart == testing$classe) / length(testing$classe)
```

```
## [1] 0.4895497
```

```
sum(pred.lda == testing$classe) / length(testing$classe)
```

```
## [1] 0.6987256
```

```
sum(pred.rf == testing$classe) / length(testing$classe)
```

```
## [1] 0.9906542
```


Step5. The final prediction results

From the out of sample error estimation, we select the model random forest with the higher accuracy. Then, applying this model to do prediction. Got the result below.

```
#pred.rpart <- predict(rpart,final_testing)  
#pred.rpart  
pred.rf <- predict(rpart,final_testing)  
pred.rf
```

```
## [1] C A C A A C C A A A C C C A C A A A A C  
## Levels: A B C D E
```