Coursera_Practical ML

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#Assignment

#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. The aim of this project is to predict the manner in which participants perform a barbell lift. The data comes from http://groupware.les.inf.puc-rio.br/har (http://groupware.les.inf.puc-rio.br/har) wherein 6 participants were asked to perform the same set of exercises correctly and incorrectly with accelerometers placed on the belt, forearm, arm, and dumbell.

#For the purpose of this project, the following steps would be followed:

#1.Import the Data #2.Removing Columns that are not relevant #3.Using Random Forest #4.Predicting Test Set Output for 20 cases

```
library(caret)
## Loading required package: lattice
## Loading required package: ggplot2
library(tidyverse)
## -- Attaching packages ------
                                           ----- tidyverse 1.3.1 --
## v tibble 3.1.1
                    v dplyr
                             1.0.5
## v tidyr
           1.1.3
                   v stringr 1.4.0
## v readr
           1.4.0
                    v forcats 0.5.1
## v purrr
           0.3.4
## -- Conflicts -----conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                  masks stats::lag()
## x purrr::lift()
                  masks caret::lift()
library(future)
##
## Attaching package: 'future'
```

```
## The following object is masked from 'package:caret':
##
##
       cluster
library(randomForest)
## randomForest 4.6-14
## Type rfNews() to see new features/changes/bug fixes.
##
## Attaching package: 'randomForest'
## The following object is masked from 'package:dplyr':
##
##
       combine
## The following object is masked from 'package:ggplot2':
##
##
       margin
## Loading the training and the test data from the given url
train<-read.csv(url("https://d396qusza40orc.cloudfront.net/predmachlearn/pml-training.csv"))</pre>
test<-read.csv(url("https://d396qusza40orc.cloudfront.net/predmachlearn/pml-testing.csv"))</pre>
# We check the size of the training and test sets
object.size(train)
## 22799176 bytes
object.size(test)
## 45056 bytes
# For the classification problem we find number of classes to predict
unique(train$classe)
## [1] "A" "B" "C" "D" "E"
```

```
train<-train[,-c(1:6)] # removing the id columns that are not relevant train
test<-test[,-c(1:6)] # removing the id columns that are not relevant test

var<-nearZeroVar(train) # find the variables with zero variance
var<-var[-26] # retaining the classe variable
t1<-train %>% select(-all_of(var))
t2<-select_if(t1,function(x) sum(is.na(x)) <=1) #dropping columns that contain nas</pre>
```

```
plan(multisession) # parallel processing to speed up t2$classe <-as.factor(t2$classe) # converting the classification var to factor model=randomForest(classe~., data=t2, method='class')
```

model # printing the results of the model

```
##
## Call:
    randomForest(formula = classe ~ ., data = t2, method = "class")
##
##
                  Type of random forest: classification
                        Number of trees: 500
##
## No. of variables tried at each split: 7
##
##
           OOB estimate of error rate: 0.14%
## Confusion matrix:
##
             В
                  C
                            E class.error
        Δ
                       D
## A 5579
             0
                       0
                            1 0.0001792115
                  1
                            0 0.0010534633
## B
        3 3793
                       0
## C
             5 3417
                       0
                            0 0.0014611338
## D
        0
             0
                 13 3202
                            1 0.0043532338
## E
                       4 3603 0.0011089548
```

```
vec_names<-names(t2)
vec_names<-vec_names[-length(vec_names)] # applying the same subset from train to test set colum
ns removing the classe columns
test_1<-test %>% select(all_of(vec_names))
```

base_predict<-predict(model,test_1) # creating a prediction based on model</pre>

base_predict # printing out the results of prediction

```
## 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
```