

Practical Machine Learning Class (Coursera) HAR Clustering Model/Analysis using Random Forest

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This is the submission RMD for predicting the actual test set (PML-testing.csv). This is based on training the model as in PML_Project.Rmd

Load the test data

```
library(caret)

## Loading required package: lattice
## Loading required package: ggplot2

setwd("~/My Training/PML")

currentTrainPath = 'train.RData'
currentTestPath = 'test.RData'

modFit<-readRDS(currentTrainPath)
tst = read.csv('pml-testing.csv',header=T)
```

Data Cleanup and Feature Addition

```
data<-tst

data<-data[which(data$new_window=="no"),]
```

Delete all columns containing NA

```
rmcol<-array()
j<-1
for (i in 1:ncol(data)) {
  if ((sum(is.na(data[,i])) > 0) | (sum(data[,i] == "") > 0)) {
    rmcol[j]<-i
    j<-j+1
  }
}

data<-data[,-rmcol]
rm(rmcol)

tm<-unlist(lapply(strsplit(as.character(data$cvtd_timestamp)," "),
function(x) x[2])))
time<-(lapply(strsplit(tm,":"),function(x) {return
(round((as.numeric(x[1])) + round(as.numeric(x[2])/60))))}))
```

```

data$cvtd_timestamp =
factor(NA,levels=c('midnight','morning','noon','afternoon','evening'))

data$cvtd_timestamp[time>22 | time<=6] <- 'midnight'
data$cvtd_timestamp[time>6 & time<=10] <- 'morning'
data$cvtd_timestamp[time>10 & time <= 13] <- 'noon'
data$cvtd_timestamp[time>13 & time <= 18] <- 'afternoon'
data$cvtd_timestamp[time>18 & time <= 22] <- 'evening'

rm(tm,time)

```

Predict the actual test set

```

library(caret)

nam<-names(data)
valCols<-
c(which(nam=="accel_dumbbell_x"),which(nam=="magnet_dumbbell_x"),which
(nam=="magnet_dumbbell_y"),which(nam=="magnet_dumbbell_z"),which(nam=="
roll_forearm"),which(nam=="pitch_forearm"),which(nam=="yaw_forearm"),
which(nam=="total_accel_forearm"),which(nam=="gyros_forearm_x"),which(
nam=="gyros_forearm_y"),which(nam=="gyros_forearm_z"),which(nam=="acce
l_forearm_x"),which(nam=="accel_forearm_y"),which(nam=="accel_forearm_
z"),which(nam=="magnet_forearm_x"),which(nam=="magnet_forearm_y"),whic
h(nam=="magnet_forearm_z"),which(nam=="classe"))

data_b<-data
prediction<-predict(modFit,data[,valCols])

## Loading required package: randomForest
## randomForest 4.6-7
## Type rfNews() to see new features/changes/bug fixes.

data$classe<-prediction
nam<-names(data)

finCols<-
c(which(nam=="accel_dumbbell_x"),which(nam=="magnet_dumbbell_x"),which
(nam=="magnet_dumbbell_y"),which(nam=="magnet_dumbbell_z"),which(nam=="
roll_forearm"),which(nam=="pitch_forearm"),which(nam=="yaw_forearm"),
which(nam=="total_accel_forearm"),which(nam=="gyros_forearm_x"),which(
nam=="gyros_forearm_y"),which(nam=="gyros_forearm_z"),which(nam=="acce
l_forearm_x"),which(nam=="accel_forearm_y"),which(nam=="accel_forearm_
z"),which(nam=="magnet_forearm_x"),which(nam=="magnet_forearm_y"),whic
h(nam=="magnet_forearm_z"),which(nam=="classe"))

```

```
data[,finCols]
```

```
##      accel_dumbbell_x magnet_dumbbell_x magnet_dumbbell_y  
magnet_dumbbell_z
```

```
## 1          21          523          -528  
-56  
## 2         -153         -502          388  
-36  
## 3         -141         -506          349  
41  
## 4          -51         -576          238  
53  
## 5          -18         -424          252  
312  
## 6         -138         -543          262  
96  
## 7         -145         -484          354  
97  
## 8         -140         -515          350  
53  
## 9           0         -519          348  
-32  
## 10          -7         -531          321  
-164  
## 11          -4         -556          280  
-23  
## 12         -149         -539          293  
67  
## 13          27         -328          271  
363  
## 14         -139         -523          337  
46  
## 15         -159         -511          323  
85  
## 16          43         -284          311  
368  
## 17          14          484         -558  
-19  
## 18         -20         -527          329  
-77  
## 19          22          505         -537  
-86  
## 20         185          272          403  
340
```

```
##      roll_forearm pitch_forearm yaw_forearm total_accel_forearm  
## 1          141.0          49.30          156.0           33  
## 2          109.0         -17.60          106.0           39  
## 3          131.0         -32.60           93.0           34  
## 4           0.0           0.00           0.0           43
```

## 5	-176.0	-2.16	-47.9	24
## 6	150.0	1.46	89.7	43
## 7	155.0	34.50	152.0	32
## 8	-161.0	43.60	-89.5	47
## 9	15.5	-63.50	-139.0	36
## 10	13.2	19.40	-105.0	24
## 11	137.0	-9.41	104.0	46
## 12	138.0	-19.90	70.5	36
## 13	176.0	16.70	-38.5	23
## 14	53.1	-46.20	159.0	33
## 15	152.0	26.50	-168.0	24
## 16	-176.0	16.20	-42.2	25
## 17	79.4	-2.60	109.0	30
## 18	-173.0	49.80	-133.0	25
## 19	-164.0	59.30	-149.0	23
## 20	173.0	19.20	-83.2	21
##	gyros_forearm_x	gyros_forearm_y	gyros_forearm_z	accel_forearm_x
## 1	0.74	-3.34	-0.59	-110
## 2	1.12	-2.78	-0.18	212
## 3	0.18	-0.79	0.28	154
## 4	1.38	0.69	1.80	-92
## 5	-0.75	3.10	0.80	131
## 6	-0.88	4.26	1.35	230
## 7	-0.53	1.80	0.75	-192
## 8	0.63	-0.74	0.49	-151
## 9	0.03	0.02	-0.02	195
## 10	0.02	0.13	-0.07	-212
## 11	0.05	0.05	0.11	-3
## 12	-0.02	0.83	0.38	182
## 13	0.00	-4.69	-1.26	149
## 14	0.02	0.00	0.03	232
## 15	0.64	-5.97	-1.05	-18
## 16	-0.31	-3.45	-0.66	161
## 17	0.10	-0.06	0.18	144
## 18	-0.75	4.14	0.95	-129
## 19	-1.06	2.79	0.67	-148
## 20	-1.01	3.18	1.26	41
##	accel_forearm_y	accel_forearm_z	magnet_forearm_x	
## 1	267	-149	-714	
419				
## 2	297	-118	-237	
791				
## 3	271	-129	-51	
698				
## 4	406	-39	-233	
783				
## 5	-93	172	375	-
787				
## 6	322	-144	-300	

800				
## 7	170	-175	-678	
284				
## 8	-331	-282	-109	-
619				
## 9	204	-217	0	
652				
## 10	98	-7	-403	
723				
## 11	405	-203	-248	
720				
## 12	263	-148	32	
648				
## 13	46	167	456	-
677				
## 14	106	-198	123	
555				
## 15	43	-226	-540	-
155				
## 16	-61	171	532	-
775				
## 17	201	-154	-146	
756				
## 18	-29	-202	-500	-
232				
## 19	21	-172	-614	-
46				
## 20	-100	179	70	-
703				
##	magnet_forearm_z	classe		
## 1	617	B		
## 2	873	A		
## 3	783	B		
## 4	521	A		
## 5	91	A		
## 6	884	E		
## 7	585	D		
## 8	-32	B		
## 9	469	A		
## 10	512	A		
## 11	438	B		
## 12	702	C		
## 13	29	B		
## 14	648	A		
## 15	389	E		
## 16	114	E		
## 17	708	A		
## 18	329	B		
## 19	471	B		
## 20	74	B		