

## Session 18 Segmentation Analysis

### Weight Lifting Exercise

#### 5. Problem Statement

1. Use the below given data set  
DataSet
2. Perform the below given activities:
  - a. Create classification model using different decision trees.
  - b. Verify model goodness of fit.
  - c. Apply all the model validation techniques.
  - d. Make conclusions

```
setwd("C:/Users/Seshan/Desktop/sv R related/acadgild/assignments/session17")
```

```
library(readr)
```

```
Weightlift <- read.csv("C:/Users/Seshan/Desktop/sv R related/acadgild/assignments/session 18  
Assign/Weight lift.csv",sep=',',header=F)
```

```
#problem: find out natual grouping
```

```
df<-Weightlift
```

```
#df=Weightlift <- read.csv("C:/Users/Seshan/Desktop/sv R related/acadgild/assignments/session 18  
Assign/Weightlift.csv",sep=',',header=F)
```

```
View(df)
```

```
#df = read.csv("https://archive.ics.uci.edu/ml/machine-learning-  
databases/00273/Example_WearableComputing_weight_lifting_exercises_biceps_curl_variations.csv",s  
ep=',',header=F)
```

#problem: find out natural grouping

#View(df)

head(df)

str(df)

set.seed(1234)

ind = sample(1:nrow(df),0.8\*nrow(df),replace = F)

df\_train = df[ind,-1]

df\_test = df[-ind,-1]

summary(df)

dim(df)

apply(df,2,range)

apply(df,2,summary)

df[] <- lapply(df, function(x) as.numeric(as.character(x)))

# KMeans - comes from Rcmdr library

# Kmeans- from amap library

# kmeans- from stats library

# steps in k-means clustering

#1- preprocessing the data (impute missing values, remove outliers, feature transformation)

#2- scaling or standardization of data set

#3- decide the number of clusters (value of K)

#4- iterate over the samples to create clusters

#5- decide the distance measure

#6- calculate the group accuracy

# scaling of data

df\_train1 <- scale(df\_train)

```
head(df_train1)
```

```
class(df_train1)
```

```
# screeplot approach to decide the number of clusters
```

```
km = kmeans(df_train1,1)
```

```
km$withinss
```

```
km$tot.withinss
```

```
km = kmeans(df_train1,2)
```

```
km$withinss
```

```
km$tot.withinss
```

```
km = kmeans(df_train1,3)
```

```
km$withinss
```

```
km$tot.withinss
```

```
km = kmeans(df_train1,4)
```

```
km$withinss
```

```
km$tot.withinss
```

```
km = kmeans(df_train1,5)
```

```
km$withinss
```

```
km$tot.withinss
```

```
km = kmeans(df_train1,6)
```

```
km$withinss
```

```
km$tot.withinss
```

```
km = kmeans(df_train1,7)
```

```
km$withinss
```

```
km$tot.withinss
```

```
km = kmeans(df_train1,8)
```

```
km$withinss
```

```
km$tot.withinss
```

```
km = kmeans(df_train1,9)
```

```
km$withinss
```

```
km$tot.withinss
```

```
km = kmeans(df_train1,10)
```

```
km$withinss
```

```
km$tot.withinss
```

```
dev.off()
```

```
sumsq=NULL
```

```
for (i in 1:10)
```

```
  sumsq[i] = sum(kmeans(df_train,centers=i,
```

```
    iter.max = 1000,
```

```
    nstart=i,
```

```
    algorithm='Forgy')$withinss)
```

```
plot(1:10,sumsq,type='b', main='Screeplot showing within group sum of squares')
```

```
km = kmeans(df_train1,3)
```

```
km$withinss
```

```
km$tot.withinss
```

```
class(km$cluster)
```

```
summary(km)
```

```
km$centers
```

```
as.numeric(km$cluster)
```

```
length(km$cluster)
```

```
dim(df_train)
```

```
class(df_train)
```

```
df_train$cl <- km$cluster
```

```
head(df_train)
```

```
# profiles of clusters
```

```
aggregate(df_train[,1:8],list(df_train[,9]),mean)
```

```
table(df$V1)
```

```
library(cluster)
```

```
clusplot(df_train,df_train$cl,cex=0.9,color=T,shade=T, labels=4,lines=0)
```

```
#HC clustering or Hierarchical Clustering
```

```
# distance (euclidean, manhattan, cosine distance)
```

```
# Divisive method (top down)
```

```
# Agglomerative method (bottom up)
```

```
df_train = df_train[,-9]
```

```
head(df_train)
```

```
# compute the distance matrix
```

```
d1 <- dist(df_train,method='euclidean')
```

```
summary(d1)
```

```
# HC
```

```
fit <- hclust(d1,method = 'ward.D2')
```

```
plot(fit)
```

```
# single, double, average, ward, ward.D2
```

```
# agglomerative method
```

```
fit <- agnes(d1,metric='euclidean',method = 'ward')
```

```
plot(fit)
```

```
# divisive method
```

```
fit <- diana(d1,metric='euclidean')
```

```
plot(fit)
```

```

library(ggdendro)

if(require(cluster)){

  fit<- agnes(d1, metric = "manhattan", stand = TRUE)

  dg <- as.dendrogram(fit)

  ggdendrogram(dg)

  fit <- diana(d1, metric = "manhattan", stand = TRUE)

  dg <- as.dendrogram(fit)

  ggdendrogram(dg)

}

```

```

> setwd("C:/Users/Seshan/Desktop/sv R related/acadgild/assignments/session17")
> library(readr)
>
> weightlift <- read.csv("C:/Users/Seshan/Desktop/sv R related/acadgild/assignments/session 18 Assign/Weight lift.csv",sep=',',header=F)
> #problem: find out natual grouping
> df<-weightlift
> view(df)
> head(df)

```

	V1	V2	V3	V4
1	user_name	raw_timestamp_part_1	raw_timestamp_part_2	cvtd_timestamp
2	eurico	1322489729	34670	28/11/2011 14:15
3	eurico	1322489729	62641	28/11/2011 14:15
4	eurico	1322489729	70653	28/11/2011 14:15
5	eurico	1322489729	82654	28/11/2011 14:15
6	eurico	1322489729	90637	28/11/2011 14:15

	V5	V6	V7	V8	V9	V10
1	new_window	num_window	roll_belt	pitch_belt	yaw_belt	total_accel_belt
2	no	1	3.7	41.6	-82.8	3
3	no	1	3.66	42.8	-82.5	2
4	no	1	3.58	43.7	-82.3	1
5	no	1	3.56	44.4	-82.1	1
6	no	1	3.57	45.1	-81.9	1

	V11	V12	V13	V14
1	kurtosis_roll_belt	kurtosis_picth_belt	skewness_roll_belt	skewness_roll_belt
2	-1.03566	-0.39133	0.005406	0.045115
3	-1.03566	-0.39133	0.005406	0.045115
4	-1.03566	-0.39133	0.005406	0.045115

5	-1.03566		-0.39133		0.005406		0.045115
6	-1.03566		-0.39133		0.005406		0.045115
	v15	v16	v17	v18	v19		
1	max_roll_belt	max_pitch_belt	max_yaw_belt	min_roll_belt	min_pitch_belt		
2	-4.1	20	-1	-7.25	18		
3	-4.1	20	-1	-7.25	18		
4	-4.1	20	-1	-7.25	18		
5	-4.1	20	-1	-7.25	18		
6	-4.1	20	-1	-7.25	18		
	v20	v21	v22	v23			
1	min_yaw_belt	amplitude_roll_belt	amplitude_pitch_belt	amplitude_yaw_belt			
2	-1	1.345	2	0			
3	-1	1.345	2	0			
4	-1	1.345	2	0			
5	-1	1.345	2	0			
6	-1	1.345	2	0			
	v24	v25	v26	v27			
1	var_total_accel_belt	avg_roll_belt	stddev_roll_belt	var_roll_belt			
2	0.3	121.9	0.6	0.35			
3	0.3	121.9	0.6	0.35			
4	0.3	121.9	0.6	0.35			
5	0.3	121.9	0.6	0.35			
6	0.3	121.9	0.6	0.35			
	v28	v29	v30	v31	v32		
1	avg_pitch_belt	stddev_pitch_belt	var_pitch_belt	avg_yaw_belt	stddev_yaw_belt		
2	25.75	0.35	0.1	-4.95	0.4		
3	25.75	0.35	0.1	-4.95	0.4		
4	25.75	0.35	0.1	-4.95	0.4		
5	25.75	0.35	0.1	-4.95	0.4		
6	25.75	0.35	0.1	-4.95	0.4		
	v33	v34	v35	v36	v37	v3	
8							
1	var_yaw_belt	gyros_belt_x	gyros_belt_y	gyros_belt_z	accel_belt_x	accel_belt_y	
2	0.17	2.02	0.18	0.02	-3	-1	
8							
3	0.17	1.96	0.14	0.05	-2	-1	
3							
4	0.17	1.88	0.08	0.05	-2	-	
6							
5	0.17	1.8	0.03	0.08	-6	-	
5							
6	0.17	1.77	0	0.13	-4	-	
9							
	v39	v40	v41	v42	v43	v44	
1	accel_belt_z	magnet_belt_x	magnet_belt_y	magnet_belt_z	roll_arm	pitch_arm	
2	22	387	525	-267	132	-43.7	
3	16	405	512	-254	129	-45.3	
4	8	409	511	-244	125	-46.8	
5	7	422	513	-221	120	-48.1	
6	0	418	508	-208	115	-49.1	
	v45	v46	v47	v48	v49		
1	yaw_arm	total_accel_arm	var_accel_arm	avg_roll_arm	stddev_roll_arm		
2	-53.6	38	65.0977	76.22175	16.1039		
3	-49	38	65.0977	76.22175	16.1039		
4	-43.7	35	65.0977	76.22175	16.1039		
5	-38.1	35	65.0977	76.22175	16.1039		
6	-31.7	34	65.0977	76.22175	16.1039		
	v50	v51	v52	v53	v54		
1	var_roll_arm	avg_pitch_arm	stddev_pitch_arm	var_pitch_arm	avg_yaw_arm		
2	259.3599	-10.1695	10.66725	113.7978	19.0615		
3	259.3599	-10.1695	10.66725	113.7978	19.0615		
4	259.3599	-10.1695	10.66725	113.7978	19.0615		
5	259.3599	-10.1695	10.66725	113.7978	19.0615		



6	259.3599	-10.1695	10.66725	113.7978	19.0615	
	V55	V56	V57	V58	V59	V60
1	stddev_yaw_arm	var_yaw_arm	gyros_arm_x	gyros_arm_y	gyros_arm_z	accel_arm_x
2	35.8809	1287.463	2.65	-0.61	-0.02	143
3	35.8809	1287.463	2.79	-0.64	-0.11	146
4	35.8809	1287.463	2.91	-0.69	-0.15	156
5	35.8809	1287.463	3.08	-0.72	-0.23	158
6	35.8809	1287.463	3.2	-0.77	-0.25	163
	V61	V62	V63	V64	V65	
1	accel_arm_y	accel_arm_z	magnet_arm_x	magnet_arm_y	magnet_arm_z	
2	30	-346	556	-205	-374	
3	35	-339	599	-206	-335	
4	44	-307	613	-198	-319	
5	52	-305	646	-186	-268	
6	55	-288	670	-175	-241	
	V66	V67	V68	V69		
1	kurtosis_roll_arm	kurtosis_pitch_arm	kurtosis_yaw_arm	skewness_roll_arm		
2	-1.18224	-0.96912	-0.86977	0.12353		
3	-1.18224	-0.96912	-0.86977	0.12353		
4	-1.18224	-0.96912	-0.86977	0.12353		
5	-1.18224	-0.96912	-0.86977	0.12353		
6	-1.18224	-0.96912	-0.86977	0.12353		
	V70	V71	V72	V73	V74	
1	skewness_pitch_arm	skewness_yaw_arm	max_roll_arm	max_pitch_arm	max_yaw_arm	
2	-0.10319	0.059765	8.45	77.25	38	
3	-0.10319	0.059765	8.45	77.25	38	
4	-0.10319	0.059765	8.45	77.25	38	
5	-0.10319	0.059765	8.45	77.25	38	
6	-0.10319	0.059765	8.45	77.25	38	
	V75	V76	V77	V78	V79	
9						
1	min_roll_arm	min_pitch_arm	min_yaw_arm	amplitude_roll_arm	amplitude_pitch_arm	
2	-33.6	-58.6	10	36.945	121.	
5						
3	-33.6	-58.6	10	36.945	121.	
5						
4	-33.6	-58.6	10	36.945	121.	
5						
5	-33.6	-58.6	10	36.945	121.	
5						
6	-33.6	-58.6	10	36.945	121.	
5						
	V80	V81	V82	V83		
1	amplitude_yaw_arm	roll_dumbbell	pitch_dumbbell	yaw_dumbbell		
2	27	51.23553997	11.69884724	104.2647274		
3	27	55.82441814	9.645819033	100.2280531		
4	27	55.4698307	6.875243852	101.0841063		
5	27	55.94485974	11.07929719	99.78455638		
6	27	55.21173932	11.42683324	100.4225829		
	V84	V85	V86			
1	kurtosis_roll_dumbbell	kurtosis_pitch_dumbbell	skewness_roll_dumbbell			
2	-0.09595	-0.4422	0.0819			
3	-0.09595	-0.4422	0.0819			
4	-0.09595	-0.4422	0.0819			
5	-0.09595	-0.4422	0.0819			
6	-0.09595	-0.4422	0.0819			
	V87	V88	V89	V90		
0						
1	skewness_pitch_dumbbell	max_roll_dumbbell	max_pitch_dumbbell	max_yaw_dumbbell		
1						
2	-0.216	41.85	133	-0.		
1						
3	-0.216	41.85	133	-0.		

1				
4	-0.216	41.85	133	-0.
1				
5	-0.216	41.85	133	-0.
1				
6	-0.216	41.85	133	-0.
1				
	v91	v92	v93	v9
4				
1	min_roll_dumbbell	min_pitch_dumbbell	min_yaw_dumbbell	amplitude_roll_dumbbell
1				
2	-26.75	20.2	-0.1	55.7
1				
3	-26.75	20.2	-0.1	55.7
1				
4	-26.75	20.2	-0.1	55.7
1				
5	-26.75	20.2	-0.1	55.7
1				
6	-26.75	20.2	-0.1	55.7
1				
	v95	v96	v97	
1	amplitude_pitch_dumbbell	amplitude_yaw_dumbbell	total_accel_dumbbell	
2	54.74	0	4	
3	54.74	0	4	
4	54.74	0	4	
5	54.74	0	5	
6	54.74	0	4	
	v98	v99	v100	v101
1	var_accel_dumbbell	avg_roll_dumbbell	stddev_roll_dumbbell	var_roll_dumbbell
2	2.41635	-5.11805	17.058	291.001
3	2.41635	-5.11805	17.058	291.001
4	2.41635	-5.11805	17.058	291.001
5	2.41635	-5.11805	17.058	291.001
6	2.41635	-5.11805	17.058	291.001
	v102	v103	v104	v105
1	avg_pitch_dumbbell	stddev_pitch_dumbbell	var_pitch_dumbbell	avg_yaw_dumbbell
2	13.9312	14.1062	199.0775	64.7063
3	13.9312	14.1062	199.0775	64.7063
4	13.9312	14.1062	199.0775	64.7063
5	13.9312	14.1062	199.0775	64.7063
6	13.9312	14.1062	199.0775	64.7063
	v106	v107	v108	v109
1	stddev_yaw_dumbbell	var_yaw_dumbbell	gyros_dumbbell_x	gyros_dumbbell_y
2	13.5747	184.5578	-0.31	0.16
3	13.5747	184.5578	-0.31	0.14
4	13.5747	184.5578	-0.31	0.16
5	13.5747	184.5578	-0.31	0.16
6	13.5747	184.5578	-0.31	0.14
	v110	v111	v112	v113
1	gyros_dumbbell_z	accel_dumbbell_x	accel_dumbbell_y	accel_dumbbell_z
2	0.08	5	21	37
3	0.07	4	22	35
4	0.05	3	23	37
5	0.07	5	24	38
6	0.07	5	23	37
	v114	v115	v116	v117
1	magnet_dumbbell_x	magnet_dumbbell_y	magnet_dumbbell_z	roll_forearm
2	-471	191	277	-111
3	-472	184	281	-112
4	-468	190	275	-114
5	-469	184	285	-115
6	-468	189	292	-117
	v118	v119	v120	v121

1	pitch_forearm	yaw_forearm	kurtosis_roll_forearm	kurtosis_pitch_forearm
2	26.5	138	-1.09475	-0.97525
3	26.2	138	-1.09475	-0.97525
4	26	137	-1.09475	-0.97525
5	25.8	137	-1.09475	-0.97525
6	25.5	137	-1.09475	-0.97525
	v122	v123	v124	
1	skewness_roll_forearm	skewness_pitch_forearm	max_roll_forearm	
2	-0.05065	0.17285	49.6	
3	-0.05065	0.17285	49.6	
4	-0.05065	0.17285	49.6	
5	-0.05065	0.17285	49.6	
6	-0.05065	0.17285	49.6	
	v125	v126	v127	v128
1	max_pitch_forearm	max_yaw_forearm	min_roll_forearm	min_pitch_forearm
2	168	-1.1	4.65	-168.5
3	168	-1.1	4.65	-168.5
4	168	-1.1	4.65	-168.5
5	168	-1.1	4.65	-168.5
6	168	-1.1	4.65	-168.5
	v129	v130	v131	
1	min_yaw_forearm	amplitude_roll_forearm	amplitude_pitch_forearm	
2	-1.1	32.2	341.5	
3	-1.1	32.2	341.5	
4	-1.1	32.2	341.5	
5	-1.1	32.2	341.5	
6	-1.1	32.2	341.5	
	v132	v133	v134	v135
1	amplitude_yaw_forearm	total_accel_forearm	var_accel_forearm	avg_roll_forearm
2	0	30	14.0772	27.85936
3	0	31	14.0772	27.85936
4	0	32	14.0772	27.85936
5	0	33	14.0772	27.85936
6	0	34	14.0772	27.85936
	v136	v137	v138	v139
1	stddev_roll_forearm	var_roll_forearm	avg_pitch_forearm	stddev_pitch_forearm
2	45.16342	2749.163	25.35597	8.906695
3	45.16342	2749.163	25.35597	8.906695
4	45.16342	2749.163	25.35597	8.906695
5	45.16342	2749.163	25.35597	8.906695
6	45.16342	2749.163	25.35597	8.906695
	v140	v141	v142	v143
1	var_pitch_forearm	avg_yaw_forearm	stddev_yaw_forearm	var_yaw_forearm
2	79.33451	17.09505	74.27584	5541.956
3	79.33451	17.09505	74.27584	5541.956
4	79.33451	17.09505	74.27584	5541.956
5	79.33451	17.09505	74.27584	5541.956
6	79.33451	17.09505	74.27584	5541.956
	v144	v145	v146	v147
1	gyros_forearm_x	gyros_forearm_y	gyros_forearm_z	accel_forearm_x
2	-0.05	-0.37	-0.43	-170
3	-0.06	-0.37	-0.59	-178
4	-0.05	-0.27	-0.72	-182
5	0.02	-0.24	-0.79	-185
6	0.08	-0.27	-0.82	-188
	v148	v149	v150	v151
1	accel_forearm_y	accel_forearm_z	magnet_forearm_x	magnet_forearm_y
2	155	184	-1160	1400
3	164	182	-1150	1410
4	172	185	-1130	1400
5	182	188	-1120	1400
6	195	188	-1100	1400
	v152			
1	magnet_forearm_z			

```

2          -876
3          -871
4          -863
5          -855
6          -843
>
> str(df)
'data.frame': 4025 obs. of 152 variables:
 $ V1 : Factor w/ 6 levels "adelmo","carlitos",...: 6 3 3 3 3 3 3 3 3 3 ...
 $ V2 : Factor w/ 89 levels "1322489729","1322489730",...: 89 1 1 1 1 1 1 1 1 1
1 ...
 $ V3 : Factor w/ 3789 levels "100232","100295",...: 3789 1038 2291 2620 3101
3435 310 403 597 696 ...
 $ V4 : Factor w/ 8 levels "2/12/2011 13:35",...: 8 2 2 2 2 2 2 2 2 2 ...
 $ V5 : Factor w/ 3 levels "new_window","no",...: 1 2 2 2 2 2 2 2 2 2 ...
 $ V6 : Factor w/ 92 levels "1","10","11",...: 92 1 1 1 1 1 1 1 1 1 ...
 $ V7 : Factor w/ 870 levels "-0.03","-0.04",...: 870 665 663 662 660 661 655
652 643 618 ...
 $ V8 : Factor w/ 822 levels "-27.8","-27.9",...: 822 403 404 405 406 407 409
411 415 417 ...
 $ V9 : Factor w/ 936 levels "-0.02","-0.03",...: 936 650 648 647 645 643 643
643 646 649 ...
 $ V10 : Factor w/ 28 levels "0","1","10","11",...: 28 21 13 2 2 2 2 21 22 13 .
..
 $ V11 : Factor w/ 90 levels "-0.076054","-0.120447",...: 90 30 30 30 30 30 30
30 30 30 ...
 $ V12 : Factor w/ 79 levels "-0.06016","-0.108371",...: 79 15 15 15 15 15 15 1
5 15 15 ...
 $ V13 : Factor w/ 90 levels "-0.011683","-0.037647",...: 90 45 45 45 45 45 45
45 45 45 ...
 $ V14 : Factor w/ 81 levels "-0.045472","-0.04606",...: 81 40 40 40 40 40 40 4
0 40 40 ...
 $ V15 : Factor w/ 66 levels "-0.2","-0.4",...: 66 24 24 24 24 24 24 24 24 24 .
..
 $ V16 : Factor w/ 15 levels "10","17","19",...: 15 4 4 4 4 4 4 4 4 4 ...
 $ V17 : Factor w/ 35 levels "-0.1","-0.2",...: 35 10 10 10 10 10 10 10 10 10 .
..
 $ V18 : Factor w/ 65 levels "-1","-1.5","-10.5",...: 65 27 27 27 27 27 27 27 2
7 27 ...
 $ V19 : Factor w/ 14 levels "0","1","13","15",...: 14 7 7 7 7 7 7 7 7 7 ...
 $ V20 : Factor w/ 35 levels "-0.1","-0.2",...: 35 10 10 10 10 10 10 10 10 10 .
..
 $ V21 : Factor w/ 64 levels "0","0.1","0.2",...: 64 23 23 23 23 23 23 23 23 23
...
 $ V22 : Factor w/ 11 levels "0","1","2","21",...: 11 3 3 3 3 3 3 3 3 3 ...
 $ V23 : Factor w/ 2 levels "0","amplitude_yaw_belt": 2 1 1 1 1 1 1 1 1 1 ...
 $ V24 : Factor w/ 27 levels "0","0.1","0.2",...: 27 4 4 4 4 4 4 4 4 4 ...
 $ V25 : Factor w/ 62 levels "-0.7","-1","-1.6",...: 62 26 26 26 26 26 26 26 26
26 ...
 $ V26 : Factor w/ 33 levels "0","0.1","0.2",...: 33 7 7 7 7 7 7 7 7 7 ...
 $ V27 : Factor w/ 39 levels "0","0.1","0.2",...: 39 5 5 5 5 5 5 5 5 5 ...
 $ V28 : Factor w/ 55 levels "-30.6","-31.8",...: 55 10 10 10 10 10 10 10 10 10
..
 $ V29 : Factor w/ 23 levels "0","0.1","0.2",...: 23 5 5 5 5 5 5 5 5 5 ...
 $ V30 : Factor w/ 24 levels "0","0.1","0.2",...: 24 2 2 2 2 2 2 2 2 2 ...
 $ V31 : Factor w/ 70 levels "-0.6","-0.8",...: 70 25 25 25 25 25 25 25 25 25 .
..
 $ V32 : Factor w/ 21 levels "0","0.1","0.2",...: 21 5 5 5 5 5 5 5 5 5 ...
 $ V33 : Factor w/ 56 levels "0","0.01","0.02",...: 56 17 17 17 17 17 17 17 17
17 ...
 $ V34 : Factor w/ 136 levels "-0.02","-0.03",...: 136 135 134 133 132 130 129
131 129 128 ...
 $ V35 : Factor w/ 51 levels "-0.02","-0.03",...: 51 37 35 31 28 26 2 4 4 2 ...
 $ V36 : Factor w/ 99 levels "-0.02","-0.03",...: 99 49 51 51 53 56 58 57 62 68

```

```

...
$ V37 : Factor w/ 154 levels "-1","-10","-11",...: 154 25 14 14 55 36 84 84 90
1 ...
$ V38 : Factor w/ 113 levels "-1","-10","-11",...: 113 9 5 21 19 25 25 14 16 1
0 ...
$ V39 : Factor w/ 214 levels "-1","-10","-11",...: 214 155 148 212 204 140 135
138 139 137 ...
$ V40 : Factor w/ 247 levels "-1","-10","-11",...: 247 192 202 204 208 206 213
215 216 218 ...
$ V41 : Factor w/ 187 levels "428","435","437",...: 187 63 51 50 52 47 49 47 4
2 46 ...
$ V42 : Factor w/ 372 levels "-112","-115",...: 372 83 70 60 38 28 18 12 5 4 .
..
$ V43 : Factor w/ 1238 levels "-0.02","-0.05",...: 1238 671 668 664 659 654 64
9 643 1226 1182 ...
$ V44 : Factor w/ 1694 levels "-0.01","-0.05",...: 1694 517 533 546 557 565 57
0 573 571 564 ...
$ V45 : Factor w/ 1670 levels "-0.18","-0.23",...: 1670 453 407 371 318 278 22
8 158 48 336 ...
$ V46 : Factor w/ 60 levels "1","10","11",...: 60 32 32 29 29 28 27 22 21 20 .
..
$ V47 : Factor w/ 83 levels "0","0.0278","0.1308",...: 83 62 62 62 62 62 62 62
62 62 ...
$ V48 : Factor w/ 90 levels "-128","-128.4898",...: 90 74 74 74 74 74 74 74 74
74 ...
$ V49 : Factor w/ 89 levels "0","0.05","0.2323",...: 89 21 21 21 21 21 21 21 2
1 21 ...
$ V50 : Factor w/ 89 levels "0","0.0025","0.054",...: 89 30 30 30 30 30 30 30
30 30 ...
$ V51 : Factor w/ 90 levels "-10.1695","-10.902",...: 90 1 1 1 1 1 1 1 1 1 ...
$ V52 : Factor w/ 90 levels "0","0.0548","0.2209",...: 90 20 20 20 20 20 20 20
20 20 ...
$ V53 : Factor w/ 90 levels "0","0.003","0.0488",...: 90 19 19 19 19 19 19 19
19 19 ...
$ V54 : Factor w/ 89 levels "-0.0558","-1.7763",...: 89 51 51 51 51 51 51 51 5
1 51 ...
$ V55 : Factor w/ 86 levels "0","0.1924","0.3087",...: 86 41 41 41 41 41 41 41
41 41 ...
$ V56 : Factor w/ 86 levels "0","0.037","0.0953",...: 86 15 15 15 15 15 15 15
15 15 ...
$ V57 : Factor w/ 543 levels "-0.02","-0.03",...: 543 457 466 473 484 491 498
510 512 504 ...
$ V58 : Factor w/ 318 levels "-0.02","-0.03",...: 318 38 40 43 45 48 52 52 52
52 ...
$ V59 : Factor w/ 213 levels "-0.02","-0.03",...: 213 1 7 9 14 15 18 19 13 7 .
..
$ V60 : Factor w/ 707 levels "-1","-10","-100",...: 707 339 342 353 355 361 35
8 363 350 339 ...
$ V61 : Factor w/ 419 levels "-1","-10","-100",...: 419 343 348 358 367 370 37
4 383 387 395 ...
$ V62 : Factor w/ 646 levels "-1","-10","-100",...: 646 258 250 219 218 198 18
5 136 128 115 ...
$ V63 : Factor w/ 1151 levels "-1","-10","-100",...: 1151 892 937 952 988 1015
1042 1070 1074 1091 ...
$ V64 : Factor w/ 817 levels "-1","-10","-100",...: 817 110 111 102 89 79 78 6
6 57 38 ...
$ V65 : Factor w/ 1038 levels "-1","-10","-100",...: 1038 245 214 201 159 138
95 24 7 290 ...
$ V66 : Factor w/ 88 levels "-0.11926","-0.19002",...: 88 33 33 33 33 33 33 33
33 33 ...
$ V67 : Factor w/ 84 levels "-0.10176","-0.15381",...: 84 28 28 28 28 28 28 28
28 28 ...
$ V68 : Factor w/ 82 levels "-0.06791","-0.12096",...: 82 26 26 26 26 26 26 26
26 26 ...

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$ V69 : Factor w/ 88 levels "-0.00696","-0.01884",...: 88 44 44 44 44 44 44 44
44 44 ...
$ V70 : Factor w/ 84 levels "-0.01247","-0.09627",...: 84 3 3 3 3 3 3 3 3 3 ..
.
$ V71 : Factor w/ 82 levels "-0.0046","-0.008",...: 82 41 41 41 41 41 41 41 41 41
41 ...
$ V72 : Factor w/ 83 levels "-0.7","-1.1",...: 83 80 80 80 80 80 80 80 80 80 .
..
$ V73 : Factor w/ 80 levels "-1.9","-122",...: 80 63 63 63 63 63 63 63 63 63 .
..
$ V74 : Factor w/ 39 levels "19","20","21",...: 39 17 17 17 17 17 17 17 17 17
...
$ V75 : Factor w/ 88 levels "-10","-11.5",...: 88 14 14 14 14 14 14 14 14 14 .
..
$ V76 : Factor w/ 84 levels "-10.9","-100",...: 84 43 43 43 43 43 43 43 43 43
...
$ V77 : Factor w/ 30 levels "1","10","11",...: 30 2 2 2 2 2 2 2 2 2 ...
$ V78 : Factor w/ 88 levels "0","0.1","0.57",...: 88 42 42 42 42 42 42 42 42 4
2 ...
$ V79 : Factor w/ 78 levels "0","1","100.2",...: 78 13 13 13 13 13 13 13 13 13
...
$ V80 : Factor w/ 41 levels "0","1","12","13",...: 41 17 17 17 17 17 17 17 17 17
17 ...
$ V81 : Factor w/ 3653 levels "-0.949086746",...: 3653 2993 3032 3029 3034 302
6 3017 3028 3052 3017 ...
$ V82 : Factor w/ 3572 levels "-0.738381151",...: 3572 1153 3542 3326 1118 112
9 1313 1328 3378 1313 ...
$ V83 : Factor w/ 3602 levels "-10.50214329",...: 3602 698 627 644 3598 633 63
9 3591 3572 639 ...
$ V84 : Factor w/ 90 levels "-0.0292","-0.0312",...: 90 5 5 5 5 5 5 5 5 5 ...
$ V85 : Factor w/ 90 levels "-0.0122","-0.0334",...: 90 17 17 17 17 17 17 17 17 1
7 17 ...
$ V86 : Factor w/ 90 levels "-0.0369","-0.0649",...: 90 45 45 45 45 45 45 45 4
5 45 ...
$ V87 : Factor w/ 90 levels "-0.0084","-0.0126",...: 90 8 8 8 8 8 8 8 8 8 ...
$ V88 : Factor w/ 85 levels "-18","-22.8",...: 85 40 40 40 40 40 40 40 40 40 .
..
$ V89 : Factor w/ 80 levels "-17.6","-19.4",...: 80 28 28 28 28 28 28 28 28 28
...
$ V90 : Factor w/ 44 levels "-0.1","-0.2",...: 44 1 1 1 1 1 1 1 1 1 ...
$ V91 : Factor w/ 83 levels "-1.2","-123.3",...: 83 10 10 10 10 10 10 10 10 10
...
$ V92 : Factor w/ 85 levels "-113.8","-118.5",...: 85 52 52 52 52 52 52 52 52
52 ...
$ V93 : Factor w/ 44 levels "-0.1","-0.2",...: 44 1 1 1 1 1 1 1 1 1 ...
$ V94 : Factor w/ 90 levels "0","0.54","0.9",...: 90 58 58 58 58 58 58 58 58 5
8 ...
$ V95 : Factor w/ 90 levels "0","0.61","1.02",...: 90 75 75 75 75 75 75 75 75
75 ...
$ V96 : Factor w/ 2 levels "0","amplitude_yaw_dumbbell": 2 1 1 1 1 1 1 1 1 1
...
$ V97 : Factor w/ 34 levels "1","10","11",...: 34 28 28 28 29 28 28 28 28 28 .
..
$ V98 : Factor w/ 87 levels "0","0.0196","0.0204",...: 87 47 47 47 47 47 47 47
47 47 ...
$ V99 : Factor w/ 90 levels "-10.4132","-11.0964",...: 90 30 30 30 30 30 30 30
30 30 ...
[list output truncated]
> set.seed(1234)
> ind = sample(1:nrow(df),0.8*nrow(df),replace = F)
> df_train = df[ind,-1]
> df_test = df[-ind,-1]
> summary(df)
      v1              v2              v3              v4

```

adelmo	: 311	1323084336:	61	160329 :	3	5/12/2011	14:23	:1585
carlitos	:1580	1323095007:	61	160332 :	3	5/12/2011	11:25	:1243
eurico	: 88	1322832938:	60	188310 :	3	5/12/2011	14:22	: 456
jeremy	: 4	1323084327:	60	248321 :	3	5/12/2011	11:23	: 337
pedro	:2041	1323084346:	60	512311 :	3	2/12/2011	13:35	: 311
user_name:	1	1323084354:	60	536294 :	3	28/11/2011	14:15	: 88
v5 v6 v7 v8 v9								
new_window:	1	29	: 61	123	: 520	26.1	: 175	-93.2 : 150
no	:3936	79	: 61	122	: 411	26.2	: 161	-93.1 : 140
yes	: 88	20	: 60	124	: 382	25.9	: 159	-93.3 : 140
		37	: 60	121	: 252	26	: 132	-94.1 : 104
		4	: 60	125	: 201	26.4	: 125	-93.8 : 96
		43	: 60	129	: 115	25.8	: 112	-94.4 : 64
v10 v11 v12 v13								
20	:1042	-1.03566	:3936	-0.39133	:3944	0.005406	:3936	
3	: 826	-0.076054:	1	-0.15095 :	2	-0.011683:	1	
19	: 780	-0.120447:	1	-2.060105:	2	-0.037647:	1	
2	: 296	-0.157538:	1	0.988872 :	2	-0.068863:	1	
21	: 231	-0.313326:	1	-0.06016 :	1	-0.130813:	1	
4	: 217	-0.395556:	1	-0.108371:	1	-0.160041:	1	
v14 v15 v16 v17 v18								
0.045115	:3944	-4.1	:3939	20	:3959	-1	:3941	-7.25 :3936
-1.717824:	2	-93.1	: 7	21	: 23	-0.7	: 9	-93.2 : 5
-0.045472:	1	-2.9	: 3	3	: 12	-1.1	: 9	-93.4 : 4
-0.04606 :	1	-91.2	: 3	4	: 9	-1.4	: 7	-93.3 : 3
-0.059758:	1	-0.8	: 2	5	: 8	-1.5	: 7	-94 : 3
-0.104399:	1	-1.5	: 2	22	: 2	-1.3	: 6	-94.4 : 3
v19 v20 v21 v22								
18	:3954	-1	:3941	1.345	:3936	2	:3949	
19	: 22	-0.7	: 9	0.1	: 7	1	: 32	
1	: 14	-1.1	: 9	0.5	: 5	3	: 16	
3	: 12	-1.4	: 7	0.2	: 3	0	: 7	
2	: 6	-1.5	: 7	0.3	: 3	4	: 7	
20	: 5	-1.3	: 6	0.7	: 3	5	: 5	
v23 v24 v25 v26								
0	:4024	0.3	:3948	121.9	:3936	0.6	:3941	
amplitude_yaw_belt:	1	0.2	: 20	1.4	: 6	0.5	: 10	
		0	: 10	1.1	: 5	0.1	: 8	
		0.5	: 7	123	: 5	0	: 7	
		0.1	: 6	122.1	: 3	0.4	: 7	
		0.4	: 4	124	: 3	0.2	: 5	
v27 v28 v29 v30 v31								
0.35	:3936	25.75	:3936	0.35	:3936	0.1	:3953	-4.95 :3936
0	: 19	25.9	: 6	0.1	: 15	0	: 29	-93.3 : 6
0.2	: 10	26	: 6	0.2	: 15	0.2	: 10	-3.1 : 3
0.1	: 8	26.2	: 4	0.3	: 11	0.8	: 4	-4.6 : 3
0.3	: 7	7.4	: 4	0.5	: 10	0.3	: 3	-92.9 : 3
0.6	: 5	26.4	: 3	0.7	: 5	0.4	: 3	-2.2 : 2
v32 v33 v34 v35 v36								
0.4	:3947	0.17	:3939	0.02	: 242	-0.03	:1025	-0.44 : 420
0.1	: 11	0	: 13	-0.43	: 222	-0.02	: 824	-0.43 : 370
0.3	: 10	0.01	: 5	0	: 214	0	: 742	-0.46 : 361
0	: 9	0.03	: 4	-0.42	: 198	-0.05	: 389	-0.02 : 339
0.5	: 9	0.08	: 3	-0.4	: 179	0.02	: 283	-0.41 : 231
0.2	: 6	0.12	: 3	-0.45	: 164	0.03	: 114	-0.03 : 215
v37 v38 v39 v40 v41								
-42	: 270	70	: 431	-177	: 148	1	: 229	584 : 178
-41	: 260	69	: 353	-175	: 141	0	: 216	583 : 176
-43	: 225	71	: 345	-178	: 140	-1	: 207	582 : 157
-40	: 211	3	: 258	-172	: 133	2	: 197	581 : 135
-20	: 187	4	: 256	-174	: 133	4	: 177	585 : 128
-19	: 186	68	: 225	21	: 131	-2	: 173	580 : 125
v42 v43 v44 v45 v46								
-370	: 87	-131	: 76	20.7	: 22	-162	: 119	34 : 413

-368	:	83	-130	:	48	15.3	:	17	-161	:	76	25	:	179
-375	:	83	-133	:	48	17.7	:	17	-164	:	63	3	:	116
-376	:	81	-134	:	48	-26.4	:	13	-163	:	57	29	:	114
-373	:	79	-132	:	47	13.8	:	13	-165	:	23	23	:	107
-369	:	78	117	:	47	13.9	:	13	126	:	22	19	:	104
v47			v48			v49			v50			v51		
65.0977	:	3936	76.22175	:	3936	16.1039	:	3936	259.3599	:	3936	-10.1695	:	3936
0	:	8	-128	:	1	0	:	2	0	:	2	-10.902	:	1
0.0278	:	1	-128.4898	:	1	0.05	:	1	0.0025	:	1	-11.7236	:	1
0.1308	:	1	-129.6863	:	1	0.2323	:	1	0.054	:	1	-12.2027	:	1
0.1616	:	1	-130.7826	:	1	0.417	:	1	0.1739	:	1	-14.164	:	1
0.1883	:	1	-131.2292	:	1	0.4247	:	1	0.1804	:	1	-14.9221	:	1
v52			v53			v54			v55					
10.66725	:	3936	113.7978	:	3936	19.0615	:	3936	35.8809	:	3936			
0	:	1	0	:	1	-161	:	2	0	:	5			
0.0548	:	1	0.003	:	1	-0.0558	:	1	0.1924	:	1			
0.2209	:	1	0.0488	:	1	-1.7763	:	1	0.3087	:	1			
0.2218	:	1	0.0492	:	1	-105.7773	:	1	0.3594	:	1			
0.297	:	1	0.0882	:	1	-115.4516	:	1	0.4871	:	1			
v56			v57			v58			v59			v60		
1287.463	:	3936	0.02	:	190	-0.02	:	258	-0.02	:	243	-289	:	109
0	:	5	0	:	128	-0.03	:	84	0	:	172	-288	:	85
0.037	:	1	-0.02	:	39	0	:	73	-0.03	:	117	-290	:	59
0.0953	:	1	-0.06	:	31	-0.05	:	52	-0.07	:	84	-287	:	51
0.1292	:	1	-0.05	:	30	-0.22	:	51	0.02	:	83	1	:	45
0.2373	:	1	-0.22	:	25	-0.24	:	45	0.16	:	70	-1	:	43
v61			v62			v63			v64			v65		
111	:	122	-123	:	89	-367	:	43	335	:	49	510	:	41
110	:	108	-122	:	79	-368	:	42	336	:	40	512	:	41
109	:	77	-124	:	71	-369	:	33	337	:	37	513	:	41
-10	:	41	-125	:	46	-372	:	30	338	:	36	511	:	37
6	:	39	-121	:	35	-370	:	29	339	:	36	515	:	37
-9	:	38	-88	:	31	-371	:	26	334	:	35	514	:	33
v66			v67			v68			v69					
-1.18224	:	3938	-0.96912	:	3942	-0.86977	:	3944	0.12353	:	3938			
-0.11926	:	1	-0.10176	:	1	-0.06791	:	1	-0.00696	:	1			
-0.19002	:	1	-0.15381	:	1	-0.12096	:	1	-0.01884	:	1			
-0.20488	:	1	-0.15426	:	1	-0.1697	:	1	-0.03359	:	1			
-0.34389	:	1	-0.16444	:	1	-0.20332	:	1	-0.08596	:	1			
-0.36227	:	1	-0.18233	:	1	-0.21951	:	1	-0.1009	:	1			
v70			v71			v72			v73			v74		
-0.10319	:	3942	0.059765	:	3944	8.45	:	3936	77.25	:	3936	38	:	3944
-0.01247	:	1	-0.0046	:	1	-10.1	:	2	-161	:	3	34	:	11
-0.09627	:	1	-0.008	:	1	-13.9	:	2	127	:	3	39	:	6
-0.14513	:	1	-0.00863	:	1	-3	:	2	-162	:	2	45	:	4
-0.14758	:	1	-0.05777	:	1	-6.9	:	2	-164	:	2	46	:	4
-0.17734	:	1	-0.08516	:	1	0	:	2	180	:	2	32	:	3
v75			v76			v77			v78			v79		
-33.6	:	3936	-58.6	:	3936	10	:	3942	36.945	:	3936	121.5	:	3936
-4	:	2	-138	:	2	4	:	8	1.3	:	2	0	:	5
-65.5	:	2	-161	:	2	34	:	7	45.2	:	2	1	:	4
-10	:	1	-162	:	2	6	:	6	0	:	1	125.2	:	2
-11.5	:	1	-164	:	2	3	:	5	0.1	:	1	138.8	:	2
-12.7	:	1	-180	:	2	5	:	5	0.57	:	1	158.5	:	2
v80			v81			v82			v83					
27	:	3941	13.2272917	:	58	0	:	93	-84.40838906	:	58			
0	:	8	0	:	39	-70.91576678	:	58	0	:	13			
29	:	5	13.05217456	:	11	-70.49400371	:	11	-84.87393888	:	11			
31	:	5	13.07948887	:	9	-70.63995378	:	9	-84.64918975	:	9			
25	:	4	13.35451266	:	9	-70.67116245	:	9	-84.69053461	:	9			
16	:	3	55.82441814	:	9	9.645819033	:	9	100.2280531	:	9			
v84			v85			v86			v87			v88		
-0.09595	:	3936	-0.4422	:	3936	0.0819	:	3936	-0.216	:	3936	41.85	:	3936
-0.0292	:	1	-0.0122	:	1	-0.0369	:	1	-0.0084	:	1	-70	:	2



-0.0312 :	1	-0.0334 :	1	-0.0649 :	1	-0.0126 :	1	-70.1 :	2
-0.0363 :	1	-0.0605 :	1	-0.0876 :	1	-0.0902 :	1	38 :	2
-0.0947 :	1	-0.0668 :	1	-0.096 :	1	-0.1013 :	1	48.4 :	2
-0.0972 :	1	-0.1238 :	1	-0.1135 :	1	-0.163 :	1	49.9 :	2
v89		v90		v91		v92		v93	
133 :3936		-0.1 :3938		-26.75 :3936		20.2 :3936		-0.1 :3938	
-84.3 : 3		-0.7 : 9		-2.4 : 2		-85.3 : 2		-0.7 : 9	
124 : 2		-0.9 : 5		-70.9 : 2		-85.4 : 2		-0.9 : 5	
133.9 : 2		-0.6 : 4		-71 : 2		-85.5 : 2		-0.6 : 4	
134.6 : 2		-0.8 : 4		10.6 : 2		0 : 2		-0.8 : 4	
137 : 2		0.2 : 4		10.9 : 2		21.9 : 2		0.2 : 4	
v94		v95				v96		v97	
55.71 :3936		54.74 :3936		0 :4024		9 : 513			
0 : 1		0 : 1		amplitude_yaw_dumbbell: 1		10 : 482			
0.54 : 1		0.61 : 1				8 : 328			
0.9 : 1		1.02 : 1				37 : 307			
0.96 : 1		1.1 : 1				11 : 227			
1.08 : 1		1.2 : 1				5 : 217			
v98		v99		v100		v101		v102	
2.41635 :3936		-5.11805 :3936		17.058 :3936		291.001 :3936		13.9312 :3936	
0 : 4		-10.4132 : 1		0 : 1		0 : 1		-0.6827 : 1	
0.0196 : 1		-11.0964 : 1		0.1811 : 1		0.0328 : 1		-1.7762 : 1	
0.0204 : 1		-110.9328 : 1		0.2034 : 1		0.0414 : 1		-12.1577 : 1	
0.0213 : 1		-12.4734 : 1		0.3015 : 1		0.0909 : 1		-17.0387 : 1	
0.0217 : 1		-12.9177 : 1		0.3396 : 1		0.1153 : 1		-19.3215 : 1	
v103		v104		v105		v106		v107	
14.1062 :3936		199.0775 :3936		64.7063 :3936		13.5747 :3936		184.5578 :3936	
0 : 1		0 : 1		-101.7805 : 1		0 : 1		0 : 1	
0.1799 : 1		0.0324 : 1		-105.6502 : 1		0.2172 : 1		0.0472 : 1	
0.2384 : 1		0.0568 : 1		-15.4101 : 1		0.256 : 1		0.0656 : 1	
0.2425 : 1		0.0588 : 1		-17.5064 : 1		0.2815 : 1		0.0793 : 1	
0.2687 : 1		0.0722 : 1		-30.8185 : 1		0.2864 : 1		0.082 : 1	
v108		v109		v110		v111		v112	
0 : 263		-0.02 : 327		0 : 247		-234 : 102		48 : 134	
0.47 : 126		-0.1 : 134		-0.02 : 169		5 : 97		47 : 90	
0.48 : 114		-0.08 : 129		-0.26 : 106		23 : 96		46 : 63	
0.45 : 109		-0.11 : 122		-0.23 : 103		17 : 95		1 : 54	
-0.02 : 103		-0.13 : 107		-0.25 : 103		18 : 94		-9 : 51	
-0.03 : 102		-0.06 : 104		-0.28 : 97		0 : 93		2 : 51	
v113		v114		v115		v116		v117	
-272 : 93		-552 : 77		290 : 77		-73 : 86		0 : 311	
-270 : 75		-558 : 46		295 : 43		-68 : 42		141 : 77	
84 : 67		-554 : 38		-533 : 37		-69 : 42		139 : 71	
-271 : 66		-555 : 38		292 : 35		-65 : 36		134 : 65	
79 : 61		533 : 37		-517 : 32		-71 : 35		140 : 61	
85 : 59		521 : 34		-522 : 32		-70 : 33		133 : 57	
v118		v119		v120		v121		v122	
0 : 312		0 : 311		-1.09475 :3944		-0.97525 :3944		-0.05065 :3944	
-63.8 : 66		107 : 101		-1.3846 : 2		-0.0259 : 1		-0.009 : 1	
-63.5 : 53		108 : 85		-0.0699 : 1		-0.0918 : 1		-0.011 : 1	
-63.7 : 52		106 : 79		-0.0781 : 1		-0.1289 : 1		-0.0252 : 1	
-63.9 : 52		102 : 68		-0.1168 : 1		-0.1574 : 1		-0.0525 : 1	
-63.6 : 43		105 : 65		-0.1804 : 1		-0.2494 : 1		-0.0705 : 1	
v123		v124		v125		v126		v127	
0.17285 :3944		49.6 :3938		168 :3938		-1.1 :3948		4.65 :3936	
-0.0428 : 1		0 : 7		176 : 9		-1.3 : 14		0 : 7	
-0.0673 : 1		-0.2 : 2		0 : 7		-1.5 : 6		-63.9 : 3	
-0.0732 : 1		-63.4 : 2		174 : 7		-0.7 : 5		7.3 : 3	
-0.14 : 1		-63.6 : 2		180 : 6		-0.9 : 5		-63.7 : 2	
-0.2117 : 1		53.4 : 2		171 : 4		-1 : 5		22.7 : 2	
v128		v129		v130		v131			
-168.5 :3936		-1.1 :3948		32.2 :3936		341.5 :3936			
-177 : 7		-1.3 : 14		0 : 8		354 : 9			
-178 : 7		-1.5 : 6		0.3 : 5		0 : 8			

```

0      : 7 -0.7 : 5 23.9 : 2 1 : 7
-175   : 6 -0.9 : 5 41.6 : 2 350 : 7
-176   : 6 -1 : 5 0.15 : 1 351 : 7
                                v132 v133 v134 v135
0      :4024 36 : 593 14.0772:3936 27.85936 :3936
amplitude_yaw_forearm: 1 35 : 366 0 : 7 0 : 7
                                : 34 : 343 0.02083: 1 -136.38298: 1
                                : 37 : 242 0.03701: 1 -14.5 : 1
                                : 33 : 226 0.25391: 1 -145.13953: 1
                                : 28 : 167 0.26768: 1 -18.14035 : 1
                                v136 v137 v138 v139
45.16342:3936 2749.163:3936 25.35597:3936 8.906695:3936
0      : 7 0 : 7 0 : 7 0 : 8
0.04082 : 1 0.00167 : 1 -0.00125: 1 0.06088 : 1
0.12963 : 1 0.0168 : 1 -0.59022: 1 0.06652 : 1
0.15744 : 1 0.02479 : 1 -0.84531: 1 0.07474 : 1
0.31045 : 1 0.09638 : 1 -1.01095: 1 0.07814 : 1
                                v140 v141 v142 v143
79.33451:3936 17.09505 :3936 74.27584:3936 5541.956:3936
0      : 8 0 : 7 0 : 8 0 : 8
0.00371 : 1 -1.07213 : 1 0.05774 : 1 0.00333 : 1
0.00442 : 1 -100.77442: 1 0.20841 : 1 0.04343 : 1
0.00559 : 1 -12.26034 : 1 0.39038 : 1 0.15239 : 1
0.00611 : 1 -13.03158 : 1 0.44344 : 1 0.19664 : 1
                                v144 v145 v146 v147 v148
0.02 : 225 -0.02 : 173 -0.02 : 207 192 : 82 204 : 78
0 : 132 0 : 171 -0.03 : 154 193 : 67 205 : 68
0.03 : 112 -0.03 : 85 0 : 138 191 : 55 203 : 59
0.06 : 92 0.02 : 82 -0.05 : 111 194 : 44 202 : 52
0.08 : 89 0.03 : 58 -0.07 : 69 190 : 36 206 : 45
-0.02 : 76 -0.05 : 42 0.02 : 69 195 : 26 207 : 33
                                v149 v150 v151 v152
-214 : 119 -10 : 38 655 : 51 472 : 37
-213 : 98 -12 : 37 656 : 47 471 : 33
-215 : 98 -11 : 35 653 : 44 469 : 32
-216 : 70 -13 : 34 751 : 43 470 : 30
-191 : 53 -9 : 30 654 : 41 468 : 27
-188 : 52 -14 : 28 749 : 40 467 : 25
[ reached getOption("max.print") -- omitted 1 row ]
> dim(df)
[1] 4025 152

$ v29 : Factor w/ 23 levels "0","0.1","0.2",...: 23 5 5 5 5 5 5 5 5 5 ...
$ v30 : Factor w/ 24 levels "0","0.1","0.2",...: 24 2 2 2 2 2 2 2 2 2 ...
$ v31 : Factor w/ 70 levels "-0.6","-0.8",...: 70 25 25 25 25 25 25 25 25 25 ...
...
$ v32 : Factor w/ 21 levels "0","0.1","0.2",...: 21 5 5 5 5 5 5 5 5 5 ...
$ v33 : Factor w/ 56 levels "0","0.01","0.02",...: 56 17 17 17 17 17 17 17 17 17 ...
$ v34 : Factor w/ 136 levels "-0.02","-0.03",...: 136 135 134 133 132 130 129 131 129 128 ...
$ v35 : Factor w/ 51 levels "-0.02","-0.03",...: 51 37 35 31 28 26 2 4 4 2 ..
.
$ v36 : Factor w/ 99 levels "-0.02","-0.03",...: 99 49 51 51 53 56 58 57 62 6 8 ...
$ v37 : Factor w/ 154 levels "-1","-10","-11",...: 154 25 14 14 55 36 84 84 9 0 1 ...
$ v38 : Factor w/ 113 levels "-1","-10","-11",...: 113 9 5 21 19 25 25 14 16 10 ...
$ v39 : Factor w/ 214 levels "-1","-10","-11",...: 214 155 148 212 204 140 13 5 138 139 137 ...

```

```

$ V40 : Factor w/ 247 levels "-1","-10","-11",...: 247 192 202 204 208 206 21
3 215 216 218 ...
$ V41 : Factor w/ 187 levels "428","435","437",...: 187 63 51 50 52 47 49 47
42 46 ...
$ V42 : Factor w/ 372 levels "-112","-115",...: 372 83 70 60 38 28 18 12 5 4
...
$ V43 : Factor w/ 1238 levels "-0.02","-0.05",...: 1238 671 668 664 659 654 6
49 643 1226 1182 ...
$ V44 : Factor w/ 1694 levels "-0.01","-0.05",...: 1694 517 533 546 557 565 5
70 573 571 564 ...
$ V45 : Factor w/ 1670 levels "-0.18","-0.23",...: 1670 453 407 371 318 278 2
28 158 48 336 ...
$ V46 : Factor w/ 60 levels "1","10","11",...: 60 32 32 29 29 28 27 22 21 20
...
$ V47 : Factor w/ 83 levels "0","0.0278","0.1308",...: 83 62 62 62 62 62 62 6
2 62 62 ...
$ V48 : Factor w/ 90 levels "-128","-128.4898",...: 90 74 74 74 74 74 74 74 7
4 74 ...
$ V49 : Factor w/ 89 levels "0","0.05","0.2323",...: 89 21 21 21 21 21 21 21
21 21 ...
$ V50 : Factor w/ 89 levels "0","0.0025","0.054",...: 89 30 30 30 30 30 30 30
30 30 ...
$ V51 : Factor w/ 90 levels "-10.1695","-10.902",...: 90 1 1 1 1 1 1 1 1 1 ..
.
$ V52 : Factor w/ 90 levels "0","0.0548","0.2209",...: 90 20 20 20 20 20 20 20
0 20 20 ...
$ V53 : Factor w/ 90 levels "0","0.003","0.0488",...: 90 19 19 19 19 19 19 19
19 19 ...
$ V54 : Factor w/ 89 levels "-0.0558","-1.7763",...: 89 51 51 51 51 51 51 51
51 51 ...
$ V55 : Factor w/ 86 levels "0","0.1924","0.3087",...: 86 41 41 41 41 41 41 41
1 41 41 ...
$ V56 : Factor w/ 86 levels "0","0.037","0.0953",...: 86 15 15 15 15 15 15 15
15 15 ...
$ V57 : Factor w/ 543 levels "-0.02","-0.03",...: 543 457 466 473 484 491 498
510 512 504 ...
$ V58 : Factor w/ 318 levels "-0.02","-0.03",...: 318 38 40 43 45 48 52 52 52
52 ...
$ V59 : Factor w/ 213 levels "-0.02","-0.03",...: 213 1 7 9 14 15 18 19 13 7
...
$ V60 : Factor w/ 707 levels "-1","-10","-100",...: 707 339 342 353 355 361 3
58 363 350 339 ...
$ V61 : Factor w/ 419 levels "-1","-10","-100",...: 419 343 348 358 367 370 3
74 383 387 395 ...
$ V62 : Factor w/ 646 levels "-1","-10","-100",...: 646 258 250 219 218 198 1
85 136 128 115 ...
$ V63 : Factor w/ 1151 levels "-1","-10","-100",...: 1151 892 937 952 988 101
5 1042 1070 1074 1091 ...
$ V64 : Factor w/ 817 levels "-1","-10","-100",...: 817 110 111 102 89 79 78
66 57 38 ...
$ V65 : Factor w/ 1038 levels "-1","-10","-100",...: 1038 245 214 201 159 138
95 24 7 290 ...
$ V66 : Factor w/ 88 levels "-0.11926","-0.19002",...: 88 33 33 33 33 33 33 3
3 33 33 ...
$ V67 : Factor w/ 84 levels "-0.10176","-0.15381",...: 84 28 28 28 28 28 28 2
8 28 28 ...

```

```

$ V68 : Factor w/ 82 levels "-0.06791","-0.12096",...: 82 26 26 26 26 26 26 2
6 26 26 ...
$ V69 : Factor w/ 88 levels "-0.00696","-0.01884",...: 88 44 44 44 44 44 44 4
4 44 44 ...
$ V70 : Factor w/ 84 levels "-0.01247","-0.09627",...: 84 3 3 3 3 3 3 3 3 3 .
..
$ V71 : Factor w/ 82 levels "-0.0046","-0.008",...: 82 41 41 41 41 41 41 41 4
1 41 ...
$ V72 : Factor w/ 83 levels "-0.7","-1.1",...: 83 80 80 80 80 80 80 80 80 80
...
$ V73 : Factor w/ 80 levels "-1.9","-122",...: 80 63 63 63 63 63 63 63 63 63
...
$ V74 : Factor w/ 39 levels "19","20","21",...: 39 17 17 17 17 17 17 17 17 17
...
$ V75 : Factor w/ 88 levels "-10","-11.5",...: 88 14 14 14 14 14 14 14 14 14
...
$ V76 : Factor w/ 84 levels "-10.9","-100",...: 84 43 43 43 43 43 43 43 43 43
...
$ V77 : Factor w/ 30 levels "1","10","11",...: 30 2 2 2 2 2 2 2 2 2 ...
$ V78 : Factor w/ 88 levels "0","0.1","0.57",...: 88 42 42 42 42 42 42 42 42
42 ...
$ V79 : Factor w/ 78 levels "0","1","100.2",...: 78 13 13 13 13 13 13 13 13 1
3 ...
$ V80 : Factor w/ 41 levels "0","1","12","13",...: 41 17 17 17 17 17 17 17 17
17 ...
$ V81 : Factor w/ 3653 levels "-0.949086746",...: 3653 2993 3032 3029 3034 30
26 3017 3028 3052 3017 ...
$ V82 : Factor w/ 3572 levels "-0.738381151",...: 3572 1153 3542 3326 1118 11
29 1313 1328 3378 1313 ...
$ V83 : Factor w/ 3602 levels "-10.50214329",...: 3602 698 627 644 3598 633 6
39 3591 3572 639 ...
$ V84 : Factor w/ 90 levels "-0.0292","-0.0312",...: 90 5 5 5 5 5 5 5 5 5 ...
$ V85 : Factor w/ 90 levels "-0.0122","-0.0334",...: 90 17 17 17 17 17 17 17
17 17 ...
$ V86 : Factor w/ 90 levels "-0.0369","-0.0649",...: 90 45 45 45 45 45 45 45
45 45 ...
$ V87 : Factor w/ 90 levels "-0.0084","-0.0126",...: 90 8 8 8 8 8 8 8 8 8 ...
$ V88 : Factor w/ 85 levels "-18","-22.8",...: 85 40 40 40 40 40 40 40 40 40
...
$ V89 : Factor w/ 80 levels "-17.6","-19.4",...: 80 28 28 28 28 28 28 28 28 2
8 ...
$ V90 : Factor w/ 44 levels "-0.1","-0.2",...: 44 1 1 1 1 1 1 1 1 1 ...
$ V91 : Factor w/ 83 levels "-1.2","-123.3",...: 83 10 10 10 10 10 10 10 10 1
0 ...
$ V92 : Factor w/ 85 levels "-113.8","-118.5",...: 85 52 52 52 52 52 52 52 52
52 ...
$ V93 : Factor w/ 44 levels "-0.1","-0.2",...: 44 1 1 1 1 1 1 1 1 1 ...
$ V94 : Factor w/ 90 levels "0","0.54","0.9",...: 90 58 58 58 58 58 58 58 58
58 ...
$ V95 : Factor w/ 90 levels "0","0.61","1.02",...: 90 75 75 75 75 75 75 75 75
75 ...
$ V96 : Factor w/ 2 levels "0","amplitude_yaw_dumbbell": 2 1 1 1 1 1 1 1 1 1
...
$ V97 : Factor w/ 34 levels "1","10","11",...: 34 28 28 28 29 28 28 28 28 28
...
$ V98 : Factor w/ 87 levels "0","0.0196","0.0204",...: 87 47 47 47 47 47 47 4
7 47 47 ...

```

```
$ v99 : Factor w/ 90 levels "-10.4132","-11.0964",...: 90 30 30 30 30 30 30 3  
0 30 30 ...
```

```
>
> set.seed(1234)
> ind = sample(1:nrow(df), 0.8*nrow(df), replace = F)
> df_train = df[ind, -1]
> df_test = df[-ind, -1]
>
> summary(df)
```

	v10	v11	v12	v13
20	:1042	-1.03566 :3936	-0.39133 :3944	0.005406 :3936
3	: 826	-0.076054: 1	-0.15095 : 2	-0.011683: 1
19	: 780	-0.120447: 1	-2.060105: 2	-0.037647: 1
2	: 296	-0.157538: 1	0.988872 : 2	-0.068863: 1
21	: 231	-0.313326: 1	-0.06016 : 1	-0.130813: 1
4	: 217	-0.395556: 1	-0.108371: 1	-0.160041: 1

	v19		v20		v21		v22
18	:3954	-1	:3941	1.345	:3936	2	:3949
19	: 22	-0.7	: 9	0.1	: 7	1	: 32
1	: 14	-1.1	: 9	0.5	: 5	3	: 16
3	: 12	-1.4	: 7	0.2	: 3	0	: 7
2	: 6	-1.5	: 7	0.3	: 3	4	: 7
20	: 5	-1.3	: 6	0.7	: 3	5	: 5

V27		V28		V29		V30		V31	
0.35	:3936	25.75	:3936	0.35	:3936	0.1	:3953	-4.95	:3936
0	: 19	25.9	: 6	0.1	: 15	0	: 29	-93.3	: 6
0.2	: 10	26	: 6	0.2	: 15	0.2	: 10	-3.1	: 3
0.1	: 8	26.2	: 4	0.3	: 11	0.8	: 4	-4.6	: 3

0.3	:	7	7.4	:	4	0.5	:	10	0.3	:	3	-92.9	:	3
0.6	:	5	26.4	:	3	0.7	:	5	0.4	:	3	-2.2	:	2
		v32			v33			v34			v35			v36
0.4	:	3947	0.17	:	3939	0.02	:	242	-0.03	:	1025	-0.44	:	420
0.1	:	11	0	:	13	-0.43	:	222	-0.02	:	824	-0.43	:	370
0.3	:	10	0.01	:	5	0	:	214	0	:	742	-0.46	:	361
0	:	9	0.03	:	4	-0.42	:	198	-0.05	:	389	-0.02	:	339
0.5	:	9	0.08	:	3	-0.4	:	179	0.02	:	283	-0.41	:	231
0.2	:	6	0.12	:	3	-0.45	:	164	0.03	:	114	-0.03	:	215
		v37			v38			v39			v40			v41
-42	:	270	70	:	431	-177	:	148	1	:	229	584	:	178
-41	:	260	69	:	353	-175	:	141	0	:	216	583	:	176
-43	:	225	71	:	345	-178	:	140	-1	:	207	582	:	157
-40	:	211	3	:	258	-172	:	133	2	:	197	581	:	135
-20	:	187	4	:	256	-174	:	133	4	:	177	585	:	128
-19	:	186	68	:	225	21	:	131	-2	:	173	580	:	125
		v42			v43			v44			v45			v46
-370	:	87	-131	:	76	20.7	:	22	-162	:	119	34	:	413
-368	:	83	-130	:	48	15.3	:	17	-161	:	76	25	:	179
-375	:	83	-133	:	48	17.7	:	17	-164	:	63	3	:	116
-376	:	81	-134	:	48	-26.4	:	13	-163	:	57	29	:	114
-373	:	79	-132	:	47	13.8	:	13	-165	:	23	23	:	107
-369	:	78	117	:	47	13.9	:	13	126	:	22	19	:	104
		v47			v48			v49			v50			v51
65.0977	:	3936	76.22175	:	3936	16.1039	:	3936	259.3599	:	3936	-10.1695	:	3936
0	:	8	-128	:	1	0	:	2	0	:	2	-10.902	:	1
0.0278	:	1	-128.4898	:	1	0.05	:	1	0.0025	:	1	-11.7236	:	1
0.1308	:	1	-129.6863	:	1	0.2323	:	1	0.054	:	1	-12.2027	:	1
0.1616	:	1	-130.7826	:	1	0.417	:	1	0.1739	:	1	-14.164	:	1
0.1883	:	1	-131.2292	:	1	0.4247	:	1	0.1804	:	1	-14.9221	:	1
		v52			v53			v54			v55			
10.66725	:	3936	113.7978	:	3936	19.0615	:	3936	35.8809	:	3936			
0	:	1	0	:	1	-161	:	2	0	:	5			
0.0548	:	1	0.003	:	1	-0.0558	:	1	0.1924	:	1			
0.2209	:	1	0.0488	:	1	-1.7763	:	1	0.3087	:	1			
0.2218	:	1	0.0492	:	1	-105.7773	:	1	0.3594	:	1			
0.297	:	1	0.0882	:	1	-115.4516	:	1	0.4871	:	1			
		v56			v57			v58			v59			v60
1287.463	:	3936	0.02	:	190	-0.02	:	258	-0.02	:	243	-289	:	109
0	:	5	0	:	128	-0.03	:	84	0	:	172	-288	:	85
0.037	:	1	-0.02	:	39	0	:	73	-0.03	:	117	-290	:	59
0.0953	:	1	-0.06	:	31	-0.05	:	52	-0.07	:	84	-287	:	51
0.1292	:	1	-0.05	:	30	-0.22	:	51	0.02	:	83	1	:	45
0.2373	:	1	-0.22	:	25	-0.24	:	45	0.16	:	70	-1	:	43
		v61			v62			v63			v64			v65
111	:	122	-123	:	89	-367	:	43	335	:	49	510	:	41
110	:	108	-122	:	79	-368	:	42	336	:	40	512	:	41
109	:	77	-124	:	71	-369	:	33	337	:	37	513	:	41
-10	:	41	-125	:	46	-372	:	30	338	:	36	511	:	37
6	:	39	-121	:	35	-370	:	29	339	:	36	515	:	37
-9	:	38	-88	:	31	-371	:	26	334	:	35	514	:	33
		v66			v67			v68			v69			
-1.18224	:	3938	-0.96912	:	3942	-0.86977	:	3944	0.12353	:	3938			
-0.11926	:	1	-0.10176	:	1	-0.06791	:	1	-0.00696	:	1			
-0.19002	:	1	-0.15381	:	1	-0.12096	:	1	-0.01884	:	1			
-0.20488	:	1	-0.15426	:	1	-0.1697	:	1	-0.03359	:	1			
-0.34389	:	1	-0.16444	:	1	-0.20332	:	1	-0.08596	:	1			

-0.36227: 1	-0.18233: 1	-0.21951: 1	-0.1009 : 1	
v70	v71	v72	v73	v74
-0.10319:3942	0.059765:3944	8.45 :3936	77.25 :3936	38 :3944
-0.01247: 1	-0.0046 : 1	-10.1 : 2	-161 : 3	34 : 11
-0.09627: 1	-0.008 : 1	-13.9 : 2	127 : 3	39 : 6
-0.14513: 1	-0.00863: 1	-3 : 2	-162 : 2	45 : 4
-0.14758: 1	-0.05777: 1	-6.9 : 2	-164 : 2	46 : 4
-0.17734: 1	-0.08516: 1	0 : 2	180 : 2	32 : 3
v75	v76	v77	v78	v79
-33.6 :3936	-58.6 :3936	10 :3942	36.945 :3936	121.5 :3936
-4 : 2	-138 : 2	4 : 8	1.3 : 2	0 : 5
-65.5 : 2	-161 : 2	34 : 7	45.2 : 2	1 : 4
-10 : 1	-162 : 2	6 : 6	0 : 1	125.2 : 2
-11.5 : 1	-164 : 2	3 : 5	0.1 : 1	138.8 : 2
-12.7 : 1	-180 : 2	5 : 5	0.57 : 1	158.5 : 2
v80	v81	v82	v83	
27 :3941	13.2272917 : 58	0 : 93	-84.40838906: 58	
0 : 8	0 : 39	-70.91576678: 58	0 : 13	
29 : 5	13.05217456: 11	-70.49400371: 11	-84.87393888: 11	
31 : 5	13.07948887: 9	-70.63995378: 9	-84.64918975: 9	
25 : 4	13.35451266: 9	-70.67116245: 9	-84.69053461: 9	
16 : 3	55.82441814: 9	9.645819033 : 9	100.2280531 : 9	
v84	v85	v86	v87	v88
-0.09595:3936	-0.4422:3936	0.0819 :3936	-0.216 :3936	41.85 :3936
-0.0292 : 1	-0.0122: 1	-0.0369: 1	-0.0084: 1	-70 : 2
-0.0312 : 1	-0.0334: 1	-0.0649: 1	-0.0126: 1	-70.1 : 2
-0.0363 : 1	-0.0605: 1	-0.0876: 1	-0.0902: 1	38 : 2
-0.0947 : 1	-0.0668: 1	-0.096 : 1	-0.1013: 1	48.4 : 2
-0.0972 : 1	-0.1238: 1	-0.1135: 1	-0.163 : 1	49.9 : 2
v89	v90	v91	v92	v93
133 :3936	-0.1 :3938	-26.75 :3936	20.2 :3936	-0.1 :3938
-84.3 : 3	-0.7 : 9	-2.4 : 2	-85.3 : 2	-0.7 : 9
124 : 2	-0.9 : 5	-70.9 : 2	-85.4 : 2	-0.9 : 5
133.9 : 2	-0.6 : 4	-71 : 2	-85.5 : 2	-0.6 : 4
134.6 : 2	-0.8 : 4	10.6 : 2	0 : 2	-0.8 : 4
137 : 2	0.2 : 4	10.9 : 2	21.9 : 2	0.2 : 4
v94	v95	v96	v97	
55.71 :3936	54.74 :3936	0 :4024	9 : 513	
0 : 1	0 : 1	amplitude_yaw_dumbbell: 1	10 : 482	
0.54 : 1	0.61 : 1		8 : 328	
0.9 : 1	1.02 : 1		37 : 307	
0.96 : 1	1.1 : 1		11 : 227	
1.08 : 1	1.2 : 1		5 : 217	
v98	v99	v100	v101	v102
2.41635:3936	-5.11805 :3936	17.058 :3936	291.001:3936	13.9312 :3936
0 : 4	-10.4132 : 1	0 : 1	0 : 1	-0.6827 : 1
0.0196 : 1	-11.0964 : 1	0.1811 : 1	0.0328 : 1	-1.7762 : 1
0.0204 : 1	-110.9328: 1	0.2034 : 1	0.0414 : 1	-12.1577: 1
0.0213 : 1	-12.4734 : 1	0.3015 : 1	0.0909 : 1	-17.0387: 1
0.0217 : 1	-12.9177 : 1	0.3396 : 1	0.1153 : 1	-19.3215: 1
v103	v104	v105	v106	v107
14.1062:3936	199.0775:3936	64.7063 :3936	13.5747:3936	184.5578:3936
0 : 1	0 : 1	-101.7805: 1	0 : 1	0 : 1
0.1799 : 1	0.0324 : 1	-105.6502: 1	0.2172 : 1	0.0472 : 1
0.2384 : 1	0.0568 : 1	-15.4101 : 1	0.256 : 1	0.0656 : 1
0.2425 : 1	0.0588 : 1	-17.5064 : 1	0.2815 : 1	0.0793 : 1
0.2687 : 1	0.0722 : 1	-30.8185 : 1	0.2864 : 1	0.082 : 1

v108		v109		v110		v111		v112	
0	: 263	-0.02	: 327	0	: 247	-234	: 102	48	: 134
0.47	: 126	-0.1	: 134	-0.02	: 169	5	: 97	47	: 90
0.48	: 114	-0.08	: 129	-0.26	: 106	23	: 96	46	: 63
0.45	: 109	-0.11	: 122	-0.23	: 103	17	: 95	1	: 54
-0.02	: 103	-0.13	: 107	-0.25	: 103	18	: 94	-9	: 51
-0.03	: 102	-0.06	: 104	-0.28	: 97	0	: 93	2	: 51
v113		v114		v115		v116		v117	
-272	: 93	-552	: 77	290	: 77	-73	: 86	0	: 311
-270	: 75	-558	: 46	295	: 43	-68	: 42	141	: 77
84	: 67	-554	: 38	-533	: 37	-69	: 42	139	: 71
-271	: 66	-555	: 38	292	: 35	-65	: 36	134	: 65
79	: 61	533	: 37	-517	: 32	-71	: 35	140	: 61
85	: 59	521	: 34	-522	: 32	-70	: 33	133	: 57
v118		v119		v120		v121		v122	
0	: 312	0	: 311	-1.09475:3944		-0.97525:3944		-0.05065:3944	
-63.8	: 66	107	: 101	-1.3846	: 2	-0.0259	: 1	-0.009	: 1
-63.5	: 53	108	: 85	-0.0699	: 1	-0.0918	: 1	-0.011	: 1
-63.7	: 52	106	: 79	-0.0781	: 1	-0.1289	: 1	-0.0252	: 1
-63.9	: 52	102	: 68	-0.1168	: 1	-0.1574	: 1	-0.0525	: 1
-63.6	: 43	105	: 65	-0.1804	: 1	-0.2494	: 1	-0.0705	: 1
v123		v124		v125		v126		v127	
0.17285:3944		49.6	:3938	168	:3938	-1.1	:3948	4.65	:3936
-0.0428:	1	0	: 7	176	: 9	-1.3	: 14	0	: 7
-0.0673:	1	-0.2	: 2	0	: 7	-1.5	: 6	-63.9	: 3
-0.0732:	1	-63.4	: 2	174	: 7	-0.7	: 5	7.3	: 3
-0.14	: 1	-63.6	: 2	180	: 6	-0.9	: 5	-63.7	: 2
-0.2117:	1	53.4	: 2	171	: 4	-1	: 5	22.7	: 2
v128		v129		v130		v131			
-168.5	:3936	-1.1	:3948	32.2	:3936	341.5	:3936		
-177	: 7	-1.3	: 14	0	: 8	354	: 9		
-178	: 7	-1.5	: 6	0.3	: 5	0	: 8		
0	: 7	-0.7	: 5	23.9	: 2	1	: 7		
-175	: 6	-0.9	: 5	41.6	: 2	350	: 7		
-176	: 6	-1	: 5	0.15	: 1	351	: 7		
		v132		v133		v134		v135	
0		:4024		36	: 593	14.0772:3936		27.85936	:3936
amplitude_yaw_forearm:	1			35	: 366	0	: 7	0	: 7
				34	: 343	0.02083:	1	-136.38298:	1
				37	: 242	0.03701:	1	-14.5	: 1
				33	: 226	0.25391:	1	-145.13953:	1
				28	: 167	0.26768:	1	-18.14035	: 1
v136		v137		v138		v139			
45.16342:3936		2749.163:3936		25.35597:3936		8.906695:3936			
0	: 7	0	: 7	0	: 7	0	: 8		
0.04082	: 1	0.00167	: 1	-0.00125:	1	0.06088	: 1		
0.12963	: 1	0.0168	: 1	-0.59022:	1	0.06652	: 1		
0.15744	: 1	0.02479	: 1	-0.84531:	1	0.07474	: 1		
0.31045	: 1	0.09638	: 1	-1.01095:	1	0.07814	: 1		
v140		v141		v142		v143			
79.33451:3936		17.09505	:3936	74.27584:3936		5541.956:3936			
0	: 8	0	: 7	0	: 8	0	: 8		
0.00371	: 1	-1.07213	: 1	0.05774	: 1	0.00333	: 1		
0.00442	: 1	-100.77442:	1	0.20841	: 1	0.04343	: 1		
0.00559	: 1	-12.26034	: 1	0.39038	: 1	0.15239	: 1		
0.00611	: 1	-13.03158	: 1	0.44344	: 1	0.19664	: 1		
v144		v145		v146		v147		v148	



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0.02 : 225 -0.02 : 173 -0.02 : 207 192 : 82 204 : 78
0 : 132 0 : 171 -0.03 : 154 193 : 67 205 : 68
0.03 : 112 -0.03 : 85 0 : 138 191 : 55 203 : 59
0.06 : 92 0.02 : 82 -0.05 : 111 194 : 44 202 : 52
0.08 : 89 0.03 : 58 -0.07 : 69 190 : 36 206 : 45
-0.02 : 76 -0.05 : 42 0.02 : 69 195 : 26 207 : 33
      v149      v150      v151      v152
-214 : 119 -10 : 38 655 : 51 472 : 37
-213 : 98 -12 : 37 656 : 47 471 : 33
-215 : 98 -11 : 35 653 : 44 469 : 32
-216 : 70 -13 : 34 751 : 43 470 : 30
-191 : 53 -9 : 30 654 : 41 468 : 27
-188 : 52 -14 : 28 749 : 40 467 : 25
[ reached getOption("max.print") -- omitted 1 row ]
> dim(df)
[1] 4025 152
> # outlier definition
> # x > Q3+1.5*IQR - positive side outlier
> # x < Q1-1.5*IQR - negative or lower side outlier
> #par(mfrow=c(2,3))
> #(boxplot(df$v24)$out);(boxplot(df$v25)$out);(boxplot(df$v26)$out);(boxplot
(df$v27)$out);(boxplot(df$v28)$out)
> #(boxplot(df$v29)$out);(boxplot(df$v30)$out);(boxplot(df$v31)$out);(boxplot
(df$v32)$out)
>
>
> apply(df,2,range)
      v1      v2      v3
[1,] "adelmo" "1322489729" "100232"
[2,] "user_name" "raw_timestamp_part_1" "raw_timestamp_part_2"
      v4      v5      v6      v7      v8
[1,] "2/12/2011 13:35" "new_window" "1" "-0.03" "-27.8"
[2,] "cvtd_timestamp" "yes" "num_window" "roll_belt" "pitch_belt"
      v9      v10      v11      v12
[1,] "-0.02" "0" "-0.076054" "-0.06016"
[2,] "yaw_belt" "total_accel_belt" "kurtosis_roll_belt" "kurtosis_pitch_belt"
      v13      v14      v15      v16
[1,] "-0.011683" "-0.045472" "-0.2" "10"
[2,] "skewness_roll_belt" "skewness_roll_belt" "max_roll_belt" "max_pitch_belt"
      v17      v18      v19      v20
[1,] "-0.1" "-1" "0" "-0.1"
[2,] "max_yaw_belt" "min_roll_belt" "min_pitch_belt" "min_yaw_belt"
      v21      v22      v23
[1,] "0" "0" "0"
[2,] "amplitude_roll_belt" "amplitude_pitch_belt" "amplitude_yaw_belt"
      v24      v25      v26      v27
[1,] "0" "-0.7" "0" "0"
[2,] "var_total_accel_belt" "avg_roll_belt" "stddev_roll_belt" "var_roll_belt"
      v28      v29      v30      v31
[1,] "-30.6" "0" "0" "-0.6"
[2,] "avg_pitch_belt" "stddev_pitch_belt" "var_pitch_belt" "avg_yaw_belt"
      v32      v33      v34      v35
[1,] "0" "0" "-0.02" "-0.02"
[2,] "stddev_yaw_belt" "var_yaw_belt" "gyros_belt_x" "gyros_belt_y"
      v36      v37      v38      v39

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[1,] "-0.02"      "-1"      "-1"      "-1"
[2,] "gyros_belt_z" "accel_belt_x" "accel_belt_y" "accel_belt_z"
    v40          v41          v42          v43          v44
[1,] "-1"      "428"      "-112"      "-0.02"      "-0.01"
[2,] "magnet_belt_x" "magnet_belt_y" "magnet_belt_z" "roll_arm" "pitch_arm"
    v45          v46          v47          v48
[1,] "-0.18"      "1"      "0"      "-128"
[2,] "yaw_arm" "total_accel_arm" "var_accel_arm" "avg_roll_arm"
    v49          v50          v51          v52
[1,] "0"      "0"      "-10.1695"      "0"
[2,] "stddev_roll_arm" "var_roll_arm" "avg_pitch_arm" "stddev_pitch_arm"
    v53          v54          v55          v56          v57
[1,] "0"      "-0.0558"      "0"      "0"      "-0.02"
[2,] "var_pitch_arm" "avg_yaw_arm" "stddev_yaw_arm" "var_yaw_arm" "gyros_arm_
x"
    v58          v59          v60          v61          v62
[1,] "-0.02"      "-0.02"      "-1"      "-1"      "-1"
[2,] "gyros_arm_y" "gyros_arm_z" "accel_arm_x" "accel_arm_y" "accel_arm_z"
    v63          v64          v65          v66
[1,] "-1"      "-1"      "-1"      "-0.11926"
[2,] "magnet_arm_x" "magnet_arm_y" "magnet_arm_z" "kurtosis_roll_arm"
    v67          v68          v69
[1,] "-0.10176"      "-0.06791"      "-0.00696"
[2,] "kurtosis_pitch_arm" "kurtosis_yaw_arm" "skewness_roll_arm"
    v70          v71          v72          v73
[1,] "-0.01247"      "-0.0046"      "-0.7"      "-1.9"
[2,] "skewness_pitch_arm" "skewness_yaw_arm" "max_roll_arm" "max_pitch_arm"
    v74          v75          v76          v77
[1,] "19"      "-10"      "-10.9"      "1"
[2,] "max_yaw_arm" "min_roll_arm" "min_pitch_arm" "min_yaw_arm"
    v78          v79          v80
[1,] "0"      "0"      "0"
[2,] "amplitude_roll_arm" "amplitude_pitch_arm" "amplitude_yaw_arm"
    v81          v82          v83          v84
[1,] "-0.949086746"      "-0.738381151"      "-10.50214329"      "-0.0292"
[2,] "roll_dumbbell" "pitch_dumbbell" "yaw_dumbbell" "kurtosis_roll_dumbbell"
    v85          v86
[1,] "-0.0122"      "-0.0369"
[2,] "kurtosis_pitch_dumbbell" "skewness_roll_dumbbell"
    v87          v88          v89
[1,] "-0.0084"      "-18"      "-17.6"
[2,] "skewness_pitch_dumbbell" "max_roll_dumbbell" "max_pitch_dumbbell"
    v90          v91          v92
[1,] "-0.1"      "-1.2"      "-113.8"
[2,] "max_yaw_dumbbell" "min_roll_dumbbell" "min_pitch_dumbbell"
    v93          v94          v95
[1,] "-0.1"      "0"      "0"
[2,] "min_yaw_dumbbell" "amplitude_roll_dumbbell" "amplitude_pitch_dumbbell"
    v96          v97          v98
[1,] "0"      "1"      "0"
[2,] "amplitude_yaw_dumbbell" "total_accel_dumbbell" "var_accel_dumbbell"
    v99          v100          v101
[1,] "-10.4132"      "0"      "0"
[2,] "avg_roll_dumbbell" "stddev_roll_dumbbell" "var_roll_dumbbell"
    v102          v103          v104
[1,] "-0.6827"      "0"      "0"
[2,] "avg_pitch_dumbbell" "stddev_pitch_dumbbell" "var_pitch_dumbbell"

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      v105          v106          v107
[1,] "-101.7805"      "0"          "0"
[2,] "avg_yaw_dumbbell" "stddev_yaw_dumbbell" "var_yaw_dumbbell"
      v108          v109          v110
[1,] "-0.02"         "-0.02"         "-0.02"
[2,] "gyros_dumbbell_x" "gyros_dumbbell_y" "gyros_dumbbell_z"
      v111          v112          v113
[1,] "-1"            "-1"            "-1"
[2,] "accel_dumbbell_x" "accel_dumbbell_y" "accel_dumbbell_z"
      v114          v115          v116          v117
[1,] "-1"            "-459"         "-1"            "-111"
[2,] "magnet_dumbbell_x" "magnet_dumbbell_y" "magnet_dumbbell_z" "roll_forear
m"
      v118          v119          v120
[1,] "-0.05"         "-100"         "-0.0699"
[2,] "pitch_forearm" "yaw_forearm" "kurtosis_roll_forearm"
      v121          v122          v123
[1,] "-0.0259"       "-0.009"       "-0.0428"
[2,] "kurtosis_picth_forearm" "skewness_roll_forearm" "skewness_pitch_forearm"
"
      v124          v125          v126
[1,] "-0.2"          "-143"         "-0.1"
[2,] "max_roll_forearm" "max_picth_forearm" "max_yaw_forearm"
      v127          v128          v129
[1,] "-0.9"          "-144"         "-0.1"
[2,] "min_roll_forearm" "min_pitch_forearm" "min_yaw_forearm"
      v130          v131          v132
[1,] "0"              "0"              "0"
[2,] "amplitude_roll_forearm" "amplitude_pitch_forearm" "amplitude_yaw_forear
m"
      v133          v134          v135
[1,] "10"             "0"             "-136.38298"
[2,] "total_accel_forearm" "var_accel_forearm" "avg_roll_forearm"
      v136          v137          v138
[1,] "0"              "0"              "-0.00125"
[2,] "stddev_roll_forearm" "var_roll_forearm" "avg_pitch_forearm"
      v139          v140          v141
[1,] "0"              "0"              "-1.07213"
[2,] "stddev_pitch_forearm" "var_pitch_forearm" "avg_yaw_forearm"
      v142          v143          v144          v145
[1,] "0"              "0"              "-0.02"         "-0.02"
[2,] "stddev_yaw_forearm" "var_yaw_forearm" "gyros_forearm_x" "gyros_forearm_
y"
      v146          v147          v148          v149
[1,] "-0.02"         "-1"          "-10"         "-1"
[2,] "gyros_forearm_z" "accel_forearm_x" "accel_forearm_y" "accel_forearm_z"
      v150          v151          v152
[1,] "-1"            "-1"            "-100"
[2,] "magnet_forearm_x" "magnet_forearm_y" "magnet_forearm_z"
>
> apply(df,2,summary)
      v1          v2          v3          v4          v5          v6
Length "4025"      "4025"      "4025"      "4025"      "4025"      "4025"
Class  "character" "character" "character" "character" "character" "character"
"
Mode    "character" "character" "character" "character" "character" "character"
"

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Length	v7	v8	v9	v10	v11	v12
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
"	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
"						
Length	v13	v14	v15	v16	v17	v18
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
"	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
"						
Length	v19	v20	v21	v22	v23	v24
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
"	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
"						
Length	v25	v26	v27	v28	v29	v30
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
"	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
"						
Length	v31	v32	v33	v34	v35	v36
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
"	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
"						
Length	v37	v38	v39	v40	v41	v42
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
"	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
"						
Length	v43	v44	v45	v46	v47	v48
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
"	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
"						
Length	v49	v50	v51	v52	v53	v54
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
"	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
"						
Length	v55	v56	v57	v58	v59	v60
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
"	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
"						
Length	v61	v62	v63	v64	v65	v66
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"

Class	"character"	"character"	"character"	"character"	"character"	"character"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
Length	v67	v68	v69	v70	v71	v72
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
Length	v73	v74	v75	v76	v77	v78
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
Length	v79	v80	v81	v82	v83	v84
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
Length	v85	v86	v87	v88	v89	v90
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
Length	v91	v92	v93	v94	v95	v96
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
Length	v97	v98	v99	v100	v101	v102
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
Length	v103	v104	v105	v106	v107	v108
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
Length	v109	v110	v111	v112	v113	v114
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"
Mode	"character"	"character"	"character"	"character"	"character"	"character"
Length	v115	v116	v117	v118	v119	v120
Class	"4025"	"4025"	"4025"	"4025"	"4025"	"4025"

```

Mode    "character" "character" "character" "character" "character" "character" "character"
"
      V121      V122      V123      V124      V125      V126
Length "4025"      "4025"      "4025"      "4025"      "4025"      "4025"
Class  "character" "character" "character" "character" "character" "character" "character"
"
Mode    "character" "character" "character" "character" "character" "character" "character"
"
      V127      V128      V129      V130      V131      V132
Length "4025"      "4025"      "4025"      "4025"      "4025"      "4025"
Class  "character" "character" "character" "character" "character" "character" "character"
"
Mode    "character" "character" "character" "character" "character" "character" "character"
"
      V133      V134      V135      V136      V137      V138
Length "4025"      "4025"      "4025"      "4025"      "4025"      "4025"
Class  "character" "character" "character" "character" "character" "character" "character"
"
Mode    "character" "character" "character" "character" "character" "character" "character"
"
      V139      V140      V141      V142      V143      V144
Length "4025"      "4025"      "4025"      "4025"      "4025"      "4025"
Class  "character" "character" "character" "character" "character" "character" "character"
"
Mode    "character" "character" "character" "character" "character" "character" "character"
"
      V145      V146      V147      V148      V149      V150
Length "4025"      "4025"      "4025"      "4025"      "4025"      "4025"
Class  "character" "character" "character" "character" "character" "character" "character"
"
Mode    "character" "character" "character" "character" "character" "character" "character"
"
      V151      V152
Length "4025"      "4025"
Class  "character" "character"
Mode    "character" "character"
> df[] <- lapply(df, function(x) as.numeric(as.character(x)))
There were 50 or more warnings (use warnings() to see the first 50)
> # KMeans - comes from Rcmdr library
> # kmeans- from amap library
> # kmeans- from stats library
>
> # steps in k-means clustering
> #1- preprocessing the data (impute missing values, remove outliers, feature
transformation)
> #2- scaling or standardization of data set
> #3- decide the number of clusters (value of k)
> #4- iterate over the samples to create clusters
> #5- decide the distance measure
> #6- calculate the group accuracy
>
> # scaling of data
> df_train1 <- scale(df_train)
Error in colMeans(x, na.rm = TRUE) : 'x' must be numeric
>
> head(df_train1)
      V2      V3      V4      V5      V6      V7

```

```

21  1.27299870 -0.6409610 -0.37105619 -1.03850116  1.9078725  1.0812457
111 -1.89747567  1.2570894 -2.16243759  0.02067199  0.5274917  1.3655732
108 -0.36101501 -0.5330150 -0.37105619  0.92853470 -1.1434956 -1.4777024
110 -1.71456368 -0.8928350  1.26455292  0.17198244 -0.4169794  0.6705503
150  0.07797375  1.4010174 -0.05951160  0.62591379  0.9634014 -1.4303145
177  0.18772094  0.2226069 -0.02056852  0.17198244  1.4719628 -1.0512111
      v8      v9      v10      v11      v12      v13
21  1.1479385 -1.0099319  0.8615852  0.1924509  0.60908324  1.5417428
111  0.5540868 -1.0099319  3.4201656 -1.0084453 -0.83737711  0.2967363
108 -0.2712665  0.9086010  0.0490630 -0.8337695 -0.28431874 -0.2427665
110  0.8963064 -0.6102376  1.5358057 -1.1176177  0.05602487  0.9192395
150 -0.6436820 -0.2105432 -0.7980346  1.8300367 -1.60315024 -1.7506077
177 -1.3583172  1.3082953 -0.2448280  1.7863678 -1.47552138 -1.3494389
      v14
21  0.07490999
111 -0.62907105
108 -0.86803709
110 -0.24801710
150 -0.66782230
177  0.26866624
>
> class(df_train1)
[1] "matrix"
> # screeplot approach to decide the number of clusters
> km = kmeans(df_train1,1)
> km$withinss
[1] 1833
> km$tot.withinss
[1] 1833
>
> km = kmeans(df_train1,2)
> km$withinss
[1] 685.8406 600.9150
> km$tot.withinss
[1] 1286.756
>
> km = kmeans(df_train1,3)
> km$withinss
[1] 346.0231 298.1925 336.7207
> km$tot.withinss
[1] 980.9363
>
> km = kmeans(df_train1,4)
> km$withinss
[1] 275.1233 210.9994 209.2629 216.7761
> km$tot.withinss
[1] 912.1616
>
> km = kmeans(df_train1,5)
> km$withinss
[1] 176.7833 119.2848 138.2213 309.2322 112.0491
> km$tot.withinss
[1] 855.5707
>
> km = kmeans(df_train1,6)
> km$withinss
[1] 172.84625 114.91540 137.94960 143.94848  94.49965 149.72160

```

```

> km$tot.withinss
[1] 813.881
>
> km = kmeans(df_train1,7)
> km$withinss
[1] 74.27123 78.06204 79.70444 58.01375 81.96519 298.19248 96.74012
> km$tot.withinss
[1] 766.9492
>
> km = kmeans(df_train1,8)
> km$withinss
[1] 33.42481 196.45068 124.67682 58.35021 126.40989 34.71128 77.80059
[8] 70.22846
> km$tot.withinss
[1] 722.0527
>
> km = kmeans(df_train1,9)
> km$withinss
[1] 44.271295 78.062037 110.262048 107.569538 4.597058 47.484979 138.221
277
[8] 37.955215 104.487311
> km$tot.withinss
[1] 672.9108
>
> km = kmeans(df_train1,10)
> km$withinss
[1] 50.42743 57.36509 74.27123 52.76306 39.50401 81.96519 233.91216
[8] 23.12942 30.22351 33.42481
> km$tot.withinss
[1] 676.9859
> dev.off()
null device
1
> sumsq=NULL
> for (i in 1:10)
+   sumsq[i] = sum(kmeans(df_train,centers=i,
+                       iter.max = 1000,
+                       nstart=i,
+                       algorithm='Forgy')$withinss)
Error in do_one(nmeth) : NA/NaN/Inf in foreign function call (arg 1)
In addition: warning message:
In storage.mode(x) <- "double" : NAs introduced by coercion
> plot(1:10,sumsq,type='b', main='Screeplot showing within group sum of squares')
> km = kmeans(df_train1,3)
> km$withinss
[1] 346.0231 336.7207 298.1925
> km$tot.withinss
[1] 980.9363
>
> class(km$cluster)
[1] "integer"
>
> summary(km)

```

	Length	Class	Mode
cluster	142	-none-	numeric
centers	39	-none-	numeric



```

totss      1      -none- numeric
withinss   3      -none- numeric
tot.withinss 1      -none- numeric
betweenss  1      -none- numeric
size       3      -none- numeric
iter       1      -none- numeric
ifault     1      -none- numeric
>
> km$centers
      V2      V3      V4      V5      V6      V7
1 -1.0688290 -0.3745716 -0.45159755  0.2834020 -0.69107412  0.03476227
2  0.7873041 -0.3366416  0.30126292 -0.6685043  0.63578373  0.81778071
3  0.1178079  0.7627367  0.08674128  0.5102454 -0.07309503 -0.99715372
      V8      V9      V10      V11      V12      V13
1  0.1176195 -0.02704718  0.05220622 -0.94949223  0.4882226  0.4014300
2  0.9257427 -0.57705540  0.59737684  0.08451446  0.5280108  0.7409713
3 -1.2053248  0.70608916 -0.75462324  0.82885315 -1.0992526 -1.2652089
      V14
1 -0.8160752
2  1.0661743
3 -0.4577762
>
> as.numeric(km$cluster)
[1] 2 1 1 1 3 3 2 2 1 1 1 1 2 3 2 3 3 2 2 2 2 3 2 2 2 1 1 3 3 3 1 2 2 2 2 1
2
[38] 3 3 1 1 1 3 3 3 1 1 1 2 1 2 3 1 3 2 3 1 1 3 1 1 2 3 3 2 3 2 2 2 1 2 2 1
1
[75] 1 2 1 2 2 3 1 2 3 2 1 3 2 3 2 3 3 2 2 1 2 3 3 2 1 2 1 2 2 2 1 2 1 3 2 1
2
[112] 3 1 3 1 2 3 2 1 3 3 2 2 1 3 1 1 2 3 3 3 3 3 2 2 1 3 2 3 3 3 1
>
> length(km$cluster)
[1] 142
>
> dim(df_train)
[1] 3220 151
>
> class(df_train)
[1] "data.frame"
>
> df_train$c1 <- km$cluster
Error in `.$<-.data.frame`(`*tmp*`, c1, value = c(`21` = 2L, `111` = 1L, :
replacement has 142 rows, data has 3220
>
> head(df_train)
      V2      V3      V4 V5 V6      V7      V8      V9 V10      V11
458 1323084232 976322 5/12/2011 11:23 no 12 1.52 8.1 -94.3 3 -1.03566
2505 1323094981 340386 5/12/2011 14:23 no 57 124 26.1 -2.1 19 -1.03566
2452 1323094980 324269 5/12/2011 14:23 no 56 123 26.4 -3.24 19 -1.03566
2508 1323094981 352685 5/12/2011 14:23 no 57 124 26.1 -2.1 20 -1.03566
3462 1323095007 104321 5/12/2011 14:23 no 79 123 26.7 -3.73 20 -1.03566
2575 1323094982 972320 5/12/2011 14:23 no 58 123 26.3 -3.51 20 -1.03566
      V12      V13      V14 V15 V16 V17      V18 V19 V20      V21 V22 V23 V24
458 -0.39133 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3
2505 -0.39133 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3
2452 -0.39133 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3
2508 -0.39133 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3

```

[illegible]

2575 8.45 77.25 38 -33.6 -58.6 10 36.945 121.5 27 -6.488101883 30.7383386  
6

	v83	v84	v85	v86	v87	v88	v89	v90	v91	v92
458	-85.05658246	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2
2505	125.8836028	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2
2452	130.2009178	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2
2508	118.1308197	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2
3462	126.7219718	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2
2575	125.6198141	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2

	v93	v94	v95	v96	v97	v98	v99	v100	v101	v102	v103
458	-0.1	55.71	54.74	0	37	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
2505	-0.1	55.71	54.74	0	8	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
2452	-0.1	55.71	54.74	0	9	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
2508	-0.1	55.71	54.74	0	9	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
3462	-0.1	55.71	54.74	0	9	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
2575	-0.1	55.71	54.74	0	9	2.41635	-5.11805	17.058	291.001	13.9312	14.1062

	v104	v105	v106	v107	v108	v109	v110	v111	v112	v113	v114
458	199.0775	64.7063	13.5747	184.5578	0	-0.02	0	-234	47	-272	-553
2505	199.0775	64.7063	13.5747	184.5578	0.55	-0.05	-0.23	12	-21	75	534
2452	199.0775	64.7063	13.5747	184.5578	0.42	-0.08	-0.56	24	-5	89	513
2508	199.0775	64.7063	13.5747	184.5578	0.56	-0.13	-0.23	32	1	78	531
3462	199.0775	64.7063	13.5747	184.5578	0.32	-0.47	0.28	15	-22	85	513
2575	199.0775	64.7063	13.5747	184.5578	0.42	-0.08	-0.52	28	-6	88	524

	v115	v116	v117	v118	v119	v120	v121	v122	v123	v124	v125
458	300	-68	27	-63.8	-151	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
2505	-517	-92	112	65.2	116	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
2452	-534	-63	88	46.2	83.6	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
2508	-518	-88	125	73.7	130	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
3462	-532	-67	140	48.6	164	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
2575	-522	-64	89.4	18.5	73.4	-1.09475	-0.97525	-0.05065	0.17285	49.6	168

	v126	v127	v128	v129	v130	v131	v132	v133	v134	v135	v136
458	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	36	14.0772	27.85936	45.16342
2505	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	35	14.0772	27.85936	45.16342
2452	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	37	14.0772	27.85936	45.16342
2508	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	32	14.0772	27.85936	45.16342
3462	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	27	14.0772	27.85936	45.16342
2575	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	33	14.0772	27.85936	45.16342

	v137	v138	v139	v140	v141	v142	v143	v144	v1
--	------	------	------	------	------	------	------	------	----

45  
458 2749.163 25.35597 8.906695 79.33451 17.09505 74.27584 5541.956 0.03 -0.05  
05  
2505 2749.163 25.35597 8.906695 79.33451 17.09505 74.27584 5541.956 0.47 -2.47  
47  
2452 2749.163 25.35597 8.906695 79.33451 17.09505 74.27584 5541.956 -0.53 2.81  
81  
2508 2749.163 25.35597 8.906695 79.33451 17.09505 74.27584 5541.956 0.16 -2.39  
39  
3462 2749.163 25.35597 8.906695 79.33451 17.09505 74.27584 5541.956 0.51 -3.53  
53  
2575 2749.163 25.35597 8.906695 79.33451 17.09505 74.27584 5541.956 -0.39 1.4  
.4

	v146	v147	v148	v149	v150	v151	v152
458	-0.02	190	202	-213	-18	662	462
2505	-0.3	-88	283	-178	-654	499	491
2452	0.57	-6	335	-140	-556	713	680
2508	-0.44	-120	230	-176	-687	399	476
3462	-1.1	-118	174	-155	-707	214	558

```

2575  0.05  110  269 -147 -308  790  709
> # profiles of clusters
> aggregate(df_train[,1:8],list(df_train[,9]),mean)
  Group.1 v2 v3 v4 v5 v6 v7 v8 v9
1         0 NA NA NA NA NA NA NA NA
2         1 NA NA NA NA NA NA NA NA
3        10 NA NA NA NA NA NA NA NA
4        11 NA NA NA NA NA NA NA NA
5        12 NA NA NA NA NA NA NA NA
6        13 NA NA NA NA NA NA NA NA
7        14 NA NA NA NA NA NA NA NA
8        15 NA NA NA NA NA NA NA NA
9        16 NA NA NA NA NA NA NA NA
10       17 NA NA NA NA NA NA NA NA
11       18 NA NA NA NA NA NA NA NA
12       19 NA NA NA NA NA NA NA NA
13         2 NA NA NA NA NA NA NA NA
14       20 NA NA NA NA NA NA NA NA
15       21 NA NA NA NA NA NA NA NA
16       22 NA NA NA NA NA NA NA NA
17       23 NA NA NA NA NA NA NA NA
18       24 NA NA NA NA NA NA NA NA
19       25 NA NA NA NA NA NA NA NA
20       26 NA NA NA NA NA NA NA NA
21         3 NA NA NA NA NA NA NA NA
22         4 NA NA NA NA NA NA NA NA
23         5 NA NA NA NA NA NA NA NA
24         6 NA NA NA NA NA NA NA NA
25         7 NA NA NA NA NA NA NA NA
26         8 NA NA NA NA NA NA NA NA
27         9 NA NA NA NA NA NA NA NA
There were 50 or more warnings (use warnings() to see the first 50)
>
> table(df$v1)
< table of extent 0 >
>
> library(cluster)
>
> clusplot(df_train,df_train$c1,cex=0.9,color=T,shade=T, labels=4,lines=0)
Error in is.list(s.x.2d) :
cannot use 'cor = TRUE' with a constant variable
> #HC clustering or Hierarchical Clustering
> # distance (euclidean, manhattan, cosine distance)
>
> # Divisive method (top down)
> # Agglomerative method (bottom up)
>
>
>
> df_train = df_train[,-9]
> head(df_train)
      v2      v3      v4 v5 v6  v7  v8  v9      v11      v1
2
458 1323084232 976322 5/12/2011 11:23 no 12 1.52  8.1 -94.3 -1.03566 -0.3913
3
2505 1323094981 340386 5/12/2011 14:23 no 57 124 26.1 -2.1 -1.03566 -0.3913
3

```

2452 1323094980 324269 5/12/2011 14:23 no 56 123 26.4 -3.24 -1.03566 -0.3913

3

2508 1323094981 352685 5/12/2011 14:23 no 57 124 26.1 -2.1 -1.03566 -0.3913

3

3462 1323095007 104321 5/12/2011 14:23 no 79 123 26.7 -3.73 -1.03566 -0.3913

3

2575 1323094982 972320 5/12/2011 14:23 no 58 123 26.3 -3.51 -1.03566 -0.3913

3

V13 V14 V15 V16 V17 V18 V19 V20 V21 V22 V23 V24 V25 V26

458 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3 121.9 0.6

2505 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3 121.9 0.6

2452 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3 121.9 0.6

2508 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3 121.9 0.6

3462 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3 121.9 0.6

2575 0.005406 0.045115 -4.1 20 -1 -7.25 18 -1 1.345 2 0 0.3 121.9 0.6

V27 V28 V29 V30 V31 V32 V33 V34 V35 V36 V37 V38 V39 V40 V4

1

458 0.35 25.75 0.35 0.1 -4.95 0.4 0.17 0.02 0 -0.02 -21 3 22 -9 59

9

2505 0.35 25.75 0.35 0.1 -4.95 0.4 0.17 -0.43 -0.05 -0.44 -42 69 -171 1 57

9

2452 0.35 25.75 0.35 0.1 -4.95 0.4 0.17 -0.47 -0.03 -0.46 -33 68 -169 1 58

2

2508 0.35 25.75 0.35 0.1 -4.95 0.4 0.17 -0.43 -0.03 -0.46 -44 71 -172 2 57

9

3462 0.35 25.75 0.35 0.1 -4.95 0.4 0.17 -0.59 0 -0.56 -45 70 -177 -9 58

3

2575 0.35 25.75 0.35 0.1 -4.95 0.4 0.17 -0.37 -0.03 -0.41 -43 70 -174 1 58

7

V42 V43 V44 V45 V46 V47 V48 V49 V50 V51

458 -317 -129 20.7 -161 34 65.0977 76.22175 16.1039 259.3599 -10.1695

2505 -366 118 -23.4 90.1 17 65.0977 76.22175 16.1039 259.3599 -10.1695

2452 -365 139 -24.7 62.3 24 65.0977 76.22175 16.1039 259.3599 -10.1695

2508 -378 120 -22.1 79.6 18 65.0977 76.22175 16.1039 259.3599 -10.1695

3462 -372 102 -27.8 110 15 65.0977 76.22175 16.1039 259.3599 -10.1695

2575 -377 146 -1.77 72.3 23 65.0977 76.22175 16.1039 259.3599 -10.1695

V52 V53 V54 V55 V56 V57 V58 V59 V60 V61 V

62

458 10.66725 113.7978 19.0615 35.8809 1287.463 -0.02 0 -0.02 -289 109 -1

23

2505 10.66725 113.7978 19.0615 35.8809 1287.463 -2.28 0.88 -0.38 76 46 1

42

2452 10.66725 113.7978 19.0615 35.8809 1287.463 1.67 -1.01 0.11 176 61 1

40

2508 10.66725 113.7978 19.0615 35.8809 1287.463 -2.52 0.9 -0.43 102 37 1

44

3462 10.66725 113.7978 19.0615 35.8809 1287.463 -2.57 0.98 0.02 24 44 1

37

2575 10.66725 113.7978 19.0615 35.8809 1287.463 1.33 -1.01 -0.18 161 67 1

38

V63 V64 V65 V66 V67 V68 V69 V70 V71 V72

458 -373 335 511 -1.18224 -0.96912 -0.86977 0.12353 -0.10319 0.059765 8.45

2505 8 349 575 -1.18224 -0.96912 -0.86977 0.12353 -0.10319 0.059765 8.45

2452 305 341 490 -1.18224 -0.96912 -0.86977 0.12353 -0.10319 0.059765 8.45

2508 139 310 578 -1.18224 -0.96912 -0.86977 0.12353 -0.10319 0.059765 8.45

3462 -165 378 538 -1.18224 -0.96912 -0.86977 0.12353 -0.10319 0.059765 8.45

2575 151 391 512 -1.18224 -0.96912 -0.86977 0.12353 -0.10319 0.059765 8.45

	V73	V74	V75	V76	V77	V78	V79	V80		V81	V82
458	77.25	38	-33.6	-58.6	10	36.945	121.5	27	13.02493063	-70.31756075	
2505	77.25	38	-33.6	-58.6	10	36.945	121.5	27	-26.97447709	15.28721363	
2452	77.25	38	-33.6	-58.6	10	36.945	121.5	27	-5.418908377	26.30018238	
2508	77.25	38	-33.6	-58.6	10	36.945	121.5	27	1.186057995	38.92878663	
3462	77.25	38	-33.6	-58.6	10	36.945	121.5	27	-24.95711462	16.92074714	
2575	77.25	38	-33.6	-58.6	10	36.945	121.5	27	-6.488101883	30.73833866	
	V83	V84	V85	V86	V87	V88	V89	V90	V91	V92	
458	-85.05658246	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2	
2505	125.8836028	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2	
2452	130.2009178	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2	
2508	118.1308197	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2	
3462	126.7219718	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2	
2575	125.6198141	-0.09595	-0.4422	0.0819	-0.216	41.85	133	-0.1	-26.75	20.2	
	V93	V94	V95	V96	V97	V98	V99	V100	V101	V102	V103
458	-0.1	55.71	54.74	0	37	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
2505	-0.1	55.71	54.74	0	8	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
2452	-0.1	55.71	54.74	0	9	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
2508	-0.1	55.71	54.74	0	9	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
3462	-0.1	55.71	54.74	0	9	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
2575	-0.1	55.71	54.74	0	9	2.41635	-5.11805	17.058	291.001	13.9312	14.1062
	V104	V105	V106	V107	V108	V109	V110	V111	V112	V113	V114
458	199.0775	64.7063	13.5747	184.5578	0	-0.02	0	-234	47	-272	-553
2505	199.0775	64.7063	13.5747	184.5578	0.55	-0.05	-0.23	12	-21	75	534
2452	199.0775	64.7063	13.5747	184.5578	0.42	-0.08	-0.56	24	-5	89	513
2508	199.0775	64.7063	13.5747	184.5578	0.56	-0.13	-0.23	32	1	78	531
3462	199.0775	64.7063	13.5747	184.5578	0.32	-0.47	0.28	15	-22	85	513
2575	199.0775	64.7063	13.5747	184.5578	0.42	-0.08	-0.52	28	-6	88	524
	V115	V116	V117	V118	V119	V120	V121	V122	V123	V124	V125
458	300	-68	27	-63.8	-151	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
2505	-517	-92	112	65.2	116	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
2452	-534	-63	88	46.2	83.6	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
2508	-518	-88	125	73.7	130	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
3462	-532	-67	140	48.6	164	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
2575	-522	-64	89.4	18.5	73.4	-1.09475	-0.97525	-0.05065	0.17285	49.6	168
	V126	V127	V128	V129	V130	V131	V132	V133	V134	V135	V136
458	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	36	14.0772	27.85936	45.16342
2505	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	35	14.0772	27.85936	45.16342
2452	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	37	14.0772	27.85936	45.16342
2508	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	32	14.0772	27.85936	45.16342
3462	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	27	14.0772	27.85936	45.16342
2575	-1.1	4.65	-168.5	-1.1	32.2	341.5	0	33	14.0772	27.85936	45.16342
	V137	V138	V139	V140	V141	V142	V143	V144	V1		
45											
458	2749.163	25.35597	8.906695	79.33451	17.09505	74.27584	5541.956	0.03	-0.		
05											
2505	2749.163	25.35597	8.906695	79.33451	17.09505	74.27584	5541.956	0.47	-2.		
47											
2452	2749.163	25.35597	8.906695	79.33451	17.09505	74.27584	5541.956	-0.53	2.		
81											
2508	2749.163	25.35597	8.906695	79.33451	17.09505	74.27584	5541.956	0.16	-2.		
39											
3462	2749.163	25.35597	8.906695	79.33451	17.09505	74.27584	5541.956	0.51	-3.		
53											
2575	2749.163	25.35597	8.906695	79.33451	17.09505	74.27584	5541.956	-0.39	1		
.4											
	V146	V147	V148	V149	V150	V151	V152				

```

458 -0.02 190 202 -213 -18 662 462
2505 -0.3 -88 283 -178 -654 499 491
2452 0.57 -6 335 -140 -556 713 680
2508 -0.44 -120 230 -176 -687 399 476
3462 -1.1 -118 174 -155 -707 214 558
2575 0.05 110 269 -147 -308 790 709
>
> # compute the distance metrix
> d1 <- dist(df_train,method='euclidean')
Warning message:
In dist(df_train, method = "euclidean") : NAs introduced by coercion
> summary(d1)
      Min.   1st Qu.   Median     Mean   3rd Qu.    Max.
 16.5 169114.5 330259.8 365890.1 543588.0 1164944.1
> # HC
> fit <- hclust(d1,method = 'ward.D2')
> plot(fit)
>
> # single, double, average, ward, ward.D2
>
> # agglomerative method
> fit <- agnes(d1,metric='euclidean',method = 'ward')
> plot(fit)
Hit <Return> to see next plot:
Hit <Return> to see next plot: # divisive method
> fit <- diana(d1,metric='euclidean')
> plot(fit)
Hit <Return> to see next plot:
Hit <Return> to see next plot:

```

```
library(ggdendro)
```

```
if(require(cluster)){
```

```
  fit<- agnes(d1, metric = "manhattan", stand = TRUE)
```

```
  dg <- as.dendrogram(fit)
```

```
  ggdendrogram(dg)
```

```
  fit <- diana(d1, metric = "manhattan", stand = TRUE)
```

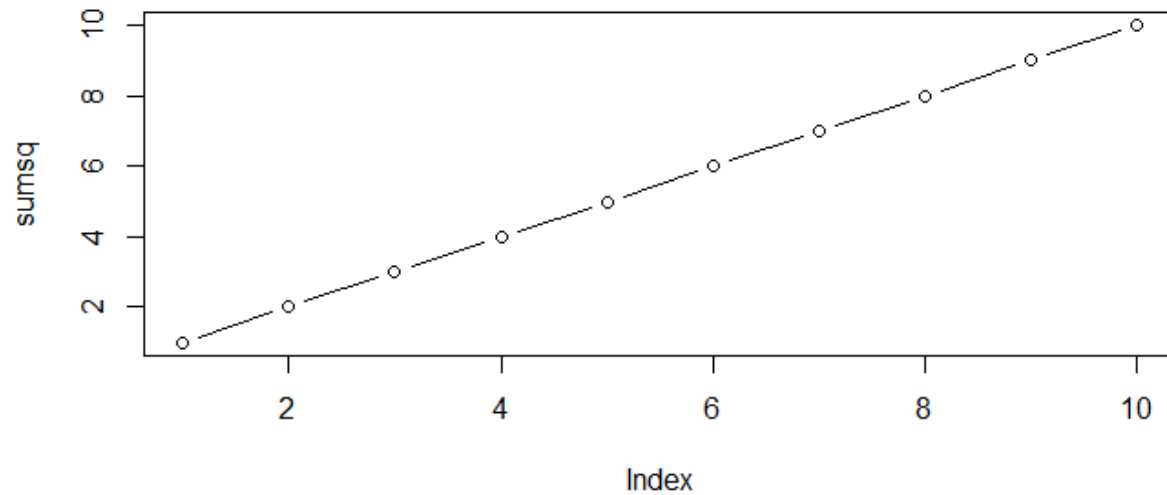
```
  dg <- as.dendrogram(fit)
```

```
  ggdendrogram(dg)
```

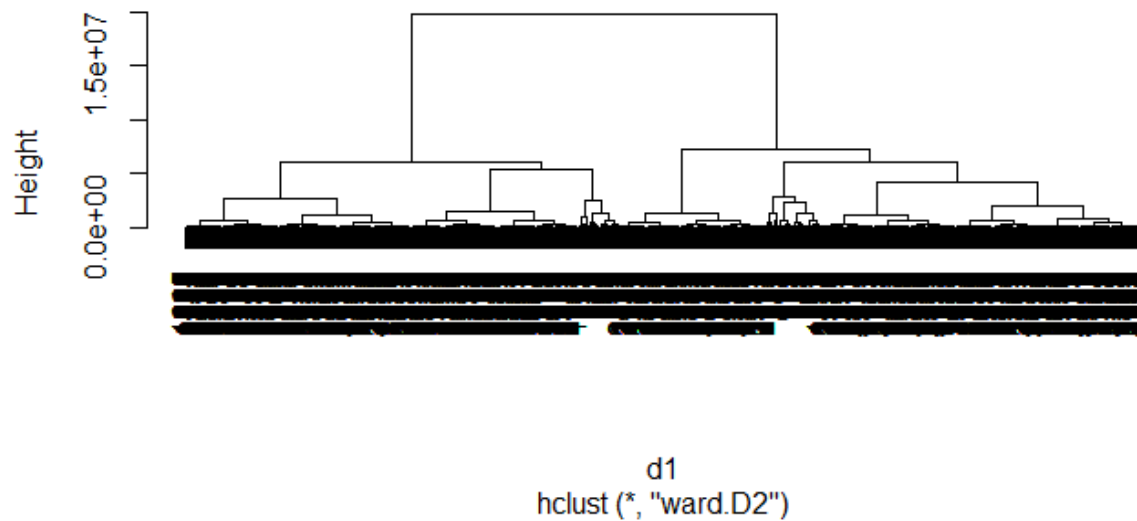
```
}
```

A **Scree Plot** is a simple line segment **plot** that shows the fraction of total variance in the data as explained or represented by each PC. The PCs are ordered, and by definition are therefore assigned a number label, by decreasing order of contribution to total variance.

**Screeplot showing within group sum of squares**

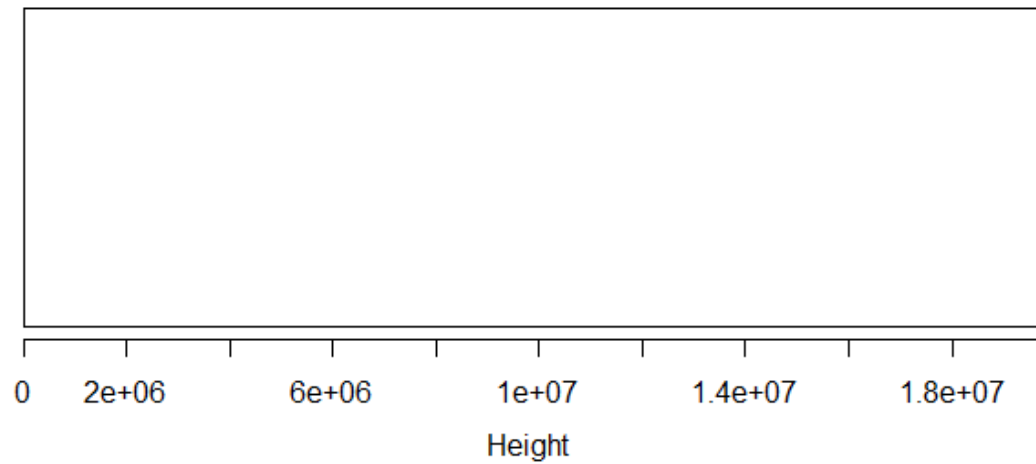


**Cluster Dendrogram**





**Banner of `agnes(x = d1, metric = "euclidean", method = "ward")`**



Agglomerative Coefficient = 1

Conclusion:-

The term *cluster analysis* encompasses a number of different algorithms and methods for grouping objects of similar kind into respective categories. In other words cluster analysis is an exploratory data analysis tool which aims at sorting different objects into groups in a way that the degree of association between two objects is maximal if they belong to the same group and minimal otherwise. Given the above, cluster analysis can be used to discover structures in data without providing an explanation/interpretation. In other words, cluster analysis simply discovers structures in data without explaining why they exist.

## ***k*-Means Clustering**

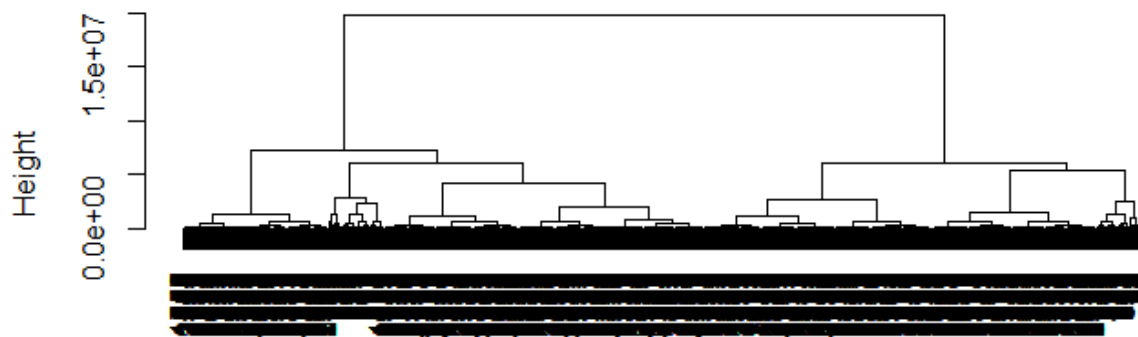
In general, the *k*-means method will produce exactly *k* different clusters of greatest possible distinction. It should be mentioned that the best number of clusters *k* leading to the greatest separation (distance) is not known as *a priori* and must be computed from the data

### **INTERPRETATION OF RESULTS**

Usually, as the result of a *k*-means clustering analysis, we would examine the means for each cluster on each dimension to assess how distinct our *k* clusters are. Ideally, we would obtain very different means for most, if not all dimensions, used in the analysis. The magnitude of the *F* values from the analysis of variance performed on each dimension is another indication of how well the respective dimension discriminates between clusters.

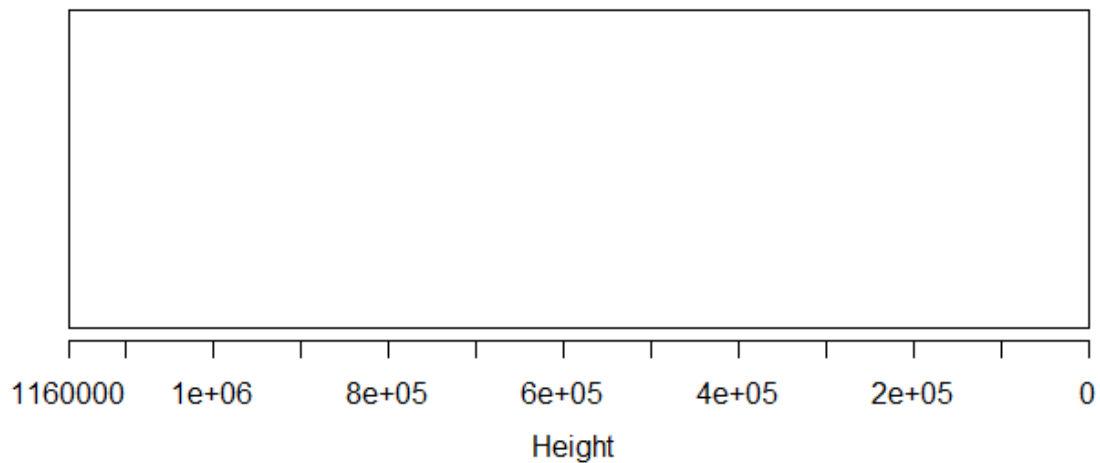
**Euclidean distance.** This is probably the most commonly chosen type of distance. It simply is the geometric distance in the multidimensional space.

### Dendrogram of `agnes(x = d1, metric = "euclidean", method = "ward")`

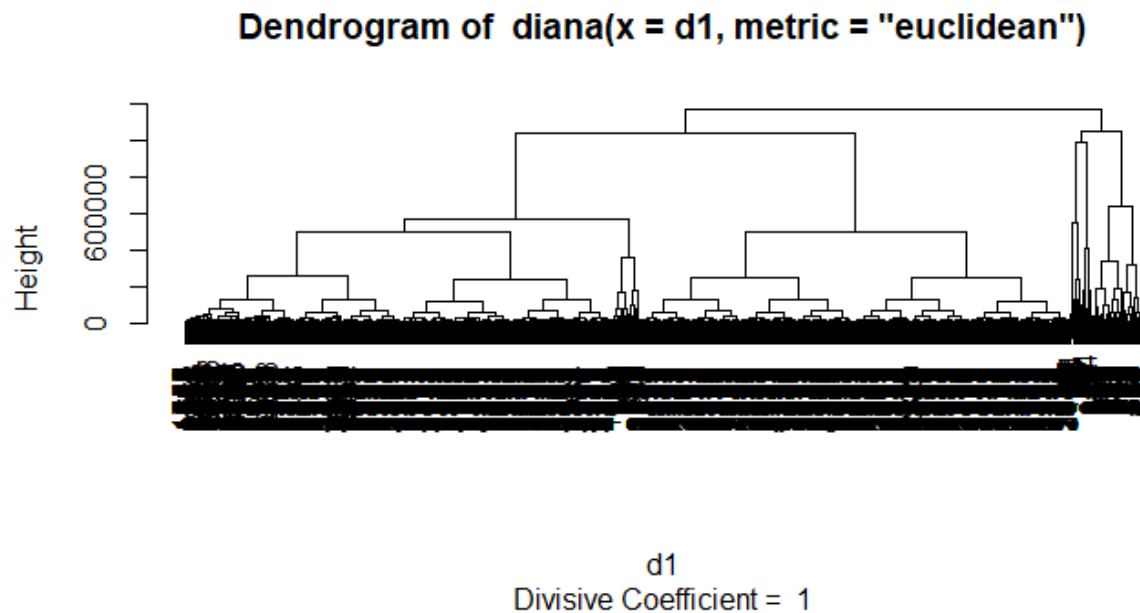


`d1`  
Agglomerative Coefficient = 1

### Banner of `diana(x = d1, metric = "euclidean")`



Divisive Coefficient = 1



A **dendrogram** or **tree diagram** allows to illustrate the **hierarchical organisation** of several entities. For example, we often use it to make family trees. It is constituted of a root node, which give birth to several nodes that end by giving leaf nodes (the

bottom of the tree). Dendrogram can be made with 2 types of dataset. **i/ a numeric matrix** where several variables describe the features of individuals. We can then calculate the distance between individuals and cluster them. **ii/ A hierarchical**

gg dendroplot given below provides a good output. As the data is large it needs lot of cleaning and removal of missing data

