

MVA_Ass5.R

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```
##### Assignment 5 #####
## Applying Cluster Analysis

#Getting working directory
getwd()

## [1] "/Users/mihikagupta/Desktop/SEM_2/MVA"

#Setting directory to load data set
setwd("/Users/mihikagupta/Desktop/SEM_2/MVA")

#Reading the data into a data frame
#df <- read.csv(file = 'US_Acc_June20.csv')
num <- read.csv(file = 'num.csv')
# Performing clustering on the first 500 records for now to achieve easy and quick results and test the
num<-num[1:500,]
attach(num)
# Printing first few columns of data set for inference
#head(df)

## Setting random seed to shuffle data before splitting
set.seed(23)

#Checking number of rows
#rows<-sample(nrow(df))

#Shuffling the data
#mva<-df[rows,]

#Taking the required number of instances from the shuffled data to reduce any biases
#mva<-mva[950000:1000000,]

#Checking the structure of the data set
#str(mva)

# Checking the number of rows and columns in the current uncleaned dataset
#ncol(mva)
#nrow(mva)

# Printing all the column names to find and filter the relevant and irrelevant attributes
#names<-names(mva)
#names
```

```
## DATA CLEANING ##

#Dropping the surplus attributes which do not contribute to the analysis
#mva <- mva[-c(1:3,7:10,13,14,19,21:23,33,47:49)]

#Checking for any null values in the present data set
# is.na(mva[,])

#Checking which rows have all the values filled and complete
# complete.cases(mva)

#Making a new dataframe with only the rows that have complete information and all values filled
#Mva<-na.omit(mva)
#Mva<-Mva[!(is.na(Mva$Sunrise_Sunset) | Mva$Sunrise_Sunset==""), ]
#Mva<- Mva[complete.cases(Mva),]
#Verifying for missing values in the new dataframe
#complete.cases(Mva)
#unique(Mva$Sunrise_Sunset)
#Checking the number of rows and columns in the new CLEANED dataframe
#ncol(Mva)
#nrow(Mva)

# Creating new dataframe with only the numerical attributes to perform statistical functions
#num<-Mva[,c(1,4,11:15,17,18)]
#write.csv(num,"/Users/mihikagupta/Desktop/SEM_2/MVA/num.csv", row.names = FALSE)
#Scaling the dataset before applying clustering algorithms
```

```
nrow(num)
```

```
## [1] 500
```

```
ncol(num)
```

```
## [1] 9
```

```
##### Checking to see for any correlation that might interfere with the data #####
cor(num)
```

```
##          Severity Distance.mi. Temperature.F. Wind_Chill.F.
## Severity      1.00000000  0.060325232   -0.026274516   -0.03165259
## Distance.mi.    0.06032523  1.000000000   -0.010229758   -0.01679232
## Temperature.F. -0.02627452 -0.010229758    1.000000000    0.99372259
## Wind_Chill.F.  -0.03165259 -0.016792324    0.993722592    1.00000000
## Humidity...     0.08687654  0.004473866   -0.452662114   -0.43537980
## Pressure.in.    0.06567115 -0.008218439    0.055390478    0.05040739
## Visibility.mi.  -0.01784787 -0.041477419    0.256766004    0.26208308
## Wind_Speed.mph. -0.06205678  0.009728557    0.004761887   -0.05069851
## Precipitation.in. 0.03617412  0.101174144   -0.104794100   -0.10133502
## Humidity...     0.086876542  0.065671149   -0.01784787   -0.062056779
## Severity      0.086876542  0.065671149   -0.01784787   -0.062056779
## Distance.mi.    0.004473866 -0.008218439   -0.04147742    0.009728557
## Temperature.F. -0.452662114  0.055390478    0.25676600    0.004761887
## Wind_Chill.F.  -0.435379799  0.050407389    0.26208308   -0.050698506
## Humidity...     1.000000000  0.233181450   -0.39892391   -0.137092830
## Pressure.in.    0.233181450  1.000000000   -0.26428984   -0.014390090
## Visibility.mi.  -0.398923914 -0.264289840    1.00000000    0.014696688
```

```
## Wind_Speed.mph.    -0.137092830 -0.014390090      0.01469669      1.000000000
## Precipitation.in.  0.244133880  0.007382003     -0.24473243      0.020686860
##                    Precipitation.in.
## Severity           0.036174121
## Distance.mi.       0.101174144
## Temperature.F.     -0.104794100
## Wind_Chill.F.      -0.101335015
## Humidity...        0.244133880
## Pressure.in.       0.007382003
## Visibility.mi.     -0.244732428
## Wind_Speed.mph.    0.020686860
## Precipitation.in.  1.000000000
```

```
# Removing the wind chills attribute #
# since "wind chills had a high correlation with temperature, it is removed
num<-num[-c(4)]
```

```
# Checking the new dimensions
nrow(num)
```

```
## [1] 500
```

```
ncol(num)
```

```
## [1] 8
```

```
# Scaling the new data set for better accuracies in clustering
num<-scale(num)
```

```
# Determine number of clusters
```

```
wss <- (nrow(num)-1)*sum(apply(num,2,var))
```

```
for (i in 2:15) wss[i] <- sum(kmeans(num,
                                     centers=i)$withinss)
```

```
plot(wss)
```

```
#here we observe that after 8-10 clusters, the distance remains almost constant, so we fix this number
```

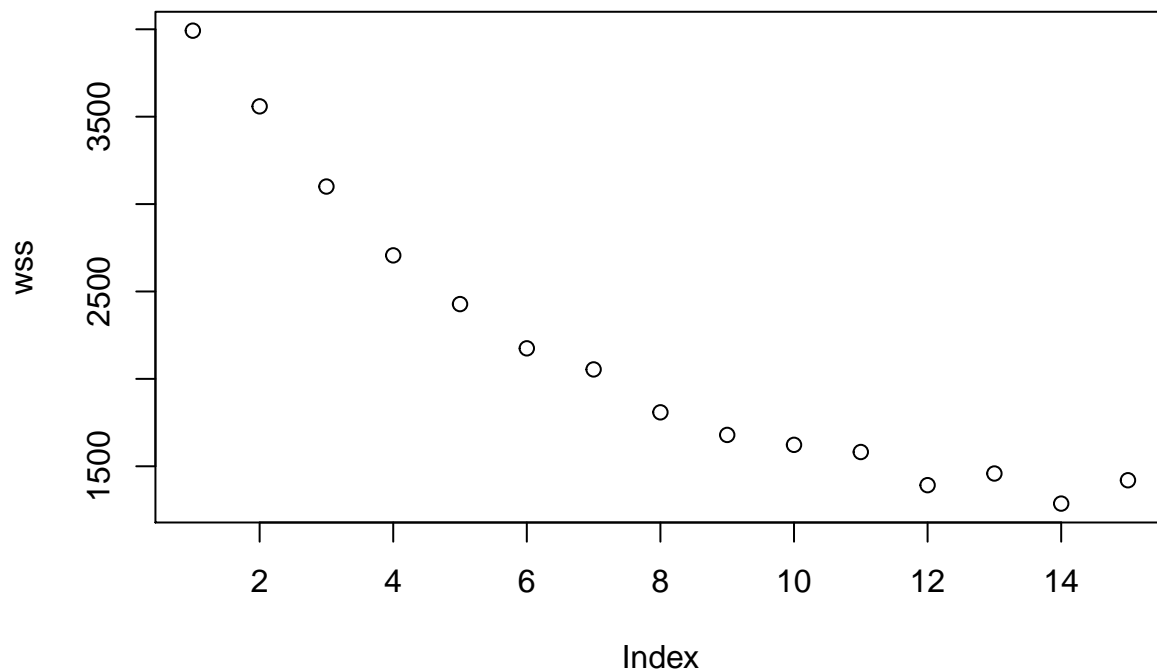
```
## Loading the cluster library ###
```

```
library(cluster)
```

```
library(factoextra)
```

```
## Loading required package: ggplot2
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```



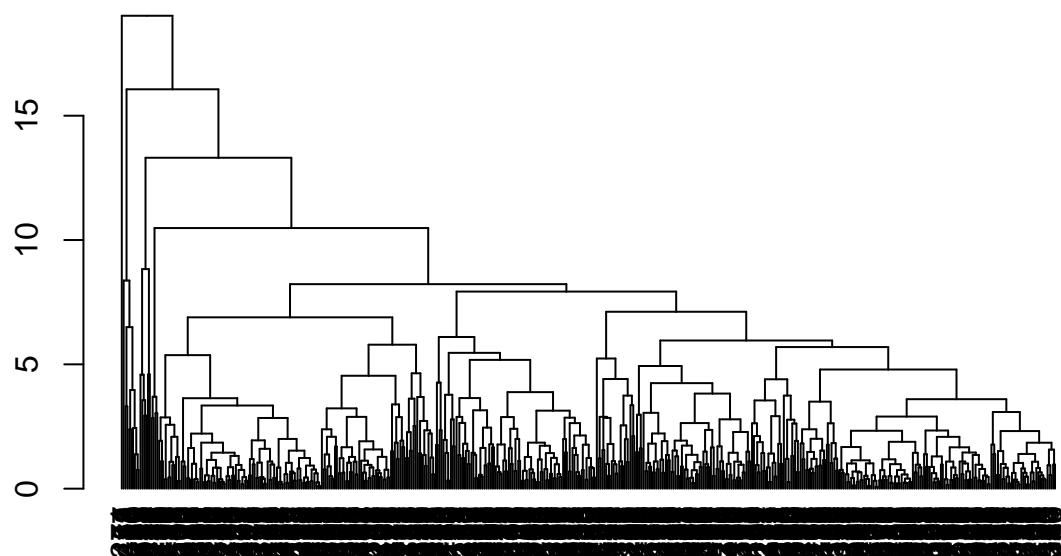
```
# Understanding different methods

# producing a distance matrix of the dataset
numdist<-dist(num,method = "euclidean")

#Sliced dendrograms for better visualizations

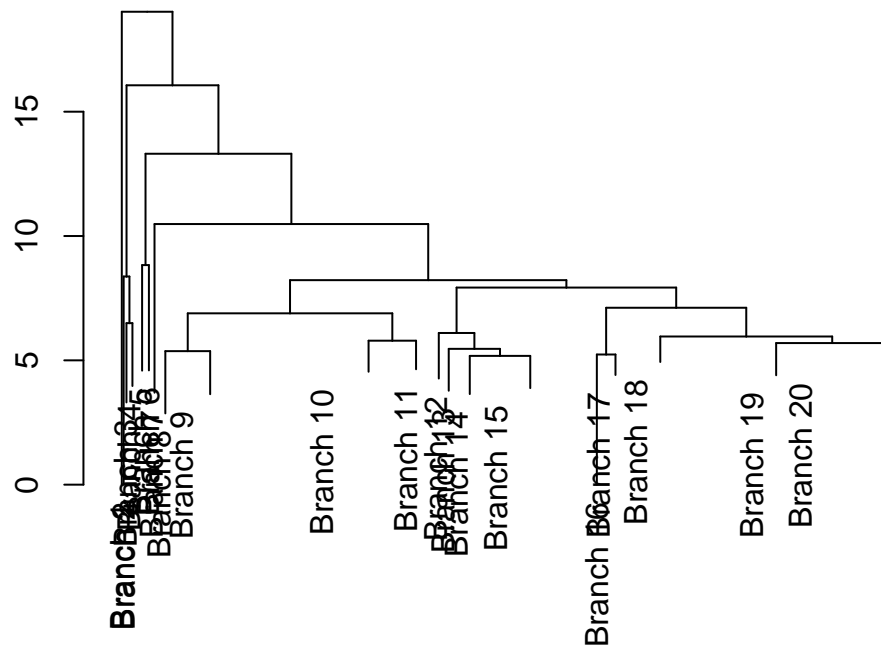
hc <- hclust(dist(num))
hcd <- as.dendrogram(hc)
plot(hcd, main="Main")
```

Main



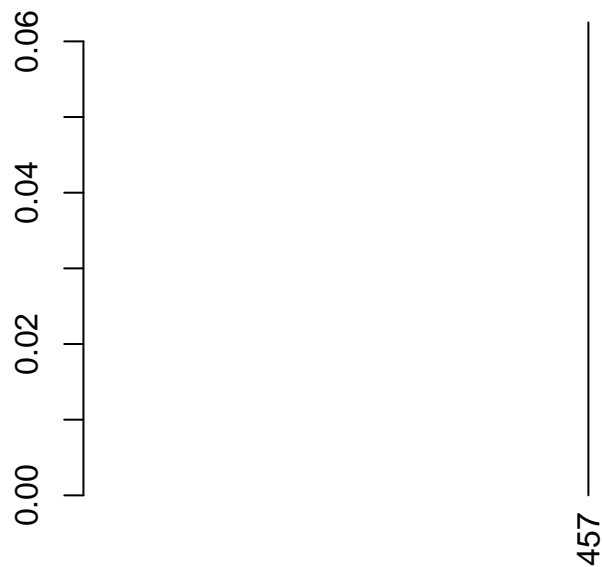
```
plot(cut(hcd, h=5)$upper,
     main="Upper tree of cut at h=5")
```

Upper tree of cut at h=5



```
plot(cut(hcd, h=5)$lower[[2]],
     main="Second branch of lower tree with cut at h=5")
```

Second branch of lower tree with cut at h=5

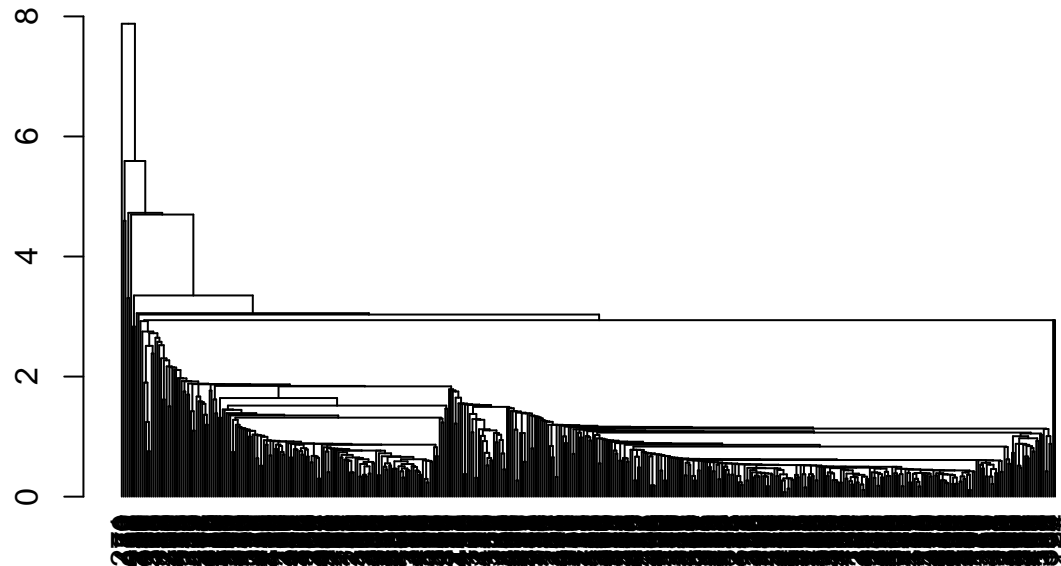


#Single

```
numsingle<- hclust(numdist, method = "single")
```

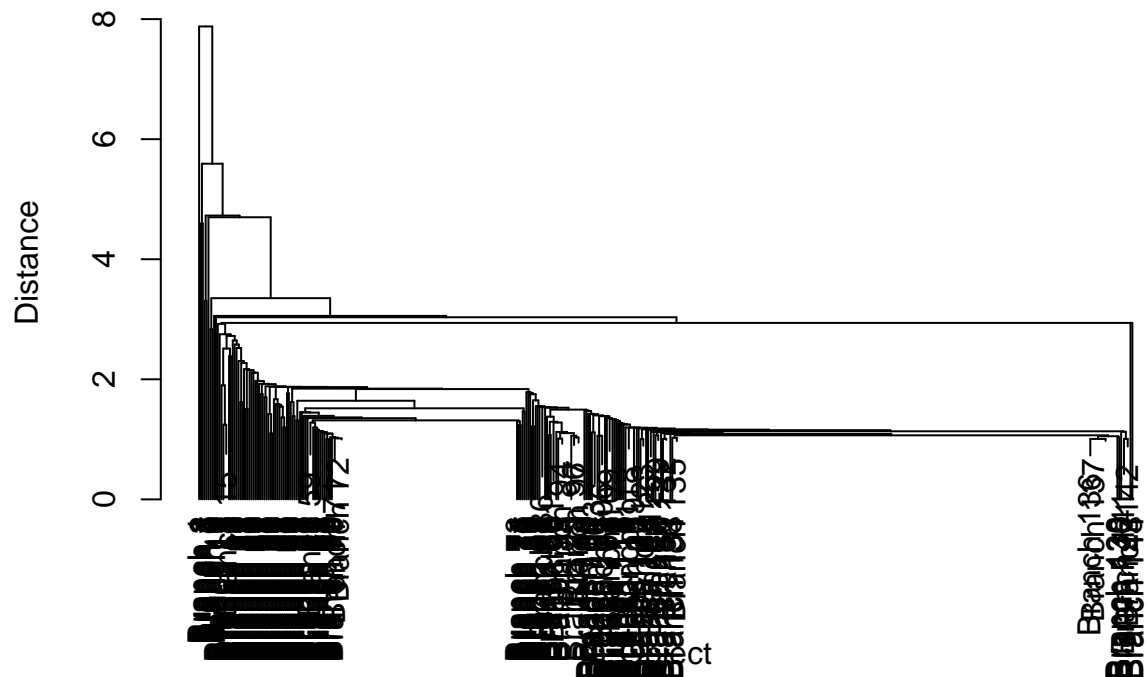
```
single<-as.dendrogram(numsingle)
plot(single,main="Main")
```

Main



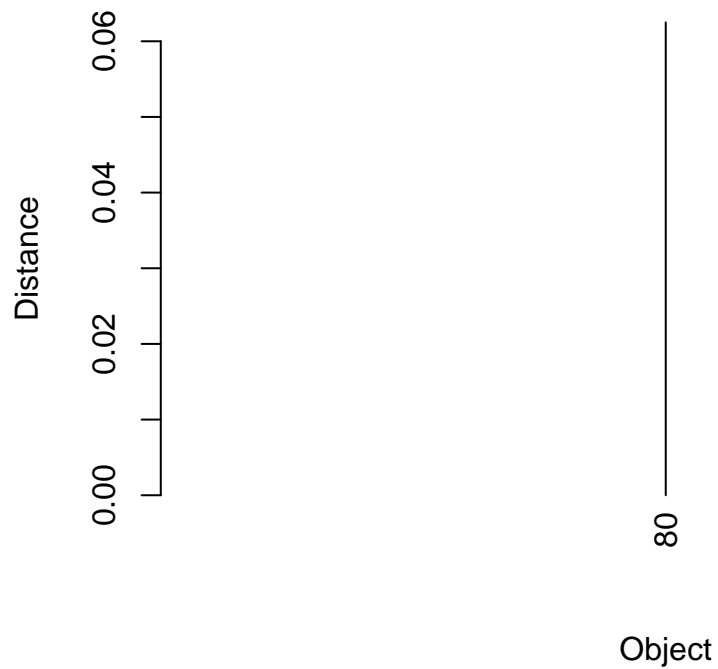
```
plot(cut(single,h=1)$upper,xlab="Object",ylab="Distance",
     main="Dendrogram. Nearest neighbor linkage")
```

Dendrogram. Nearest neighbor linkage



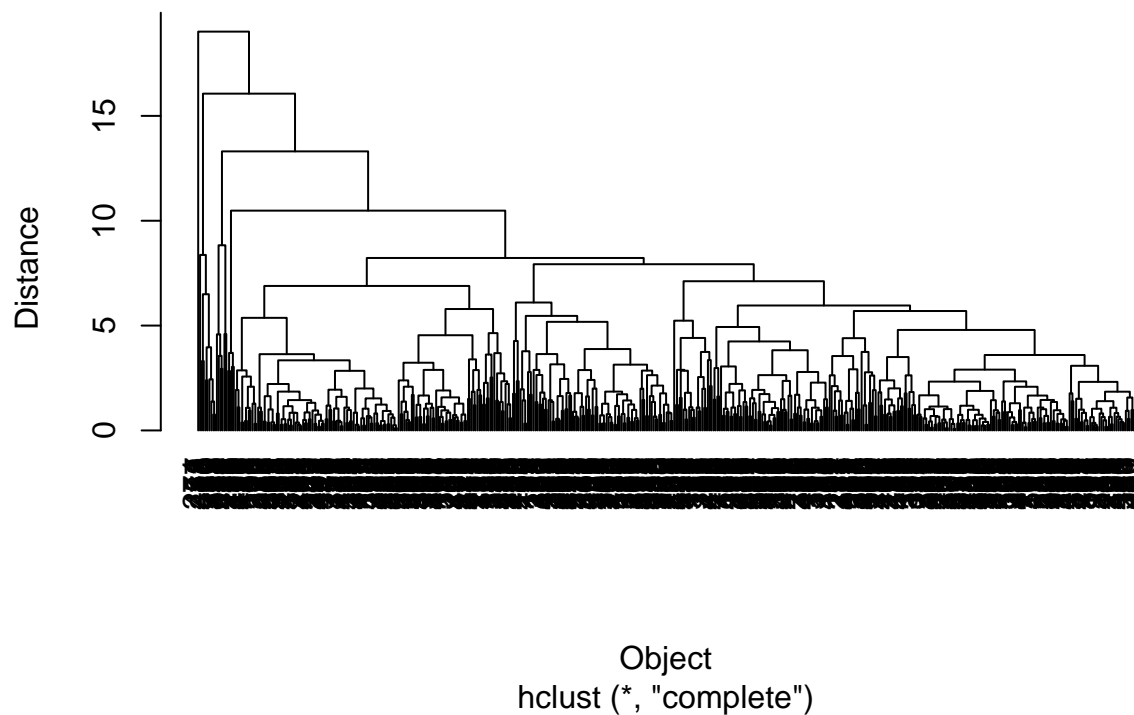
```
plot(cut(single,h=1)$lower[[2]],xlab="Object",ylab="Distance",
     main="Dendrogram. Nearest neighbor linkage")
```

Dendrogram. Nearest neighbor linkage



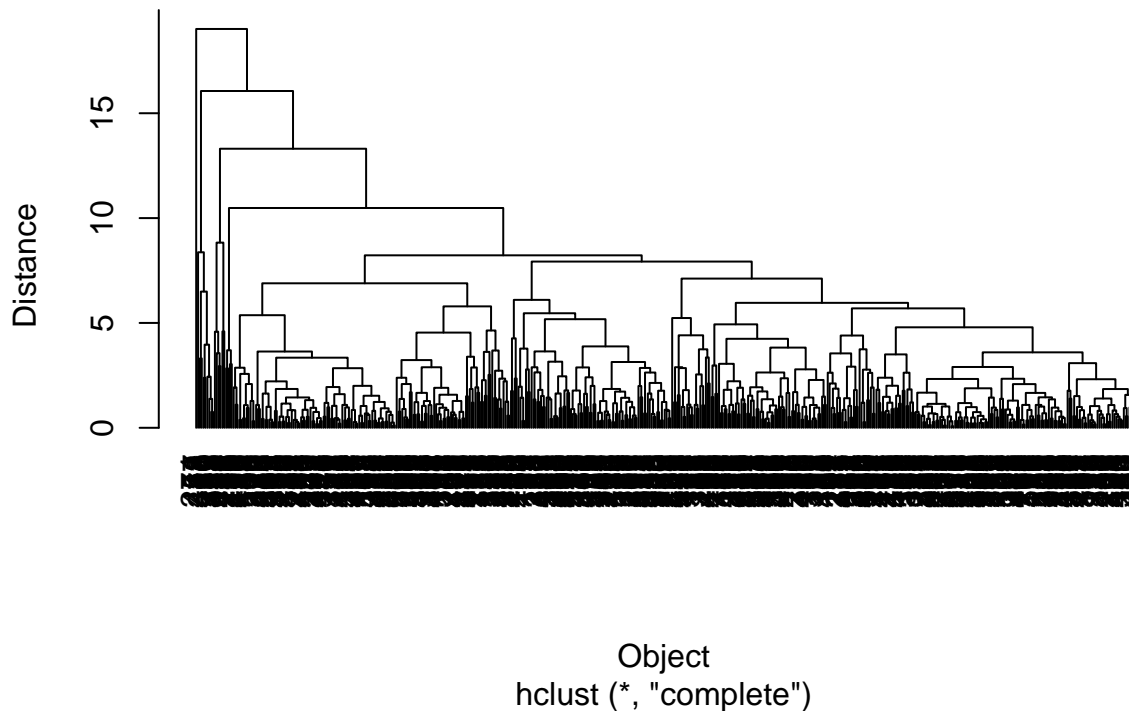
```
# Complete link  
numcomplete<-hclust(numdist)  
plot(numcomplete, hang=-1,xlab="Object",ylab="Distance",  
      main="Dendrogram. Farthest neighbor linkage")
```

Dendrogram. Farthest neighbor linkage



```
# Average linkage
numaverage<-hclust(numdist)
plot(numaverage, hang=-1,xlab="Object",ylab="Distance",
     main="Dendrogram. Group average linkage")
```

Dendrogram. Group average linkage



```
# Single link clustering
km.res<-kmeans(num,centers = 10,iter.max = 10,nstart = 3)
print(km.res)
```

```
## K-means clustering with 10 clusters of sizes 4, 10, 12, 66, 92, 16, 21, 118, 63, 98
##
```

```
## Cluster means:
```

##	Severity	Distance.mi.	Temperature.F.	Humidity...	Pressure.in.
## 1	0.3743416	-0.286109267	-0.23287818	-1.05967239	-3.80953332
## 2	0.1908408	0.826887066	-0.40688757	1.14095972	0.14273235
## 3	0.2214243	4.773803896	0.08429905	-0.03386252	0.02608219
## 4	0.1797195	-0.085547380	-1.30850887	1.02826788	0.23463854
## 5	-0.4434336	-0.167237674	0.90139423	-1.06906195	0.21588584
## 6	-0.3137863	-0.250997367	-1.28100876	0.48714091	-3.06267571
## 7	-0.3683996	-0.182746036	0.62953348	-1.57746213	-2.77625202
## 8	-0.5898150	-0.227487660	-0.11392558	0.54886337	0.37576459
## 9	-0.6014165	-0.126760851	0.10716318	-0.29267170	0.17486842
## 10	1.4603665	-0.007091685	0.21826522	0.02784033	0.30711587

##	Visibility.mi.	Wind_Speed.mph.	Precipitation.in.
## 1	7.35699377	0.15304465	-0.220774169
## 2	-1.26889254	0.11862939	6.138093483
## 3	0.06005446	-0.06922055	-0.071924271
## 4	-1.34121619	0.12019372	0.082338351


```
## 5      0.22570256      -0.22477502      -0.220774169
## 6     -0.19420836      -0.41695804       0.002500678
## 7      0.37677791      -0.05590513      -0.220774169
## 8      0.13776297      -0.62129863      -0.114813225
## 9      0.20867873       1.67441751      -0.152728501
## 10     0.16416466      -0.12807181      -0.173385222
```

```
##
```

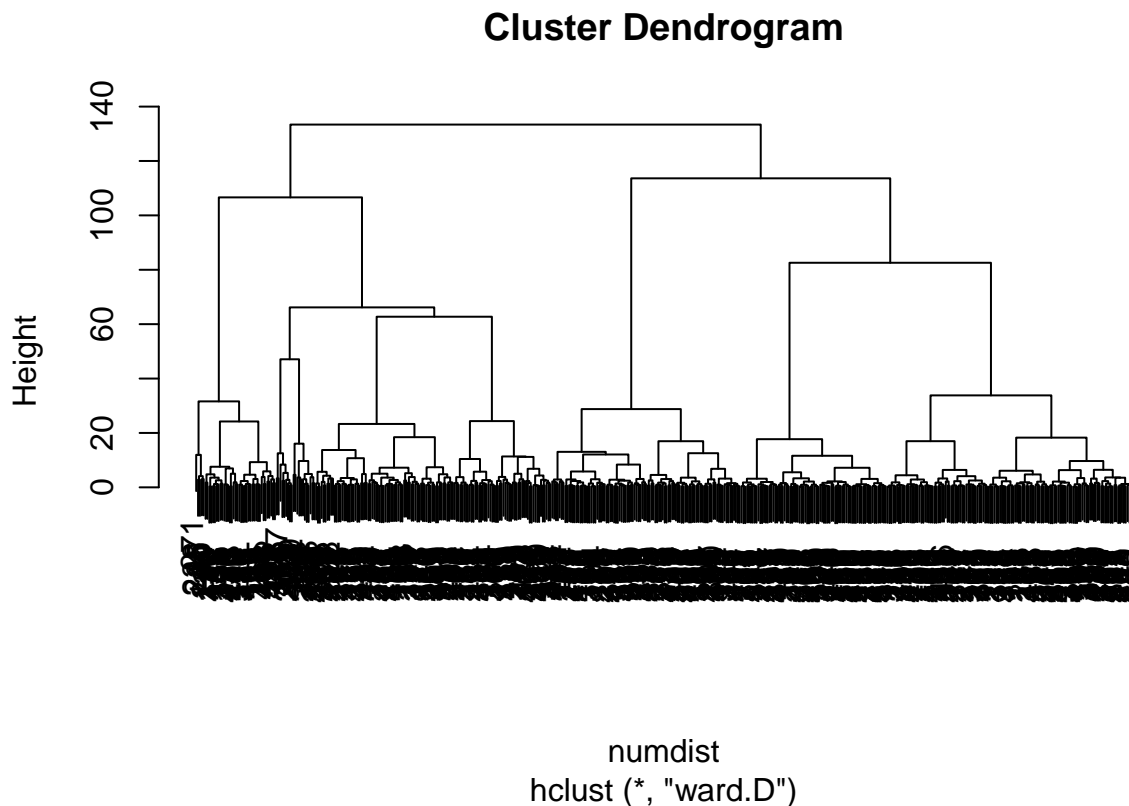
```
## Clustering vector:
```

```
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20
##  9  5 10  8  9  4  4 10  8 10  5 10  5  5  6  2  8  8  9  8
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
## 10  5  8  5  7 10  8  8  8  5 10  2  4 10  8  7 10 10  3  8
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
##  5  4  4  7  8  4  8  5  5  8  9  8  8  4  9  1  5 10  6 10
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
##  6  4  9  9  4 10  4  7 10  5  9  8 10 10  9  4  8  8  7  3
## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
##  5  9  4  8  9 10  4  2  4 10  5 10  8  4 10 10  5  4  6  8
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
##  5  5  9  9  5  9 10  5 10  7  7  8  8 10  5  8  4  7  8  8
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
##  9  8 10  8  8  9  7  8  5  8  9  8  8  9  4  5  3  9  8  5
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
##  7  8  4 10  5  5  4  5  5  4 10  5  8  5  7  7  5 10  4  8
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
##  4 10  8  9  4  4  7 10  8  4  9  5  5  8  9 10  9  9  4  8
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
##  6  9  6  9  3 10  8  9  9 10  4  5  8 10 10  9 10 10  9 10
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
##  4  4  8  4  4 10 10  5  8  5 10  3  7 10 10  5  5  4  9  9
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
##  7  3  5  3  9  4  8  8  4  4  5  8  8  8  8  8  9  1  8  8
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
##  8  5  9 10  5  1  8  8  8 10 10  5  5  4  6 10  4  8  8  4
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
##  8  8  9 10 10  8 10 10  9  8  1 10  2 10  8  4  4  8  4  9
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
##  9  8  9  5  8 10  5  6  5  7  5  5  9  5  5  5  4  8  8  8
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 10 10  2  8 10  8  8  9  4  9  6 10  8 10 10  5  8  4  9  7
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 10  4  8 10  5 10  4 10  9  6  5  5  8  4  5  7  5  9 10 10
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
##  3  5 10  8  5  8 10  8  9  8  8  3  5  9 10 10  8  4 10 10
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
##  9  8  5  8  5  9  9  8  5  9  5  8  2  4 10 10 10  4 10  4
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
##  8  5  8  5  3  4  5  5  8 10  5  5  8 10  5  8  8 10 10  8
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 10  6  5 10  8  9  6 10  7 10 10  9  2  5 10  5  8  5  8 10
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
##  9  8  9  5  8  7  9  8  8  4  8  4  5  5  7  6  8  3  9 10
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 10  6  6  5  8  6 10  9  5  9  4  4  5  4 10  4  2  5  5 10
```

```
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
##   8   5   4   4   8   2  10   8   4   5   8   3  10   9   8   8   4   8   4   2
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
##   5  10   9   4   5   5   8   8   8   5   5  10  10   9   5   4   8   5  10   8
##
## Within cluster sum of squares by cluster:
## [1]  57.42334 113.61006 145.09099 265.51365 143.53003  53.86346  71.31594
## [8] 186.17083 197.33970 251.18168
## (between_SS / total_SS =  62.8 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"
##
# Ward Hierarchical Clustering
fit <- hclust(numdist, method="ward")
```

```
## The "ward" method has been renamed to "ward.D"; note new "ward.D2"
```

```
# display dendrogram
plot(fit)
```



```
# using the agnes function for all options in a single line code
?agnes
# Agglomerative Nesting (Hierarchical Clustering)
(agnum <- agnes(num, metric="euclidean", stand=TRUE, method = "single"))

## Call:      agnes(x = num, metric = "euclidean", stand = TRUE, method = "single")
## Agglomerative coefficient:  0.9225946
```

```

## Order of objects:
## [1] 1 189 64 308 421 82 85 263 367 177 171 2 148 173 353 335 462 11
## [19] 459 382 434 458 22 388 296 403 449 49 81 387 102 291 154 30 172 470
## [37] 57 105 145 29 116 122 348 132 393 169 270 468 350 419 372 35 40 248
## [55] 53 120 193 77 351 93 100 445 208 418 242 217 491 245 325 210 444 498
## [73] 252 304 500 112 431 228 300 234 240 364 306 400 488 422 428 478 395 72
## [91] 84 489 119 239 487 152 363 495 153 160 298 368 104 283 483 220 384 23
## [109] 45 78 405 497 307 437 461 465 417 50 174 278 285 247 133 139 142 266
## [127] 28 357 313 396 227 333 233 97 216 284 24 235 475 101 115 369 392 481
## [145] 236 163 289 371 13 485 14 345 192 391 41 51 223 342 134 269 280 131
## [163] 237 113 203 258 389 47 397 20 146 70 157 5 108 317 370 294 365 433
## [181] 414 295 275 140 232 91 490 52 62 425 471 106 439 486 231 476 243 293
## [199] 354 474 249 125 282 187 17 406 310 366 344 63 429 361 332 196 412 138
## [217] 261 346 323 175 221 320 316 259 209 130 241 225 7 42 204 161 427 164
## [235] 494 281 6 165 18 362 33 327 159 378 65 218 452 191 469 229 334 358
## [253] 430 135 463 147 477 126 149 329 349 129 338 484 87 121 432 201 456 279
## [271] 178 4 230 128 27 381 383 448 3 26 197 404 398 312 37 190 176 411
## [289] 482 86 95 492 379 162 376 151 441 206 326 12 328 360 399 408 69 158
## [307] 109 31 410 48 60 198 253 195 265 359 287 8 274 455 355 493 21 123
## [325] 74 305 214 250 211 268 66 264 420 186 38 200 301 394 347 467 424 453
## [343] 447 34 58 96 267 343 377 401 302 319 90 356 168 277 286 499 297 315
## [361] 43 166 67 260 374 170 117 479 202 257 386 226 454 339 460 254 318 10
## [379] 144 390 73 92 76 276 309 188 251 219 340 114 256 324 194 207 321 15
## [397] 44 59 330 442 181 311 68 110 435 127 183 446 118 36 111 155 336 103
## [415] 167 156 426 46 124 450 54 75 136 337 215 272 89 79 409 213 55 182
## [433] 141 290 180 199 107 415 473 244 179 496 184 99 443 299 150 451 292 331
## [451] 9 39 212 385 222 185 314 19 262 61 205 416 288 407 352 436 402 380
## [469] 464 94 423 98 71 322 341 83 480 375 440 16 32 273 303 373 466 25
## [487] 143 56 238 246 137 224 438 255 80 472 457 88 413 271
## Height (summary):
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.09727 0.57043 0.92079 1.30261 1.59595 15.08708
##
## Available components:
## [1] "order" "height" "ac" "merge" "diss" "call"
## [7] "method" "order.lab" "data"

#View(agn.num)

# Description of cluster merging
agn.num$merge

## [,1] [,2]
## [1,] -78 -405
## [2,] -403 -449
## [3,] -169 -270
## [4,] -461 -465
## [5,] -72 -84
## [6,] -53 -120
## [7,] -35 -40
## [8,] -294 -365
## [9,] -148 -173
## [10,] -296 2
## [11,] -102 -291
## [12,] -112 -431

```

```

## [13,] -217 -491
## [14,]   8 -433
## [15,] -306 -400
## [16,] -382 -434
## [17,]  -64 -308
## [18,] -239 -487
## [19,]  -81 -387
## [20,]   -2   9
## [21,]  -57 -105
## [22,] -174 -278
## [23,] -228 -300
## [24,] -227 -333
## [25,]   1 -497
## [26,] -335 -462
## [27,]   6 -193
## [28,]  15 -488
## [29,] -307 -437
## [30,] -106 -439
## [31,] -425 -471
## [32,]  11 -154
## [33,] -392 -481
## [34,] -201 -456
## [35,] -132 -393
## [36,]  -95 -492
## [37,] -113 -203
## [38,]   36 -379
## [39,]   29   4
## [40,]  -11 -459
## [41,]   25   39
## [42,] -240 -364
## [43,]   27  -77
## [44,]   42   28
## [45,]  -26 -197
## [46,]   12   23
## [47,] -122 -348
## [48,] -208 -418
## [49,]   40   16
## [50,]  -30 -172
## [51,]  -49   19
## [52,]   51   32
## [53,] -164 -494
## [54,]  -50   22
## [55,] -139 -142
## [56,] -210 -444
## [57,]  -22 -388
## [58,]   17 -421
## [59,] -214 -250
## [60,]   47   35
## [61,] -369   33
## [62,]   57   10
## [63,] -160 -298
## [64,] -245 -325
## [65,]  -36 -111
## [66,] -363 -495

```

```

## [67,] 46 -234
## [68,] -100 -445
## [69,] -133 55
## [70,] 62 52
## [71,] -104 -283
## [72,] -347 -467
## [73,] -37 -190
## [74,] -216 -284
## [75,] -428 -478
## [76,] -156 -426
## [77,] 56 -498
## [78,] -13 -485
## [79,] 43 -351
## [80,] -152 66
## [81,] 70 50
## [82,] 3 -468
## [83,] 67 44
## [84,] -12 -328
## [85,] 7 -248
## [86,] 48 -242
## [87,] 83 -422
## [88,] -51 -223
## [89,] 5 -489
## [90,] 84 -360
## [91,] -93 68
## [92,] -350 -419
## [93,] -69 -158
## [94,] -1 -189
## [95,] 60 82
## [96,] 13 64
## [97,] 86 96
## [98,] 20 -353
## [99,] 54 -285
## [100,] 87 75
## [101,] 49 -458
## [102,] 79 91
## [103,] -235 -475
## [104,] -119 18
## [105,] -116 95
## [106,] 71 -483
## [107,] 41 -417
## [108,] 73 -176
## [109,] -192 -391
## [110,] 99 -247
## [111,] -269 -280
## [112,] 105 92
## [113,] 81 -470
## [114,] -29 112
## [115,] -82 -85
## [116,] -45 107
## [117,] 45 -404
## [118,] 69 -266
## [119,] 113 21
## [120,] 108 -411

```

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## [121,] 101 119
## [122,] -60 -198
## [123,] 106 -220
## [124,] 110 118
## [125,] -23 116
## [126,] -399 -408
## [127,] -263 -367
## [128,] 85 102
## [129,] 114 -372
## [130,] 129 128
## [131,] 77 -252
## [132,] 125 124
## [133,] -8 -274
## [134,] -28 -357
## [135,] 132 134
## [136,] 121 -145
## [137,] -163 -289
## [138,] 63 -368
## [139,] 122 -253
## [140,] 97 131
## [141,] 130 140
## [142,] 90 126
## [143,] 120 -482
## [144,] 141 -304
## [145,] 136 144
## [146,] -153 138
## [147,] 133 -455
## [148,] -355 -493
## [149,] 145 -500
## [150,] 149 100
## [151,] 150 -395
## [152,] 135 -313
## [153,] -151 -441
## [154,] 151 89
## [155,] 152 -396
## [156,] 154 104
## [157,] -381 -383
## [158,] 156 80
## [159,] 143 -86
## [160,] 158 146
## [161,] -14 -345
## [162,] 155 24
## [163,] -70 -157
## [164,] -59 -330
## [165,] 98 26
## [166,] 58 115
## [167,] 160 123
## [168,] 166 127
## [169,] 117 -398
## [170,] 167 -384
## [171,] 88 -342
## [172,] 171 -134
## [173,] 170 162
## [174,] 153 -206

```

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## [175,] 169 -312
## [176,] -229 -334
## [177,] -3 175
## [178,] -101 -115
## [179,] 37 -258
## [180,] 78 161
## [181,] -41 172
## [182,] 173 -233
## [183,] 182 -97
## [184,] 183 74
## [185,] 177 159
## [186,] -52 -62
## [187,] 94 168
## [188,] 184 -24
## [189,] 187 -177
## [190,] -200 -301
## [191,] 188 103
## [192,] 191 178
## [193,] 176 -358
## [194,] 192 61
## [195,] 165 194
## [196,] -31 -410
## [197,] 189 -171
## [198,] -293 -354
## [199,] 137 -371
## [200,] -195 -265
## [201,] 195 -236
## [202,] 201 199
## [203,] 180 109
## [204,] 147 148
## [205,] 202 203
## [206,] 197 205
## [207,] 142 93
## [208,] 38 -162
## [209,] 206 181
## [210,] 208 -376
## [211,] 209 111
## [212,] 185 210
## [213,] 211 -131
## [214,] 213 -237
## [215,] 214 179
## [216,] 215 -389
## [217,] 139 200
## [218,] -310 -366
## [219,] 198 -474
## [220,] -48 217
## [221,] -110 -435
## [222,] 212 174
## [223,] 207 -109
## [224,] -181 -311
## [225,] 216 -47
## [226,] -38 190
## [227,] -264 -420
## [228,] 225 -397

```

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## [229,] -218 -452
## [230,]  228  -20
## [231,]  230 -146
## [232,]  231  163
## [233,]  232   -5
## [234,]  233 -108
## [235,] -261 -346
## [236,]  234 -317
## [237,] -129 -338
## [238,]  223  196
## [239,]  -90 -356
## [240,]  222 -326
## [241,]   14 -414
## [242,]  236 -370
## [243,]  220 -359
## [244,]  226 -394
## [245,] -424 -453
## [246,]  242  241
## [247,]  -21 -123
## [248,]  -73  -92
## [249,]  239 -168
## [250,]  238  243
## [251,] -243  219
## [252,]  240  250
## [253,]  -42 -204
## [254,] -212 -385
## [255,]  186   31
## [256,]  -91 -490
## [257,]  246 -295
## [258,]  164 -442
## [259,]  -67 -260
## [260,]  257 -275
## [261,]  252 -287
## [262,]  260 -140
## [263,]  262 -232
## [264,]  261  204
## [265,] -117 -479
## [266,]  263  256
## [267,]  264  247
## [268,]  267  -74
## [269,] -257 -386
## [270,] -196 -412
## [271,]  244   72
## [272,]  -65  229
## [273,]  268 -305
## [274,]  273   59
## [275,] -125 -282
## [276,] -292 -331
## [277,]  266  255
## [278,]  -66  227
## [279,]  277   30
## [280,]  279 -486
## [281,]  280 -231
## [282,]  274 -211

```



```

## [283,] 281 -476
## [284,] 258 224
## [285,] 282 -268
## [286,] -226 -454
## [287,] 249 -277
## [288,] 285 278
## [289,] 283 251
## [290,] 275 -187
## [291,] -273 -303
## [292,] 288 -186
## [293,] 292 271
## [294,] 289 -249
## [295,] -256 -324
## [296,] -58 -96
## [297,] -221 -320
## [298,] -127 -183
## [299,] 293 245
## [300,] 294 290
## [301,] -6 -165
## [302,] 253 -161
## [303,] 299 -447
## [304,] 303 -34
## [305,] -43 -166
## [306,] 235 -323
## [307,] 305 259
## [308,] 300 -17
## [309,] 308 -406
## [310,] 272 -191
## [311,] -343 -377
## [312,] 311 -401
## [313,] -267 312
## [314,] 304 296
## [315,] 314 313
## [316,] 307 -374
## [317,] 309 218
## [318,] 315 -302
## [319,] -202 269
## [320,] 318 -319
## [321,] 317 -344
## [322,] 316 -170
## [323,] 310 -469
## [324,] 320 287
## [325,] -68 221
## [326,] -39 254
## [327,] -339 -460
## [328,] 325 298
## [329,] -159 -378
## [330,] 323 193
## [331,] -103 -167
## [332,] 321 -63
## [333,] 284 328
## [334,] 265 319
## [335,] 332 -429
## [336,] -44 333

```

```
## [337,] -286 -499
## [338,] 324 337
## [339,] -18 -362
## [340,] 34 -279
## [341,] 322 334
## [342,] 338 -297
## [343,] 335 -361
## [344,] -130 -241
## [345,] 343 -332
## [346,] 342 -315
## [347,] 346 341
## [348,] 345 270
## [349,] -33 -327
## [350,] 347 286
## [351,] -254 -318
## [352,] 297 -316
## [353,] 349 329
## [354,] 348 -138
## [355,] 350 327
## [356,] 354 306
## [357,] 336 -446
## [358,] 355 351
## [359,] 339 353
## [360,] 359 330
## [361,] 358 -10
## [362,] -135 -463
## [363,] -4 -230
## [364,] 361 -144
## [365,] -27 157
## [366,] 356 -175
## [367,] 360 -430
## [368,] 367 362
## [369,] 368 -147
## [370,] -7 302
## [371,] 366 352
## [372,] 371 -259
## [373,] 369 -477
## [374,] 65 -155
## [375,] 301 373
## [376,] 326 -222
## [377,] 364 -390
## [378,] 370 -427
## [379,] 372 -209
## [380,] -15 357
## [381,] 379 344
## [382,] 381 -225
## [383,] 382 378
## [384,] 383 53
## [385,] 380 -118
## [386,] 374 -336
## [387,] 384 -281
## [388,] 387 375
## [389,] -79 -409
## [390,] 388 -126
```

```
## [391,] -329 -349
## [392,]  385  386
## [393,]  392  331
## [394,]  390 -149
## [395,]  394  391
## [396,]  395  237
## [397,]  363 -128
## [398,]  397  365
## [399,]  389 -213
## [400,]  -76 -276
## [401,]  377  248
## [402,]  401  400
## [403,] -114  295
## [404,]  396 -484
## [405,]  404  -87
## [406,]  -99 -443
## [407,]  398 -448
## [408,]  402 -309
## [409,]  405 -121
## [410,]  -32  291
## [411,]  408 -188
## [412,]  411 -251
## [413,] -415 -473
## [414,]  409 -432
## [415,]  414  340
## [416,]  415 -178
## [417,]  412 -219
## [418,]  417 -340
## [419,]  418  403
## [420,] -215 -272
## [421,] -136 -337
## [422,] -124 -450
## [423,]  416  407
## [424,]  419 -194
## [425,]  424 -207
## [426,] -107  413
## [427,]  425 -321
## [428,]  393   76
## [429,]  -55 -182
## [430,]  423  427
## [431,]  430  428
## [432,]  431  -46
## [433,]  432  422
## [434,] -380 -464
## [435,]  433  -54
## [436,]  435  -75
## [437,]  436  421
## [438,]  437  420
## [439,]  438  -89
## [440,]  439  399
## [441,]  440  429
## [442,] -141 -290
## [443,]  441  442
## [444,]  443 -180
```

```

## [445,] 444 -199
## [446,] 426 -244
## [447,] 445 446
## [448,] 447 -179
## [449,] 376 -185
## [450,] -288 -407
## [451,] -9 449
## [452,] -150 -451
## [453,] 448 -496
## [454,] 453 -184
## [455,] 454 406
## [456,] 455 -299
## [457,] 456 452
## [458,] 457 276
## [459,] 458 451
## [460,] -19 -262
## [461,] -375 -440
## [462,] 459 -314
## [463,] 462 460
## [464,] -83 -480
## [465,] 463 -61
## [466,] 465 -205
## [467,] 466 -416
## [468,] 467 450
## [469,] 468 -352
## [470,] 469 -436
## [471,] 470 -402
## [472,] 471 434
## [473,] 472 -94
## [474,] 473 -423
## [475,] 474 -98
## [476,] -238 -246
## [477,] -373 -466
## [478,] -16 410
## [479,] 475 -71
## [480,] 479 -322
## [481,] 480 -341
## [482,] -56 476
## [483,] 481 464
## [484,] 483 461
## [485,] 478 477
## [486,] -137 -224
## [487,] -88 -413
## [488,] 484 485
## [489,] 488 -25
## [490,] 486 -438
## [491,] 489 -143
## [492,] 491 482
## [493,] 492 490
## [494,] 493 -255
## [495,] -80 -472
## [496,] 494 495
## [497,] 496 -457
## [498,] 497 487

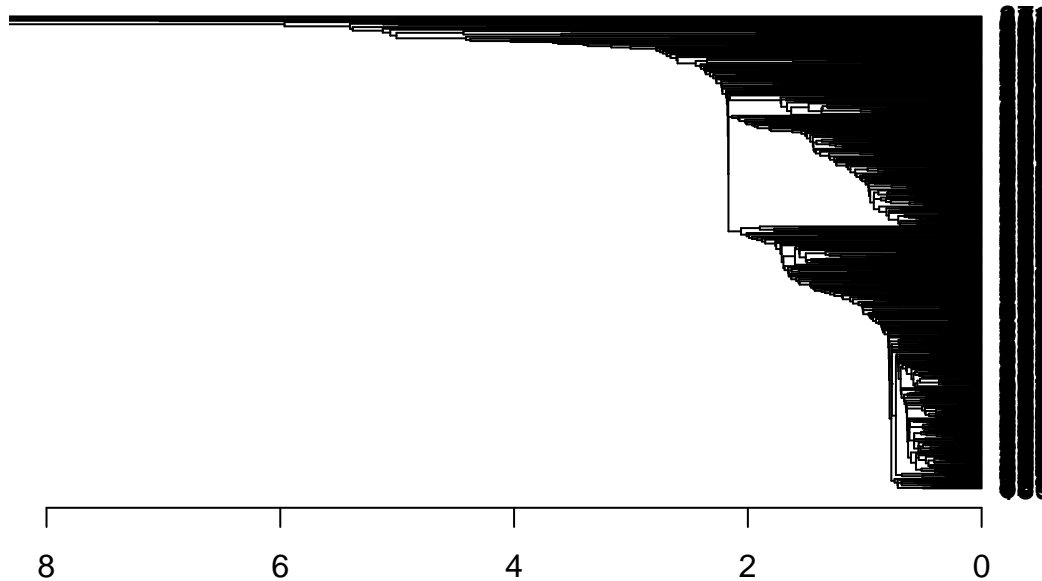
```

```
## [499,] 498 -271
```

```
#Dendrogram
```

```
plot(as.dendrogram(agn.num), xlab= "US Accidents Dendrogram",xlim=c(8,0),  
     horiz = TRUE,main="Dendrogram for us accident records")
```

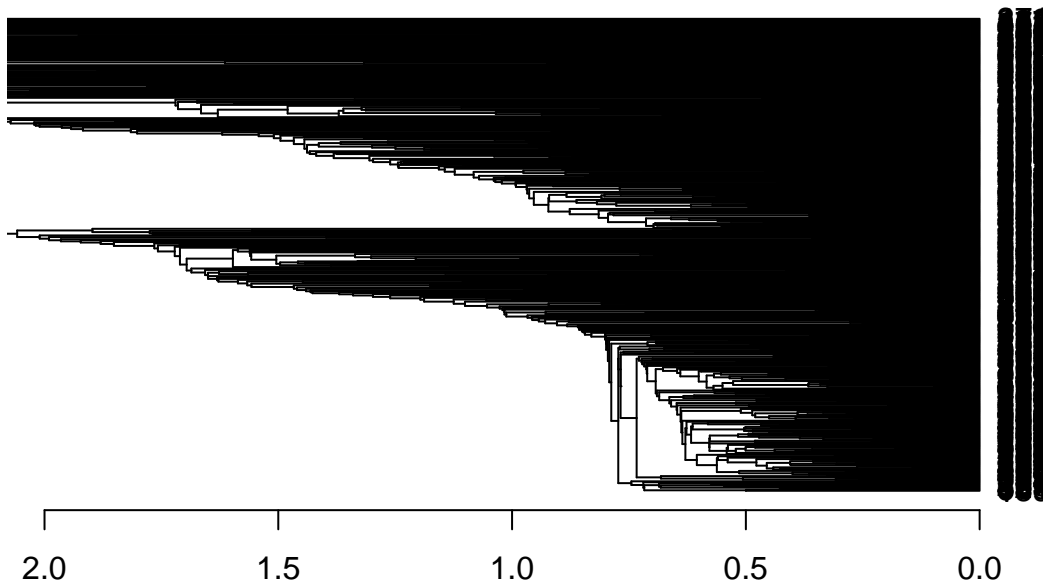
Dendrogram for us accident records



US Accidents Dendrogram

```
plot(as.dendrogram(agn.num), xlab= "US Accidents Dendrogram",xlim=c(2,0),  
     horiz = TRUE,main="Dendrogram for us accident records")
```

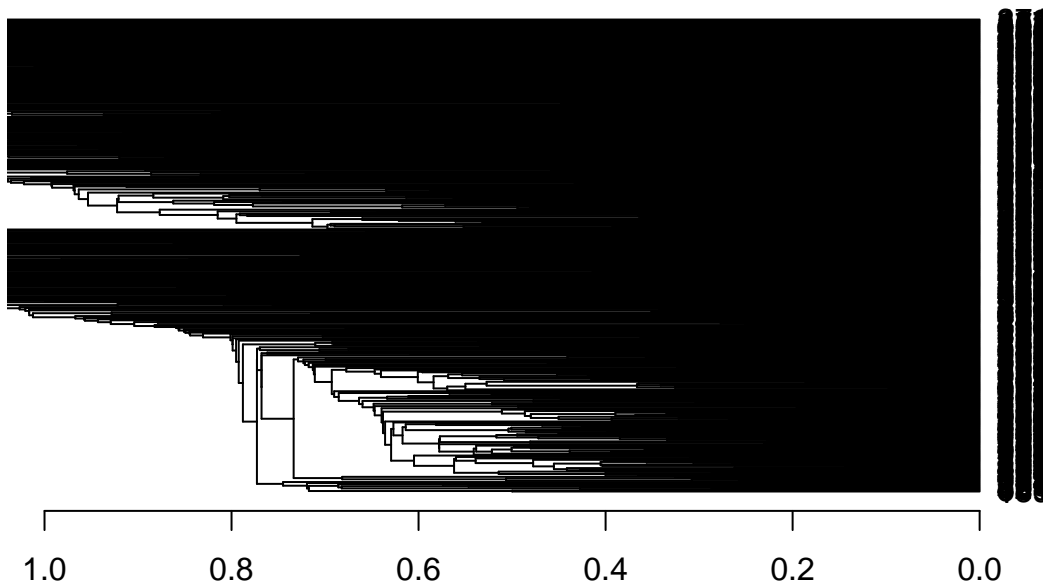
Dendrogram for us accident records



US Accidents Dendrogram

```
plot(as.dendrogram(agn.num), xlab= "US Accidents Dendrogram",xlim=c(1,0),  
     horiz = TRUE,main="Dendrogram for us accident records")
```

Dendrogram for us accident records



US Accidents Dendrogram

```
#Interactive Plots,  
#plot(agn.num,ask=TRUE)
```

```

#plot(agn.num, which.plots=2)

#K-Means Clustering

# K-means, k=2, 3, 4, 5, 6
# Centers (k's) are numbers thus, 10 random sets are chosen

# Computing the percentage of variation accounted for. Two clusters
matstd.num<- scale(num[,1:8])
(kmeans2.num <- kmeans(matstd.num,2,nstart = 10))

## K-means clustering with 2 clusters of sizes 252, 248
##
## Cluster means:
##      Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
## 1  0.1486065  0.02752796    -0.6182097  0.7626184  0.09499899
## 2 -0.1510034 -0.02797196     0.6281808 -0.7749187 -0.09653123
##  Visibility.mi. Wind_Speed.mph. Precipitation.in.
## 1    -0.3542998    -0.2346050      0.2130170
## 2     0.3600143     0.2383889     -0.2164527
##
## Clustering vector:
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20
##  2  2  2  1  2  1  1  2  1  2  2  1  2  2  1  1  1  1  2  1
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
##  1  2  1  2  2  2  1  1  1  2  2  1  1  1  1  2  1  2  1  1
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
##  2  1  1  2  1  1  1  2  2  1  2  1  2  1  2  2  2  1  1  2
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
##  1  1  1  2  1  1  1  2  1  2  1  1  1  1  1  1  2  1  2  2
## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
##  2  2  1  1  2  2  1  1  1  1  2  1  2  1  1  1  2  1  1  2
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
##  2  2  2  2  2  2  1  2  1  2  2  2  2  2  2  1  1  2  1  2
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
##  1  1  2  2  2  2  2  1  2  1  2  1  1  2  1  2  1  1  1  2
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
##  2  1  1  2  2  2  1  2  2  1  2  2  1  2  2  2  2  1  1  1
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
##  1  1  2  2  1  1  2  1  1  1  2  2  2  1  2  2  1  1  1  1
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
##  1  2  2  1  1  2  1  2  2  2  1  2  2  1  2  2  2  2  2  1
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
##  1  1  2  1  1  1  2  2  1  2  1  2  2  2  1  2  2  1  2  2
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
##  2  2  2  1  2  1  1  1  1  1  2  1  1  1  2  1  2  2  2  2
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
##  2  2  2  2  2  2  1  1  1  2  1  2  2  1  1  2  1  2  2  1
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
##  1  2  2  1  2  1  1  2  2  1  2  1  1  2  1  1  1  1  1  2
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300

```

```

## 1 1 2 2 1 1 2 1 2 2 2 2 2 2 2 2 1 1 1 1
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 2 1 1 1 2 1 1 2 1 2 1 2 1 1 2 2 1 1 2 2
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 1 1 1 2 2 1 1 1 2 1 2 2 1 1 2 2 2 2 1 2
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 1 2 1 1 2 1 2 1 2 1 2 2 2 2 2 2 1 1 1 1
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
## 1 1 2 1 2 2 2 1 2 2 2 2 1 1 1 1 1 1 1 1
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
## 1 2 1 2 2 1 2 2 1 1 2 2 1 2 2 1 1 2 2 1
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 1 1 2 2 1 2 1 2 2 2 2 2 1 2 2 2 1 2 1 1
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
## 2 1 2 2 1 2 1 2 1 1 2 1 2 2 2 1 1 1 2 1
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 1 1 1 2 2 1 2 1 2 2 1 1 2 1 2 1 1 2 2 1
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
## 1 2 1 1 1 1 2 1 1 2 1 2 1 2 2 1 1 2 1 1
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 2 1 2 1 2 2 2 1 1 2 2 1 2 2 2 1 1 2 1 1
##
## Within cluster sum of squares by cluster:
## [1] 1824.705 1546.653
## (between_SS / total_SS = 15.5 %)
##
## Available components:
##
## [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
## [6] "betweenss" "size" "iter" "ifault"
perc.var.2 <- round(100*(1 - kmeans2.num$betweenss/kmeans2.num$totss),1)
names(perc.var.2) <- "Perc. 2 clus"
perc.var.2

## Perc. 2 clus
## 84.5

# Computing the percentage of variation accounted for. Three clusters
(kmeans3.num <- kmeans(matstd.num,3,nstart = 10))

## K-means clustering with 3 clusters of sizes 10, 241, 249
##
## Cluster means:
## Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
## 1 0.1908408 0.826887066 -0.4068876 1.1409597 0.1427323
## 2 -0.1776836 -0.030426898 0.6445745 -0.7904393 -0.1152943
## 3 0.1643106 -0.003758989 -0.6075244 0.7192220 0.1058578
## Visibility.mi. Wind_Speed.mph. Precipitation.in.
## 1 -1.2688925 0.1186294 6.13809348
## 2 0.3609564 0.2647871 -0.21484488
## 3 -0.2983999 -0.2610441 -0.03856755
##
## Clustering vector:
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20

```



```

## 2 2 2 3 2 3 3 2 3 3 2 3 2 2 3 1 3 3 2 3
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
## 3 2 3 2 2 2 3 3 3 2 2 1 3 3 3 2 3 2 3 3
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
## 2 3 3 2 3 3 3 2 2 3 2 3 2 3 2 2 2 3 3 2
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
## 3 3 3 2 3 3 3 2 3 2 3 3 3 3 3 3 2 3 2 2
## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
## 2 2 3 3 2 2 3 1 3 3 2 3 2 3 3 3 2 3 3 2
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
## 2 2 2 2 2 2 3 2 3 2 2 2 2 2 2 3 3 2 3 3
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
## 3 3 2 2 2 2 2 3 2 3 2 3 3 2 3 2 3 3 3 2
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
## 2 3 3 2 2 2 3 2 2 3 2 2 3 2 2 2 2 3 3 3
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
## 3 3 2 2 3 3 2 3 3 3 2 2 2 3 2 2 3 2 3 3
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
## 3 2 2 3 3 2 3 2 2 3 3 2 3 3 2 2 2 2 2 3
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
## 3 3 2 3 3 3 2 2 3 2 3 2 2 2 3 2 2 3 2 2
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
## 2 2 2 3 2 3 3 3 3 3 2 3 3 3 2 3 2 2 2 2
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
## 2 2 2 2 2 2 3 3 3 2 3 2 2 3 3 2 3 2 2 3
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
## 3 2 2 3 2 3 3 2 2 3 2 3 1 2 3 3 3 3 3 2
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
## 3 3 2 2 3 3 2 3 2 2 2 2 2 2 2 2 3 3 3 3
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 3 3 1 3 2 3 3 2 3 2 3 2 3 3 2 2 3 3 2 2
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 3 3 3 2 2 3 3 3 2 3 2 2 3 3 2 2 2 2 3 2
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 3 2 3 3 2 3 2 3 2 3 2 2 2 2 2 3 3 3 3 3
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
## 3 3 2 3 2 2 2 3 2 2 2 2 1 3 3 3 3 3 3 3
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
## 3 2 3 2 2 3 2 2 3 3 2 2 3 2 2 3 3 2 2 3
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 3 3 2 2 3 2 3 2 2 2 2 2 1 2 3 2 3 2 3 3
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
## 2 3 2 2 3 2 3 3 3 3 2 3 2 2 2 3 3 3 2 3
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 3 3 3 2 2 3 2 3 2 2 3 3 2 3 2 3 1 2 2 3
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
## 3 2 3 3 3 1 2 3 3 2 3 2 3 2 2 3 3 2 3 1
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 2 3 2 3 2 2 3 3 3 2 2 3 2 2 2 3 3 2 3 3
##
## Within cluster sum of squares by cluster:
## [1] 113.6101 1519.3207 1353.0941
## (between_SS / total_SS = 25.2 %)
##

```

```
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"
```

```
perc.var.3 <- round(100*(1 - kmeans3.num$betweenss/kmeans3.num$totss),1)
```

```
names(perc.var.3) <- "Perc. 3 clus"
perc.var.3
```

```
## Perc. 3 clus
##      74.8
```

```
# Computing the percentage of variation accounted for. Four clusters
(kmeans4.num<- kmeans(matstd.num,4,nstart = 10))
```

```
## K-means clustering with 4 clusters of sizes 41, 10, 226, 223
```

```
##
```

```
## Cluster means:
```

```
##      Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
## 1 -0.2746246  0.004734917   -0.09888672  -0.8244867  -2.9857488
## 2  0.1908408  0.826887066   -0.40688757   1.1409597   0.1427323
## 3  0.2119515  0.023318575   -0.61459811   0.7418733   0.2796961
## 4 -0.1728692 -0.061582961    0.65929329  -0.6514306   0.2590899
## Visibility.mi. Wind_Speed.mph. Precipitation.in.
## 1      0.8790505      -0.07988788      -0.15106885
## 2     -1.2688925      0.11862939      6.13809348
## 3     -0.3308795     -0.26397385     -0.03267005
## 4      0.2306127      0.27689327     -0.21436628
```

```
##
```

```
## Clustering vector:
```

```
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20
##  4  4  4  3  4  3  3  4  3  3  4  3  4  4  1  2  3  3  4  3
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
##  3  4  3  4  1  4  3  3  3  4  4  2  3  3  4  1  3  4  3  4
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
##  4  3  3  1  3  3  3  4  4  3  4  3  4  3  4  1  4  3  1  4
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
##  1  3  3  4  3  3  3  1  3  4  3  3  3  3  3  3  4  3  1  4
## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
##  4  4  3  3  4  4  3  2  3  3  4  3  4  3  3  3  4  3  1  4
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
##  4  4  1  4  4  4  3  4  3  1  1  4  4  4  4  3  3  1  3  4
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
##  3  3  4  4  4  4  1  3  4  3  4  3  3  4  3  4  3  3  3  4
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
##  1  3  3  4  4  4  3  4  4  3  4  4  3  4  1  1  4  3  3  3
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
##  3  3  4  4  3  3  1  3  3  3  4  4  4  3  4  4  3  4  3  3
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
##  1  4  1  3  3  4  3  4  4  4  3  4  4  3  4  4  4  4  4  3
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
##  3  3  4  3  3  3  4  4  3  4  3  4  1  4  3  4  4  3  4  4
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
##  1  4  4  3  4  3  3  3  3  3  4  3  3  3  4  3  4  1  4  4
```

```
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
## 4 4 4 4 4 1 3 4 3 4 3 4 4 3 1 4 3 4 4 3
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
## 3 4 4 3 4 3 3 4 4 3 1 3 2 4 4 3 3 3 3 4
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
## 3 3 4 4 3 3 4 3 4 1 4 4 4 4 4 4 3 3 3 3
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 3 3 2 3 4 3 3 4 3 4 1 4 3 3 4 4 3 3 4 1
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 3 3 3 4 4 3 3 3 4 1 4 4 3 3 4 1 4 4 3 4
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 3 4 3 3 4 3 4 3 4 3 4 4 4 4 4 3 3 3 3 3
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
## 3 3 4 3 4 4 4 3 4 4 4 4 2 3 3 3 3 3 3
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
## 3 4 3 4 4 3 4 4 3 3 4 4 3 4 4 3 3 4 4 3
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 3 1 4 4 3 4 1 4 1 4 4 4 2 4 3 4 3 4 4 3
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
## 4 3 4 4 3 1 3 4 3 3 4 3 4 4 1 1 3 3 4 3
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 3 3 1 4 4 1 4 3 4 4 3 3 4 3 4 3 2 4 4 3
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
## 3 4 3 3 3 2 4 3 3 4 3 1 3 4 4 3 3 4 3 2
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 4 3 4 3 4 4 4 3 3 4 4 3 4 4 4 3 3 4 3 3
##
```

```
## Within cluster sum of squares by cluster:
```

```
## [1] 528.9751 113.6101 1118.6016 854.2970
```

```
## (between_SS / total_SS = 34.5 %)
```

```
##
```

```
## Available components:
```

```
##
```

```
## [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
```

```
## [6] "betweenss" "size" "iter" "ifault"
```

```
perc.var.4 <- round(100*(1 - kmeans4.num$betweenss/kmeans4.num$totss),1)
```

```
names(perc.var.4) <- "Perc. 4 clus"
```

```
perc.var.4
```

```
## Perc. 4 clus
```

```
## 65.5
```

```
# Computing the percentage of variation accounted for. Five clusters
```

```
(kmeans5.num <- kmeans(matstd.num,5,nstart = 10))
```

```
## K-means clustering with 5 clusters of sizes 37, 158, 16, 169, 120
```

```
##
```

```
## Cluster means:
```

```
## Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
```

```
## 1 -0.24559346 -0.21842156 -0.2094344 -0.7652081 -3.09931925
```

```
## 2 -0.57800420 -0.16871146 -0.5525952 0.7239109 0.25129884
```

```
## 3 0.03027763 2.91283925 -0.3041773 0.7462929 -0.04163422
```

```
## 4 -0.43458194 -0.08902735 0.7309719 -0.8053124 0.22195759
```

```
## 5 1.44476274 0.02648502 -0.1967358 0.3174324 0.31770759
```

```

## Visibility.mi. Wind_Speed.mph. Precipitation.in.
## 1      0.9466474      -0.03995983      -0.14353314
## 2      -0.2441005      -0.32441259      -0.05798137
## 3      -0.9723333       0.06162912       3.97679296
## 4       0.2277958       0.37654018      -0.21231879
## 5      -0.1616520      -0.09904711      -0.11062524
##
## Clustering vector:
## 1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20
## 4  4  5  2  4  2  2  5  2  5  4  5  4  4  1  3  2  2  4  2
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
## 5  4  2  4  1  5  2  2  2  4  5  3  2  5  2  1  5  5  2  2
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
## 4  2  5  1  2  5  2  4  4  2  4  2  2  2  4  1  4  5  1  5
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
## 1  2  2  4  2  5  5  1  5  4  2  2  5  5  2  5  4  2  1  3
## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
## 4  4  5  2  4  5  2  3  2  5  4  5  4  2  5  5  4  5  1  4
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
## 4  4  1  4  4  4  5  4  5  1  1  2  4  5  4  2  5  1  2  2
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
## 2  2  5  4  4  4  1  2  4  2  4  2  2  4  2  4  3  2  2  4
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
## 1  2  3  5  4  4  2  4  4  5  5  4  2  4  1  1  4  5  2  2
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
## 2  5  4  4  2  5  1  5  2  5  4  4  4  2  4  5  2  4  2  2
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
## 1  4  1  2  5  4  2  4  4  5  2  4  2  5  5  4  5  4  4  5
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
## 2  5  4  2  2  5  4  4  2  4  5  4  1  5  5  4  4  2  5  4
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
## 4  4  4  3  4  5  2  2  2  2  4  2  2  2  4  2  4  1  2  2
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
## 4  4  4  5  4  1  2  2  2  5  5  4  4  5  1  4  5  2  4  5
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
## 2  2  4  5  5  2  5  4  4  2  1  5  3  4  2  5  5  2  2  4
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
## 2  2  4  4  2  5  4  2  4  1  4  4  4  4  4  4  5  2  2  2
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 5  5  3  2  4  2  2  4  5  4  1  5  2  5  5  4  2  5  4  4
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 5  2  2  5  4  5  2  5  4  2  4  4  2  2  4  1  4  4  5  5
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 5  4  5  2  4  2  5  2  4  2  4  4  4  4  5  5  2  2  5  5
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
## 2  2  4  2  4  4  4  2  4  4  4  2  3  5  5  5  5  2  5  5
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
## 2  4  2  4  4  5  4  4  2  5  4  4  2  5  4  2  2  5  5  2
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 5  1  4  5  2  4  1  5  1  5  5  4  3  4  5  4  2  4  2  5
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
## 4  2  4  4  2  1  2  2  2  2  2  2  4  4  1  1  2  3  4  5
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 5  2  1  4  4  1  5  2  4  4  5  2  4  5  4  2  3  4  4  5

```

```

## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
##    2    4    2    5    2    3    5    2    2    4    2    3    5    4    2    2    2    2    5    3
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
##    4    5    4    2    4    4    2    2    2    4    4    5    5    4    4    2    2    4    5    2
##
## Within cluster sum of squares by cluster:
## [1] 416.0750 507.6989 452.8908 550.5556 441.7780
## (between_SS / total_SS = 40.7 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"
perc.var.5 <- round(100*(1 - kmeans5.num$betweenss/kmeans5.num$totss),1)
names(perc.var.5) <- "Perc. 5 clus"
perc.var.5

## Perc. 5 clus
##      59.3

# Computing the percentage of variation accounted for. Six clusters
(kmeans6.num <- kmeans(matstd.num,6,nstart = 10))

## K-means clustering with 6 clusters of sizes 6, 39, 107, 172, 166, 10
##
## Cluster means:
##      Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
## 1  0.3743416   6.70831691   -0.3958477  0.002130805   -0.3831819
## 2 -0.2608534  -0.22001382   -0.1216707 -0.829591959   -3.0212702
## 3  1.4461919   0.01938662    0.2930756 -0.100668833    0.2974282
## 4 -0.6178428  -0.11510118    0.6152846 -0.618210029    0.2721188
## 5 -0.2557514  -0.13382650   -0.7590297  0.831538521    0.2413979
## 6  0.1908408   0.82688707   -0.4068876  1.140959724    0.1427323
##      Visibility.mi. Wind_Speed.mph. Precipitation.in.
## 1   -0.09061981   -0.06205071    0.07692563
## 2    0.91111568   -0.10506978   -0.14749422
## 3    0.17411628   -0.01951409   -0.17737121
## 4    0.22824898    0.31311561   -0.21246627
## 5   -0.48307310   -0.29207316   -0.00341745
## 6   -1.26889254    0.11862939    6.13809348
##
## Clustering vector:
##    1    2    3    4    5    6    7    8    9   10   11   12   13   14   15   16   17   18   19   20
##    4    4    3    5    4    5    5    3    5    3    4    3    4    4    2    6    5    5    4    5
##   21   22   23   24   25   26   27   28   29   30   31   32   33   34   35   36   37   38   39   40
##    3    4    5    4    2    3    5    5    5    4    3    6    5    3    4    2    3    3    5    4
##   41   42   43   44   45   46   47   48   49   50   51   52   53   54   55   56   57   58   59   60
##    4    5    5    2    5    5    5    3    4    5    4    5    4    5    4    2    4    3    2    3
##   61   62   63   64   65   66   67   68   69   70   71   72   73   74   75   76   77   78   79   80
##    2    5    4    4    5    3    5    2    3    4    5    5    3    3    5    5    4    5    2    1
##   81   82   83   84   85   86   87   88   89   90   91   92   93   94   95   96   97   98   99   100
##    4    4    5    5    4    3    5    6    5    3    4    3    4    5    3    3    4    5    2    4
##  101  102  103  104  105  106  107  108  109  110  111  112  113  114  115  116  117  118  119  120
##    4    4    2    4    4    4    3    4    3    2    2    4    4    3    4    5    5    2    5    4

```

```
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
## 5 5 3 4 4 4 2 5 4 5 4 5 5 4 5 4 1 5 5 4
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
## 2 5 5 3 4 4 5 4 4 5 3 4 5 4 2 2 4 3 5 5
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
## 5 3 4 4 5 5 2 3 5 5 4 4 4 5 4 3 4 4 5 5
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
## 2 4 2 5 3 3 5 3 4 3 5 4 4 3 3 4 3 3 4 3
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
## 5 5 4 5 5 3 3 4 5 4 3 4 2 3 3 4 4 5 3 4
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
## 2 4 4 1 4 5 5 5 5 5 4 5 5 5 4 5 4 2 4 4
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
## 4 4 4 3 4 2 5 4 5 3 3 4 3 5 2 3 5 4 4 5
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
## 5 4 4 3 3 5 3 3 4 5 2 3 6 3 4 5 3 5 5 4
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
## 5 5 4 4 5 3 3 5 4 2 4 4 4 4 4 4 5 5 5 5
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 3 5 6 5 3 5 5 4 5 4 2 3 5 3 3 4 5 5 3 2
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 3 5 5 3 4 3 5 3 4 5 4 4 5 5 4 2 4 4 3 3
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 1 4 3 5 4 5 3 5 4 5 4 3 4 4 3 3 5 5 3 3
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
## 5 5 4 5 4 4 4 5 4 4 4 4 6 5 3 3 3 5 3 5
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
## 5 4 5 4 4 5 4 4 5 5 4 4 5 3 4 5 5 3 3 5
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 3 2 4 3 5 4 2 3 2 3 3 4 6 4 3 4 5 4 4 3
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
## 4 5 4 3 5 2 5 4 5 5 4 5 4 4 2 2 5 1 4 3
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 3 5 2 4 4 2 3 5 4 4 5 5 3 5 3 5 6 4 4 3
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
## 5 4 5 5 5 6 3 5 5 4 5 1 3 4 4 5 5 4 5 6
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 4 3 4 5 4 4 4 5 5 4 4 3 3 4 4 5 5 4 3 5
##
```

```
## Within cluster sum of squares by cluster:
## [1] 83.39929 433.56834 321.26540 484.60073 663.93247 113.61006
## (between_SS / total_SS = 47.4 %)
```

```
##
```

```
## Available components:
```

```
##
```

```
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"
```

```
perc.var.6 <- round(100*(1 - kmeans6.num$betweenss/kmeans6.num$totss),1)
names(perc.var.6) <- "Perc. 6 clus"
perc.var.6
```

```
## Perc. 6 clus
## 52.6
```

```
# Computing the percentage of variation accounted for. Seven clusters
(kmeans7.num <- kmeans(matstd.num,7,nstart = 10))
```

```
## K-means clustering with 7 clusters of sizes 149, 115, 12, 10, 114, 64, 36
##
## Cluster means:
##      Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
## 1 -0.5801087 -0.18488754 -0.51774848 0.71735931 0.26045732
## 2 1.4354547 -0.03825598 -0.18625016 0.35355089 0.32930384
## 3 0.2214243 4.77380390 0.08429905 -0.03386252 0.02608219
## 4 0.1908408 0.82688707 -0.40688757 1.14095972 0.14273235
## 5 -0.3661001 -0.15176808 0.86834199 -0.97347784 0.20572274
## 6 -0.5144903 -0.13288908 0.14990304 -0.30313761 0.17807339
## 7 -0.2373277 -0.21667704 -0.19345008 -0.78252376 -3.14631893
## Visibility.mi. Wind_Speed.mph. Precipitation.in.
## 1 -0.28630855 -0.44910686 -0.05534099
## 2 -0.17746808 -0.10177788 -0.10583616
## 3 0.06005446 -0.06922055 -0.07192427
## 4 -1.26889254 0.11862939 6.13809348
## 5 0.23112072 -0.25450445 -0.22077417
## 6 0.21745525 1.71383032 -0.15379171
## 7 0.96589378 -0.06683060 -0.14138756
##
## Clustering vector:
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## 6 5 2 1 6 1 1 2 1 2 5 2 5 5 7 4 1 1 6 1
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
## 2 5 1 5 7 2 1 1 1 5 2 4 1 2 1 7 2 2 3 1
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
## 5 1 2 7 1 2 1 5 5 1 6 1 1 1 6 7 5 2 7 5
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
## 7 1 6 6 1 2 2 7 2 5 6 1 2 2 6 2 5 1 7 3
## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
## 5 6 2 1 6 2 1 4 1 2 5 2 5 1 2 2 5 2 7 5
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
## 5 5 6 6 5 6 2 5 2 7 7 1 5 2 5 1 2 7 1 1
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
## 6 1 2 5 1 6 7 1 5 1 6 1 1 6 1 5 3 6 1 5
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
## 7 1 1 2 5 5 1 5 5 2 2 5 1 5 7 7 5 2 1 1
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
## 1 2 5 6 1 2 7 2 1 2 6 5 5 1 6 2 6 6 1 1
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
## 7 6 7 1 3 6 1 6 6 2 1 5 1 2 2 6 2 5 6 2
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
## 1 2 5 1 1 2 5 5 1 5 2 3 7 2 2 5 5 1 6 6
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
## 5 3 5 3 6 2 1 1 1 1 5 1 1 1 5 1 5 7 1 1
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
## 1 5 6 2 5 7 1 1 1 2 2 5 5 2 7 6 2 1 5 2
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
## 1 1 6 2 2 1 2 6 6 1 7 2 4 2 1 2 2 1 1 6
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
## 6 1 6 5 1 2 5 1 5 7 5 5 6 5 5 5 2 1 1 1
```

```

## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 2 2 4 1 5 1 1 6 2 6 7 2 1 2 2 5 1 2 6 5
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 2 1 1 2 5 2 1 2 6 1 5 5 1 1 5 7 5 6 2 5
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 3 5 2 1 5 1 2 1 6 1 5 3 5 6 5 2 1 1 2 2
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
## 1 1 5 1 5 6 6 1 5 6 5 1 4 2 2 2 2 1 2 2
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
## 1 5 1 5 3 2 5 5 1 2 5 5 1 2 5 1 1 2 2 1
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 2 7 5 2 1 6 7 2 7 2 2 6 4 5 2 5 1 5 1 2
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
## 6 1 6 5 1 7 6 1 1 1 1 1 5 5 7 7 1 3 6 2
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 2 1 7 5 5 7 2 6 5 6 2 1 5 2 5 1 4 5 5 2
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
## 1 5 1 2 1 4 2 1 1 5 1 3 2 6 5 1 1 1 2 4
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 5 2 6 6 5 5 1 1 1 5 5 2 2 6 5 1 1 5 2 1
##
## Within cluster sum of squares by cluster:
## [1] 435.1724 393.6987 145.0910 113.6101 218.7422 213.0439 410.8366
## (between_SS / total_SS = 51.6 %)
##
## Available components:
##
## [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
## [6] "betweenss" "size" "iter" "ifault"
perc.var.7 <- round(100*(1 - kmeans7.num$betweenss/kmeans7.num$totss),1)
names(perc.var.7) <- "Perc. 7 clus"
perc.var.7

## Perc. 7 clus
## 48.4

# Computing the percentage of variation accounted for. nine clusters
(kmeans9.num <- kmeans(matstd.num,9,nstart = 10))

## K-means clustering with 9 clusters of sizes 124, 10, 6, 34, 51, 138, 65, 68, 4
##
## Cluster means:
## Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
## 1 -0.6023561 -0.08516217 0.8282891 -0.831521130 0.2257714
## 2 0.1908408 0.82688707 -0.4068876 1.140959724 0.1427323
## 3 0.3743416 6.70831691 -0.3958477 0.002130805 -0.3831819
## 4 -0.3272790 -0.21259279 -0.2576174 -0.657605737 -3.0097296
## 5 -0.5791428 -0.07604851 -0.4314103 -0.061386834 0.1601000
## 6 -0.5697566 -0.19017656 -0.4453709 0.739837550 0.2981513
## 7 1.4047690 0.02233057 -0.6647676 0.856224807 0.3343610
## 8 1.4267725 -0.01345174 0.5897004 -0.534381518 0.2853177
## 9 0.3743416 -0.28610927 -0.2328782 -1.059672391 -3.8095333
## Visibility.mi. Wind_Speed.mph. Precipitation.in.
## 1 0.22878186 -0.07870325 -0.21213127

```



```

## 2    -1.26889254      0.11862939      6.13809348
## 3    -0.09061981     -0.06205071      0.07692563
## 4      0.11893137     -0.13037512     -0.11570365
## 5      0.03447782      1.86638459     -0.11570365
## 6     -0.24831306     -0.54486185     -0.06286384
## 7     -0.47343178     -0.12836626     -0.01742231
## 8      0.21579340      0.01639583     -0.22077417
## 9      7.35699377      0.15304465     -0.22077417
##
## Clustering vector:
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20
##  1  1  8  6  5  6  6  8  6  8  1  8  1  1  4  2  6  6  1  6
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
##  8  1  6  1  4  8  6  6  6  1  8  2  6  7  6  4  8  8  6  6
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
##  1  6  7  4  6  7  6  8  1  6  1  6  6  6  5  9  1  7  4  8
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
##  4  6  5  5  6  7  7  4  7  1  5  6  8  7  5  7  1  6  4  3
## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
##  1  5  7  6  5  8  6  2  6  7  1  8  1  6  7  7  1  7  4  1
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
##  1  1  5  1  1  1  8  1  8  4  4  6  1  8  1  6  7  4  6  6
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
##  5  6  8  1  1  5  4  6  1  6  5  6  6  1  6  1  3  5  6  1
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
##  4  6  6  8  1  1  6  1  1  7  8  1  6  1  4  4  1  7  6  6
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
##  5  7  1  5  6  7  4  7  6  7  5  1  1  6  5  8  5  5  6  6
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
##  4  5  4  5  7  8  6  5  1  8  6  1  6  7  8  5  8  8  5  8
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
##  5  7  1  6  6  8  8  1  6  1  7  1  4  8  7  1  1  6  5  5
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
##  1  1  1  3  5  7  6  6  6  6  1  6  6  6  1  6  1  9  6  6
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
##  1  1  5  8  1  9  6  6  6  8  7  1  8  7  4  8  7  6  1  7
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
##  6  1  1  7  8  6  8  8  1  6  9  8  2  8  6  7  7  6  5  1
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
##  5  6  1  1  6  7  8  4  1  4  1  1  5  1  1  1  7  6  6  6
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
##  8  7  2  6  8  6  6  5  7  5  4  8  6  7  8  1  6  7  8  1
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
##  7  5  6  8  1  7  6  7  5  4  1  1  6  6  1  4  1  1  7  8
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
##  3  1  8  6  1  6  8  6  5  6  1  8  1  5  8  7  6  6  8  7
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
##  5  6  1  6  1  5  1  6  1  1  1  6  2  7  7  7  7  6  7  7
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
##  6  1  6  1  1  7  1  1  6  7  1  1  6  8  1  6  6  8  8  6
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
##  7  4  1  8  6  5  4  8  4  8  8  5  2  1  8  1  6  1  6  7
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
##  5  6  5  8  6  4  5  6  6  6  6  6  1  1  4  4  6  3  1  7

```

```

## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
##   8   6   4   1   1   4   8   5   1   1   7   6   8   7   8   5   2   1   1   7
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
##   6   1   6   7   6   2   8   6   6   1   6   3   8   5   1   6   6   6   7   2
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
##   1   7   5   5   1   1   6   6   6   1   1   7   8   5   1   6   6   1   7   6
##
## Within cluster sum of squares by cluster:
## [1] 250.36466 113.61006 83.39929 179.46177 199.22258 353.07040 224.68787
## [8] 152.04291 57.42334
## (between_SS / total_SS = 59.6 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"
perc.var.9 <- round(100*(1 - kmeans9.num$betweenss/kmeans9.num$totss),1)
names(perc.var.9) <- "Perc. 9 clus"
perc.var.9

## Perc. 9 clus
##      40.4

# Computing the percentage of variation accounted for. ten clusters
(kmeans10.num <- kmeans(matstd.num,10,nstart = 10))

## K-means clustering with 10 clusters of sizes 96, 35, 8, 96, 55, 4, 51, 7, 115, 33
##
## Cluster means:
##      Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
## 1 -0.46670363 -0.08373242  0.87768023 -1.062821807  0.19339608
## 2 -0.33344712 -0.21013380 -0.29371012 -0.608316049 -2.97569649
## 3  0.37434157  6.31879532 -0.47930387  0.219890444 -0.26850890
## 4  1.46387740  0.06231273  0.29775845 -0.002818278  0.31919043
## 5  0.19084080 -0.10589224 -1.07687868  1.159414485  0.24698057
## 6  0.37434157 -0.28610927 -0.23287818 -1.059672391 -3.80953332
## 7 -0.61512337 -0.15293349  0.44074977 -0.478062544  0.23422267
## 8 -0.01887437 -0.24935445 -0.23813527  1.221790457  0.21277893
## 9 -0.59103205 -0.19851669 -0.08773703  0.587538822  0.37029042
## 10 0.01290066 -0.05443549 -1.49362185  0.320180946  0.08257993
##      Visibility.mi. Wind_Speed.mph. Precipitation.in.
## 1      0.2268724      -0.262806373      -0.22077417
## 2      0.1227841      -0.149317966      -0.11870567
## 3     -0.2466753      -0.085711197       1.43145970
## 4      0.1596071     -0.192900382     -0.17239795
## 5     -1.6414298      0.002868981      0.20791354
## 6      7.3569938      0.153044647     -0.22077417
## 7      0.2284551      1.713540361     -0.21376947
## 8     -1.3495186      0.204667532       6.71988394
## 9      0.1302443     -0.649728626     -0.12136832
## 10     0.1285428      1.054098644     -0.06921791
##
## Clustering vector:
##   1   2   3   4   5   6   7   8   9  10  11  12  13  14  15  16  17  18  19  20

```

```

## 7 1 4 9 10 5 10 4 9 4 1 4 1 1 2 8 9 9 7 9
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
## 4 1 9 1 2 4 9 9 9 1 4 8 5 4 9 2 4 4 9 9
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
## 1 10 5 2 9 5 9 1 1 9 7 9 9 5 7 6 1 10 2 4
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
## 2 9 10 7 5 4 5 2 4 1 10 9 4 4 10 5 9 9 2 3
## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
## 1 7 5 9 7 4 5 3 5 10 1 4 9 5 4 10 1 5 2 9
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
## 1 1 10 7 1 7 4 1 4 2 2 9 9 4 1 9 5 2 9 9
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
## 10 9 4 9 10 7 2 9 7 9 7 9 9 7 5 1 3 10 9 1
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
## 2 9 5 4 1 1 5 1 1 5 4 1 9 1 2 2 1 4 5 9
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
## 10 4 1 7 5 5 2 10 9 5 7 1 1 9 7 4 7 7 5 9
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
## 2 7 2 10 4 4 10 7 7 4 5 1 9 4 4 7 4 4 7 4
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
## 5 5 9 10 5 4 4 1 9 1 4 1 2 4 4 1 1 5 7 7
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
## 1 1 7 3 7 5 9 9 5 5 1 9 9 9 9 9 1 6 9 9
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
## 9 1 7 4 1 6 9 9 9 4 4 1 4 10 2 4 5 9 9 5
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
## 9 9 7 4 4 9 4 4 7 9 6 4 8 4 9 5 10 9 5 7
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
## 10 10 7 1 9 4 1 2 1 2 1 1 7 1 1 1 10 9 10 9
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 4 4 8 9 4 9 9 7 5 7 2 4 9 4 4 1 9 10 7 1
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 4 10 9 4 1 4 5 4 7 2 1 1 9 5 1 2 1 7 4 4
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 3 1 4 9 1 9 4 9 7 9 9 4 1 7 4 10 9 5 4 4
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
## 10 9 1 9 1 7 7 9 1 7 1 9 8 5 4 4 4 5 4 5
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
## 9 1 9 1 1 5 1 1 9 4 1 1 9 4 1 9 9 4 4 9
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 4 2 1 4 9 10 2 4 2 4 4 7 3 1 4 1 9 1 9 4
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
## 7 9 7 1 9 2 10 9 9 5 9 5 1 1 2 2 9 3 7 4
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 4 2 2 1 9 2 4 10 1 7 5 5 1 5 4 5 8 1 1 4
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
## 9 1 5 10 9 8 4 9 5 1 9 3 4 7 9 9 5 9 5 5
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 1 4 7 10 1 1 9 9 9 1 1 4 4 7 1 5 9 1 4 9
##
## Within cluster sum of squares by cluster:
## [1] 180.70262 185.78922 138.76397 254.56009 191.38862 57.42334 137.29726
## [8] 46.64630 180.59779 105.38141
## (between_SS / total_SS = 63.0 %)

```

```
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"
perc.var.10<- round(100*(1 - kmeans10.num$betweenss/kmeans10.num$totss),1)
names(perc.var.10) <- "Perc. 10 clus"
perc.var.10

## Perc. 10 clus
##          37

# Computing the percentage of variation accounted for. twelve clusters
(kmeans12.num <- kmeans(matstd.num,12,nstart = 10))

## K-means clustering with 12 clusters of sizes 9, 4, 5, 33, 115, 54, 1, 34, 50, 52, 92, 51
##
## Cluster means:
##      Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
## 1  0.06850695  0.95055333  -0.34736573  1.14431910  0.18227478
## 2  0.83309350 -0.28610927   0.12197480 -1.07047039 -3.80360195
## 3 -0.17616074  7.24676722  -0.08673133 -0.09001212 -0.41560673
## 4  0.01290066 -0.05443549  -1.49362185  0.32018095  0.08257993
## 5 -0.59103205 -0.19851669  -0.08773703  0.58753882  0.37029042
## 6  1.59768005  0.21341075  -0.09531789  0.59482095  0.39873295
## 7 -0.54316228 -0.28610927  -0.67973009 -1.04887439 -3.72451710
## 8 -0.38124984 -0.20789923  -0.32410401 -0.59408810 -2.95437016
## 9 -0.65326274 -0.15026998  0.41794847 -0.46319095  0.24095574
## 10 0.12732130 -0.09580825  -1.09908342  1.17136024  0.22531519
## 11 -0.54316228 -0.07610161  0.85510906 -1.02493100  0.19310596
## 12 1.32782597 -0.07237667   0.72318904 -0.74568353  0.22460843
##      Visibility.mi. Wind_Speed.mph. Precipitation.in.
## 1  -1.20847934   0.12436527   6.44770127
## 2   4.77400640   0.19606372  -0.22077417
## 3  -0.15949947  -0.01903164   0.13646559
## 4   0.12854276   1.05409864  -0.06921791
## 5   0.13024433  -0.64972863  -0.12136832
## 6   0.03374625  -0.43551998  -0.13477201
## 7  13.16871534   0.06700651  -0.22077417
## 8   0.04296116  -0.16074152  -0.11570365
## 9   0.22794864   1.70517273  -0.21362937
## 10 -1.68122674   0.03788590   0.23264552
## 11  0.22570256  -0.27901646  -0.22077417
## 12  0.23858447   0.10749504  -0.22077417
##
## Clustering vector:
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20
##  9 11 12  5  4 10  4 12  5 12 11  6 11 11  8  1  5  5  9  5
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
##  6 11  5 11  2 12  5  5  5 11 12  1 10  6  5  8  6 12  5  5
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
## 11  4 10  8  5 10  5 12 11  5  9  5  5 10  9  2 11  4  8 12
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
##  8  5  4  9 10  6 10  8  6 11  4  5  6  6  4 10  5  5  8  3
```

```

## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
## 11 9 10 5 9 12 10 1 10 4 11 6 5 10 6 4 11 10 8 5
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
## 11 11 4 9 11 9 6 11 6 8 8 5 5 12 11 5 6 8 5 5
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
## 4 5 12 5 4 9 8 5 9 5 9 5 5 9 10 11 3 4 5 11
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
## 8 5 10 12 11 11 10 11 11 10 12 11 5 11 8 8 11 6 10 5
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
## 4 6 11 9 10 10 8 4 5 10 9 11 11 5 9 12 9 9 10 5
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
## 8 9 8 4 6 12 4 9 9 6 10 11 5 6 12 9 12 12 9 6
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
## 10 10 5 4 10 6 12 11 5 11 6 11 8 12 6 11 11 10 9 9
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
## 11 11 9 3 9 10 5 5 10 10 11 5 5 5 5 5 11 2 5 5
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
## 5 11 9 12 11 2 5 5 5 12 6 11 12 4 8 12 10 5 5 10
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
## 5 5 9 6 12 5 6 12 9 5 7 6 1 12 5 10 4 5 10 9
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
## 4 4 9 11 5 6 12 8 11 8 11 11 9 11 11 11 4 5 4 5
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 12 6 1 5 12 5 5 9 10 9 8 12 5 6 12 11 5 4 12 11
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 6 4 5 12 11 6 10 6 9 8 11 11 5 10 11 8 11 9 6 12
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 6 11 6 5 11 5 12 5 9 5 5 12 11 9 12 4 5 10 6 6
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
## 4 5 11 5 11 9 9 5 11 9 11 5 1 10 6 6 6 10 6 10
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
## 5 11 5 11 11 10 11 11 5 6 11 11 5 12 11 5 5 12 12 5
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 6 8 11 12 5 4 8 12 8 12 12 9 1 11 6 11 5 11 5 6
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
## 9 5 9 12 5 8 4 5 5 10 5 10 11 11 8 8 5 3 9 6
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 6 8 8 11 5 8 12 4 11 9 10 10 12 6 12 10 1 11 11 6
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
## 5 11 10 4 5 1 12 5 10 11 5 3 6 9 5 5 10 5 6 10
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 11 6 9 4 11 11 5 5 5 11 11 6 12 9 11 10 5 11 6 5
##
## Within cluster sum of squares by cluster:
## [1] 101.46760 16.79062 62.15281 105.38141 180.59779 142.70698 0.00000
## [8] 173.43423 131.38535 181.13044 158.43653 103.36376
## (between_SS / total_SS = 66.0 %)
##
## Available components:
##
## [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
## [6] "betweenss" "size" "iter" "ifault"

```

```
perc.var.12 <- round(100*(1 - kmeans12.num$betweenss/kmeans12.num$totss),1)
names(perc.var.12) <- "Perc. 12 clus"
perc.var.12
```

```
## Perc. 12 clus
##          34
```

```
# Computing the percentage of variation accounted for. fifteen clusters
(kmeans15.num <- kmeans(matstd.num,15,nstart = 10))
```

```
## K-means clustering with 15 clusters of sizes 53, 40, 42, 4, 8, 49, 51, 47, 52, 15, 28, 16, 67, 21, 7
##
```

```
## Cluster means:
```

```
##      Severity Distance.mi. Temperature.F. Humidity... Pressure.in.
## 1  1.56882772  0.141662554    -0.0604856   0.5973111   0.41042712
## 2  -0.72666305 -0.175460984     0.6542543  -0.4517451   0.27954915
## 3  -0.54316228 -0.064107068     1.2816619  -1.2473519   0.14913445
## 4   0.37434157 -0.286109267    -0.2328782  -1.0596724  -3.80953332
## 5   0.37434157  6.318795315    -0.4793039   0.2198904  -0.26850890
## 6  -0.61806056  0.008826541     0.4789735   0.5897770   0.37788875
## 7   1.32782597 -0.072376673     0.7231890  -0.7456835   0.22460843
## 8  -0.58220500 -0.239371355     0.1949587  -0.8264816   0.32311312
## 9   0.12732130 -0.095808248    -1.0990834   1.1713602   0.22531519
## 10 -0.42082843 -0.034253472    -0.3327627  -0.4384275   0.13429663
## 11  0.11219761 -0.063486737    -1.5901439   0.3163011   0.04557095
## 12 -0.31378632 -0.250997367    -1.2810088   0.4871409  -3.06267571
## 13 -0.54316228 -0.236139937    -0.3870597   0.6394813   0.35578931
## 14 -0.36839964 -0.182746036     0.6295335  -1.5774621  -2.77625202
## 15 -0.01887437 -0.249354450    -0.2381353   1.2217905   0.21277893
```

```
##      Visibility.mi. Wind_Speed.mph. Precipitation.in.
```

```
## 1      0.0295947    -0.43850816    -0.133149323
## 2      0.2344061     1.18980426    -0.220774169
## 3      0.1922788    -0.09687567    -0.220774169
## 4      7.3569938     0.15304465    -0.220774169
## 5     -0.2466753    -0.08571120     1.431459701
## 6      0.1114506    -0.09804626    -0.002055951
## 7      0.2385845     0.10749504    -0.220774169
## 8      0.2372914    -0.54075229    -0.220774169
## 9     -1.6812267     0.03788590     0.232645521
## 10     0.1676789     3.02098272    -0.101694250
## 11     0.1338541     0.96917788    -0.131464230
## 12     -0.1942084    -0.41695804     0.002500678
## 13     0.1573984    -0.94747308    -0.194114486
## 14     0.3767779    -0.05590513    -0.220774169
## 15     -1.3495186     0.20466753     6.719883942
```

```
##
```

```
## Clustering vector:
```

```
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20
##  2  3  7  6  2  9 11  7 13  7  8  1  3  3 12 15 13 13  2 13
## 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
##  1  8 13  6 14  7  6 13 13  3  7 15  9  1  6 14  1  7  6  6
## 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
##  2 11  9 14 13  9 13  7  3 13  2 13  6  9  2  4  3 11 12  7
## 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
```

```

## 12 13 11 2 9 1 9 14 1 3 10 6 1 1 10 9 6 13 14 5
## 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
## 3 2 9 6 2 7 9 5 9 11 6 1 6 9 1 11 3 9 12 6
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
## 3 3 11 2 3 2 1 8 1 14 14 8 6 7 3 13 1 14 13 6
## 121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140
## 11 13 7 8 8 10 14 13 2 6 2 13 13 2 9 8 5 11 13 8
## 141 142 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157 158 159 160
## 14 13 9 7 3 3 9 3 3 9 7 8 13 3 14 14 3 1 9 13
## 161 162 163 164 165 166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
## 11 1 8 10 9 9 14 11 13 9 2 3 3 13 2 7 6 10 9 6
## 181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 196 197 198 199 200
## 12 10 12 11 1 7 11 10 2 1 9 3 6 1 7 10 7 7 10 1
## 201 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219 220
## 9 9 6 11 9 1 7 8 8 6 1 6 14 7 1 8 8 9 10 2
## 221 222 223 224 225 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
## 14 3 2 5 10 9 13 13 9 9 3 13 13 13 8 13 8 4 8 8
## 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 256 257 258 259 260
## 6 8 2 7 8 4 13 6 13 7 1 6 7 11 12 7 9 6 6 9
## 261 262 263 264 265 266 267 268 269 270 271 272 273 274 275 276 277 278 279 280
## 6 6 2 1 7 13 1 7 2 13 4 1 15 7 6 9 11 13 9 2
## 281 282 283 284 285 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
## 11 11 2 8 13 1 7 12 8 14 3 3 2 8 8 8 11 13 6 13
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320
## 7 1 15 6 7 13 13 2 9 2 12 7 13 1 7 3 13 11 7 14
## 321 322 323 324 325 326 327 328 329 330 331 332 333 334 335 336 337 338 339 340
## 1 11 6 7 3 1 9 1 10 12 3 6 13 9 3 14 3 2 1 7
## 341 342 343 344 345 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
## 5 2 1 13 3 6 7 13 10 6 6 7 3 2 7 11 13 9 1 1
## 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375 376 377 378 379 380
## 11 13 8 13 8 2 2 6 3 2 8 6 15 9 1 1 1 9 1 9
## 381 382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399 400
## 6 8 6 8 6 9 2 8 6 1 3 8 13 7 8 13 13 7 7 13
## 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
## 1 12 8 7 13 11 12 7 14 7 7 10 5 8 1 3 13 8 6 1
## 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 436 437 438 439 440
## 2 13 2 7 13 14 11 6 13 9 8 9 8 8 14 12 13 5 2 1
## 441 442 443 444 445 446 447 448 449 450 451 452 453 454 455 456 457 458 459 460
## 1 12 12 8 6 12 7 11 8 2 9 9 7 1 7 9 15 8 8 1
## 461 462 463 464 465 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
## 13 3 9 11 13 15 7 13 9 3 13 5 1 2 6 13 9 8 1 9
## 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 496 497 498 499 500
## 8 1 2 11 3 3 8 13 6 3 8 1 7 10 3 9 13 6 1 13
##
## Within cluster sum of squares by cluster:
## [1] 121.62749 62.46288 60.03897 57.42334 138.76397 91.40352 103.36376
## [8] 43.43998 181.13044 43.54383 82.30302 53.86346 69.90284 71.31594
## [15] 46.64630
## (between_SS / total_SS = 69.3 %)
##
## Available components:
##
## [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
## [6] "betweenss" "size" "iter" "ifault"

```

```
perc.var.15 <- round(100*(1 - kmeans15.num$betweenss/kmeans15.num$totss),1)
names(perc.var.15) <- "Perc. 15 clus"
perc.var.15
```

```
## Perc. 15 clus
##          30.7
```

```
# Saving four k-means clusters in a list
```

```
clus1 <- matrix(names(kmeans4.num$cluster[kmeans4.num$cluster == 1]),
               ncol=1, nrow=length(kmeans4.num$cluster[kmeans4.num$cluster == 1]))
colnames(clus1) <- "Cluster 1"
clus2 <- matrix(names(kmeans4.num$cluster[kmeans4.num$cluster == 2]),
               ncol=1, nrow=length(kmeans4.num$cluster[kmeans4.num$cluster == 2]))
colnames(clus2) <- "Cluster 2"
clus3 <- matrix(names(kmeans4.num$cluster[kmeans4.num$cluster == 3]),
               ncol=1, nrow=length(kmeans4.num$cluster[kmeans4.num$cluster == 3]))
colnames(clus3) <- "Cluster 3"
clus4 <- matrix(names(kmeans4.num$cluster[kmeans4.num$cluster == 4]),
               ncol=1, nrow=length(kmeans4.num$cluster[kmeans4.num$cluster == 4]))
colnames(clus4) <- "Cluster 4"
list(clus1,clus2,clus3,clus4)
```

```
## [[1]]
##      Cluster 1
## [1,] "15"
## [2,] "25"
## [3,] "36"
## [4,] "44"
## [5,] "56"
## [6,] "59"
## [7,] "61"
## [8,] "68"
## [9,] "79"
## [10,] "99"
## [11,] "103"
## [12,] "110"
## [13,] "111"
## [14,] "118"
## [15,] "127"
## [16,] "141"
## [17,] "155"
## [18,] "156"
## [19,] "167"
## [20,] "181"
## [21,] "183"
## [22,] "213"
## [23,] "221"
## [24,] "238"
## [25,] "246"
## [26,] "255"
## [27,] "271"
## [28,] "290"
## [29,] "311"
## [30,] "320"
```



```

## [31,] "330"
## [32,] "336"
## [33,] "402"
## [34,] "407"
## [35,] "409"
## [36,] "426"
## [37,] "435"
## [38,] "436"
## [39,] "443"
## [40,] "446"
## [41,] "472"
##
## [[2]]
##      Cluster 2
## [1,] "16"
## [2,] "32"
## [3,] "88"
## [4,] "273"
## [5,] "303"
## [6,] "373"
## [7,] "413"
## [8,] "457"
## [9,] "466"
## [10,] "480"
##
## [[3]]
##      Cluster 3
## [1,] "4"
## [2,] "6"
## [3,] "7"
## [4,] "9"
## [5,] "10"
## [6,] "12"
## [7,] "17"
## [8,] "18"
## [9,] "20"
## [10,] "21"
## [11,] "23"
## [12,] "27"
## [13,] "28"
## [14,] "29"
## [15,] "33"
## [16,] "34"
## [17,] "37"
## [18,] "39"
## [19,] "42"
## [20,] "43"
## [21,] "45"
## [22,] "46"
## [23,] "47"
## [24,] "50"
## [25,] "52"
## [26,] "54"
## [27,] "58"

```

```
## [28,] "62"
## [29,] "63"
## [30,] "65"
## [31,] "66"
## [32,] "67"
## [33,] "69"
## [34,] "71"
## [35,] "72"
## [36,] "73"
## [37,] "74"
## [38,] "75"
## [39,] "76"
## [40,] "78"
## [41,] "83"
## [42,] "84"
## [43,] "87"
## [44,] "89"
## [45,] "90"
## [46,] "92"
## [47,] "94"
## [48,] "95"
## [49,] "96"
## [50,] "98"
## [51,] "107"
## [52,] "109"
## [53,] "116"
## [54,] "117"
## [55,] "119"
## [56,] "121"
## [57,] "122"
## [58,] "128"
## [59,] "130"
## [60,] "132"
## [61,] "133"
## [62,] "135"
## [63,] "137"
## [64,] "138"
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## [66,] "142"
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## [68,] "147"
## [69,] "150"
## [70,] "153"
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## [73,] "160"
## [74,] "161"
## [75,] "162"
## [76,] "165"
## [77,] "166"
## [78,] "168"
## [79,] "169"
## [80,] "170"
## [81,] "174"
```

```
## [82,] "177"  
## [83,] "179"  
## [84,] "180"  
## [85,] "184"  
## [86,] "185"  
## [87,] "187"  
## [88,] "191"  
## [89,] "194"  
## [90,] "200"  
## [91,] "201"  
## [92,] "202"  
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## [96,] "209"  
## [97,] "211"  
## [98,] "215"  
## [99,] "218"  
## [100,] "224"  
## [101,] "226"  
## [102,] "227"  
## [103,] "228"  
## [104,] "229"  
## [105,] "230"  
## [106,] "232"  
## [107,] "233"  
## [108,] "234"  
## [109,] "236"  
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## [113,] "254"  
## [114,] "257"  
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## [116,] "261"  
## [117,] "264"  
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## [119,] "267"  
## [120,] "270"  
## [121,] "272"  
## [122,] "276"  
## [123,] "277"  
## [124,] "278"  
## [125,] "279"  
## [126,] "281"  
## [127,] "282"  
## [128,] "285"  
## [129,] "286"  
## [130,] "288"  
## [131,] "297"  
## [132,] "298"  
## [133,] "299"  
## [134,] "300"  
## [135,] "301"
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[140,] "309"
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[178,] "386"
[179,] "389"
[180,] "390"
[181,] "393"
[182,] "396"
[183,] "397"
[184,] "400"
[185,] "401"
[186,] "405"
[187,] "415"
[188,] "417"
[189,] "420"

```

## [190,] "422"
## [191,] "425"
## [192,] "427"
## [193,] "429"
## [194,] "430"
## [195,] "432"
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## [220,] "488"
## [221,] "489"
## [222,] "492"
## [223,] "496"
## [224,] "497"
## [225,] "499"
## [226,] "500"
##
## [[4]]
##      Cluster 4
##      [1,] "1"
##      [2,] "2"
##      [3,] "3"
##      [4,] "5"
##      [5,] "8"
##      [6,] "11"
##      [7,] "13"
##      [8,] "14"
##      [9,] "19"
##     [10,] "22"
##     [11,] "24"
##     [12,] "26"
##     [13,] "30"
##     [14,] "31"

```

```
## [15,] "35"
## [16,] "38"
## [17,] "40"
## [18,] "41"
## [19,] "48"
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## [66,] "157"
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## [68,] "164"
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```
## [69,] "171"
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## [113,] "253"
## [114,] "256"
## [115,] "258"
## [116,] "259"
## [117,] "262"
## [118,] "263"
## [119,] "265"
## [120,] "268"
## [121,] "269"
## [122,] "274"
```

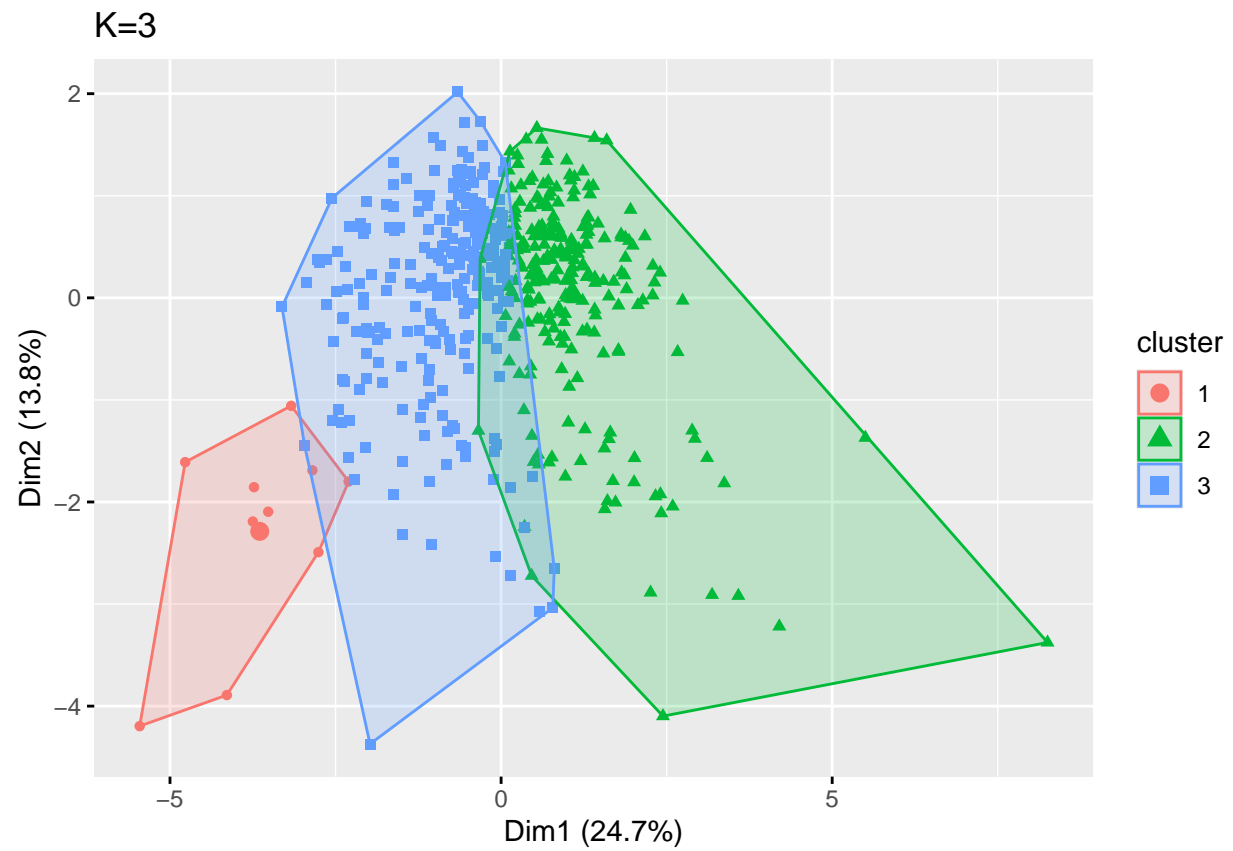
[123,] "275"
[124,] "280"
[125,] "283"
[126,] "284"
[127,] "287"
[128,] "289"
[129,] "291"
[130,] "292"
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[173,] "391"
[174,] "392"
[175,] "394"
[176,] "395"


```
## [177,] "398"
## [178,] "399"
## [179,] "403"
## [180,] "404"
## [181,] "406"
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## [204,] "455"
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## [207,] "462"
## [208,] "467"
## [209,] "470"
## [210,] "474"
## [211,] "475"
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## [214,] "483"
## [215,] "485"
## [216,] "486"
## [217,] "487"
## [218,] "490"
## [219,] "491"
## [220,] "493"
## [221,] "494"
## [222,] "495"
## [223,] "498"
```

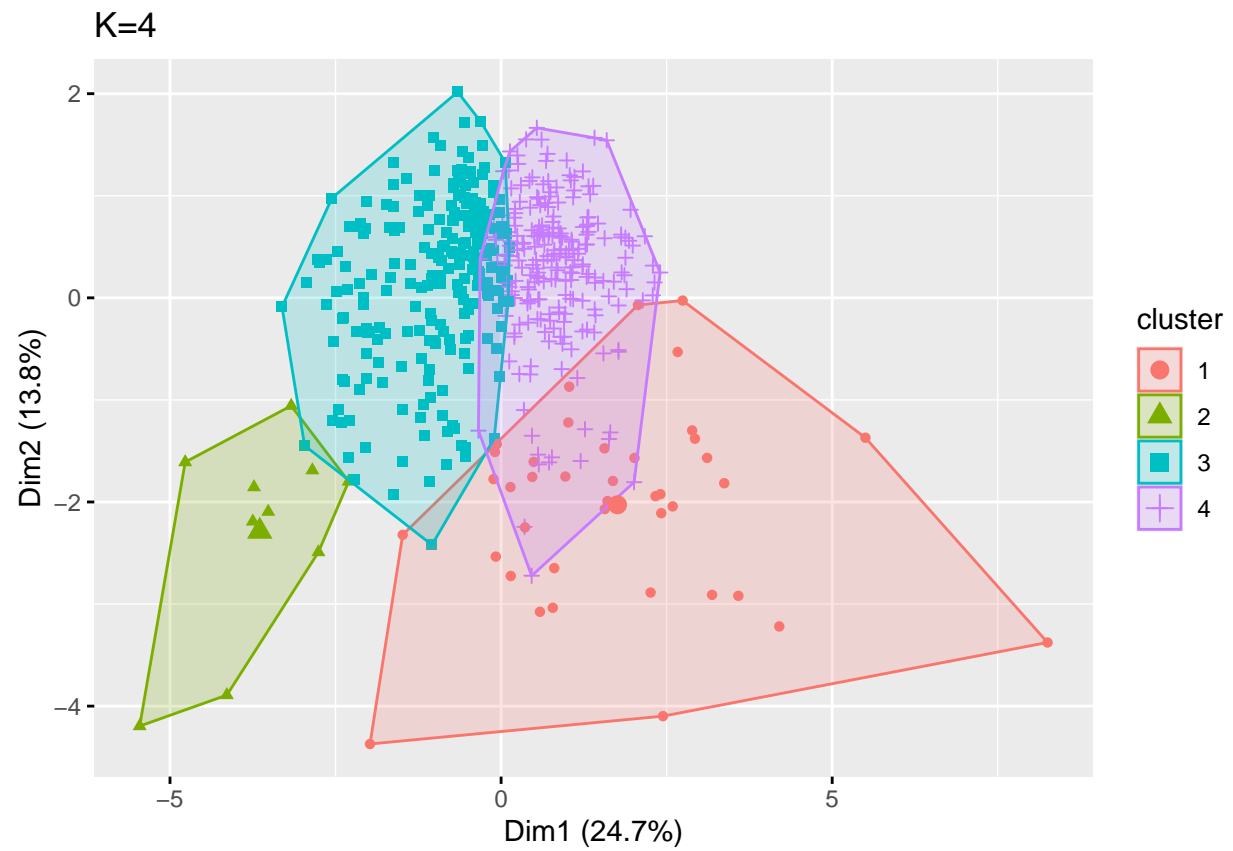
```
# Comparing the effect of number of clusters
fviz_cluster(kmeans2.num,geom = "point",data=matstd.num)+ggtitle("K=2")
```



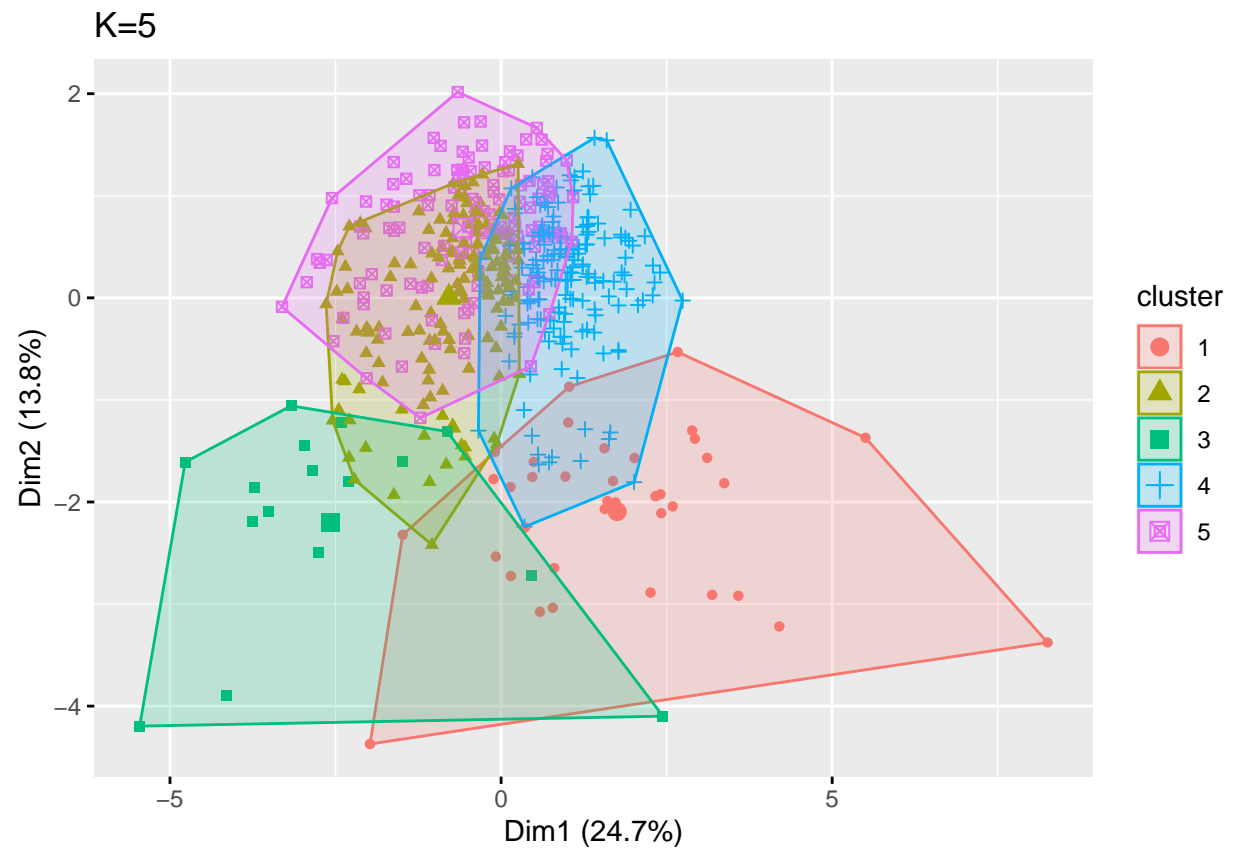
```
fviz_cluster(kmeans3.num,geom = "point",data=matstd.num)+ggtitle("K=3")
```



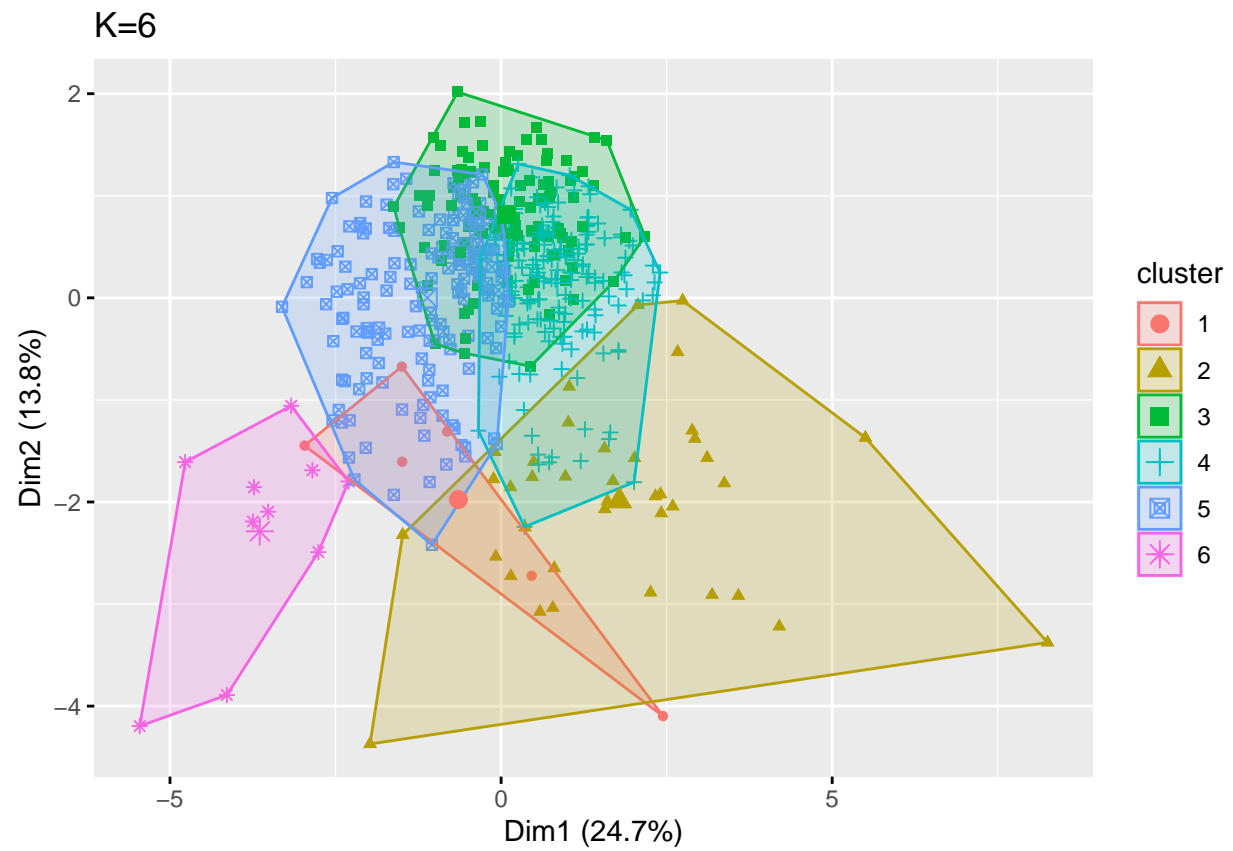
```
fviz_cluster(kmeans4.num,geom = "point",data=matstd.num)+ggtitle("K=4")
```



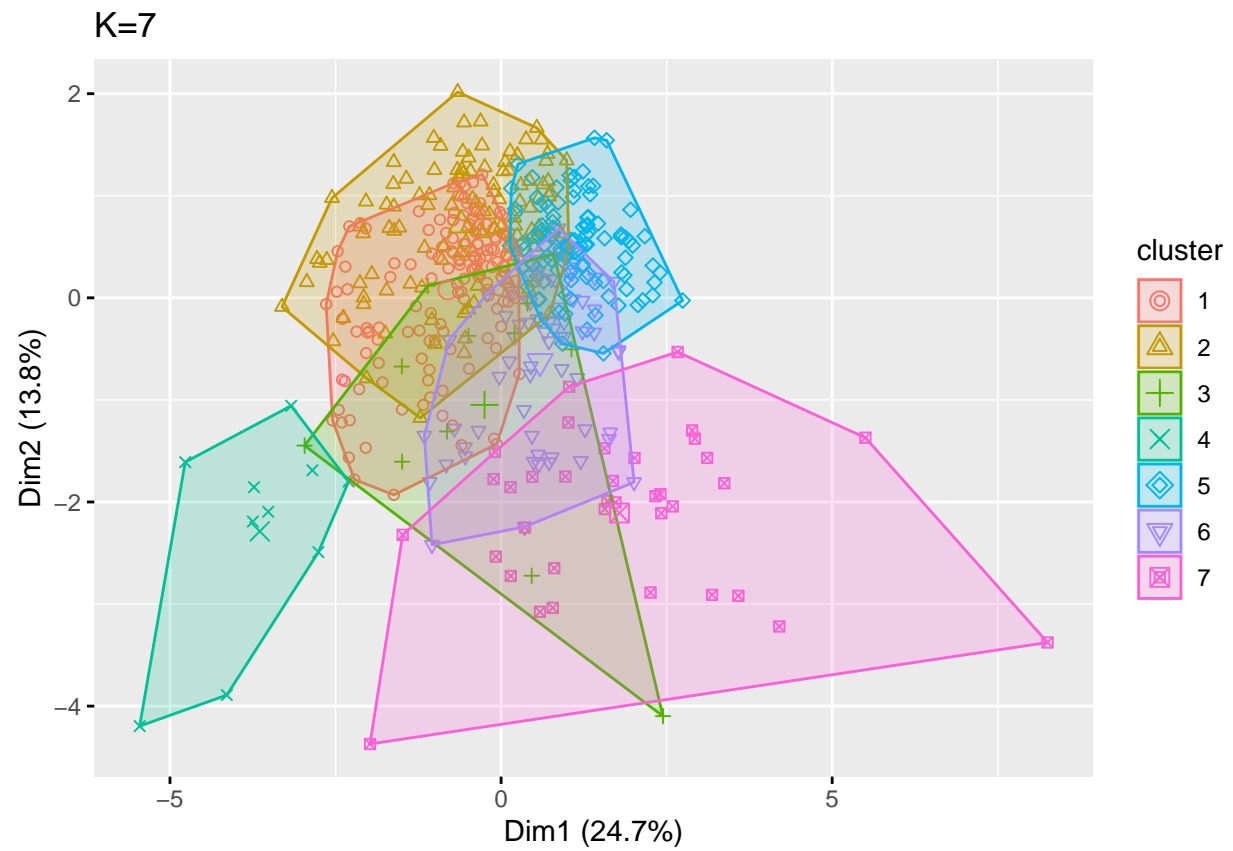
```
fviz_cluster(kmeans5.num,geom = "point",data=matstd.num)+ggtitle("K=5")
```



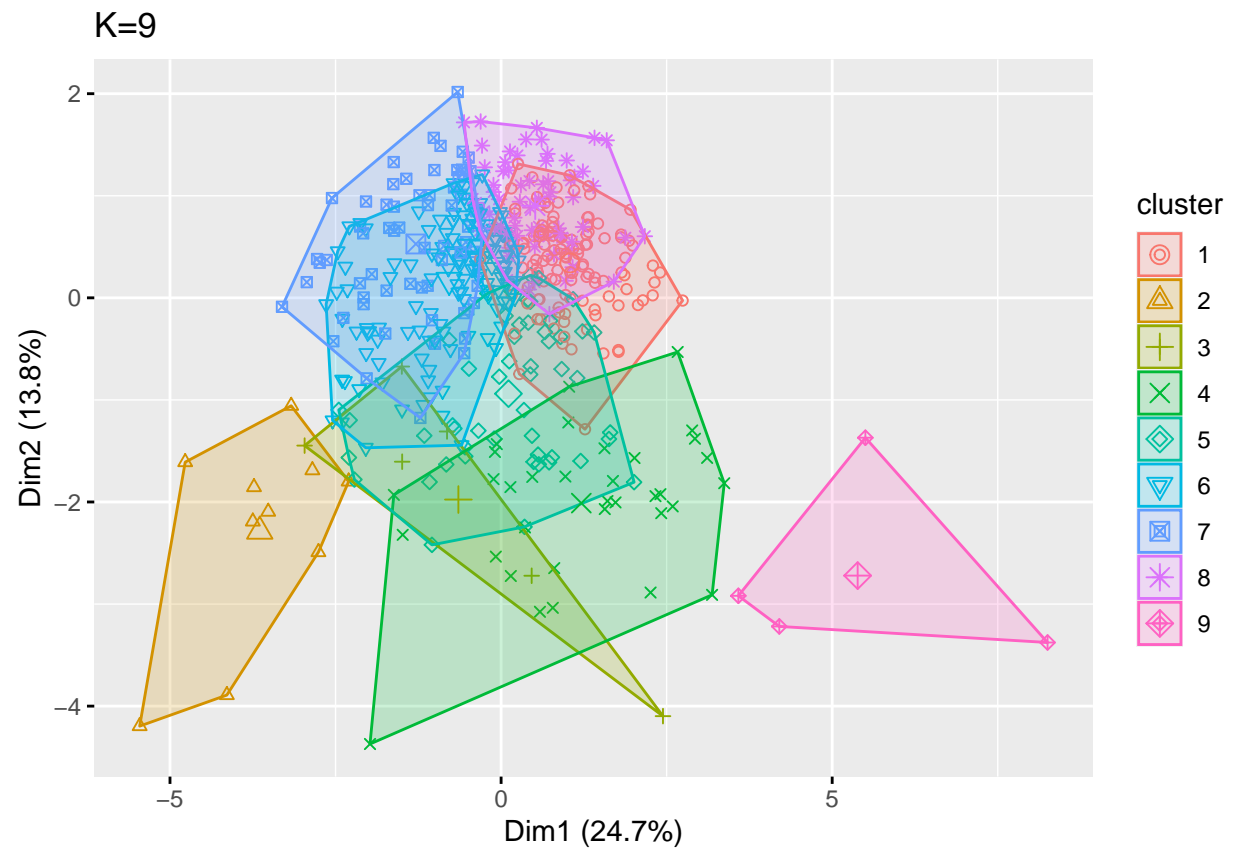
```
fviz_cluster(kmeans6.num,geom = "point",data=matstd.num)+ggtitle("K=6")
```



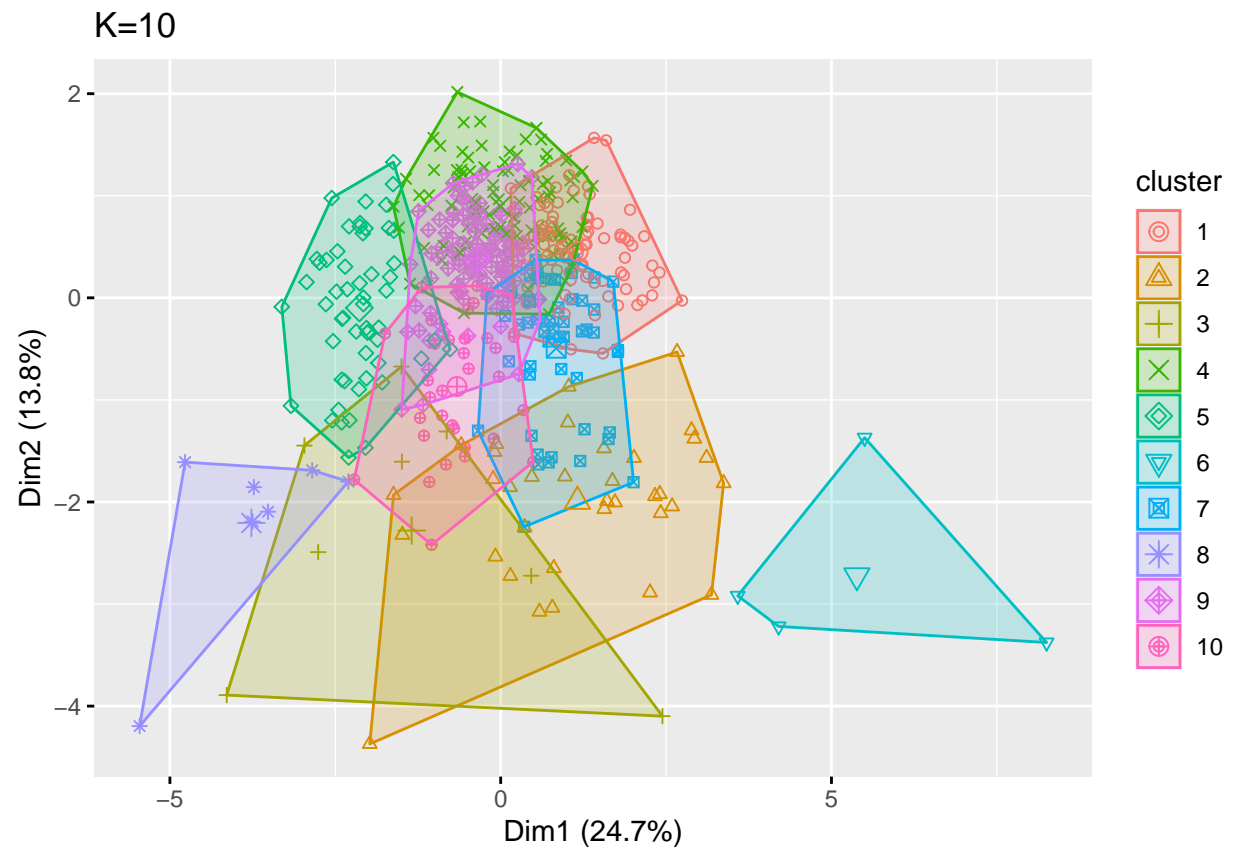
```
fviz_cluster(kmeans7.num,geom = "point",data=matstd.num)+ggtitle("K=7")
```



```
fviz_cluster(kmeans9.num,geom = "point",data=matstd.num)+ggtitle("K=9")
```



```
fviz_cluster(kmeans10.num,geom = "point",data=matstd.num)+ggtitle("K=10")
```

```
fviz_cluster(kmeans12.num,geom = "point",data=matstd.num)+ggtitle("K=12")
```



```
fviz_cluster(kmeans15.num,geom = "point",data=matstd.num)+ggtitle("K=15")
```

