Clustering Algorithm

Doe Emmanuel

2/18/2021

```
#####Clustering with k-means -----
## Example: Finding Teen Market Segments ----
## Step 2: Exploring and preparing the data ----
teens <- read.csv("C:\\Users\\EMMANUEL\\Desktop\\NIIT\\MACHINE LEARNING\\snsdata.csv")
str(teens)
                 30000 obs. of 40 variables:
## 'data.frame':
                     $ gradyear
               : int
                     "M" "F" "M" "F" ...
   $ gender
               : chr
##
   $ age
               : num
                     19 18.8 18.3 18.9 19 ...
                     7 0 69 0 10 142 72 17 52 39 ...
##
   $ friends
               : int
## $ basketball : int 0000000000...
## $ football
              : int 0 1 1 0 0 0 0 0 0 0 ...
## $ soccer
              : int 0000000000...
## $ softball
               : int 000000100...
  $ volleyball : int 0000000000...
## $ swimming
               : int
                     0 0 0 0 0 0 0 0 0 0 ...
##
   $ cheerleading: int
                     0 0 0 0 0 0 0 0 0 0 ...
##
   $ baseball : int 0000000000...
## $ tennis
              : int 0000000000...
## $ sports
              : int
                     0 0 0 0 0 0 0 0 0 0 ...
##
   $ cute
              : int
                     0 1 0 1 0 0 0 0 0 1 ...
## $ sex
              : int 0000110200...
## $ sexy
              : int
                     0 0 0 0 0 0 0 1 0 0 ...
##
   $ hot
              : int
                     0 0 0 0 0 0 0 0 0 1 ...
##
   $ kissed
              : int
                     0 0 0 0 5 0 0 0 0 0 ...
##
              : int
                    1000100000...
  $ dance
   $ band
              : int
                     0 0 2 0 1 0 1 0 0 0 ...
##
   $ marching
              : int
                     0 0 0 0 0 1 1 0 0 0 ...
##
   $ music
              : int
                     0 2 1 0 3 2 0 1 0 1 ...
##
                     0 2 0 1 0 0 0 1 0 1 ...
  $ rock
              : int
##
   $ god
               : int
                     0 1 0 0 1 0 0 0 0 6 ...
                     0 0 0 0 0 0 0 0 0 0 ...
##
   $ church
               : int
##
   $ jesus
              : int 00000000000...
## $ bible
              : int
                     0000000000...
## $ hair
                     0600100001...
               : int
## $ dress
               : int
                     0 4 0 0 0 1 0 0 0 0 ...
              : int 0000000000...
## $ blonde
  $ mall
               : int 0 1 0 0 0 0 2 0 0 0 ...
   $ shopping
               : int 0000210001...
```

```
## $ clothes : int 0 0 0 0 0 0 0 0 0 ...
## $ hollister : int 0 0 0 0 0 0 2 0 0 0 ...
## $ abercrombie : int 0000000000...
## $ die
                : int 0000000000...
## $ death
                : int 0010000000...
## $ drunk
                : int 0000110000...
## $ drugs
                : int 0000100000...
# look at missing data for female variable
table(teens$gender)
##
##
      F
## 22054 5222
table(teens$gender, useNA = "ifany")
##
##
      F
            M <NA>
## 22054 5222 2724
# look at missing data for age variable
summary(teens$age)
##
     Min. 1st Qu. Median
                            Mean 3rd Qu.
                                           Max.
                                                  NA's
    3.086 16.312 17.287 17.994 18.259 106.927
                                                  5086
# eliminate age outliers
teens$age <- ifelse(teens$age >= 13 & teens$age < 20,
                  teens$age, NA)
summary(teens$age)
##
     Min. 1st Qu. Median
                           Mean 3rd Qu.
                                           Max.
                                                  NA's
                  17.27 17.25
                                                  5523
##
    13.03
          16.30
                                  18.22
                                          20.00
# reassign missing gender values to "unknown"
teens$female <- ifelse(teens$gender == "F" &</pre>
                       !is.na(teens$gender), 1, 0)
teens$no_gender <- ifelse(is.na(teens$gender), 1, 0)</pre>
# check our recoding work
table(teens$gender, useNA = "ifany")
##
##
      F
            M <NA>
## 22054 5222 2724
```

```
table(teens$female, useNA = "ifany")
##
##
       0
## 7946 22054
table(teens$no_gender, useNA = "ifany")
##
##
       0
## 27276 2724
# finding the mean age by cohort
mean(teens$age) # doesn't work
## [1] NA
mean(teens$age, na.rm = TRUE) # works
## [1] 17.25243
# age by cohort
aggregate(data = teens, age ~ gradyear, mean, na.rm = TRUE)
     gradyear
##
## 1
         2006 18.65586
## 2
         2007 17.70617
## 3
         2008 16.76770
         2009 15.81957
## 4
# create a vector with the average age for each gradyear, repeated by person
ave_age <- ave(teens$age, teens$gradyear,</pre>
               FUN = function(x) mean(x, na.rm = TRUE))
teens$age <- ifelse(is.na(teens$age), ave_age, teens$age)</pre>
# check the summary results to ensure missing values are eliminated
summary(teens$age)
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                               Max.
                    17.24
                            17.24 18.21
##
     13.03 16.28
                                              20.00
## Step 3: Training a model on the data ----
interests <- teens[5:40]</pre>
interests_z <- as.data.frame(lapply(interests, scale))</pre>
set.seed(2345)
teen_clusters <- kmeans(interests_z, 5)</pre>
```

```
## Step 4: Evaluating model performance ----
# look at the size of the clusters
teen_clusters$size
```

[1] 1038 601 4066 2696 21599

look at the cluster centers
teen_clusters\$centers

```
##
      basketball
                    football
                                  soccer
                                           softball volleyball
                                                                  swimming
## 1
     0.362160730 0.37985213 0.13734997 0.1272107 0.09247518
                                                                0.26180286
## 2 -0.094426312 0.06691768 -0.09956009 -0.0379725 -0.07286202
                                                                0.04578401
## 3 0.003980104 0.09524062 0.05342109 -0.0496864 -0.01459648
                                                                 0.32944934
     1.372334818 1.19570343 0.55621097 1.1304527 1.07177211
                                                                 0.08513210
## 5 -0.186822093 -0.18729427 -0.08331351 -0.1368072 -0.13344819 -0.08650052
     cheerleading
                    baseball
                                  tennis
                                              sports
                                                             cute
## 1
       0.2159945 0.25312305 0.11991682 0.77040675 0.475265034 2.043945661
      -0.1070370 -0.11182941 0.04027335 -0.10638613 -0.027044898 -0.042725567
## 2
## 3
       0.5142451 -0.04933628 0.06703386 -0.05435093 0.796948359 -0.003156716
## 4
      0.0400367 1.09279737 0.13887184 1.08316097 -0.005291962 -0.033193640
## 5
      -0.1092056 \ -0.13616893 \ -0.03683671 \ -0.15903307 \ -0.171452198 \ -0.092301138
##
                          hot
                                   kissed
                                                 dance
                                                              band marching
            sexv
## 1 0.547956598 0.314845390 3.02610259 0.455501275 0.39009330 -0.0105463
## 2 -0.027913348 -0.035027022 -0.04581067 0.050772118 4.09723438 5.2196105
## 3 0.266741598 0.623263396 -0.01284964 0.650572336 -0.03301257 -0.1131486
## 4 0.003036966 0.009046774 -0.08755418 -0.001993853 -0.07317758 -0.1039509
## 5 -0.076149916 -0.132614350 -0.13080557 -0.145524147 -0.11740538 -0.1104553
##
          music
                      rock
                                   god
                                           church
                                                        jesus
## 1 1.21014015
                1.2014998 0.41743650 0.1621804 0.12698409
                                                               0.07464400
## 2 0.51624366 0.1865286 0.09706027 0.0675347 0.05333966
## 3 0.24527495 0.1166274 0.32867738 0.5195729
                                                  0.26142784
                                                               0.23946855
## 4 0.07102323
                0.1565155 0.04902918 0.1320602 0.01776986 0.01719220
## 5 -0.12755935 -0.1044230 -0.09075500 -0.1239664 -0.05901846 -0.05243708
                     dress
                                blonde
                                               mall
           hair
                                                       shopping
                                                                     clothes
## 1 2.59053048 0.5312082 0.36322464 0.622896285
                                                    0.27607550 1.245121599
## 2 -0.05146837 0.0492724 -0.01238629 -0.087713363 -0.03710273 -0.004395251
## 3 0.35590025 0.5837827 0.03301526 0.808620531 1.07073115 0.616207360
## 4 0.01714820 -0.0653358 0.03690938 -0.004723697 0.03497875 0.016201064
## 5 -0.19220150 -0.1286412 -0.02793327 -0.179127117 -0.21816580 -0.177738408
      hollister abercrombie
                                     die
                                               death
                                                           drunk
                                                                      drugs
## 1 0.31525537
                  0.4131560 1.712160983 0.94713629 1.83371069
## 2 -0.16788599 -0.1413652 0.008941101 0.05464759 -0.08699556 -0.06414588
                 0.7935060 0.062399295 0.12642222 0.03594162 -0.05888141
## 3 0.85951603
## 4 -0.08381546 -0.0861708 -0.067312427 -0.01611162 -0.06891763 -0.08795059
## 5 -0.16182051 -0.1545430 -0.085876102 -0.06882571 -0.08386703 -0.10777278
## Step 5: Improving model performance ----
# apply the cluster IDs to the original data frame
teens$cluster <- teen_clusters$cluster</pre>
# look at the first five records
teens[1:5, c("cluster", "gender", "age", "friends")]
```

```
##
     cluster gender
                     age friends
## 1
          5
                  M 18.982
                                 7
## 2
           3
                  F 18.801
                                 0
## 3
           5
                  M 18.335
                                69
## 4
           5
                  F 18.875
                                 0
## 5
           1
               <NA> 18.995
                                10
# mean age by cluster
aggregate(data = teens, age ~ cluster, mean)
##
     cluster
                  age
## 1
           1 17.09319
## 2
           2 17.38488
## 3
           3 17.03773
## 4
           4 17.03759
## 5
           5 17.30265
# proportion of females by cluster
aggregate(data = teens, female ~ cluster, mean)
##
     cluster
                female
## 1
          1 0.8025048
## 2
           2 0.7237937
## 3
           3 0.8866208
## 4
           4 0.6984421
## 5
           5 0.7082735
# mean number of friends by cluster
aggregate(data = teens, friends ~ cluster, mean)
     cluster friends
##
## 1
           1 30.66570
## 2
           2 32.79368
## 3
           3 38.54575
## 4
           4 35.91728
## 5
           5 27.79221
```

R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see http://rmarkdown.rstudio.com.