

HW4

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```
#install.packages('tidyverse') # data manipulation
#install.packages('cluster')   # clustering algorithms
#install.packages('factoextra') # clustering algorithms & visualization
#install.packages('gridExtra')
```

```
library(tidyverse, warn.conflicts = FALSE) # data manipulation
```

```
## Warning: replacing previous import 'lifecycle::last_warnings' by
## 'rlang::last_warnings' when loading 'pillar'
```

```
## Warning: replacing previous import 'lifecycle::last_warnings' by
## 'rlang::last_warnings' when loading 'tibble'
```

```
## Warning: replacing previous import 'lifecycle::last_warnings' by
## 'rlang::last_warnings' when loading 'hms'
```

```
## -- Attaching packages ----- tidyverse 1.3.1 --
```

```
## v ggplot2 3.3.5    v purrr   0.3.4
## v tibble  3.1.4    v dplyr  1.0.7
## v tidyr   1.1.4    v stringr 1.4.0
## v readr   2.0.1    v forcats 0.5.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
library(cluster) # clustering algorithms
```

```
## Warning: package 'cluster' was built under R version 4.1.2
```

```
library(factoextra) # clustering algorithms & visualization
```

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

```
library(gridExtra) # subfigure layout package
```

```
##
## Attaching package: 'gridExtra'

## The following object is masked from 'package:dplyr':
##
##      combine

read the data file

Essay = read.csv('/Users/shivangi/Desktop/Applied Machine learning/Week 4/HW4-data-fedPapers.csv')
str(Essay)

## 'data.frame':    85 obs. of  72 variables:
## $ author       : chr  "dispt" "dispt" "dispt" "dispt" ...
## $ filename     : chr  "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_fed_52.txt" ...
## $ a            : num  0.28 0.177 0.339 0.27 0.303 0.245 0.349 0.414 0.248 0.442 ...
## $ all          : num  0.052 0.063 0.09 0.024 0.054 0.059 0.036 0.083 0.04 0.062 ...
## $ also         : num  0.009 0.013 0.008 0.016 0.027 0.007 0.007 0.009 0.007 0.006 ...
## $ an           : num  0.096 0.038 0.03 0.024 0.034 0.067 0.029 0.018 0.04 0.075 ...
## $ and          : num  0.358 0.393 0.301 0.262 0.404 0.282 0.335 0.478 0.356 0.423 ...
## $ any          : num  0.026 0.063 0.008 0.056 0.04 0.052 0.058 0.046 0.034 0.037 ...
## $ are          : num  0.131 0.051 0.068 0.064 0.128 0.111 0.087 0.11 0.154 0.093 ...
## $ as           : num  0.122 0.139 0.203 0.111 0.148 0.252 0.073 0.074 0.161 0.1 ...
## $ at           : num  0.017 0.114 0.023 0.056 0.013 0.015 0.116 0.037 0.047 0.031 ...
## $ be           : num  0.411 0.393 0.474 0.365 0.344 0.297 0.378 0.331 0.289 0.379 ...
## $ been         : num  0.026 0.165 0.015 0.127 0.047 0.03 0.044 0.046 0.027 0.025 ...
## $ but          : num  0.009 0 0.038 0.032 0.061 0.037 0.007 0.055 0.027 0.037 ...
## $ by           : num  0.14 0.139 0.173 0.167 0.209 0.186 0.102 0.092 0.168 0.174 ...
## $ can          : num  0.035 0 0.023 0.056 0.088 0 0.058 0.037 0.047 0.056 ...
## $ do           : num  0.026 0.013 0 0 0 0 0.015 0.028 0 0 ...
## $ down         : num  0 0 0.008 0 0 0.007 0 0 0 0 ...
## $ even         : num  0.009 0.025 0.015 0.024 0.02 0.007 0.007 0.018 0 0.006 ...
## $ every        : num  0.044 0 0.023 0.04 0.027 0.007 0.087 0.064 0.081 0.05 ...
## $ for          : num  0.096 0.076 0.098 0.103 0.141 0.067 0.116 0.055 0.127 0.1 ...
## $ from         : num  0.044 0.101 0.053 0.079 0.074 0.096 0.08 0.083 0.074 0.124 ...
## $ had          : num  0.035 0.101 0.008 0.016 0 0.022 0.015 0.009 0.007 0 ...
## $ has          : num  0.017 0.013 0.015 0.024 0.054 0.015 0.036 0.037 0.02 0.019 ...
## $ have         : num  0.044 0.152 0.023 0.143 0.047 0.119 0.044 0.074 0.074 0.044 ...
## $ her          : num  0 0 0 0 0 0.007 0 0.034 0.025 ...
## $ his          : num  0.017 0 0 0.024 0.02 0.067 0 0.018 0.02 0.05 ...
## $ if.          : num  0 0.025 0.023 0.04 0.034 0.03 0.029 0 0 0.025 ...
## $ in.          : num  0.262 0.291 0.308 0.238 0.263 0.401 0.189 0.267 0.248 0.274 ...
## $ into         : num  0.009 0.025 0.038 0.008 0.013 0.037 0 0.037 0.013 0.037 ...
## $ is           : num  0.157 0.038 0.15 0.151 0.189 0.26 0.167 0.083 0.208 0.23 ...
## $ it           : num  0.175 0.127 0.173 0.222 0.108 0.156 0.102 0.165 0.134 0.131 ...
## $ its          : num  0.07 0.038 0.03 0.048 0.013 0.015 0 0.046 0.02 0.019 ...
## $ may          : num  0.035 0.038 0.12 0.056 0.047 0.074 0.08 0.092 0.027 0.106 ...
## $ more         : num  0.026 0 0.038 0.056 0.067 0.045 0.08 0.064 0.06 0.081 ...
## $ must         : num  0.026 0.013 0.083 0.071 0.013 0.015 0.044 0.018 0.027 0.068 ...
## $ my           : num  0 0 0 0 0 0.007 0 0 0 0 ...
## $ no           : num  0.035 0 0.03 0.032 0.047 0.059 0.022 0.018 0.02 0.044 ...
## $ not          : num  0.114 0.127 0.068 0.087 0.128 0.134 0.102 0.101 0.094 0.106 ...
## $ now          : num  0 0 0 0 0 0.007 0 0.007 0.012 ...
```

```
## $ of      : num  0.9 0.747 0.858 0.802 0.869 ...
## $ on      : num  0.14 0.139 0.15 0.143 0.054 0.141 0.051 0.083 0.127 0.118 ...
## $ one     : num  0.026 0.025 0.03 0.032 0.047 0.052 0.073 0.046 0.06 0.031 ...
## $ only    : num  0.035 0 0.023 0.048 0.027 0.022 0.007 0.046 0.02 0.012 ...
## $ or      : num  0.096 0.114 0.06 0.064 0.081 0.074 0.153 0.037 0.154 0.081 ...
## $ our     : num  0.017 0 0 0.016 0.027 0.03 0.051 0 0.007 0.025 ...
## $ shall   : num  0.017 0 0.008 0.016 0 0.015 0.007 0 0.02 0 ...
## $ should  : num  0.017 0.013 0.068 0.032 0 0.03 0.007 0 0 0.012 ...
## $ so      : num  0.035 0.013 0.038 0.04 0.027 0.007 0.051 0.018 0.04 0.05 ...
## $ some    : num  0.009 0.063 0.03 0.024 0.067 0.045 0.007 0.028 0.027 0.025 ...
## $ such    : num  0.026 0 0.045 0.008 0.027 0.015 0.015 0 0.013 0.031 ...
## $ than    : num  0.009 0 0.023 0 0.047 0.03 0.109 0.055 0.067 0.044 ...
## $ that    : num  0.184 0.152 0.188 0.238 0.162 0.208 0.233 0.165 0.208 0.218 ...
## $ the     : num  1.43 1.25 1.49 1.33 1.19 ...
## $ their   : num  0.114 0.165 0.053 0.071 0.027 0.089 0.109 0.083 0.154 0.081 ...
## $ then    : num  0 0 0.015 0.008 0.007 0.007 0.015 0.009 0.007 0.012 ...
## $ there   : num  0.009 0 0.015 0 0.007 0.007 0.036 0.028 0.02 0 ...
## $ things  : num  0.009 0 0 0 0 0 0 0 0 0.012 ...
## $ this    : num  0.044 0.051 0.075 0.103 0.094 0.126 0.08 0.11 0.067 0.093 ...
## $ to      : num  0.507 0.355 0.361 0.532 0.485 0.445 0.56 0.34 0.49 0.498 ...
## $ up      : num  0 0 0 0 0 0 0.007 0 0 0 ...
## $ upon    : num  0 0.013 0 0 0 0 0 0 0 0 ...
## $ was     : num  0.009 0.051 0.008 0.087 0.027 0.007 0.015 0.018 0.027 0 ...
## $ were    : num  0.017 0 0.015 0.079 0.02 0.03 0.029 0.009 0.007 0 ...
## $ what    : num  0 0 0.008 0.008 0.02 0.015 0.015 0.009 0.02 0.025 ...
## $ when    : num  0.009 0 0 0.024 0.007 0.037 0.007 0 0.02 0.012 ...
## $ which   : num  0.175 0.114 0.105 0.167 0.155 0.186 0.211 0.175 0.201 0.199 ...
## $ who     : num  0.044 0.038 0.008 0 0.027 0.045 0.022 0.018 0.04 0.031 ...
## $ will    : num  0.009 0.089 0.173 0.079 0.168 0.111 0.145 0.267 0.154 0.106 ...
## $ with    : num  0.087 0.063 0.045 0.079 0.074 0.089 0.073 0.129 0.027 0.081 ...
## $ would   : num  0.192 0.139 0.068 0.064 0.04 0.037 0.073 0.037 0.04 0.031 ...
## $ your    : num  0 0 0 0 0 0 0 0 0 0 ...
```

```
typeof(Essay)
```

```
## [1] "list"
```

To remove any missing value that might be present in the data

```
Essay = na.omit(Essay)
```

Remove the label information

```
Essay.unlabeled = Essay[, which(!names(Essay) %in% c('author', 'filename'))]
head(Essay.unlabeled)
```

```
##      a  all also  an  and  any  are  as  at  be  been  but  by
## 1 0.280 0.052 0.009 0.096 0.358 0.026 0.131 0.122 0.017 0.411 0.026 0.009 0.140
## 2 0.177 0.063 0.013 0.038 0.393 0.063 0.051 0.139 0.114 0.393 0.165 0.000 0.139
## 3 0.339 0.090 0.008 0.030 0.301 0.008 0.068 0.203 0.023 0.474 0.015 0.038 0.173
## 4 0.270 0.024 0.016 0.024 0.262 0.056 0.064 0.111 0.056 0.365 0.127 0.032 0.167
## 5 0.303 0.054 0.027 0.034 0.404 0.040 0.128 0.148 0.013 0.344 0.047 0.061 0.209
```

```
## 6 0.245 0.059 0.007 0.067 0.282 0.052 0.111 0.252 0.015 0.297 0.030 0.037 0.186
##      can      do down even every for. from had has have her his if.
## 1 0.035 0.026 0.000 0.009 0.044 0.096 0.044 0.035 0.017 0.044 0 0.017 0.000
## 2 0.000 0.013 0.000 0.025 0.000 0.076 0.101 0.101 0.013 0.152 0 0.000 0.025
## 3 0.023 0.000 0.008 0.015 0.023 0.098 0.053 0.008 0.015 0.023 0 0.000 0.023
## 4 0.056 0.000 0.000 0.024 0.040 0.103 0.079 0.016 0.024 0.143 0 0.024 0.040
## 5 0.088 0.000 0.000 0.020 0.027 0.141 0.074 0.000 0.054 0.047 0 0.020 0.034
## 6 0.000 0.000 0.007 0.007 0.007 0.067 0.096 0.022 0.015 0.119 0 0.067 0.030
##      in. into is it its may more must my no not now of
## 1 0.262 0.009 0.157 0.175 0.070 0.035 0.026 0.026 0 0.035 0.114 0 0.900
## 2 0.291 0.025 0.038 0.127 0.038 0.038 0.000 0.013 0 0.000 0.127 0 0.747
## 3 0.308 0.038 0.150 0.173 0.030 0.120 0.038 0.083 0 0.030 0.068 0 0.858
## 4 0.238 0.008 0.151 0.222 0.048 0.056 0.056 0.071 0 0.032 0.087 0 0.802
## 5 0.263 0.013 0.189 0.108 0.013 0.047 0.067 0.013 0 0.047 0.128 0 0.869
## 6 0.401 0.037 0.260 0.156 0.015 0.074 0.045 0.015 0 0.059 0.134 0 0.876
##      on one only or our shall should so some such than that
## 1 0.140 0.026 0.035 0.096 0.017 0.017 0.017 0.035 0.009 0.026 0.009 0.184
## 2 0.139 0.025 0.000 0.114 0.000 0.000 0.013 0.013 0.063 0.000 0.000 0.152
## 3 0.150 0.030 0.023 0.060 0.000 0.008 0.068 0.038 0.030 0.045 0.023 0.188
## 4 0.143 0.032 0.048 0.064 0.016 0.016 0.032 0.040 0.024 0.008 0.000 0.238
## 5 0.054 0.047 0.027 0.081 0.027 0.000 0.000 0.027 0.067 0.027 0.047 0.162
## 6 0.141 0.052 0.022 0.074 0.030 0.015 0.030 0.007 0.045 0.015 0.030 0.208
##      the their then there things this to up upon was were what when
## 1 1.425 0.114 0.000 0.009 0.009 0.044 0.507 0 0.000 0.009 0.017 0.000 0.009
## 2 1.254 0.165 0.000 0.000 0.000 0.051 0.355 0 0.013 0.051 0.000 0.000 0.000
## 3 1.490 0.053 0.015 0.015 0.000 0.075 0.361 0 0.000 0.008 0.015 0.008 0.000
## 4 1.326 0.071 0.008 0.000 0.000 0.103 0.532 0 0.000 0.087 0.079 0.008 0.024
## 5 1.193 0.027 0.007 0.007 0.000 0.094 0.485 0 0.000 0.027 0.020 0.020 0.007
## 6 1.469 0.089 0.007 0.007 0.000 0.126 0.445 0 0.000 0.007 0.030 0.015 0.037
##      which who will with would your
## 1 0.175 0.044 0.009 0.087 0.192 0
## 2 0.114 0.038 0.089 0.063 0.139 0
## 3 0.105 0.008 0.173 0.045 0.068 0
## 4 0.167 0.000 0.079 0.079 0.064 0
## 5 0.155 0.027 0.168 0.074 0.040 0
## 6 0.186 0.045 0.111 0.089 0.037 0
```

Since we don't want the clustering algorithm to depend on an arbitrary variable unit, we start by scaling/standardizing the data using the R function `scale`:

```
Essay.unlabeled = scale(Essay.unlabeled)
head(Essay.unlabeled)
```

```
##      a      all      also      an      and      any
## 1 -0.1723368 -0.03534435 0.15950160 0.93187600 -0.23253417 -0.68518464
## 2 -1.5135000 0.43010589 0.63520811 -1.02557977 0.07316143 0.93870812
## 3 0.5959023 1.57257465 0.04057497 -1.29557367 -0.73038129 -1.47518652
## 4 -0.3025468 -1.22012677 0.99198800 -1.49806910 -1.07101353 0.63148517
## 5 0.1271463 0.04928297 2.30018092 -1.16057672 0.16923719 -0.07073873
## 6 -0.6280719 0.26085126 -0.07835166 -0.04685188 -0.89633033 0.45592919
##      are      as      at      be      been      but      by
## 1 1.5654231 -0.05418106 -1.0404489 1.26866739 -1.0162916 -1.3159303 0.2605755
## 2 -0.7567578 0.36449075 2.6603712 1.06077823 3.1791960 -1.8238448 0.2402926
```

```

## 3 -0.2632943 1.94066694 -0.8115322 1.99627946 -1.3483086 0.3206833 0.9299109
## 4 -0.3794034 -0.32508634 0.4475097 0.73739509 2.0322282 -0.0179264 0.8082135
## 5 1.4783413 0.58614052 -1.1930600 0.49485773 -0.3824410 1.6186872 1.6600949
## 6 0.9848779 3.14742684 -1.1167544 -0.04796397 -0.8955582 0.2642484 1.1935884
## can do down even every for.
## 1 -0.02062228 3.0053860 -0.4239542 -0.2431406 1.09472692 0.06360606
## 2 -1.27268916 1.0262730 -0.4239542 1.3777967 -1.30239174 -0.50550078
## 3 -0.44990235 -0.9528399 1.7936523 0.3647109 -0.04935244 0.12051674
## 4 0.73061785 -0.9528399 -0.4239542 1.2764881 0.87680704 0.26279345
## 5 1.87536471 -0.9528399 -0.4239542 0.8712538 0.16856744 1.34409646
## 6 -1.27268916 -0.9528399 1.5164515 -0.4457577 -0.92103196 -0.76159886
## from had has have her his
## 1 -1.31170510 0.61463684 -1.1001862 -1.3396177 -0.3769551 -0.2733913
## 2 0.77813746 3.54670543 -1.2606596 1.5116901 -0.3769551 -0.6732398
## 3 -0.98172996 -0.58484577 -1.1804229 -1.8940386 -0.3769551 -0.6732398
## 4 -0.02846844 -0.22944351 -0.8193579 1.2740811 -0.3769551 -0.1087477
## 5 -0.21178797 -0.94024802 0.3841920 -1.2604147 -0.3769551 -0.2028298
## 6 0.59481793 0.03710818 -1.1804229 0.6404571 -0.3769551 0.9026339
## if. in. into is it its
## 1 -1.5128966 -0.7758578 -0.7853726 0.01246845 0.4072769 0.72662889
## 2 -0.1289512 -0.3693776 0.0471346 -2.02169914 -0.6633995 -0.34810226
## 3 -0.2396668 -0.1310961 0.7235467 -0.10718847 0.3626654 -0.61678505
## 4 0.7014161 -1.1122552 -0.8374043 -0.09009462 1.4556476 -0.01224877
## 5 0.3692692 -0.7618412 -0.5772458 0.55947150 -1.0872090 -1.18773597
## 6 0.1478379 1.1724439 0.6715150 1.77313452 -0.0165325 -1.12056527
## may more must my no not
## 1 -0.8830840 -0.70385836 -0.2855115 -0.4056847 0.13701191 0.7490012
## 2 -0.7842747 -1.63698791 -0.8122064 -0.4056847 -1.68267750 1.2015142
## 3 1.9165132 -0.27318318 2.0238430 -0.4056847 -0.12294372 -0.8521988
## 4 -0.1914188 0.37282959 1.5376631 -0.4056847 -0.01896147 -0.1908336
## 5 -0.4878468 0.76761517 -0.8122064 -0.4056847 0.76090542 1.2363229
## 6 0.4014371 -0.02195599 -0.7311764 -0.4056847 1.38479893 1.4451751
## now of on one only or
## 1 -0.8475975 -0.07882731 1.8402042 -0.6574777 0.826718310 -0.0164496
## 2 -0.8475975 -1.36832319 1.8141910 -0.7019372 -1.561133119 0.3830407
## 3 -0.8475975 -0.43280657 2.1003361 -0.4796396 0.008026391 -0.8154303
## 4 -0.8475975 -0.90477892 1.9182438 -0.3907206 1.713634554 -0.7266546
## 5 -0.8475975 -0.34009772 -0.3969308 0.2761720 0.280923697 -0.3493582
## 6 -0.8475975 -0.28110118 1.8662174 0.4984695 -0.060197935 -0.5047156
## our shall should so some such than
## 1 -0.1827435 -0.09108447 -0.4841143 0.3222109 -0.7284739 -0.1937633 -1.3055544
## 2 -0.7005166 -0.97442043 -0.6865729 -1.0471854 2.8824281 -1.7565984 -1.6416073
## 3 -0.7005166 -0.55873292 2.0972330 0.5089468 0.6757658 0.9483085 -0.7828055
## 4 -0.2132007 -0.14304541 0.2751055 0.6334373 0.2745544 -1.2757260 -1.6416073
## 5 0.1218290 -0.97442043 -1.3445634 -0.1757514 3.1499023 -0.1336542 0.1133355
## 6 0.2132007 -0.19500635 0.1738762 -1.4206571 1.6787941 -0.8549627 -0.5214311
## that the their then there things
## 1 -0.47881966 0.7661753 0.68392790 -1.0008309 -0.7880202 1.4470317
## 2 -1.02535321 -0.1422477 1.90906321 -1.0008309 -1.1961687 -0.6067331
## 3 -0.41050296 1.1114823 -0.78143002 1.4673691 -0.5159211 -0.6067331
## 4 0.44345572 0.2402462 -0.34902932 0.3155424 -1.1961687 -0.6067331
## 5 -0.85456148 -0.4663051 -1.40600880 0.1509958 -0.8787198 -0.6067331
## 6 -0.06891949 0.9999216 0.08337138 0.1509958 -0.8787198 -0.6067331
## this to up upon was were

```

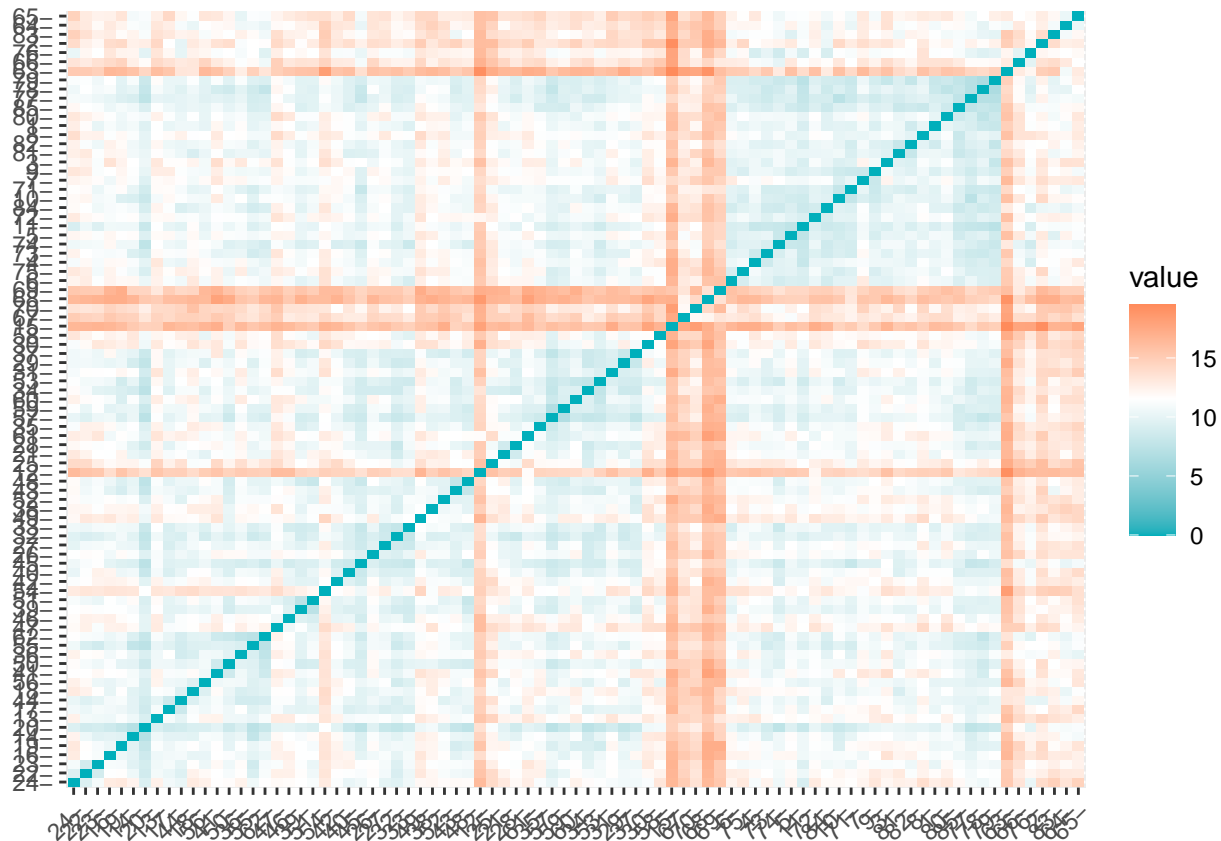
```
## 1 -1.4818540 -0.27620743 -0.5120406 -1.0657505 -0.57950415 -0.15447029
## 2 -1.2406879 -1.73456457 -0.5120406 -0.5916546 0.86621899 -0.96910375
## 3 -0.4138329 -1.67699784 -0.5120406 -1.0657505 -0.61392613 -0.25030952
## 4 0.5508314 -0.03634605 -0.5120406 -1.0657505 2.10541024 2.81654586
## 5 0.2407607 -0.48728543 -0.5120406 -1.0657505 0.04009148 -0.01071144
## 6 1.3432342 -0.87106363 -0.5120406 -1.0657505 -0.64834810 0.46848471
##      what      when      which      who      will      with      would
## 1 -1.1025238 -0.2068178 0.4006641 0.4609835 -1.3597325 0.2849510 1.2098029
## 2 -1.1025238 -0.8858546 -1.0164186 0.2198537 -0.1463229 -0.6496151 0.4995915
## 3 -0.4165986 -0.8858546 -1.2254963 -0.9857954 1.1277571 -1.3505396 -0.4518237
## 4 -0.4165986 0.9249103 0.2148172 -1.3073018 -0.2979991 -0.0265710 -0.5054246
## 5 0.6122890 -0.3577148 -0.0639532 -0.2222177 1.0519190 -0.2212723 -0.8270297
## 6 0.1835858 1.9057413 0.6562036 0.5011718 0.1873647 0.3628315 -0.8672304
##      your
## 1 -0.2087646
## 2 -0.2087646
## 3 -0.2087646
## 4 -0.2087646
## 5 -0.2087646
## 6 -0.2087646
```

```
typeof(Essay.unlabeled)
```

```
## [1] "double"
```

```
distance = factoextra::get_dist(Essay.unlabeled) # default method is euclidean

# visualization of the distance matrix
factoextra::fviz_dist(distance, gradient = list(low = "#00AFBB", mid = "white", high = "#FC4E07"))
```



k-Means Clustering

```
# the kmeans() function is available in pre-loaded {stats} package
model.r = kmeans(Essay.unlabeled, centers = 7, nstart = 25)
model.r
```

```
## K-means clustering with 7 clusters of sizes 15, 7, 4, 15, 1, 24, 19
##
## Cluster means:
##      a      all      also      an      and      any
## 1  0.57333261 -0.5543919  0.01678964  0.6708819 -0.65235612  0.4325217
## 2 -0.79362463 -0.4524361  0.14251208 -0.7845138  1.24728209 -0.9735980
## 3 -1.98551138 -0.6065787  1.52715783 -1.3208856  2.84844078  0.1377340
## 4  0.40319151  0.4554941 -0.36377557  0.1713932 -0.10326860 -0.5535177
## 5  1.27299447 -0.3315400  0.87306137 -1.2955737 -1.00114025  0.1487062
## 6 -0.04049915  0.1180427  0.39239958 -0.4391868  0.04586718 -0.2938411
## 7 -0.07639257  0.2408079 -0.64168833  0.5251089 -0.46789679  0.7885616
##      are      as      at      be      been      but
## 1 -0.70837900  0.0295533  0.4653143  0.3916831 -0.06853399 -0.4393074
## 2 -0.09327752 -0.2899038  0.5456169 -1.6483805  0.26434540 -0.5177788
## 3  0.47690082  0.9617138 -0.8019940  0.1195023 -1.31812527  1.0261202
## 4 -0.10074167 -0.3841929  0.4882060 -0.4983905  0.23329965 -0.2248546
## 5 -0.49551242 -0.8422692 -1.1167544  0.8644385 -0.89555817 -0.1307963
## 6  0.26040583  0.2721367 -0.3982103  0.1993856  0.02880736  0.1960561
```

## 7	0.26989062	-0.1151023	-0.2231761	0.3690301	0.06077789	0.2583078
##	by	can	do	down	even	every
## 1	-0.5439792	-0.3378126	0.01134333	0.53700864	0.3376953	-0.3144883
## 2	0.5851017	-1.0478282	-0.51787001	-0.42395419	-0.3444492	-0.6408493
## 3	0.4329801	0.2029611	0.60761453	-0.42395419	-0.1924863	-1.1253318
## 4	-0.2343270	-0.2161832	0.03164193	-0.10979326	0.4795273	-0.4561362
## 5	-1.0578124	0.8737112	1.33075194	-0.42395419	-1.1549178	1.1492069
## 6	0.6873613	0.0375094	-0.10283626	-0.08900320	-0.1080625	0.6952071
## 7	-0.5048367	0.6873133	0.08879849	0.04291034	-0.2804648	0.1427611
##	for.	from	had	has	have	her
## 1	0.46577490	0.08396753	0.1763074	-0.110600733	-0.014287599	-0.35522174
## 2	-0.06241046	0.38007221	2.0743247	-0.062840815	0.493365873	0.12202513
## 3	-0.05732915	0.84229929	-0.4182510	-0.859476253	-0.613590249	0.48461743
## 4	-0.53395613	-0.09201921	-0.0902443	0.651647564	0.323645173	0.08565501
## 5	-1.81444652	-0.17512406	-0.9402480	-0.578647933	-0.917201724	6.60876813
## 6	0.09561832	-0.25761785	-0.1702098	-0.145704273	-0.195574450	-0.00244272
## 7	0.06360606	0.02363289	-0.4796280	-0.008545329	-0.001503958	-0.27890987
##	his	if.	in.	into	is	it
## 1	1.0280766	0.48736590	0.5192723	-0.37952535	0.1560568	0.5143446
## 2	0.3213415	-0.49273111	-0.4514745	0.59718401	-1.0668858	-0.4594612
## 3	-0.4086342	1.76705416	-0.7022708	0.97069729	-1.2695699	0.6582167
## 4	-0.0507305	-0.11418907	-0.1301616	-0.20955512	-0.2427996	-0.7496485
## 5	-0.6732398	0.97820523	-1.0842221	4.20967065	-0.3635961	-1.4887126
## 6	-0.1979296	-0.36422189	-0.5101272	0.02545472	0.3173087	-0.1373554
## 7	-0.5184997	-0.07650691	0.7084221	-0.21302390	0.3471480	0.4683243
##	its	may	more	must	my	no
## 1	-0.2563023	-0.02454088	0.04264529	-0.3395315	-0.09861408	-0.2650528
## 2	-0.3624960	-1.23597444	-0.38085197	-0.7022372	0.03891560	-0.7988284
## 3	-0.5915960	0.01443391	1.86224792	-0.6197602	-0.40568466	-0.8248239
## 4	-0.0727024	-0.58446033	-0.42152685	-0.1315546	-0.05711806	-0.3863654
## 5	-0.6167850	1.91651322	1.59307593	0.4842733	-0.40568466	0.6569232
## 6	-0.1619835	0.27518071	0.24721600	0.2074209	-0.22413956	0.3536416
## 7	0.7549113	0.48464490	-0.34874023	0.4736115	0.49849187	0.5009498
##	not	now	of	on	one	only
## 1	0.09691833	-0.4075517	0.50271004	-0.7801919	-0.1654591	-0.15116370
## 2	-0.66820994	0.7774954	-0.64350851	0.3017094	0.1110367	-0.44030490
## 3	0.79251208	-0.6018273	-2.45765222	0.1688563	1.7433358	1.38956900
## 4	-0.46698253	0.3508250	0.87972605	-0.4489572	-0.3877567	-0.28306407
## 5	-0.60853789	-0.8475975	-0.18839232	-0.6310496	4.1886087	-1.56113312
## 6	0.25297731	-0.2741336	-0.20103444	1.0305434	0.1224162	0.22407009
## 7	0.08397193	0.2759235	-0.07306073	-0.4448498	-0.3462611	0.01161715
##	or	our	shall	should	so	some
## 1	0.3238570	-0.3979747	-0.06337197	0.51468154	0.46330021	-0.14003062
## 2	-0.4666689	-0.3263276	-0.53646395	-0.46242227	-0.14907485	0.80950286
## 3	1.8755810	1.5457052	0.03881788	0.83186670	0.92910244	-0.12665691
## 4	-0.0401231	0.6456936	0.10983116	-0.05557687	0.07737943	-0.08207787
## 5	-1.1483389	-0.7005166	1.36382183	-1.34456335	0.94466376	-1.33029091
## 6	-0.2032483	-0.1345195	-0.20150147	-0.33437925	-0.02013818	0.35814012
## 7	-0.1297553	-0.1939646	0.33554219	0.12592549	-0.59181199	-0.47859668
##	such	than	that	the	their	then
## 1	0.0867456	0.26767087	-0.03817698	0.3620776	-0.17126459	-0.2877954
## 2	-0.2366983	-0.45208682	-0.57153517	-0.4549213	1.39086872	0.3390491
## 3	1.8349169	0.87878924	0.12322121	-2.3814309	1.09831190	0.2744058
## 4	-0.3740904	-0.29988509	-0.41505741	-0.3200366	-0.41789314	-0.3316745


```

## 5 -0.8549627  3.95927391  0.98998928 -1.2153557 -1.69427594 -1.0008309
## 6 -0.2388450 -0.06402576 -0.22619021  0.3883298 -0.02973344  0.1235713
## 7  0.2744545 -0.12052588  0.77605015  0.2092105 -0.15179392  0.2029579
##      there      things      this      to      up      upon
## 1  0.2399095 -0.40896320  0.17874661  0.7228951  0.09571931  0.75040160
## 2 -0.8204129 -0.60673314 -0.56640730 -1.2219466  0.05510862 -0.72190069
## 3 -0.4025465 -0.20738998 -1.53353242 -0.3673547 -0.51204063 -1.06575048
## 4  0.1310699  1.00585259  0.36708582  0.2777124  0.56624313  0.46594405
## 5  3.5655645 -0.60673314 -1.96418610  1.0574218  1.69353979  0.02831704
## 6 -0.7105475 -0.15984913 -0.07792298 -0.7751191 -0.29148259 -1.01560572
## 7  0.8039979  0.02981385  0.30241221  0.6610193 -0.15605221  0.81143905
##      was      were      what      when      which      who
## 1  0.06762907  0.23208127  0.4007955  0.74383380 -0.5223756  0.7905275
## 2  2.01689658  1.28311817 -0.4043500  0.05186294 -0.6480435  0.6561838
## 3 -0.42460524  0.02522827  0.5051132  0.60425401 -1.3590738  0.5815484
## 4 -0.12283922 -0.07140962 -0.3365740 -0.33256531  0.2411455 -0.6107046
## 5 -0.88930196 -0.96910375 -1.1025238  2.50932955  0.1683554  0.5011718
## 6 -0.07177995 -0.06661766 -0.1022163 -0.23511095  0.2709584 -0.2942217
## 7 -0.47261484 -0.46973092  0.1790732 -0.30609214  0.3957734 -0.1608776
##      will      with      would      your
## 1 -0.07756304  0.01496527  0.5594458 -0.20876463
## 2 -1.01521083  0.22932210 -0.7293996 -0.20876463
## 3  0.82819663  0.58673800  0.5866929  0.61658391
## 4 -0.55281512  0.40696383  0.5147784 -0.13998559
## 5  0.55138758  2.73818703  1.4510068 -0.20876463
## 6  0.32387329 -0.10931904 -0.5165914 -0.06260916
## 7  0.25921133 -0.54714071 -0.1266922  0.31250813
##
## Clustering vector:
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26
##  6  2  6  6  6  6  6  6  6  6  6  6  7  4  4  5  4  1  4  4  4  7  4  4  1  7  7
## 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52
##  1  7  7  7  7  7  1  7  7  2  7  1  1  1  4  1  4  4  1  1  1  1  1  4  1  1
## 53 54 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78
##  7  7  4  4  7  7  7  7  7  4  2  2  2  2  3  3  3  3  6  6  6  6  6  2  6  6
## 79 80 81 82 83 84 85
##  6  6  6  6  6  6  6
##
## Within cluster sum of squares by cluster:
## [1] 756.1569 427.1200 278.8521 718.4455 0.0000 1108.9977 1105.4565
## (between_SS / total_SS = 25.3 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"

```

Print just the centroids

```
model.r$centers
```

```

##      a      all      also      an      and      any
## 1  0.57333261 -0.5543919  0.01678964  0.6708819 -0.65235612  0.4325217

```

## 2	-0.79362463	-0.4524361	0.14251208	-0.7845138	1.24728209	-0.9735980
## 3	-1.98551138	-0.6065787	1.52715783	-1.3208856	2.84844078	0.1377340
## 4	0.40319151	0.4554941	-0.36377557	0.1713932	-0.10326860	-0.5535177
## 5	1.27299447	-0.3315400	0.87306137	-1.2955737	-1.00114025	0.1487062
## 6	-0.04049915	0.1180427	0.39239958	-0.4391868	0.04586718	-0.2938411
## 7	-0.07639257	0.2408079	-0.64168833	0.5251089	-0.46789679	0.7885616
##	are	as	at	be	been	but
## 1	-0.70837900	0.0295533	0.4653143	0.3916831	-0.06853399	-0.4393074
## 2	-0.09327752	-0.2899038	0.5456169	-1.6483805	0.26434540	-0.5177788
## 3	0.47690082	0.9617138	-0.8019940	0.1195023	-1.31812527	1.0261202
## 4	-0.10074167	-0.3841929	0.4882060	-0.4983905	0.23329965	-0.2248546
## 5	-0.49551242	-0.8422692	-1.1167544	0.8644385	-0.89555817	-0.1307963
## 6	0.26040583	0.2721367	-0.3982103	0.1993856	0.02880736	0.1960561
## 7	0.26989062	-0.1151023	-0.2231761	0.3690301	0.06077789	0.2583078
##	by	can	do	down	even	every
## 1	-0.5439792	-0.3378126	0.01134333	0.53700864	0.3376953	-0.3144883
## 2	0.5851017	-1.0478282	-0.51787001	-0.42395419	-0.3444492	-0.6408493
## 3	0.4329801	0.2029611	0.60761453	-0.42395419	-0.1924863	-1.1253318
## 4	-0.2343270	-0.2161832	0.03164193	-0.10979326	0.4795273	-0.4561362
## 5	-1.0578124	0.8737112	1.33075194	-0.42395419	-1.1549178	1.1492069
## 6	0.6873613	0.0375094	-0.10283626	-0.08900320	-0.1080625	0.6952071
## 7	-0.5048367	0.6873133	0.08879849	0.04291034	-0.2804648	0.1427611
##	for.	from	had	has	have	her
## 1	0.46577490	0.08396753	0.1763074	-0.110600733	-0.014287599	-0.35522174
## 2	-0.06241046	0.38007221	2.0743247	-0.062840815	0.493365873	0.12202513
## 3	-0.05732915	0.84229929	-0.4182510	-0.859476253	-0.613590249	0.48461743
## 4	-0.53395613	-0.09201921	-0.0902443	0.651647564	0.323645173	0.08565501
## 5	-1.81444652	-0.17512406	-0.9402480	-0.578647933	-0.917201724	6.60876813
## 6	0.09561832	-0.25761785	-0.1702098	-0.145704273	-0.195574450	-0.00244272
## 7	0.06360606	0.02363289	-0.4796280	-0.008545329	-0.001503958	-0.27890987
##	his	if.	in.	into	is	it
## 1	1.0280766	0.48736590	0.5192723	-0.37952535	0.1560568	0.5143446
## 2	0.3213415	-0.49273111	-0.4514745	0.59718401	-1.0668858	-0.4594612
## 3	-0.4086342	1.76705416	-0.7022708	0.97069729	-1.2695699	0.6582167
## 4	-0.0507305	-0.11418907	-0.1301616	-0.20955512	-0.2427996	-0.7496485
## 5	-0.6732398	0.97820523	-1.0842221	4.20967065	-0.3635961	-1.4887126
## 6	-0.1979296	-0.36422189	-0.5101272	0.02545472	0.3173087	-0.1373554
## 7	-0.5184997	-0.07650691	0.7084221	-0.21302390	0.3471480	0.4683243
##	its	may	more	must	my	no
## 1	-0.2563023	-0.02454088	0.04264529	-0.3395315	-0.09861408	-0.2650528
## 2	-0.3624960	-1.23597444	-0.38085197	-0.7022372	0.03891560	-0.7988284
## 3	-0.5915960	0.01443391	1.86224792	-0.6197602	-0.40568466	-0.8248239
## 4	-0.0727024	-0.58446033	-0.42152685	-0.1315546	-0.05711806	-0.3863654
## 5	-0.6167850	1.91651322	1.59307593	0.4842733	-0.40568466	0.6569232
## 6	-0.1619835	0.27518071	0.24721600	0.2074209	-0.22413956	0.3536416
## 7	0.7549113	0.48464490	-0.34874023	0.4736115	0.49849187	0.5009498
##	not	now	of	on	one	only
## 1	0.09691833	-0.4075517	0.50271004	-0.7801919	-0.1654591	-0.15116370
## 2	-0.66820994	0.7774954	-0.64350851	0.3017094	0.1110367	-0.44030490
## 3	0.79251208	-0.6018273	-2.45765222	0.1688563	1.7433358	1.38956900
## 4	-0.46698253	0.3508250	0.87972605	-0.4489572	-0.3877567	-0.28306407
## 5	-0.60853789	-0.8475975	-0.18839232	-0.6310496	4.1886087	-1.56113312
## 6	0.25297731	-0.2741336	-0.20103444	1.0305434	0.1224162	0.22407009
## 7	0.08397193	0.2759235	-0.07306073	-0.4448498	-0.3462611	0.01161715

```
##           or           our           shall           should           so           some
## 1  0.3238570 -0.3979747 -0.06337197  0.51468154  0.46330021 -0.14003062
## 2 -0.4666689 -0.3263276 -0.53646395 -0.46242227 -0.14907485  0.80950286
## 3  1.8755810  1.5457052  0.03881788  0.83186670  0.92910244 -0.12665691
## 4 -0.0401231  0.6456936  0.10983116 -0.05557687  0.07737943 -0.08207787
## 5 -1.1483389 -0.7005166  1.36382183 -1.34456335  0.94466376 -1.33029091
## 6 -0.2032483 -0.1345195 -0.20150147 -0.33437925 -0.02013818  0.35814012
## 7 -0.1297553 -0.1939646  0.33554219  0.12592549 -0.59181199 -0.47859668
##           such           than           that           the           their           then
## 1  0.0867456  0.26767087 -0.03817698  0.3620776 -0.17126459 -0.2877954
## 2 -0.2366983 -0.45208682 -0.57153517 -0.4549213  1.39086872  0.3390491
## 3  1.8349169  0.87878924  0.12322121 -2.3814309  1.09831190  0.2744058
## 4 -0.3740904 -0.29988509 -0.41505741 -0.3200366 -0.41789314 -0.3316745
## 5 -0.8549627  3.95927391  0.98998928 -1.2153557 -1.69427594 -1.0008309
## 6 -0.2388450 -0.06402576 -0.22619021  0.3883298 -0.02973344  0.1235713
## 7  0.2744545 -0.12052588  0.77605015  0.2092105 -0.15179392  0.2029579
##           there           things           this           to           up           upon
## 1  0.2399095 -0.40896320  0.17874661  0.7228951  0.09571931  0.75040160
## 2 -0.8204129 -0.60673314 -0.56640730 -1.2219466  0.05510862 -0.72190069
## 3 -0.4025465 -0.20738998 -1.53353242 -0.3673547 -0.51204063 -1.06575048
## 4  0.1310699  1.00585259  0.36708582  0.2777124  0.56624313  0.46594405
## 5  3.5655645 -0.60673314 -1.96418610  1.0574218  1.69353979  0.02831704
## 6 -0.7105475 -0.15984913 -0.07792298 -0.7751191 -0.29148259 -1.01560572
## 7  0.8039979  0.02981385  0.30241221  0.6610193 -0.15605221  0.81143905
##           was           were           what           when           which           who
## 1  0.06762907  0.23208127  0.4007955  0.74383380 -0.5223756  0.7905275
## 2  2.01689658  1.28311817 -0.4043500  0.05186294 -0.6480435  0.6561838
## 3 -0.42460524  0.02522827  0.5051132  0.60425401 -1.3590738  0.5815484
## 4 -0.12283922 -0.07140962 -0.3365740 -0.33256531  0.2411455 -0.6107046
## 5 -0.88930196 -0.96910375 -1.1025238  2.50932955  0.1683554  0.5011718
## 6 -0.07177995 -0.06661766 -0.1022163 -0.23511095  0.2709584 -0.2942217
## 7 -0.47261484 -0.46973092  0.1790732 -0.30609214  0.3957734 -0.1608776
##           will           with           would           your
## 1 -0.07756304  0.01496527  0.5594458 -0.20876463
## 2 -1.01521083  0.22932210 -0.7293996 -0.20876463
## 3  0.82819663  0.58673800  0.5866929  0.61658391
## 4 -0.55281512  0.40696383  0.5147784 -0.13998559
## 5  0.55138758  2.73818703  1.4510068 -0.20876463
## 6  0.32387329 -0.10931904 -0.5165914 -0.06260916
## 7  0.25921133 -0.54714071 -0.1266922  0.31250813
```

Get cluster assignment

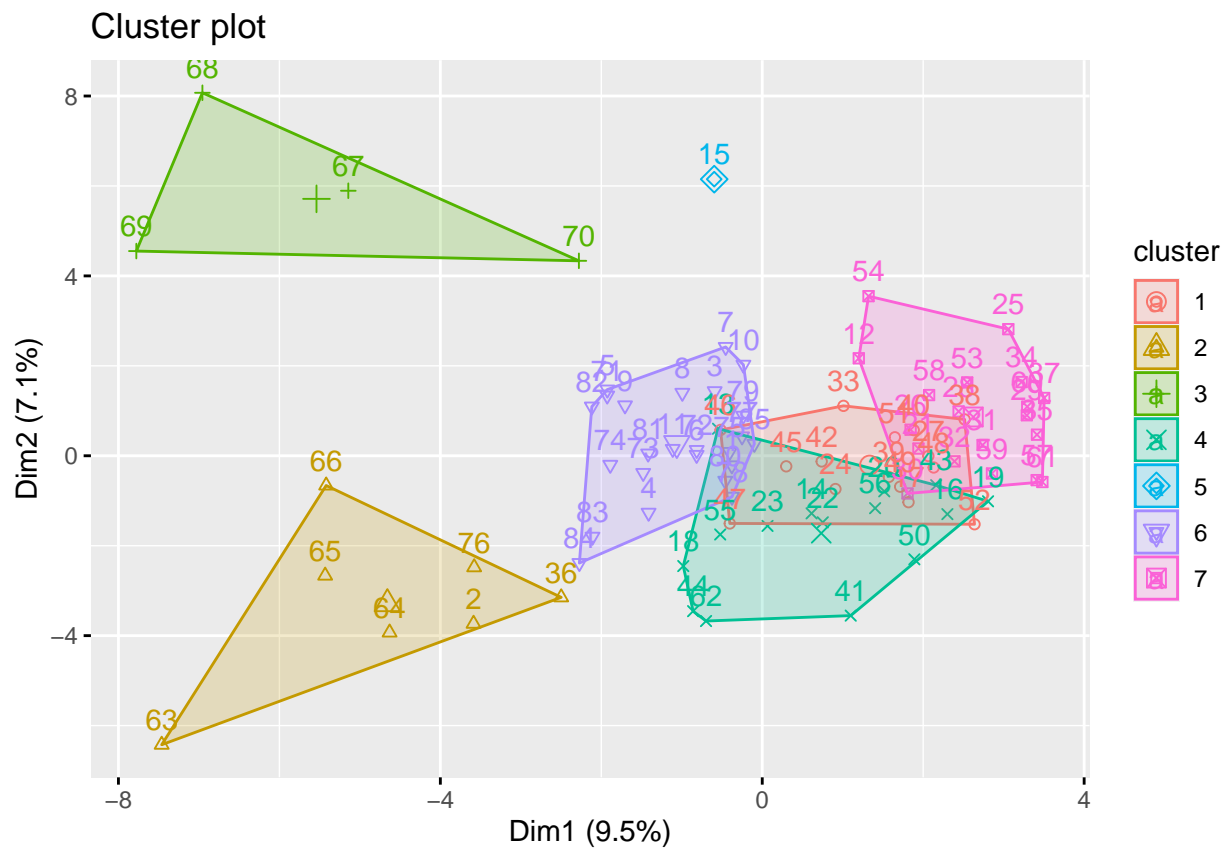
```
cluster.assignment = data.frame(Essay, model.r$cluster) |> relocate(model.r.cluster, .after = author)
glimpse(cluster.assignment)
```

```
## Rows: 85
## Columns: 73
## $ author      <chr> "dispt", "dispt", "dispt", "dispt", "dispt", "dispt", ~
## $ model.r.cluster <int> 6, 2, 6, 6, 6, 6, 6, 6, 6, 6, 6, 7, 4, 4, 5, 4, 1, 4, ~
## $ filename     <chr> "dispt_fed_49.txt", "dispt_fed_50.txt", "dispt_fed_51.~
## $ a            <dbl> 0.280, 0.177, 0.339, 0.270, 0.303, 0.245, 0.349, 0.414~
## $ all          <dbl> 0.052, 0.063, 0.090, 0.024, 0.054, 0.059, 0.036, 0.083~
```

## \$ also	<dbl> 0.009, 0.013, 0.008, 0.016, 0.027, 0.007, 0.007, 0.009~
## \$ an	<dbl> 0.096, 0.038, 0.030, 0.024, 0.034, 0.067, 0.029, 0.018~
## \$ and	<dbl> 0.358, 0.393, 0.301, 0.262, 0.404, 0.282, 0.335, 0.478~
## \$ any	<dbl> 0.026, 0.063, 0.008, 0.056, 0.040, 0.052, 0.058, 0.046~
## \$ are	<dbl> 0.131, 0.051, 0.068, 0.064, 0.128, 0.111, 0.087, 0.110~
## \$ as	<dbl> 0.122, 0.139, 0.203, 0.111, 0.148, 0.252, 0.073, 0.074~
## \$ at	<dbl> 0.017, 0.114, 0.023, 0.056, 0.013, 0.015, 0.116, 0.037~
## \$ be	<dbl> 0.411, 0.393, 0.474, 0.365, 0.344, 0.297, 0.378, 0.331~
## \$ been	<dbl> 0.026, 0.165, 0.015, 0.127, 0.047, 0.030, 0.044, 0.046~
## \$ but	<dbl> 0.009, 0.000, 0.038, 0.032, 0.061, 0.037, 0.007, 0.055~
## \$ by	<dbl> 0.140, 0.139, 0.173, 0.167, 0.209, 0.186, 0.102, 0.092~
## \$ can	<dbl> 0.035, 0.000, 0.023, 0.056, 0.088, 0.000, 0.058, 0.037~
## \$ do	<dbl> 0.026, 0.013, 0.000, 0.000, 0.000, 0.000, 0.015, 0.028~
## \$ down	<dbl> 0.000, 0.000, 0.008, 0.000, 0.000, 0.007, 0.000, 0.000~
## \$ even	<dbl> 0.009, 0.025, 0.015, 0.024, 0.020, 0.007, 0.007, 0.018~
## \$ every	<dbl> 0.044, 0.000, 0.023, 0.040, 0.027, 0.007, 0.087, 0.064~
## \$ for.	<dbl> 0.096, 0.076, 0.098, 0.103, 0.141, 0.067, 0.116, 0.055~
## \$ from	<dbl> 0.044, 0.101, 0.053, 0.079, 0.074, 0.096, 0.080, 0.083~
## \$ had	<dbl> 0.035, 0.101, 0.008, 0.016, 0.000, 0.022, 0.015, 0.009~
## \$ has	<dbl> 0.017, 0.013, 0.015, 0.024, 0.054, 0.015, 0.036, 0.037~
## \$ have	<dbl> 0.044, 0.152, 0.023, 0.143, 0.047, 0.119, 0.044, 0.074~
## \$ her	<dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.007, 0.000~
## \$ his	<dbl> 0.017, 0.000, 0.000, 0.024, 0.020, 0.067, 0.000, 0.018~
## \$ if.	<dbl> 0.000, 0.025, 0.023, 0.040, 0.034, 0.030, 0.029, 0.000~
## \$ in.	<dbl> 0.262, 0.291, 0.308, 0.238, 0.263, 0.401, 0.189, 0.267~
## \$ into	<dbl> 0.009, 0.025, 0.038, 0.008, 0.013, 0.037, 0.000, 0.037~
## \$ is	<dbl> 0.157, 0.038, 0.150, 0.151, 0.189, 0.260, 0.167, 0.083~
## \$ it	<dbl> 0.175, 0.127, 0.173, 0.222, 0.108, 0.156, 0.102, 0.165~
## \$ its	<dbl> 0.070, 0.038, 0.030, 0.048, 0.013, 0.015, 0.000, 0.046~
## \$ may	<dbl> 0.035, 0.038, 0.120, 0.056, 0.047, 0.074, 0.080, 0.092~
## \$ more	<dbl> 0.026, 0.000, 0.038, 0.056, 0.067, 0.045, 0.080, 0.064~
## \$ must	<dbl> 0.026, 0.013, 0.083, 0.071, 0.013, 0.015, 0.044, 0.018~
## \$ my	<dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.007, 0.000~
## \$ no	<dbl> 0.035, 0.000, 0.030, 0.032, 0.047, 0.059, 0.022, 0.018~
## \$ not	<dbl> 0.114, 0.127, 0.068, 0.087, 0.128, 0.134, 0.102, 0.101~
## \$ now	<dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.007, 0.000~
## \$ of	<dbl> 0.900, 0.747, 0.858, 0.802, 0.869, 0.876, 0.902, 1.029~
## \$ on	<dbl> 0.140, 0.139, 0.150, 0.143, 0.054, 0.141, 0.051, 0.083~
## \$ one	<dbl> 0.026, 0.025, 0.030, 0.032, 0.047, 0.052, 0.073, 0.046~
## \$ only	<dbl> 0.035, 0.000, 0.023, 0.048, 0.027, 0.022, 0.007, 0.046~
## \$ or	<dbl> 0.096, 0.114, 0.060, 0.064, 0.081, 0.074, 0.153, 0.037~
## \$ our	<dbl> 0.017, 0.000, 0.000, 0.016, 0.027, 0.030, 0.051, 0.000~
## \$ shall	<dbl> 0.017, 0.000, 0.008, 0.016, 0.000, 0.015, 0.007, 0.000~
## \$ should	<dbl> 0.017, 0.013, 0.068, 0.032, 0.000, 0.030, 0.007, 0.000~
## \$ so	<dbl> 0.035, 0.013, 0.038, 0.040, 0.027, 0.007, 0.051, 0.018~
## \$ some	<dbl> 0.009, 0.063, 0.030, 0.024, 0.067, 0.045, 0.007, 0.028~
## \$ such	<dbl> 0.026, 0.000, 0.045, 0.008, 0.027, 0.015, 0.015, 0.000~
## \$ than	<dbl> 0.009, 0.000, 0.023, 0.000, 0.047, 0.030, 0.109, 0.055~
## \$ that	<dbl> 0.184, 0.152, 0.188, 0.238, 0.162, 0.208, 0.233, 0.165~
## \$ the	<dbl> 1.425, 1.254, 1.490, 1.326, 1.193, 1.469, 1.259, 1.176~
## \$ their	<dbl> 0.114, 0.165, 0.053, 0.071, 0.027, 0.089, 0.109, 0.083~
## \$ then	<dbl> 0.000, 0.000, 0.015, 0.008, 0.007, 0.007, 0.015, 0.009~
## \$ there	<dbl> 0.009, 0.000, 0.015, 0.000, 0.007, 0.007, 0.036, 0.028~
## \$ things	<dbl> 0.009, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000~

```
## $ this      <dbl> 0.044, 0.051, 0.075, 0.103, 0.094, 0.126, 0.080, 0.110~
## $ to        <dbl> 0.507, 0.355, 0.361, 0.532, 0.485, 0.445, 0.560, 0.340~
## $ up        <dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.007, 0.000~
## $ upon      <dbl> 0.000, 0.013, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000~
## $ was       <dbl> 0.009, 0.051, 0.008, 0.087, 0.027, 0.007, 0.015, 0.018~
## $ were      <dbl> 0.017, 0.000, 0.015, 0.079, 0.020, 0.030, 0.029, 0.009~
## $ what      <dbl> 0.000, 0.000, 0.008, 0.008, 0.020, 0.015, 0.015, 0.009~
## $ when      <dbl> 0.009, 0.000, 0.000, 0.024, 0.007, 0.037, 0.007, 0.000~
## $ which     <dbl> 0.175, 0.114, 0.105, 0.167, 0.155, 0.186, 0.211, 0.175~
## $ who       <dbl> 0.044, 0.038, 0.008, 0.000, 0.027, 0.045, 0.022, 0.018~
## $ will      <dbl> 0.009, 0.089, 0.173, 0.079, 0.168, 0.111, 0.145, 0.267~
## $ with      <dbl> 0.087, 0.063, 0.045, 0.079, 0.074, 0.089, 0.073, 0.129~
## $ would     <dbl> 0.192, 0.139, 0.068, 0.064, 0.040, 0.037, 0.073, 0.037~
## $ your      <dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000~
```

```
fviz_cluster(model.r, data = Essay.unlabeled)
```

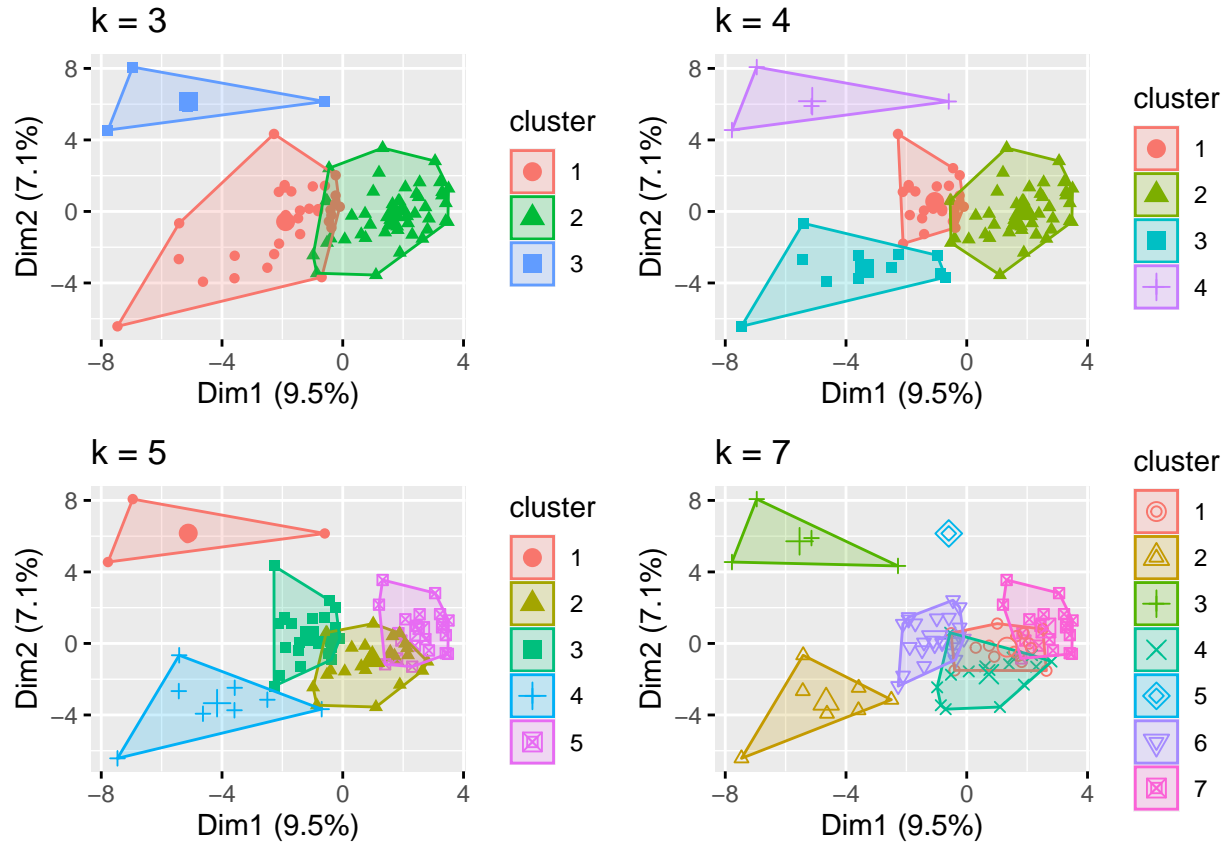


```
k3 = kmeans(Essay.unlabeled, centers = 3, nstart = 25)
k4 = kmeans(Essay.unlabeled, centers = 4, nstart = 25)
k5 = kmeans(Essay.unlabeled, centers = 5, nstart = 25)

# plots to compare
p1 = fviz_cluster(k3, geom = "point", data = Essay.unlabeled) + ggtitle("k = 3")
p2 = fviz_cluster(k4, geom = "point", data = Essay.unlabeled) + ggtitle("k = 4")
p3 = fviz_cluster(k5, geom = "point", data = Essay.unlabeled) + ggtitle("k = 5")
```

```
p4 = fviz_cluster(model.r, geom = "point", data = Essay.unlabeled) + ggtitle("k = 7")
```

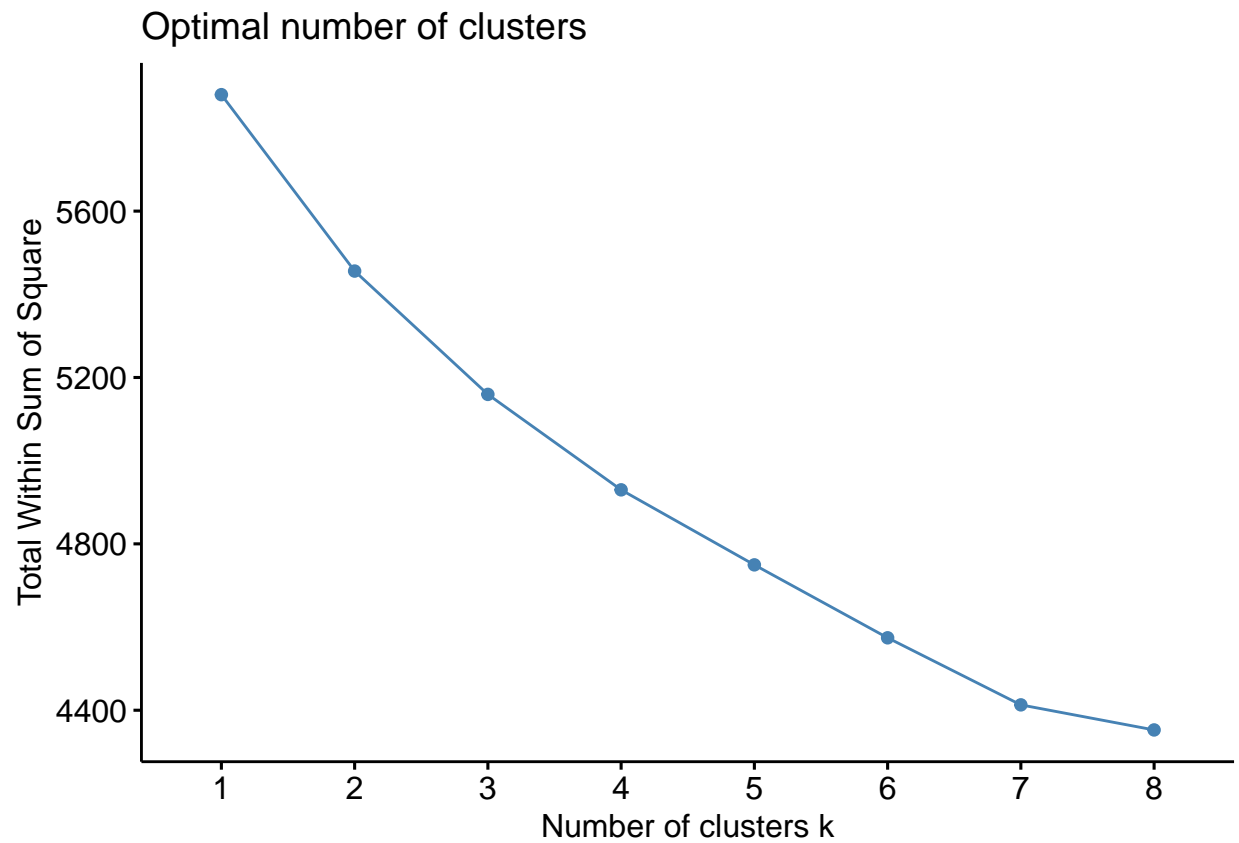
```
library(gridExtra)
grid.arrange(p1, p2, p3, p4, nrow = 2)
```



Elbow method

Fortunately, this process to compute the “Elbow method” has been wrapped up in a single function (fviz_nbclust):

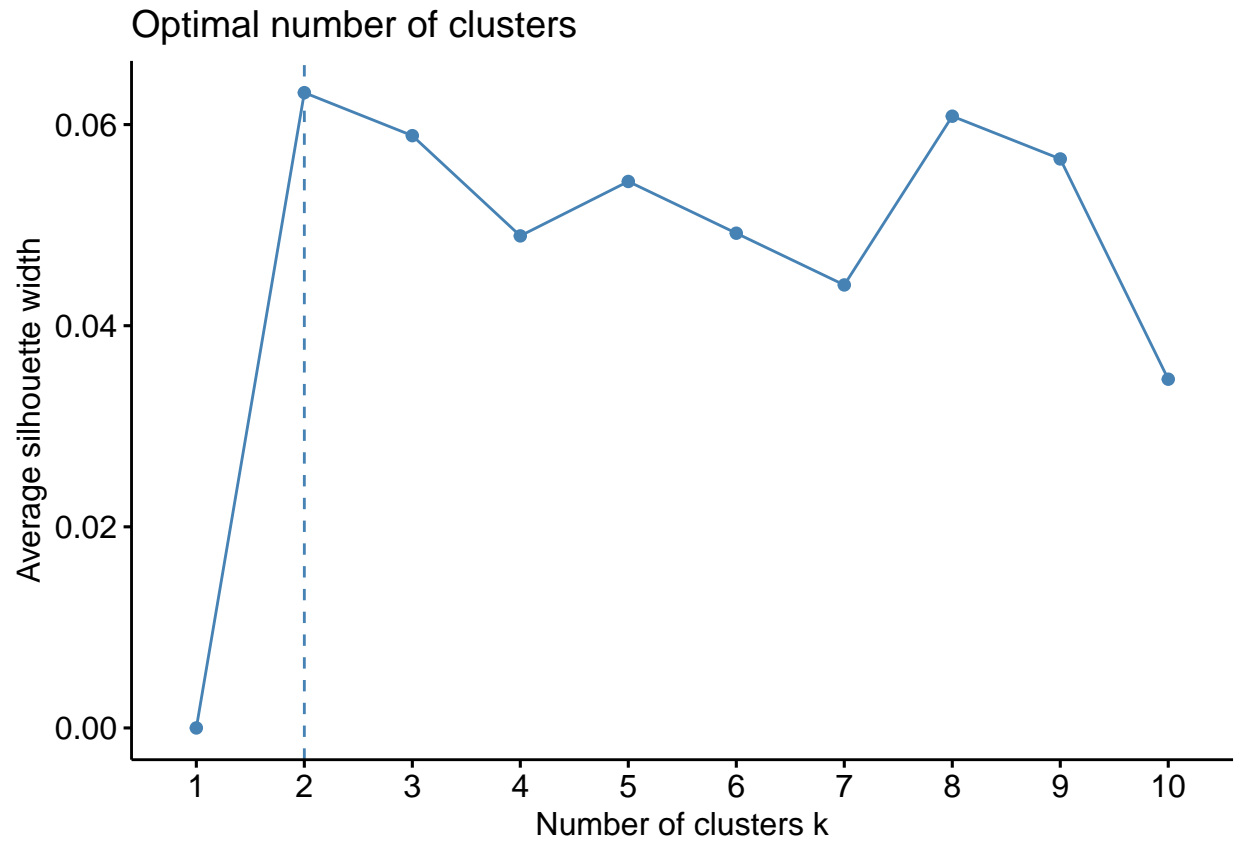
```
set.seed(123)
fviz_nbclust(Essay.unlabeled, kmeans, method = "wss", k.max = 8)
```



Average Silhouette Method

Similar to the elbow method, this process to compute the “average silhouette method” has been wrapped up in a single function

```
set.seed(123)
fviz_nbclust(Essay.unlabeled, kmeans, method = "silhouette")
```

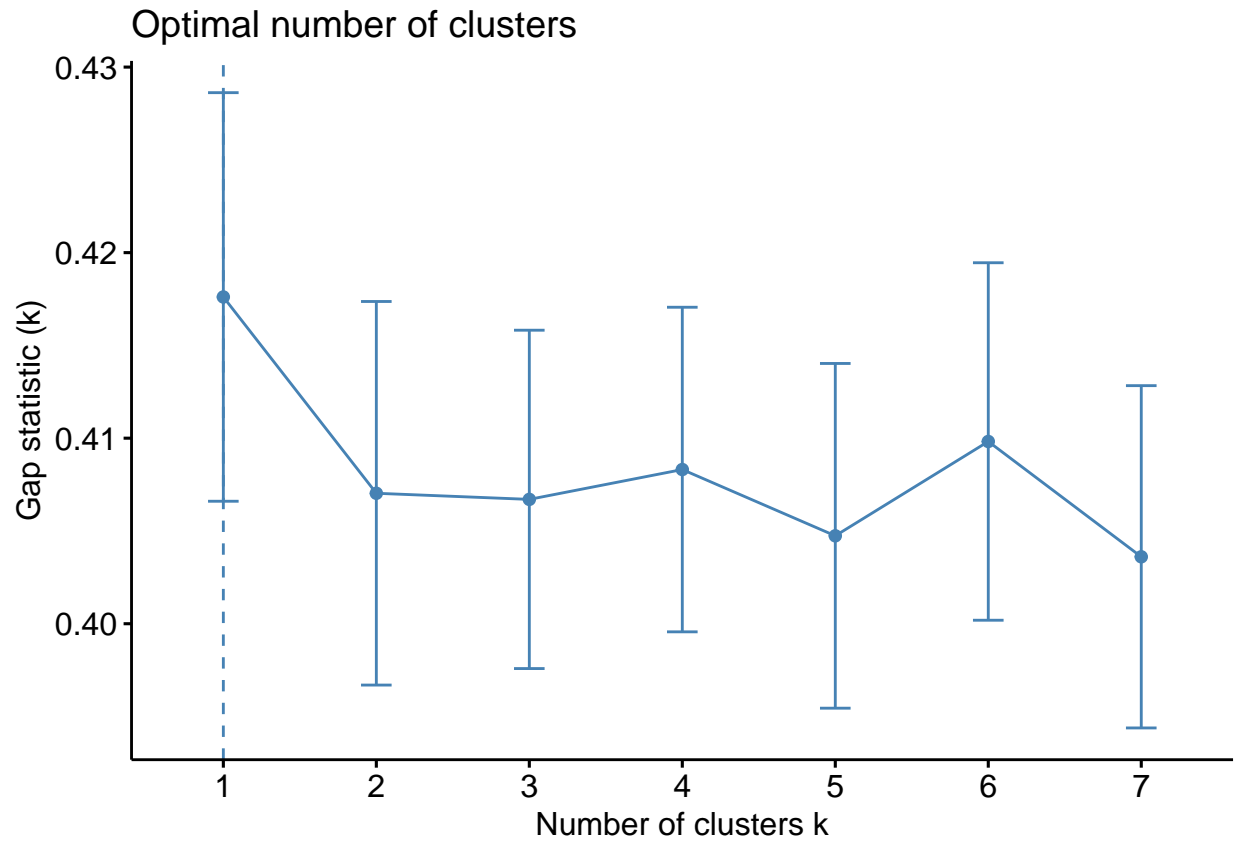


according to silhouette method, the number of clusters k should be 2 or 8. We can take 2 as we want to decide between two authors.

Gap Statistic Method

```
set.seed(123)
gap.stat = clusGap(Essay.unlabeled, FUN = kmeans, K.max = 7, B = 50)

set.seed(123)
fviz_gap_stat(gap.stat)
```

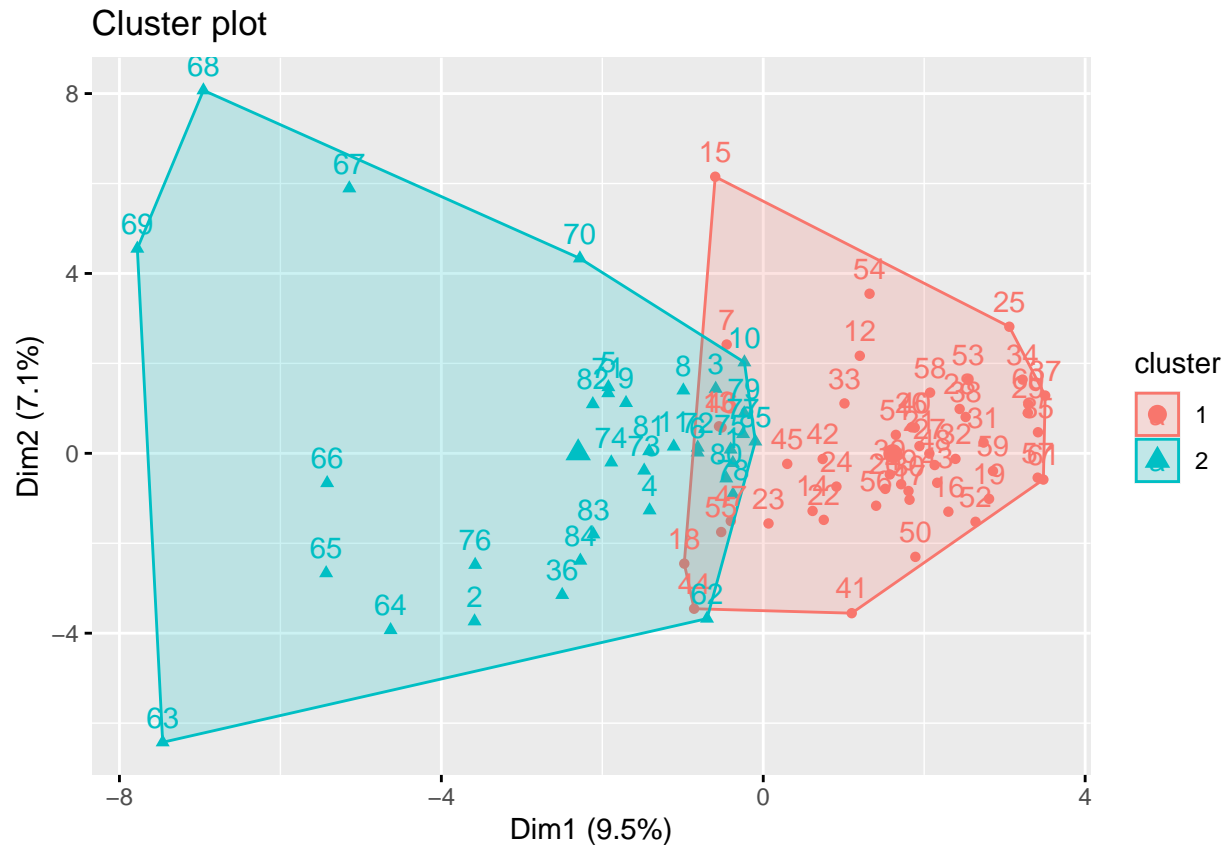
2, 5 and 7 clusters offer least gap statistic. We can use 2.

Extracting Results

```
set.seed(123)

final.res = kmeans(Essay.unlabeled, 2, nstart = 25)

fviz_cluster(final.res, data = Essay.unlabeled)
```



```
Essay.unlabeled = as.data.frame(Essay.unlabeled)
Essay.unlabeled$cluster = final.res$cluster

# Essay.unlabeled |> group_by(cluster) |> summarise_all(mean)
```

Let's inspect the clusters

```
Essay$cluster = final.res$cluster
Essay |>
  count(cluster) |>
  left_join(
    Essay |>
      select(cluster, author) |>
      group_by(cluster) |>
      slice(1:10) |> # just preview 10
      mutate(authors = str_c(author, collapse = ', ')) |> # collapse all animal names into one vector
      select(-author) |>
      ungroup() |>
      distinct()
  )
```

```
## Joining, by = "cluster"
```

```
##   cluster n
## 1       1 50
```

```
## 2      2 35
##
## 1 dispt, Hamilton, Hamilton, Hamilton, Hamilton, Hamilton, Hamilton, Hamilton, Hamilton, Hamilton
## 2      dispt, dispt, dispt, dispt, dispt, dispt, dispt, dispt, dispt, dispt, dispt
```

Conclusion

We observe that the documents in the left cluster have features that are similar to those of Hamilton, while the documents in the right cluster have traits that are more similar to those of other. We can suspect that the other author is Madison. Since the cluster with Madison includes most disputed documents, this suggests that Madison was the author for most of the disputed documents but Hamilton was the author for one of the disputed documents.