## R Code:

```
install.packages("dplyr")
install.packages("readr")
install.packages("ggplot2")
install.packages("scales")
install.packages("ggthemes")
install.packages("tidyr")
install.packages("AppliedPredictiveModeling")
install.packages("outliers")
install.packages("factoextra")
install.packages("cluster")
install.packages("ggfortify")
install.packages("fpc")
install.packages("NbClust")
library(dplyr)
library(readr)
library(ggplot2)
library(scales)
library(ggthemes)
library(tidyr)
library(NbClust)
library(cluster)
library(fpc)
library(outliers)
library(factoextra)
#----- read -----
help(read.csv)
?read.csv
```

googleplaystore <- read.csv("C:/Users/mrymh/OneDrive/Desktop/Data Mining Final submission/IT326-Dataset-section-52845-Group[2]/googleplaystore.csv")

```
#----- Rating vs.price scatter plot-----
{r echo= FALSE, message=FALSE, warning=FALSE, Bivariate_Plots}
#plot of rating vs. Price
ggplot(aes(x = Rating , y = Price), data = df)+
geom_jitter(alpha = 0.3, color = 'royalblue1')+
ylim(0,25)+
geom_line(stat = 'summary', fun.y = mean)+ ggtitle('Rating vs. Price')
#----- Plot for top genres vs. category ------
ggplot(aes (x = Genres), data = topgenres)+
geom_bar(aes(fill = Category))+
coord_flip()+
ggtitle('Top genres and Category')
#----- Pie chart for Type -----
{r echo= FALSE,message=FALSE, warning=FALSE}
#There are two types paid and unpaid
df_type = subset(df, (Type == 'Free' | Type == 'Paid'))
temp <- df_type%>%
group_by(Type)%>%
summarise(n = n())
#pie chart
```

```
ggplot(aes(x = ", y = n, fill = Type), data = temp )+
geom_bar(stat = 'identity')+
coord_polar('y', start = 0)+
theme_void()+
ggtitle('Type')
#-----preprocessing-----
# -----Replace missing values-----
googleplaystore$Rating[is.nan(googleplaystore$Rating)]<-0.0
#-----Remove outliers ------
outrev = outlier(googleplaystore$Rating, logical = TRUE)
sum(outrev)
Find_outlier = which(outrev ==TRUE, arr.ind = TRUE)
googleplaystore= googleplaystore[-Find_outlier,]
#-----
outGen = outlier(googleplaystore$Genres, logical = TRUE)
sum(outGen)
Find_outlier = which(outGen ==TRUE, arr.ind = TRUE)
googleplaystore= googleplaystore[-Find_outlier,]
outCon = outlier(googleplaystore$Content.Rating, logical = TRUE)
sum(outCon)
Find_outlier = which(outCon ==TRUE, arr.ind = TRUE)
googleplaystore= googleplaystore[-Find_outlier,]
#-----
outcat = outlier(googleplaystore$Category, logical = TRUE)
sum(outcat)
Find_outlier = which(outcat ==TRUE, arr.ind = TRUE)
googleplaystore= googleplaystore[-Find_outlier,]
```

#----- convert to numeric -----

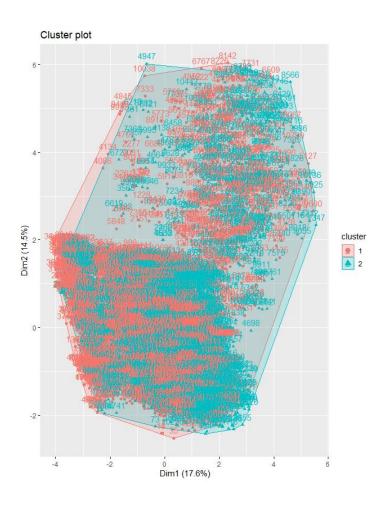
googleplaystore\$Category <- sapply(googleplaystore\$Category,as.numeric)
googleplaystore\$Reviews <- sapply(googleplaystore\$Reviews,as.numeric)
googleplaystore\$Size <- sapply(googleplaystore\$Size,as.numeric)
googleplaystore\$Installs <- sapply(googleplaystore\$Installs,as.numeric)
googleplaystore\$Type <- sapply(googleplaystore\$Type,as.numeric)
googleplaystore\$Price <- sapply(googleplaystore\$Price,as.numeric)
googleplaystore\$Content.Rating <- sapply(googleplaystore\$Content.Rating,as.numeric)
googleplaystore\$Genres <- sapply(googleplaystore\$Genres,as.numeric)
googleplaystore\$Last.Updated <- sapply(googleplaystore\$Last.Updated,as.numeric)
googleplaystore\$Current.Ver<- sapply(googleplaystore\$Current.Ver,as.numeric)
googleplaystore\$Android.Ver<- sapply(googleplaystore\$Android.Ver,as.numeric)

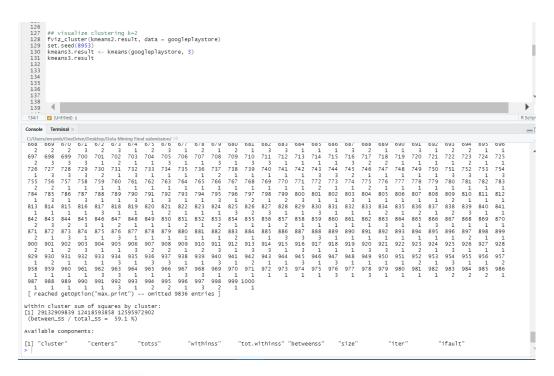
```
#-----
kmeans2.result <- kmeans(googleplaystore,2)</pre>
kmeans2.result
## visualize clustering k=2
fviz_cluster(kmeans2.result, data = googleplaystore)
set.seed(8953)
kmeans3.result <- kmeans(googleplaystore, 3)</pre>
kmeans3.result
## visualize clustering k=3
fviz_cluster(kmeans3.result, data = googleplaystore)
kmeans4.result <- kmeans(googleplaystore, 4)</pre>
kmeans4.result
## visualize clustering k=4
fviz_cluster(kmeans4.result, data = googleplaystore)
kmeans5.result <- kmeans(googleplaystore, 5)</pre>
kmeans5.result
## visualize clustering k=5
fviz_cluster(kmeans5.result, data = googleplaystore)
##-----
# group into 4 clusters
pam1.result <- pam(googleplaystore, 2)</pre>
plot(pam1.result)
fviz_cluster(pam1.result, data = googleplaystore)
pam2.result <- pam(googleplaystore, 3)</pre>
fviz_cluster(pam2.result, data = googleplaystore)
```

```
plot(pam2.result)
pam3.result <- pam(googleplaystore, 4)</pre>
fviz_cluster(pam3.result, data = googleplaystore)
plot(pam3.result)
pam4.result <- pam(googleplaystore, 5)</pre>
fviz_cluster(pam4.result, data = googleplaystore)
plot(pam4.result)
#---Evaluation
# group into 4 clusters
pam.result <- pam(googleplaystore, 2)</pre>
plot(pam.result)
pam.result <- pam(googleplaystore, 3)</pre>
plot(pam.result)
pam.result <- pam(googleplaystore, 4)</pre>
plot(pam.result)
pam.result <- pam(googleplaystore, 5)</pre>
plot(pam.result)
##for all clusters
fviz_nbclust(googleplaystore, kmeans, method = "silhouette")+ labs(subtitle = "Silhouette method")
#extra heririchal clustering
install.packages("pvcluster")
```

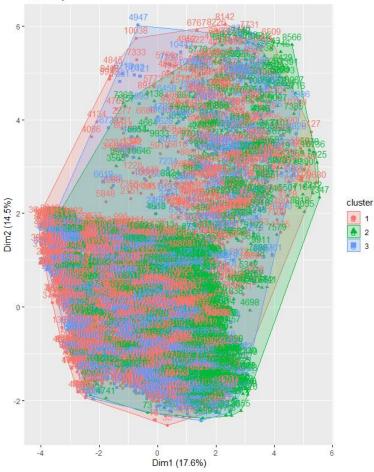
```
library(pvclust)
dd <- dist(scale(googleplaystore), method = "euclidean")</pre>
hc <- hclust(dd, method = "ward.D")</pre>
plot(hc)
set.seed(1234)
result <- pvclust(googleplaystore[1:100, 1:10], method.dist="euclidean",
          method.hclust="average", nboot=10)
plot(result)
pvrect(result)
dend <- googleplaystore[1:30,-5] %>% scale %>% dist %>%
 hclust %>% as.dendrogram %>%
 set("branches_k_color", k=3) %>% set("branches_lwd", 1.2) %>%
 set("labels_colors") %>% set("labels_cex", c(.9,1.2)) %>%
 set("leaves_pch", 19) %>% set("leaves_col", c("blue", "red"))
# plot the dend in usual "base" plotting engine:
plot(dend)
```

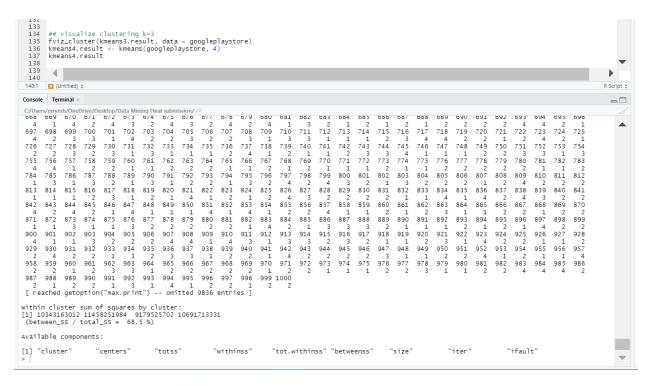
# Figures:



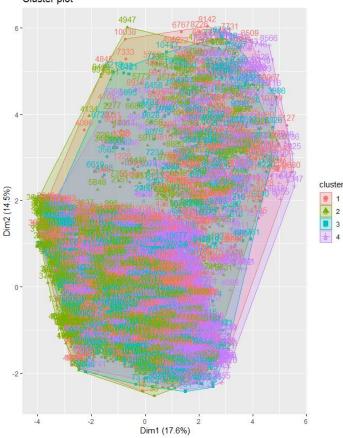


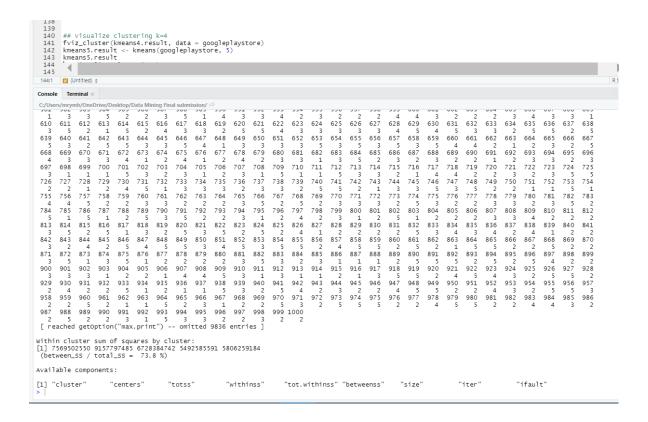


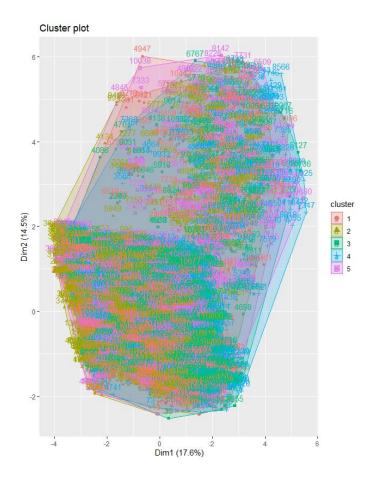








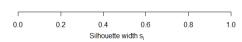




Silhouette plot of pam(x = googleplaystore, k = 2) n = 10836 2 clusters  $C_j$   $j: n_j \mid ave_{jec_j} s_i$ 

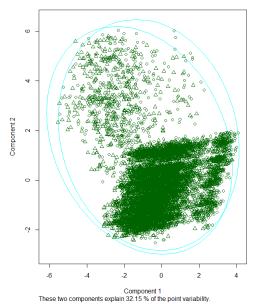
1: 5835 | 0.38

2: 5001 | 0.41



Average silhouette width: 0.4

#### clusplot(pam(x = googleplaystore, k = 2))



#### Silhouette plot of pam(x = googleplaystore, k = 3)

3 clusters C<sub>j</sub> j: n<sub>j</sub> | ave<sub>i∈Cj</sub> s<sub>i</sub> n = 10836

1: 3940 | 0.29

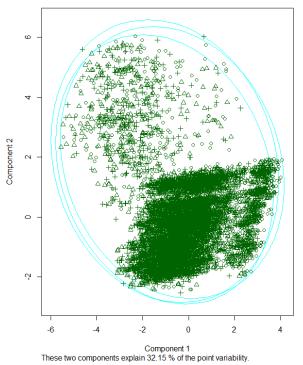
2: 3442 | 0.30

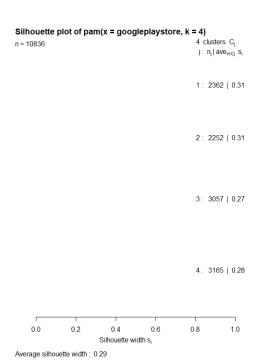
3: 3454 | 0.22



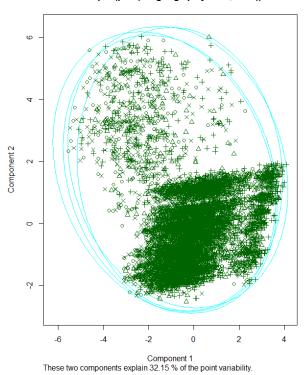
Average silhouette width: 0.27

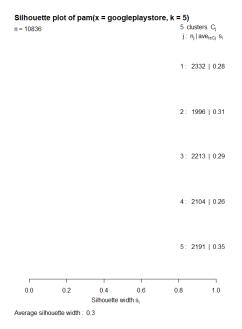
#### clusplot(pam(x = googleplaystore, k = 3))



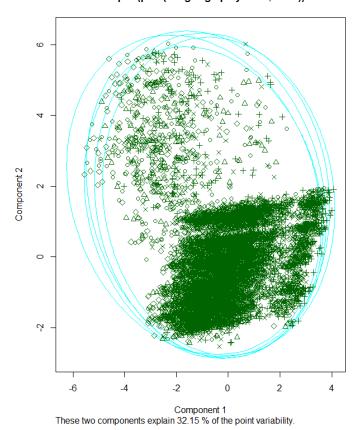


#### clusplot(pam(x = googleplaystore, k = 4))

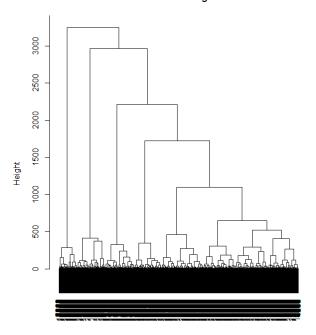




#### clusplot(pam(x = googleplaystore, k = 5))



#### **Cluster Dendrogram**



dd hclust (\*, "ward.D")

# **Output:**

406 407 408 409 410 411 412 413 414 415 416 417 418 419 420 2 2 1 1 2 2 1 2 1 1 2 1 1 1 1 1 2 1 2 2 1 1 2 2 1 1 1 2 1 526 527 528 529 530 531 532 533 534 535 536 537 538 539 540 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435 1 1 1 2 1 1 1 2 2 1 1 2 2 1 2 541 542 543 544 545 546 547 548 549 550 551 552 553 554 555 1 1 1 1 2 1 1 1 2 2 2 1 1 1 2 436 437 438 439 440 441 442 443 444 445 446 447 448 449 450 1 2 1 2 1 1 1 2 2 2 2 1 1 1 1 1 2 1 1 1 2 1 2 2 2 1 1 1 2 1 556 557 558 559 560 561 562 563 564 565 566 567 568 569 570 451 452 453 454 455 456 457 458 459 460 461 462 463 464 465 1 1 1 1 2 2 2 2 1 2 1 1 2 2 1 2 1 1 1 1 2 2 1 1 1 1 2 1 1 1 571 572 573 574 575 576 577 578 579 580 581 582 583 584 585 466 467 468 469 470 471 472 473 474 475 476 477 478 479 480 2 1 1 1 2 2 2 1 1 1 2 1 586 587 588 589 590 591 592 593 594 595 596 597 598 599 600 481 482 483 484 485 486 487 488 489 490 491 492 493 494 495 1 1 1 2 2 1 1 2 1 2 1 1 1 2 2 2 2 1 1 1 1 1 2 2 1 2 1 2 2 2 601 602 603 604 605 606 607 608 609 610 611 612 613 614 615 496 497 498 499 500 501 502 503 504 505 506 507 508 509 510 1 1 1 1 2 2 1 1 2 1 1 1 2 1 1 2 2 1 1 1 2 2 2 1 2 2 1 1 1 1 616 617 618 619 620 621 622 623 624 625 626 627 628 629 630 2 1 1 1 1 1 2 1 2 1 1 2 2 1 2 511 512 513 514 515 516 517 518 519 520 521 522 523 524 525

631 632 633 634 635 636 637 638 639 640 641 642 643 644 645 1 1 1 1 1 1 2 1 1 1 2 1 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 826 827 828 829 830 831 832 833 834 835 836 837 838 839 840 646 647 648 649 650 651 652 653 654 655 656 657 658 659 660 2 2 1 1 1 1 1 1 2 1 2 1 2 2 1 1 2 2 1 1 2 1 1 1 1 1 1 2 1 2 841 842 843 844 845 846 847 848 849 850 851 852 853 854 855 661 662 663 664 665 666 667 668 669 670 671 672 673 674 675 1 2 1 2 1 1 2 1 1 1 2 1 1 1 1 2 1 2 1 1 1 1 2 1 2 1 2 2 1 2 856 857 858 859 860 861 862 863 864 865 866 867 868 869 870 676 677 678 679 680 681 682 683 684 685 686 687 688 689 690 1 2 1 1 1 1 1 2 1 1 1 1 1 1 1 871 872 873 874 875 876 877 878 879 880 881 882 883 884 885 2 1 2 1 2 1 2 2 1 1 1 1 1 1 1 691 692 693 694 695 696 697 698 699 700 701 702 703 704 705 1 1 2 1 1 2 1 1 1 1 1 2 1 1 1 1 2 2 1 1 2 1 2 2 1 2 1 1 2 886 887 888 889 890 891 892 893 894 895 896 897 898 899 900 706 707 708 709 710 711 712 713 714 715 716 717 718 719 720 2 2 2 1 1 1 1 1 1 1 1 2 1 1 2 1 1 2 1 2 2 1 1 1 1 2 2 2 1 1 901 902 903 904 905 906 907 908 909 910 911 912 913 914 915 721 722 723 724 725 726 727 728 729 730 731 732 733 734 735 1 1 2 1 1 1 2 2 1 2 2 1 2 2 1 916 917 918 919 920 921 922 923 924 925 926 927 928 929 930 1 1 2 1 1 1 1 2 1 2 1 2 1 1 1 736 737 738 739 740 741 742 743 744 745 746 747 748 749 750 2 1 1 1 1 2 1 2 1 1 1 1 1 2 1 2 1 1 1 1 1 2 2 2 1 1 1 1 1 931 932 933 934 935 936 937 938 939 940 941 942 943 944 945 751 752 753 754 755 756 757 758 759 760 761 762 763 764 765 1 1 1 2 1 2 2 1 1 1 1 2 1 1 1 2 2 1 2 2 2 1 1 1 1 1 1 1 1 1 946 947 948 949 950 951 952 953 954 955 956 957 958 959 960 766 767 768 769 770 771 772 773 774 775 776 777 778 779 780 1 2 1 1 1 1 2 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 961 962 963 964 965 966 967 968 969 970 971 972 973 974 975 781 782 783 784 785 786 787 788 789 790 791 792 793 794 795 1 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 1 2 1 1 2 1 1 1 2 2 976 977 978 979 980 981 982 983 984 985 986 987 988 989 990 796 797 798 799 800 801 802 803 804 805 806 807 808 809 810 1 1 2 1 1 1 1 2 2 2 1 1 1 1 1 1 2 1 2 2 1 1 2 1 1 1 1 2 1 991 992 993 994 995 996 997 998 999 1000 811 812 813 814 815 816 817 818 819 820 821 822 823 824 825 1 2 1 2 1 1 1 1 1 1

[ reached getOption("max.print") -- omitted 9836 entries ]

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

[1] 35675021689 33475326394

(between\_SS / total\_SS = 47.8 %)

#### **Available components:**

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

- > ## visualize clustering k=2
- > fviz\_cluster(kmeans2.result, data = googleplaystore)
- > set.seed(8953)
- > kmeans3.result <- kmeans(googleplaystore, 3)
- > kmeans3.result

K-means clustering with 3 clusters of sizes 3069, 4644, 3123

#### **Cluster means:**

App Category Rating Reviews Size Installs Type Price

1 6867.025 18.59759 3.490974 1152.814 215.1854 10.190942 2.185728 87.33529

2 2148.195 18.59281 3.442442 2685.810 192.6906 10.423127 2.141473 88.20457

3 7078.711 19.04323 4.016971 4395.316 227.2856 9.988793 2.119757 88.63305

Content.Rating Genres Last.Updated Current.Ver Android.Ver

- 1 3.491365 65.31248 621.7908 1277.691 19.94037
- 2 3.399871 65.71835 652.2808 1195.169 18.77239
- 3 3.536663 66.64521 611.2389 1464.661 20.57797

#### Clustering vector:

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

> plot(kmeans3.result, data = googleplaystore)

'x' is a list, but does not have components 'x' and 'y'

> googleplaystore\$Category <- sapply(googleplaystore\$Category,as.numeric)

Error in xy.coords(x, y, xlabel, ylabel, log):

> ##-----

> ##Data mining task

> #-----

- > googleplaystore\$Reviews <- sapply(googleplaystore\$Reviews,as.numeric)
- > googleplaystore\$Size <- sapply(googleplaystore\$Size,as.numeric)
- > googleplaystore\$Installs <- sapply(googleplaystore\$Installs,as.numeric)
- > googleplaystore\$Type <- sapply(googleplaystore\$Type,as.numeric)
- > googleplaystore\$Price <- sapply(googleplaystore\$Price,as.numeric)
- > googleplaystore\$Content.Rating <- sapply(googleplaystore\$Content.Rating,as.numeric)
- > googleplaystore\$Genres <- sapply(googleplaystore\$Genres,as.numeric)
- > googleplaystore\$Last.Updated <- sapply(googleplaystore\$Last.Updated,as.numeric)
- > googleplaystore\$Current.Ver<- sapply(googleplaystore\$Current.Ver,as.numeric)
- > googleplaystore\$Android.Ver<- sapply(googleplaystore\$Android.Ver,as.numeric)
- > googleplaystore\$App<- sapply(googleplaystore\$App,as.numeric)
- > #-----
- > kmeans2.result <- kmeans(googleplaystore,2)
- > kmeans2.result

K-means clustering with 2 clusters of sizes 5601, 5235

#### **Cluster means:**

App Category Rating Reviews Size Installs Type Price

- $1\ 7237.547\ 18.80950\ 3.830673\ 2858.093\ 225.4133\ 10.06017\ 2.152473\ 87.99125$
- 2 2410.779 18.63247 3.398262 2622.594 191.5056 10.41624 2.142693 88.17880

31 32 33 34 35 36 37 38 39 40 41 42 43 44 45

46 47 48 49 50 51 52 53 54 55 56 57 58 59 60

1 2 2 2 1 1 1 2 1 1 2 1 1 1 1

2 1 1 1 1 1 1 2 1 1 1 2 1 1

Content.Rating Genres Last.Updated Current.Ver Android.Ver

- 1 3.526335 65.79147 615.4306 1393.747 20.37475
- 2 3.399809 65.95511 649.3486 1191.854 18.81987

#### **Clustering vector:**

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15

1 2 1 1 1 1 1 1 1 1 1 1 1 1 2

16 17 18 19 20 21 22 23 24 25 26 27 28 29 30

1 1 2 2 1 1 2 1 1 1 1 2 2 1 2

```
61 62 63 64 65 66 67 68 69 70 71 72 73 74 75
                                                               346 347 348 349 350 351 352 353 354 355 356 357 358 359 360
2 2 2 2 2 2 1 1 1 1 2 1 2 1 1
                                                               1 1 1 2 2 1 1 2 1 1 2 1 1 2 1
                                                               361 362 363 364 365 366 367 368 369 370 371 372 373 374 375
76 77 78 79 80 81 82 83 84 85 86 87 88 89 90
1 1 1 2 1 1 1 1 1 1 2 2 1 2 1
                                                               2 1 1 1 1 1 1 1 1 2 1 1 1 1 2
91 92 93 94 95 96 97 98 99 100 101 102 103 104 105
                                                               376 377 378 379 380 381 382 383 384 385 386 387 388 389 390
1 1 1 2 1 2 1 1 1 1 1 2 1 2 1
                                                               2 1 1 1 1 1 1 1 1 1 1 1 2 1
106 107 108 109 110 111 112 113 114 115 116 117 118 119 120
                                                               391 392 393 394 395 396 397 398 399 400 401 402 403 404 405
2 1 1 1 1 1 2 1 1 2 1 1 2 1 1
                                                               1 1 1 1 1 1 2 1 2 1 1 1 1 1 1
121 122 123 124 125 126 127 128 129 130 131 132 133 134 135
                                                               406 407 408 409 410 411 412 413 414 415 416 417 418 419 420
2 1 1 1 1 2 1 2 1 1 1 2 2 2 1
                                                                1 2 1 2 2 1 1 2 2 1 1 1 2 1
136 137 138 139 140 141 142 143 144 145 146 147 148 149 150
                                                               421 422 423 424 425 426 427 428 429 430 431 432 433 434 435
1 1 1 1 1 2 2 1 2 2 2 1 1 2 2
                                                               1 1 1 1 2 1 1 1 2 2 2 1 1 1 2
151 152 153 154 155 156 157 158 159 160 161 162 163 164 165
                                                               436 437 438 439 440 441 442 443 444 445 446 447 448 449 450
2 2 1 2 1 1 1 1 2 2 1 1 1 2 2
                                                               1 2 1 1 1 2 1 2 2 2 1 1 1 2 1
166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
                                                               451 452 453 454 455 456 457 458 459 460 461 462 463 464 465
1 1 2 2 2 2 2 2 1 2 1 2 1 1
                                                                2 1 1 1 1 2 2 1 1 1 1 2 1 1 1
                                                               466 467 468 469 470 471 472 473 474 475 476 477 478 479 480
181 182 183 184 185 186 187 188 189 190 191 192 193 194 195
1 2 1 2 2 1 2 1 1 1 2 1 2 1 1
                                                                1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
196 197 198 199 200 201 202 203 204 205 206 207 208 209 210
                                                               481 482 483 484 485 486 487 488 489 490 491 492 493 494 495
1 1 2 1 2 1 2 1 1 2 1 2 1 1 1
                                                               2 2 1 1 1 1 1 2 2 1 2 1 2 2 2
211 212 213 214 215 216 217 218 219 220 221 222 223 224 225
                                                               496 497 498 499 500 501 502 503 504 505 506 507 508 509 510
1 1 2 1 2 2 1 1 1 1 2 1 1 1 2
                                                                2 2 1 1 1 2 2 2 1 2 2 1 1 1 1
226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
                                                               511 512 513 514 515 516 517 518 519 520 521 522 523 524 525
1 1 1 1 1 1 1 2 1 1 1 2 1 1 1
                                                                2 2 1 1 2 2 1 2 1 1 2 1 1 1 1
241 242 243 244 245 246 247 248 249 250 251 252 253 254 255
                                                               526 527 528 529 530 531 532 533 534 535 536 537 538 539 540
1 2 1 1 1 1 1 2 2 1 2 2 1 1 1
                                                                1 1 1 2 1 1 1 2 2 1 1 2 2 1 2
256 257 258 259 260 261 262 263 264 265 266 267 268 269 270
                                                               541 542 543 544 545 546 547 548 549 550 551 552 553 554 555
2 1 1 2 1 1 1 2 2 1 2 1 1 1 1
                                                               1 2 1 2 1 1 1 2 2 2 2 1 1 1 1
271 272 273 274 275 276 277 278 279 280 281 282 283 284 285
                                                               556 557 558 559 560 561 562 563 564 565 566 567 568 569 570
2 1 1 1 1 1 2 2 1 2 2 1 1 1 1
                                                               1 1 1 1 2 2 2 2 1 2 1 1 2 2 1
                                                               571 572 573 574 575 576 577 578 579 580 581 582 583 584 585
286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
1 1 1 2 1 1 1 2 1 1 1 1 1 1 1
                                                                2 1 1 1 2 2 2 1 1 1 2 1 1 1 1
301 302 303 304 305 306 307 308 309 310 311 312 313 314 315
                                                               586 587 588 589 590 591 592 593 594 595 596 597 598 599 600
1 1 1 1 1 1 1 2 1 1 2 2 2 2 2 2
                                                               1 1 1 2 2 1 1 2 1 2 1 1 1 2 2
316 317 318 319 320 321 322 323 324 325 326 327 328 329 330
                                                               601 602 603 604 605 606 607 608 609 610 611 612 613 614 615
2 1 1 1 2 1 2 2 1 2 1 2 1 2 1
                                                               1 1 1 1 2 2 1 1 2 1 1 1 2 1 1
331 332 333 334 335 336 337 338 339 340 341 342 343 344 345
                                                               616 617 618 619 620 621 622 623 624 625 626 627 628 629 630
1 2 1 1 1 1 1 1 1 1 1 1 2
                                                               2 1 1 1 1 1 2 1 2 1 1 2 2 1 2
```

631 632 633 634 635 636 637 638 639 640 641 642 643 644 645 1 1 1 1 1 1 2 1 1 1 2 1 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 826 827 828 829 830 831 832 833 834 835 836 837 838 839 840 646 647 648 649 650 651 652 653 654 655 656 657 658 659 660 2 2 1 1 1 1 1 1 2 1 2 1 2 2 1 1 2 2 1 1 2 1 1 1 1 1 1 2 1 2 841 842 843 844 845 846 847 848 849 850 851 852 853 854 855 661 662 663 664 665 666 667 668 669 670 671 672 673 674 675 1 2 1 2 1 1 2 1 1 1 2 1 1 1 1 2 1 2 1 1 1 1 2 1 2 1 2 2 1 2 856 857 858 859 860 861 862 863 864 865 866 867 868 869 870 676 677 678 679 680 681 682 683 684 685 686 687 688 689 690 1 2 1 1 1 1 1 2 1 1 1 1 1 1 1 871 872 873 874 875 876 877 878 879 880 881 882 883 884 885 2 1 2 1 2 1 2 2 1 1 1 1 1 1 1 691 692 693 694 695 696 697 698 699 700 701 702 703 704 705 1 1 2 1 1 2 1 1 1 1 1 2 1 1 1 1 2 2 1 1 2 1 2 2 1 2 1 1 2 886 887 888 889 890 891 892 893 894 895 896 897 898 899 900 706 707 708 709 710 711 712 713 714 715 716 717 718 719 720 2 2 2 1 1 1 1 1 1 1 1 2 1 1 2 1 1 2 1 2 2 1 1 1 1 2 2 2 1 1 901 902 903 904 905 906 907 908 909 910 911 912 913 914 915 721 722 723 724 725 726 727 728 729 730 731 732 733 734 735 1 1 2 1 1 1 2 2 1 2 2 1 2 2 1 916 917 918 919 920 921 922 923 924 925 926 927 928 929 930 1 1 2 1 1 1 1 2 1 2 1 2 1 1 1 736 737 738 739 740 741 742 743 744 745 746 747 748 749 750 2 1 1 1 1 2 1 2 1 1 1 1 1 2 1 2 1 1 1 1 1 2 2 2 1 1 1 1 1 931 932 933 934 935 936 937 938 939 940 941 942 943 944 945 751 752 753 754 755 756 757 758 759 760 761 762 763 764 765 1 1 1 2 1 2 2 1 1 1 1 2 1 1 1 2 2 1 2 2 2 1 1 1 1 1 1 1 1 1 946 947 948 949 950 951 952 953 954 955 956 957 958 959 960 766 767 768 769 770 771 772 773 774 775 776 777 778 779 780 1 2 1 1 1 1 2 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 961 962 963 964 965 966 967 968 969 970 971 972 973 974 975 781 782 783 784 785 786 787 788 789 790 791 792 793 794 795 1 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 1 2 1 1 2 1 1 1 2 976 977 978 979 980 981 982 983 984 985 986 987 988 989 990 796 797 798 799 800 801 802 803 804 805 806 807 808 809 810 1 1 2 1 1 1 1 2 2 2 1 1 1 1 1 1 2 1 2 2 1 1 2 1 1 1 1 2 1 991 992 993 994 995 996 997 998 999 1000 811 812 813 814 815 816 817 818 819 820 821 822 823 824 825 1 2 1 2 1 1 1 1 1 1

[ reached getOption("max.print") -- omitted 9836 entries ]

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

[1] 35675021689 33475326394

(between\_SS / total\_SS = 47.8 %)

#### **Available components:**

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

- > ## visualize clustering k=2
- > fviz\_cluster(kmeans2.result, data = googleplaystore)
- > set.seed(8953)
- > kmeans3.result <- kmeans(googleplaystore, 3)
- > kmeans3.result

K-means clustering with 3 clusters of sizes 3069, 4644, 3123

#### **Cluster means:**

App Category Rating Reviews Size Installs Type Price

1 6867.025 18.59759 3.490974 1152.814 215.1854 10.190942 2.185728 87.33529

2 2148.195 18.59281 3.442442 2685.810 192.6906 10.423127 2.141473 88.20457

3 7078.711 19.04323 4.016971 4395.316 227.2856 9.988793 2.119757 88.63305

Content.Rating Genres Last.Updated Current.Ver Android.Ver

- 1 3.491365 65.31248 621.7908 1277.691 19.94037
- 2 3.399871 65.71835 652.2808 1195.169 18.77239
- 3 3.536663 66.64521 611.2389 1464.661 20.57797

### Clustering vector:

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

3 3 2 1 1 3 3 2 2 1 3 3 3 3 3

991 992 993 994 995 996 997 998 999 1000

1 2 1 2 1 3 3 1 3 3

[ reached getOption("max.print") -- omitted 9836 entries ]

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

[1] 13261402224 27056449783 13830733105

886 887 888 889 890 891 892 893 894 895 896 897 898 899 900

(between SS / total SS = 59.1 %)

#### Available components:

- [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
- [6] "betweenss" "size" "iter" "ifault"
- > ## visualize clustering k=3
- > fviz\_cluster(kmeans3.result, data = googleplaystore)
- > kmeans4.result <- kmeans(googleplaystore, 4)
- > kmeans4.result

K-means clustering with 4 clusters of sizes 2841, 2747, 2497, 2751

#### **Cluster means:**

App Category Rating Reviews Size Installs Type Price

1 7097.168 18.98381 4.022527 4409.852 228.8508 9.94333 2.119676 88.72826

2 2554.132 18.44266 3.062723 1071.518 185.2756 10.36767 2.147434 88.01274

3 2264.517 18.82739 3.767121 4294.469 197.3444 10.48018 2.140969 88.26712

4 7387.924 18.64268 3.634206 1287.648 222.8957 10.17012 2.183206 87.31516

Content.Rating Genres Last.Updated Current.Ver Android.Ver

1 3.532207 66.35903 607.2925 1473.073 20.62443

2 3.397161 65.54132 654.5497 1129.056 18.74081

3 3.400881 66.37205 641.9828 1257.099 18.86304

4 3.522356 65.23955 625.2163 1315.974 20.16176

### Clustering vector:

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15	2 2 4 3 1 1 1 1 2 2 4 1 4 2 3
4 3 1 4 1 4 4 1 2 4 4 1 1 2 2	166 167 168 169 170 171 172 173 174 175 176 177 178 179 180
16 17 18 19 20 21 22 23 24 25 26 27 28 29 30	1 1 3 2 2 2 3 3 1 3 4 2 2 4 4
1 1 3 2 4 1 3 1 1 4 4 2 2 4 2	181 182 183 184 185 186 187 188 189 190 191 192 193 194 195
31 32 33 34 35 36 37 38 39 40 41 42 43 44 45	4 3 1 3 2 4 3 4 1 4 3 1 2 1 4
4 2 3 2 1 1 1 1 4 1 2 4 4 4 4	196 197 198 199 200 201 202 203 204 205 206 207 208 209 210
46 47 48 49 50 51 52 53 54 55 56 57 58 59 60	4 4 3 1 3 1 2 1 1 2 1 2 4 1 1
2 4 1 1 1 4 4 4 2 1 4 1 2 4 1	211 212 213 214 215 216 217 218 219 220 221 222 223 224 225
61 62 63 64 65 66 67 68 69 70 71 72 73 74 75	1 4 3 4 2 2 4 4 1 4 3 1 1 4 3
3 3 3 2 3 3 4 1 1 1 2 4 3 1 1	226 227 228 229 230 231 232 233 234 235 236 237 238 239 240
76 77 78 79 80 81 82 83 84 85 86 87 88 89 90	4 4 1 1 1 1 4 2 1 4 4 2 4 1 1
4 1 1 3 4 4 4 4 1 2 3 4 2 1	241 242 243 244 245 246 247 248 249 250 251 252 253 254 255
91 92 93 94 95 96 97 98 99 100 101 102 103 104 105	1 2 4 4 1 4 1 3 2 1 3 3 1 1 4
1 1 1 3 1 2 4 2 4 1 4 2 1 2 1	256 257 258 259 260 261 262 263 264 265 266 267 268 269 270
106 107 108 109 110 111 112 113 114 115 116 117 118 119 120	2 4 1 2 1 4 1 3 2 1 2 4 1 1 1
3 1 1 4 4 1 3 4 4 3 4 1 2 1 1	271 272 273 274 275 276 277 278 279 280 281 282 283 284 285
121 122 123 124 125 126 127 128 129 130 131 132 133 134 135	2 4 4 1 4 1 3 2 1 3 3 1 1 4 1
3 1 4 4 4 2 1 2 1 1 1 3 2 3 1	286 287 288 289 290 291 292 293 294 295 296 297 298 299 300
136 137 138 139 140 141 142 143 144 145 146 147 148 149 150	1 1 4 2 1 4 4 2 4 1 1 4 4 4 1
1 1 4 1 4 2 3 1 3 2 3 4 1 2 2	301 302 303 304 305 306 307 308 309 310 311 312 313 314 315
151 152 153 154 155 156 157 158 159 160 161 162 163 164 165	4 1 1 4 4 1 1 3 1 4 2 3 2 2 3

4 4 3 4 4 3 1 1 1 1 1 4 2 1 4

1 4 4 3 2 1 4 2 1 2 1 1 1 2 2

886 887 888 889 890 891 892 893 894 895 896 897 898 899 900 946 947 948 949 950 951 952 953 954 955 956 957 958 959 960 3 3 3 1 4 4 4 1 4 1 4 2 1 1 2 1 3 4 4 1 1 2 4 1 4 4 2 1 1 4 901 902 903 904 905 906 907 908 909 910 911 912 913 914 915 961 962 963 964 965 966 967 968 969 970 971 972 973 974 975 4 4 3 1 1 1 2 2 4 2 3 4 3 3 1 1 3 3 4 1 1 1 1 1 4 1 1 4 4 4 916 917 918 919 920 921 922 923 924 925 926 927 928 929 930 976 977 978 979 980 981 982 983 984 985 986 987 988 989 990 3 1 4 4 1 3 4 2 1 1 4 4 1 1 2 1 1 3 4 4 1 1 2 2 2 1 1 4 1 1 931 932 933 934 935 936 937 938 939 940 941 942 943 944 945 991 992 993 994 995 996 997 998 999 1000 1 1 4 3 1 3 3 4 1 1 4 2 1 1 1 4 3 4 2 4 1 1 4 1 1

[ reached getOption("max.print") -- omitted 9836 entries ]

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

[1] 11453483107 10557233161 9307249817 10354205077

(between\_SS / total\_SS = 68.5 %)

#### Available components:

- [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
- [6] "betweenss" "size" "iter" "ifault"
- > ## visualize clustering k=4
- > fviz\_cluster(kmeans4.result, data = googleplaystore)
- > kmeans5.result <- kmeans(googleplaystore, 5)
- > kmeans5.result

K-means clustering with 5 clusters of sizes 2508, 1934, 2224, 2196, 1974

#### **Cluster means:**

App Category Rating Reviews Size Installs Type Price

1 7260.772 19.02153 4.036722 4543.264 227.2791 9.940191 2.117225 88.76874

2 1555.544 18.55377 3.116960 1215.774 188.3283 10.394519 2.153568 87.72854

3 4850.503 18.40558 3.384442 1528.921 204.2963 10.265737 2.157374 88.12770

4 2319.006 18.75729 3.809107 4518.211 199.0437 10.499545 2.134791 88.35291

5 8135.481 18.83435 3.648126 1352.247 222.5805 10.108916 2.184397 87.20213

#### Content.Rating Genres Last.Updated Current.Ver Android.Ver

- 3.522727 66.28230 607.4725 1476.742 20.58413
- 2 3.385212 64.86194 662.0936 1199.965 18.90228
- 3.495504 65.44829 629.5544 1187.478 19.28552 3
- 3.413934 66.15710 644.0301 1250.067 18.74362
- 3.493414 66.49240 622.0456 1334.971 20.46960

#### **Clustering vector:**

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15

5 4 1 5 1 5 5 1 3 3 5 1 1 1 2 211 212 213 214 215 216 217 218 219 220 221 222 223 224 225 1 5 4 5 3 3 3 5 1 3 4 1 1 5 4 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 1 1 2 3 3 1 4 1 1 3 3 2 2 5 3 226 227 228 229 230 231 232 233 234 235 236 237 238 239 240 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 5 5 1 1 1 1 5 3 1 5 5 2 5 3 1 5 2 4 3 3 1 1 3 5 1 3 5 5 5 5 241 242 243 244 245 246 247 248 249 250 251 252 253 254 255 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 1 3 3 5 1 5 1 4 2 1 4 2 1 1 3

256 257 258 259 260 261 262 263 264 265 266 267 268 269 270 2 3 1 1 1 5 5 5 2 1 5 1 3 5 1

61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 2 5 1 2 1 5 1 4 2 1 2 5 3 1 1

271 272 273 274 275 276 277 278 279 280 281 282 283 284 285 4 4 3 3 4 2 5 1 1 1 3 5 2 1 1

76 77 78 79 80 81 82 83 84 85 86 87 88 89 90 3 3 5 1 5 1 4 2 1 4 2 1 1 3 1

5 1 1 4 5 5 5 5 5 1 2 4 5 2 1 286 287 288 289 290 291 292 293 294 295 296 297 298 299 300

91 92 93 94 95 96 97 98 99 100 101 102 103 104 105 1 1 5 3 1 5 5 3 5 1 1 5 3 3 1

1 1 1 2 1 2 3 3 3 1 5 2 1 2 1 301 302 303 304 305 306 307 308 309 310 311 312 313 314 315

106 107 108 109 110 111 112 113 114 115 116 117 118 119 120

4 1 1 5 5 1 4 5 5 3 5 1 2 3 1 316 317 318 319 320 321 322 323 324 325 326 327 328 329 330

121 122 123 124 125 126 127 128 129 130 131 132 133 134 135 2 3 3 5 2 1 2 3 1 2 3 4 5 3 3

4 3 5 3 5 3 1 3 1 1 1 4 3 4 3 331 332 333 334 335 336 337 338 339 340 341 342 343 344 345

136 137 138 139 140 141 142 143 144 145 146 147 148 149 150 1 4 5 1 1 1 1 3 1 3 1 3 5 5 3

1 1 3 1 5 3 4 1 4 2 4 5 3 2 3 346 347 348 349 350 351 352 353 354 355 356 357 358 359 360

151 152 153 154 155 156 157 158 159 160 161 162 163 164 165 1 1 1 4 4 5 1 2 1 1 2 5 5 3 3

3 3 3 4 1 1 1 1 3 2 5 1 5 2 4 361 362 363 364 365 366 367 368 369 370 371 372 373 374 375

166 167 168 169 170 171 172 173 174 175 176 177 178 179 180 4 1 1 5 1 1 1 5 1 4 5 3 1 1 2

1 1 4 3 3 3 2 4 1 4 5 3 3 5 5 376 377 378 379 380 381 382 383 384 385 386 387 388 389 390

181 182 183 184 185 186 187 188 189 190 191 192 193 194 195 2 3 1 5 5 3 1 1 1 3 5 3 5 4 1

196 197 198 199 200 201 202 203 204 205 206 207 208 209 210

5 4 1 2 2 5 3 5 1 5 4 1 3 1 5 391 392 393 394 395 396 397 398 399 400 401 402 403 404 405

1 5 5 5 3 3 4 3 4 3 1 3 5 3 3

5 1 1 3 3 3 1 4 3 5 2 4 2 2 3

5 3 4 1 4 1 3 1 1 2 1 2 3 1 1

691 692 693 694 695 696 697 698 699 700 701 702 703 704 705

1 4 4 5 1 3 4 1 1 5 3 1 5 5 5

406 407 408 409 410 411 412 413 414 415 416 417 418 419 420

4 1 2 1 3 3 4 3 5 1 3 1 3 1 1

1 1 2 5 5 1 1 2 2 3 1 1 5 1 1

3 4 5 3 3 1 1 3 1 1

[ reached getOption("max.print") -- omitted 9836 entries ]

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

[1] 9152948774 5495589969 6808452033 7569502550 5727822822

(between\_SS / total\_SS = 73.8 %)

#### **Available components:**

```
[1] "cluster" "centers"
                                       "withinss"
                           "totss"
                                                   "tot.withinss"
[6] "betweenss" "size"
                            "iter"
                                      "ifault"
> ## visualize clustering k=5
> fviz_cluster(kmeans5.result, data = googleplaystore)
> ##-----
> # group into 4 clusters
> pam1.result <- pam(googleplaystore, 2)
> plot(pam1.result)
Hit <Return> to see next plot: fviz_cluster(pam1.result, data = googleplaystore)
Hit <Return> to see next plot: pam2.result <- pam(googleplaystore, 3)
> fviz_cluster(pam2.result, data = googleplaystore)
> plot(pam2.result)
Hit <Return> to see next plot: pam3.result <- pam(googleplaystore, 4)
Hit <Return> to see next plot: fviz cluster(pam3.result, data = googleplaystore)
> plot(pam3.result)
Hit <Return> to see next plot: pam4.result <- pam(googleplaystore, 5)
Hit <Return> to see next plot: fviz_cluster(pam4.result, data = googleplaystore)
> plot(pam4.result)
Hit <Return> to see next plot: #---Evaluation
```

```
Hit <Return> to see next plot: ##-----
> # group into 4 clusters
> pam.result <- pam(googleplaystore, 2)
> plot(pam.result)
Hit <Return> to see next plot: pam.result <- pam(googleplaystore, 3)
Hit <Return> to see next plot:
> pam.result <- pam(googleplaystore, 3)
> plot(pam.result)
Hit <Return> to see next plot: pam.result <- pam(googleplaystore, 4)
Hit <Return> to see next plot: plot(pam.result)
> pam.result <- pam(googleplaystore, 5)
> plot(pam.result)
Hit <Return> to see next plot: fviz_nbclust(googleplaystore, kmeans, method = "silhouette")+
labs(subtitle = "Silhouette method")
Hit <Return> to see next plot:
> install.packages("pvcluster")
WARNING: Rtools is required to build R packages but is not currently installed. Please download and
install the appropriate version of Rtools before proceeding:
https://cran.rstudio.com/bin/windows/Rtools/
Installing package into 'C:/Users/Mashari Aljasser/Documents/R/win-library/3.6'
(as 'lib' is unspecified)
> library(pvclust)
> dd <- dist(scale(googleplaystore), method = "euclidean")</pre>
> hc <- hclust(dd, method = "ward.D")
> plot(hc)
> set.seed(1234)
> result <- pvclust(googleplaystore[1:100, 1:10], method.dist="euclidean",
```

```
method.hclust="average", nboot=10)
Bootstrap (r = 0.5)... Done.
Bootstrap (r = 0.6)... Done.
Bootstrap (r = 0.7)... Done.
Bootstrap (r = 0.8)... Done.
Bootstrap (r = 0.9)... Done.
Bootstrap (r = 1.0)... Done.
Bootstrap (r = 1.1)... Done.
Bootstrap (r = 1.2)... Done.
Bootstrap (r = 1.3)... Done.
Bootstrap (r = 1.4)... Done.
> plot(result)
> pvrect(result)
> result <- pvclust(googleplaystore[1:100, 1:10], method.dist="euclidean",
            method.hclust="average", nboot=10)
Bootstrap (r = 0.5)... Done.
Bootstrap (r = 0.6)... Done.
Bootstrap (r = 0.7)... Done.
Bootstrap (r = 0.8)... Done.
Bootstrap (r = 0.9)... Done.
Bootstrap (r = 1.0)... Done.
Bootstrap (r = 1.1)... Done.
Bootstrap (r = 1.2)... Done.
Bootstrap (r = 1.3)... Done.
Bootstrap (r = 1.4)... Done.
> plot(result)
> pvrect(result)
> dend <- googleplaystore[1:30,-5] %>% scale %>% dist %>%
```

- + hclust %>% as.dendrogram %>%
- + set("branches\_k\_color", k=3) %>% set("branches\_lwd", 1.2) %>%
- + set("labels\_colors") %>% set("labels\_cex", c(.9,1.2)) %>%
- + set("leaves\_pch", 19) %>% set("leaves\_col", c("blue", "red"))
- > # plot the dend in usual "base" plotting engine:
- > plot(dend)

>