314 A1-vignette

```
library(imager)
library(tidyverse)
library(tidymodels)
library(dplyr)
library(sp)
library(scales)
library(cowplot)
library(ggplot2)
#install.packages("devtools")
#library(devtools)
#devtools::install_github("sharlagelfand/dmc")
library(dmc)

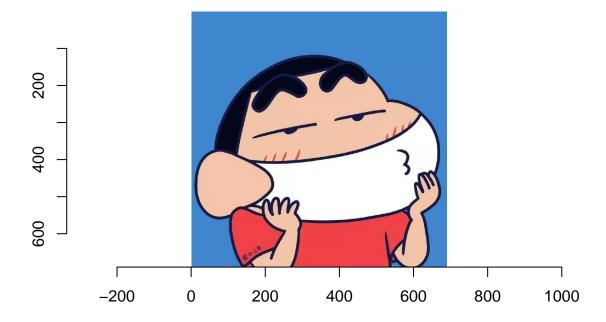
source('functions.R')
```

The tasks for this project is to write several R functions that will take an image and create a cross-stitch pattern. I write the function in 'function.R', and I will test the function in following parts.

First, I load the image by function imager::load.image(). The following is the picture I use.

```
set.seed(8477)
image_file_name<-'Crayon.jpg'
im<-imager::load.image(image_file_name)
plot(im,main="Image")</pre>
```

Image



The second step is to gain the tibble of information derived from the k_means. We use the function process_image() and store the result in 'gain_clusters'. Each row provide information for k-means clustering of each k.

k: k value from 2-8

totss: The total sum of squares.

betweenss: The between-cluster sum of squares, i.e. totss - tot.withinss.

tot.withinss: Total within-cluster sum of squares, i.e. sum(withinss).

totss, tot. withinss, betweenss, and iter (1 value) contain information about the full clustering

centers(nested column): A matrix of cluster centers. tidy_dat: contains coordinate,RGB information abour original data points. centers and tidy_dat contain information about each cluster

```
gain_clusters<-process_image(image_file_name,k=2:8)
gain_clusters</pre>
```

```
# A tibble: 7 x 8
##
##
         k kclust
                     totss tot.withinss betweenss
                                                     iter centres
                                                                       tidy_dat
                                                                       t>
##
     <int> <list>
                      <dbl>
                                   <dbl>
                                              <dbl> <int> <list>
## 1
         2 <kmean~ 139964.
                                  59752.
                                            80212.
                                                        1 <tibble [2~ <tibble [476,1~
                                                        3 <tibble [3~ <tibble [476,1~
## 2
         3 <kmean~ 139964.
                                  29067.
                                           110897.
## 3
         4 <kmean~ 139964.
                                   9042.
                                           130922.
                                                        3 <tibble [4~ <tibble [476,1~
                                                        3 <tibble [5~ <tibble [476,1~
## 4
         5 <kmean~ 139964.
                                   2082.
                                           137882.
## 5
         6 <kmean~ 139964.
                                   1510.
                                           138454.
                                                        4 <tibble [6~ <tibble [476,1~
## 6
         7 <kmean~ 139964.
                                           138845.
                                                        3 <tibble [7~ <tibble [476,1~
                                   1119.
```

The third step is to produce and plot a scree plot. We use the function scree plot() and recieve one graph with two plots.

The within-cluster sum of squares used to measure the variability of the observations within each cluster

The left graph shows that the total within-cluster sum of squares is decreasing slowly after k=4. After k=5, the total within-cluster sum of squares is close to the results in k=5. The value at k=5 is good.

The right graph shows the change of ratio(=tot.withinss(k)/tot.withinss(k-1)).

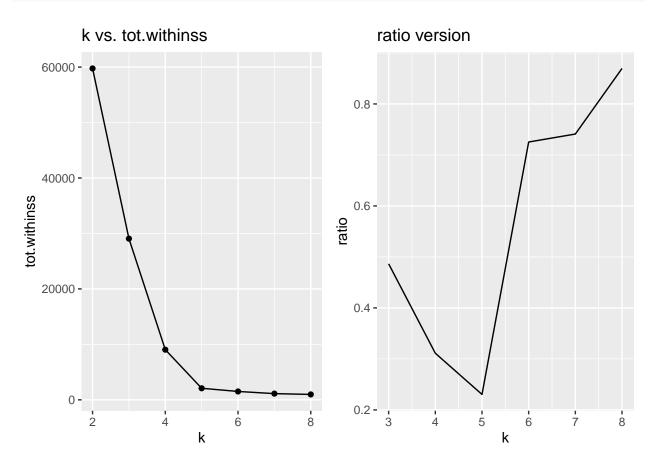
When k=7, ratio=tot.withinss(7)/tot.withinss(6)=0.74. The tot.withinss(k=7) is 72% of tot.withinss(k=6).

When k=6, ratio=tot.withinss(6)/tot.withinss(5)=0.72. The tot.withinss(k=6) is 72% of tot.withinss(k=5).

When k=5, ratio=tot.withinss(5)/tot.withinss(4)=0.22. The tot.withinss(k=5) is 22% of tot.withinss(k=4).

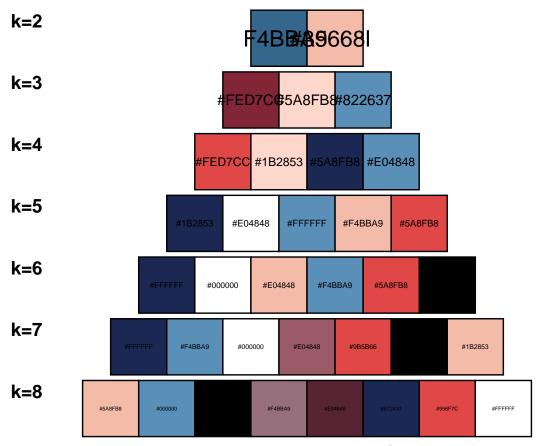
From this the number of clusters seems to be 6 or 7 (because increase trend becomes slow). And the tot.withinss is close when k=5,6 and 7. So we cannot have a certain conclusion that 6 is the right number of clusters. We will check it in following colour strips.

scree_plot(gain_clusters)



We use function colour_strips() to produce colour strips with the DMC colour. Visually, when k=6, color black and color dark blue is too similar (dark tones). Let's see what happens if we choose k-5, 5 colors are clearly unlike each other. So K=5 is much better choice.

colour_strips(gain_clusters)



So now we have 5 clusters we need to put it into make_pattern() function, and produce a cross stitch pattern that can be followed. I graph 2 versions, one with color and one without color.

