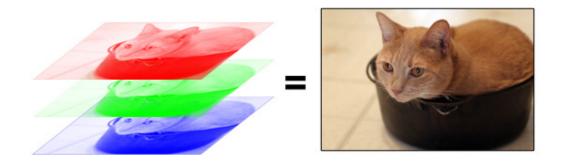
ImageCompression

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1 Image compression by k-means clustering

1.0.1 3-channel (RGB) color images in 24-bit color representation allows for more than 16 million (2²⁴) different colors. In the computer, an RGB image is stored as a stack of (3) matrices, hence each pixel of the image is a 3-dimensional vector reflecting the mixture of the colors Red, Green and Blue:



- 1.0.2 By doing k-means clustering of the 3-dimensional (RGB) pixel vectors, we can find just a few (k) useful color combinations (the resulting cluster centers) to replace the original pixel vectors in a compressed version of the image.
- 1.0.3 See also the VMLS k-means notes on clustering applications.
- 2 Candidate color images for doing pixel-clustering







```
[1]: # import Pkg; Pkg.add("Images"); Pkg.add("ImageMagick")
# import Pkg; Pkg.add("HTTP")
using Images # Image loading, saving, manipulation
using HTTP # Internet access
```

[2]: using Plots # Precompiles on every startup (~20 secondss)
gr() # Needs modules Plots and GR to be installed, may need a rebuild of GR

→with ']build GR'
default(size=(600, 450), fmt = :png) # Default plot size, change output format

→to png

2.1 Load an image either locally or from the internet

```
[3]: # Load image stored on a local folder adress:
     # myimage = "C:/Users/ulfin/Dropbox/MATH310/Forelesninger_2021/Clustering/
     → Everyones_a_little_bit_racist_sometimes.jpg";
     # Ximq = load(myimage);
     # Load image from the Internet:
     #imageadress = "https://cdn.images.express.co.uk/img/dynamic/151/590x/secondary/
      \rightarrow spacex-launch-why-starman-tesla-roadster-david-bowie-falcon-heavy-1225205.
     #imageadress = "http://pressarchive.theoldglobe.org/_img/pressphotos/
      →pre2008%20photos/aveQ5.jpg";
     #imageadress = "https://vgc.no/drfront/images/2018/02/12/c=1114,366,1920,1048;
     \rightarrow w=262; h=143;384858.jpq";
     #imageadress = "https://www.dagbladet.no/images/73342156.jpg?
      \rightarrow imageId=73342156\Imx=15.602322206096\Imy=10.807860262009\Imcropw=72.
      \hookrightarrow 060957910015 @croph=61.764705882353 @width=912 @height=521 @compression=80";
     imageadress = "http://lynski.no/wp-content/uploads/2016/09/
      →Anders_H%C3%B8st_OSM_Gautefall_seier.jpg"
     #imageadress = "https://upload.wikimedia.org/wikipedia/en/7/7d/
     →Lenna_%28test_image%29.png";
     #imageadress = "http://www.johnloomis.org/ece563/notes/basics/components/
      →mandrill/Mandrill.jpg";
                                        # download("http://pressarchive.theoldglobe.
      →org/_img/pressphotos/pre2008%20photos/aveQ5.jpg", "myimage.png"); Ximg =
     → load("myimage.png");
     myimage = download(imageadress); # Needs package ImageMagick
     Ximg = load(myimage);
     original = plot(Ximg, title = string("The original image to be compressed by ⊔
      \rightarrowk-means color clustering"), size = (1000,420))
     display(original)
```

The original image to be compressed by k-means color clustering



```
[4]: # Find image size and prepare to reshape the pixel-data for clustering
n,m = size(Ximg);
nm = n*m
```

[4]: 497025

2.1.1 We make a three-column matrix X where each row is an RGB-vector of a pixel position in the original image

```
[12]: mat = channelview(Ximg); # Convert from image format to 3 x n x m (0-1).

X = float( reshape( permutedims(mat, (2,3,1)), (nm, 3) )); # Channels last,

→vectorize image dims, convert to float
```

2.1.2 Now we can cluster the RGB-vectors of the pixels in X

```
[13]: include("mykmeans.jl")
      ## Cluster RGB-pixel values (in X) into k color clusters by the k-means_\sqcup
       \hookrightarrow algorithm
      k = 16; # The number of clusters
      Otime begin
      Cid, Ccenters, J, cs = mykmeans(X,k); # This will take some time...
                                                           a vector of cluster labels for
                                              # Cid:
       \hookrightarrow the rows in X.
                                              # Ccenters: the resulting k cluster centers.
                                              # J:
                                                           the clustering objective
       → function values
                                                           the cluster sizes (number of
                                              # cs:
       → members in each cluster)
```

```
#Cs = uint8(Ccenters); # Convert cluster centers into uint8-format
end
```

7.703864 seconds (588.67 k allocations: 4.807 GiB, 8.79% gc time)

[13]: ([16; 16; ...; 4; 4], Float32[0.4045972 0.3435841 0.32962012; 0.29181877 0.24158487 0.22087784; ...; 0.33571938 0.4275605 0.109497234; 0.6044407 0.5926064 0.6641352], Any[0.019008199030849958, 0.011868108709077407, 0.010860277039459403, 0.010529063215637012, 0.010386475345327977, 0.010305557434096224, 0.01025533710710975, 0.01022521472291614, 0.010205729021143868, 0.010189910339176631 ... 0.010027353521130307, 0.010007958393216729, 0.009991486094020048, 0.009975415860585209, 0.009959413093148971, 0.009944156954977072, 0.00992869443267328, 0.009913173777409632, 0.009902186340214156, 0.009892212254359484], [44394.0; 42369.0; ...; 9591.0; 18082.0])

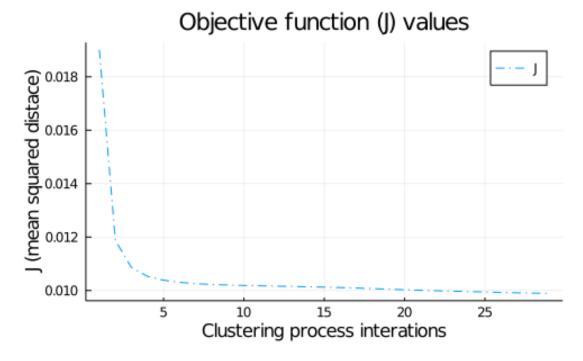
[14]: # Plotting the objective function values reflecting the clustering process

Jplot = plot(J, linestyle = :dashdot, title = "Objective function (J) values",

ylabel = "J (mean squared distace)", xlabel = "Clustering process

→interations", label = "J", size = (500, 300))

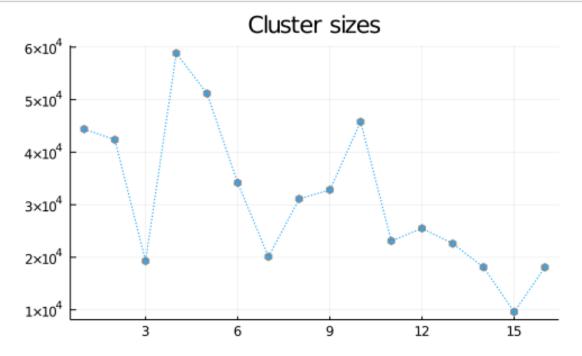
display(Jplot)



```
[15]: # Plotting the cluster sizes:
csplot = plot(cs, line = (:dot, 1), marker = ([:hex :d], 3, 0.8, Plots.

→stroke(3, :gray)), title = "Cluster sizes", label = "", size = (500, 300))
```

display(csplot)



2.1.3 Reshape and display cluster labels as associated image

```
[16]: cl = reshape(Cid,(n,m)); # cl is an image (n x m - matrix) viewing the cluster

⇒labels

#print(string("Image view of the cluster labels for ", k, " clusters"))

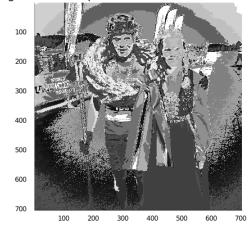
#Gray.(cl/k)

labelplot = plot(Gray.(cl/k), title = string("Image view of the pixelwise

⇒cluster labels for ", k, " clusters"), size = (1000,420))

display(labelplot)
```

Image view of the pixelwise cluster labels for 16 clusters



2.1.4 Show the compressed image by inserting the (k) cluster center vectors to display the colors

```
[17]: # print(string("The compressed image based on k=", k, " color clusters"))

# Use cluster-IDs (Cid) as lookup in cluster centers (Ccenters), reshape,

→ permute and convert to RGB

# colorview(RGB, permutedims( reshape(Ccenters[Cid,:],(n, m, 3)), (3,1,2)))

cmpplot = plot(colorview(RGB, permutedims( reshape(Ccenters[Cid,:],(n, m, 3)),

→ (3,1,2))), title = string("The compressed image based on k=", k, " color

→ clusters"), size = (1000,420))

display(cmpplot)
```

The compressed image based on k=16 color clusters



2.1.5 The residual image (difference between original and compressed images based on the clustering)

```
[18]: #print(string("The residual image based on k=", k, " color clusters"))

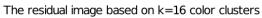
#Ximg - colorview(RGB, permutedims( reshape(Ccenters[Cid,:],(n, m, 3)),

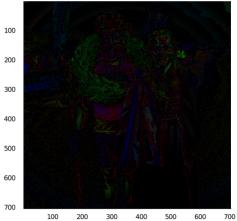
\( \times(3,1,2)))

resplot = plot(Ximg-colorview(RGB, permutedims( reshape(Ccenters[Cid,:],(n, m, \( \times \) \)

\( \times 3)), (3,1,2))), title = string("The residual image based on k=", k, " color \( \times \) \( \times \) clusters"), size = (1000,420))

display(resplot)
```





[]: