



# **Applied Mathematics and Statistics**

## **Project Report** **Color Compression**

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## I. Project description

### 1. Idea

- One image may contain many pixels. And one pixel contains 3 numbers R, G, B, which are in range [ 0, 255 ]. These 3 numbers would make up a color. So we have 256<sup>3</sup> colors, which make the file heavy.

- This project is for the purpose of reducing the number of colors on the image down to k\_cluster.

- We first pick k pixels in the image as centroids and then separate others to these clusters. And every loop, we update the new centroids and separate again. This process repeats until the centroids do not change or maximum iterations are reached.

- We finally have a new image with less colors.

### 2. Libraries and modules needed for project

- Numpy
- Module pyplot from matplotlib
- Module Image from PIL

### 3. Enviroment

- Jupyter Notebook

## II. Algorithms

**Step 1:** Use PIL to read image

**Step 2:** Reshape original image from 3 dimension to 2 dimension

**Step 3:** Kmean

- Random init K centroid:

- 'random' --> centroid has `c` channels, with `c` is initial random in [0,255]

- 'in\_pixels' --> centroid is a random pixels of original image

- Update centroids:

- Relabel each pixel with label of nearest centroid

- Find new value for centroid by calculate with mean of all pixel in that centroid (use np.linalg.norm)

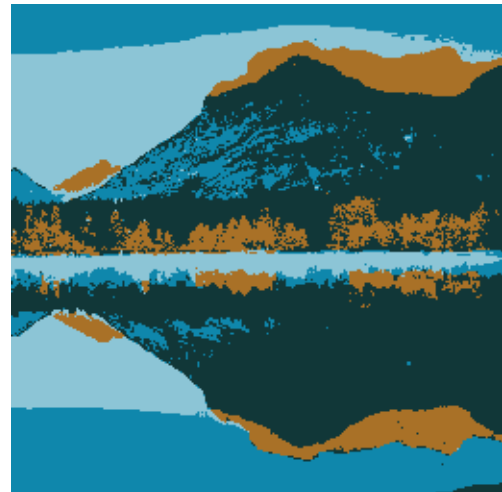
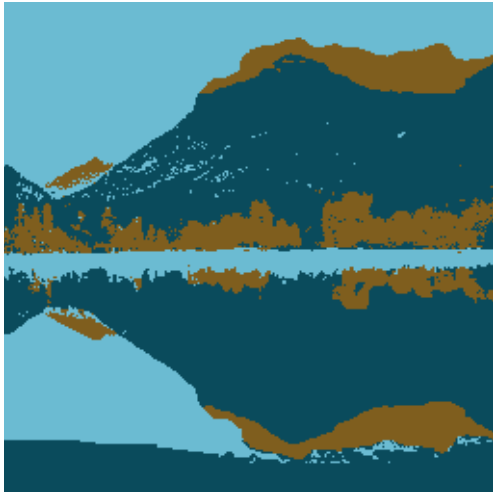
- Until reach max\_iteration

## III. Result

**Original image:**



**Result (in 3, 5, 7, 9 centroids):**



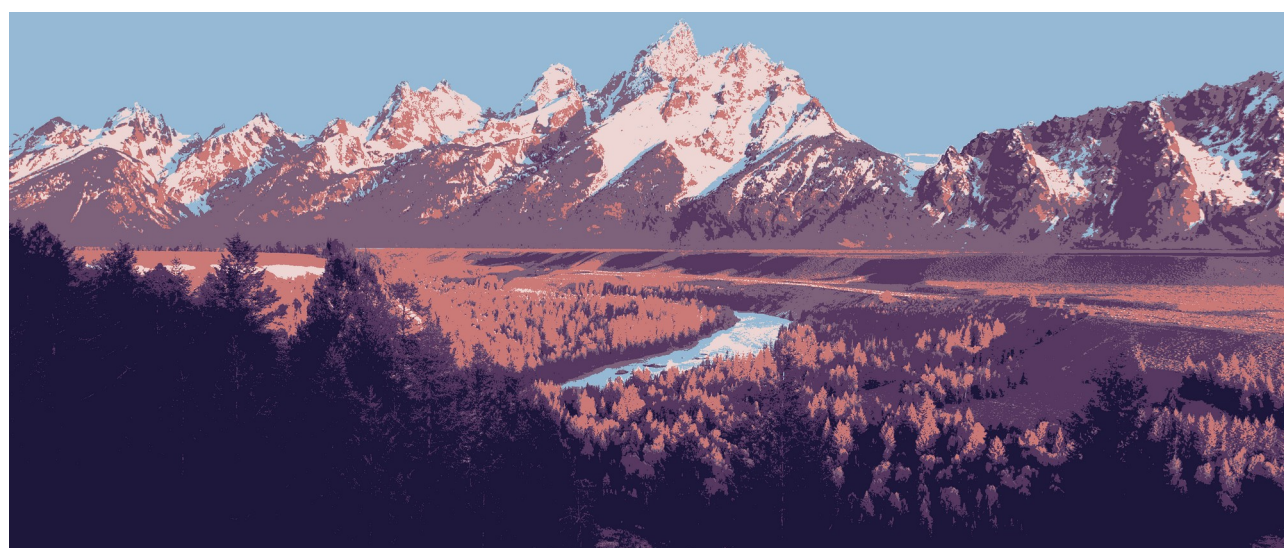
**Try with higher resolution image (2506x1080):**

**Original image:**





**Result (3, 5, 7 centroids):**



#### **IV. Comment**

- These picture are landscape, it natural usually have more color than other type of image
- In 3 centroids, we can only see the 'shape' of the picture, just near to black-white picture, hard to imagine how the real image look like
- When have more centroids (5, 7) or even 9 centroids, it very easy to guest how original image look like but also can save lot of memory to store the image. The higher k\_cluster is the more beautiful the image is
- Though being compressed, we still see the structure of the origin image.

#### **V. Reference**

<https://www.geeksforgeeks.org/python-pil-image-open-method/>

<https://ai538393399.wordpress.com/2020/09/29/k-means-clustering-algorithm-without-libraries>

<https://towardsdatascience.com/create-your-own-k-means-clustering-algorithm-in-python-d7d4c9077670>