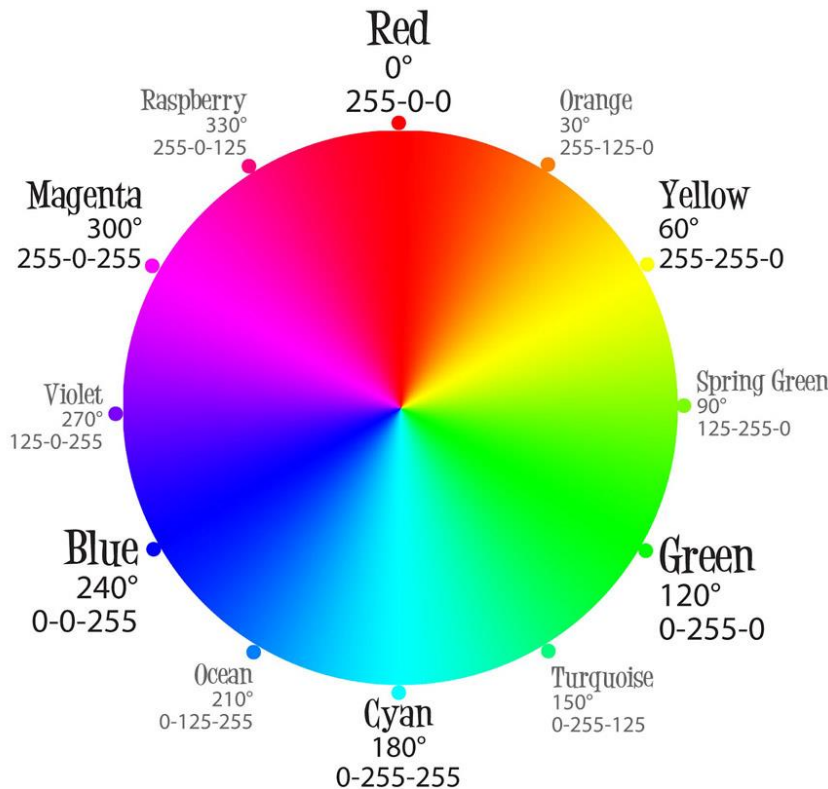




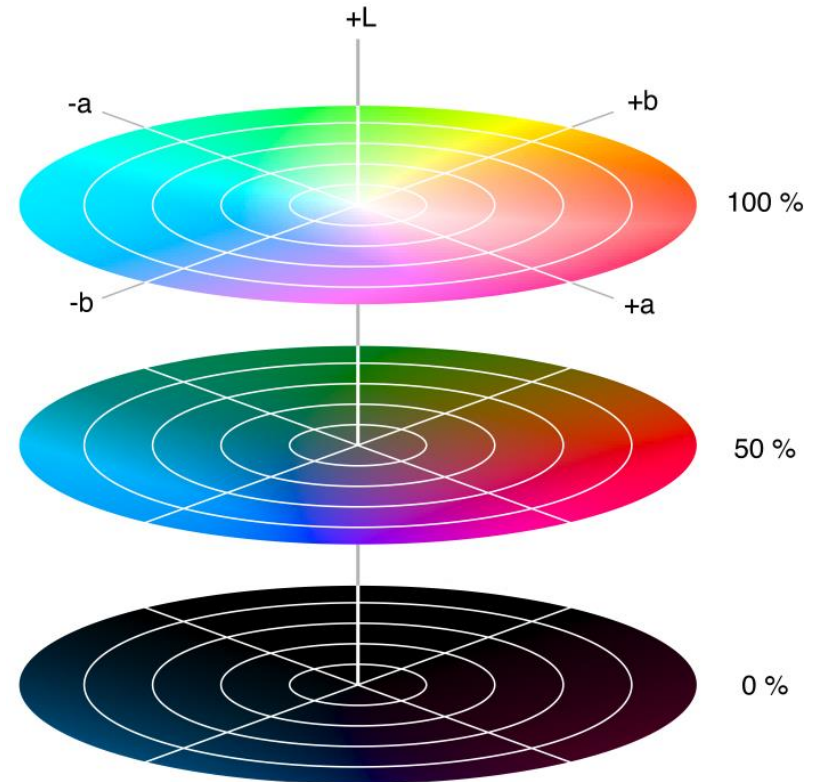
Cryo-Images Project

■ Lab Color Space

- $L^*a^*b^*$ color space was modeled after a color-opponent theory stating that two colors cannot be red and green at the same time or yellow and blue at the same time.
- As shown below, L^* indicates lightness, a^* is the red/green coordinate, and b^* is the yellow/blue coordinate. Deltas for L^* (ΔL^*), a^* (Δa^*) and b^* (Δb^*) may be positive (+) or negative (-). The total difference, Delta E (ΔE^*), however, is always positive.



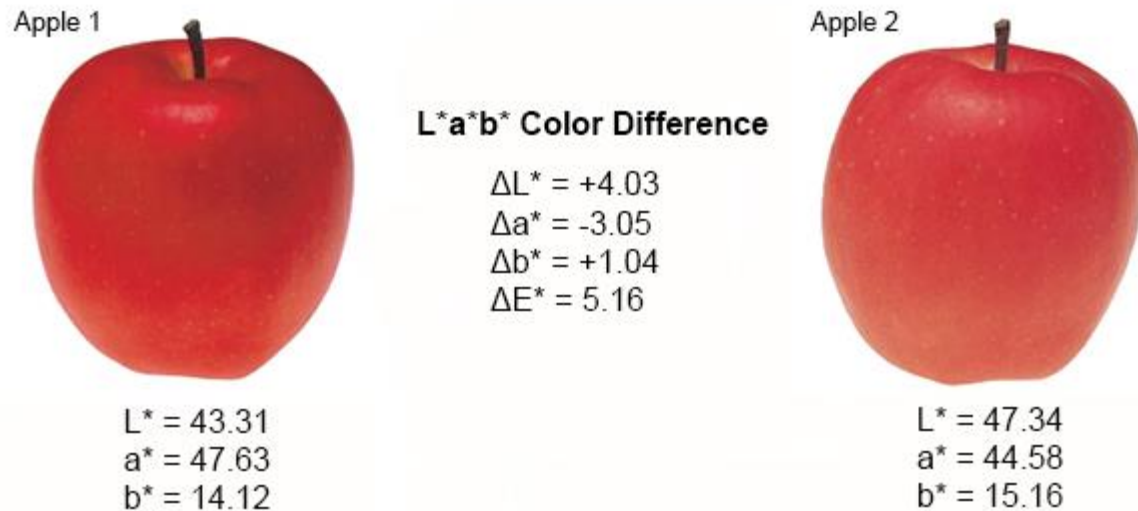
$L^*a^*b^*$ color space



■ RGB and Lab Color Space

- L*a*b* color space was modeled after a color-opponent theory stating that two colors cannot be red and green at the same time or yellow and blue at the same time.
- To determine the total color difference between all three coordinates, the following formula is used:

$$\Delta E^* = [\Delta L^{*2} + \Delta a^{*2} + \Delta b^{*2}]^{1/2}$$

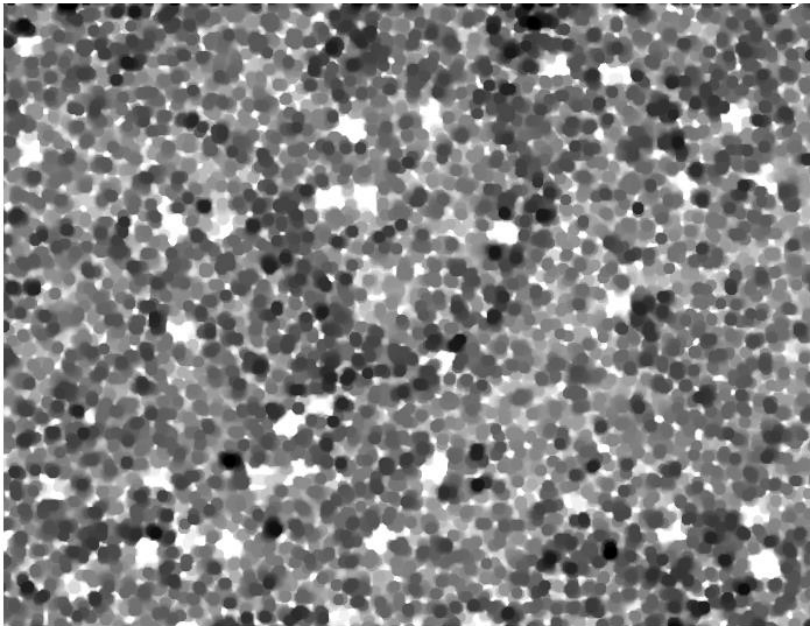


- Looking at the L*a*b* values for each apple in Figure 1, we can objectively determine that the apples don't match in color. These values tell us that Apple 2 (sample) is lighter, less red, and more yellow in color than Apple 1 (standard). If we put the values of $\Delta L^* = +4.03$, $\Delta a^* = -3.05$, and $\Delta b^* = +1.04$ into the color difference equation, it can be determined that the total color difference between the two apples is 5.16.

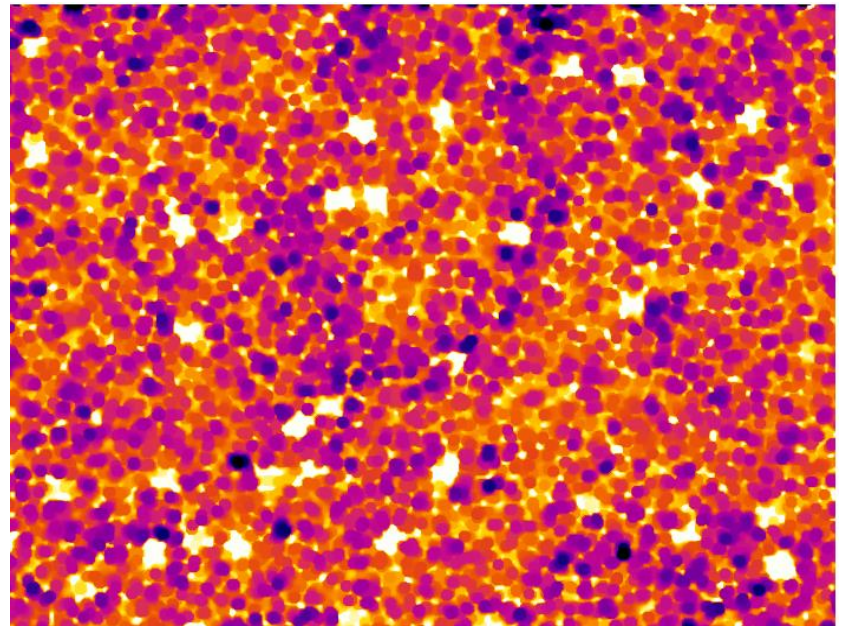
■ Colored (RGB) CryoEM Image

- To convert the CryoEM from grayscale to RGB color space then to the Lab space we used the following steps.
 - **Step1:** Load an Coloring Map
 - **Step2:** Build a **LOOK UP** Table based on the MIN and the MAX color value in the grayscale image to scale the gray image to the color map.
 - **Step3:** Generate the RGB image from the scaled map

Original Grayscale CryoEM Image

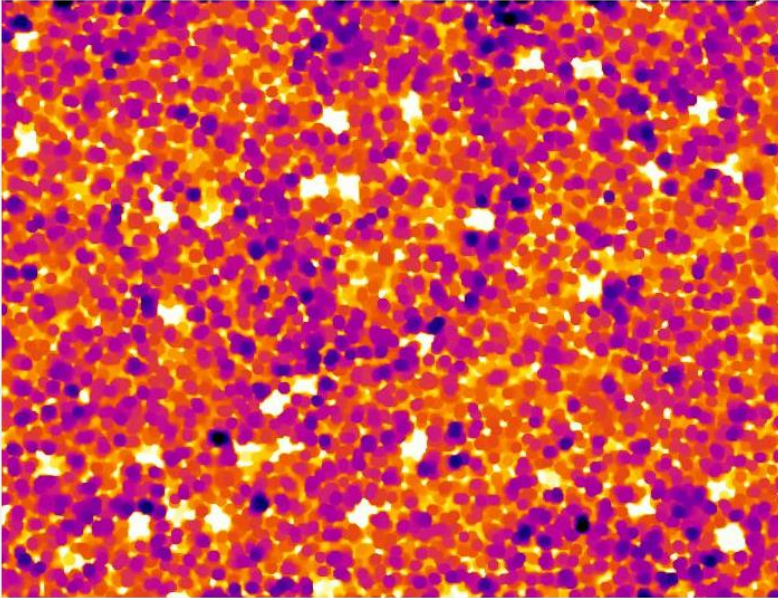


Colored CryoEM Image

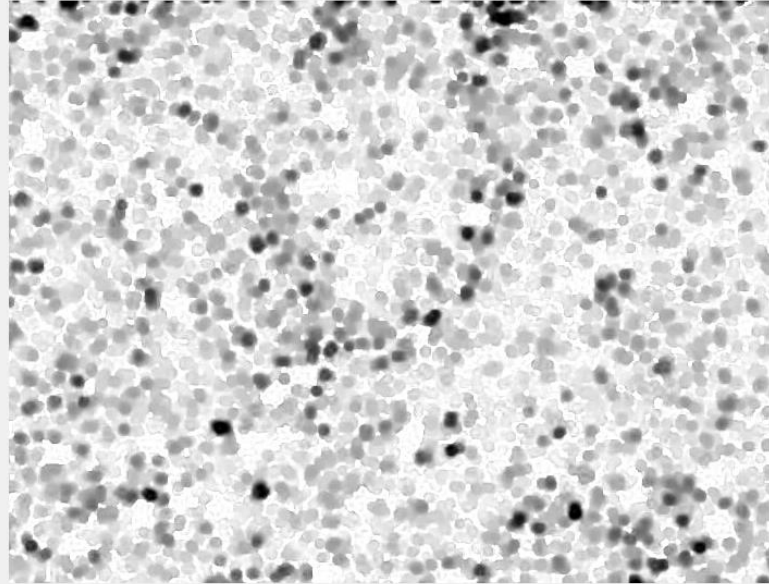


■ CryoEM Image in RGB Color Space

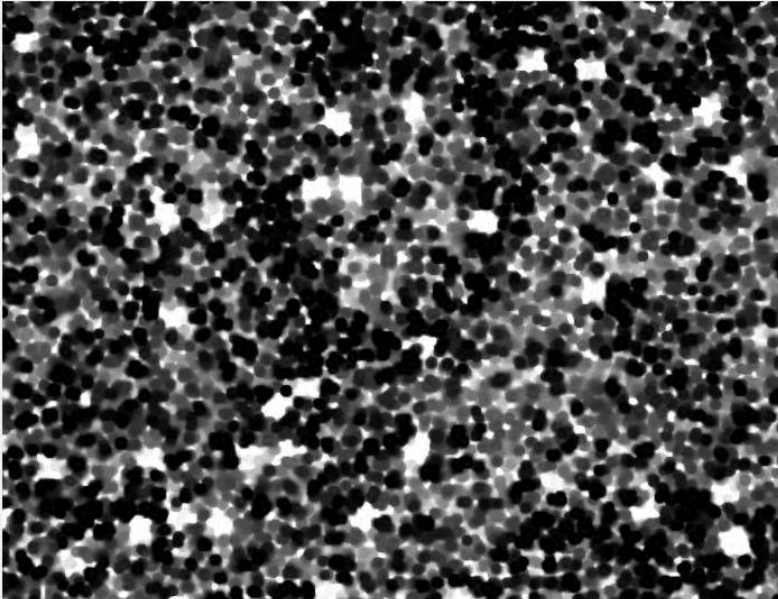
Original CryoEM RGB Color Space



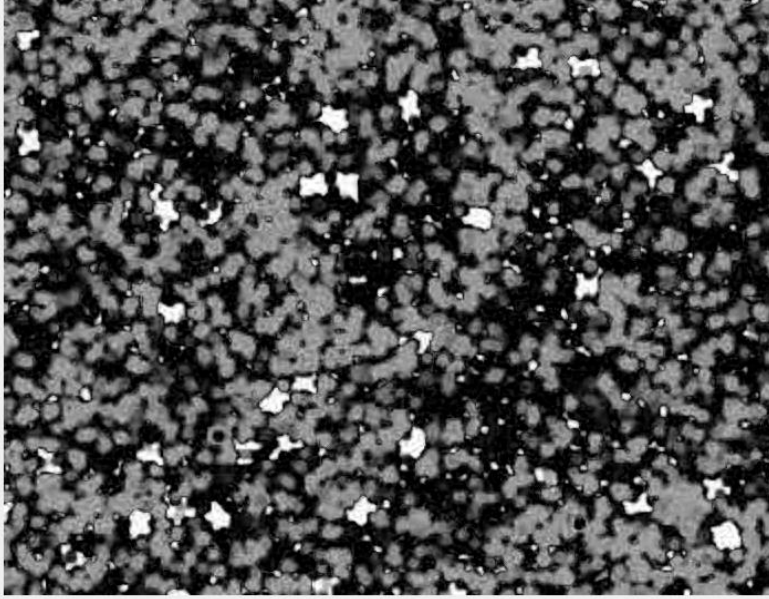
Red Color Space



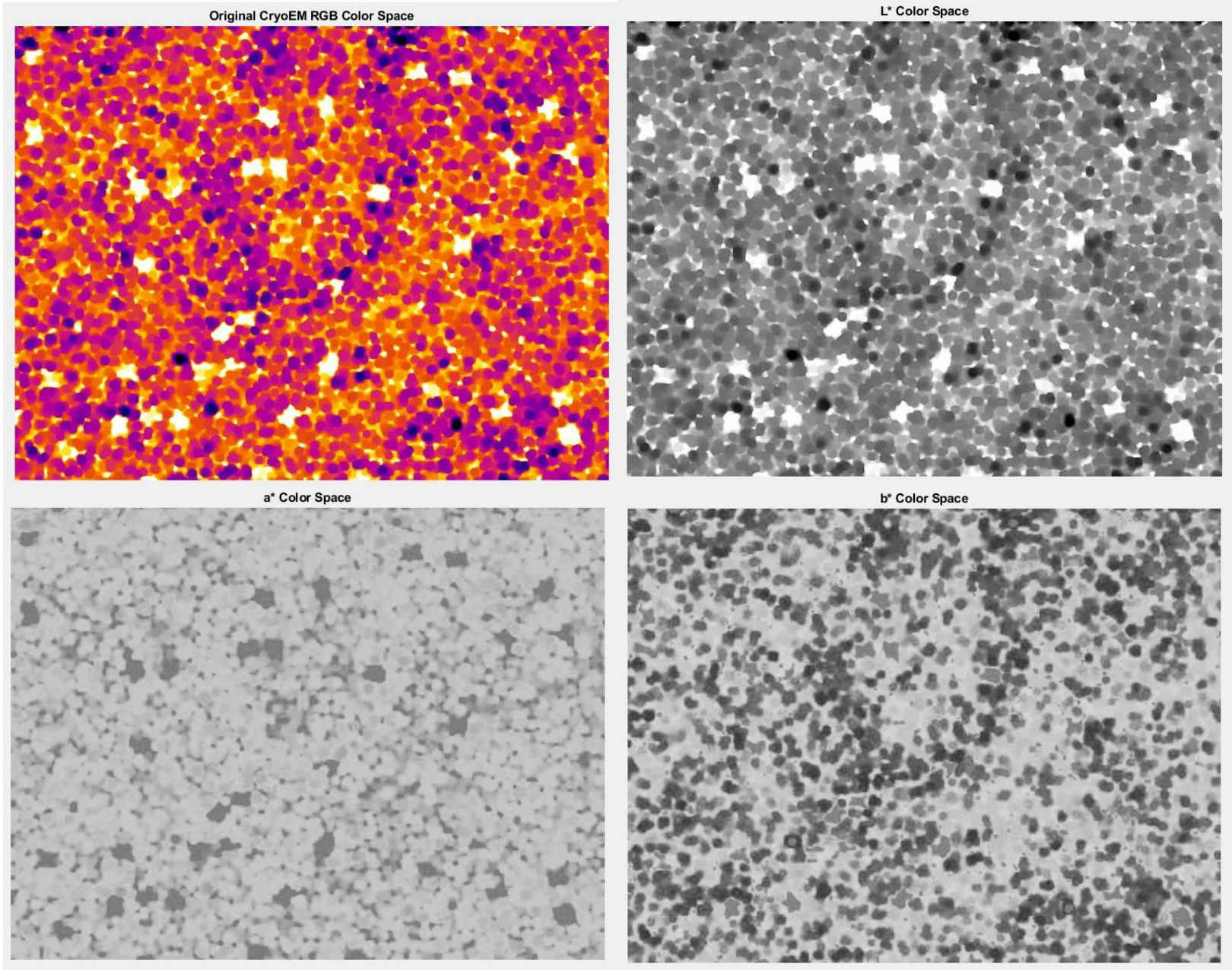
Green Color Space



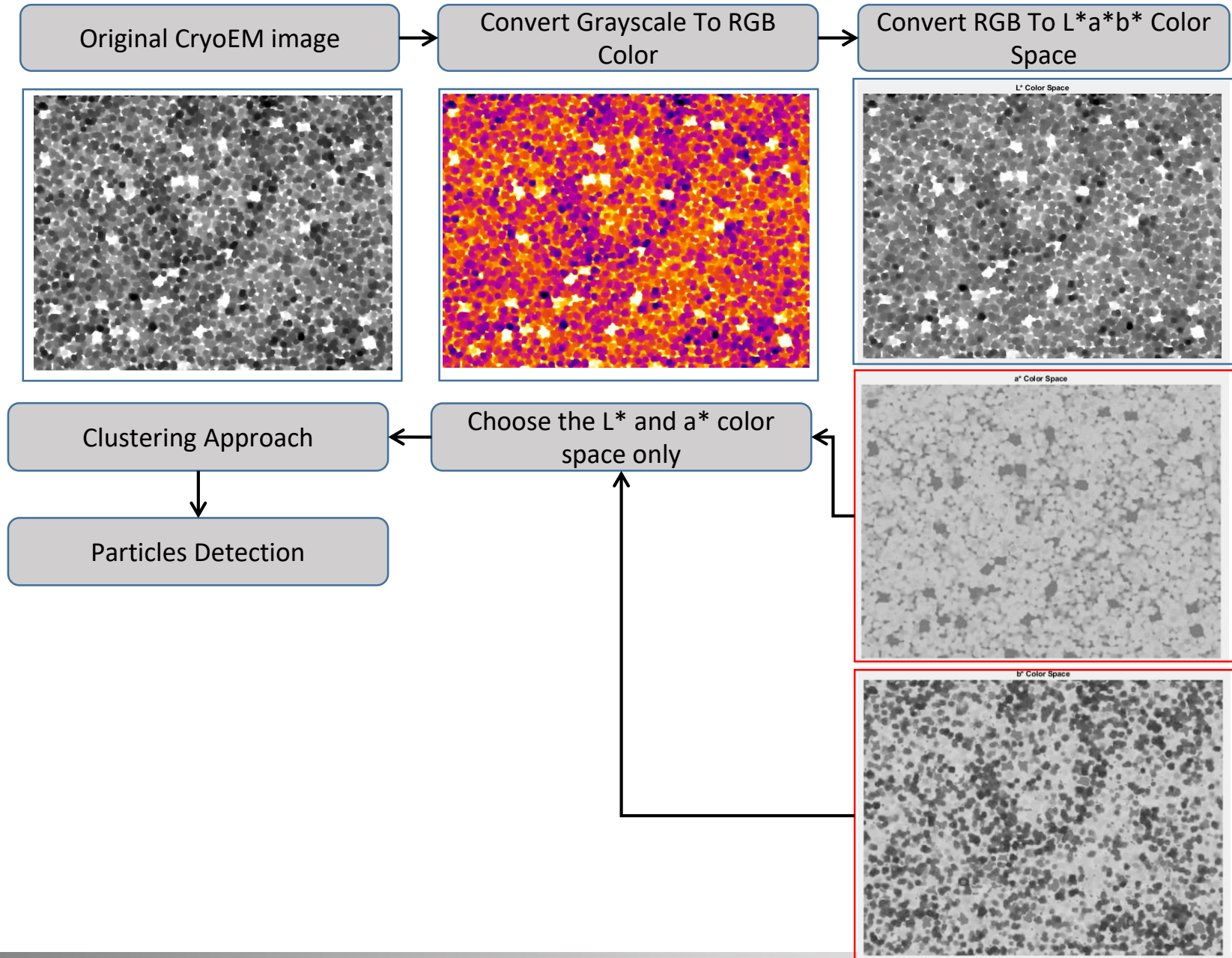
Blue Color Space



■ CryoEM Image in L*a*b* Color Space



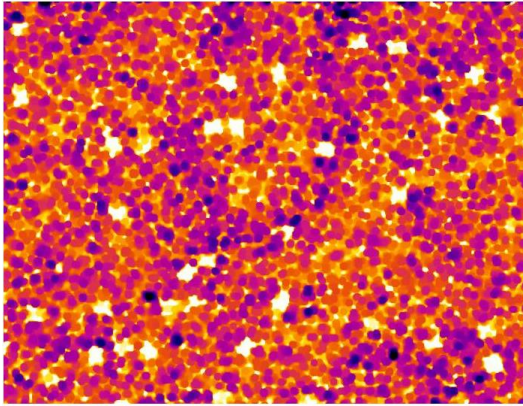
■ CryoEM Image Based L*a*b* Clustering Approach



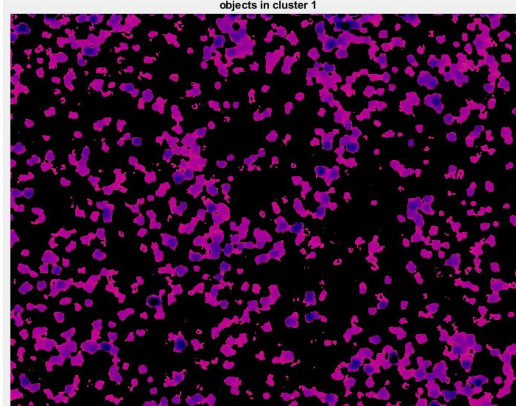
▪ Self-Organizing Map (SOM)

- The Self-Organizing Map (SOM) is one of the most popular neural network models. It belongs to the category of competitive learning networks.
- SOM is based on **unsupervised learning**, which means that no human intervention is needed during the learning and that little needs to be known about the characteristics of the input data.
- Basically, the **SOM** is used for clustering data without knowing the class memberships of the input data.
- The **SOM** algorithm provides a topology preserving mapping from the high dimensional space to map units.
 - The property of topology preserving means that the mapping preserves the **relative distance between** the points.
 - Points that **are near each other** in the input space are mapped to nearby map units in the SOM.
 - The SOM can thus serve as a cluster analyzing tool of high-dimensional data.
 - SOM has the capability to generalize that means the network can recognize or characterize inputs it has **never encountered** before.
 - A new input is **assimilated** with the map unit it is mapped to.
- The SOM clustering consist many stages such as:
 - **Initialization** : Choose random values for the initial weight vectors W_j .
 - **Sampling** : Draw a sample training input vector x from the input space.
 - **Matching** : Find the winning neuron $I(x)$ with weight vector closest to input vector.
 - **Updating** : Apply the weight update equation $\Delta W_{ji} = \eta(t)T_{j,I(x)}(t)(x_i - w_{ji})$
 - **Continuation** : keep returning to step 2 until the feature map stops changing.

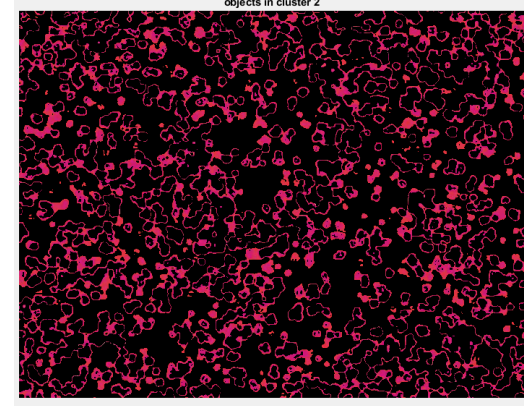
■ CryoEM Image Clustering using SOM



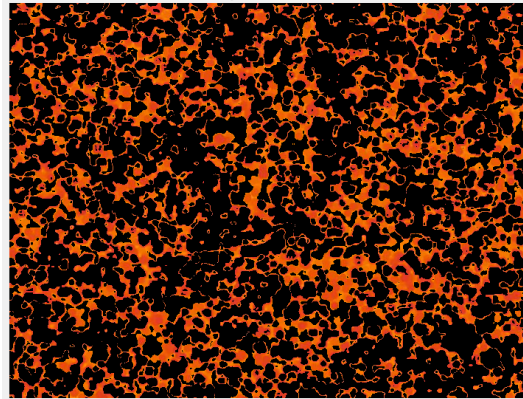
objects in cluster 3



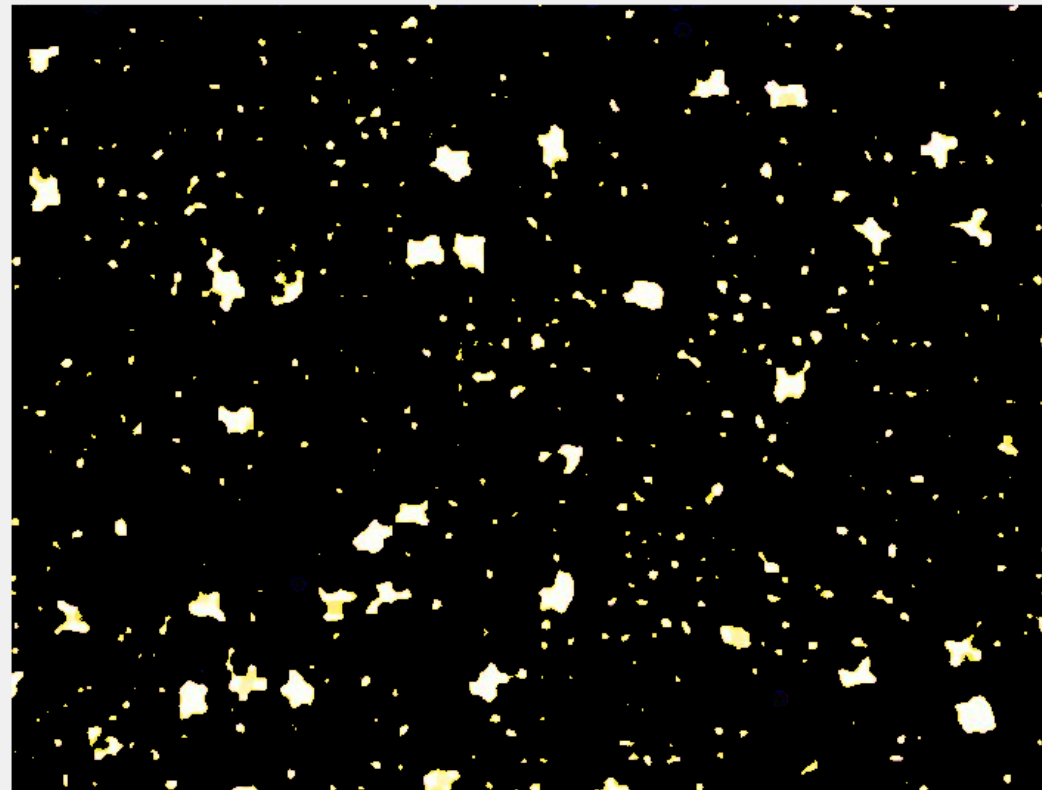
objects in cluster 1



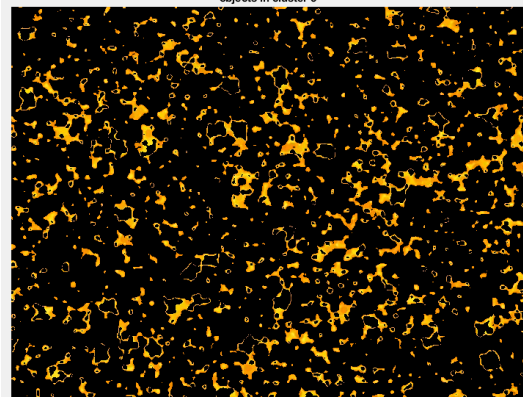
objects in cluster 2



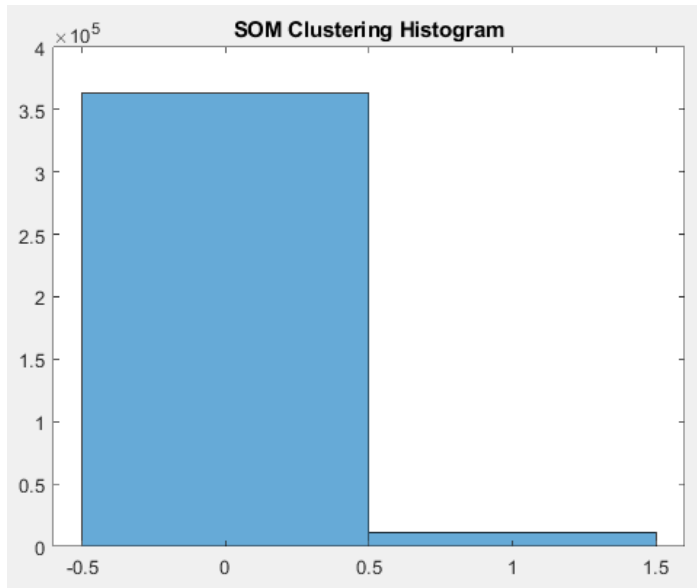
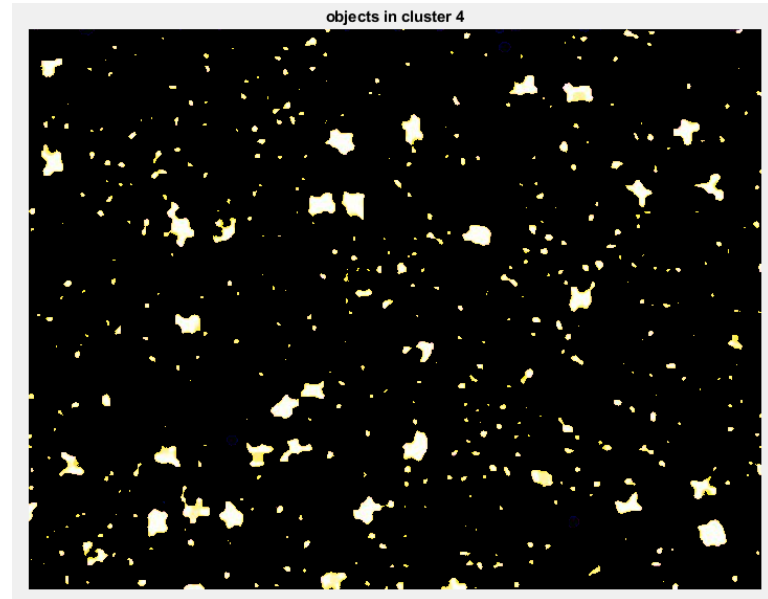
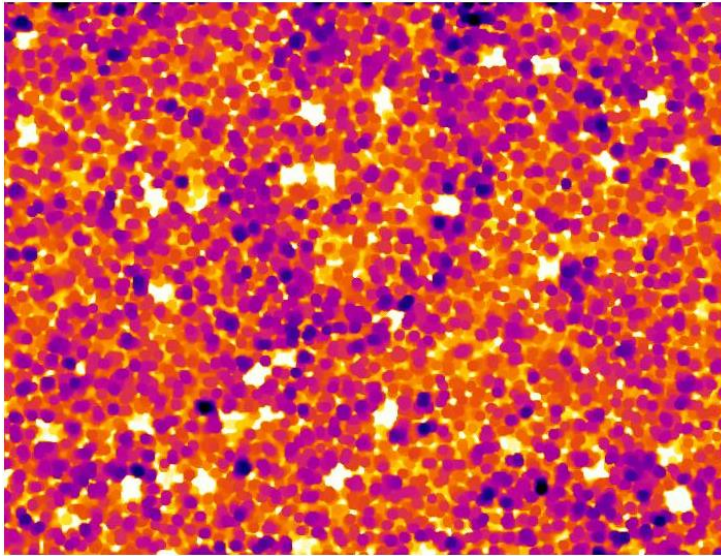
objects in cluster 5



objects in cluster 4

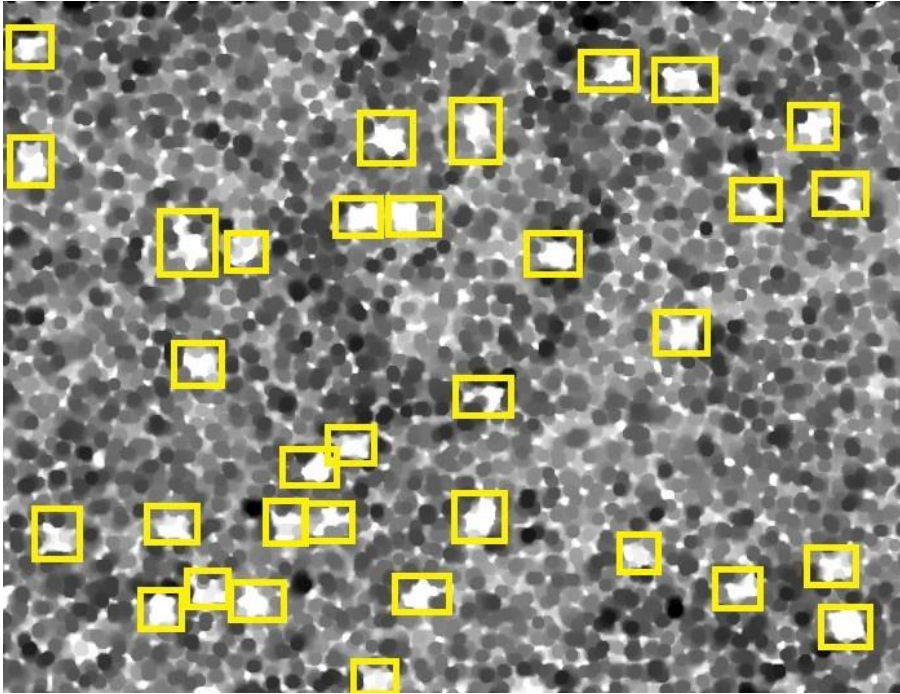


■ CryoEM Image Clustering using SOM

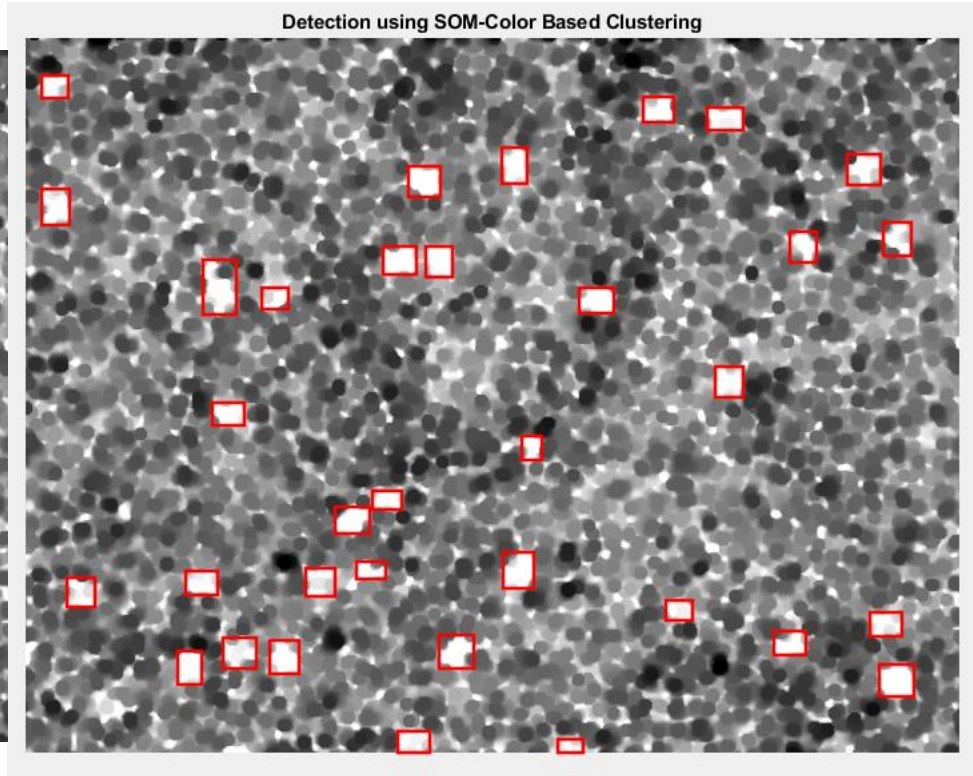


■ CryoEM Image Clustering using SOM

33 Particles as a GT

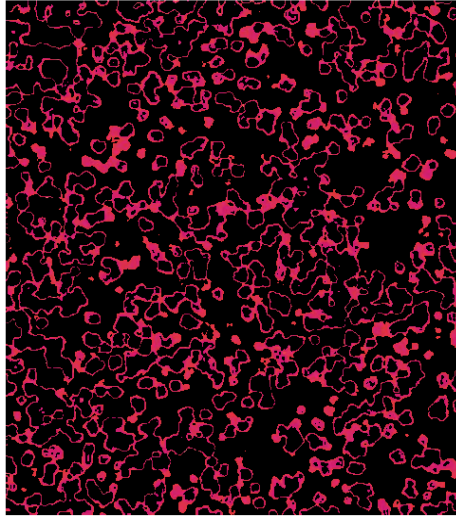


34 Particles Detected

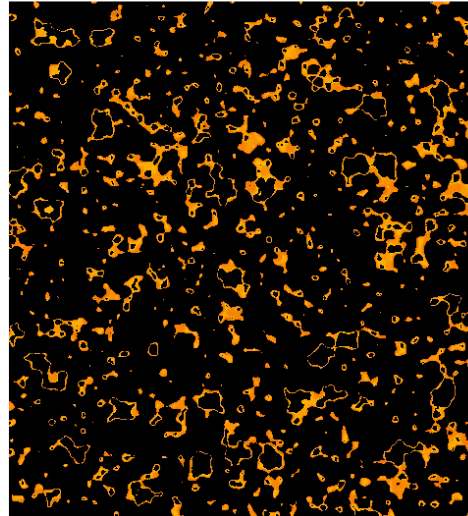


■ CryoEM Image Clustering using SOM

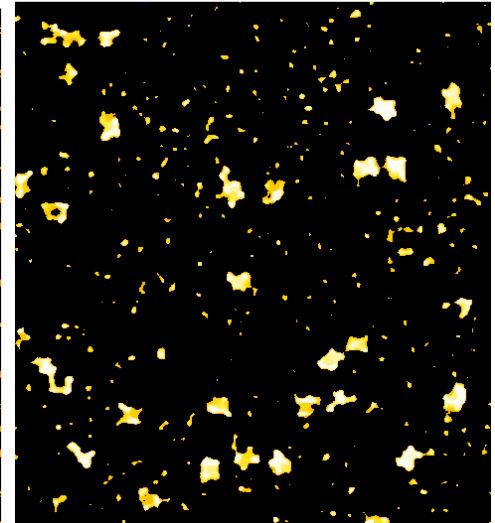
objects in cluster 1



objects in cluster 2



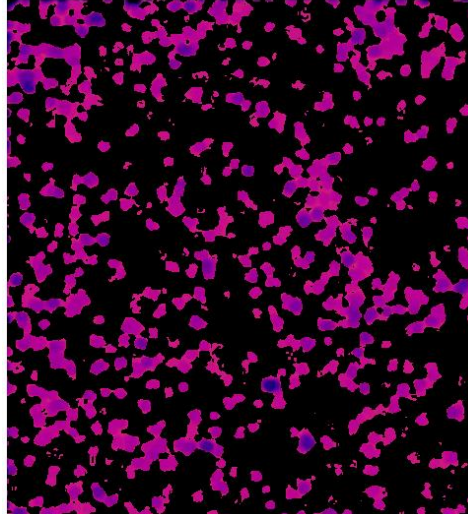
objects in cluster 3



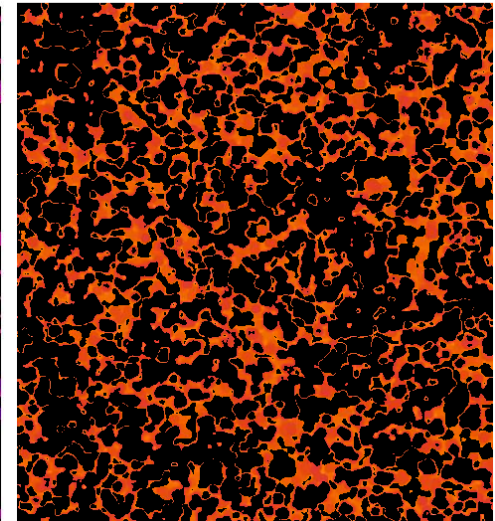
Clustered Image



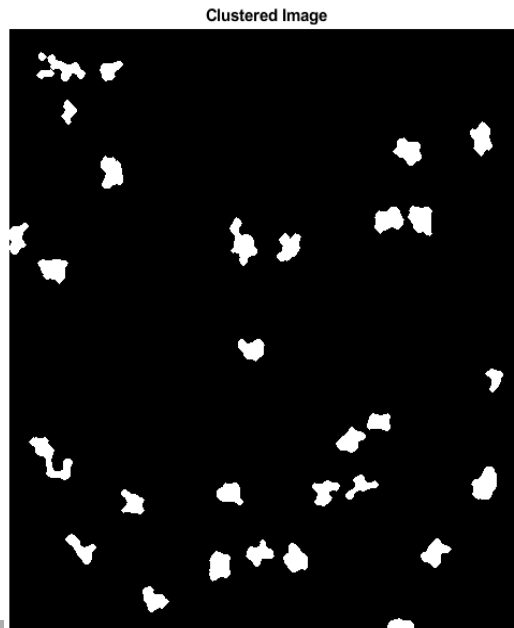
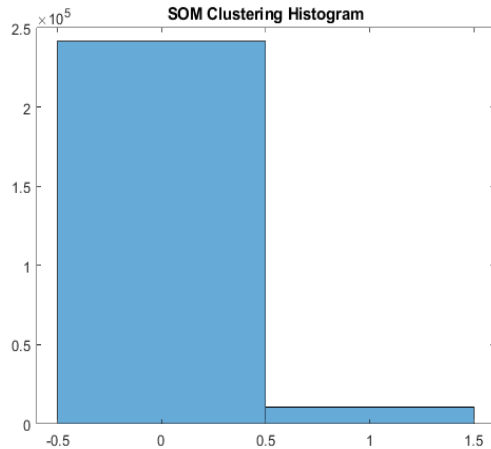
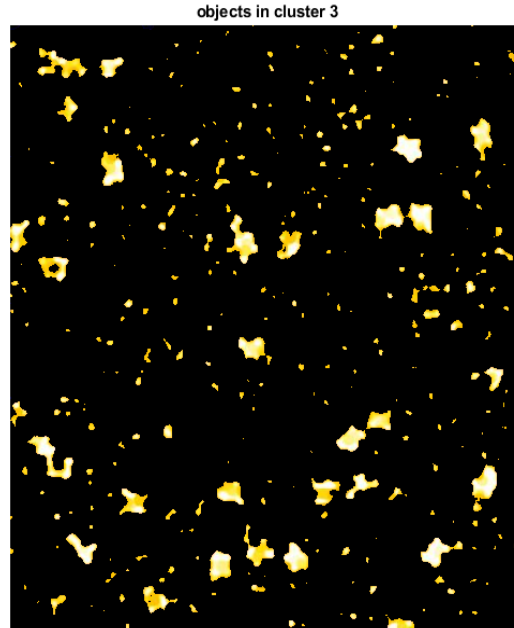
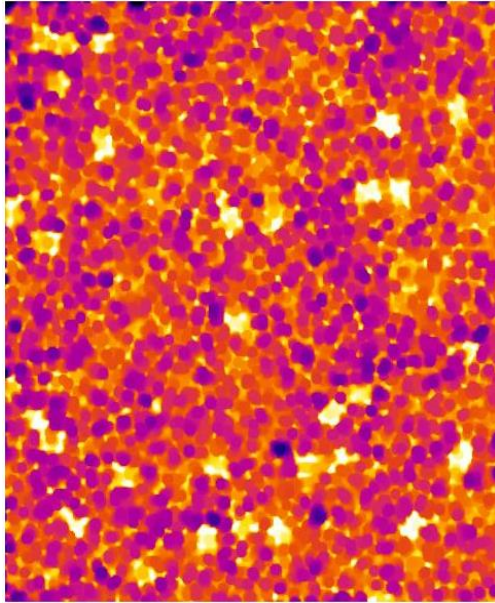
objects in cluster 5



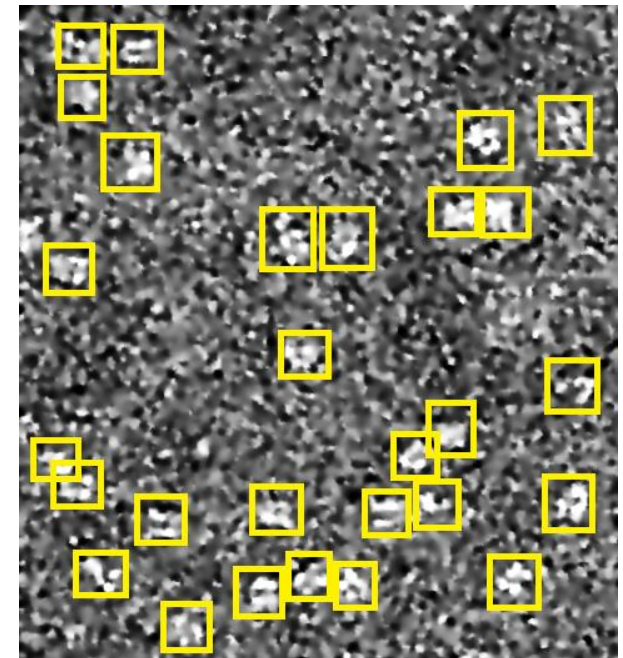
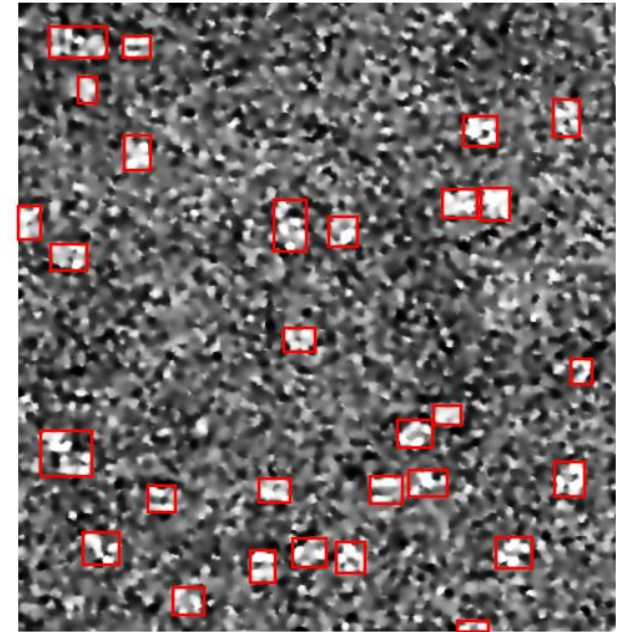
objects in cluster 4



■ CryoEM Image Clustering using SOM



Detection using SOM-Color Based Clustering



■ CryoEM Image Clustering using SOM

