

k-Means Clustering on Satellite Images

1. Initialize corresponding k color values for k clusters.
e.g. $color = [[0,255,255], [255,255,0]]$ for 2 clusters
2. Input satellite images, $imageInput$ (512, 512,4).
3. Initialize $clusteredImage$ as the output image.
4. Select k random x,y co-ordinates and extract pixel values from $imageInput$ to initialize $initCentroids$ ($k,4$), where each centroid has dimension (1X1X4).
5. $classData = clustering(k, centroids)$
$classData$ will contain pixel information and assigned class information.
Explanation of $clustering(k, centroids)$
 - a. $classData = classifyPixels(k, centroids)$
#use Euclidean distance to find distance between the pixel values of centroids and each pixel in $imageInput$; assign pixel to the centroid with minimum distance
 - b. $centroids_{up} = updateCentroids(classData)$
#use $classData$ to derive the mean of each clusters
 - c. if stopping criterion is satisfied:
 - i. return $classData$
 - d. else:
 - i. $clustering(k, centroids_{up})$
6. For each pixel in $classData$,
 - a. $clusteredImage[classData \rightarrow pixel] = color[classData \rightarrow pixel \rightarrow class]$
7. Display $clusteredImage$

Stopping criteria:

1. If $centroids$ and $centroids_{up}$ are same
2. Euclidean distance between the pixel values of $centroids$ and $centroids_{up}$ is less than a threshold value
3. Number of updates crosses a threshold value

Evaluation of clustering - silhouette coefficient

1. Initialize silhouette array s (512,512).
2. For each pixel p in *classData*, $s_p = \text{calSilhouette}(p)$
Explanation of $\text{calSilhouette}(p)$
 - a. calculate the mean distance of the point from all the points in the same cluster.
For example, if p is assigned cluster 1, then $a_p = d_1 = \text{mean}(\text{dist}(\text{imageInput}_p, \text{imageInput}_q))$, where q represent the points assigned to cluster 1.
Remember the distance is to be calculated based on pixel values, not positions.
 - b. calculate the means of the distances from clusters other than the assigned cluster, i.e. $d_m = \{d_2, d_3, d_4\}$.
 - c. $b_p = \min(d_m)$
 - d. $s_p = (b_p - a_p) / \max(b_p, a_p)$
3. $\text{silhouette_coeff} = \max(s)$

Assignment 10

1. Implement k-means clustering on satellite images for $k=2,3,4,5$. Check what changes if we use k-medoids instead of k-means.
2. Plot silhouette coefficients for $k=2,3,4,5$

Outputs expected:

1. Clustered images with $k=2,3,4,5$.
2. Silhouette coefficient comparison plot for $k=2,3,4,5$.

Note: implement user defined functions for Euclidean distance.