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- Assignment 1
- Computer Vision
- 9051pai2023

Q1. Write a Python program that takes any one of the RGB image from the given database as input, converts it to

a quantized image by using k-Means image quantization at varying 'k' values i.e. 1, 2, 4, 8, 16. Also, calculates the Mean Squared Error (MSE). Your program should include the following steps:

Read an RGB image as input.(DONE)

used Opencv

• Implement a function to perform k-Means image quantization.

Color quantization is critical for displaying images with many colors on devices that can only display a limited number of colors, usually due to memory limitations, and enables efficient compression of certain types of images. Color quantization, 2023.

- Convert the input RGB image to a quantized image using the k-Means algorithm by varying 'k'.
- Calculate the Mean Squared Error (MSE) between the original RGB image and quantized image.
- Analyze and discuss the effectiveness and limitations of k-Means image quantization in the context of the calculated MSE.

Defaulting to user installation because normal site-packages is not writeable

Requirement already satisfied: scikit-image in /Users/christhaliyath/Lib rary/Python/3.9/lib/python/site-packages (0.22.0)

Requirement already satisfied: numpy>=1.22 in /Users/christhaliyath/Libr ary/Python/3.9/lib/python/site-packages (from scikit-image) (1.24.1) Requirement already satisfied: scipy>=1.8 in /Users/christhaliyath/Libra ry/Python/3.9/lib/python/site-packages (from scikit-image) (1.12.0) Requirement already satisfied: networkx>=2.8 in /Users/christhaliyath/Library/Python/3.9/lib/python/site-packages (from scikit-image) (3.2.1) Requirement already satisfied: pillow>=9.0.1 in /Users/christhaliyath/Library/Python/3.9/lib/python/site-packages (from scikit-image) (9.4.0) Requirement already satisfied: imageio>=2.27 in /Users/christhaliyath/Library/Python/3.9/lib/python/site-packages (from scikit-image) (2.34.0) Requirement already satisfied: tifffile>=2022.8.12 in /Users/christhaliyath/Library/Python/3.9/lib/python/site-packages (from scikit-image) (2024.2.12)

Requirement already satisfied: packaging>=21 in /Users/christhaliyath/Li brary/Python/3.9/lib/python/site-packages (from scikit-image) (22.0) Requirement already satisfied: lazy_loader>=0.3 in /Users/christhaliyath /Library/Python/3.9/lib/python/site-packages (from scikit-image) (0.3) Defaulting to user installation because normal site-packages is not writ eable

Requirement already satisfied: scikit-learn in /Users/christhaliyath/Lib rary/Python/3.9/lib/python/site-packages (1.4.1.post1)

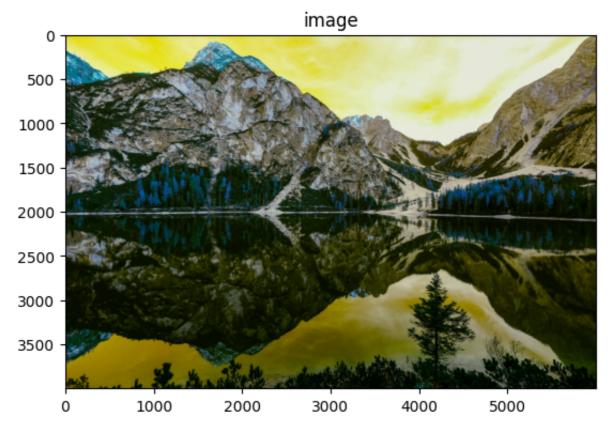
Requirement already satisfied: numpy<2.0,>=1.19.5 in /Users/christhaliya th/Library/Python/3.9/lib/python/site-packages (from scikit-learn) (1.24.1)

Requirement already satisfied: scipy>=1.6.0 in /Users/christhaliyath/Lib rary/Python/3.9/lib/python/site-packages (from scikit-learn) (1.12.0) Requirement already satisfied: joblib>=1.2.0 in /Users/christhaliyath/Library/Python/3.9/lib/python/site-packages (from scikit-learn) (1.3.2) Requirement already satisfied: threadpoolctl>=2.0.0 in /Users/christhaliyath/Library/Python/3.9/lib/python/site-packages (from scikit-learn) (3.4.0)

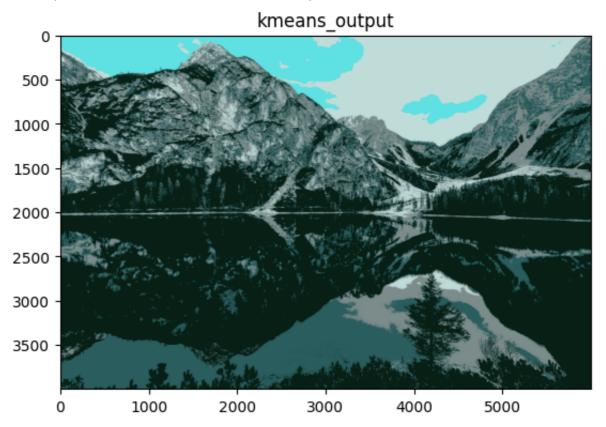
In []: import cv2

```
from cv2 import cvtColor , COLOR_BGR2RGB , TERM_CRITERIA_MAX_ITER , KMEAN
        import numpy as np
        from skimage import measure
        from skimage.metrics import structural similarity as ssim
        from matplotlib import pyplot as plt
        from sklearn.cluster import MiniBatchKMeans
        import numpy as np
        from numpy import unique , float32 , uint8
        import argparse
        import cv2
        #from enum import unique
        #Read an RGB image as input.(DONE)
        def read_image(path_to_image):
            image=cv2.imread(path_to_image)
            return image
        def display_image(title,image):
            #cv2.imshow("input_image",image)
            plt.imshow(image)
            plt.title(title)
            plt.show()
In [ ]: def mse(imageA, imageB):
            # the 'Mean Squared Error' between the two images is the
            # sum of the squared difference between the two images;
            # NOTE: the two images must have the same dimension
            err = np.sum((imageA.astype("float") - imageB.astype("float")) ** 2)
            err /= float(imageA.shape[0] * imageA.shape[1])
            # return the MSE, the lower the error, the more "similar"
            # the two images are
        def diff remove bg(img0, img, img1):
            d1 = diff(img0, img)
            d2 = diff(img, img1)
            return cv2.bitwise_and(d1, d2)
        #Implement a function to perform k-Means image quantization.
        def k_means_image_quantization(img):
            # Convert it from BGR to RGB
            img_RGB = cvtColor(img, COLOR_BGR2RGB)
            # Reshape image to an Mx3 array
            img data = img RGB.reshape(-1, 3)
            # Find the number of unique RGB values
```

```
print(len(unique(img_data, axis=0)), 'unique RGB values out of', img_
   # Specify the algorithm's termination criteria
    criteria = (TERM_CRITERIA_MAX_ITER + TERM_CRITERIA_EPS, 10, 1.0)
   # Run the k-means clustering algorithm on the pixel values
    compactness, labels, centers = kmeans(data=img_data.astype(float32),
   # Apply the RGB values of the cluster centers to all pixel labels
    colours = centers[labels].reshape(-1, 3)
   # Find the number of unique RGB values
    print(len(unique(colours, axis=0)), 'unique RGB values out of', img_d
   # Reshape array to the original image shape
    img_colours = colours.reshape(img_RGB.shape)
   # Display the quantized image
   #imshow(img colours.astype(uint8))
   display_image("kmeans_output",img_colours.astype(uint8))
   #show()
    return img_colours
#Convert the input RGB image to a quantized image using the k-Means algor
def vary_k_in_k_means(image):
   pass
image=read_image("Assignment_1_database/Q1/RGB image 2.jpg")
display_image("image",image)
k_means_image=k_means_image_quantization(image)
```



377518 unique RGB values out of 24000000 pixels 5 unique RGB values out of 24000000 pixels

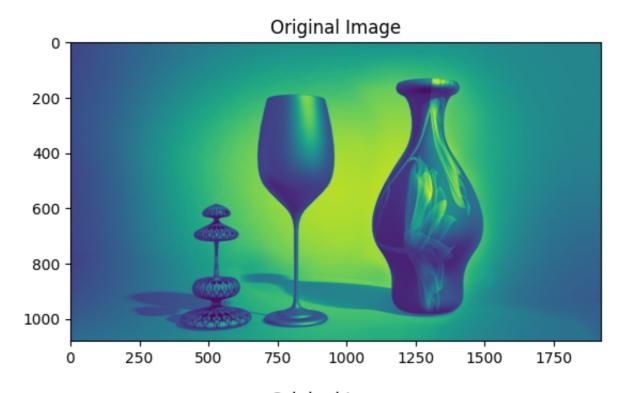


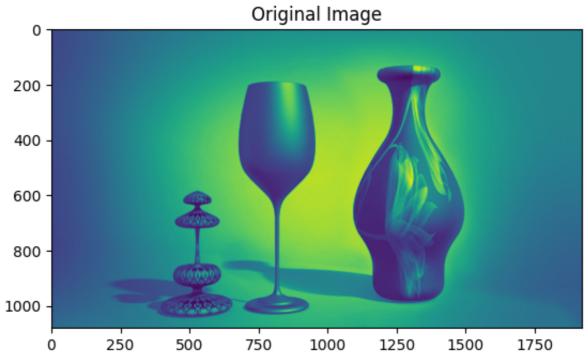
```
In []: #Analyze and discuss the effectiveness and limitations of k-Means image q
        m = mse(image, k means image)
        #s = ssim(image, k_means_image)
        #s = measure.compare_ssim(image, k_means_image)
        s = measure.metrics.structural_similarity(image, k_means_image)
        #s=measure.
        print ("mse: %s, ssim: %s" % (m, 0))
                                                   Traceback (most recent call la
        AttributeError
        st)
        /Users/christhaliyath/MTECH/IIT jammu mtech artificial intelligence mach
        ine_learning/SECOND_SEM/Computer Vision (Electives)
                                                                Dr. Harkeerat Ka
        ur/assignment/chris sunny thaliyath assignment 1.ipynb Cell 8 line 5
              <a href='vscode-notebook-cell:/Users/christhaliyath/MTECH/IIT_jamm</pre>
        u_mtech_artificial_intelligence_machine_learning/SECOND_SEM/Computer%20V
        ision%20%28Electives%29%09Dr.%20Harkeerat%20Kaur/assignment/chris_sunny_
        thaliyath assignment 1.ipynb#X10sZmlsZQ%3D%3D?line=1'>2</a> m = mse(imag
        e, k means image)
              <a href='vscode-notebook-cell:/Users/christhaliyath/MTECH/IIT_jamm</pre>
        u_mtech_artificial_intelligence_machine_learning/SECOND_SEM/Computer%20V
        ision%20%28Electives%29%09Dr.%20Harkeerat%20Kaur/assignment/chris_sunny_
        thaliyath_assignment_1.ipynb#X10sZmlsZQ%3D%3D?line=2'>3</a> #s = ssim(im
        age, k_means_image)
              <a href='vscode-notebook-cell:/Users/christhaliyath/MTECH/IIT jamm</pre>
        u mtech artificial intelligence machine learning/SECOND SEM/Computer%20V
        ision%20%28Electives%29%09Dr.%20Harkeerat%20Kaur/assignment/chris_sunny_
        thaliyath_assignment_1.ipynb#X10sZmlsZQ%3D%3D?line=3'>4</a> #s = measure
        .compare_ssim(image, k_means_image)
        ----> <a href='vscode-notebook-cell:/Users/christhaliyath/MTECH/IIT_jamm
        u mtech artificial intelligence machine learning/SECOND SEM/Computer%20V
        ision%20%28Electives%29%09Dr.%20Harkeerat%20Kaur/assignment/chris sunny
        thaliyath_assignment_1.ipynb#X10sZmlsZQ%3D%3D?line=4'>5</a> s = measure.
        metrics.structural_similarity(image, k_means_image)
              <a href='vscode-notebook-cell:/Users/christhaliyath/MTECH/IIT_jamm</pre>
        u_mtech_artificial_intelligence_machine_learning/SECOND_SEM/Computer%20V
        ision%20%28Electives%29%09Dr.%20Harkeerat%20Kaur/assignment/chris_sunny_
        thaliyath assignment 1.ipynb#X10sZmlsZQ%3D%3D?line=5'>6</a> #s=measure.
              <a href='vscode-notebook-cell:/Users/christhaliyath/MTECH/IIT jamm</pre>
        u_mtech_artificial_intelligence_machine_learning/SECOND_SEM/Computer%20V
        ision%20%28Electives%29%09Dr.%20Harkeerat%20Kaur/assignment/chris_sunny_
        thaliyath_assignment_1.ipynb#X10sZmlsZQ%3D%3D?line=6'>7</a> print ("mse:
        %s, ssim: %s" % (m, 0))
        File ~/Library/Python/3.9/lib/python/site-packages/lazy_loader/__init__.
        py:89, in attach.<locals>. getattr (name)
             87
                    return attr
             88 else:
                    raise AttributeError(f"No {package_name} attribute {name}")
```

AttributeError: No skimage.measure attribute metrics

Q2. Write a code to find the connected components present in an image as shown in the given example. Utilize the provided image in the database 'Connected component' to identify the components, emphasizing accuracy.

```
In []: import cv2
        import cv2
        import numpy as np
        # Load the image (replace 'connected components.jpg' with your image file
        image = cv2.imread("Assignment 1 database/Q2/Connected component.jpg", 0)
        # Binarize the image
        _, binary = cv2.threshold(image, 127, 255, cv2.THRESH_BINARY)
        # Find connected components
        num_labels, labels, stats, centroids = cv2.connectedComponentsWithStats(b)
        # Create an output image to draw the labeled components
        output = np.zeros_like(binary, dtype=np.uint8)
        # Iterate through the connected components
        for label in range(1, num labels):
            # Get the bounding box coordinates for the current component
            x, y, w, h, area = stats[label]
            # Draw a rectangle around the connected component
            cv2.rectangle(output, (x, y), (x + w, y + h), 255, 2)
            # Add the label number to the center of the component
            text = str(label)
            cv2.putText(output, text, (x + w // 2, y + h // 2), cv2.FONT_HERSHEY_
        # Display the original image and the output with labeled components
        #cv2.imshow('Original Image', image)
        display_image('Original Image', image)
        #cv2.imshow('Connected Components', output)
        display_image('Original Image', image)
        def find_connected_components(image):
          Finds connected components in a binary image.
          Args:
              image: The binary image as a NumPy array.
          Returns:
              labeled_image: The image with connected components labeled.
              num_components: The number of connected components.
          # Apply connected component labeling using OpenCV (adapt function call
          ret, labeled
```





Q3. From the attached database of standard test images, select the gray scale 'lena_gray_512.tif' image of dimension pixels. Perform the following operations:

- (a) Reduce the image size by down sampling to 256x256, name this image as 'reduce_lena_256.tif'.
- (b) Compare the reduced image 'reduced_lena_256.tif' with the given image in the database 'lena_gray_256.tif' by finding the PSNR values between them. Also plot the image showing MSE difference for the same.
- (c) Now again resize the computed image 'reduced_lena_256.tif' to original dimension pixel using these methods:
- Nearest Neighbour Interpolation
- · Bilinear Interpolation
- Bicubic Interpolation

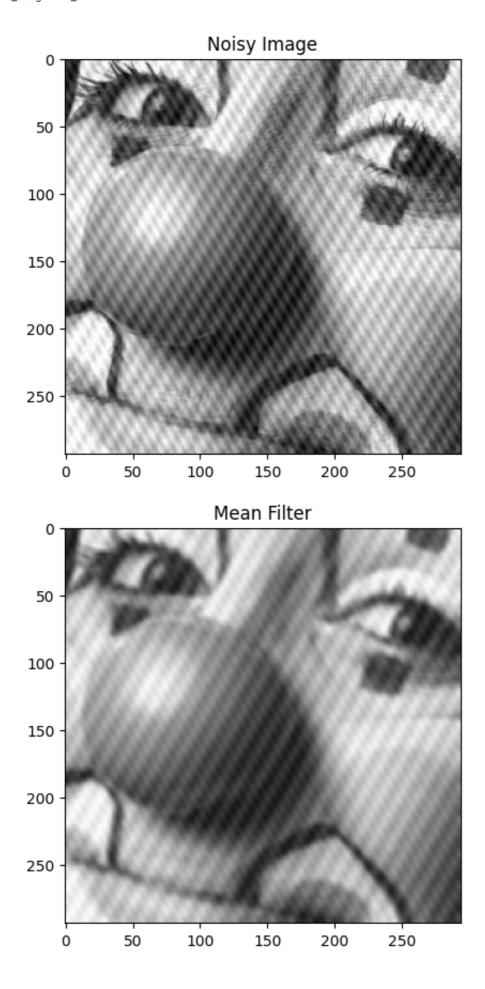
```
In []: import cv2
        import numpy as np
        def down_sample(image, new_size=(256, 256)):
          Downsamples a grayscale image using averaging.
          Args:
              image: The original grayscale image as a NumPy array.
              new size: The desired output image size (default: (256, 256)).
          Returns:
              down_sampled_image: The downsampled grayscale image.
          h, w = image.shape
          # Ensure new size is within original image dimensions
          new_h, new_w = min(new_size[0], h), min(new_size[1], w)
          down_sampled_image = np.zeros((new_h, new_w), dtype=image.dtype)
          # Average pixel values in blocks for downsampling
          for y in range(0, h, 2):
            for x in range(0, w, 2):
              block = image[y:y+2, x:x+2] # Extract 2x2 block
              down_sampled_image[y // 2, x // 2] = np.mean(block) # Average valu
```

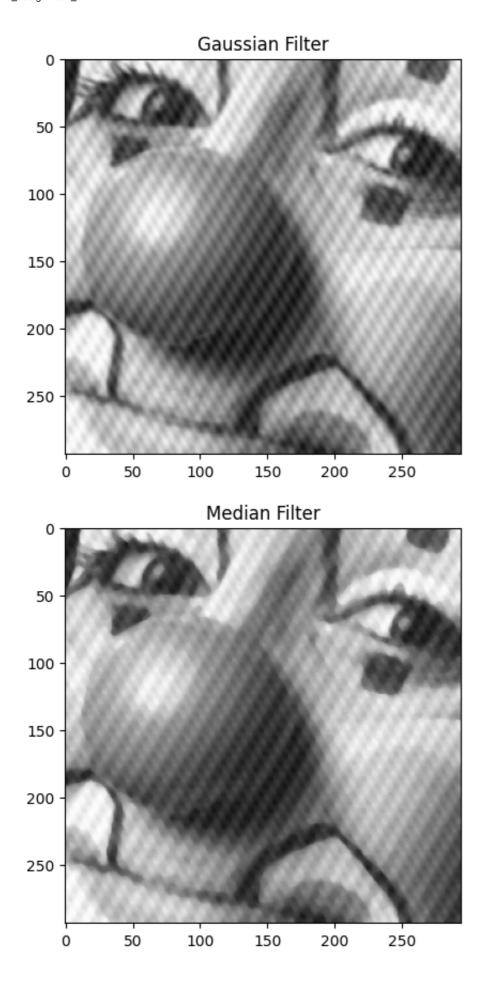
```
return down sampled image
def calculate_psnr(image1, image2):
 Calculates the Peak Signal-to-Noise Ratio (PSNR) between two images.
      image1: The first image as a NumPy array.
      image2: The second image as a NumPy array.
 Returns:
     psnr: The PSNR value in dB.
 mse = np.mean(np.square(image1 - image2))
 if mse == 0: # Avoid division by zero
    return float('inf')
 max_intensity = 255.0 # Assuming 8-bit images
 psnr = 10 * np.log10(max_intensity**2 / mse)
 return psnr
def plot_mse_difference(image1, image2, title="MSE Difference"):
 Calculates and plots the pixel-wise Mean Squared Error (MSE) difference
 Args:
     image1: The first image as a NumPy array.
      image2: The second image as a NumPy array.
      title: The title for the plot (default: "MSE Difference").
 .....
 mse diff = np.square(image1 - image2)
 display_image(title, mse_diff.astype(np.uint8)) # Convert to uint8 for
def up_sample(image, scale_factor=2, method="bilinear"):
 Upsamples a grayscale image using various interpolation methods.
 Args:
      image: The original grayscale image as a NumPy array.
      scale_factor: The upsampling factor (default: 2).
      method: The interpolation method (default: "bilinear", options: "ne
 Returns:
     up_sampled_image: The upsampled grayscale image.
 h, w = image.shape
 new_h, new_w = int(h * scale_factor), int(w * scale_factor)
 if method == "nearest":
    up_sampled_image = cv2.resize(image, (new_w, new_h), interpolation=cv
 elif method == "bilinear":
    up_sampled_image = cv2.resize(image, (new_w, new_h), interpolation=cv
 elif method == "bicubic":
    up_sampled_image = cv2.resize(image, (new_w, new_h), interpolation=cv
```

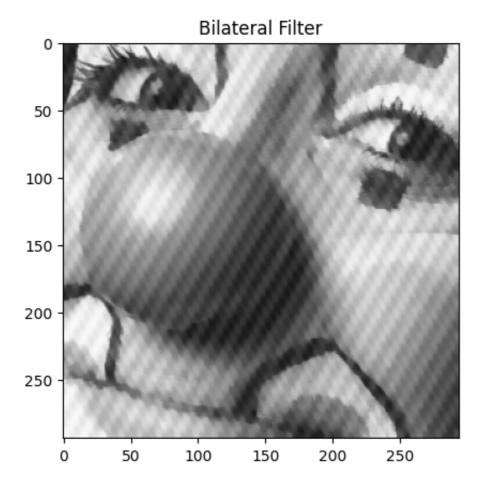
```
else:
    raise ValueError(f"Invalid interpolation method: {method}")
return up_
```

Q4. Develop codes in python to obtain the filtered imaged from the noisy image given to you as input 'Noisy image' in the database. Show the steps involved to be applied for filtering and explain why?

```
In []: import cv2
        import numpy as np
        import os
        def load images from folder(folder):
            images = []
            for filename in os.listdir(folder):
                img = cv2.imread(os.path.join(folder,filename))
                if img is not None:
                    images.append(img)
            return images
        def question_function():
            #images=load_images_from_folder("Assignment_1_database/Q4/Noisy_image
            images = [cv2.imread("Assignment_1_database/Q4/Noisy image.jpg")]
            for image in images:
                noisy_image = image
                # Apply different filtering techniques
                filtered_mean = cv2.blur(noisy_image, (5, 5))
                filtered_gaussian = cv2.GaussianBlur(noisy_image, (5, 5), 0)
                filtered_median = cv2.medianBlur(noisy_image, 5)
                filtered_bilateral = cv2.bilateralFilter(noisy_image, 9, 75, 75)
                # Display the images
                #cv2.imshow('Noisy Image', noisy_image)
                display_image('Noisy Image', image)
                #cv2.imshow('Mean Filter', filtered mean)
                display_image('Mean Filter', filtered_mean)
                #cv2.imshow('Gaussian Filter', filtered_gaussian)
                display_image('Gaussian Filter', filtered_gaussian)
                #cv2.imshow('Median Filter', filtered_median)
                display_image('Median Filter', filtered_median)
                #cv2.imshow('Bilateral Filter', filtered_bilateral)
                display_image('Bilateral Filter', filtered_bilateral)
            # cv2.waitKey(0)
            # cv2.destroyAllWindows()
        question_function()
```



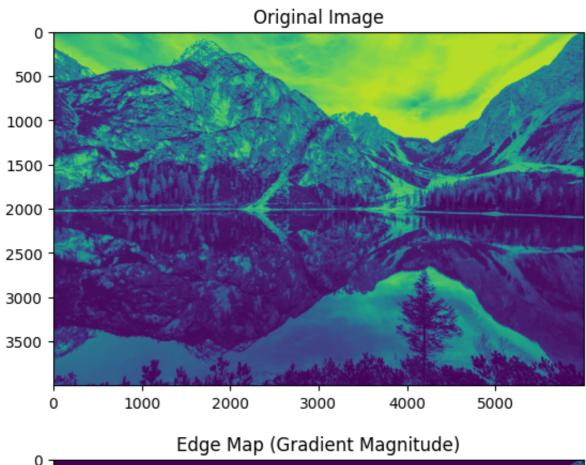


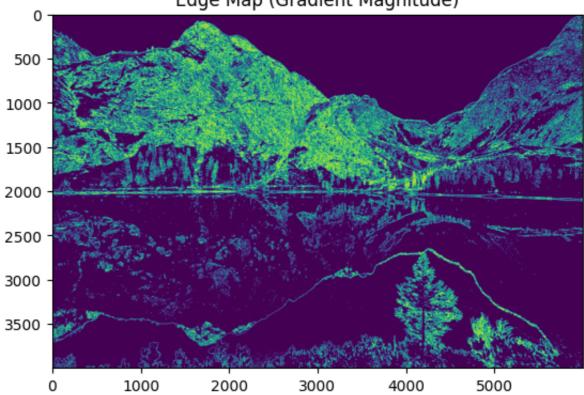


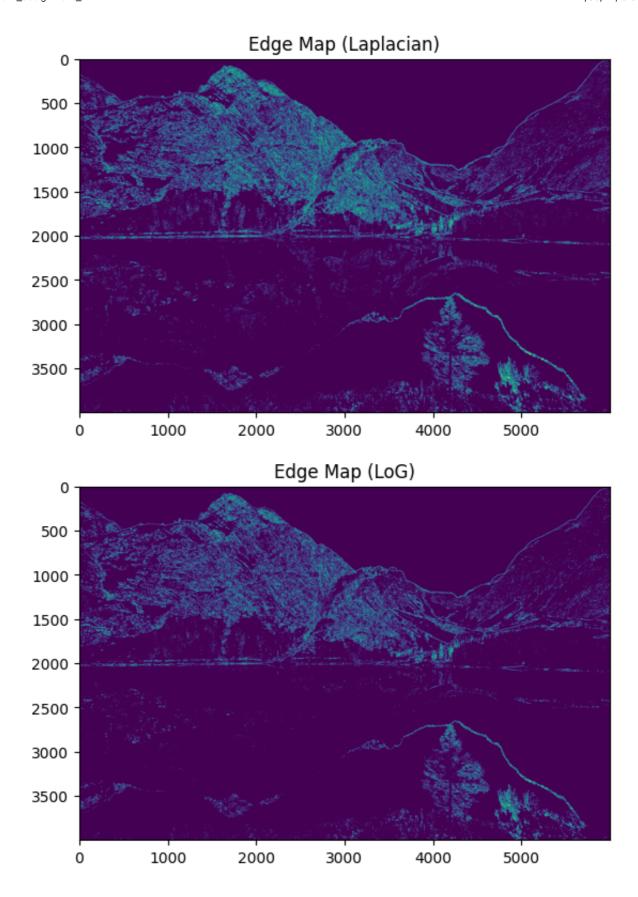
Q5. Take any one image from the given database and compute##:

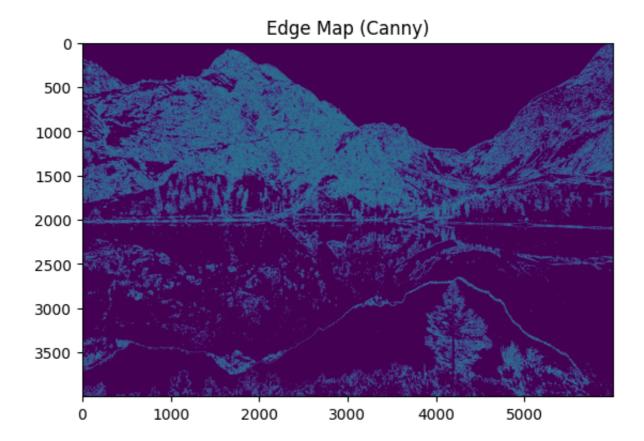
- First order derivative along 'x'
- First order derivative along 'y'
- Gradient image magnitude and followed by thresholding to output edge map.
- · Edge map computed using second order derivative
- Edge map using LoG and Canny edge detector

```
In []: import cv2
        import numpy as np
        # Load the image (replace 'image.jpg' with your image file)
        image = cv2.imread("Assignment_1_database/Q1/RGB image 2.jpg", 0) # Read
        # Compute first-order derivatives
        sobelx = cv2.Sobel(image, cv2.CV_64F, 1, 0, ksize=3)
        sobely = cv2.Sobel(image, cv2.CV_64F, 0, 1, ksize=3)
        # Compute gradient magnitude and threshold for edge map
        magnitude = np.sqrt(sobelx**2 + sobely**2)
        edge_map_magnitude = np.uint8(magnitude > 100) # Adjust the threshold va
        # Compute second-order derivative for edge map
        laplacian = cv2.Laplacian(image, cv2.CV_64F)
        edge_map_laplacian = np.uint8(np.abs(laplacian) > 50) # Adjust the thres
        # Compute LoG and Canny edge maps
        blurred = cv2.GaussianBlur(image, (3, 3), 0)
        log = cv2.Laplacian(blurred, cv2.CV_64F)
        edge_map_log = np.uint8(np.abs(log) > 30) # Adjust the threshold value a
        edge_map_canny = cv2.Canny(image, 100, 200) # Adjust the threshold value
        # Display the results
        display_image('Original Image', image)
        display_image('Edge Map (Gradient Magnitude)', edge_map_magnitude)
        display_image('Edge Map (Laplacian)', edge_map_laplacian)
        display_image('Edge Map (LoG)', edge_map_log)
        display_image('Edge Map (Canny)', edge_map_canny)
```









Q6. Apply the bit plane slicing of the following image size 3x3, for an 8-bit image "0" is encoded in 00000000 and "255" is encoded in 11111111

167 133 111

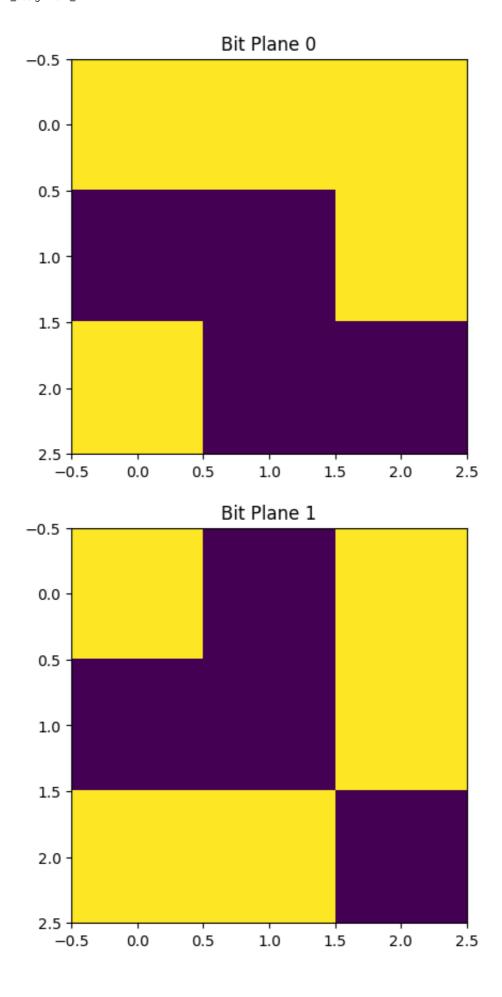
144 140 135

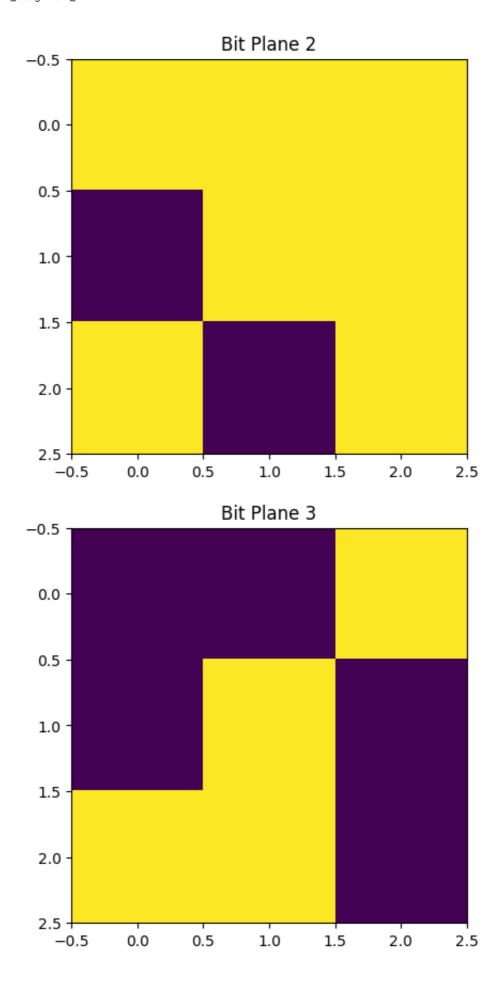
159 154 148

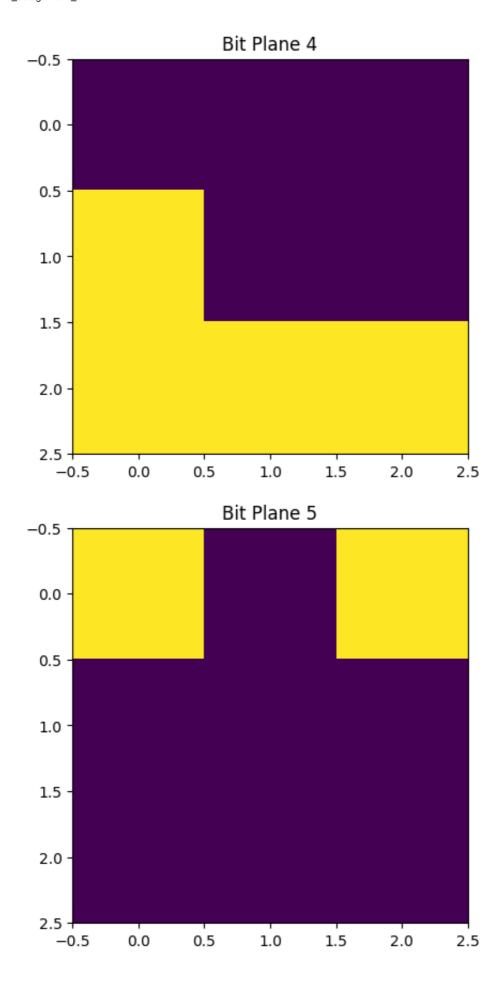
Implement the bit-plane slicing process using a programming language of your choice (e.g., Python with OpenCV).

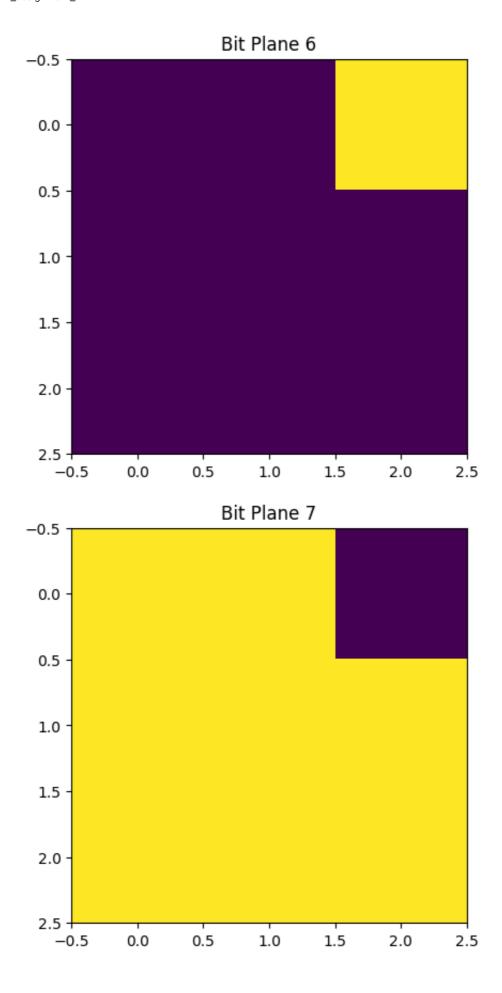
Provide the code snippet and explain each step in the implementation.

```
In [ ]: import cv2
        import numpy as np
        # Input image
        img = np.array([[167, 133, 111],
                         [144, 140, 135],
                         [159, 154, 148]], dtype=np.uint8)
        # Function to extract a single bit plane
        def get_bit_plane(img, bit_pos):
            # Create a mask to extract the desired bit plane
            mask = 1 << bit_pos</pre>
            # Apply the mask to extract the bit plane
            bit_plane = np.bitwise_and(img, mask)
            # Shift the extracted bit plane to the right-most position
            bit_plane = np.right_shift(bit_plane, bit_pos)
            return bit_plane
        # Extract and display each bit plane
        for bit_pos in range(8):
            bit_plane = get_bit_plane(img, bit_pos)
            display_image(f"Bit Plane {bit_pos}", bit_plane)
            #cv2.waitKey(0)
        cv2.destroyAllWindows()
```









REFERENCES

- https://gist.github.com/mstfldmr/45d6e47bb661800b982c39d30215bc88
- https://pyimagesearch.com/2014/07/07/color-quantization-opencv-using-k-means-clustering/
- https://gist.github.com/gonzalo123/df3e43477f8627ecd1494d138eae03ae
- https://stackoverflow.com/questions/55178229/importerror-cannot-import-name-structural-similarity-error

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