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Certificate This is to certify that Mr. Omkar Bharat Khandare Student of MCA Course, First year, Semester 2, Roll No MC2223044 has successfully complete the required number of practical in subject of Image Processing as prescribed by the University of Mumbai and our supervision during the academic-year 2022-2023.

Practical In-Charge	Internal Examiner
Date:	Date:
External Examiner.	Director
Date:	IMCOST
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OpenCv

- OpenCV is a programming library/package that has been created especially for allowing programmers to enter the world of Computer Vision. The primary developer of the OpenCV package is Intel Corporation, and the package was released to the public during the year 1999-2000.
- OpenCV stands for Open-Source Computer Vision (Library). It is the most commonly used, popular, and well-documented Computer Vision library. It is open-source, which means that one does not require a license to utilize the software.

NumPy

- NumPy is a Python library used for working with arrays.
- It also has functions for working in domain of linear algebra, fourier transform, and matrices.
- NumPy was created in 2005 by Travis Oliphant. It is an open source project and you
 can use it freely.
 NumPy stands for Numerical Python.
- In Python we have lists that serve the purpose of arrays, but they are slow to process.
 - NumPy aims to provide an array object that is up to 50x faster than traditional Python lists.
 - The array object in NumPy is called ndarray, it provides a lot of supporting functions that make working with ndarray very easy.
 - Arrays are very frequently used in data science, where speed and resources are very important.

Steps to install OpenCV

- Open Anaconda Prompt
 Start Menu / Anaconda 7 Anaconda Prompt
- In Anaconda Prompt, type commands to install necessary libraries: pip install opency-python

- Title: Program to display image using read and write operation.
- imread() helps us read an image
- imshow() displays an image in a window
- imwrite() writes an image into the file directory

Theory:

Program to display image using read and write operation.

Reading, displaying, and writing images are basic to image processing and computer vision. Even when cropping, resizing, rotating, or applying different filters to process images.

For reading an image, use the imread() function in OpenCV. Here's the syntax:

imread(filename, flags)

It takes two arguments:

The first argument is the image name, which requires a fully qualified pathname to the file.

The second argument is an optional flag that lets you specify how the image should be represented.

OpenCV offers several options for this flag, but those that are most common include:

cv2.IMREAD_UNCHANGED or -1 cv2.IMREAD_GRAYSCALE

or 0

cv2.IMREAD_COLOR or 1

Source Code:

```
import numpy as np import cv2 print("This program read
```

and write image") # Load an color image

in grayscale

 $img = cv2.imread ('C:\NUsers\NIM160\NDesktop\NPython\ Practicals\NPractical \ Practical \ New York \ Practical \$

 $1 \land bike.jpg', 1)$

#Display the image

cv2.imshow('image',img)

#key binding function

cv2.waitKey(0)

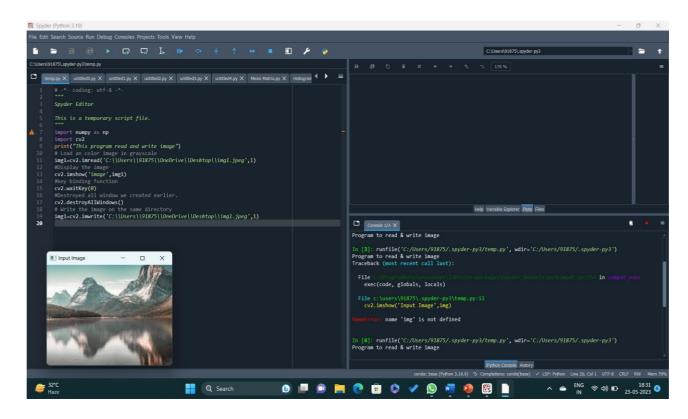
#Destroyed all window we created earlier. cv2.destroyAllWindows()

Write the image on the same directory

 $cv2.imwrite ("C:\Users\IM160\Desktop\Python Practicals\Practical Practical Practical$

1\\grayscale.jpg", img);

Output:



- Title: Program to enhance image arithmetic and logical operations.
- The input images can be subjected to arithmetic operations such as addition, subtraction, multiplication and bitwise operations (AND, OR, NOT, XOR).

 These operations can help to improve the qualities of the input photos. •

 Image Addition:
- Python OpenCV provides cv2.addWeighted() method to add two images. This method calculates the weighted sum of two arrays. It returns an image in the output, which is a combination of the corresponding pixel values of the input images Image Subtraction:
- In Python OpenCV, the cv2.subtract() method is used to subtract two images. Both images should be of same size and type.
- Image Multiplication:
- In Python OpenCV, cv2.multiply() method is used to multiply two images.
- OpenCV Bitwise AND
- Python OpenCV provides cv2.bitwise_and() method to perform bitwise AND logical operation. It combines corresponding pixels of two image buffers by a bitwise AND operation. It is commonly used for detecting differences in the input images. It returns an image highlighting the target regions with a binary mask.
- OpenCV Bitwise OR
- Python OpenCV provides cv2.bitwise_or() method to perform bitwise OR logical operation. It combines corresponding pixels of two image buffers by a bitwise OR operation. The bitwise OR operator are useful for processing

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binary-valued images (0 or 1) to detect objects which have moved between frames.

- OpenCV Bitwise XOR
- Python OpenCV provides cv2.bitwise_xor() method to perform bitwise XOR logical operation. It combines corresponding pixels of two image buffers by a bitwise XOR operation. It is also used for processing binary-valued images (0 or 1) to detect objects which have moved between frames.
- OpenCV Bitwise NOT
- Bitwise NOT or complement is a unary operation that performs logical negation on each bit. It is forming the ones' complement of the given binary value. It means the bits that are 1 become 0, and those that are 0 become 1. Python OpenCV provides cv2.bitwise_not() method to perform a bitwise NOT operation on each pixel of the input image. It inverts the image representation.
- Title: Program to enhance image arithmetic and logical operations.
 2.1 Program to perform addition operation on image.

Theory:

Program to enhance image using image arithmetic and logical operations.

Arithmetic Operations like Addition, Subtraction, and Bitwise Operations(AND, OR, NOT, XOR) can be applied to the input images. These operations can be helpful in enhancing the properties of the input images. The Image arithmetics are important for analyzing the input image properties. The operated images can be further used as an enhanced input image, and many more operations can be applied for clarifying, thresholding, dilating etc of the image.

Addition of Image:

We can add two images by using function **cv2.add()**. This directly adds up image pixels in the two images.

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Syntax: cv2.add(img1, img2)

Subtraction of Image:

Just like addition, we can subtract the pixel values in two images and merge them with the help of cv2.subtract(). The images should be of equal size and depth.

Syntax: cv2.subtract(src1, src2)

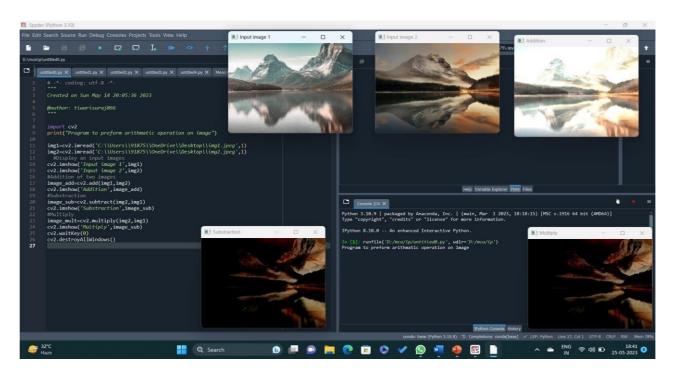
Source Code:

```
# -*- coding: utf-8 -*-
** ** **
Created on Sun May 14 20:05:36 2023
@author: OMKAR """ import cv2
print("Program to preform arithmatic operation on image")
img1=cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img1.jpeg',1)
img2=cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg',1)
#Displey an input images
cv2.imshow('Input image 1',img1)
cv2.imshow('Input image 2',img2)
#Addition of two images
image_add=cv2.add(img1,img2)
cv2.imshow('Addition',image add)
#Substraction image sub=cv2.subtract(img2,img1)
cv2.imshow('Substraction',image sub)
```

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```
#Multiply
image_mult=cv2.multiply(img2,img1)
cv2.imshow('Multiply',image_sub)
cv2.waitKey(0) cv2.destroyAllWindows()
```

Output:



2.2 Program to perform logical operation on image. Source

Code:

```
# -*- coding: utf-8 -*-
```

```
Created on Sun May 14 21:25:08 2023
@author: OMKAR
** ** **
import cv2
print("Program to perform arithmetic operations on image")
# reading an image img1=cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img1.jpeg',1)
img2=cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg',1)
#Displaying
                                 image
                     an
cv2.imshow('image
                               1',img1)
cv2.imshow('image 2',img2) # logical
bitwise
                and
                             operations
image and=cv2.bitwise and(img1,img2
) cv2.imshow('bitwise And',image and)
                       OR
                              operation
    logical
             bitwise
```

logical bitwise NOT operation

image_or=cv2.bitwise_or(img1,img2)

cv2.imshow('bitwise Or',image or)

logical bitwise XOR operation

image xor=cv2.bitwise xor(img1,img2)

cv2.imshow('bitwise Xor',image xor) #

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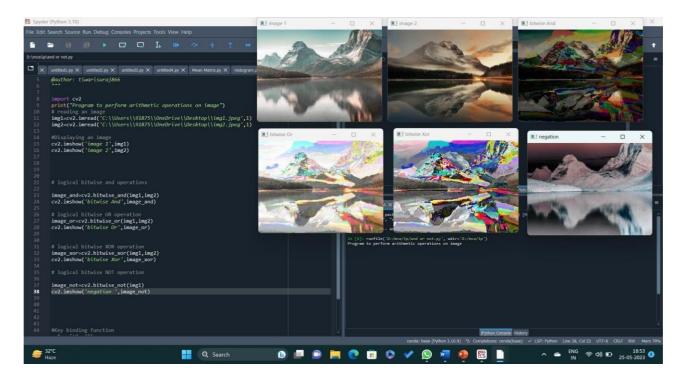
cv2.imshow('negation ',image_not)

#Key binding function cv2.waitKey(0)

#Destroy all previous window

cv2.destroyAllWindows()

Output :



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- Title: Program to Implement Image Negative.
- Image negation is produced by subtracting each pixel from the maximum intensity value.
- For e.g. for an 8-bit image, the maximum intensity value is 28 -1 = 255. Thus each pixel is subtracted from 255 to produces the resulting image.
- Thus the transformation function used in image negative is:

$$s = T(r) = L-1-r$$

Where, L-1 is maximum intensity value

s is output pixel value.

r is input pixel value.

Theory:

For grayscale images, light areas appear dark and vice versa. For color images, colors are replaced by their complementary colors. Thus, red areas appear cyan, greens appear magenta, and blues appear yellow, and vice versa.

Image negative is produced by subtracting each pixel from the maximum intensity value. e.g. for an 8-bit image, the max intensity value is $2^8-1=255$, thus each pixel is subtracted from 255 to produce the output image.

Program to implement Image Negative.

$$s = T(r) = L - 1 - r$$

Thus, the transformation function used in image negative is

Where L-1 is the max intensity value and s, and r are the output and input pixel values respectively.

Source Code:

```
# -*- coding: utf-8 -*-
"""

Created on Sun May 14 20:31:33 2023

@author: OMKAR
"""

import cv2

print("Program to preform arithmatic operation on image")

img2=cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg',1
) cv2.imshow('Input Image',img2)

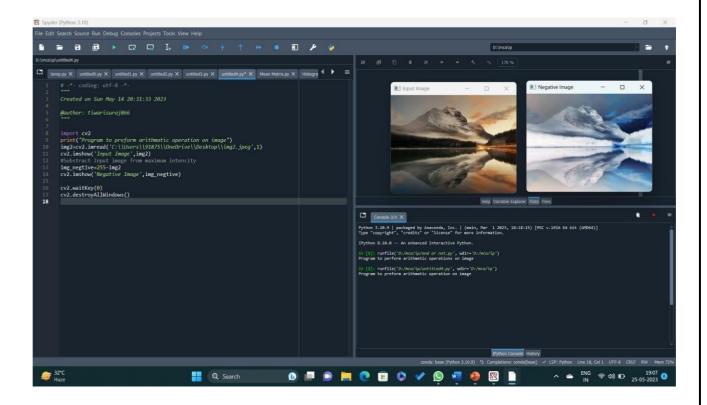
#Substract Input image from maximum intencity

img_negtive=255-img2 cv2.imshow('Negative

Image',img_negtive) cv2.waitKey(0)

cv2.destroyAllWindows()
```

Output:



- Title: Program to implement Thesholding of an image
- Thresholding is a technique in OpenCV, which is the assignment of pixel values in relation to the threshold value provided. In thresholding, each pixel value is compared with the threshold value. If the pixel value is smaller than the threshold, it is set to 0, otherwise, it is set to a maximum value (generally 255). Thresholding is a very popular segmentation technique, used for separating an object considered as a foreground from its background. A threshold is a value which has two regions on its either side i.e. below the threshold or above the threshold.

In Computer Vision, this technique of thresholding is done on grayscale images. So initially, the image has to be converted in grayscale color space.

- If f(x, y) < T then f(x, y) = 0
- else f(x, y) = 255
- where f(x, y) = Coordinate Pixel Value
- T = Threshold Value.
- In OpenCV with Python, the function **cv2.threshold** is used for thresholding.
- Syntax:
- cv2.threshold(source, thresholdValue, maxVal, thresholdingTechnique)
- Parameters:
 - -> source: Input Image array (must be in Grayscale).
 - -> thresholdValue: Value of Threshold below and above which pixel values will change accordingly.
 - -> maxVal: Maximum value that can be assigned to a pixel.
 - -> thresholding Technique: The type of thresholding to be applied.
- The different Simple Thresholding Techniques are:

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- **cv2.THRESH_BINARY**: If pixel intensity is greater than the set threshold, value set to 255, else set to 0 (black).
- **cv2.THRESH_BINARY_INV**: Inverted or Opposite case of cv2.THRESH_BINARY.
- **cv.THRESH_TRUNC**: If pixel intensity value is greater than threshold, it is truncated to the threshold. The pixel values are set to be the same as the threshold. All other values remain the same.
- **cv.THRESH_TOZERO**: Pixel intensity is set to 0, for all the pixels intensity, less than the threshold value. **cv.THRESH_TOZERO_INV**: Inverted or Opposite case of cv2.THRESH_TOZERO.

Theory:

1. Program to implement Thresholding of an Image.

Thresholding is a technique in OpenCV, which is the assignment of pixel values in relation to the threshold value provided. In thresholding, each pixel value is compared with the threshold value. If the pixel value is smaller than the threshold, it is set to 0, otherwise, it is set to a maximum value (generally 255). Thresholding is a very popular segmentation technique, used for separating an object considered as a foreground from its background. A threshold is a value which has two regions on its either side i.e. below the threshold or above the threshold. In Computer Vision, this technique of thresholding is done on grayscale images. So initially, the image has to be converted in grayscale color space.

where

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f (x, y) = Coordinate Pixel Value T

= Threshold Value.

In OpenCV with Python, the function cv2.threshold is used for thresholding.

Syntax: cv2.threshold(source, thresholdValue, maxVal, thresholdingTechnique)

Parameters:

-> **source**: Input Image array (must be in Grayscale).

-> **thresholdValue**: Value of Threshold below and above which pixel values will change accordingly.

-> maxVal: Maximum value that assigned can be pixel. -> thresholdingTechnique: The type of thresholding applied. The different Simple Thresholding Techniques are:

cv2.THRESH_BINARY: If pixel intensity is greater than the set threshold, value set to 255, else set to 0 (black). **cv2.THRESH_BINARY_INV**: Inverted or Opposite case of cv2.THRESH_BINARY.

cv.THRESH_TRUNC: If pixel intensity value is greater than threshold, it is truncated to the threshold. The pixel values are set to be the same as the threshold. All other values remain the same.

cv.THRESH_TOZERO: Pixel intensity is set to 0, for all the pixels intensity, less than the threshold value.

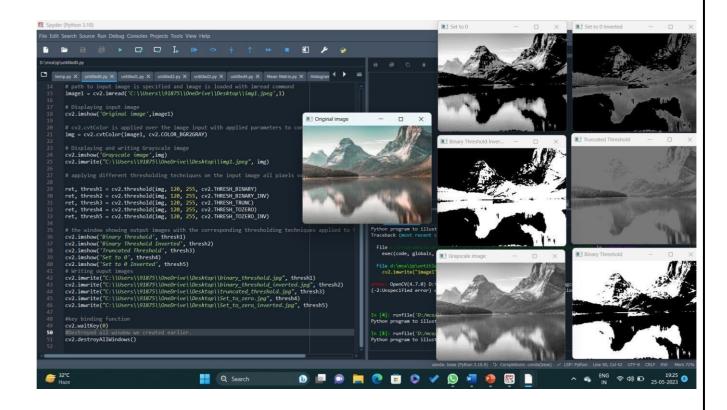
cv.THRESH_TOZERO_INV: Inverted or Opposite case of cv2.THRESH_TOZERO.

Source Code:

```
# -*- coding: utf-8 -*-
       Created on Sun May 14 20:05:36 2023
       @author: OMKAR
       # organizing imports import
       cv2
       import numpy as np
print("Python program to illustrate simple the sholding type on image")
       # path to input image is specified and image is loaded with imread command image1
       = cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img1.jpeg',1)
       # Displaying input image
       cv2.imshow('Original image',image1)
       # cv2.cvtColor is applied over the image input with applied parameters to convert the
       image in grayscale img = cv2.cvtColor(image1, cv2.COLOR_BGR2GRAY)
       # Displaying and writing Grayscale image cv2.imshow('Grayscale
       image',img)
       cv2.imwrite("C:\\Users\\91875\\OneDrive\\Desktop\\img1.jpeg", img)
       # applying different thresholding techniques on the input image all pixels value above
       120 will be set to 255 ret, thresh1 = cv2.threshold(img, 120, 255,
       cv2.THRESH BINARY) ret, thresh2 = cv2.threshold(img, 120, 255,
       cv2.THRESH BINARY INV) ret, thresh3 = cv2.threshold(img, 120, 255,
       cv2.THRESH TRUNC) ret, thresh4 = cv2.threshold(img, 120, 255,
       cv2.THRESH TOZERO) ret, thresh5 = cv2.threshold(img, 120, 255,
       cv2.THRESH TOZERO INV) # the window showing output images with
       the corresponding thresholding techniques applied to the input images
```

```
cv2.imshow('Binary Threshold', thresh1) cv2.imshow('Binary Threshold
Inverted', thresh2) cv2.imshow('Truncated Threshold', thresh3)
cv2.imshow('Set to 0', thresh4) cv2.imshow('Set to 0 Inverted', thresh5)
#
                     Writing
                                                ouput
                                                                          images
cv2.imwrite("C:\\Users\\91875\\OneDrive\\Desktop\\binary threshold.jpg", thresh1)
cv2.imwrite("C:\\Users\\91875\\OneDrive\\Desktop\\binary threshold inverted.jpg
                                                                         thresh2)
cv2.imwrite("C:\\Users\\91875\\OneDrive\\Desktop\\truncated threshold.jpg",
thresh3)
             cv2.imwrite("C:\\Users\\91875\\OneDrive\\Desktop\\Set to zero.jpg",
thresh4)
cv2.imwrite("C:\\Users\\91875\\OneDrive\\Desktop\\Set to zero inverted.jpg",
thresh5)
#key binding function
cv2.waitKey(0)
#Destroyed all window we created earlier.
cv2.destroyAllWindows()
```

Output:



- Title: Program to implement smoothing or averaging filter in spatial domain.
 - The mean filter is an example of a linear filter. It basically replaces each pixel in the output image with the mean (average) value of the neighborhood. This has the effect of smoothing the image (reducing the amount of intensity variations between a pixel and the next), removing noise from the image, and brightening the image.
 - Thus, in mean filtering, each pixel of the image will be replaced with the mean value of its neighbors, including the pixel itself. The 3x3 kernel used for mean filtering is as shown in the figure below, although other kernel sizes could be used (i.e. 5x5):

1/9	1/9	1/9
1/9	1/9	1/9
1/9	1/9	1/9

- What the above kernel is actually trying to tell us is that we sum all the elements under the kernel and take the mean (average) of the total. An important point to mention here is that all the elements of the mean kernel should:
 - **O** sum to 1
 - **O** be the same

Theory:

Program to Implement smoothing or averaging filter in spatial domain.

<u>Spatial Filtering</u> technique is used directly on pixels of an image. Mask is usually considered to be added in size so that it has a specific center pixel. This mask is moved on the image such that the center of the mask traverses all image pixels.

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Neighborhood processing in spatial domain: Here, to modify one pixel, we consider values of the immediate neighboring pixels also. For this purpose, 3X3, 5X5, or 7X7 neighborhood mask can be considered. An example of a 3X3 mask is shown below. f(x-1, y-1) f(x-1, y) f(x-1, y+1) f(x, y-1) f(x, y) f(x, y+1) f(x+1, y-1) f(x+1, y) f(x+1, y+1)

Low Pass filtering: It is also known as the smoothing filter. It removes the high-frequency content from the image. It is also used to blur an image. A low pass averaging filter mask is as shown.

1/9 1/9 1/9

1/9 1/9 1/9

1/9 1/9 1/9

High Pass Filtering: It eliminates low-frequency regions while retaining or enhancing the high-frequency components. A high pass filtering mask is as shown.

- -1/9 -1/9 -1/9
- -1/9 8/9 -1/9
- -1/9 -1/9 -1/9

Median Filtering: It is also known as nonlinear filtering. It is used to eliminate salt and pepper noise. Here the pixel value is replaced by the median value of the neighboring pixel.

cv2.waitKey(0)

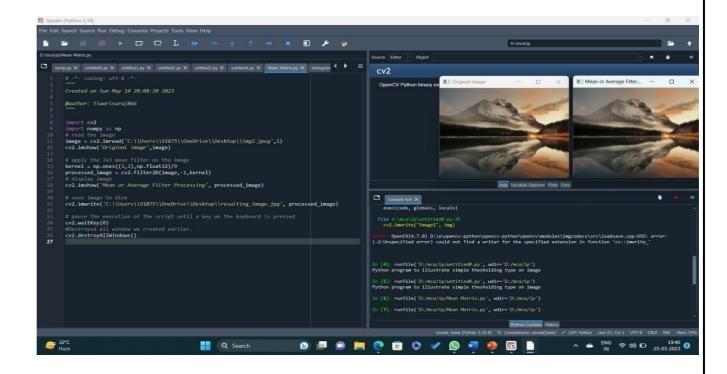
Source Code: # -*- coding: utf-8 -*-Created on Sun May 14 20:08:20 2023 @author: OMKAR """ import cv2 import numpy as np # read the image image = cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg',1) cv2.imshow('Original image',image) # apply the 3x3 mean filter on the image kernel = np.ones((3,3),np.float32)/9 processed_image = cv2.filter2D(image,-1,kernel) # display image cv2.imshow('Mean or Average Filter Processing', processed image) # save image to disk cv2.imwrite("C:\\Users\\91875\\OneDrive\\Desktop\\resulting_image.jpg", processed_image) # pause the execution of the script until a key on the keyboard is pressed

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#Destroyed all window we created earlier.

cv2.destroyAllWindows()

Output:



- Title: Program to produce the Histogram, Equalized Histogram and Equalized image of an input image.
- Histogram is considered as a graph or plot which is related to frequency of pixels in an Gray Scale Image with pixel values (ranging from 0 to 255). Grayscale image is an image in which the value of each pixel is a single sample, that is, it carries only intensity information where pixel value varies from 0 to 255. Images of this sort, also known as black-and-white, are composed exclusively of shades of gray, varying from black at the weakest intensity to white at the strongest where Pixel can be considered as a every point in an image.
- we use cv2.calcHist()(in-built function in OpenCV) to find the histogram.
 cv2.calcHist(images, channels, mask, histSize, ranges[, hist[, accumulate]])
- images: it is the source image of type uint8 or float32 represented as "[img]". channels: it is the index of channel for which we calculate histogram. For grayscale image, its value is [0] and color image, you can pass [0], [1] or [2] to calculate histogram of blue, green or red channel respectively. mask: mask image. To find histogram of full image, it is given as "None". histSize: this represents our BIN count. For full scale, we pass [256]. ranges: this is our RANGE. Normally, it is [0,256].

Theory:

Program to produce the Histogram, Equalized Histogram, and Equalized image of an input image.

Histogram equalization is a method in image processing of contrast adjustment using the image's histogram.

This method usually increases the global contrast of many images, especially when the usable data of the image is represented by close contrast values. Through this adjustment, the intensities can be better distributed on the histogram. This allows for areas of lower local contrast to gain a higher

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contrast. Histogram equalization accomplishes this by effectively spreading out the most frequent intensity values. The method is useful in images with backgrounds and foregrounds that are both

bright or both dark.

OpenCV has a function to do this, cv2.equalizeHist(). Its input is just grayscale image and output is

our histogram equalized image.

The histogram of a digital image, with intensity levels between 0 and (L-1), is a function $h(r_k) = n_k$

, where r_k is the kth intensity level and n_k is the number of pixels in the image having that intensity

level. We can also normalize the histogram by dividing it by the total number of pixels in the image.

For an N x N image, we have the following definition of a normalized histogram function:

This $p(r_k)$ function is the probability of the occurrence of a pixel with the intensity level r_k . Clearly,

The histogram of an image, as shown in the figure, consists of the x-axis representing the intensity

levels r_k and the y-axis denoting the $h(r_k)$ or the $p(r_k)$ functions.

The histogram of an image gives important information about the grayscale and contrast of the

image. If the entire histogram of an image is centered towards the left end of the x-axis, then it

implies a dark image. If the histogram is more inclined towards the right end, it signifies a white or

bright image. A narrow-width histogram plot at the center of the intensity axis shows a low-

contrast image, as it has a few levels of grayscale. On the other hand, an evenly distributed

histogram over the entire x-axis gives a high-contrast effect to the image.

In image processing, there frequently arises the need to improve the contrast of the image. In such

cases, we use an intensity transformation technique known as histogram equalization. Histogram

equalization is the process of uniformly distributing the image histogram over the entire intensity

axis by choosing a proper intensity transformation function. Hence, histogram equalization is an

intensity transformation process.

The choice of the ideal transformation function for uniform distribution of the image histogram is

mathematically explained below.

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```
Source Code:
# -*- coding: utf-8 -*-
,,,,,,
Spyder Editor
This is a temporary script file.
""" import
cv2
from matplotlib import pyplot as plt
# reads an input image img =
cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg',1)
# find frequency of pixels in range 0-255 histr =
cv2.calcHist([img],[0],None,[256],[0,256])
# show the plotting graph of an image
plt.plot(histr) plt.show()
# alternative way to find histogram of an image
plt.hist(img.ravel(),256,[0,256]) plt.show()
# To produced Equalized image
```

```
img = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY) # It converts the image into grayscale dst = cv2.equalizeHist(img) # Apply histogram equalization with the function # Display both original & Equalized images

cv2.imshow('Source image', img)

cv2.imshow('Equalized Image', dst) #

Storing the equalized image

cv2.imwrite('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg',dst)

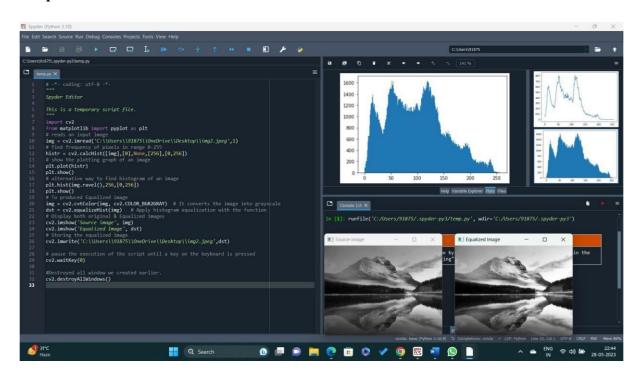
# pause the execution of the script until a key on the keyboard is pressed

cv2.waitKey(0)

#Destroyed all window we created earlier.

cv2.destroyAllWindows()
```

Output:



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- Title: Program for smooth an image using low pass filter in frequency domain.
 - As in one-dimensional signals, images also can be filtered with various low-pass filters(LPF), high-pass filters(HPF) etc. LPF helps in removing noises, blurring the images etc. OpenCV provides a function cv2.filter2D() to convolve a kernel with an image.

Theory:

Program for smooth an image using low pass filter in frequency domain.

Low pass filter removes the high frequency components that means it keeps low frequency components. It is used for smoothing the image. It is used to smoothen the image by attenuating high frequency components and preserving low frequency components.

G(u, v) = H(u, v) . F(u, v)

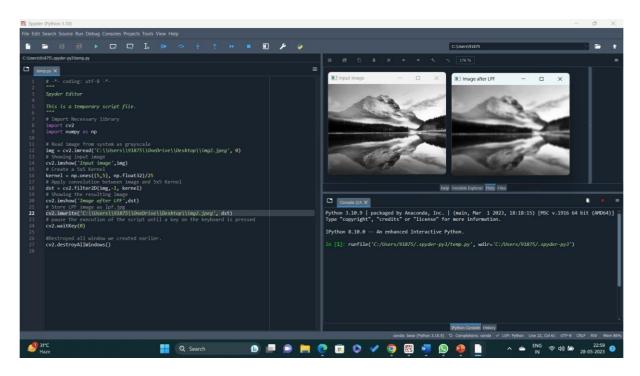
where F(u, v) is the Fourier Transform of original image and

H(u, v) is the Fourier Transform of filtering mask.

Source Code:

cv2.destroyAllWindows()

```
# -*- coding: utf-8 -*-
# Import Necessary library
import cv2 import numpy
as np
# Read image from system as grayscale
img = cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg', 0)
# Showing input image
cv2.imshow('Input image',img) #
Create a 5x5 Kernel kernel =
np.ones((5,5), np.float32)/25
# Apply convolution between image and 5x5 Kernel dst =
cv2.filter2D(img,-1, kernel) # Showing the resulting image
cv2.imshow('Image after LPF',dst) # Store LPF image as lpf.jpg
cv2.imwrite('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg', dst)
# pause the execution of the script until a key on the keyboard is pressed
cv2.waitKey(0)
#Destroyed all window we created earlier.
```



- Title: Program for smooth an image using high pass filter in frequency domain.
- A **High Pass Filter(HPF)** main advantage is used to sharpen the image by attenuating the low frequency. When the impulse response or signal is passed through a high pass filter, an HPF mainly allows high frequencies to pass through. As High pass filters are used for sharpening the images, the frequency obtained is less compared to cut-off frequency(ωc). In OpenCV and in digital image processing we also use HPF functionality to find the edges in an image.
- The reason to blur the image is to add smoothening effect to an image. By blurring the image we can filter the unwanted noise from the image. In the OpenCV library, we widely use Gaussian Filter. It employs the technique "kernel convolution".
- **Note:** 127 is added after subtracting the image with a blurred image to add the greyish look. We shall use Gaussian Blur to blur the image.
- dst = cv2.GaussianBlur(src, ksize, sigmaX[, dst[, sigmaY[, borderType=BORDER DEFAULT]]])
- hpf = img cv2.GaussianBlur(img,(21,21),3)+127

Theory:

Program for sharpen the image using high pass filter in frequency domain

High pass filter removes the low frequency components that means it keeps high frequency components. It is used for sharpening the image. It is used to sharpen the image by attenuating low frequency components and preserving high frequency components.

Mechanism of high pass filtering in frequency domain is given by: H(u, v) = 1 - H'(u, v) where H(u, v) is the Fourier Transform of high pass filtering and H'(u, v) is the Fourier Transform of low pass filtering

-*- coding: utf-8 -*-

""" Spyder

Editor

This is a temporary script file.

** ** **

Import Necessary library

import cv2 # Reading the

image

 $img = cv2.imread('C:\Users\)91875\)OneDrive\Desktop\img1.jpeg', 1)$

subtract the original image with the blurred image

after subtracting add 127 to the total result

hpf = img - cv2.GaussianBlur(img, (21, 21), 3)+127

display both original image and filtered image cv2.imshow("Original",

img) cv2.imshow("High Passed Filter", hpf) # Storing the result of HPF

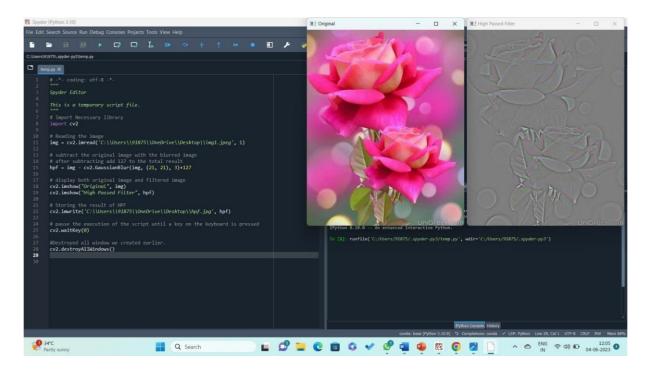
cv2.imwrite('C:\\Users\\91875\\OneDrive\\Desktop\\hpf.jpg', hpf) #

pause the execution of the script until a key on the keyboard is pressed

cv2.waitKey(0)

#Destroyed all window we created earlier.

cv2.destroyAllWindows()



- Title: Program to find DFT/FFT Forward and Inverse Transform of Image
- Fast Fourier Transform (FFT) to transform image to frequency domain.
- Digital images are not continuous so we use Fast Fourier Transform (DFT) instead of Fourier transform.

Theory:

Program to find DFT/FFT forward and Inverse Transform of Image.

Fourier Transform is a mathematical technique that helps to transform Time Domain function x(t) to Frequency Domain function $X(\omega)$. In this article, we will see how to find Fourier Transform in MATLAB.

Properties of Fourier Transform:

<u>Linearity:</u> The addition of two functions corresponding to the addition of the two frequency spectrum is called linearity. If we multiply a function by a constant, the Fourier transform of the resultant function is multiplied by the same constant. The Fourier transform of the sum of two or more functions is the sum of the Fourier transforms of the functions.

Scaling: Scaling is the method that is used to change the range of the independent variables or features of data. If we stretch a function by the factor in the time domain then squeeze the Fourier transform by the same factor in the frequency domain. **Differentiation:** Differentiating function with respect to time yields to the constant multiple of the initial function.

<u>Convolution</u>: It includes the multiplication of two functions. The Fourier transform of a convolution of two functions is the point-wise product of their respective Fourier transforms.

Frequency Shift and Time Shift: Frequency is shifted according to the coordinates. There is a duality between the time and frequency domains and frequency shift affects the time shift. The time variable shift also affects the frequency function. The time-

shifting property concludes that a linear displacement in time corresponds to a linear phase factor in the frequency domain.

Equation for DFT:
$$X(k) = \sum_{n=0}^{N-1} x[n].e^{\frac{-j2\pi kn}{N}}$$

Equation for IDFT:
$$x(n) = \sum_{k=0}^{N-1} X[k].e^{\frac{j2\pi kn}{N}}$$

```
Source Code:
# -*- coding: utf-8 -*-
Created on Sun Jun 4 12:10:55 2023
@author: OMKAR
** ** **
import cv2 as cv import numpy as
np from matplotlib import pyplot as
plt
# Create a picture
\# src = np.ones((5, 5), dtype=np.uint8)*100
src = cv.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img1.jpeg' ,0)
print("*"*100) print(src) print(src.shape) f = np.fft.fft2(src)
print("*"*100) print(f) fshift = np.fft.fftshift(f) print("*"*100)
print(fshift)
# Convert complex numbers into floating-point numbers for Fourier spectrum display fimg
= np.log(np.abs(fshift))
print(fimg)
# Inverse Fourier transform ifshift
= np.fft.ifftshift(fshift)
```

```
# Convert complex numbers into floating-point numbers for Fourier spectrum display ifimg = np.log(np.abs(ifshift)) if_img = np.fft.ifft2(ifshift) origin_img = np.abs(if_img)

# Image display plt.subplot(221), plt.imshow(src, "gray"),

plt.title('origin') plt.axis('off') plt.subplot(222), plt.imshow(fimg,

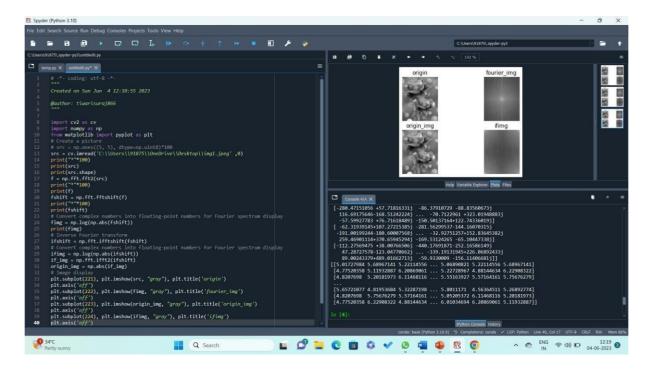
"gray"), plt.title('fourier_img') plt.axis('off') plt.subplot(223),

plt.imshow(origin_img, "gray"), plt.title('origin_img') plt.axis('off')

plt.subplot(224), plt.imshow(ifimg, "gray"), plt.title('ifimg')

plt.axis('off')

plt.show()
```



- Title: Program to find DCT forward and Inverse Transform of Image.
- Discrete Cosine Transform is used in lossy image compression because it has very strong energy compaction, i.e., its large amount of information is stored in very low frequency component of a signal and rest other frequency having very small data which can be stored by using very less number of bits (usually, at most 2 or 3 bit).

Theory:

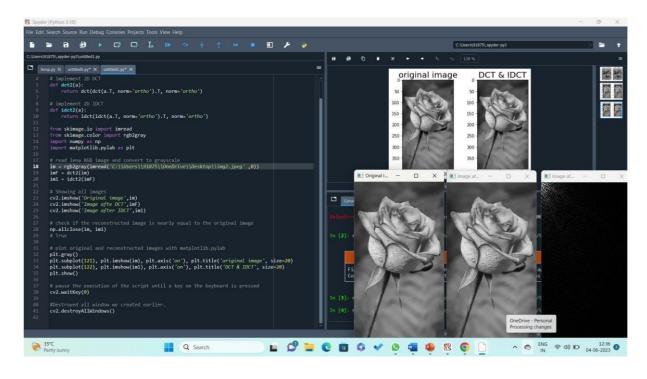
Program to find DCT forward and Inverse Transform of Image.

Discrete Cosine Transform is used in lossy image compression because it has very strong energy compaction, i.e., its large amount of information is stored in very low frequency component of a signal and rest other frequency having very small data which can be stored by using very less number of bits (usually, at most 2 or 3 bit). To perform DCT Transformation on an image, first we have to fetch image file information (pixel value in term of integer having range 0-255) which we divides in block of 8 X 8 matrix and then we apply discrete cosine transform on that block of data.

After applying discrete cosine transform, we will see that its more than 90% data will be in lower frequency component. For simplicity, we took a matrix of size 8 X 8 having all value as 255 (considering image to be completely white) and we are going to perform 2-D discrete cosine transform on that to observe the output.

```
# -*- coding: utf-8 -*-
*****
Created on Sun Jun 4 12:10:55 2023
@author: OMKAR
*****
from scipy.fftpack import dct, idct
import cv2 # implement 2D DCT
def dct2(a):
  return dct(dct(a.T, norm='ortho').T, norm='ortho')
# implement 2D IDCT
def idct2(a):
  return idct(idct(a.T, norm='ortho').T, norm='ortho')
from skimage.io import imread from skimage.color
import rgb2gray import numpy as np import
matplotlib.pylab as plt
# read lena RGB image and convert to grayscale
im = rgb2gray(imread('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg',0))
imF = dct2(im) im1 = idct2(imF)
```

```
# Showing all images
cv2.imshow('Original
                          image',im)
cv2.imshow('Image afte DCT',imF)
cv2.imshow('Image after IDCT',im1)
# check if the reconstructed image is nearly equal to the original image np.allclose(im,
im1)
# True
# plot original and reconstructed images with matplotlib.pylab plt.gray()
plt.subplot(121), plt.imshow(im), plt.axis('on'), plt.title('original image', size=20)
plt.subplot(122), plt.imshow(im1), plt.axis('on'), plt.title('DCT & IDCT',
size=20) plt.show()
# pause the execution of the script until a key on the keyboard is pressed
cv2.waitKey(0)
#Destroyed all window we created earlier.
cv2.destroyAllWindows()
```



- Title: Program to find Edges using Prewit/Sobel/Fri-chen/Robert operators.
- Edge detection involves mathematical methods to find points in an image where the brightness of pixel intensities changes distinctly.
- cv2.Sobel(original image,ddepth,xorder,yorder,kernelsize)

Theory:

Program to find Edges using Prewit/ Sobel/ Fri-chen / Robert operators.

The Prewitt operator was developed by Judith M. S. Prewitt. Prewitt operator is used for edge detection in an image. Prewitt operator detects both types of edges, these are:

Horizontal edges or along the x-axis, Vertical

Edges or along the y-axis.

Wherever there is a sudden change in pixel intensities, an edge is detected by the mask. Since the edge is defined as the change in pixel intensities, it can be calculated by using differentiation. Prewitt mask is a first-order derivative mask. In the graph representation of Prewitt-mask's result, the edge is represented by the local maxima or local minima.

Both the first and second derivative masks follow these three properties:

More weight means more edge detection.

The opposite sign should be present in the mask. (+ and -) The

Sum of the mask values must be equal to zero.

Prewitt operator provides us two masks one for detecting edges in the horizontal direction and another for detecting edges in a vertical direction.

Prewitt Operator [X-axis] = [-1 0 1; -1 0 1; -1 0 1]

Prewitt Operator [Y-axis] = [-1 -1 -1; 0 0 0; 1 1 1]

It is named after Irwin Sobel and Gary Feldman. Like the Prewitt operator <u>Sobel operator</u> is also used to detect two kinds of edges in an image:

- 1. Vertical direction
- 2. Horizontal direction

Image Processing

ACEDMIC YEAR: 2022-23

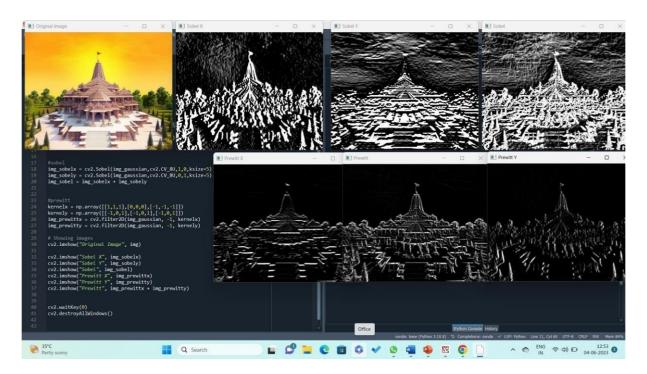
The difference between Sobel and Prewitt Operator is that in Sobel operator the coefficients of masks are adjustable according to our requirement provided they follow all properties of derivative masks.

Sobel-X Operator = [-1 0 1; -2 0 2; -1 0 1]

Sobel-Y Operator = [-1 -2 -1; 0 0 0; 1 2 1]

```
# -*- coding: utf-8 -*-
       11 11 11
       Created on Sun Jun 4 12:10:55 2023
       @author: OMKAR
       import cv2
       import numpy as np
       img = cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img3.jpeg')
       gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY) img_gaussian
       = cv2.GaussianBlur(gray,(3,3),0)
       #sobel img sobelx =
       cv2.Sobel(img gaussian,cv2.CV 8U,1,0,ksize=5) img sobely
       = cv2.Sobel(img_gaussian,cv2.CV_8U,0,1,ksize=5) img_sobel
       = img_sobelx + img_sobely
       #prewitt
kernelx = np.array([[1,1,1],[0,0,0],[-1,-1,-1]])
kernely = np.array([[-1,0,1],[-1,0,1],[-1,0,1]]) img prewittx =
       cv2.filter2D(img gaussian, -1, kernelx) img prewitty =
       cv2.filter2D(img_gaussian, -1, kernely)
       #
                 Showing
                                   images
       cv2.imshow("Original Image", img)
       cv2.imshow("Sobel X", img_sobelx)
       cv2.imshow("Sobel Y", img_sobely)
```

```
cv2.imshow("Sobel", img_sobel)
cv2.imshow("Prewitt X",
img_prewittx) cv2.imshow("Prewitt
Y", img_prewitty)
cv2.imshow("Prewitt",
img_prewittx + img_prewitty)
cv2.waitKey(0)
cv2.destroyAllWindows()
```



- Title: Program to find edges using canny Edge Detection.
- <u>Canny edge detector</u> is an edge detection operator that uses multi-stage algorithm to detect a wide range of edges in images. The main stages are:
- Filtering out noise using <u>Gaussian blur</u> algorithm.
- Finding the strength and direction of edges using <u>Sobel Filters</u>.
- Isolating the strongest edges and thin them to one-pixel wide lines by applying <u>non-maximum suppression</u>.
- Using <u>hysteresis</u> to isolate the best edges.

Theory:

Program to find edges using canny Edge Detection.

Canny Edge detection is an Algorithm consisting of 4 major steps:

Reduce Noise using Gaussian Smoothing.

Compute image gradient using Sobel filter.

Apply Non-Max Suppression or NMS to just jeep the local maxima

Finally, apply Hysteresis thresholding which that 2 threshold values T_upper and T_lower which is used in the Canny() function.

Canny Edge filter in OpenCV. Canny() Function in OpenCV is used to detect the edges in an image.

Syntax: cv2.Canny(image, T_lower, T_upper, aperture_size, L2Gradient) **Where:**

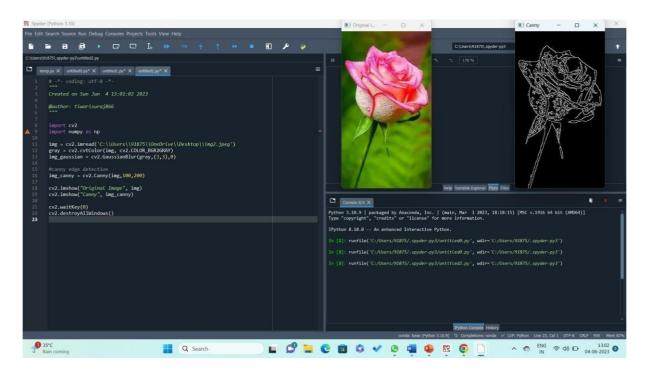
- 1. Image: Input image to which Canny filter will be applied
- 2. T_lower: Lower threshold value in Hysteresis Thresholding 3.

T_upper: Upper threshold value in Hysteresis Thresholding

- 4. aperture_size: Aperture size of the Sobel filter.
- 5. L2Gradient: Boolean parameter used for more precision in calculating Edge Gradient.

MC2223044 OMKAR KHANDARE

```
# -*- coding: utf-8 -*-
      Created on Sun Jun 4 13:01:02 2023
      @author: OMKAR
      *****
      import cv2
      import numpy as np
      img = cv2.imread('C:\\Users\\91875\\OneDrive\\Desktop\\img2.jpeg')
      gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY) img_gaussian
      = cv2.GaussianBlur(gray,(3,3),0)
      #canny edge detection
      img_canny = cv2.Canny(img, 100, 200)
      cv2.imshow("Original Image", img) cv2.imshow("Canny",
      img_canny)
cv2.waitKey(0)
      cv2.destroyAllWindows()
```



- Title: Program to implement Huffmann coding technique for image compression.
- Huffman Encoding is a Lossless Compression Algorithm used to compress the data. It is
 an algorithm developed by David A. Huffman.
 Following are the two steps in
 Huffman Coding
 - 1. Building Huffman Tree
 - 2. Assigning codes to Leaf Nodes

Theory:

Program to implement Huffman coding technique for image compression.

Huffman coding is one of the basic compression methods, that have proven useful in image and video compression standards. When applying Huffman encoding technique on an Image, the source symbols can be either pixel intensities of the Image, or the output of an intensity mapping function.

The first step of Huffman coding technique is to reduce the input image to a ordered histogram, where the probability of occurrence of a certain pixel intensity value is as prob_pixel = numpix/totalnum

where **numpix** is the number of occurrence of a pixel with a certain intensity value and **totalnum** is the total number of pixels in the input Image.

```
# -*- coding: utf-8 -*-
       Created on Sun Jun 4 13:10:04 2023
       @author: OMKAR
       *****
       import re import
       numpy as np from PIL
       import Image import
       cv2
print("Huffman Compression Program")
       print("=
       h = int(input("Enter 1 if you want to input an colour image file, 2 for default gray
       scale case:")) if h == 1:
          file = input("Enter the filename:")
          my_string =
          np.asarray(Image.open(file),np.uint8) shape =
          my string.shape a = my string print ("Enetered
          string is:",my_string) my_string =
          str(my_string.tolist())
       elif h == 2:
          array = np.arange(0, 737280, 1, np.uint8)
          my_string = np.reshape(array, (1024, 720))
          print ("Enetered string is:",my string) a =
```

MC2223044 OMKAR KHANDARE

```
my_string my_string =
  str(my string.tolist())
else:
  print("You entered invalid input")
                                               # taking user input
letters = [] only letters
= [] for letter in
my string:
  if letter not in letters:
    frequency = my string.count(letter) #frequency of each letter repetition
    letters.append(frequency) letters.append(letter) only letters.append(letter)
nodes = [] while len(letters)
> 0:
nodes.append(letters[0:2])
  letters = letters[2:]
                                       # sorting according to frequency
nodes.sort() huffman tree
= []
huffman_tree.append(nodes) node
                                              #Make each unique character as a leaf
def combine_nodes(nodes):
  pos = 0 newnode
  = [] if len(nodes)
  > 1: nodes.sort()
    nodes[pos].append("1")
                                          # assigning values 1 and 0
    nodes[pos+1].append("0") combined node1 =
    (nodes[pos][0] + nodes[pos+1][0])
    combined node2 = (nodes[pos][1] + nodes[pos+1][1]) \# combining the nodes to
generate pathways
    newnode.append(combined node1)
    newnode.append(combined node2)
```

```
newnodes=[]
            newnodes.append(newnode)
            newnodes = newnodes + nodes[2:] nodes
            = newnodes
            huffman tree.append(nodes)
            combine_nodes(nodes)
         return huffman tree
                                                    # huffman tree generation
       newnodes = combine nodes(nodes)
       huffman_tree.sort(reverse = True) print("Huffman
       tree with merged pathways:")
checklist = [] for level in
       huffman tree: for node in
       level:
            if node not in checklist:
              checklist.append(node)
            else:
              level.remove(node)
       count = 0 for level in
       huffman tree:
         print("Level", count,":",level) #print huffman tree
       count+=1 print()
       letter binary = [] if
       len(only_letters) == 1:
         lettercode = [only letters[0], "0"] letter binary.append(letter code*len(my string))
       else:
         for letter in
            only letters: code
            ="" for node in
            checklist:
              if len (node)>2 and letter in node[1]:
                                                         #genrating binary code
```

```
code = code + node[2]
     lettercode = [letter,code] letter binary.append(lettercode)
print(letter_binary)
print("Binary code generated:")
for letter in letter binary:
  print(letter[0], letter[1])
bitstring ="" for character
in my string:
  for item in letter binary:
     if character in item:
       bitstring = bitstring + item[1]
binary ="0b"+bitstring print("Your
message as binary is:")
                         # binary code generated
uncompressed file size = len(my string)*7 compressed file size = len(binary)-2
print("Your original file size was", uncompressed file size, "bits. The compressed size
is:",compressed file size) print("This is a saving of ",uncompressed file size-
compressed file size,"bits") output = open("compressed.txt","w+")
print("Compressed file generated as compressed.txt") output =
open("compressed.txt","w+") print("Decoding. .....")
output.write(bitstring)
bitstring = str(binary[2:])
uncompressed string =""
code ="" for digit in
bitstring:
              code
                        =
code+digit
  pos=0 #iterating and decoding for letter in letter binary:
     if code ==letter[1]:
       uncompressed string=uncompressed string+letter binary[pos] [0]
       code=""
```

```
pos+=1
print("Your UNCOMPRESSED data is:") if
h == 1:
  temp = re.findall(r'\d+', uncompressed string) res = list(map(int, temp)) res =
  np.array(res) res = res.astype(np.uint8) res = np.reshape(res, shape) print(res)
  print("Observe the shapes and input and output arrays are matching or not")
  print("Input image dimensions:",shape) print("Output image
  dimensions:",res.shape) data = Image.fromarray(res)
  data.save('C:\\Users\\IM160\\Desktop\\Python Practical\\Practical
  13 \cdot \text{huffman.jpg'} if a.all() == res.all():
     print("Success")
if h == 2:
  temp = re.findall(r'\d+',
  uncompressed string) res = list(map(int,
  temp)) print(res) res = np.array(res) res =
  res.astype(np.uint8) res = np.reshape(res,
  (1024, 720)) print(res)
  data = Image.fromarray(res)
  data.save('C:\\Users\\IM160\\Desktop\\Python Practicals\\Practical
  13\huffman.jpg') print("Success")
```

