
Computer Vision And Image Processing

Project 1

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1.Edge Detection - With sobel filtering

Edge detection reduces the overhead of processing huge amounts of data in image processing. Edge detection in sobel filtering is done by detecting changes in frequencies, ΔX . It works by approximately detecting the gradient of the images. The sobel filter is as follows

The steps for sobel filtering is as follows

1. apply convolution of sobel matrix - in x axis
2. apply convolution of sobel matrix - in y-axis
3. apply gradient with results of 1 and 2

the X sobel matrix

-1	0	+1
-2	0	+2
-1	0	+1

G_x

the y sobel matrix is

+1	+2	+1
0	0	0
-1	-2	-1

G_y

The output is as follows

Sobel X



Sobel Y



code

```
from lib import basics
```

```
import cv2
```

```
import numpy as np
```

```
import math
```

```
img = cv2.imread("task1.png",0)
```

```
kernel_v = np.array([[1,2,1],[0,0,0],[-1,-2,-1]])
```

```
kernel_h = np.array([[1,0,-1],[2,0,-2],[1,0,-1]])
```

```
edge_detected_image_v = basics.apply_convolution1(img, kernel_v)
```

```
edge_detected_image_h = basics.apply_convolution1(img, kernel_h)
```

```
cv2.imshow('img',img)
```

```
# cv2.imshow('blurred_image',blurred_image)
```

```
cv2.imshow('edge_detected_image_v',edge_detected_image_v)
```

```
cv2.imshow('edge_detected_image_h',edge_detected_image_h)
```

```
cv2.imshow('gradient',basics.find_gradient(edge_detected_image_v,edge_detected_image_h))
```

```
cv2.waitKey()
```

```
def apply_convolution1(image,kernel,size=3):
```

```
    rows,cols = image.shape
```

```
    new_image = np.zeros((rows,cols), dtype='uint8')
```

```
    for i in range(0,3):
```

```
        for j in range(0,3):
```

```
            total = 0
```

```
            for k in range(0,size):
```

```
                for m in range(0,size):
```

```
                    total = total + kernel[k,m] * image[i+k,j+m]
```

```
            new_image[i+math.floor(size/2)+1,j+math.floor(size/2)+1] = total
```

```
    return new_image
```

```
def find_gradient(image1,image2):
```

```
    rows,cols = image1.shape[0],image1.shape[1]
```

```
    for i in range(rows):
```

```
        for j in range(cols):
```

```
            image1[i,j] = math.sqrt(image1[i,j]**2 + image2[i,j]**2)
```

```
    return image1
```

Task 2: Key point detection

Keypoints are major points of interest in image processing

Keypoint detection is done by the method of scalar invariant fourier transform by the following steps

1. the images are downscaled to 4 octaves .
2. In each octave multiple images the gaussian blurring is applied progressively with 5 sigma values
3. the difference of gaussian for the images are found with respect to the next image in the same octave resulting in 4 difference of gaussian images for each octave
4. these difference of gaussian images are used to detect keypoints by comparing the the each other on ignoring the outermost octave.we compare with matrices above and below we get a list of keypoints. the key points of higher score are obtained by thresholding which ignore key points of no interest

output

output of keypoints

13,22

8,23

7,24

25,22

23,36

octave 4



octave 3



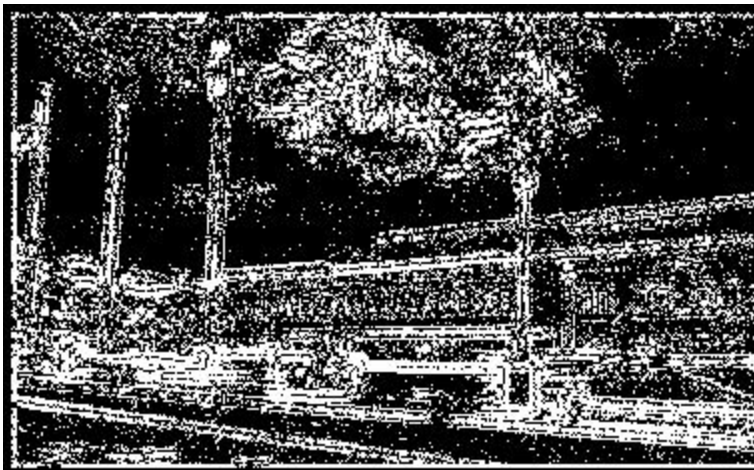
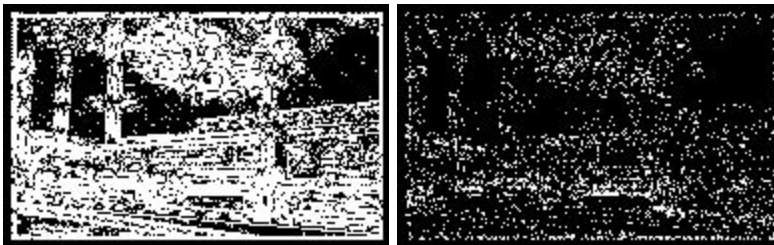
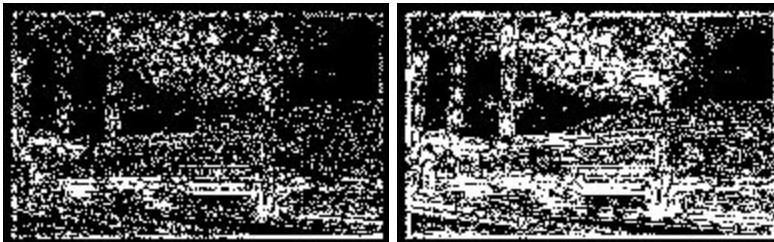
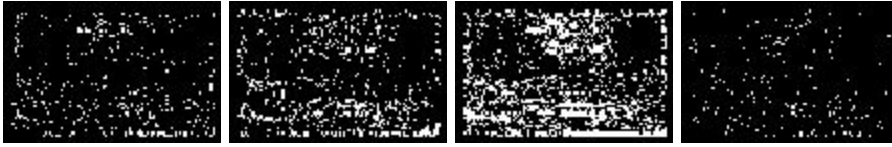
octave 2

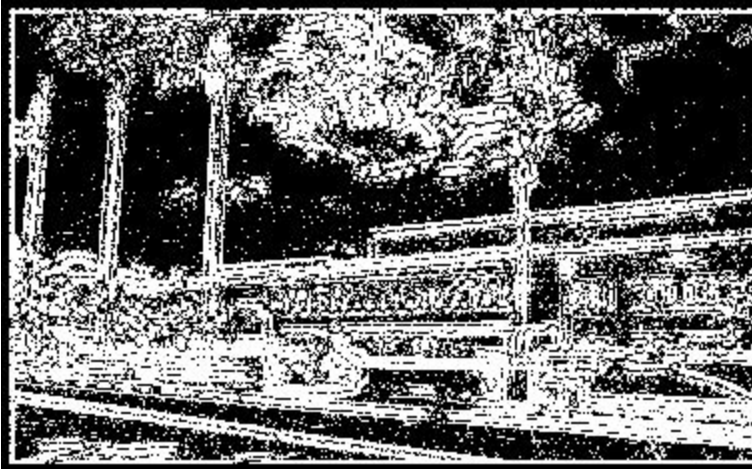


ocative 1

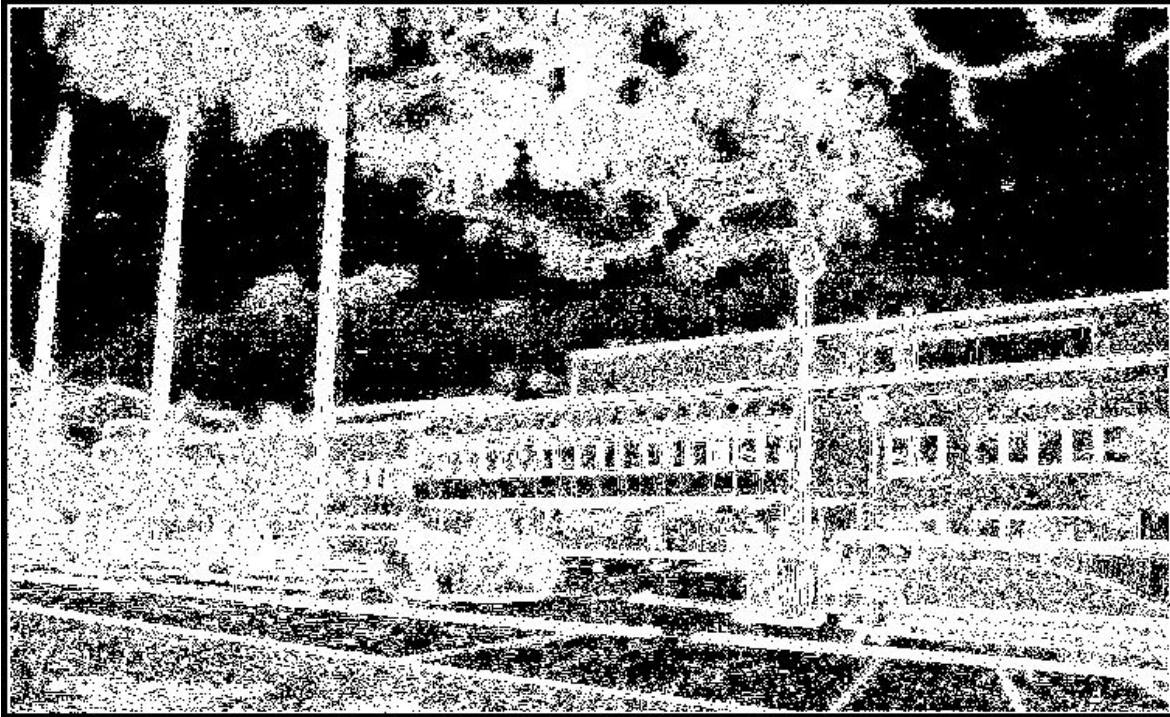


the difference of gaussian









code:

```

from lib import basics
import numpy as np
import cv2
import math

rgb_image = cv2.imread("task2.jpg")
original_image = cv2.imread("task2.jpg",0)

img_scaled = np.empty(shape=5,dtype=object)
img_scaled[0] = original_image.copy()
img_scaled[1] = np.array(basics.scale_half(img_scaled[0])).copy()
img_scaled[2] = np.array(basics.scale_half(img_scaled[1])).copy()
img_scaled[3] = np.array(basics.scale_half(img_scaled[2])).copy()

gaussian_matrix = [[1/math.sqrt(2),1,math.sqrt(2),2,2*math.sqrt(2)],
                    [math.sqrt(2),2,2*math.sqrt(2),4,4*math.sqrt(2)],
                    [2*math.sqrt(2),4,4*math.sqrt(2),8,8*math.sqrt(2)],
                    [4*math.sqrt(2),8,8*math.sqrt(2),16,16*math.sqrt(2)]
                    ]

image_array = np.empty((4,5), dtype = object)
DoG_array = np.empty((4,4),dtype = object)
for i in range(0,4):
    for j in range(0,5):
        gaussian_kernel = basics.calculate_gaussian(gaussian_matrix[i][j],7)
        image_array[i,j] =
basics.apply_convolution(basics.add_padding(img_scaled[i],7//2),gaussian_kernel,7)
        if j > 0:

```

```

        DoG_array[i,j-1] = basics.computer_difference(image_array[i,j],image_array[i,j-1])
    print(gaussian_kernel)

key_point_array = np.empty((4,2),dtype = object)
for j in range (0,2):
    key_point_array_1 = basics.find_keys(DoG_array[0][0],DoG_array[0][1],DoG_array[0][2])
    key_point_array_2 = basics.find_keys(DoG_array[0][1],DoG_array[0][2],DoG_array[0][3])
    key_point_array_21 =
basics.find_keys(DoG_array[1][0],DoG_array[1][1],DoG_array[1][2])
    key_point_array_22 =
basics.find_keys(DoG_array[1][1],DoG_array[1][2],DoG_array[1][3])
    key_point_array_31 =
basics.find_keys(DoG_array[2][0],DoG_array[2][1],DoG_array[2][2])
    key_point_array_32 =
basics.find_keys(DoG_array[2][1],DoG_array[2][2],DoG_array[2][3])
    key_point_array_41 =
basics.find_keys(DoG_array[3][0],DoG_array[3][1],DoG_array[3][2])
    key_point_array_42 =
basics.find_keys(DoG_array[3][1],DoG_array[3][2],DoG_array[3][3])

plot_image_1 = rgb_image.copy()
img_copy_1 =img_scaled[0].copy()
img_copy_2 =np.array(img_scaled[1]).copy()
img_copy_3 =np.array(img_scaled[2]).copy()
img_copy_4 =np.array(img_scaled[3]).copy()

for point in key_point_array_1:
    cv2.circle(img_scaled[0],point,1,(0,255,0))
for point in key_point_array_2:
    cv2.circle(img_copy_1,point,1,(255,0,0))

for point in key_point_array_21:
    cv2.circle(img_scaled[1],point,1,(0,255,0))
for point in key_point_array_22:
    cv2.circle(img_copy_2,point,1,(255,0,0))

for point in key_point_array_31:
    cv2.circle(img_scaled[2],point,1,(0,255,0))
for point in key_point_array_32:
    cv2.circle(img_copy_3,point,1,(255,0,0))

for point in key_point_array_41:
    cv2.circle(img_scaled[3],point,1,(0,255,0))
for point in key_point_array_42:
    cv2.circle(img_copy_4,point,1,(255,0,0))

cv2.imshow('image1',img_scaled[0])

cv2.imshow('image12',img_copy_1)

```

```

cv2.imshow('image21',img_scaled[1])
cv2.imshow('image22',img_copy_2)
cv2.imshow('image31',img_scaled[2])
cv2.imshow('image32',img_copy_3)
cv2.imshow('image41',img_scaled[3])
cv2.imshow('image42',img_copy_4)
cv2.waitKey(0)

```

```

def apply_convolution(image,kernel,size=3):
    rows,cols = image.shape
    new_image = np.zeros((rows,cols), dtype='float32')
    sum_total =0.0
    for i in range(0,size):
        for j in range(0,size):
            sum_total = sum_total + kernel[i][j]
    for i in range(0,rows-size):
        for j in range(0,cols-size):
            total = 0.0
            for k in range(0,size):
                for m in range(0,size):
                    total = total + kernel[k,m] * image[i +k,j+m]
            new_image[i+math.floor(size/2)+1,j+math.floor(size/2)+1] = abs((total)/(sum_total))
    return new_image

def scale_half(image):
    scaled_x = image[:,::2]
    scaled_xy = scaled_x[:,1::2]
    return scaled_xy

```

```

def calculate_gaussian(sigma_value,size):
    print(size)
    sum = 0.0
    print(sigma_value)
    new_kernel = np.zeros((size,size),dtype = float)
    for i in range(int(-math.floor(size/2)),int(math.floor(size/2)+1)):
        for j in range(int(-math.floor(size/2)),int(math.floor(size/2)+1))
            numerator = math.exp(-float((i*i+j*j))/float(2*sigma_value*sigma_value))
            denominator = (2*(math.pi)* math.pow(sigma_value,2))

```

```

        new_kernel[(i+math.floor(size/2)),(j+math.floor(size/2))] = numerator/denominator
        sum = sum + new_kernel[i+math.floor(size/2),j+math.floor(size/2)]
    return new_kernel

```

```

def computer_difference(image1,image2):
    rows,cols = image1.shape
    new_image = np.zeros((rows,cols),dtype = np.float)
    for i in range(0,rows):
        for j in range (0 , cols):
            new_image[i,j] = abs(int(image1[i,j])-int(image2[i,j]))
    return new_image

```

```

def find_keys(d1,d2,d3):
    keys = []
    for i in range(1,d1.shape[0]-1):
        for j in range(1,d1.shape[1]-1):
            d11 = d1[i-1:i+2,j-1:j+2] #3X3 matrix in dog1
            d22 = d2[i-1:i+2,j-1:j+2] #3X3 matrix in dog2
            d33 = d3[i-1:i+2,j-1:j+2] #3X3 matrix in dog3
            center = d22[1,1] #Center element in dog2 - d22
            d22 = np.delete(d22[1], 1) #Delete center element from the dog2 - d22
            mn1,mx1 = d11.min(),d11.max() #Min and Max element in d11
            mn2,mx2 = d22.min(),d22.max() #Min and Max element in d22
            mn3,mx3 = d33.min(),d33.max() #Min and Max element in d33
            mins = min([mn1,mn2,mn3]) #Mins between the 3 mins above
            maxs = max([mx1,mx2,mx3]) #Maxs between the 3 maxs above
            #Keypoint should be either less than the min or greater than the max
            if((center<mins) or (center>maxs)) and center > 0.2:
                keys.append((j,i))
    return keys

```

```

def add_padding(img, padding):
    img = np.pad(img, (padding, padding), mode='constant', constant_values=0)
    return img

```

Task 3: Cursor detection

Cursor detection using template matching involves a series of steps of image processing followed by multi scale template matching

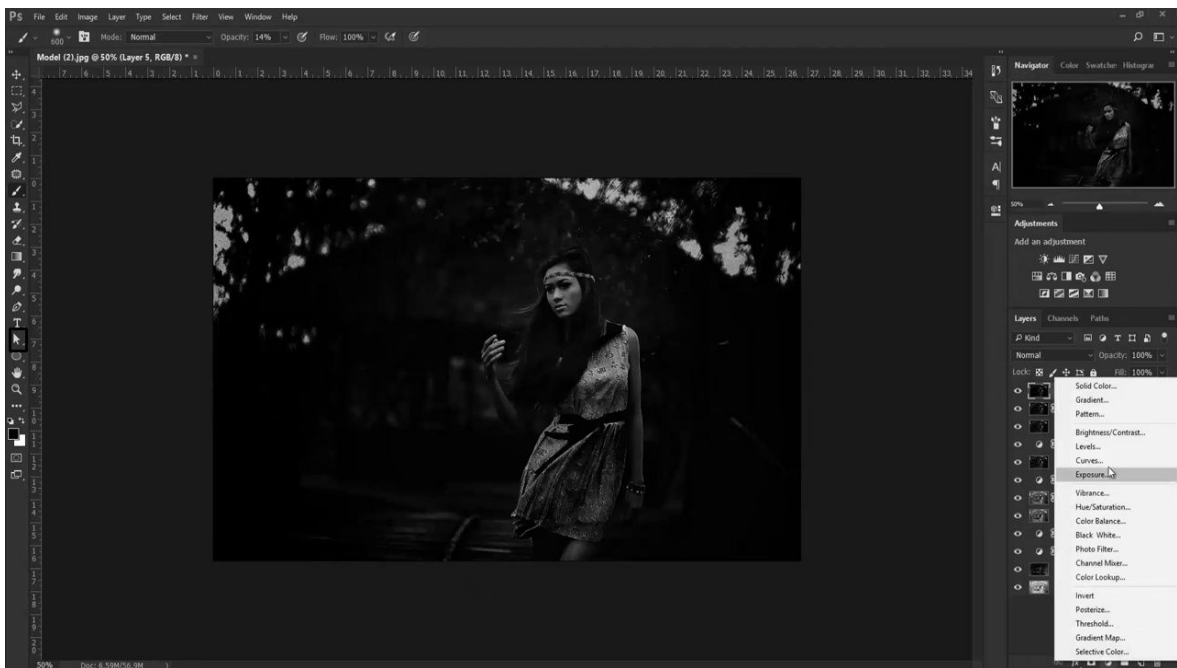
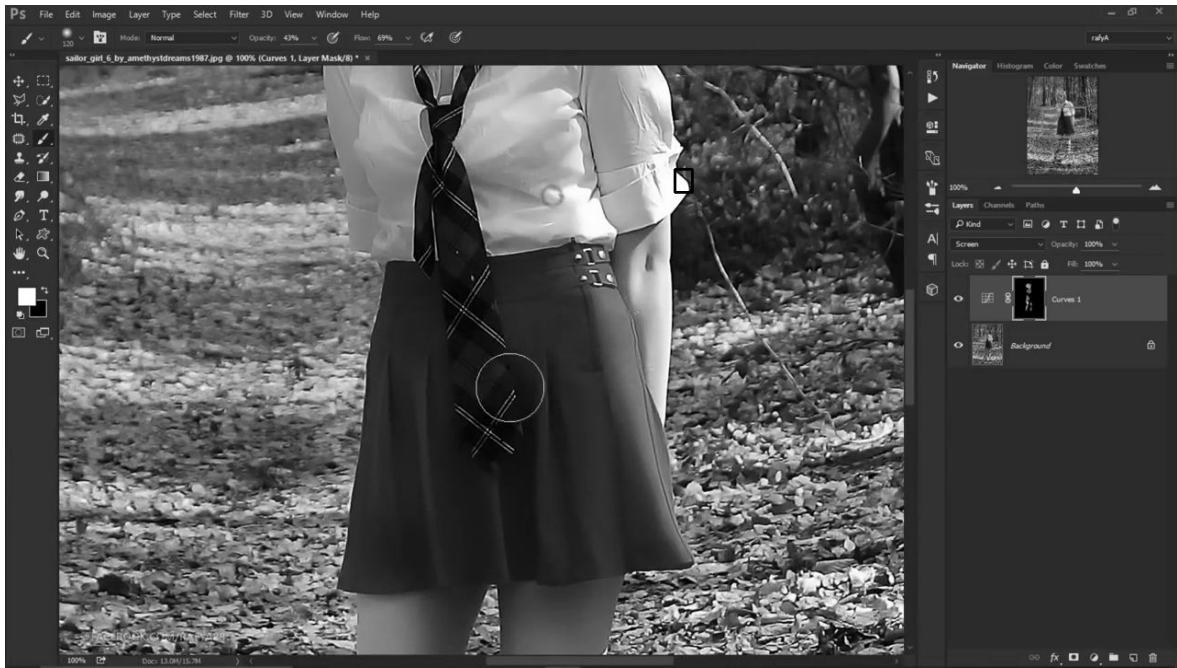
My first attempt was thresholding the images since most cursor are white. for that a lot of false positives appeared.

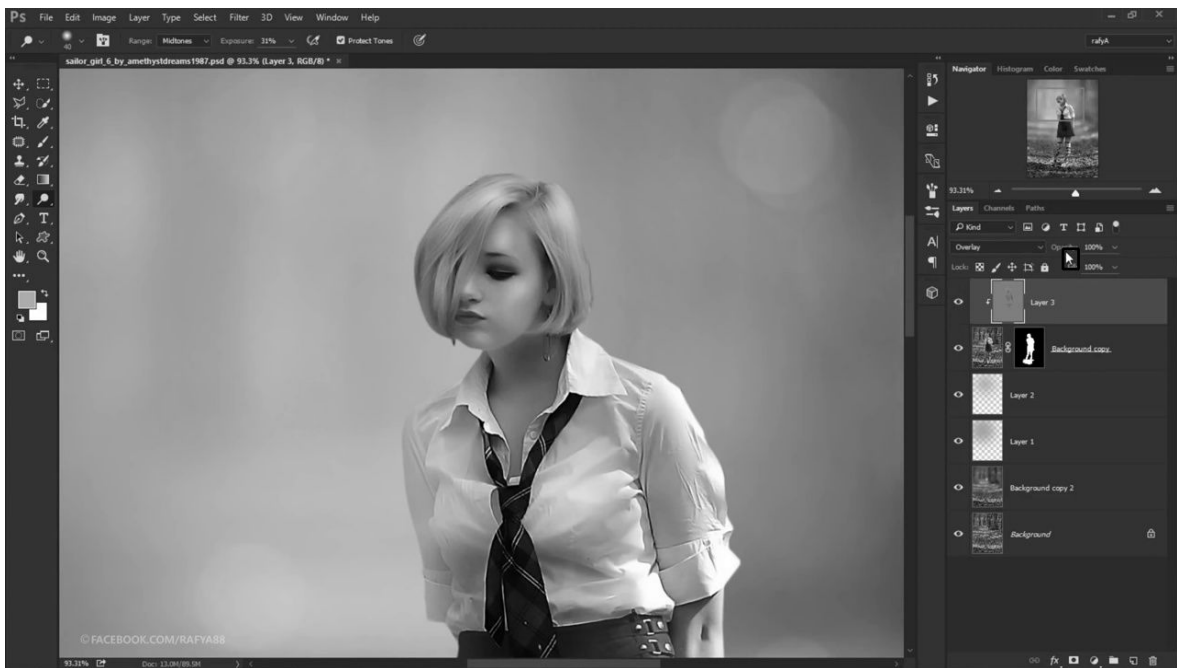
Thus the next attempt was made to do gaussian blur followed by laplacian transform on the images and the template. scale the template to multiple size and do template matching

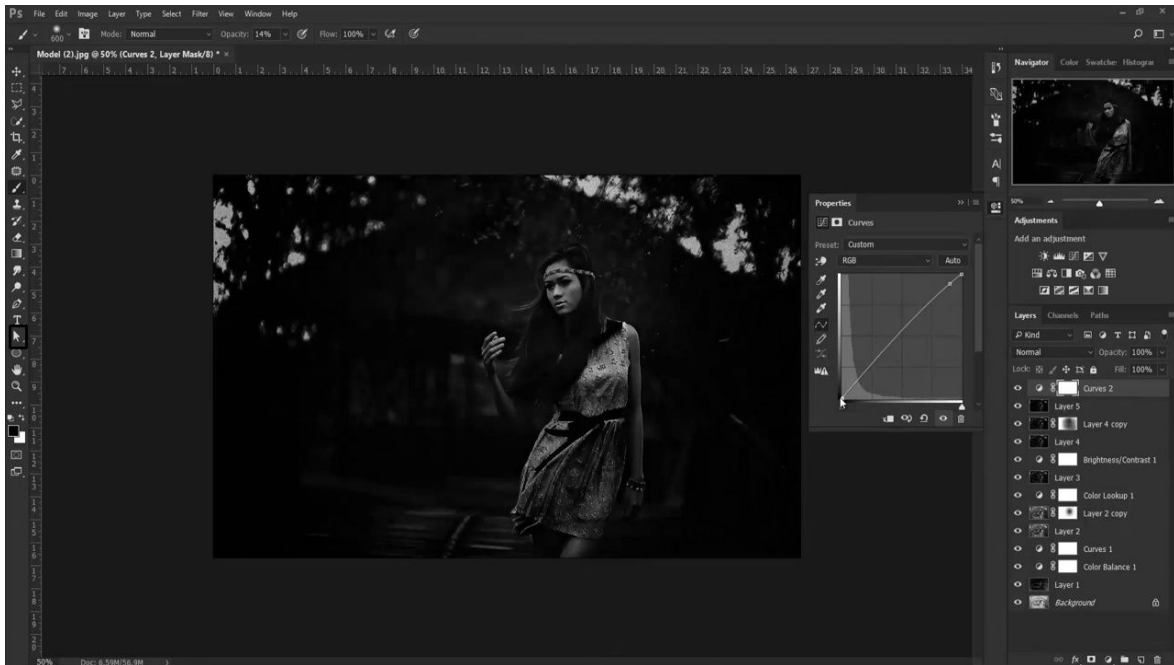
the gaussian blur would smoothen out noise in image and the laplacian transform will detect edges in the images

the final output is as followsc

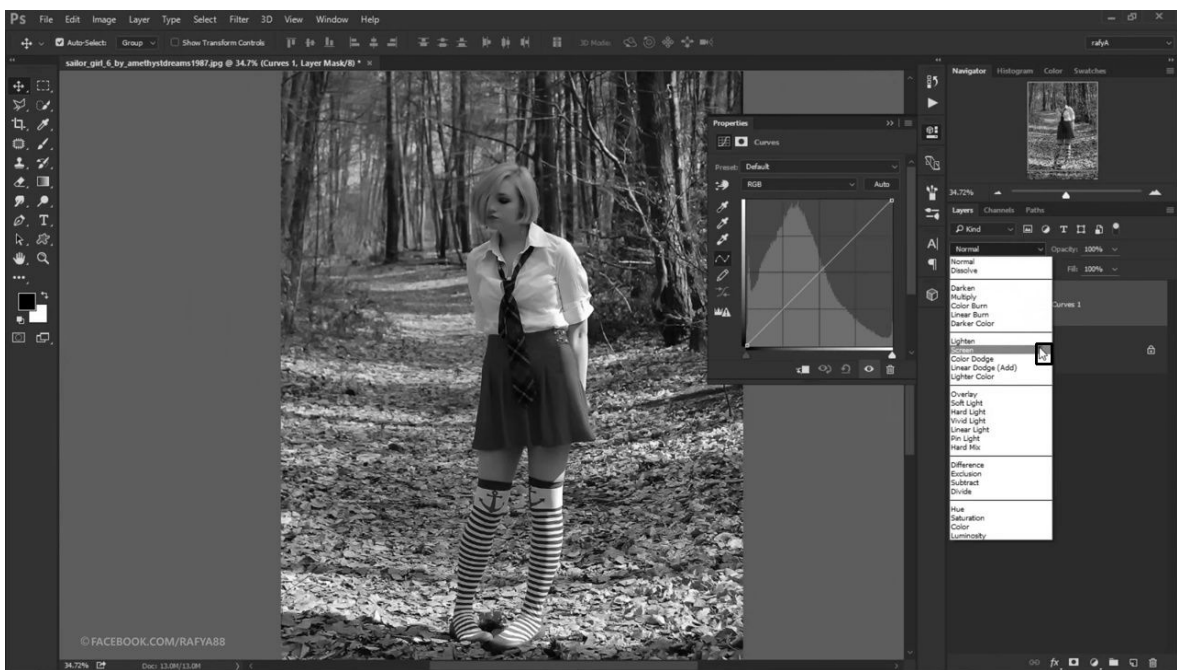


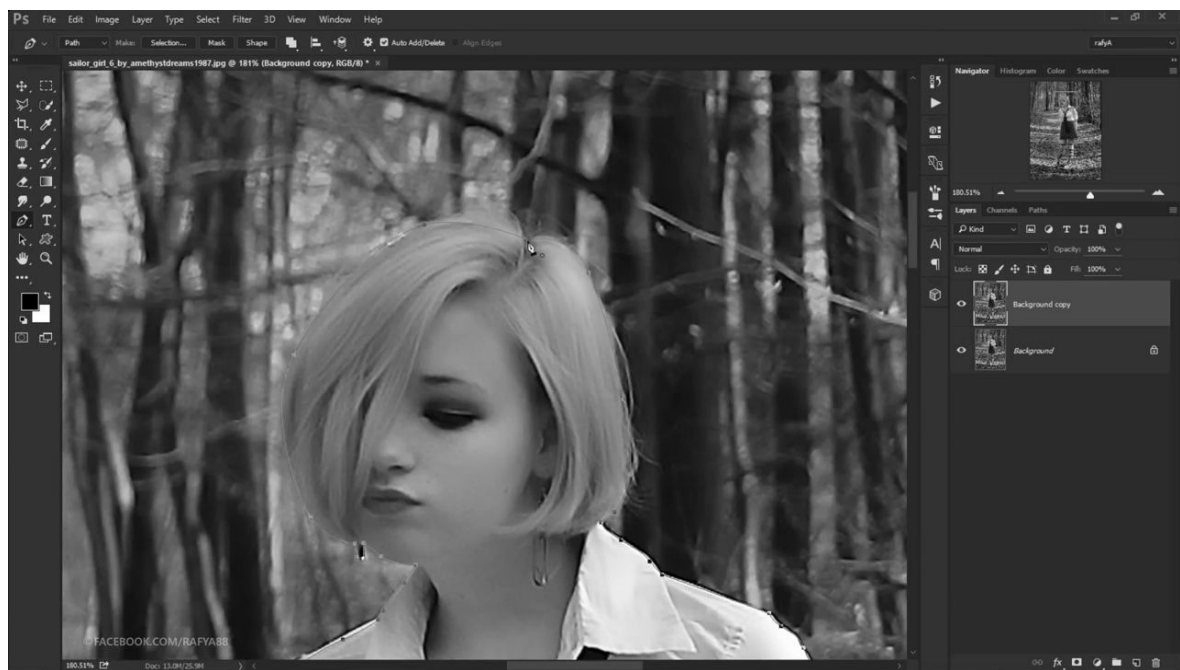


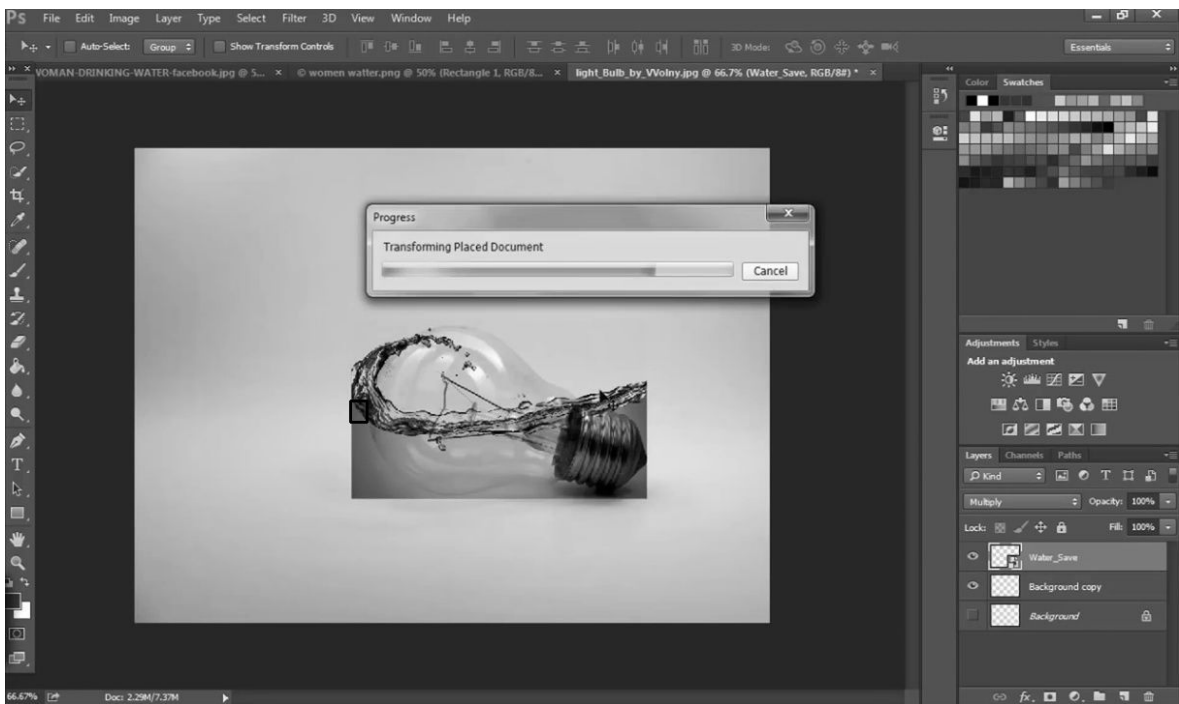


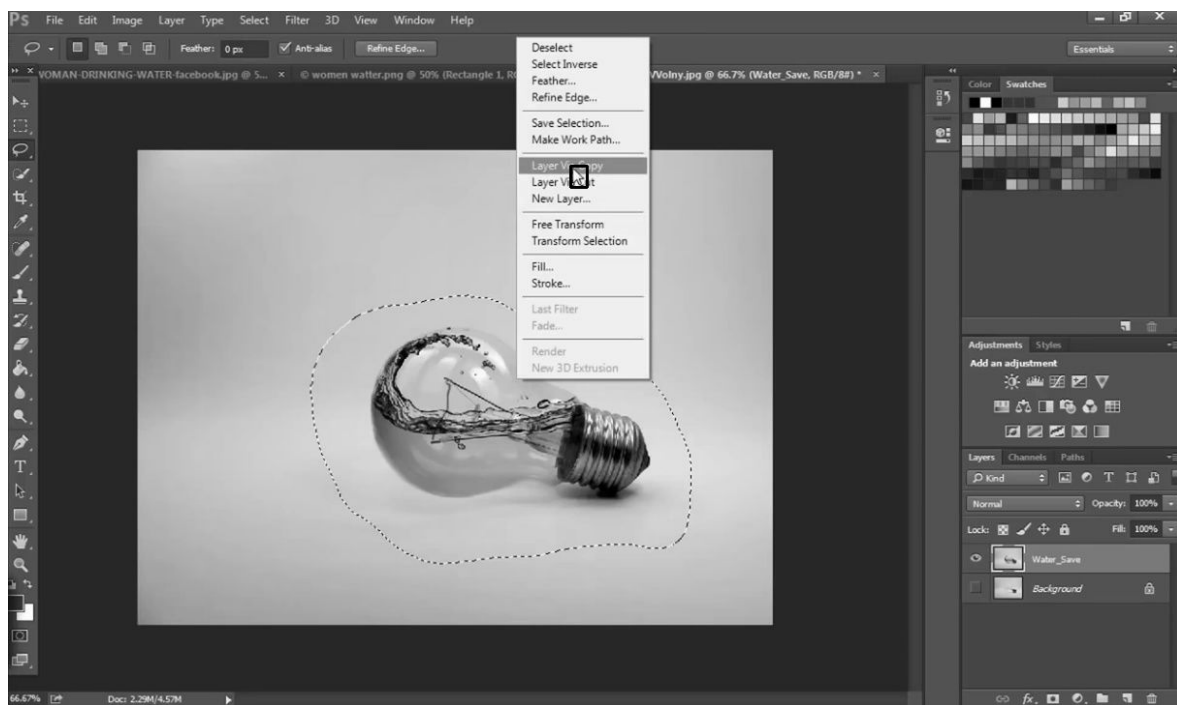


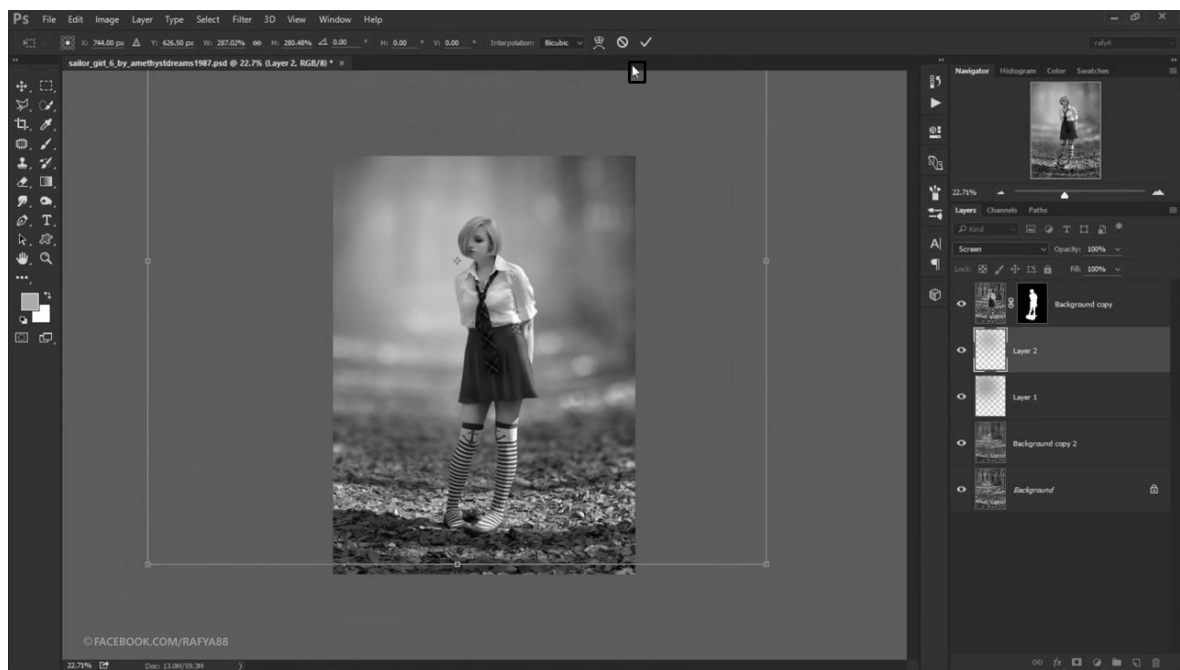
THANK YOU FOR WATCHING
LIKE AND SHARE VIDEO IF YOU LIKE

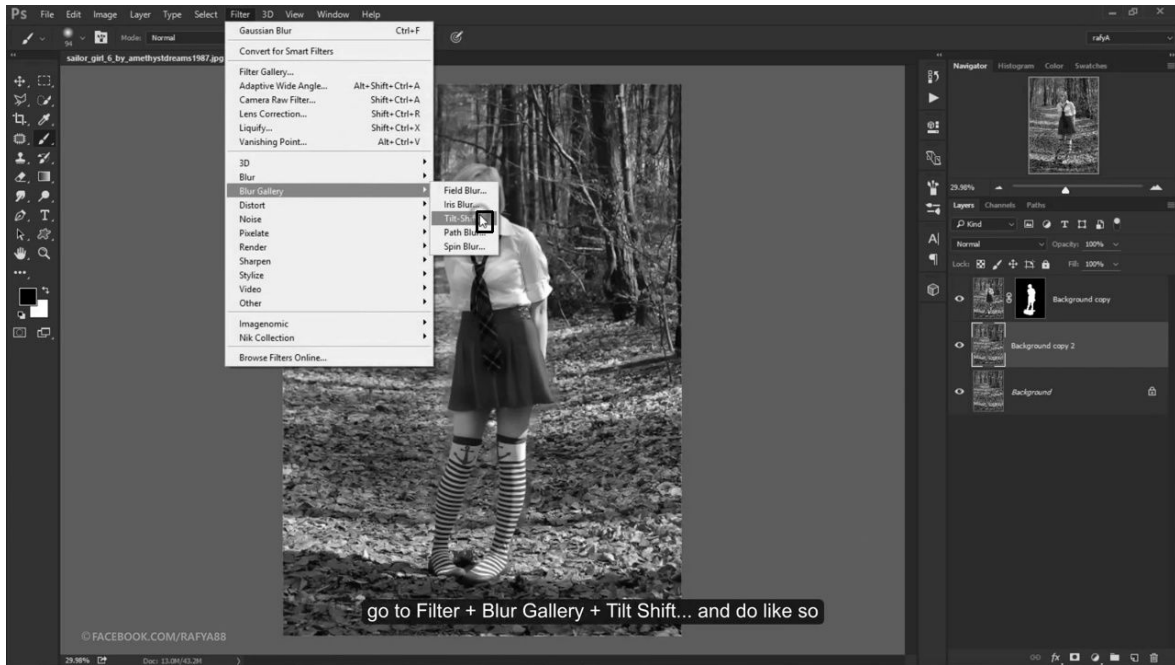


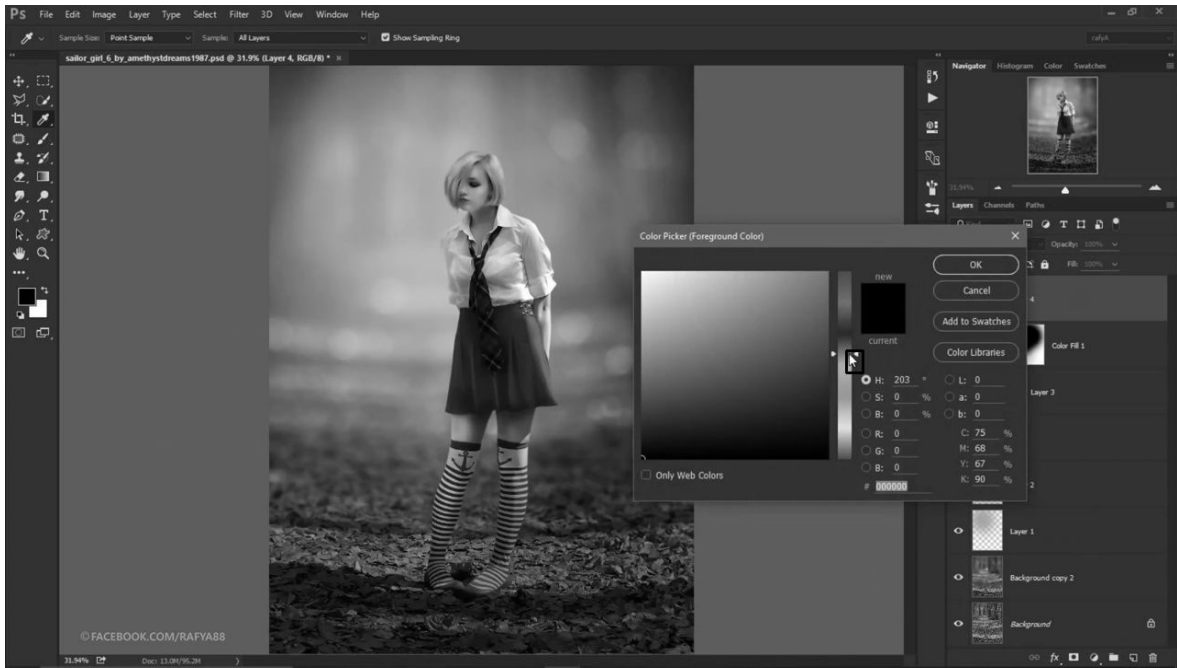


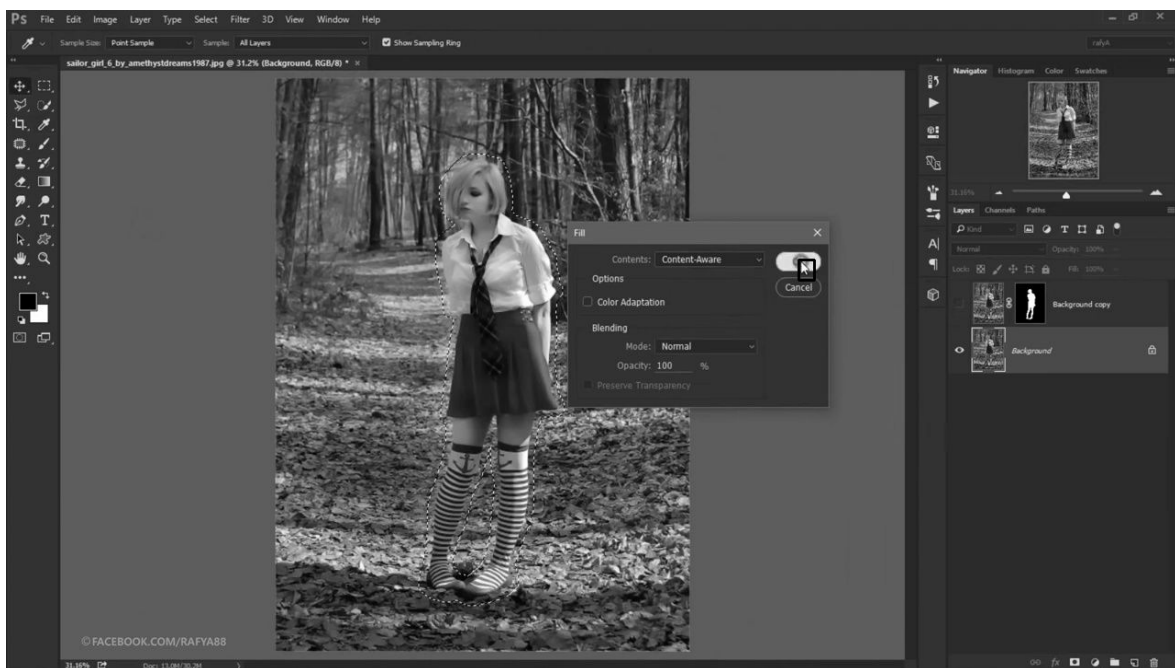












code

```
import cv2
import numpy as np
import glob
import imutils

images = [cv2.imread(file,0) for file in
glob.glob("/home/hariprasath/Desktop/ub/computer_vision_image_processing/task3/*.jpg")]
template =
cv2.imread("/home/hariprasath/Desktop/ub/computer_vision_image_processing/task3/template
/template.png",0)

blur = [cv2.GaussianBlur(image,(3,3),1) for image in images]
```

```

laplacian = [cv2.Laplacian(image,cv2.CV_32F) for image in blur]
w, h = template.shape[:-1]

for i,image in enumerate(laplacian):
    current_max = 0
    gmin_loc=(0,0)
    gv =0
    gh=0
    for scale in np.linspace(0.6, 1.0, 20)[::-1]:
        resized = imutils.resize(template, width = int(template.shape[1] * scale))

        template_gaussian = cv2.GaussianBlur(resized,(3,3),1)
        template_laplacian = cv2.Laplacian(template_gaussian,cv2.CV_32F)

        r = template_laplacian.shape[1] / float(resized.shape[1])
        result=cv2.matchTemplate(image,template_laplacian,cv2.TM_CCOEFF_NORMED)
        min_val, max_val, min_loc, max_loc = cv2.minMaxLoc(result)
        if(current_max < max_val and max_val > 0.45):
            current_max =max_val
            gmin_loc = max_loc
            gv,gh =resized.shape[:-1]
            #cv2.rectangle(images[i], gmin_loc, (gmin_loc[0] + gv, gmin_loc[1] + gh), (255,0,0), 2

    cv2.rectangle(images[i], gmin_loc, (gmin_loc[0] + gv, gmin_loc[1] + gh), (0,0,0), 2)
cv2.imshow("image",images[i])
cv2.waitKey(0)

```