

# **Assignment 1**

**Presented by:**

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## PART 1

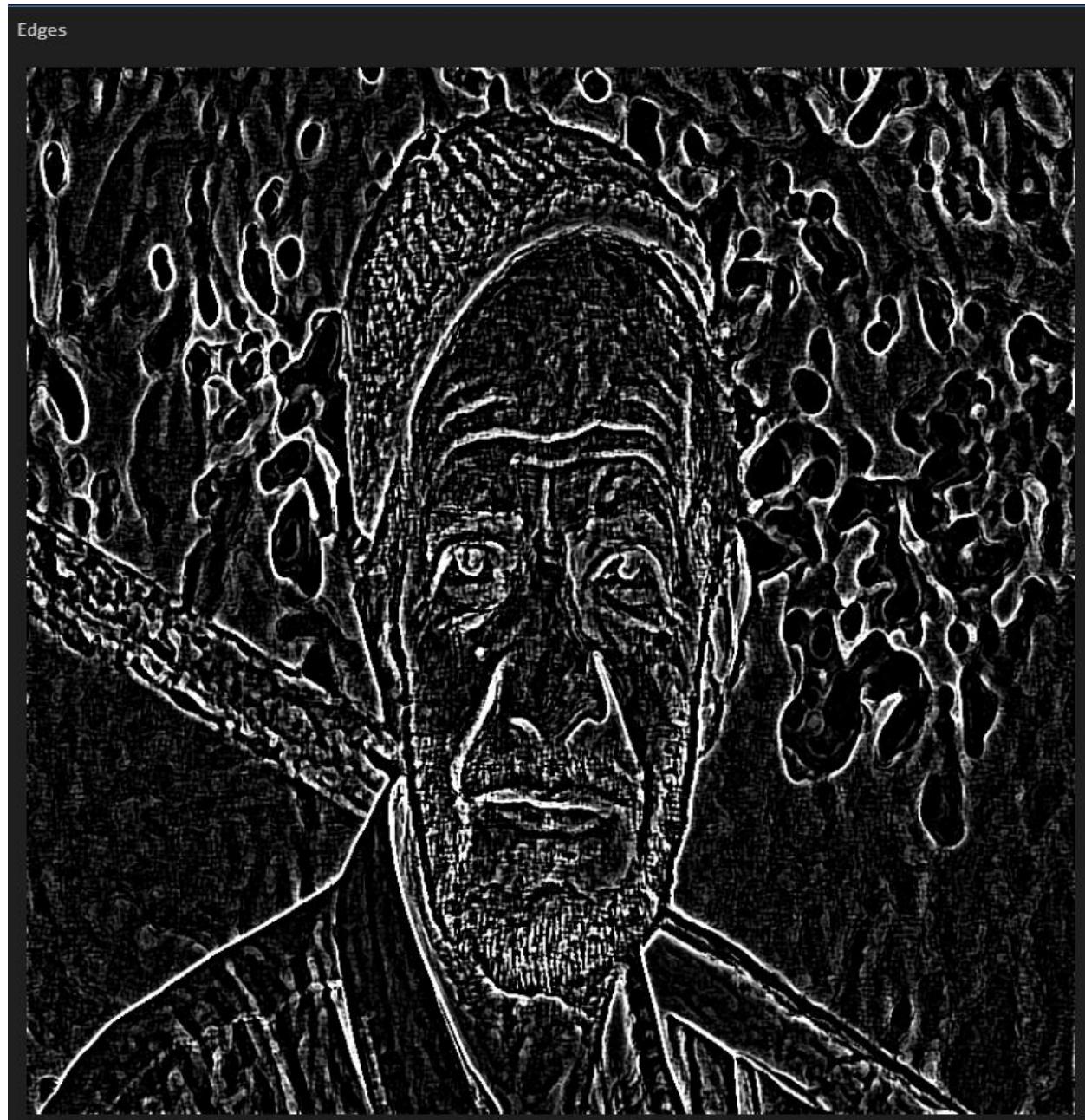
Original Image



second derivative filter which measures how fast the intensity changes in all directions  
strong positive or negative means that there is a strong edge

cv2.CV\_8U tells that the output is 8bit unsigned integer to represent values from 0-255

ksize=5 kernel size is 5 by 5



ret contain the value of the threshold which is equal to 100 here

mask contain the imaage after thresholding

if the value is greater than 100 it convert it to black and if smaller it convert it to white

cv2.THRESH\_BINARY\_INV cause the laplacian return high values for places having strong edge



Final Cartoonified Image



## PART 2

```
# Convert to grayscale and apply Gaussian smoothing
gray = cv2.cvtColor(img_bgr, cv2.COLOR_BGR2GRAY)
kernel_size = 5
sigma = 1.5

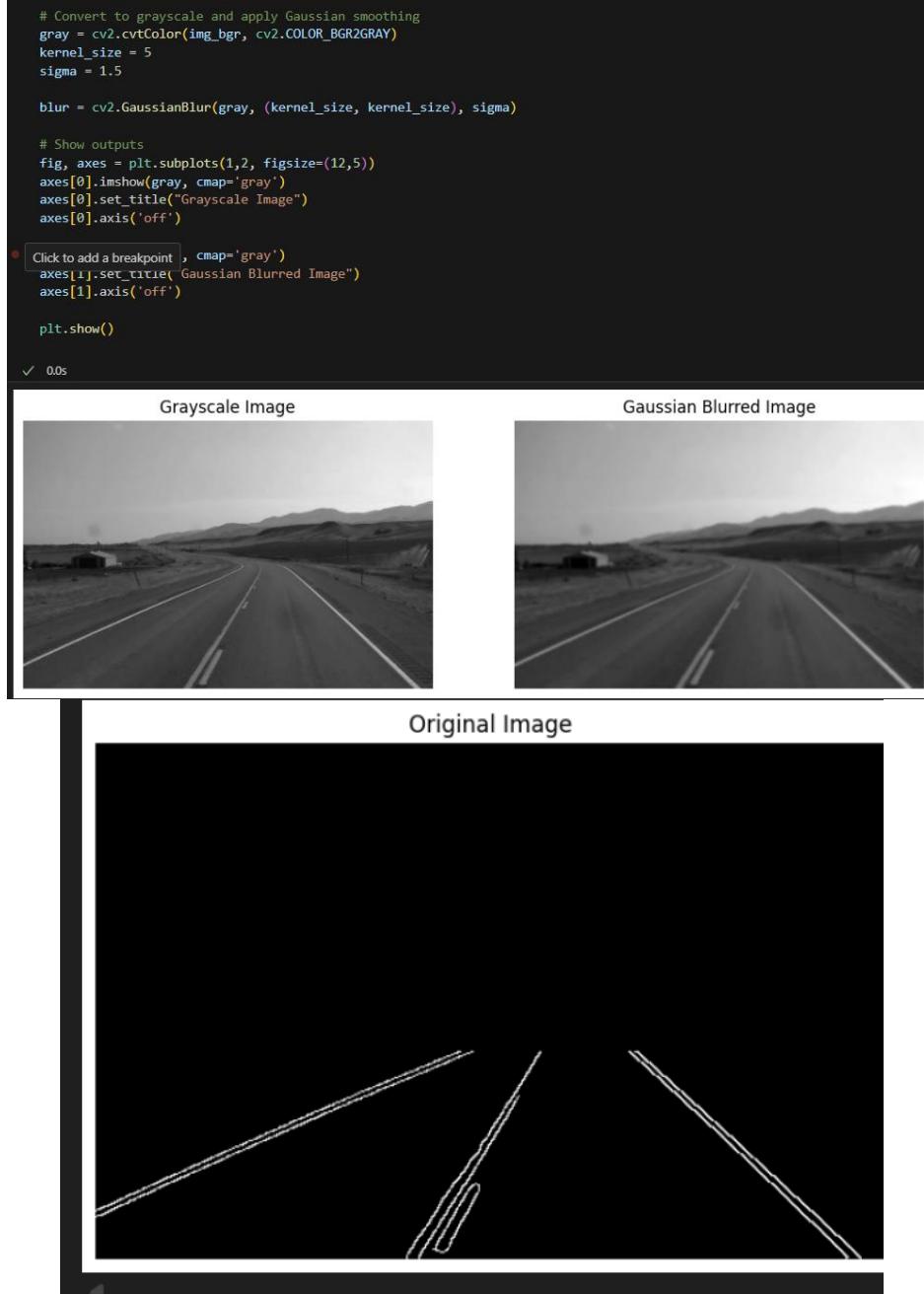
blur = cv2.GaussianBlur(gray, (kernel_size, kernel_size), sigma)

# Show outputs
fig, axes = plt.subplots(1,2, figsize=(12,5))
axes[0].imshow(gray, cmap='gray')
axes[0].set_title("Grayscale Image")
axes[0].axis('off')

axes[1].set_title("Gaussian Blurred Image")
axes[1].axis('off')

plt.show()

✓ 0.0s
```



The screenshot shows a Jupyter Notebook cell with the following code:

```
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✓ 0.0s
```

Below the code, there are three images displayed:

- Grayscale Image**: A grayscale photograph of a road curving through a landscape.
- Gaussian Blurred Image**: The same road image, but with a smooth, blurred appearance due to the Gaussian blur operation.
- Original Image**: A black image with white line segments representing the road's lanes.

```

# Apply Standard Hough Transform (ρ-θ)
rho = 1
theta = np.pi / 180
threshold = 100

lines = cv2.HoughLines(edges_roi, rho, theta, threshold)

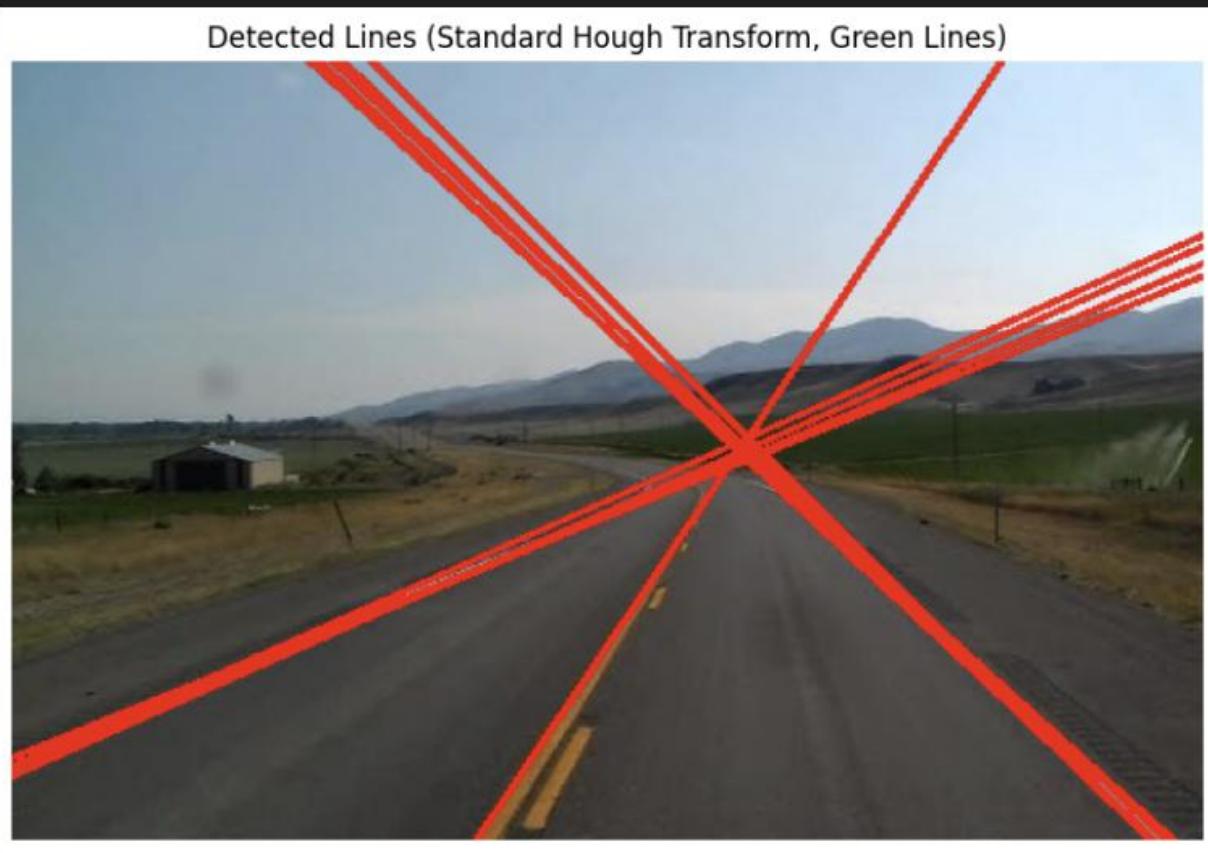
overlay_std = img_rgb.copy()

# Draw detected lines in green
if lines is not None:
    for rho, theta in lines[:,0]:
        a, b = np.cos(theta), np.sin(theta)
        x0, y0 = a * rho, b * rho
        x1, y1 = int(x0 + 800 * (-b)), int(y0 + 800 * (a))
        x2, y2 = int(x0 - 800 * (-b)), int(y0 - 800 * (a))
        cv2.line(overlay_std, (x1, y1), (x2, y2), (255, 0, 0), 2)

plt.figure(figsize=(10,6))
plt.imshow(overlay_std)
plt.title("Detected Lines (Standard Hough Transform, Green Lines)")
plt.axis("off")
plt.show()

```

✓ 0.0s





```
#Probabilistic Hough Transform
minLineLength = 40
maxLineGap = 20
houghP_thresh = 50

linesP = cv2.HoughLinesP(edges_roi, 1, np.pi/180, houghP_thresh, minLineLength=minLineLength, maxLineGap=maxLineGap)
overlay_p = img_rgb.copy()

if linesP is not None:
    for line in linesP:
        x1, y1, x2, y2 = line[0]
        cv2.line(overlay_p, (x1, y1), (x2, y2), (0, 255, 0), 4)

plt.figure(figsize=(10,5))
plt.imshow(overlay_p)
plt.title("Detected Lines (Probabilistic Hough Transform)")
plt.axis("off")
plt.show()
```



