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Computer Vision

Lab 4: Detection of Corner

Lab Task 1 – Harris Corner Detector ---

Implement the Harris Corner Detector by completing the following steps:

(a) Compute the horizontal and vertical image derivatives (I_x, I_y) using Sobel or any suitable operator.

(b) Compute the product of the derivatives:

- $A = I_x^2$
- $B = I_y^2$
- $C = I_x I_y$

(c) Apply Gaussian smoothing (or any suitable blurring method) to each of the product images A, B, C to reduce noise.

(d) Compute the Harris corner response function:



$$R = (AB - C^2) - k(A+B)^2$$

where K is a sensitivity constant (typically 0.04–0.06).

(e) Apply thresholding on R to identify strong corner points and mark them on the original image.

Code

```
def convolution2d(image, kernel):

    image_h, image_w = image.shape

    kernel_h, kernel_w = kernel.shape

    padding_h = (kernel_h - 1) // 2

    padding_w = (kernel_w - 1) // 2

    padded_image = np.pad(image, ((padding_h, padding_h), (padding_w, padding_w)), mode='constant',
    constant_values=0)

    output = np.zeros((image_h, image_w))

    for i in range(image_h):

        for j in range(image_w):
```



```
new = padded_image[i:i+kernel_h, j:j+kernel_w]

output[i, j] = np.sum(new * kernel)

return output

image = np.random.rand(8, 4)

image = cv2.imread('chess.png', cv2.IMREAD_GRAYSCALE)

def guassian_filter2d(sigma):

    size = int(2 * (np.pi * sigma))

    if size % 2 == 0:

        size += 1

    kernel = np.zeros((size, size))

    for x in range(size):

        for y in range(size):

            kernel[x, y] = (1/(2 * np.pi * sigma**2)) * np.exp(-(x - size//2)**2 + (y - size//2)**2) / (2 * sigma**2)

    return kernel

def harris_corner_detector(img, k):

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



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```
sobel_y = np.array([[-1, -2, -1], [0, 0, 0], [1, 2, 1]])  
  
a = convolution2d(img, sobel_x)  
  
b = convolution2d(img, sobel_y)  
  
A = a * a  
  
B = b * b  
  
C = a * b  
  
guass_value = guassian_filter2d(1.5)  
  
guass_A = convolution2d(A, guass_value)  
  
guass_B = convolution2d(B, guass_value)  
  
guass_C = convolution2d(C, guass_value)  
  
# Harris response  
  
R = (guass_A * guass_B - guass_C ** 2) - k * (guass_A + guass_B) ** 2
```



```
# Threshold
```

```
corners = np.zeros_like(R)
```

```
threshold = 0.50 * np.max(R)
```

```
corners[R > threshold] = 1
```

```
return corners, R
```

```
import matplotlib.pyplot as plt
```

```
corners, R = harris_corner_detector(image, k=0.04)
```

```
plt.figure(figsize=(8, 8))
```

```
plt.imshow(image, cmap='gray')
```

```
y, x = np.where(corners == 1)
```

```
plt.scatter(x, y, c='b', s=10)
```

```
plt.title('Harris Corners')
```

```
plt.axis('off')
```

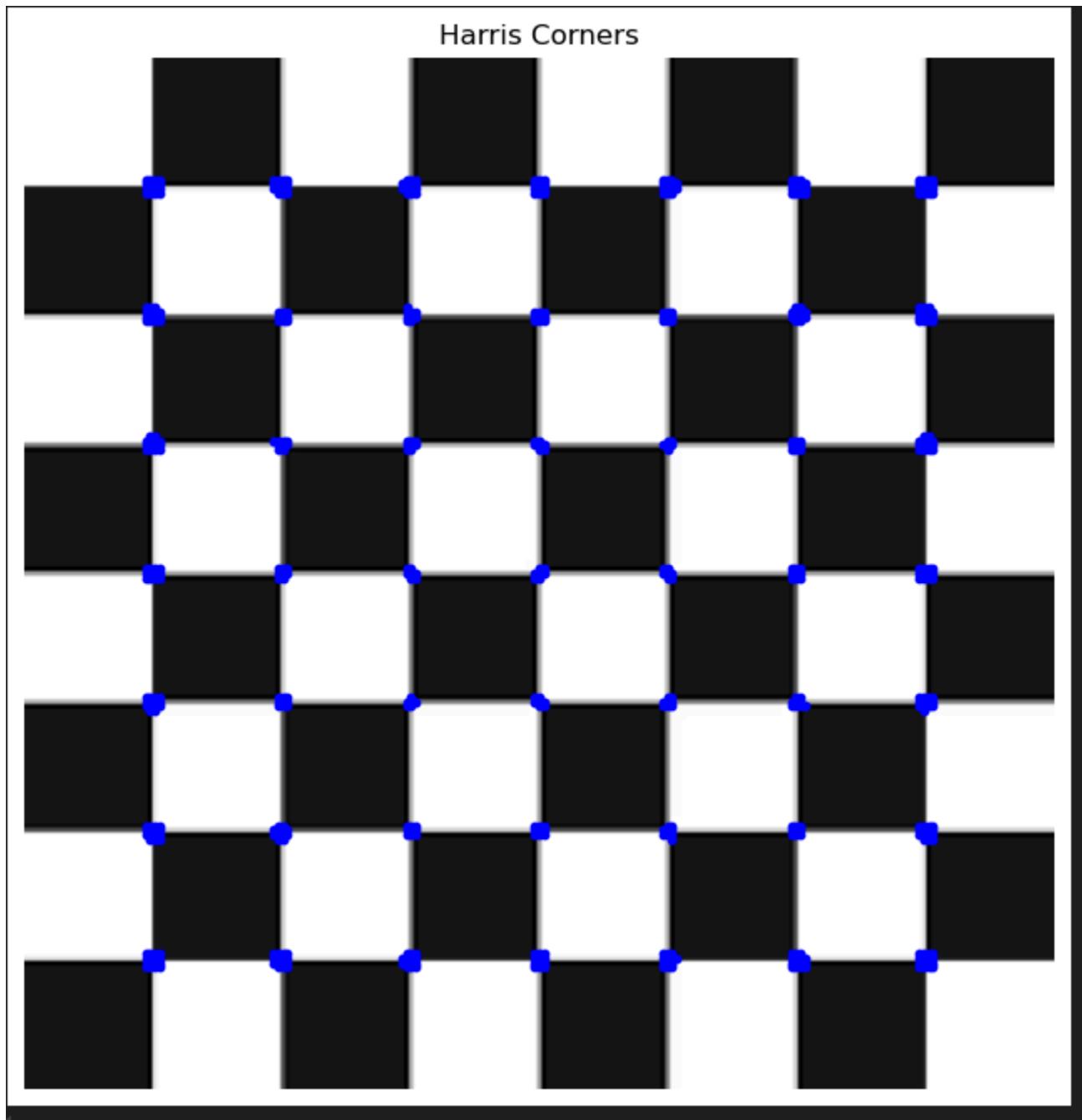
```
plt.show()
```



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Output :





Built-In Harris

```
b_corners = cv2.cornerHarris(image,3,3,0.04)
threshold = 0.30 * b_corners.max()
y, x = np.where(b_corners > threshold)
plt.figure(figsize=(8, 8))
plt.imshow(image, cmap='gray')
plt.scatter(x, y, c='b', s=10)
plt.title('Harris Corners (OpenCV) ')
plt.axis('off')
plt.show()
```

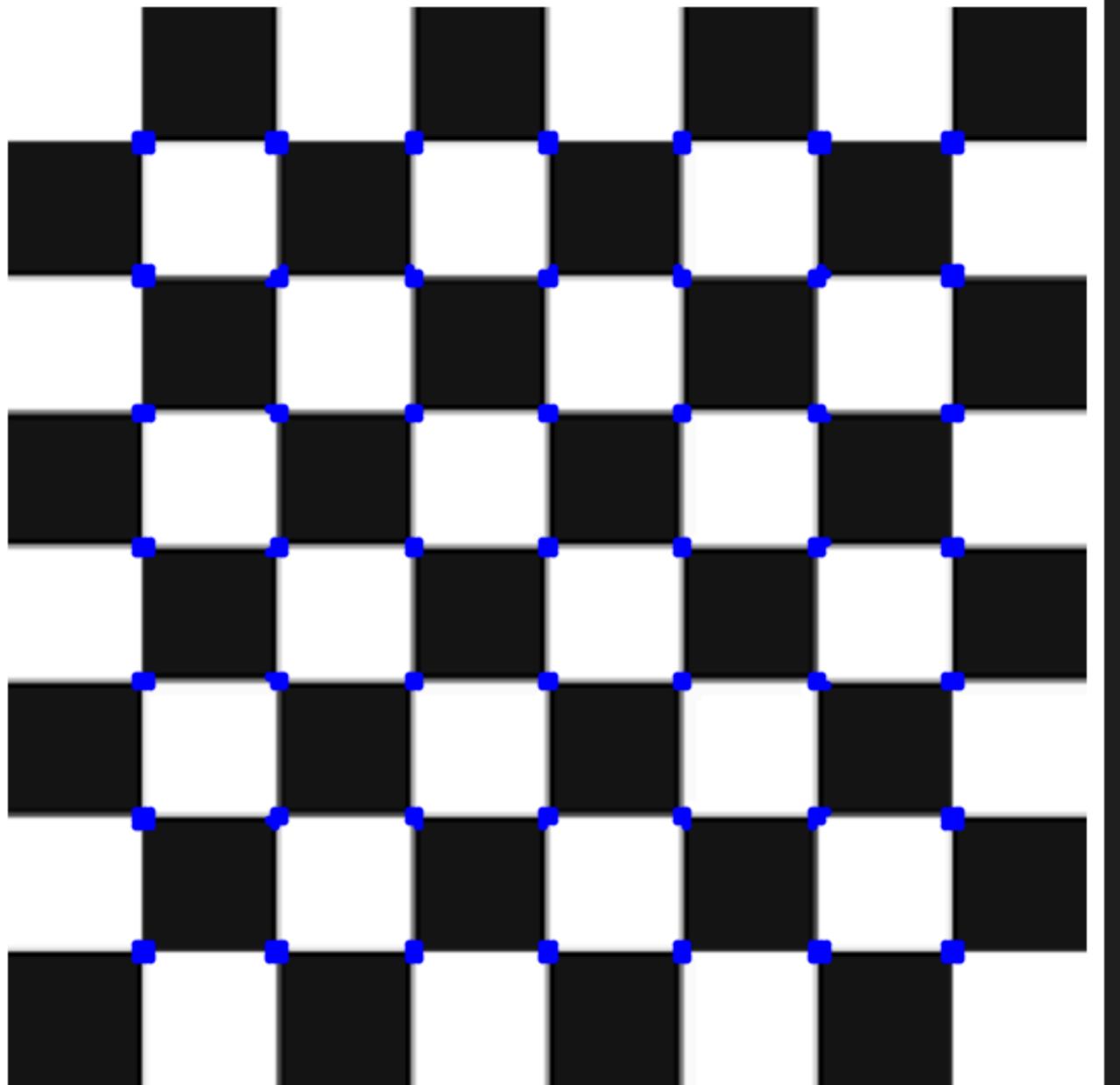
Output:



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Harris Corners (OpenCV)



Lab Task 2 -Shi Tomasi



Implement the **Shi–Tomasi corner detector** (also known as *Good Features to Track*) and compare the detected corners with those obtained using the Harris method. Explain why Shi–Tomasi often provides more stable corner points.

Code:

```
def shiTomaschi(img):
    sobel_x = np.array([[-1, 0, 1], [-2, 0, 2], [-1, 0, 1]])
    sobel_y = np.array([[-1, -2, -1], [0, 0, 0], [1, 2, 1]])
    a = convolution2d(img, sobel_x)
    b = convolution2d(img, sobel_y)
    A = a * a
    B = b * b
    C = a * b
    guass_value = gaussian_filter2d(1.5)
    guass_A = convolution2d(A, guass_value)
    guass_B = convolution2d(B, guass_value)
    guass_C = convolution2d(C, guass_value)

    # Shi-Tomasi (minimum eigenvalue)
    R = 0.5 * ((guass_A + guass_B) - np.sqrt((guass_A + guass_B) ** 2 - 4 *
(guass_A * guass_B - guass_C ** 2) + 1e-10))

    # Threshold
    corners = np.zeros_like(R)
    threshold = 0.60 * np.max(R)
    corners[R > threshold] = 1
    return corners, R

corners, R = shiTomaschi(image)

y, x = np.where(corners == 1)
plt.figure(figsize=(8, 8))
plt.imshow(image, cmap='gray')
plt.scatter(x, y, c='b', s=10)
plt.title('Shi-Tomasi Corner Detection')
plt.axis('off')
plt.show()
```



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Output:

Shi-Tomasi Corner Detection

