



National University
Of Computer and Emerging Sciences

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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):
```



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```
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 guassain_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]])
```



```
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)

# 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)

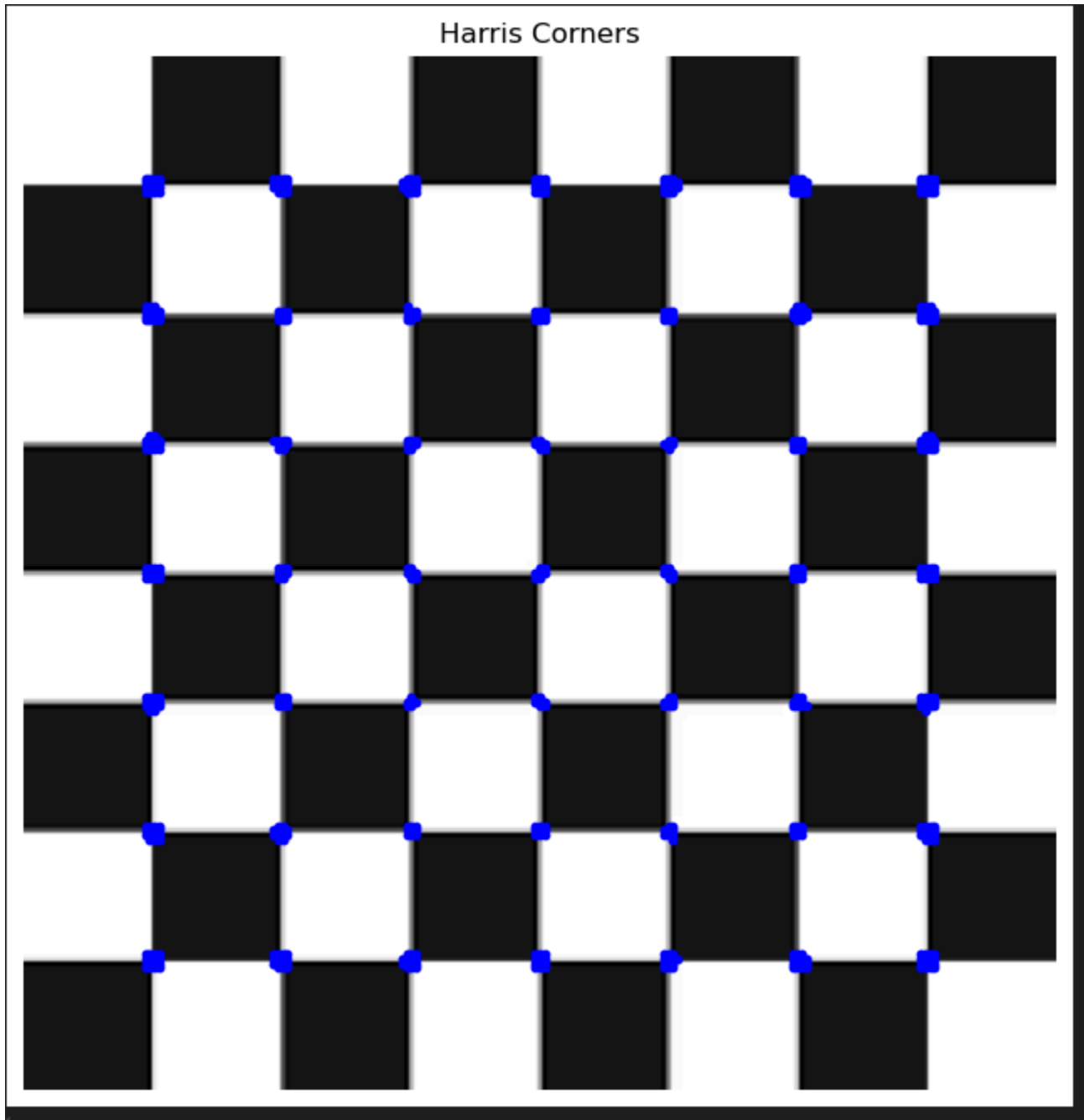
plt.title('Harris Corners')

plt.axis('off')

plt.show()
```



Output :





Built-In Harris

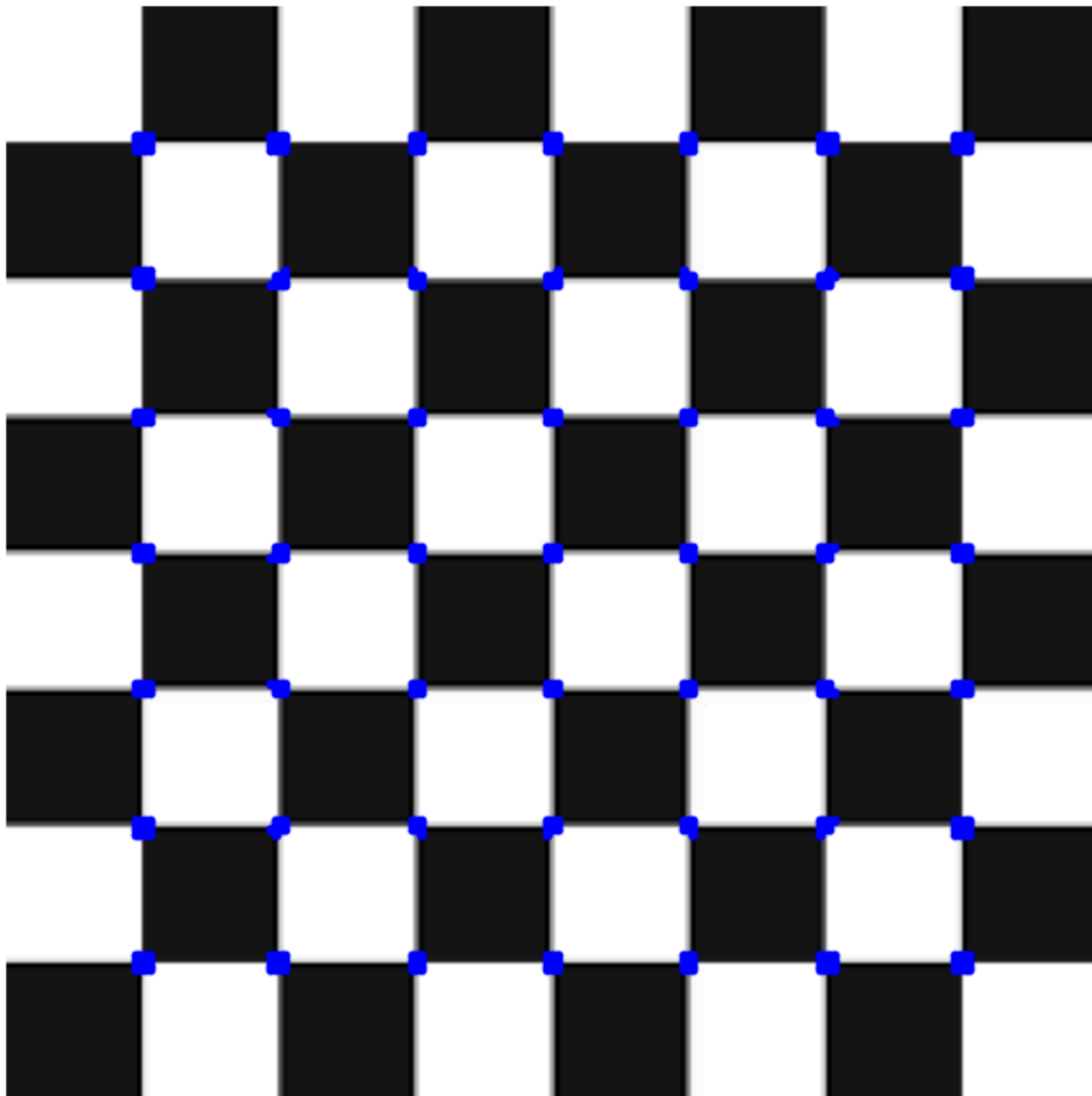
```
b_corners = cv2.cornerHarris(image,3,3,0.04)
threshold = 0.30 * b_corners.max()
y, x = np.where(b_corners > threshold)
plt.figure(figsize=(8, 8))
plt.imshow(image, cmap='gray')
plt.scatter(x, y, c='b', s=10)
plt.title('Harris Corners (OpenCV)')
plt.axis('off')
plt.show()
```

Output:



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Harris Corners (OpenCV)



Lab Task 2 -Shi Tomasi_____

Computer Vision



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:

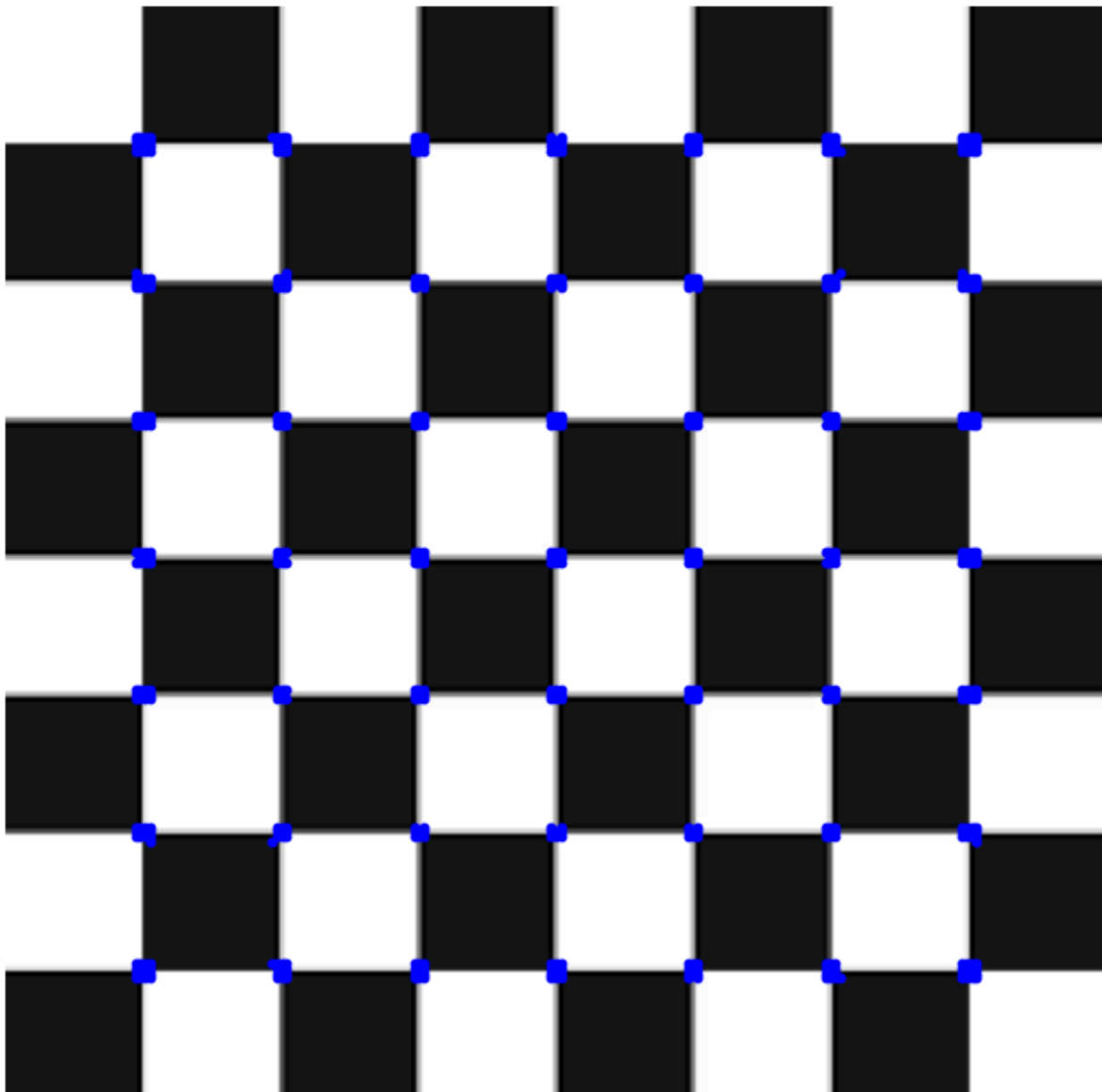
```
def shiTomashi(img):  
    sobel_x = np.array([[ -1,  0,  1], [-2,  0,  2], [-1,  0,  1]])  
    sobel_y = np.array([[ -1, -2, -1], [ 0,  0,  0], [ 1,  2,  1]])  
    a = convolution2d(img, sobel_x)  
    b = convolution2d(img, sobel_y)  
    A = a * a  
    B = b * b  
    C = a * b  
  
    guass_value = 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 = shiTomashi(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



Computer Vision