

Image Reconstruction

COMPUTER VISIONS

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Course Code

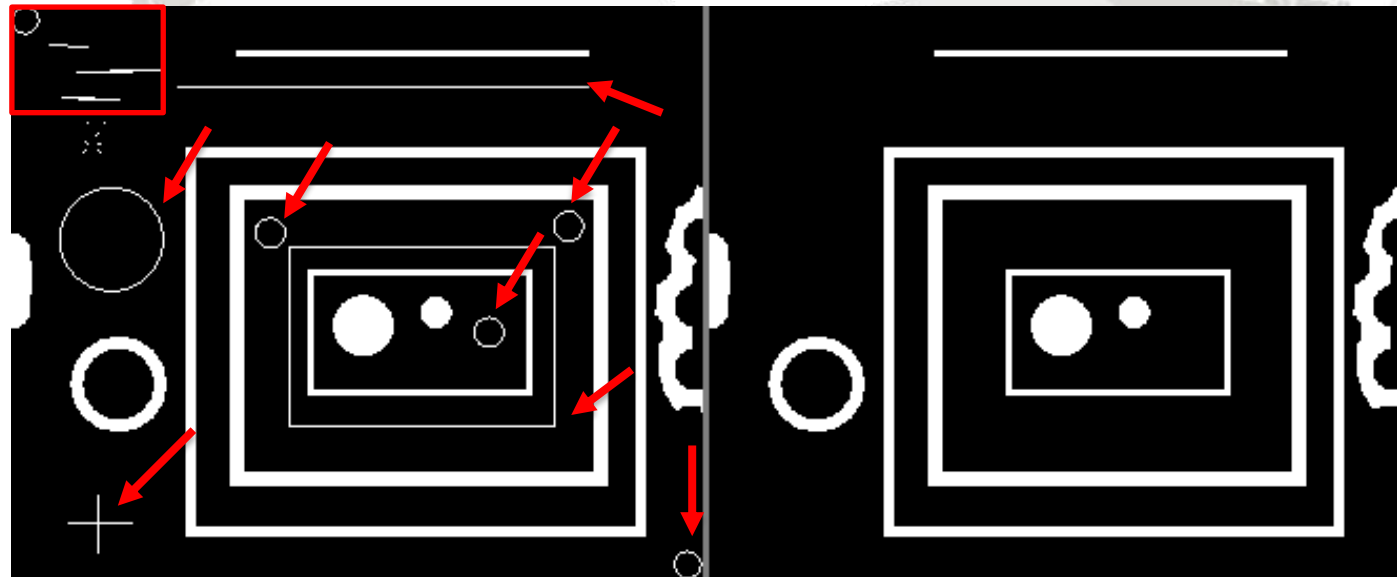
Overview

1. Morphological reconstruction
2. Hough Line Transforms
3. Hough Circle Transforms

1. Morphological Theory

Morphology is a broad set of image processing operations that process images based on shapes. **Morphological operations (kernel)** apply a **structuring element** to an input image (gray image), creating an output image of the same size.

Example Applied the opening operator



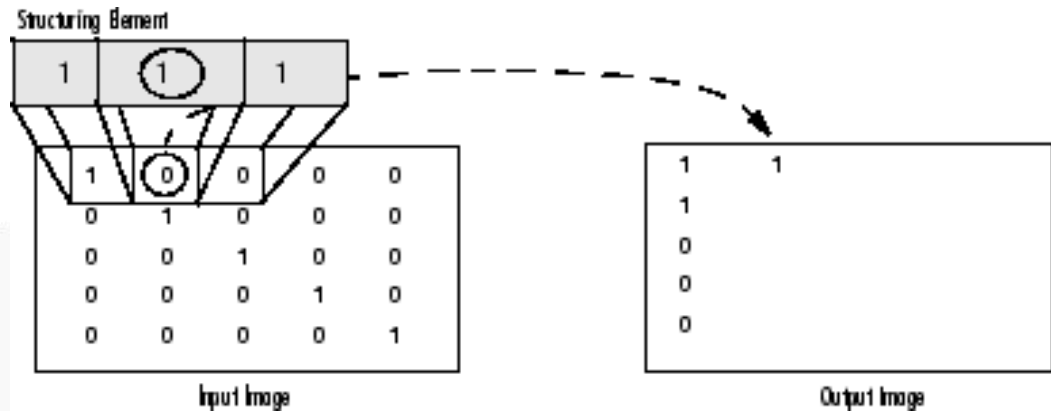
input image

output image

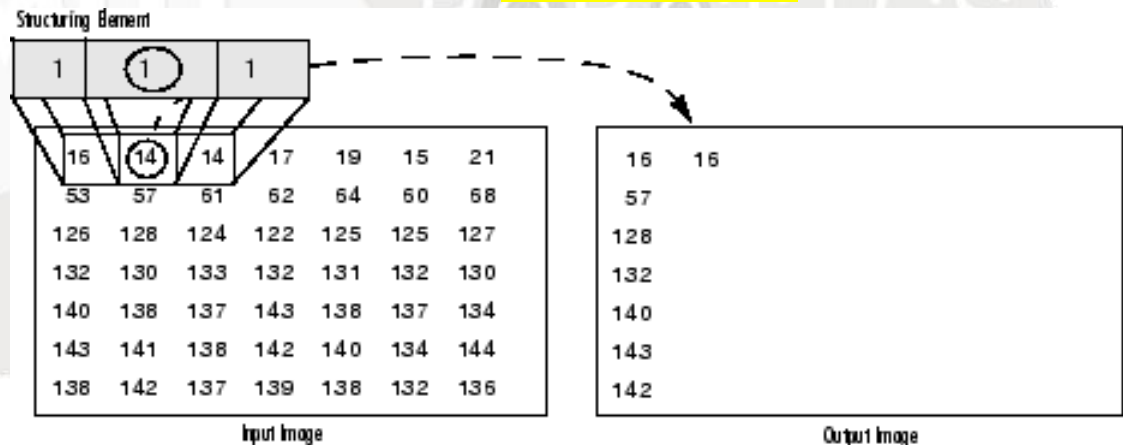
1. Morphological Theory

1. Dilation
2. Erosion
3. Opening
4. Closing
5. Gradient
6. Top Hat
7. Black Hat

Morphological Dilation of a Binary Image



Morphological Dilation of a Grayscale Image



1. Dilation (increase border)

A pixel element is “1” if **at least one pixel** under the kernel is “1”.

So it **increases** the white region in the image or size of foreground object increases.

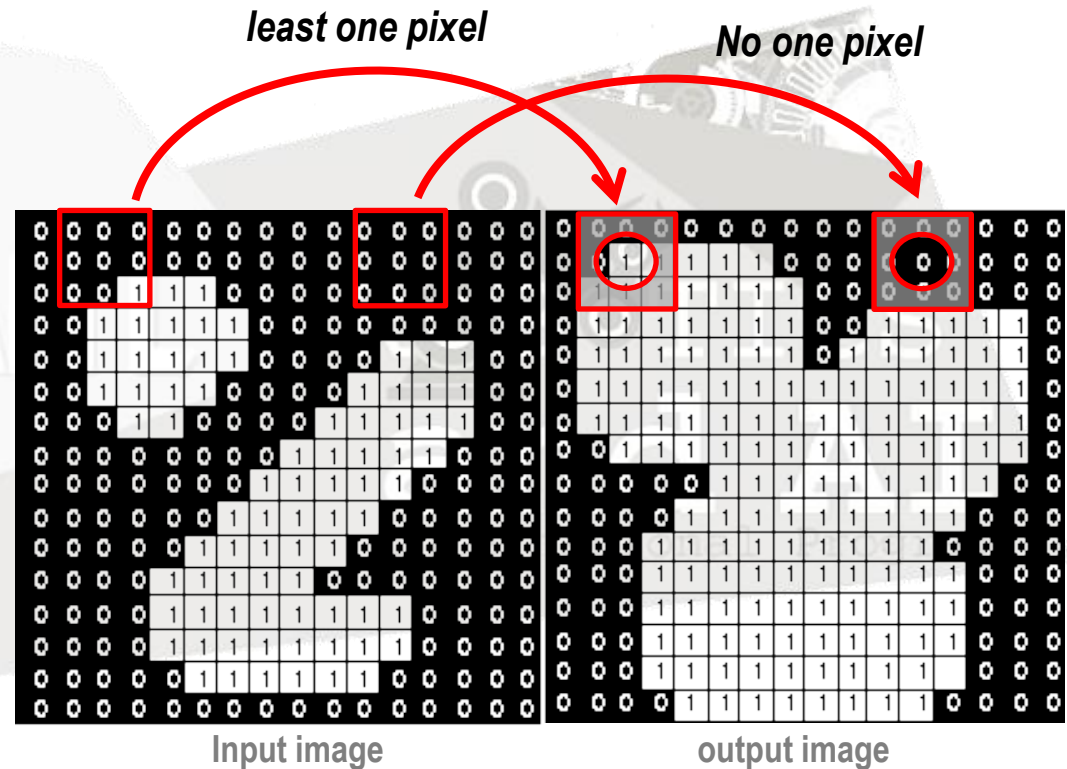
Dilation

kernel size [3x3]

1	1	1
1	1	1
1	1	1



Example dilation



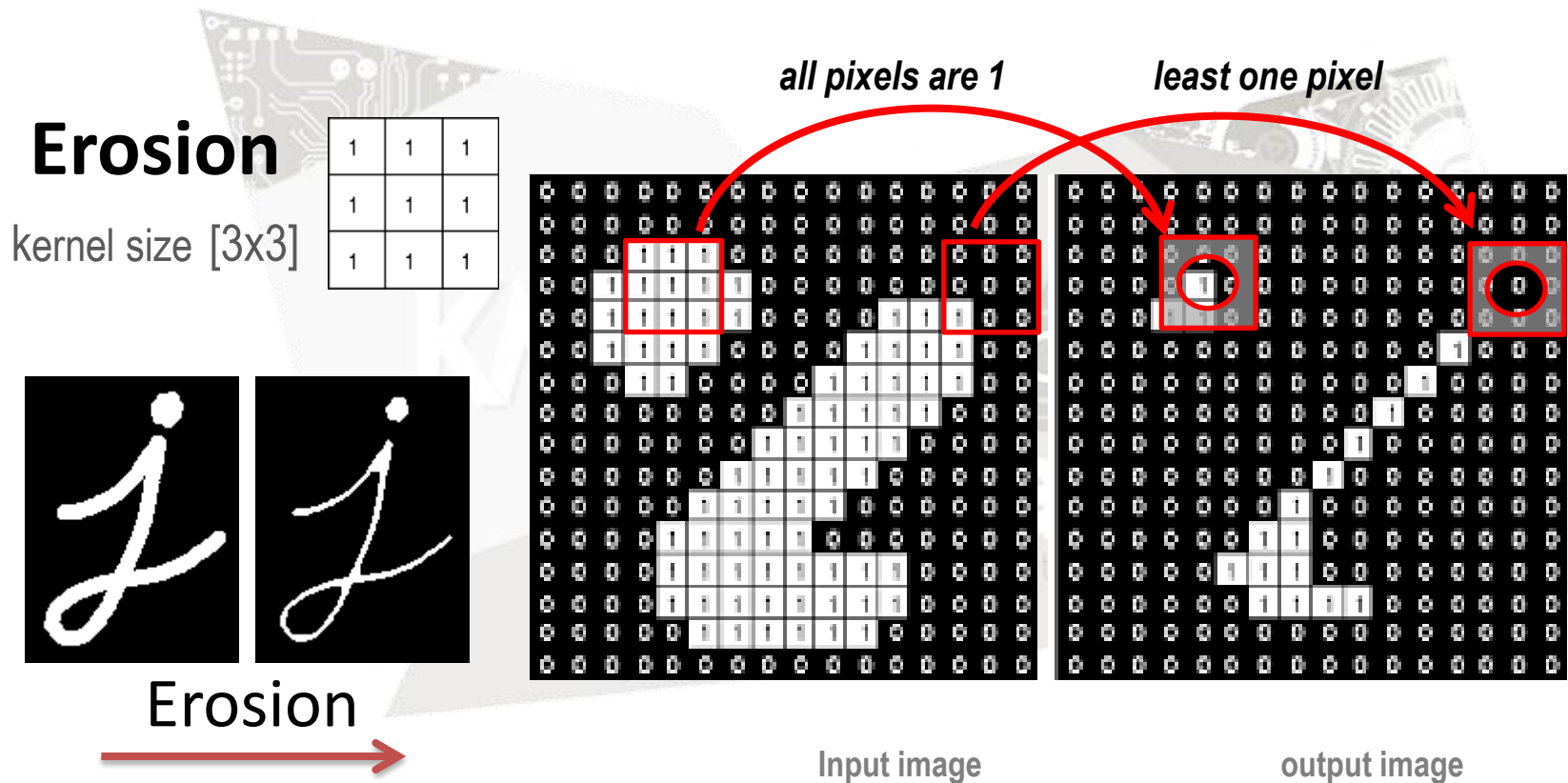
1. Dilation (increase border)



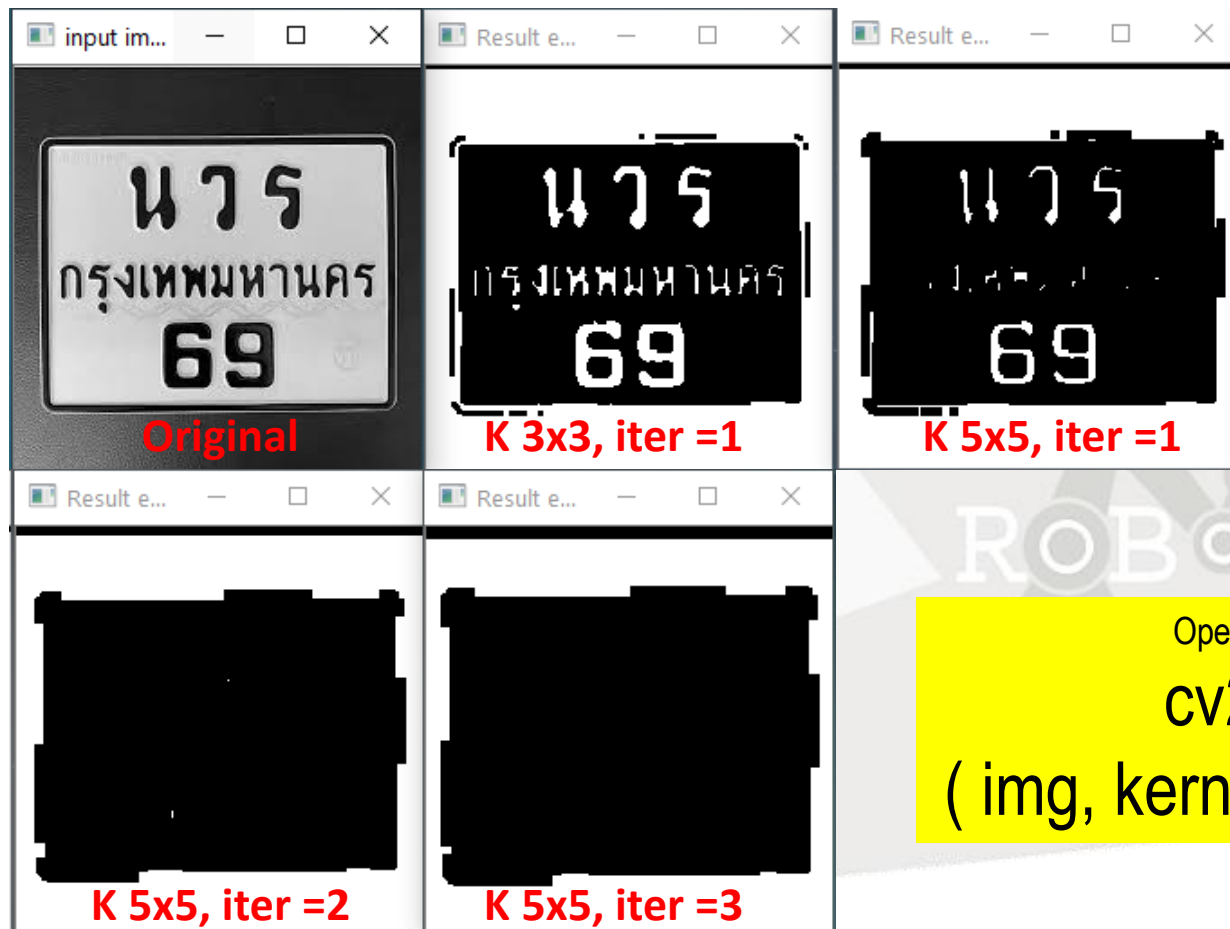
OpenCV function :
`cv2.dilate`
(img, kernel, iterations=1)

2. Erosion (decrease border)

The kernel slides through the image (as in 2D convolution). A pixel in the original image (either 1 or 0) will be considered 1 only **if all the pixels** under the kernel is 1, otherwise it is eroded (made to zero).



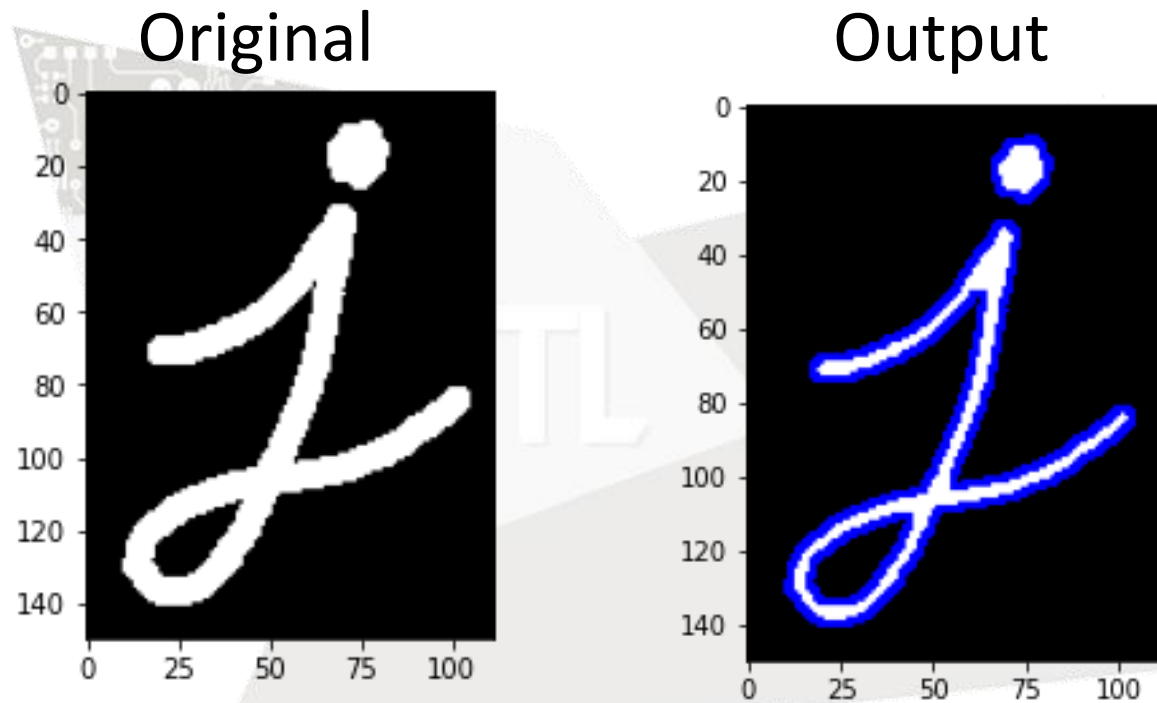
2. Erosion (decrease border)



OpenCV function :
cv2.erode
(img, kernel, iterations=1)

Quiz#1 (15 min)

1. Let's show the blue border inside object.



***** Border size 2 pixels**

3. Opening (reduce noise)

Opening is just another name of **erosion followed by dilation**

Opening

kernel size [3x3]

1	1	1
1	1	1
1	1	1

OpenCV function :

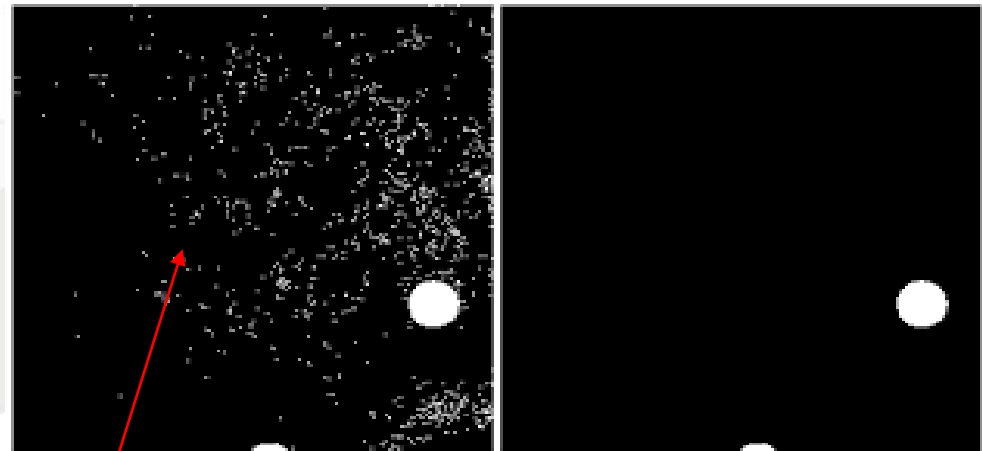
`cv2.morphologyEx()`
`cv2.MORPH_OPEN`

Example Opening



Input image

output image



noise

4. Closing (Fill holes)

Closing is reverse of Opening, **Dilation followed by Erosion**. It is useful in closing small holes inside the foreground objects, or small black points on the object.

Closing

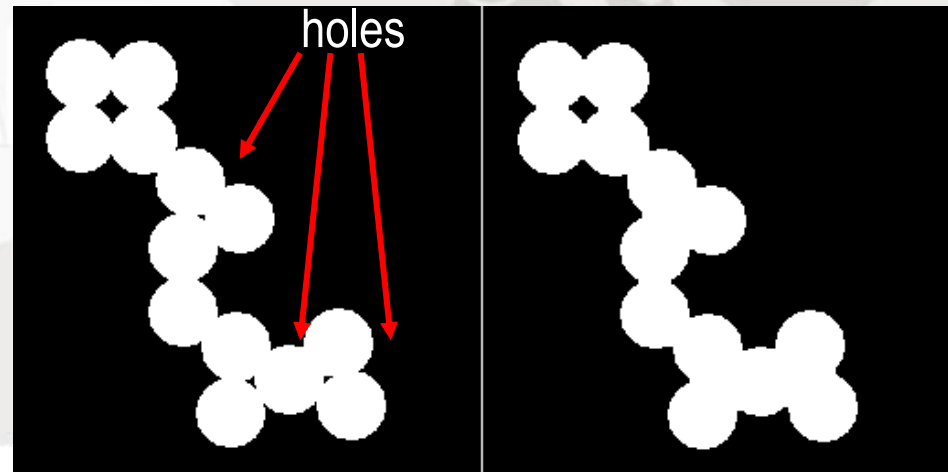
kernel size [3x3]

1	1	1
1	1	1
1	1	1



OpenCV function :

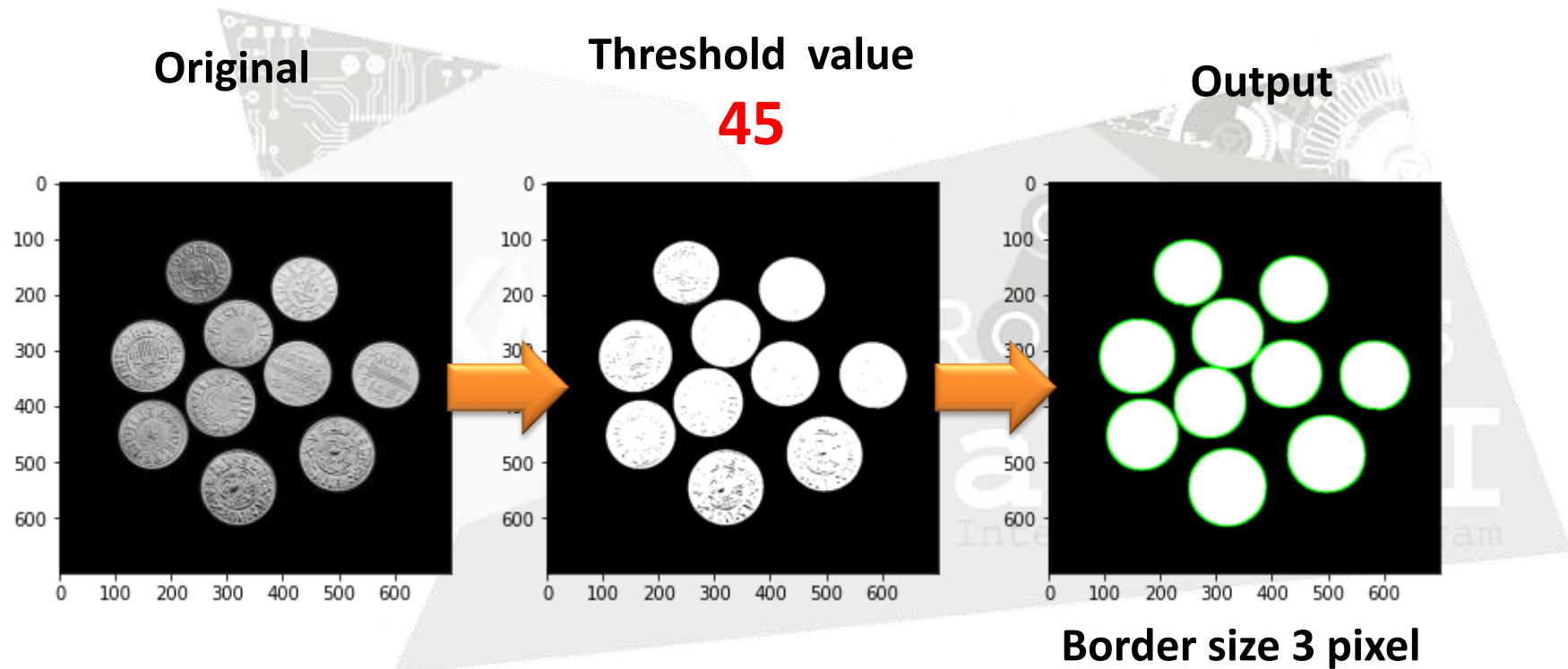
`cv2.morphologyEx()`
`cv2.MORPH_CLOSE`



Example Closing

Quiz#2 (15 min)

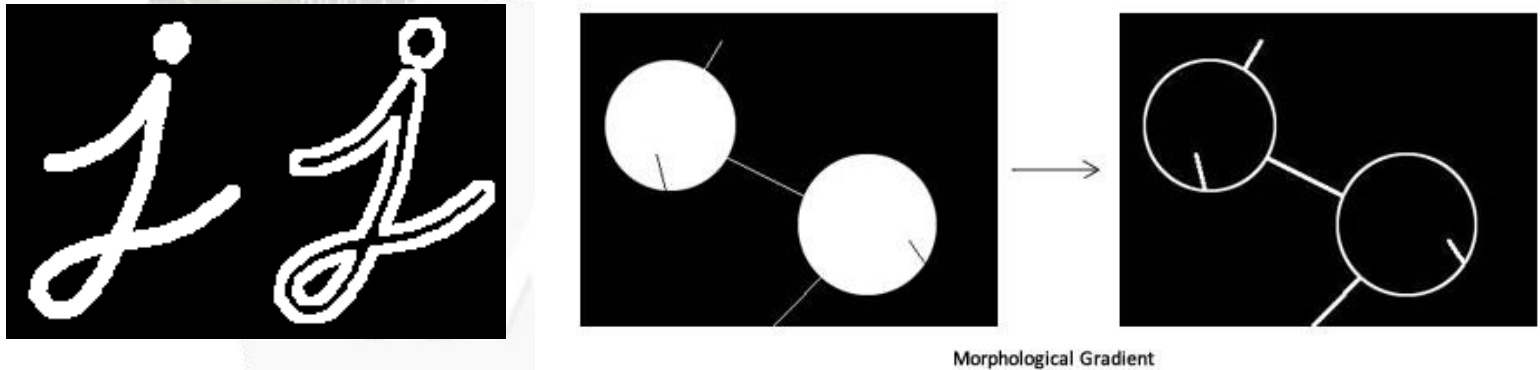
1. Let's show the code for draw the green border outside object.



5. Morphological Gradient

It is the **difference** between **dilation** and **erosion** of an image.

Below example is done for a **9x9 kernel**

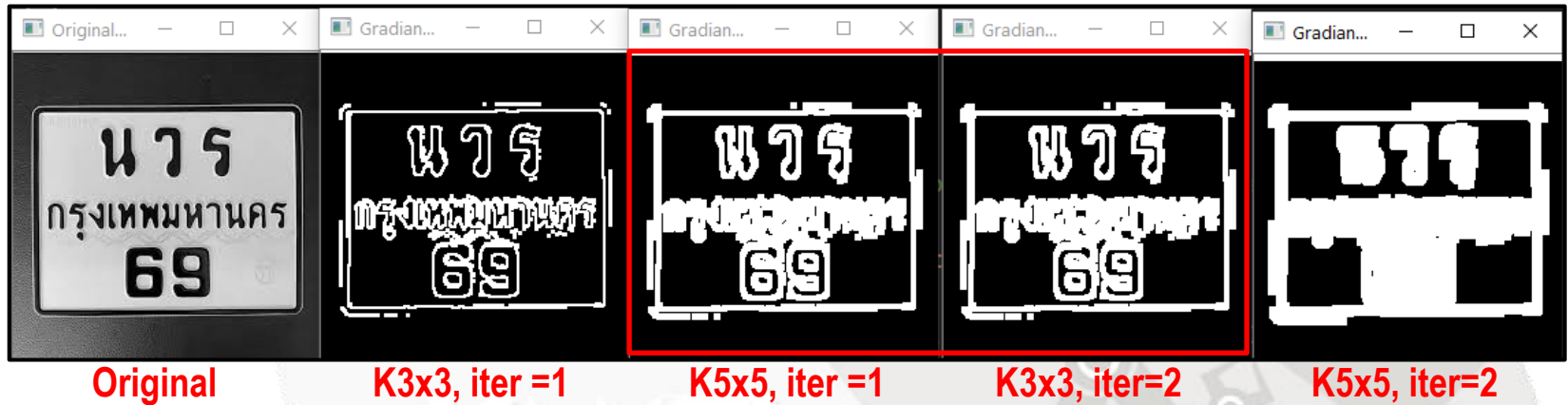


OpenCV function :

```
cv2.morphologyEx()  
cv2.MORPH_CLOSE
```

5. Morphological Gradient

Example Gradient



```
img = cv2.imread("dataset/69.jpg",0)
# threshold
_,bw = cv2.threshold(img,180,255,cv2.THRESH_BINARY)
gra1 = cv2.morphologyEx(bw,cv2.MORPH_GRADIENT,np.ones((3,3),np.uint8))
gra2 = cv2.morphologyEx(bw,cv2.MORPH_GRADIENT,np.ones((5,5),np.uint8))
gra3 = cv2.morphologyEx(bw,cv2.MORPH_GRADIENT,np.ones((3,3),np.uint8),iterations=2)
gra4 = cv2.morphologyEx(bw,cv2.MORPH_GRADIENT,np.ones((5,5),np.uint8),iterations=2)
```

6. Top Hat (noise object)

It is the difference between input image and Opening of the image..

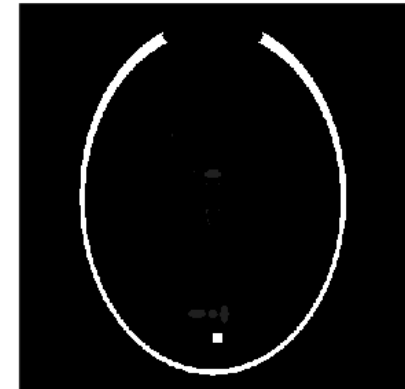
Example is done for a **9x9 kernel**



original



white tophat



OpenCV function :

```
cv2.morphologyEx()  
cv2.MORPH_TOPHAT
```

6. Top Hat (noise object)

Example Tophat



```
img = cv2.imread("dataset/69.jpg",0)
# tophat = input - closing
_,bw = cv2.threshold(img,150,255,cv2.THRESH_BINARY)
clo1 = cv2.morphologyEx(bw,cv2.MORPH_OPEN,np.ones((3,3),np.uint8),iterations=1)

tophat = cv2.subtract(bw,clo1)
tophat2 = cv2.morphologyEx(bw,cv2.MORPH_TOPHAT,np.ones((3,3),np.uint8),iterations=1)
```

7. Black Hat (Hole object)

It is the **difference** between the closing of the input image and input image.

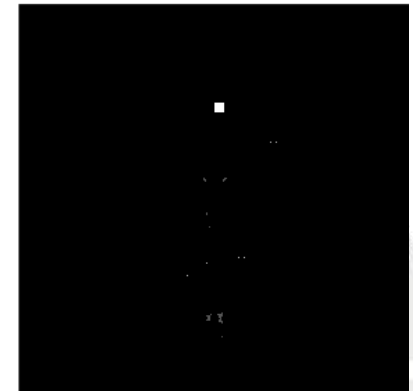
Example is done for a **9x9 kernel**



original



black tophat



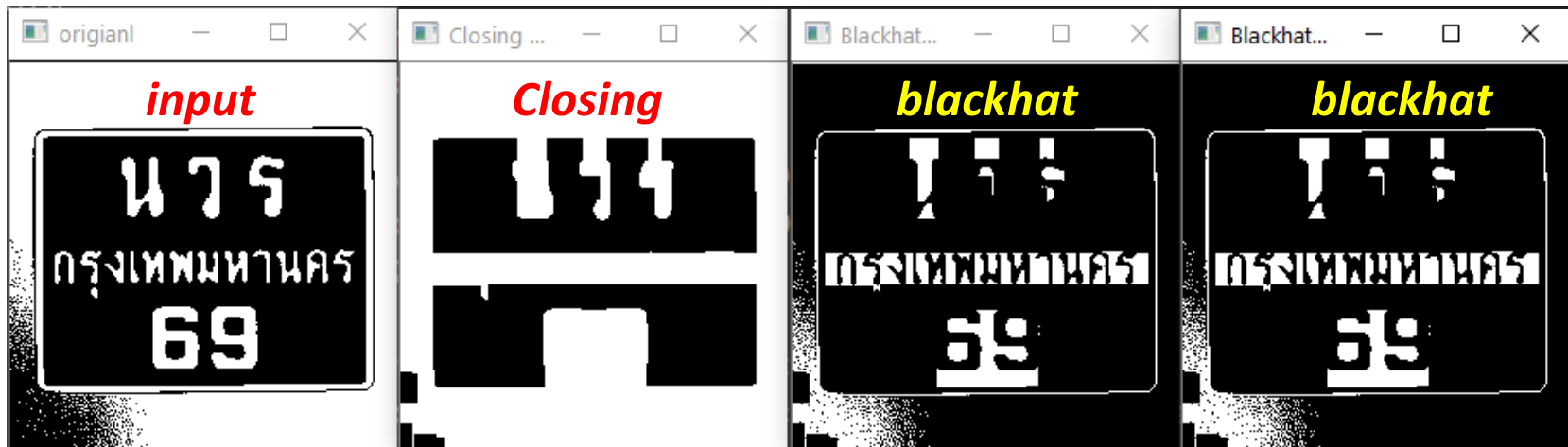
OpenCV function :

```
cv2.morphologyEx()  
cv2.MORPH_BLACKHAT
```


Example Blackhat

Closing-Input

morphologyEx



```
img = cv2.imread("dataset/69.jpg",0)
_,bw = cv2.threshold(img,100,255,cv2.THRESH_BINARY_INV)

# blackhat = closing - input
cls1 = cv2.morphologyEx(bw,cv2.MORPH_CLOSE,np.ones((5,5),np.uint8),iterations=3)
blk1 = cv2.subtract(cls1,bw)
blk2 = cv2.morphologyEx(bw,cv2.MORPH_BLACKHAT,np.ones((5,5),np.uint8),iterations=3)
```

II. Structuring Element

Rectangular Kernel

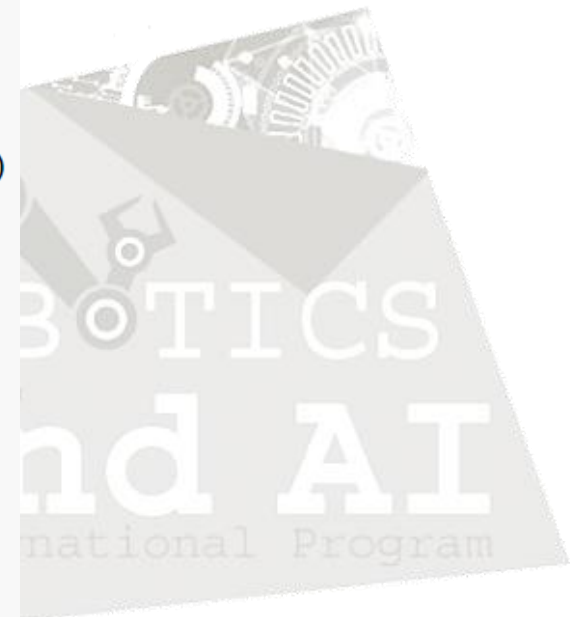
```
>>> cv2.getStructuringElement(cv2.MORPH_RECT,(5,5))
array([[1, 1, 1, 1, 1],
       [1, 1, 1, 1, 1],
       [1, 1, 1, 1, 1],
       [1, 1, 1, 1, 1],
       [1, 1, 1, 1, 1]], dtype=uint8)
```

Elliptical Kernel

```
>>> cv2.getStructuringElement(cv2.MORPH_ELLIPSE,(5,5))
array([[0, 0, 1, 0, 0],
       [1, 1, 1, 1, 1],
       [1, 1, 1, 1, 1],
       [1, 1, 1, 1, 1],
       [0, 0, 1, 0, 0]], dtype=uint8)
```

Cross-shaped Kernel

```
>>> cv2.getStructuringElement(cv2.MORPH_CROSS,(5,5))
array([[0, 0, 1, 0, 0],
       [0, 0, 1, 0, 0],
       [1, 1, 1, 1, 1],
       [0, 0, 1, 0, 0],
       [0, 0, 1, 0, 0]], dtype=uint8)
```



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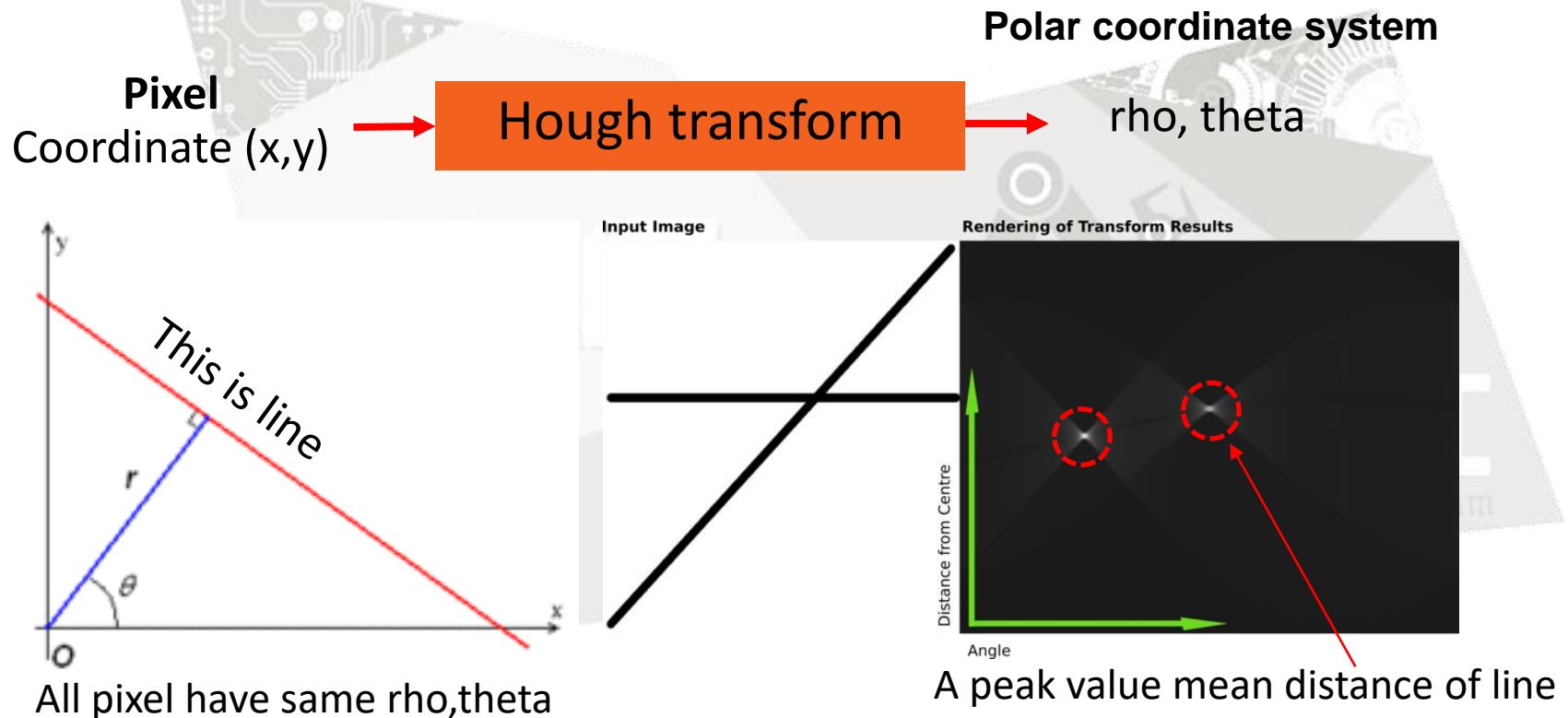
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2. Hough Line Transforms

Hough Line Transforms

The **Hough transform** is a feature extraction technique used in image analysis, computer vision, and digital image processing. The purpose of the technique is to find imperfect instances of objects within a certain class of shapes by a voting procedure.



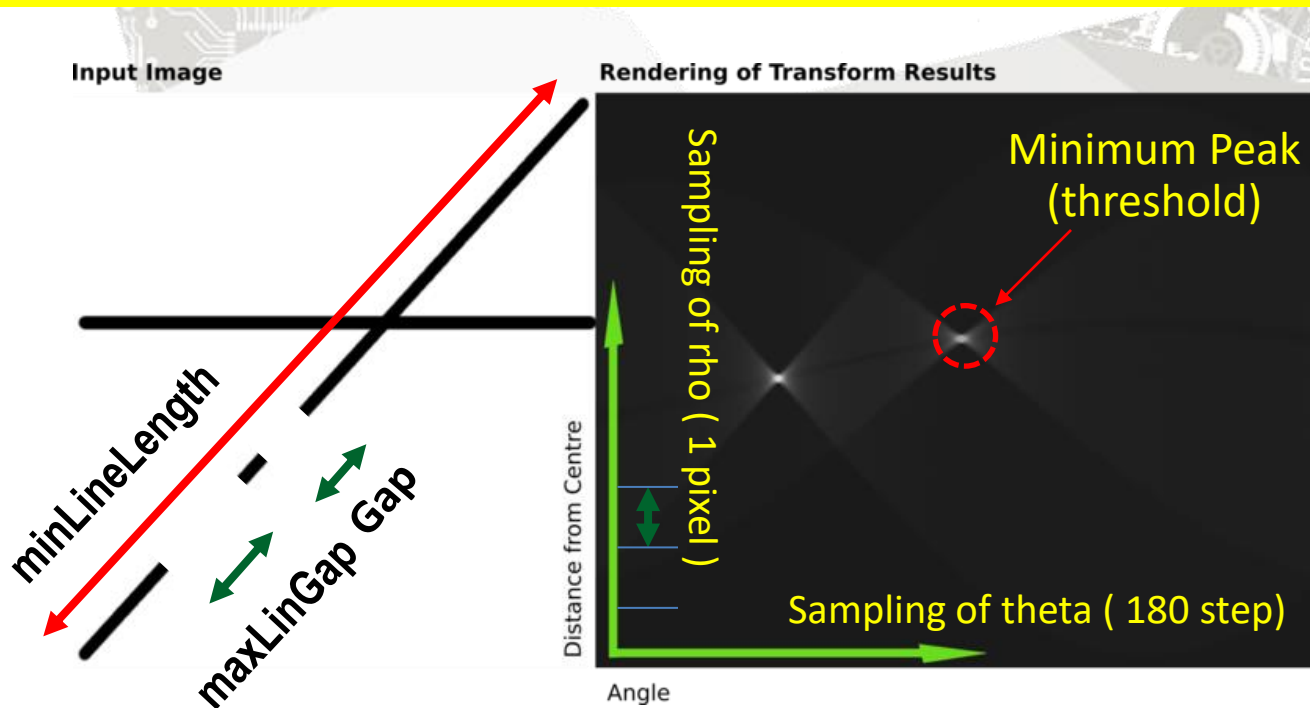
Hough Line Transforms

Hough line Parameters

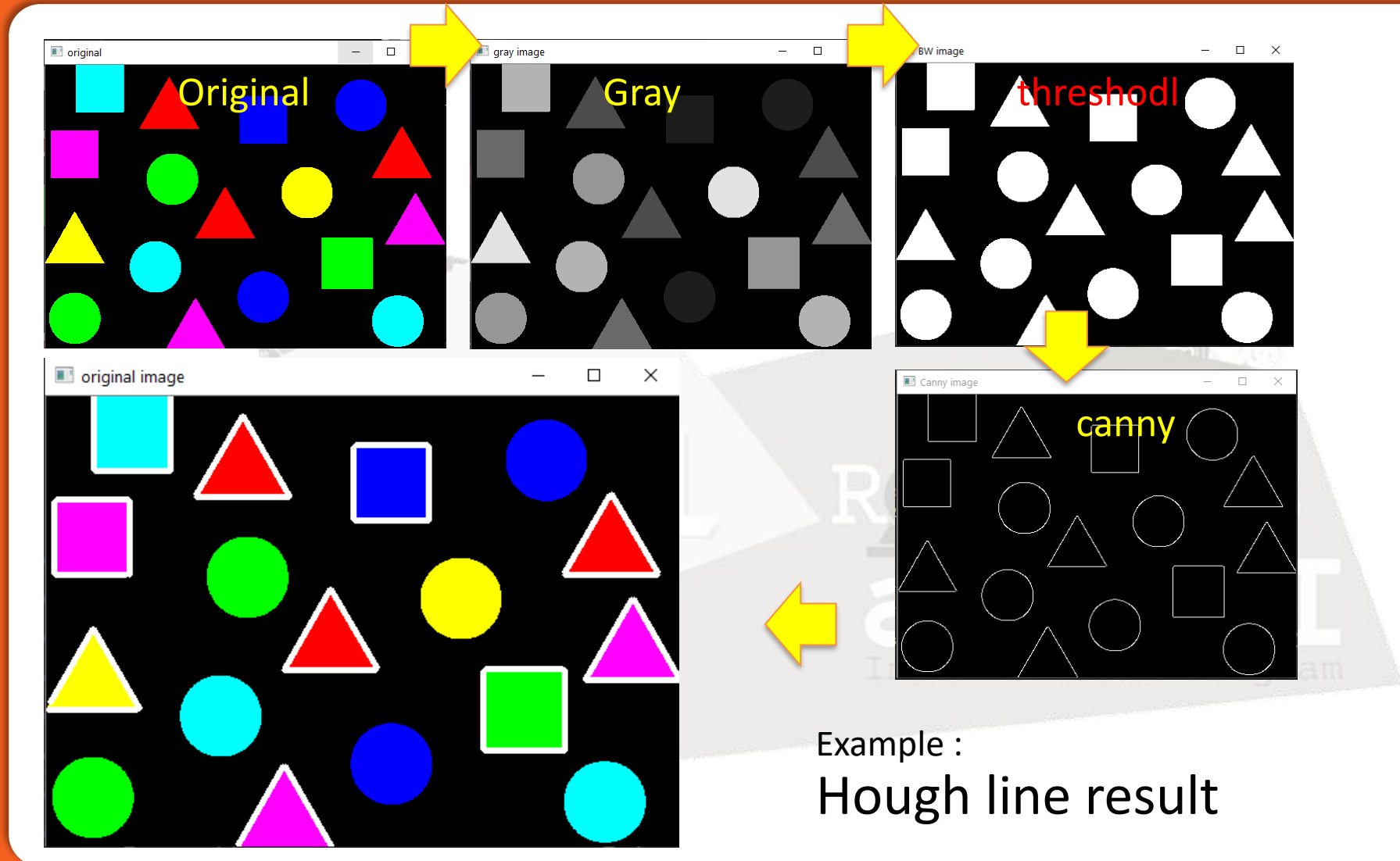
OpenCV function :

`cv2.HoughLinesP`

(dst, rho, theta, threshold, minLineLength, maxLineGap)

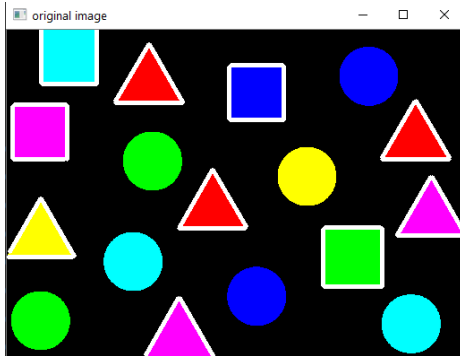


Hough Line Transforms



Example :
Hough line result

Hough Line Transforms



OpenCV function :
cv2.HoughLinesP()

```
img = cv2.imread("dataset/colorobject.png")
gray = cv2.cvtColor(img,cv2.COLOR_BGR2GRAY)
# threshold
_,thr = cv2.threshold(gray,10,255,cv2.THRESH_BINARY)
# Canny edge detection
res = cv2.Canny(thr,10,200)
# hough line
lines = cv2.HoughLinesP(res,1,np.pi/360,50,minLineLength=20,maxLineGap=10)
print(len(lines))

for i in range(len(lines)):
    for x1,y1,x2,y2 in lines[i]:
        cv2.line(img,(x1,y1),(x2,y2),(255,255,255),thickness=3) # change color to white
```

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3. Hough Circle Transforms

Hough Circle Transforms

In a two-dimensional space, a circle can be described by:

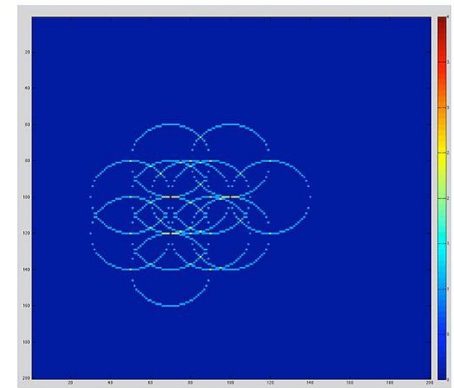
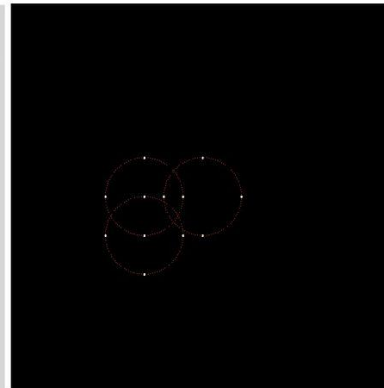
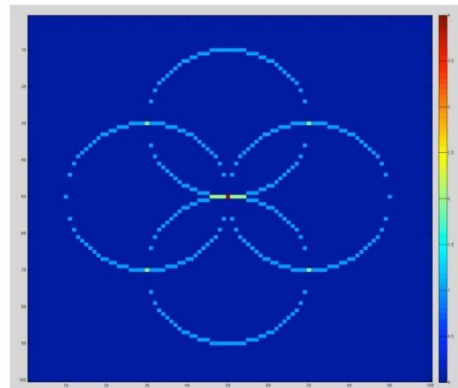
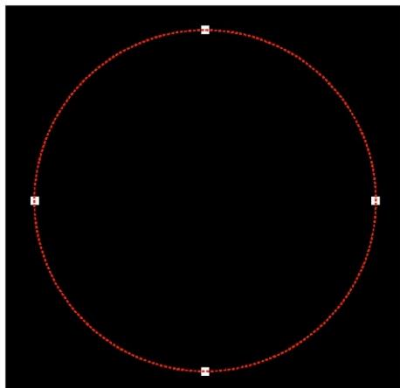
$$(x - a)^2 + (y - b)^2 = r^2$$

Pixel
Coordinate (x,y)

Hough Circle
transform

a,b,r

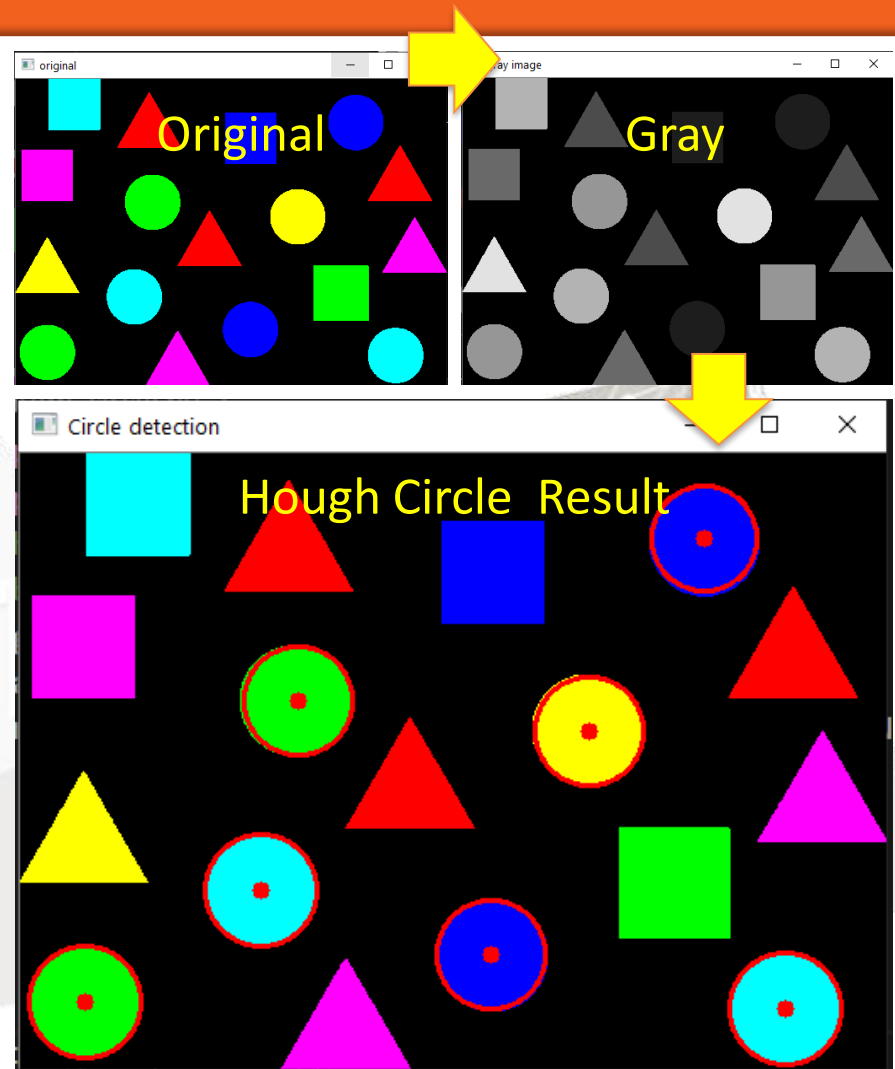
In practice, an accumulator matrix is introduced to find the intersection point in the parameter space



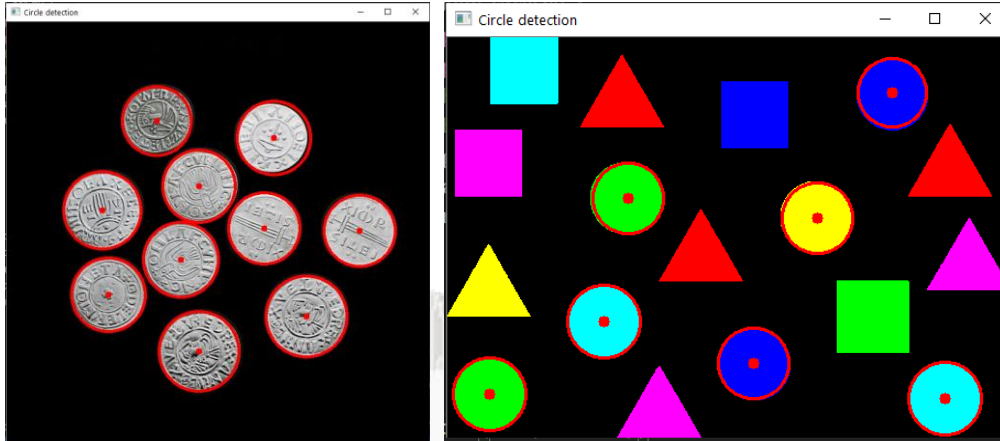
Hough Circle Transforms

Example
Circle detection 3 step :

1. BGR
2. Convert to Gray
3. HoughCircles()



Hough Circle Transforms



OpenCV function :
cv2.HoughCircles()

```
circles = cv2.HoughCircles(gray,cv2.HOUGH_GRADIENT,
                           dp=1.0,minDist=20,
                           param1=10,param2=10,
                           minRadius=30,maxRadius=40)

print(len(circles))
print("Befor around : ",circles[0])
circles = np.uint16(np.around(circles))
print("After around : ",circles[0])
for i in circles[0]:
    # index 0 : x-center , index 1 : y-center, index 2 : radius
    cv2.circle(img,(i[0],i[1]),i[2],(0,0,255),thickness=2)
    cv2.circle(img,(i[0],i[1]),5,(0,0,255),thickness=-1) # thickness -1 mean fill color
```

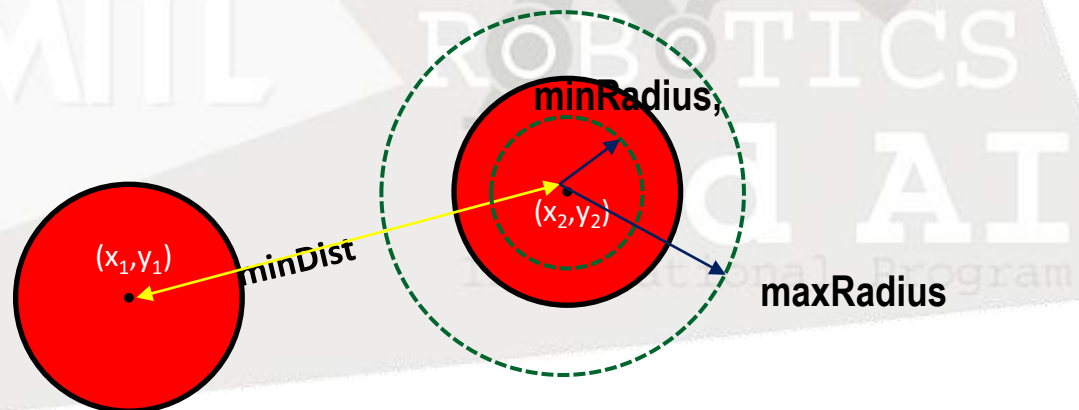
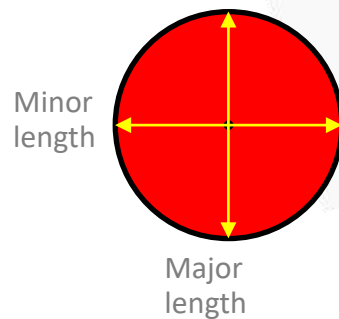
Hough Circle Transforms

OpenCV function :

```
cv2. cv2.HoughCircles(
    gray, cv2.HOUGH_GRADIENT,
    dp, minDist, param1, param2, minRadius, maxRadius)
```

$$dp = \frac{Major}{Minor}$$

param_1 = 200: Upper threshold for the internal Canny edge detector.
param_2 = 100*: Threshold for center detection.



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Q & A

Next : Segmentation