

# Computer Vision

## Ch.7 Connected Component Labeling

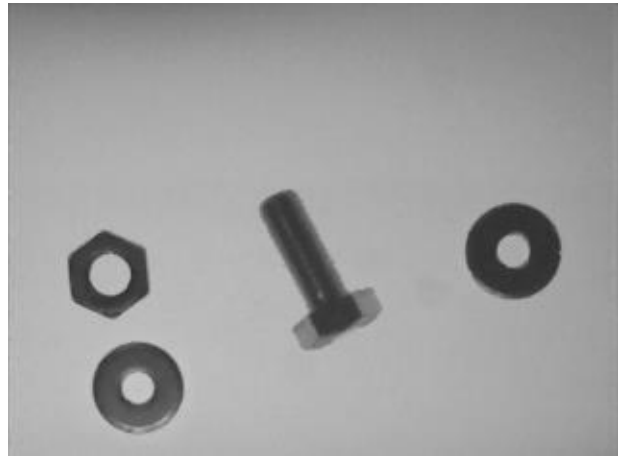
Prof. Po-Yueh Chen (陳伯岳)

E-mail: [pychen@cc.ncue.edu.tw](mailto:pychen@cc.ncue.edu.tw)

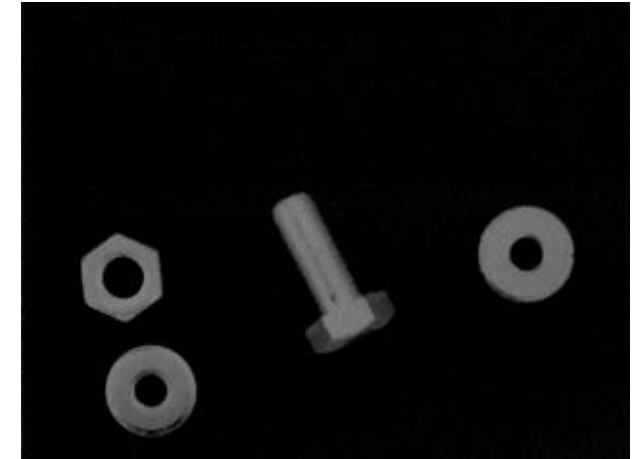
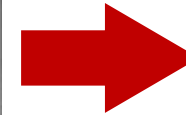
Ext: 8440

NCUE CSIE

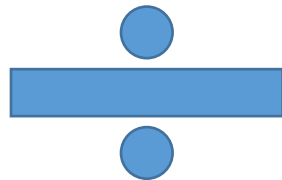
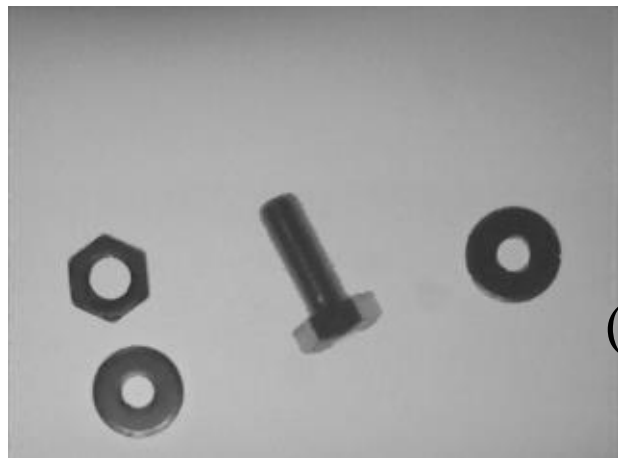
# Review (1/3)



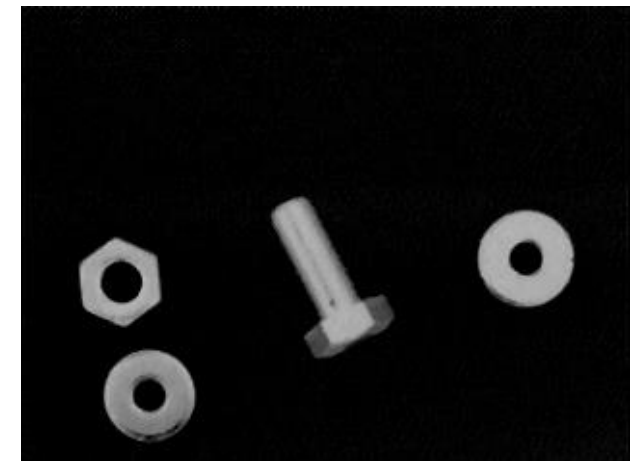
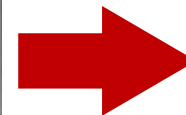
(Minus)



Subtraction



(Divided by)

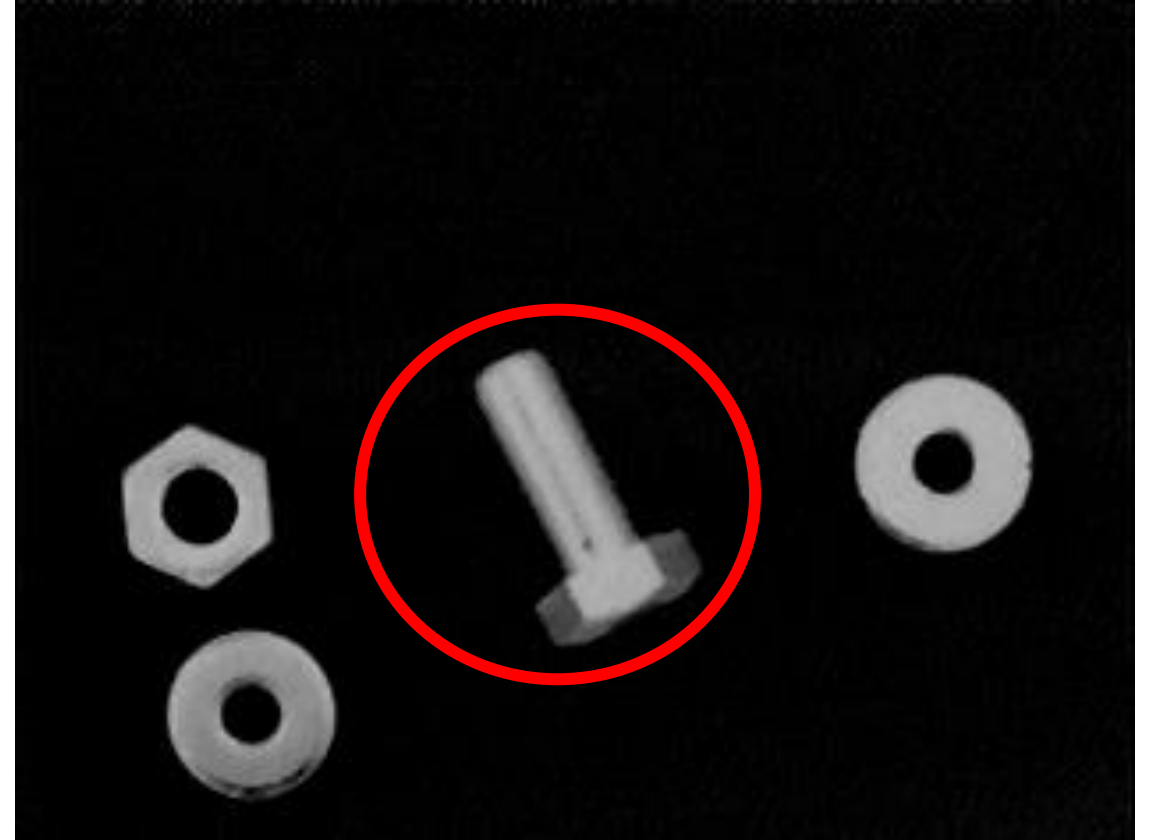


Division

# Review (2/3)



Subtraction



Division

# Review (3/3)

- ✓ In previous chapter, we have known about ...
  - Morphological operations



➤ Remove noise by dilate and erode



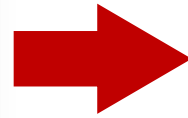
➤ Extract boundary

# Connected component labeling (1/22)

✓ In this episode, we are going to talk about ...

## ➤ Connected component labeling

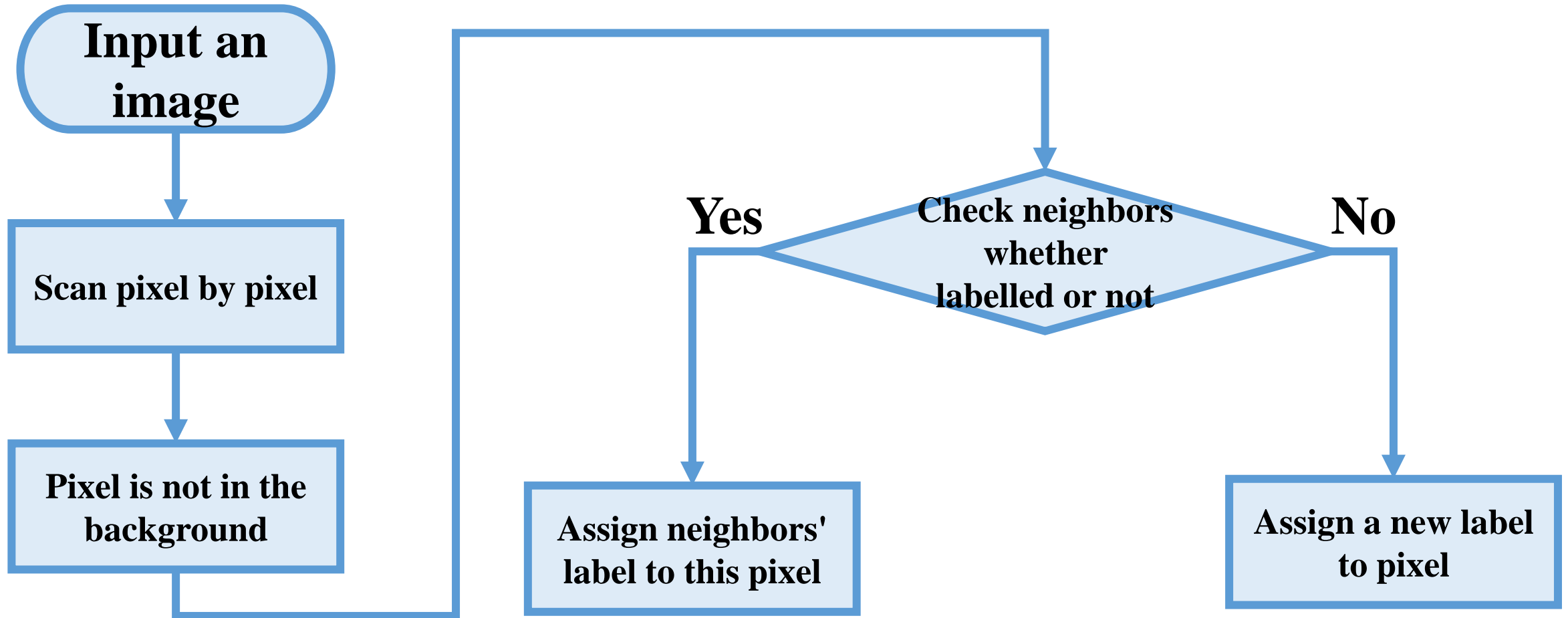
Extracting **connected components** from an image is crucial to many automated computer vision applications.



(Extracted)

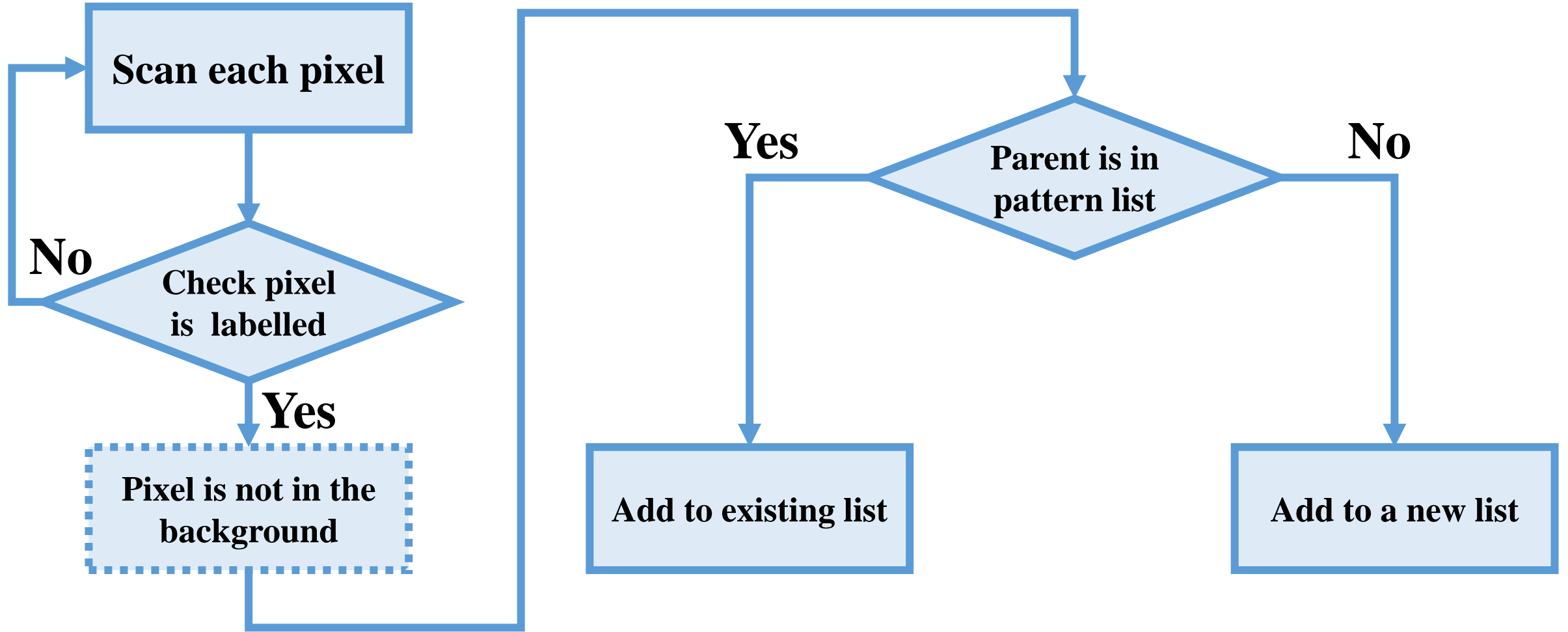


# Connected component labeling (2/22)



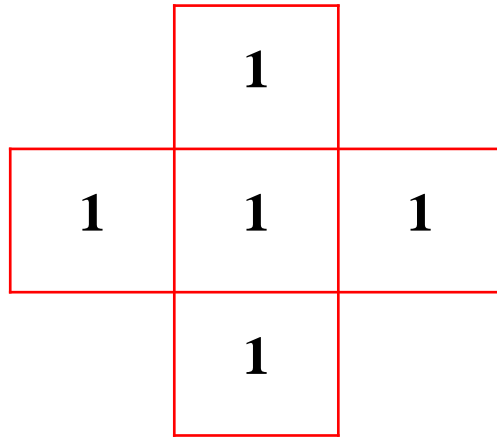
- **The first pass (assigning labels)**

# Connected component labeling (3/22)

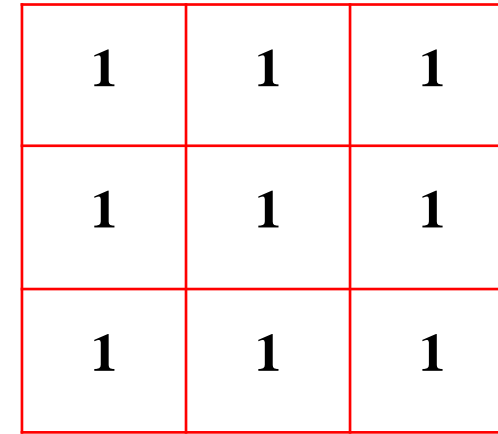


- **The second pass (aggregation)**

# Connected component labeling (4/22)



**4-neighbor**



**8-neighbor**

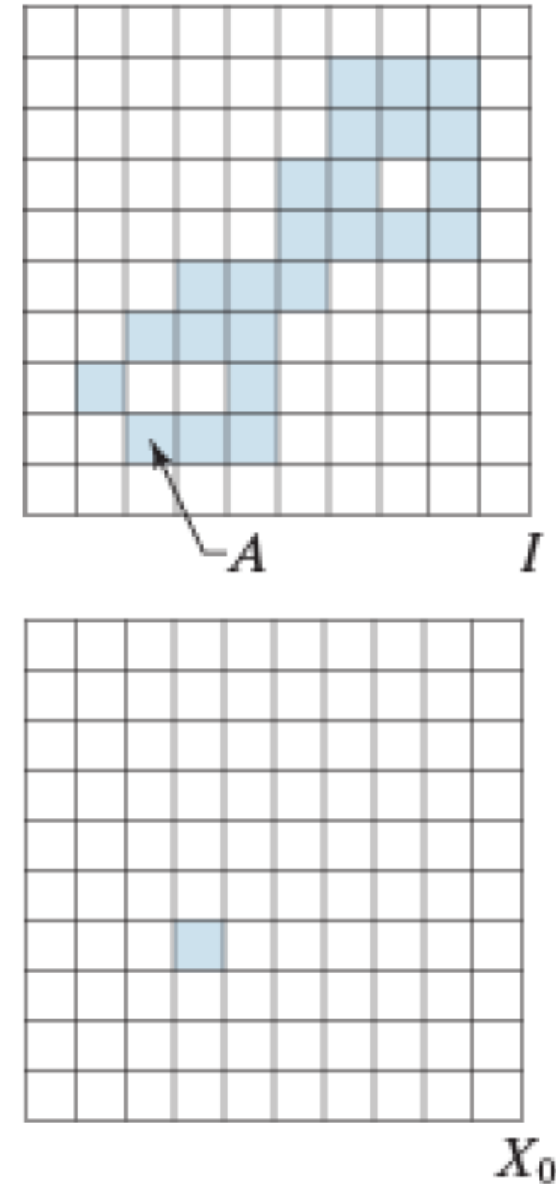
- ✓ Here, we can use two different tracing methods.
- ✓ You can check its 4-or 8-neighbors, depending on your application.
- ✓ In the following, we use **8-neighbors** to search the example image.



# Connected component labeling (5/22)

## ➤ Recursive approach

- Let  $A$  be a set of foreground pixels consisting of one or more connected components.
- Form an image  $X_0$  (of the same size as  $I$ ) whose elements are 0's, except for a known foreground pixel we set to 1.
- The objective is to start with  $X_0$  and find all the connected components in  $I$ .

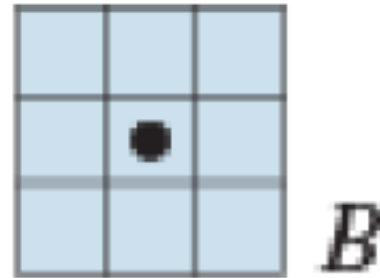


# Connected component labeling (6/22)

- We can use the following iterative procedure:

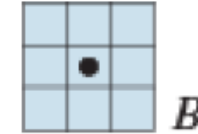
$$X_k = (X_{k-1} \oplus B) \cap A$$

- B is the kernel.



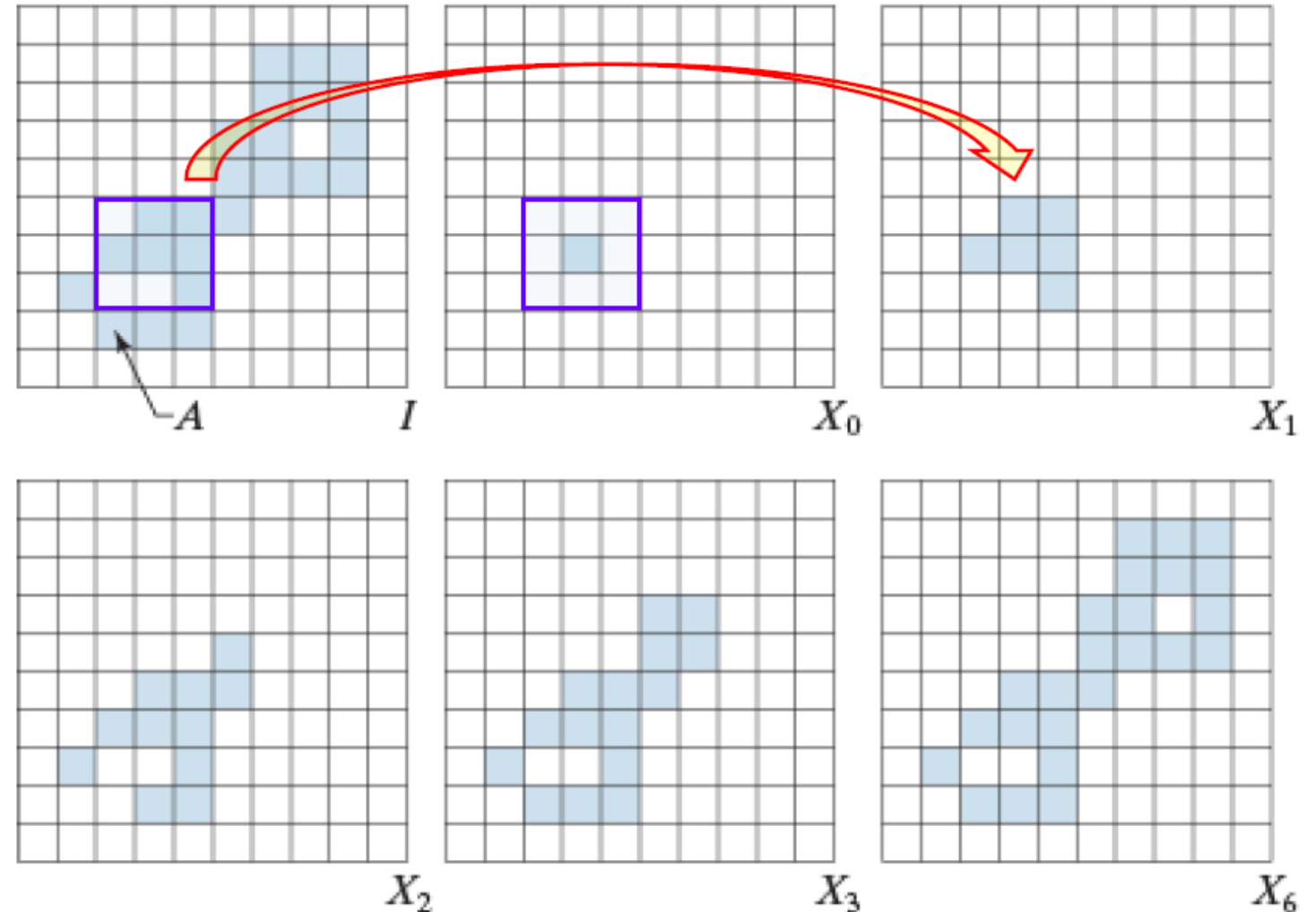
- The procedure terminates when  $X_k = X_{k-1}$ .

# Connected component labeling (7/22)

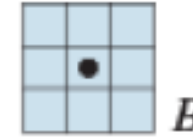


$$X_k = (X_{k-1} \oplus B) \cap A$$

- At the first iteration, dilate  $X_0$  with the kernel  $B$ , and then compute its intersection with  $A$  to get  $X_1$ .

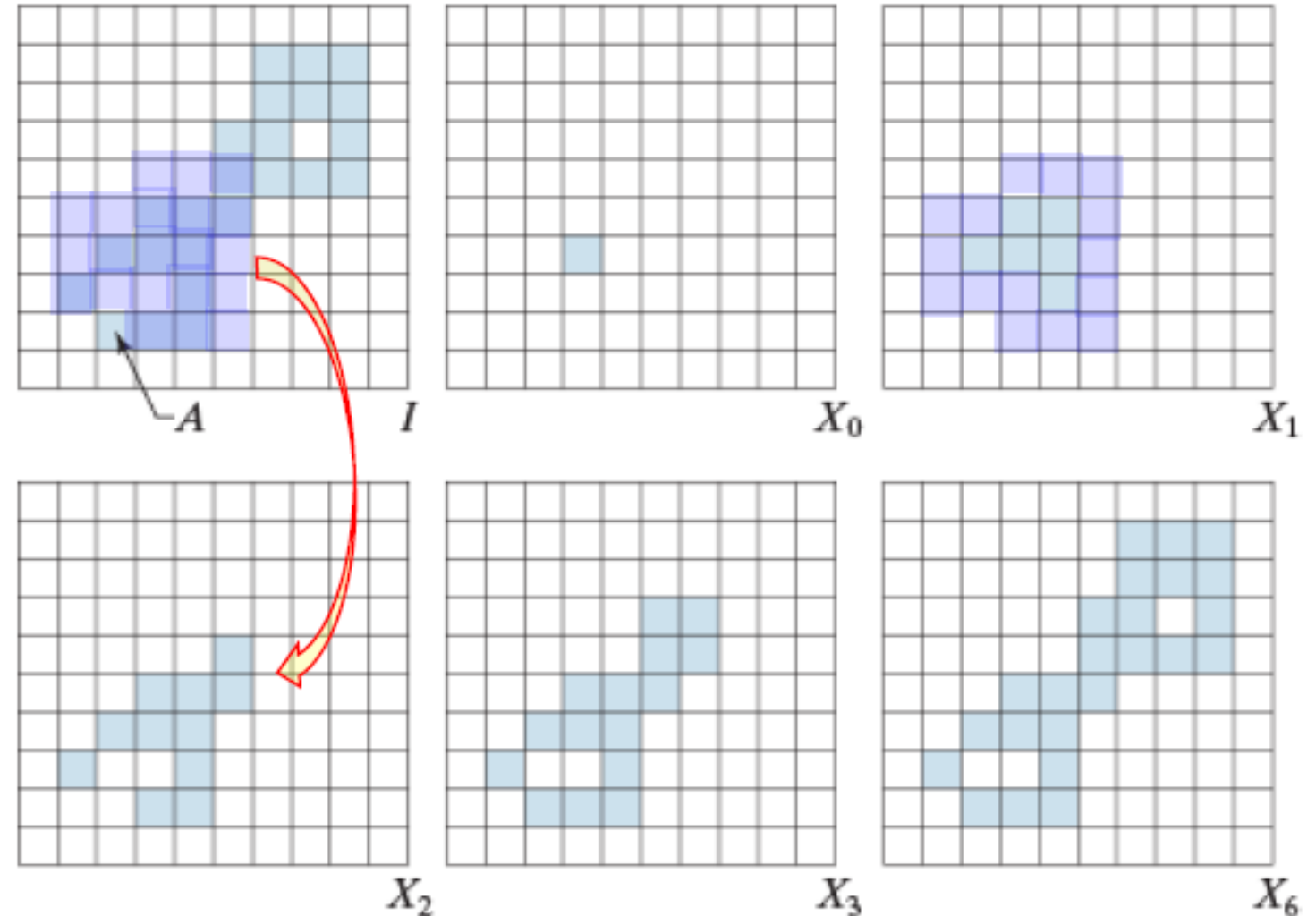


# Connected component labeling (8/22)

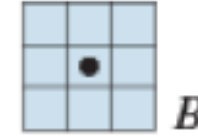


$$X_k = (X_{k-1} \oplus B) \cap A$$

- At the second iteration, dilate  $X_1$  with the kernel  $B$ , and then compute its intersection with  $A$  to get  $X_2$ .

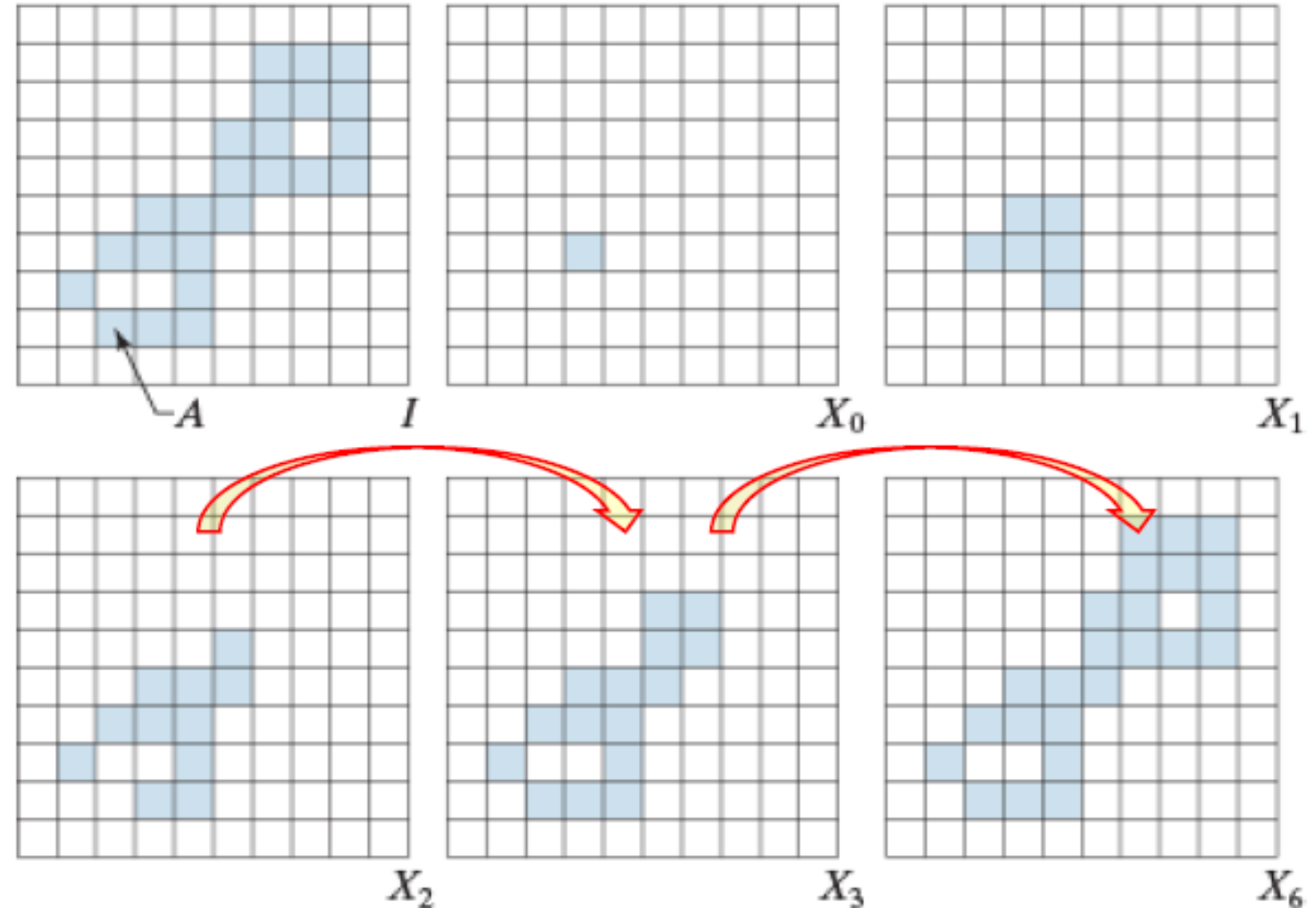


# Connected component labeling (9/22)



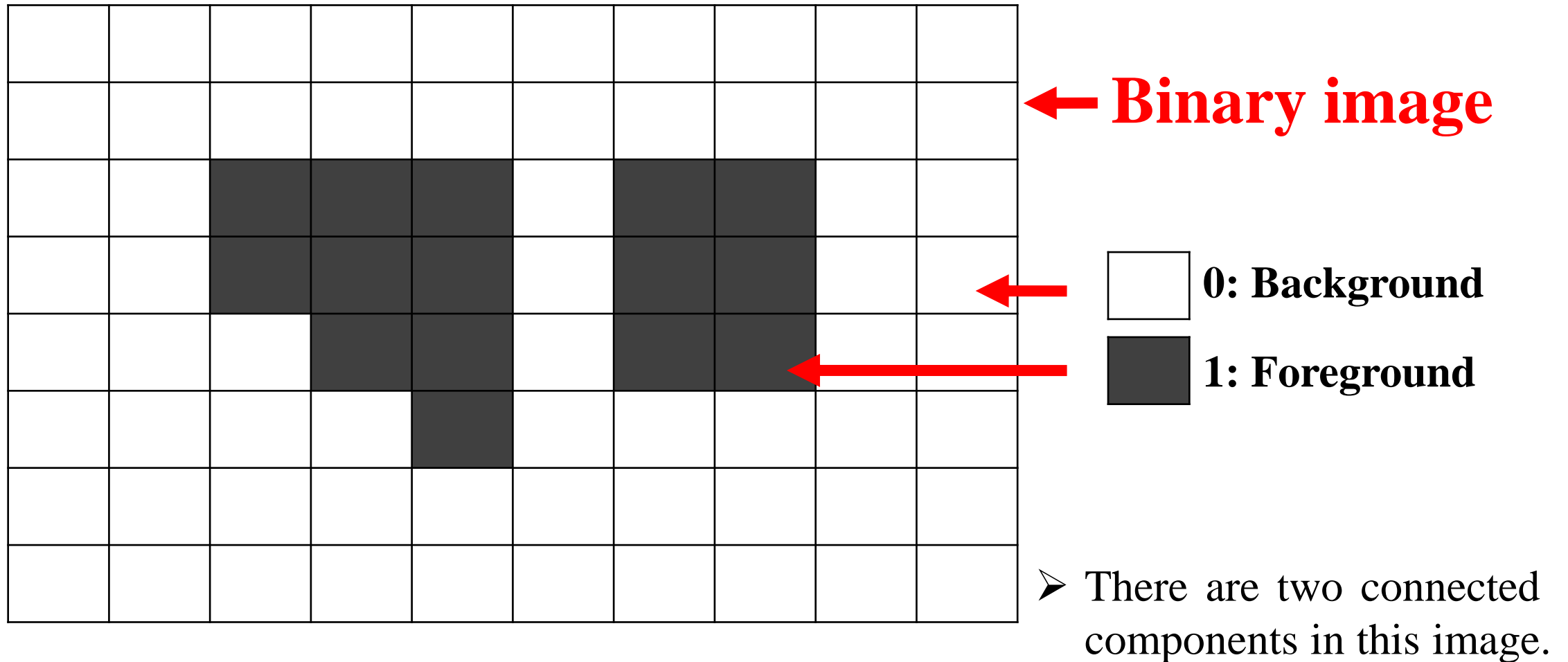
$$X_k = (X_{k-1} \oplus B) \cap A$$

- In this method, the connected component labeling can be done by dilation and intersection.
- Continue the procedure until  $X_k = X_{k-1}$ .

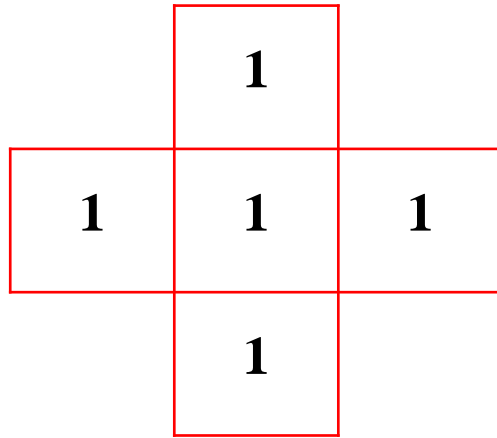


# Connected component labeling (10/22)

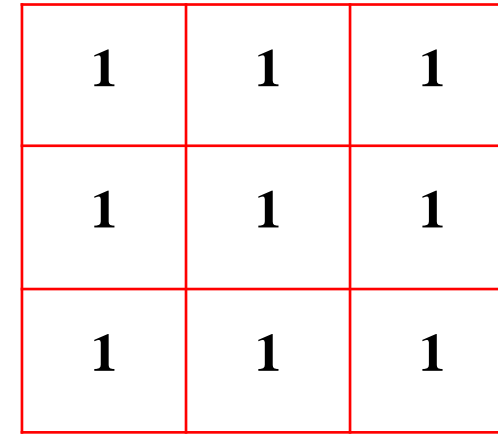
## ➤ Sequential approach



# Connected component labeling (11/22)



**4-neighbor**



**8-neighbor**

- ✓ Here, we can use two different tracing methods.
- ✓ You can check its 4-or 8-neighbors, depending on your application.
- ✓ In the following, we use **4-neighbors** to search the example image.

# Connected component labeling (12/22)

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮

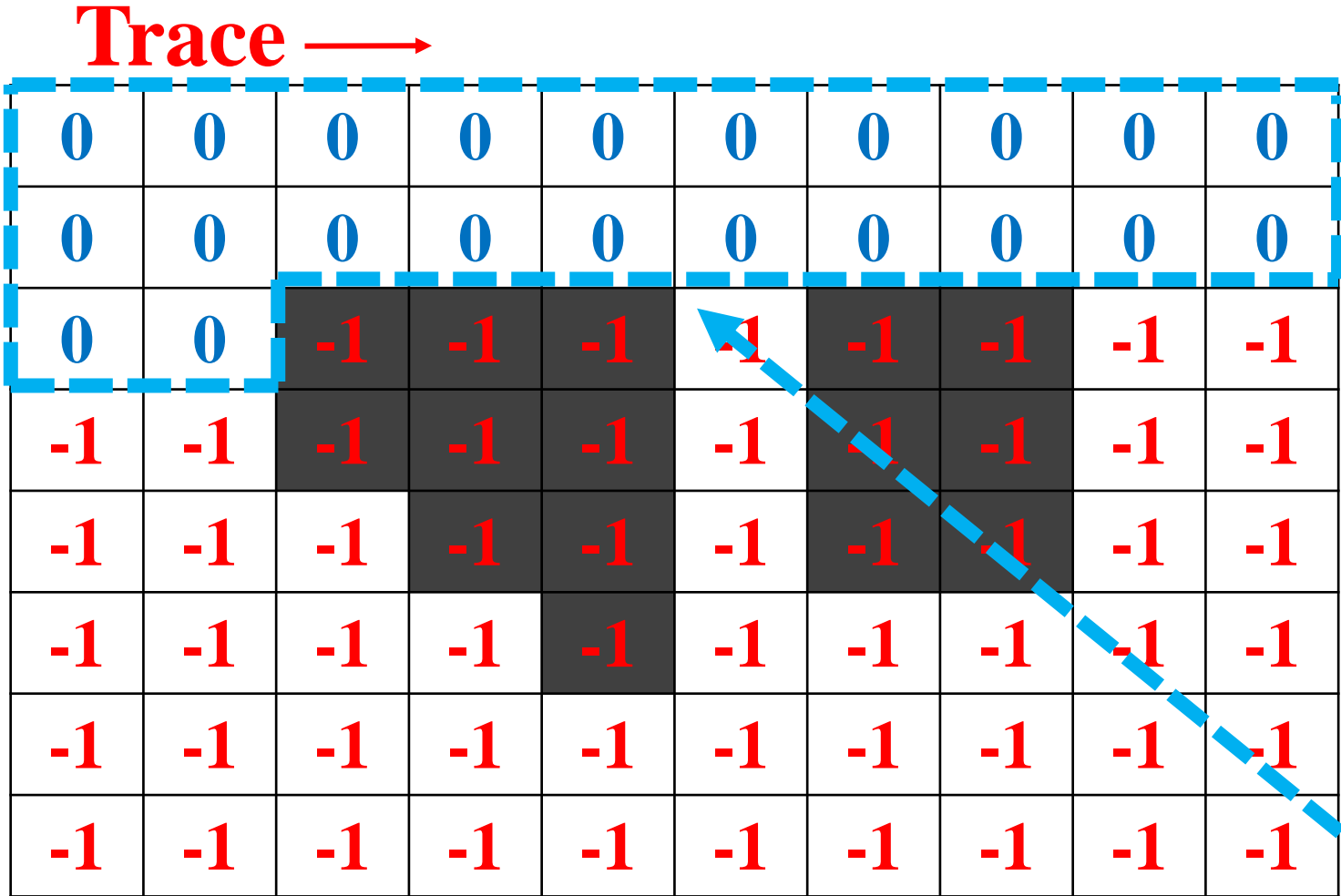
This mask is used to record the information of each traced pixel.

Form a mask (of the same size as the image) and initialize all its pixel values to -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	-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	-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	-1	-1	-1	-1	-1	-1

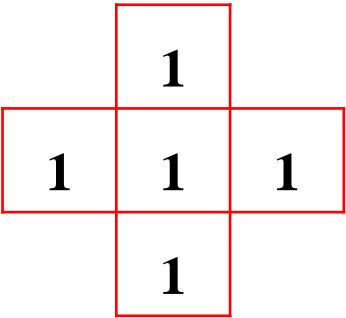


# Connected component labeling (13/22)



## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮



### 4-neighbor

In this section, due to all the pixels are background, so we label the corresponding points in the mask to 0.

# Connected component labeling (14/22)

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮

Trace →

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	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	-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	-1	-1	-1	-1	-1	-1	-1	-1

Label the corresponding point in the mask to 1, meaning this pixel belongs to **Object #1**.

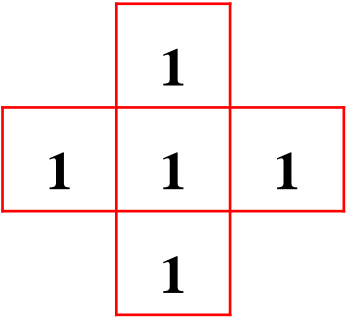
# Connected component labeling (15/22)

Trace →

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	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	-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	-1	-1	-1	-1	-1	-1	-1	-1

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮



4-neighbor

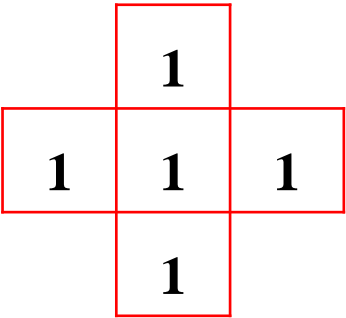
# Connected component labeling (16/22)

Trace →

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	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	-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	-1	-1	-1	-1	-1	-1	-1	-1

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮



4-neighbor

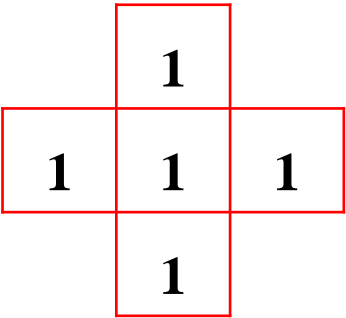
# Connected component labeling (17/22)

Trace →

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	1	1	1	0	-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	-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	-1	-1	-1	-1

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮



4-neighbor

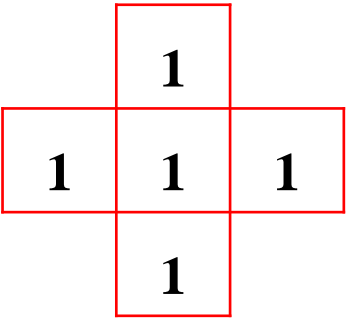
# Connected component labeling (18/22)

Trace →

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	1	1	1	0	-1	-1	-1	-1
-1	-1	-1	-1	1	0	-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	-1	-1	-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1	-1	-1	-1

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮



4-neighbor

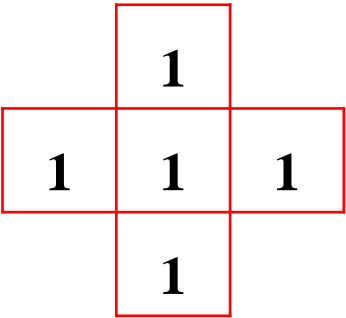
# Connected component labeling (19/22)

**Trace** 

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	1	1	1	0	-1	-1	-1	-1
-1	0	1	1	1	0	-1	-1	-1	-1
-1	-1	0	1	1	0	-1	-1	-1	-1
-1	-1	-1	0	1	0	-1	-1	-1	-1
-1	-1	-1	-1	0	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1	-1	-1	-1

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮



**4-neighbor**

# Connected component labeling (20/22)

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮

Trace →

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

We meet a foreground pixel here, and label the corresponding point in the mask to the next number, i.e., 2, meaning this pixel belongs to **Object #2**.



# Connected component labeling (21/22)

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	1	1	1	0	2	2	0	0
0	0	1	1	1	0	2	2	0	0
0	0	0	1	1	0	2	2	0	0
0	0	0	0	1	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	★

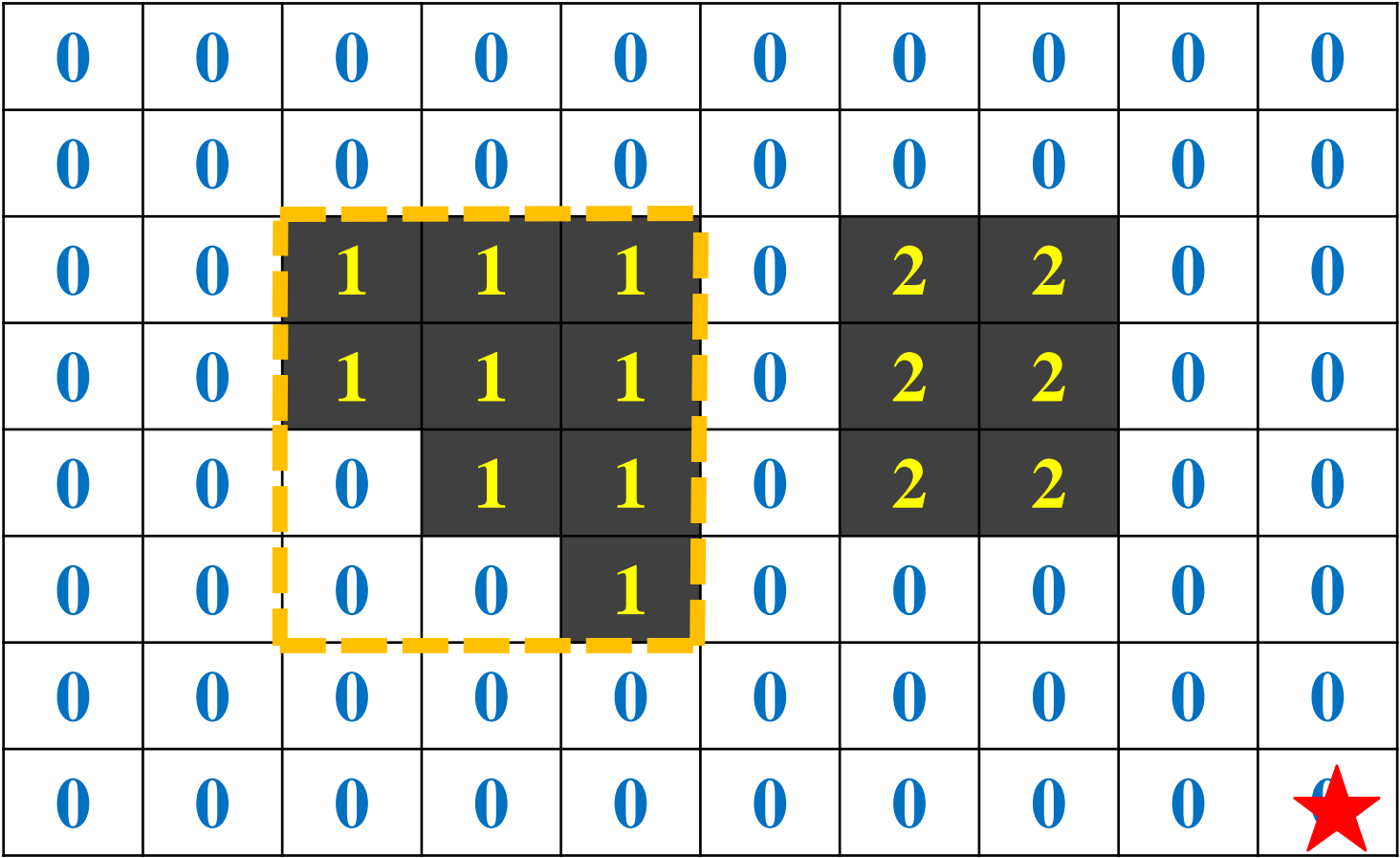
After tracing all pixels in the image, all connected components are labelled.

Trace →

# Connected component labeling (22/22)

## Labeling

- 1: Unlabeled
- 0 : Background
- 1 : Object #1
- 2 : Object #2
- ⋮



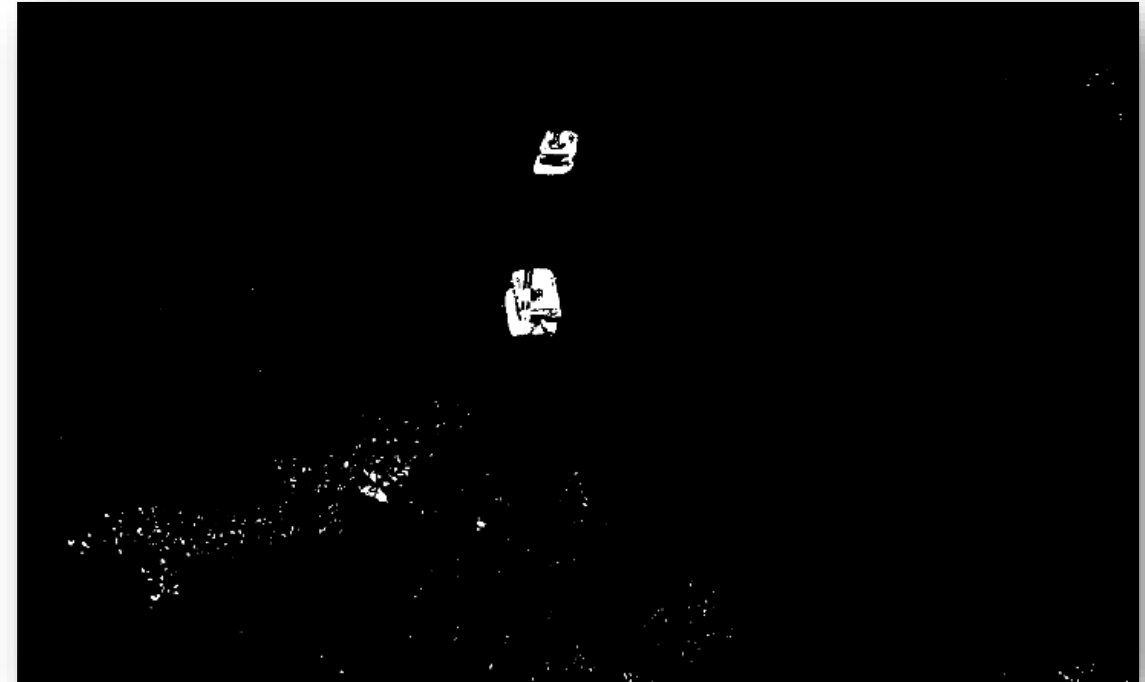
**Trace** →

When performing connected component labelling, we can also record the max  $x$ , max  $y$ , min  $x$ , min  $y$  for each object to get its bounding box (or bounding rectangle).

# Example of Connected component labeling (1/3)



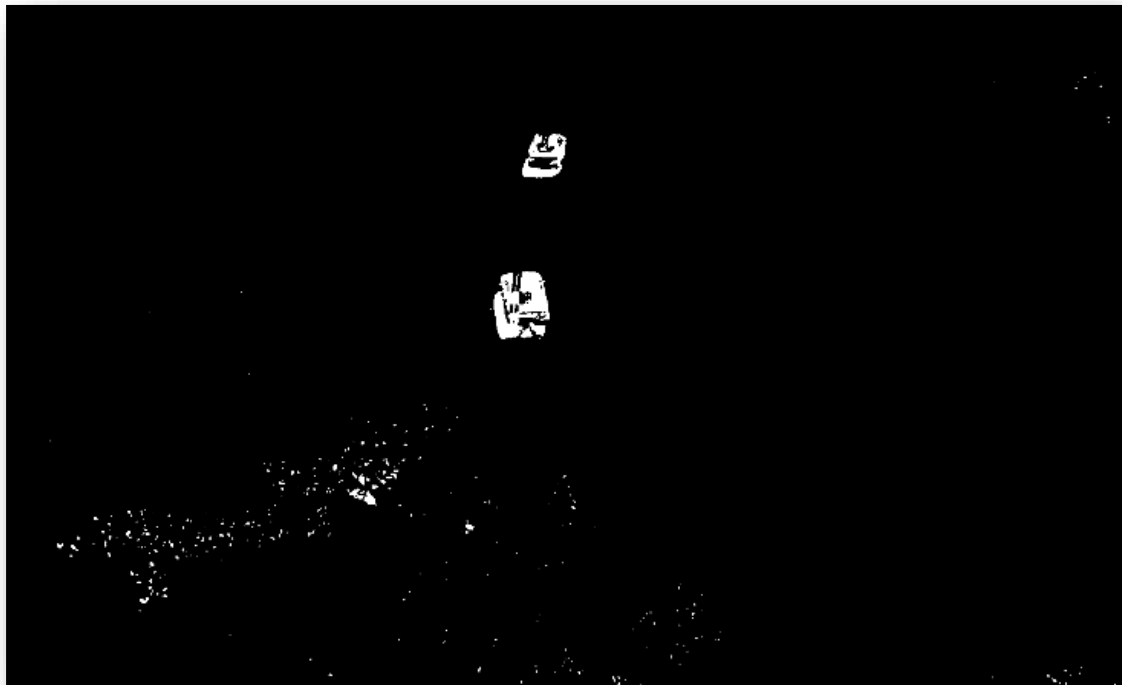
Original Video Frame



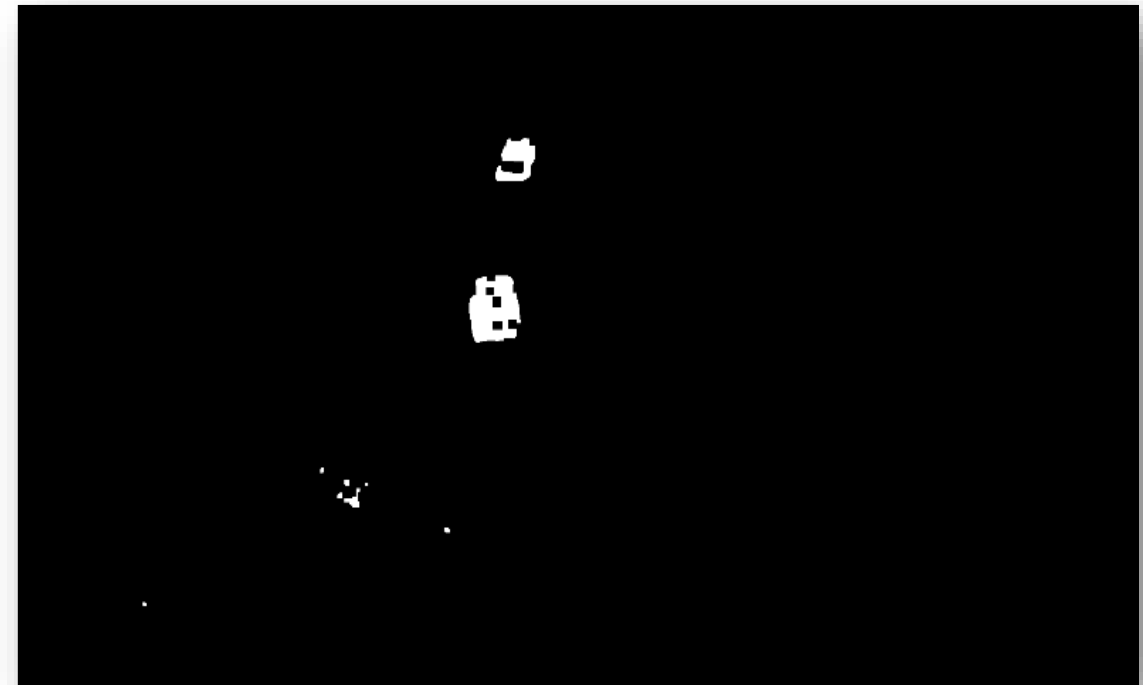
Foreground Pixels

✓ Background subtraction (in previous episode)

# Example of Connected component labeling (2/3)



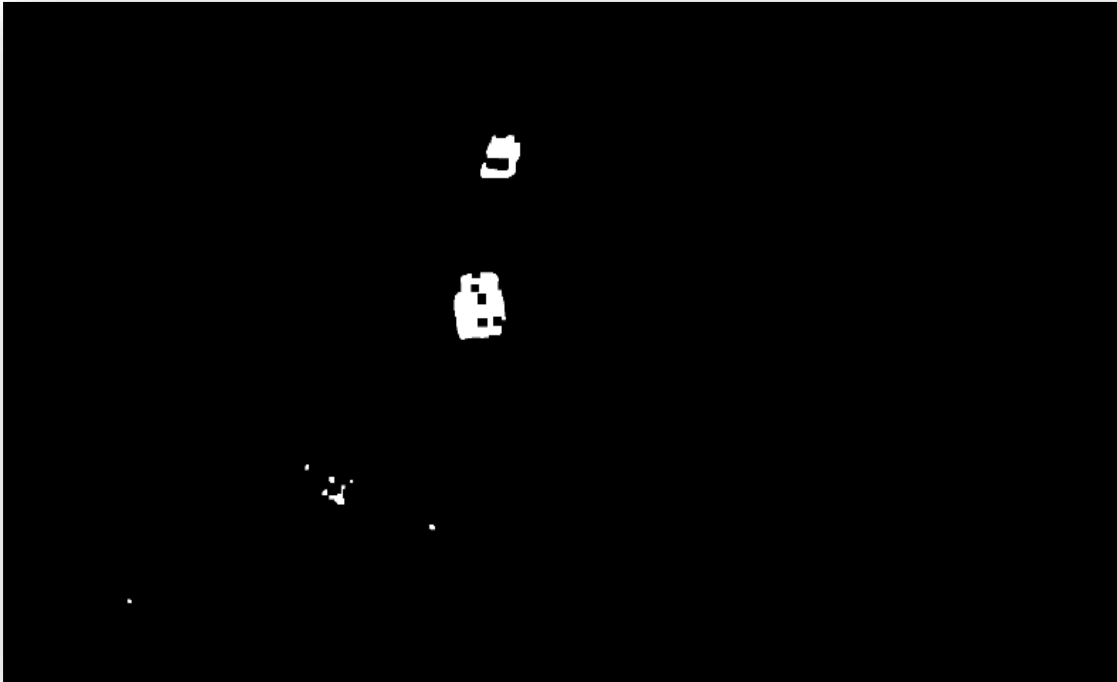
Foreground Pixels



Remove Noise

✓ **Morphological operations**

# Example of Connected component labeling (3/3)



Remove Noise



Object marked

✓ **Connected component labeling**

# OpenCV – connectedComponents(1/4)

## ➤ Code

### Syntax:

`connectedComponents(src, labels, connectivity, type);`

**src** – Input 8-bit single-channel (Gray Level).

**labels** – Output label map.

**connectivity** – 4- or 8-connected components

**type** – Output label type (CV\_32S or CV\_16U)

# OpenCV – connectedComponents(2/4)

## ➤ Demo Code

```
Mat img;
int threshval = 100; // Bar Default Value.

int main(int argc, const char** argv)
{

    img = imread("D:/Data.jpg", IMREAD_GRAYSCALE);

    imshow("Image", img);

    namedWindow("Connected Components", WINDOW_AUTOSIZE);
    createTrackbar("Threshold", "Connected Components", &threshval, 255, on_trackbar);
    on_trackbar(threshval, 0);

    waitKey(0);
    return 0;
}
```

# OpenCV – connectedComponents(3/4)

## ➤ Demo Code

```
static void on_trackbar(int, void*)
{
    Mat bw = threshval < 128 ? (img < threshval) : (img > threshval);
    Mat labelImage(img.size(), CV_32S);

    int nLabels = connectedComponents(bw, labelImage, 8);

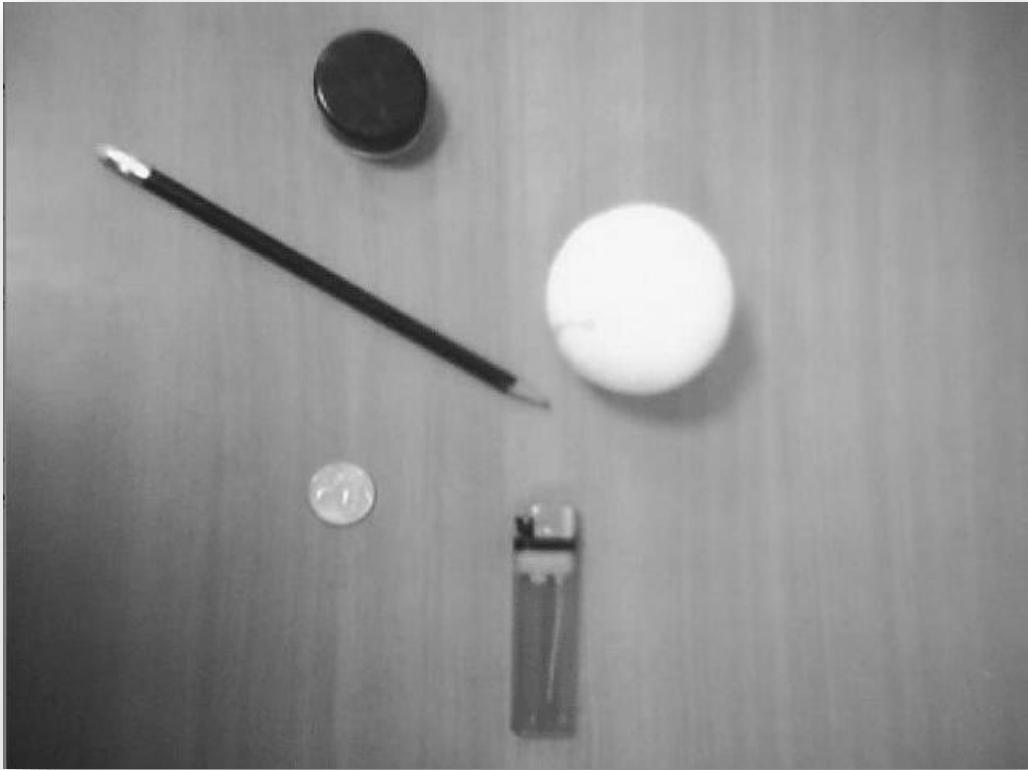
    vector<Vec3b> colors(nLabels);
    colors[0] = Vec3b(0, 0, 0); //Background
    for (int label = 1; label < nLabels; ++label)
    {
        colors[label] = Vec3b((rand() & 255), (rand() & 255), (rand() & 255));
    }

    // Draw color to the whole image.
    Mat dst(img.size(), CV_8UC3);
    for (int r = 0; r < dst.rows; ++r) {
        for (int c = 0; c < dst.cols; ++c) {
            int label = labelImage.at<int>(r, c);
            Vec3b &pixel = dst.at<Vec3b>(r, c);
            pixel = colors[label];
        }
    }
    imshow("Connected Components", dst);
}
```

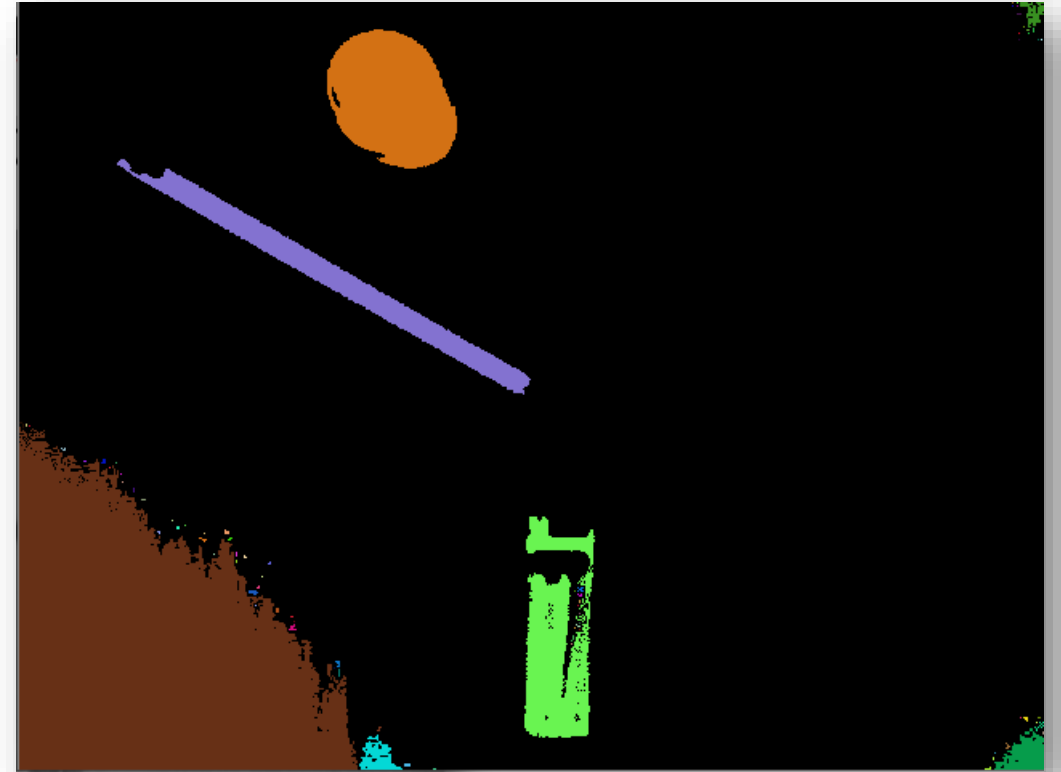


# OpenCV – connectedComponents(4/4)

## ➤ Demo Result



Original Image



Connected Component Results

# Practice

# OpenCV – findContours (1/4)

## ➤ Code

### Syntax:

`findContours(src, contours, hierarchy, mode, method, offset);`

**src** – Source, an 8-bit single-channel image.

**contours** – Detected contours. Each contour is stored as a **vector of points**.

**hierarchy** – Optional output vector, containing information about the image topology.

It has as many elements as the number of contours.

**mode** – Contour retrieval mode

**method** – Contour approximation method.

**offset** – Optional offset by which every contour point is shifted.

# OpenCV – findContours (2/4)

White foreground regions are on a black background.

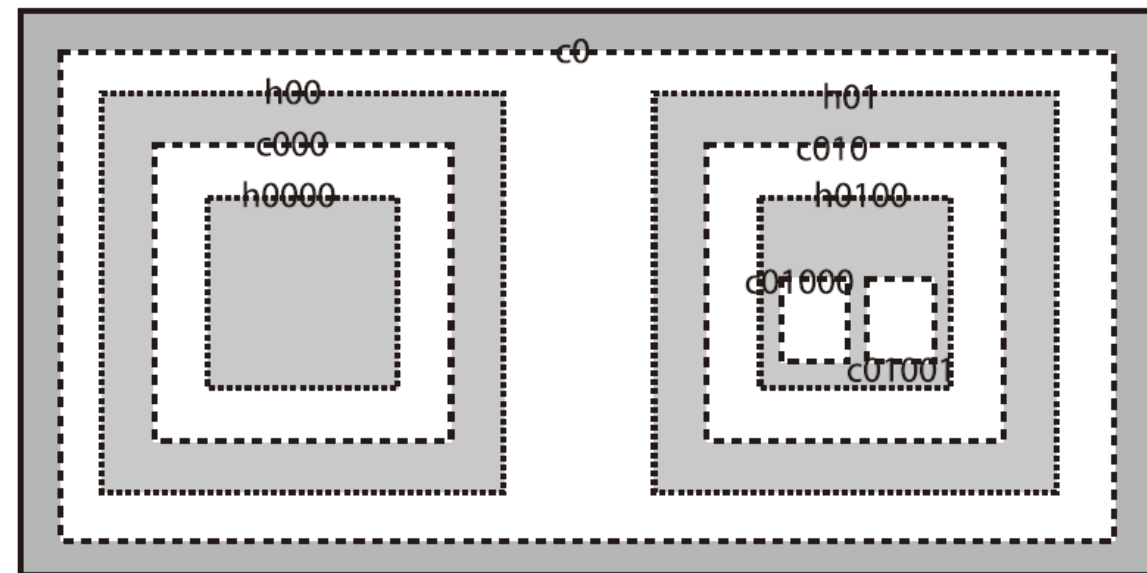
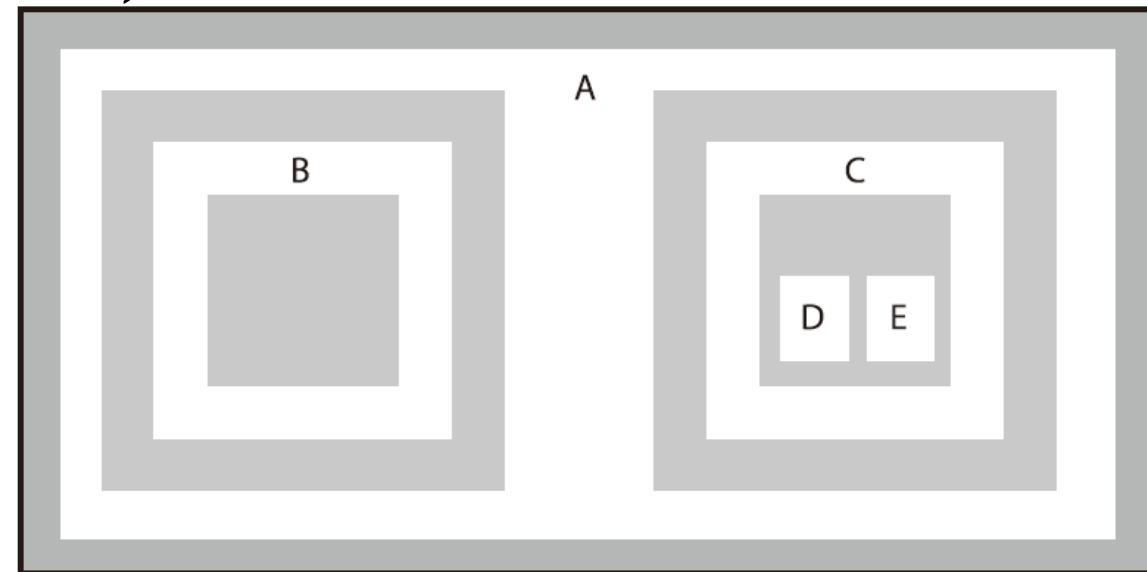
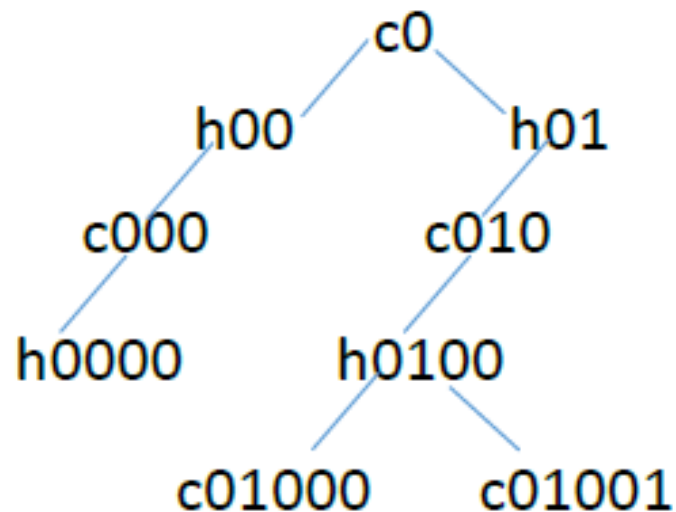
c: contour

h: hole

Contours (dashed lines): exterior boundaries

Contours (dotted lines): interior boundaries

The contour hierarchy can be represented as a tree:



# OpenCV – findContours (3/4)

## ➤ Contour mode

### Syntax:

findContours(src, contours, hierarchy, **mode**, method, offset)

➤ **mode** – Contour retrieval mode.

**CV\_RETR\_EXTERNAL**: Retrieves only the extreme outer contours.

**CV\_RETR\_LIST** : Retrieves all of the contours without establishing any hierarchical relationships.

**CV\_RETR\_CCOMP** : Retrieves all of the contours and organizes them into a **two-level hierarchy**.

At the top level, there are **external boundaries** of the components.

At the second level, there are boundaries of the holes.

If there is another contour inside a hole of a connected component, it is still put at the **top** level.

**CV\_RETR\_TREE** : Retrieves all of the contours and reconstructs a full hierarchy of nested contours.

# OpenCV – findContours (4/4)

## ➤ Contour method

### Syntax:

➤ findContours(src, contours, hierarchy, mode, **method**, offset)

**method** – Contour approximation method.

**CV\_CHAIN\_APPROX\_NONE:** Stores absolutely all the contour points.

- That is, any 2 subsequent points  $(x1, y1)$  and  $(x2, y2)$  of the contour will be either horizontal, vertical or diagonal neighbors, that is,  $\max(\text{abs}(x1-x2), \text{abs}(y2-y1)) == 1$ .

**CV\_CHAIN\_APPROX\_SIMPLE:** Compresses horizontal, vertical, and diagonal segments and leaves only their end points. For example, an up-right rectangular contour is encoded with 4 points.

**CV\_CHAIN\_APPROX\_TC89\_L1, CV\_CHAIN\_APPROX\_TC89\_KCOS:** Applies one of the flavors of the Teh-Chin chain approximation algorithm. See the OpenCV doc.[TehChin89] for more the details.

# OpenCV – drawContours (1/2)

## ➤ Code

### Syntax:

`drawContours(src, contours, index, color, thickness, lineType, hierarchy, maxLevel, offset);`

**src** – Destination image where you would like to draw contours.

**contours** – All the input contours (obtained from `findContours`).

- Each contour is stored as a point vector.

**index** – Parameter indicating a contour to draw. If it is negative, all the input contours are drawn.

**color** – Color of the contours.

**thickness** – Thickness of lines the contours are drawn with. If it is negative (e.g., `thickness = CV_FILLED` ), the contour interiors are drawn.

**lineType** – Line connectivity. (**8**: 8-connected, **4**: 4-connected, **CV\_AA**: anti-aliased line)

# OpenCV – drawContours (1/2)

## ➤ Code

### Syntax:

EX: drawContours(drawing, contours, (int)i, color, 2, LINE\_8, hierarchy, 0);

drawContours(src, contours, index, color, thickness, lineType, hierarchy, maxLevel, offset);

**hierarchy** – Optional information about hierarchy.

It is only needed if you want to draw only some of the contours (see maxLevel).

**maxLevel** – Maximal level for drawn contours.

- If it is 0, only the contour specified by contour index is drawn.
- If it is 1, the function draws the contour(s) and all the nested contours. (E.g., contours 2 and 3 are drawn.)
- If it is 2, the function draws the contours, all the nested contours, all the nested-to-nested contours, and so on. (E.g., contours 2, 3, 4 are drawn.)
- This parameter is only taken into account when there is hierarchy available



# OpenCV – Canny Edge Detector

## ➤ Code

### Syntax:

`Canny(src, edges, threshold1, threshold2, apertureSize, L2gradient=false );`

**src** – Source, an 8-bit single-channel image.

**edges** – output edge map; it has the same size and type as image .

**threshold1** – First threshold for the hysteresis procedure.

**threshold2** – Second threshold for the hysteresis procedure.

**apertureSize** – Aperture size for the Sobel() operator.

**L2gradient** – A flag, indicating whether a more accurate  $L_2$  norm  $= \sqrt{(dI/dx)^2 + (dI/dy)^2}$  should be used to calculate the image gradient magnitude ( `L2gradient=true` ), or whether  $L_1$  the default norm  $= |dI/dx| + |dI/dy|$  is enough ( `L2gradient=false` ).

# Give it a try (1/3)

## ➤ Demo Code

```
int main(int argc, char** argv)
{
    Mat src = imread("D:/Data.jpg"); // Load source image

    cvtColor(src, src_gray, COLOR_BGR2GRAY);
    blur(src_gray, src_gray, Size(3, 3)); // Convert image to gray and blur it to get clear contour.

    const char* source_window = "Source";
    namedWindow(source_window);
    imshow(source_window, src);

    const int max_thresh = 255;
    createTrackbar("Canny Thresh:", source_window, &thresh, max_thresh, thresh_callback);
    thresh_callback(0, 0);

    waitKey();
    return 0;
}
```

# Give it a try (2/3)

## ➤ Demo Code

```
void thresh_callback(int, void*)
{
    Mat canny_output; // Detect edges using Canny
    Canny(src_gray, canny_output, thresh, thresh*2); // Canny Edge Detector to reinforce the contour result.

    vector<vector<Point>> contours; // Find contours
    vector<Vec4i> hierarchy;
    findContours(canny_output, contours, hierarchy, RETR_TREE,
                CHAIN_APPROX_SIMPLE);

    Mat drawing = Mat::zeros(canny_output.size(), CV_8UC3); // Declare background to zero.
    for (size_t i = 0; i < contours.size(); i++) // Draw contours for loop
    {
        Scalar color = Scalar(255, 255, 255);
        drawContours(drawing, contours, (int)i, color, 2, LINE_8, hierarchy, 0);
    }

    imshow("Contours", drawing);
}
```

# Give it a try (3/3)

## ➤ Demo Result



Original Image



Result of Contours

# OpenCV – boundingRect

## ➤ Code

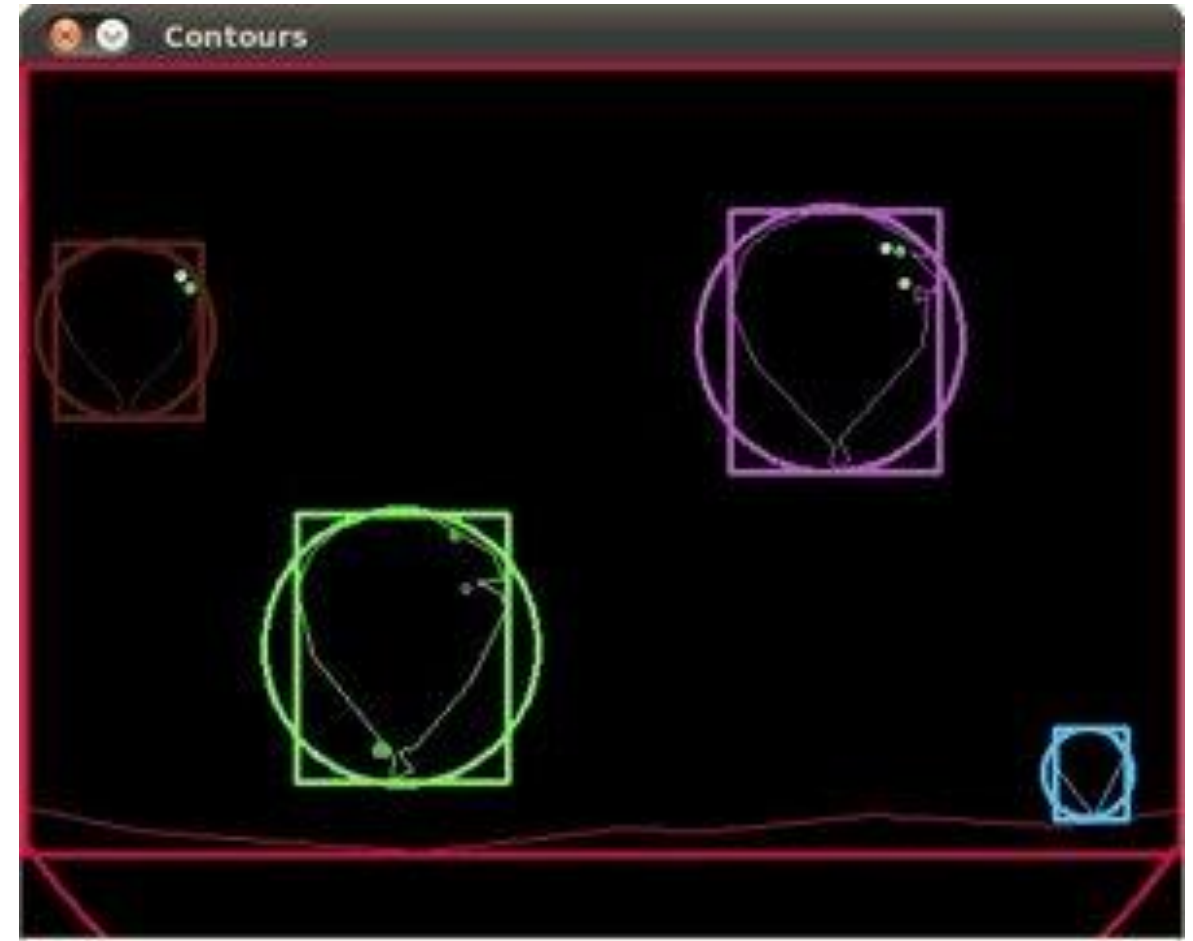
### Syntax:

`Rect boundingRect(InputArray points);`

**points** – Input 2D point set, stored in `std::vector` or `Mat`.

- ✓ The function calculates and returns the minimal up-right bounding rectangle for the specified point set.

# Give it a try



✓ You can try this on ➤ [https://docs.opencv.org/2.4/doc/tutorials/imgproc/shapedescriptors/bounding\\_rects\\_circles/bounding\\_rects\\_circles.html](https://docs.opencv.org/2.4/doc/tutorials/imgproc/shapedescriptors/bounding_rects_circles/bounding_rects_circles.html)

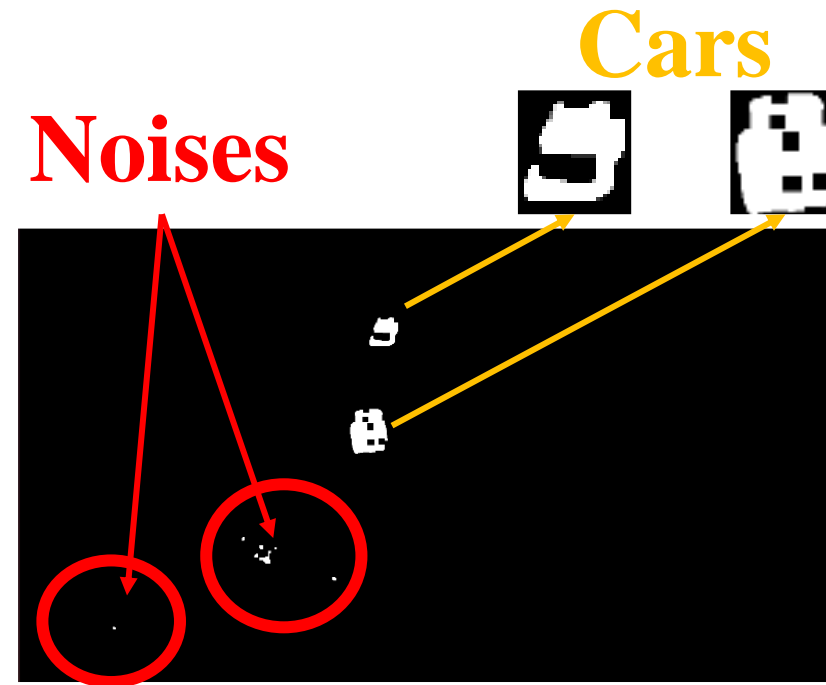
# Demo

# Practice



# Basic object filtering (1/5)

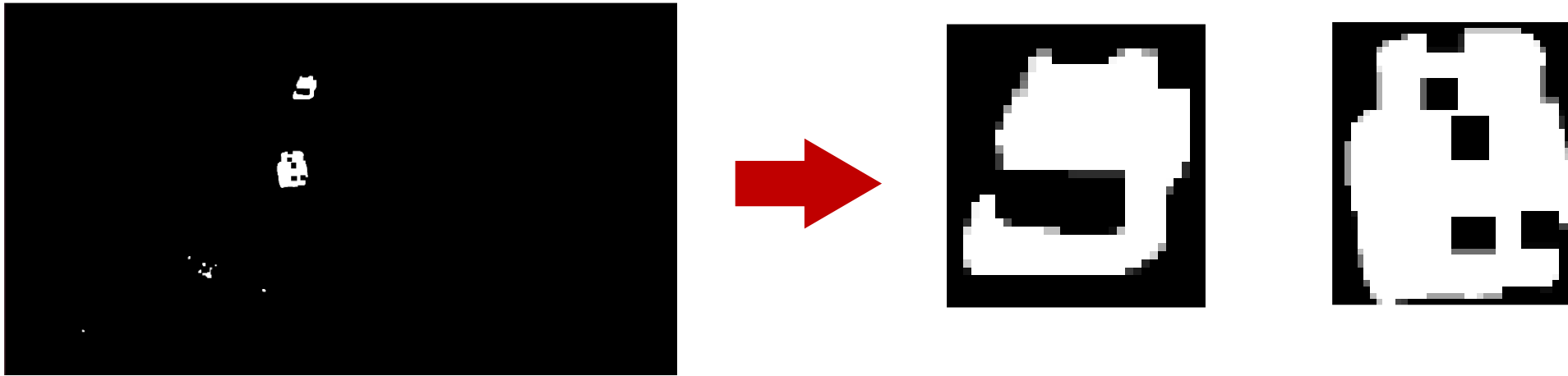
- After morphological operations, there may be still some **noises** which do not belong to foreground objects.
- Object filtering is required to eliminate such noises.
  - ✓ Object size
  - ✓ Shape
  - ✓ Compactness



# Basic object filtering (2/5)

## ➤ Object size

- ✓ Area of the bounding box: width  $\times$  height
- ✓ Number of foreground pixels



- Filter out the objects which are too small or too large.

# Basic object filtering (3/5)

## ➤ Shape

- ✓ Aspect ratio of the bounding box.  
→  $R = \text{width} / \text{height}$ .



Stand  
 $R=0.3$



Squat  
 $R=0.7$

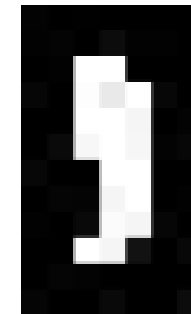
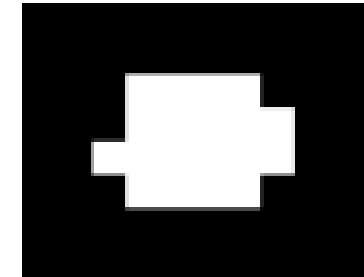
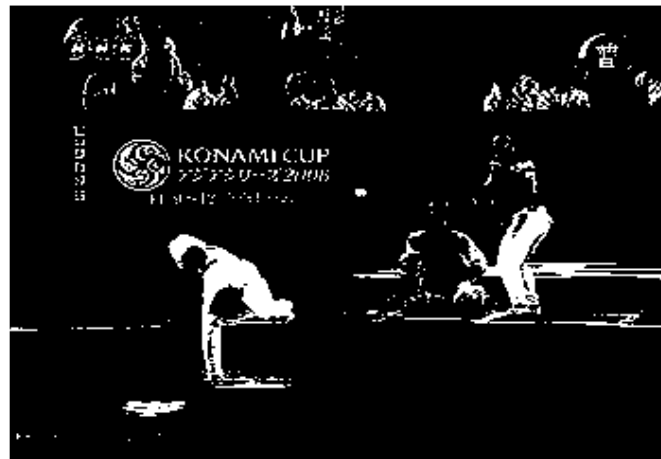


Lie down  
 $R=5.5$

# Basic object filtering (4/5)

## ➤ Shape

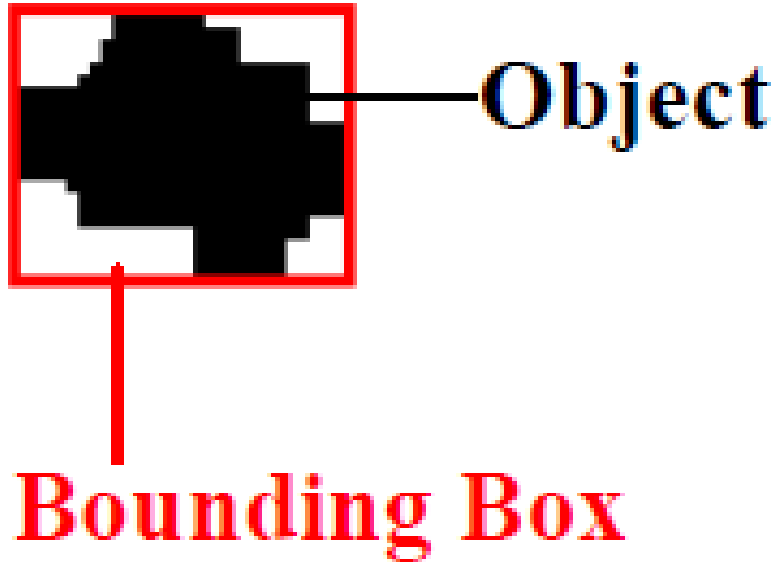
- ✓ Ball detection (Aspect ratio = 1)



# Basic object filtering (5/5)

## ➤ Compactness

- ✓ Object size / Bounding box area



*Thanks!*

*Any questions?*