# University of Moratuwa

# EN2550 - Fundamentals of Image Processing and Machine Vision

# Assignment III

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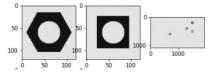
GitHub link - https://github.com/KavinduWeerasinghe/Assignment-3

The Tasks given in the assignment leads to object detection in an array of frames. To do the job this has divided into two parts,

- (a) Connected Component Analysis (Analysis of objects using contours in a single frame) and
- (b) Detecting Objects on a Synthetic Conveyor (Detecting objects considering a frame at a time in the given video)

Explanation: The procedure to get the contours is as follows...

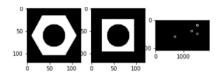
Original Images...



1. Convert the images into binary state and apply threshold.

```
hexnut_template = cv.cvtColor(hexnut_template,cv.COLOR_RGB2GRAY)
squarenut_template = cv.cvtColor(squarenut_template,cv.COLOR_RGB2GRAY)
conveyor_f100 = cv.cvtColor(coveyor_f100,cv.COLOR_RGB2GRAY)

ret1,th1 = cv.threshold(hexnut_template,0,255,cv.THRESH_BINARY_INV+cv.THRESH_OTSU)
ret2,th2 = cv.threshold(squarenut_template,0,255,cv.THRESH_BINARY_INV+cv.THRESH_OTSU)
ret3,th3 = cv.threshold(conveyor_f100,0,255,cv.THRESH_BINARY_INV+cv.THRESH_OTSU)
```



In the given code Otsu's thresholding<sup>1</sup> has been applied. Also, the values are inverted so that the background becomes black. The applied threshold value is 20.0.

Covering the small holes inside the foreground.

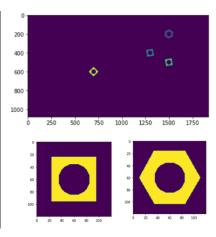
```
1 kernel=np.ones((3,3),np.uint8)
2 closing_1==cv.morphologyEx(th1,cv.MORPH_CLOSE, kernel)
3 closing_2 = cv.morphologyEx(th2, cv.MORPH_CLOSE, kernel)
4 closing_3 = cv.morphologyEx(th3, cv.MORPH_CLOSE, kernel)
```

In the given code cv.morphologyEx has been used which uses Dilation<sup>2</sup> and Erosion<sup>3</sup> in the mentioned order. This is used to remove small black spots on the foreground and white spots in the

background.

3. Connected Component Analysis





<sup>&</sup>lt;sup>1</sup> The threshold value is selected automatically by the function.

<sup>&</sup>lt;sup>2</sup> Increases the foreground of the image.

<sup>&</sup>lt;sup>3</sup> Erodes away the foreground

Given the binary image and the connectivity (provided these are 4-connected images), cv.connectedComponentsWithStats function returns,

num\_lables: Number of labels with 0 representing the background.

lables: mapping of the image's pixels according to the labels.

stats: An array consisting of several attributes of the given component.

centroids: Centroids of the selected components.

The components are color coded from the array given in lables.

The results interpreted from stats are as follows...

# Calculations of image - 1

Number of connected components = 2

# Statistics...

- 1. The leftmost (x) coordinate = 10
- 2. The topmost (y) coordinate = 16
- 3. The horizontal size of the bounding box = 101
- 4. The vertical size of the bounding box = 88
- 5. The total area (in pixels) = 4728

Centroids = [[59.33684864 59.63513234]

[59.83375635 59.22356176]]

## Calculations of image - 2

Number of connected components = 2

#### Statistics...

- 1. The leftmost (x) coordinate = 24
- 2. The topmost (y) coordinate = 24
- 3. The horizontal size of the bounding box = 72
- 4. The vertical size of the bounding box = 72
- 5. The total area (in pixels) = 3227

Centroids = [[59.5875772 59.5875772]

[59.19677719 59.19677719]]

#### Calculations of image - 3

Number of connected components = 5

#### Statistics...

- 1. The leftmost (x) coordinate = 1454
- 2. The topmost (y) coordinate = 150
- 3. The horizontal size of the bounding box = 92
- 4. The vertical size of the bounding box = 100
- 5. The total area (in pixels) = 4636
- 1. The leftmost (x) coordinate = 1459
- 2. The topmost (y) coordinate = 459
- 3. The horizontal size of the bounding box = 82
- 4. The vertical size of the bounding box = 82
- 5. The total area (in pixels) = 3087

Centroids = [[ 957.36323524 540.44416273]

 $[1499.24201898\ 199.28515962]$ 

 $[1299.18302559\ \ \, 399.18302559]$ 

 $[1499.18302559\ 499.18302559]$ 

[ 700. 600.

- 1. The leftmost (x) coordinate = 1259
- 2. The topmost (y) coordinate = 359
- 3. The horizontal size of the bounding box = 82
- 4. The vertical size of the bounding box = 82
- 5. The total area (in pixels) = 3087
- 1. The leftmost (x) coordinate = 650
- 2. The topmost (y) coordinate = 550
- 3. The horizontal size of the bounding box = 101
- 4. The vertical size of the bounding box = 101
- 5. The total area (in pixels) = 3144

Note that the first centroid in each data set is related to label 0 i.e., the background

## 4. Contour Analysis



The function cv.findContours finds and returns the respective contours with the hierarchy of contours. The retrieval mode used is, RETR EXTERNAL and the approximation method is CHAIN APPROX NONE.

RETR\_EXTERNAL: retrieves only the extreme outer contours. i.e., All the child contours and the parent contours of the outermost contours are set to -1.

CHAIN\_APPROX\_NONE: All the points belonging to the contour are stored.

# Then the task is Detecting Objects on a Synthetic Conveyor.

Explanation: The procedure to the second part of the problem is as follows...

1. Test using a single frame

The base contour is the contour used to compare with the extracted contours in conveyor\_100. This is done using cv.matchshapes function which returns similarity between the two contours calculated using hu moments. For the same sample returned value is 0. Then a threshold is applied to the ret value and measured if the respective contour is the same as the hexnut template. If so a bounding box is drawn on conveyor\_100 and Count is increased by one which is ultimately printed as the result.

By changing Base\_Ctr to Contours[1][0] the whole code can be changed to check with squarenut\_template.

# 2. Implement for the video.

This function returns the number of matching contours in a given frame and draws a bounding box on the frame around the detected objects. The procedure is same as before...

- Convert the images into binary state and apply threshold
- Covering the small holes inside the foreground
- Contour Analysis
- Apply a threshold and selecting the matching labels.

```
f = 0
frame_array = []
Running Sum=("Prev":0,"Total":0)
while cap.isOpened():
    ret, frame = cap.read()
    if not ret:
        print("Can't receive frame (stream end?). Exiting.")
        break

    f += 1
        text = 'Frame:' + str(f)
        Count, frame-Count_Ctr(frame)
        cv.putText(frame,text , (100, 100), cv.FONT_HERSHEY_COMPLEX, 1, (0,250,0), 1, cv.LINE_AA)
    text = 'Current Contours:' + str(Count)
        cv.putText(frame,text , (100, 130), cv.FONT_HERSHEY_COMPLEX, 1, (250,0,0), 1, cv.LINE_AA)
    if Running_Sum["Prov"]=Count:
        Total=Running_Sum["Total"]
    else:
        added_val=max(Count-Running_Sum["Prev"],0)
        Running_Sum["Total"] +=added_val
        Total=Running_Sum["Total"]
Running_Sum["Total"]
Running_Sum["Total"]
Running_Sum["Total"]
Running_Sum["Total"]
cv.mishow("conveyor", frame)
    if cv.waitKey(1) == ord('q'):
        break
    frame_array.append(frame)
```

This process is looped for every frame using the code to the left. Total values are calculated through a simple dictionary containing the previous number of frames and the total up to previous frame.

The resultant video can be found here. https://github.com/KavinduWeerasinghe/Assignment-3/blob/main/Result%20-%20Low%20Size.m4v