

# EN2550\_Assignment5

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## 1 EN2550 2021: Object Counting on a Convey Belt

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In this assignment, you will be counting and tracking the hexagonal nuts on a moving convey belt.

### 1.0.1 Let's first import required libraries

```
[2]: import cv2 as cv
import numpy as np
import matplotlib.pyplot as plt
from google.colab.patches import cv2_imshow

%matplotlib inline
```

### 1.0.2 Let's load and visualize the template image and the convey belt snapshot at a given time.

```
[ ]: template_im = cv.imread(r'template.png', cv.IMREAD_GRAYSCALE)
belt_im = cv.imread(r'belt.png', cv.IMREAD_GRAYSCALE)

fig, ax = plt.subplots(1,2,figsize=(10,10))
ax[0].imshow(template_im, cmap='gray')
ax[1].imshow(belt_im, cmap='gray')
plt.show()
```

## 1.1 Part-I :

Before going into the implementation, let's play with some functions.

### 1.1.1 Otsu's thresholding

Please read [thresholding](#) to get an idea about different types of thresholding and how to use them. (Please use `cv.THRESH_BINARY_INV`).

```
[ ]: th_t, img_t = cv.threshold(template_im, 0, 255, cv.THRESH_BINARY_INV + cv.  
    ↪ THRESH_OTSU)  
    #th_b, img_b = "< Your code to apply thresholding to the belt_im >"
```

### 1.1.2 Morphological closing

Carry out morphological closing to remove small holes inside the foreground. Use a  $3 \times 3$  kernel. See [closing](#) for a guide.

```
[ ]: kernel = #"< 3x3 matrix with all ones, with uint8 dtype >"  
    closing_t = cv.morphologyEx(img_t, cv.MORPH_CLOSE, kernel)  
    #closing_b = "< Your code to apply morphological closing for belt >"
```

### 1.1.3 Connected component analysis

Apply the `connectedComponentsWithStats` function ([see this](#)).

```
[ ]: retval_t, labels_t, stats_t, centroids_t = cv.  
    ↪ connectedComponentsWithStats(closing_t)  
    retval_b, labels_b, stats_b, centroids_b = #"< Your code to perform connected_  
    ↪ component analysis for closing_b >"
```

- How many connected components are detected in each image?
- What are the statistics? Interpret these statistics.
- What are the centroids?

### 1.1.4 Contour analysis

Use `findContours` function to retrieve the *extreme outer* contours. ([see](#) for help and [see](#) for information.)

Display these contours.

```
[ ]: contours_t, hierarchy_t = cv.findContours(closing_t, cv.RETR_TREE, cv.  
    ↪ CHAIN_APPROX_SIMPLE)  
    contours_b, hierarchy_b = #"< Your code to perform counter analysis for_  
    ↪ closing_b >"
```

```
[ ]: # Visualizing contours  
    im_contours_belt = np.zeros((belt_im.shape[0], belt_im.shape[1], 3), np.uint8)
```

```

conts = cv.drawContours(im_contours_belt, contours_b, -1, (0,255,0), 3).
↳astype('uint8')
plt.imshow(conts)

```

### 1.1.5 Count the number of matching hexagonal nuts in belt.png.

Use the `matchShapes` function as shown in [examples](#) to match contours in the belt image with that in the template.

Get an idea about the value output by the `cv.matchShapes` when both the template and the reference image have the same shape. Understand the given code snippet.

```

[ ]: label = 1 # remember that the label of the background is 0
belt = ((labels_b >= label)*255).astype('uint8')
belt_cont, template_hierarchy = cv.findContours(belt, cv.RETR_EXTERNAL, cv.
↳CHAIN_APPROX_SIMPLE)
for j,c in enumerate(belt_cont):
    print(cv.matchShapes(contours_t[0], c, cv.CONTOURS_MATCH_I1, 0.0))

```

## 1.2 Part - II

### 1.2.1 Frame tracking through image moments.

Use the `cv.contourArea()`, see [this](#) and calculate the the area of the `contours_b[1]`

```

[ ]: ca = '<Your code goes here>'

```

Use the `cv.moments` to extract the x and y coordinates of the centroid of `contours_b[1]`.

```

[ ]: M = '<Your code goes here>'
cx, cy = '<Your code goes here>'

```

Make a variable called `count` to represent the number of contours and set it to the value 1. Make an np array `[cx, cy, ca, count]` and name this as `object_prev_frame`

```

[ ]: object_prev_frame = '< Your code here >'

```

Similarly, you can create the `object_curr_frame`(to describe the current values) and define the threshold `delta_x` to check whether the corresponding element of both the `object_curr_frame` and `object_prev_frame` are less than the `delta_x`. You can set `delta_x` as 15 or so. (Here the `delta_x` can be thought of as the movement of the `cx` from frame to frame)

```

[ ]: delta_x = '< Your code goes here >'

```

## 1.3 Part - III

- 1.3.1 1. Implement the function `get_indexed_image`, which takes an image as the input, performs thresholding, closing, and connected component analysis and return retval, labels, stats, centroids. (Grading)

```
[ ]: def get_indexed_image(im):  
    """ Thresholding, closing, and connected component analysis lumped  
    """  
  
    '< Your code goes here. Approximately 4 lines >'  
  
    return retval, labels, stats, centroids
```

- 1.3.2 2. Implement the function `is_new`, which checks the dissimilarity between 2 vectors. (Grading)

```
[ ]: def is_new(a, b, delta, i):  
    """ Vector Dissimilarity with an Array of Vectors  
    Checks if vector b is similar to a one or more vectors in a outside the_  
    →tolerances specified in delta.  
    vector i specifies which elements in b to compare with those in a.  
    """  
  
    'Check whether the absolute different between all the elements of ith_  
    →column of each array is greater than the ith delta value (See thee example_  
    →in the next cell)'  
  
    return None
```

```
[ ]: # check is_new expected answer False  
  
a = np.array([[1.36100e+03, 5.53000e+02, 5.99245e+04, 2.00000e+00],  
             [7.61000e+02, 4.53000e+02, 5.99385e+04, 1.00000e+00],  
             [1.55200e+03, 2.43000e+02, 6.00585e+04, 3.00000e+00]])  
b = np.array([7.51000e+02, 4.53000e+02, 5.99385e+04, 3.00000e+00])  
delta = np.array([delta_x])  
i = np.array([0])  
  
assert is_new(a, b, delta, i) == False, " Check the function "
```

**1.3.3 3.** If the array `a` is in the shape of (number of nuts , `len(object_prev_frame)`) ( i.e. array `a` is made by stacking all the `object_prev_frame` for each frame. If `b` is in the form of [`cx`, `cy`, `ca`, `count`], write the function `prev_index` to find the index of a particular nut in the previous frame. (Grading)

```
[ ]: def prev_index(a, b, delta, i):
      """ Returns Previous Index
      Returns the index of the appearance of the object in the previous frame.
      (See thee example in the next cell)
      """
      index = -1
      '< Your code goes here >'
      return index
```

```
[ ]: # check prev_index expected answer 1
a = np.array([[1.36100e+03, 5.53000e+02, 5.99245e+04, 2.00000e+00],
              [7.61000e+02, 4.53000e+02, 5.99385e+04, 1.00000e+00],
              [1.55200e+03, 2.43000e+02, 6.00585e+04, 3.00000e+00]])
b = np.array([7.51000e+02, 4.53000e+02, 5.99385e+04, 3.00000e+00])
delta = np.array([delta_x])
i = np.array([0])

assert prev_index(a,b,delta,i) == 1, " Check the function "
```

You can use following code snippet load and access each frame of a video

```
[ ]: cap = cv.VideoCapture('conveyor_many_frame.mp4') # give the correct path here
while cap.isOpened():
    ret, frame = cap.read()
    if not ret:
        print("Can't receive frame (stream end?). Exiting ...")
        break
    cv2.imshow(frame)
    if cv.waitKey(1) == ord('q'):
        break

cap.release()
cv.destroyAllWindows()
```

**1.3.4 3.** Implement a code to detect hexagonal nuts in a moving convey belt. (Grading)

## 1.4 Steps:

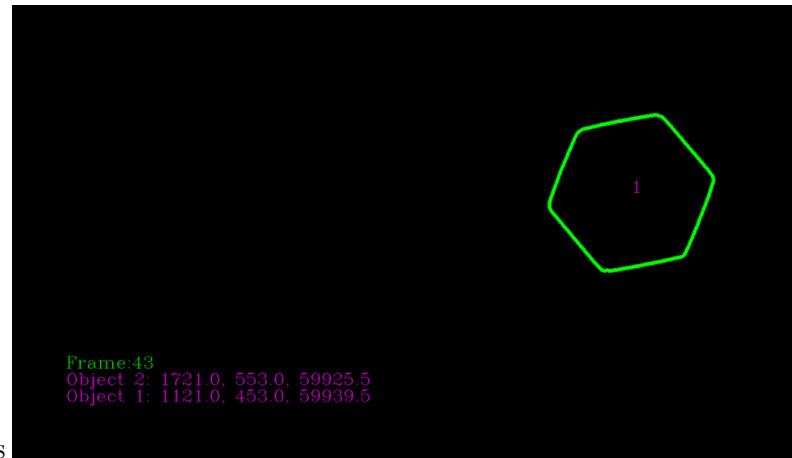
1. Use the above code snippet to access each frame and remember to convert the frame into grey scale. Name the variable as `grey`

2. Call `get_indexed_image` and extract `retval`, `labels`, `stats`, `centroids`.
3. Find contours of all nuts present in a given frame of the belt.
4. Initiate a 3-D array with zeros to draw contours. Call this `im_contours_belt`
5. Draw each contour. Use `cv.drawContours`. [See this](#)

## 1.5 Object detection and tracking

For each contour of the belt frame,

1. Use `is_new` and `prev_index` functions to track each frame and get the indices of each nut.
2. Write a code to detect and track hexagonal nuts in each frame.
3. You may refer, [annotation](#) to understand how to add texts and labels to each frame.



4. Output for a random frame would be as follows

**Hint:** If you are thresholding on areas (template and contour) you can use 500 as the threshold. You can set the matching threshold to be 0.5 and experiment

```
[ ]: '< Your code goes here >'
```