

# Image Segmentation & Contours

# Finding Contours

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Contours can be explained simply as a curve joining all the continuous points (along the boundary), having same color or intensity.

The contours are a useful tool for shape analysis and object detection and recognition.

- For better accuracy, use binary images. So before finding contours, apply threshold or canny edge detection.
- Since OpenCV 3.2, findContours() no longer modifies the source image but returns a modified image as the first of three return parameters.
- In OpenCV, finding contours is like finding white object from black background. So remember, object to be found should be white and background should be black.

# Finding Contours

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There are three arguments in **cv2.findContours()** function,

1. first one is source image,
  2. second is contour retrieval mode, there are 4 options:
    - a. [RETR\\_LIST](#),
    - b. [RETR\\_EXTERNAL](#),
    - c. [RETR\\_CCOMP](#)
    - d. [RETR\\_TREE](#)
  3. third is contour approximation method.
- It outputs a modified image, the contours and hierarchy. contours is a Python list of all the contours in the image.
  - Each individual contour is a Numpy array of (x,y) coordinates of boundary points of the object.



# Drawing contours

To draw the contours, **cv2.drawContours** function is used.

1. Its first argument is source image,
2. Second argument is the contours which should be passed as a Python list,
3. Third argument is index of contours (useful when drawing individual contour.

To draw all contours, pass -1)

4. Remaining arguments are color, thickness.



# Sorting contours

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Once we find the contours of an image, if the order is important we will need to sort them.

**We can sort contours by:**

- **Orientation:** for example left to right
- **Area:** by finding the area inside the contours we can sort to get the biggest ones or smaller ones (normally discarding small contours is a way of getting rid of noise in the image)

# Moments

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Image moments help you to calculate some features like center of mass of the object and area of the object.

$$C_x = \frac{M_{10}}{M_{00}} \text{ and } C_y = \frac{M_{01}}{M_{00}}$$

- Read about Image moment here:
  - [https://en.wikipedia.org/wiki/Image\\_moment](https://en.wikipedia.org/wiki/Image_moment)



# Contour Perimeter

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It is also called arc length.

It can be found out using *cv2.arcLength()* function.

- First argument is the contour
- Second argument specify whether shape is a closed contour (if passed True), or just a curve.

*perimeter = cv2.arcLength(cnt,True)*

# Contour Area

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Contour area is given by the function:

1. *cv2.contourArea()*

*a. It receives only one parameter: the contour*

2. From moments: **M['m00']**.



# Approximating Contours

- It approximates a contour shape to another shape with less number of vertices depending upon the precision we specify.
- Suppose you are trying to find a square in an image, but due to some problems in the image, you didn't get a perfect square, but a "bad shape" you can approximate the shape with the function `cv2.approxPolyDP` the second argument is called epsilon, which is maximum distance from contour to approximated contour.
- It is an accuracy parameter. A wise selection of epsilon is needed to get the correct output. Depending on the epsilon value you will get different results as shown here



# Advice for choosing the Epsilon value

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- Small values give precise approximation
- Large values give more generic approximation
- A good rule of thumb is less than 5% of the contour perimeter

```
accuracy = 0.03  
perimeter = cv2.arcLength(c, True)  
epsilon = accuracy * perimeter  
approx = cv2.approxPolyDP(c, epsilon, True)
```

# Counting lines in a polygon

Once we got the image contour polygon approximations we can count the number of lines in each polygon by simply doing this:

*len(approximation)*

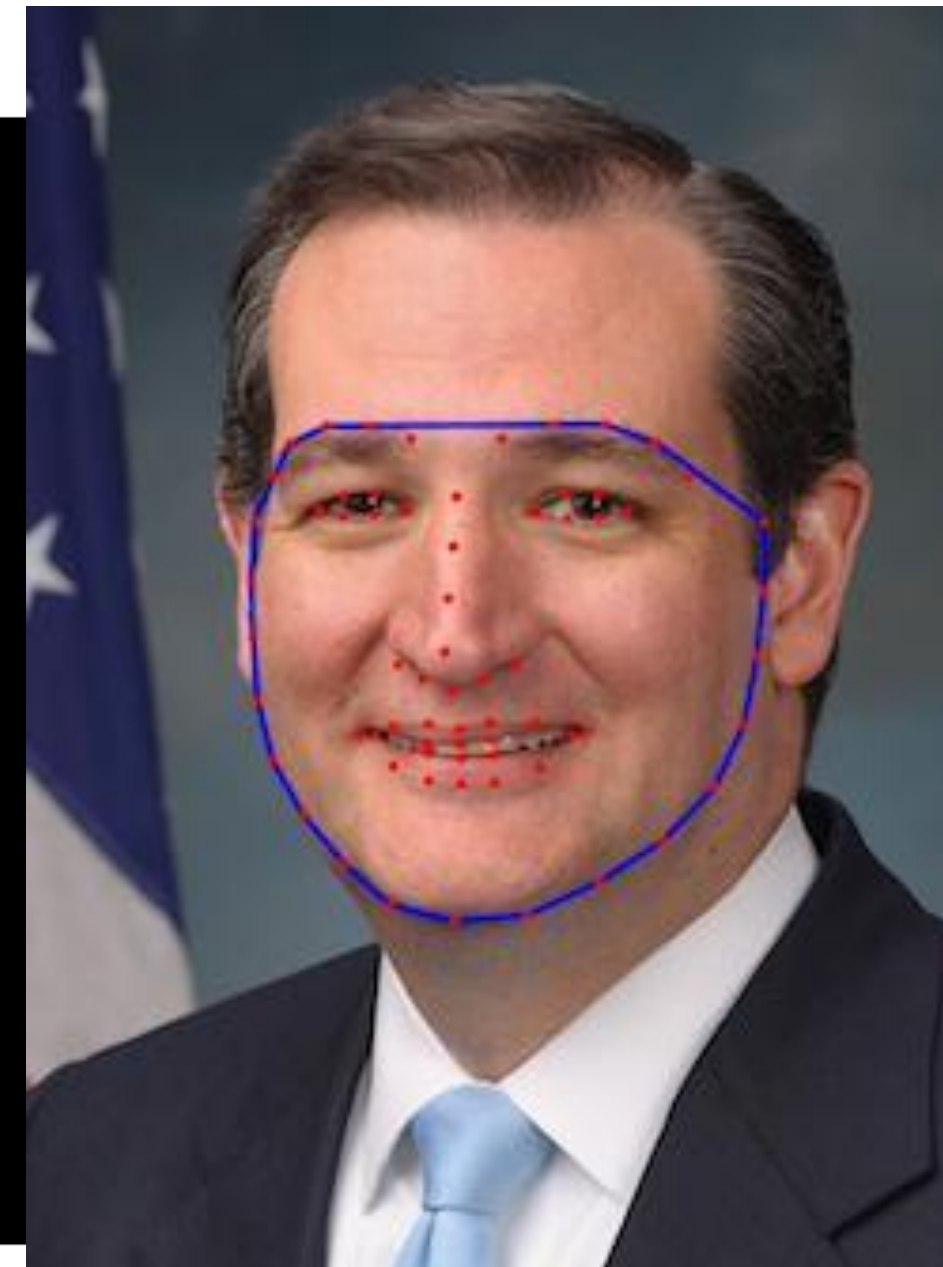
```
approx = cv2.approxPolyDP(c, epsilon, True)

cv2.drawContours(house, [approx], 0, (0, 255, 0), 2)
showImg(house, f'Found {len(approx)} lines in this polygon')
```



# Convex Hull

The *Convex Hull* of a shape or a group of points is a tight fitting convex boundary around the points or the shape.



<https://www.learnopencv.com/convex-hull-using-opencv-in-python-and-c/>

[https://docs.opencv.org/3.1.0/dd/d49/tutorial\\_py\\_contour\\_features.html](https://docs.opencv.org/3.1.0/dd/d49/tutorial_py_contour_features.html)

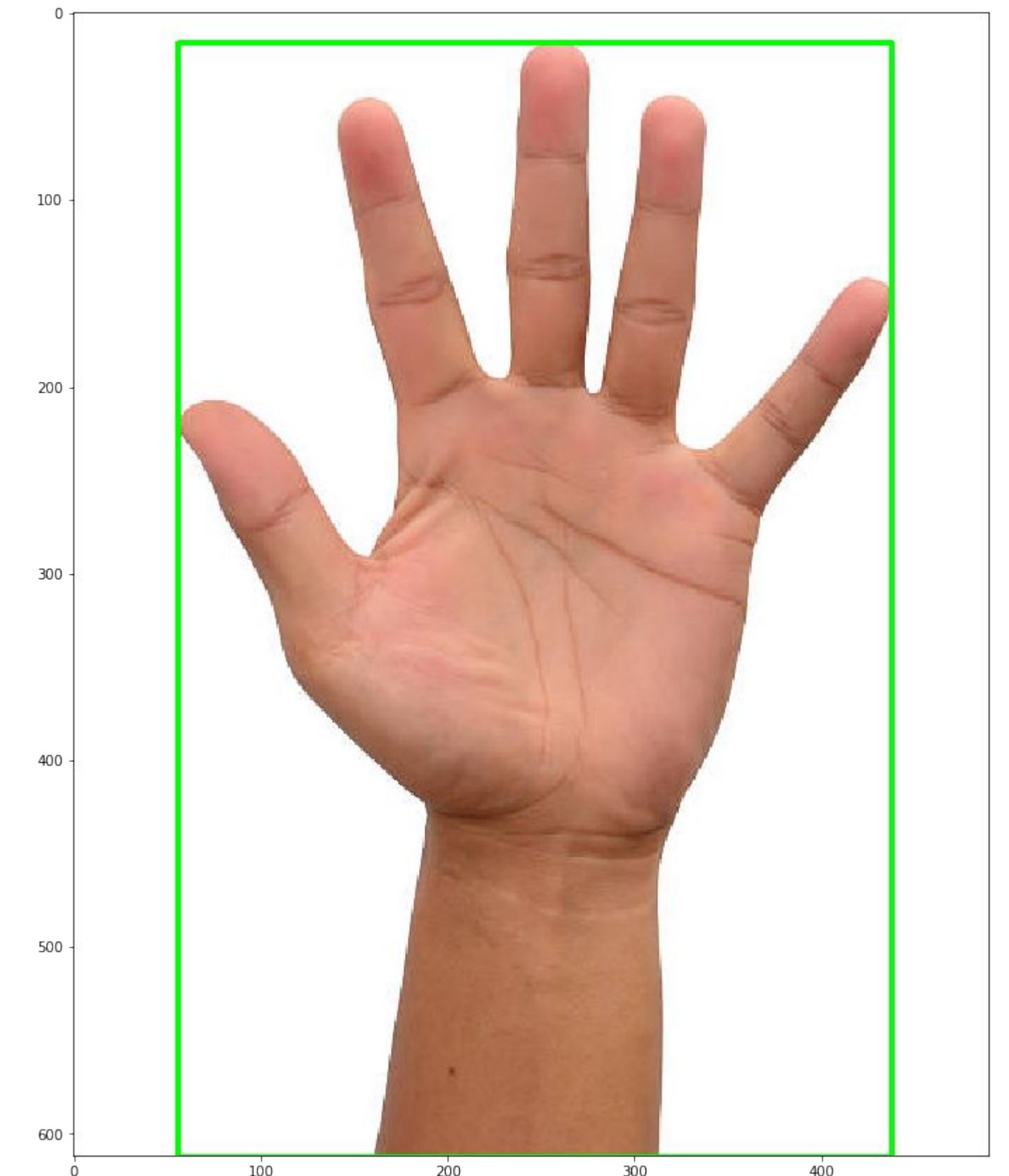
# Bounding Rectangle

Is the minimum fitting rectangle around an object, it does not take into consideration rotation the rectangle coordinates can be found with:

```
x,y,w,h = cv2.boundingRect(cnt)
```

Where *cnt* are the contours of the image

Interesting fact: it also works with a threshed image

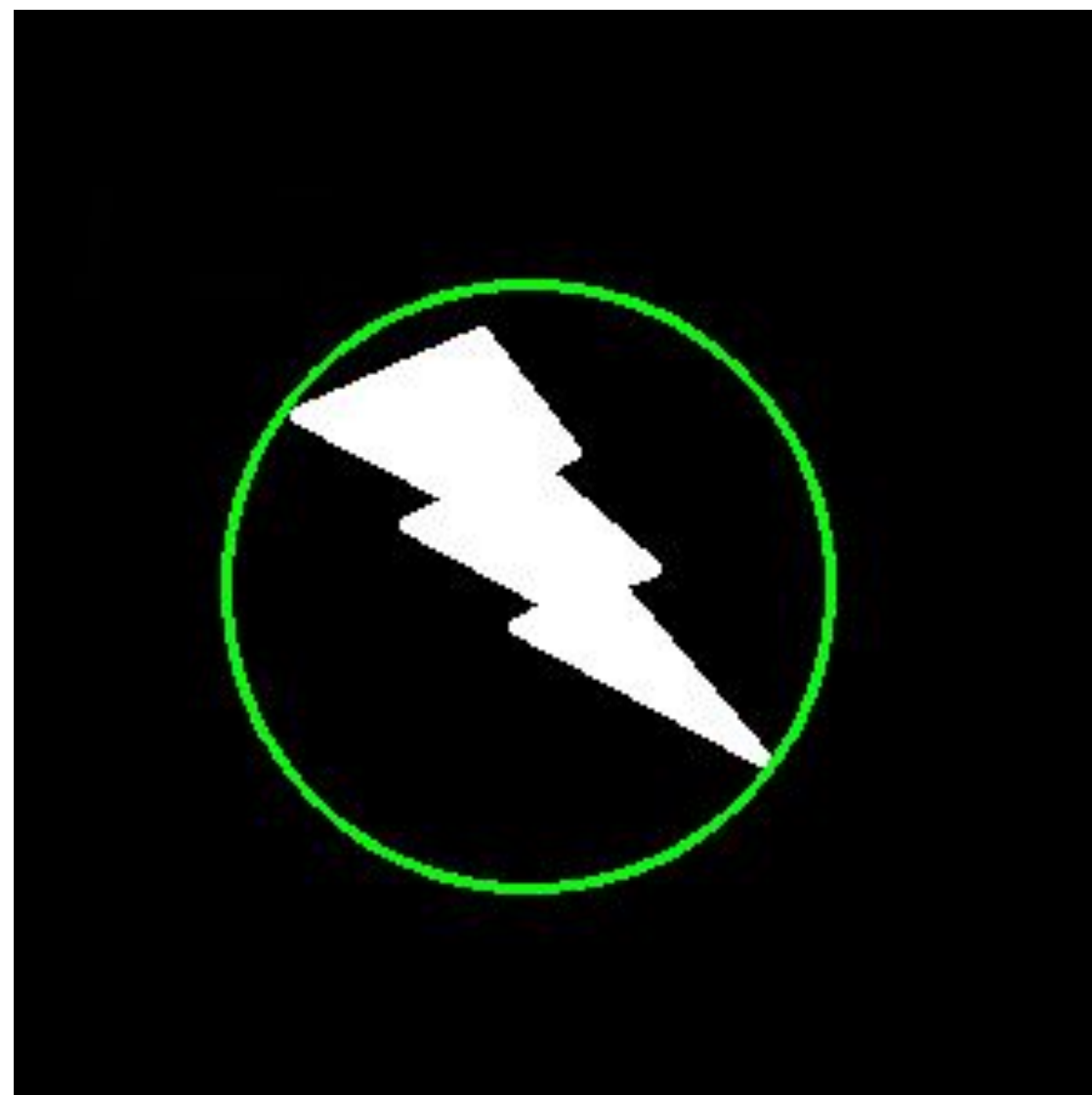




# Minimum Enclosing Circle

The circumcircle of an object using the function `cv2.minEnclosingCircle()`

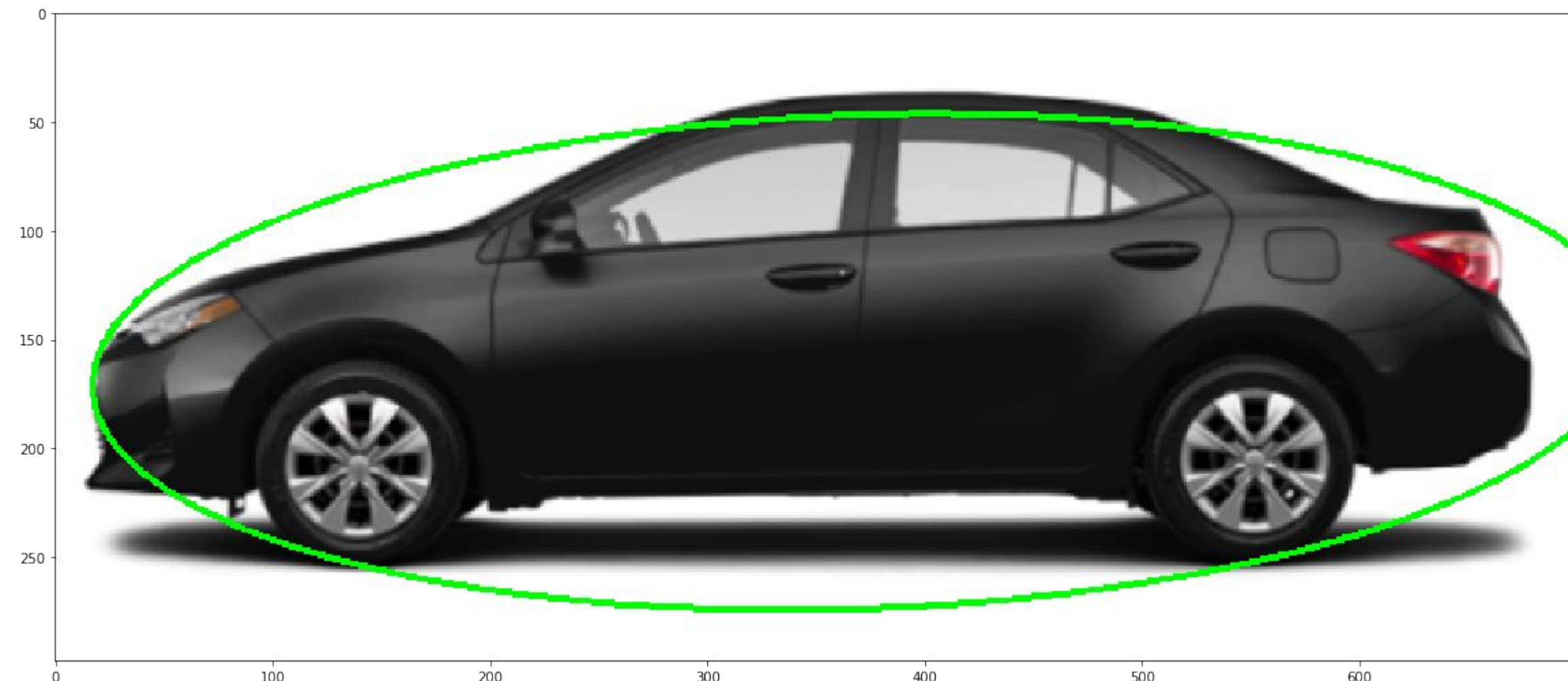
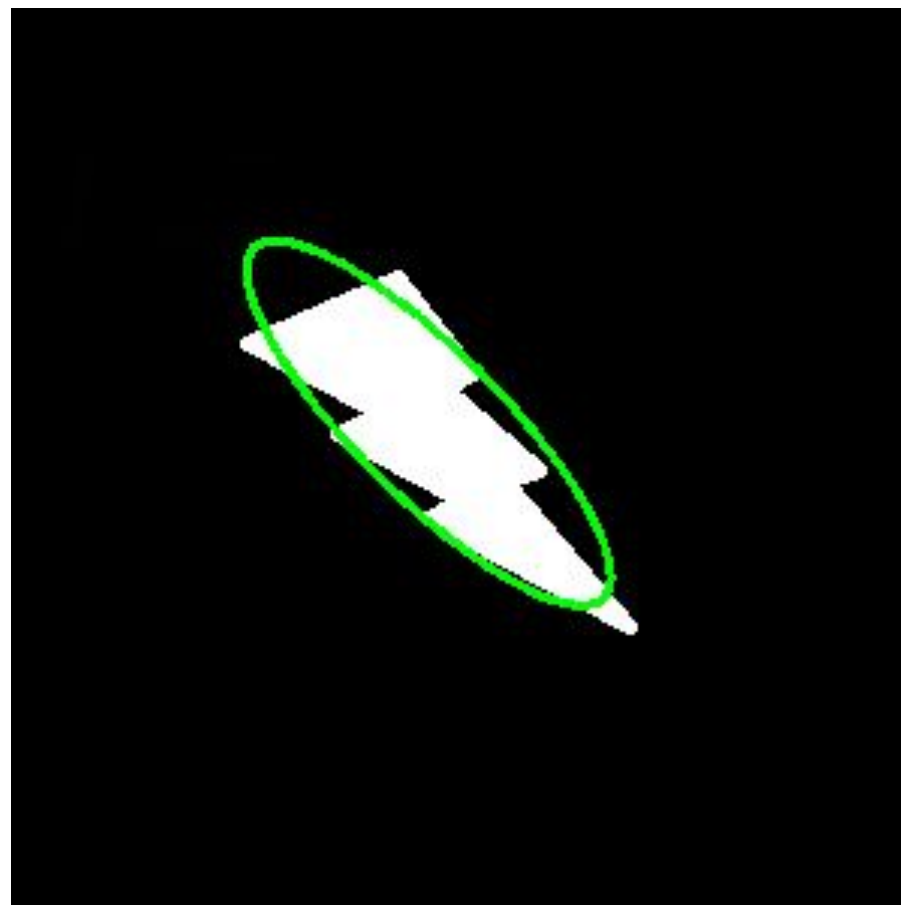
It is a circle which completely covers the object with minimum area.





# Fitting an ellipse

- To find the fitting ellipse of an object, it will be rotated as needed:  
*ellipse = cv2.fitEllipse(cnt)*
- It also receives the object contour as an input parameter.
- Make sure it is one contour and not an array of contours, you will get an error otherwise

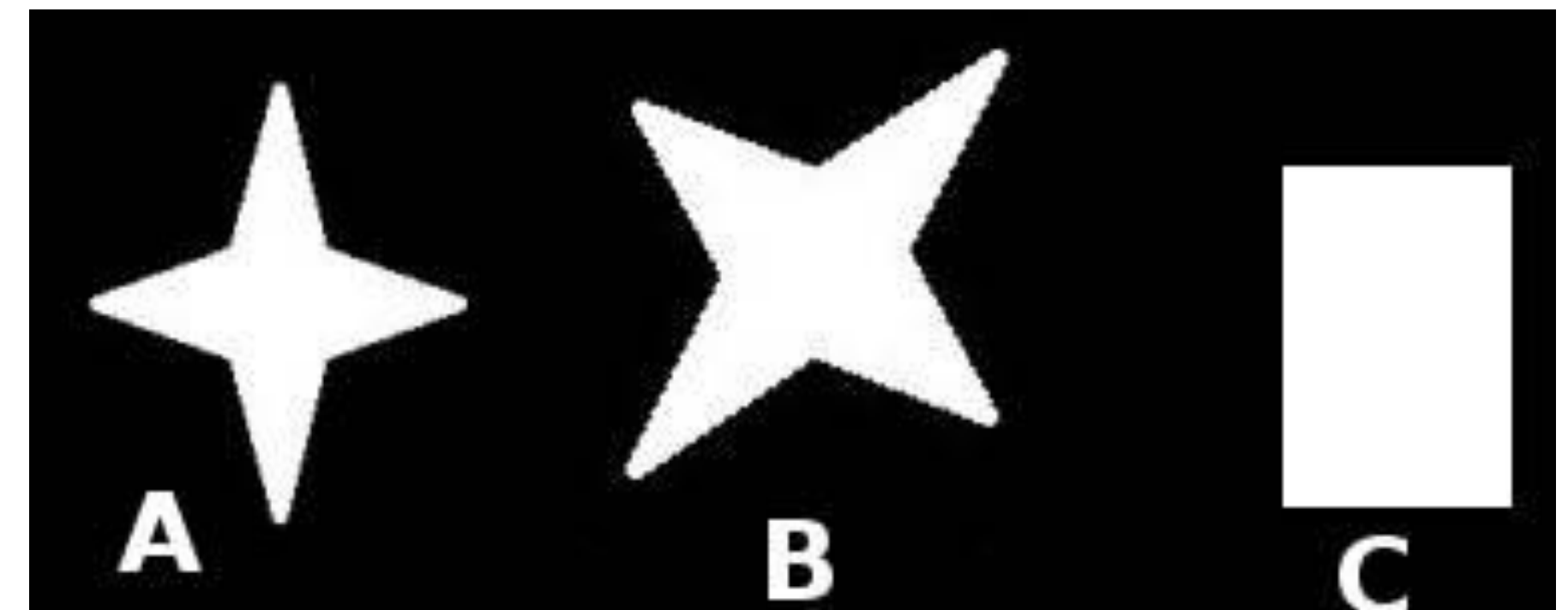


# Matching contour shapes

OpenCV comes with a function `cv2.matchShapes()` which enables us to compare two shapes, or two contours and returns a metric showing the similarity.

The lower the result, the better match it is. It is calculated based on the hu-moment values.

Example: using A from the image as the template we get the following scores when comparing them



Matching Image A with itself = 0.0

Matching Image A with Image B = 0.001946

Matching Image A with Image C = 0.326911

# Line Detection with Hough Transform

We can detect lines with Hough Line Transform:

- The Hough Line Transform is a transform used to detect straight lines.
- To apply the Transform, first an edge detection pre-processing is desirable

*lines = cv.HoughLines(dst, 1, np.pi / 180, 150)*

- *lines*: A vector that will store the parameters  $(r, \theta)$  of the detected lines In [ ]:
- *dst*: Output of the edge detector. It should be a grayscale image (although in fact it is a binary one)
- *rho*: The resolution of the parameter  $r$  in pixels. We use 1 pixel.
- *theta*: The resolution of the parameter  $\theta$  in radians. We use 1 degree ( $CV\_PI/180$ )
- *threshold*: The minimum number of intersections to *"detect"* a line

## Probabilistic Hough Transform (adds 2 new parameters)

*lines = cv2.HoughLinesP(dst, 1, np.pi/180, 100, minLineLength, maxLineGap)*

- *minLineLength* - Minimum length of line. Line segments shorter than this are rejected.
- *maxLineGap* - Maximum allowed gap between line segments to treat them as single line.



# Circle Detection

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OpenCV implements another Hough Transform version to detect Circles:

*cv2.HoughCircles(image, method, dp, minDist)*

- *src\_gray*: Input image (grayscale)
- *circles*: A vector that stores sets of 3 values:  $x_{\{c\}}$ ,  $y_{\{c\}}$ ,  $r$  for each detected circle.
- *method*: Define the detection method. Currently CV\_HOUGH\_GRADIENT <sup>In [ ]:</sup> this is the only one available in OpenCV
- *dp = 1*: The inverse ratio of resolution
- *min\_dist = src\_gray.rows/8*: Minimum distance between detected centers
- *param\_1 = 200*: Upper threshold for the internal Canny edge detector
- *param\_2 = 100\**: Threshold for center detection.
- *min\_radius = 0*: Minimum radio to be detected. If unknown, put zero as default.
- *max\_radius = 0*: Maximum radius to be detected. If unknown, put zero as default

# You may also want to check

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- Blob detection
- Fitting a line
- Fitting with a rotated rectangle

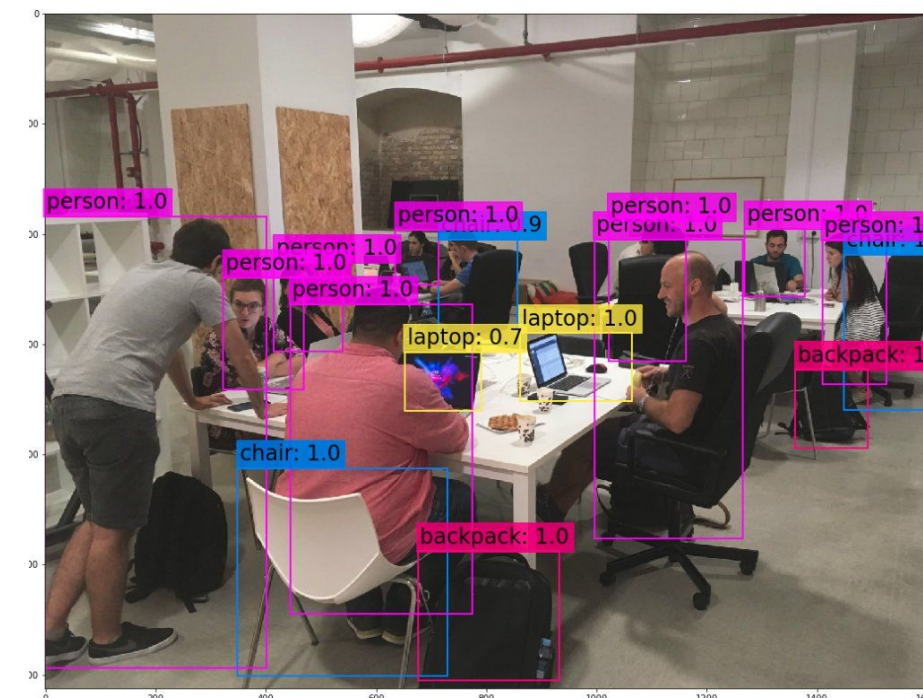
In [ ]:

## Business Case for Sustainability



# George Studenko

# Software Engineer and AI Enthusiast



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