

# Orientation Computation

$$\tan 2\phi \triangleq \frac{b}{a-c}$$

$$a = \iint_{\Omega} (x - \bar{x})^2 B(x, y) dx dy \quad \dots (2)$$

$$b = \iint_{\Omega} 2(x - \bar{x})(y - \bar{y}) B(x, y) dx dy \quad \dots (3)$$

$$c = \iint_{\Omega} (y - \bar{y})^2 B(x, y) dx dy \quad \dots (4)$$

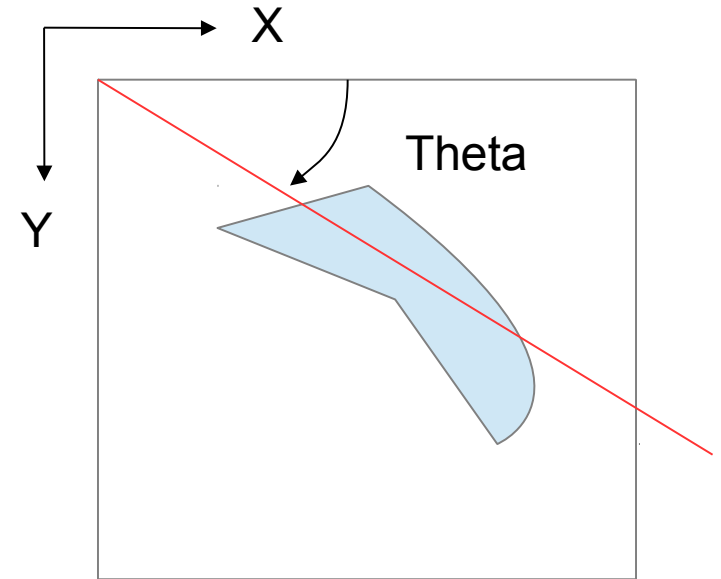
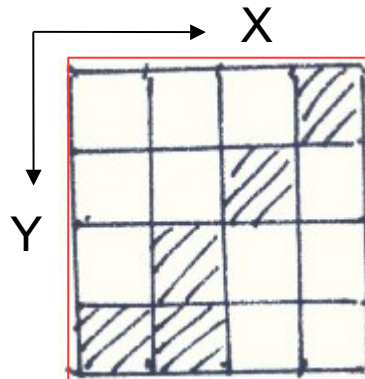
Example: See my handout

$$a = 7$$

$$b = -8$$

$$c = 6$$

$$\text{Theta} = -41.4375$$



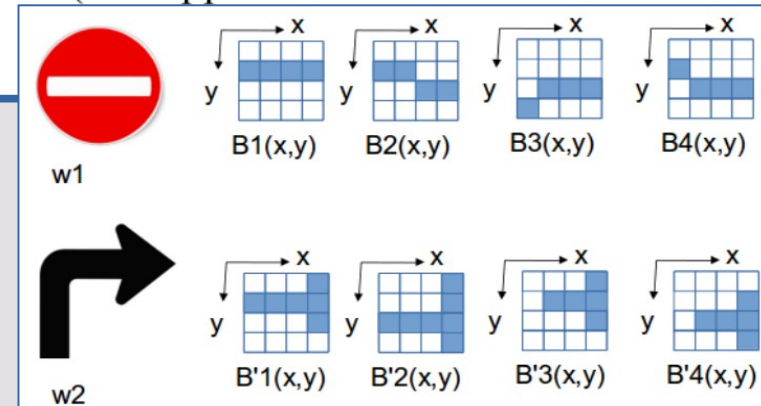
Reference: Robot Vision, by  
BPK, Horn, Chapter 3, pp. 46-  
64

Note: my hand calculation use  
integer, when have access to  
computer, use Float!  
( $\bar{x} = 2.8$  changed to 3, and  
 $\bar{y} = 2.4$  changed to 2)

# Computation of Moments

**QUESTION 3** (15 Points) Given two traffic signs and their binarized images taken from different conditions as shown in the following figure, design a machine learning technique by answering the following questions:

5.1 (5 pts) Based on given 2 classes of image, find moments  $m_{01}$ ,  $m_{10}$  for each of the image, and form feature vector space with your computation result (see Appendix for  $m_{pq}$  definition if needed).



	$w_{11}$	$w_{12}$	$w_{13}$	$w_{14}$	$w_{21}$	$w_{22}$	$w_{23}$	$w_{24}$
$u_y/m_{01}$	2/5	2/5	3/5	2/5	2/10	(18/6)/0	(8/5)/(2/5)	3/0
$u_x/m_{10}$	2.5/0	2.5/0	2.5/0	2.5/0	(18/6)/0	(22/7)/(1/7)	(17/5)/0	(17/5)/0

$$m_{10,23} = (4 - \frac{17}{5}) + (2 - \frac{17}{5}) + (3 - \frac{17}{5}) + (4 - \frac{17}{5})$$

$$= \frac{6}{5} + (-\frac{7}{5}) + (-\frac{2}{5}) + \frac{3}{5} = 0$$
  

$$m_{10,24} = (4 - \frac{17}{5}) \times 2 + (2 - \frac{17}{5}) + (3 - \frac{17}{5}) + (4 - \frac{17}{5}) = 0$$
  

(2) PART II. K-means Cluster Algorithm.  
 Use  $(u_x, u_y)$  to form feature vectors.  
 Then, apply OpenCV. K-mean

For Class 2,  $w_{aj}, j=1,2,3,4$

$$u_{x21} = [4 + (1+2+3+4) + 4] \frac{1}{A} = 18/6$$

$$u_{x22} = [4 \times 3 + (1+2+3+4)] \frac{1}{A} = 22/7$$

$$u_{x23} = [4 \times 2 + (2+3+4)] \frac{1}{A} = 17/5$$

$$u_{x24} = [4 \times 2 + (2+3+4)] \frac{1}{A} = 17/5$$
  

$$m_{10,21} = [(4 - \frac{18}{6}) + (1 - \frac{18}{6}) + (2 - \frac{18}{6}) + (3 - \frac{18}{6}) + (4 - \frac{18}{6})] \frac{1}{A}$$

$$= [\frac{6}{6} - \frac{12}{6} - \frac{6}{6} + 0 + \frac{6}{6} + \frac{6}{6}] \frac{1}{6} = 0$$
  

$$m_{10,22} = [(4 - \frac{22}{7}) \times 3 + (1 - \frac{22}{7}) + (2 - \frac{22}{7}) + (3 - \frac{22}{7}) + (4 - \frac{22}{7})] \frac{1}{A}$$

$$= (-\frac{45}{7} - \frac{15}{7} - \frac{8}{7} - \frac{1}{7} + \frac{6}{7}) \frac{1}{A} = (-\frac{63}{7}) \frac{1}{A} = (-9) \frac{1}{A}$$

# Python Example For Moments

First, let's find contours, by openCV.org definition, “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.”

Note: In OpenCV, object to be found should be white and background should be black when applying contour finding function.

```
cv2.findContours(thresh,cv2.RETR_TREE,cv2.CHAIN_APPROX_SIMPLE)
```

The arguments: the 1st is source image, 2nd is contour retrieval mode, 3rd is contour approximation method. And it outputs 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.

```
im = cv2.imread('test.jpg')
imgray = cv2.cvtColor(im,cv2.COLOR_BGR2GRAY)
ret,thresh = cv2.threshold(imgray,127,255,0)
im2, contours, hierarchy = cv2.findContours(thresh,cv2.RETR_TREE,cv2.CHAIN_APPROX_SIMPLE)
```

# Contours

actly what a `contour` is. A `contour` is a list of points that represent, in one way or another, a curve in an image. This representation can be different depending on the circumstance at hand. There are many ways to represent a curve. `Contours` are represented in OpenCV by sequences in which every entry in the sequence encodes information about the location of the next point on the curve. We will dig into the details of such

Reference: Learning OpenCV, pp. 250

The function `cvFindContours()` computes `contours` from binary images. It can take images created by `cvCanny()`, which have edge pixels in them, or images created by functions like `cvThreshold()` or `cvAdaptiveThreshold()`, in which the edges are implicit as boundaries between positive and negative regions.\*

# Contours Mode Variable

```
findContours( image_canny, contours, hierarchy,  
→ RETR_CCOMP, CHAIN_APPROX_SIMPLE );
```

The mode variable can be set to any of four options: CV\_RETR\_EXTERNAL, CV\_RETR\_LIST, CV\_RETR\_CCOMP, or CV\_RETR\_TREE. The value of mode indicates to cvFindContours() exactly what **contours** we would like found and how we would like the result presented to us. In particular, the manner in which the tree node variables (h\_prev, h\_next, v\_prev, and v\_next) are used to “hook up” the found **contours** is determined by the value of mode. In Figure 8-3, the resulting topologies are shown for all four possible values of mode. In every case, the structures can be thought of as “levels” which are related by the “horizontal” links (h\_next and h\_prev), and those levels are separated from one another by the “vertical” links (v\_next and v\_prev).

Retrieves only the extreme outer contours. It sets hierarchy[i][2]=hierarchy[i][3]=-1 for all the contours.

CV\_RETR\_EXTERNAL

first = c0

CV\_RETR\_TREE

first = c0

CV\_RETR\_CCOMP

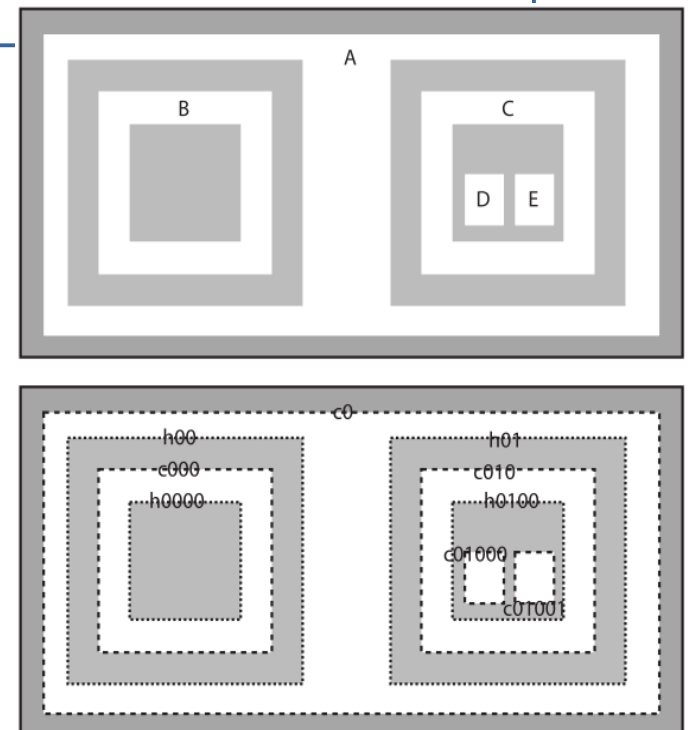
first = c01000→c01001→c010→c000→c0  
h0100 h0000 h01→h00

h00→h01  
c000 c010  
h0000 h0100  
c01000→c01001

CV\_RETR\_LIST

first = c01000→c01001→h0100→c010→c000→h01→h00→c0

retrieves all contours without  
any hierarchical relationships.



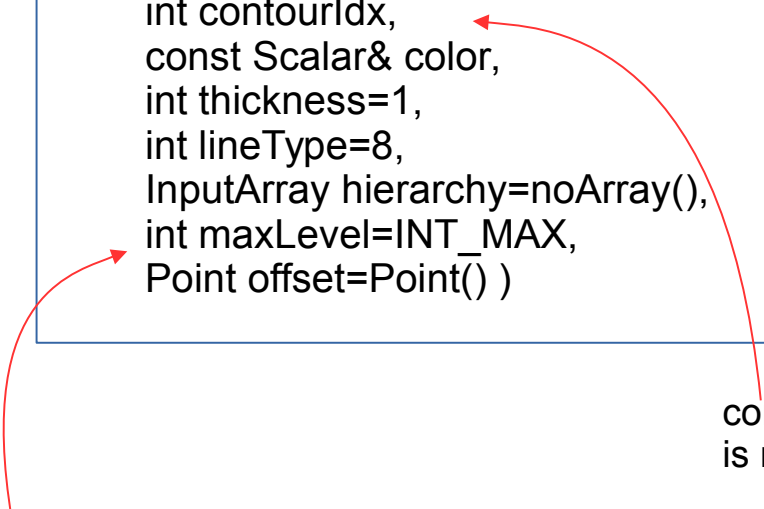
CV\_RETR\_CCOMP  
Retrieves all the contours  
into two-level hierarchy,  
top-level for external  
boundaries and the 2nd  
level for the holes.



# Contours Mode Variable

<http://opencvexamples.blogspot.com/2013/09/find-contour.html>

```
void drawContours(  
    InputOutputArray image,  
    InputArrayOfArrays contours,  
    int contourIdx,  
    const Scalar& color,  
    int thickness=1,  
    int lineType=8,  
    InputArray hierarchy=noArray(),  
    int maxLevel=INT_MAX,  
    Point offset=Point() )
```



hierarchy – Output vector, containing contour topology. It has as many elements as the number of contours. For each i-th contour contours[i], the elements hierarchy[i][0], hierarchy[i][1], hierarchy[i][2], and hierarchy[i][3] are set to 0-based indices in contours of the next and previous contours at the same hierarchical level, the first child contour and the parent contour, respectively. If for the contour i there are no next, previous, parent, or nested contours, the corresponding elements of hierarchy[i] will be negative.

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

maxLevel – Maximal level for drawn contours. If it is 0, only the specified contour is drawn. If it is 1, the function draws the contour(s) and all the nested contours. If it is 2, the function draws the contours, all the nested contours, all the nested-to-nested contours, and so on. This parameter is only taken into account when there is hierarchy available.

# Canny Edge Detector

[https://en.wikipedia.org/wiki/Canny\\_edge\\_detector](https://en.wikipedia.org/wiki/Canny_edge_detector)

For  $(2K+1)$  by  $(2K+1)$  Gaussian kernel:

$$H_{ij} = \frac{1}{2\pi\sigma^2} \exp\left(-\frac{(i - (k+1))^2 + (j - (k+1))^2}{2\sigma^2}\right);$$

Canny edge detection algorithm in 5 steps:

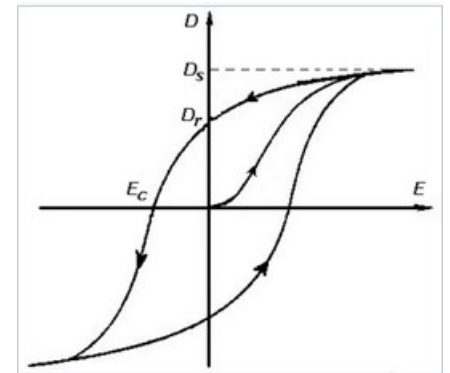
1. Apply Gaussian filter to remove noise;
2. Compute intensity gradients;
3. Apply non-maximum suppression to get rid of spurious response to edge detection;
4. Apply double thresholds to potential edges
5. Track edge by hysteresis: Finalize the detection of edges by suppressing weak and not connected edges.

$$G = \sqrt{G_x^2 + G_y^2}$$

$$\Theta = \text{atan2}(G_y, G_x), \quad \text{Edge gradient}$$

The edge direction is rounded to one of 4 angles ( $0^\circ$ ,  $45^\circ$ ,  $90^\circ$  and  $135^\circ$ ). An edge direction will be set to a specific angle values, for instance  $\theta$  in  $[0^\circ, 22.5^\circ]$  or  $[157.5^\circ, 180^\circ]$  maps to  $0^\circ$ .

Hysteresis is the dependence of the state of a system on its history. For example, a magnet may have more than one possible magnetic moment in a given magnetic field, depending on its past. Plots of a single component of the moment often form a loop or hysteresis curve, different values of one variable depending on the direction of change of another variable.

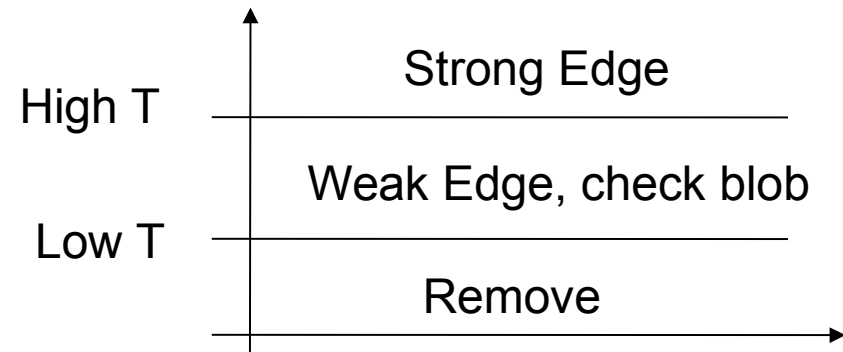


<https://en.wikipedia.org/wiki/Hysteresis>

# Double Threshold And Blob

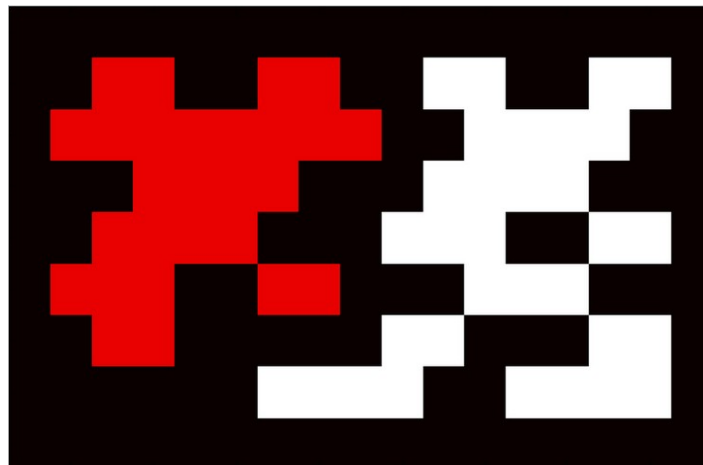
[https://en.wikipedia.org/wiki/Canny\\_edge\\_detector](https://en.wikipedia.org/wiki/Canny_edge_detector)

Double threshold: some edge pixels are caused by noise and color variation, e.g., spurious responses, so remove those with weak gradient and keep those with a high gradient by selecting high and low threshold values. If an edge pixel's gradient value is higher than the high threshold value, it is marked as a strong edge pixel. If an edge pixel's gradient value is smaller than the high threshold but higher than the low threshold, it is marked as a weak edge pixel. If an edge pixel's value is smaller than the low threshold value, it will be suppressed.



Blob algorithm: similar to flood-fill algorithm, based on 4-connected or 8-connected neighbors

[https://en.wikipedia.org/wiki/Connected-component\\_labeling](https://en.wikipedia.org/wiki/Connected-component_labeling)

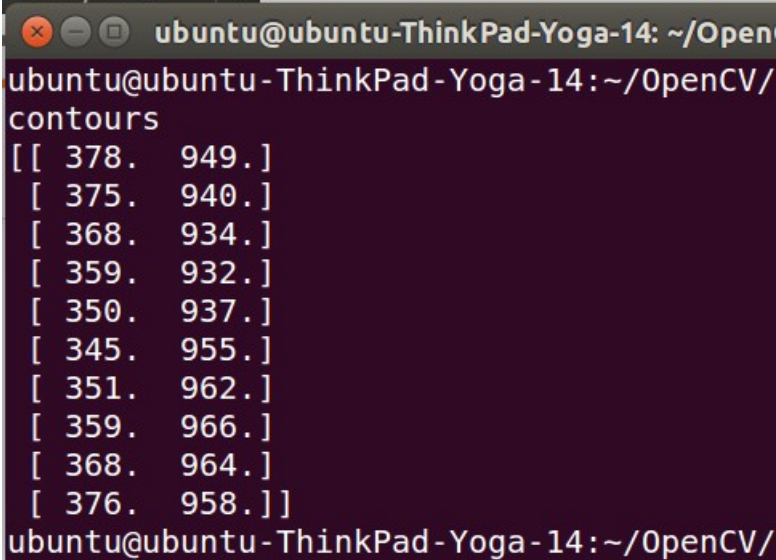




# Contours Data Type In Python

<https://stackoverflow.com/questions/20928944/create-contour-from-scratch-in-python-opencv-cv2>

```
#-----#  
# program: contour-test.py #  
# tested by: HL #  
import cv2, numpy  
contour = numpy.array( [  
    (378, 949), (375, 940), (368, 934),  
    (359, 932), (350, 937), (345, 955),  
    (351, 962), (359, 966), (368, 964),  
    (376, 958) ], numpy.float32 )  
cv2.isContourConvex(contour)  
print ('contours')  
print (contour)
```



A terminal window screenshot showing the execution of the program. The window title is 'ubuntu@ubuntu-ThinkPad-Yoga-14: ~/OpenCV/'. The prompt is 'ubuntu@ubuntu-ThinkPad-Yoga-14:~/OpenCV/'. The user has entered 'contours' and the output is a list of 12 points in a 2xN array format, where each row represents a point with its x and y coordinates. The points are: (378, 949), (375, 940), (368, 934), (359, 932), (350, 937), (345, 955), (351, 962), (359, 966), (368, 964), (376, 958), (378, 949), and (375, 940).

```
ubuntu@ubuntu-ThinkPad-Yoga-14: ~/OpenCV/  
contours  
[[ 378.  949.]  
 [ 375.  940.]  
 [ 368.  934.]  
 [ 359.  932.]  
 [ 350.  937.]  
 [ 345.  955.]  
 [ 351.  962.]  
 [ 359.  966.]  
 [ 368.  964.]  
 [ 376.  958.]]  
ubuntu@ubuntu-ThinkPad-Yoga-14:~/OpenCV/
```

# Compute Contours Features

[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)

## 1. Moments

```
1 import cv2
2 import numpy as np
3
4 img = cv2.imread('star.jpg',0)
5 ret,thresh = cv2.threshold(img,127,255,0)
6 contours,hierarchy = cv2.findContours(thresh, 1, 2)
7
8 cnt = contours[0]
9 M = cv2.moments(cnt)
10 print M
```

## 2. Contour Area

```
area = cv2.contourArea(cnt)
```

## 3. Contour Perimeter

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

## 4. Contour Approximation

```
1 epsilon = 0.1*cv2.arcLength(cnt,True)
2 approx = cv2.approxPolyDP(cnt,epsilon,True)
```

## 5. Convex Hull

 Convexity defects

checks a curve for convexity defects and corrects it

```
hull = cv2.convexHull(cnt)
```

## 6. Checking Convexity

```
k = cv2.isContourConvex(cnt)
```

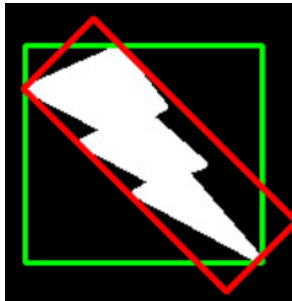


## 7.a. Straight Bounding Rectangle

```
1 x,y,w,h = cv2.boundingRect(cnt)
2 cv2.rectangle(img,(x,y),(x+w,y+h),(0,255,0),2)
```

## 7.b. Rotated Rectangle

```
1 rect = cv2.minAreaRect(cnt)
2 box = cv2.boxPoints(rect)
3 box = np.int0(box)
4 cv2.drawContours(img,[box],0,(0,0,255),2)
```

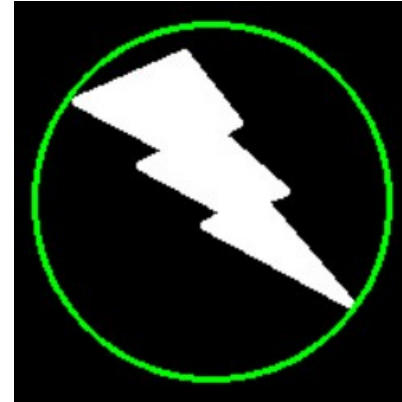


# Compute Contours Features

[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)

## 8. Minimum Enclosing Circle

```
1 (x,y),radius = cv2.minEnclosingCircle(cnt)
2 center = (int(x),int(y))
3 radius = int(radius)
4 cv2.circle(img,center,radius,(0,255,0),2)
```



## 9. Fitting an Ellipse

```
1 ellipse = cv2.fitEllipse(cnt)
2 cv2.ellipse(img,ellipse,(0,255,0),2)
```



<http://nicky.vanforeest.com/misc/fitEllipse/fitEllipse.html>

## 10. Fitting a Line

```
1 rows,cols = img.shape[:2]
2 [vx,vy,x,y] = cv2.fitLine(cnt, cv2.DIST_L2,0,0.01,0.01)
3 lefty = int((-x*vy/vx) + y)
4 righty = int(((cols-x)*vy/vx)+y)
5 cv2.line(img,(cols-1,righty),(0,lefty),(0,255,0),2)
```



# From Contour Find Shapes

[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)

## 4. Contour Approximation

```
1 epsilon = 0.1*cv2.arcLength(cnt,True)
2 approx = cv2.approxPolyDP(cnt,epsilon,True)
```



## 5. Convex Hull

Convexity defects  
checks a curve for convexity defects and corrects it

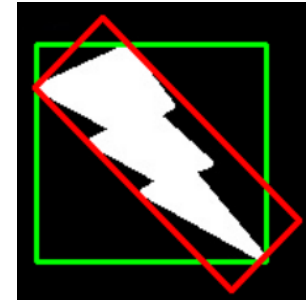
```
hull = cv2.convexHull(cnt)
```

## 7.a. Straight Bounding Rectangle

```
1 x,y,w,h = cv2.boundingRect(cnt)
2 cv2.rectangle(img,(x,y),(x+w,y+h),(0,255,0),2)
```

## 7.b. Rotated Rectangle

```
1 rect = cv2.minAreaRect(cnt)
2 box = cv2.boxPoints(rect)
3 box = np.int0(box)
4 cv2.drawContours(img,[box],0,(0,0,255),2)
```



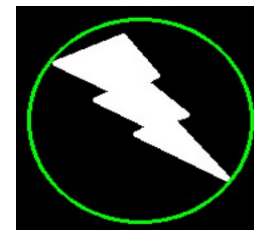
## 9. Fitting an Ellipse

```
1 ellipse = cv2.fitEllipse(cnt)
2 cv2.ellipse(img,ellipse,(0,255,0),2)
```



## 8. Minimum Enclosing Circle

```
1 (x,y),radius = cv2.minEnclosingCircle(cnt)
2 center = (int(x),int(y))
3 radius = int(radius)
4 cv2.circle(img,center,radius,(0,255,0),2)
```



# Contour-Shapes Properties

[http://opencv-python-tutroals.readthedocs.io/en/latest/py\\_tutorials/py\\_imgproc/py\\_contours/py\\_contour\\_properties/py\\_contour\\_properties.html](http://opencv-python-tutroals.readthedocs.io/en/latest/py_tutorials/py_imgproc/py_contours/py_contour_properties/py_contour_properties.html)

## 11. Aspect Ratio

$$\text{Aspect Ratio} = \frac{\text{Width}}{\text{Height}}$$

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

## 12. Extent

$$\text{Extent} = \frac{\text{Object Area}}{\text{Bounding Rectangle Area}}$$

```
area = cv2.contourArea(cnt)
x,y,w,h = cv2.boundingRect(cnt)
rect_area = w*h
extent = float(area)/rect_area
```

## 13. Solidity

$$\text{Solidity} = \frac{\text{Contour Area}}{\text{Convex Hull Area}}$$

```
area = cv2.contourArea(cnt)
hull = cv2.convexHull(cnt)
hull_area = cv2.contourArea(hull)
solidity = float(area)/hull_area
```

## 14. Equivalent Diameter

$$\text{Equivalent Diameter} = \sqrt{\frac{4 \times \text{Contour Area}}{\pi}}$$

```
area = cv2.contourArea(cnt)
equi_diameter = np.sqrt(4*area/np.pi)
```

## 15. Orientation

Following method also gives the Major Axis and Minor Axis lengths.

```
(x,y),(MA,ma),angle = cv2.fitEllipse(cnt)
```

# Contour Mask And Pixel Points

[http://opencv-python-tutroals.readthedocs.io/en/latest/py\\_tutorials/py\\_imgproc/py\\_contours/py\\_contour\\_properties/py\\_contour\\_properties.html](http://opencv-python-tutroals.readthedocs.io/en/latest/py_tutorials/py_imgproc/py_contours/py_contour_properties/py_contour_properties.html)

```
min_val, max_val, min_loc, max_loc = cv2.minMaxLoc(imgray, mask = mask)
```

## 16. Mask and Pixel Points

All the points comprises that object (contour)

```
mask = np.zeros(imgray.shape, np.uint8)
cv2.drawContours(mask, [cnt], 0, 255, -1)
pixelpoints = np.transpose(np.nonzero(mask))
#pixelpoints = cv2.findNonZero(mask)
```

Above, “two methods, one using Numpy functions, next one using OpenCV function (last commented line) are given to do the same. Results are also same, but with a slight difference. Numpy gives coordinates in (row, column) format, while OpenCV gives coordinates in (x,y) format. So basically the answers will be interchanged. Note that, row = x and column = y.”

## 17. Maximum Value, Minimum Value and their locations

## 18. Mean Color or Mean Intensity

```
mean_val = cv2.mean(im, mask = mask)
```

## 19. Extreme Points

```
leftmost = tuple(cnt[cnt[:, :, 0].argmin()][0])
rightmost = tuple(cnt[cnt[:, :, 0].argmax()][0])
topmost = tuple(cnt[cnt[:, :, 1].argmin()][0])
bottommost = tuple(cnt[cnt[:, :, 1].argmax()][0])
```



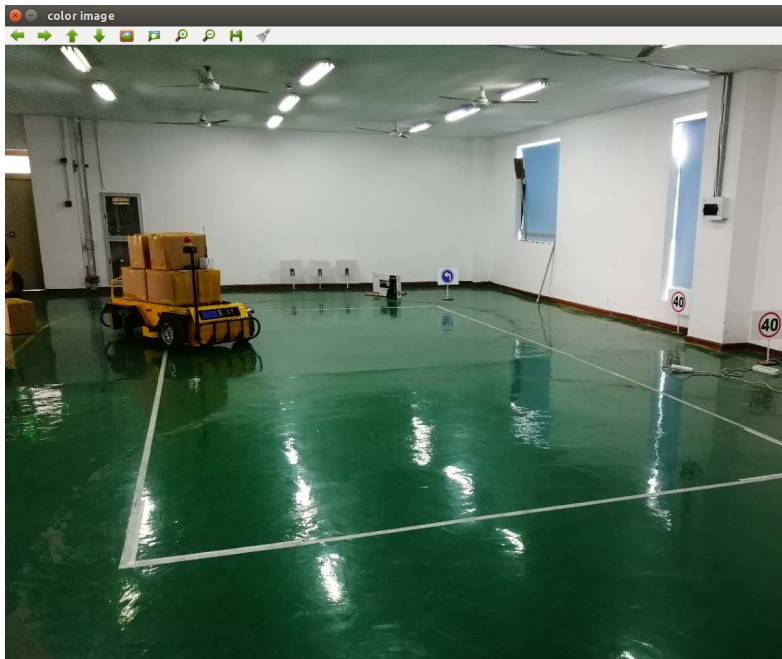
# Example Separation of Floor Track



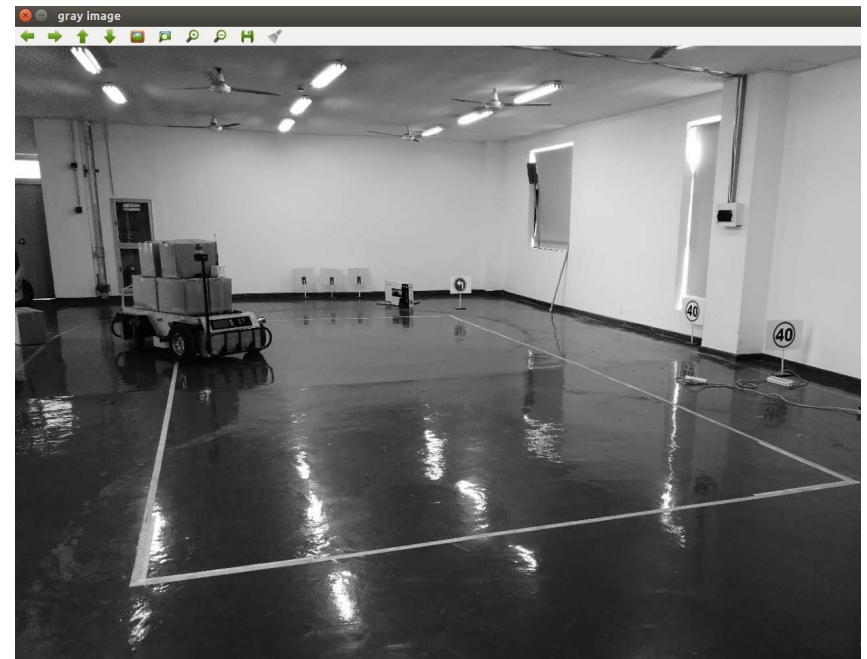
# From Shapes & Colors Find ROI And Remove Reflections

	Ceiling Lights (class w1)	Window Lights (class w2)
Shape	Rectangles x1	Rectangles x1
	Ellipses x2	Ellipses x2
	Circles x3	Circles x3
Location	Anywhere x4 smaller part image x5	Anywhere x4 smaller part image x5
Color	white x6	white x6
Repeated Pattern	maybe x7	maybe x7

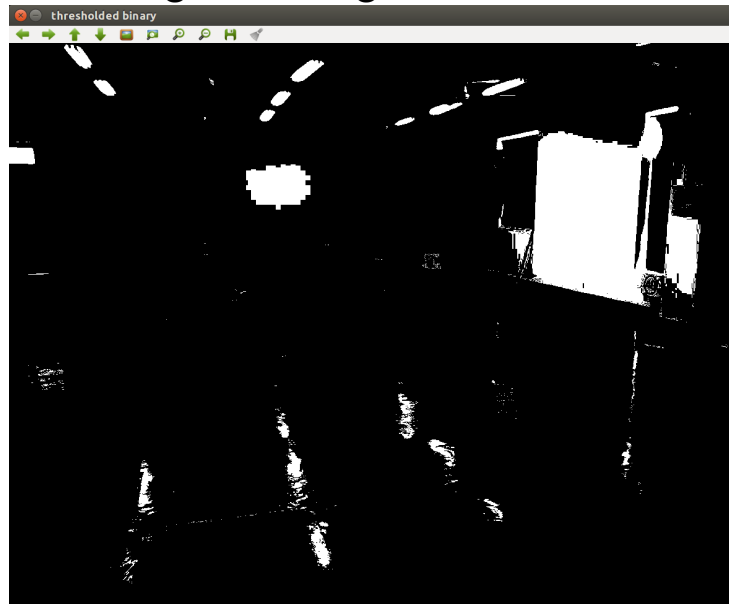
# Team Homework Separation of Floor Track



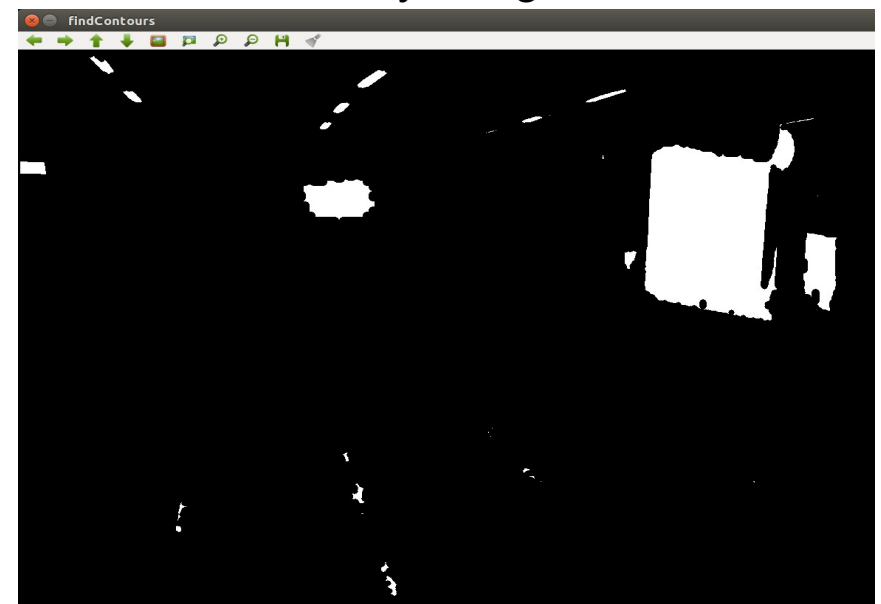
Original image



Gray-image



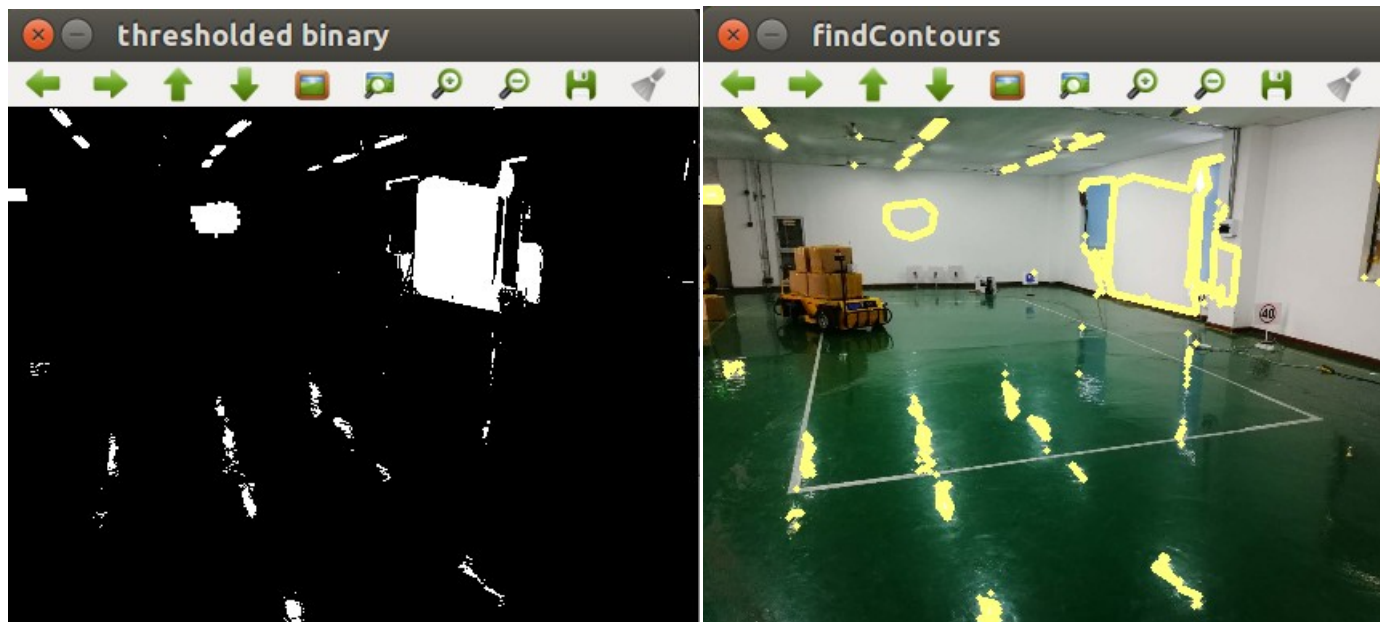
thresholdbinary



findcontour

Difference =  
Original – high  
intensity  
thresholded  
image  
>> image with  
removal of high  
intensity region

# Reflection Removal Based On Threshold



# Reflection Removal Based On Adaptive Threshold

```
img_blr4 = cv2.GaussianBlur(img, (21,21), 36, 47)
```

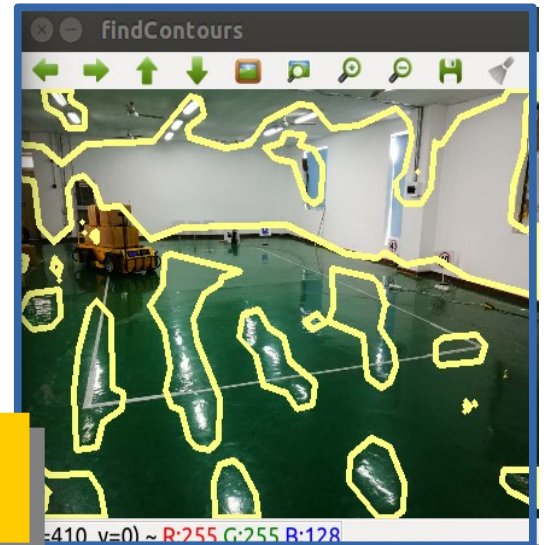
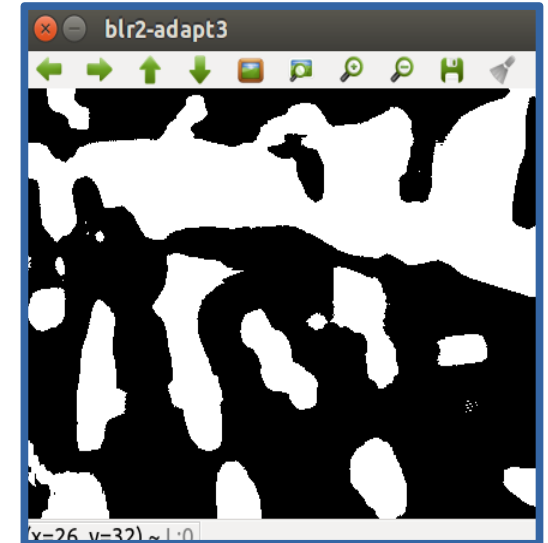
```
img_blr4_gray = cv2.cvtColor(img_blr4, cv2.COLOR_BGR2GRAY)
```

```
thresh3 = cv2.adaptiveThreshold(img_blr4_gray,255,\  
                                cv2.ADAPTIVE_THRESH_GAUSSIAN_C,\  
                                cv2.THRESH_BINARY,233,0)
```

```
cv2.imshow('blr4-adapt3',thresh3)
```

```
_,contours,hierarchy = cv2.findContours(thresh3, \  
                                         cv2.RETR_TREE, cv2.CHAIN_APPROX_SIMPLE)  
contours = [cv2.approxPolyDP(cnt, 3, True) for cnt in contours]
```

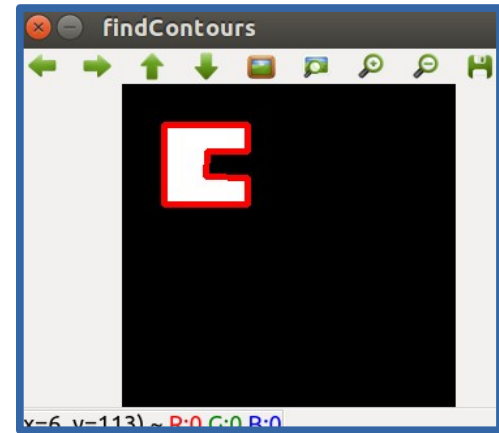
```
cv2.drawContours(img, contours, -1, (128,255,255),3)
```



Is the track being removed as well?

# Contour Attributes

```
_, contours, hierarchy = cv2.findContours(thresh, /  
                                         cv2.RETR_TREE, cv2.CHAIN_APPROX_SIMPLE)  
contours = [cv2.approxPolyDP(cnt, 3, True) for cnt in contours]  
cv2.drawContours(img, contours, -1, (0,0,255),3)
```





# Inference Engine

1

Table 1. Attribute Table

	Ceiling Lights (class w1)	Window Lights (class w2)
Shape	Rectangles x1	Rectangles x1
	Ellipses x2	Ellipses x2
	Circles x3	Circles x3
Location	Anywhere x4 smaller part image x5	Anywhere x4 smaller part image x5
Color	white x6	white x6
Repeated Pattern	maybe x7	maybe x7

3

So decision function

$$f(X) = x1 \ x5 \ x6 + x2 \ x6 + x3 \ x6 + x5 \ x6 \ \dots (1)$$

C/c++ implementation of the inference engine (switching function)

4

Table 2. Identification Table

2

	x1 rect	x2 elli	x3 cir	x4 loc	x5 sml	x6 wht	x7 rep	f(X)
x1 x5 x6	1	D	D	D	1	1	D	1
x2 x6	D	1	D	D	D	1	D	1
x3 x6	D	D	1	D	D	1	D	1
x5 x6	D	D	D	D	1	1	D	1

Define primary implicant, removal of any of its column will result in the mis-identification of f(X)

# No: C/C++ Inference Engine

```
#include<stdio.h>
int And(int a, int b);
int Or(int a, int b);
int Not(int a);
void main()
{
    ///where main body of code will go
}
int And(int a, int b)
{
    int output;
    if(a==0 && b==0)
        output=0;
    if(a==1 && b==0)
        output=0;
    if(a==0 && b==1)
        output=0;
    if(a==1 && b==1)
        output=1;
    return (output);
}
```

Simplify it 1.  
as boolean;  
2. logically  
as &&

```
int Or(int a, int b)
{
    int output;
    if(a==0 && b==0)
        output=0;
    if(a==1 && b==0)
        output=1;
    if(a==0 && b==1)
        output=1;
    if(a==1 && b==1)
        output=1;
    return (output);
}
```

```
int Not(int a)
{
    int output;
    if(a==0 )
        output=1;
    if(a==1 )
        output=0;
    return (output);
}
```

In fact C/C++ support all the boolean logic operators, so build inference engine should be straight forward

Simplify it 1.  
as boolean;

```
int And(int a, int b)
{
    return a && b;
}
```

```
return Not(And(a, b));
```

Build NAND,  
NOR, XOR etc

# C/C++ Bitwise Operators

Operators	Meaning of operators
&	Bitwise AND
	Bitwise OR
^	Bitwise exclusive OR
~	Bitwise complement
<<	Shift left
>>	Shift right

```
// C Program to demonstrate the working of logical operators
#include <stdio.h>
int main()
{
    int a = 5, b = 5, c = 10, result;
    result = (a == b) && (c > b);
    printf("(a == b) && (c > b) equals to %d \n", result);

    result = (a == b) && (c < b);
    printf("(a == b) && (c < b) equals to %d \n", result);

    result = (a == b) || (c < b);
    printf("(a == b) || (c < b) equals to %d \n", result);

    result = (a != b) || (c < b);
    printf("(a != b) || (c < b) equals to %d \n", result);

    result = !(a != b);
    printf("!(a != b) equals to %d \n", result);

    result = !(a == b);
    printf("!(a == b) equals to %d \n", result);
    return 0;
}
```

# C/C++ Inference Engine

```
//-----Inference Engine to find reflection spots---//
//-----April 7, 2018, by HL, version 0x0.1; -----//
#include <stdio.h>
#include <stdbool.h>
#define dimension 100
bool  x[dimension], f_identification;
int  item;

int main()
{
    printf("Inference Engine to identify reflections \n");
    printf("x1 rectangle? 1 for Y or 0 for N \n");
    scanf("%i",&item);
    if (item == 1) x[1] = true;
    if (item == 0) x[1] = false;

    printf("x2 ellips? 1 for Y or 0 for N \n");
    scanf("%i",&item);
    if (item == 1) x[2] = true;
    if (item == 0) x[2] = false;

    printf("x3 circle? 1 for Y or 0 for N \n");
    scanf("%i",&item);
    if (item == 1) x[3] = true;
    if (item == 0) x[3] = false;

    printf("x4 location? 1 for Y or 0 for N \n");
    scanf("%i",&item);
    if (item == 1) x[4] = true;
    if (item == 0) x[4] = false;
```

```
    printf("x5 small size? 1 for Y or 0 for N \n");
    scanf("%i",&item);
    if (item == 1) x[5] = true;
    if (item == 0) x[5] = false;

    printf("x6 white color? 1 for Y or 0 for N \n");
    scanf("%i",&item);
    if (item == 1) x[6] = true;
    if (item == 0) x[6] = false;

    printf("x7 repetative? 1 for Y or 0 for N \n");
    scanf("%i",&item);
    if (item == 1) x[7] = true;
    if (item == 0) x[7] = false;

    f_identification = (x[1] && x[5] && x[6])
                        || (x[2] && x[6])
                        || (x[3] && x[6])
                        || (x[5] && x[6]);

    if (f_identification){
        printf("The object is reflection\n");}
    else {
        printf("The object is not reflection\n");}

    return 0;
}
```

# OpenCV Contours For Shapes

Table 3 (based on Table 2) openCV functions

	x1 <u>rect</u>	x2 <u>elli</u>	x3 <u>cir</u>	x4 <u>loc</u>	x5 <u>sml</u>	x6 <u>wht</u>	x7 rep
x1 x5 x6	1	D	D	D	1	1	D
x2 x6	D	1	D	D	D	1	D
x3 x6	D	D	1	D	D	1	D
x5 x6	D	D	D	D	1	1	D

Rectangle detection (size, location and color, as well as total number);

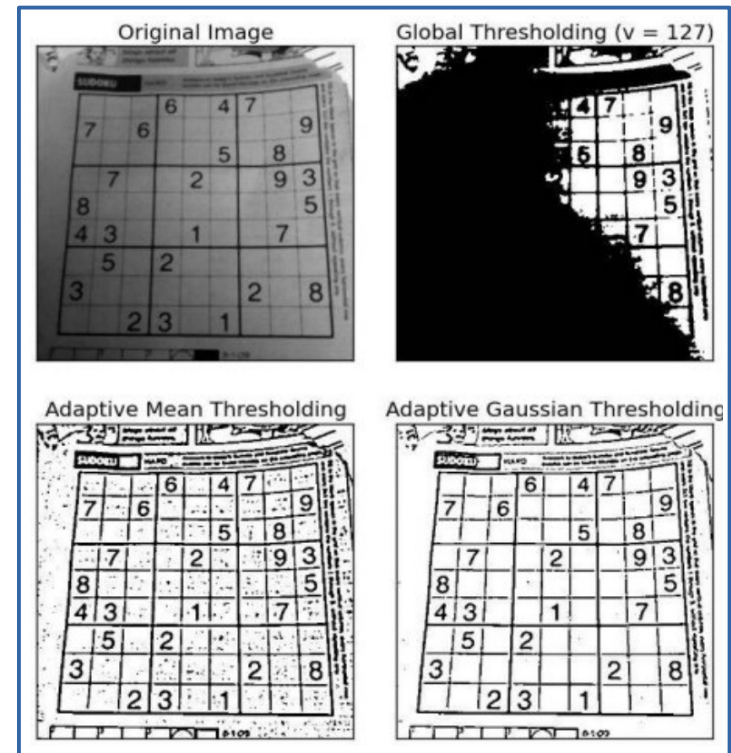
ellips detection (size, location and color, as well as total number);

Circle detection (size, location and color, as well as total number);

# Adaptive Threshold

[https://docs.opencv.org/3.3.0/d7/d4d/tutorial\\_py\\_thresholding.html](https://docs.opencv.org/3.3.0/d7/d4d/tutorial_py_thresholding.html)

```
thresh2 = cv2.adaptiveThreshold(img_gray,255,\  
    cv2.ADAPTIVE_THRESH_MEAN_C,\  
    cv2.THRESH_BINARY,33,0)  
thresh3 = cv2.adaptiveThreshold(img_gray,255,\  
    cv2.ADAPTIVE_THRESH_GAUSSIAN_C,\  
    cv2.THRESH_BINARY,33,0)
```



`cv2.adaptiveThreshold(src, maxValue, adaptiveMethod, thresholdType, blockSize, C[, dst]) → dst`

`src` – Source 8-bit single-channel image.

`dst` – Destination image of the same size and the same type as `src`.

`maxValue` – Non-zero value assigned to the pixels for which the condition is satisfied.

`adaptiveMethod` – `ADAPTIVE_THRESH_MEAN_C` or `ADAPTIVE_THRESH_GAUSSIAN_C`.

`thresholdType` – `THRESH_BINARY` or `THRESH_BINARY_INV`.

`blockSize` – Size of a pixel neighborhood 3, 5, 7, and so on.

`C` – Constant subtracted from the mean or weighted mean, positive may be zero or negative.



# cv:: void adaptiveThreshold( )

```
void adaptiveThreshold(  
    InputArray src,  
    OutputArray dst,  
    double maxValue,  
    int adaptiveMethod,  
    int thresholdType,  
    int blockSize, double  
    C)
```

## Parameters:

src – Source 8-bit single-channel image.

dst – Destination image of the same size and the same type as src .

maxValue – value assigned to pixels for condition satisfied.

adaptiveMethod –

ADAPTIVE\_THRESH\_MEAN\_C or

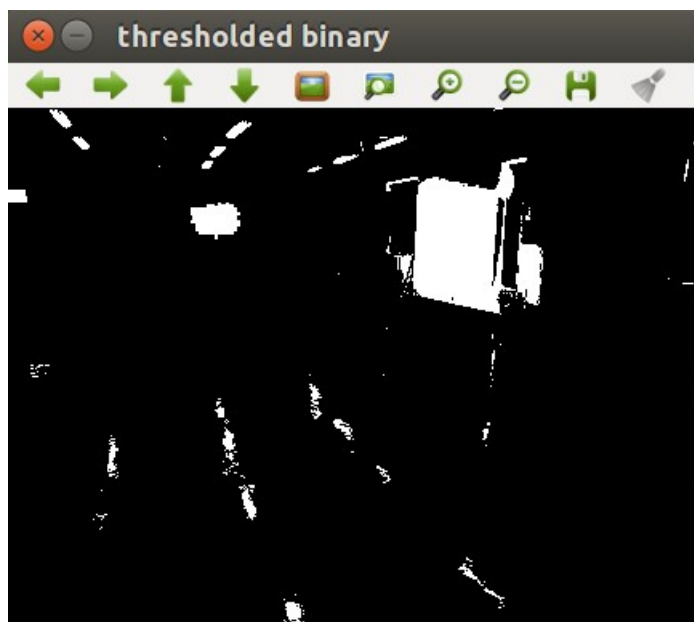
ADAPTIVE\_THRESH\_GAUSSIAN\_C .

thresholdType – THRESH\_BINARY or THRESH\_BINARY\_INV .

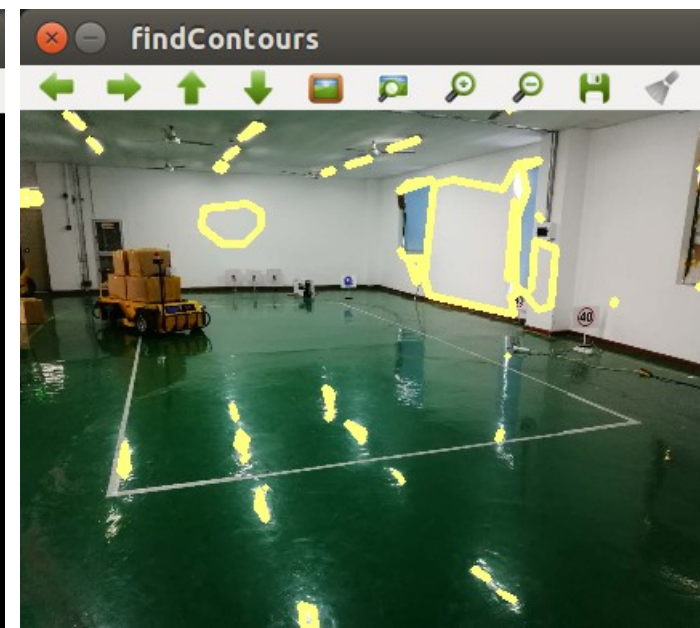
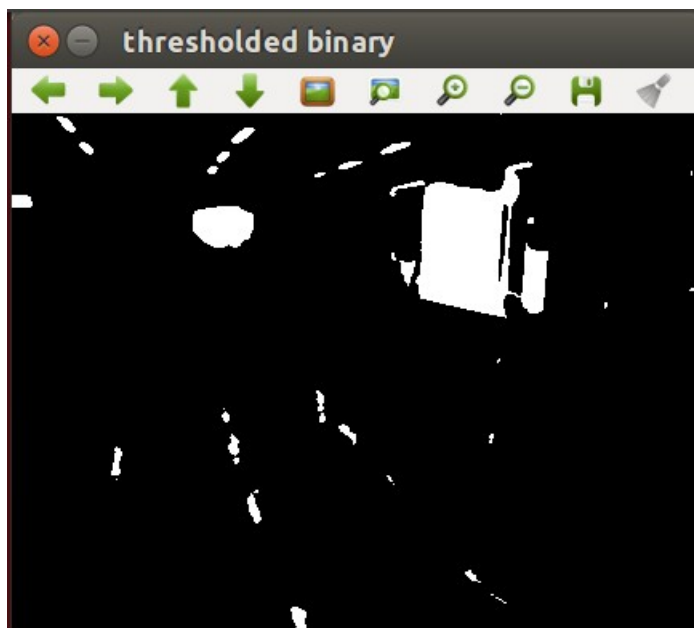
blockSize – kernel size, 3, 5, 7, and so on.

C – Constant subtracted from the mean or weighted mean. Normally, positive, can be zero or negative as well.

# With Or W/O Gaussian Blur Binrization+Contour



Binrization+Contour



GaussianBlur+  
Binrization+Con  
tour

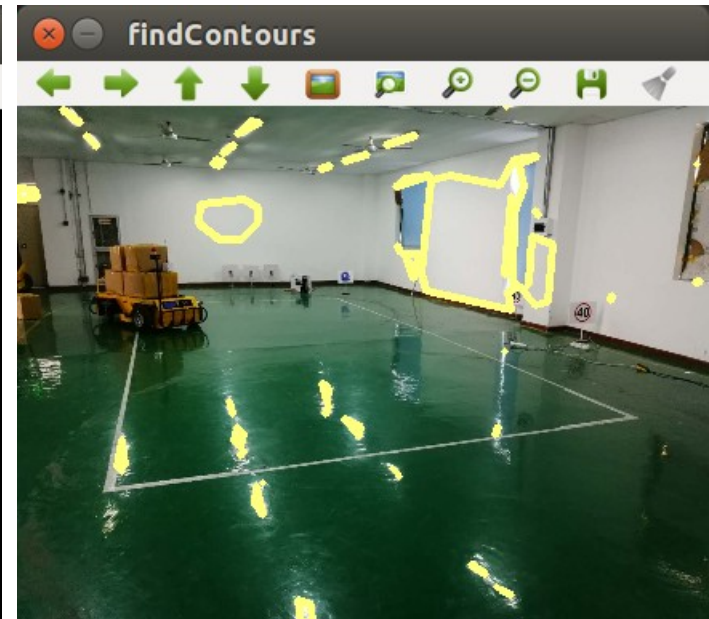
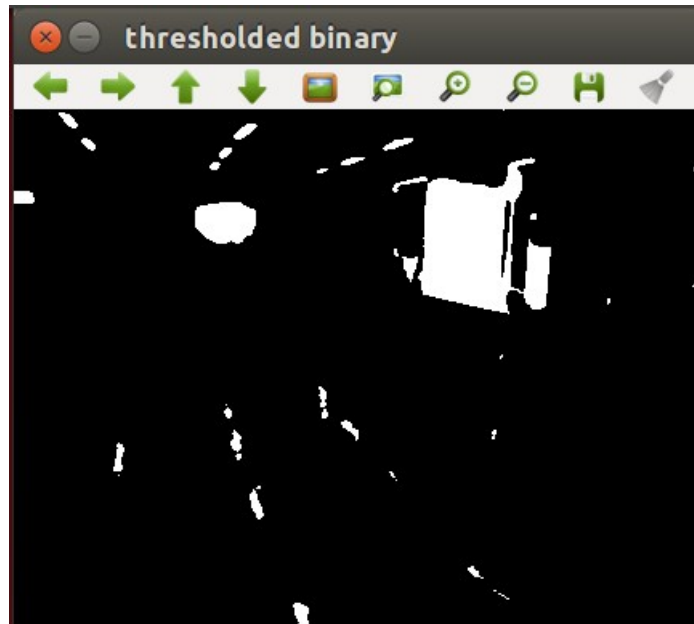
```
img_blr0 = cv2.GaussianBlur(img, (3, 3), 2, 3)
img_gray = cv2.cvtColor(img_blr0, cv2.COLOR_BGR2GRAY)
ret,thresh = cv2.threshold(img_gray,200,255,0)
```

# Surgical Removal

[http://opencv-python-tutroals.readthedocs.io/en/latest/py\\_tutorials/py\\_core/py\\_basic\\_ops/py\\_basic\\_ops.html](http://opencv-python-tutroals.readthedocs.io/en/latest/py_tutorials/py_core/py_basic_ops/py_basic_ops.html)

```
>>> px = img[100,100]
>>> print px
[157 166 200]

# accessing only blue pixel
>>> blue = img[100,100,0]
>>> print blue
157
```



# cv::inRange( ) Thresholding Colour Images

[https://docs.opencv.org/2.4/modules/core/doc/operations\\_on\\_arrays.html#inrange](https://docs.opencv.org/2.4/modules/core/doc/operations_on_arrays.html#inrange)

```
C++: void  
inRange(  
    InputArray src,  
    InputArray lowerb,  
    InputArray upperb,  
    OutputArray dst)
```

src – first input array.

lowerb – inclusive lower boundary array or scalar.

upperb – inclusive upper boundary array or scalar.

dst – output array, CV\_8U type.

$$\text{dst}(I) = \text{lowerb}(I)_0 \leq \text{src}(I)_0 \leq \text{upperb}(I)_0 \wedge \text{lowerb}(I)_1 \leq \text{src}(I)_1 \leq \text{upperb}(I)_1$$

## Finding Lane Lines with Colour Thresholds

<https://medium.com/@tjosh.owoyemi/finding-lane-lines-with-colour-thresholds-beb542e0d839>

Joshua Owoyemi Self-driving Car Engineer, PhD Candidate in Computer Vision and Robot Manipulation, Sharing technology insights.

# My ColorPicker.cpp

```
//-----*
// program: colorPicker.cpp; Coded by: HL on line *
// source. *
// purpose: hsv color picking *
// last update: April 28, 2018. *
//-----*
#include "opencv2/opencv.hpp"
#include <iostream>
using namespace cv;
using namespace std;

Mat im_hsv;
void pick_color(int e, int x, int y, int s, void *)
{
    if (e==1) // left mouse down
    {
        Vec3b p = im_hsv.at<Vec3b>(y, x); //pixel value
        cerr << int(p[0]) << " " << int(p[1]) << " " << int(p[2]) << endl;
    }
}
```

```
int main( int argc, char** argv )
{
    namedWindow("hsv");
    setMouseCallback("hsv", pick_color);

    if (argc<2) return -1;
    Mat im_bgr = imread(argv[1]);
    if (im_bgr.empty()) return -2;

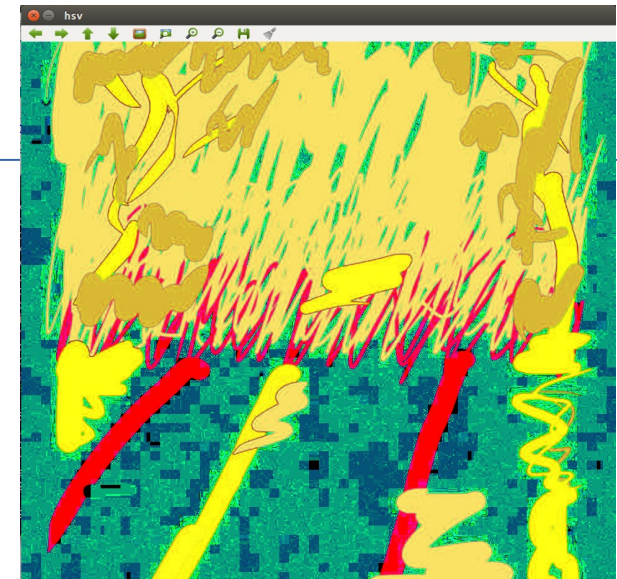
    cvtColor(im_bgr, im_hsv, COLOR_BGR2HSV);
    imshow("hsv", im_hsv);
    waitKey();

    return 0;
}
```

```
ubuntu@ubuntu-ThinkPad-Yoga-14: ~/Documents/SJSU/CMPE297/CMPE297Vi
ts/source/cpp$ ./main art-road1.jpg
init done
opengl support available
24 247 255
100 226 249
17 255 255
```

```
~/Documents/SJSU/CMPE297/CMPE297Vi
deoAnalytics/lec/lec5-binary-image/lec5-2-
Contours-Moments/source/cpp$ ./main art-
road1.jpg
```

Right click to pick pixel color



Display image in hsv space



# My ColorPicker.py

/Documents/SJSU/CMPE297/CMPE297VideoAnalytics/lec/lec5-binary-image/lec5-2-Contours-Moments/source/py\$

opencv hsv color picker

How to define the “lower” and “upper” range of a color?

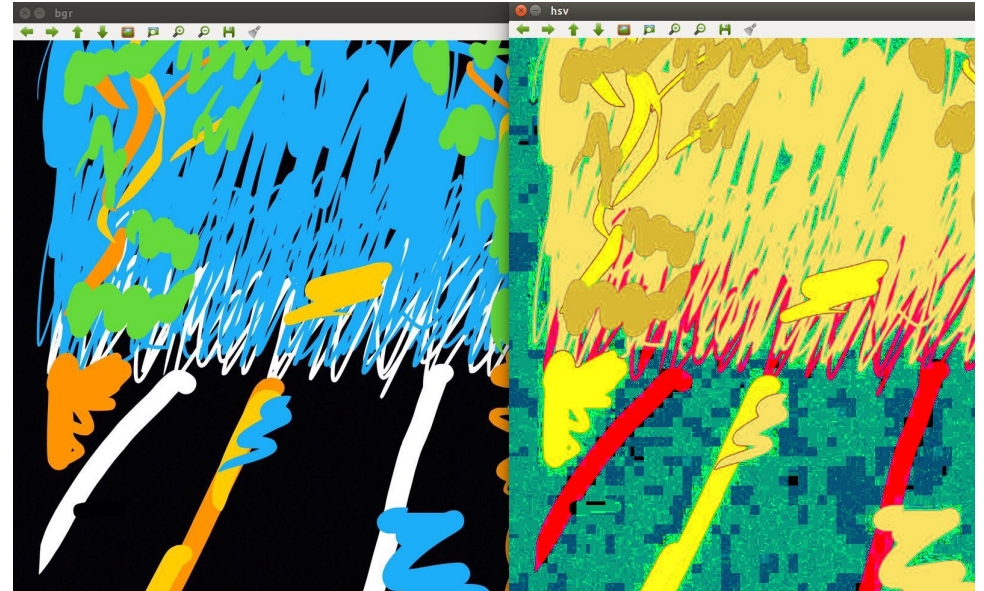
<http://answers.opencv.org/question/134248/how-to-define-the-lower-and-upper-range-of-a-color/>

```
#-----*
# program: colorPicker.py;                *
# reference code: see Harry Li's PPT for  *
#      original source;                   *
# date: April 28, 2018; status: tested;   *
#-----*

import cv2
import numpy as np

image_hsv = None # global
pixel = (20,60,80) # some default

# mouse callback function
def pick_color(event,x,y,flags,param):
    if event == cv2.EVENT_LBUTTONDOWN:
        pixel = image_hsv[y,x]
```



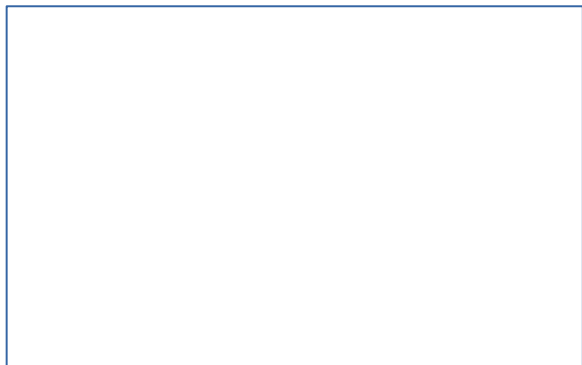


# C++: void log(InputArray src, OutputArray dst)

[https://docs.opencv.org/2.4/modules/core/doc/operations\\_on\\_arrays.html#inrange](https://docs.opencv.org/2.4/modules/core/doc/operations_on_arrays.html#inrange)

<https://www.learnopencv.com/high-dynamic-range-hdr-imaging-using-opencv-cpp-python/>

High dynamic imaging



The function log calculates the natural logarithm of the absolute value of every element of the input array

$$\text{dst}(I) = \begin{cases} \log|\text{src}(I)| & \text{if } \text{src}(I) \neq 0 \\ C & \text{otherwise} \end{cases}$$

where C is a large negative number (about -700 in the current implementation). The maximum relative error is about 7e-6 for single-precision input and less than 1e-10 for double-precision input. Special values (NaN, Inf) are not handled.

# OpenCV Version 2.1 C++ Overview

IP/10 Computer Vision & Machine Learning

April 28, 2018. HL.

1) OpenCV.2.1 C++.

### 1.1 Data Type Definition.

## 1.2 Matrix Ops

1.2 Matrix Ops

- (1) Add/sub/mult/division
- (2) Abs/Absdiff etc

## 2) Image Processing

## Filtering 2.1 filter2D();

$$\frac{\partial I(x,y)}{\partial x}, \frac{\partial I(x,y)}{\partial y}$$

boxToler); Lowercase  
C → C+

LowCase  
 $C \rightarrow C +$

$$f(x,y) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu_1)^2 + (y-\mu_2)^2}{2\sigma^2}}$$

GaussianBlur( );

$$\frac{\partial I}{\partial x} = \frac{1}{2}(F.D + B.D) \text{ along } x$$
$$\frac{1}{2} \begin{array}{|c|c|c|} \hline 1 & & 1 \\ \hline 1 & \times & 1 \\ \hline 1 & & 1 \\ \hline \end{array}$$
$$\rightarrow \frac{1}{2} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix}$$

Similarly,  $\frac{\partial I}{\partial y}$ .

K-by-k.

"me"

"mean" filter:

$$\frac{1}{K^2} \begin{vmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 1 & 3 & 6 \end{vmatrix}$$

2.

$$\nabla^2 f(x, y)$$

	1	2	1
--	---	---	---

$k^2 -$   
Input @

→ 1  
Content of  
the kernel

median  
 $\frac{2^2 + 2^2}{2}$

## < Ranking Center Point

SP-7

50-70 Questions  
Coding/math. For Page 1

# Contours Trees

Contours

- 1° Not always equal to Boundary
- 2° OpenCV Implementation  
List of Points.
- 3° Contours has to Be Computed on binary Image

Canny (Edge) Threshold Adaptive. → B(x,y) → Contours.

Topological relationship of each contour and mapping from a given contour image to tree structure

4° `findContours( )`; *memorize*

Version 1.0. for C.

`CvArr → CvMat`  
`IplImage`

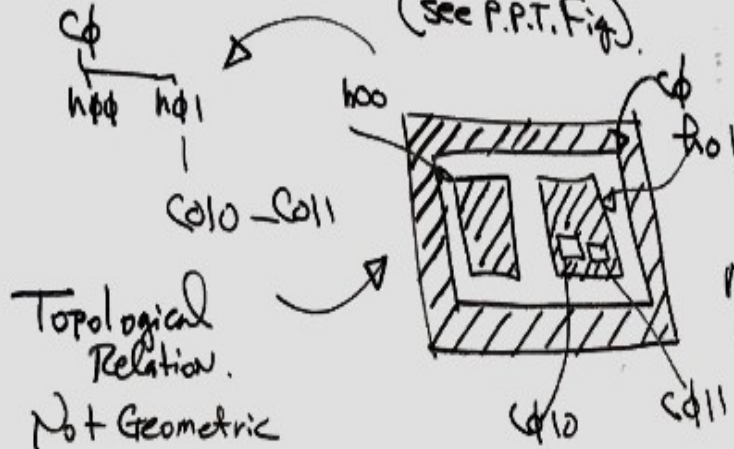
Version 2.0. for C++

`Mat → CvArr`  
`CvMat`  
`IplImage`

Example:

\* Requirement 1: CV-RETR-TREE

(see P.P.T. Fig.)

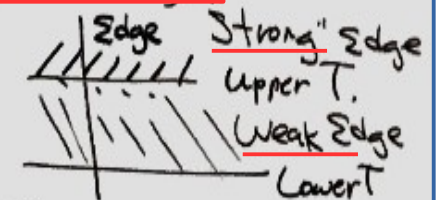
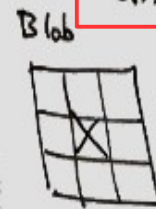


Requirement 2: C++ Implementation  
`CMakeLists.txt`  
`LabelCannyContours.CPP.`

`cvtColor( )`; HW

`GaussianBlur( )`;

`Canny( )`;



`drawContours( )`;

`findContour( )`; only works on binary image, one of the 3 images, Canny, Threshold and adaptiveThreshold

# Full Stack Embedded Software Developer

