

2021 Fall

Robotics HW3 – Robot Vision

Team Members and Division of Work

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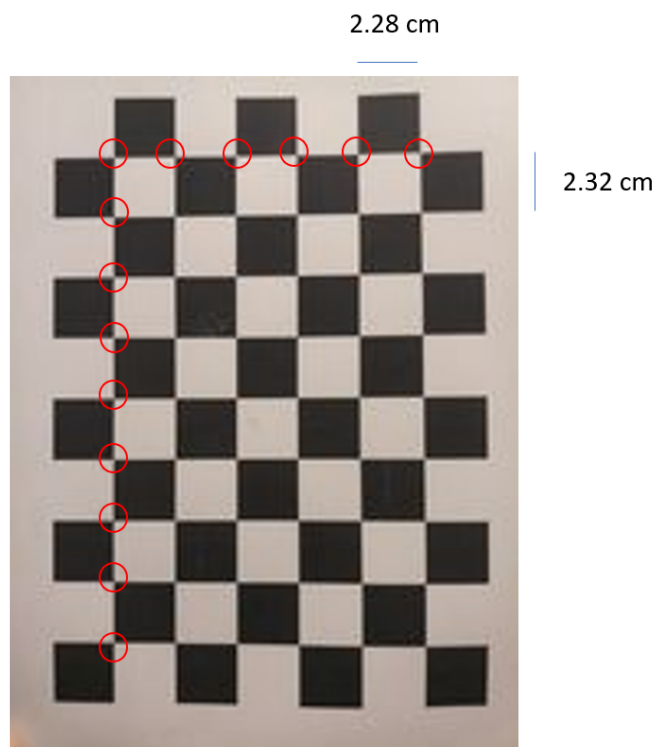
Part A: Camera Calibration

The camera we use:

The rear camera of Samsung Galaxy J7+ (SM-C710F/DS)

The checkerboard we use:

With 9*6 inner corners like following picture:



The calibration process:

Please refer to hw3_a.py for complete calibration code.

Brief explanation of the code is given below:

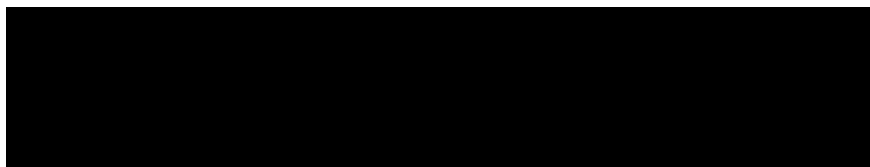
1. `cv.findChessboardCorners(image, patternSize[, corners[, flags]]) -> retval, corners`
is used to find the corner on chessboard in the image.
2. `cv.cornerSubPix(image, corners, winSize, zeroZone, criteria) -> corners`
is used to increase the accuracy of the above result.
3. `cv.calibrateCamera(objectPoints, imagePoints, imageSize, cameraMatrix, distCoeffs[, rvecs[, tvecs[, flags[, criteria]]]]`

is used to calculate intrinsic parameters, extrinsic parameters, distortion coefficients based on the above corner result

The Intrinsic Parameters:

The calibration result shows that the camera matrix of Samsung Galaxy J7+ is:

```
kuei@kuei-System-Product-Name:~/Documents/Course/Robotics/hw3$ python index.py
020.jpg
007.jpg
015.jpg
017.jpg
004.jpg
010.jpg
018.jpg
016.jpg
001.jpg
008.jpg
019.jpg
005.jpg
009.jpg
012.jpg
003.jpg
006.jpg
014.jpg
011.jpg
002.jpg
013.jpg
[[ 809.00256253    0.          222.04532571]
 [    0.          806.40691337  517.96448883]
 [    0.           0.           1.          ]]
kuei@kuei-System-Product-Name:~/Documents/Course/Robotics/hw3$
```



The Intrinsic Parameters are thus:

$$(f_x, f_y) = (809.00, 806.41)$$
$$(c_x, c_y) = (222.05, 517.97)$$

where each parameter is defined as following [1][2]:

Intrinsic parameters	Unit	Definition
f_x	Pixel	Focal length multiplied by conversion parameters between length and pixel on the x direction
f_y	Pixel	Focal length multiplied by conversion parameters between length and pixel on the y direction
c_x	Pixel	X Coordinate of the image center with respect to the bottom left of the image
c_y	pixel	Y Coordinate of the image center with respect to the bottom left of the image

Undistort the pictures:





The following command is used to undistort the original picture.







```
cv.undistort( src, cameraMatrix, distCoeffs[, dst[, newCameraMatrix]] ) -> dst
```







The effect is to mitigate the radial distortion and the tangent distortion of the original picture.







The 21 pictures:







The original 21 pictures and their undistorted version are list below.
Among them, the pictures No. 1~20 are used for calibration.







Picture No.	Original Picture	Undistorted Version
1		
2		

3		
4		
5		

6		
7		
8		

9		
10		
11		

12		
13		
14		

15		
16		
17		

18



19



20



21

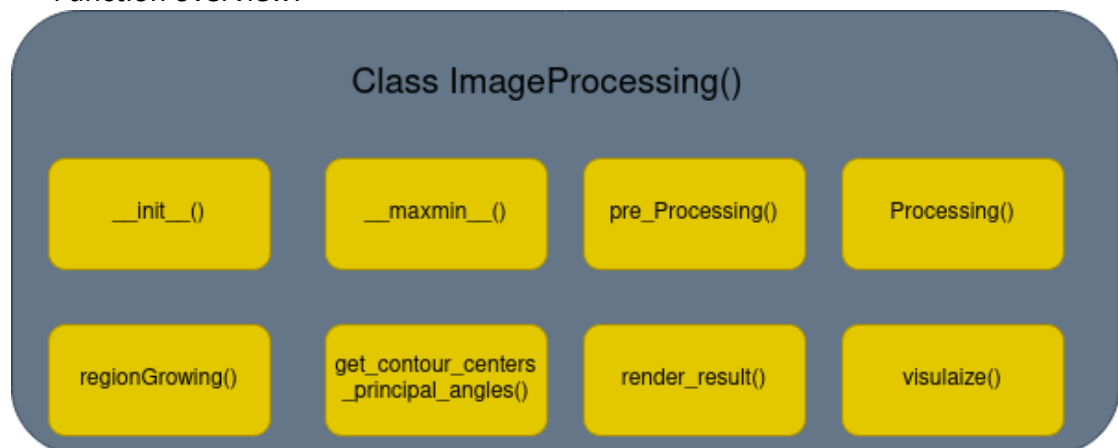


Part B: Object Detection

Process:

- Step 1: Image pre-processing
- Step 2: Binary Labeling using Region Growing Algorithm
- Step 3: Find contour
- Step 4: Get centroids and principal angles
- Step 5: render result to result image

Function overview:

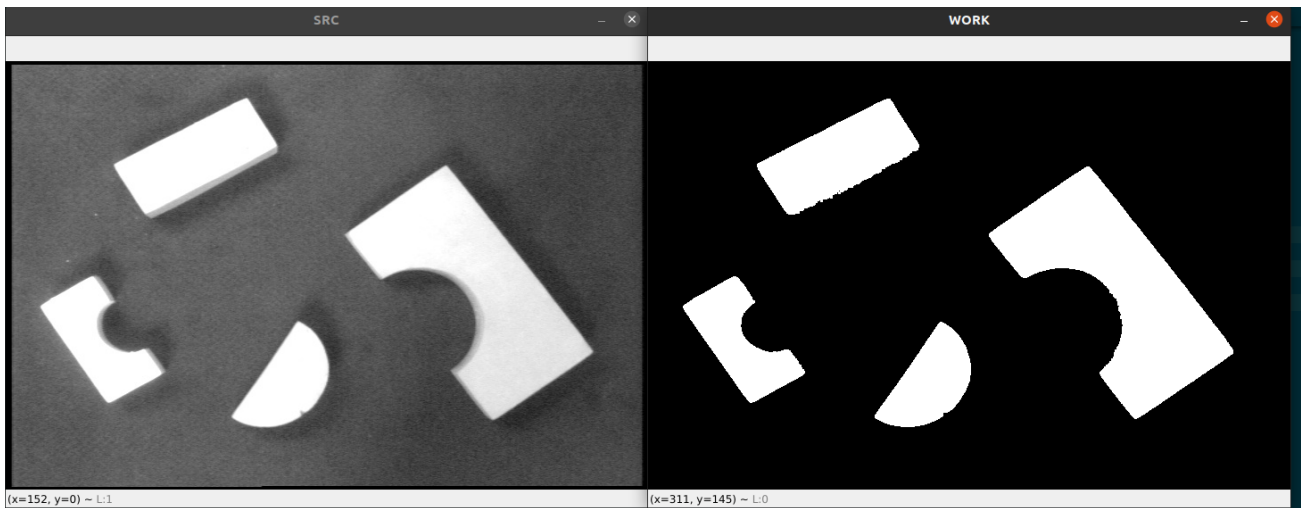


Implement:

Step 1: Image pre-processing

Key Function

```
ret, self.workIMG =  
cv.threshold(self.workIMG, 180, 255, cv.THRESH_BINARY)  
cv.erode(self.workIMG, self.kernel, self.workIMG)  
cv.dilate(self.workIMG, self.kernel, self.workIMG)
```



Step 2: Binary Labeling using Region Growing Algorithm

1. init row = 1, col = 1
2. access each element $I(\text{row}, \text{col})$, if $I(\text{row}, \text{col}) = 255$
 - a. $i = i + 1$
 - b. call region growing algorithm
 - c. if not all element has been travialied, goto step 2

Region growing algorithm:

1. set $I(k,j) = i$, push (k,j) push(0,0)
2. if $(j < n)$ and $I(k,j+1) = 255$
 - a. $I(k,j+1) = i$
 - b. push (K,j+1)
3. if $(k > 1)$ and $I(K-1,j) = 255$
 - a. $I(K-1,j) = i$
 - b. push(K-1,j)
4. if $(j > 1)$ and $I(k,j-1) = 255$
 - a. $I(k,j-1) = 1$
 - b. push (k,j-1)
5. if $(k < m)$ and $I(K+1,j) = 255$
 - a. $I(K+,j) = i$
 - b. push(K+1,j)
6. $k,j = \text{pop}()$, if $(k,j) \neq (0,0)$, goto step 2
7. pop()

```

pixel = 0
for row in range(self.workIMG.shape[0]):
    for col in range(self.workIMG.shape[1]):
        if(self.workIMG[row,col] == 255):
            # print(row,col)
            pixel = pixel + 1
            self.regionGrowing(row,col,pixel)

```

```

def regionGrowing(self,k,j,i):
    # k: ROW
    # j: COL
    # m: ROW number
    # n: COL number
    # i: pixel
    #init parameters
    points = []
    m = self.workIMG.shape[0]
    n = self.workIMG.shape[1]

    #step 1
    self.workIMG[k,j] = i
    points.append([k,j])
    points.append([0,0])

    #step 2
    #Kernel Right
    if (j<n) and self.workIMG[k,j+1] == 255:
        self.workIMG[k,j+1] = i
        points.append([k,j+1])

    #step 3
    #Kernel Up
    if (k>1) and self.workIMG[k-1,j] == 255:
        self.workIMG[k-1,j] = i
        points.append([k-1,j])

    #step 4
    #Kernel Left
    if (j>1) and self.workIMG[k,j-1] == 255:
        self.workIMG[k,j-1] = i
        points.append([k,j-1])

    #step 5
    #Kernel Down
    if (k<m) and self.workIMG[k+1,j] == 255:
        self.workIMG[k+1,j] = i
        points.append([k+1,j])

```

```

#Step 6
while(True):
    # print(len(points))
    k,j = points.pop()
    if(k == 0 and j == 0):
        points.pop()
        break

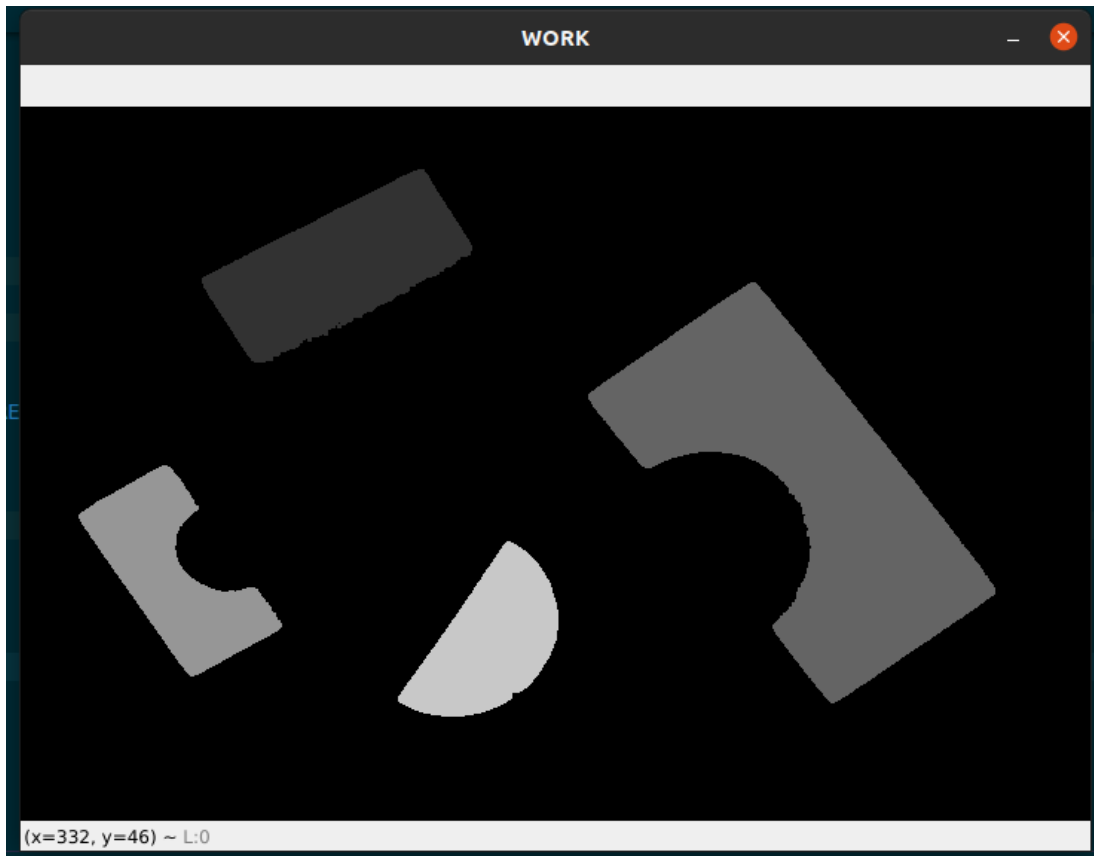
#step 2
#Kernel Right
if (j<n) and self.workIMG[k,j+1] == 255:
    self.workIMG[k,j+1] = i
    points.append([k,j+1])

#step 3
#Kernel Up
if (k>1) and self.workIMG[k-1,j] == 255:
    self.workIMG[k-1,j] = i
    points.append([k-1,j])

#step 4
#Kernel Left
if (j>1) and self.workIMG[k,j-1] == 255:
    self.workIMG[k,j-1] = i
    points.append([k,j-1])

#step 5
#Kernel Down
if (k<m) and self.workIMG[k+1,j] == 255:
    self.workIMG[k+1,j] = i
    points.append([k+1,j])

```



After segmentation, and rendered for visualization

Step 3: Find contour

key function:

```
self.contour, _ =
cv.findContours(self.workIMG, cv.RETR_EXTERNAL, cv.CHAIN_APPROX_SIMPLE)
```

Step 4: Get centroids and principal angles

1. Calculate image moments
2. Calculate centroid from image moments
3. Calculate principal angle

Code(part)

```
for i, c in enumerate(contours):
    self.M = cv.moments(c)
    # print(M)

    center = (int(self.M["m10"] / self.M["m00"]),
int(self.M["m01"] / self.M["m00"]))

    principal_angles[i] =
(math.atan2(2*self.M['mu11'],
self.M['mu02']-self.M['mu20'])/2)*180/math.pi
```

```
centers[i] = center
```

Step 5:render result to result image

In this step, we draw line which penetrate the principal center,

The default offset pixel is 100,

$p1 = (x,y) - (100,100 \cdot \tan(\text{atan2}(\mu_{11}, \mu_{20} - \mu_{02})/2))$

$p1 = (x,y) + (100,100 \cdot \tan(\text{atan2}(\mu_{11}, \mu_{20} - \mu_{02})/2))$

```
def render_result(self, line_length = 100):

    for i in range(len(self.contour)):

cv.drawContours(self.resultIMG, self.contour, i, (255, 0, 255), 5)

cv.circle(self.resultIMG, (self.centers[i][0], self.centers[i][1]),
2, (255, 0, 0), 8)

        # print(self.centers[i][0])
        cv.line(self.resultIMG,
                (self.centers[i][0] - line_length, self.centers[i][1]
- int(line_length * math.tan(math.atan2(2*self.M['mu11'],
self.M['mu20'] - self.M['mu02'])/2))),
                (self.centers[i][0] + line_length, self.centers[i][1]
+ int(line_length * math.tan(math.atan2(2*self.M['mu11'],
self.M['mu20'] - self.M['mu02'])/2))),
                (255, 125, 64), 2)

        #
cv.line(self.resultIMG, (100, 200), (200, 300), (255, 255, 0), 10)

        for row in range(self.workIMG.shape[0]):
            for col in range(self.workIMG.shape[1]):
                self.workIMG[row, col] = self.workIMG[row, col] * 50
```

```
Reading File: /home/robot/Desktop/HW/images/er7-4.jpg
robot@robot-GS75-Stealth-10SE:~/Desktop/HW$ /bin/python3 /home/robot/Desktop/HW/hm_3_b.py ./images/er7-4.jpg
Reading File: /home/robot/Desktop/HW/images/er7-4.jpg
Centers [[315 359]
 [102 313]
 [527 253]
 [213 107]]
Principal Angles [[-37.08449929]
 [ 36.56504275]
 [ 39.17740366]
 [-64.0709341 ]]
robot@robot-GS75-Stealth-10SE:~/Desktop/HW$
```


The result are:

Input cmd: `/bin/python3 /home/robot/Desktop/HW/hm_3_b.py ./images/er7-4.jpg`

output:

Reading File: `/home/robot/Desktop/HW/images/er7-4.jpg`

Centers `[[315 359]`

`[102 313]`

`[527 253]`

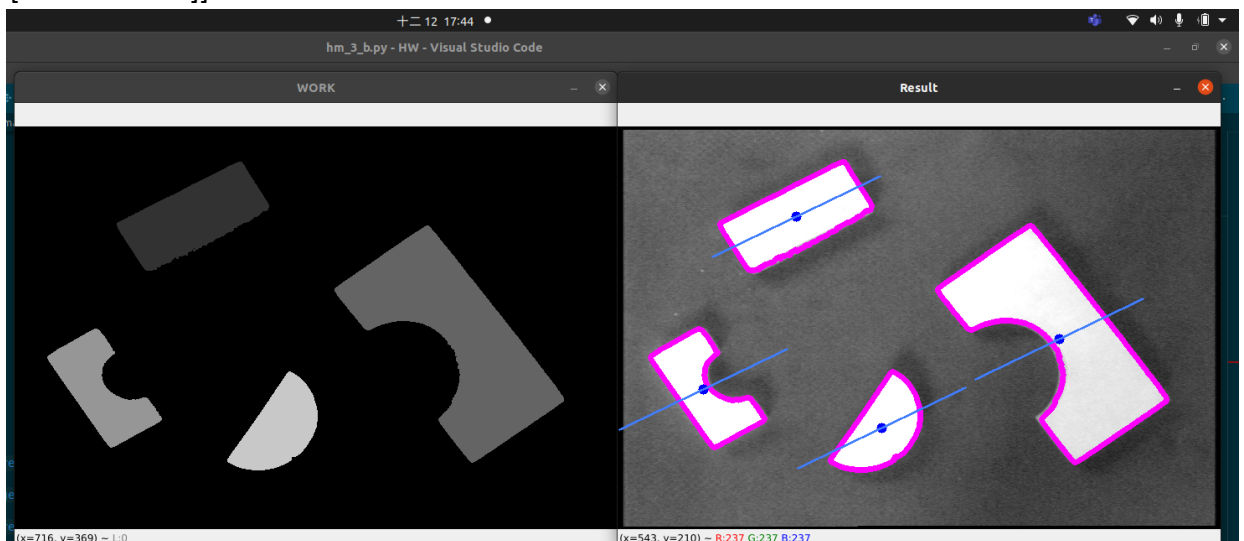
`[213 107]]`

principal Angles `[[-37.08449929]`

`[36.56504275]`

`[39.17740366]`

`[-64.0709341]]`



The code also can be accessed here:

<https://github.com/Runnlion/NTU-Robotic-2021-Fall/tree/main/Team2/HM3>

Reference:

[1] 一文帶你搞懂相機內參外參(Intrinsics & Extrinsics)

<https://zhuanlan.zhihu.com/p/389653208>

[2] Camera Calibration相機校正

<https://medium.com/image-processing-and-ml-note/camera-calibration%E7%9B%B8%E6%A9%9F%E6%A0%A1%E6%AD%A3-1d94ffa7cbb4>

[3] OpenCV

<https://opencv.org/>