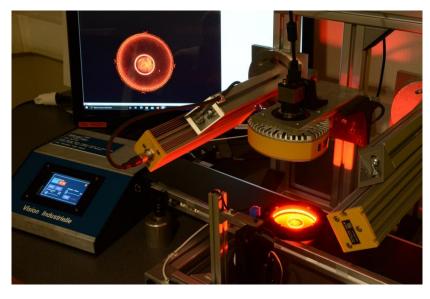


SC 19 - Machine Vision

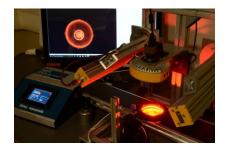
Image Processing

Julien VILLEMEJANE









At the end of this training, the learners will be able to:

Use basic building blocks of OpenCV

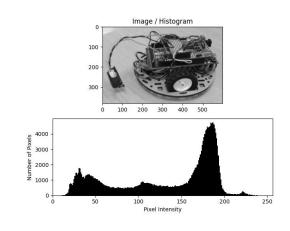
Open an image Create the histogram of an image

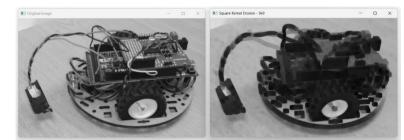
Apply basic transforms on images

- blur, filters
- erosion, dilation, opening, closing

Display the contour of objects

SC19 – Image Processing

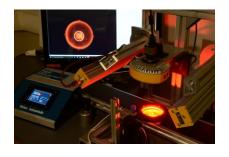












What is image processing for ?

Why is image processing required in a machine vision chain?

For industrial inspection applications?

What are the **main processes**?
What is the goal of each of them?
What is the ideal **workflow**?





https://www.youtube.com/@firstprinciplesofcomputerv3258



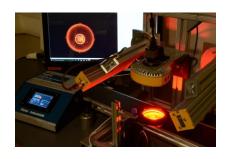










Image from the camera

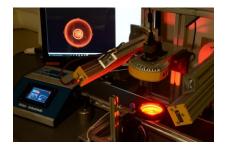
- Noise
- Bad contrast
- Inhomogeneous Lighting

- ...

Desired image with objects with well-defined contours

- Homogeneous zones
- Transition zones





SC19 – Image Processing Steps for processing an image

Acquisition

Pre Processing

Segmentation

Feature Extraction

Classification

Decision

Light, Camera...

Noise Reduction / Filtering
Contrast Enhancement
Normalization

Thresholding
Edge Detection
Region of Interest Selection

Geometric Features
Texture Analysis
Color Analysis

Object detection
Template Matching
Classification

Tolerances, pass/fail, real-time feedback...

Ensuring clarity and reducing unwanted information Making features of interest stand out Standardizing the image scale or intensity

Isolating objects of ROI / Separating product from background Identifying boundaries and contours
Focusing only on relevant portions of the image

Extracting measurements (size, shape, position...) Recognizing patterns, symbols, points of interest

Identifying and labelling objects
Checking correct shape, size or orientation
Categorizing objects to specific groups (defective or non-def.)

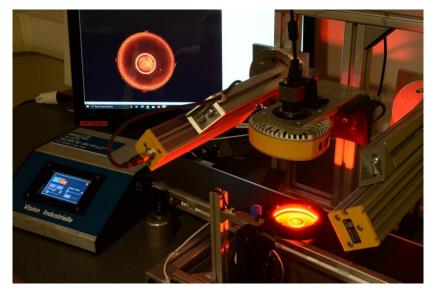




SC 19 – Machine Vision

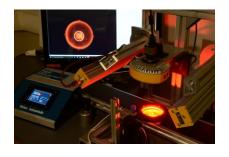
Image Processing with OpenCV

Julien VILLEMEJANE









SC19 – Image Processing What is OpenCV?

Open Source Computer Vision Library

An open source **computer vision** and machine learning software **library**

supporting multiple programming languages such as Python, C++, Java, and MATLAB

Image Processing

Filtering, Edge detection, Image transformations...

Object Recognition

Detecting objects in images and videos

CV Algorithms

Motion tracking, 3D reconstruction, Augmented reality

Machine Learning

Image classification, Pattern recognition, Scene Understanding



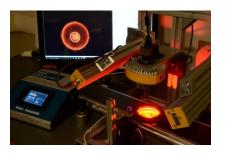
https://opencv.org











SC19 – Image Processing Installation of OpenCV

Installing OpenCV for Python 3

pip install opencv-python

Testing OpenCV importation in a script

import cv2
Cv2.__version__

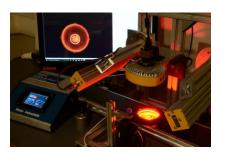
LEnsE. TECH

https://iogs-lense-training.github.io/image-processing/



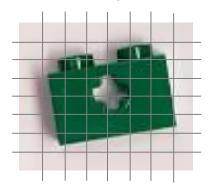
https://opencv.org



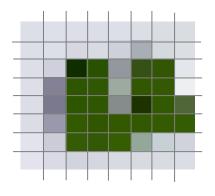


Digital Images

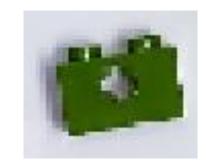
Continuous image



Digital image: projection of the continuous image







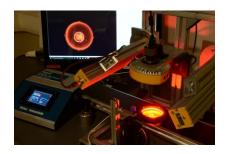
Digital Image

Picture represented in numerical form

so that it can be **stored**, **processed**, and **displayed** by a computer or digital device.

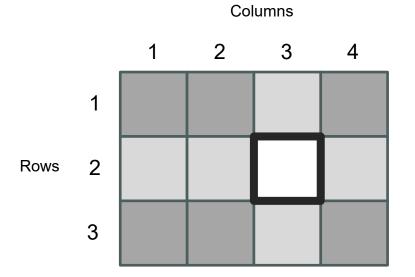
8 x 8 grid 16 x 16 grid 32 x 32 grid





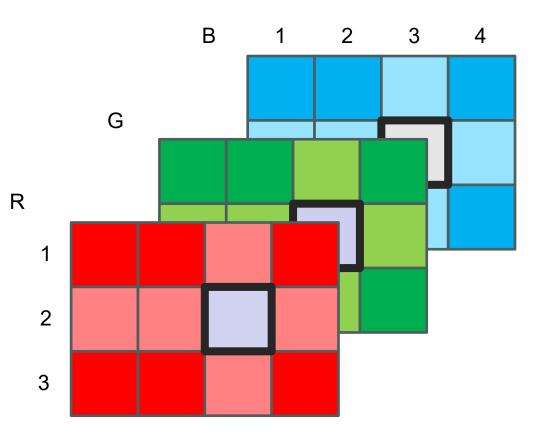
Nb of pixels = $h \times v$

GrayScale



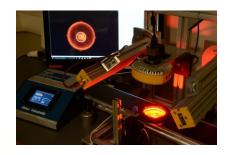
SC19 – Image Processing

Digital Images





GrayScale

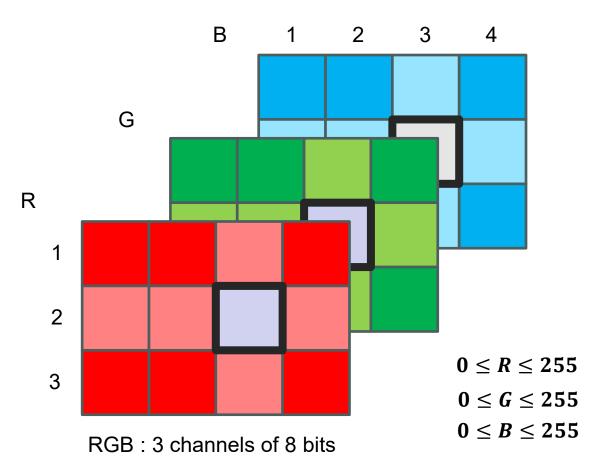


Nb of pixels = $h \times v$

Each pixel is converted into **n bits**.

SC19 – Image Processing

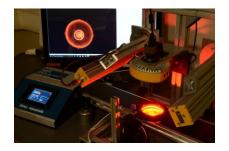
Digital Images / RGB





R=200, G=100, G=50

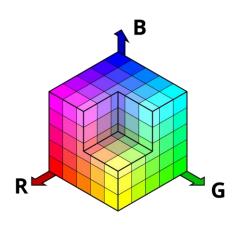




Digital Images / Color spaces

RGB

Used primarily in **electronic displays** like computer screens, cameras, and scanners. The combination of these three primary colors at various intensities can produce any color.

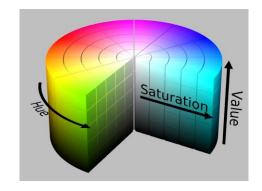


HSV

Used in **image editing**. It separates image's color from its brightness.

Hue: type of color

Saturation: intensity of the color **Value**: Brightness of the color



Color Space

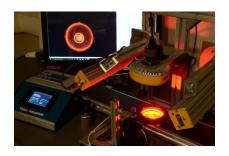
Model for **representing colors** in a consistent and reproducible way

Each color space uses a different method for organizing and describing color, depending on the purpose or application



Images Source : Wikipedia





Digital Images / Color spaces

Table 9 from

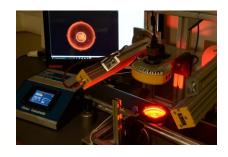
Segmentation of Images by Color Features: A Survey - Scientific Figure on ResearchGate. Available from: https://www.researchgate.net/figure/Advantages-and-disadvantages-of-color-spaces tbl7 323632019 [accessed 10 Oct 2024]

Color Space

Model for **representing colors** in a consistent and reproducible way

Advantages	Disadvantages
Convenient for image acquisition and	Non-uniform illumination sensitive;
displaying;	Differences between colors is not linear
Based on human color perception;	Non removable singularities
Robust before non-uniform illumination;	
The chromaticity is decoupled from	
the intensity	
Efficient in measuring small color	Singularity problem as other
difference;	
The chromaticity is decoupled from	nonlinear transformations
the intensity;	
Efficient coding color information for	Due to the linear transformation,
TV signal.	correlation between the component
	channels exists, although not as
	high as the RGB space
	Convenient for image acquisition and displaying; Based on human color perception; Robust before non-uniform illumination; The chromaticity is decoupled from the intensity Efficient in measuring small color difference; The chromaticity is decoupled from the intensity; Efficient coding color information for



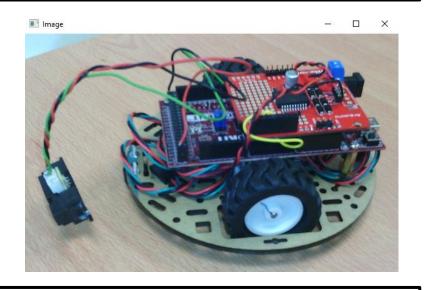


OpenCV / Open and display an image

Acquisition

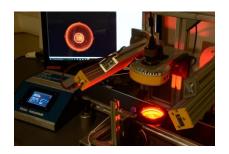
import cv2

image_rgb = cv2.imread('path/to/image.png')
image gray = cv2.imread('path/to/image.png', cv2.IMREAD_GRAYSCALE)



cv2.*imshow*('*Image* ', image_rgb) cv2.*waitKey*(0)





SC19 – Image Processing Pre-processing

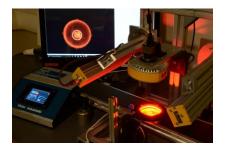
Acquisition

Pre Processing

Noise Reduction / Filtering Contrast Enhancement Normalization

Ensuring clarity and reducing unwanted information Making features of interest stand out Standardizing the image scale or intensity

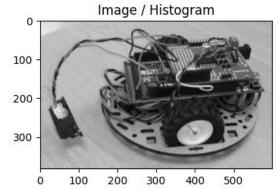


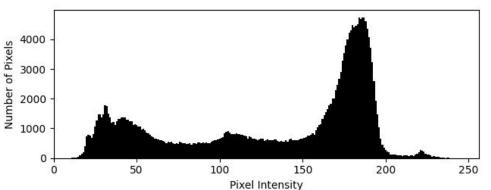


OpenCV / Histogram of an image

Acquisition

Pre Processing



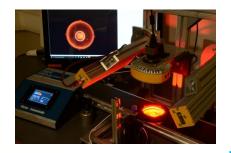


Histogram

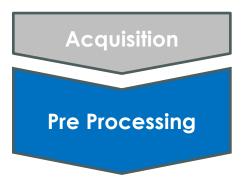
Graphical representation that shows the distribution of pixel intensity values in an image

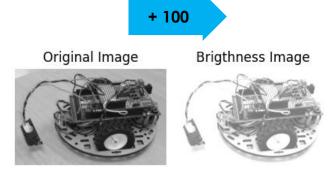
cv2.calcHist([image], [chan], Mask, [bins_nb], [min, max])

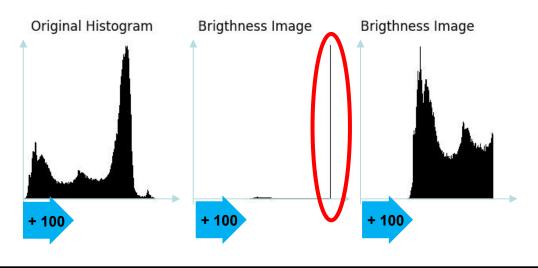




OpenCV / Contrast and Brightness

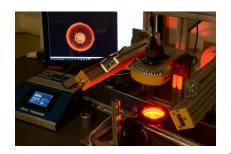




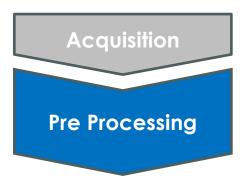


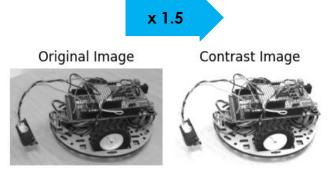
new_img = cv2.convertScaleAbs(image, beta=100)

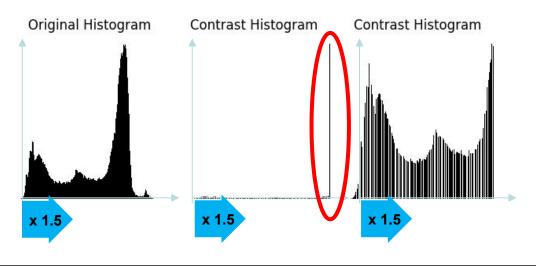




OpenCV / Contrast and Brightness





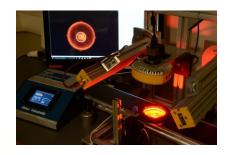


new_img = cv2.convertScaleAbs(image, alpha=1.5)



Continuing Education

Formation Continue





Convolution (filter)

kernel

-1	0	-2
1	5	1
-2	0	-1

original image

5	8	4	2	3	1	5
9	5	1	8	7	6	2
5	7	1	5	6	8	7
5	8	2	8	4	3	3
5	6	6	7	2	5	1

SC19 – Image Processing

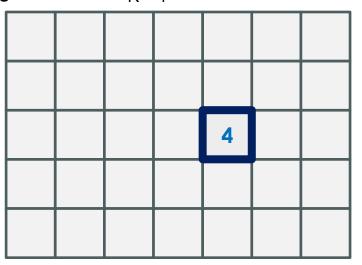
OpenCV / Convolution

5	8	4	2	3	1	5
9	5	1	8 x -1	7 × 0	6 x -2	2
5	7	1	5 × 1	6 x 5	8 x 1	7
5	8	2	8 × -2	4 x 0	3 x -1	3
5	6	6	7	2	5	1

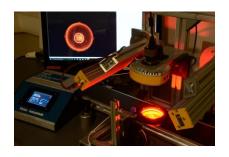
filtered image

$$R = -8 + 0 - 12 + 5 + 30 + 8 - 16 + 0 - 3$$

 $R = 4$







OpenCV / Convolution Kernel

Acquisition

Pre Processing

kernel = cv2.*getStructuringElement*(cv2.MORPH_xx, (M,N))

Cross Kernel

0	0	1	0	0
0	0	1	0	0
1	1	1	1	1
0	0	1	0	0
0	0	1	0	0

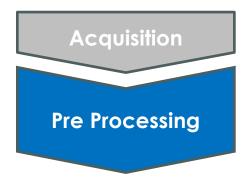
cv2.**MORPH_CROSS**

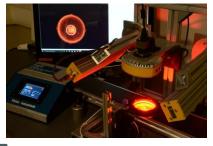
Rect Kernel

1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1

cv2.**MORPH_RECT**

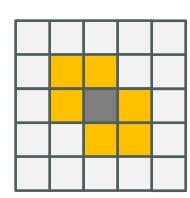






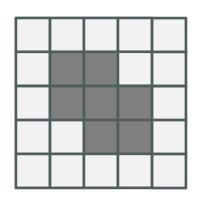
Original pixels

Removed pixels

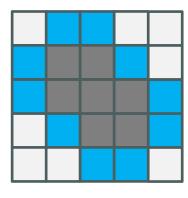


SC19 - Image Processing

OpenCV / Erosion and Dilation



Added pixels



kernel

0	1	0
1	1	1
0	1	0

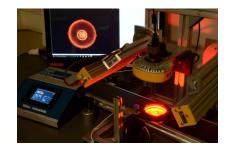
Erosion

Shrinking the foreground by removing pixels to the boundaries of objects

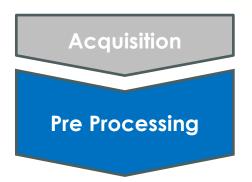
Dilation

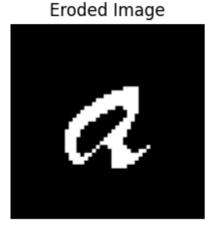
Enlarging the foreground by **adding pixels** to the boundaries of objects

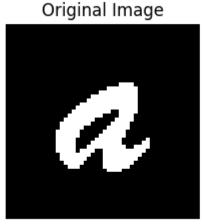


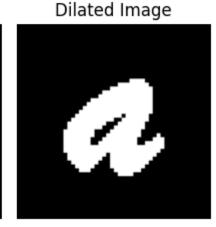


OpenCV / Erosion and Dilation









kernel

0	1	0
1	1	1
0	1	0

Erosion

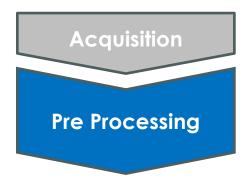
Shrinking the foreground by removing pixels to the boundaries of objects

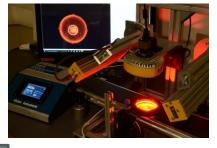
Dilation

Enlarging the foreground by **adding pixels** to the boundaries of objects

eroded_image = cv2.**erode**(image, kernel, iterations=1) dilated_image = cv2.**dilate**(image, kernel, iterations=1)







Original pixels

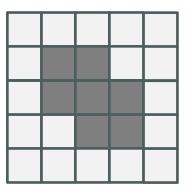
Removed pixels

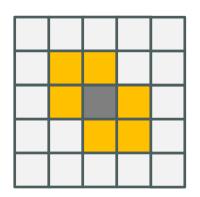
SC19 – Image Processing

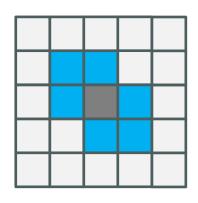
OpenCV / Opening and Closing morphological transforms



Added pixels







kernel

0	1	0
1	1	1
0	1	0

Opening

Erosion then **Dilation**

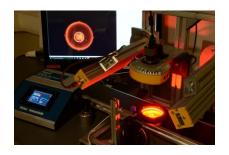
Removing small objects

Closing

Dilation then **Erosion**

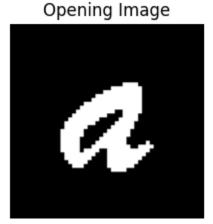
Filling in small holes

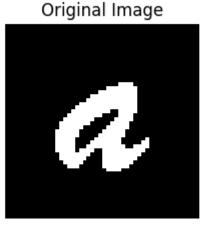


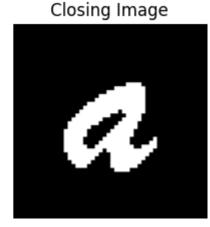


OpenCV / Opening and Closing morphological transforms









kernel

0	1	0
1	1	1
0	1	0

Opening

Erosion then **Dilation**

Removing small objects

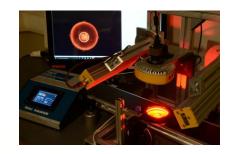
Closing

Dilation then **Erosion**

Filling in small holes

opening_image = cv2.morphologyEx(image, cv2.MORPH_OPEN, kernel) closing_image = cv2.morphologyEx(image, cv2.MORPH_CLOSE, kernel)

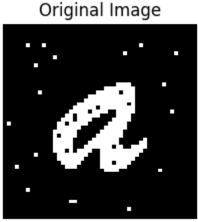


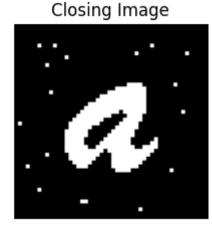


OpenCV / Opening and Closing morphological transforms









kernel

0	1	0
1	1	1
0	1	0

Opening

Erosion then **Dilation**

Removing small objects, in the background

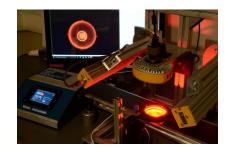
Closing

Dilation then **Erosion**

Filling in small holes in the foreground

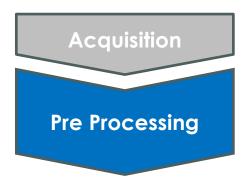
opening_image = cv2.morphologyEx(image, cv2.MORPH_OPEN, kernel) closing image = cv2.morphologyEx(image, cv2.MORPH CLOSE, kernel)

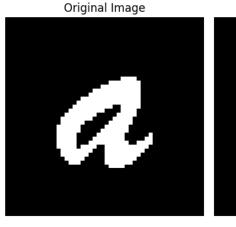


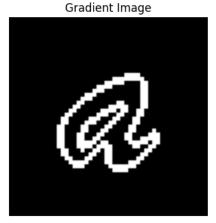


OpenCV / Gradient

morphological transforms







kernel

0	1	0
1	1	1
0	1	0

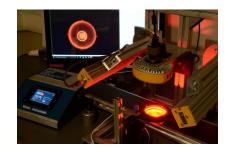
Gradient

Difference between a dilation and an erosion

Unknown pixels classification : background or foreground ?

gradient_image = cv2.morphologyEx(image, cv2.MORPH_GRADIENT, kernel)





OpenCV / Blur and Mean morphological transforms

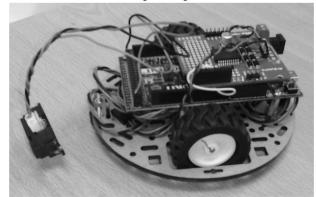
Acquisition

Pre Processing

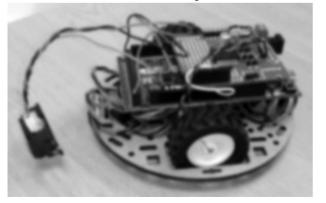
 $kernel_size = (N,M)$

blurred_image_gauss = cv2.GaussianBlur(image, kernel_size, 0) blurred_image_box = cv2.blur(image, kernel_size)

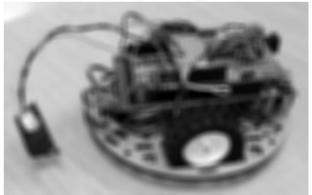
Original Image



Gaussian Blur Image



Median/Box Blur Image



Gaussian Kernel (x 1/273)

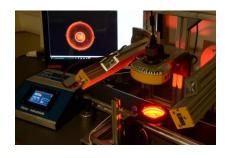
Mean Kernel (x 1/(N*M))

1/9	1/9	1/9
1/9	1/9	1/9
1/9	1/9	1/9

1	4	7	4	1	ı
4	16	26	16	4	
7	26	41	26	7	
4	16	26	16	4	
1	4	7	4	1	

Removing irrelevant details





OpenCV / Sobel

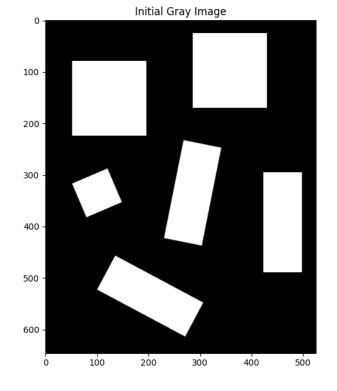


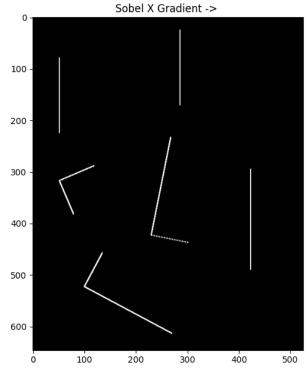
sobel_kernel_x1 = np.array([[-1, 0, 1], [-2, 0, 2], [-1, 0, 1]])

sobelx_1 = cv2.filter2D(image, -1, sobel_kernel_x1)

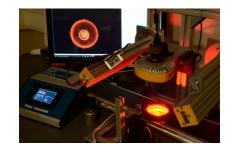
kernel

-1	0	1
-2	0	2
-1	0	1









OpenCV / Sobel

Acquisition

Pre Processing

sobelx = cv2.Sobel(image_gray, cv2.CV_64F, 1, 0, ksize=3)

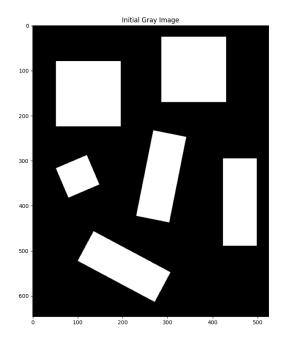
sobely = cv2.Sobel(image_gray, cv2.CV_64F, 0, 1, ksize=3)

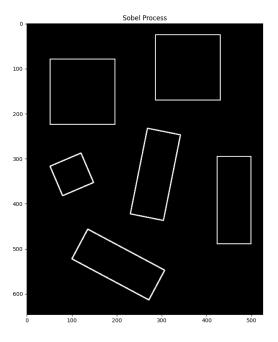
magnitude = cv2.magnitude(sobelx, sobely)

magnitude = cv2.convertScaleAbs(magnitude)

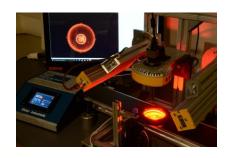
kernel

-1	0	1
-2	0	2
-1	0	1









Goal of processing an image

Acquisition

Pre Processing

Segmentation

Feature Extraction

https://docs.opencv.org/4.x/d3/db4/tutorial_py_watershed.html

Thresholding
Edge Detection
Region of Interest Selection

Geometric Features
Texture Analysis
Color Analysis

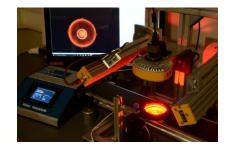
Isolating objects of ROI / Separating product from background Identifying boundaries and contours
Focusing only on relevant portions of the image

Extracting measurements (size, shape, position...)
Recognizing patterns, symbols, points of interest



Continuing Education

Formation Continue



SC19 – Image Processing

Segmentation and Feature extraction

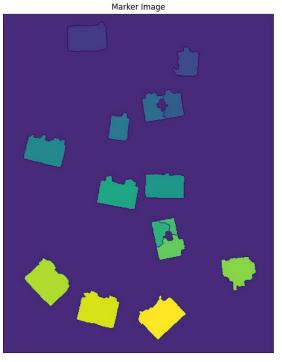
Acquisition

Pre Processing

Segmentation

Feature Extraction

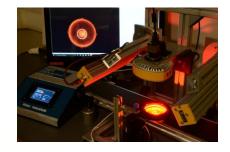






Watershed method





Segmentation and Feature extraction



Pre Processing

Segmentation

Feature Extraction

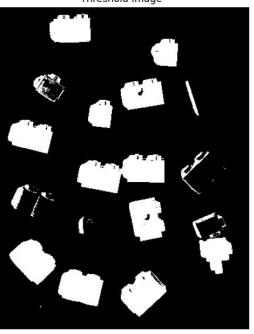
Original RGB Image



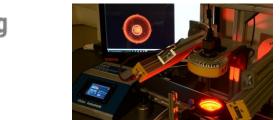
Gray Image



Threshold Image







Segmentation and Feature extraction

Acquisition

Pre Processing

Segmentation

Feature Extraction

Sure BG Image



Sure FG Image



Original Image



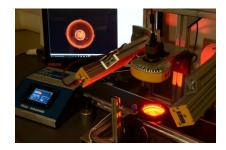
Labelling Image



Watershed method / Second step - Sure BG/FG image

Watershed method / Third step - Labelling





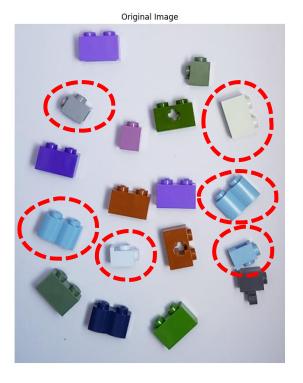
Segmentation and Feature extraction

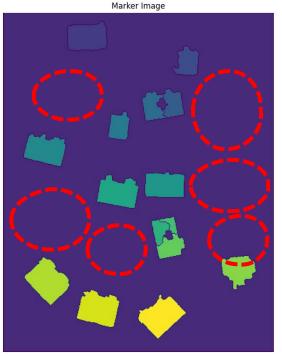
Acquisition

Pre Processing

Segmentation

Feature Extraction







Watershed method



Continuing Education

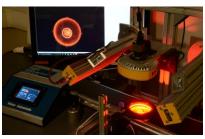
Formation Continue

Acquisition

Pre Processing

Segmentation

Feature Extraction



SC19 – Image Processing

Fourier Transform and Filtering

