

MAE 6291

Internet of Things for Engineers

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Week 8 [03/19/2025]

- Setting up the Edge Compute framework
- Practical Introduction to Image processing and Filtering for Edge Compute Applications
- Guest lecture: Designing Hardware to Improve Your Software - A Use Case in Sound by Jacob Whitton
- In-class Raspberry Pi Lab with PiNOIR camera
- Practical Introduction to OpenCV library in Python on the Raspberry Pi 4B
- Sobel, Laplacian, and Gaussian filtering on Raspberry Pi 4B
- Edge detection using thresholding and Otsu's method on Raspberry Pi 3B+

git clone https://github.com/gwu-mae6291-iot/spring2025_codes.git

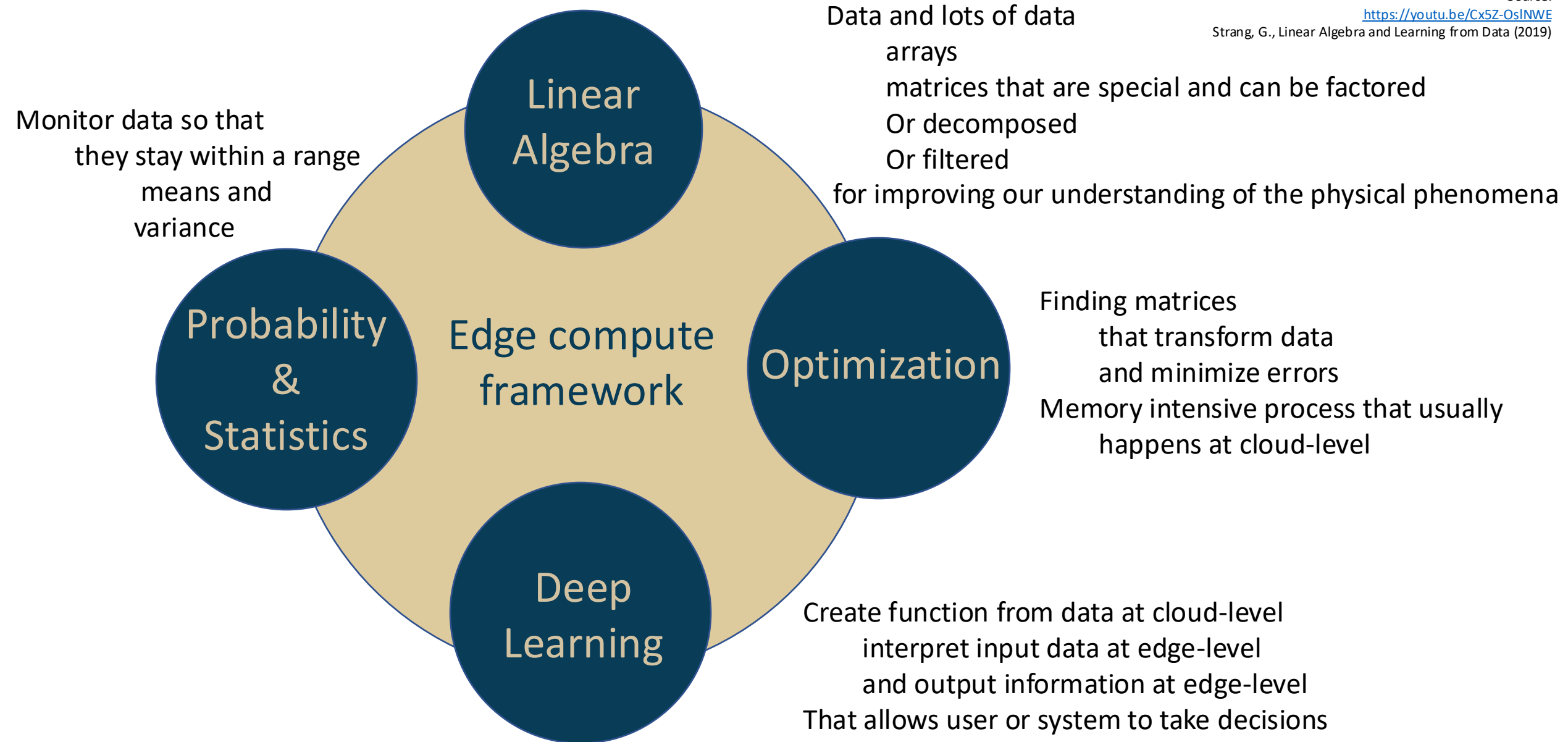


School of Engineering
& Applied Science

Spring 2025

THE GEORGE WASHINGTON UNIVERSITY

Photo: Kartik Bulusu



Review of the building blocks:

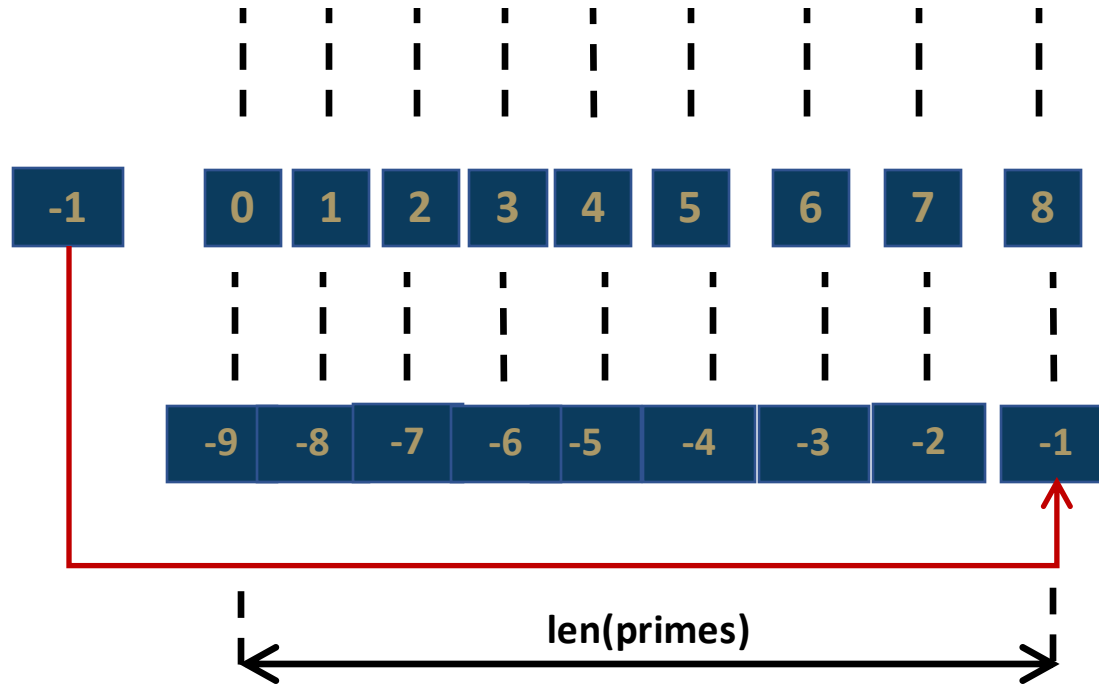
1. list – Python object-type
2. Matrix operations in Python



Indexing and Slicing Lists

Retrieve list-elements with a range of values

```
>>> primes = [2, 3, 5, 7, 9, 11, 13, 17, 19]
```



start *stop*

```
>>> primes[2:5]  
[5, 7, 9]
```

start *stop* *step*

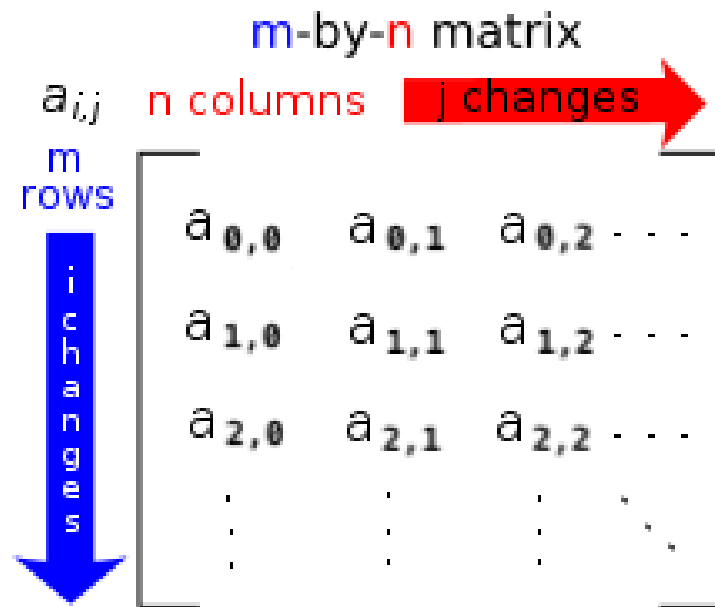
```
>>> primes[0:7:2]  
[2, 5, 9, 13]
```

start *stop* *step*

```
>>> primes[8:2:-2]  
[19, 13, 9]
```

start: at the index value
step: up or down at the increment value (default = 1)
stop: at the index value but not including it





Source: [http://en.wikipedia.org/wiki/Matrix_\(mathematics\)](http://en.wikipedia.org/wiki/Matrix_(mathematics))

The **ORDER** of a matrix

- $A_{m \times n}$ is $m \times n$
- Read as “m-by-n”

a_{ij} is called an **ELEMENT**

- at the i^{th} row and j^{th} column of A

Bookkeeping in a Matrix

Python:

```
>>> import numpy as np
>>> A = np.matrix([[ -1,  2], [ 3,  4]])
>>> A[0,0]
>>> A[0,:]
>>> A[:,0]
>>> A[1,0]
```



`A[row-0:row-M, column0:columnN]`



Matrix scalar operations

$$A = \begin{bmatrix} -1 & 2 \\ 3 & 4 \end{bmatrix} \text{ \& } s = 6$$

- Matrix, **A** has **m** rows and **m** columns
- The **ORDER** of matrix, **A** ??
- The **ORDER** of the scalar, **s** ??

Scalar Multiplication and Division

- Each element a_{ij}
- Is either **multiplied** with or **divided** by **s**

$$\begin{cases} A_{(m \times m)} * s_{(1 \times 1)} = D_{(m \times m)} \\ A_{(m \times m)} * s^{-1}_{(1 \times 1)} = F_{(m \times m)} \end{cases}$$

$$\begin{bmatrix} -1 & 2 \\ 3 & 4 \end{bmatrix} * 6 = \begin{bmatrix} -6 & 12 \\ 18 & 24 \end{bmatrix}$$

$$\begin{bmatrix} -1 & 2 \\ 3 & 4 \end{bmatrix} * \left(\frac{1}{6}\right) = \begin{bmatrix} -\frac{1}{6} & \frac{1}{3} \\ \frac{1}{2} & \frac{2}{3} \end{bmatrix}$$

Python:

```
>>> import numpy as np
>>> A = np.matrix([[ -1,  2],[3,  4]])
>>> B1 = A * 6
>>> B2 = A * (1/6)
>>> len(B1)
>>> np.shape(B2)
```



Think of

Array, A as an image
Scalar, s as brightness



Created by Guilherme Appolinário
from the Noun Project



Matrix-matrix operations

$$\begin{cases} A \times B = C \\ (m \times n) \quad (n \times p) \quad (m \times p) \end{cases}$$

- Matrix, **A** has **m** rows and **n** columns
- Matrix, **A** has **n** rows and **p** columns
- The **ORDER** of matrix, **A** ??
- The **ORDER** of matrix, **B** ??

Matrix Multiplication

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}_{(2 \times 2)} \begin{bmatrix} b_{11} \\ b_{21} \end{bmatrix}_{(2 \times 1)} = \begin{bmatrix} a_{11}b_{11} + a_{12}b_{21} \\ a_{21}b_{11} + a_{22}b_{21} \end{bmatrix}_{(2 \times 1)}$$

$$\begin{bmatrix} -1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 4 \\ -2 \end{bmatrix} = \begin{bmatrix} -4 - 4 \\ 12 - 8 \end{bmatrix} = \begin{bmatrix} -8 \\ 4 \end{bmatrix}$$

Python:

```
>>> import numpy as np
>>> A = np.matrix([[ -1, 2], [3, 4]])
>>> B = np.matrix([[4], [-2]])
>>> C = np.dot(A, B)
>>> len(C)
>>> np.shape(C)
```



Think of

Array, A as an image

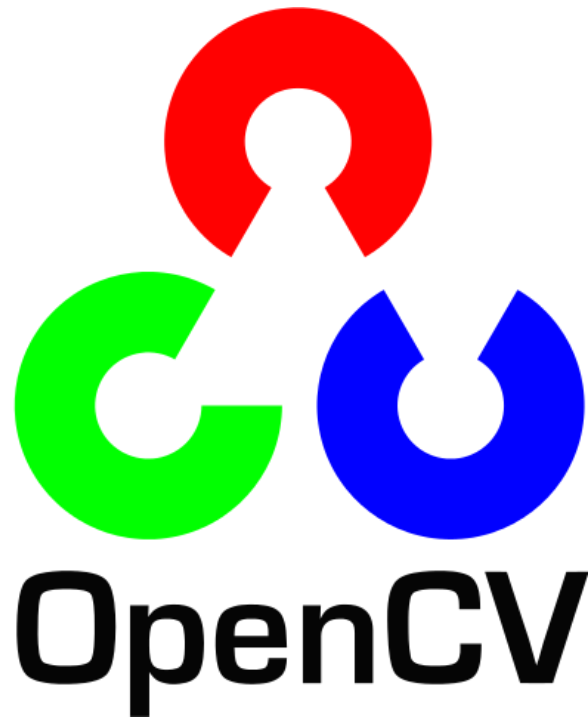
Array, B as a transformation



Created by Guilherme Appolinário
from the Noun Project

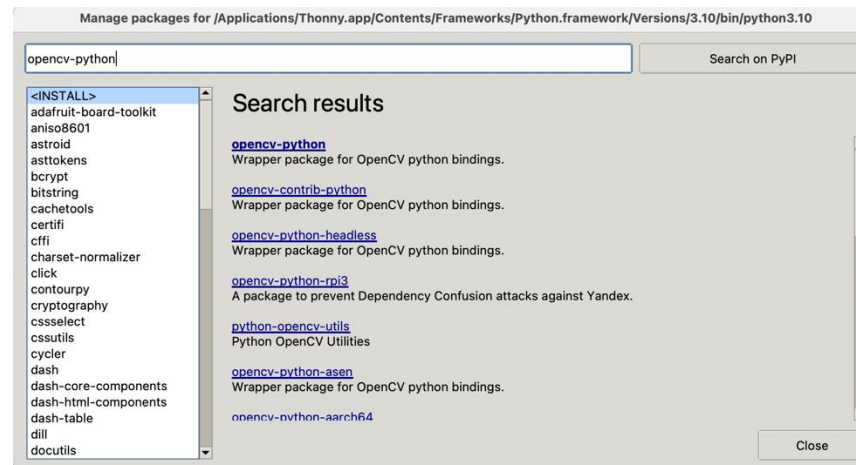


Explore Image Processing with OpenCV - Python library



OpenCV (Open Source Computer Vision Library) is a [library](#) of programming functions mainly for [real-time computer vision](#).^[1]

Originally developed by [Intel](#), it was later supported by [Willow Garage](#), then Itseez (which was later acquired by Intel^[2]). The library is [cross-platform](#) and licensed as [free and open-source software](#) under [Apache License](#) 2. Starting in 2011, OpenCV features GPU acceleration for real-time operations.^[3]



>>> sudo pip install opencv-python

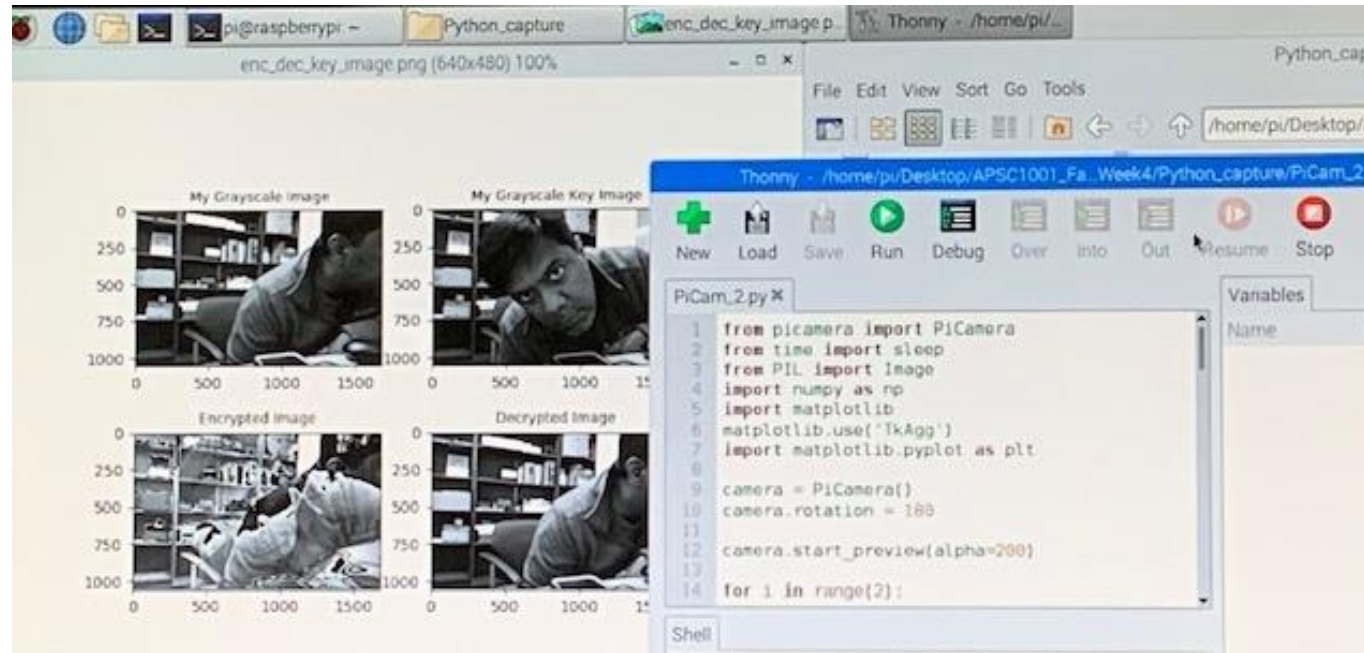
Let's mess with the PiCamera

Graded in-class lab
Download codes from shared-drive and demonstrate
[10 points]



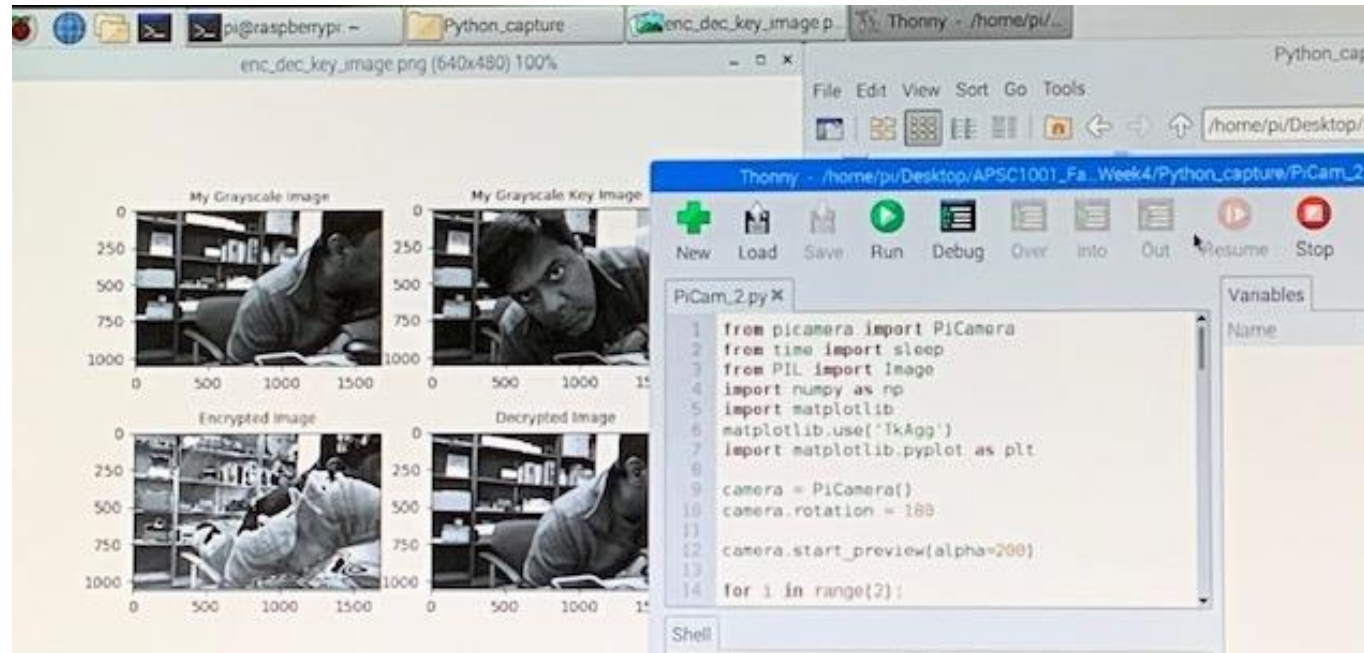
Use Raspberry Pi and PiCamera

1. To acquire images
2. To filter and transform image data
3. To detect objects using low-level ML functions



Use Raspberry Pi and PiCamera

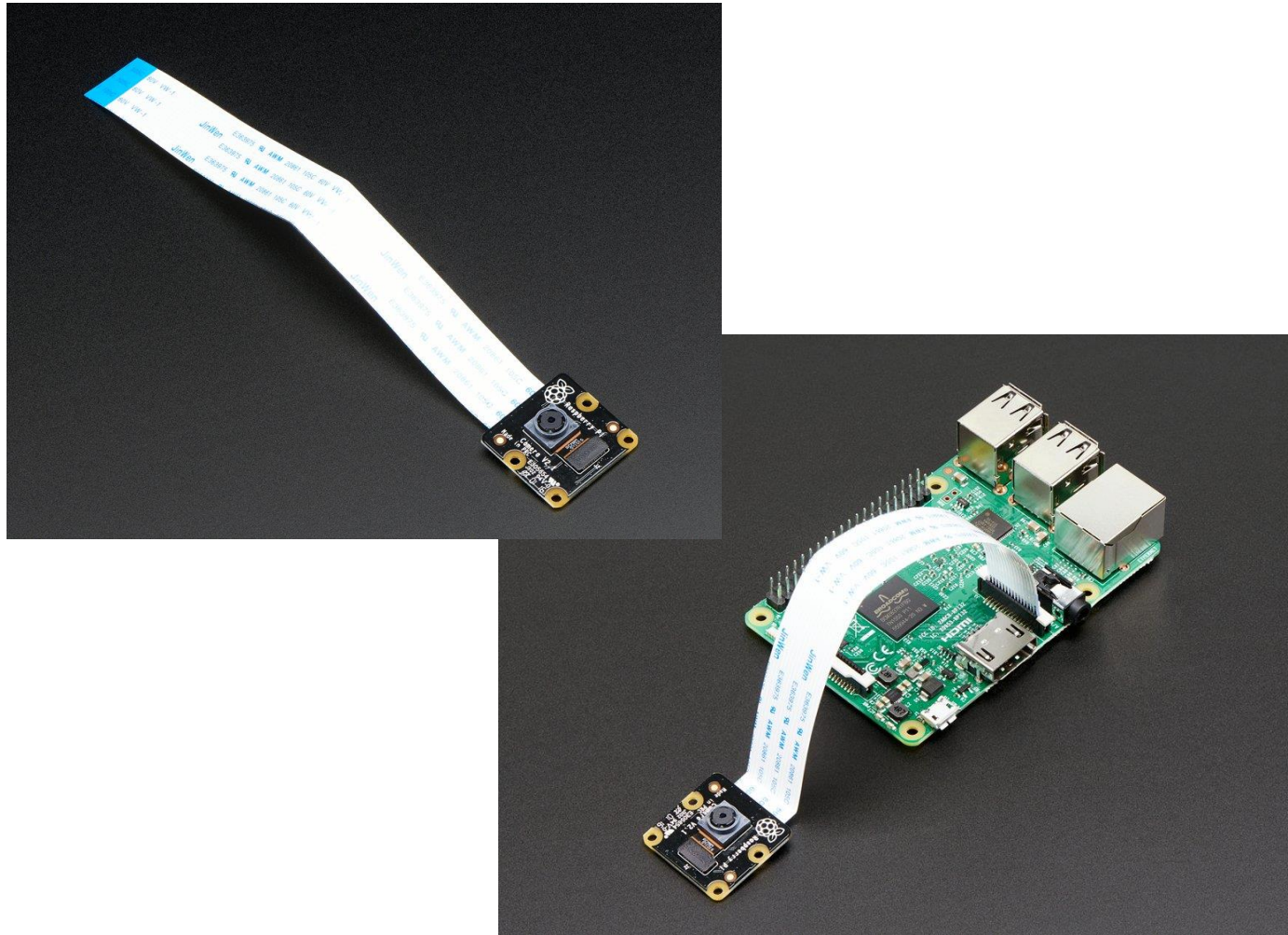
1. To acquire images
2. To perform on-board encryption-decryption
3. To filter and transform image data
4. To detect objects using low-level ML functions



Broad specs of the Pi NoIR Camera

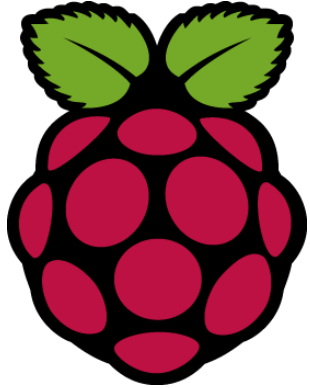
Source:

<https://www.adafruit.com/product/3100#description>



- 8 megapixel native resolution high quality Sony IMX219 image sensor
- 3280 x 2464 pixel static images
- Capture video at
 - 1920 x 1080 p30
 - 1280 x 720 p60
 - 640 x 480 p90 resolutions
- No Infrared (NoIR) filter
 - Infrared photographs or photographing objects in low light (twilight) conditions





Picamera2



Raspberry Pi hardware and Camera hardware interactions
Low-level image acquisition and processing

Computer vision on the Raspberry Pi



Picamera2



```
# import the necessary packages
```

```
from picamera2 import Picamera2
import time
import cv2
```

Picamera2

```
# initialize the camera and grab a reference
# to the raw camera capture
```

```
camera = Picamera2()
camera.resolution = (320, 240)
rawCapture = camera.capture_array("main")
```

```
# allow the camera to warmup
time.sleep(0.1)
```

Picamera2

```
# grab an image from the camera
```

```
camera.capture(rawCapture, format="bgr")
image = rawCapture.array
```

Skeleton of the OpenCV library-based Python program

```
#start the camera and display a preview
camera.start(show_preview=True)
```

```
# grab an image from the camera
camera.capture(rawCapture, format="bgr")
image = rawCapture.array
```



```
# display the image on screen
# and wait for a keypress
```

```
cv2.imshow("Image", image)
cv2.imwrite("savedImage.png", image)
cv2.waitKey(0)
```

```
cv2.destroyAllWindows()
camera.stop()
exit()
```



Goal of the lab segment

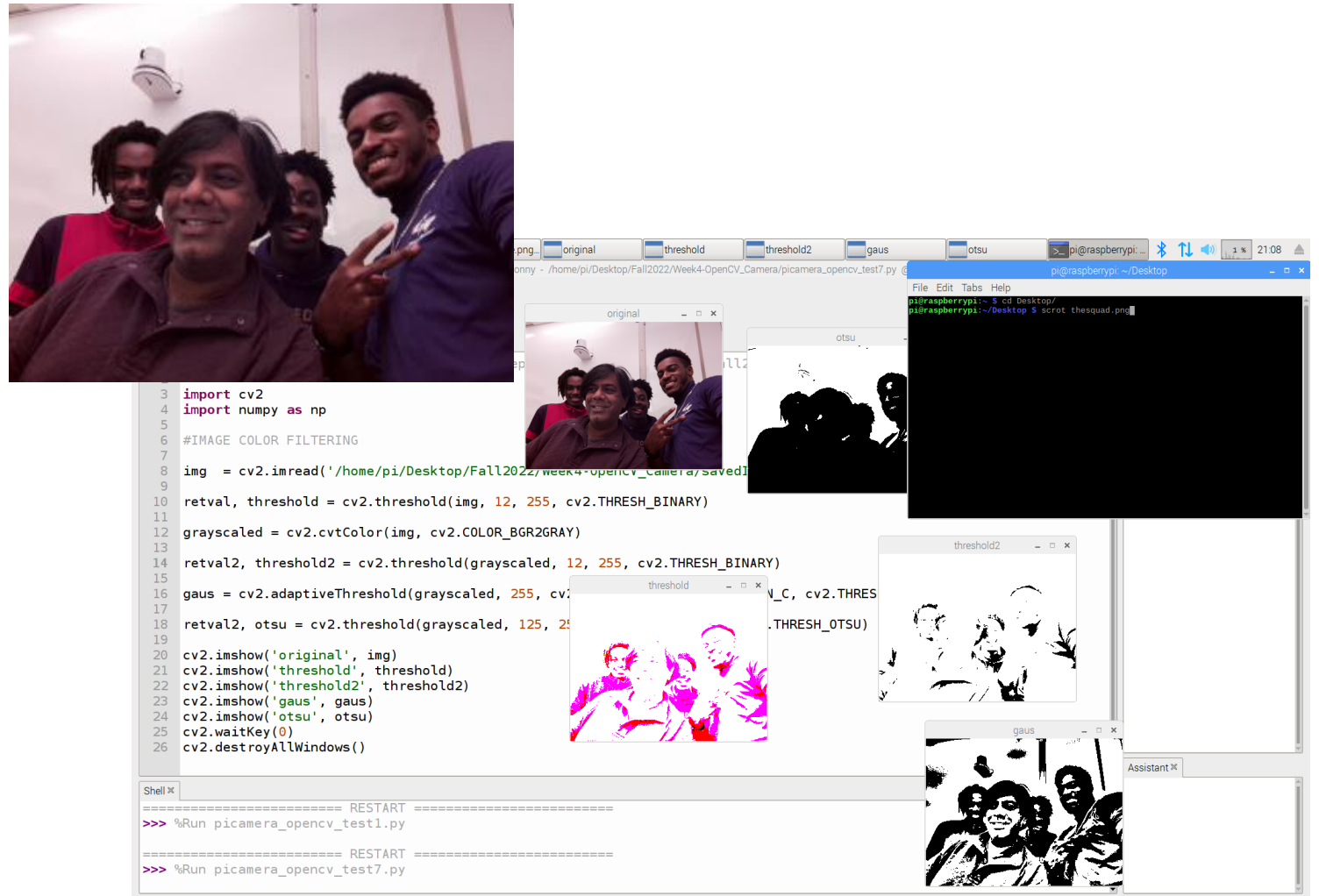
Co-work

- Observe, ask and try in groups

Make

- Build-a-hack
- Use Pi NoIR Camera to acquire an images
- import OpenCV library

Perform basic image processing functions using OpenCV



Very basics of data encryption-decryption and applications

Encryption is the transformation of data into some **unreadable form**.

Decryption is the reverse of encryption; it is the transformation of encrypted data back into some **intelligible form**.

Image encryption - process of encoding image with the help of an encryption algorithm in such a way that unauthorized users can't access it.

Authorization entails a "key".



Created by Round Icons
from the Noun Project



Building up the vocabulary: Shift cipher

Assign each letter in the alphabet a number

- Start from 0
- I have row matrix: $A_{1 \times 26}$

Message (X) = T E A C H

- Convert the letter into the number that matches its order in the alphabet starting from 0
- I now have a row matrix: $X_{1 \times 5} = [19 \ 4 \ 0 \ 2 \ 7]$

To encrypt assign a shift key (K) = 4

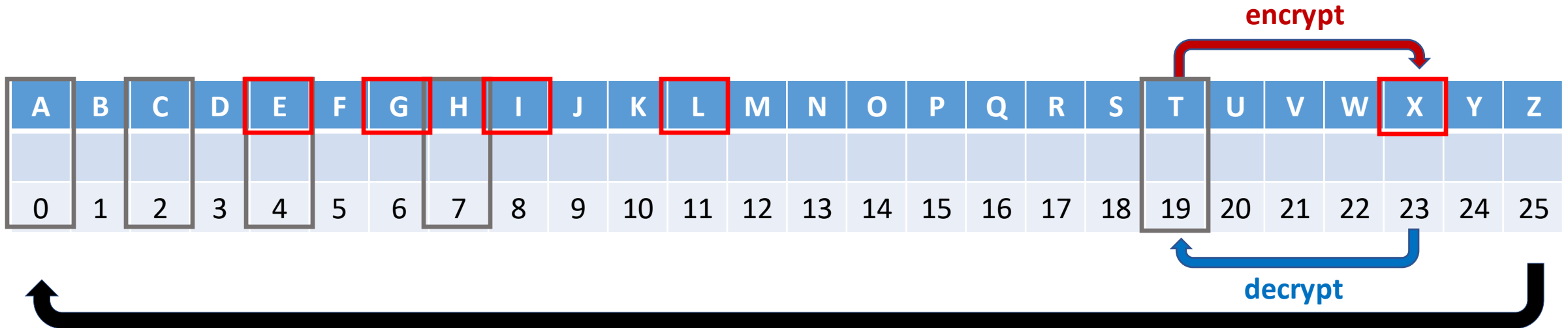
- Must be an integer from 0 to 25
- Map each letter to a different letter using the shift key
- $Y = (X + KJ)$ where J is a vector-of-ones i.e., $[1 \ 1 \ 1 \ 1 \ \dots]$
- I have a new row matrix: $Y_{1 \times 5} = [23 \ 8 \ 4 \ 6 \ 11]$

Encrypted message (Y) = X I E G L

To decrypt apply the same shift key (K) = 4

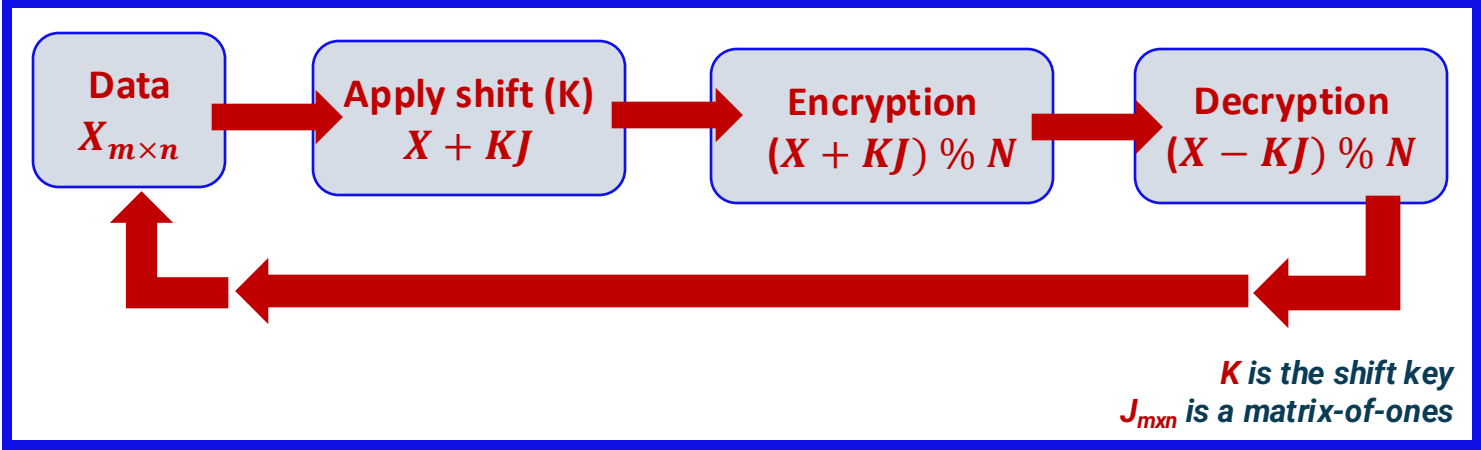
- Map each encrypted letter to a different letter using the shift key
- $X = (Y - KJ)$ where J is a vector-of-ones i.e., $[1 \ 1 \ 1 \ 1 \ \dots]$
- I have a new row matrix: $X_{1 \times 5} = [19 \ 4 \ 0 \ 2 \ 7]$

Decrypted Message (X) = T E A C H



Building up the vocabulary: Modulo

In computing, the **modulo operation** finds the **remainder after division** of one number by another.



Encryption

	T	E	A	C	H
X	19	4	0	2	7
K	20	20	20	20	20
X + K	39	24	20	22	27
(X + K)%N	13	24	20	22	1
	N	Y	U	W	B

Decryption

	N	Y	U	W	B
Y	13	24	20	22	1
K	20	20	20	20	20
Y - K	-7	4	0	2	-19
(Y - K)%N	19	4	0	2	7
	T	E	A	C	H

The in-class programming exercise will demonstrate these operations on images using Python



Apply Python to process the image



```
import cv2
```

```
import PIL
```

```
Perform encryption and decryption
```



Apply OpenCV + Python to filter the image

Simple Thresholding

The basic Thresholding technique is Binary Thresholding.

For every pixel, the same threshold value is applied.

If the pixel value is smaller than the threshold, it is set to 0, otherwise, it is set to a maximum value.



`cv2.threshold(source, thresholdValue, maxVal, thresholdingTechnique)`

Parameters:

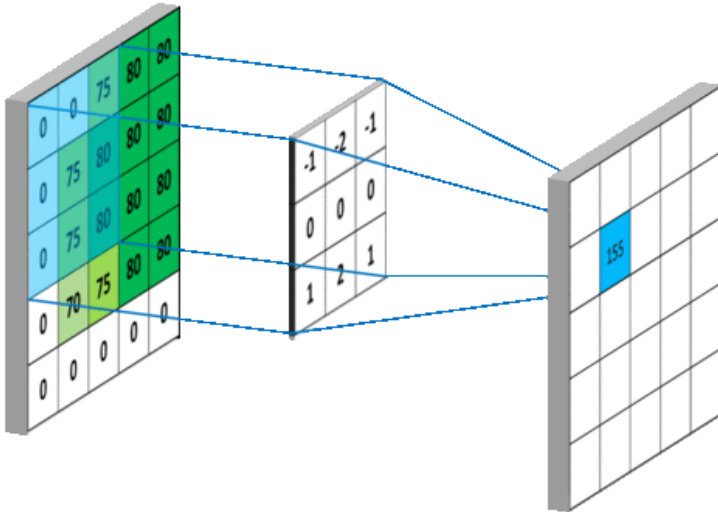
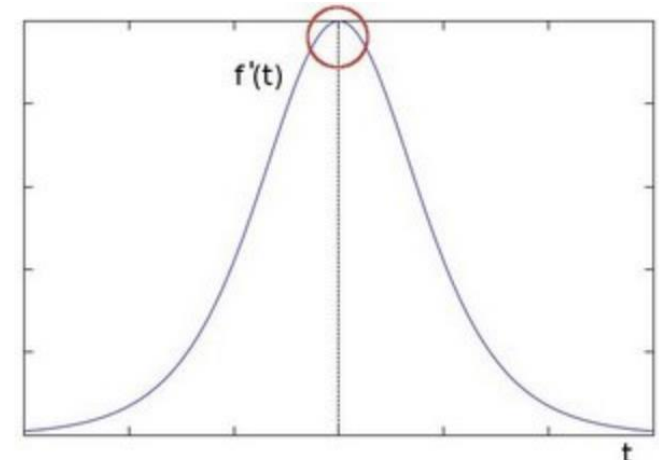
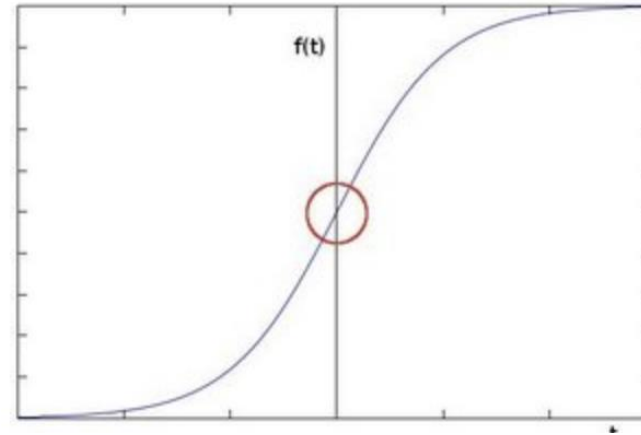
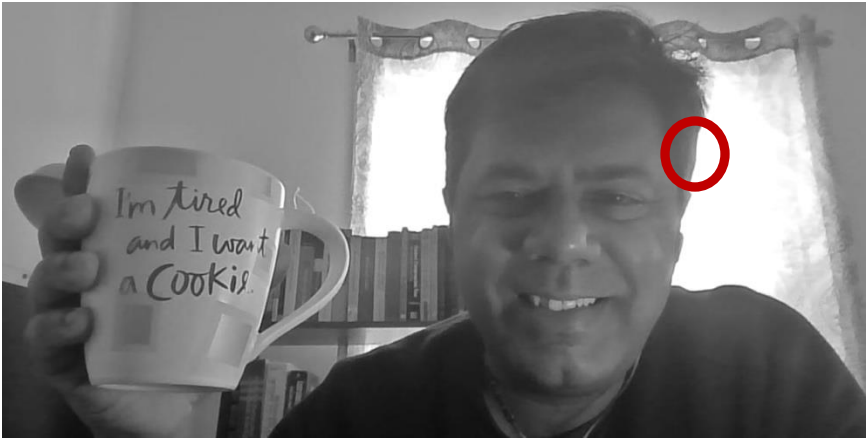
- **source:** Input Image array (must be in Grayscale).
- **thresholdValue:** Value of Threshold below and above which pixel values will change accordingly.
- **maxVal:** Maximum value that can be assigned to a pixel.
- **thresholdingTechnique:** The type of thresholding to be applied.



Apply OpenCV + Python to detect edges

Source:

<https://mlnotebook.github.io/post/CNN1/>
https://docs.opencv.org/3.4/d2/d2c/tutorial_sobel_derivatives.html



```
laplacian = cv2.Laplacian(frame, cv2.CV_64F)
sobelx = cv2.Sobel(frame, cv2.CV_64F, 1, 0, ksize=5)
sobely = cv2.Sobel(frame, cv2.CV_64F, 0, 1, ksize=5)
edges = cv2.Canny(frame, 100, 200)
```

$$G_x = \begin{bmatrix} -1 & 0 & +1 \\ -2 & 0 & +2 \\ -1 & 0 & +1 \end{bmatrix} * I$$

$$G_y = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ +1 & +2 & +1 \end{bmatrix} * I$$

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



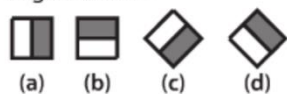
Apply OpenCV + Python to detect objects

```
face_cascade = cv2.CascadeClassifier("/home/pi/opencv-3.4.1/data/haarcascades/haarcascade_frontalface_default.xml")
```

Haar features

OpenCV's algorithm is currently using the following Haar-like features which are the input to the basic classifiers:

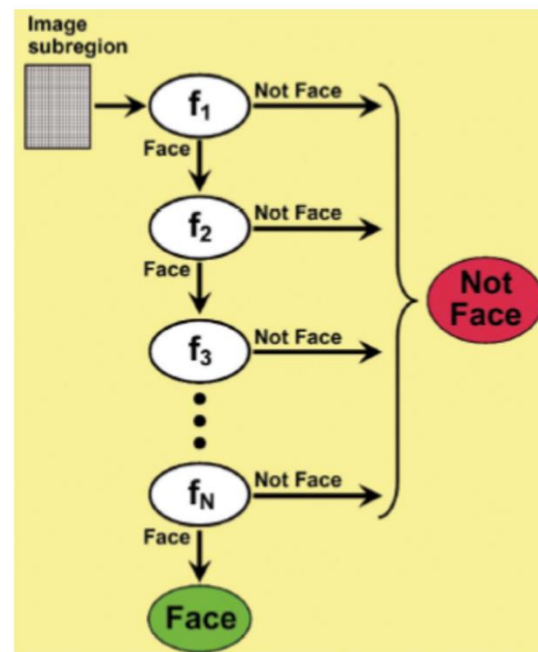
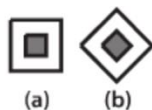
1. Edge features



2. Line features



3. Center-surround features



OpenCV's pre-trained classifiers

OpenCV already contains many pre-trained classifiers for face, eyes, smile etc. Those XML files are stored in **opencv/data/haarcascades/** folder:

```
~/OpenCV/opencv/data/haarcascades$ ls
```

```
haarcascade_eye_tree_eyeglasses.xml  haarcascade_mcs_lefttear.xml
haarcascade_eye.xml                  haarcascade_mcs_lefteye.xml
haarcascade_frontalface_alt2.xml      haarcascade_mcs_mouth.xml
haarcascade_frontalface_alt_tree.xml  haarcascade_mcs_nose.xml
haarcascade_frontalface_alt.xml       haarcascade_mcs_righttear.xml
haarcascade_frontalface_default.xml   haarcascade_mcs_righteye.xml
haarcascade_fullbody.xml              haarcascade_mcs_upperbody.xml
haarcascade_lefteye_2splits.xml       haarcascade_profileface.xml
haarcascade_lowerbody.xml             haarcascade_righteye_2splits.xml
haarcascade_mcs_eyepair_big.xml       haarcascade_smile.xml
haarcascade_mcs_eyepair_small.xml     haarcascade_upperbody.xml
```



Source:
https://www.bogotobogo.com/python/OpenCV_Python/python_opencv3_Image_Object_Detection_Face_Detection_Haar_Cascade_Classifiers.php
Navneet Dalal and Bill Triggs. [Histograms of oriented gradients for human detection](#). In *Computer Vision and Pattern Recognition, 2005. CVPR 2005. IEEE Computer Society Conference on*, volume 1, pages 886–893. IEEE, 2005.
<https://www.intel.com/content/www/us/en/docs/ipp/developer-reference/2021-7/histogram-of-oriented-gradients-hog-descriptor.html>

Apply OpenCV + Python to detect humans

```
hog = cv2.HOGDescriptor()  
hog.setSVMDetector(cv2.HOGDescriptor_getDefaultPeopleDetector())
```

Implementation of HOG (Histogram of Oriented Gradients) descriptor and object detector.

Histogram of oriented gradients (HOG) is a feature descriptor

- used to detect objects in computer vision and image processing.

The HOG descriptor technique counts

- occurrences of gradient orientation in localized portions of an image - detection window, or region of interest (ROI).

