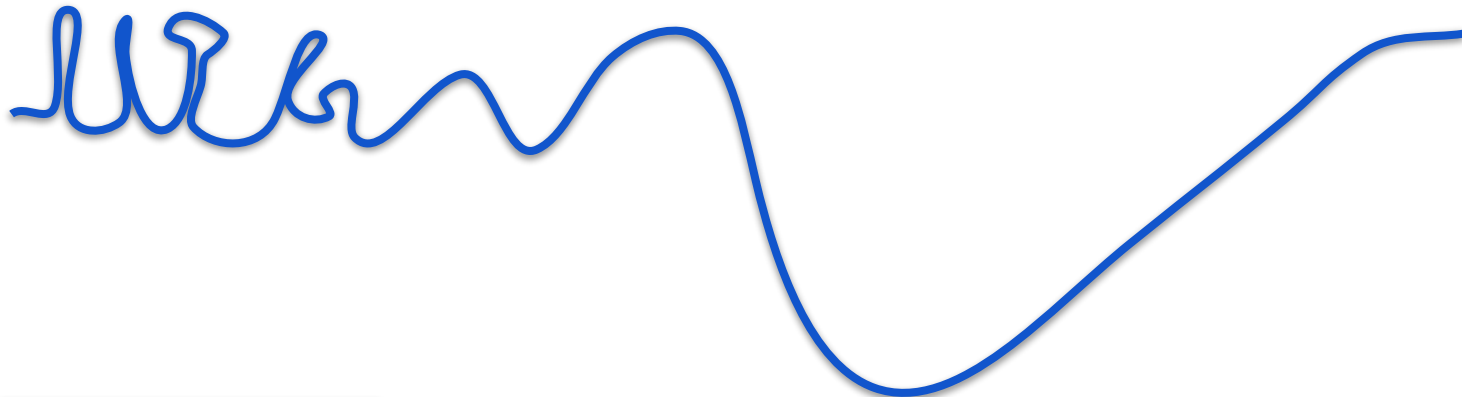


Computing with Signals



DA 623

Jan - May 2024

IIT Guwahati

Instructors: Neeraj Sharma

Lecture-17

Raise your hand to explain what information is this following code snippet providing?

```
1 import matplotlib.pyplot as plt
2 import numpy as np
3
4 im = plt.imread('./rohini_godbole.jpeg')
5 print('Datatype: ', type(im))
6 print('Size of data:', im.shape)
```

```
Datatype: <class 'numpy.ndarray'>
```

```
Size of data: (500, 600, 3)
```

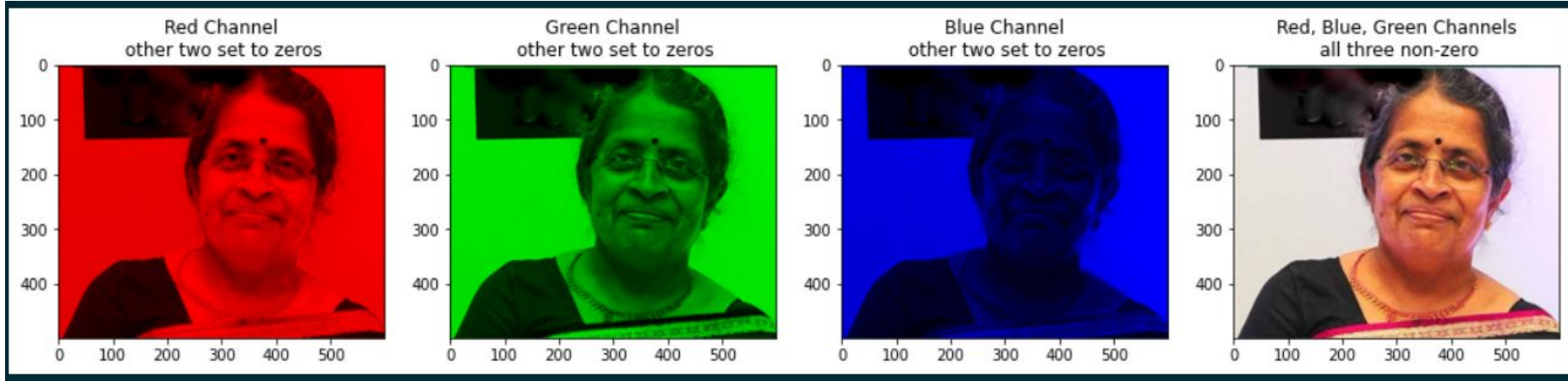
Raise your hand to explain what information is this following code snippet providing?

```
1 import matplotlib.pyplot as plt
2 import numpy as np
3
4 im = plt.imread('./rohini_godbole.jpeg')
5 print('Datatype: ', type(im))
6 print('Size of data:', im.shape)
```

```
Datatype: <class 'numpy.ndarray'>
Size of data: (500, 600, 3)
```

It is a numpy array - $M \times N \times C$, where

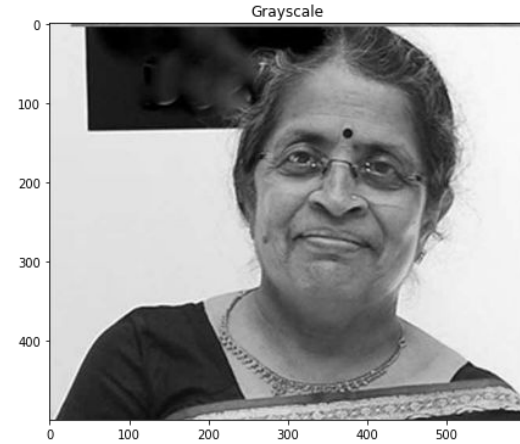
- M is the number of rows (image height)
- N is number of columns (image width)
- C is number of layers (image channels)



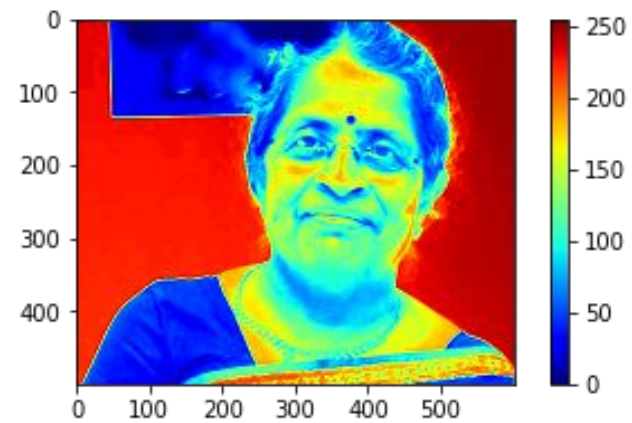
Rohini Godbole is an Indian physicist and academic specializing in elementary particle physics: field theory and phenomenology. She is currently a professor at the Centre for High Energy Physics, Indian Institute of Science, Bangalore.

Raise your hand to explain what information is this following code snippet providing?

```
1 def rgb2gray(im):  
2     """  
3     Converts RGB image to grayscale image  
4     """  
5     # write one line to compute the mean across the channel  
6     im_temp = im.mean(axis=2)  
7     return im_temp  
8  
9     # convert color image to grayscale  
10    im_gs = rgb2gray(im)  
11  
12    # plot image  
13    fig, ax = plt.subplots(nrows=1, ncols=1, figsize=[7,7])  
14    ax.imshow(im_gs, cmap='gray')  
15    ax.set_title('Grayscale')  
16    plt.show()
```



Choose colorbar with care
It impacts visualization



Perspective | [Open access](#) | [Published: 28 October 2020](#)

The misuse of colour in science communication

[Fabio Crameri](#) , [Grace E. Shephard](#) & [Philip J. Heron](#)

[Nature Communications](#) **11**, Article number: 5444 (2020) | [Cite this article](#)

291k Accesses | **353** Citations | **1323** Altmetric | [Metrics](#)

Choose colorbar with care
It impacts visualization



Perspective | [Open access](#) | [Published: 28 October 2020](#)

The misuse of colour in science communication

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[Nature Communications](#) **11**, Article number: 5444 (2020) | [Cite this article](#)

291k Accesses | **353** Citations | **1323** Altmetric | [Metrics](#)

Raise your hand to explain what information is this following code snippet providing?

- Choose `thres`
- Create `im_bw` from `im_gs` by setting all values $> \text{thres}$ to 255 and rest to 0.

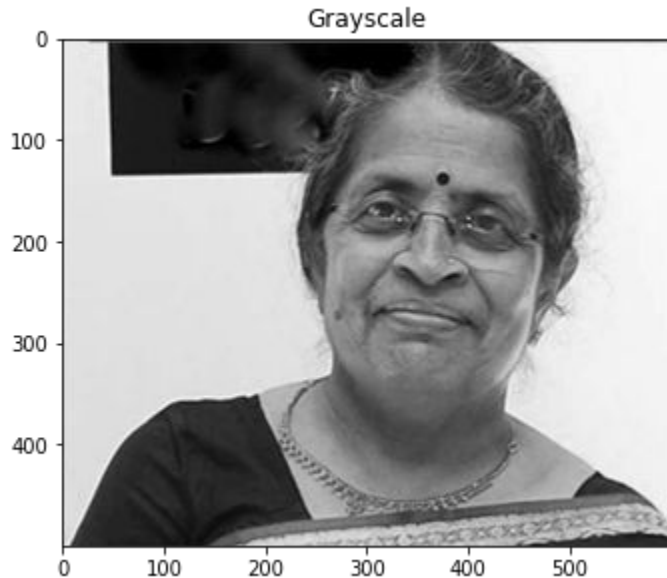
Raise your hand to explain what information is this following code snippet providing?

- Choose `thres`
- Create `im_bw` from `im_gs` by setting all values $> \text{thres}$ to 255 and rest to 0.

```
1  thres = 255//2
2  im_bw = np.zeros(im_gs.shape)
3  im_bw[im_gs>thres] = 255
```

Raise your hand to explain what information is this following code snippet providing?

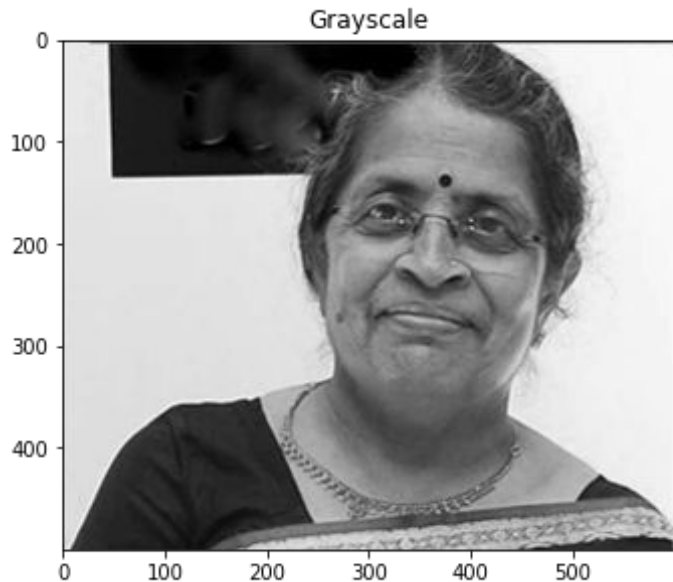
- Choose `thres`
- Create `im_bw` from `im_gs` by setting all values $> \text{thres}$ to 255 and rest to 0.



The black and white image is losing a lot of details.

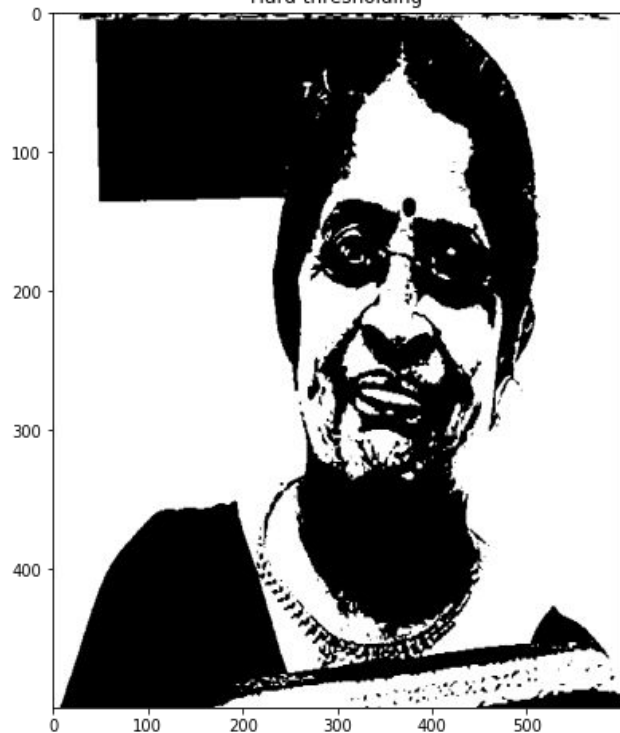
Can we improvise the black and white image?

- Choose `thres`
- Create `im_bw` from `im_gs` by setting all values $> \text{thres}$ to 255 and rest to 0.



```
# make binary image using random switching the 0 intensity pixels
im_rd = np.zeros(im_gs.shape)
im_rd[im_gs>thres] = 255
im_random = np.random.randint(0,2,im_gs.shape)
im_rd[im_gs<=thres] = 255*im_random[im_gs<=thres]
```

Black and White
Hard thresholding



Black and white
Random Dithering

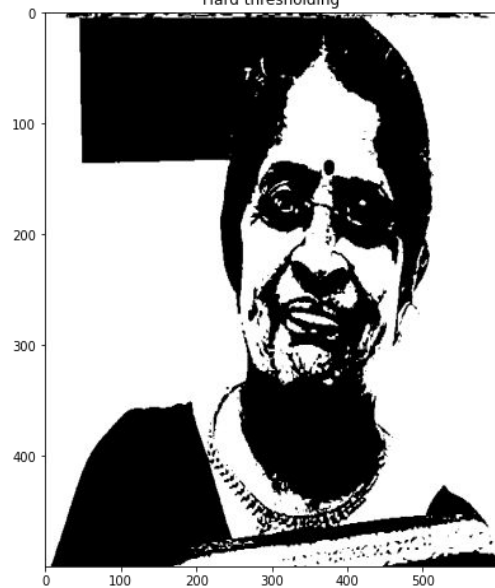


The black and white image is losing a lot of details.

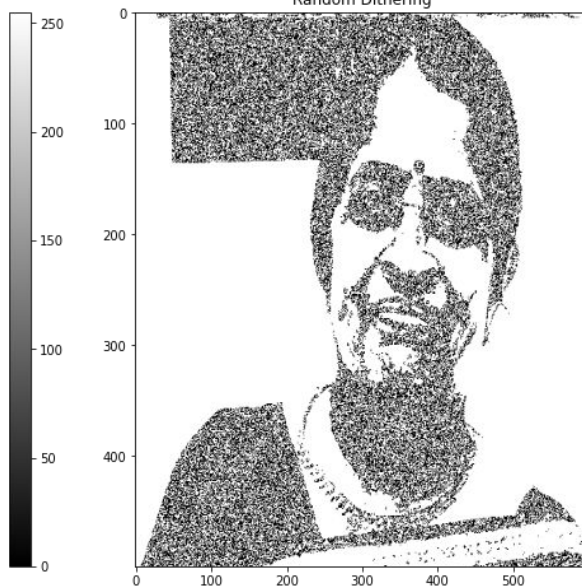
Can we improvise the black and white image?

```
2 def floyd_steinberg(im_gs):
3     """
4     Implements the Floyd steinberg algorithm to create binary image from grayscale image
5     """
6     thres = 255//2
7     im_temp = im_gs.copy()
8     for row in range(0,im_gs.shape[0]-1):
9         for col in range(1,im_gs.shape[1]-1):
10             old = im_temp[row, col]
11             if im_temp[row, col] > thres:
12                 im_temp[row, col] = 255
13             else:
14                 im_temp[row, col] = 0
15
16             quant_error = old - im_temp[row, col]
17             im_temp[row, col+1] = im_temp[row, col+1] + quant_error * 7 / 16
18             im_temp[row+1, col-1] = im_temp[row+1, col-1] + quant_error * 3 / 16
19             im_temp[row+1, col] = im_temp[row+1, col] + quant_error * 5 / 16
20             im_temp[row+1, col+1] = im_temp[row+1, col+1] + quant_error * 1 / 16
21
22     im_temp[im_temp<thres] = 0
23     im_temp[im_temp>=thres] = 255
24
25     return im_temp
```

Black and White
Hard thresholding



Black and white
Random Dithering



Black and white
Floyd Steinberg Dithering





Floyd–Steinberg
dithering

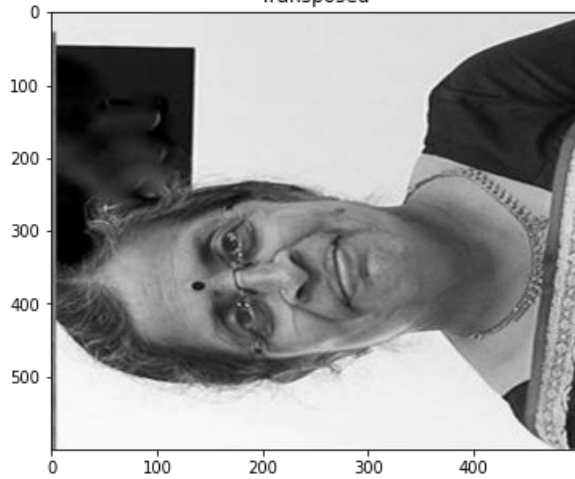
The black and white image is losing a lot of details.

Can we improvise the black and white image?

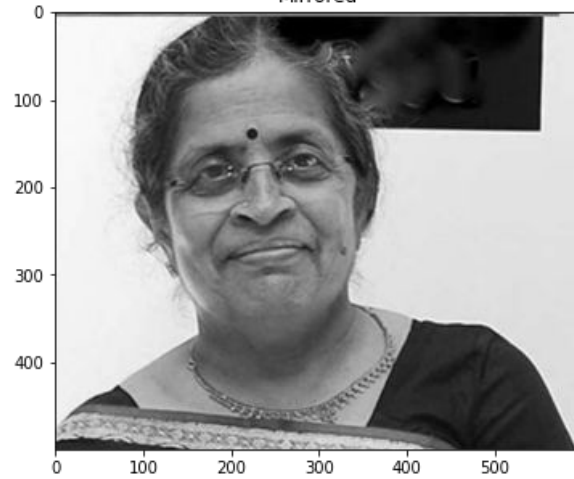
```
2 def floyd_steinberg(im_gs):
3     """
4     Implements the Floyd steinberg algorithm to create binary image from grayscale image
5     """
6     thres = 255//2
7     im_temp = im_gs.copy()
8     for row in range(0,im_gs.shape[0]-1):
9         for col in range(1,im_gs.shape[1]-1):
10             old = im_temp[row, col]
11             if im_temp[row, col] > thres:
12                 im_temp[row, col] = 255
13             else:
14                 im_temp[row, col] = 0
15
16             quant_error = old - im_temp[row, col]
17             im_temp[row, col+1] = im_temp[row, col+1] + quant_error * 7 / 16
18             im_temp[row+1, col-1] = im_temp[row+1, col-1] + quant_error * 3 / 16
19             im_temp[row+1, col] = im_temp[row+1, col] + quant_error * 5 / 16
20             im_temp[row+1, col+1] = im_temp[row+1, col+1] + quant_error * 1 / 16
21
22     im_temp[im_temp<thres] = 0
23     im_temp[im_temp>=thres] = 255
24
25     return im_temp
```


Simple image manipulation.

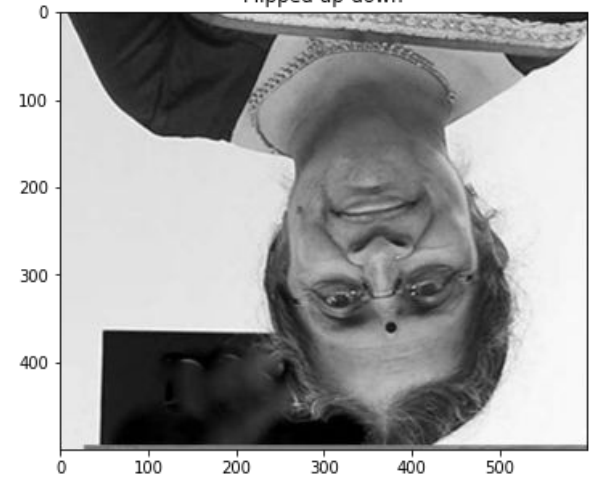
Transposed



Mirrored



Flipped up-down



Some of the slides in the following material is borrowed from the following excellent course material:



course by Prof. Shree K. Nayar, Columbia University



Reference: <https://fpcv.cs.columbia.edu/Monographs>

(used here for educational purpose)

Binary Images

Binary Images

Binary Image: Can have only two values (0 or 1).
Simple to process and analyze.



Making Binary Images

Binary Image $b(x, y)$: Usually obtained from Gray-level image $g(x, y)$ by Thresholding.

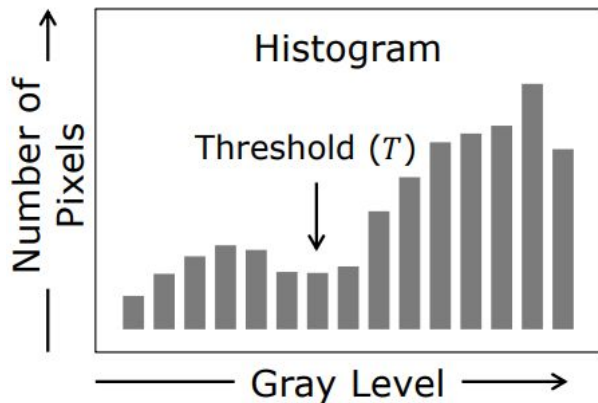
Characteristic Function:

$$b(x, y) = \begin{cases} 0, & g(x, y) < T \\ 1, & g(x, y) \geq T \end{cases}$$

Selecting a Threshold (T)

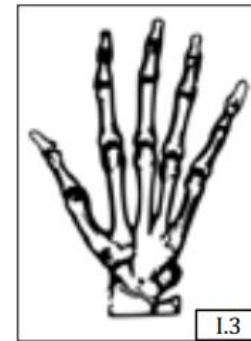
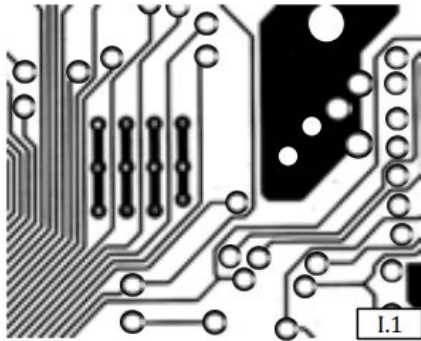


Gray Image $g(x,y)$



Binary Image $b(x,y)$

Examples of Binary Images



Capturing a Binary Images



Backlighting

Reference:
<https://fpcv.cs.columbia.edu/Monographs>

Binary Images

Binary Image: Can have only two values (0 or 1).
Simple to process and analyze.

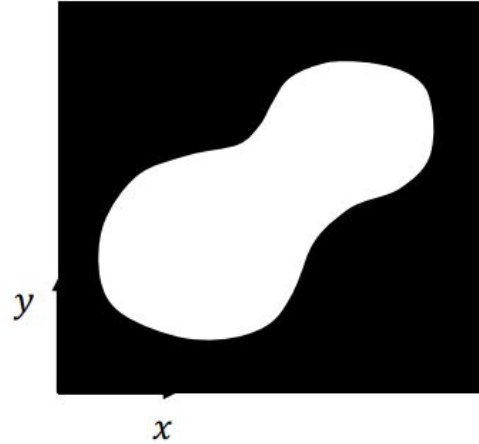
Topics:

- (1) Geometric Properties
- (2) Segmenting Binary Images
- (3) Iterative Modification

Geometric Properties of Binary Images

Assume:

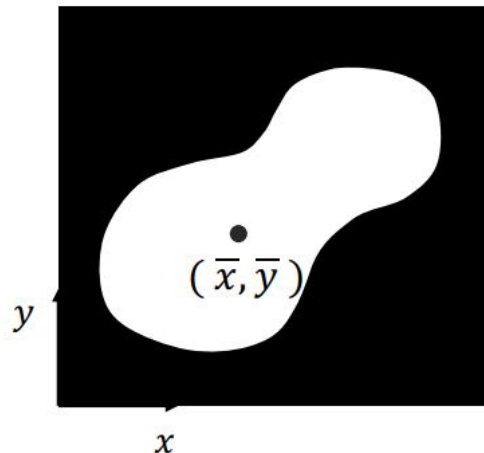
- $b(x, y)$ is continuous
- Only one object



Area and Position

Area: (Zeroth Moment)

$$A = \iint_I b(x, y) dx dy$$

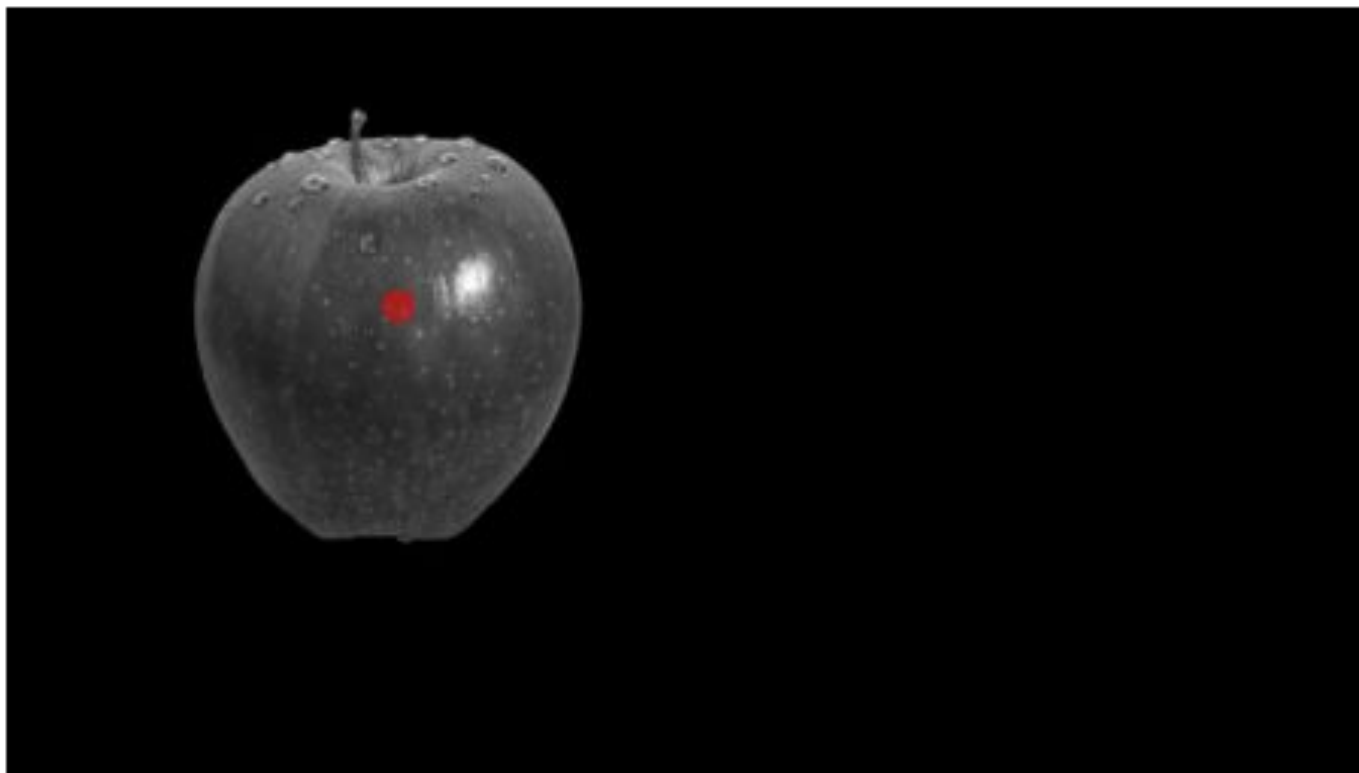


Position: Center of Area (First Moment)

$$\bar{x} = \frac{1}{A} \iint_I x b(x, y) dx dy \quad , \quad \bar{y} = \frac{1}{A} \iint_I y b(x, y) dx dy$$

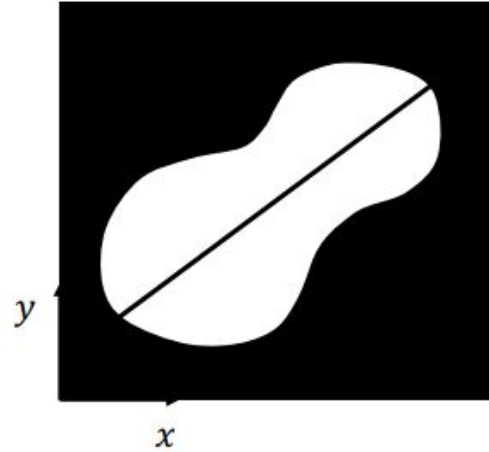
Reference:

<https://fpcv.cs.columbia.edu/Monographs>










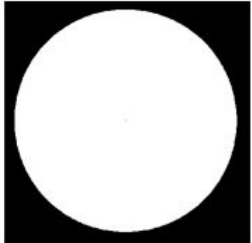
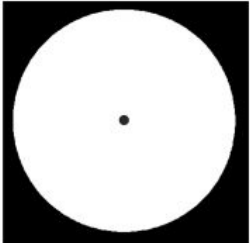
Orientation

Difficult to define!



Use: Axis of Least Second Moment

Examples

Gray Image	Binary Image	Orientation	Roundedness
			0.19
			0.49
			1.0

Given a binary image of an object, we can

- compute its area, location, and its maximum and minimum moments
- the area and the two moments are useful features because they are not affected by translation and rotation of the object
- use case: distinguish between a set of objects

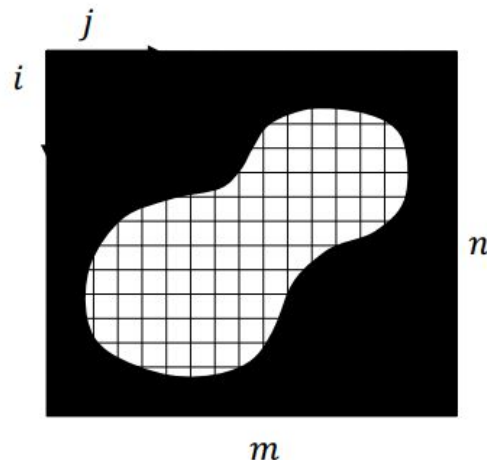
After an object is recognized, the position and the orientation are used to enable a robot to pick up the object.

Discrete Binary Images

b_{ij} : Value at cell (pixel) in row i and column j .

Assume pixel area = 1.

Area:
$$A = \sum_{i=1}^n \sum_{j=1}^m b_{ij}$$



Position: Center of Area (First Moment)

$$\bar{x} = \frac{1}{A} \sum_{i=1}^n \sum_{j=1}^m i b_{ij} \qquad \bar{y} = \frac{1}{A} \sum_{i=1}^n \sum_{j=1}^m j b_{ij}$$

Discrete Binary Images

Second Moments:

$$a' = \sum_{i=1}^n \sum_{j=1}^m i^2 b_{ij} \quad b' = 2 \sum_{i=1}^n \sum_{j=1}^m ij b_{ij} \quad c' = \sum_{i=1}^n \sum_{j=1}^m j^2 b_{ij}$$

Note: a' , b' , c' are second moments w.r.t origin.

a , b , c (w.r.t. center) can be found from a' , b' , c' , \bar{x} , \bar{y} , A

Hint: Expand $a = \sum_{i=1}^n \sum_{j=1}^m (i - \bar{x})^2 b_{ij}$ and represent in terms of a' , \bar{x} , A .

Multiple Objects



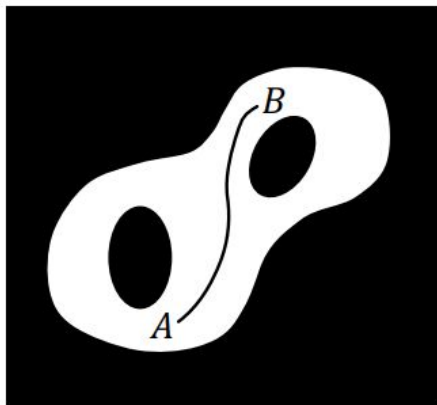
Need to Segment image into separate Components

Non-Trivial!

Reference:
<https://fpcv.cs.columbia.edu/Monographs>

Connected Component

Maximal Set of Connected Points



A and B are connected if path exists between A and B
along which $b(x, y)$ is constant.

Connected Component Labeling

Region Growing Algorithm

- (a) Find Unlabeled “Seed” point with $b = 1$.
If not found, Terminate.
- (b) Assign New Label to seed point
- (c) Assign Same Label to its Neighbors with $b = 1$
- (d) Assign Same Label to Neighbors of Neighbors with $b = 1$. Repeat until no more Unlabeled Neighbors with $b=1$.
- (e) Go to (a)

Pixel Processing

Image as a Function



$f(x, y)$ is the image intensity at position (x, y)

Pixel (Point) Processing

Transformation T of intensity f at each pixel to intensity g :

$$g(x, y) = T(f(x, y))$$

Point Processing



Original (f)



Darken ($f - 128$)



Lighten ($f + 128$)



Invert ($255 - f$)

Pixel Processing



Original (f)



Low Contrast ($f/2$)



High Contrast ($f * 2$)



Gray ($0.3f_R + 0.6f_G + 0.1f_B$)

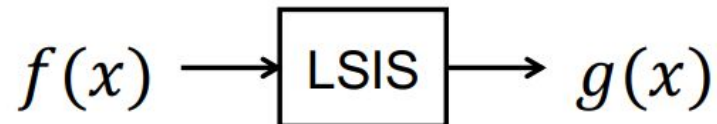
Image Processing

Image as a Function



$f(x,y)$ is the image intensity at position (x,y)

Linear Shift Invariant System (LSIS)



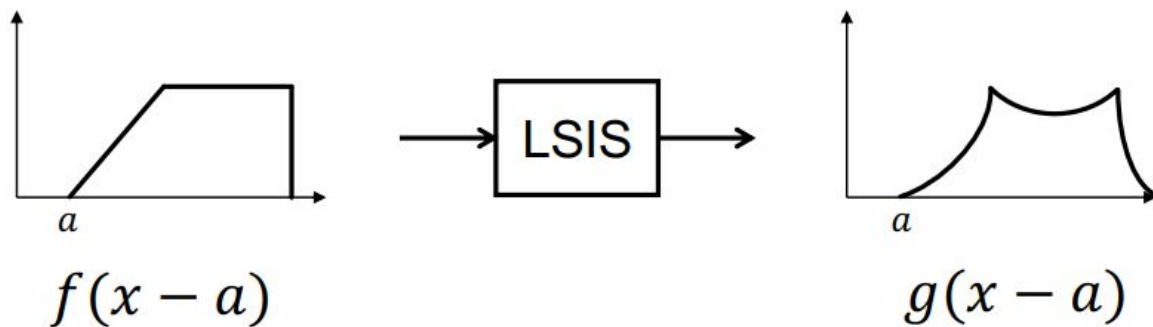
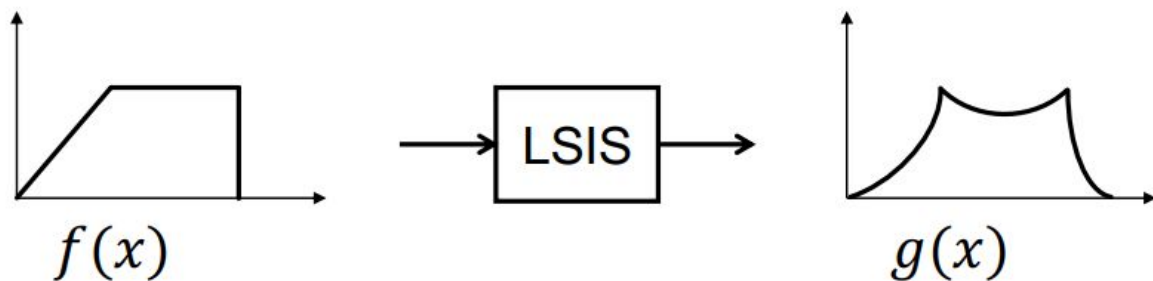
Study of Linear Shift Invariant Systems (LSIS)
leads to useful image processing algorithms.

LSIS: Linearity

$$f_1 \longrightarrow \boxed{\text{LSIS}} \longrightarrow g_1 \qquad f_2 \longrightarrow \boxed{\text{LSIS}} \longrightarrow g_2$$

$$\alpha f_1 + \beta f_2 \longrightarrow \boxed{\text{LSIS}} \longrightarrow \alpha g_1 + \beta g_2$$

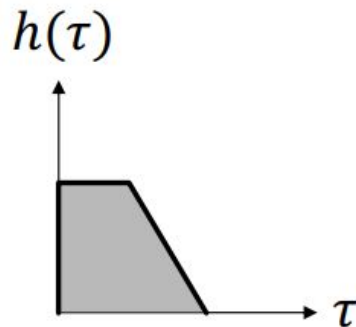
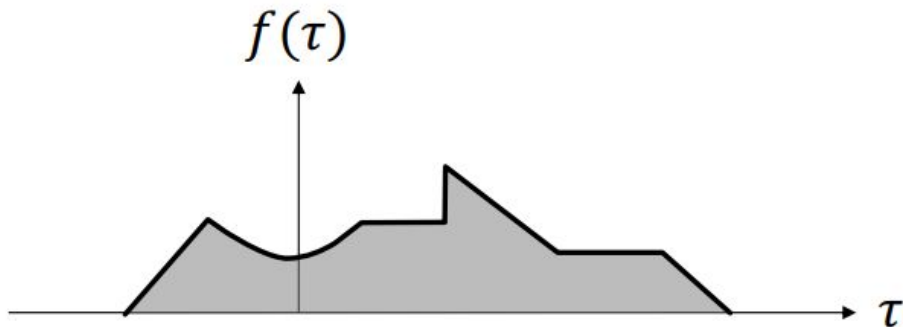
LSIS: Shift Invariance



Convolution

Convolution of two functions $f(x)$ and $h(x)$

$$g(x) = f(x) * h(x) = \int_{-\infty}^{\infty} f(\tau)h(x - \tau) d\tau$$



Reference:
<https://fpcv.cs.columbia.edu/Monographs>

Convolution is LSIS

Linearity:

$$\text{Let: } g_1(x) = \int_{-\infty}^{\infty} f_1(\tau)h(x - \tau) d\tau \quad \text{and} \quad g_2(x) = \int_{-\infty}^{\infty} f_2(\tau)h(x - \tau) d\tau$$

Then:

$$\begin{aligned} & \int_{-\infty}^{\infty} (\alpha f_1(\tau) + \beta f_2(\tau))h(x - \tau) d\tau \\ &= \alpha \int_{-\infty}^{\infty} f_1(\tau)h(x - \tau) d\tau + \beta \int_{-\infty}^{\infty} f_2(\tau)h(x - \tau) d\tau \\ &= \alpha g_1(x) + \beta g_2(x) \end{aligned}$$

Convolution is LSIS

Shift Invariance:

Let: $g(x) = \int_{-\infty}^{\infty} f(\tau)h(x - \tau) d\tau$

Then:

$$\int_{-\infty}^{\infty} f(\tau - a)h(x - \tau) d\tau$$

$$= \int_{-\infty}^{\infty} f(\mu)h(x - a - \mu) d\mu \quad \boxed{1} \quad (\text{Substituting } \mu = \tau - a)$$

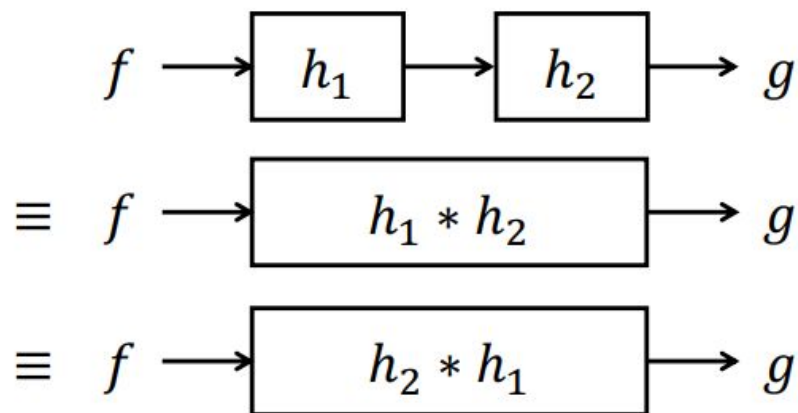
$$= g(x - a)$$

Properties of Convolution

Commutative $a * b = b * a$

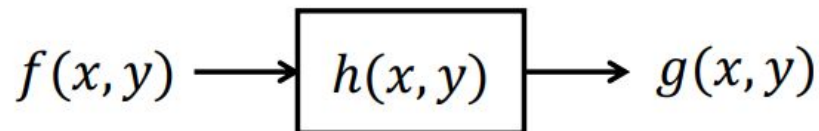
Associative $(a * b) * c = a * (b * c)$

Cascaded System



2D Convolution

LSIS:



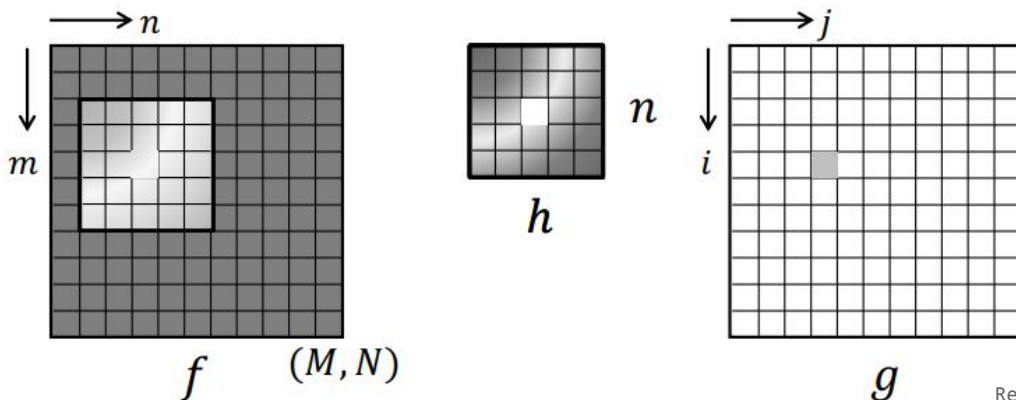
Convolution:

$$g(x, y) = \iint_{-\infty}^{\infty} f(\tau, \mu) h(x - \tau, y - \mu) d\tau d\mu$$

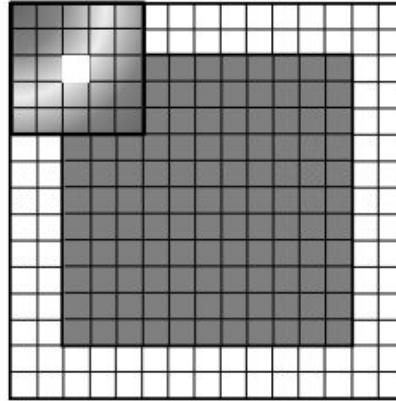
Convolution with Discrete Images

$$f[i,j] \longrightarrow \boxed{h[i,j]} \longrightarrow g[i,j]$$

$$g[i,j] = \sum_{m=1}^M \sum_{n=1}^N f[m,n] \underbrace{h[i-m, j-n]}_{\text{"Mask," "Kernel," "Filter"}}$$



Border Problem



Solution:

- Ignore border
- Pad with constant value
- Pad with reflection

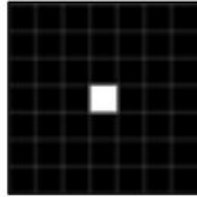
Example: Impulse Filter

Input



$f(x, y)$

*



$\delta(x, y)$

=

Output



$f(x, y)$

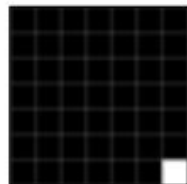
Example: Image Shift

Input



$f(x, y)$

*



=

Output



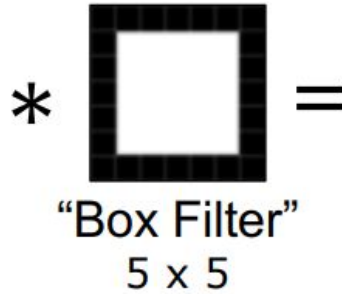
$f(x - u, y - v)$

Example: Averaging

Input



$f(x, y)$



$a(x, y)$

Output



$g(x, y)$

Result Image is saturated. Why?

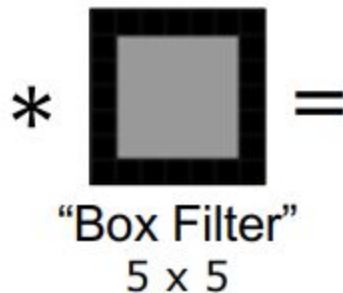
Reference:
<https://fpcv.cs.columbia.edu/Monographs>

Example: Averaging

Input



$f(x, y)$



$a(x, y)$

Output



$g(x, y)$

Sum of all the filter (kernel) weights should be 1.

Reference:

<https://fpcv.cs.columbia.edu/Monographs>