

## 211651 DIGITAL IMAGE PROCESSING

Punnarai Siricharoen, Ph.D.



# LECTURE 01 INTRODUCTION

Punnarai Siricharoen, Ph.D.

## LECTURE 01 - OBJECTIVES

1. To introduce Digital Imaging and Image processing overview
2. To describe the example fields areas of digital imaging and applications
3. To use python for read/write/show an image

## LECTURE 01 - OUTLINE

1. About this course (Syllabus)
2. Introduce Digital Imaging, applications and Challenges
3. Principal areas of digital imaging and applications
4. Python Exercises for read/write/show an image

# **ABOUT THE COURSE**

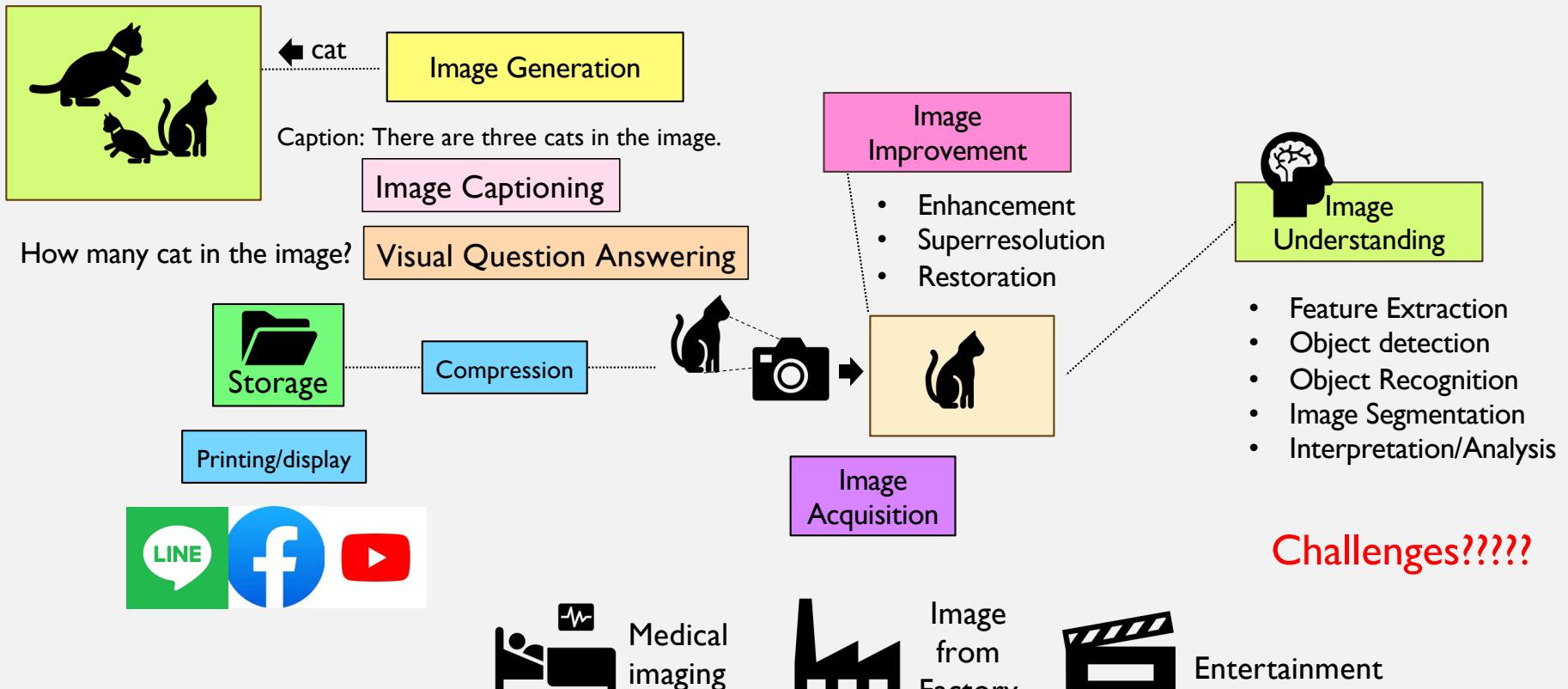
## COMMUNICATIONS

- Onsite Lecture (Online sometimes) at 17-02 (Sec1), 18-16 (Sec5)
- MCV course number : 2110651
- password : **2110651** for course materials
- Discord : <https://discord.gg/Wg3nhjCh>



WHAT **TOPICS / IDEAS** ARE YOU  
WANT TO LEARN IN THIS CLASS?

# IMAGE PROCESSING / DIGITAL IMAGING



Not to mention image processing in terms of computer graphics which is another topic.

# TENTATIVE SCHEDULE (2110651)

Week / Date Sec 1,5	Topic	Assignment HW (2 weeks due)
1 -7,9/8	Introduction to image processing and software preparation, Examples of fields using digital image processing	
2 - 14,16/8	Image Acquisition / Sampling and Quantization / Arithmetic Operations	
3 - 21,23/8	Histogram and Spatial Filtering	HW #1 Spatial Domain
4 - 28,30/8	Frequency Domain & Filtering in FD	
5 - 4,6/9	Wavelet and Multiresolution Processing	
6 - 11,13/9	Image Restoration	HW #2 Frequency & Research paper
7- 18,20/9	Image Compression	
8	Midterm exam (1 A4 paper, writing) <b>Sat 27/9/2568 9-12</b>	

9- 2,4/10	Fundamental Image Segmentation I	In-class discussion
10 – 9,11/10	Research discussion / Morphology Feature Extraction and Analysis	<b>Project Assignment</b>
11-16,18/10	Object Recognition (traditional /modern)	
12-25/10	Object Recognition II <b>(23 Oct Holiday, watch video from 25 Oct) - Hybrid</b>	HW #3 Segmentation + Image Augmentation
13-30,1/11	Image Segmentation II	Paper review presentation for project
14-6,8/11	Object <u>Detection</u> : Paper discussion & applications	
15-13,15/11	Image Generative Adversarial and other Imaging Applications – discussion (read and discuss)	
16-22/11	Invited Speaker (20 Nov no class, join a talk on 22 Nov)	
17	- Final exam <b>Sat 6/12/2568 9-12</b> - Project Presentation - <b>Thu 11/12/2568 18 – 21</b>	

## **IMPORTANT DATES (211065I)**

- Midterm – SAT 27/9/2568 @9-12
- Final exam – SAT 6/12/2568 @9-12
- Project presentation – SAT 11/12/2568 @18 – 21

**You cannot take the exam outside the predefined date & time!**

## TEACHING METHODS

- Lecture (+ in-class Lab) : Wednesday 13:00 – 16:00



- Research discussion
- Projects

# 2110651 DIGITAL IMAGING

- **Textbook:**

- Rafael C. Gonzalez and Richard E. Woods, Digital Image Processing, Addison-Wesley

- **Grading Policy**

- In-class Exercises: **5% (0.5 each class + bonus)**
- Homework = **10%**
- Project + presentation : **25%**
- Midterm Exam : **30%**
- Final Exam : **30%**

Score	Grade
$\geq 85$	A
$\geq 75$	B+
$\geq 70$	B
$\geq 65$	C+
$\geq 60$	C
$\geq 55$	D+
$\geq 50$	D
$< 50$	F

## CONTACT ME

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**Email:** [punnarai.s@chula.ac.th](mailto:punnarai.s@chula.ac.th)

**TAs:**



TA Print

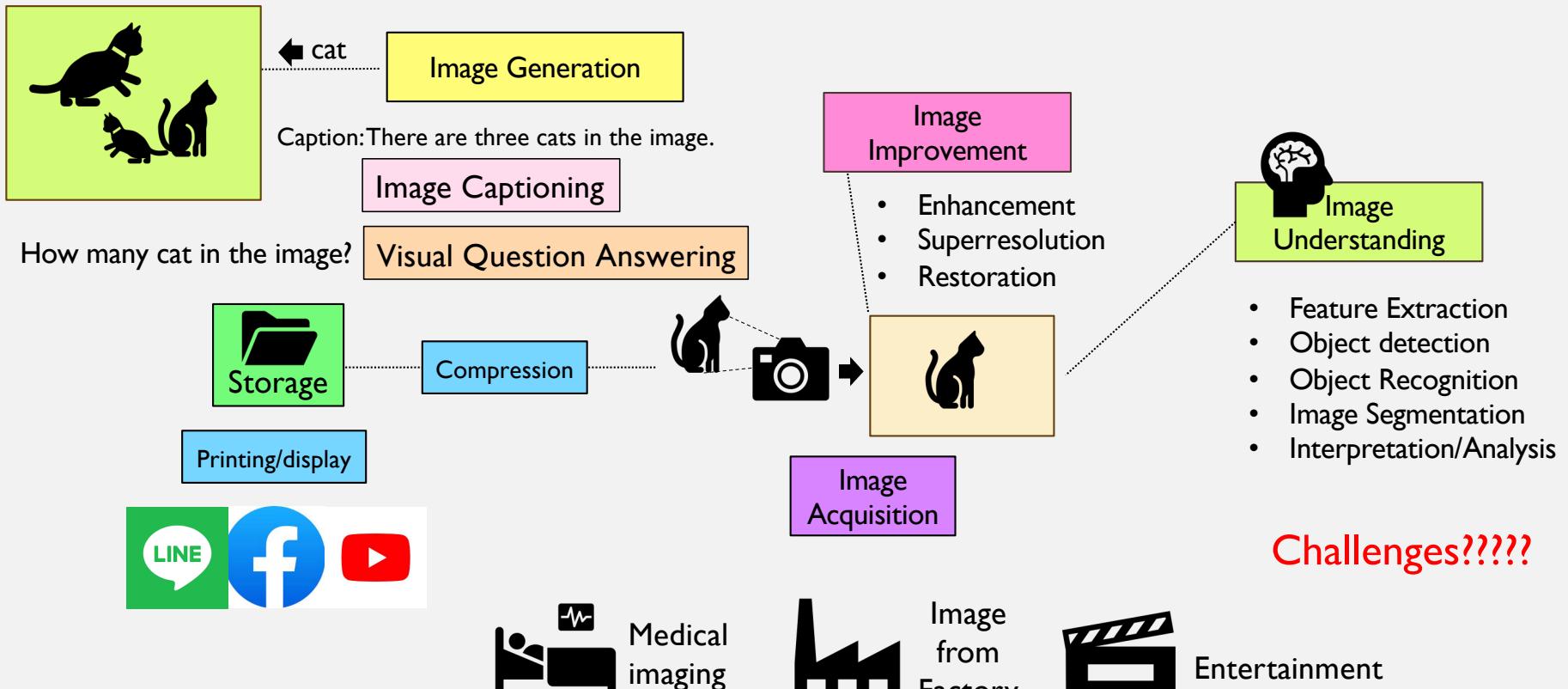


TA Tar

## WHAT IS DIGITAL IMAGING / IMAGE PROCESSING FOR YOU?

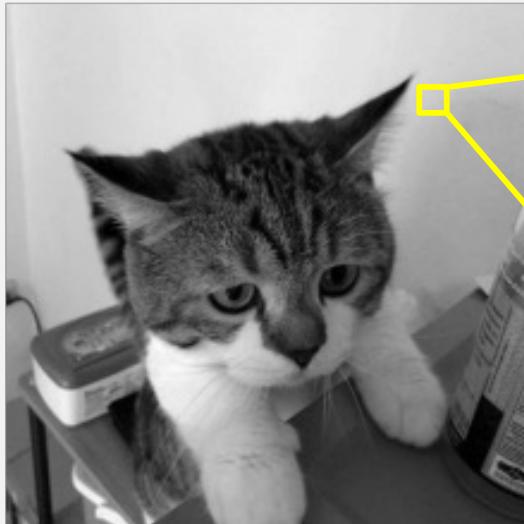
Write a description / draw a picture  
in a paper

# IMAGE PROCESSING / DIGITAL IMAGING



Not to mention image processing in terms of computer graphics which is another topic.

## DIGITAL IMAGE



What we see

170	172	171	174	178	183	185	183	182	181
170	172	171	173	178	182	184	182	182	181
170	171	171	172	177	182	184	182	182	182
170	170	171	172	176	182	183	182	183	182
170	169	171	171	175	182	182	183	183	182
169	167	171	170	175	183	182	183	183	182
168	166	170	169	174	183	181	183	183	182
168	165	170	168	173	183	181	183	182	182
171	168	169	172	176	180	181	184	181	179
168	170	172	170	174	182	183	181	181	179

What a computer sees

An image is denoted by two-dimensional function of the form:  $f(x, y)$   
f – amplitude at the spatial dimension (x,y)

# DIGITAL IMAGE



What we see

	147	140	149	151	157	163	166	164	163	162
169	170	160	173	176	193	195	193	197	181	62
183	185	184	185	189	193	193	191	189	188	31
183	185	184	184	189	192	192	190	189	188	32
183	184	184	183	188	192	192	190	189	189	32
183	183	184	183	187	192	191	190	190	189	32
183	182	184	182	186	192	190	191	190	189	32
182	180	184	181	186	193	190	191	190	189	32
181	179	183	180	185	193	189	191	190	189	32
181	178	183	179	184	193	189	191	189	189	79
184	181	182	183	187	190	189	192	188	186	79
181	183	185	181	185	192	191	189	188	186	

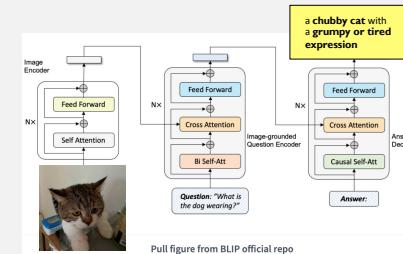
What a computer sees

# EVOLUTION

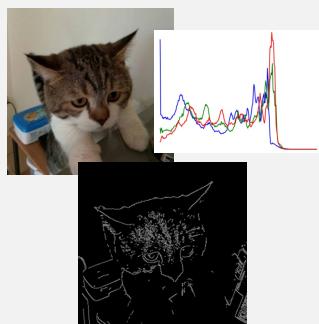
1980s  
Rule-based  
Edge, Threshold,  
Template



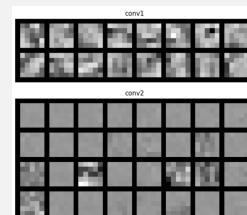
2012+  
Deep Learning  
CNN, ResNet,  
YOLO



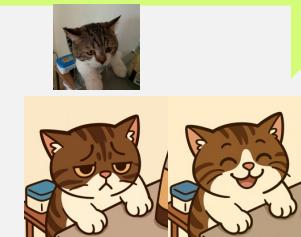
2023+  
Vision + Generation  
+ Action  
DALL·E, SAM,  
Robotics with vision



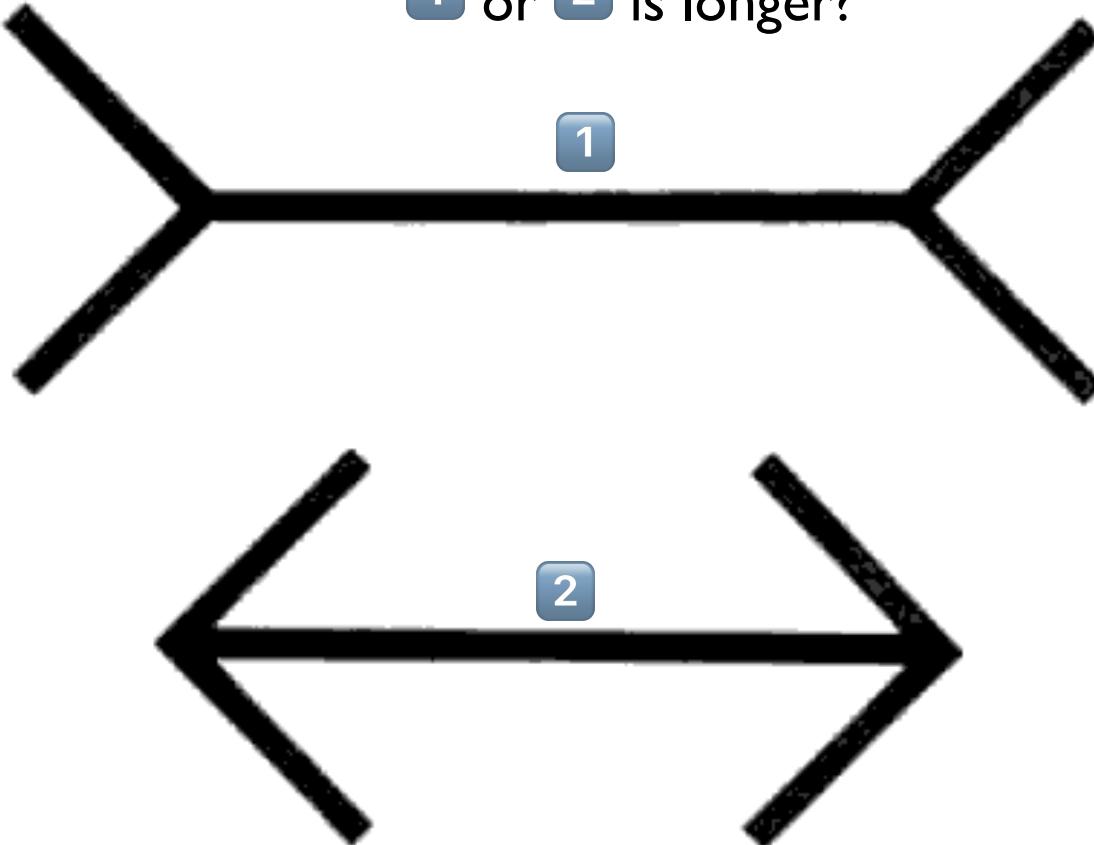
2000s  
Feature-based  
Haar, HOG, SIFT



2020+  
Multimodal /  
Vision + Language  
CLIP, BLIP,  
Flamingo, GPT-4o

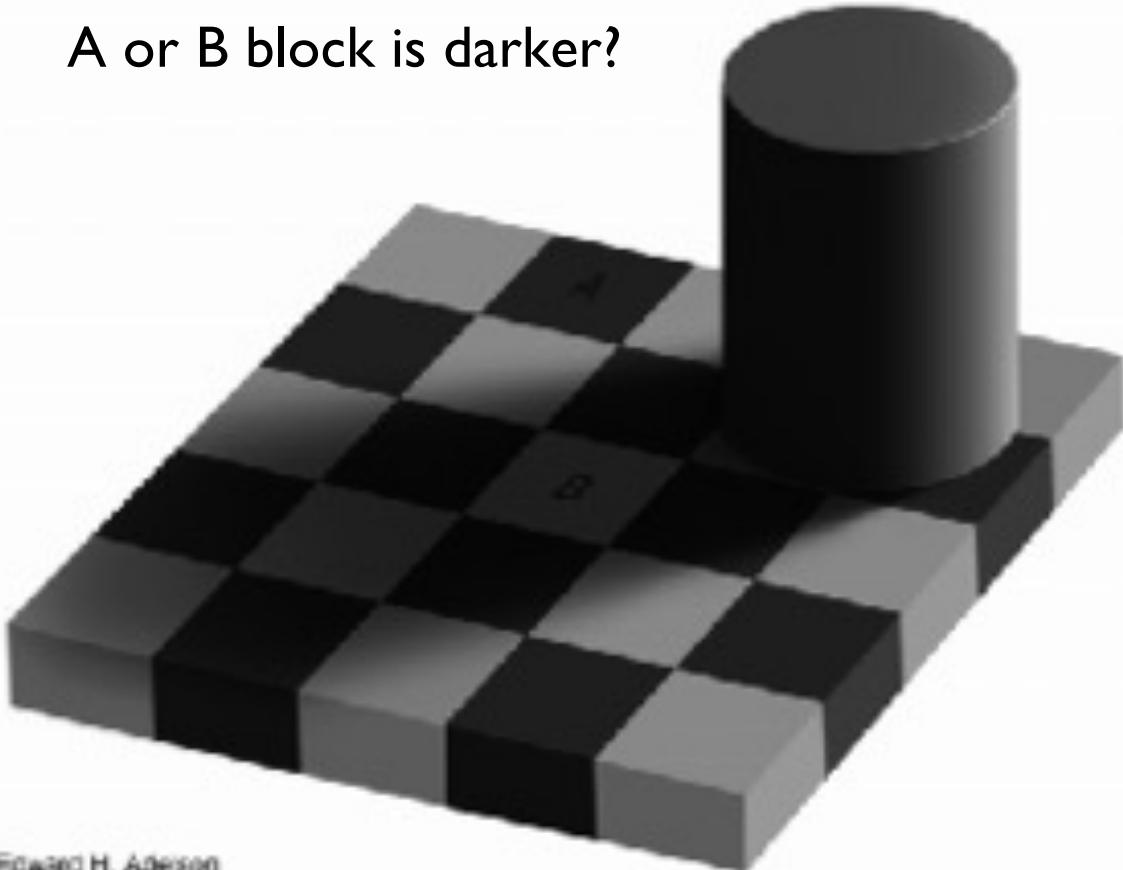


1 or 2 is longer?



Müller-Lyer illusion

A or B block is darker?



Edward H. Adelson

Checker Shadow Illusion by Edward Adelson

How many red X?

X	X	X	X	X	X	X
X	X	X	X	X	X	X
X	X	X	X	X	X	X
X	X	X	X	X	X	X
X	X	X	X	X	X	X
X	X	X	X	X	X	X
X	X	X	X	X	X	X
X	X	X	X	X	X	X
X	X	X	X	X	X	X
X	X	X	X	X	X	X

How many red X?

O	X	O	X	O	X	X
X	O	X	X	X	O	X
O	X	X	O	X	X	O
X	X	O	X	O	O	X
O	X	X	O	X	X	X
X	O	X	X	X	O	X
O	X	X	O	X	X	O
X	O	X	X	X	O	X
X	X	X	O	O	X	X
X	O	X	X	X	O	X

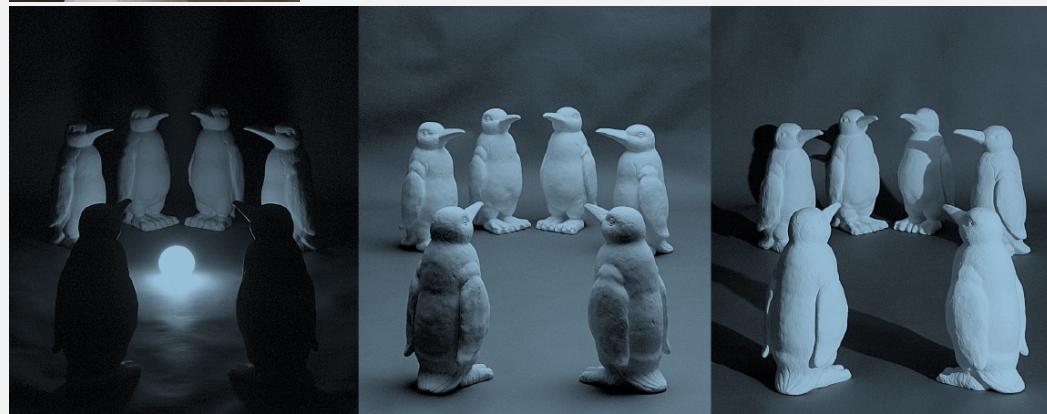
**WHAT ARE CHALLENGES  
FOR COMPUTER TO SEE  
THINGS??**

## CHALLENGES

- Point of View



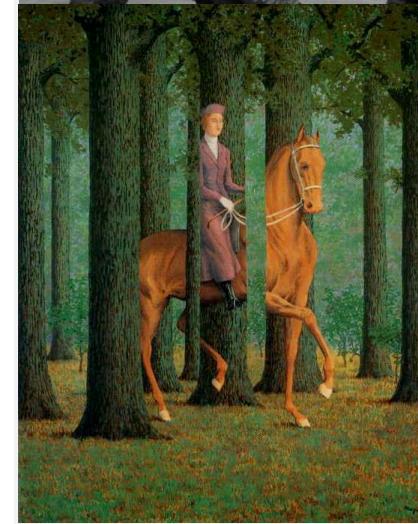
- Illumination



Wikipedia, Academic Gallery, Pinterest  
slide credit: Fei-Fei, Fergus & Torralba

# CHALLENGES

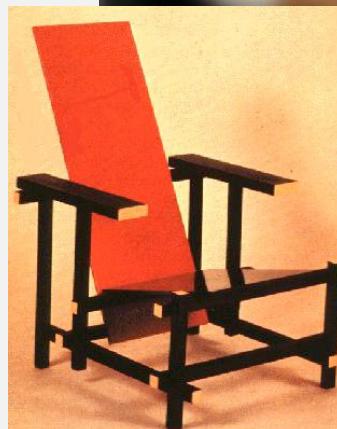
- Scale
- Deformation
- Occlusion
- Background Clutter



slide credit: Fei-Fei, Fergus & Torralba

## CHALLENGES

- Ambiguity
- Motion
- In-class Variations

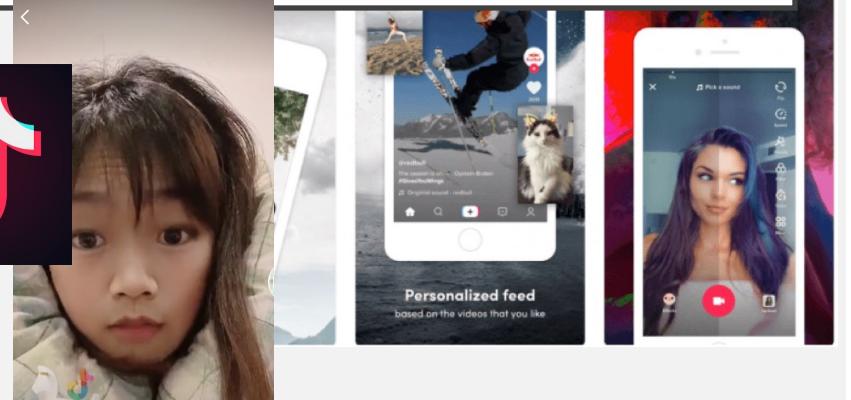


slide credit: Fei-Fei, Fergus & Torralba

HAVE YOU EXPERIENCED IMAGE  
PROCESSING IN YOUR DAILY LIFE?

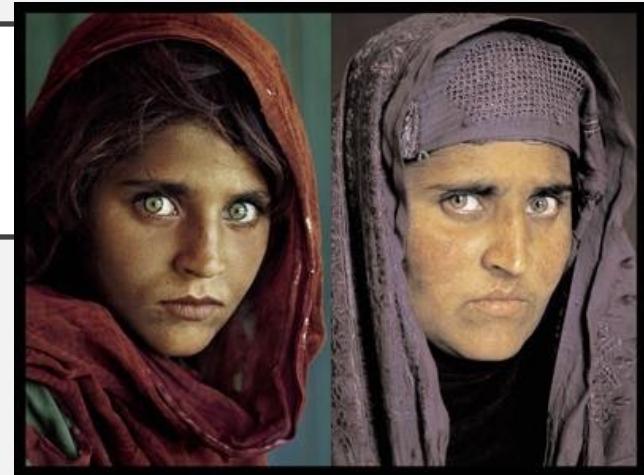
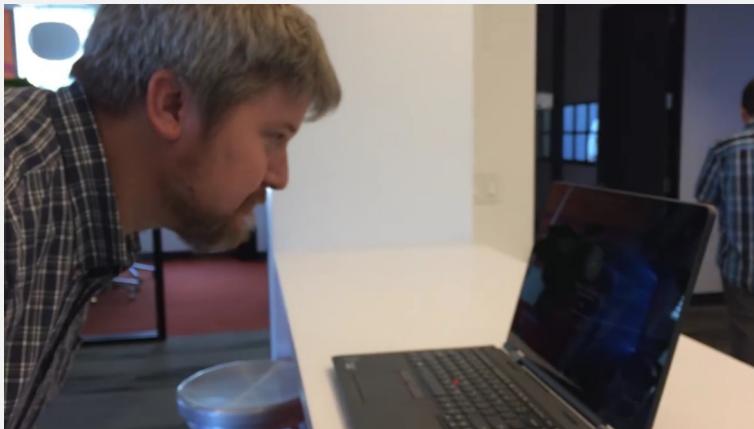
# FACE DETECTION

- Imaging applications in Daily Life
  - Face Detection



# BIOMETRICS APPLICATIONS

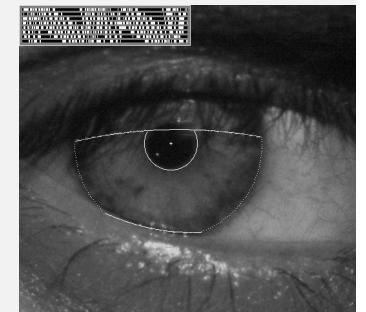
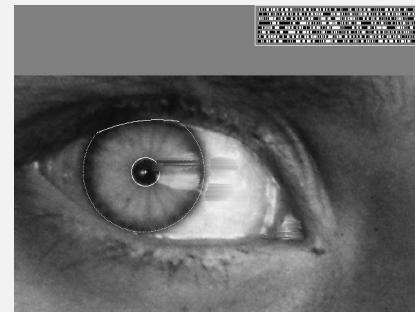
- Imaging applications in Daily Life
  - Biometrics



1984

2002

“How the Afghan Girl was Identified by Her Iris Patterns” <https://www.cl.cam.ac.uk/~jgd1000/afghan.html>



<https://blog.rewatechnology.com/fix-iphone-x-face-id-not-working/>

## HANDS-FREE SELFIE



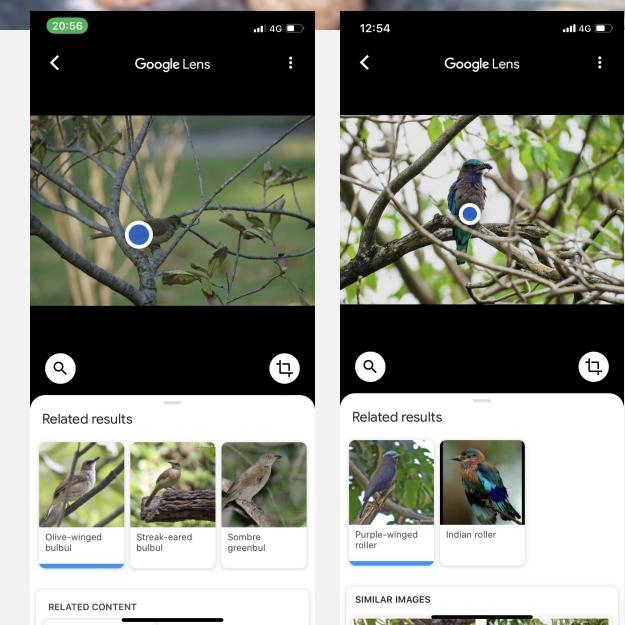
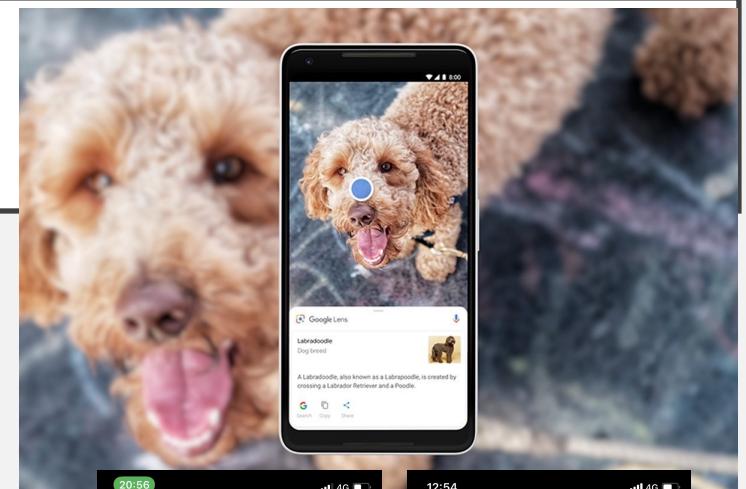
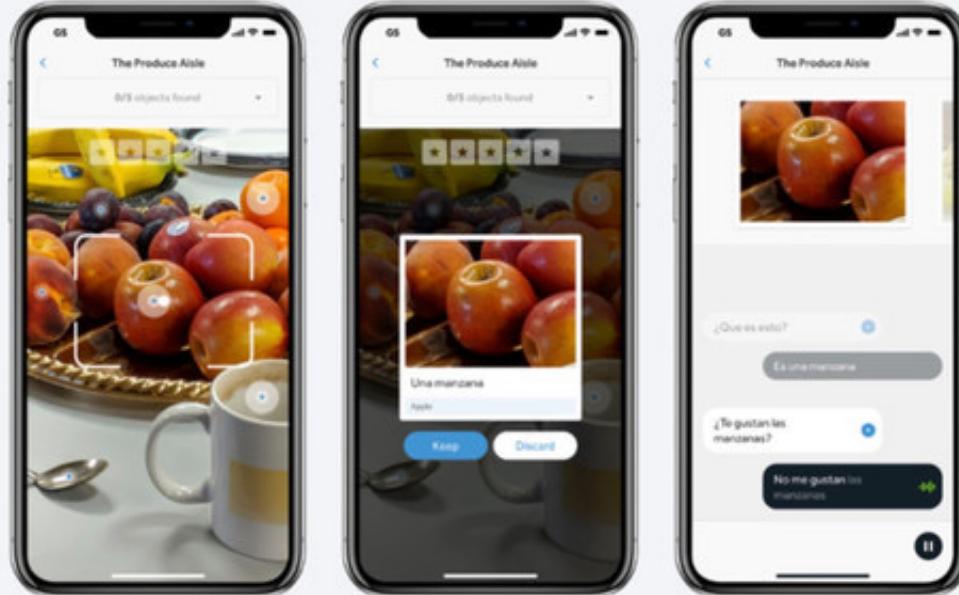
**Take a HANDS-FREE selfie: App snaps photos from up to 10ft away using a simple swipe of the arm**

<https://www.dailymail.co.uk/sciencetech/article-2674946/Take-HANDS-FREE-selfie-App-snaps-photos-10ft-away-using-simple-swipe-arm.html>

Handsome guy photo created by cookie\_studio - [www.freepik.com](http://www.freepik.com)

# OBJECT CLASSIFICATION

- Imaging applications in Daily Life
  - Object Recognition on your mobile phones



# OCR – OPTICAL CHARACTER RECOGNITION



อาจารย์คณะวิศวฯ จุฬาฯ นำเทคโนโลยี AI Deep Tech พัฒนาโปรแกรมสแกนเอกสารและรูปภาพเป็นข้อความ (OCR) awan ภาษาไทยแม่นยำกว่า 90% UTC จุฬาฯ พร้อม spin-off สู่ตลาดในนามบริษัท Eikonnex AI จำกัด

Assoc. Prof. Dr. Thanarat  
Chalidabhonges

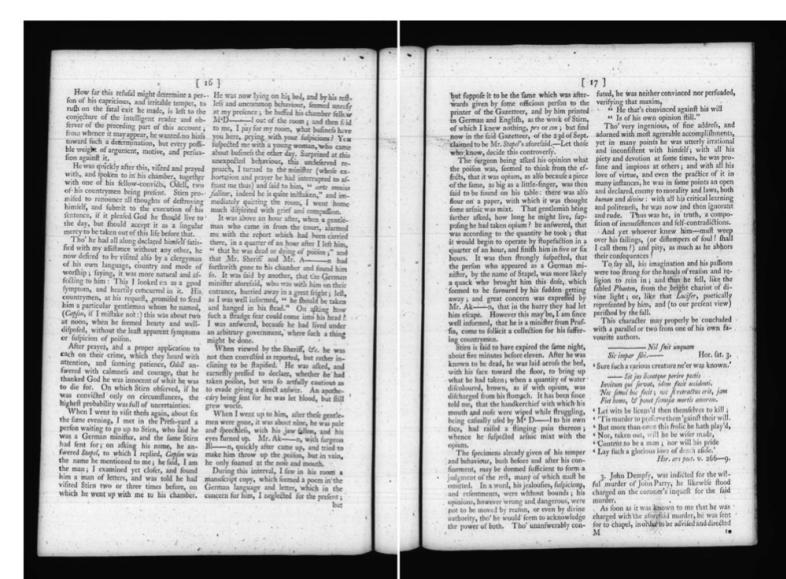


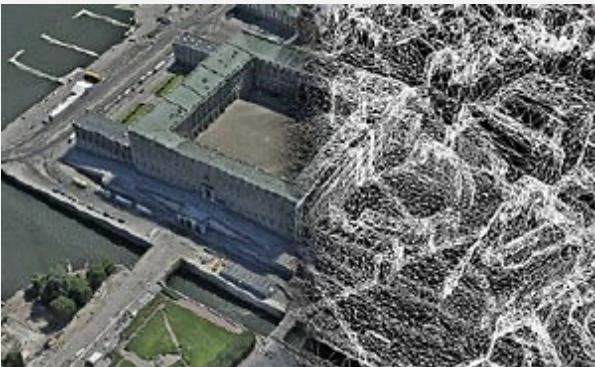
Figure 2: Ordinary's Accounts September 15, 1760 (# 3421-2)

**Character** Left: AWS 73.97%; Azure 35.26%; GCP 76.64%  
**Error Rate** Right: AWS 75.90%; Azure 12.17%; GCP 77.36%

The Old Bailey and OCR: Benchmarking AWS, Azure, and GCP with 180,000 Page Images, William Ughetta, Brian W. Kernighan, Proceedings of the ACM Symposium on Document Engineering 2020



## 3D FROM IMAGES



<https://grail.cs.washington.edu/rome/>  
Building Rome in a Day: Agarwal et al. 2009

# KINECT



<https://news.microsoft.com/en-gb/announcement/xbox-360-kinect-launched/>

Azure Kinect DK

Build for mixed reality using AI sensors

Buy now

Kinect DK ▾ Product overview Features Industries Customer stories FAQ Pricing > Documentation > More ▾

**Kinect with spatial data**

Azure Kinect is a cutting-edge spatial computing developer kit with sophisticated computer vision and speech models, advanced AI sensors, and a range of powerful SDKs that can be connected to Azure cognitive services.

Using Azure Kinect, manufacturing, retail, healthcare, and media enterprises are leveraging spatial data and context to enhance operational safety, increase performance, improve outcomes, and revolutionize the customer experience.

Health Manufacturing Media Retail Robotics

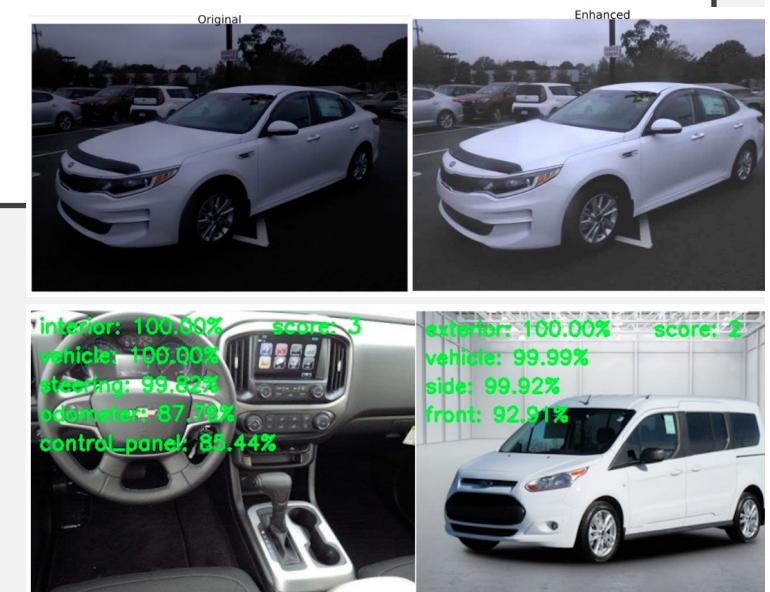
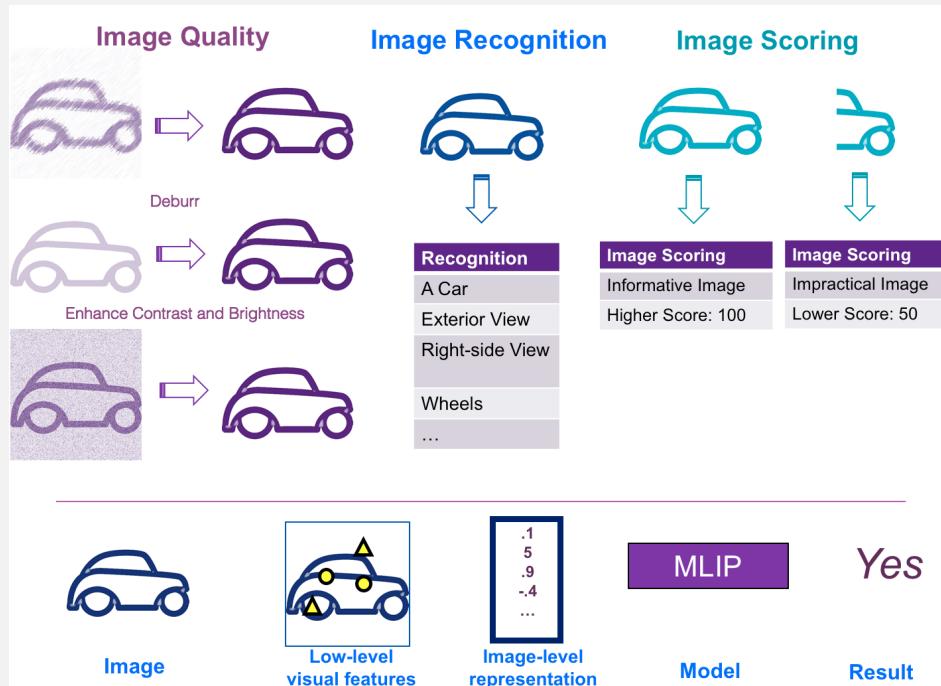
**Release timeline**

2006	PrimeSense technology shown at GDC
2007	
2008	
2009	"Project Natal" announced
2010	Kinect for Xbox 360 released
2011	Non-commercial SDK released
2012	Commercial SDK released
2013	Kinect for Windows released
2014	Kinect for Xbox One released with console
2015	Kinect 2 for Windows released
2016	
2017	Discontinuation of Kinect for Xbox hardware
2018	
2019	Azure Kinect released

- 1-MP depth sensor
- 7-microphone array
- 12-MP RGB video camera
- Accelerometer and gyroscope (IMU) (Orientation & spatial tracking)
- External sync pins

<https://azure.microsoft.com/en-us/services/kinect-dk/#industries>

# DIGITAL MARKETING



How to choose the most representative one among these image candidates needs a strategic image scoring schema

## 'Scoring':

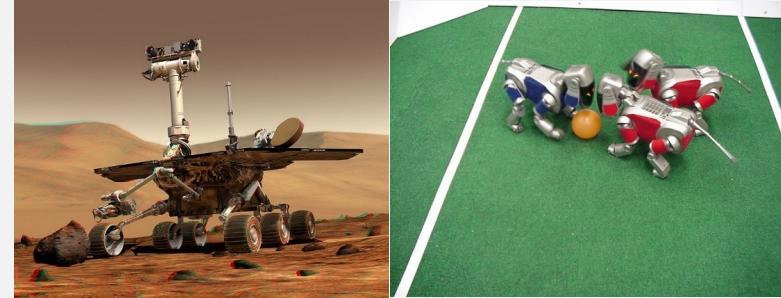
Select a representative image among image candidates.



<https://tech.cars.com/applications-of-machine-learning-image-processing-in-digital-marketing-982ee296dc8a>

## OTHERS

- Other Applications
  - Augmented Reality And Virtual Reality
  - Mobile Robots
  - Industrial Robots
  - Autocars –Tesla / Uber Bought Cmu’s Lab
  - Medical Imaging



NASA's Mars Spirit Rover

[http://en.wikipedia.org/wiki/Spirit\\_rover](http://en.wikipedia.org/wiki/Spirit_rover)

<http://www.robocup.org/>



Vision-guided robots position nut runners on wheels

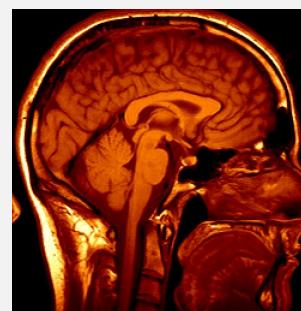
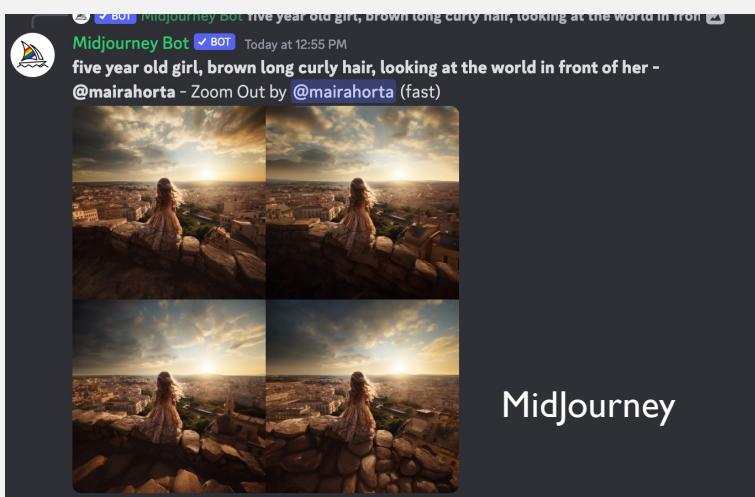


Image guided surgery [Grimson et al., MIT](#)  
3D imaging MRI, CT

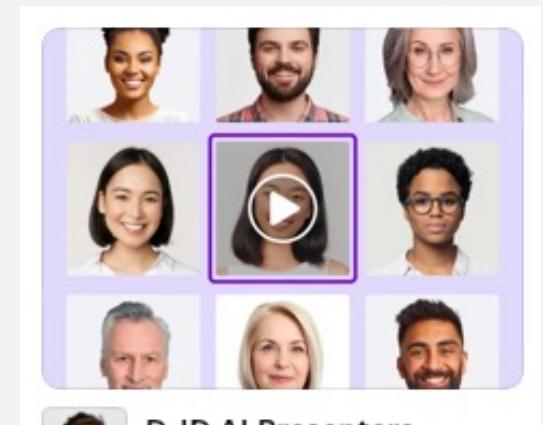
# IMAGE GENERATIVE AI



MidJourney



Canva



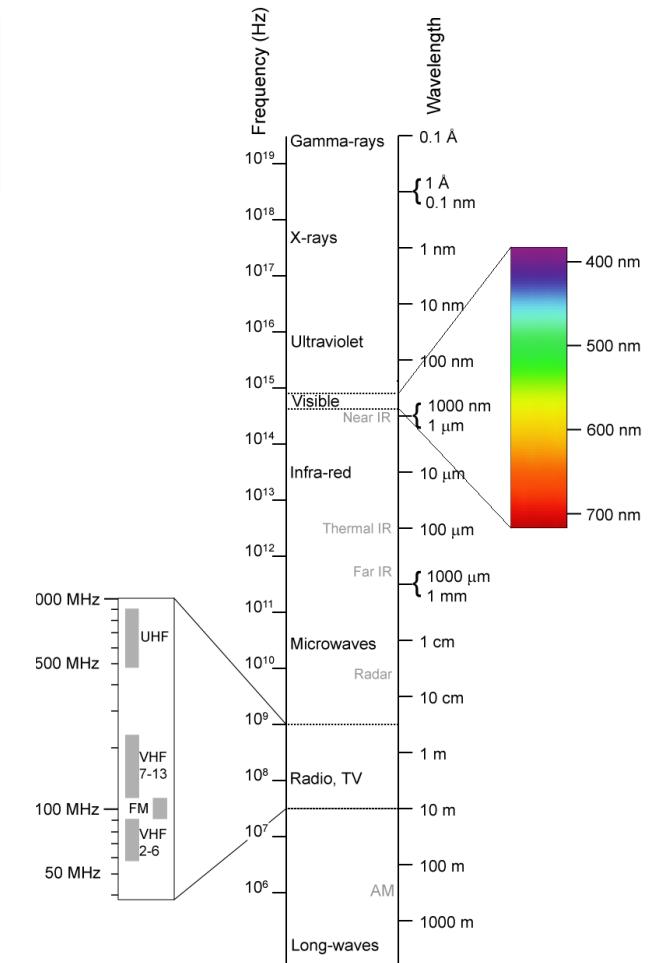
D-ID AI Presenters  
Instantly add a talking head  
video to your designs

Canva/Heygen

## **EXAMPLES OF FIELDS THAT USE DIGITAL IMAGE PROCESSING**

## EXAMPLES OF FIELDS THAT USE DIGITAL IMAGE PROCESSING

- To categorize by the sources of the image
  - Electromagnetic Energy Spectrum → X-ray and visual bands
  - Others: Acoustic, ultrasonic and electronic (electron beams used in microscopy)
  - Synthetic images for modeling and visualization



Images from Gonzalez & Woods, Digital Image Processing, second edition

## GAMMA-RAY IMAGING

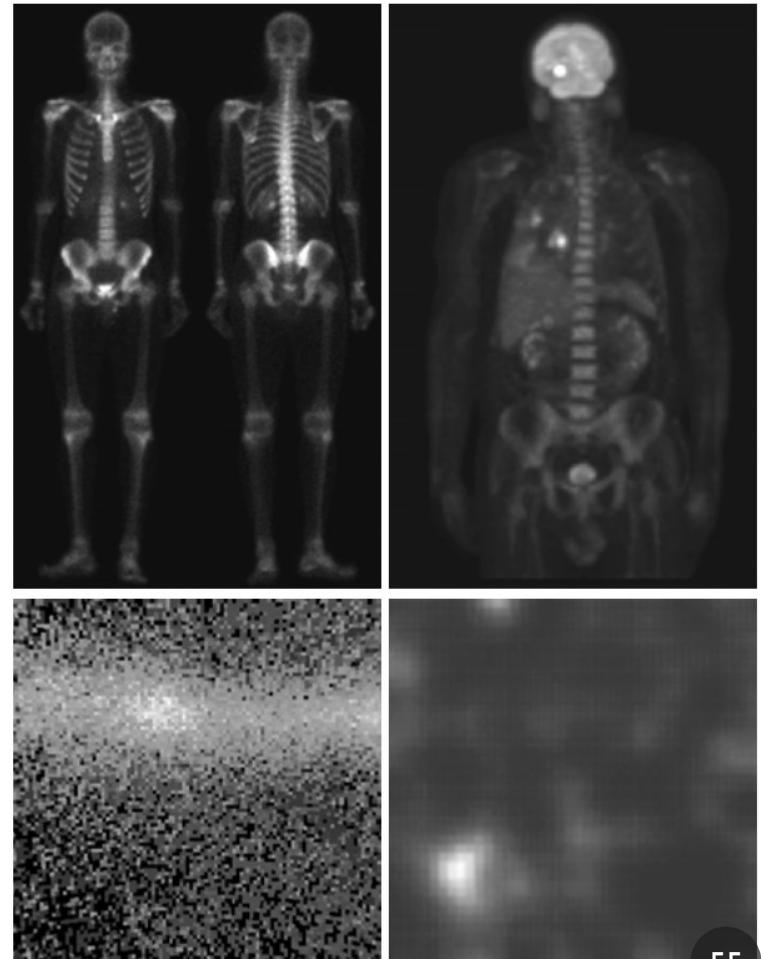
- Nuclear medicine and astronomical observations
  - (a) Locate bone pathology, such as infections, tumors
  - (b) Positron emission tomography (PET) using radioactive isotope
  - (c) Cygnus Loop in gamma ray bands
  - (d) gamma radiation from a valve in a nuclear reactor



PET scan machine from  
<https://www.healthline.com/health/pet-scan>

a  
b  
c  
d

**FIGURE 1.6**  
Examples of gamma-ray imaging. (a) Bone scan. (b) PET image. (c) Cygnus Loop. (d) Gamma radiation (bright spot) from a reactor valve.  
(Images courtesy of (a) G.E. Medical Systems, (b) Dr. Michael E. Casey, CTI PET Systems, (c) NASA, (d) Professors Zhong He and David K. Wehe, University of Michigan.)

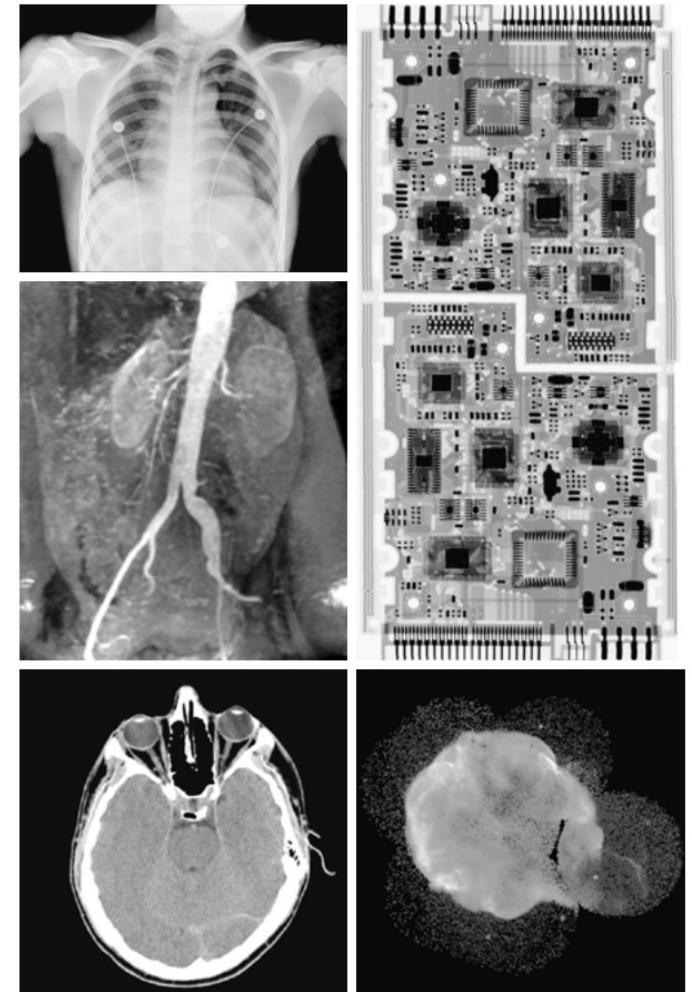
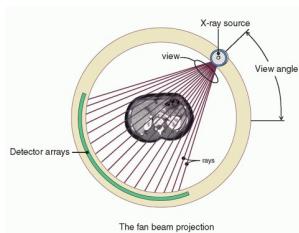


Images from Gonzalez & Woods, Digital Image Processing, second edition

## X-RAY IMAGING

- Medical diagnostics, astronomy, industrial imaging
- X-ray generated by replacing patient between X-ray source and a film (phosphor screen) sensitive to X-ray energy
- Angiography (Contrast enhancement of blood vessels)
- Computerized axial tomography (CAT) – slice taken perpendicularly through the patient -> 3D rendition

Images from Gonzalez & Woods, Digital Image Processing, second edition  
[https://en.wikipedia.org/wiki/CT\\_scan](https://en.wikipedia.org/wiki/CT_scan)  
<https://radiologykey.com/computed-tomography-15/>



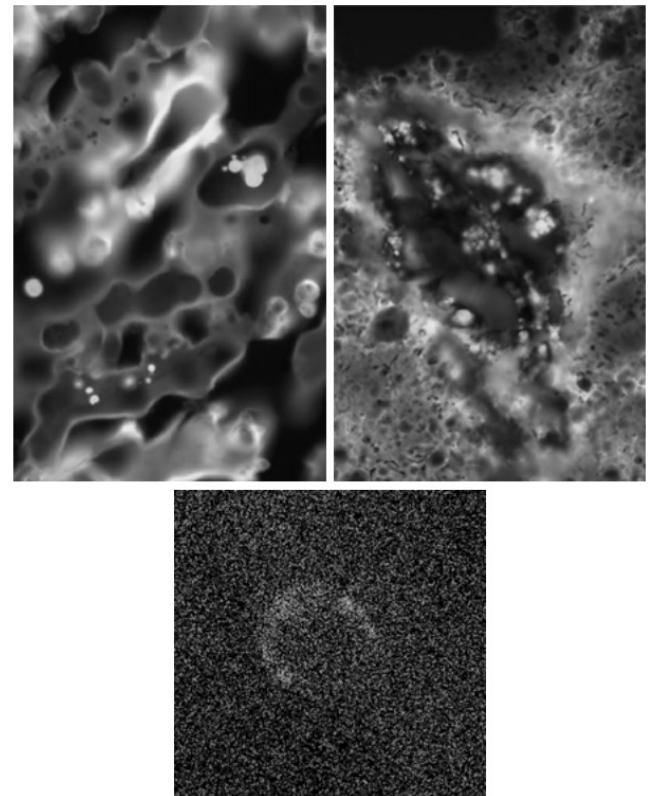
**FIGURE 1.7** Examples of X-ray imaging. (a) Chest X-ray. (b) Aortic angiogram. (c) Head CT. (d) Circuit boards. (e) Cygnus Loop. (Images courtesy of (a) and (c) Dr. David R. Pickens, Dept. of Radiology & Radiological Sciences, Vanderbilt University Medical Center, (b) Dr. Thomas R. Gest, Division of Anatomical Sciences, University of Michigan Medical School, (d) Mr. Joseph E. Pascente, Lixi, Inc., and (e) NASA.)

## IMAGING IN ULTRAVIOLET BAND

- Industrial inspection, microscopy, lasers, biological imaging, astronomical observations
- Fluorescent Microscopy – studying materials, such as corn smut (disease of the corn)
- Cygnus Loop again in Ultraviolet band

a b  
c

**FIGURE 1.8**  
Examples of ultraviolet imaging.  
(a) Normal corn.  
(b) Smut corn.  
(c) Cygnus Loop.  
(Images courtesy of (a) and  
(b) Dr. Michael W. Davidson,  
Florida State University,  
(c) NASA.)

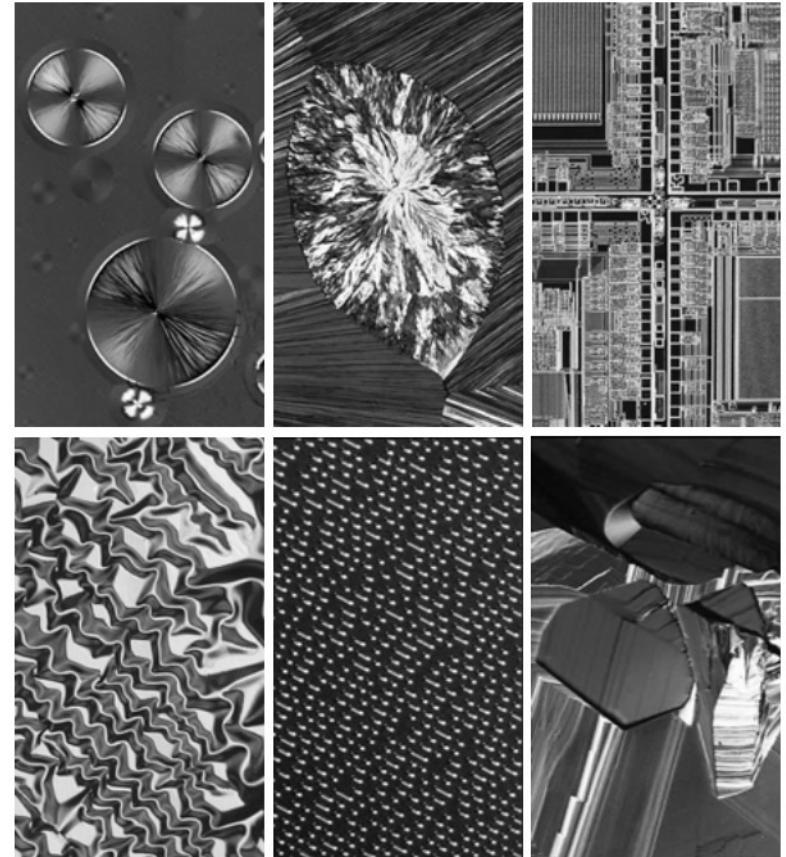


## IMAGING IN VISIBLE AND INFRARED BANDS

- Light microscopy imaging
  - Pharmaceuticals and microinspection to materials characterization



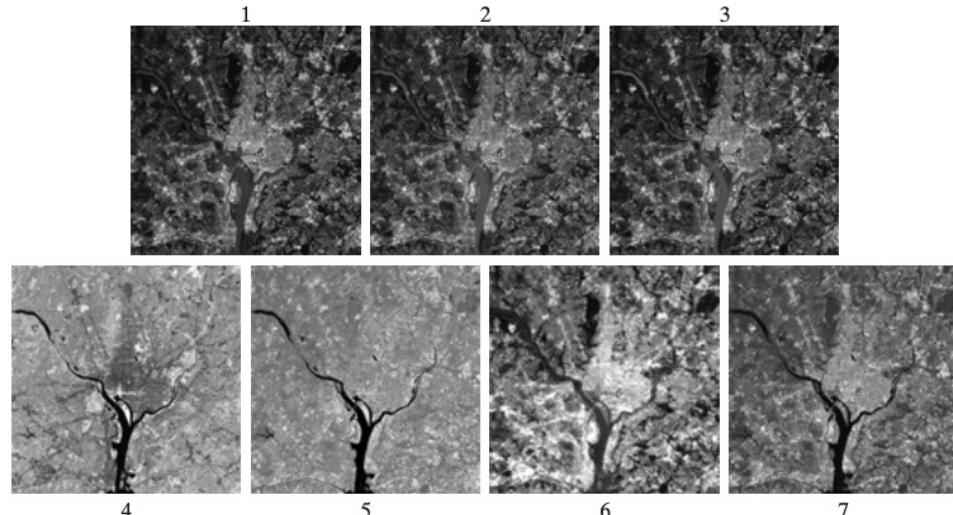
<https://www.carlroth.com>



a  
b  
c  
d  
e  
f

**FIGURE 1.9** Examples of light microscopy images. (a) Taxol (anticancer agent), magnified 250×. (b) Cholesterol—40×. (c) Microprocessor—60×. (d) Nickel oxide thin film—600×. (e) Surface of audio CD—1750×. (f) Organic superconductor—450×. (Images courtesy of Dr. Michael W. Davidson, Florida State University.)

# IMAGING IN VISIBLE AND INFRARED BANDS



**FIGURE 1.10** LANDSAT satellite images of the Washington, D.C. area. The numbers refer to the thematic bands in Table 1.1. (Images courtesy of NASA.)

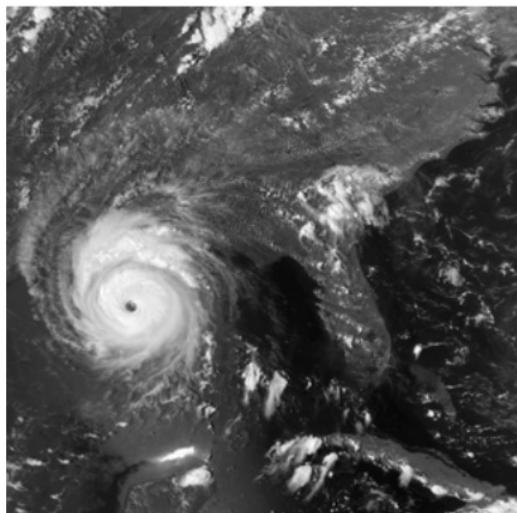
Images from Gonzalez & Woods, Digital Image Processing, second edition

- Remote sensing – Thematic bands in NASA's LANDSAT satellite
  - Multispectrum imaging

Thematic bands of NASA are as LANDSAT satellite.

Band No.	Name	Wavelength ( $\mu\text{m}$ )	Characteristics and use
1	Visible blue	0.45–0.52	Maximum water penetration
2	Visible green	0.52–0.60	Good for measuring plant vigor
3	Visible blue	0.63–0.69	Vegetation discrimination
4	Near infrared	0.76–0.90	Biomass and shoreline mapping
5	Middle infrared	1.55–1.75	Moisture content of soil
6	Thermal infrared	10.4–12.5	Soil moisture, thermal mapping
7	Middle infrared	2.08–2.35	Mineral mapping

## IMAGING IN VISIBLE AND INFRARED BANDS

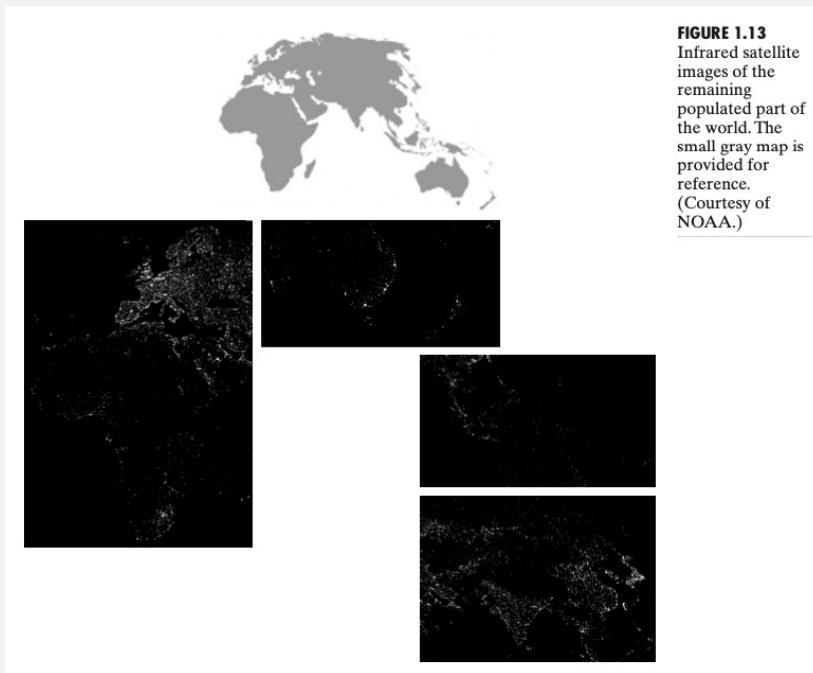


**FIGURE 1.11**  
Multispectral  
image of  
Hurricane  
Andrew taken by  
NOAA GEOS  
(Geostationary  
Environmental  
Operational  
Satellite) sensors.  
(Courtesy of  
NOAA.)

- Weather observation – multispectral imaging
  - Sensors in visible and infrared bands
- Nighttime lights of the world dataset (next page)

Images from Gonzalez & Woods, Digital Image Processing, second edition

# IMAGING IN VISIBLE AND INFRARED BANDS



**FIGURE 1.13**  
Infrared satellite images of the remaining populated part of the world. The small gray map is provided for reference.  
(Courtesy of NOAA.)



**FIGURE 1.12**  
Infrared satellite images of the Americas. The small gray map is provided for reference.  
(Courtesy of NOAA.)

Part of the *Nighttime Lights of the World* dataset

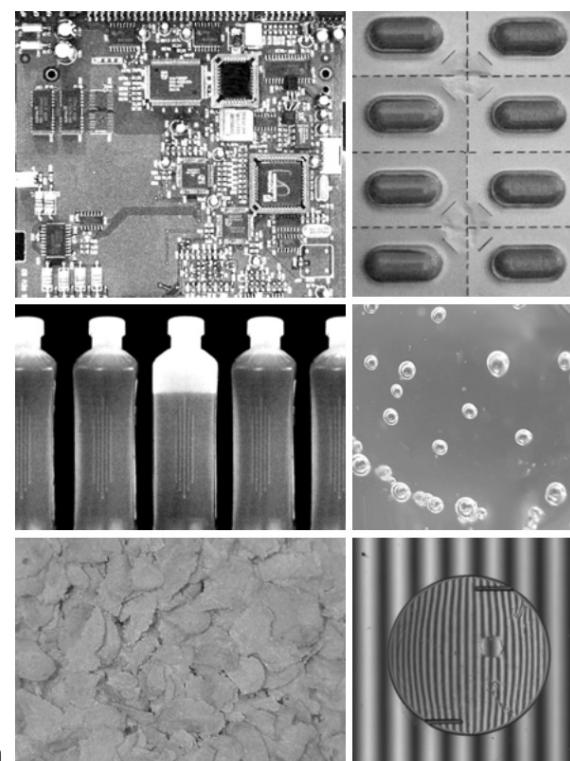
Images from Gonzalez & Woods, Digital Image Processing, second edition

# IMAGING IN VISIBLE AND INFRARED BANDS



**FIGURE 1.15**  
Some additional examples of imaging in the visual spectrum.  
(a) Thumb print.  
(b) Paper currency.  
(c) and  
(d). Automated license plate reading. (Figure  
(a) courtesy of the  
National Institute  
of Standards and  
Technology.  
Figures (c) and  
(d) courtesy of  
Dr. Juan Herrera,  
Perceptics  
Corporation.)

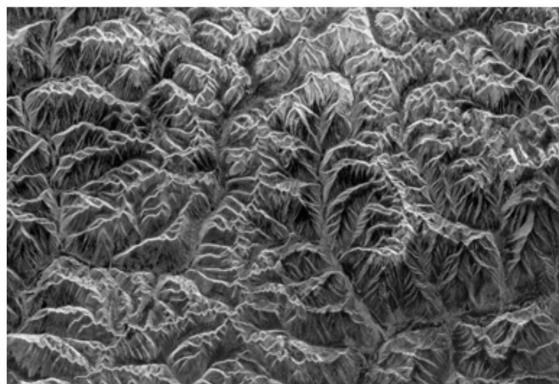
**FIGURE 1.14**  
Some examples of manufactured goods often checked using digital image processing. (a) A circuit board controller.  
(b) Packaged pills.  
(c) Bottles.  
(d) Bubbles in clear-plastic product.  
(e) Cereal.  
(f) Image of intraocular implant.  
(Fig. (f) courtesy of Mr. Pete Sites, Perceptics Corporation.)



Images from Gonzalez & Woods, Digital Image Processing, second edition

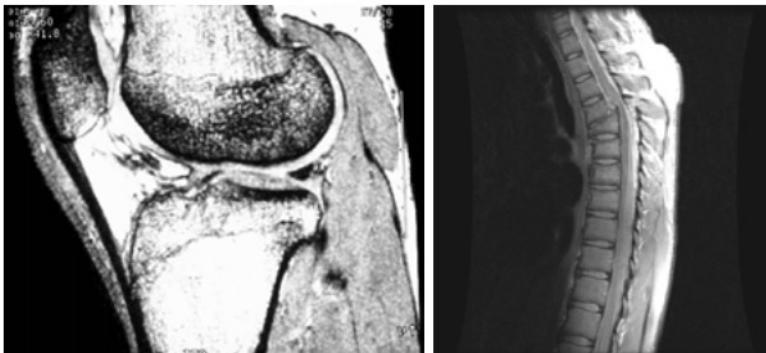
## MICROWAVE BAND

**FIGURE 1.16**  
Spaceborne radar  
image of  
mountains in  
southeast Tibet.  
(Courtesy of  
NASA.)



- Radar – to collect data over virtually any region anytime regardless of weather lighting conditions
- See through vegetation, ice and dry sand
- Explore inaccessible regions of Earth's surface

## IMAGING IN RADIO BAND



a b

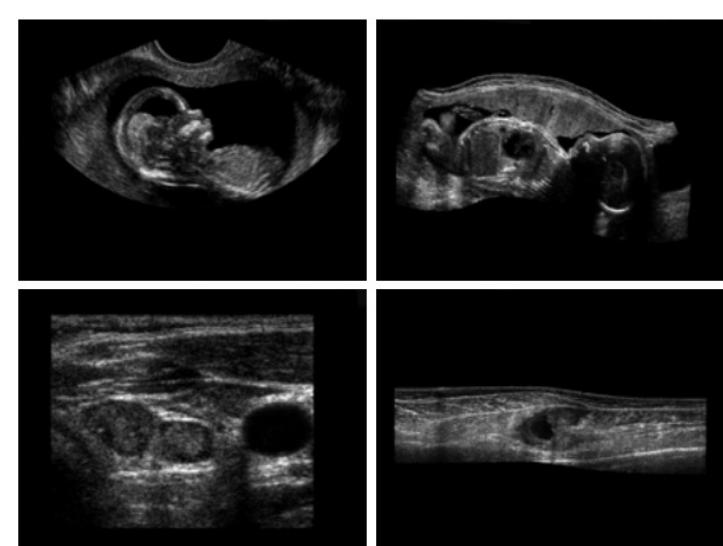
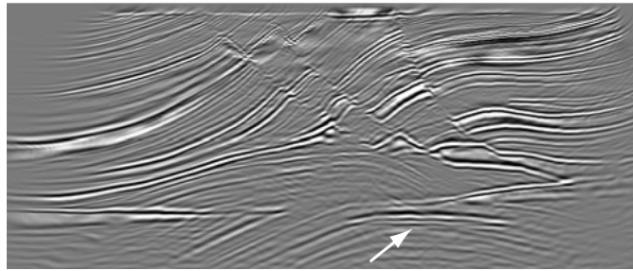
**FIGURE 1.17** MRI images of a human (a) knee, and (b) spine. (Image (a) courtesy of Dr. Thomas R. Gest, Division of Anatomical Sciences, University of Michigan Medical School, and (b) Dr. David R. Pickens, Department of Radiology and Radiological Sciences, Vanderbilt University Medical Center.)

- Magnetic Resonance Imaging (MRI) – placing a patient in a powerful magnet and passes radio waves through his/her body in short pulses

## IMAGING IN OTHER MODALITIES

- Imaging using sound – geological exploration, industry and medicine

**FIGURE 1.19**  
Cross-sectional image of a seismic model. The arrow points to a hydrocarbon (oil and/or gas) trap.  
(Courtesy of Dr. Curtis Ober, Sandia National Laboratories.)

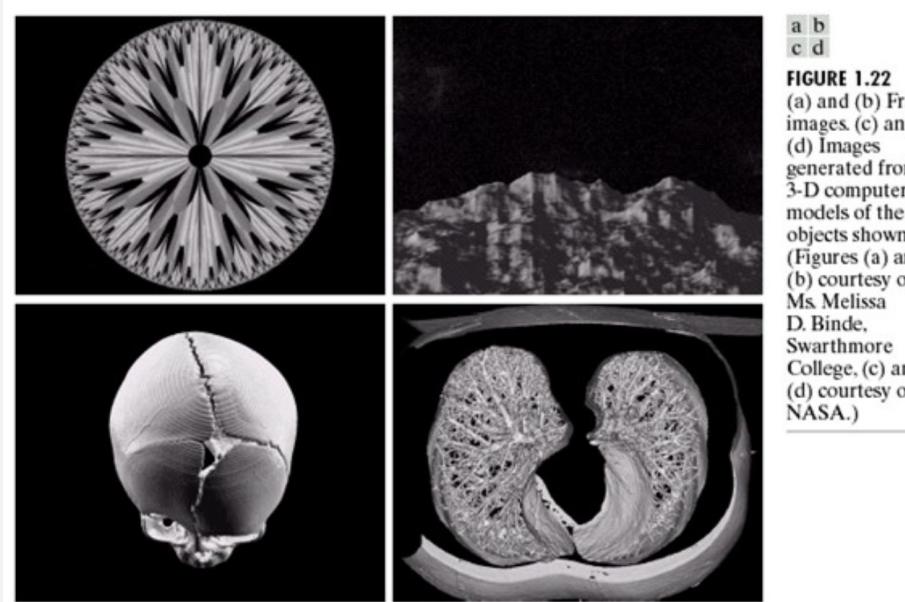
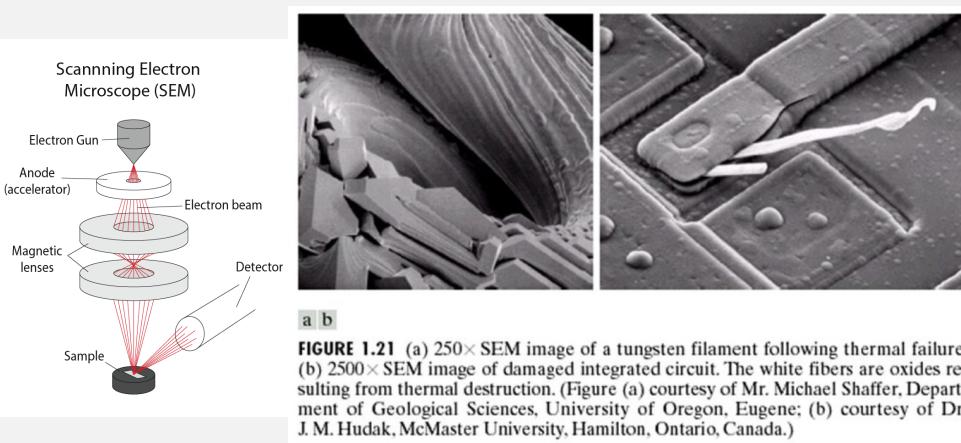


a  
b  
c  
d

**FIGURE 1.20**  
Examples of ultrasound imaging.  
(a) Baby.  
(b) Another view of baby.  
(c) Thyroids.  
(d) Muscle layers showing lesion.  
(Courtesy of Siemens Medical Systems, Inc., Ultrasound Group.)

## OTHER IMAGING MODALITIES

- Electron microscopy
- Computer-generated image



Images from  
- Gonzalez & Woods, Digital Image Processing, second edition  
- <https://www.eng-atoms.msm.cam.ac.uk/RoyalSocDemos/SEM>

## REFERENCES

- Chapter I, Rafael C. Gonzalez and Richard E. Woods, Digital Image Processing, Addison- Wesley

## EXAMPLE #1: READ AND SHOW IMAGE

Warm up!

- Read and show an image
  - **scikit-image** is a collection of algorithms for image processing.

```
from skimage import io
img = io.imread("kitty.jpg")
io.imshow(img)
io.show()

import matplotlib.pyplot as plt
plt.imshow(img)
plt.show()
```

`skimage.io.imread(fname, as_gray=False, plugin=None, **plugin_args)` [so]

Load an image from file.

**Parameters**

- `fname` : string  
Image file name, e.g. `test.jpg` or URL.
- `as_gray` : bool, optional  
If True, convert color images to gray-scale (64-bit floats). Images that are already in gray-scale format are not converted.
- `plugin` : str, optional  
Name of plugin to use. By default, the different plugins are tried (starting with imageio) until a suitable candidate is found. If not given and fname is a tiff file, the tifffile plugin will be used.

**Returns**

- `img_array` : ndarray  
The different color bands/channels are stored in the third dimension, such that a gray-image is MxN, an RGB-image MxNx3 and an RGBA-image MxNx4.

**Other Parameters**

- `plugin_args` : keywords  
Passed to the given plugin.

<https://scikit-image.org/docs/dev/>

# GET PIXEL VALUES

```
from skimage import io
import numpy as np

img = io.imread("kitty.png")

img[0, 0]
array([184, 169, 148], dtype=uint8)

img[0, 0, 0:2]
array([184, 169], dtype=uint8)
```

**NumPy** is the fundamental package for scientific computing in Python. At the core of the **NumPy** package, is the **ndarray** object.  
<http://numpy.org>

**Matplotlib** is a comprehensive library for creating static, animated, and interactive visualizations in Python.  
<https://matplotlib.org>

## matplotlib.pyplot.imshow

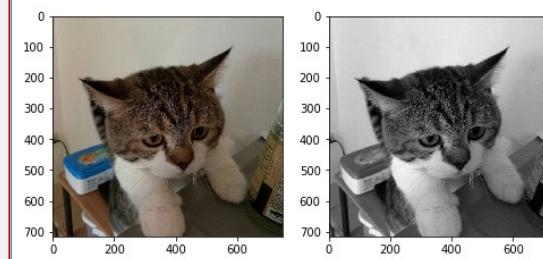
```
matplotlib.pyplot.imshow(X, cmap=None, norm=None, aspect=None, interpolation=None, alpha=None, vmin=None, vmax=None,
origin=None, extent=None, *, filternorm=True, filterrad=4.0, resample=None, url=None, data=None, **kwargs) [source]
```

Display data as an image, i.e., on a 2D regular raster.

## EXERCISE #1 READ AND SHOW IMAGE

- Compared RGB image and gray image

```
from skimage import io, color
import matplotlib.pyplot as plt
img = io.imread("kitty.png")
gray = color.rgb2gray(                )  
  
fig = plt.figure(figsize=(8, 4))
fig.add_subplot(1, 2, 1)
plt.imshow(img)
fig.add_subplot(1, 2, 2)
plt.imshow(                )
io.show()
```



In [9]:

### rgb2gray

`skimage.color.rgb2gray(rgb, *, channel_axis=-1)`[\[source\]](#)

#### Parameters

`rgb : (..., 3, ...) array_like`

The image in RGB format. By default, the final dimension denotes channels.

#### Returns

`out : ndarray`

The luminance image - an array which is the same size as the input array, but with the channel dimension removed.

#### Raises

`ValueError`

If `rgb` is not at least 2-D with shape (... , 3, ...).

<https://scikit-image.org/docs/dev/>

## QUESTIONS

- What is your image size?
- What is your image data type?
- What is value at kitty image at [50, 50, 2]?
- Which one is Red/Green/blue?
- What is the maximum value of each RGB and what color is that?
- What is the minimum value of each RGB and what color is that?

## EXERCISE #2 READ AND SHOW IMAGE USING CV2

- Read an image using opencv

```
import cv2  
image2 = cv2.imread("kitty.jpg")  
  
plt.imshow(image2)  
plt.show()
```

- What do you see?
- Add this code:

```
image2 = cv2.cvtColor(image2, cv2.COLOR_BGR2RGB)
```

## EXAMPLE #3 DICOM FILE

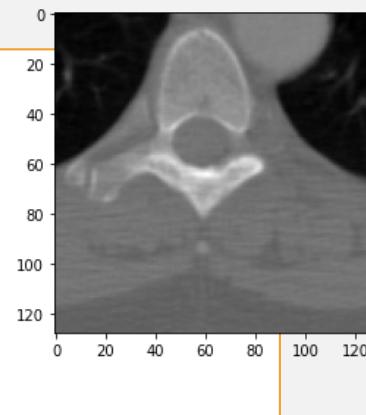
- Digital Imaging and Communications in Medicine (DICOM) format
  - Image and patient's data

```
import matplotlib.pyplot as plt
from pydicom import dcmread
from pydicom.data import get_testdata_file

fpath = get_testdata_file('CT_small.dcm')
ds = dcmread(fpath)

print(f"Patient's Name....: {ds.PatientName}")
print(f"Patient ID.....: {ds.PatientID}")
print(f"Modality.....: {ds.Modality}")
print(f"Study Date.....: {ds.StudyDate}")
print(f"Image size.....: {ds.Rows} x {ds.Columns}")
print(f"Pixel Spacing....: {ds.PixelSpacing}")

# plot the image using matplotlib
plt.imshow(ds.pixel_array, cmap=plt.cm.gray)
plt.show()
```



[https://pydicom.github.io/pydicom/stable/auto\\_examples/index.html](https://pydicom.github.io/pydicom/stable/auto_examples/index.html)

- What is the image size?
- What is the datatype of pixel\_array ?
- Min/max?
- Try with “MR\_small.dcm”

## Python For Data Science Cheat Sheet

### NumPy Basics

Learn Python for Data Science interactively at [www.DataCamp.com](http://www.DataCamp.com)



### NumPy

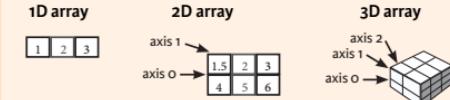
The NumPy library is the core library for scientific computing in Python. It provides a high-performance multidimensional array object, and tools for working with these arrays.

Use the following import convention:

```
>>> import numpy as np
```



### NumPy Arrays



### Creating Arrays

```
>>> a = np.array([1,2,3])
>>> b = np.array([(1,5,2,3), (4,5,6)], dtype = float)
>>> c = np.array([(1,5,2,3), (4,5,6)], [(3,2,1), (4,5,6)]),
      dtype = float)
```

### Initial Placeholders

```
>>> np.zeros((3,4))
>>> np.ones((2,3),dtype=np.int16)
>>> d = np.arange(10,25)
>>> np.linspace(0,2,9)
>>> e = np.full((2,2),7)
>>> f = np.eye(2)
>>> np.random.random((2,2))
>>> np.empty((3,2))
```

Create an array of zeros  
Create an array of ones  
Create an array of evenly spaced values (step value)  
Create an array of evenly spaced values (number of samples)  
Create a constant array  
Create a 2x2 identity matrix  
Create an array with random values  
Create an empty array

### I/O

#### Saving & Loading On Disk

```
>>> np.save('my_array', a)
>>> np.savetxt('array.npz', a, b)
>>> np.load('my_array.npy')
```

#### Saving & Loading Text Files

```
>>> np.loadtxt("myfile.txt")
>>> np.genfromtxt("myfile.csv", delimiter=',')
>>> np.savetxt("myarray.txt", a, delimiter=" ")
```

### Data Types

```
>>> np.int64
>>> np.float32
>>> np.complex
>>> np.bool_
>>> np.object_
>>> np.string_
>>> np.unicode_
```

Signed 64-bit integer types  
Standard double-precision floating point  
Complex numbers represented by 128 floats  
Boolean type storing TRUE and FALSE values  
Python object type  
Fixed-length string type  
Fixed-length unicode type

### Inspecting Your Array

<code>&gt;&gt;&gt; a.shape</code>	Array dimensions
<code>&gt;&gt;&gt; len(a)</code>	Length of array
<code>&gt;&gt;&gt; b.ndim</code>	Number of array dimensions
<code>&gt;&gt;&gt; e.size</code>	Number of array elements
<code>&gt;&gt;&gt; b.dtype</code>	Data type of array elements
<code>&gt;&gt;&gt; b.dtype.name</code>	Name of data type
<code>&gt;&gt;&gt; b.astype(int)</code>	Convert an array to a different type

### Asking For Help

```
>>> np.info(np.ndarray.dtype)
```

### Array Mathematics

#### Arithmetic Operations

<code>&gt;&gt;&gt; g = a - b</code>	Subtraction
<code>&gt;&gt;&gt; np.subtract(a,b)</code>	Subtraction
<code>&gt;&gt;&gt; b + a</code>	Addition
<code>&gt;&gt;&gt; np.add(b,a)</code>	Addition
<code>&gt;&gt;&gt; a / b</code>	Division
<code>&gt;&gt;&gt; np.divide(a,b)</code>	Division
<code>&gt;&gt;&gt; a * b</code>	Multiplication
<code>&gt;&gt;&gt; np.multiply(a,b)</code>	Multiplication
<code>&gt;&gt;&gt; np.exp(b)</code>	Exponentiation
<code>&gt;&gt;&gt; np.sqrt(b)</code>	Square root
<code>&gt;&gt;&gt; np.sin(a)</code>	Print sines of an array
<code>&gt;&gt;&gt; np.cos(b)</code>	Element-wise cosine
<code>&gt;&gt;&gt; np.log(a)</code>	Element-wise natural logarithm
<code>&gt;&gt;&gt; e.dot(f)</code>	Dot product

#### Comparison

<code>&gt;&gt;&gt; a == b</code>	Element-wise comparison
<code>&gt;&gt;&gt; array([[False, True, True], [True, False, False]], dtype=bool)</code>	Element-wise comparison
<code>&gt;&gt;&gt; a &lt; 2</code>	Element-wise comparison
<code>&gt;&gt;&gt; array([[True, False, False]], dtype=bool)</code>	Element-wise comparison
<code>&gt;&gt;&gt; np.array_equal(a, b)</code>	Array-wise comparison

#### Aggregate Functions

<code>&gt;&gt;&gt; a.sum()</code>	Array-wise sum
<code>&gt;&gt;&gt; a.min()</code>	Array-wise minimum value
<code>&gt;&gt;&gt; b.max(axis=0)</code>	Maximum value of an array row
<code>&gt;&gt;&gt; c.cumsum(axis=1)</code>	Cumulative sum of the elements
<code>&gt;&gt;&gt; a.mean()</code>	Mean
<code>&gt;&gt;&gt; b.median()</code>	Median
<code>&gt;&gt;&gt; a.corrcoef()</code>	Correlation coefficient
<code>&gt;&gt;&gt; np.std(b)</code>	Standard deviation

### Copying Arrays

<code>&gt;&gt;&gt; h = a.view()</code>	Create a view of the array with the same data
<code>&gt;&gt;&gt; np.copy(a)</code>	Create a copy of the array
<code>&gt;&gt;&gt; h = a.copy()</code>	Create a deep copy of the array

### Sorting Arrays

<code>&gt;&gt;&gt; a.sort()</code>	Sort an array
<code>&gt;&gt;&gt; c.sort(axis=0)</code>	Sort the elements of an array's axis

### Subsetting, Slicing, Indexing

Also see Lists

#### Subsetting

<code>&gt;&gt;&gt; a[2]</code>	1 2 3 3	Select the element at the 2nd index
<code>&gt;&gt;&gt; b[1,2]</code>	1 2 3 4 5 6 6 0	Select the element at row 1 column 2 (equivalent to <code>b[1][2]</code> )

#### Slicing

<code>&gt;&gt;&gt; a[0:2]</code>	1 2 3 1 2 3	Select items at index 0 and 1
<code>&gt;&gt;&gt; b[0:2,1]</code>	1 2 3 4 5 6 4 5 6	Select items at rows 0 and 1 in column 1

#### Reversed

<code>&gt;&gt;&gt; a[::-1]</code>	1 2 3 1 2 3	Reversed array a
-----------------------------------	----------------	------------------

#### Boolean Indexing

<code>&gt;&gt;&gt; a[a&lt;2]</code>	1 2 3 1	Select elements from a less than 2
<code>&gt;&gt;&gt; b[[1, 0, 1, 0], [0, 1, 2, 0]]</code>	1 2 3 4 5 6 4 5 6 1 2 3 1 2 3	Select elements (1,0),(0,1),(1,2) and (0,0)
<code>&gt;&gt;&gt; b[[1, 0, 1, 0]][:, [0, 1, 2, 0]]</code>	1 2 3 1 2 3 1 2 3 1 2 3	Select a subset of the matrix's rows and columns

### Array Manipulation

#### Transposing Array

<code>&gt;&gt;&gt; i = np.transpose(b)</code>	i.T	Permute array dimensions
---	-----	--------------------------

#### Changing Array Shape

<code>&gt;&gt;&gt; b.ravel()</code>	g.reshape(3,-2)	Flatten the array
-------------------------------------	-----------------	-------------------

#### Adding/Removing Elements

<code>&gt;&gt;&gt; h.resize((2, 6))</code>	h.append(g)	Return a new array with shape (2,6)
<code>&gt;&gt;&gt; np.append(h,g)</code>	h.insert(a, 1, 5)	Append items to an array
<code>&gt;&gt;&gt; np.insert(a, 1, 5)</code>	h.delete(a,[1])	Insert items in an array
<code>&gt;&gt;&gt; np.delete(a,[1])</code>		Delete items from an array

#### Combining Arrays

<code>&gt;&gt;&gt; np.concatenate((a,d),axis=0)</code>	a[[ 1,  2,  3, 10, 15, 20]]	Concatenate arrays
<code>&gt;&gt;&gt; np.vstack((a,b))</code>	array([[ 1,  2,  3, 10, 15, 20], [ 1,  2,  3, 10, 15, 20]])	Stack arrays vertically (row-wise)

<code>&gt;&gt;&gt; np.r_[e,f]</code>	array([[ 1,  2,  3, 10, 15, 20], [ 1,  2,  3, 10, 15, 20]])	Stack arrays vertically (row-wise)
<code>&gt;&gt;&gt; np.hstack((e,f))</code>	array([[ 1,  2,  3, 10, 15, 20], [ 1,  2,  3, 10, 15, 20]])	Stack arrays horizontally (column-wise)

<code>&gt;&gt;&gt; np.column_stack((a,d))</code>	array([[ 1,  2,  3, 10, 15, 20], [ 1,  2,  3, 10, 15, 20]])	Create stacked column-wise arrays
--	---	-----------------------------------

<code>&gt;&gt;&gt; np.c_[a,d]</code>	array([[ 1,  2,  3, 10, 15, 20], [ 1,  2,  3, 10, 15, 20]])	Create stacked column-wise arrays
--------------------------------------	---	-----------------------------------

<code>&gt;&gt;&gt; np.hsplit(a,3)</code>	[array([1, 2, 3]), array([4, 5, 6]), array([7, 8, 9])]	Split the array horizontally at the 3rd index
<code>&gt;&gt;&gt; np.vsplit(c,2)</code>	[array([[ 1,  2,  3, 10, 15, 20], [ 1,  2,  3, 10, 15, 20]]), array([[ 3,  4,  5,  6,  7,  8], [ 3,  4,  5,  6,  7,  8]])]	Split the array vertically at the 2nd index



# NUMPY

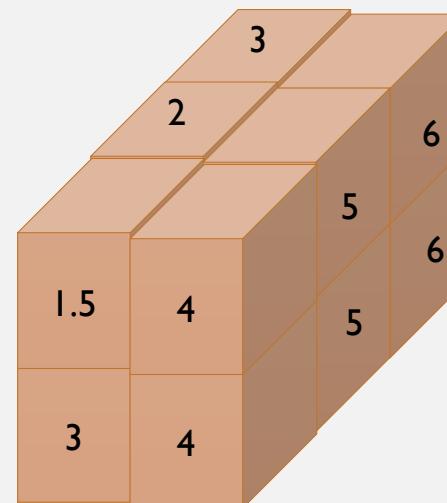
Warm up!

```
c = np.array([[[(1.5,2,3), (4,5,6)], [(3,2,1), (4,5,6)]], dtype = float)
```

```
array([[[ 1.5,  2.,  3.],
       [ 4.,  5.,  6.]],
      [[ 3.,  2.,  1.],
       [ 4.,  5.,  6.]]])
```

```
c[0,0]
Out[31]: array([ 1.5,  2.,  3.])
```

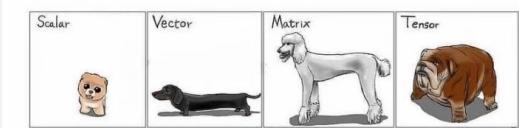
```
c[0,1]
Out[32]: array([ 4.,  5.,  6.])
```



Scalar   Vector   Matrix   Tensor

Scalar	Vector	Matrix	Tensor	Tensor
--------	--------	--------	--------	--------

Difference Among Scalar, Vector, Matrix, Tensor



## EXERCISE #4

- Create 150 x 150 black image

```
import numpy as np
import matplotlib.pyplot as plt

a = np.zeros((150,150),dtype=int)
plt.imshow(a,cmap='gray')
plt.show()
```

- Create an image of 100x100 which has red color on the half left and cyan color on the half right and display the image