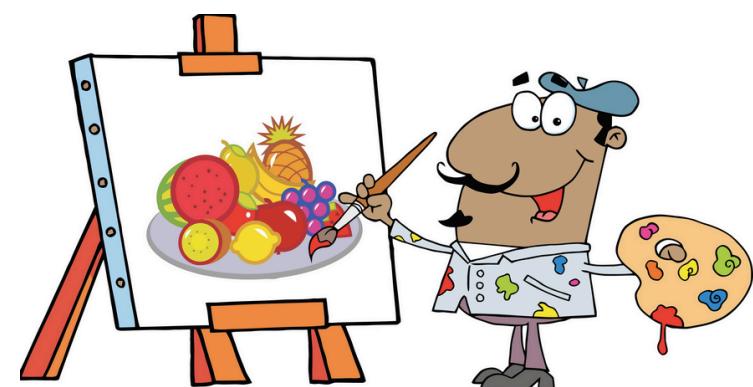
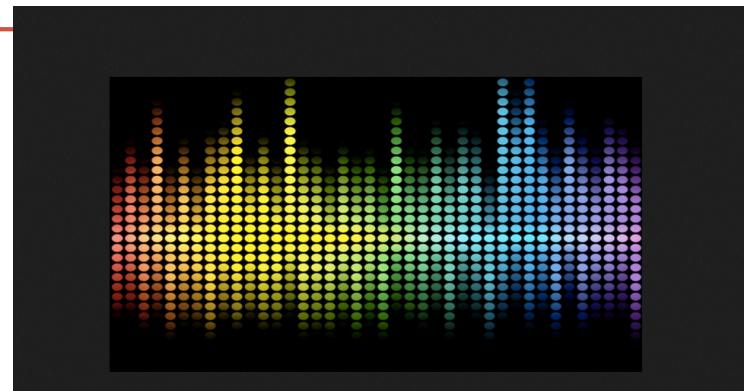


# DIGITAL IMAGE PROCESSING

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Lecture 1  
Introduction

Tammy Riklin Raviv  
Electrical and Computer Engineering  
Ben-Gurion University of the Negev



# Introduction to Digital Image Processing

- Lecturer: Dr. Tammy Riklin Raviv
- Teaching assistants: Assaf Arbelle and Boris Kodner
- No.: 361-1-4751
- Time: Wednesday 14:00-17:00
- Location: Building 28 Room 204
- Prerequisites: Digital Signal Processing

Introduction to Stochastic Processes

- Course website:

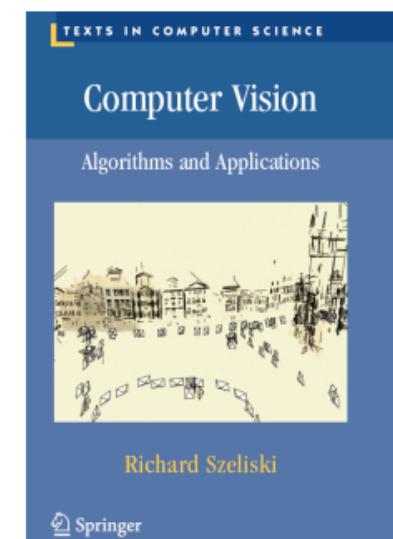
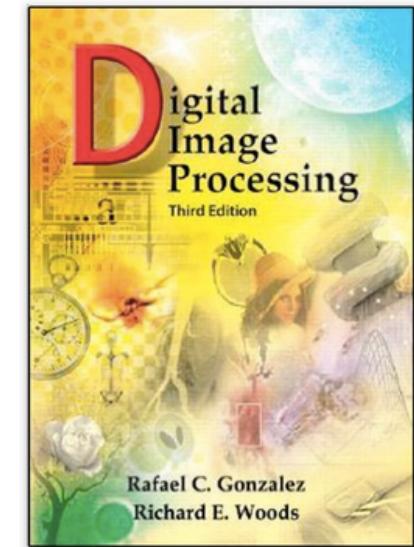
<http://www.ee.bgu.ac.il/~rrtammy/DIP-2017/DIP>

# Course Objectives

- The primary objective of this course is to provide the students the necessary computational tools to:
  - Understand the main principles of image processing and computer vision.
  - Be familiar with different classical and commonly used algorithms and understand their mathematical foundation
  - Implement (Matlab) and test commonly used image analysis algorithms
  - Develop critical reading of computer vision and digital signal processing and analysis literature

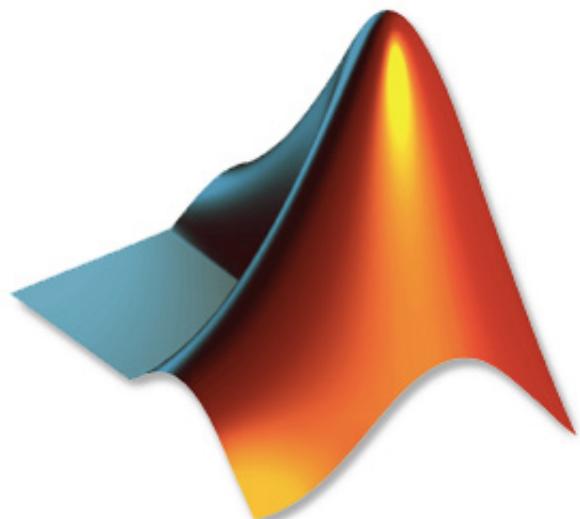
# Course Resources

- Szeliski, Richard. Computer vision: algorithms and applications. Springer Science & Business Media, 2010.
- Gonzalez, Rafael C., and Richard E. Woods. "Image processing." Digital image processing 2 (2007).
- Forsyth, David A., and Jean Ponce. "A modern approach." Computer vision: a modern approach (2003).
- Duda, Richard O., Peter E. Hart, and David G. Stork. Pattern classification. John Wiley & Sons, 2012.
- Bishop, C. "Pattern Recognition and Machine Learning (Information Science and Statistics), 1st edn. 2006. corr. 2nd printing edn." Springer, New York(2007).



# What should I do in order to succeed in the course?

- 6 Matlab assignments assignments, each 5% -> 30%
- Bonus assignments
- Final project and presentation 70%



**MATLAB**  
The Language of Technical Computing



# The instructor

- Tammy Riklin Raviv,

**Research interests:**

**Signal processing:** Biomedical Image Analysis, Computer Vision, Machine Learning

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**Telephone:** 086428812

**Fax:** 08-647 2949

**E-mail:** [rrtammy@ee.bgu.ac.il](mailto:rrtammy@ee.bgu.ac.il)

**Office:** 212/33

**Reception hours:**

please email

- Personal web page:

<http://www.ee.bgu.ac.il/~rrtammy/>

# The Final Project

- A list of possible projects will be distributed around Passover
- Students can choose a subject out of this list or come up with their own ideas.
- It is the students responsibility to schedule a meeting with the teaching assistants to discuss the project of their choice.
- Students can work with a single or two team mates.
- Students are expected to based their project on a scientific publication, make sure they understand it and are able to implement it using Matlab.
- All students should present their final project. Date TBD.

## List of topics (Tentative)

Overview on digital image processing,

Visual Perception

What is an image?

Sampling, quantization,

Histogram processing

Color image processing

Edge detection

Frequency domain analysis, Fourier transform

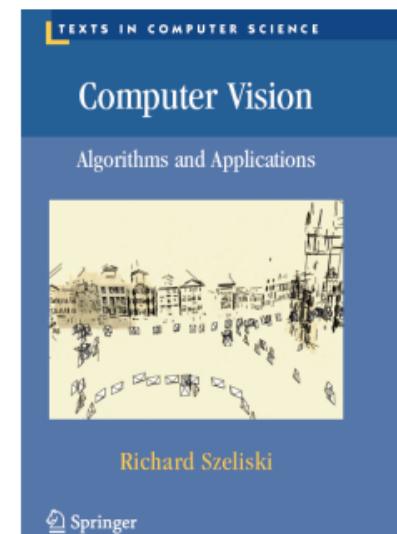
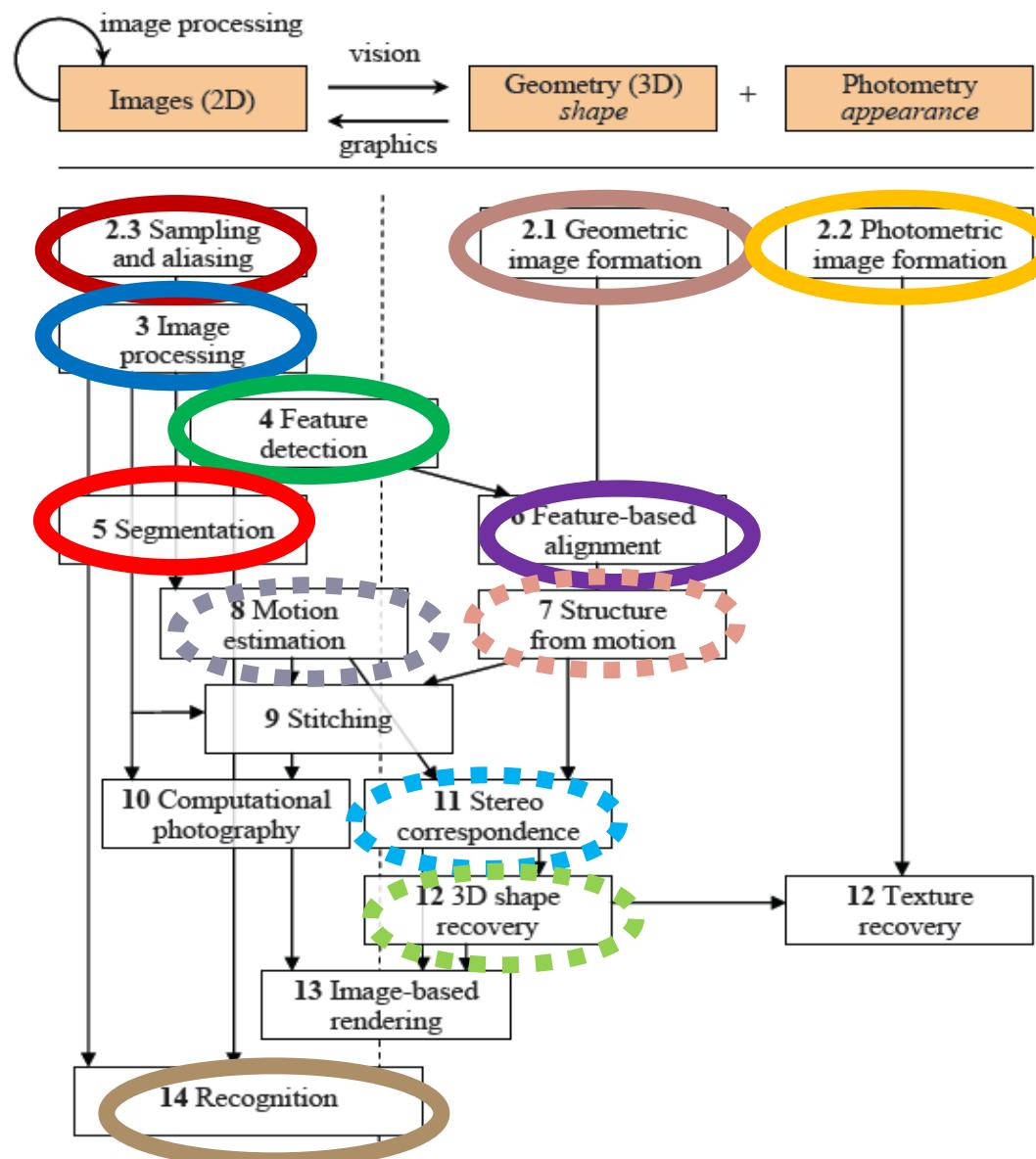
2D shape representation: Hough transform, pyramids, quad trees

Imaging geometry: Scaling, rotation, camera model, pose estimation

# List of topics (tentative)

- Photometry, shape from shading
- Image segmentation
- representation and descriptors, SIFTs and Hogs
- Stereo and Motion
- Face detection

# A graphical view on the syllabus



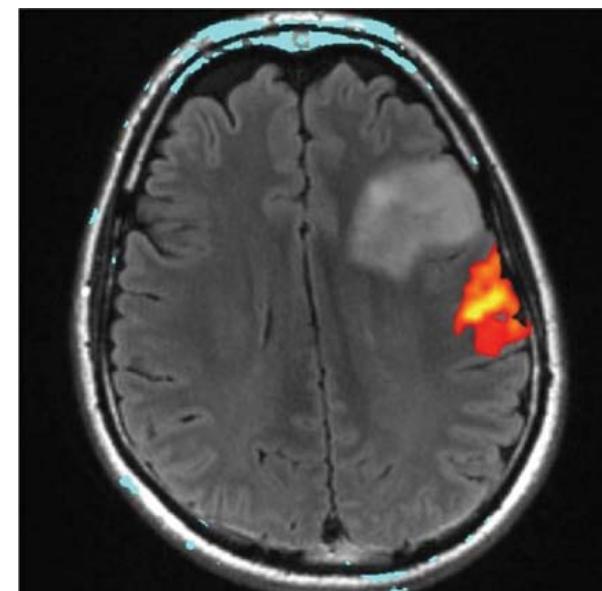
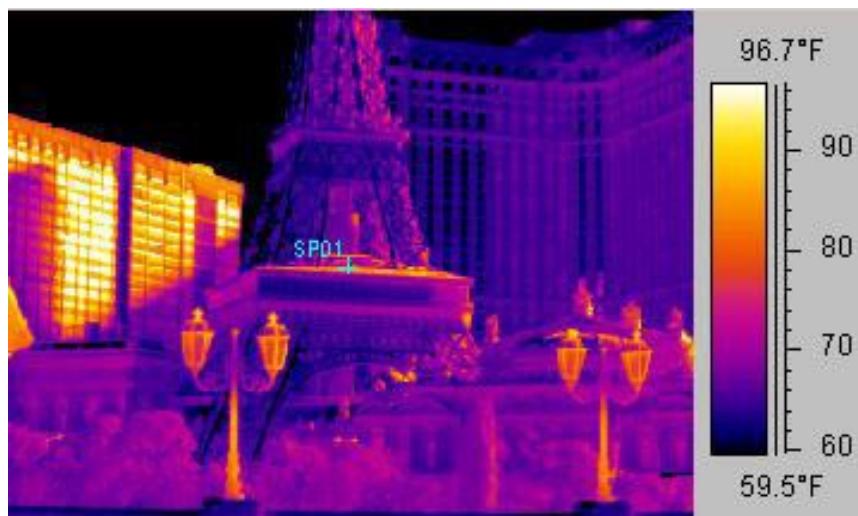
# The Rest of Today's Class

- Brief Overview
- Human Vision and Visual Perception
- What is an Image?

# The Rest of Today's Class

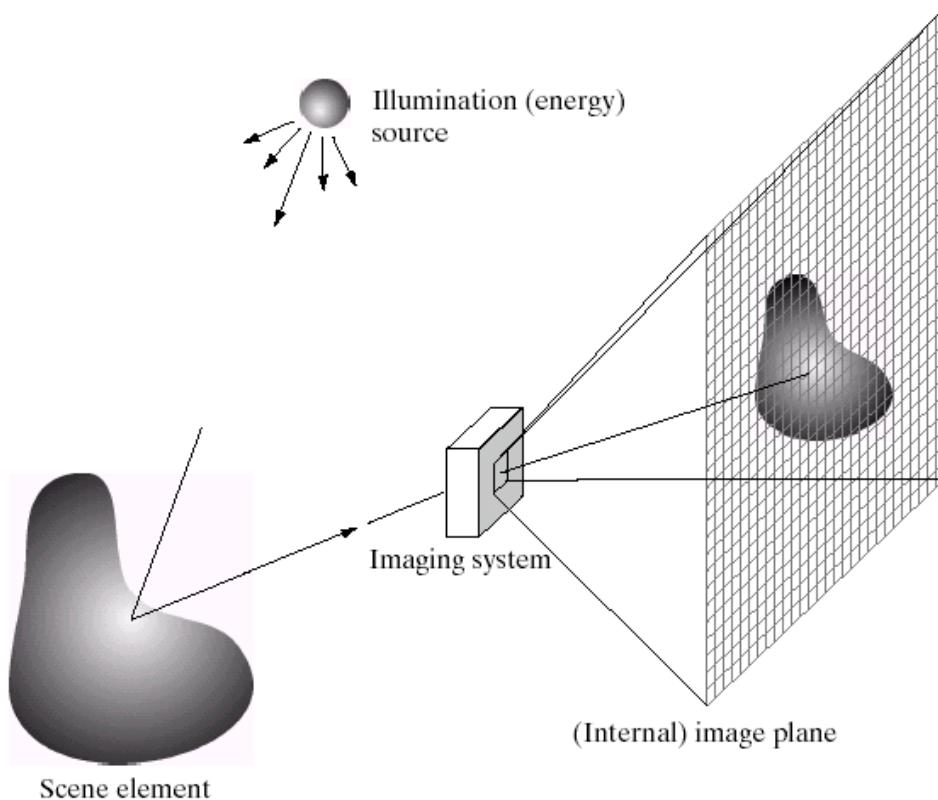
- Brief Overview
- Human Vision and Visual Perception
- What is an Image?

# A brief overview: What is an Image?

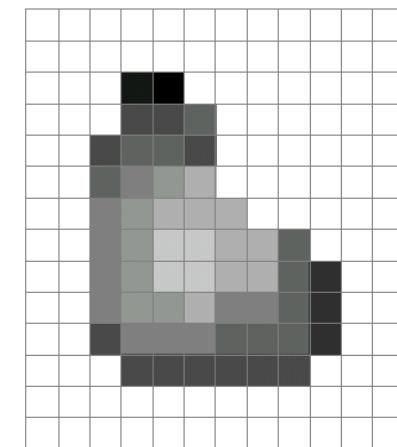


Brown's CV course

# A Brief Overview: Image Formation



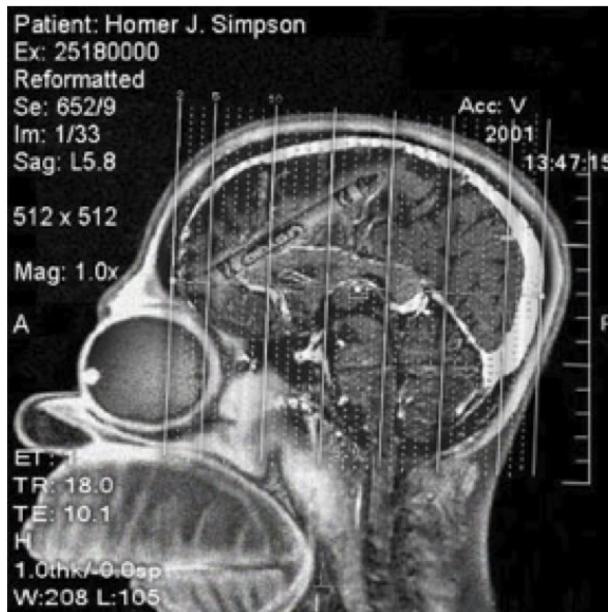
Camera Sensor



Output Image



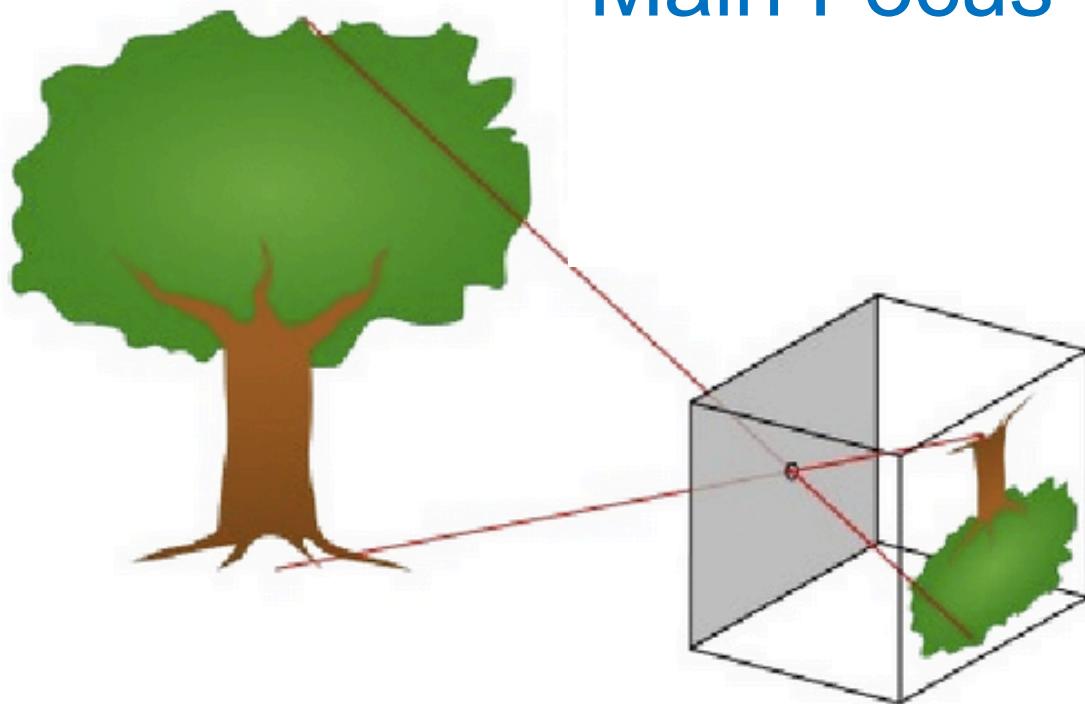
# A Brief Overview: Image Formation



See: Introduction to Medical Imaging  
Magnetic Resonance Imaging

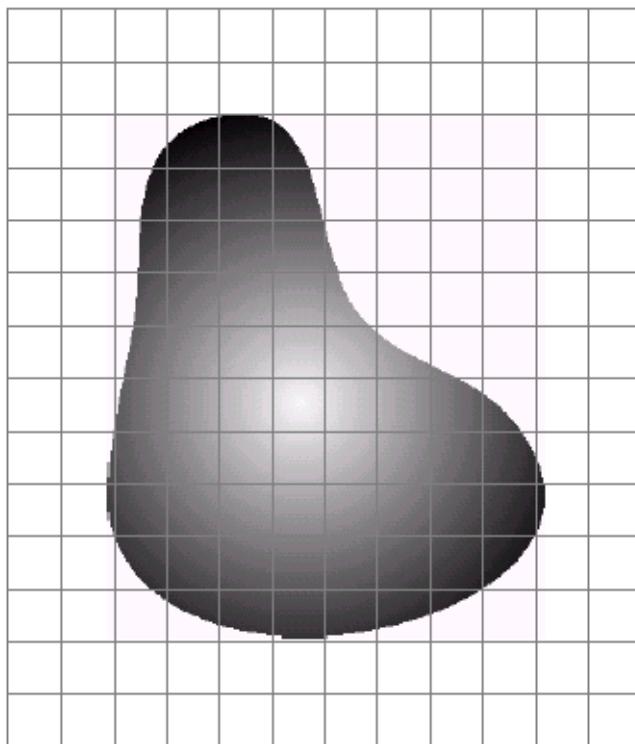
# A Brief Overview: Image Formation

Main Focus

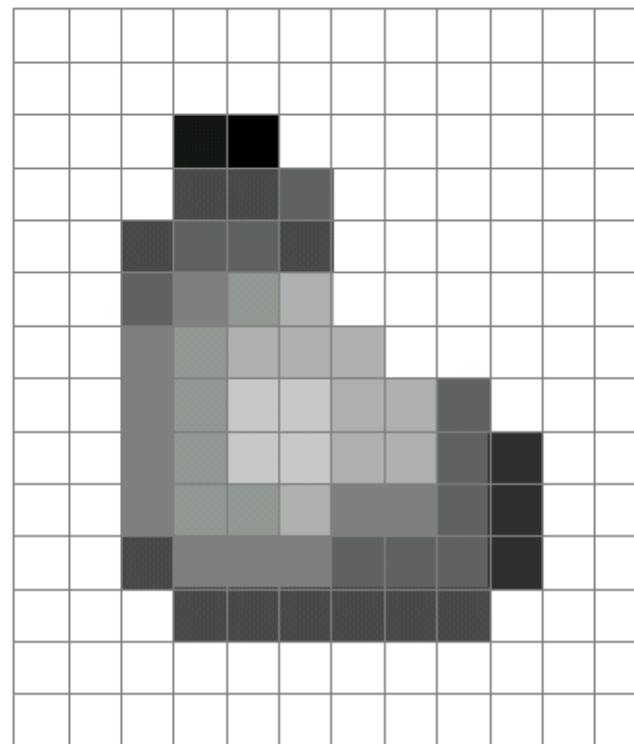


# A Brief Overview: Image Formation

## Sensor Array

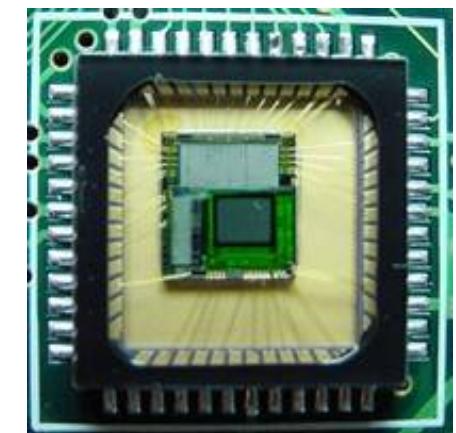


a



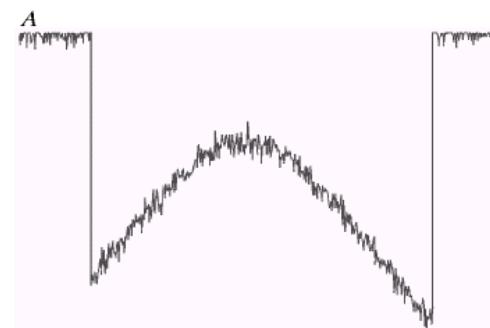
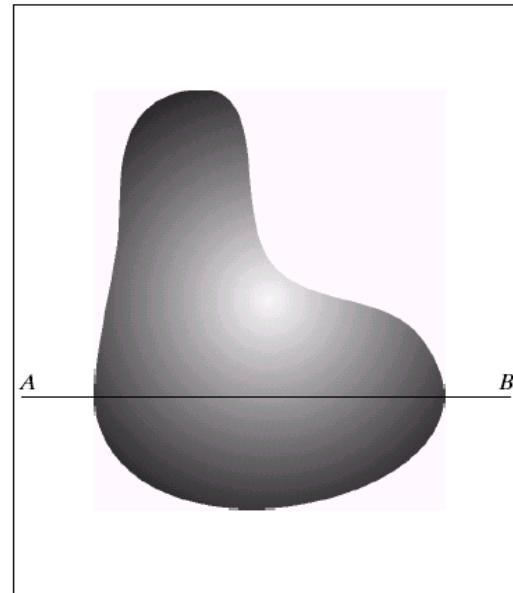
b

**FIGURE 2.17** (a) Continuous image projected onto a sensor array. (b) Result of image sampling and quantization.

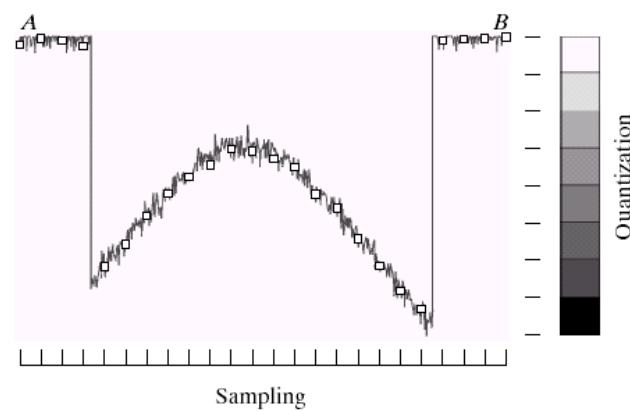


CMOS sensor

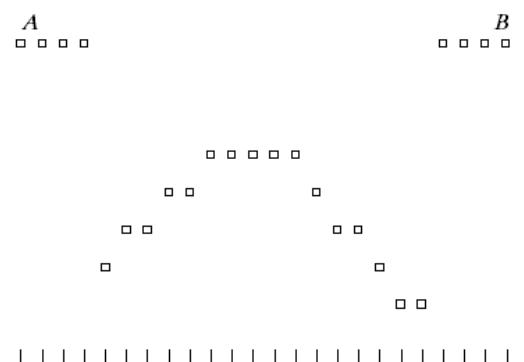
# A Brief Overview: Image Processing



## Quantization



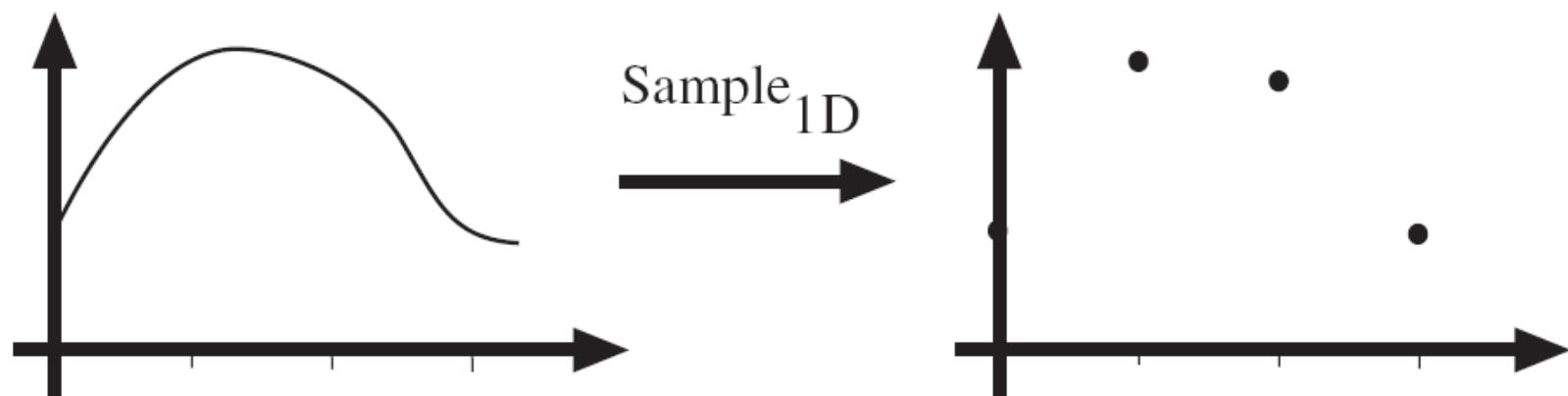
a b  
c d



**FIGURE 2.16** Generating a digital image. (a) Continuous image. (b) A scan line from  $A$  to  $B$  in the continuous image, used to illustrate the concepts of sampling and quantization. (c) Sampling and quantization. (d) Digital scan line.

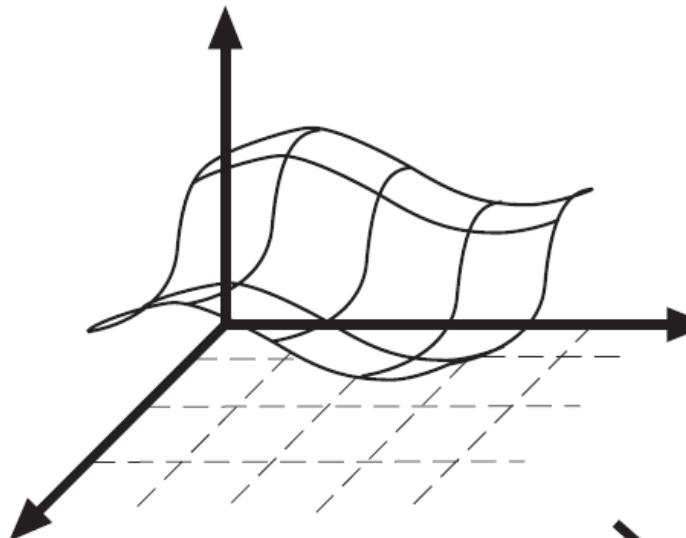
# A Brief Overview: Image Processing

## Images as 2D Signals



Sampling

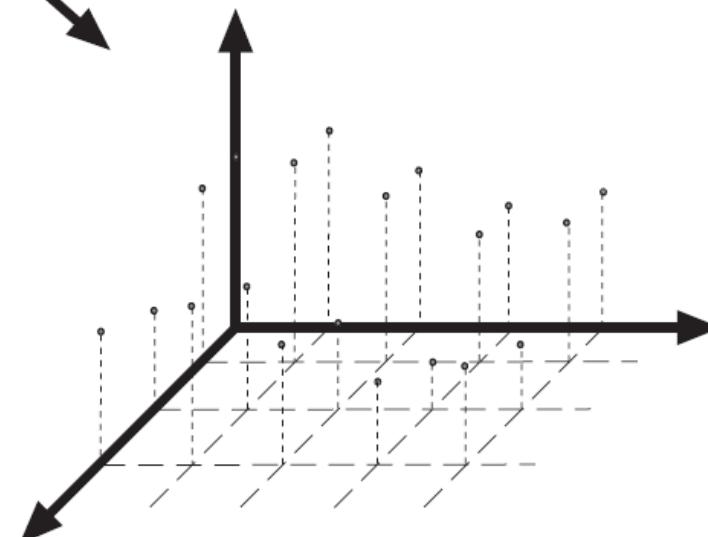
# A Brief Overview: Image Processing



Sample<sub>2D</sub>

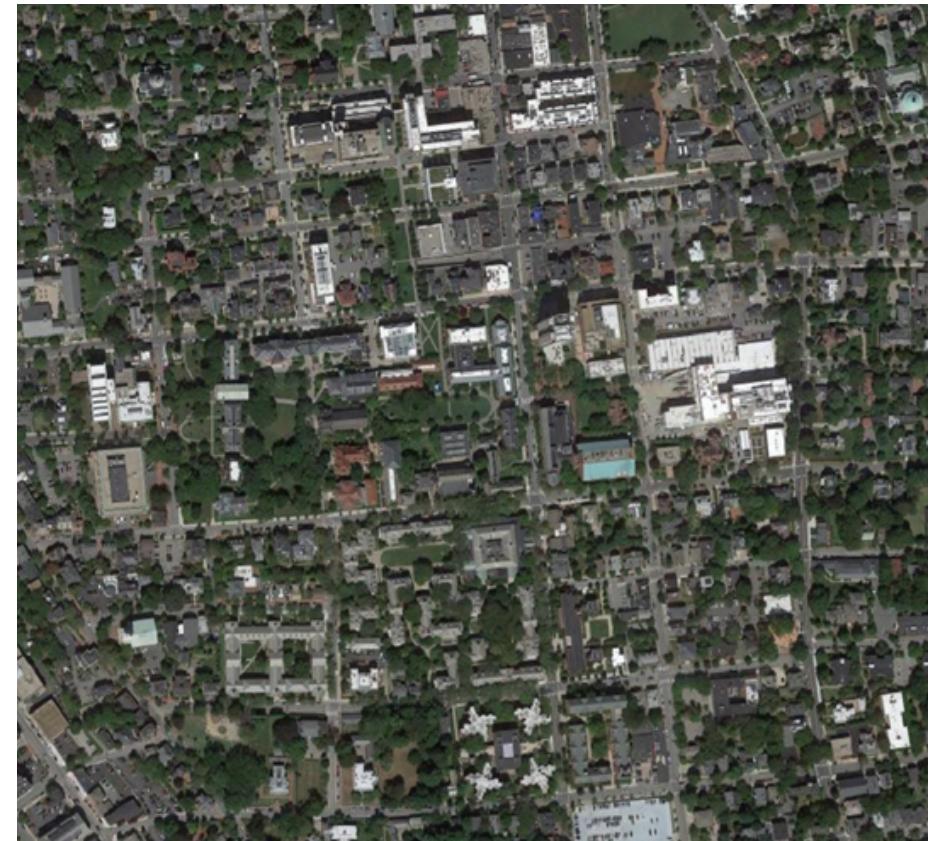
Images as 2D Signals

Sampling



# Resolution – geometric vs. spatial resolution

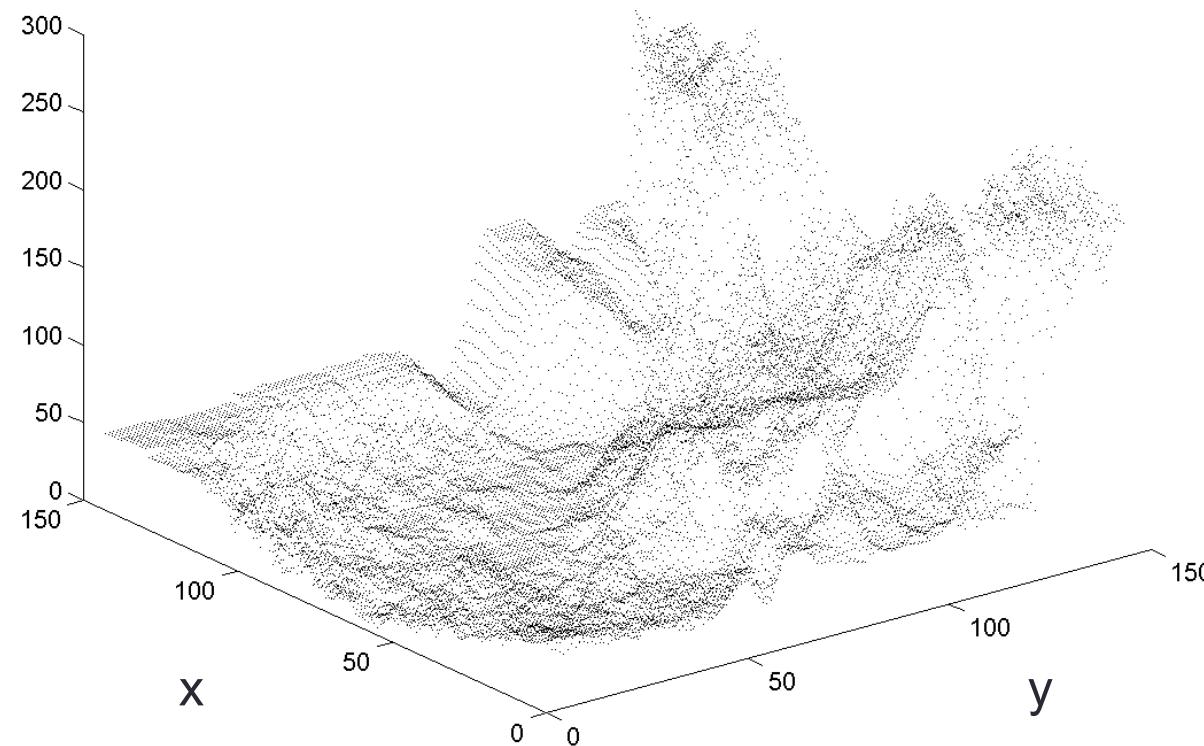
Both images are ~500x500 pixels



# A Brief Overview: Image Representation

## Grayscale Digital Image

Brightness  
or intensity



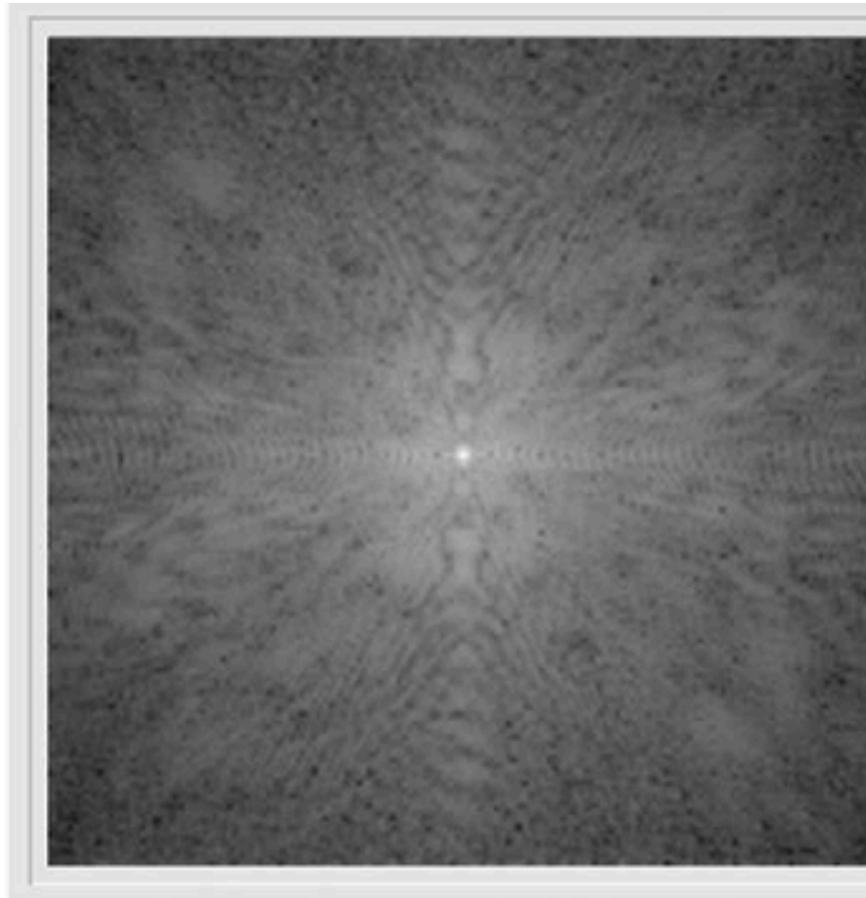
# A Brief Overview: Light and Color



# A Brief Overview: Edge detection



# A Brief Overview: Frequency Analysis

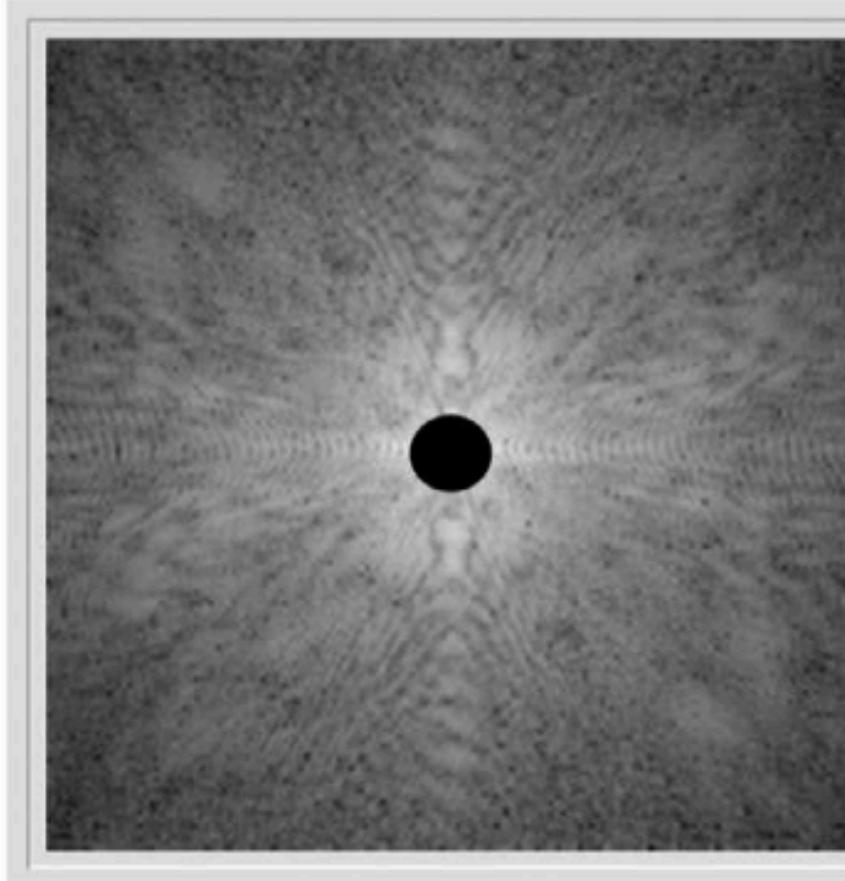


Fourier domain



Image domain

# A Brief Overview: Frequency Analysis

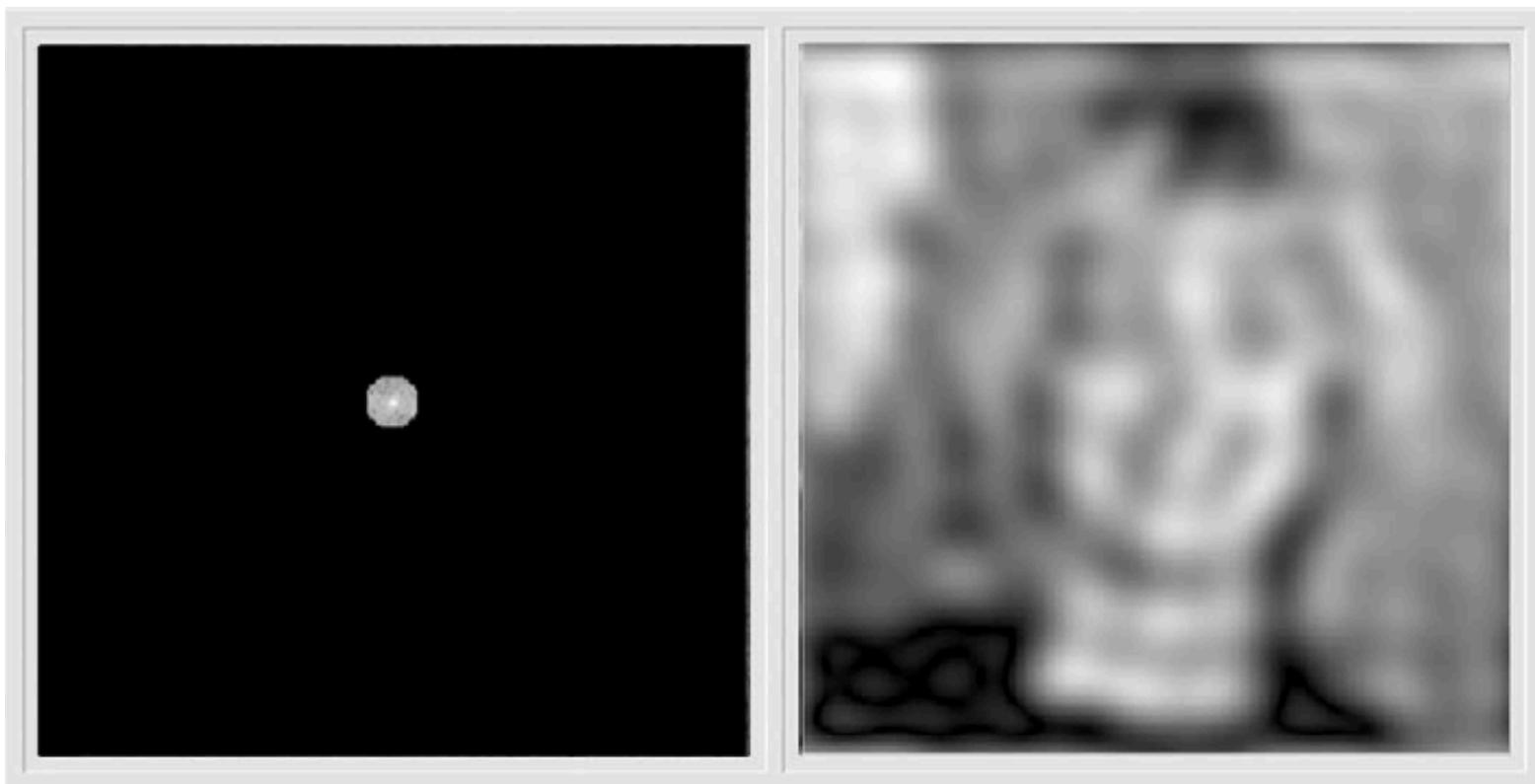


Fourier domain



Image domain

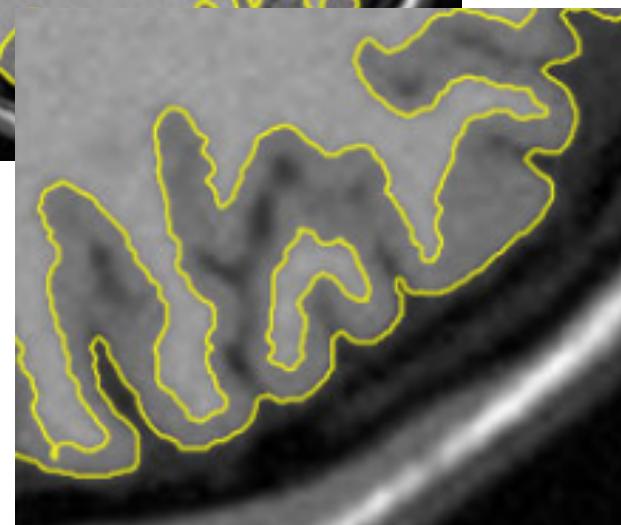
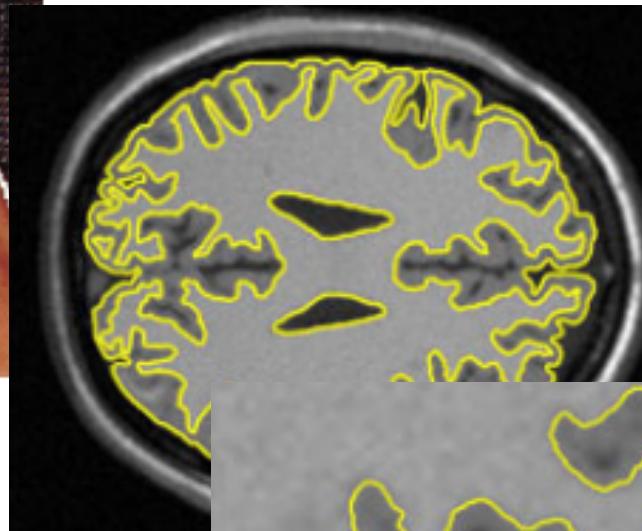
# A Brief Overview: Frequency Analysis



Fourier domain

Image domain

# A Brief Overview: Image Segmentation



# A Brief Overview: Image Segmentation



Texture  
segmentation

# A Brief Overview: Image Segmentation



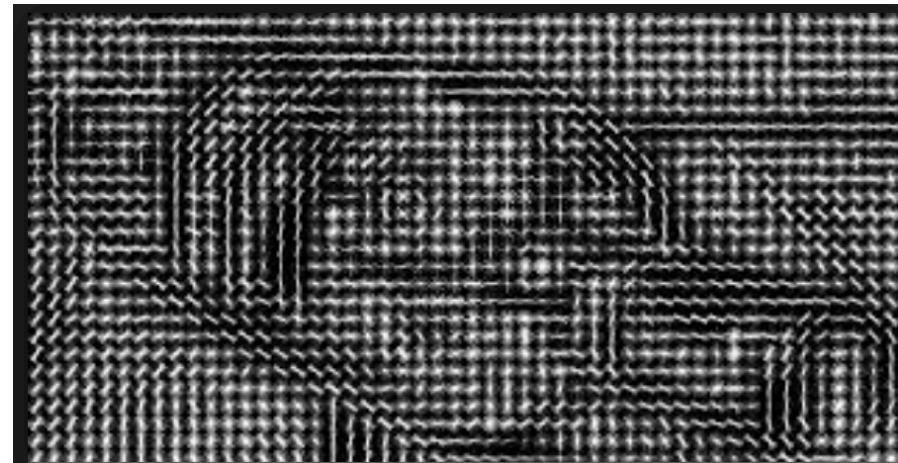
Prior based  
segmentation

# A Brief Overview: Image Segmentation

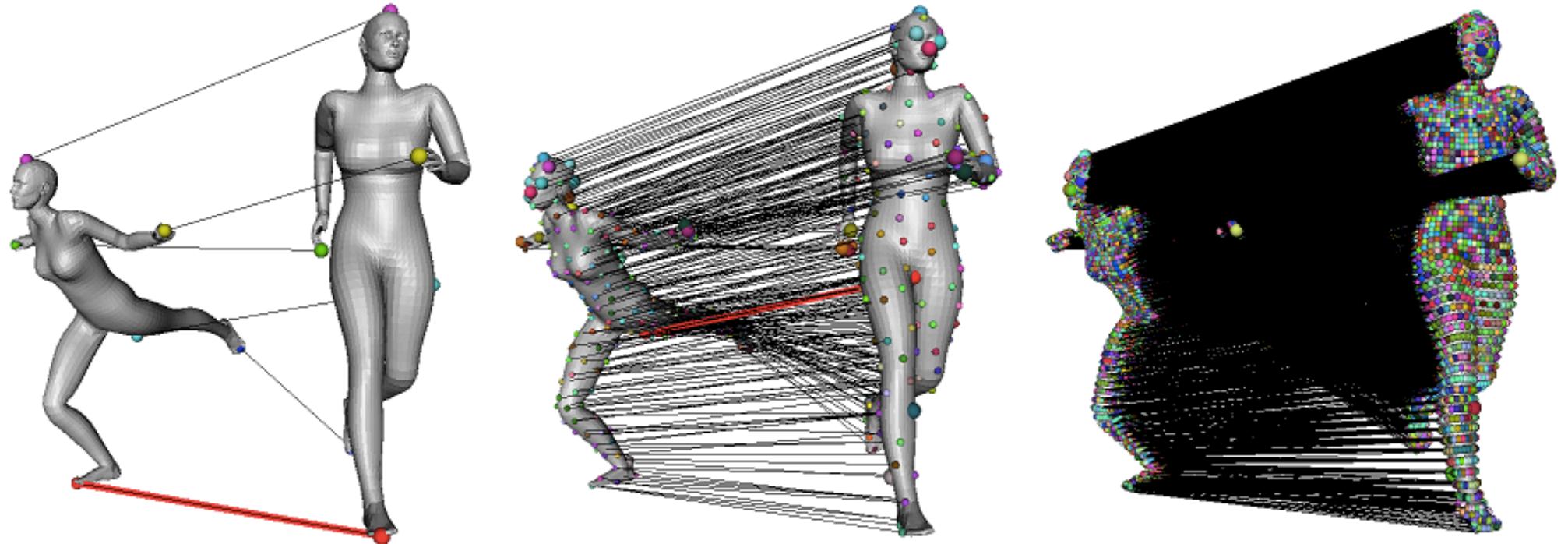


Symmetry based  
segmentation

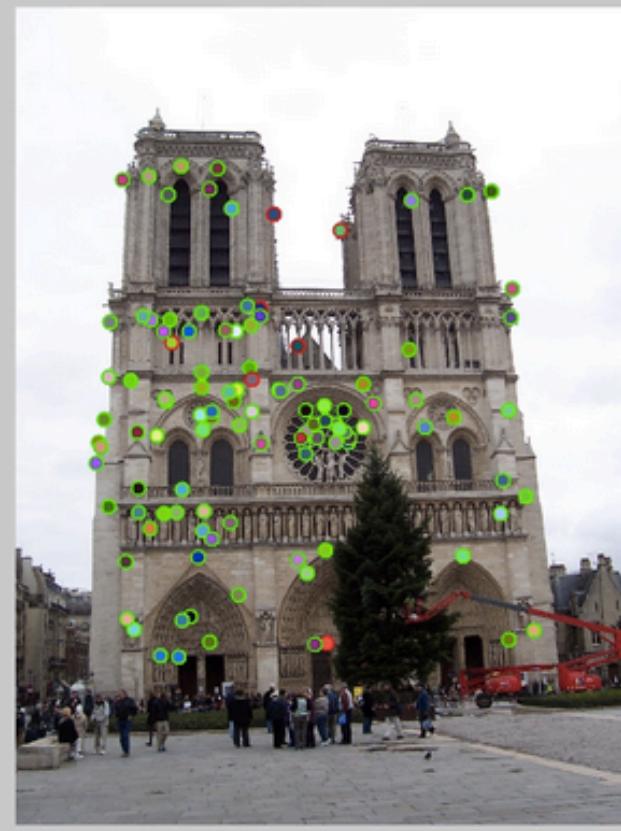
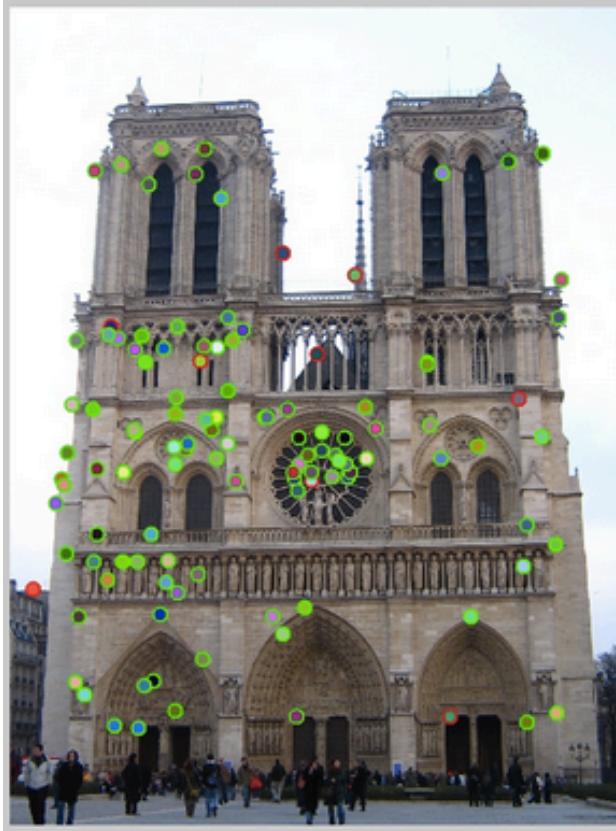
# A Brief Overview: Feature Detection



# A Brief Overview: Correspondences



# A Brief Overview: Correspondences



# A Brief Overview: Stereo and 3D reconstruction



# A Brief Overview: Object Detection and Recognition



# A Brief Overview: Motion



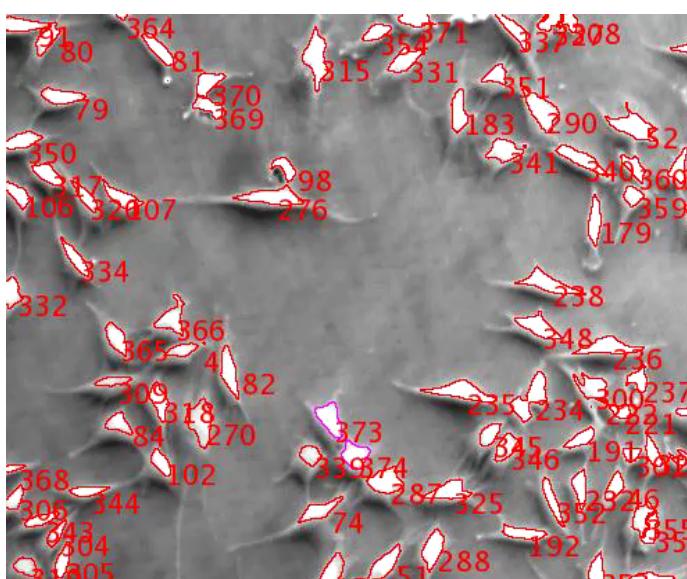
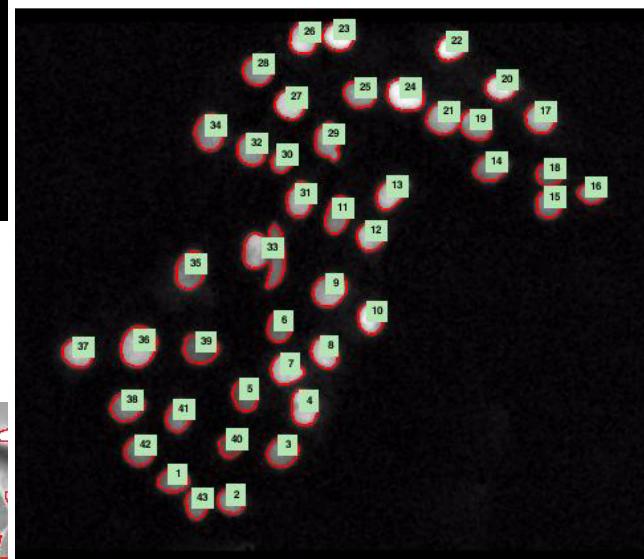
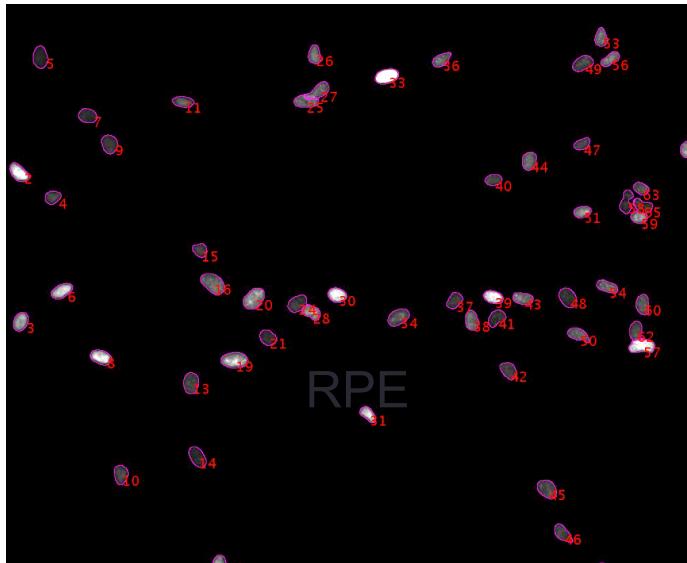
Optical Flow

# A Brief Overview: Tracking

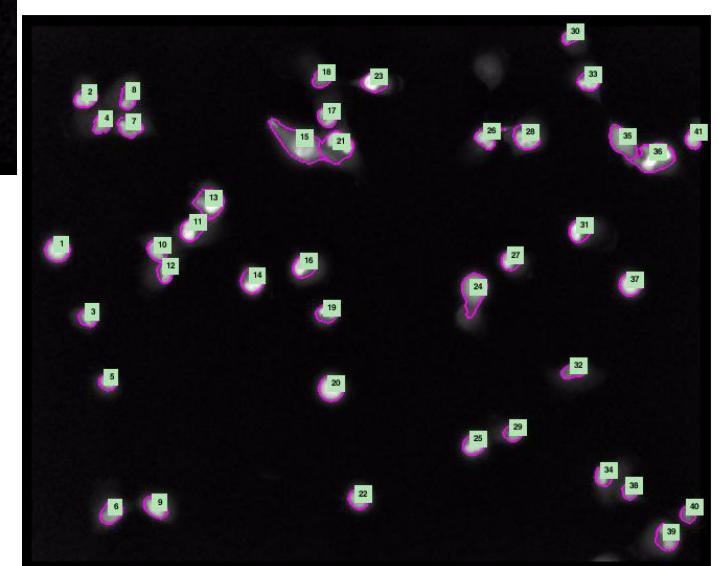


Hyun Tae Na Thesis

# A Brief Overview: Tracking



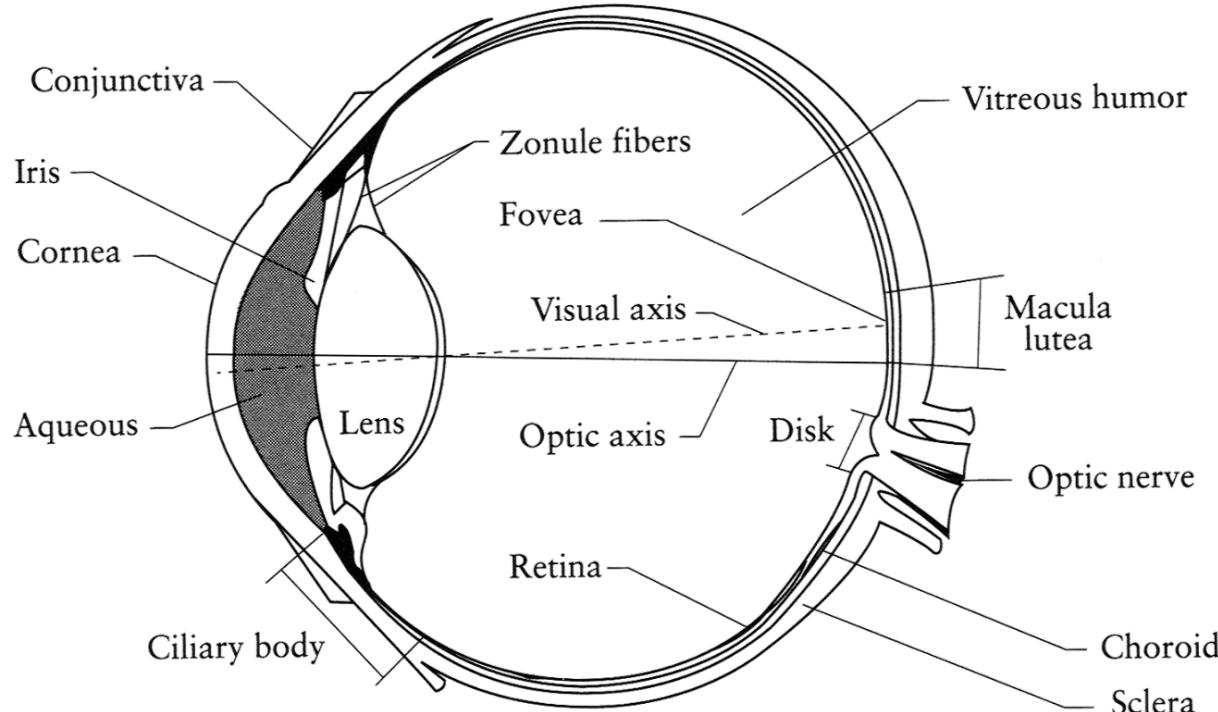
Arbelle's Thesis



# The Rest of Today's Class

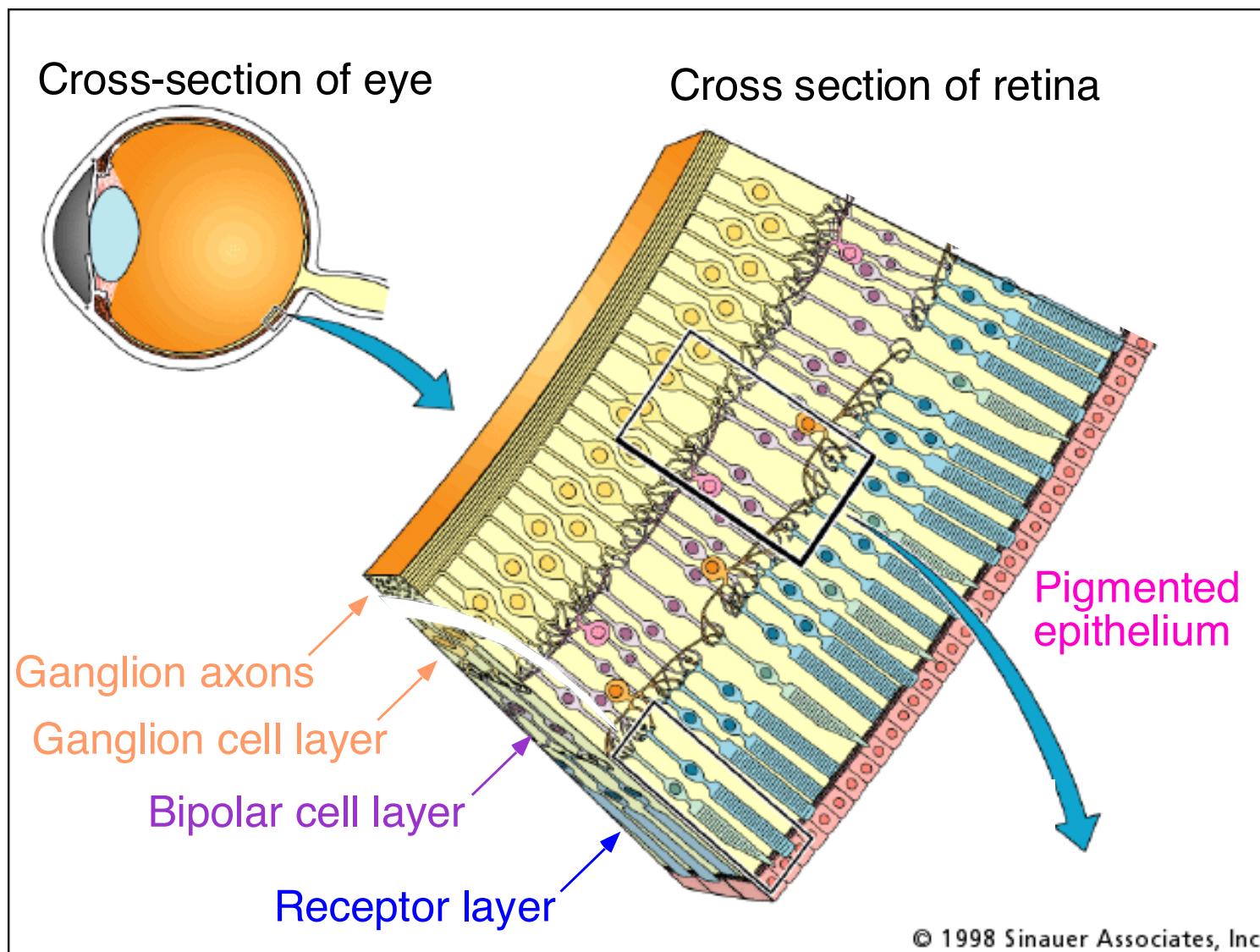
- Brief Overview
- Human Vision and Visual Perception
- What is an Image?

# Human Vision/ Visual Perception

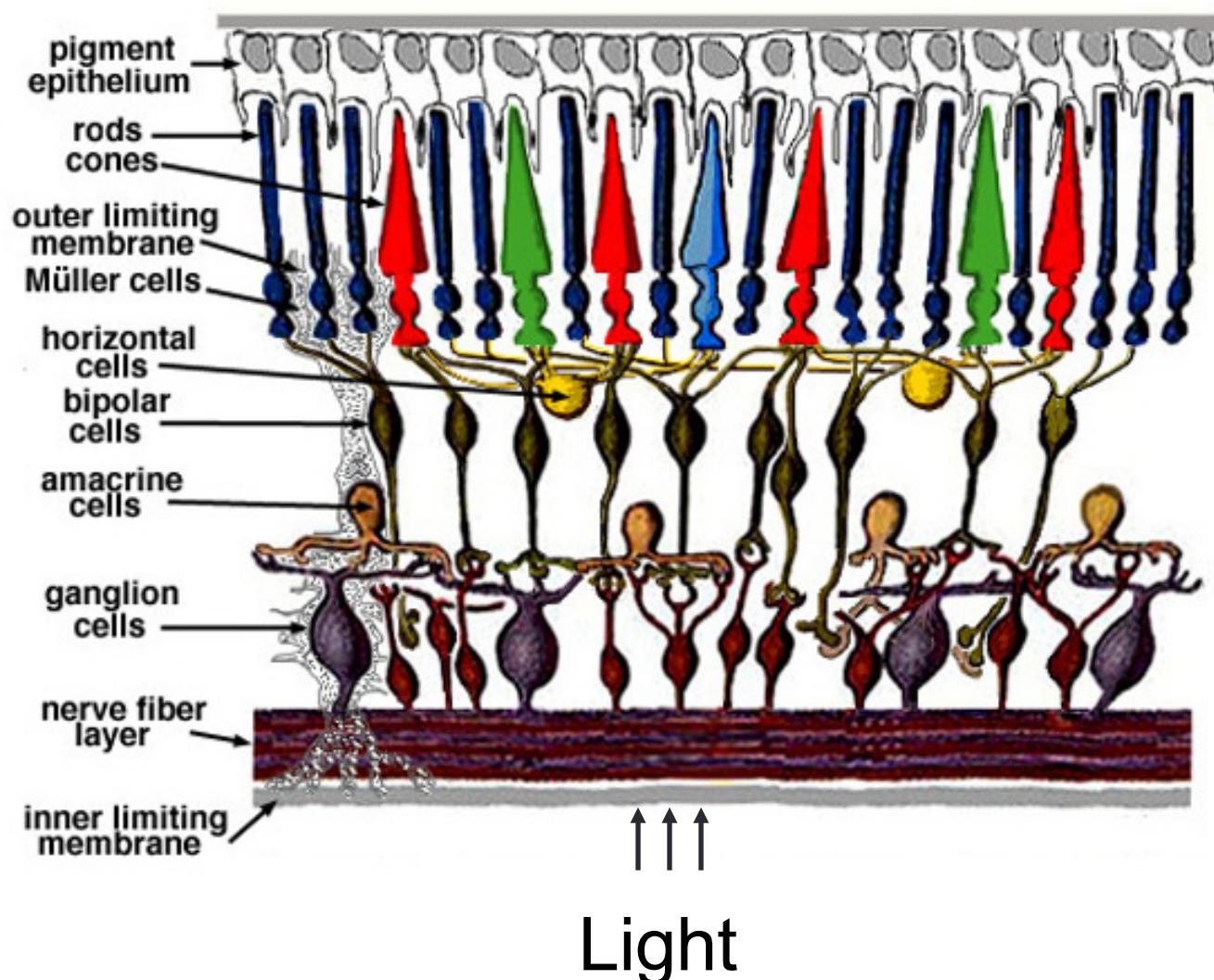


- The human eye is a camera
  - **Iris** - colored annulus with radial muscles
  - **Pupil** - the hole (aperture) whose size is controlled by the iris
  - What's the sensor?
    - photoreceptor cells (rods and cones) in the **retina**

# Human Vision: the Retina



# Human Vision: Retina up-close



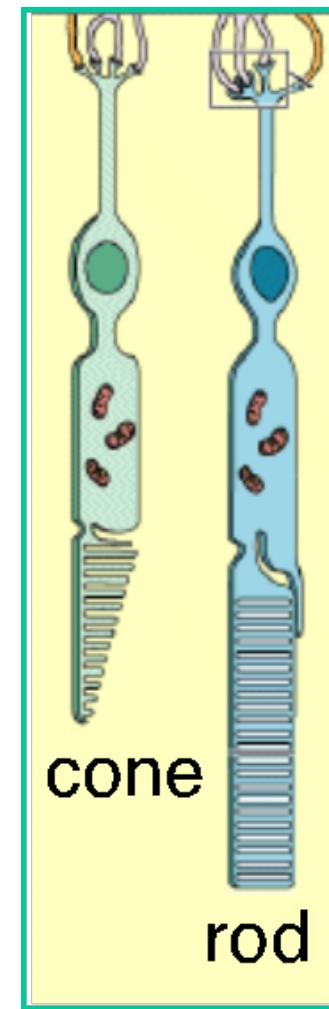
# Human Vision: Retina up-close

## Cones

cone-shaped  
less sensitive  
operate in high light  
color vision

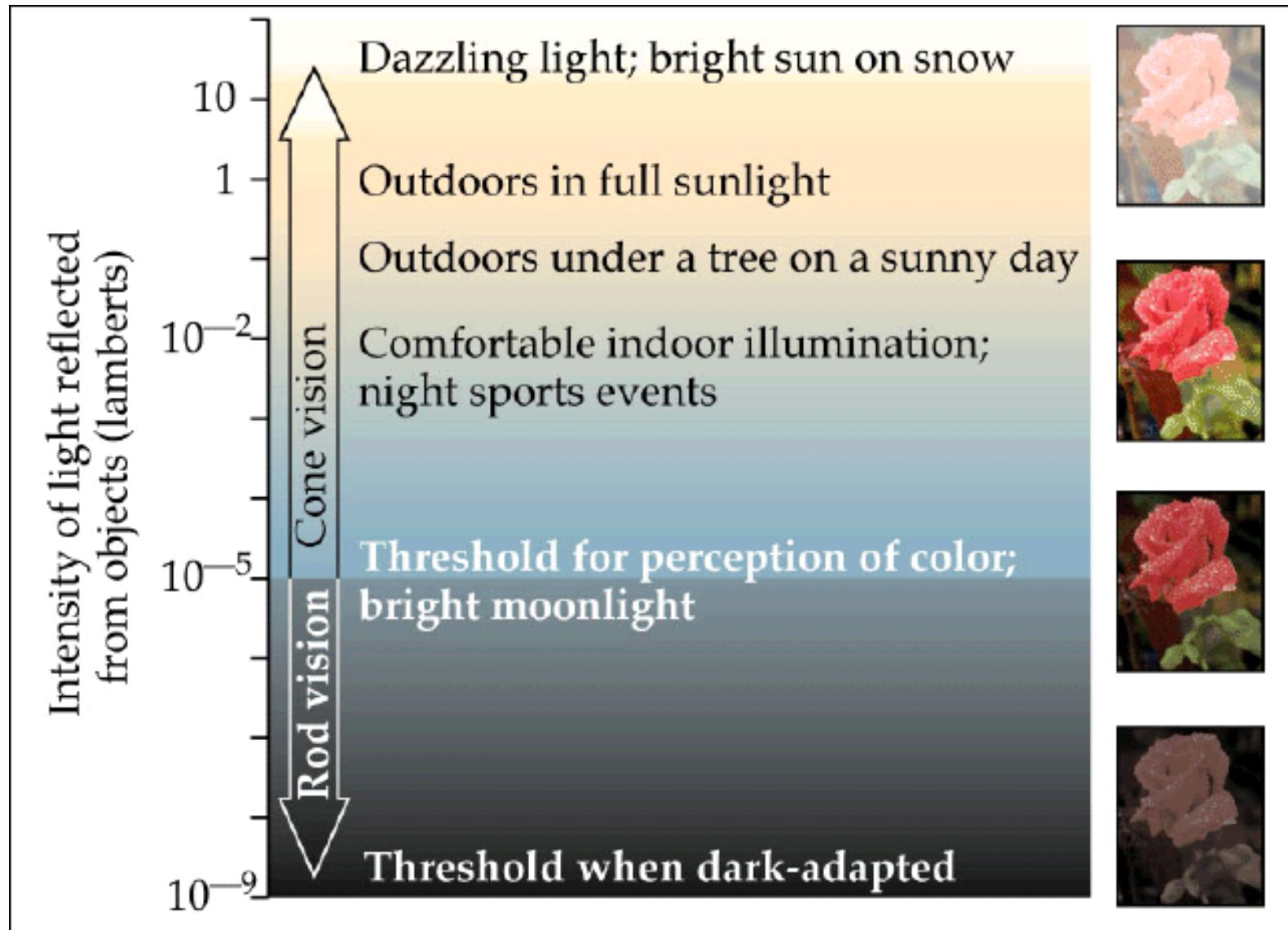
## Rods

rod-shaped  
highly sensitive  
operate at night  
gray-scale vision

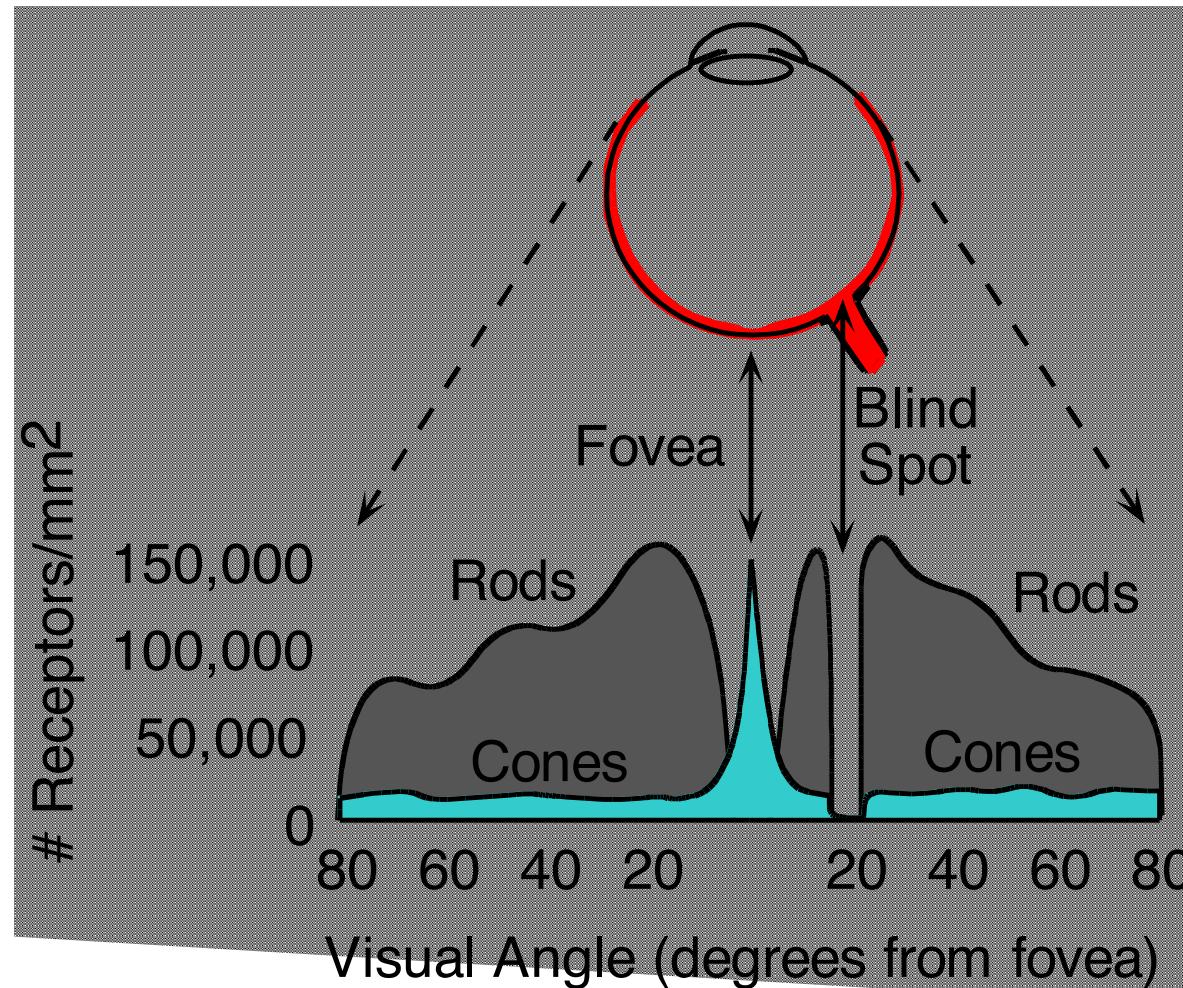


Two types of light-sensitive receptors

# Rod / Cone sensitivity



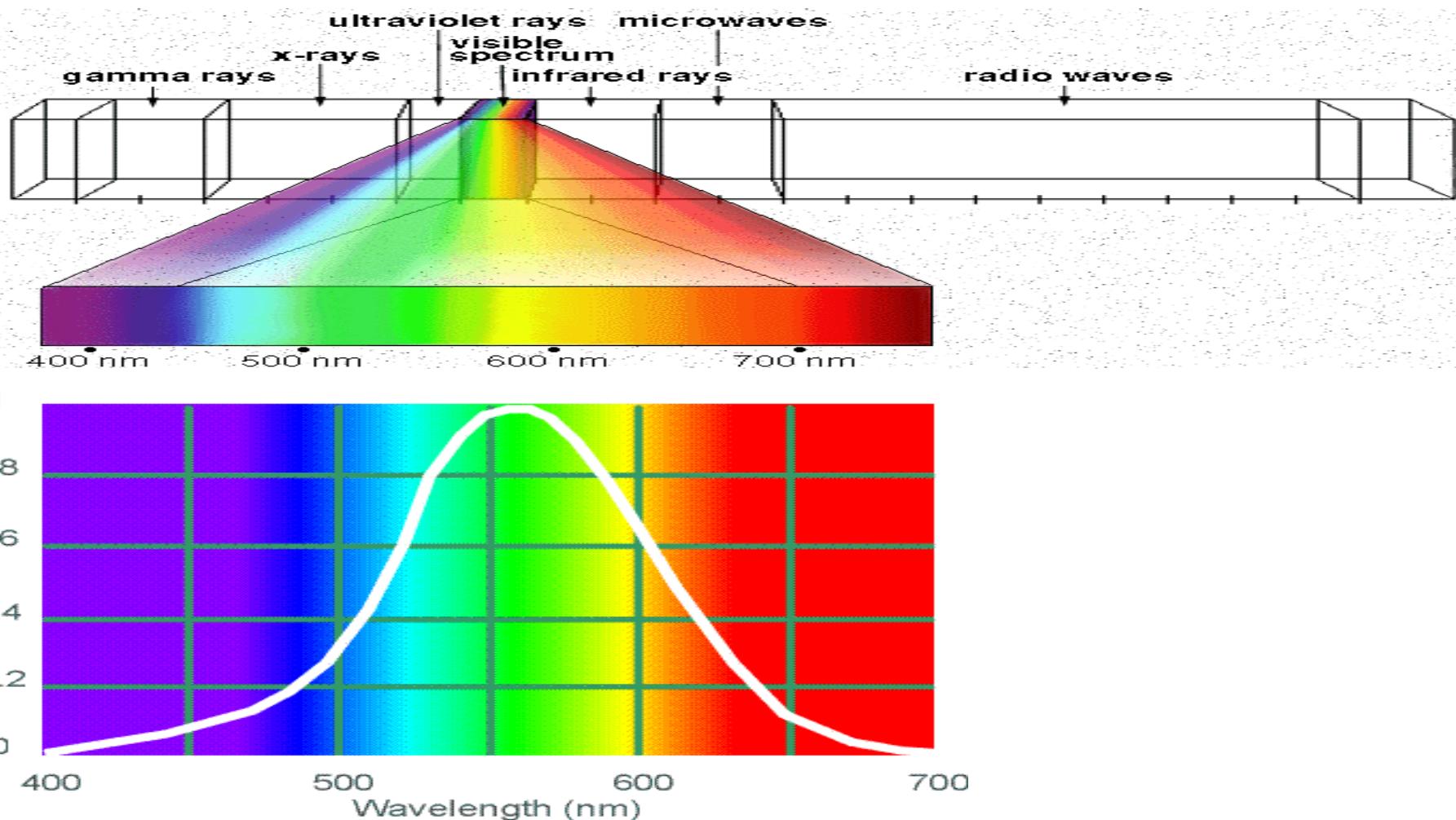
# Distribution of Rods and Cones



Night Sky: why are there more stars off-center?

Averted vision: [http://en.wikipedia.org/wiki/Averted\\_vision](http://en.wikipedia.org/wiki/Averted_vision)

# Electromagnetic Spectrum

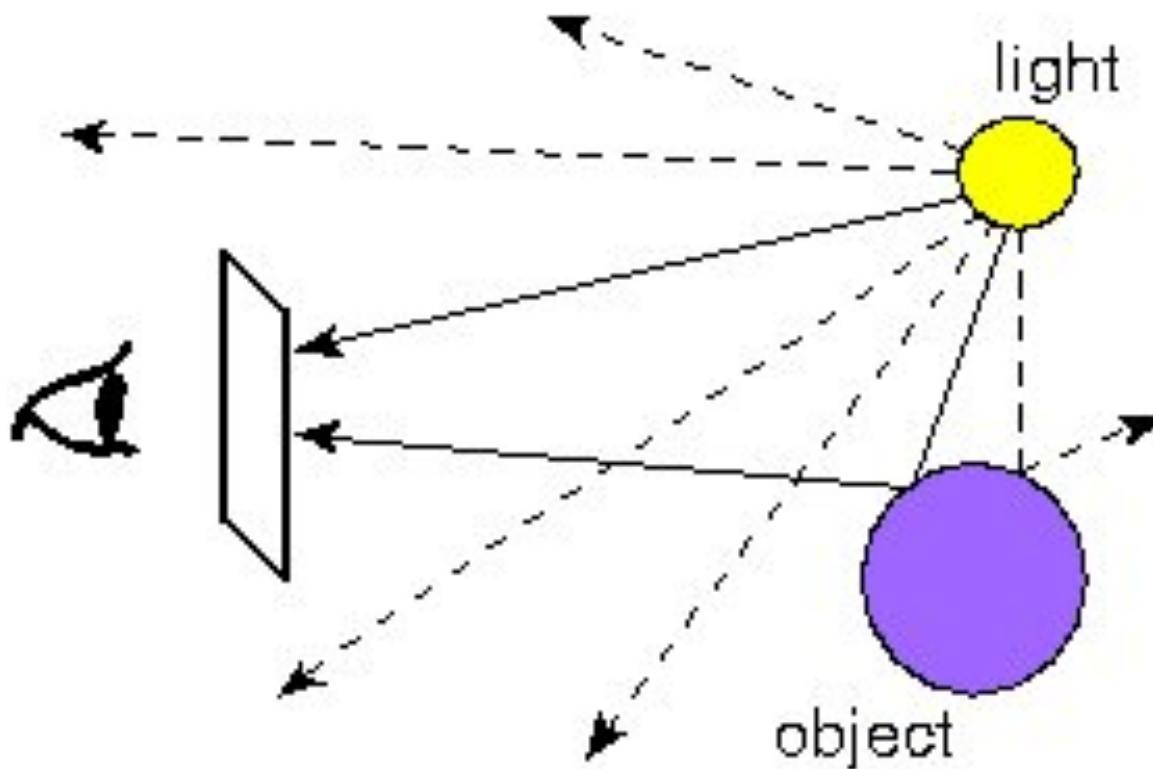


Human Luminance Sensitivity Function

<http://www.yorku.ca/eye/photopik.htm>

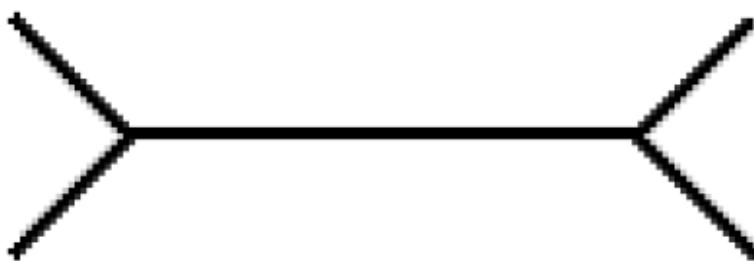
# Human Visual Perception

What can we learn from human visual perception?

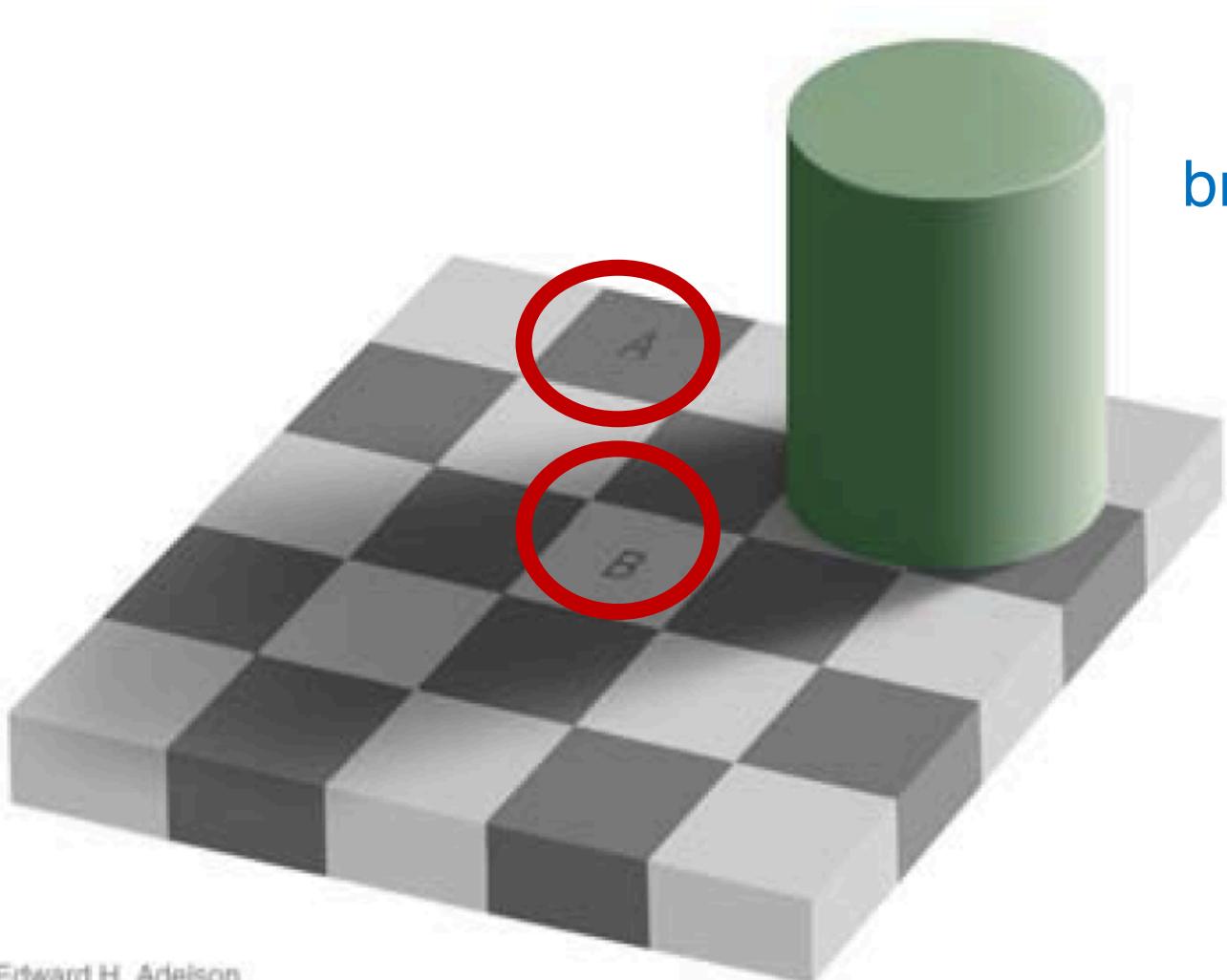


# Human Perception

Muller-Lyer illusion



# Human Perception

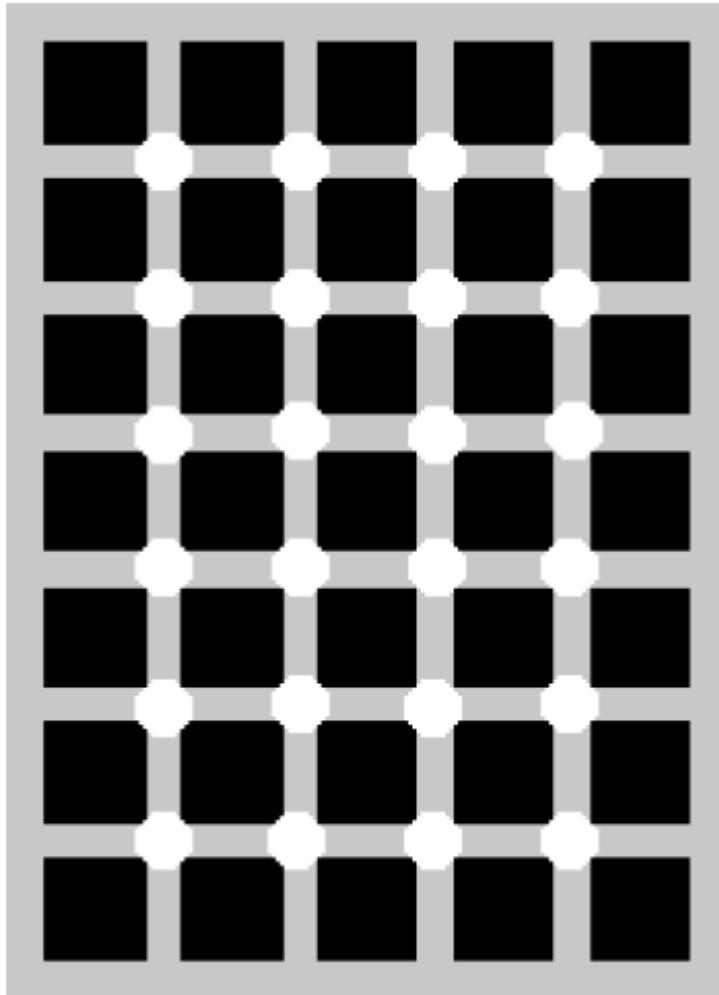


brightness constancy

Edward H. Adelson

Ted Adelson, [http://web.mit.edu/persci/  
people/adelson/checkershadow\\_illusion.html](http://web.mit.edu/persci/people/adelson/checkershadow_illusion.html)

# Human Perception



Hermann grid illusion

Hany Farid, <http://www.cs.dartmouth.edu/farid/illusions/hermann.html>

# Human Perception

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

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 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

Count the red X

pop-out effect (Treisman 1985)

# The Rest of Today's Class

- Brief Overview
- Human Vision and Visual Perception
- What is an Image?

# What is an Image?



Bela Borsodi

This Image  
is taken from  
Brown's  
Computer  
Vision Course

# How would it look through the “computer’s eyes” ?



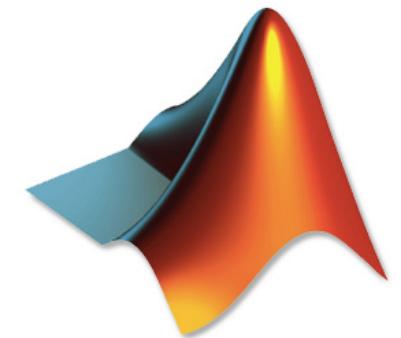
# Why is this an image?



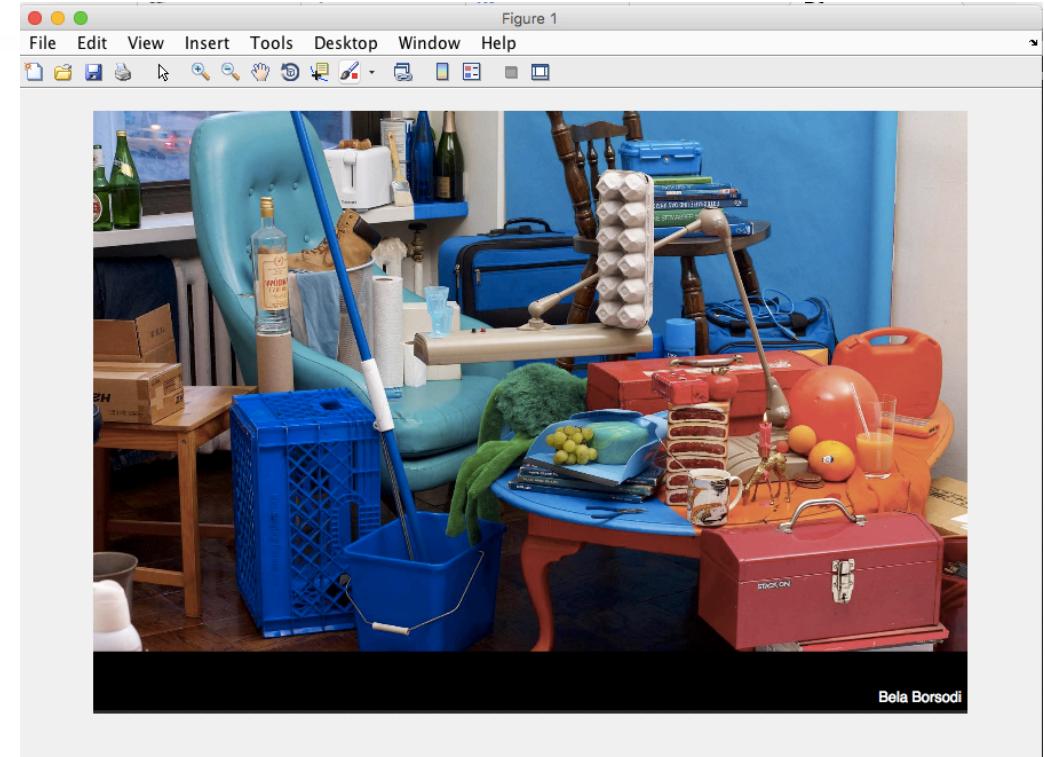
Bela Borsodi

# Hello Word !

```
>>  
>>  
>> myFirstImage = imread('someImage.png');  
>> whos  
Name          Size            Bytes  Class  
  
ans            526x764x3      1205592  uint8  
myFirstImage   526x764x3      1205592  uint8  
  
>> imshow(myFirstImage);  
|
```

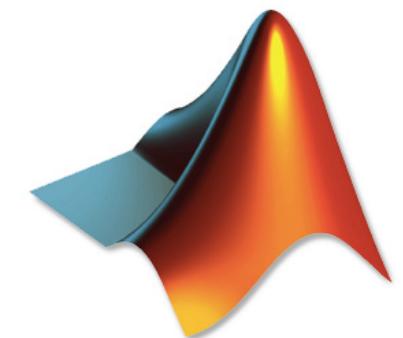


**MATLAB**  
The Language of Technical Computing

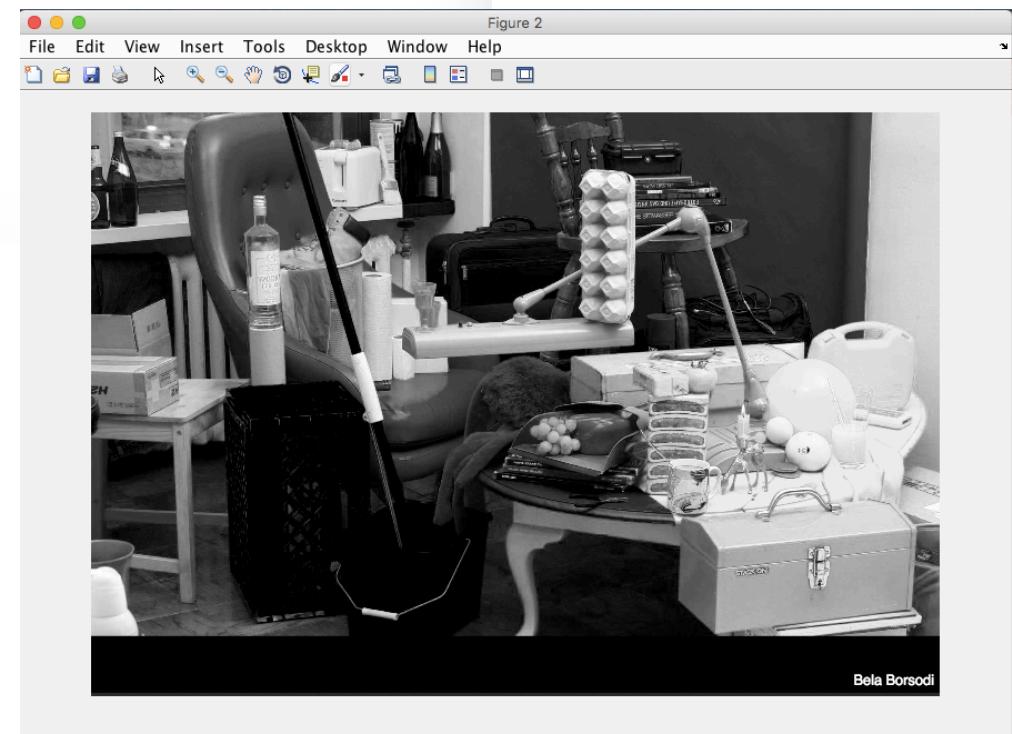


# Hello Word !

```
>>  
>> myFirstImage = imread('someImage.png');  
>> whos  
 Name          Size            Bytes  Class  
 ans           526x764x3        1205592  uint8  
 myFirstImage  526x764x3        1205592  uint8  
  
>> imshow(myFirstImage);  
>> I1 = myFirstImage(:,:,1);  
>> figure;imshow(I1)  
fx >>
```



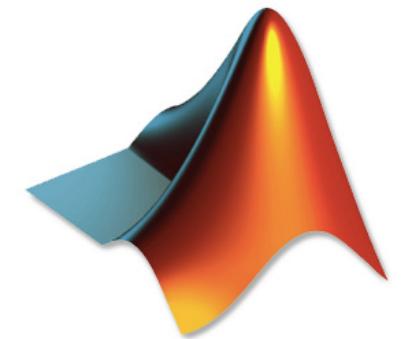
**MATLAB**  
The Language of Technical Computing



Bela Borsodi

# Hello Word !

```
>> I1 = myFirstImage(:,:,1);  
>> figure;imshow(I1)  
>> colorbar  
x >> |
```



**MATLAB**  
The Language of Technical Computing



# Hello World ☺ RGB

```
>> imshow(myFirstImage);
>> I1 = myFirstImage(:,:,:1);
>> figure;imshow(I1)
>> I2 = myFirstImage(:,:,:2);
>> figure;imshow(I2)
>> I3 = myFirstImage(:,:,:3);
>> figure;imshow(I3)
```

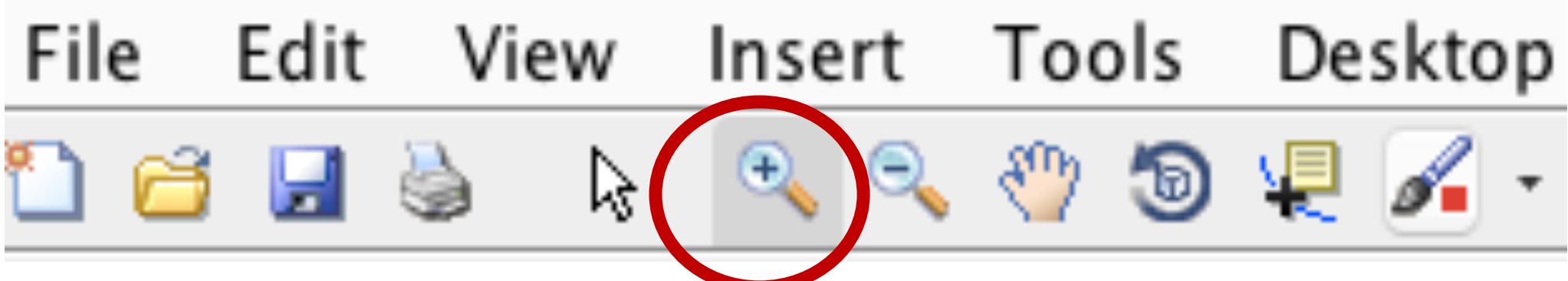


Bela Borsodi

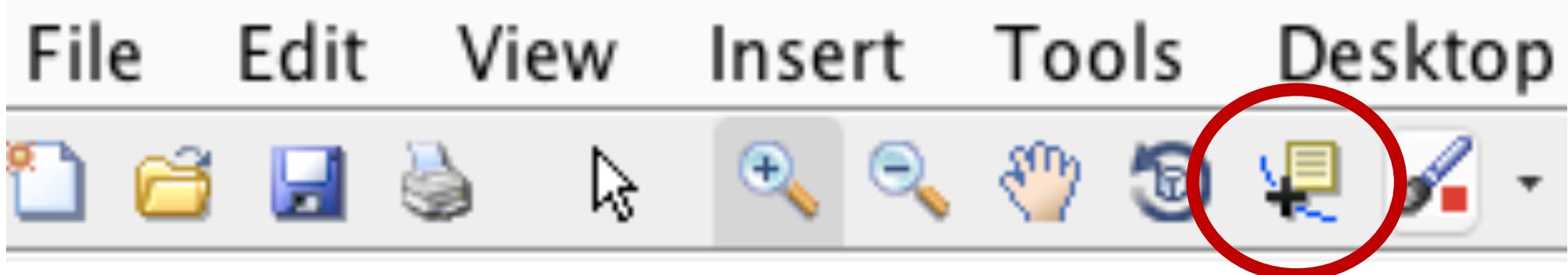


Bela Borsodi

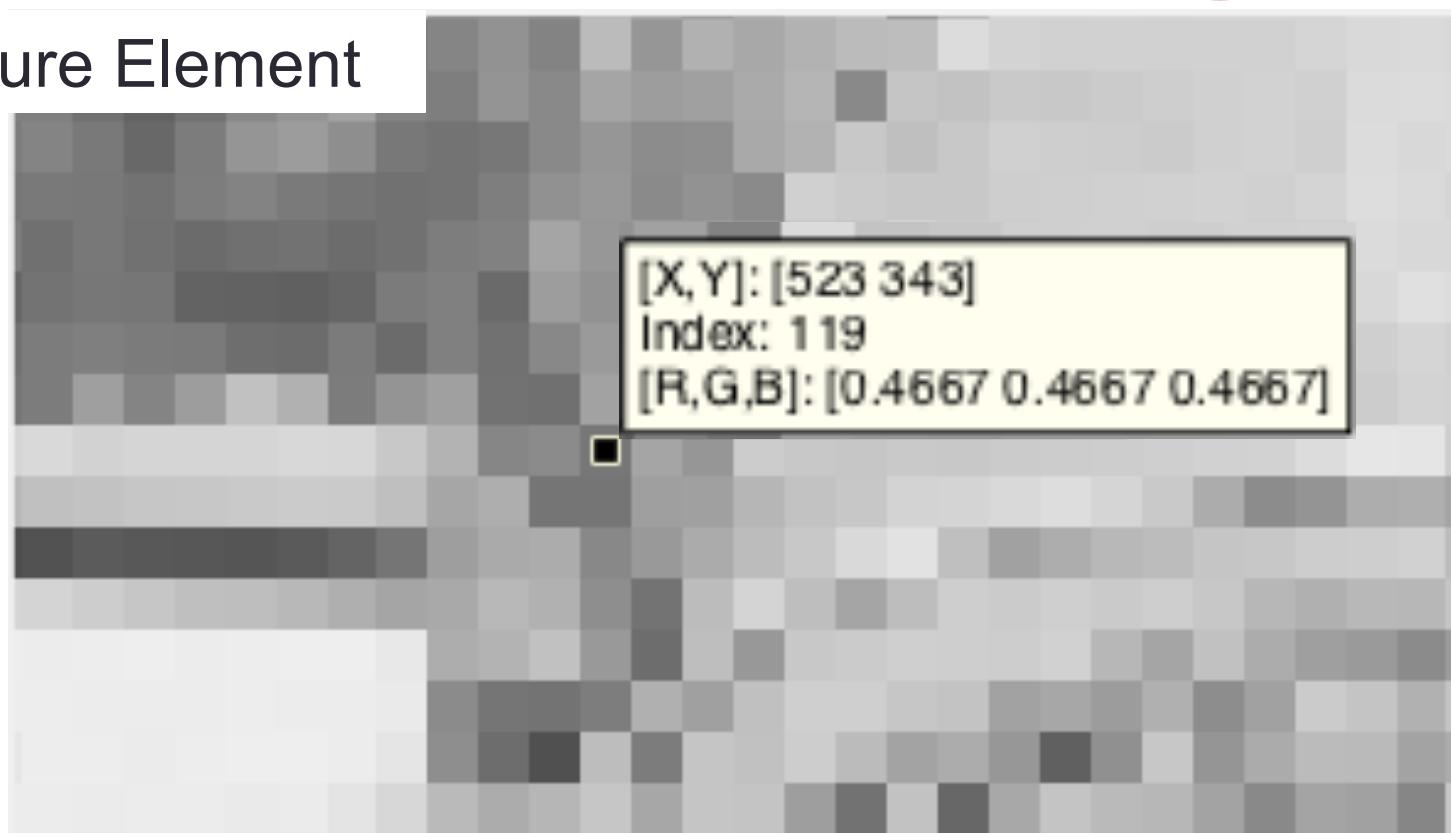
# Let's zoom in



# Let's Pixel



Pixel = Picture Element



# Let's Pixel



220	218	214	224	224	179	223	208	230	237	210
211	217	191	215	192	187	221	213	232	223	231
212	219	188	229	181	189	221	212	208	230	235
217	214	220	198	190	194	205	187	221	236	222
202	210	175	154	187	221	200	206	244	229	226
173	196	218	137	201	226	195	241	241	238	242
201	207	210	199	180	183	207	245	247	249	245
208	206	211	199	170	184	216	252	250	244	219
204	177	211	211	189	178	206	244	238	217	224
207	137	203	212	193	151	207	227	221	230	236
203	159	207	209	208	143	209	221	231	236	235

# Let's count

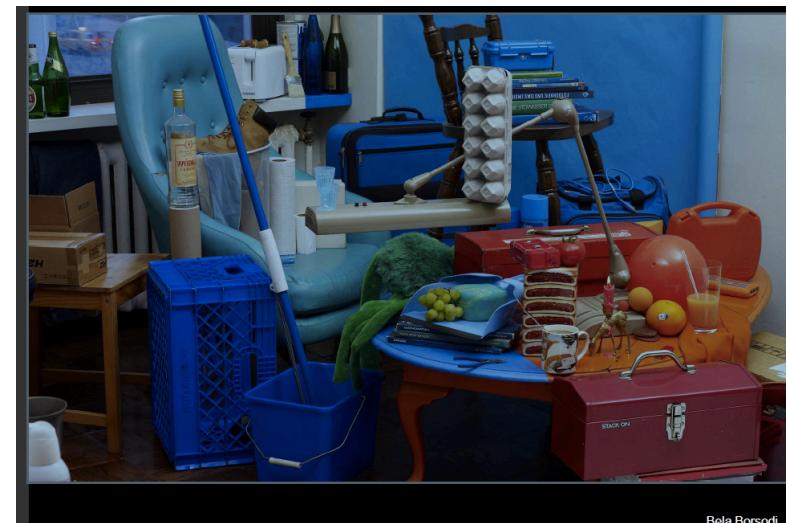
1 Pixel = 8 bits (UINT 8) = 1 Byte

```
>> whos myFirstImage
```

Name	Size	Bytes	Class
myFirstImage	526x764x3	1,205,592	uint8

myFirstImage worth maybe 1000 words

but costs much more ....



Bela Borsodi

# What is an Image?

An image  $I$  is a two dimensional (2D) function  
that maps the image domain  $\Omega$  to  $[0, 255]$

$$I: \Omega \rightarrow [0, 255]$$

or (for RGB)

$$I(\mathbf{x}) = I(x, y) = \vec{v}, \quad v \in [0, 255]^3$$

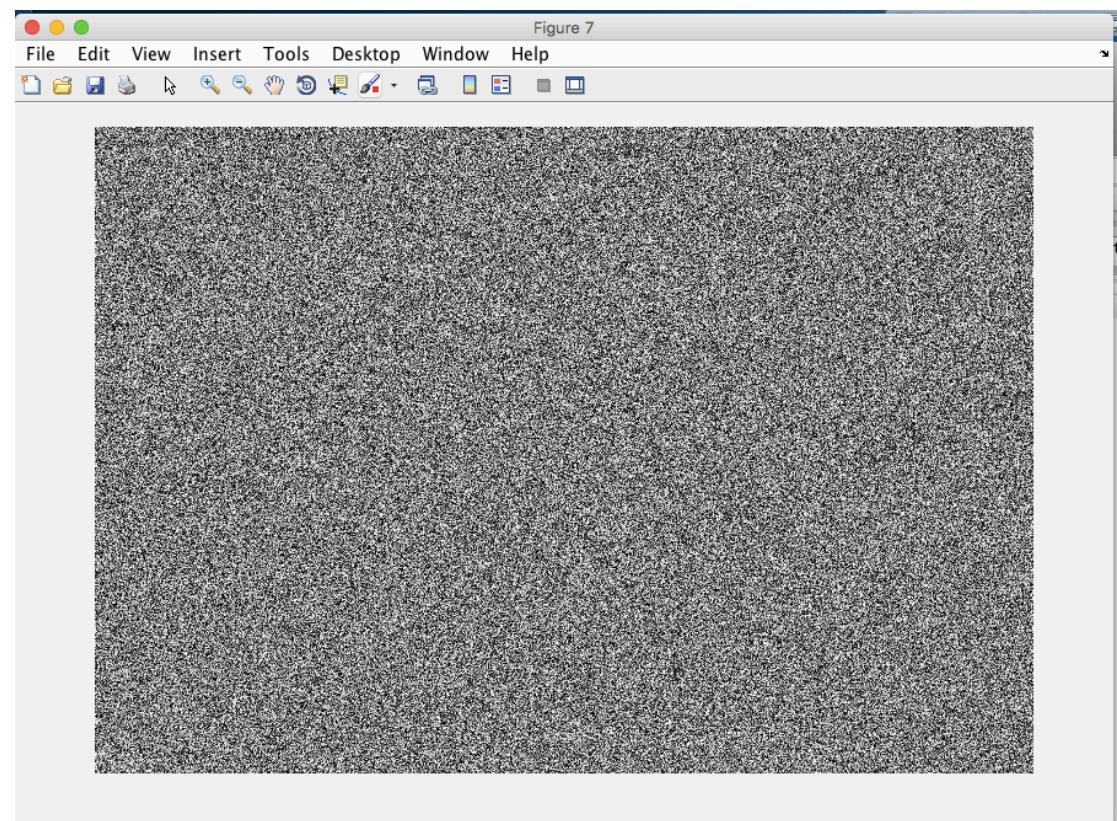
and the value of a single pixel is:

$$I(\mathbf{x}) = I(x, y) = v, \quad v \in [0, 255] \quad \text{Gray Level Pixel}$$

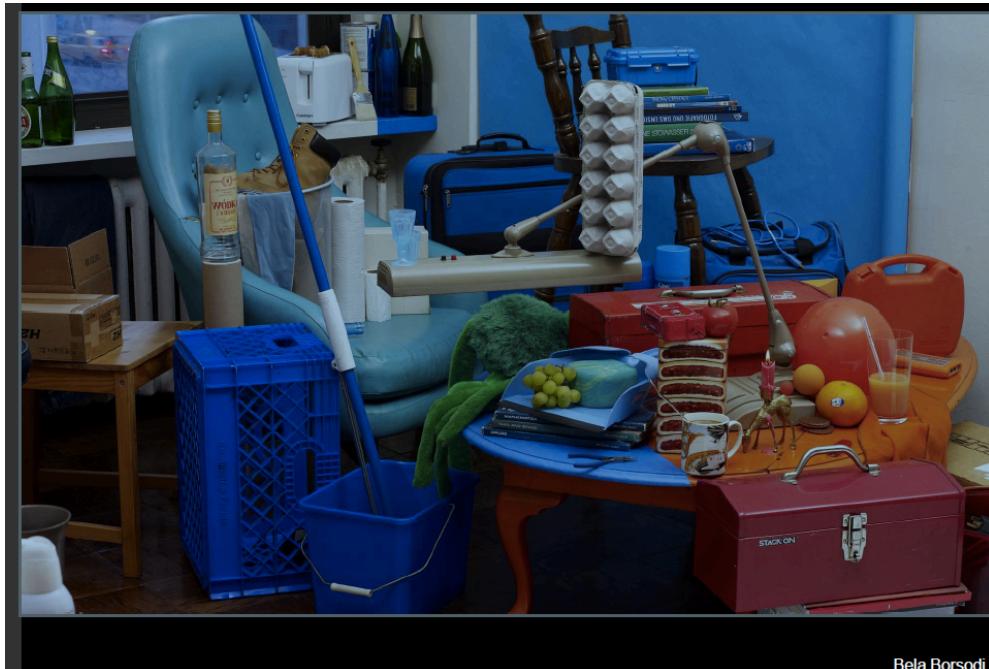
$$I(\mathbf{x}) = I(x, y) = (v_R, v_G, v_B) \quad \text{RGB pixel}$$

# Is this an image?

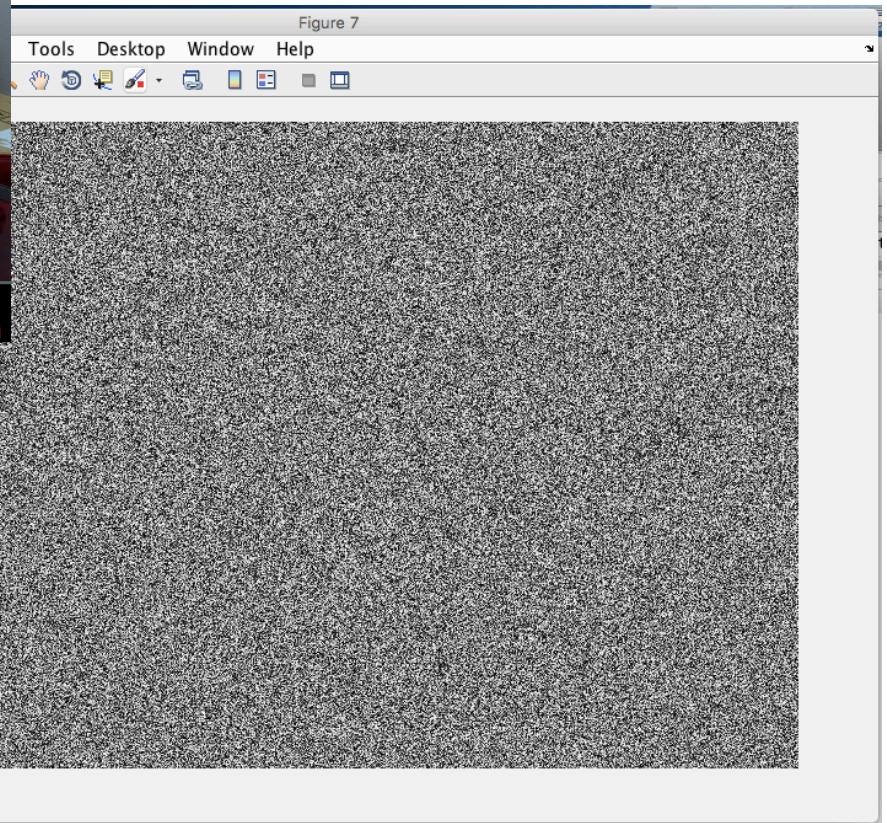
```
>>  
>> I = uint8(255*rand(526,764));  
>> I = uint8(255*rand(size(I1)));  
>> figure ; imshow(I);  
>>
```



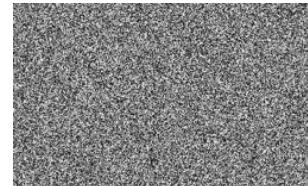
# What's the difference?



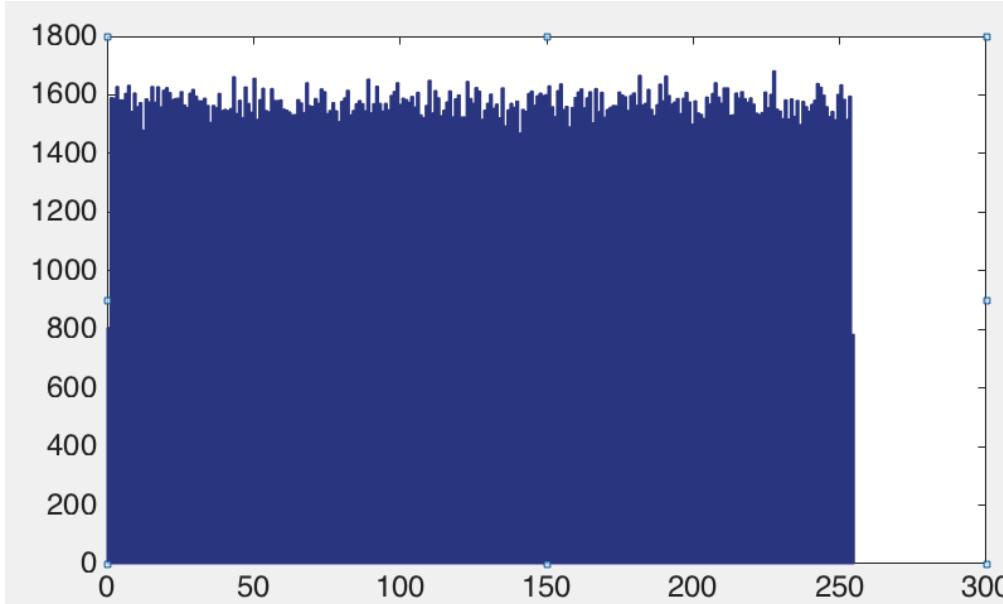
Bela Borsodi



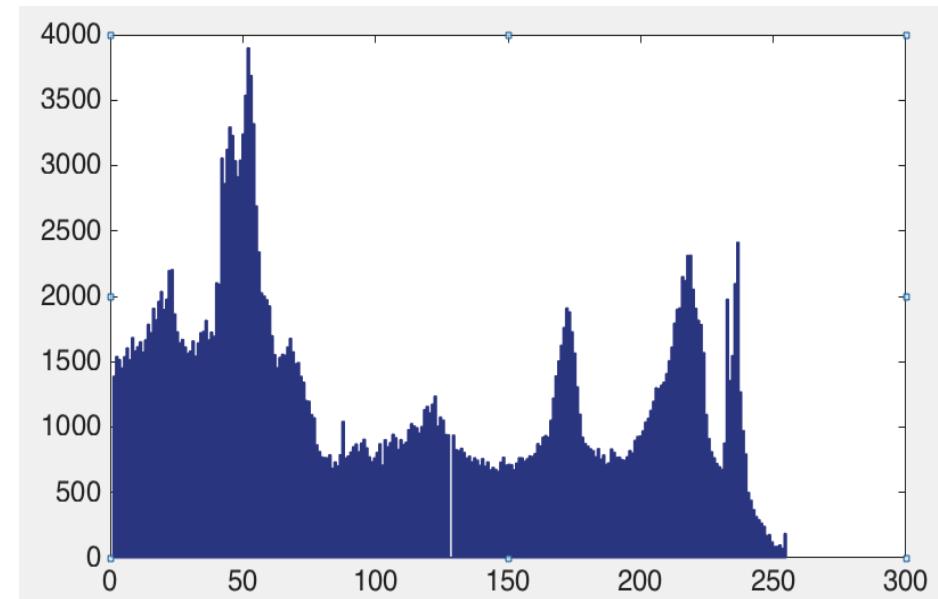
# Building an histogram



```
>>  
>> I=double(I);  
>> figure;hist(I(:,256);  
>>
```



```
>>  
>> I1 = double(I1);  
>> figure;hist(I1(I1~=0),256)  
>>
```

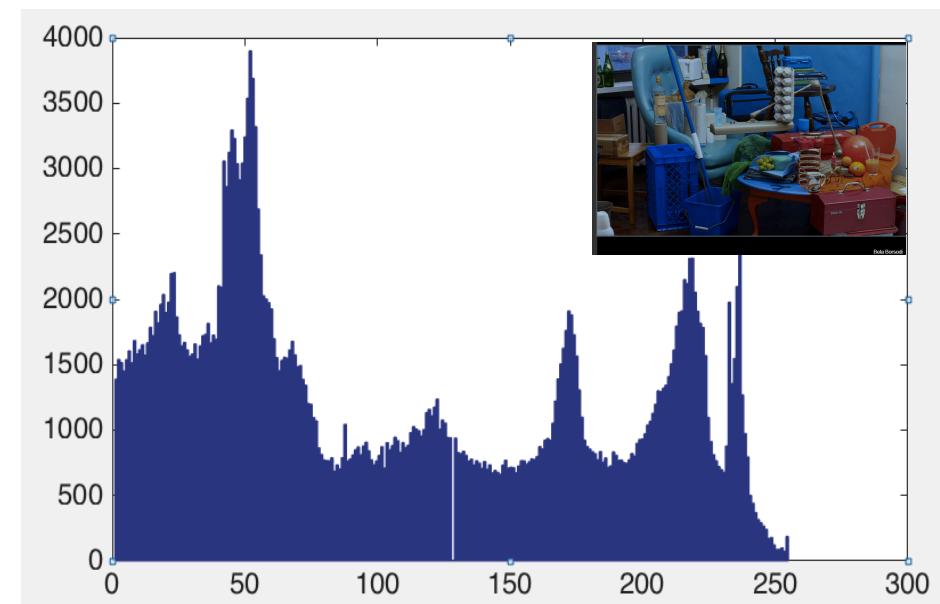
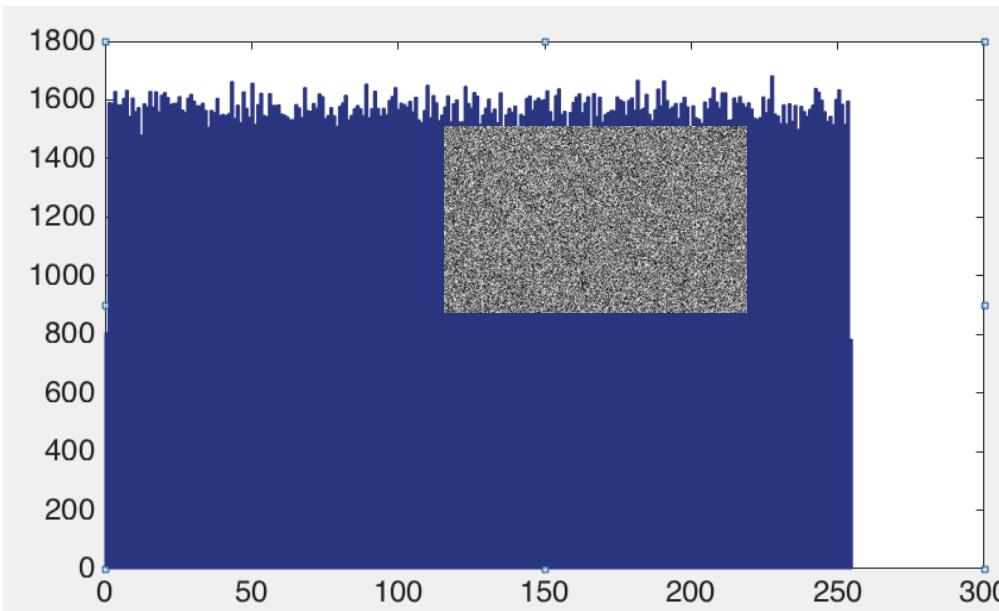


# Can we measure the differences?



```
>>  
>> I=double(I);  
>> figure;hist(I(:,256);  
>>
```

```
>> I1 = double(I1);  
>> figure;hist(I1(I1~=0),256)  
>>
```



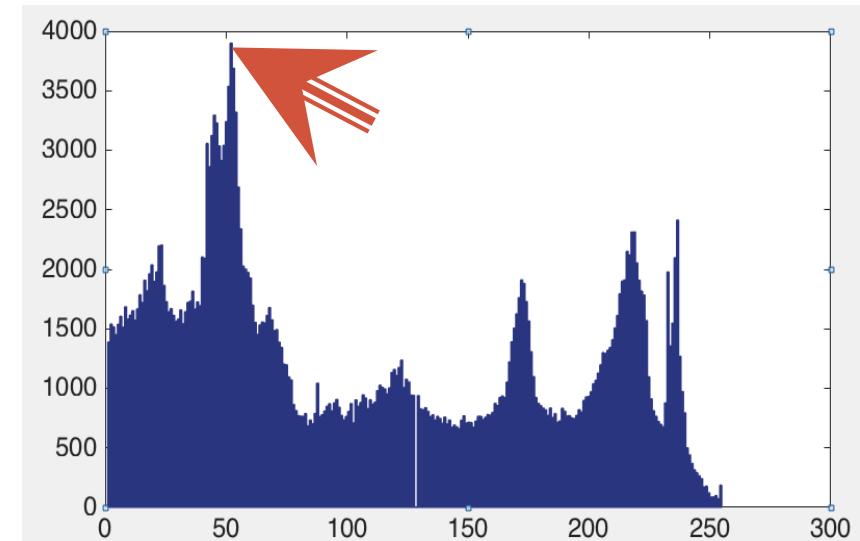
# A bit on information theory (in the context of images)

{0, 1}

Choose an arbitrary pixel in an image,  
can you guess its value?



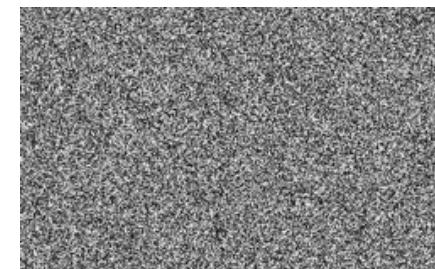
Well, we can build an histogram and gamble on the value with the highest frequency



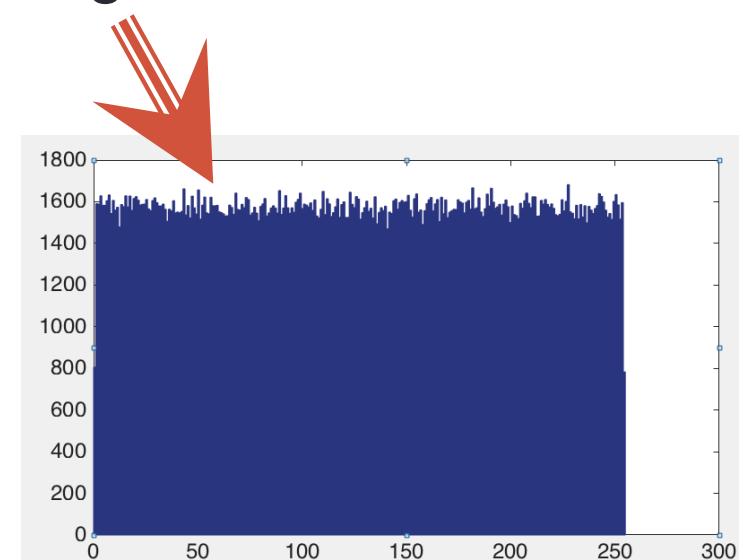
# A bit on information theory

{0, 1}

Choose an arbitrary pixel in an image,  
can you guess its value?



Well, we can build an histogram and gamble on the value with the highest frequency



# A bit on information theory (in the context of images)

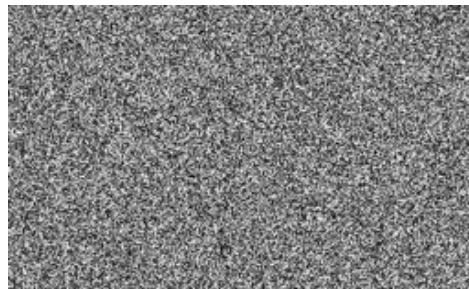
By normalizing an histogram, one can get the probability  $p_i$  for the occurrence of the i-th value.  
The **Shannon entropy** (measured in bits) is given by:

$$H = - \sum_i p_i \log_2(p_i)$$

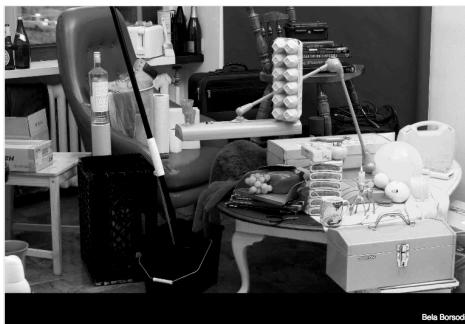
where  $-\log_2(p_i)$  is the self-information,  
which is the entropy contribution of an individual pixel.

# Entropy of an Image

What does it mean ? Does it mean anything?



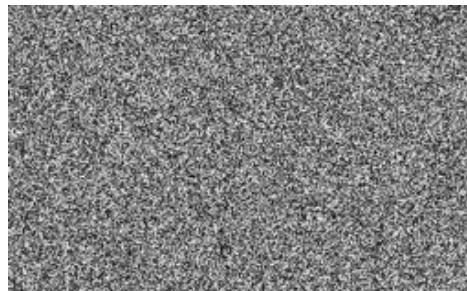
Entropy = 7.98



Entropy = 6.98

# Entropy of an Image

What does it mean ? Does it mean anything?



Entropy = 7.98



Entropy = 6.98



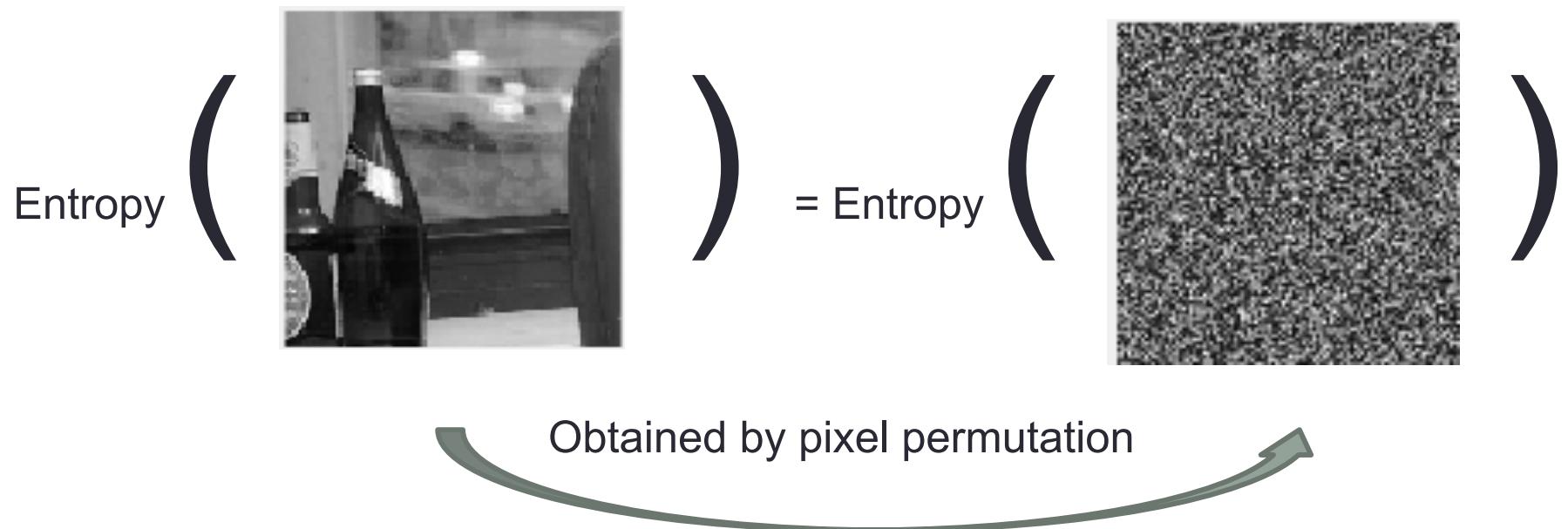
Entropy = 0

But

```
>>
>> Isame = uint8(100*ones(size(I1)));
>> figure; imshow(Isame)
```

# Entropy of an Image

What does it mean ? Does it mean anything?



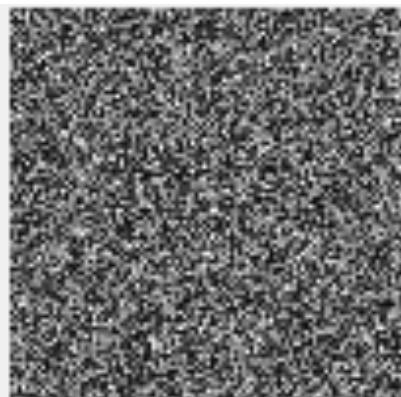
# Entropy of an Image

What does it mean ? Does it mean anything?

SmallI1



SmallI1im



```
>>
>> myFirstImage = imread('someImage.png');
>> I1 = myFirstImage(:,:,1);
>> smallI1 = I1(1:100,1:100);
>> figure;imshow(smallI1)
>> randOrd = randperm(numel(smallI1));
>> permSI1 = smallI1(randOrd);
>> whos
      Name          Size            Bytes  Class
      I1            526x764        401864  uint8
      myFirstImage   526x764x3    1205592  uint8
      permSI1       1x10000        10000  uint8
      randOrd       1x10000        80000  double
      smallI1       100x100         10000  uint8

>> permSI1im = reshape(permSI1,size(smallI1));
>> figure;imshow(permSI1im)
```

# Next-door neighbors



Bela Borsodi



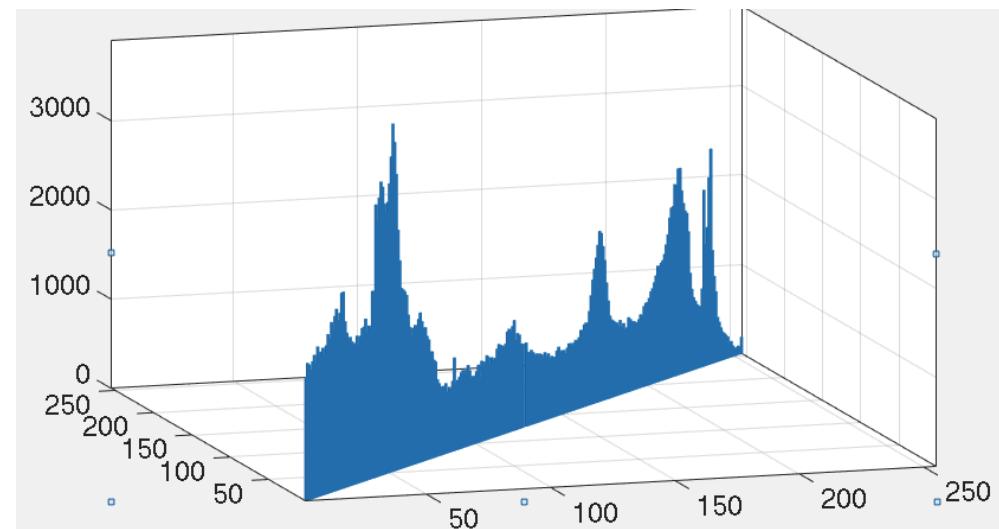
Download from  
Dreamstime.com

3541890  
Nikha | Dreamstime.com

You have chosen a pixel and you know its value.

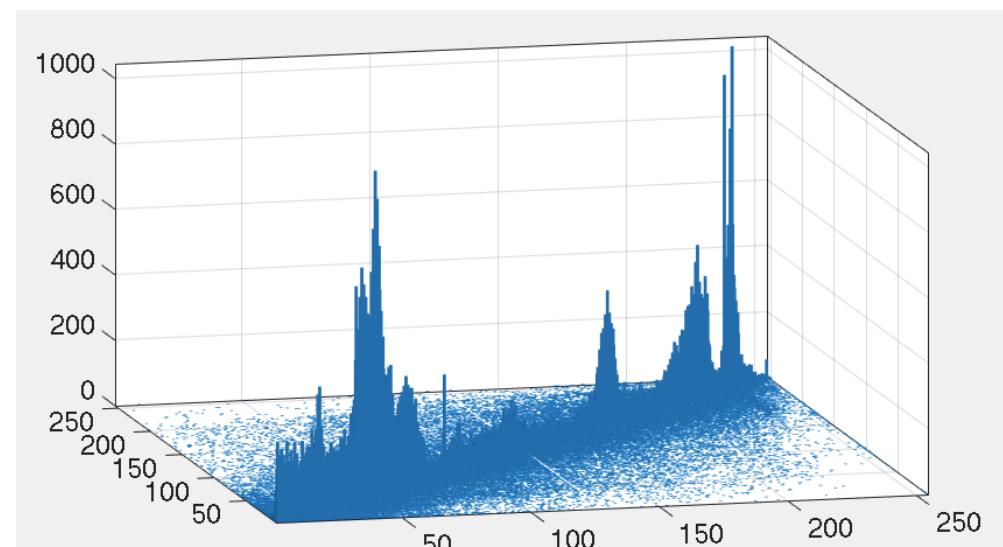
What can you say about the value of its next-door neighbor?

# Next-door neighbors



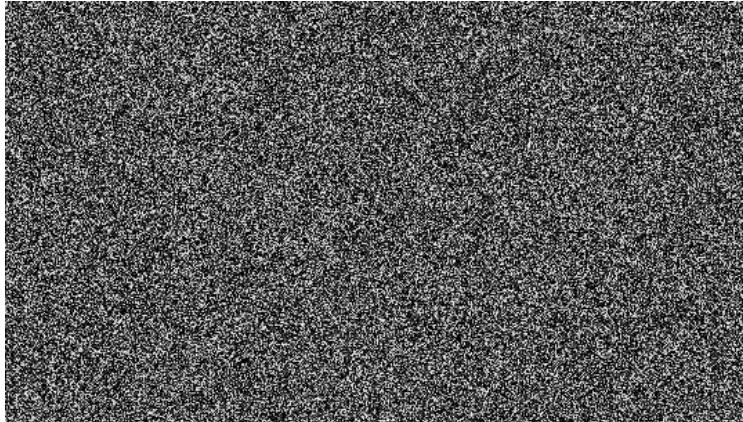
```
figure;histogram2(I1(I1~=0),I1(I1~=0),256)
```

# Next-door neighbors

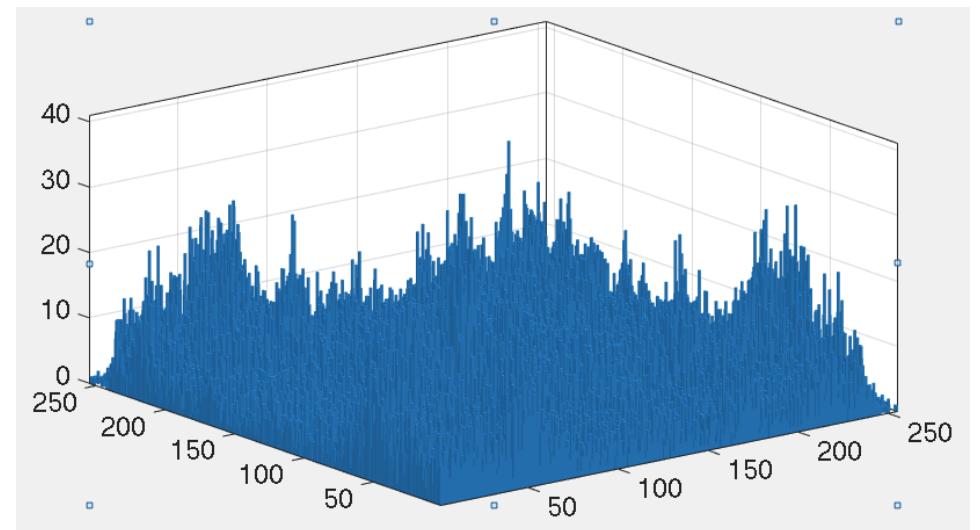


```
>> figure;histogram2(I1L(I1L~=0 & I1R~=0),I1R(I1L~=0 & I1R~=0),256)
```

# Next-door neighbors



Random permutation of the  
**red channel** of myFirstImage



```
>> figure; histogram2(permI1L(permI1L~=0 & permI1R~=0),permI1R(permI1L~=0 & permI1R~=0),256)
```

# Mutual Information

(a little bit more on information theory)

- The Mutual Information of two random variables is a measure of the variables' mutual dependence.
- The most common unit of measurement of mutual information is the bit.

# Mutual Information

(a little bit more on information theory)

- The Mutual Information of two random variables is a measure of the variables' mutual dependence.

$$\mathcal{I}(X; Y) = \sum_{y \in Y} \sum_{x \in X} p(x, y) \log \left( \frac{p(x, y)}{p(x)p(y)} \right)$$

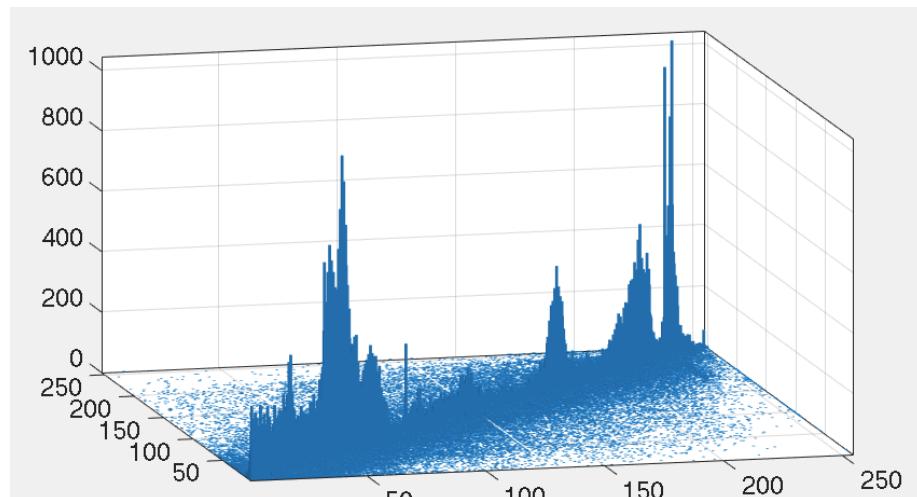
$p(x, y)$  is the joint probability function of  $X$  and  $Y$ .

$p(x), p(y)$  are the marginal probability distribution functions of  $X$  and  $Y$  (respectively).

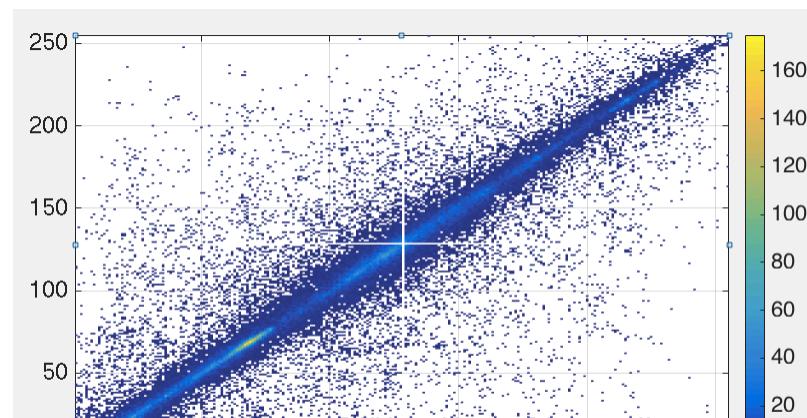
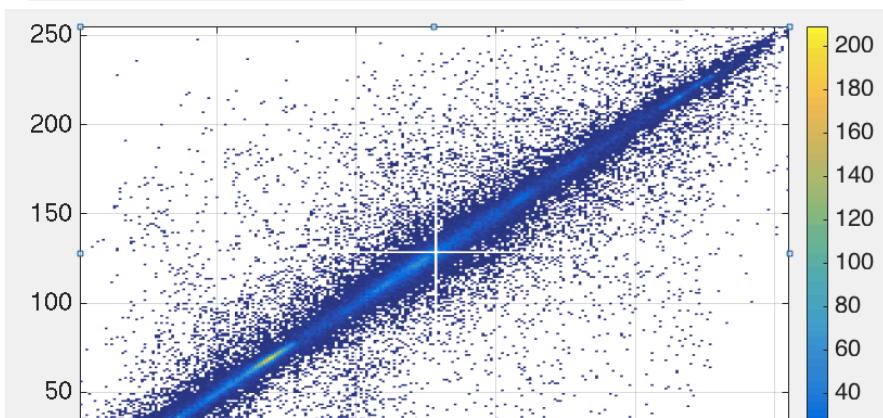
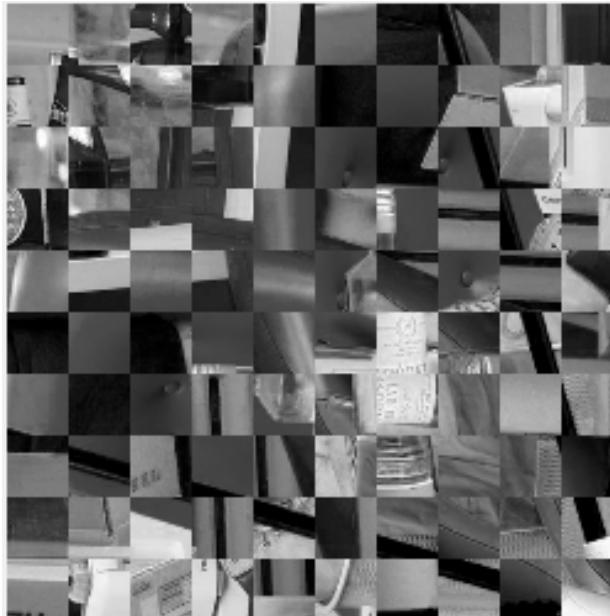
# Mini-assignment #1 (bonus)

- Read an image (any image)
- Present one of its RGB channels –  $I_1$
- Permute  $I_1$  and present it.
- Present the histogram of  $I_1$ .
- Calculate its entropy
- Calculate the Mutual Information between  $I_1$  pixels and their respective left-neighbors
- Calculate the Mutual Information between the permutation image's pixels and their respective left-neighbors

# Is it enough?

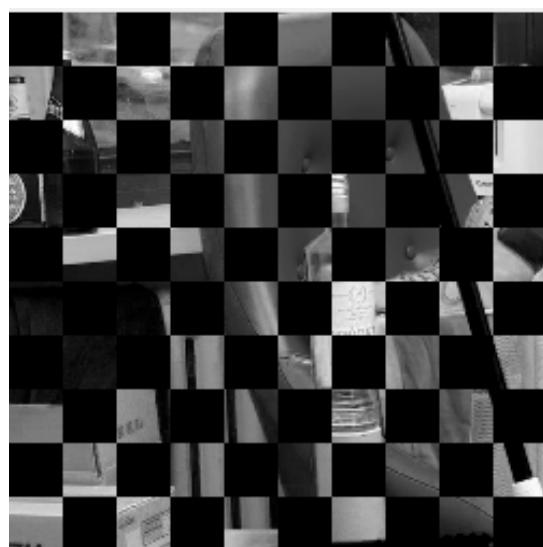
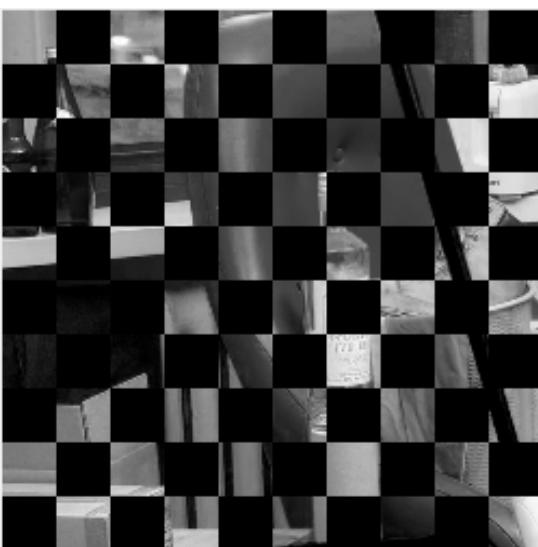
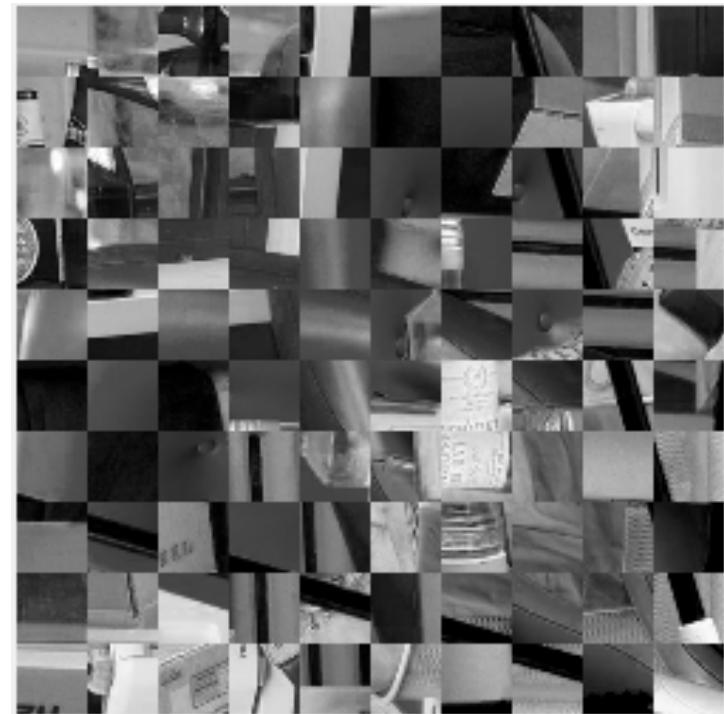
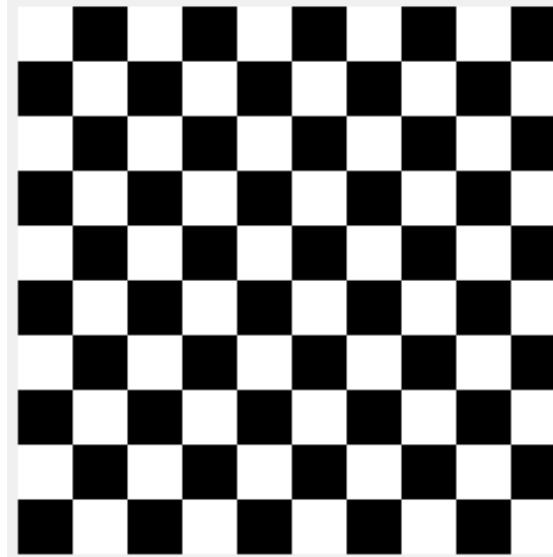


# Is it enough?

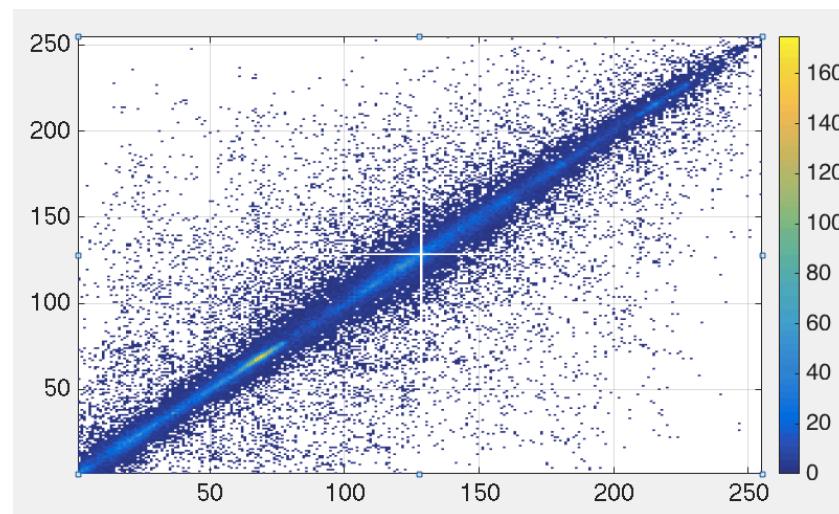
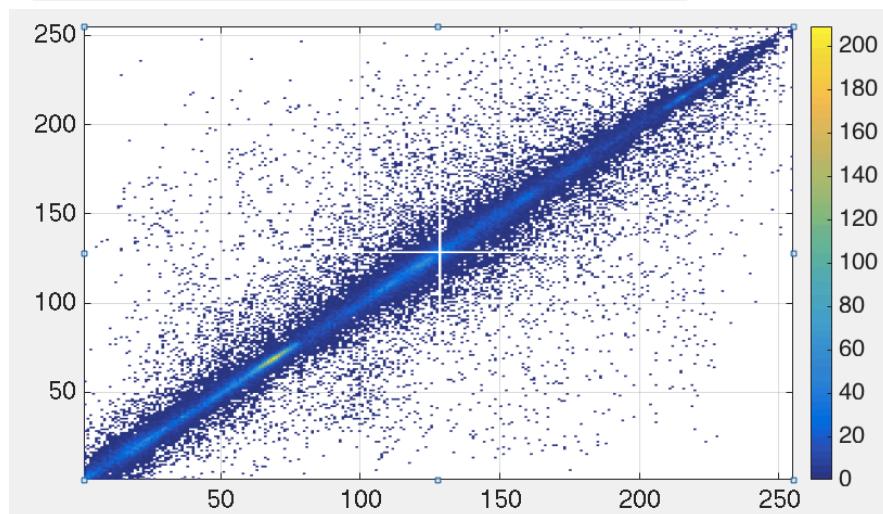


```
>> figure; histogram2(I1L(I1L~=0 & I1R~=0),I1R(I1L~=0 & I1R~=0),256,'DisplayStyle','tile')  
>> colorbar
```

# Is it enough?



# not yet there



# Next Class

Sampling

Quantization

Histogram Processing

