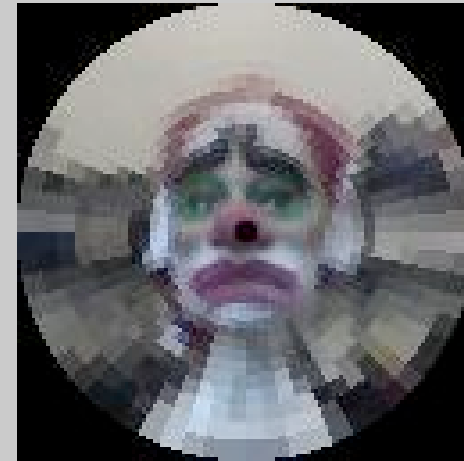
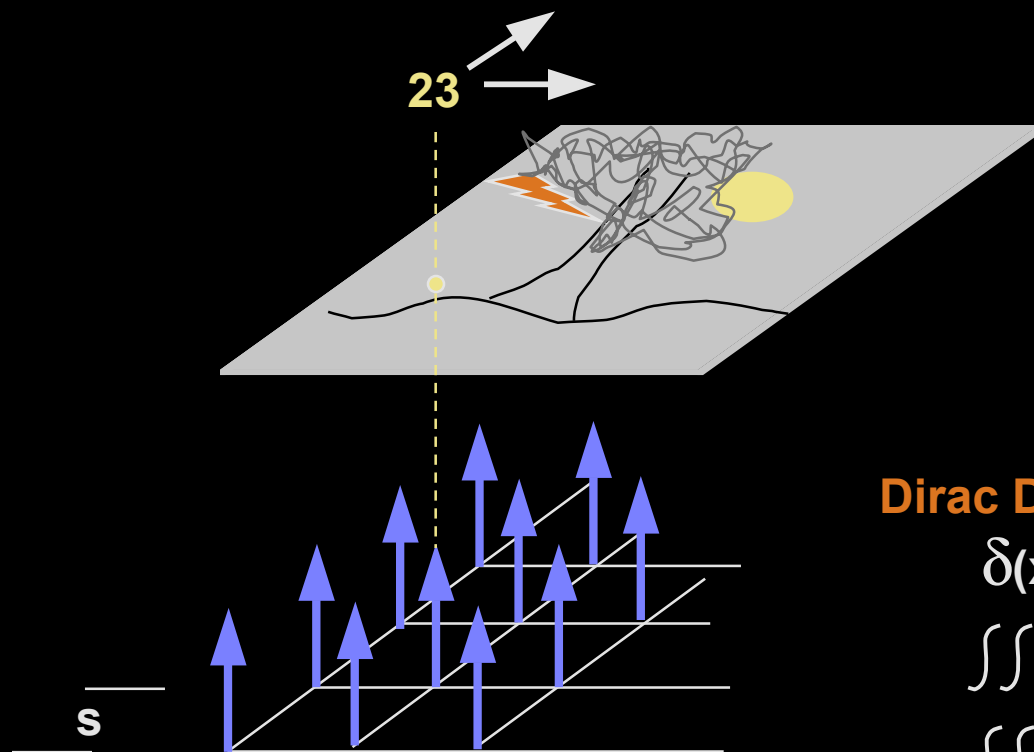


Cartesian image ----- Log-Polar representation ----- Retinal representation



## ■ Rough Idea: Ideal Case



"Digitized Image"

"Continuous Image"

Dirac Delta Function 2D "Comb"

$$\delta(x,y) = 0 \text{ for } x \neq 0, y \neq 0$$

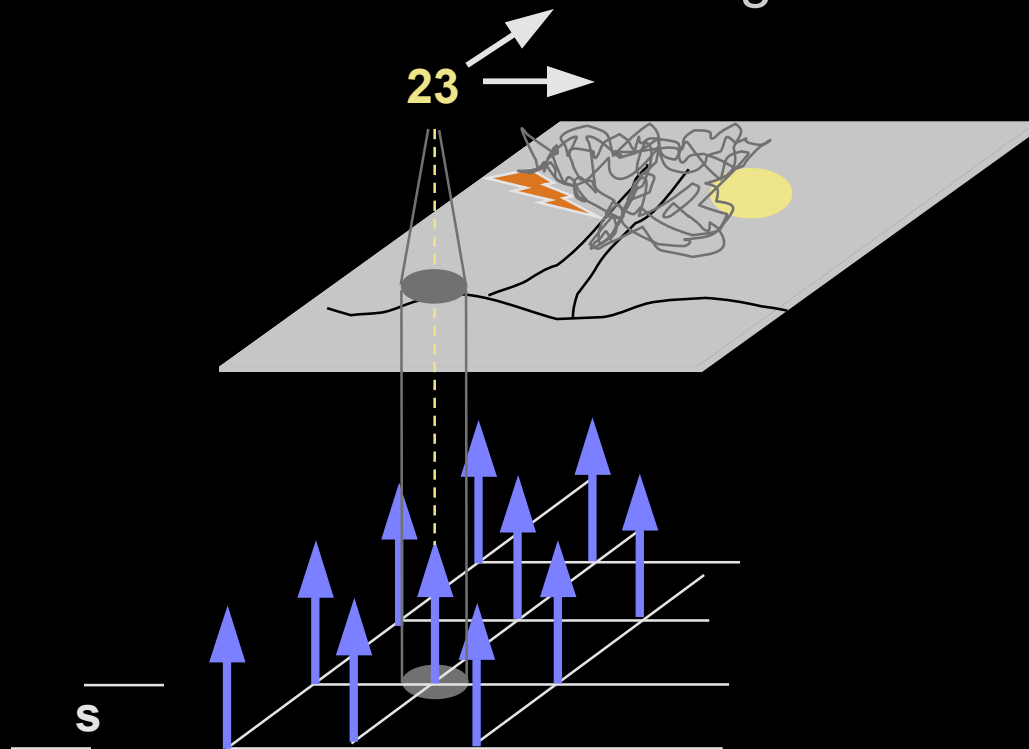
$$\iint \delta(x,y) dx dy = 1$$

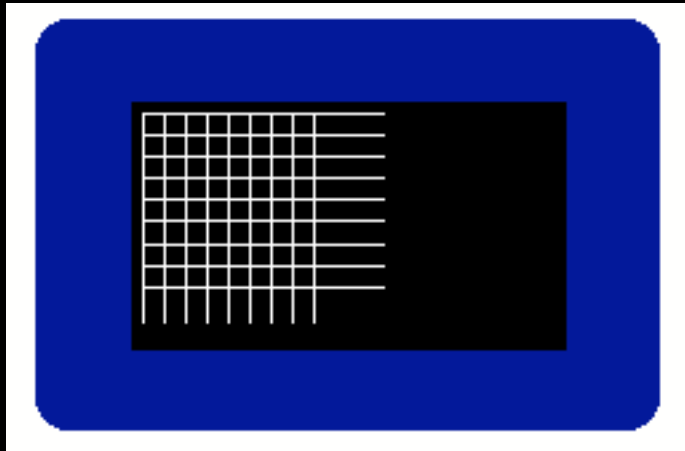
$$\iint f(x,y) \delta(x-a,y-b) dx dy = f(a,b)$$

$$\delta(x-ns,y-ns) \text{ for } n = 1 \dots 32 \text{ (e.g.)}$$

## ■ Rough Idea: Actual Case

- Can't realize an ideal point function in real equipment
- "Delta function" equivalent has an area
- Value returned is the average over this area





Digitized 35mm Slide or Film

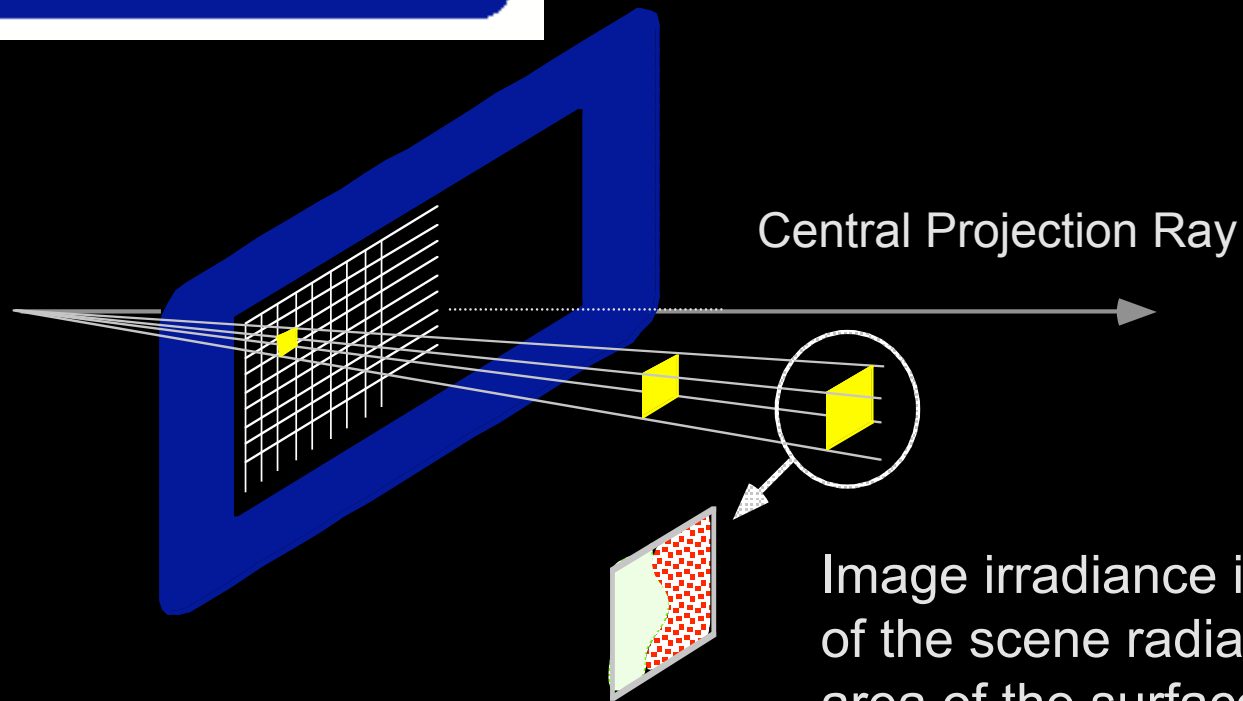
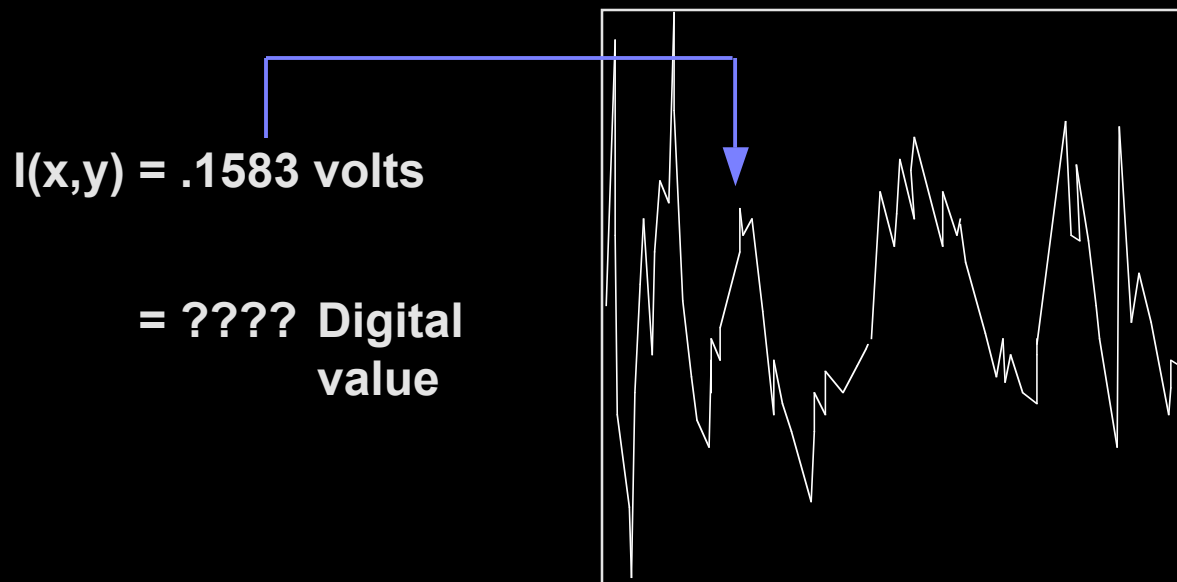


Image irradiance is the average of the scene radiance over the area of the surface intersecting the solid angle!



- Goal: determine a mapping from a continuous signal (e.g. analog video signal) to one of  $K$  discrete (digital) levels.



- $I(x,y)$  = continuous signal:  $0 \leq I \leq M$
- Want to quantize to K values  $0, 1, \dots, K-1$
- K usually chosen to be a power of 2:

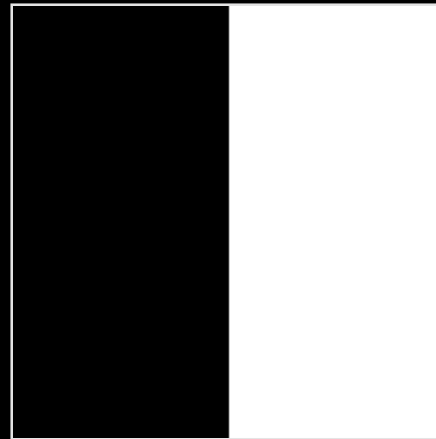
K: #Levels	#Bits
2	1
4	2
8	3
16	4
32	5
64	6
128	7
256	8

- Mapping from input signal to output signal is to be determined.
- Several types of mappings: uniform, logarithmic, etc.

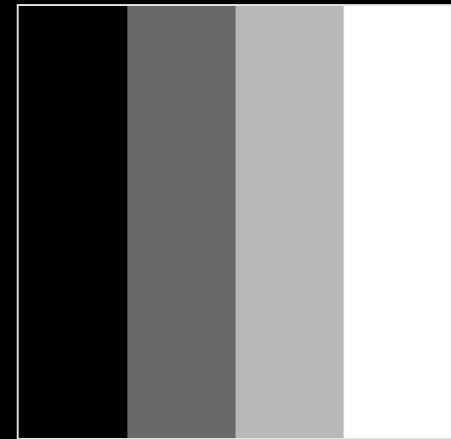
Original



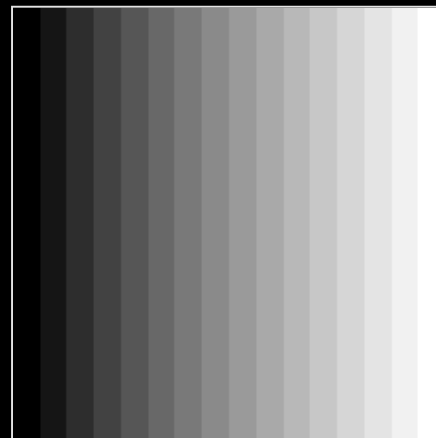
Linear Ramp



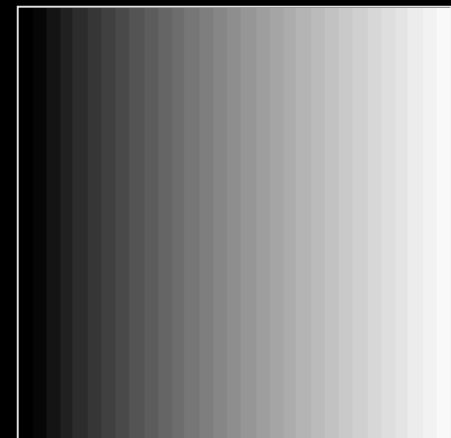
K=2



K=4

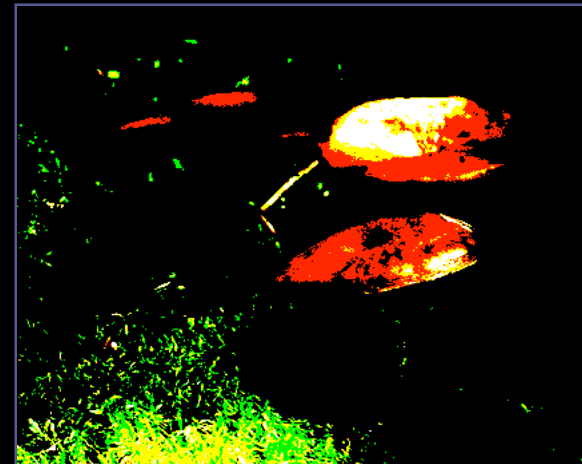


K=16



K=32

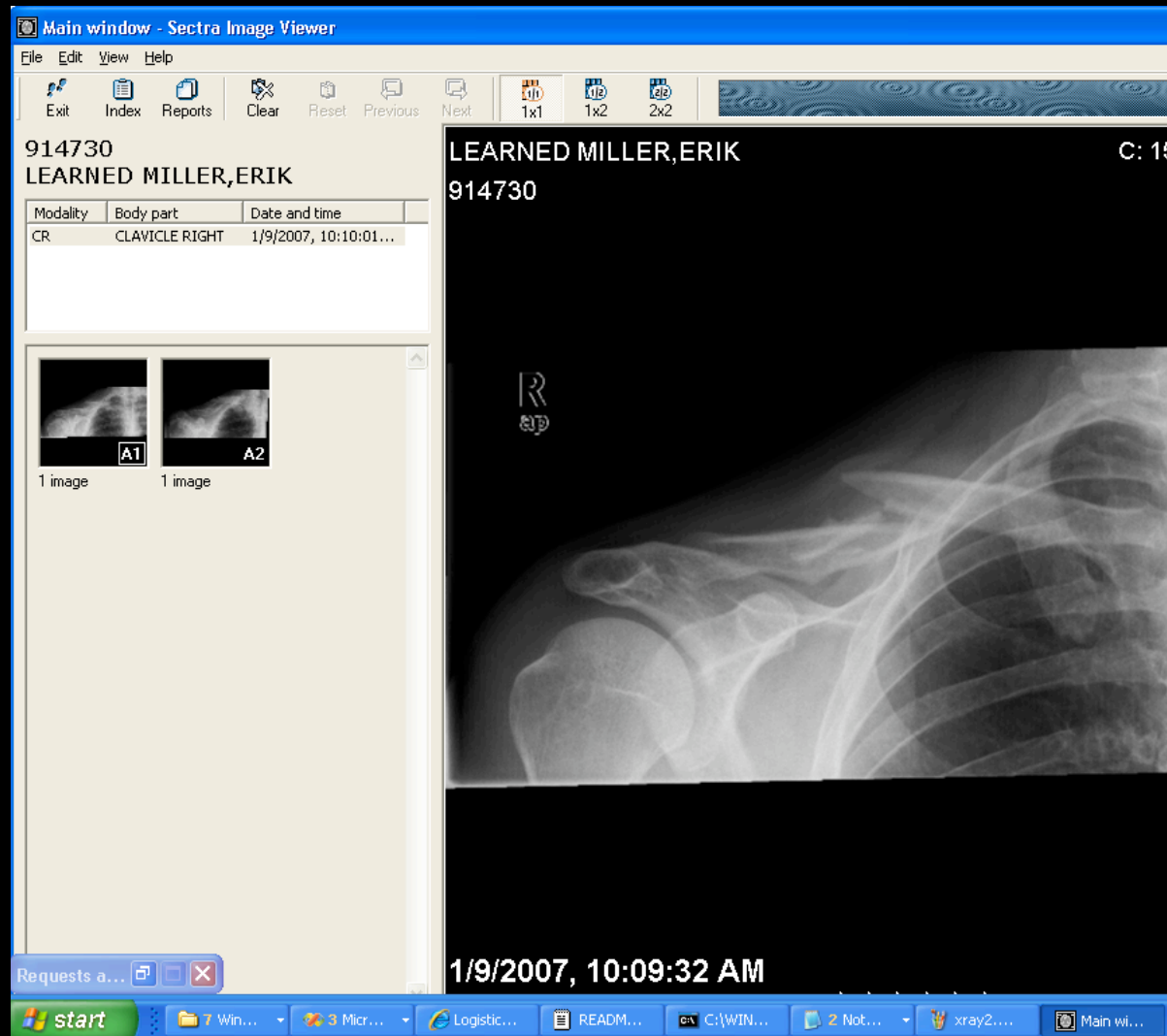




**K=2 (each color)**



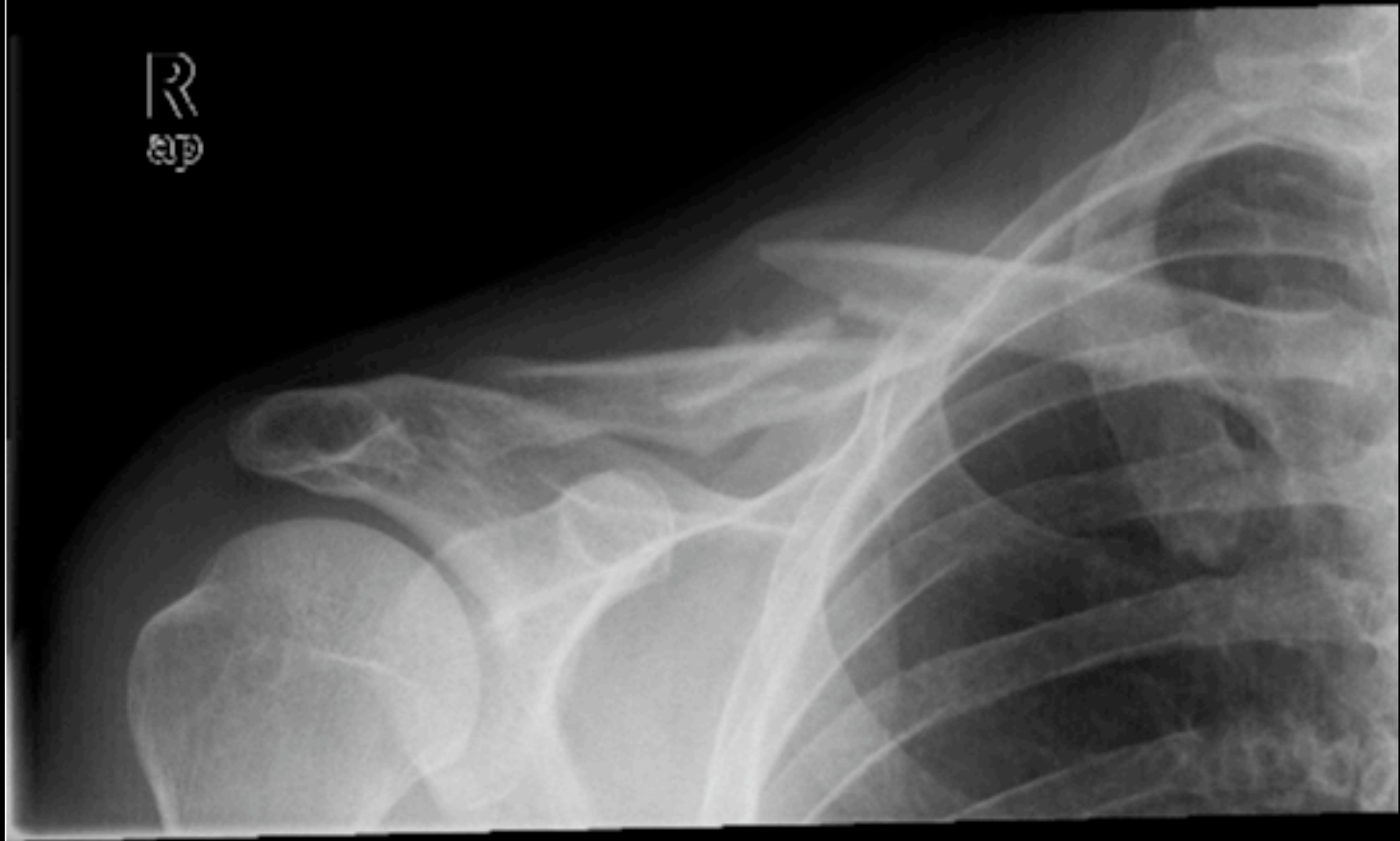
**K=4 (each color)**

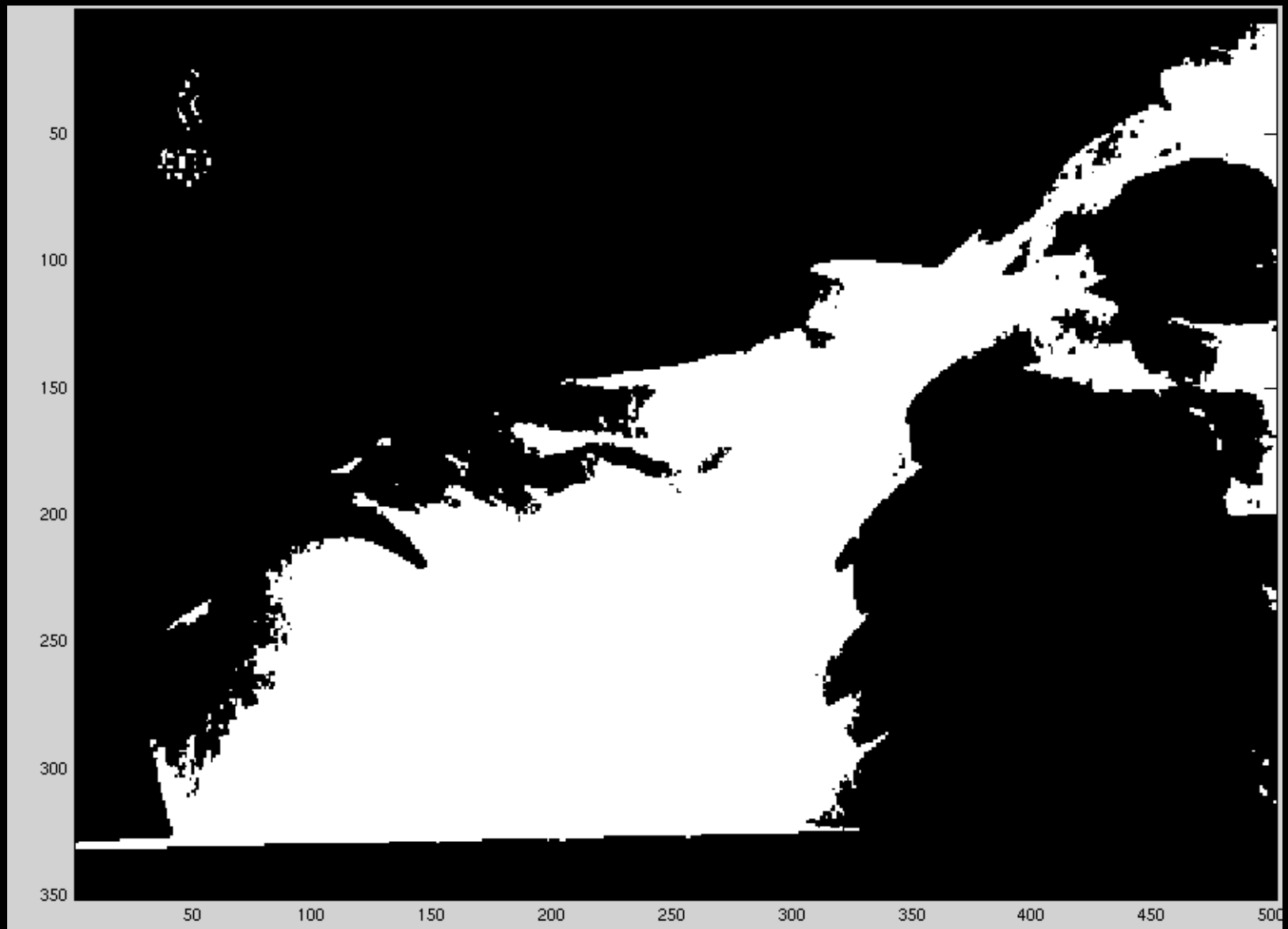


Introduction to

Computer Vision

# Digital X-rays: 8 is enough?

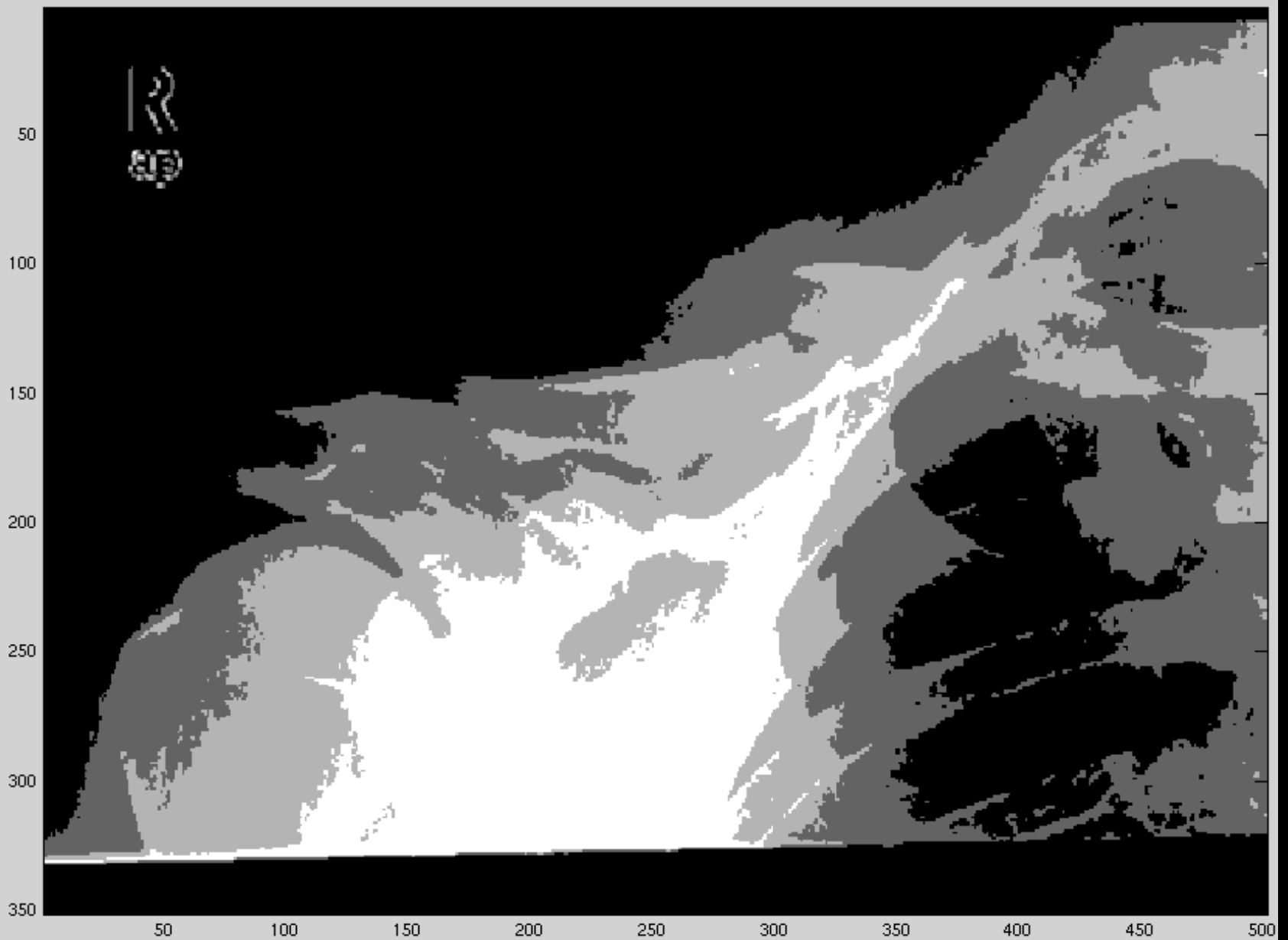




Introduction to

Computer Vision

# Digital X-rays: 2 bits





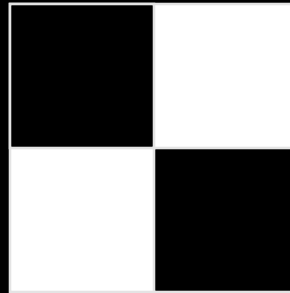
Introduction to

Computer Vision

# Digital X-rays: 8 is enough?

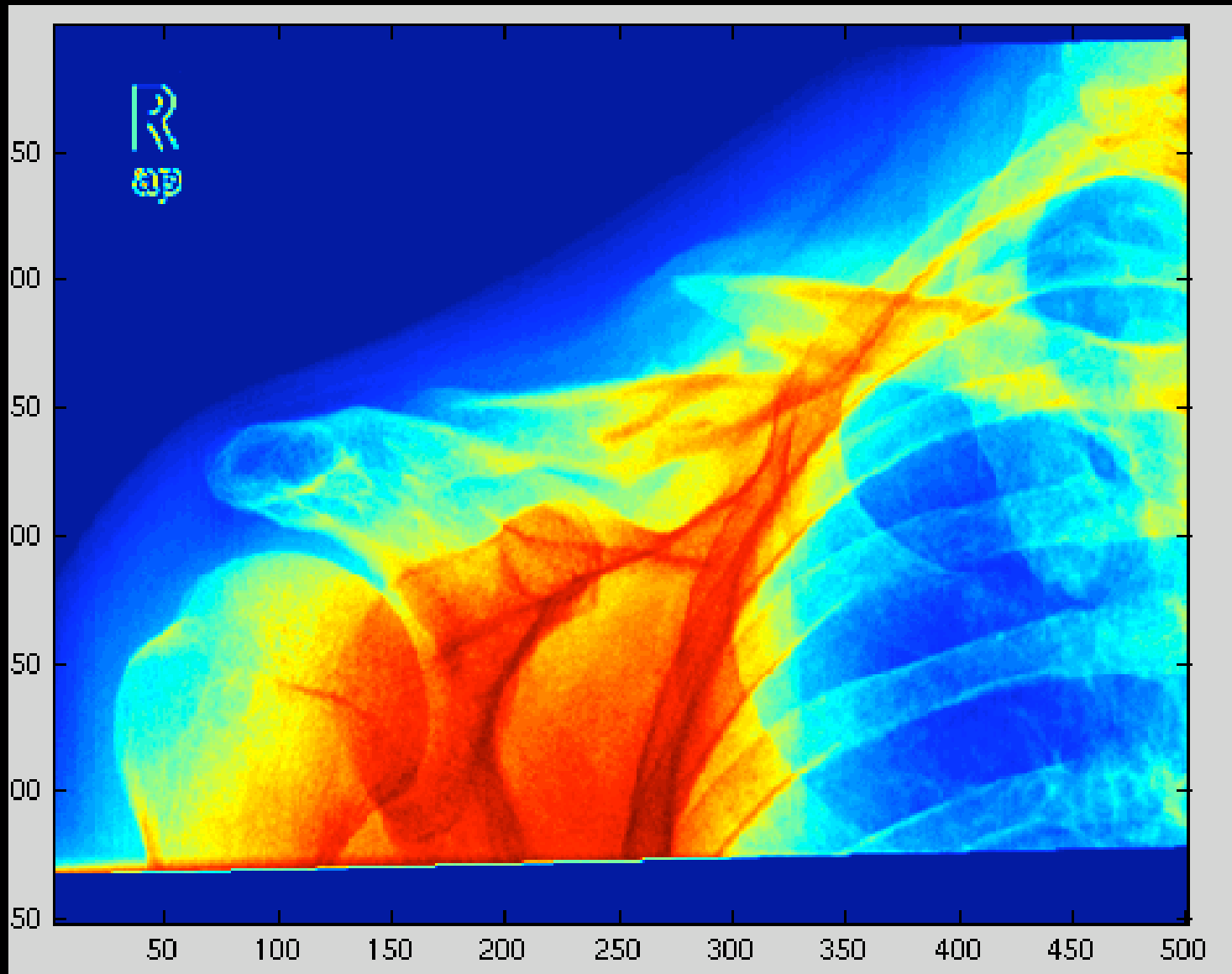


- More gray levels can be simulated with more resolution.
- A “gray” pixel:



- Doubling the resolution in each direction adds at least four new gray levels. But maybe more?



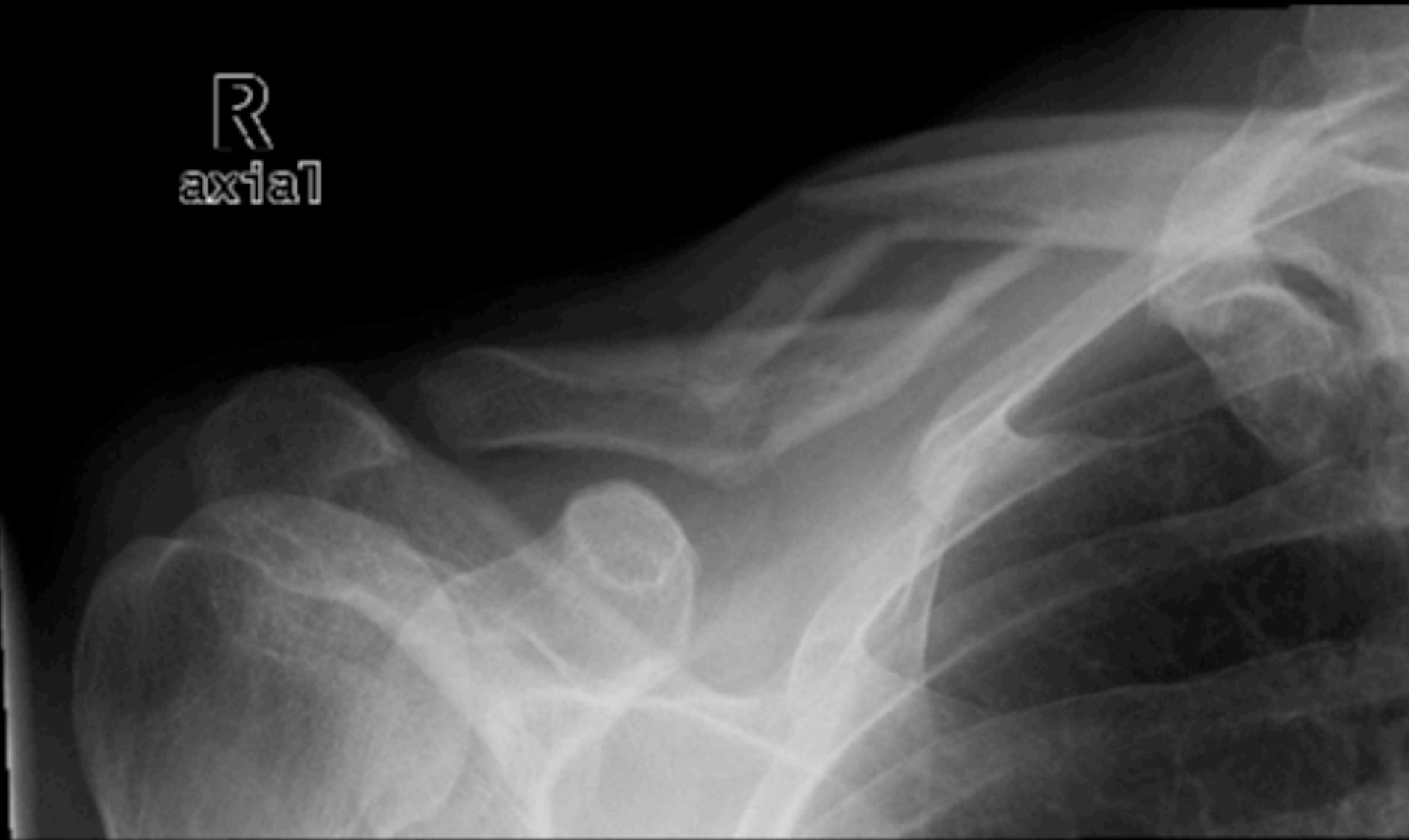


Introduction to

Computer Vision

# Digital X-rays: 8 is enough?

R  
axial



Introduction to

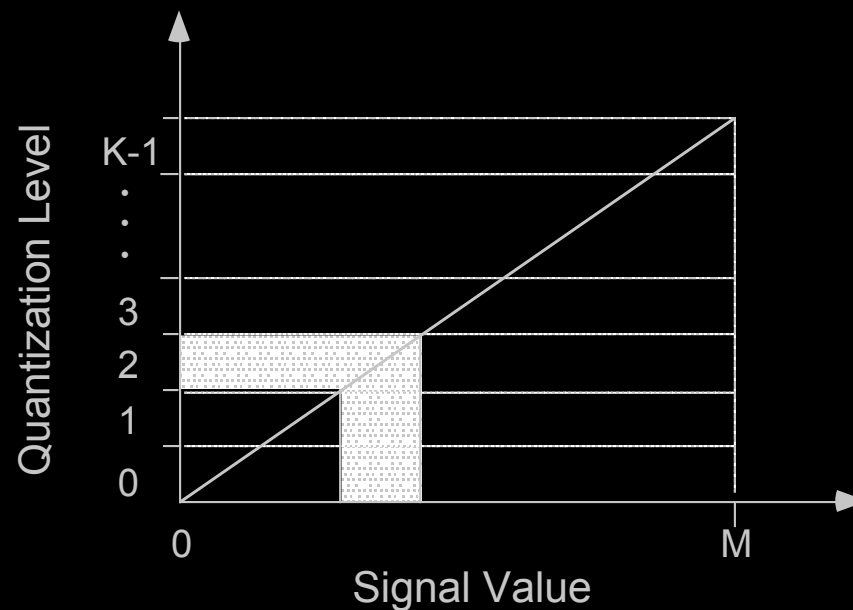
Computer Vision

MRI

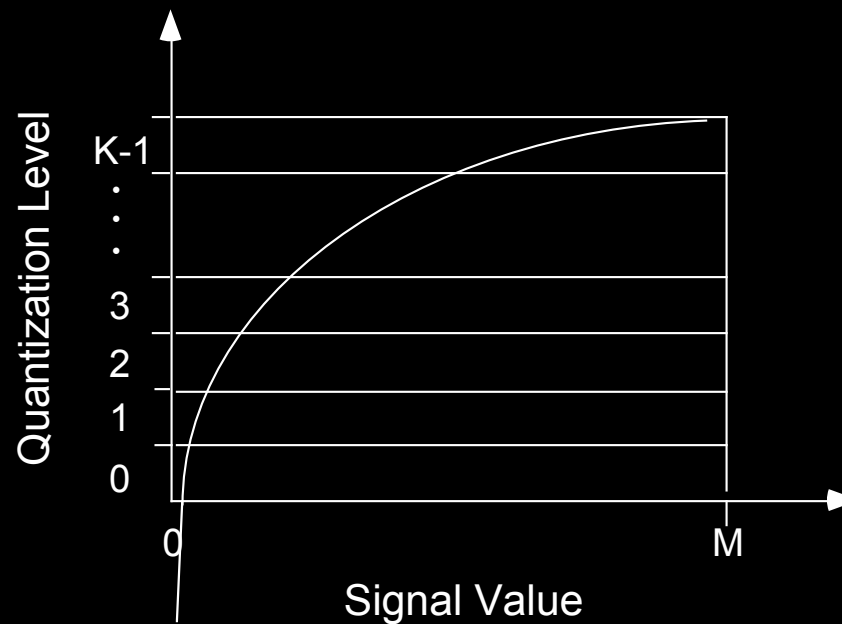


2. MRI Axi

- Uniform sampling divides the signal range  $[0-M]$  into  $K$  equal-sized intervals.
- The integers  $0, \dots, K-1$  are assigned to these intervals.
- All signal values within an interval are represented by the associated integer value.
- Defines a mapping:



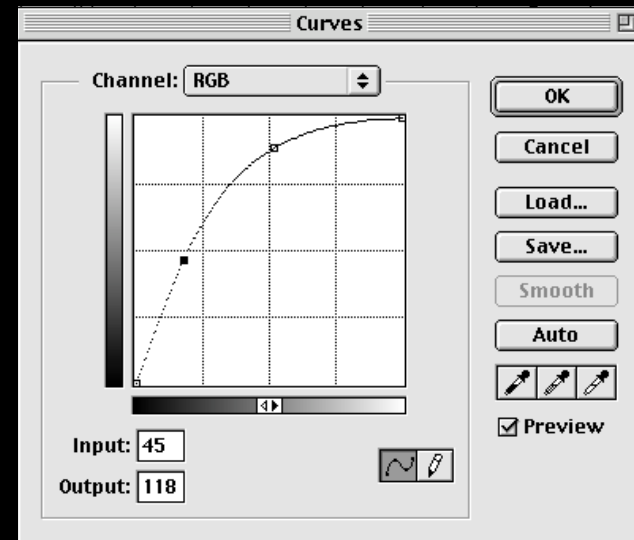
- Signal is  $\log I(x,y)$ .
- Effect is:



- Detail enhanced in the low signal values at expense of detail in high signal values.

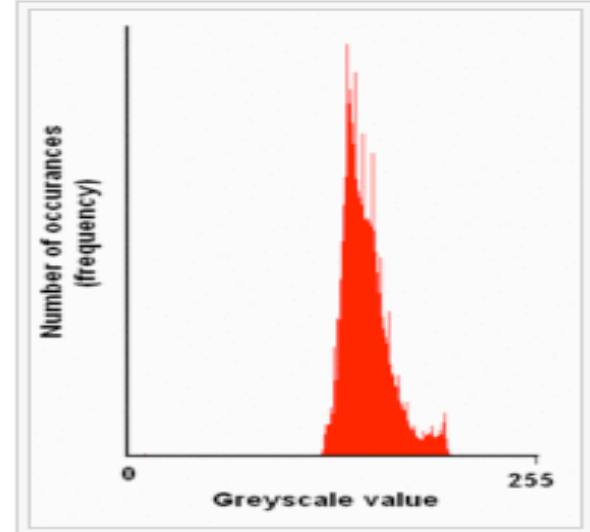
# Logarithmic Quantization

## Quantization Curve





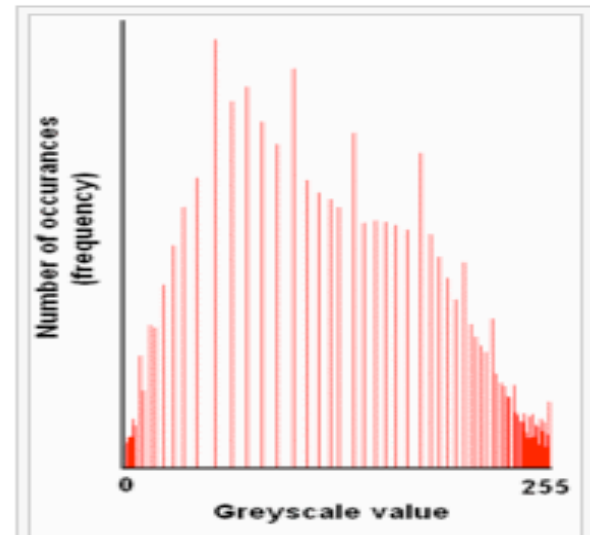
An unequalized image



Corresponding histogram



Same image after histogram equalization



Corresponding histogram

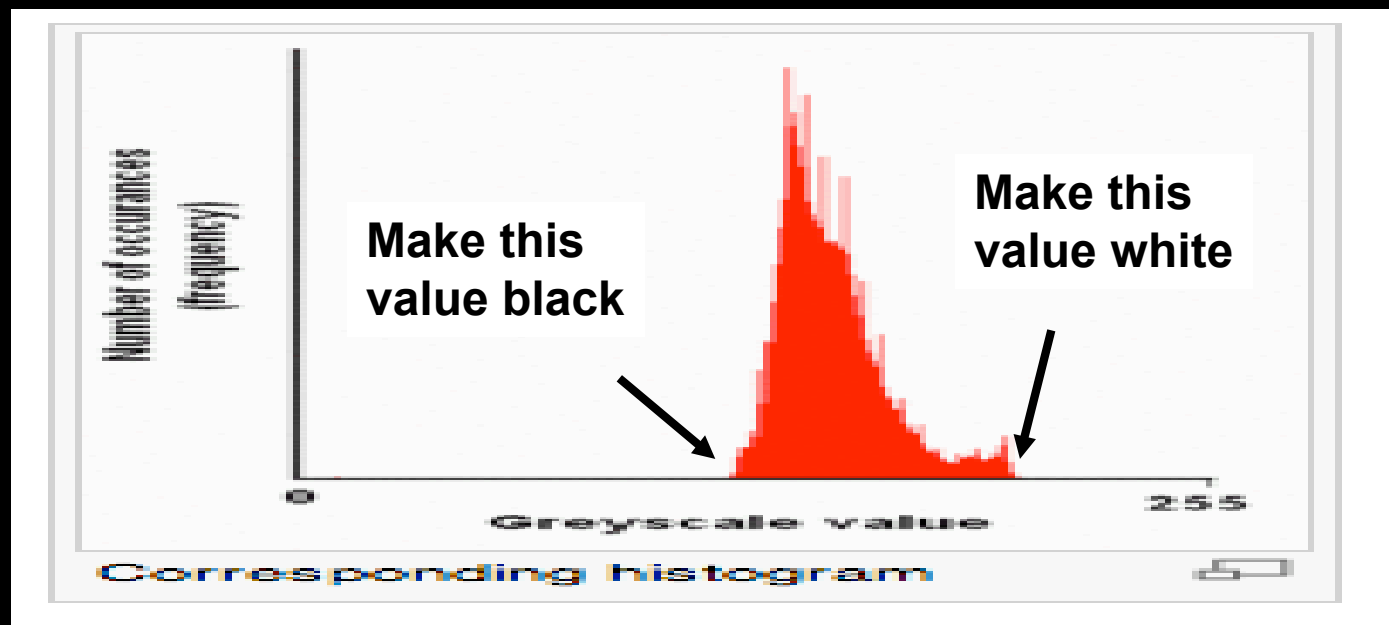
- Two methods:
  - Change the data (histogram equalization)
  - Use a look up table (brightness or color remapping)



Maps Brightness Value -> RGB Color

- 0 -> (1, 0, 0)
- 1 -> (0, 1, 0)
- 2 -> (0, 0, 1)
- 3 -> (0, 1, 1)
- ...
- 255 -> (1, 1, 1)

- Two methods:
  - Change the data.
  - Use a look up table.



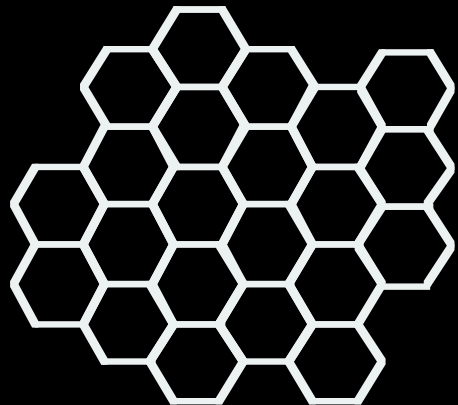
Maps Brightness Value -> RGB Color

- 0 -> (0, 0, 0)
- 1 -> (0, 0, 0)
- 2 -> (0, 0, 0)
- 3 -> (0, 0, 0)
- ...
- 130 -> (0,0,0)
- 131 -> (.01, .01, .01)
- 132 -> (.02,.02,.02)
- ...
- 200 -> (1,1,1)
- 201 -> (1,1,1)
- ...
- 255 -> (1, 1, 1)

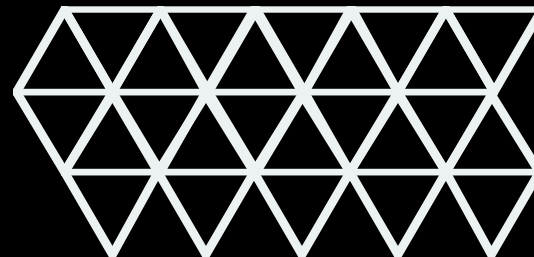


An unequalized image

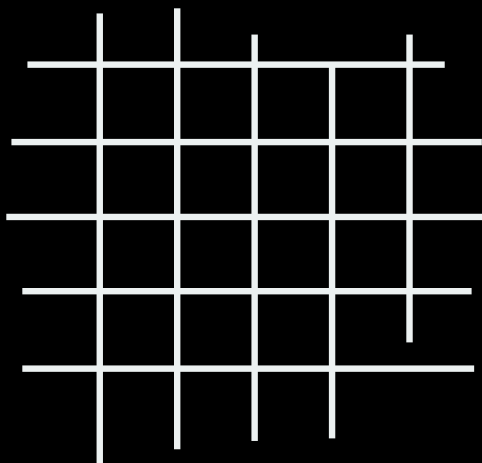




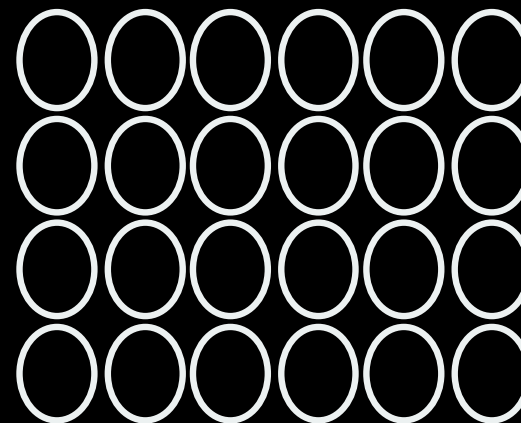
Hexagonal



Triangular

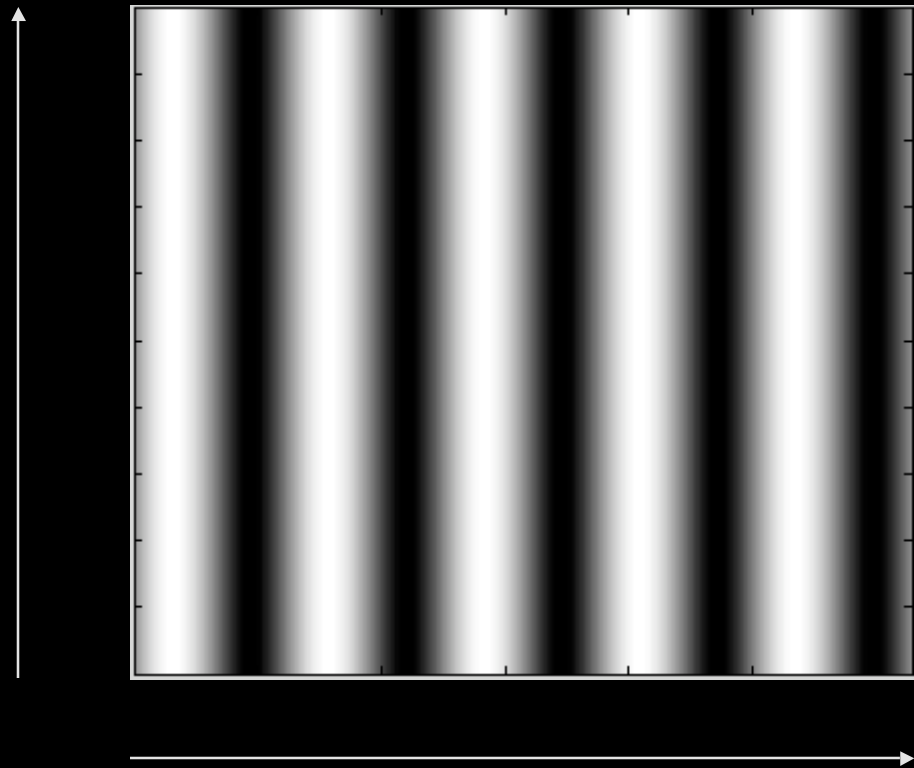


Rectangular



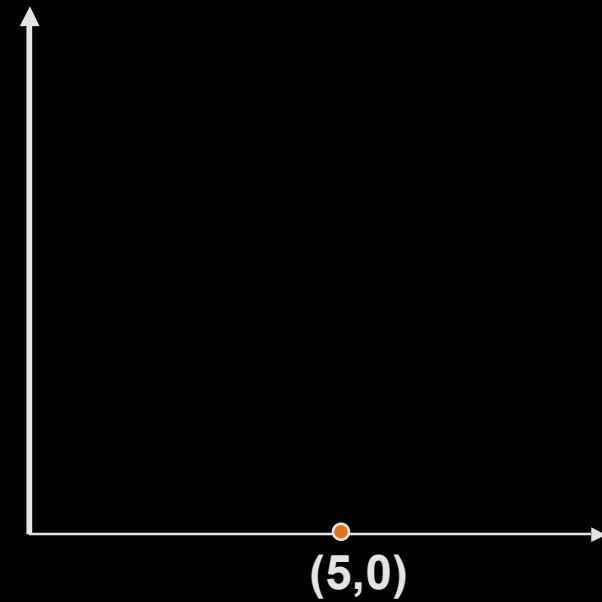
Typical

Image



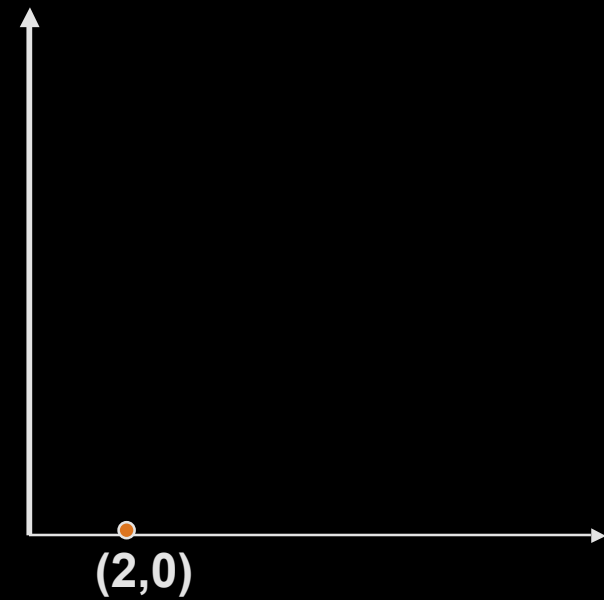
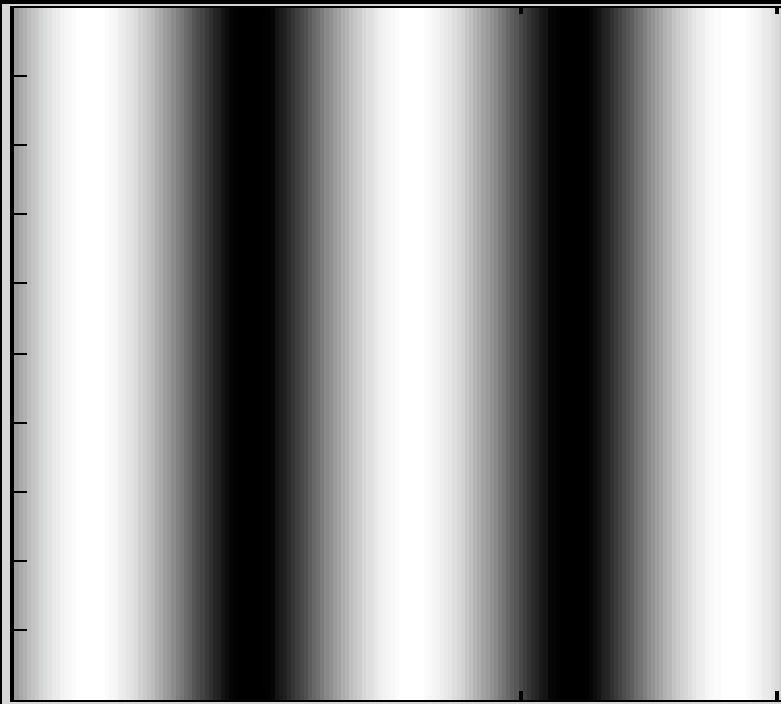
one "unit" of distance

Fourier Power Spectrum

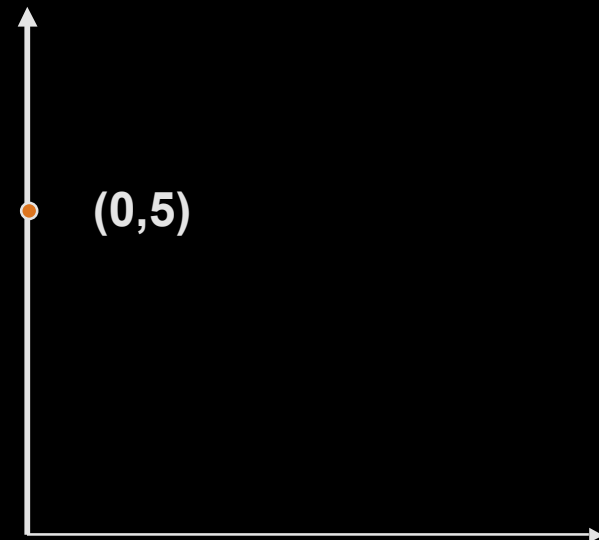
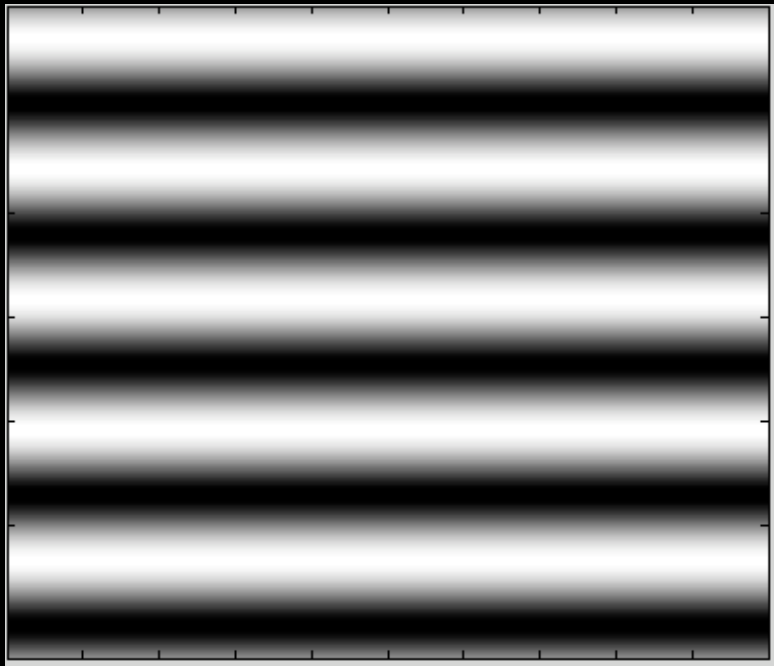


(5,0)

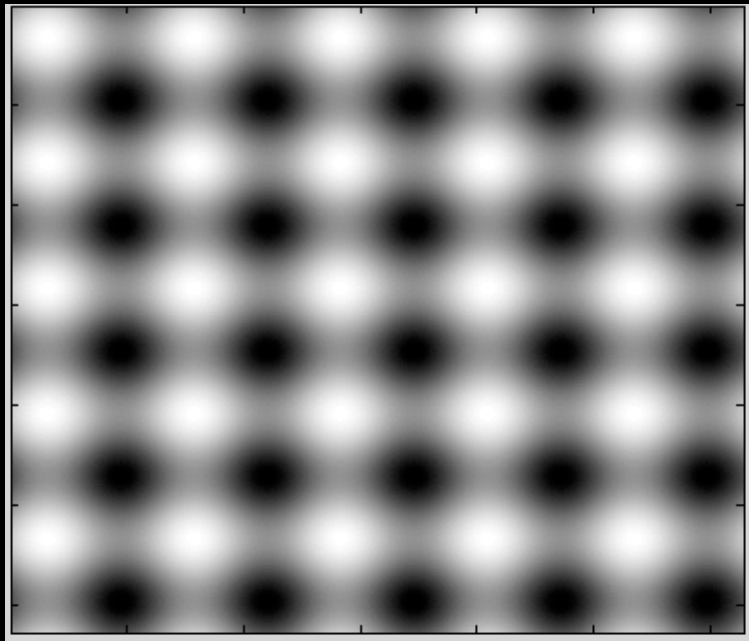
## Fourier Power Spectrum



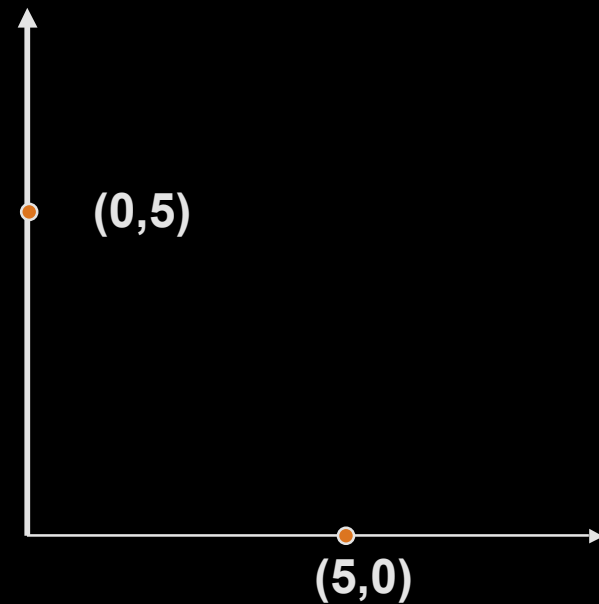
Fourier Power Spectrum







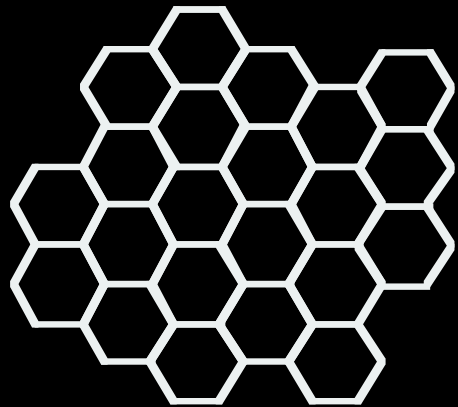
Fourier Power Spectrum



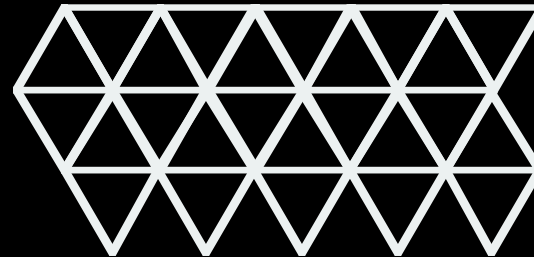
## Fourier Power Spectrum



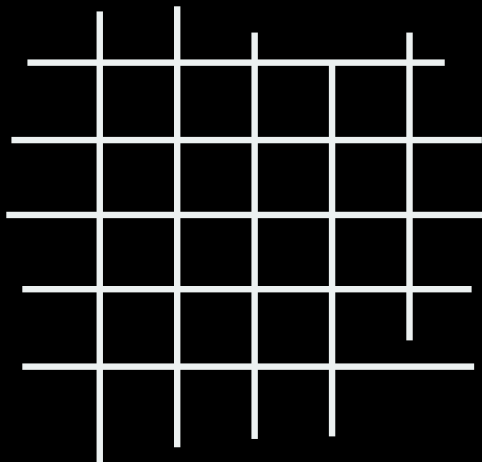
- Every sampling scheme captures some spatial frequencies but not others:
  - Low frequency sampling doesn't capture the picket fence
  - High frequency does.
- Which two-dimensional sampling scheme is most "efficient"?



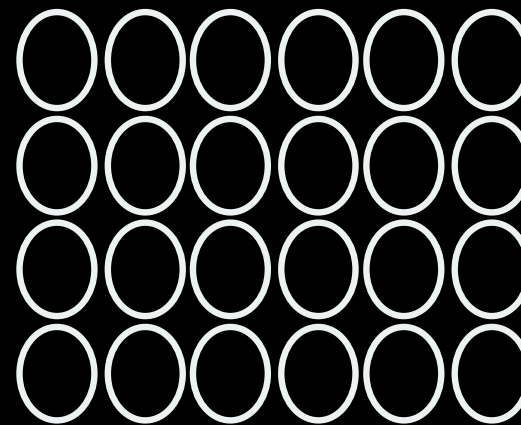
Hexagonal



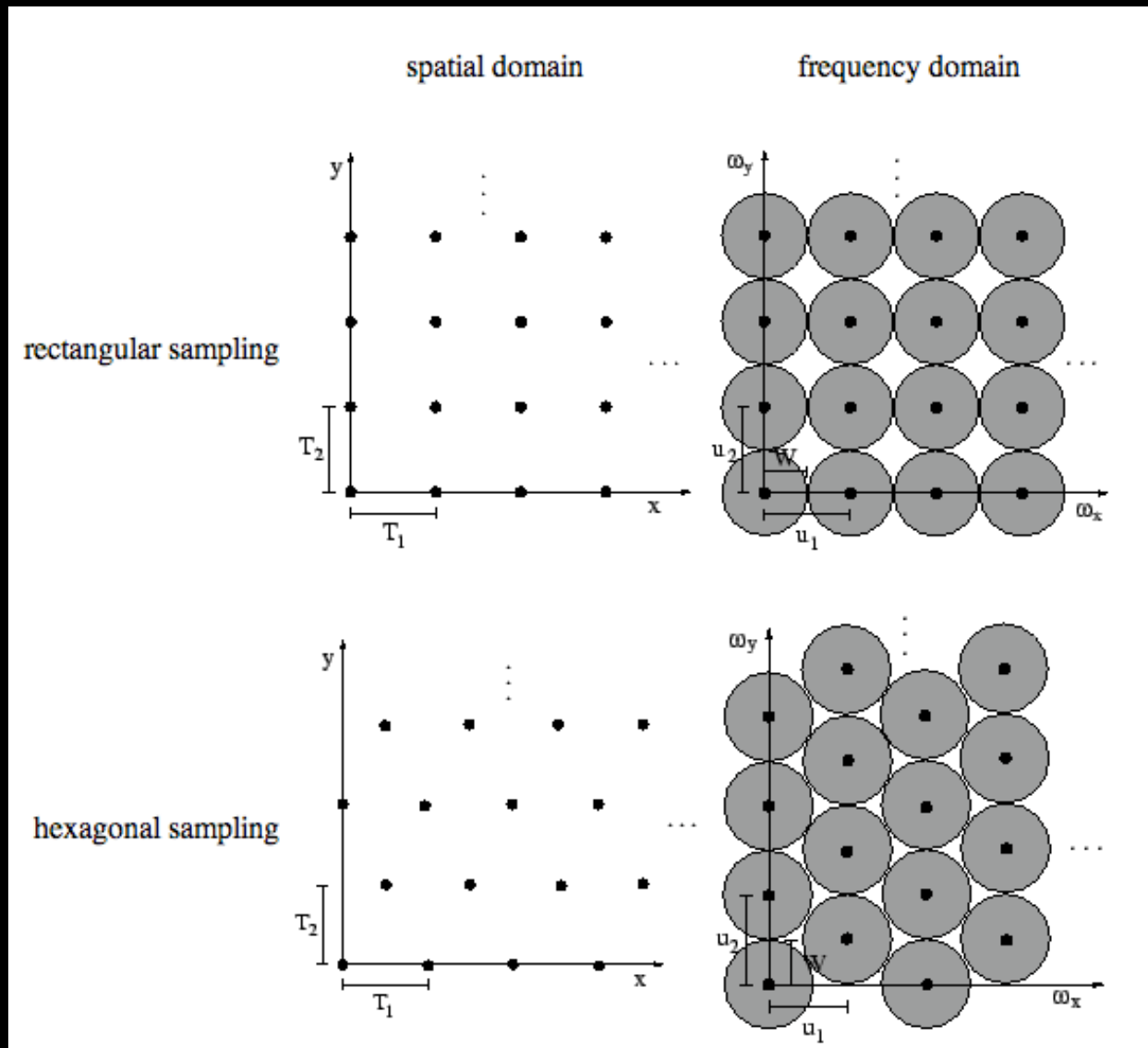
Triangular



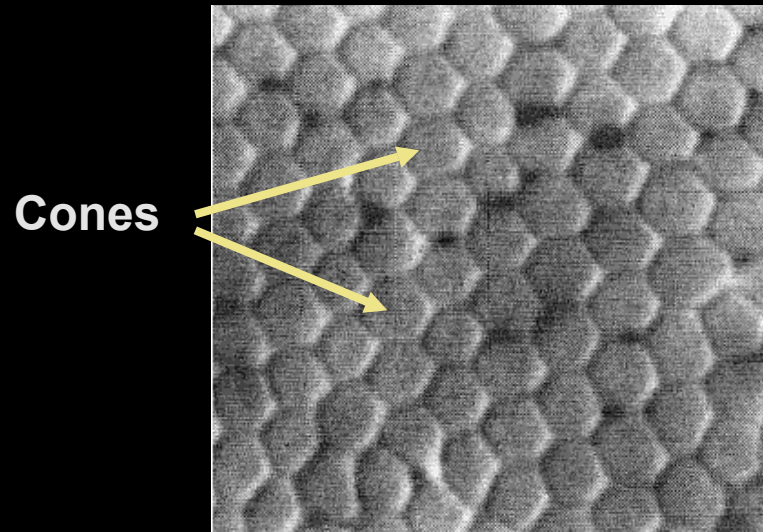
Rectangular



Typical

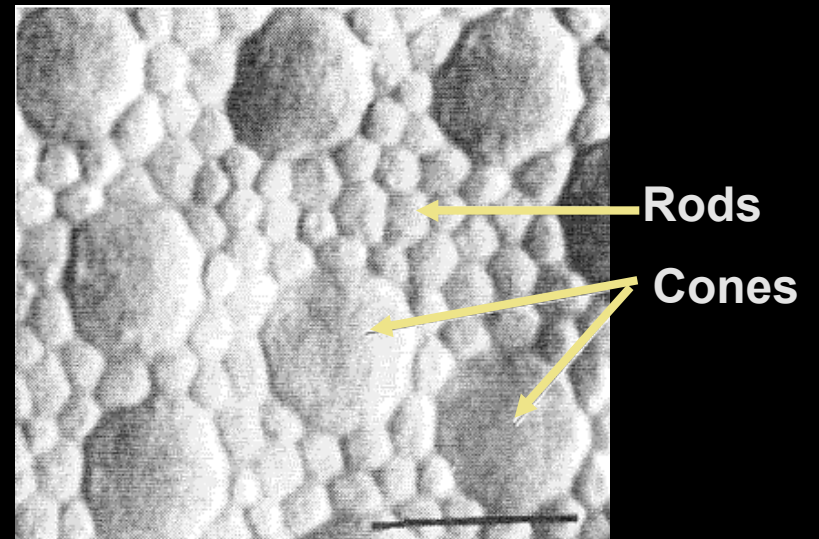


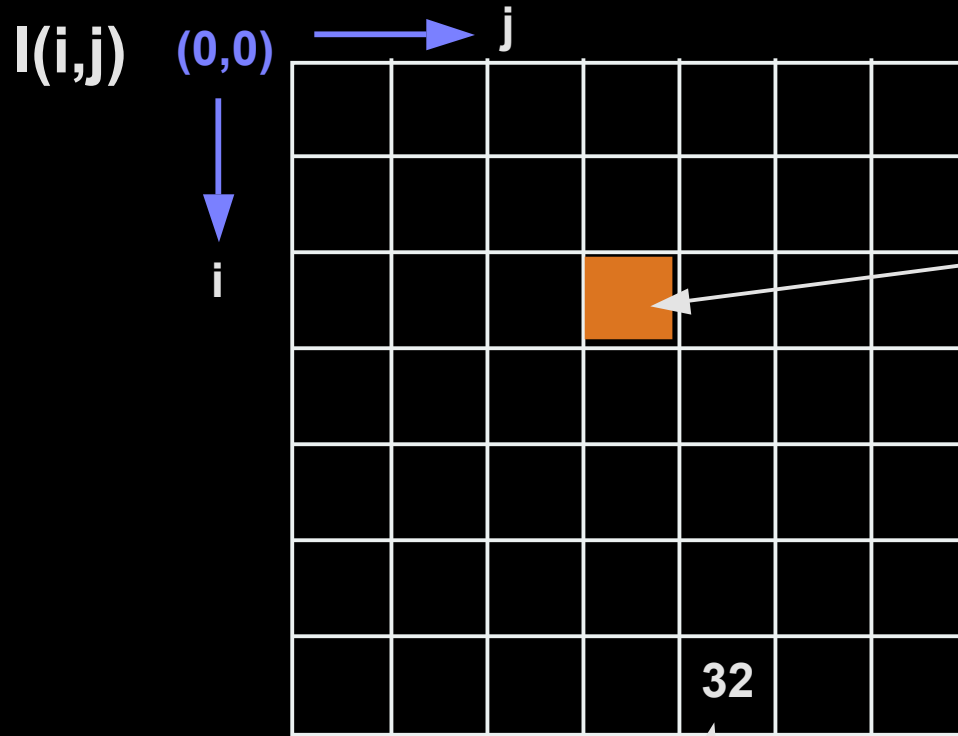
**Cones in the fovea**



**All of them are cones!**

**Moving outward from fovea**





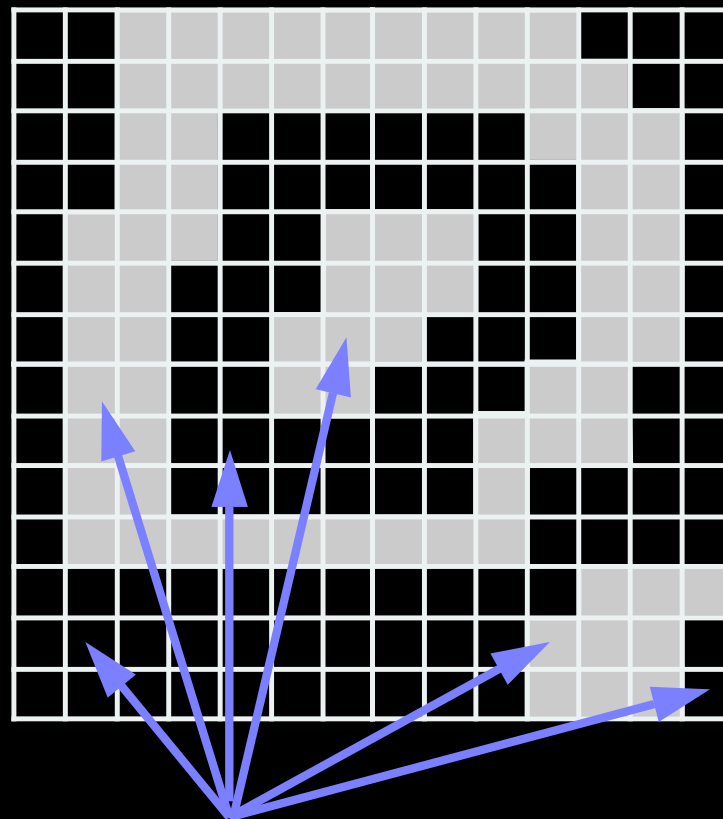
Picture Element or Pixel

Pixel value  $I(i,j) =$

- 0,1 Binary Image
- 0 - K-1 Gray Scale Image
- Vector: Multispectral Image

- Neighborhood
- Connectedness
- Distance Metrics

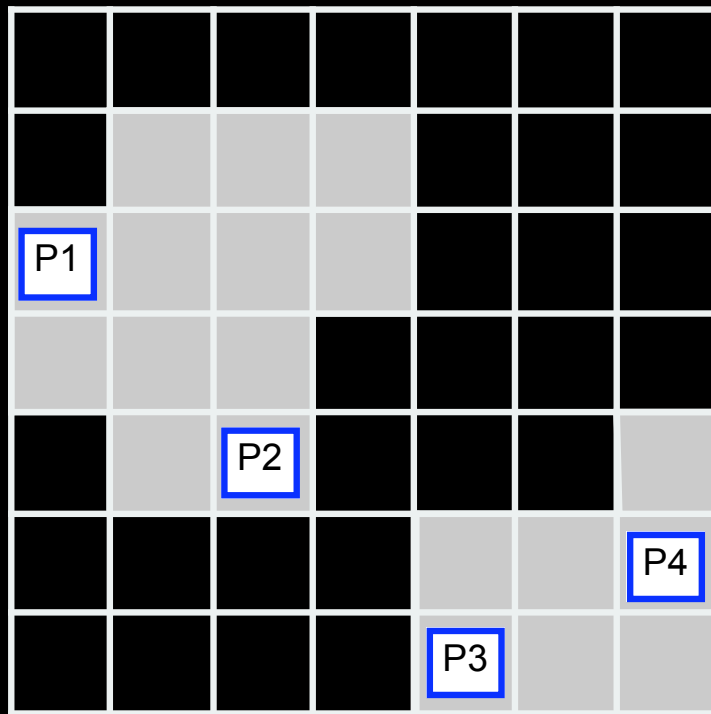
- Binary image with multiple 'objects'
- Separate 'objects' must be labeled individually



6 Connected Components



- Two points in an image are 'connected' if a path can be found for which the value of the image function is the same all along the path.



$P_1$  connected to  $P_2$

$P_3$  connected to  $P_4$

$P_1$  not connected to  $P_3$  or  $P_4$

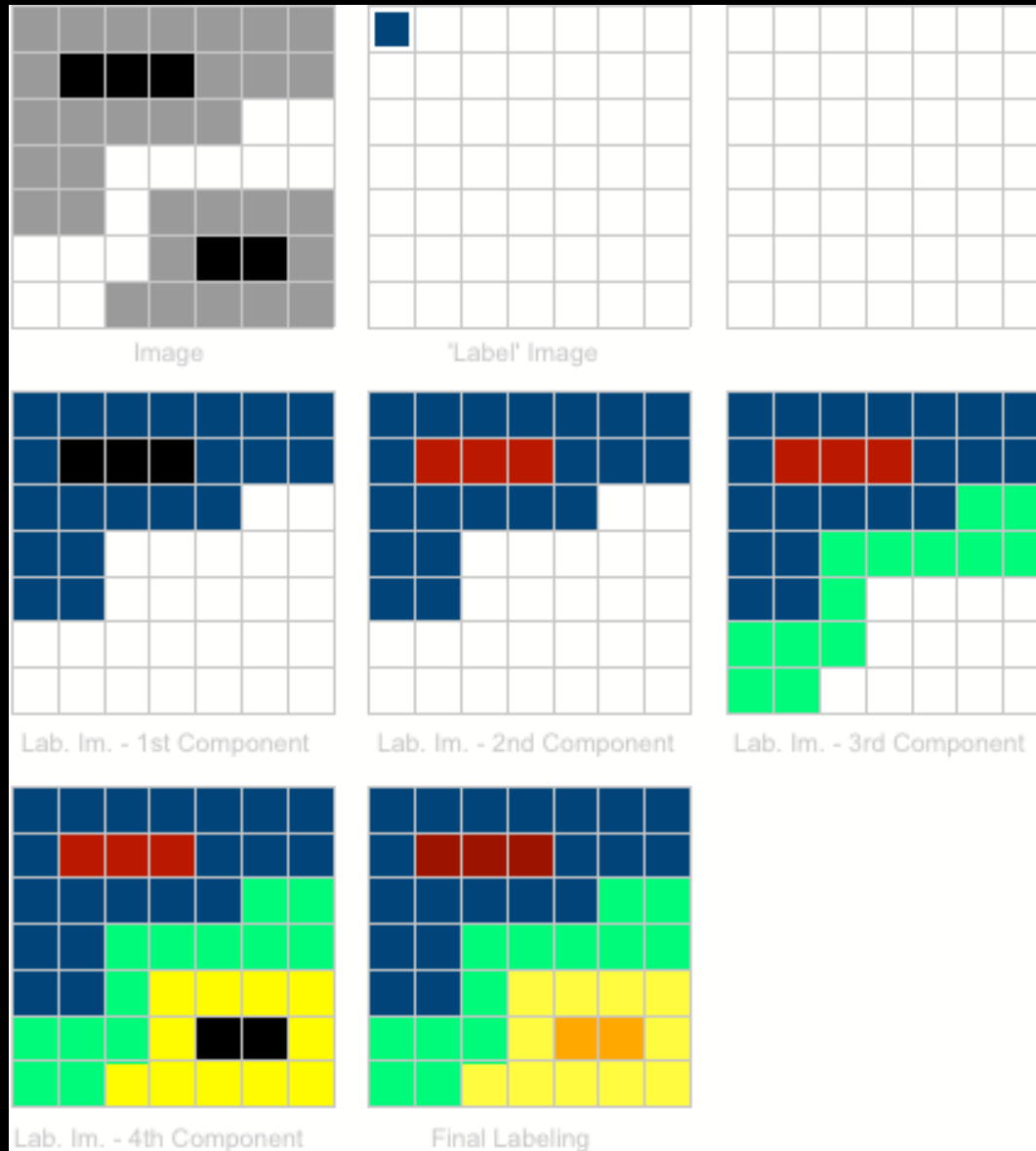
$P_2$  not connected to  $P_3$  or  $P_4$

$P_3$  not connected to  $P_1$  or  $P_2$

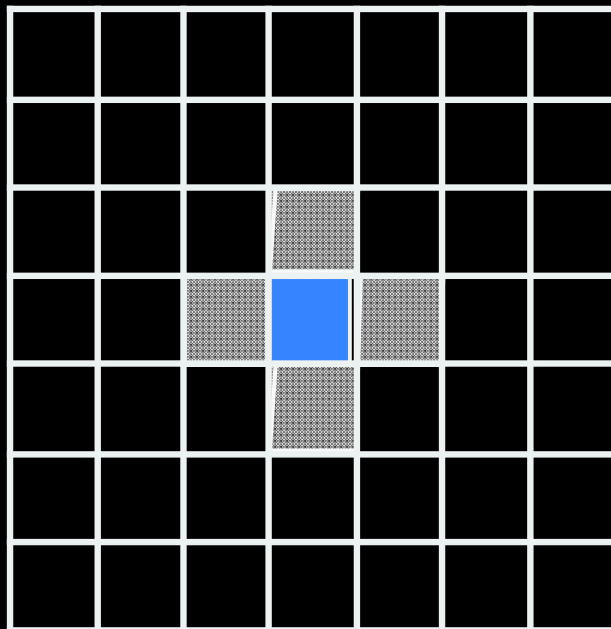
$P_4$  not connected to  $P_1$  or  $P_2$

- Pick any pixel in the image and assign it a label
- Assign same label to any neighbor pixel with the same value of the image function
- Continue labeling neighbors until no neighbors can be assigned this label
- Choose another label and another pixel not already labeled and continue
- If no more unlabeled image points, stop.

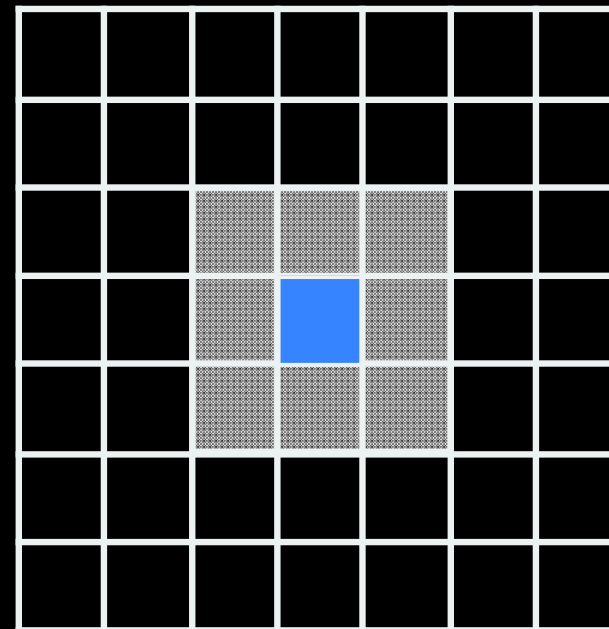
**Who's my neighbor?**



- Consider the definition of the term 'neighbor'
- Two common definitions:

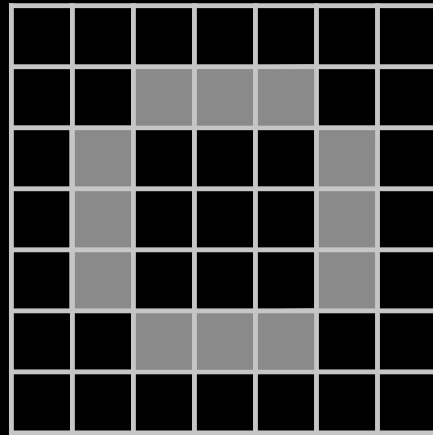


Four Neighbor



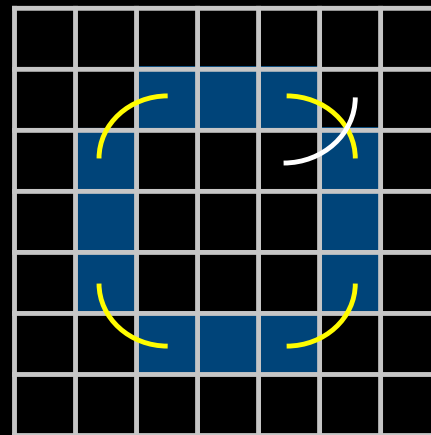
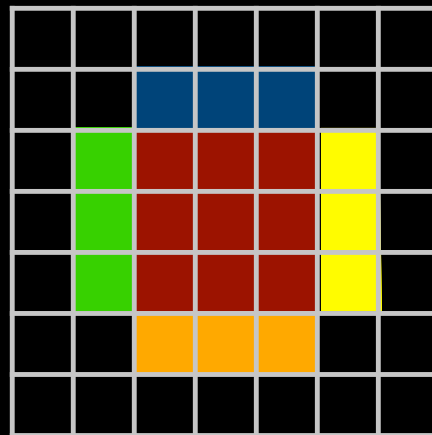
Eight Neighbor

- Consider what happens with a closed curve.
- One would expect a closed curve to partition the plane into two connected regions.



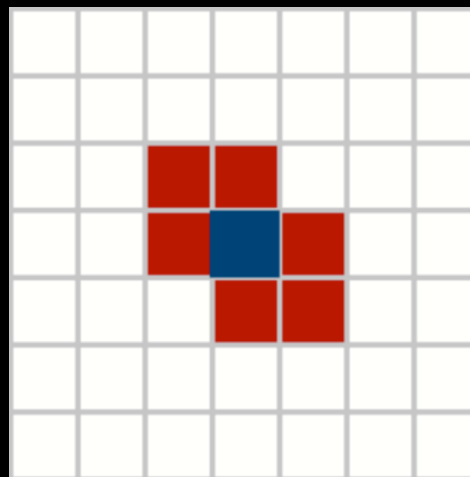
4-neighbor  
connectedness

8-neighbor  
connectedness

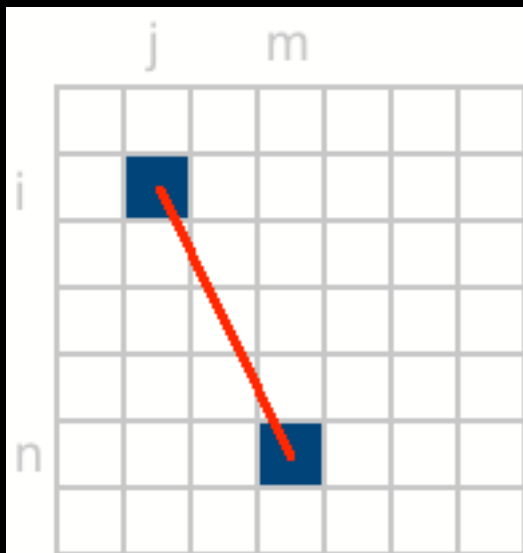


Neither neighborhood definition satisfactory!

- Use 4-neighborhood for object and 8-neighborhood for background
  - requires a-priori knowledge about which pixels are object and which are background
- Use a six-connected neighborhood:

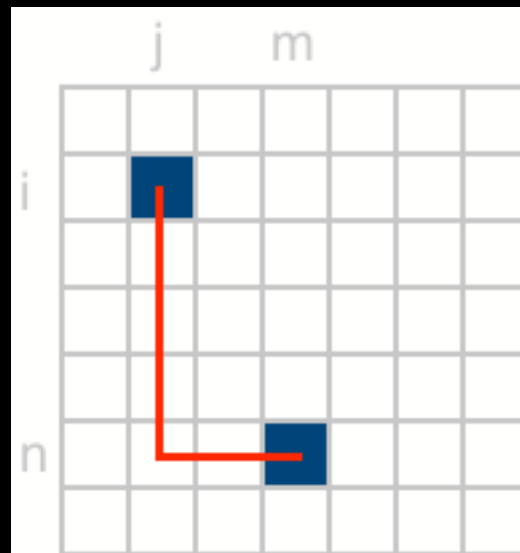


- Alternate distance metrics for digital images



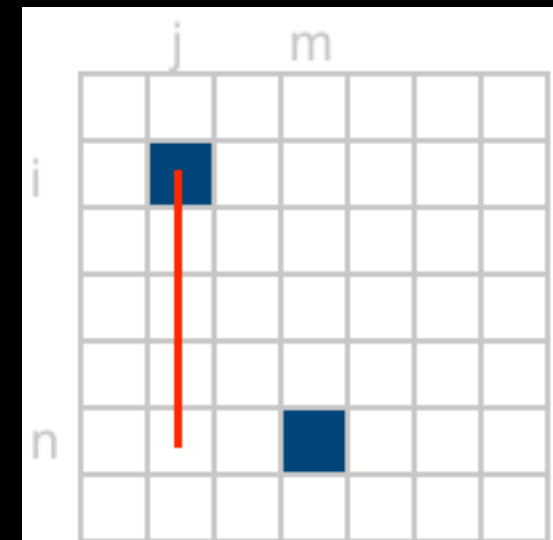
**Euclidean Distance**

$$= \sqrt{(i-n)^2 + (j-m)^2}$$



**City Block Distance**

$$= |i-n| + |j-m|$$



**Chessboard Distance**

$$= \max[ |i-n|, |j-m| ]$$