

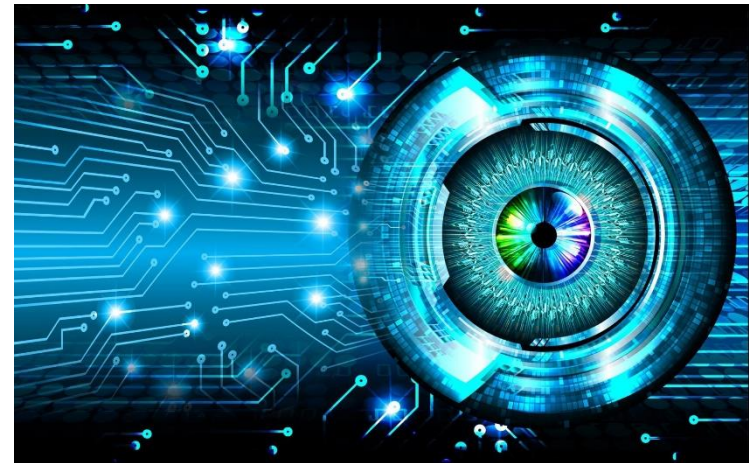


# Introduction to Computer Vision



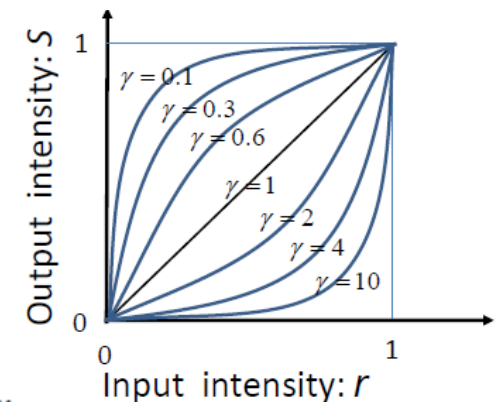
# Computer Vision

- Image data representation
- Image statistics
- Color models
- Thresholding
- Image restoration
- Depth sensing
- Camera model
- Feature extraction
- Image topic classification
- Time series visual feedback
- April tags for robotic application
- Numerous application in robotics, autonomous driving



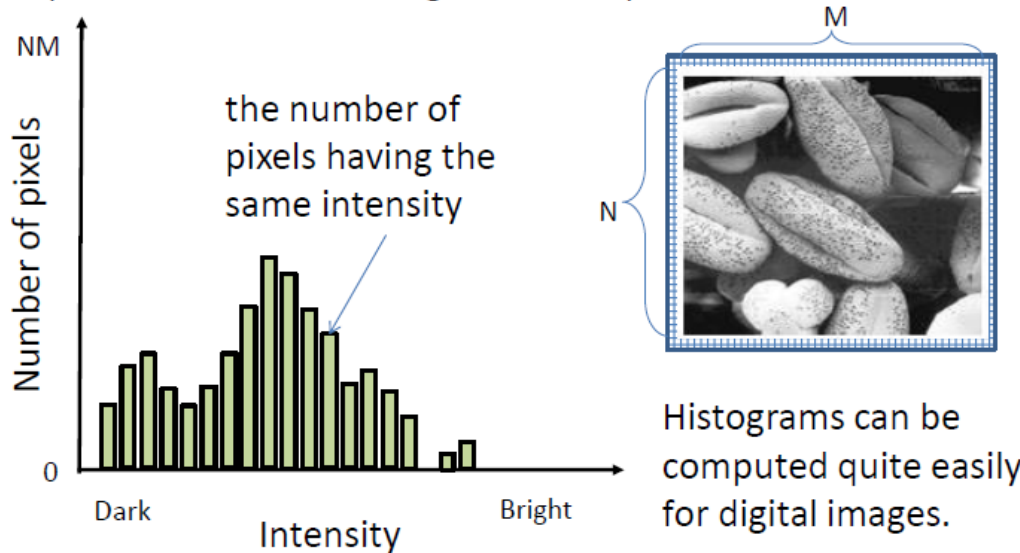
# Image Representation and Transformation

- Multi-dimensional matrix with color channel and spatial position`
- Bitmap file: large size, simple structure, uncompressed, lossless
- Joint Photographic Experts Group: lossy compression 24 bit color
- Graphics Interchange Format: LZW Lossless compression
- Portable Network Graphics: Open source, lossless compression
- Prefer to work with simple 2D \* 3 color channel matrix data format
- Intensity Gamma transformation for contrast

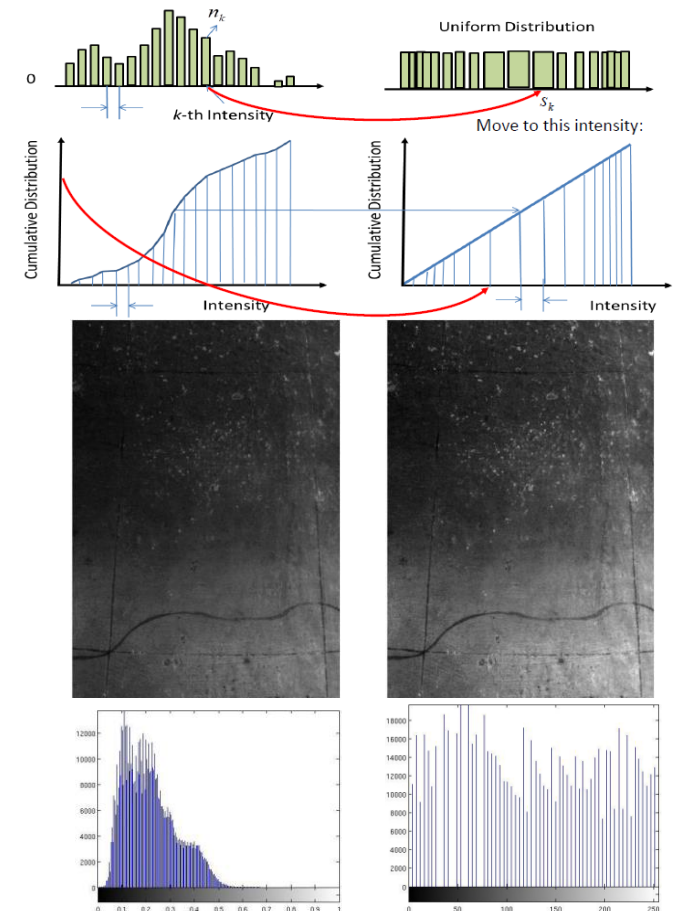


# Histogram Processing

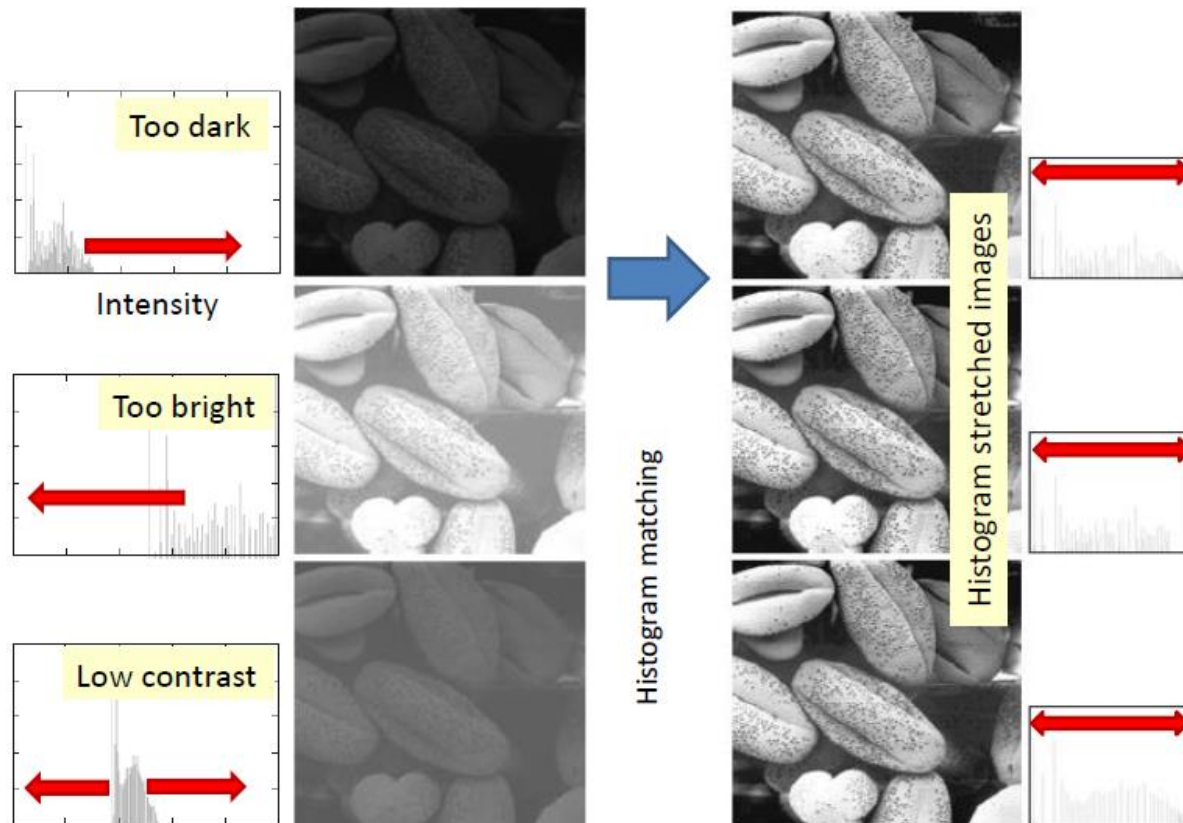
The histogram of a digital image indicates how the intensity of pixels distributes among the whole pixels.



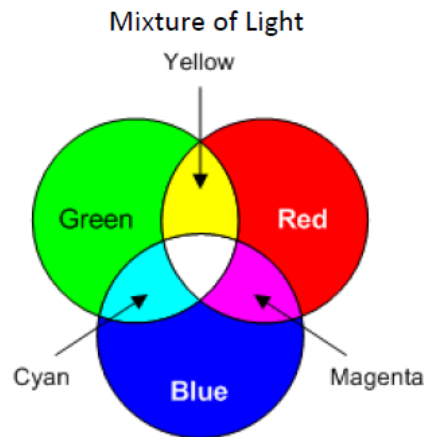
We learn from a histogram the darkness, brightness, and contrast of an image.



# Histogram Processing



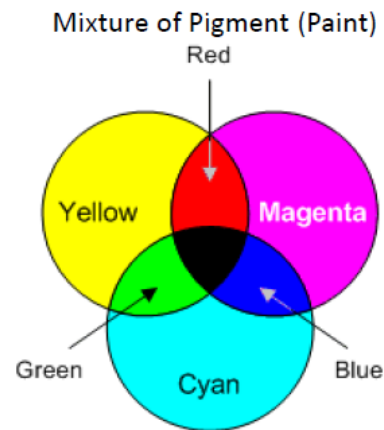
# Color Model



Red, Green, and Blue (RGB) – Primary colors of light (Additive Primaries)

Color Monitors and Video Cameras

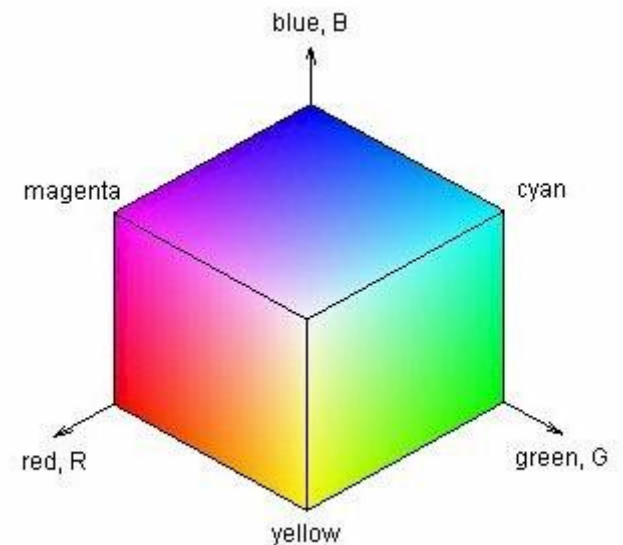
Magenta, Yellow, and Cyan – Secondary colors of light (Subtractive primaries)



Magenta, Yellow, and Cyan – Primary colors of pigment

Printers

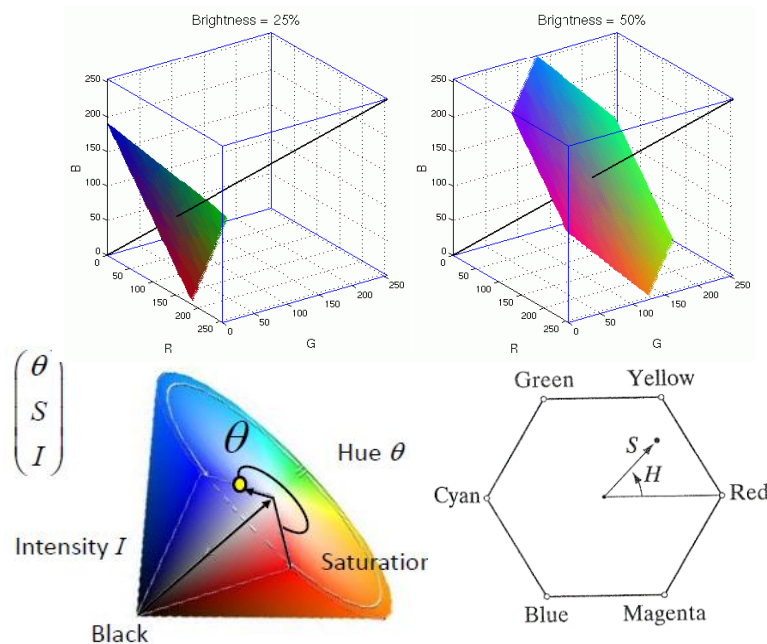
Red, Green, and Blue  
Secondary colors of pigment



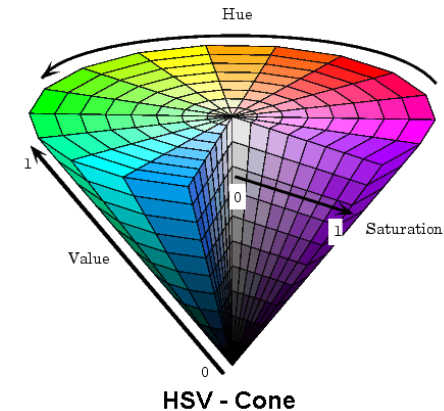
RGB color cube

# Color Model HIS, HSV

## Hue, saturation, intensity



## Hue, saturation, value



### RGB to HSI Conversion

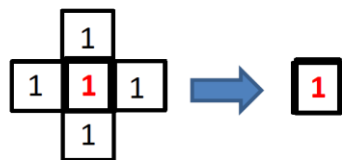
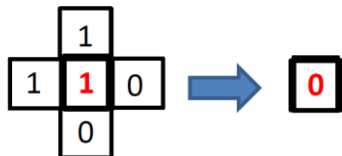
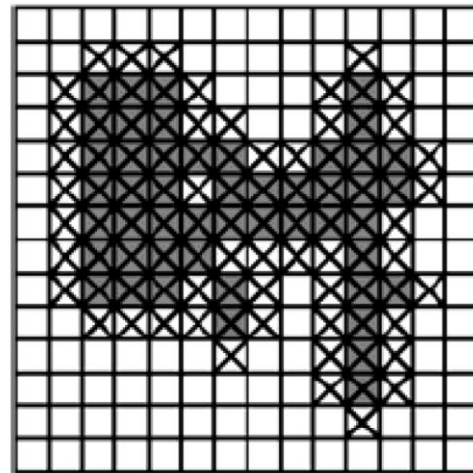
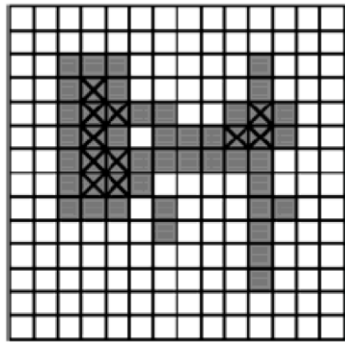
$$H = \begin{cases} \theta: & B \leq G \\ 2\pi - \theta: & B > G \end{cases} \quad \theta = \cos^{-1} \left( \frac{\frac{1}{2}[(R-G) + (R-B)]}{\sqrt{(R-G)^2 + (R-B)(G-B)}} \right)$$

$$I = \frac{1}{3}(R + G + B)$$

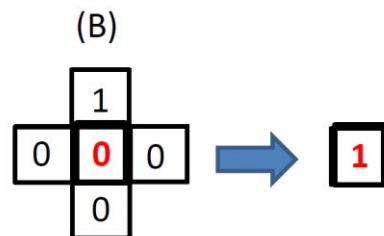
$$S = 1 - \frac{3}{(R + G + B)} \min(R, G, B)$$



# Erosion and Dilation

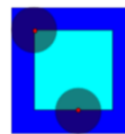


Erosion

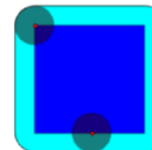


Dilation

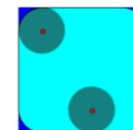
Erosion



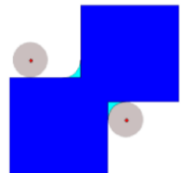
Dilation



Opening

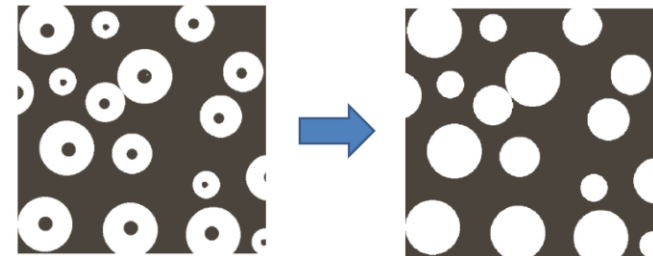


Closing



Opening = Erosion first, followed by dilation

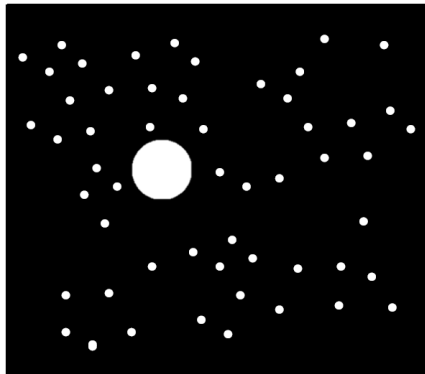
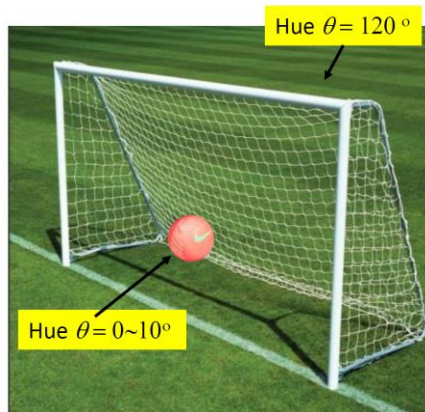
Closing = Dilation first, followed by erosion



Closing fills small holes.

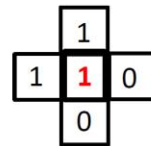


# Color Thresholding and De-noising

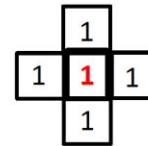


Simple rules for eliminating noise

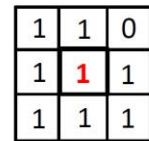
- Spatial Filtering: Erosion, Opening;  
Unless all the adjacent 4 or 8 pixels are 1s, replace the middle pixel by 0.



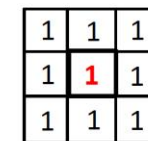
0



1



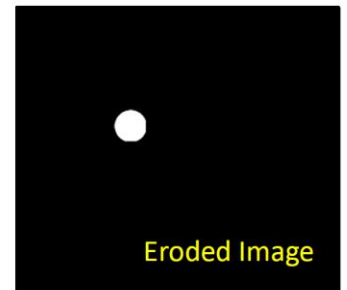
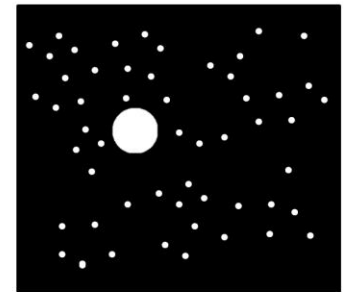
0



1

Erosion can eliminate isolated dots. The segmented ball gets smaller.

Difference Image

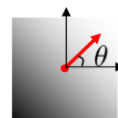
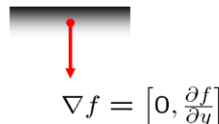
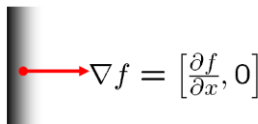


Eroded Image

# Kernel Based Edge Detection

- Originated from idea of gradient
- Kernel convolution

$$\nabla f = \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]$$



$$\nabla f = \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]$$

-1	0	1
-1	0	1
-1	0	1

Vertical

-1	-1	-1
0	0	0
1	1	1

Horizontal

Prewitt Operator

-1	0	1
-2	0	2
-1	0	1

-1	-2	-1
0	0	0
1	2	1

Sobel Operator

$$\theta = \tan^{-1} \left( \frac{\frac{\partial f}{\partial y}}{\frac{\partial f}{\partial x}} \right)$$

$$\|\nabla f\| = \sqrt{\left( \frac{\partial f}{\partial x} \right)^2 + \left( \frac{\partial f}{\partial y} \right)^2}$$



orig

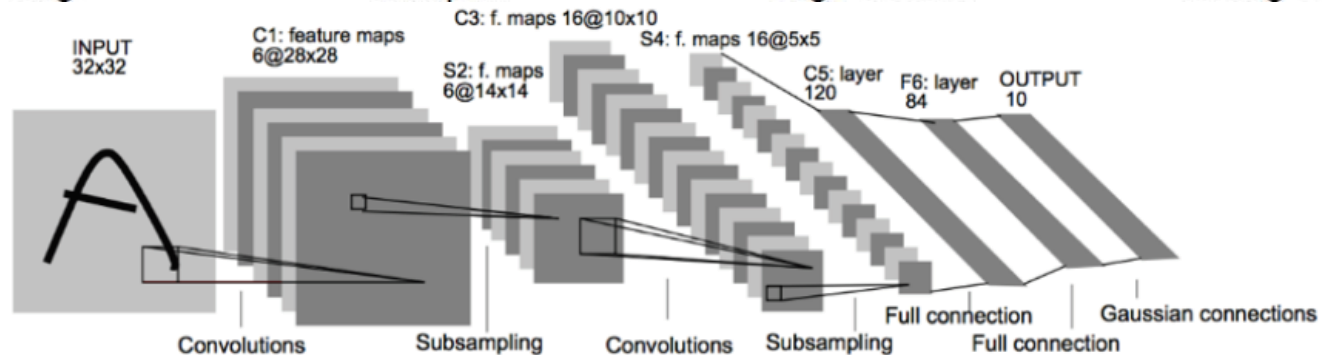


edge detect



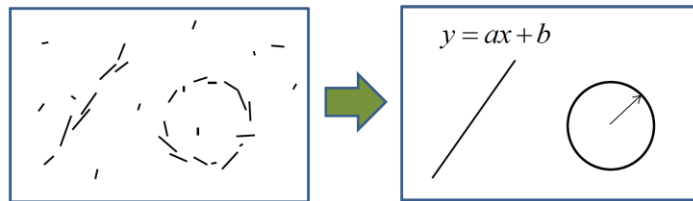
strong edges

# Convolutional Neural Network



# Hough Transformation

- Question: how to convert segmented edges to meaningful geometry?
- Utilize Hough transformation for parameterized primitive shapes

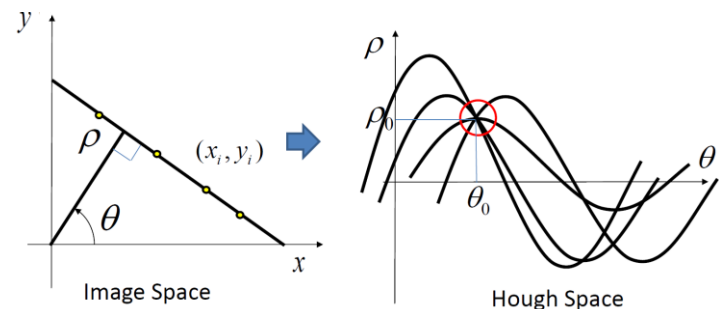
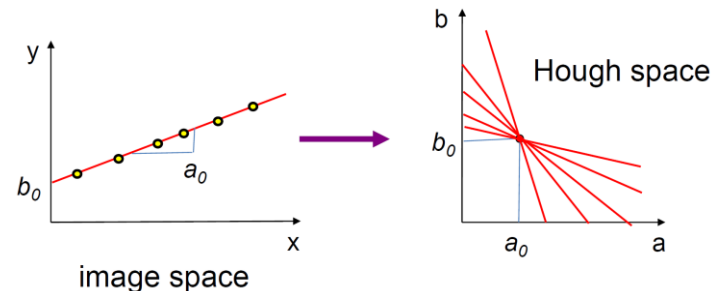


At each point of the (discrete) parameter space, find the number of lines that pass through it

– Use an array of counters

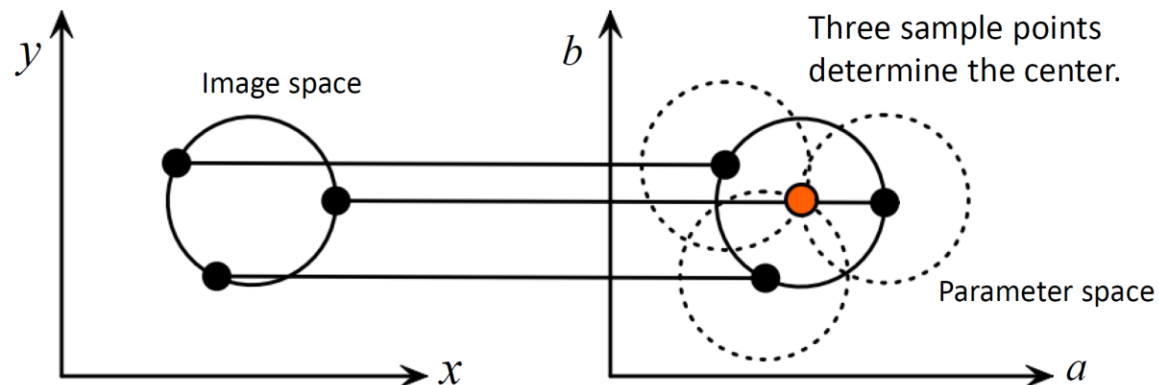
The more lines, the more edges are collinear in the image space.

Peaks in the counter array (bright points) in the Hough space found by thresholding.



# Hough Transformation for Circle

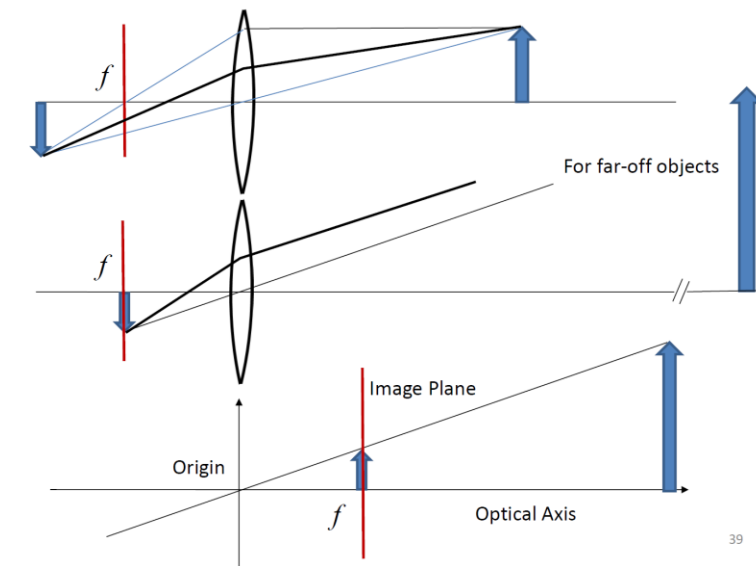
- Let  $R$  be a known radius, and  $(a, b)$  are unknown center coordinates.
- If a point at  $(x_i, y_i)$  is on the circle, it must satisfy
 
$$x_i = a + R \cos \theta, y_i = b + R \sin \theta$$
- In the parameter space,  $a$  and  $b$  must be on a circle given by
 
$$a = x_i - R \cos \theta, b = y_i - R \sin \theta$$



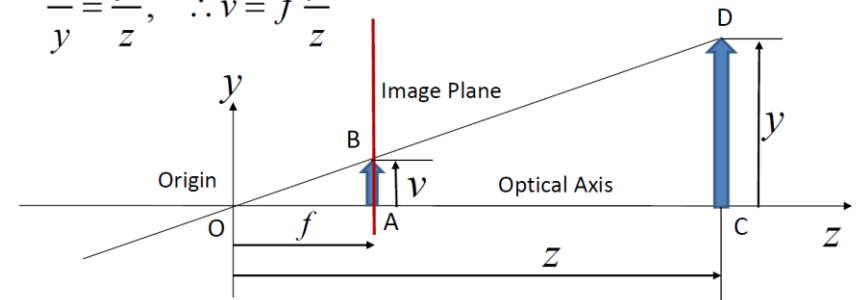
Each point in geometric space (left) generates a circle in parameter space (right). The circles in parameter space intersect at the  $(a, b)$  that is the center in geometric space.

# Camera Model

- For traditional, we know image size ( $v$ ) and focus length ( $f$ )
- With depth sensing (e.g. Kinect, Lidar), we also have depth ( $z$ )
- Can compute  $y$  directly given these information



$$\frac{v}{y} = \frac{f}{z}, \therefore v = f \frac{y}{z}$$



**Kinect**

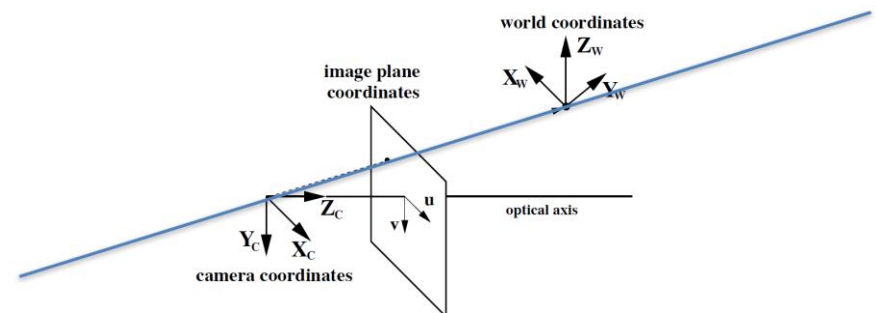


640 x 480 pixels  
640 x 480 pixels  
30 Hz USB 2

RGB Camera  
Infra-Red Depth Sensor

# April Tags

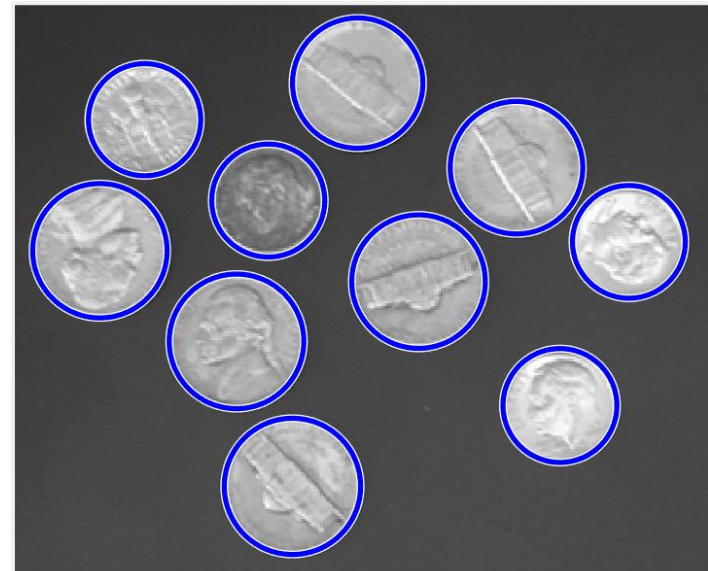
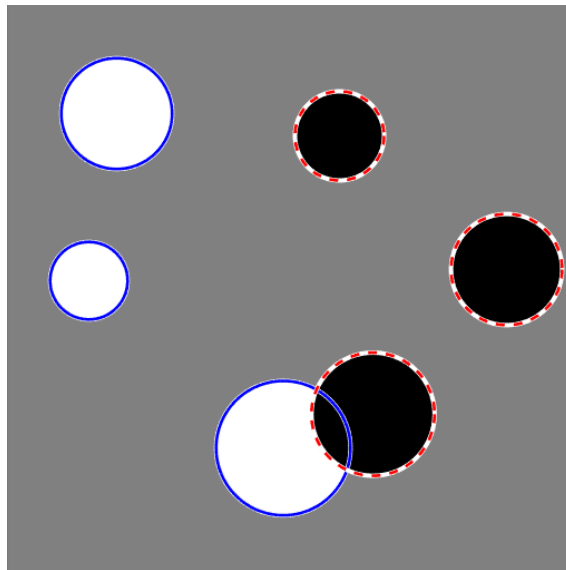
- Appearance similar to QR code
- Widely used in mobile robot application
- 3D translation and rotation transformation matrix
- Determine orientation and position of camera to a plane





# Hough Transformation Demo

- Matlab Hough transformation implementation





# Thank You!