



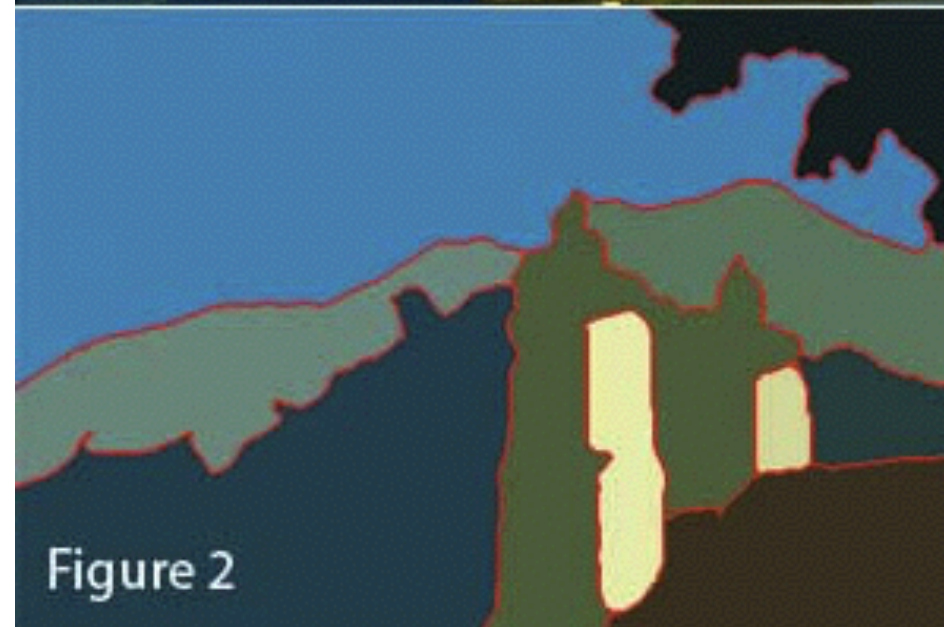
EE 604

Digital Image Processing

Image segmentation

- Segmentation is the process of
 - separating objects from background
 - partitioning an image into coherent regions
- In general, it is a difficult task
- Often a first step to image analysis in applications such as, object recognition.

Segmentation examples



source: <http://csl.illinois.edu/>

Types of Segmentation

- Partition based on predefined criteria
 - Partition based on discontinuity (isolated points, lines, edges)
 - Partition based on similarity (in color, structure, shape)
- Partition with or without manual intervention

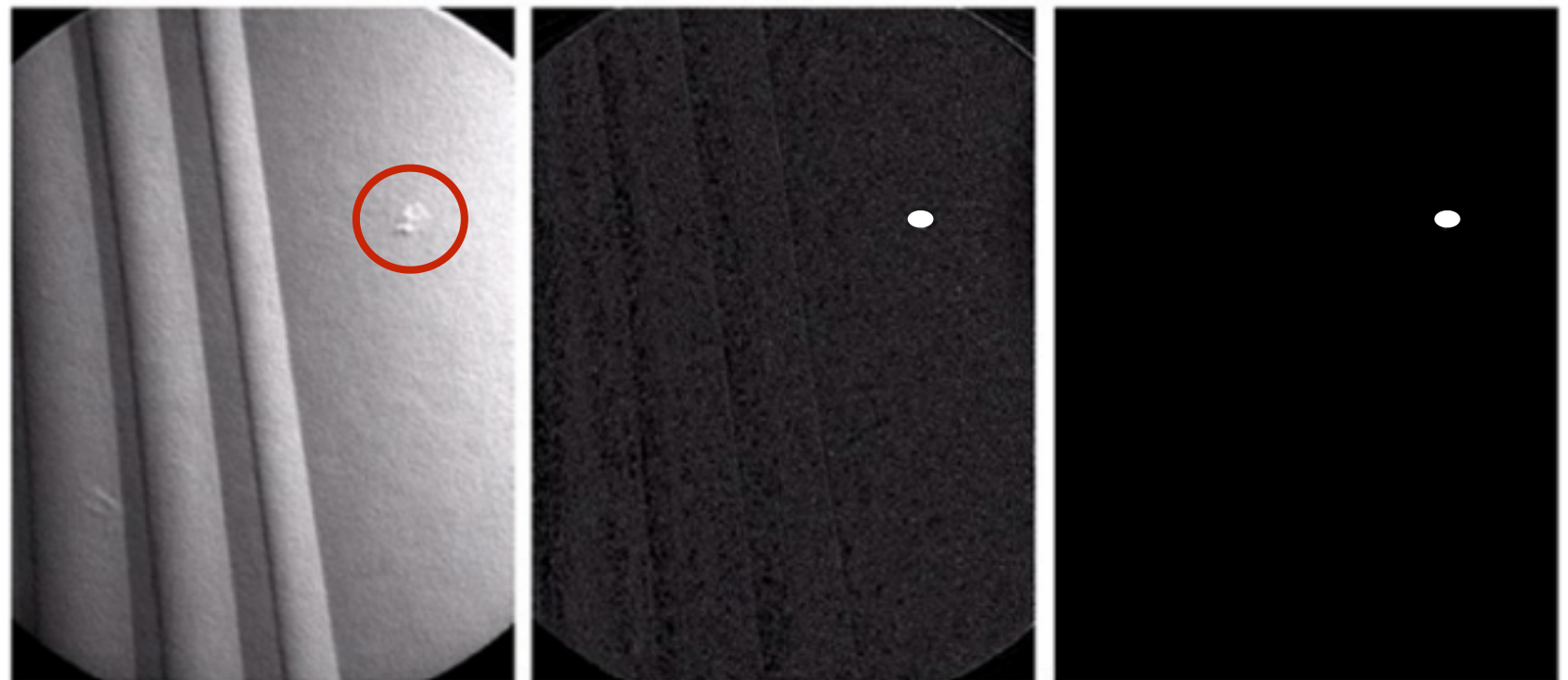
Machine Learning Perspective

- Can be seen as a supervised or unsupervised learning problem
 - Foreground-background segmentation can be seen as a binary classification of each pixel (supervised, weakly-supervised)
- Can be seen as a data clustering problem (unsupervised)
 - image segments = clusters in a suitable feature space

Basics

- Detect gray-level discontinuities
 - isolated points, lines, edges
- **Isolated point detection:**

-1	-1	-1
-1	8	-1
-1	-1	-1



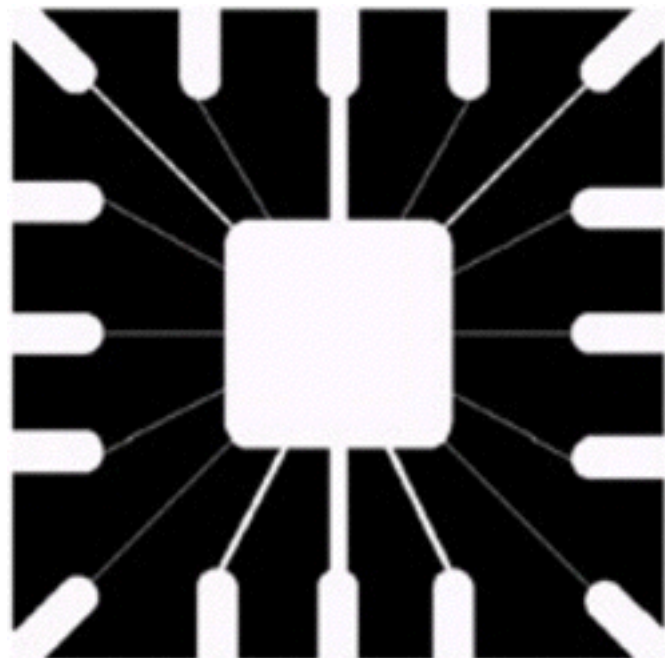
Basics

- Detect gray-level discontinuities
 - isolated points, lines, edges
- Line detection:

-1	-1	-1	-1	-1	2	-1	2	-1	2	-1	-1
2	2	2	-1	2	-1	-1	2	-1	-1	2	-1
-1	-1	-1	2	-1	-1	-1	2	-1	-1	-1	2
Horizontal			+45°			Vertical			-45°		

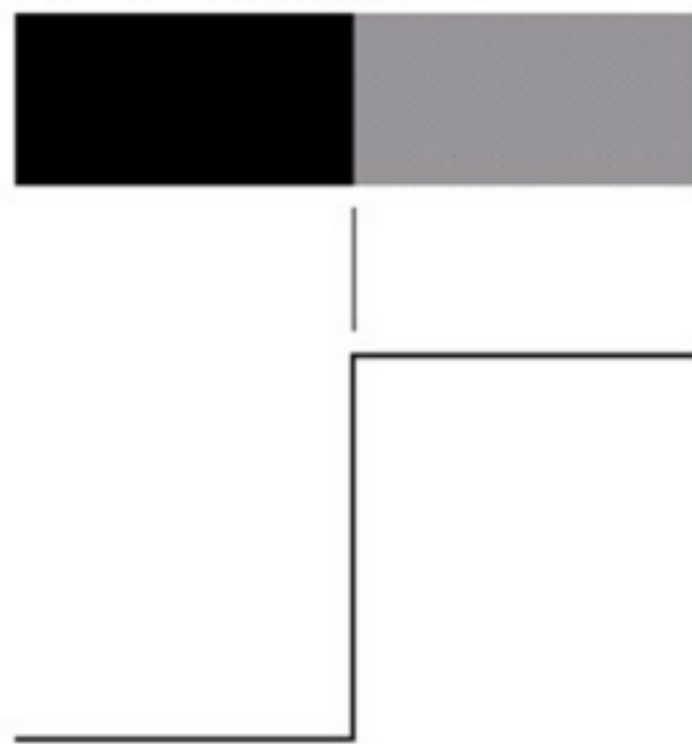
Basics

- Detect gray-level discontinuities
 - isolated points, lines, edges
- Line detection:

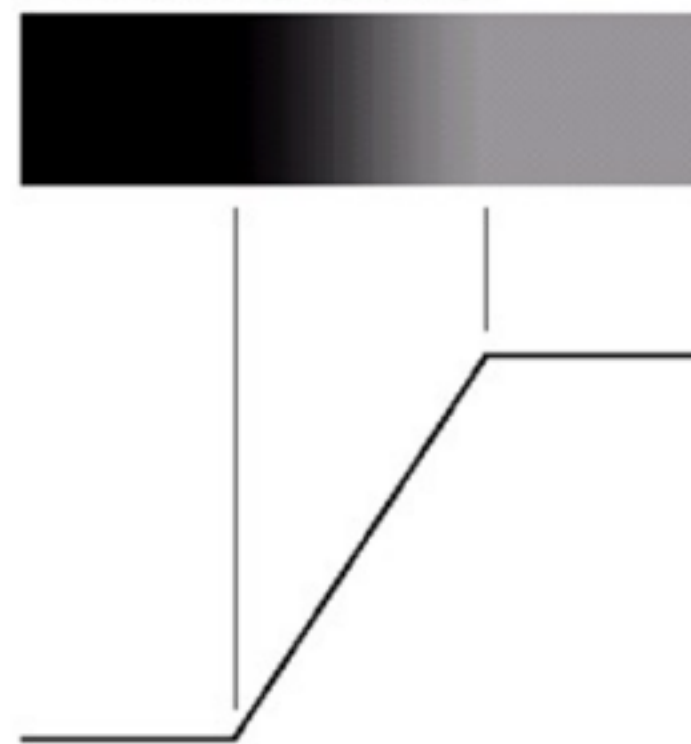


Basics

- Detect gray-level discontinuities
 - isolated points, lines, edges
- Edge detection:



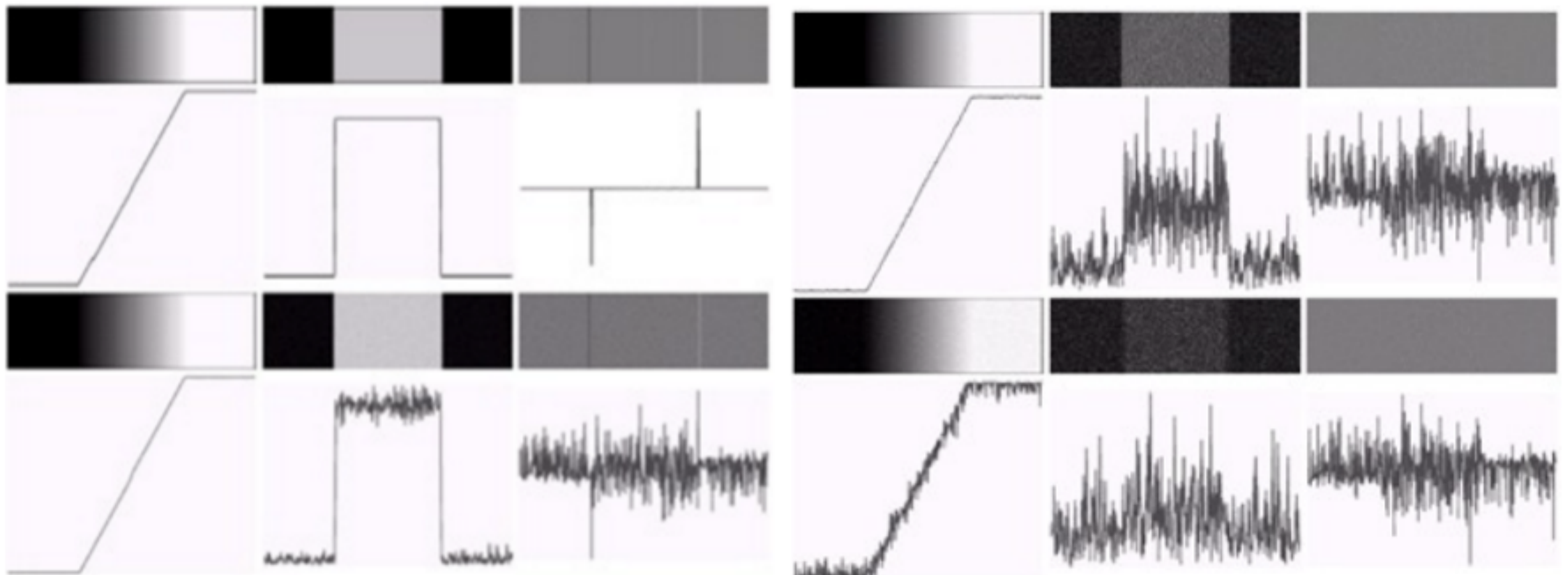
gray-level profile



gray-level profile

Basics

- Detect gray-level discontinuities
 - isolated points, lines, edges
- Edge detection:



Hough transform

- Detects geometric shapes, such as lines and circles, in images, when the parametric equation is known.
- Introduced in 1962 by [Hough 1962], and was first used to find lines in images in [Duda 1972].
- Robust detection under noise and partial occlusion
- Requires edge detection as a preprocessing step

Hough transform



Can we achieve these?

- Find the n most prominent lines
- Find lines in a particular orientation

Hough transform



Before applying Hough transform

- Apply edge detector (any derivative operator)
- Compute a magnitude edge map
- Binarize the edge map by thresholding

Hough transform



Edge map as input image to Hough transform

Before applying Hough transform

- Apply edge detector (any derivative operator)
- Compute a magnitude edge map
- Binarize the edge map by thresholding

Hough transform

- Assume that we have performed edge detection and binarization.
- We have n pixels (white) in the image that may partially describe edges and boundaries.
- We wish to find a subset of pixels that form straight lines.
- Consider a point (x_i, y_i) in the image which lies in a line $y_i = ax_i + b$
 - Infinitely many lines pass through (x_i, y_i) .
 - They satisfy the equation for different values of (a, b) .

Hough transform

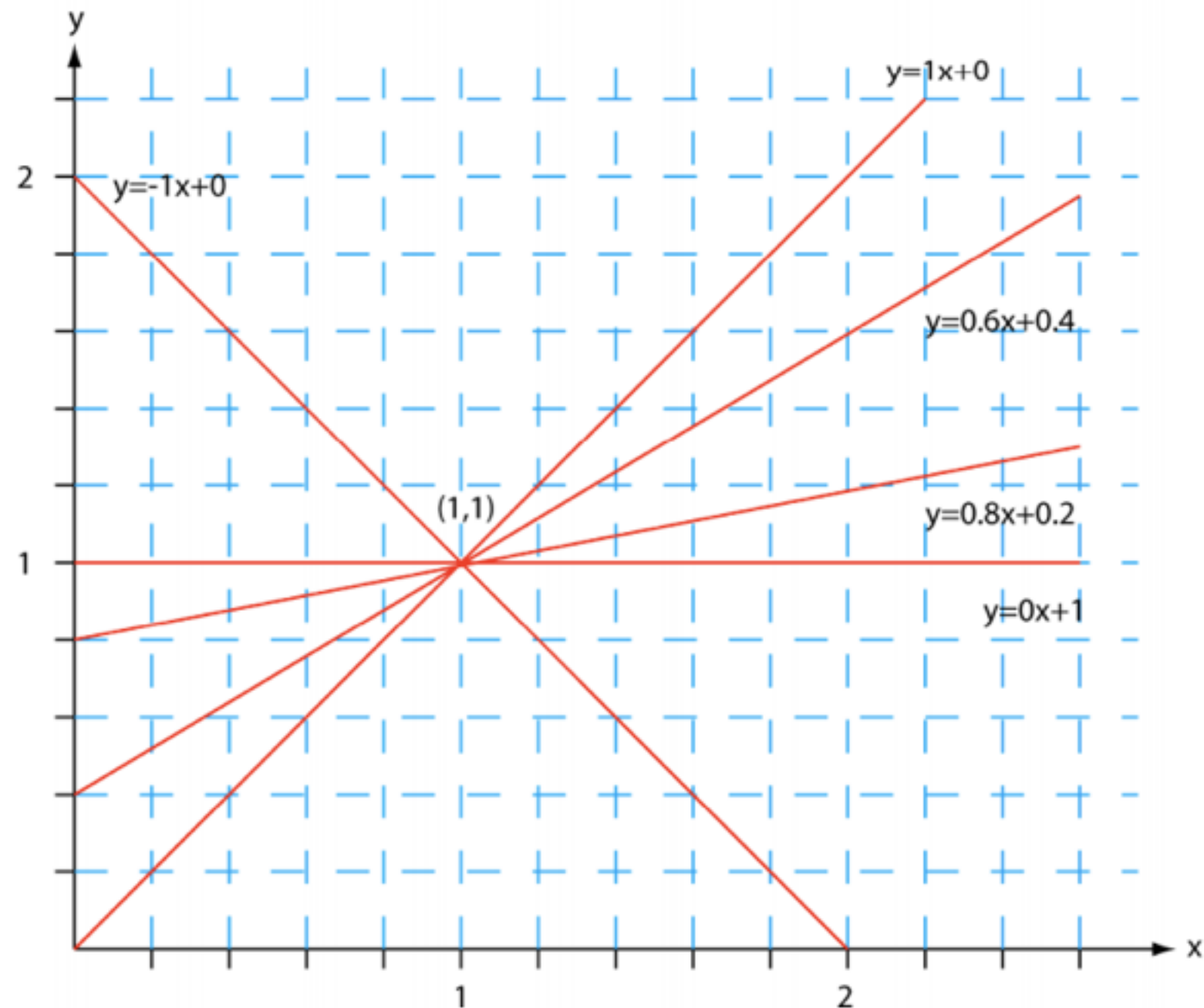
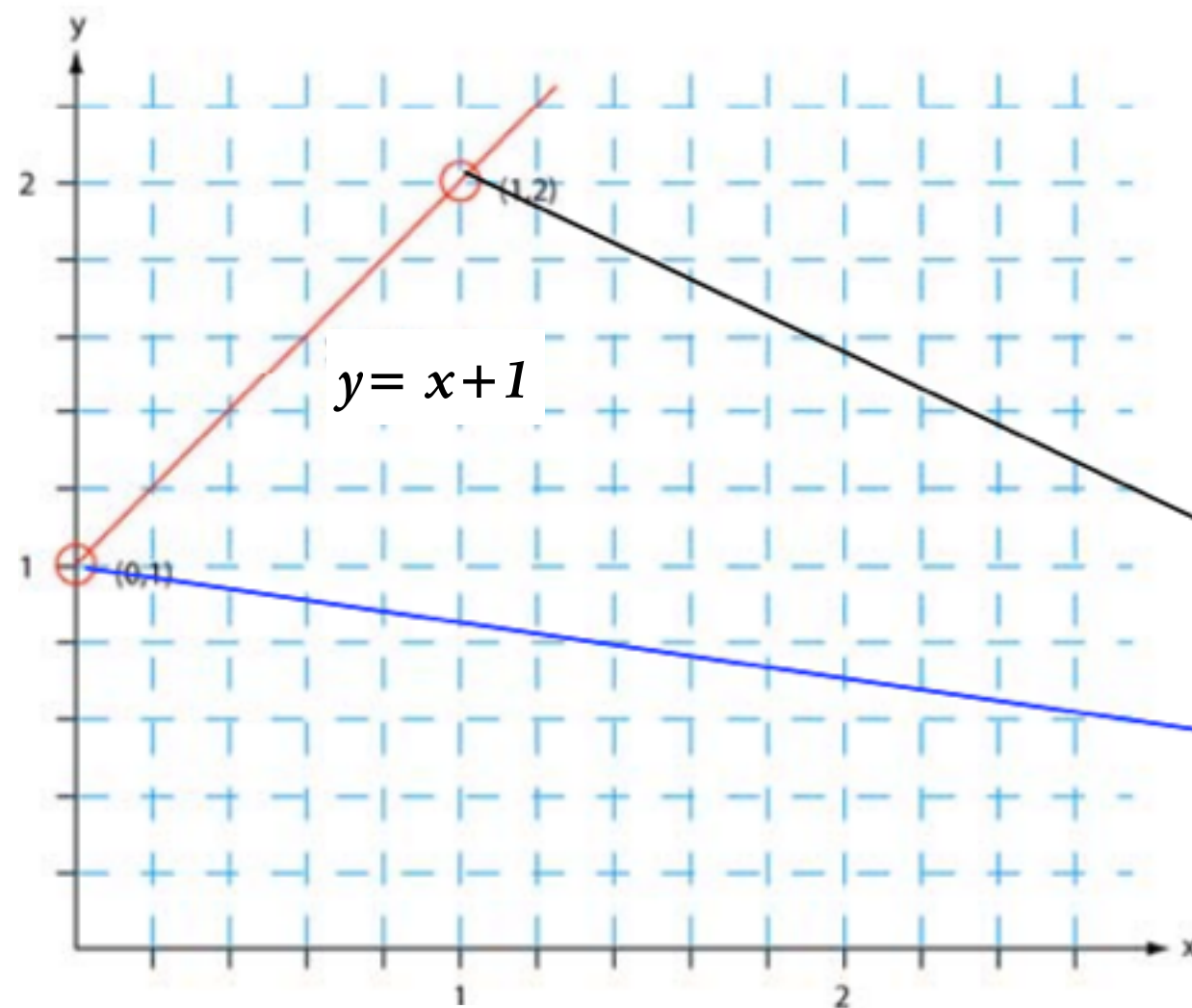


Image source: <https://www.uio.no/studier/emner/matnat/ifi/INF4300/>

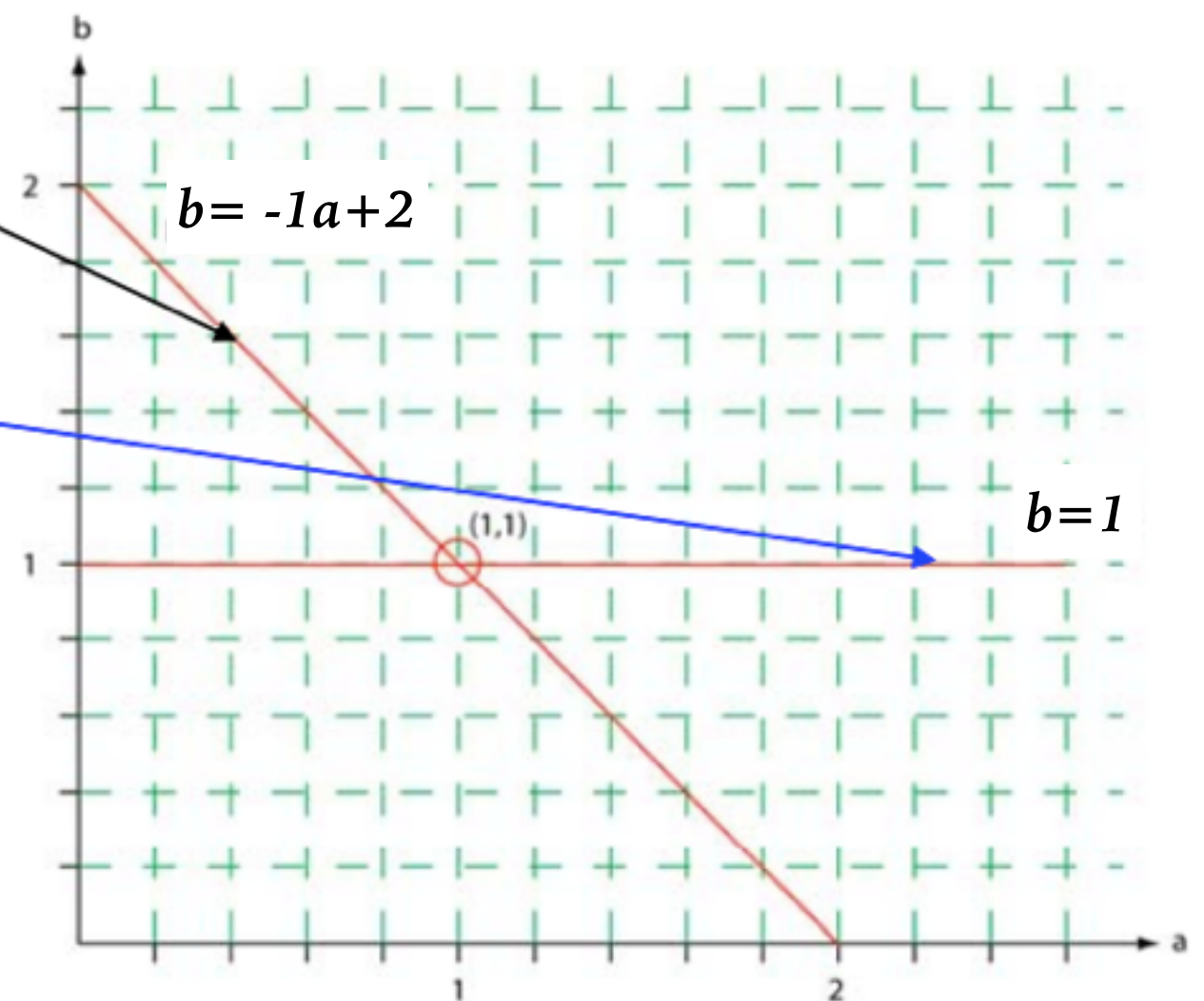
Hough transform

- The equation can obviously be rewritten as $b = -ax_i + y_i$
- Now consider x and y as parameters and a and b as variables.
- This is a line in ab -space parameterized by x and y .
- A single point in xy -space corresponds to a line in ab -space.
- Another point in xy -space will give rise to another line in ab -space.

Hough transform

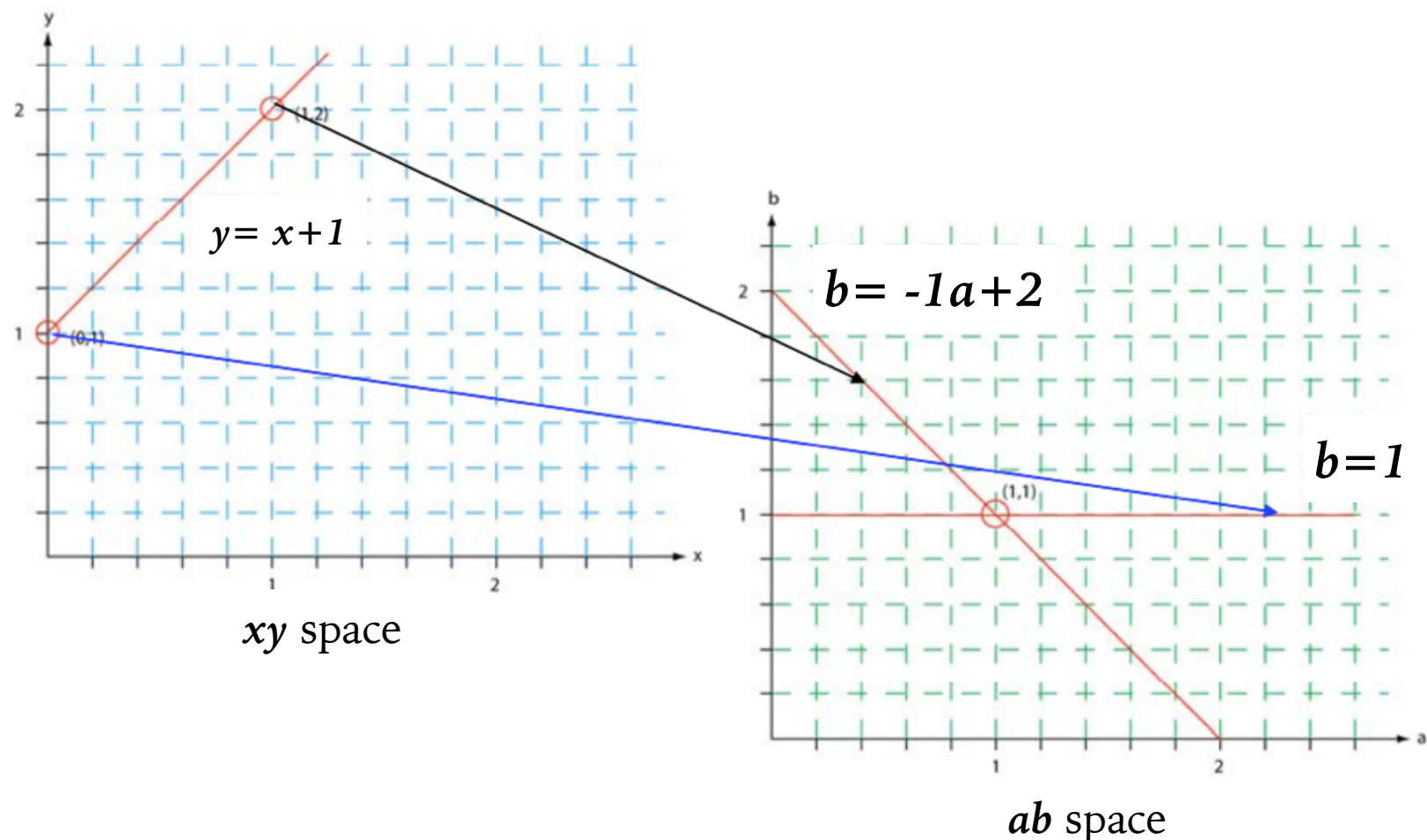


xy space



ab space

Hough transform



- All points on $y = x + 1$ will reflect as intersecting lines at $(1,1)$ in ab -space
- What if we count the number of intersections at each (a,b) location?

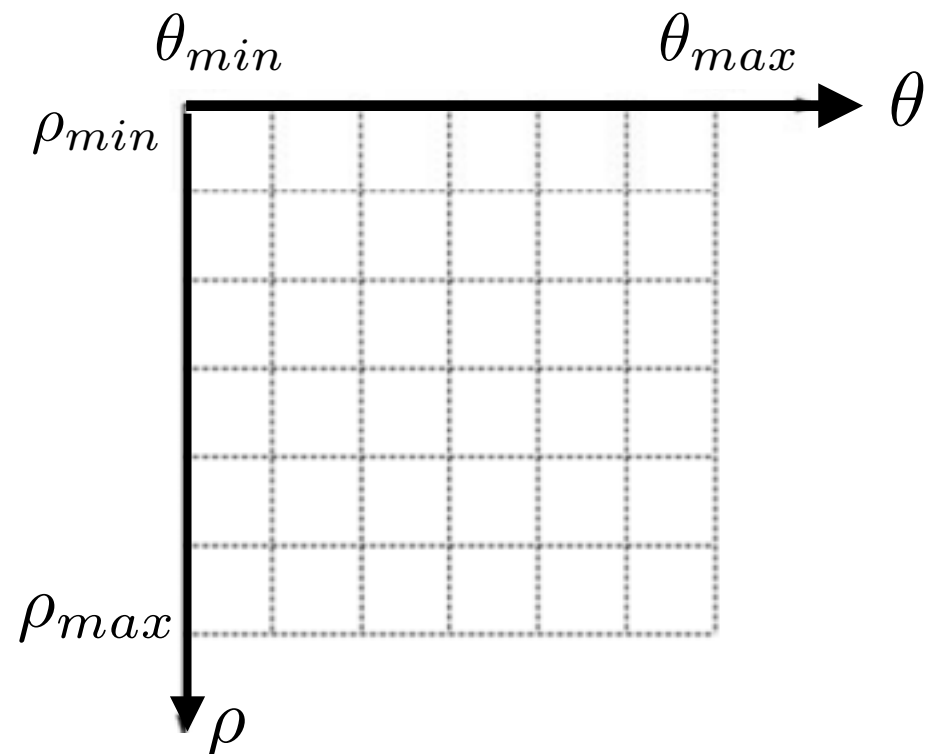
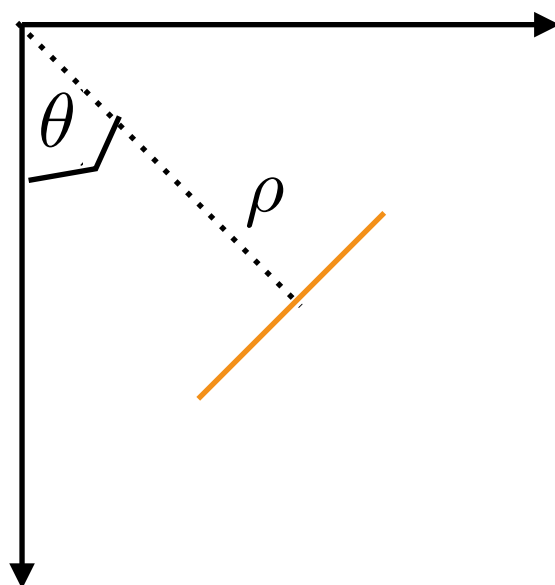
Hough transform

Key idea: Move from pixel space to parameter space

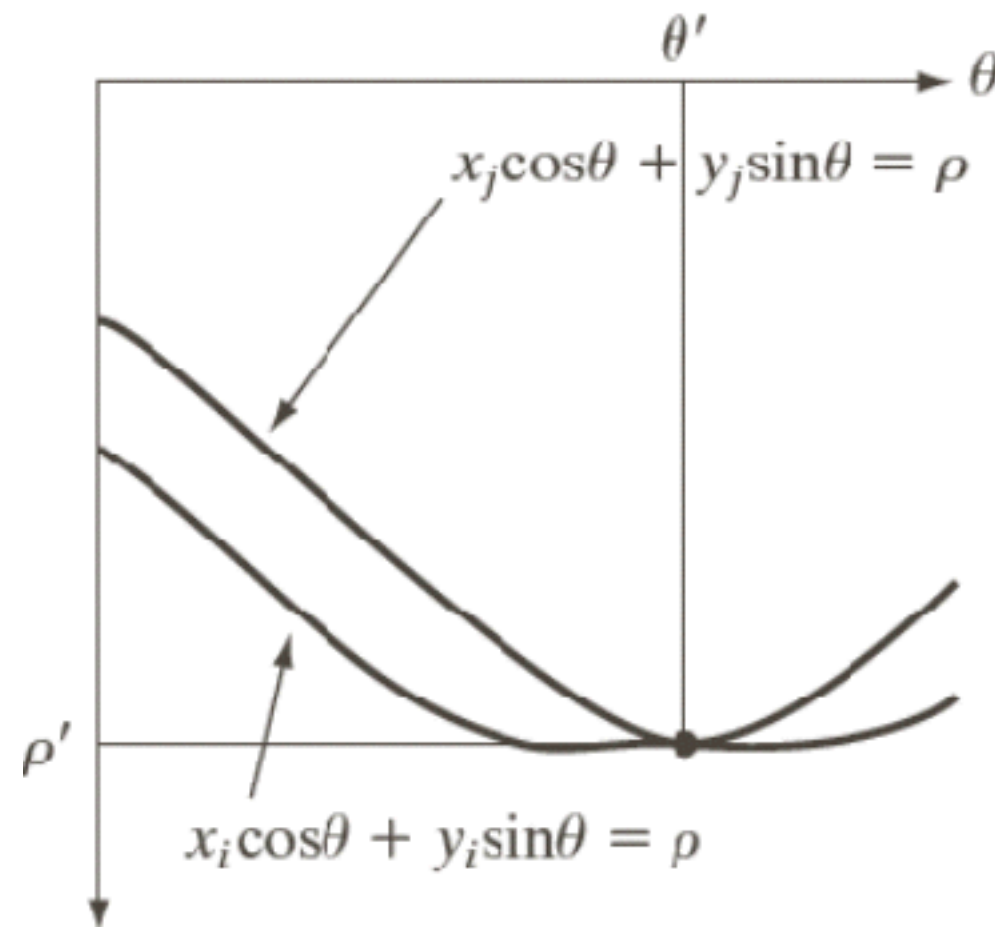
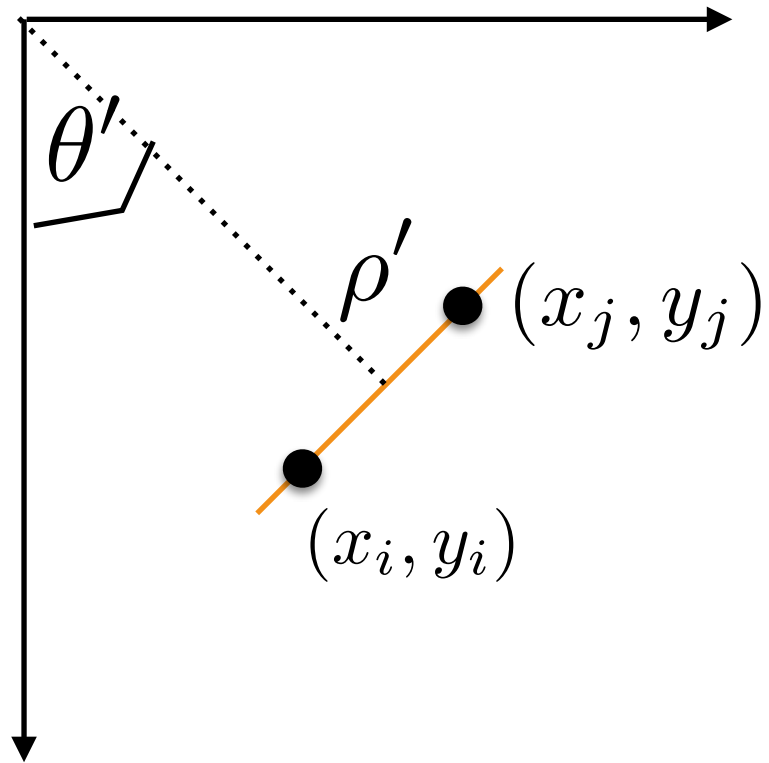
- In practice, we use the polar representation of lines

$$x_i \cos \theta + y_i \sin \theta = \rho$$

- Instead of *ab*-space, we thus have a (ρ, θ) space (Hough space)
- We will discretize the (ρ, θ) space to create an accumulator matrix A



Hough transform

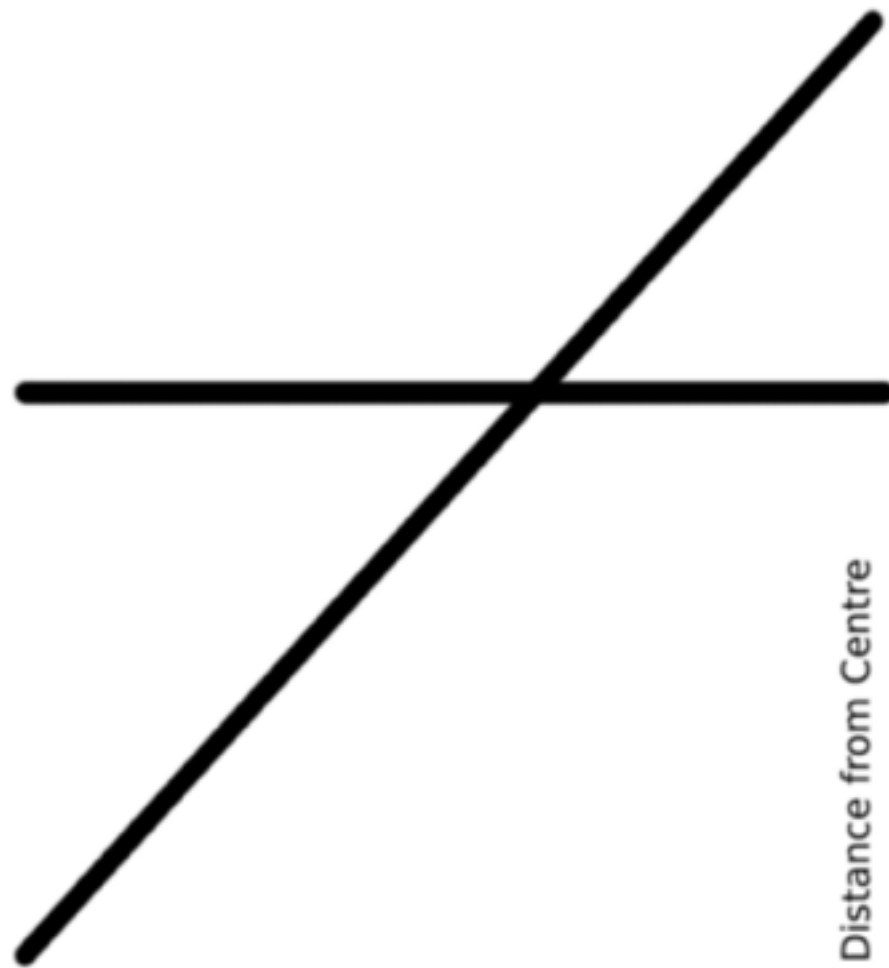


Hough transform

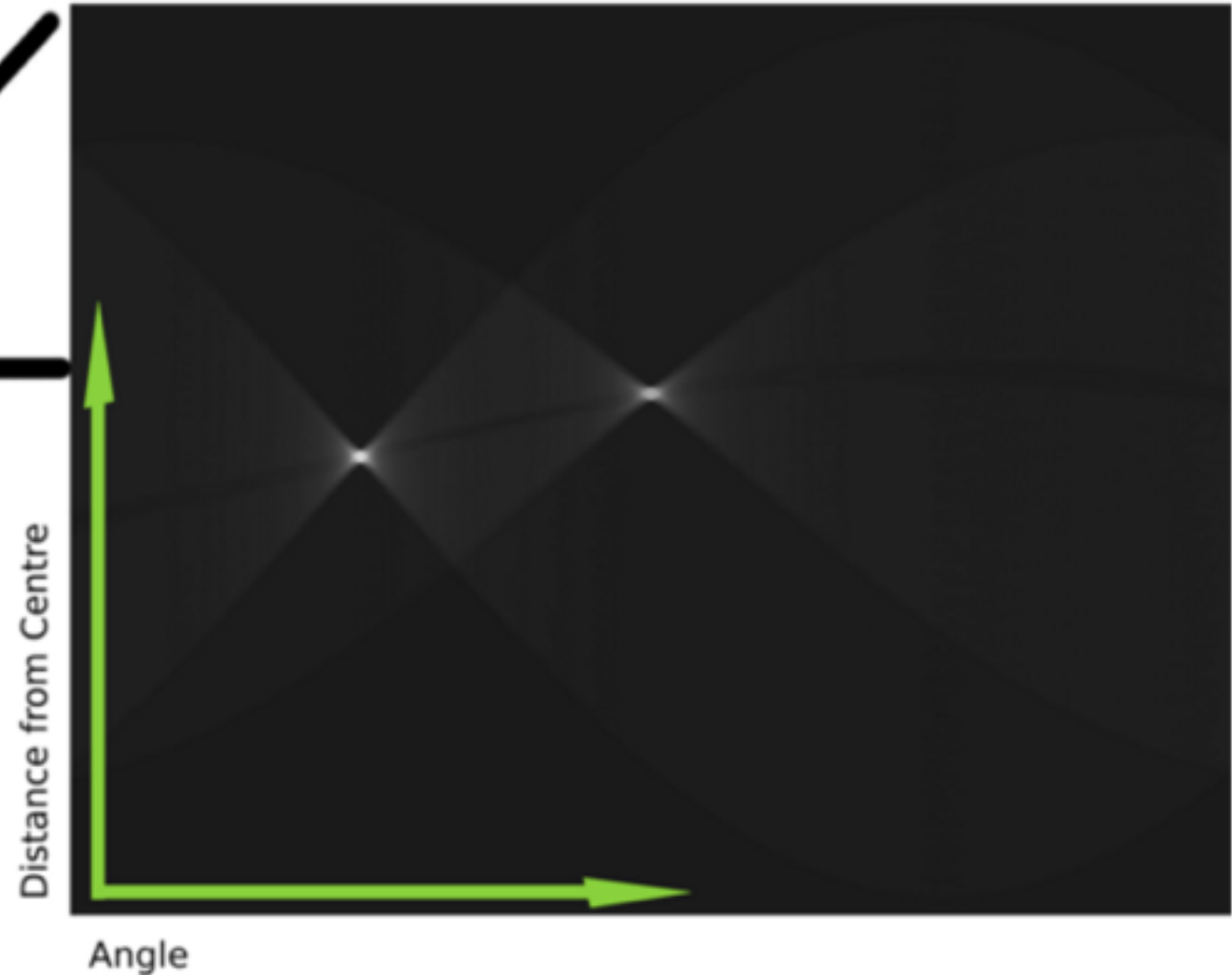
- Discretize (ρ, θ) space to create accumulator matrix A , where $A(i,j)$ is a cell corresponding to i -th ρ value and j -th θ value
 - range of ρ ?
 - range of θ ?
- Initialize: $A(i,j) = 0$ for all i and j
- For each pixel (x,y) :
 - vary θ through all possible values and compute ρ
 - For each (ρ_j, θ_j) pair, increment cell $A(i,j)$
- Find the largest values in Hough space

Hough transform

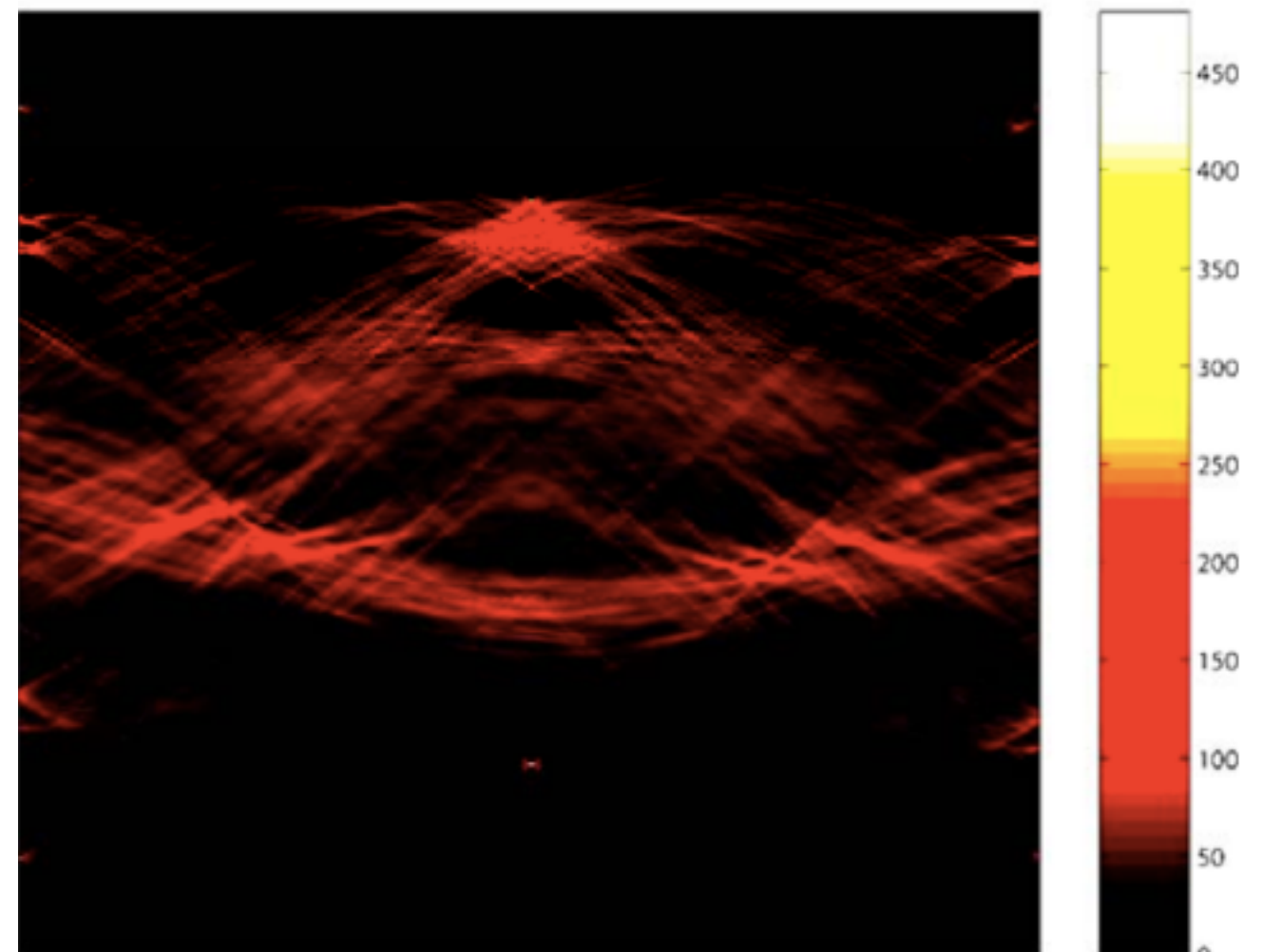
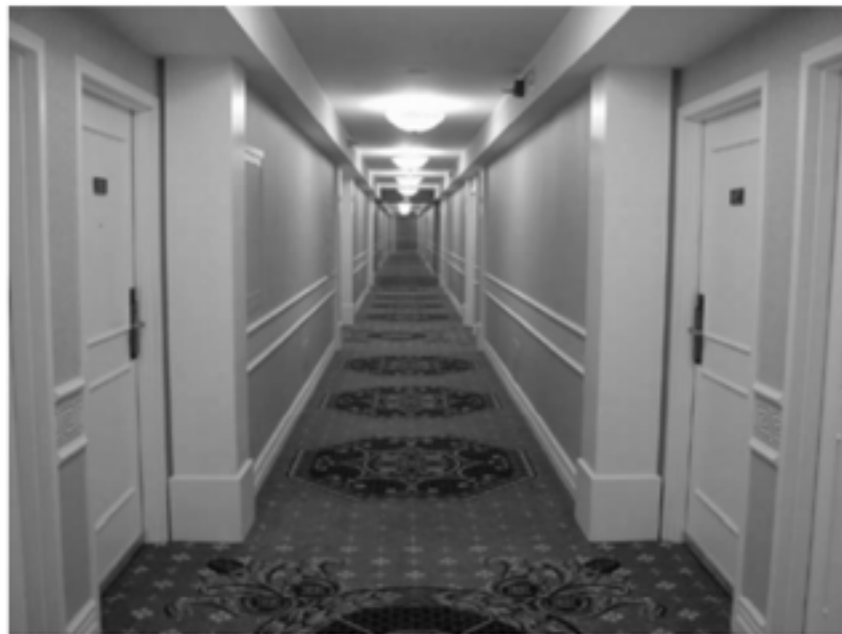
Input Image



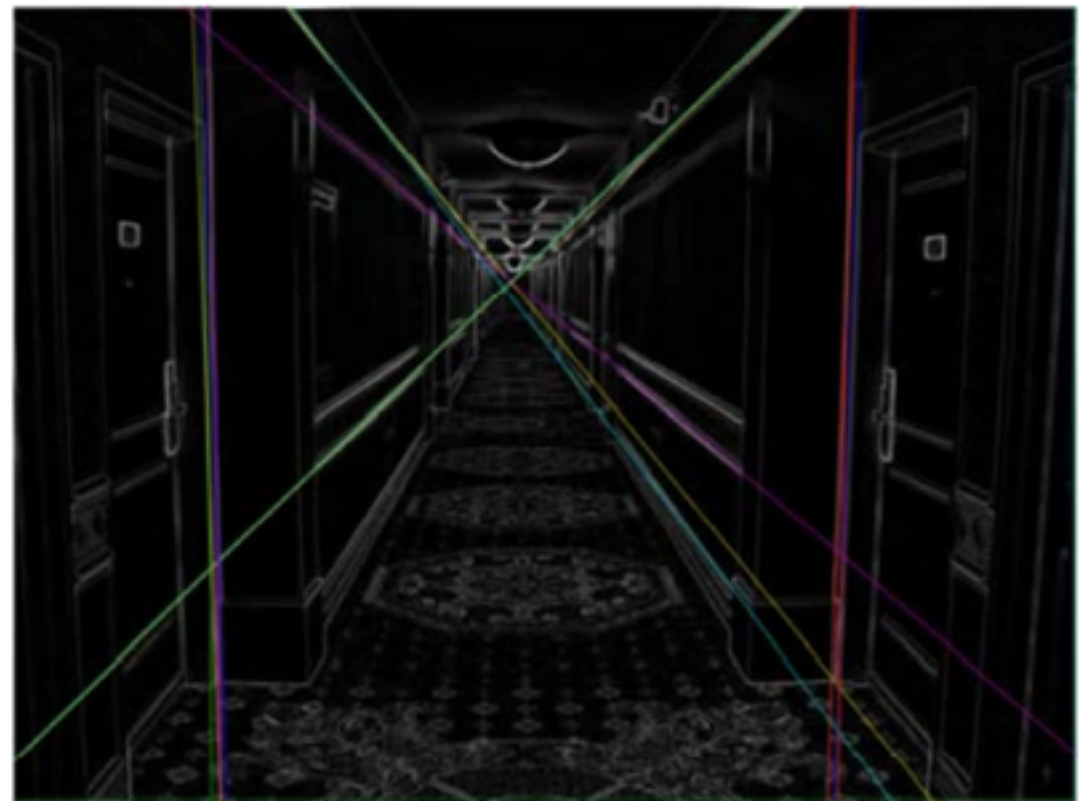
Rendering of Transform Results



Hough transform



Hough transform



Hough Transform

- Robust to occlusion, noise, missing data
- Simple implementation
- Can be extended to find other parametric shapes
- Complex shapes will give rise to high-dimensional Hough space - computationally expensive.
- Looks for only one type of shape at a time
- Co-linear line segments can not be separated