

# Lecture 12b

# Parametric Transforms

sec. 8.5.2 & ch. 11 of *Machine Vision* by Wesley E. Snyder & Hairong Qi

Spring 2024

16-725 (CMU RI) : BioE 2630 (Pitt)

Dr. John Galeotti



The content of these slides by John Galeotti, © 2012 - 2024 Carnegie Mellon University (CMU), was made possible in part by NIH NLM contract# HHSN276201000580P, and is licensed under a Creative Commons Attribution-NonCommercial 3.0 Unported License. To view a copy of this license, visit <http://creativecommons.org/licenses/by-nc/3.0/> or send a letter to Creative Commons, 171 2nd Street, Suite 300, San Francisco, California, 94105, USA. Permissions beyond the scope of this license may be available either from CMU or by emailing [itk@galeotti.net](mailto:itk@galeotti.net).  
**The most recent version of these slides may be accessed online via <http://itk.galeotti.net/>**

# Snyder ch. 11: Parametric Transforms

- Goal: Detect geometric features in an image
- Method: Exchange the role of variables and parameters
- References: Snyder 11 & ITK Software Guide book 2, 4.4

# Geometric Features?

- For now, think of geometric features as shapes that can be graphed from an equation.
- Line:  $y = mx + b$
- Circle:  $R^2 = (x - x_{\text{center}})^2 + (y - y_{\text{center}})^2$

(variables are shown in **bold purple**, parameters are in black)

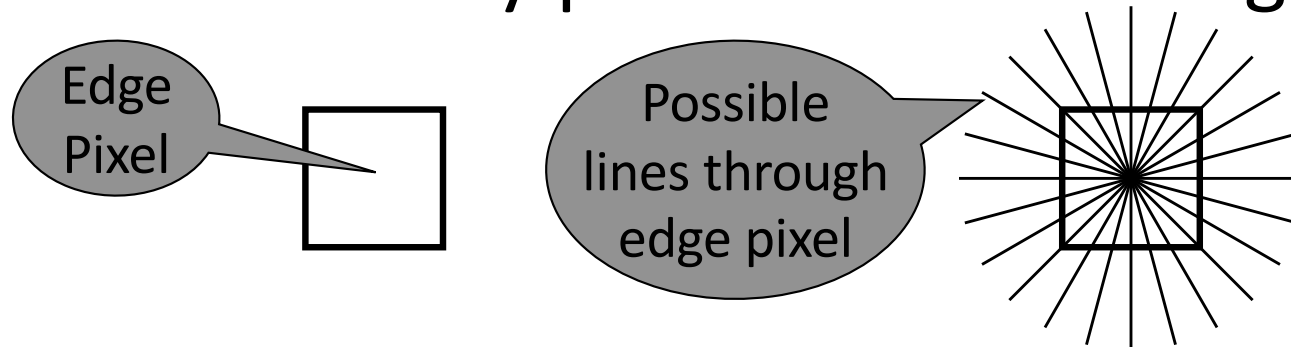
# Why Detect Geometric Features?

- Guide segmentation methods
  - Automated initialization!
- Prepare data for registration methods
- Recognize anatomical structures

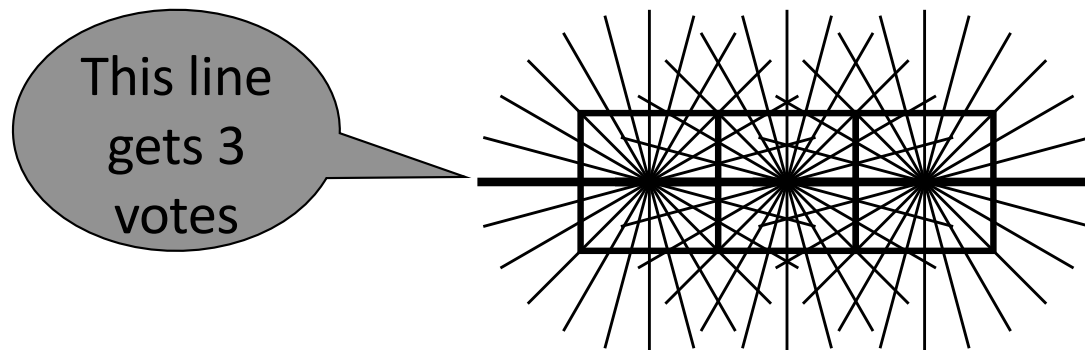
From the ITK Software Guide v 2.4, by Luis Ibáñez, et al., p. 596

# How do we do this again?

- Actually, each edge pixel “votes”
- If we are looking for lines, each edge pixel votes for every possible line through itself:



- Example: 3 collinear edge pixels:



# How to Find All Possible Shapes for each Edge Pixel

- Exchange the role of variables and parameters:
- Example for a line:  $y = \mathbf{mx} + \mathbf{b}$   
(variables are shown in **bold purple**)
- Each edge pixel in the image:
  - Has its own  $(x, y)$  coordinates
  - Establishes its own equation of  $(\mathbf{m}, \mathbf{b})$

This is the set of all possible shapes through that edge point

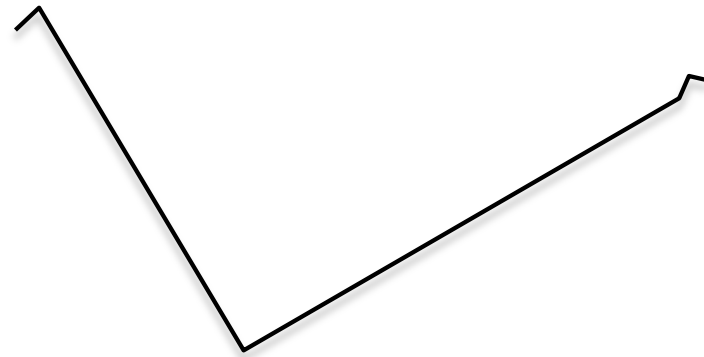
# How to Implement Voting

- With an accumulator
  - Think of it as an image in parameter space
  - Its axes are the new variables (which were formally parameters)
  - But, writing to a pixel increments (rather than overwriting) that pixel's value.
- Graph each edge pixel's equation on the accumulator (in parameter space)
- Maxima in the accumulator are located at the parameters that fit the shape to the image.

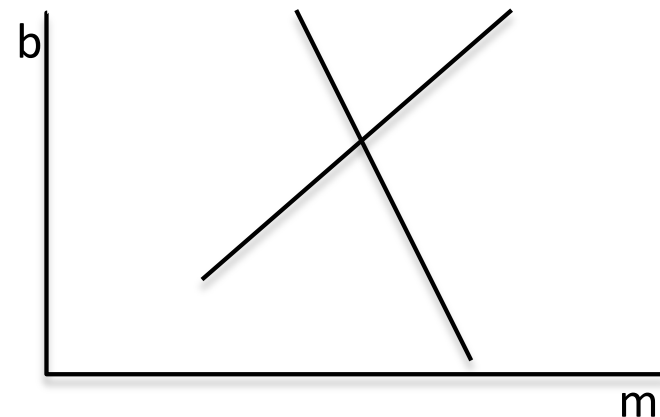
# Example 1: Finding Lines

- If we use  $y = mx + b$
- Then each edge pixel results in a line in parameter space:  
 $b = -mx + y$

Edge Detection Results  
(contains 2 dominant line segments)



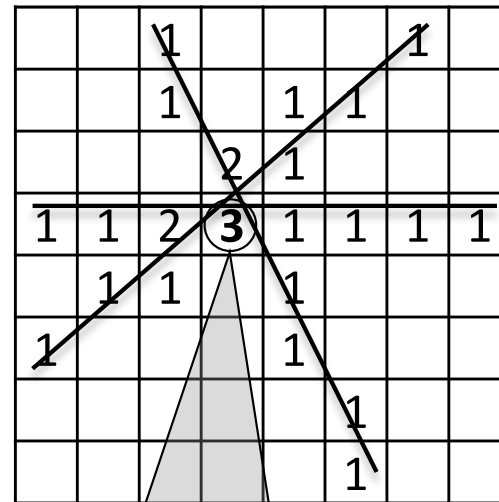
Accumulator Intermediate Result  
(after processing 2 edge pixels)





# Example 1: Finding Lines

- A closer look at the accumulator after processing 2 and then 3 edge pixels
- The votes from each edge pixel are graphed as a line in parameter space
- Each accumulator cell is incremented each time an edge pixel votes for it
  - I.e., each time a line in parameter space passes through it



Each of these edge pixels could have come from this line

# Example 2: Finding Lines... A Better Way

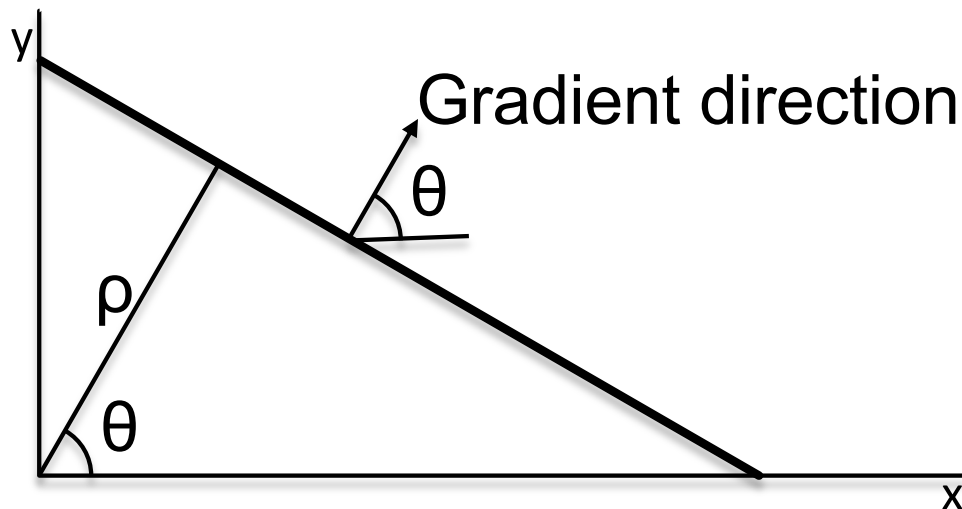
- What's wrong with the previous example?
  - Consider vertical lines:  $m = \infty$
  - My computer doesn't like infinite-width accumulator images. Does yours?
- For parametric transforms, we need a different line equation, one with a bounded parameter space.

# Example 2: Finding Lines... A Better Way

- A better line equation for parameter voting:

$$\rho = x \cos \theta + y \sin \theta$$

- $\rho \leq$  the input image diagonal size
  - But, to make math easy,  $\rho$  can be - too.
- $\theta$  is bounded within  $[0, 2\pi]$



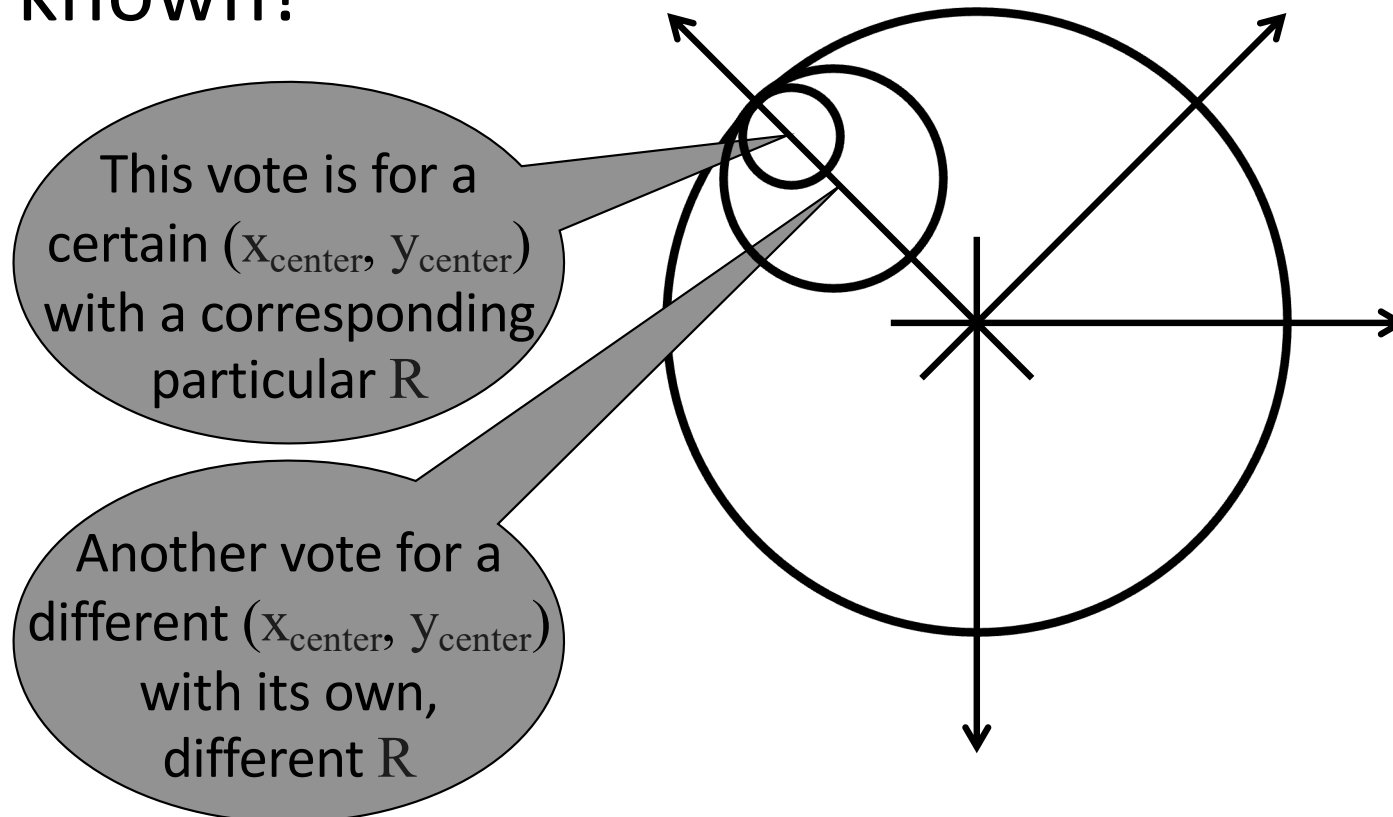
See *Machine Vision* Fig. 11.5 for example of final accumulator for 2 noisy lines

# Computational Complexity

- This can be really slow
  - Each edge pixel yields a lot of computation
  - The parameter space can be huge
- Speed things up:
  - Only consider parameter combinations that make sense...
  - Each edge pixel has an apx. direction attached to its gradient, after all.

# Example 3: Finding Circles

- Equation:  $R^2 = (x - x_{\text{center}})^2 + (y - y_{\text{center}})^2$
- Must vote for 3 parameters if  $R$  is not known!



# Example 4: General Shapes

- What if our shape is weird, but we can draw it?
  - Being able to draw it implies we know how big it will be
- See Snyder 11.4 for details
- Main idea:
  - For each boundary point, record its coordinates in a local reference frame (e.g., at the shape's center-of-gravity).
  - Itemize the list of boundary points (on our drawing) by the direction of their gradient