

Lecture 12b

Parametric Transforms

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

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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)

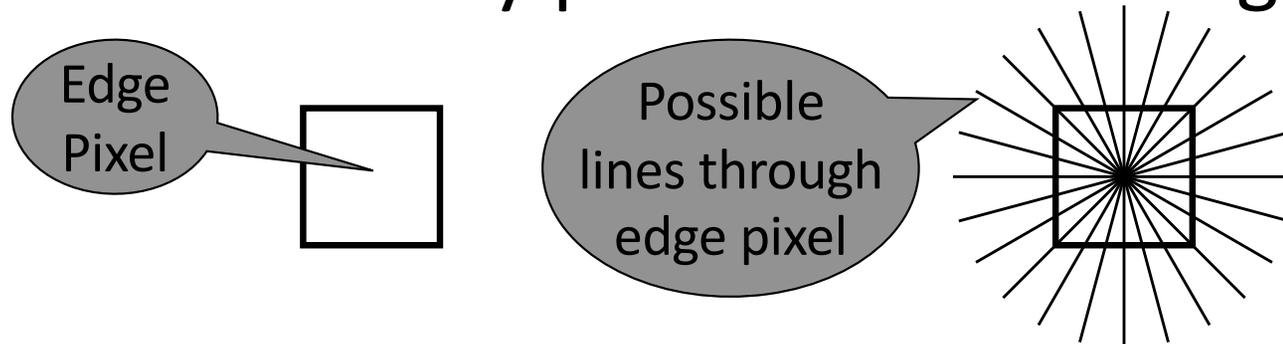
Why Detect Geometric Features?

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

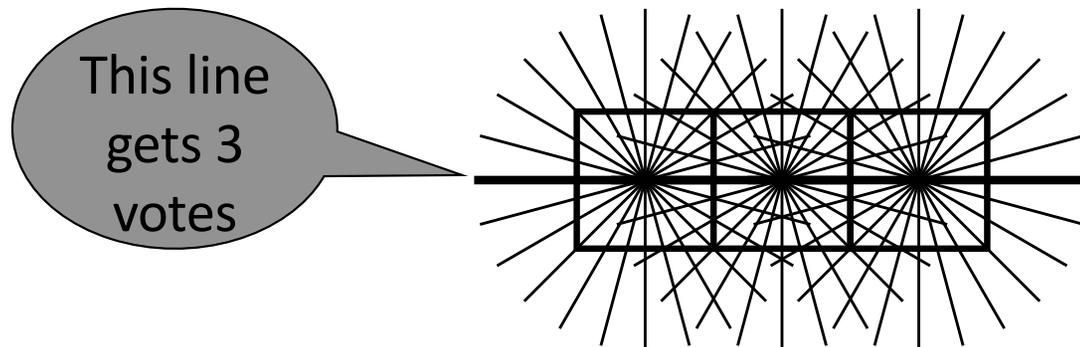
From the ITK Software Guide v 2.4, by Luis Ibáñez, et al., p. 596

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

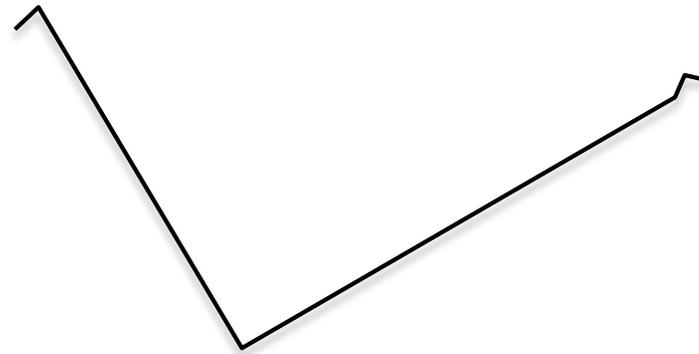
How to Implement Voting

- 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.

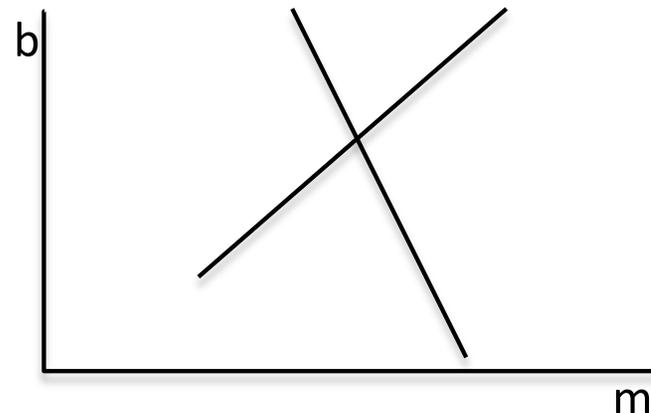
Example 1: Finding Lines

- 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)

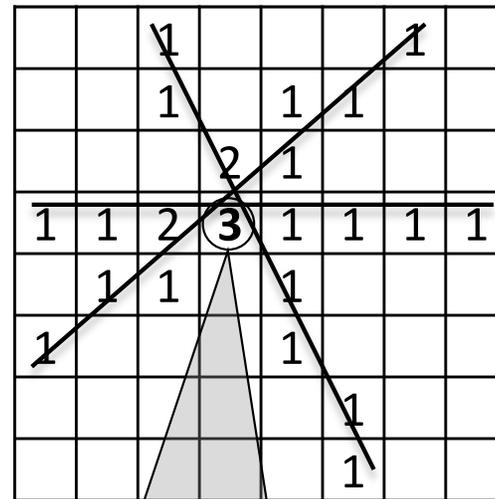


Accumulator Intermediate Result
(after processing 2 edge pixels)



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

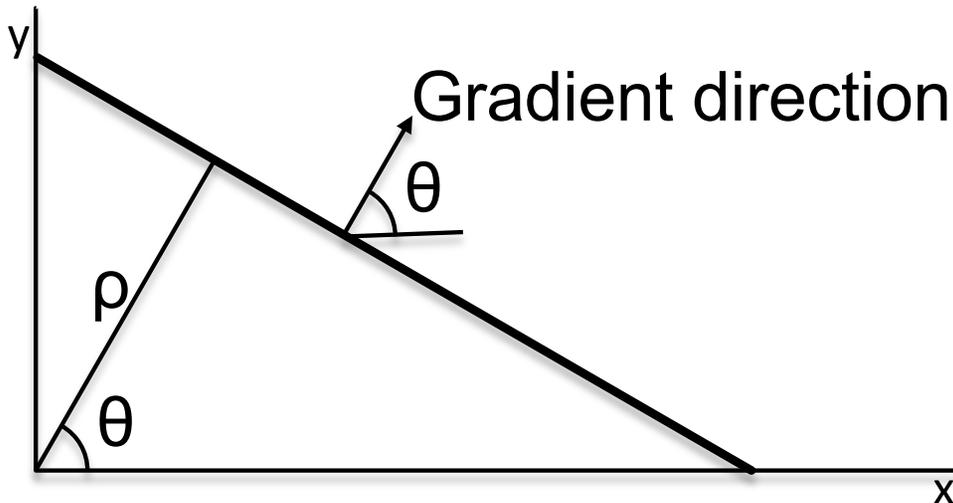
- What's wrong with the previous example?
 - Consider vertical lines: $m = \infty$
 - My computer doesn't like infinite-width accumulator images. Does yours?
- For parametric transforms, we need a different line equation, one with a bounded parameter space.

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, ρ can be - too.
- θ is bounded within $[0, 2\pi]$



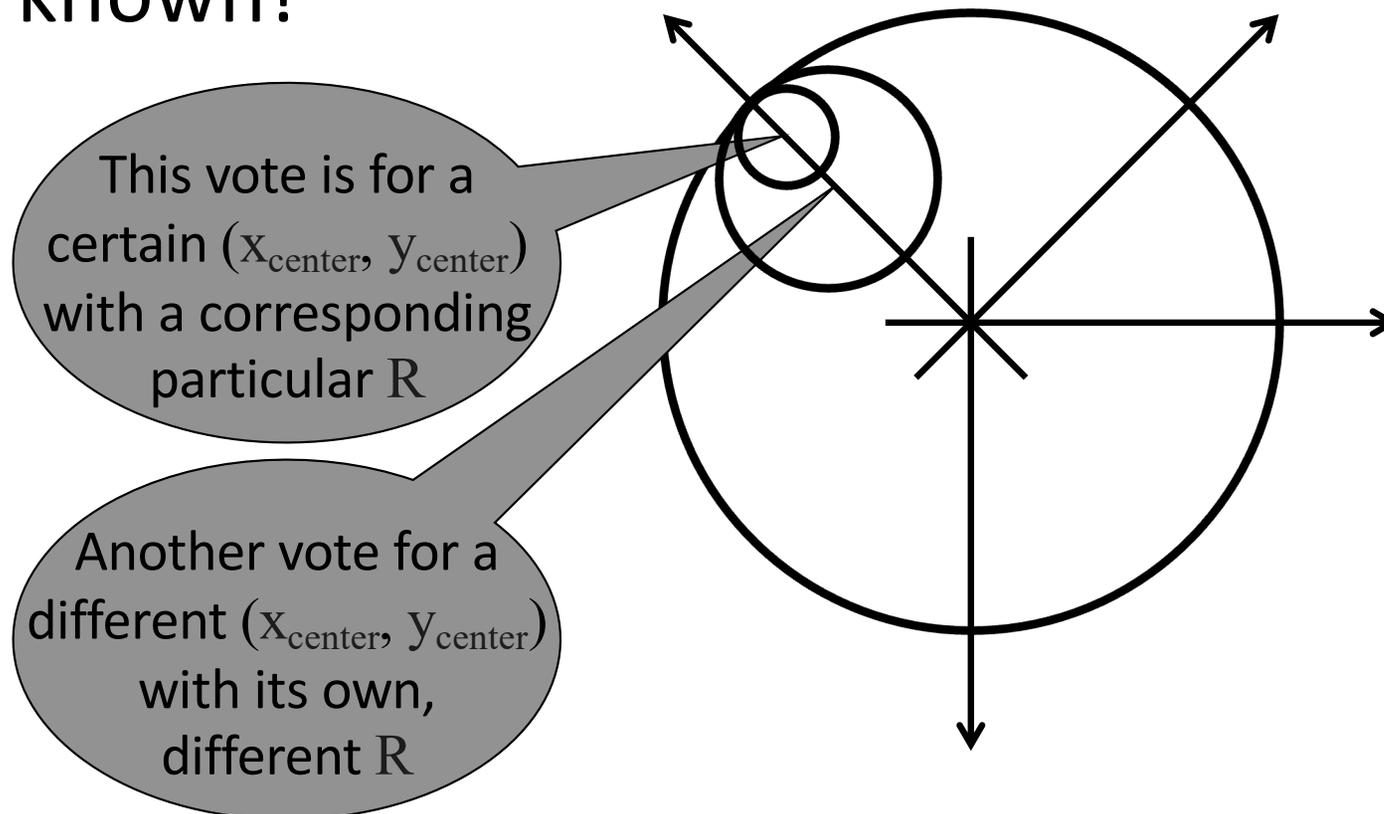
See *Machine Vision* Fig. 11.5 for example of final accumulator for 2 noisy lines

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