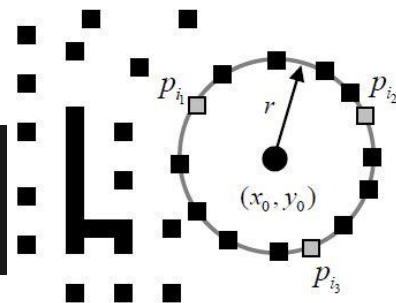


Mobile Device Camera Connector (Tabiscope)

Circle Detection with Learning Automata (LA)

Paper Seminar by Kyle Wong

600.446 Computer Integrated Surgery II
Project 7

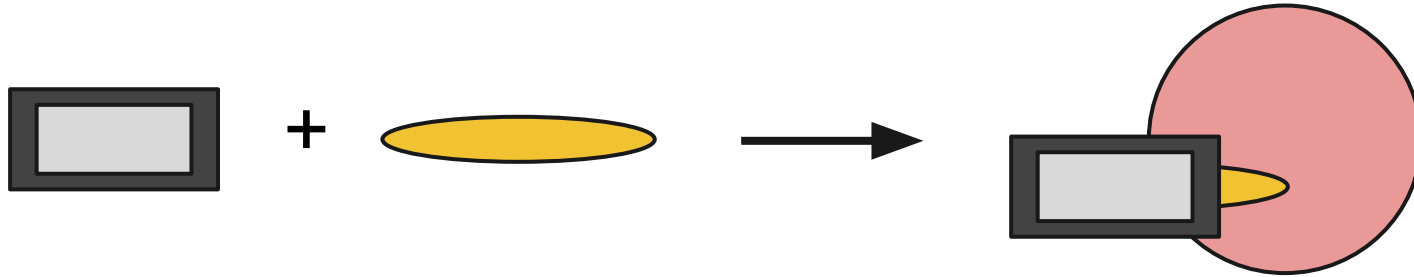


Partners: Daniel Ahn, Deepak Lingam
Mentors: Dr. Amit Kochhar, Kevin Olds



Project Overview

- Design a low cost endoscopic adapter

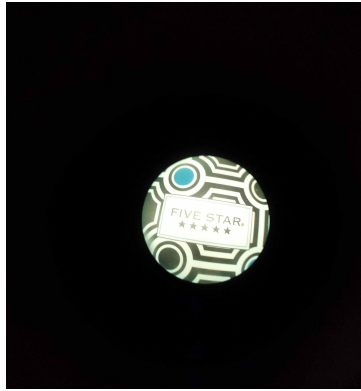


- Create a system for Android devices



Design Challenge

- Real-time image processing method for Circle Detection for auto-zoom, auto-focus, auto-brightness etc.



goal:
automatic



Paper Selection

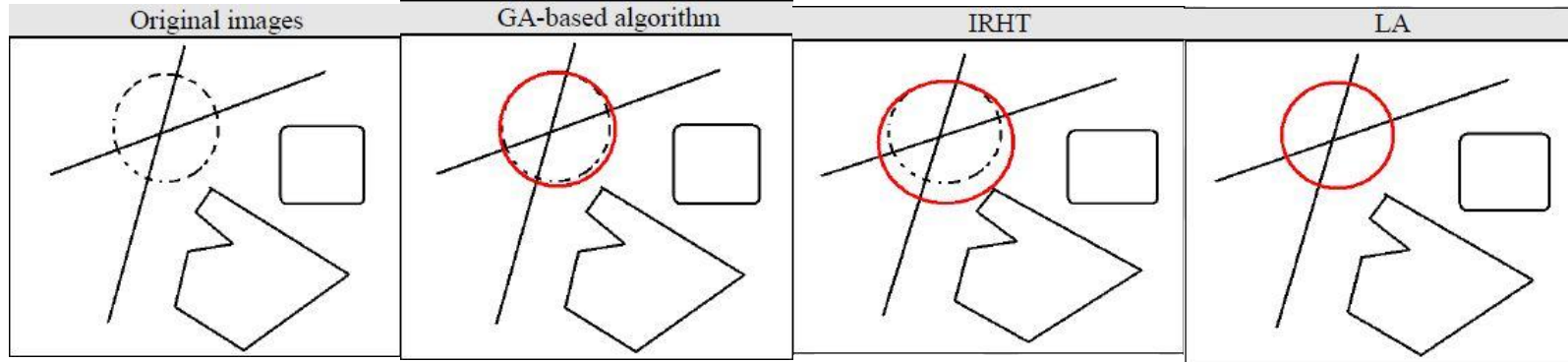
Cuevas, E., Wario, F., Zaldivar, D., & Pérez-Cisneros, M. (2013). **Circle detection on images using learning automata.** In *Artificial Intelligence, Evolutionary Computing and Metaheuristics* (pp. 545-570). Springer Berlin Heidelberg



Cuevas '13. *Circle detection on images using learning automata.*

Summary of Problem and Results

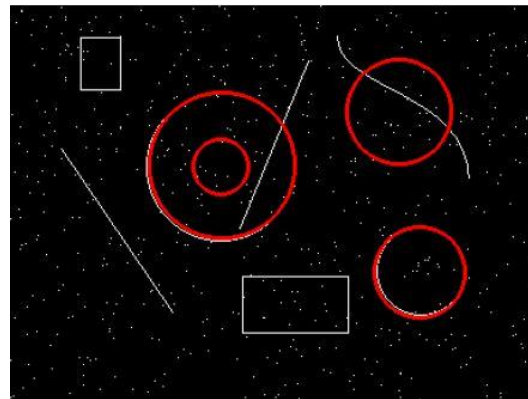
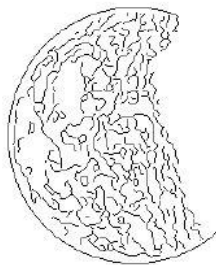
- Circle detection quickly and accurately
- Create a set of synthetic and natural images for comparison
- Compare Learning Automata (LA) with Iterative Randomised Hough Transform (IRHT) and Genetic Algorithms (GA)



Cuevas '13. Circle detection on images using learning automata.

Significance

- LA algorithm works for occluded circles and multiple circles and is robust to noise and fast → widespread use



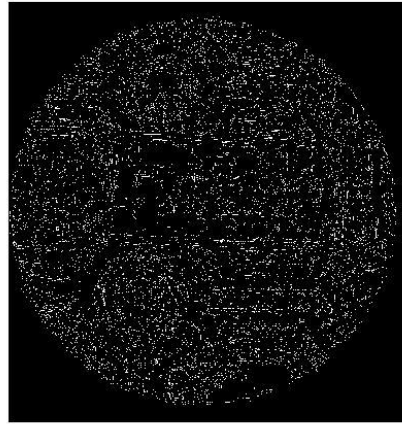
Cuevas '13. *Circle detection on images using learning automata.*

Background - LA

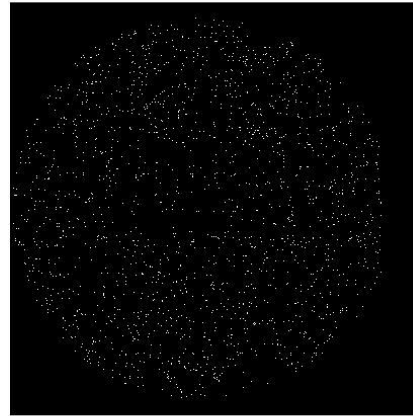
- **Learning** Automata (LA) optimize through a **learning** process
- Model actions applied to an environment with a probability density function (pdf)
- Pair actions with reinforcement signal to update the pdf to select the next action
- Iterate until optimal action is found (threshold reached, or number of iterations is done)

Background - Preprocess

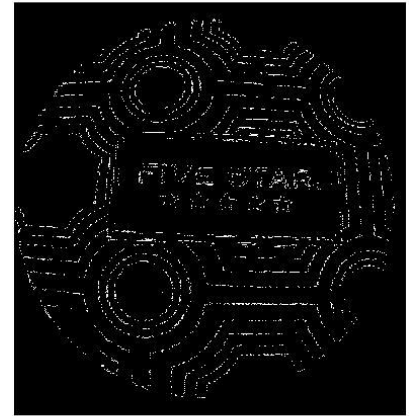
- pre-process with Canny Algorithm to get single-pixel edge map; take only a fraction (about 5%) randomly



Matlab - Canny Method



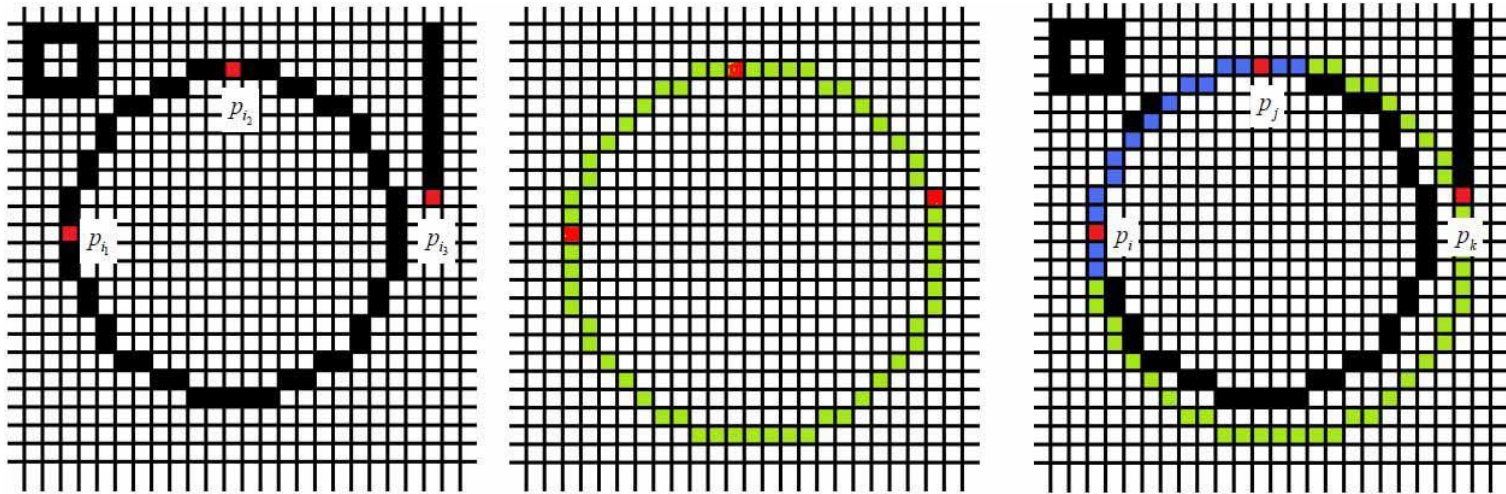
Random Sample



Matlab - Sobel Method

Background - Sample

- sample combinations of 3 points and check



Cuevas '13. Circle detection on images using learning automata.

Background 3 Circle-find

- Circle calculation

$$x_0 = \frac{\det(\mathbf{A})}{4((x_{i_2} - x_{i_1})(y_{i_3} - y_{i_1}) - (x_{i_3} - x_{i_1})(y_{i_2} - y_{i_1}))}$$

$$y_0 = \frac{\det(\mathbf{B})}{4((x_{i_2} - x_{i_1})(y_{i_3} - y_{i_1}) - (x_{i_3} - x_{i_1})(y_{i_2} - y_{i_1}))}$$

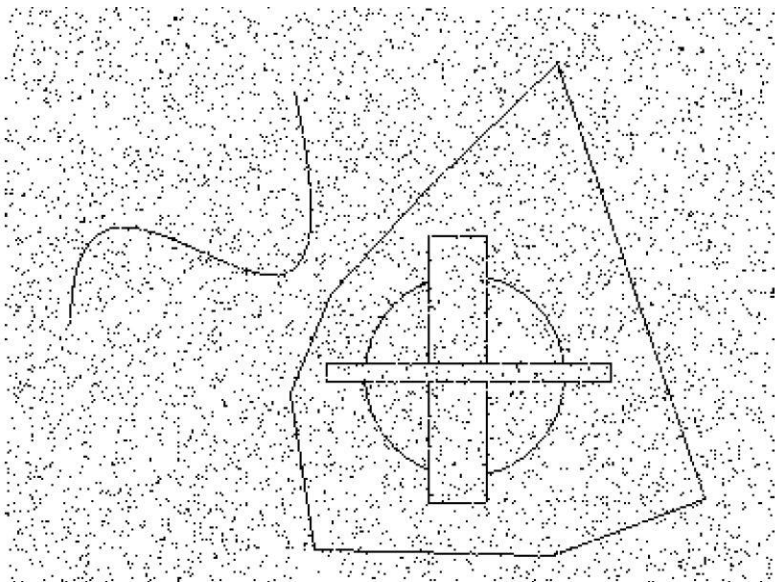
$$\mathbf{A} = \begin{bmatrix} x_{i_2}^2 + y_{i_2}^2 - (x_{i_1}^2 + y_{i_1}^2) & 2(y_{i_1} - y_{i_1}) \\ x_{i_3}^2 + y_{i_3}^2 - (x_{i_1}^2 + y_{i_1}^2) & 2(y_{i_3} - y_{i_1}) \end{bmatrix}$$

$$\mathbf{B} = \begin{bmatrix} 2(x_{i_2} - x_{i_1}) & x_{i_2}^2 + y_{i_2}^2 - (x_{i_1}^2 + y_{i_1}^2) \\ 2(x_{i_3} - x_{i_1}) & x_{i_3}^2 + y_{i_3}^2 - (x_{i_1}^2 + y_{i_1}^2) \end{bmatrix}$$

$$r = \sqrt{(x_0 - x_d)^2 + (y_0 - y_d)^2}$$

Methods

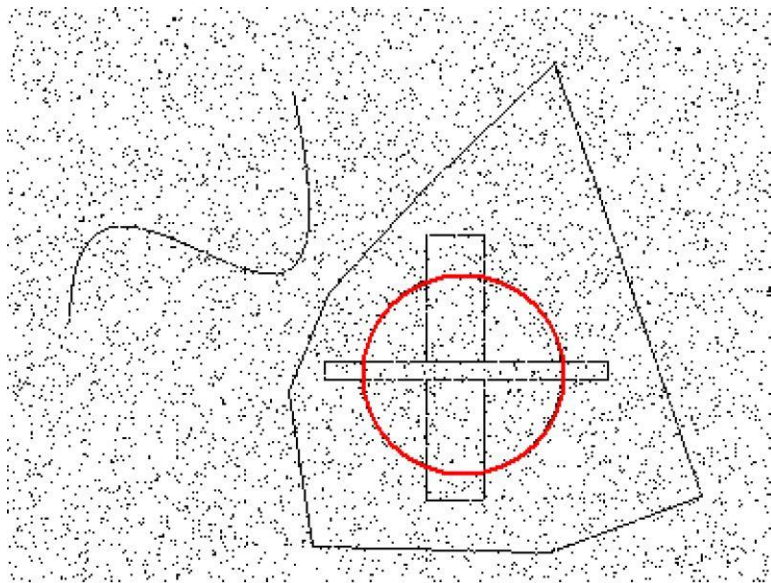
- Synthetic images with noise
- Real life images



Cuevas '13. *Circle detection on images using learning automata.*

Methods - LA

- Synthetic images with noise
- Real life images



Cuevas '13. *Circle detection on images using learning automata.*

Results

Image	Averaged execution time \pm standard deviation, s			Averaged $E_s \pm$ standard deviation		
	GA	IRHT	LA	GA	IRHT	LA
<i>Synthetic images</i>						
(a)	2.23 \pm (0.41)	1.71 \pm (0.51)	0.21 \pm (0.22)	0.41 \pm (0.044)	0.33 \pm (0.052)	0.22 \pm (0.033)
(b)	3.15 \pm (0.39)	2.80 \pm (0.65)	0.36 \pm (0.24)	0.51 \pm (0.038)	0.37 \pm (0.032)	0.26 \pm (0.041)
(c)	3.02 \pm (0.63)	4.11 \pm (0.71)	0.64 \pm (0.19)	0.71 \pm (0.036)	0.77 \pm (0.044)	0.42 \pm (0.011)
<i>Natural images</i>						
(a)	2.02 \pm (0.32)	3.11 \pm (0.41)	0.31 \pm (0.12)	0.45 \pm (0.051)	0.41 \pm (0.029)	0.25 \pm (0.037)
(b)	2.11 \pm (0.31)	3.04 \pm (0.29)	0.57 \pm (0.13)	0.87 \pm (0.071)	0.71 \pm (0.051)	0.54 \pm (0.071)
(c)	2.50 \pm (0.39)	2.80 \pm (0.17)	0.51 \pm (0.11)	0.67 \pm (0.081)	0.61 \pm (0.048)	0.31 \pm (0.015)

Cuevas '13. Circle detection on images using learning automata.

Results - LA Best

Image	Averaged execution time \pm standard deviation, s			Averaged $E_s \pm$ standard deviation		
	GA	IRHT	LA	GA	IRHT	LA
<i>Synthetic images</i>						
(a)	2.23 \pm (0.41)	1.71 \pm (0.51)	0.21 \pm (0.22)	0.41 \pm (0.044)	0.33 \pm (0.052)	0.22 \pm (0.033)
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Cuevas '13. Circle detection on images using learning automata.

Assessment

Positive

- Thorough explanation and dataset results

Negative

- Self-constructed accuracy (arbitrary and biased?)

$$E_s = \eta(|x_{true} - x_D| + |y_{true} - y_D|) + \mu|r_{true} - r_D|$$

E_s = Error score

η = weight for accuracy of the center (chosen 0.05)

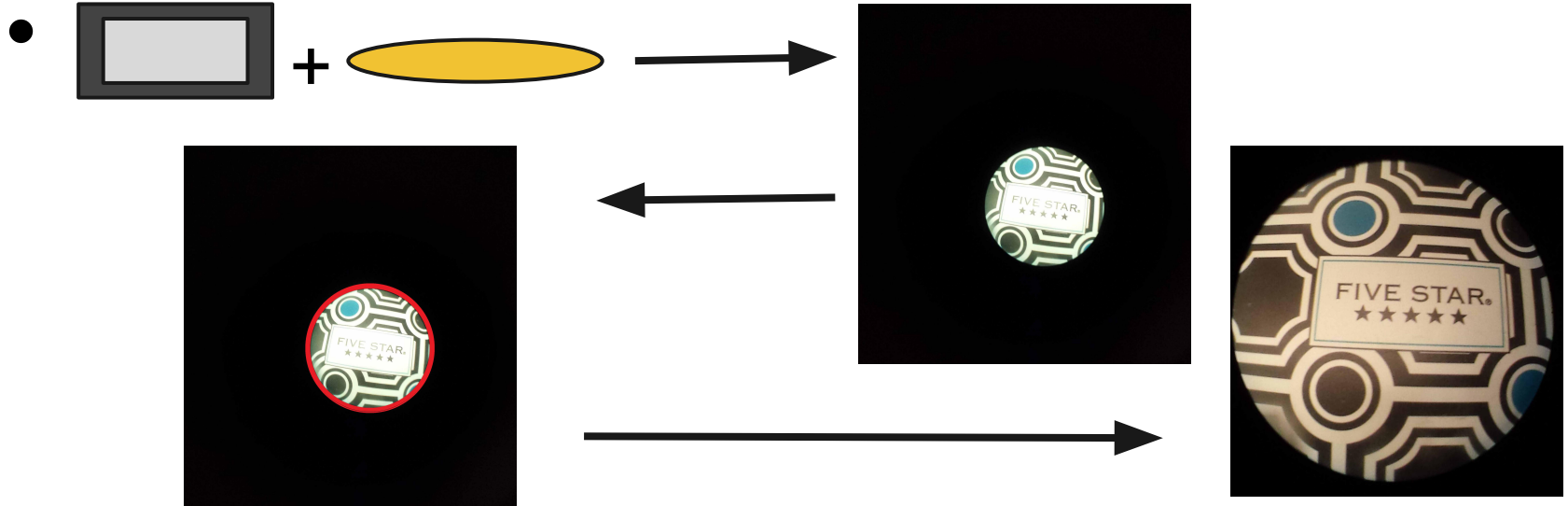
μ = weight for accuracy of the radius (chosen 0.1)

Future Work

- Compare to other methods like supposedly fast Randomized Circle Detection (accuracy and speed tradeoff)
- Compare to basic Circular Hough Transform (for a baseline)
- Implement in real-time to show capabilities (using real-time Canny edge detector)

Relevance to our Project

- Rapid and accurate circle detection in Endoscopic image

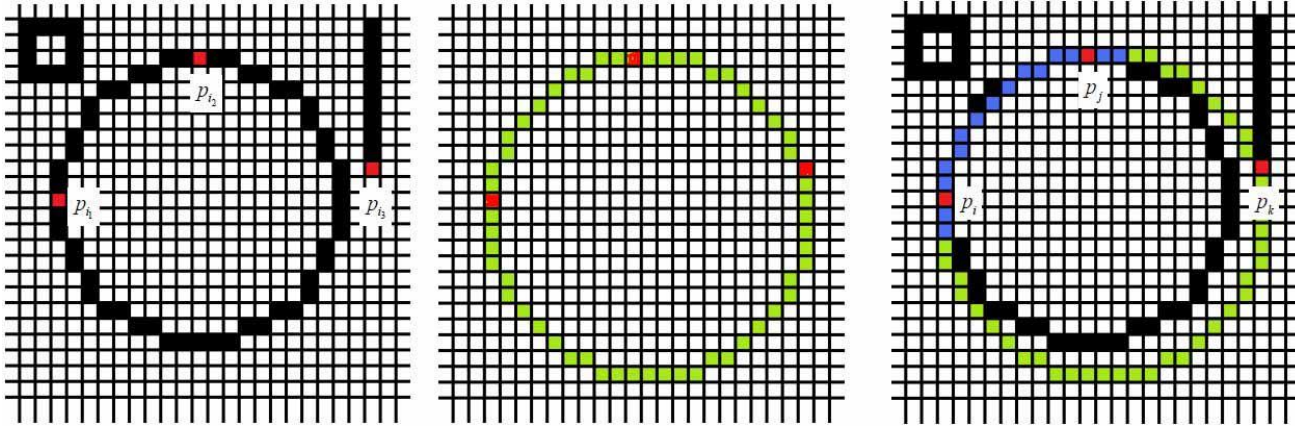


Conclusions

- Circle Detection and shape recognition are still being researched
- Speed, accuracy, memory use, and robustness are vital considerations
- Having a standard of measuring these is essential for benchmarking
 - Learning Automata for Circle Detection may be of use for our Endoscope application

Circle Detection with Learning Automata (LA)

- Questions and Feedback?



Cuevas '13. Circle detection on images using learning automata.