

Image Edge Detection Algorithm with a Single Grid System of Coupled FitzHugh-Nagumo Elements

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Key words: image processing, edge detection
application of nonlinear elements, FitzHugh-Nagumo model
initial conditions for FHN model

Outline

- Introduction: Motivation and Approach
- Background:
 - Nonlinear phenomena and pattern formation in nature
 - Image edge detection and previous algorithms in image processing
- Our Previous Edge Detection Algorithm with cFHN
 - FitzHugh-Nagumo (FHN) model
 - Grid system of coupled FHN (cFHN) and the initial conditions
- Proposed Edge Detection Algorithm with cFHN
 - The initial conditions
- Experimental Results
 - Artificial images with/without noise
 - Real images
- Conclusion

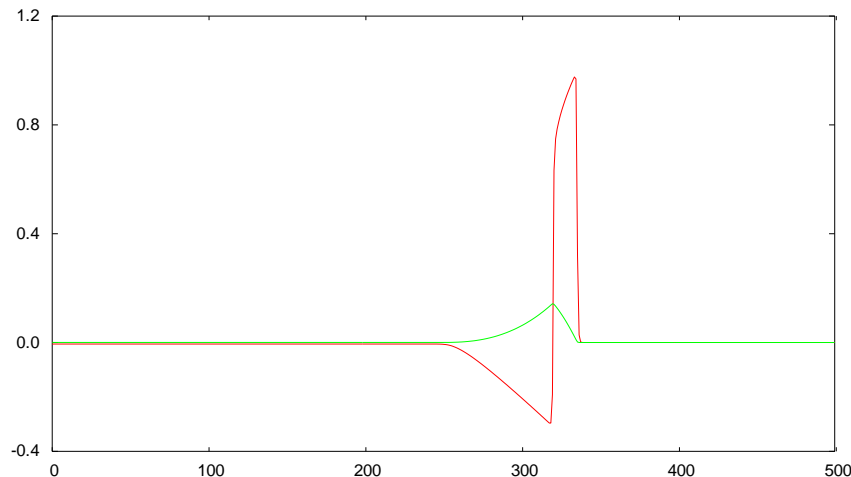
Introduction: Motivation and Approach

- Motivation
 - Biological visual system
 - Bio-inspired image processing
 - edge detection, segmentation and stereo disparity detection
- Approach
 - Coupled FitzHugh-Nagumo (cFHN) elements on a grid system
 - Reaction-diffusion system (diffusively coupled elements)
- Our previous edge detection algorithm
 - does not work for gray level image
 - noise vulnerability or not robust to noise

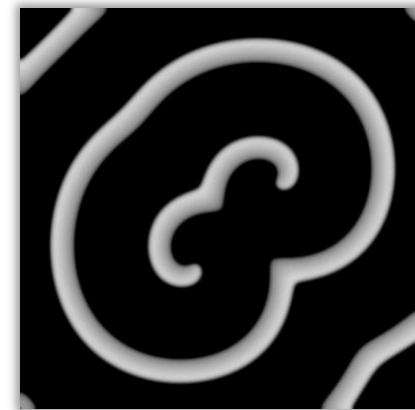
Background:

Nonlinear Phenomena & Pattern Formation

- Nonlinear elements in nature
 - Biological response to external stimuli: FitzHuhg-Nagumo model
 - Nonlinear oscillation or excitation in chemical reaction system
- Reaction-diffusion system
 - System of diffusively coupled nonlinear elements in space
 - Patterns: traveling pulses in 1D space and spiral waves in 2D space
 - Information transmission & information processing



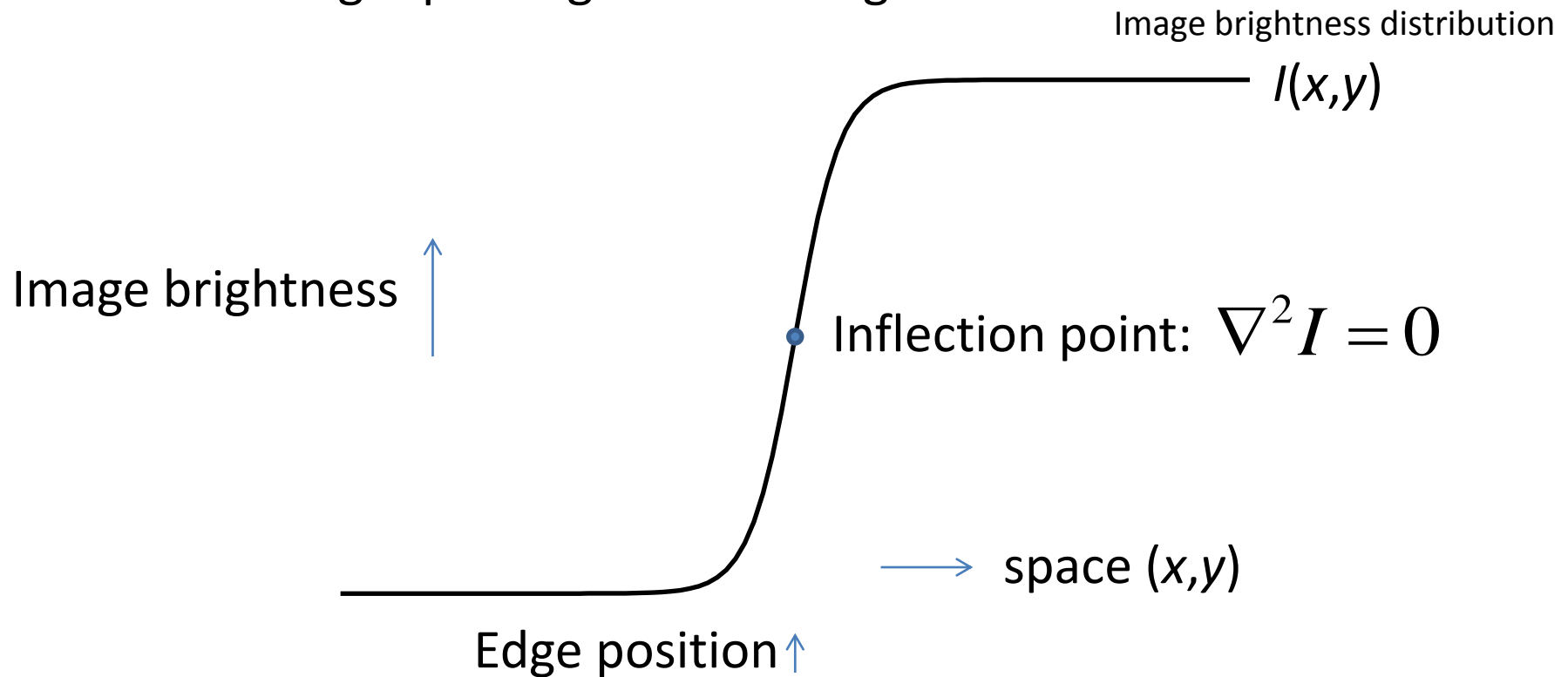
1D pulse propagating in space



2D spiral waves

Background: Definition of Image Edge

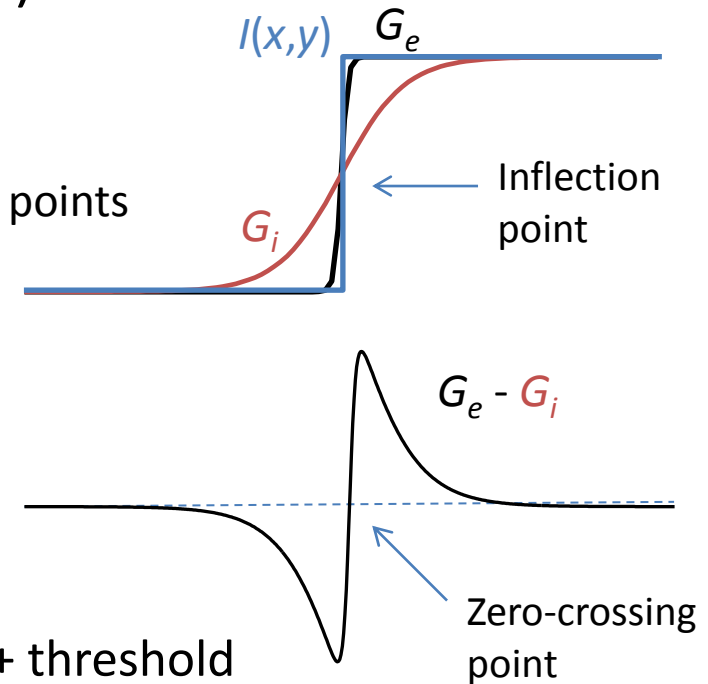
- Point having rapid brightness change



Background: Previous Edge Detection Algorithms

- Algorithms by Marr and Hildreth (1980)

- LoG: Laplacian-of-the-Gaussian
 - Gaussian: noise reduction
 - Laplacian operator: detection of inflection points
- DoG: Difference-of-two-Gaussians
 - G_e : excitation (blurred)
 - G_i : inhibition (more blurred)
- Detecting zero-crossing points



- Algorithm by Canny (1986)

- Gaussian smoothing + gradient operator + threshold
- assumption: continuity of edges

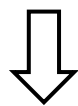
Our Previous Edge Detection Algorithm: FitzHugh-Nagumo (FHN) Model

(a) Uni-stable element

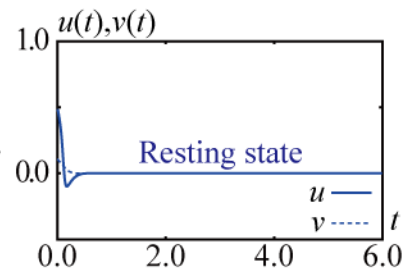
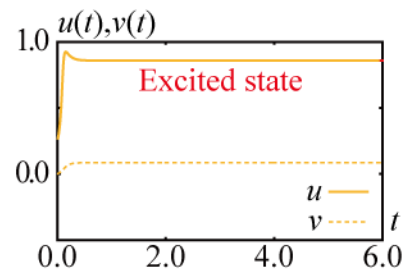
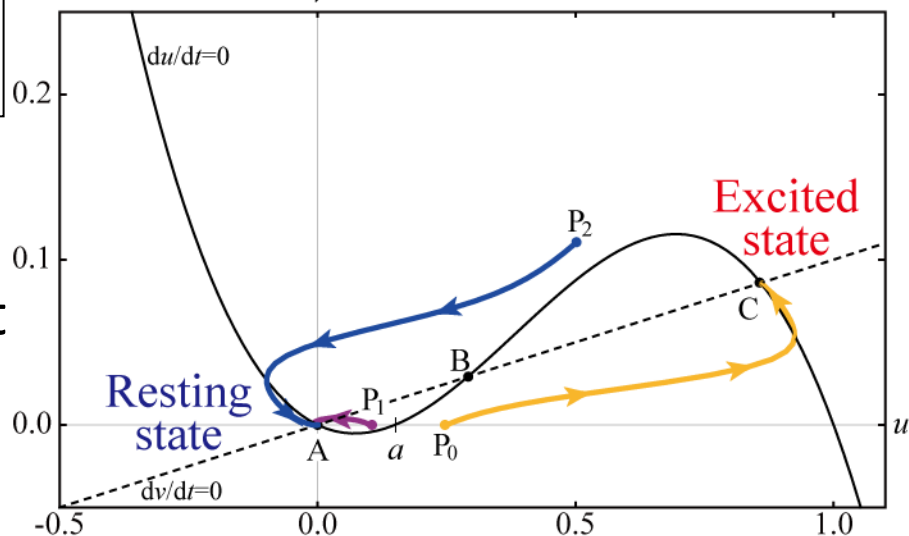
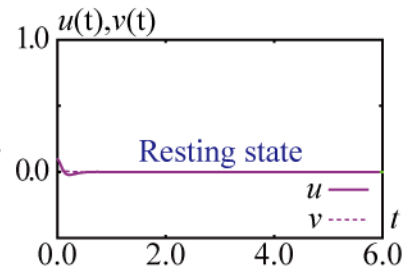
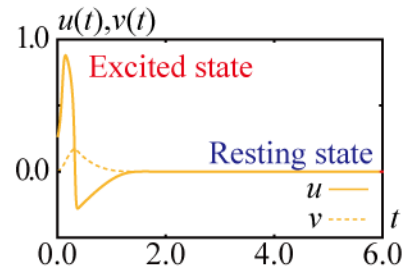
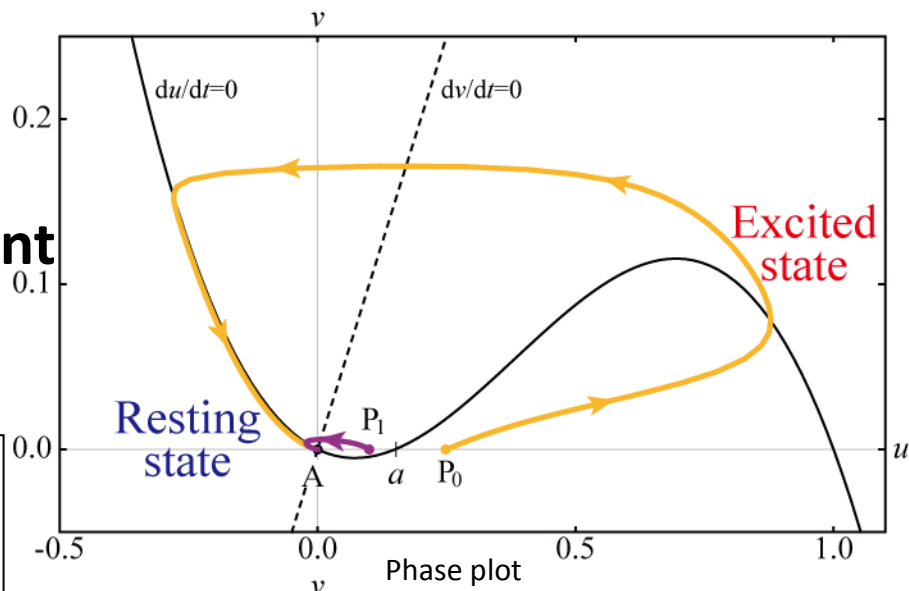
FHN model

$$\frac{du}{dt} = \frac{1}{\varepsilon} [u(u-a)(1-u) - v]$$

$$\frac{dv}{dt} = u - bv$$



(b) Bi-stable element



Our Previous Edge Detection Algorithm: Single Grid System of Coupled FHN (cFHN)

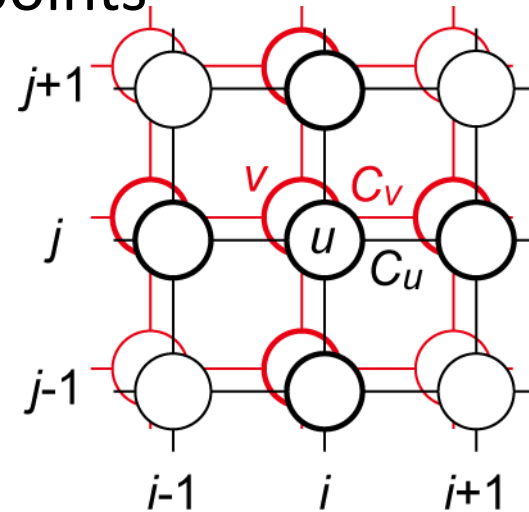
- Uni-stable elements placed at image grid points

- Nomura et al., *J. Phys. Soc. Jpn.*, 2003
- Kurata et al., *Phys. Rev. E*, 2009

$$\frac{du_{i,j}}{dt} = C_u \left[\overline{u_{i,j}} - 4u_{i,j} \right] + \frac{1}{\varepsilon} \left[u_{i,j} (u_{i,j} - a)(1 - u_{i,j}) - v_{i,j} \right]$$

$$\frac{dv_{i,j}}{dt} = C_v \left[\overline{v_{i,j}} - 4v_{i,j} \right] + u_{i,j} - bv_{i,j}$$

Spatial coupling $\overline{u_{i,j}}$ $\overline{v_{i,j}}$: averages in the nearest four points.

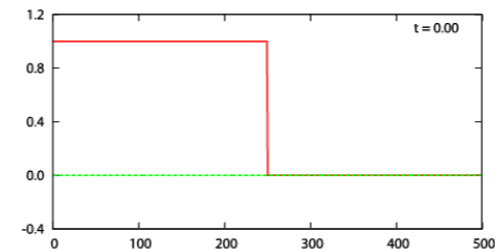


- The initial conditions:

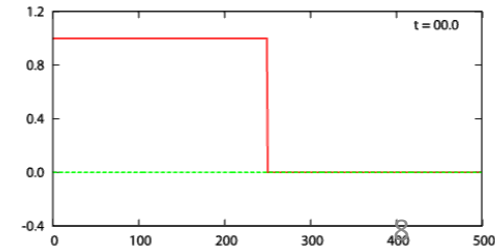
$$u_{i,j} = I_{i,j}, \quad v_{i,j} = 0$$

- Strong inhibition: $C_u \ll C_v$
 \Rightarrow Stationary pulses at edge positions
- Weak inhibition: $C_u > C_v$
 \Rightarrow Propagating pulses

u and v
($C_u \ll C_v$)



u and v
($C_u > C_v$)

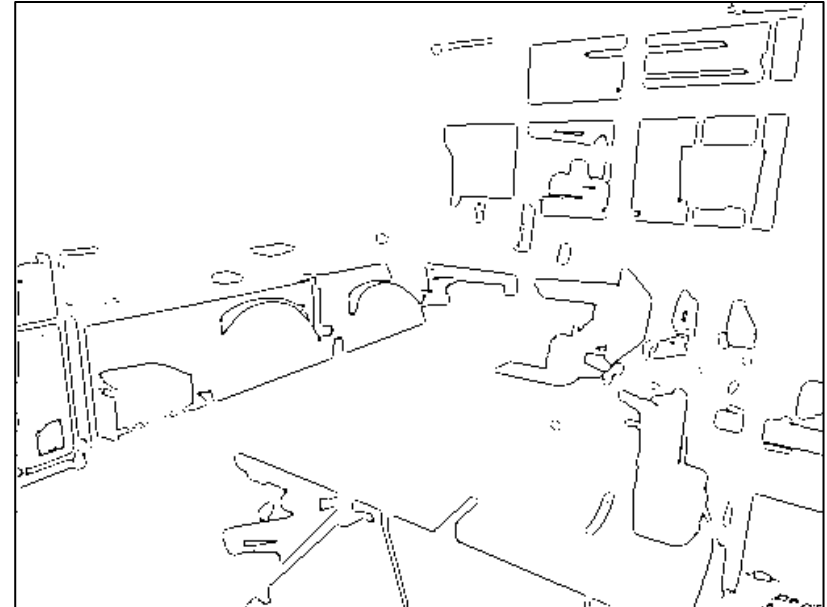


Our Previous Algorithm for Edge Detection: Example of Edge Detection with cFHN

- Example:



Initial condition of $u_{i,j}$



Result of edge detection ($\alpha=0.1$)

- Threshold for the initial condition u_0 & Self-organized pulse
=> **Previous algorithm does not work for gray level images**

Proposed Algorithm: cFHN & Initial Conditions

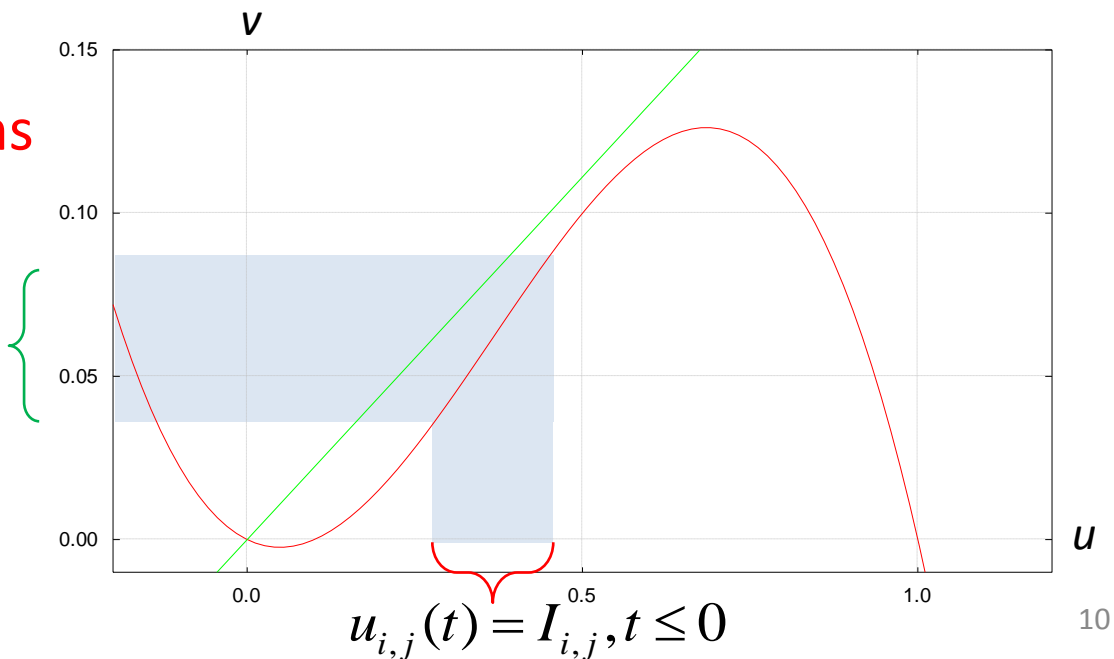
- Coupled FHN elements: delaying computation of $u_{i,j}$

$$\frac{du_{i,j}}{dt} = C_u \left[\overline{u_{i,j}} - 4u_{i,j} \right] + \frac{1}{\varepsilon} \underbrace{\left[u_{i,j} (u_{i,j} - a)(1 - u_{i,j}) - v_{i,j} \right]}_{f(u_{i,j}, v_{i,j})}, \quad \underline{t > 0}$$

$$\frac{dv_{i,j}}{dt} = C_v \left[\overline{v_{i,j}} - 4v_{i,j} \right] + u_{i,j} - bv_{i,j}, \quad \underline{t > -\tau}$$

- The initial conditions

$$v_{i,j}(t = -\tau) = f(I_{i,j}, 0)$$



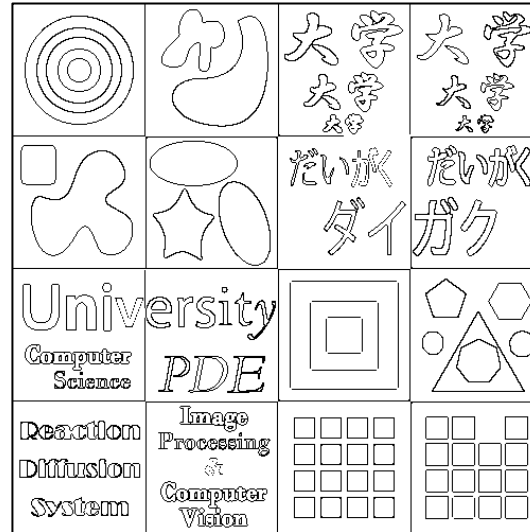
Experimental Results: Artificial Noiseless Image



The Image

500×500 pixels

256 brightness levels

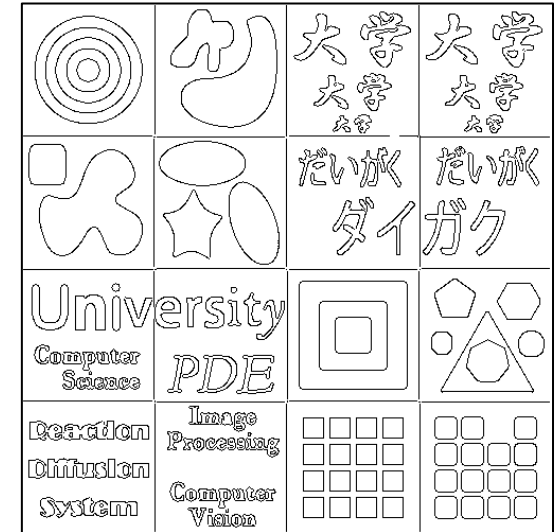


Proposed Algorithm

$C_u=4, C_v=12, a=0.1, b=4.5,$

$\varepsilon=1.0 \times 10^{-3}, \tau=5.0 \times 10^{-4}$

$\delta t=1.0 \times 10^{-4}$



Canny Algorithm

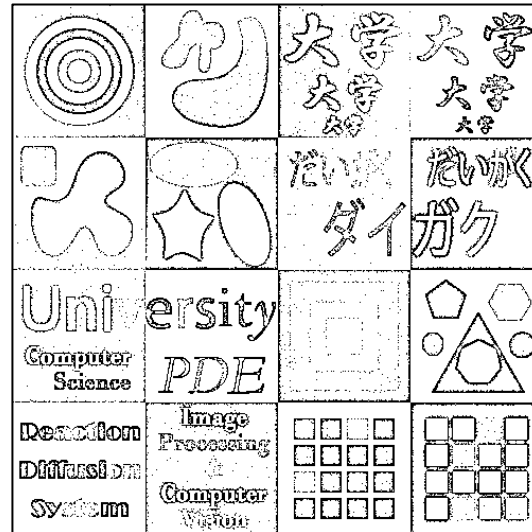
$\sigma=0.40, \theta_l=0.10, \theta_h=0.20$

Experimental Results: Artificial Noisy Image



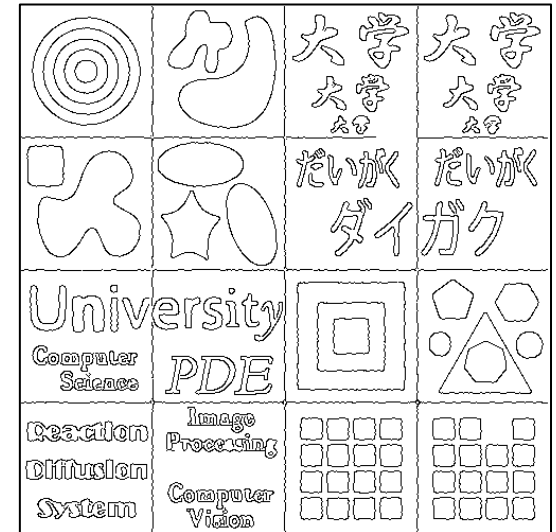
The Image

500×500 pixels
 256 brightness levels
 S.D. of noise=10



Proposed Algorithm

$C_u=4, C_v=12, a=0.1, b=4.5,$
 $\varepsilon=1.0 \times 10^{-3}, \tau=0.1$
 $\delta t=1.0 \times 10^{-4}$



Canny Algorithm

$\sigma=1.20$
 $\theta_l=0.40, \theta_h=0.70$

Experimental Results:

Quantitative Results with P, R and F measures

Algorithm	Image	P	R	F
Our Previous Algorithm	(a)	0.989	0.906	0.946
	(b)	0.747	0.908	0.819
Proposed Algorithm	(a)	0.999	0.979	0.989
	(b)	0.825	0.945	0.881
Canny Algorithm	(a)	1.000	0.975	0.987
	(b)	0.999	0.965	0.982

Image:

(a) Noiseless image

(b) Noisy image

(s.d.=10.0)

red is the best performance

- Algorithms:

- Our Previous Algorithm (Nomura et al., 2011)

- Proposed Algorithm

- Canny Algorithm (Canny, 1986)

- Evaluation measures:

M_o : obtained edge map

M_t : true edge map

$$P = \frac{|M_t \cap M_o|}{|M_o|}, R = \frac{|M_t \cap M_o|}{|M_t|}, F = \frac{2PR}{P + R}$$

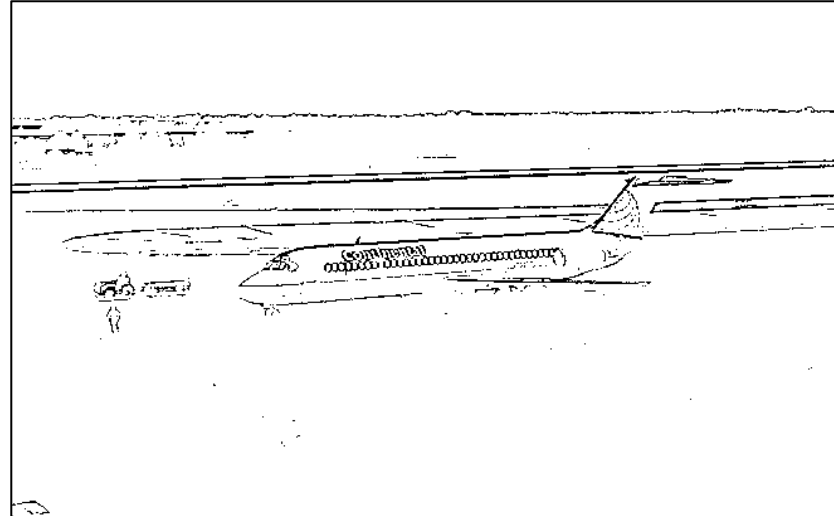
Experimental Results: Real Image (1/2)



The image

659×409 pixels, 256 brightness levels

http://marathon.csee.usf.edu/edge/edgecompare_main.html



Proposed Algorithm

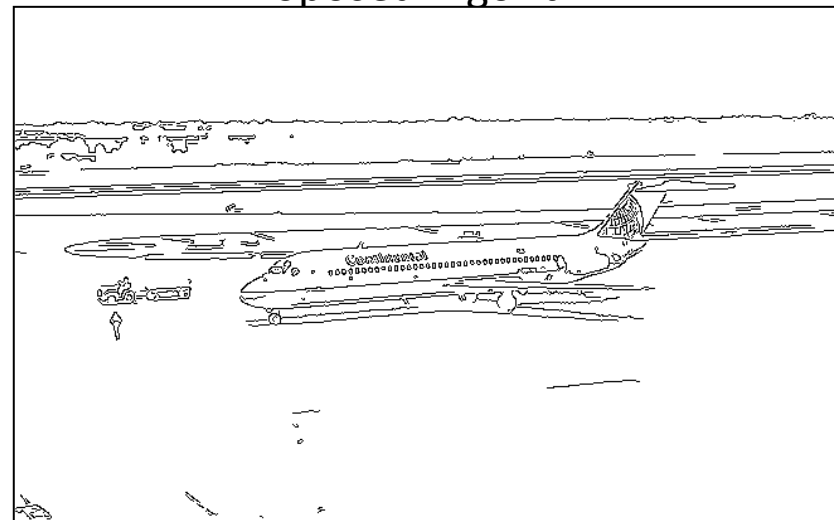
$C_u=4$
 $C_v=12$
 $\alpha=0.1$
 $b=4.5$
 $\varepsilon=1.0 \times 10^{-3}$
 $\tau=0.1$
 $\delta t=1.0 \times 10^{-4}$



Proposed Algorithm



Canny Algorithm

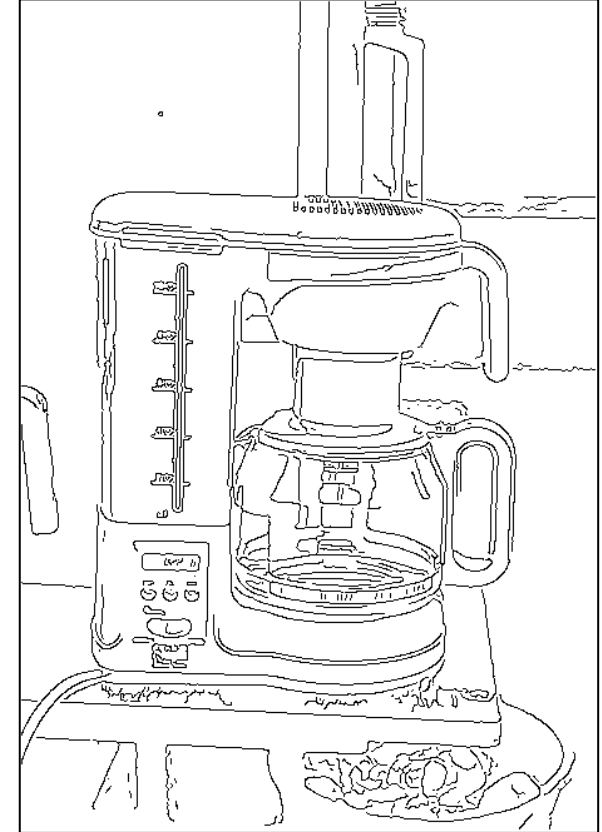
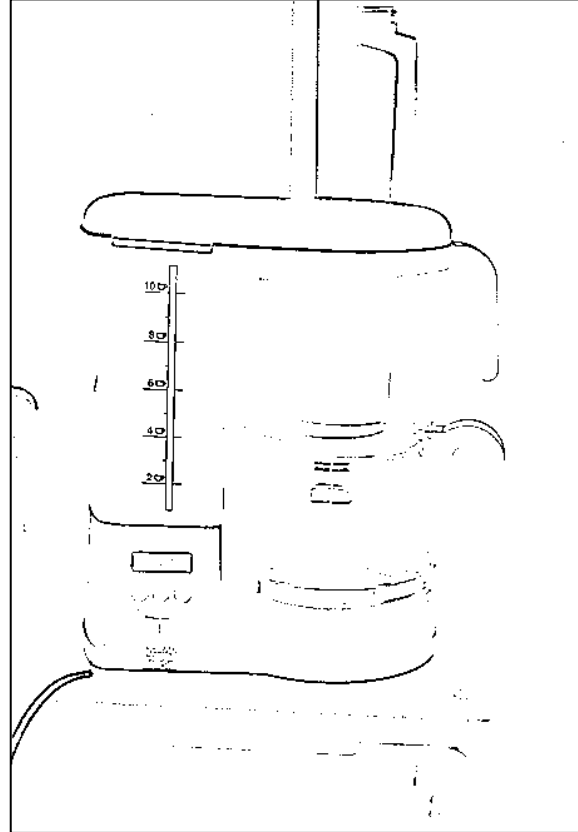
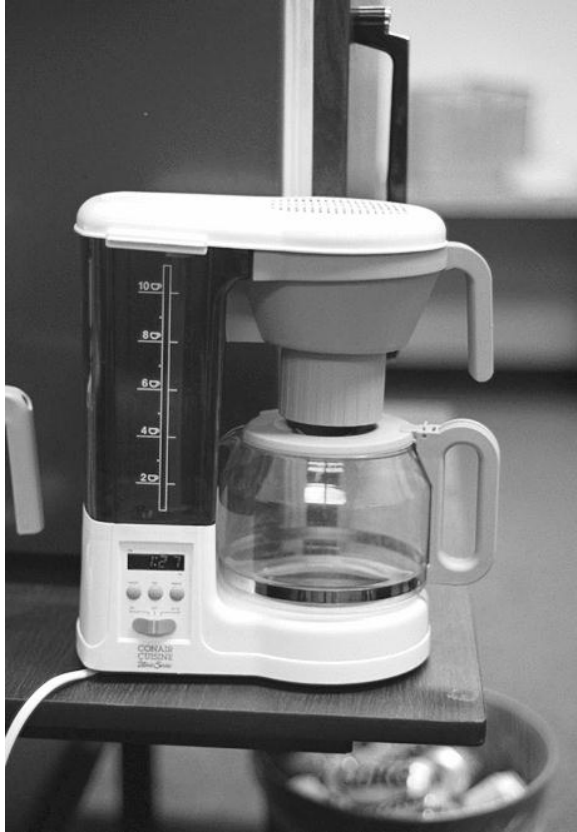


Canny Algorithm

$\sigma=0.6$
 $\theta_l=0.5$
 $\theta_h=0.9$

http://marathon.csee.usf.edu/edge/edgecompare_main.html

Experimental Results: Real Image (2/2)



The image

461×665 pixels

256 brightness levels

http://marathon.csee.usf.edu/edge/edgecompare_main.html

Proposed Algorithm

$C_u=4, C_v=12, a=0.1, b=4.5$

$\varepsilon=1.0 \times 10^{-3}, \tau=0.1$

$\delta t=1.0 \times 10^{-4}$

Canny Algorithm

$\sigma=1.2, \theta_l=0.3, \theta_h=0.8$

http://marathon.csee.usf.edu/edge/edgecompare_main.html

Conclusion

- Grid system of coupled FitzHugh-Nagumo elements for image edge detection
 - Reconsidering initial conditions for $u_{i,j}$ and $v_{i,j}$
 - Delaying computation of $u_{i,j}$
- Experiments for artificial and real gray level images
- The proposed algorithm achieved better performance than our previous algorithm.
- Future topics:
 - Noise robustness
 - Detection of blurred edges and edge strength evaluation