

Active contours without edges

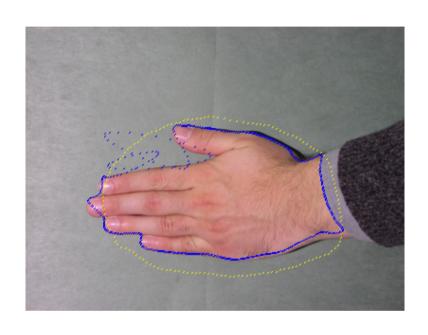
Tony F. Chan & Luminita A. Vese

A logic framework for active contours on multi-channel images

Berta Sandberg & Tony F. Chan

Presented by: Sarah Nadi & Karim Ali

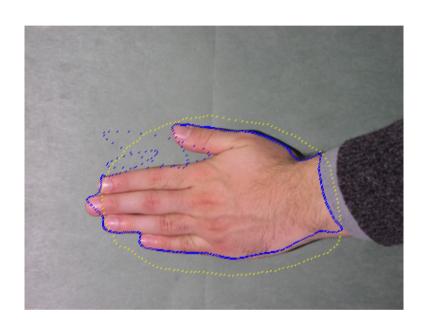
Oct. 28Th 2010 – CS 870



Detecting objects in an image



Combining multiple channels

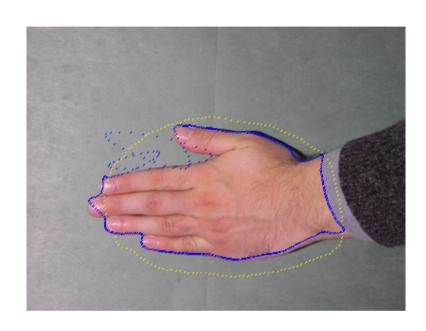


& **A**

Combining multiple channels

Detecting objects in an image

Chan-Vese model

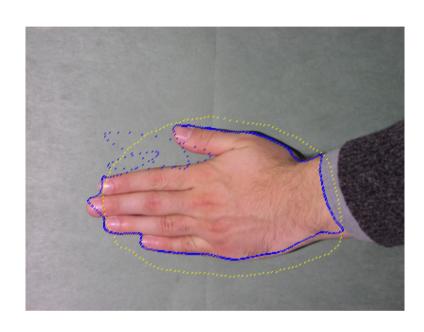


Detecting objects in an image

Chan-Vese model



Combining multiple channels



Detecting objects in an image

Chan-Vese model



Combining multiple channels

Sandberg-Chan framework

Outline

- Active contours without edges
 - Motivation
 - Model
 - Level set formulation
 - Experimental results
- Logic framework for multichannel images
 - Motivation
 - Logic operations
 - Level set formulation
 - Experimental results
- Conclusions

Active Contour Models

Overview

- Detect objects in an image
- Use image gradient to evaluate the stopping condition

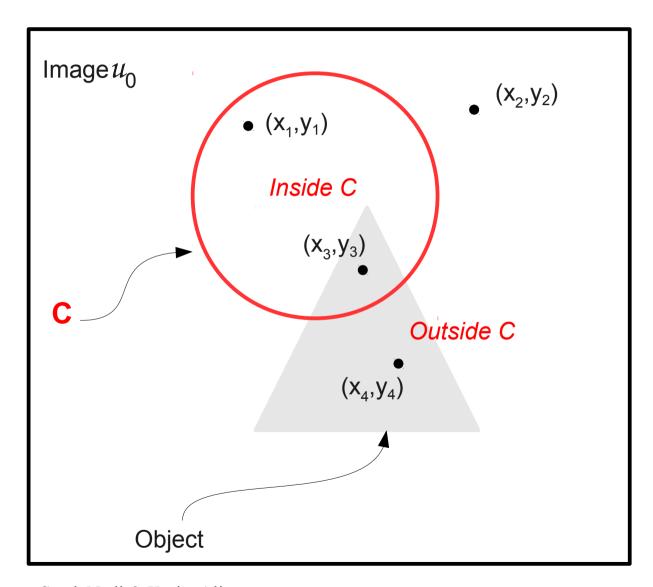
Limitations

- Can only detect edges defined by a gradient
 - Edges can be smoothed
 - Contour may pass through the boundary

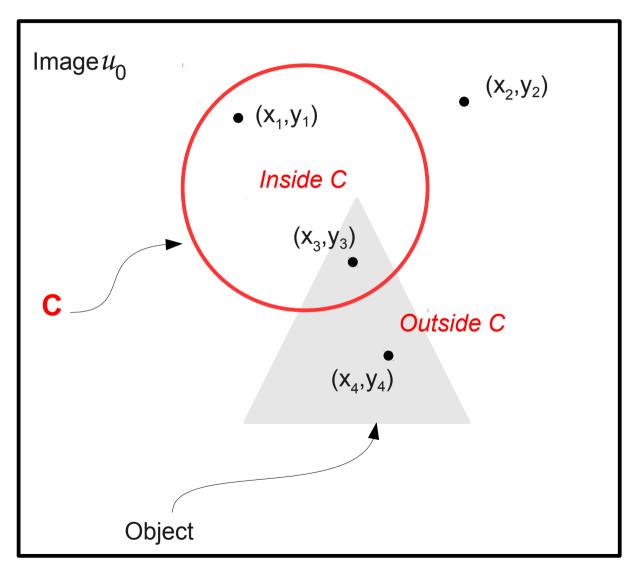
Active Contours Without Edges (Chan-Vese)

- Region based technique
- Stopping is based on Mumford-Shah segmentation techniques
 - No longer depends on gradient
 - Detects smooth or discontinuous boundaries
- Interior contours are automatically detected
- Initial curve can be anywhere in the image

Region Based View



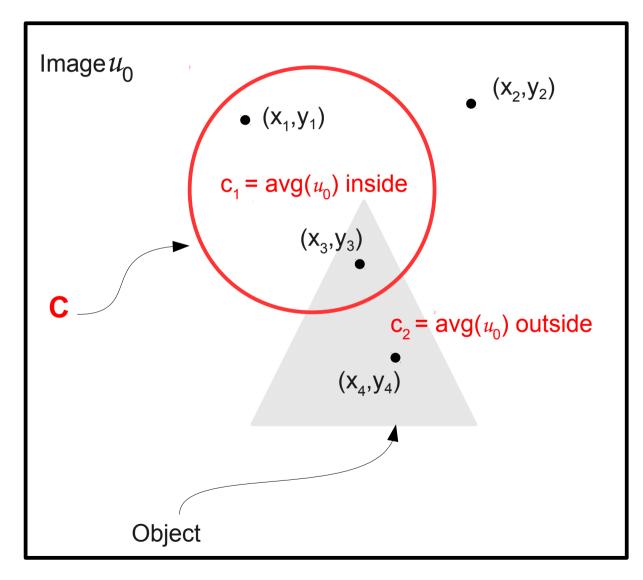
Region Based View



To fit C to our object, we want:

 (x_1,y_1) to be **outside** C (x_2,y_2) to be **outside** C (x_3,y_3) to be **inside** C (x_4,y_4) to be **inside** C

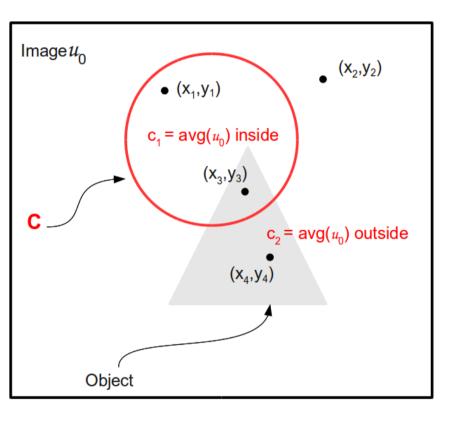
Region Based View



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Curve Fitting



Fitting inside:

$$F_1(C) = \int_{inside(C)} |u_0(x, y) - c_1|^2 dx dy$$

Fitting outside:

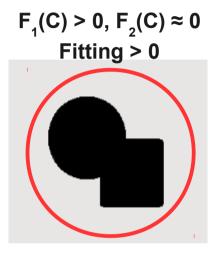
$$F_2(C) = \int_{outside(C)} |u_0(x, y) - c_2|^2 dx dy$$

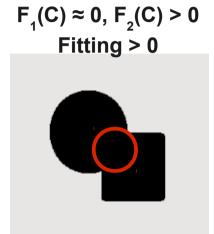
Overall fitting of C:

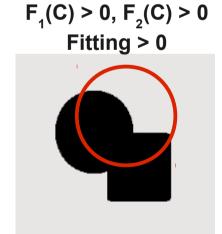
$$F_1(C) + F_2(C) = \int_{inside(C)} |u_0(x, y) - c_1|^2 dx dy + \int_{outside(C)} |u_0(x, y) - c_2|^2 dx dy$$

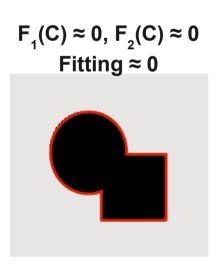
Intuition

- C should stop evolving
 when C = Object
- To achieve that:
 Fitting ≈ 0









Energy Function

$$F(c_1, c_2, C) = \lambda_1 \int_{inside(C)} |u_0(x, y) - c_1|^2 dx dy$$
$$+ \lambda_2 \int_{outside(C)} |u_0(x, y) - c_2|^2 dx dy$$

Energy Function

$$F(c_1, c_2, C) = \mu \cdot Length(C)$$

$$+ \nu \cdot Area(inside(C))$$

$$+ \lambda_1 \int_{inside(C)} |u_0(x, y) - c_1|^2 dx dy$$

$$+ \lambda_2 \int_{outside(C)} |u_0(x, y) - c_2|^2 dx dy$$

Energy Function

$$F(c_1, c_2, C) = \mu \cdot Length(C)$$

$$+ \nu \cdot Area(inside(C))$$

$$+ \lambda_1 \int_{inside(C)} |u_0(x, y) - c_1|^2 dx dy$$

$$+ \lambda_2 \int_{outside(C)} |u_0(x, y) - c_2|^2 dx dy$$

Look at minimization problem:

$$\inf_{c_1,c_2,C} F(c_1,c_2,C)$$

$$F(c_1, c_2, C) = \mu \cdot Length(C)$$

$$+ \nu \cdot Area(inside(C))$$

$$+ \lambda_1 \int_{inside(C)} |u_0(x, y) - c_1|^2 dx dy$$

$$+ \lambda_2 \int_{outside(C)} |u_0(x, y) - c_2|^2 dx dy$$

$$F(c_{1}, c_{2}, C) = \mu \int_{\Omega} \delta_{0}(\phi(x, y)) |\nabla \phi(x, y)| dx dy$$

$$+ \nu \cdot Area(inside(C))$$

$$+ \lambda_{1} \int_{inside(C)} |u_{0}(x, y) - c_{1}|^{2} dx dy$$

$$+ \lambda_{2} \int_{outside(C)} |u_{0}(x, y) - c_{2}|^{2} dx dy$$

$$F(c_{1}, c_{2}, C) = \mu \int_{\Omega} \delta_{0}(\phi(x, y)) |\nabla \phi(x, y)| dx dy$$

$$+ \nu \int_{\Omega} H(\phi(x, y)) dx dy$$

$$+ \lambda_{1} \int_{inside(C)} |u_{0}(x, y) - c_{1}|^{2} dx dy$$

$$+ \lambda_{2} \int_{outside(C)} |u_{0}(x, y) - c_{2}|^{2} dx dy$$

$$F(c_{1}, c_{2}, C) = \mu \int_{\Omega} \delta_{0}(\phi(x, y)) |\nabla \phi(x, y)| dx dy$$

$$+ \nu \int_{\Omega} H(\phi(x, y)) dx dy$$

$$+ \lambda_{1} \int_{\Omega} |u_{0}(x, y)| - c_{1}|^{2} H(\phi(x, y)) dx dy$$

$$+ \lambda_{2} \int_{outside(C)} |u_{0}(x, y) - c_{2}|^{2} dx dy$$

$$F(c_{1}, c_{2}, C) = \mu \int_{\Omega} \delta_{0}(\phi(x, y)) | \nabla \phi(x, y) | dx dy + \nu \int_{\Omega} H(\phi(x, y)) dx dy + \lambda_{1} \int_{\Omega} |u_{0}(x, y)| - c_{1}|^{2} H(\phi(x, y)) dx dy + \lambda_{2} \int_{\Omega} |u_{0}(x, y)| - c_{2}|^{2} (1 - H(\phi(x, y))) dx dy$$

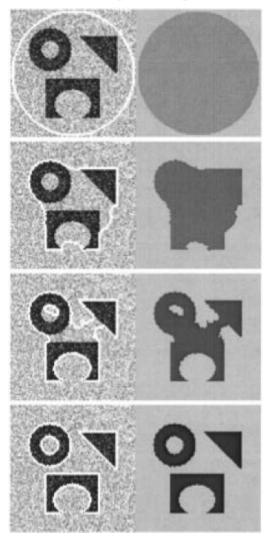
Chan-Vese Technique

$$\frac{\partial \phi}{\partial t} = \delta_{\varepsilon}(\phi) \left[\mu \operatorname{div}\left(\frac{\nabla \phi}{|\nabla \phi|}\right) - \nu - \lambda_{1}(u_{0} - c_{1})^{2} + \lambda_{2}(u_{0} - c_{2})^{2} \right]$$

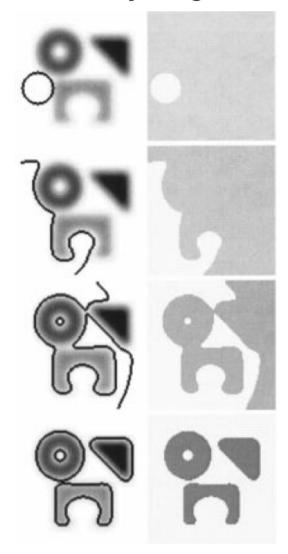
- Initialize Φ^0 by Φ_0 , n = 0
- Compute $c1(\Phi^n)$ and $c2(\Phi^n)$
- Solve the PDE to obtain Φ n+1
- Reinitialize Φ locally to the signed distance function to the curve (optional)
- Check whether the solution is stationary. If not n = n + 1 and repeat

Experimental Results

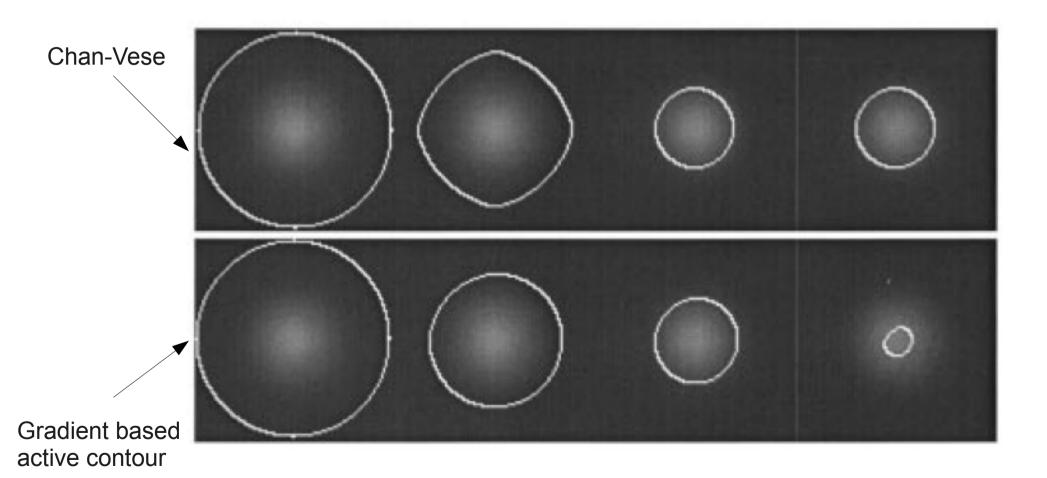
Noisy Image



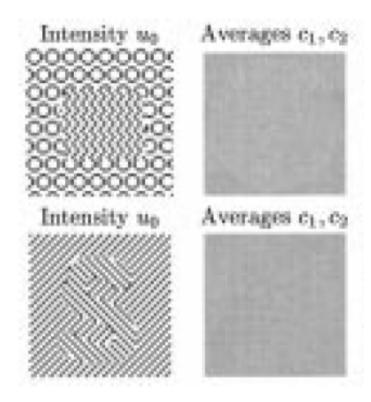
Blurry Image



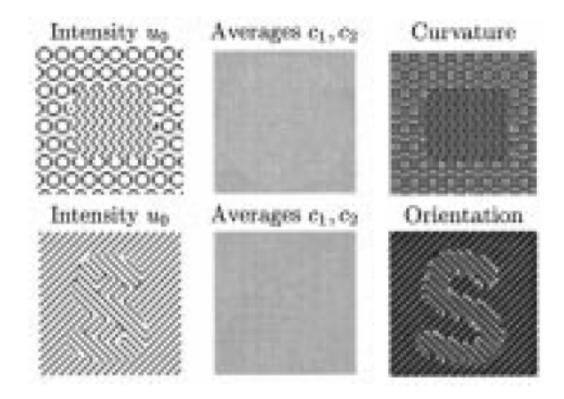
Comparing Models



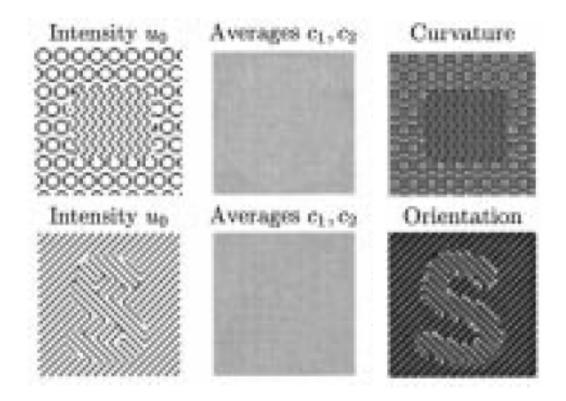
Limitations



Limitations



Limitations

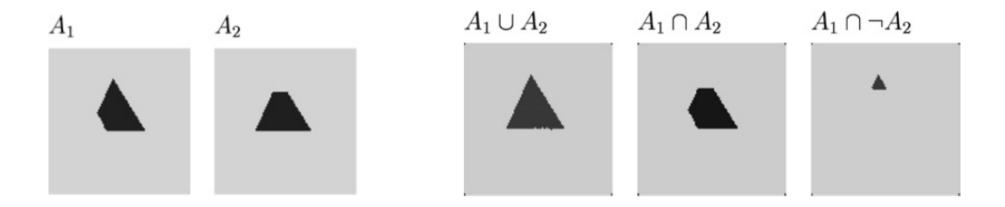


Or implement general case of Mumford-Shah function

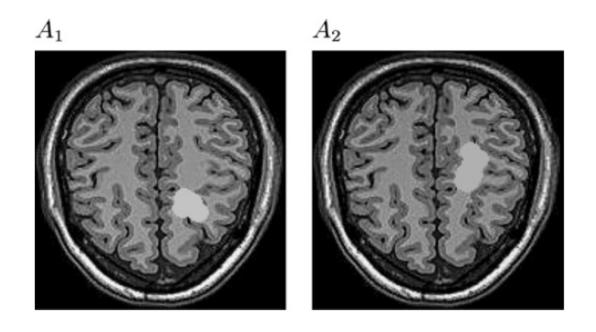
Applying the Chan-Vese Model to Multi-channel Images

The Sandberg-Chan Framework

 Allows the user to produce any logical combination of object channels



Motivating Example

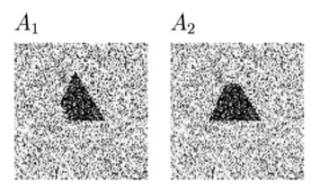


$$eg A_1 \cap A_2$$
 New tumor growth $A_1 \cap \neg A_2$ Tumor shrinking

Existing Techniques

- 1)Segment each channel independently followed by bitwise operations
 - Costly
- 2)Perform logic operations on channels forming a combined image followed by segmentation
 - Requires a lot of information about intensity

Technique 1



Segmentation Channel by Channel with Active Contours without edges:



Bitwise Logic Operations



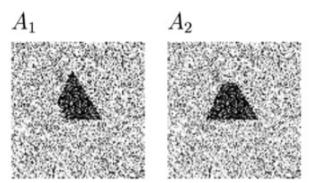




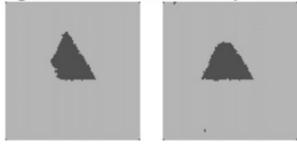
$$A_1 \cap \neg A_2$$



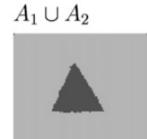
Technique 1



Segmentation Channel by Channel with Active Contours without edges:



Bitwise Logic Operations



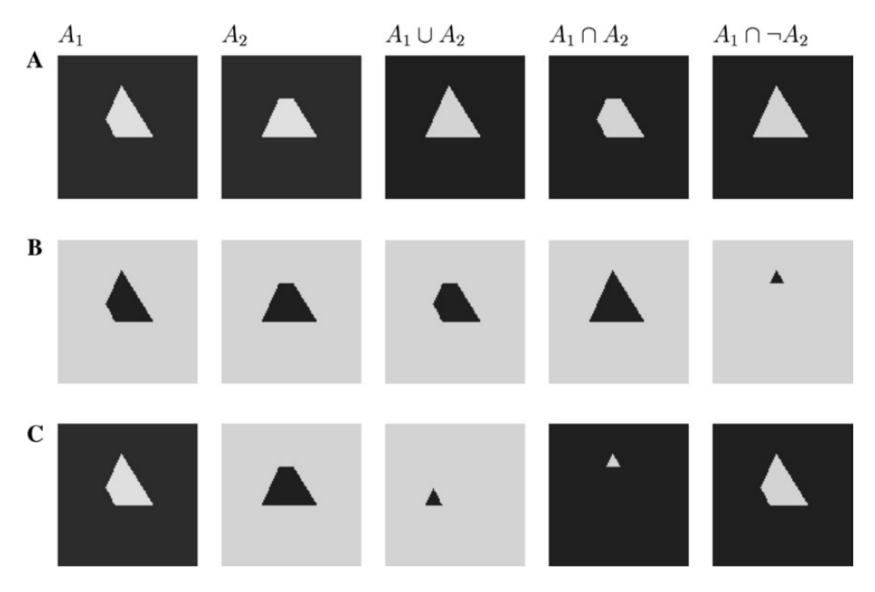




$$A_1 \cap \neg A_2$$



Technique 2

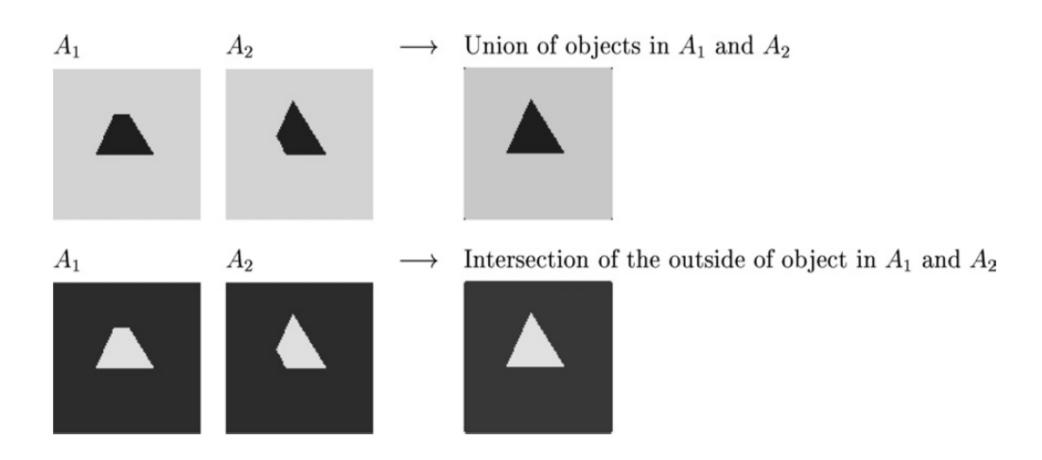


Proposed Solution

- Use the Chan-Vese model, and compare the contour fitting on each channel
 - Start with the same contour on each channel
 - Fit contour to the object on ALL channels according to the logic operator & based on regions



Region Based Logic Operations



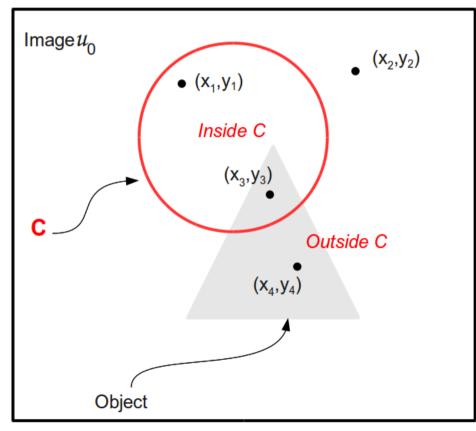
Logic Variables

$$z_i^{\text{in}}(u_0^i, x, y, C) = \begin{cases} 0 & \text{if } (x, y) \in C \text{ and } (x, y) \text{ inside the object in channel } i, \\ 1 & \text{otherwise,} \end{cases}$$

 $z_i^{\text{out}}(u_0^i, x, y, C) = \begin{cases} 1 & \text{if } (x, y) \not\in C \text{ and } (x, y) \text{ is inside the object in channel } i, \\ 0 & \text{otherwise.} \end{cases}$

Note:

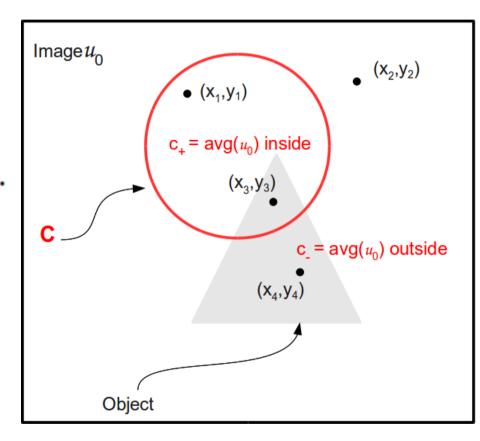
0 is true because we want to minimize the fitting term



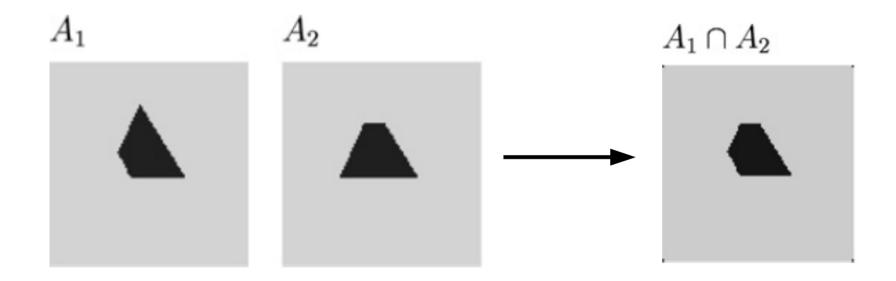
Chan-Vese Formulation

$$z_i^{\text{in}}(u_0^i, x, y, C) = \frac{|u_0^i(x, y) - c_+^i|^2}{\max_{(x, y) \in u_0^i} u_0^i},$$

$$z_i^{\text{out}}(u_0^i, x, y, C) = \frac{|u_0^i(x, y) - c_-^i|^2}{\max_{(x, y) \in u_0^i} u_0^i}.$$



Concentrate on Intersection



Interpolation Functions

$$f_{\cup} = (z_1 \cdot z_2)^{1/2}$$
 $f_{\cap} = 1 - ((1 - z_1)(1 - z_2))^{1/2}$

Intersection of inside
$$f_{A_1\cap A_2}(x,y) = 1 - \sqrt{(1-z_1^{\rm in}(x,y))(1-z_2^{\rm in}(x,y))} + \sqrt{z_1^{\rm out}(x,y)z_2^{\rm out}(x,y))}$$

General Case

$$L_1(A_1) \cap L_2(A_2) \cap \cdots \cap L_n(A_n)$$

$$F_{L_1(A_1)\cap\cdots\cap L_n(A_n)} = \mu|C| + \lambda \left[\int_{\text{inside}(C)} \left(1 - \left(\prod_{i=1}^n \left(1 - l_i(z_i^{\text{in}}) \right) \right)^{1/n} \right) dx + \int_{\text{outside}(C)} \left(\prod_{i=1}^n l_i(z_i^{\text{out}}) \right)^{1/n} dx \right].$$

$$l_i(z_i^{\text{in}}) = \begin{cases} z_i^{\text{in}} & \text{if } L_i(A_i) = A_i, \\ z_i^{\text{in}'} & \text{if } L_i(A_i) = \neg A_i. \end{cases}$$

Level Set Formulation

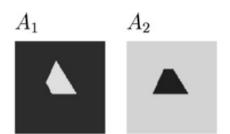
Objective function

$$F(\phi, c^+, c^-) = \mu |C(\phi)| + \lambda \left[\int_{\Omega} f_{\text{in}}(z_1^{\text{in}}, \dots, z_n^{\text{in}}) H(\phi) + f_{\text{out}}(z_1^{\text{out}}, \dots, z_n^{\text{out}}) (1 - H(\phi)) dx \right].$$

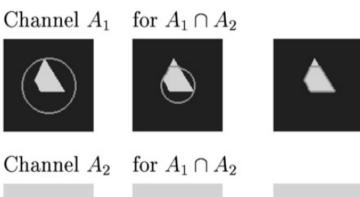
We want to minimize F

$$\frac{\partial \phi_{L_1(A_1) \cap \dots \cap L_n(A_n)}}{\partial t} = \delta_{\epsilon}(\phi) \left[\mu \nabla \cdot \left(\frac{\nabla \phi}{|\nabla \phi|} \right) - \lambda \left(1 - \left(\prod_{i=1}^n \left(1 - l_i(z_i^{\text{in}}) \right) \right)^{1/n} + \left(\prod_{i=1}^n l_i(z_i^{\text{out}}) \right)^{1/n} \right) \right].$$

Experiments – 2 Channels

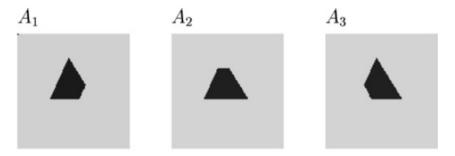


Time Evolution showing contour C evolving to desired solution in both Channels.





Experiments – 3 Channels

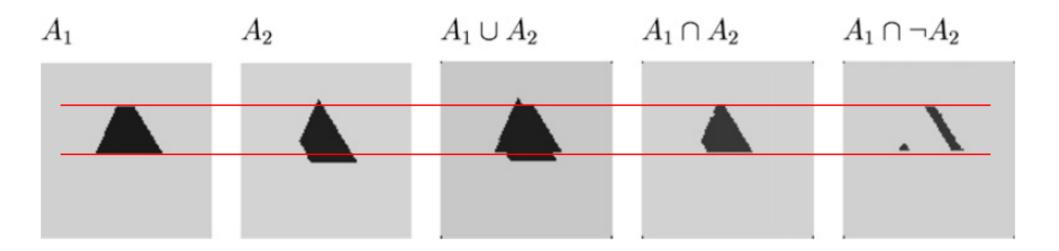


$$A_1 \cap A_2 \cap A_3$$

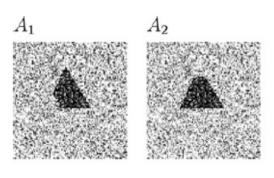
Experiments – Unregistered Images



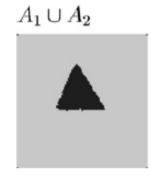
Experiments – Unregistered images

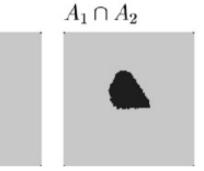


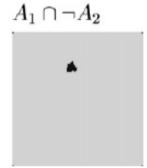
Experiments - Noise



Region based logic model









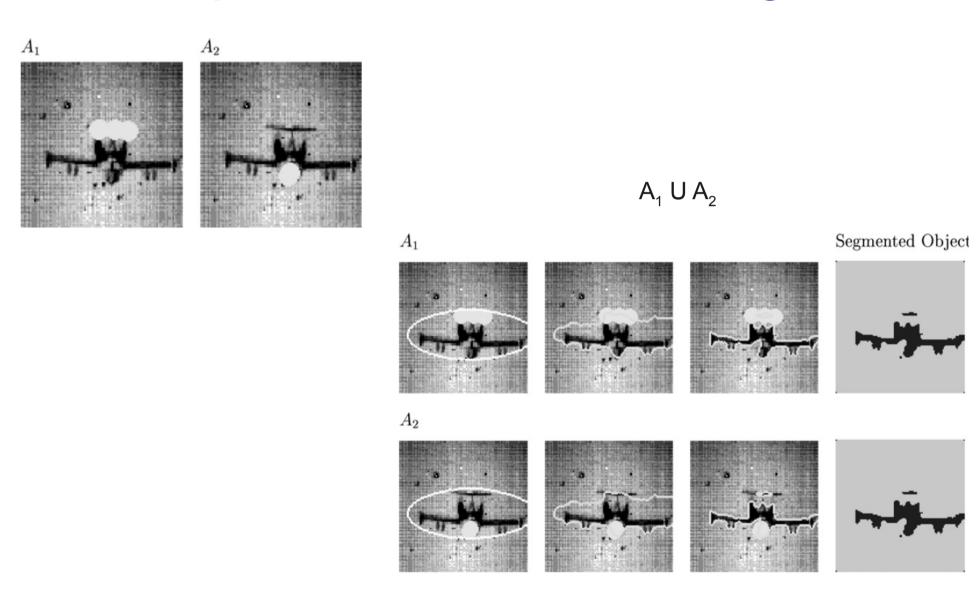






Channel by channel case

Experiments - Real Images



Conclusions

- Chan-Vese model
 - Detects objects without edge information
 - Compares image intensities inside & outside the contour
 - Can detect "holes" in objects
- Sandberg-Chan logic framework
 - Can combine multiple images using different logic operations
 - Performs segmentation & logic operation at the same step

Thank you

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