

# Image Segmentation

Ben Appleton

University of Queensland

# The goal of segmentation

- General segmentation:
  - Break an image into component objects
- Object segmentation:
  - Find a particular object of interest

# Segmentation methodology

- Identify distinguishing features
  - Colour
  - Texture
  - Shape
- Use prior knowledge
  - Position, orientation
  - Relation to other objects (eg. a number plate is found toward the bottom of a car)

# Classical methods

- Point-segmentation
  - Thresholding
  - Feature clustering
- Boundary methods
  - Edge detection
  - Laplacian of Gaussian
- Region methods
  - Region growing/splitting/merging
  - Watershed

# Image Simplification

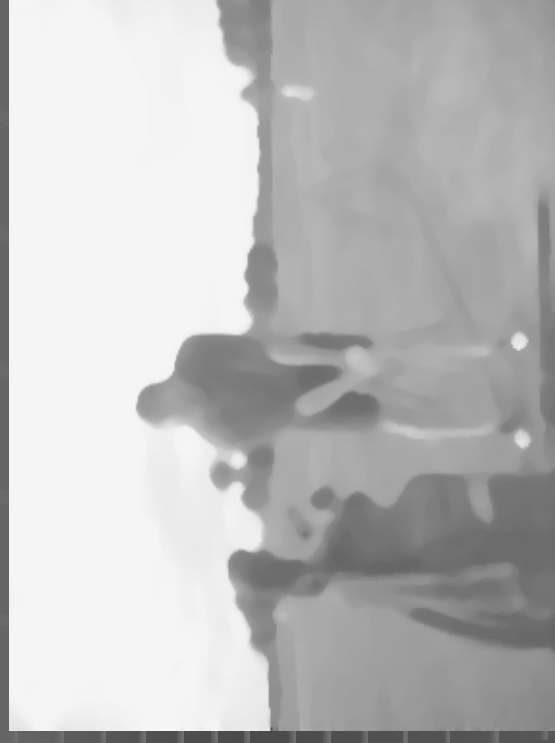
- Simplifies the segmentation problem by ‘cartooning’ the image

Median filtering	$I^{n+1}(x, y) = \text{median}_{(x', y') \in W(x, y)} \{I^n(x', y')\}$
Non-linear diffusion	$\frac{\partial I}{\partial t} = \text{div} \left( g(\ \nabla I\ ) \frac{\nabla I}{\ \nabla I\ } \right)$
Reaction-diffusion	$\frac{\partial I}{\partial t} = \text{div} \left( g(\ \nabla I\ ) \frac{\nabla I}{\ \nabla I\ } \right) + \alpha(I_0 - I)$

Original image



Median filtered



Non-linear diffusion

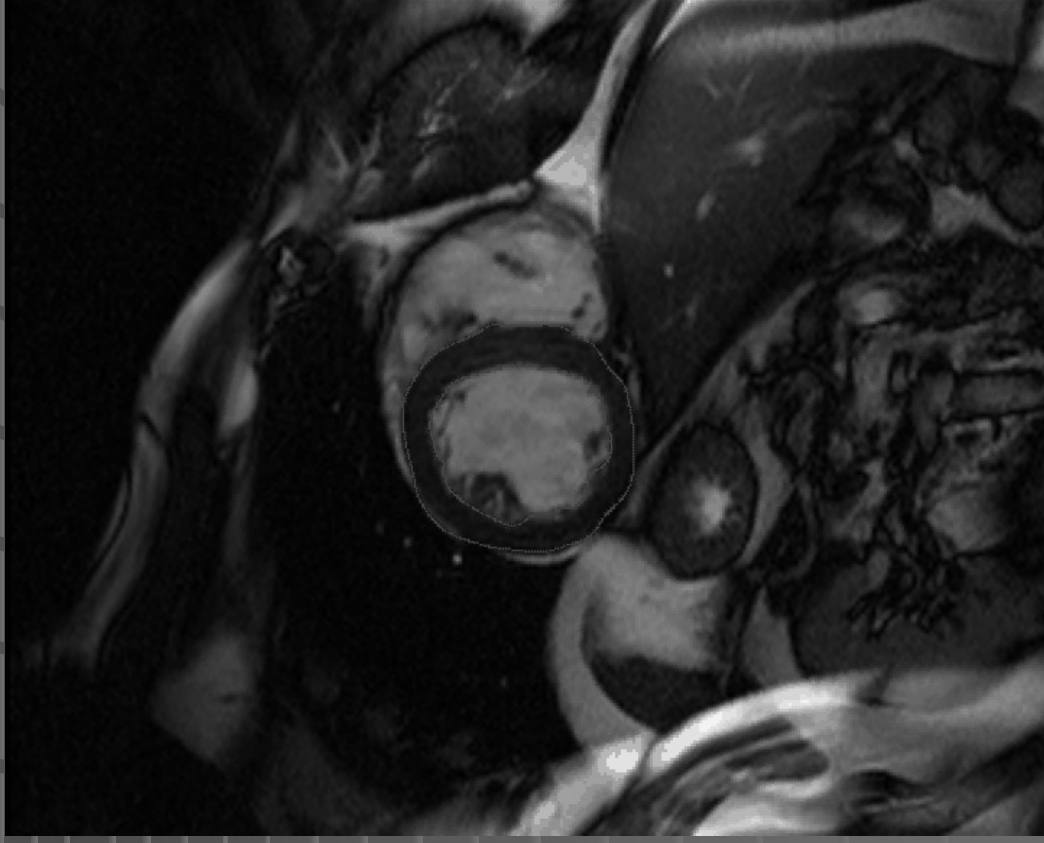


Reaction-diffusion



# Active Contours

- Snakes
- Active shapes
- Level sets

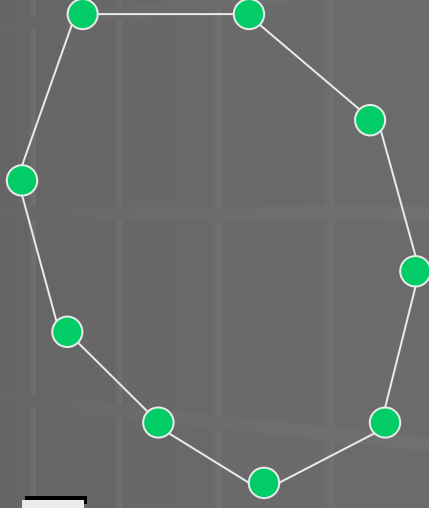


# Active Contours

- What are active contours for?
  - ‘Evolving’ surfaces
- Applications in image analysis
  - Segmentation
  - Multi-view reconstruction
  - Object tracking

# Snakes

- Represent a surface explicitly as a polytope (polygon or polyhedron)
- Behaves as a network of point masses connected by springs and thin-plate splines
- External forces derived from image push toward goal



# Benefits of Snakes

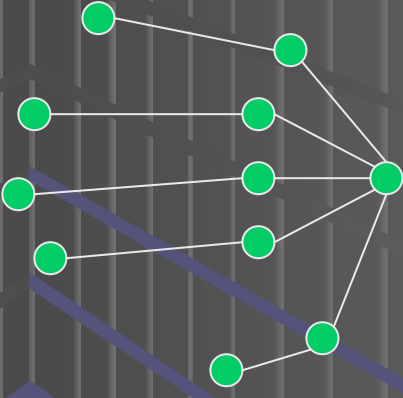
- Simple idea
- Widely accepted in medical image analysis
- Fast

# Active Shapes (Cootes)

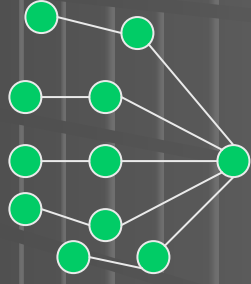
- Allows snakes to learn object shapes
- Manually segment several images
- Determine the vertex covariance matrix
  - Eigenvectors are dominant modes of shape variation

# Active Shapes

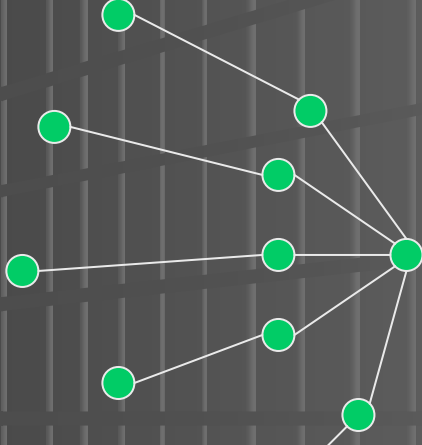
Relaxed



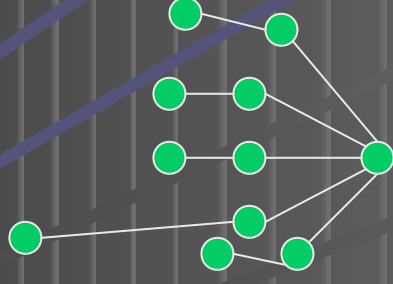
Clenched



Splayed

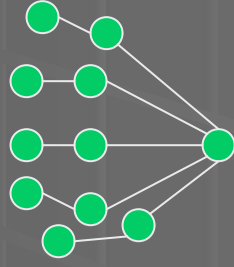


Pointing

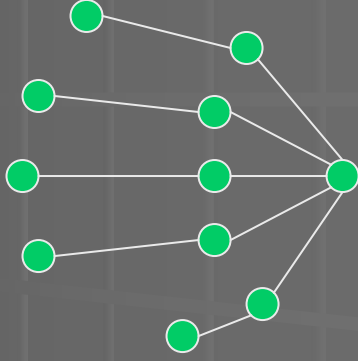


First mode (eigenvector)

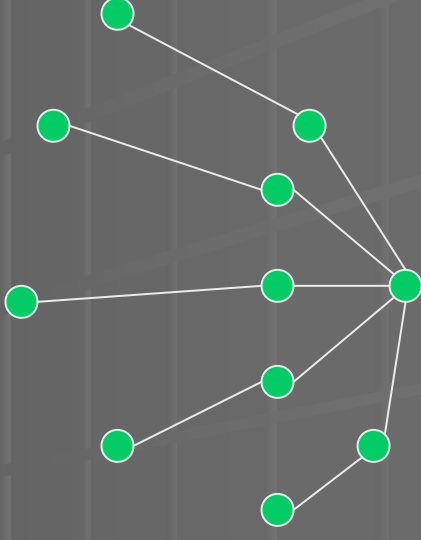
-3 $\sigma$



0 $\sigma$

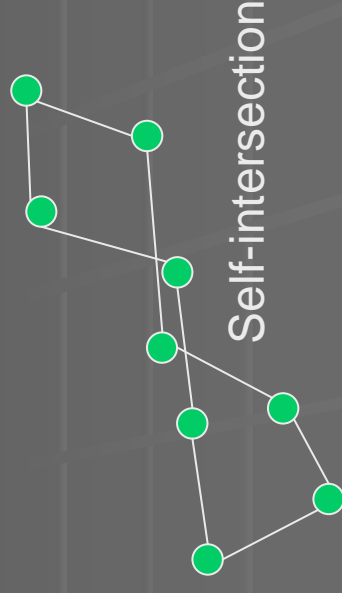
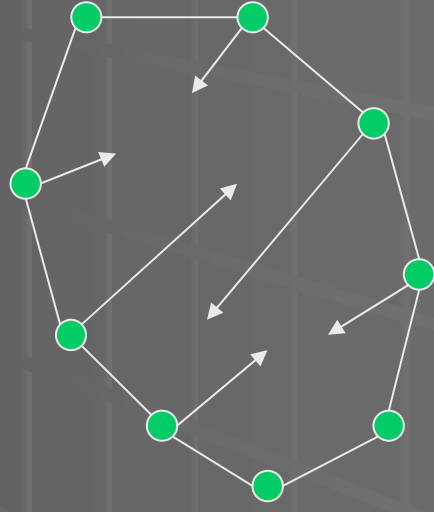


3 $\sigma$



# Problems with Snakes – Topology

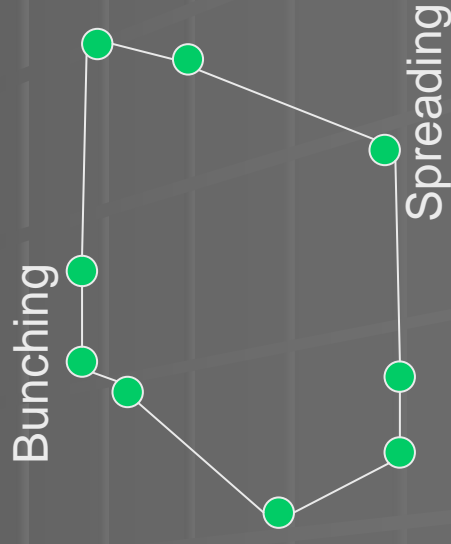
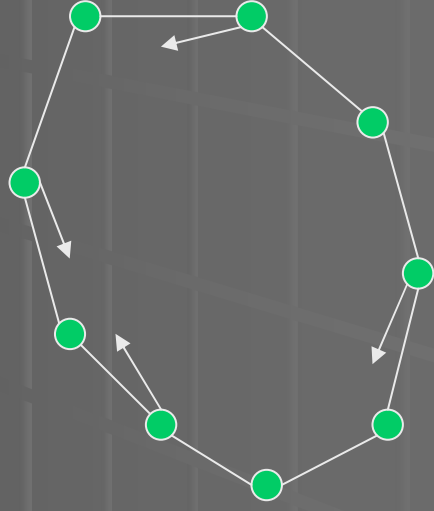
- Cannot change topology easily
- Self-intersection difficult to detect
- Complexity grows with dimensionality



# Problems with Snakes -

## Stability

- Surface points bunch up or spread out
- Produces instabilities and inaccuracies



# Level Sets - Theory

- Implicit representation of surfaces
- Advantages of level sets
- Simple example

# Level Sets

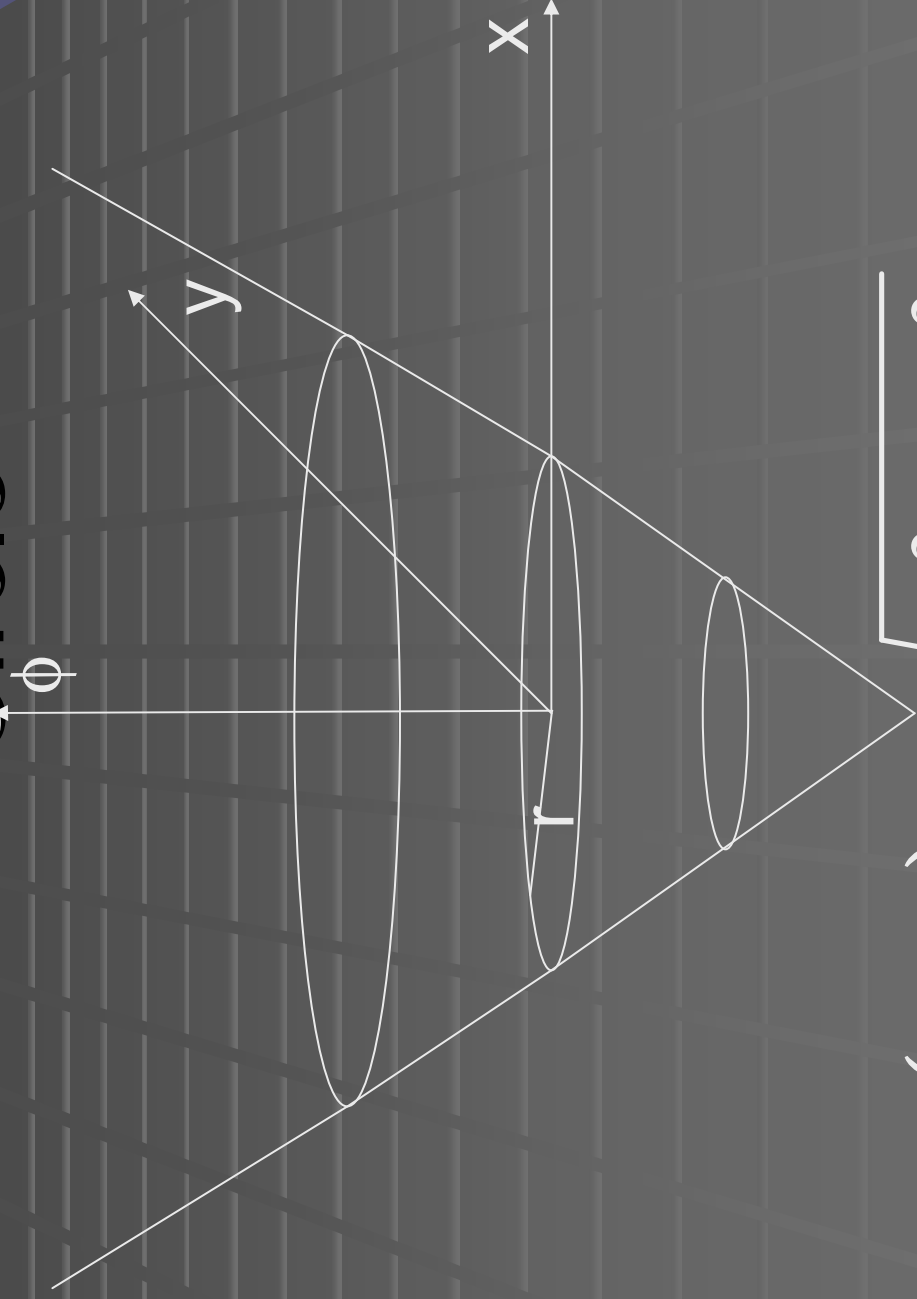
- Represent a surface implicitly as the zero-level contour of a function

$$\phi(C) = 0$$

- We are free to choose  $\phi$  off the contour

# Example – Representing a

## Circle



$$\phi(x, y) = \sqrt{x^2 + y^2} - r$$

# Advantages of Level Sets

- Topology is handled implicitly
- Extension to higher dimensions is simple
- Stability is assured
- Easy to manipulate the implicit contour via the fundamental partial differential equation

# The Fundamental PDE of Level Sets

Manipulate  $\phi$  to indirectly move

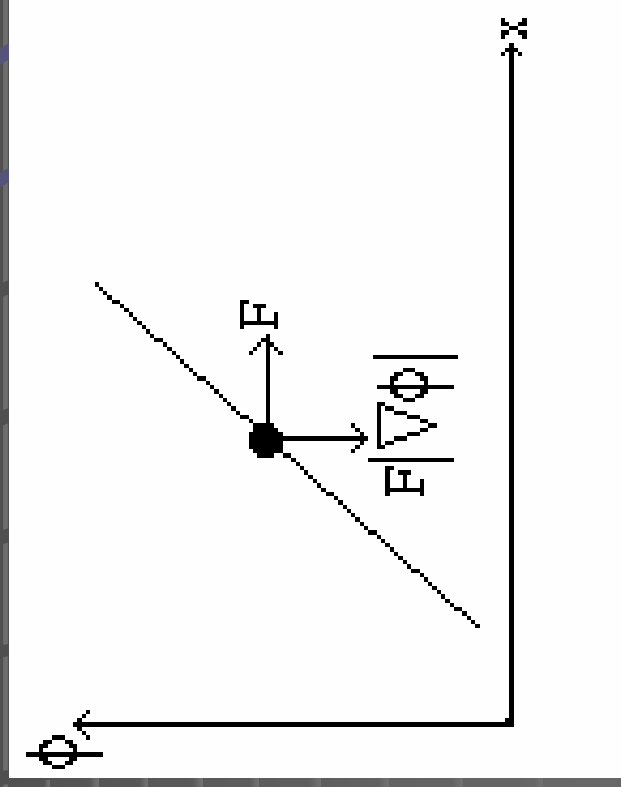
$C$ :

$$\phi(C) = 0$$

$$\frac{d\phi(C)}{dt} = \frac{\partial C}{\partial t} \cdot \nabla \phi + \frac{\partial \phi}{\partial t} = 0$$

$$\therefore \frac{\partial \phi}{\partial t} = -F |\nabla \phi|$$

where  $F$  is the speed function normal to the curve



# Example – Evolving a Circle

- Level Set representation of a circle  $\phi(x, y) = \sqrt{x^2 + y^2} - r$
- Setting  $F = 1$  causes the circle to expand uniformly
- $|\nabla\phi| = 1$  by choice of representation, so we obtain the level set evolution equation:
- unconditional stability  $\frac{\partial\phi}{\partial t} = -1$

Explicit solution:  $\phi(x, y, t) = \sqrt{x^2 + y^2} - r - t$

which means that the circle has radius  $r + t$  at time  $t$ , as expected

# Level Sets - Evolution

- Speed functions
- Two speed functions for segmentation

# Speed Functions

- Surface evolution solely dependent on  $F$
- Design  $F$  directly or derive by variational calculus
- $F$  a function of geometric variables such as normal direction, curvature and so on

# Segmentation via Malladi-Sethian-Vemuri Function

$$F = \pm 1 - \epsilon \kappa(C) + \beta V |\nabla G_\sigma * I| \cdot N \rightarrow$$

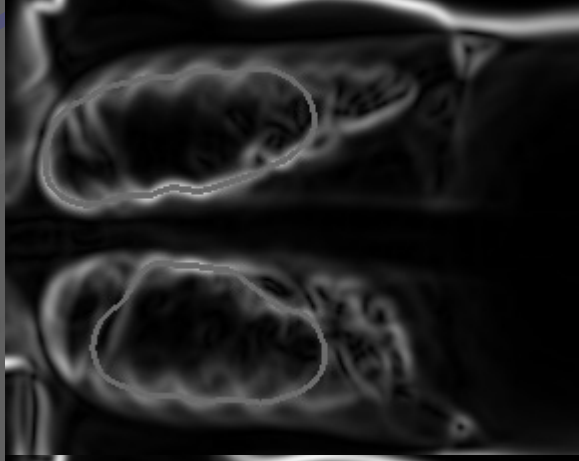
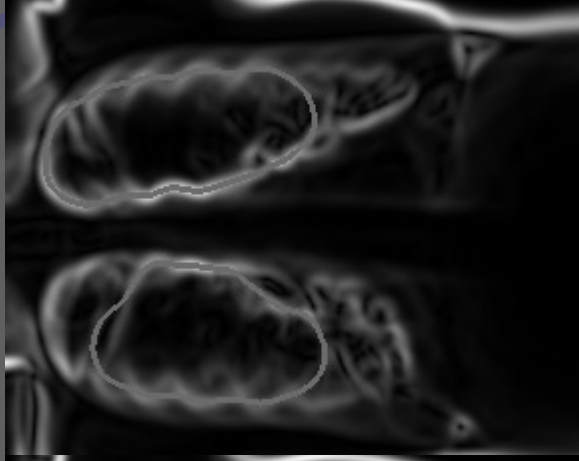
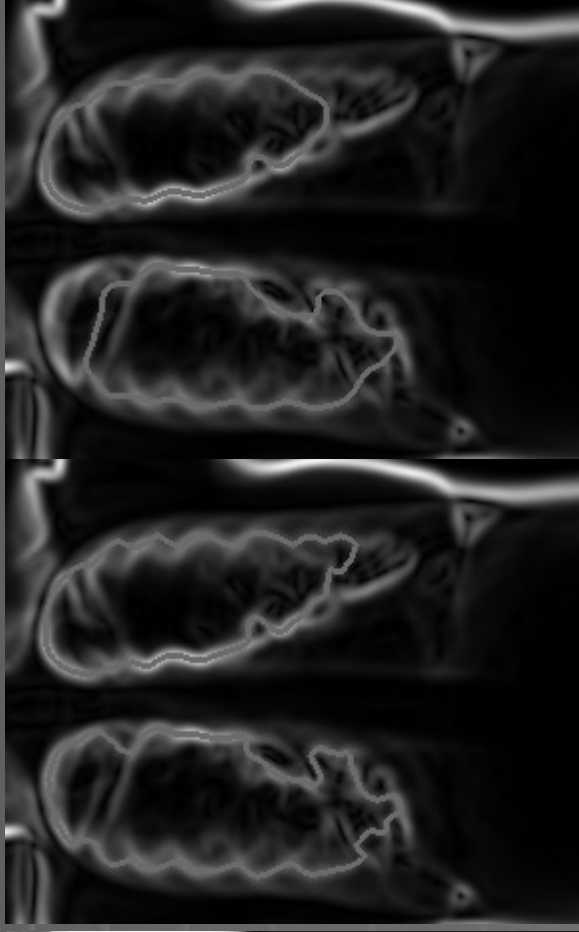
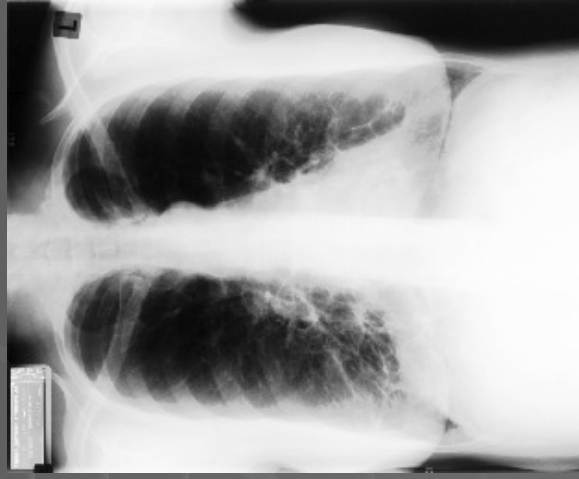
Inflation /deflation
Regularisation
Edge-attraction force

Lung x-ray

Viscosity  $\epsilon = 0.5$

Viscosity  $\epsilon = 2$

Viscosity  $\epsilon = 5$



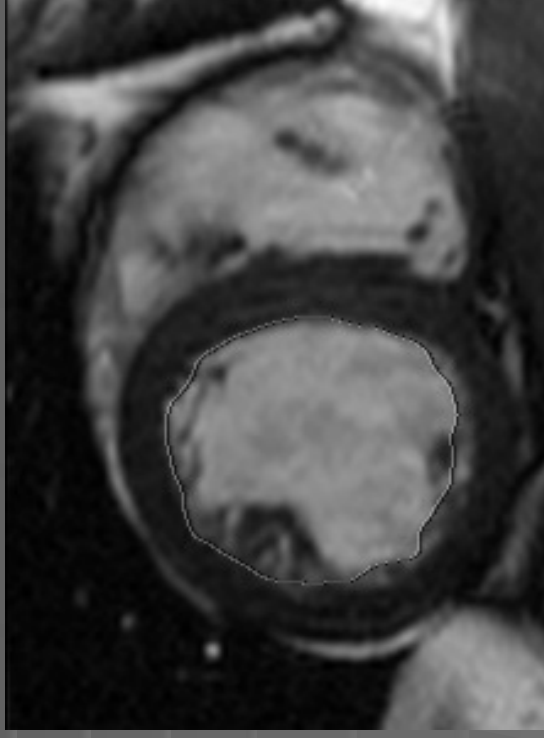
# Segmentation via Geodesic Active Contours

Gradient metric:

$$g = \frac{1}{1 + |\nabla G_\sigma * I|} + \varepsilon$$

- Minimise energy:  $E(C) = \int_C g \cdot ds$
- Speed function:  $F = -\nabla g \cdot \vec{N} - g(\kappa - \alpha)$
- Implicit scheme:  $\partial_t \phi = \alpha g + \nabla \cdot (g \nabla \phi)$

MRI of the left ventricle in a heart

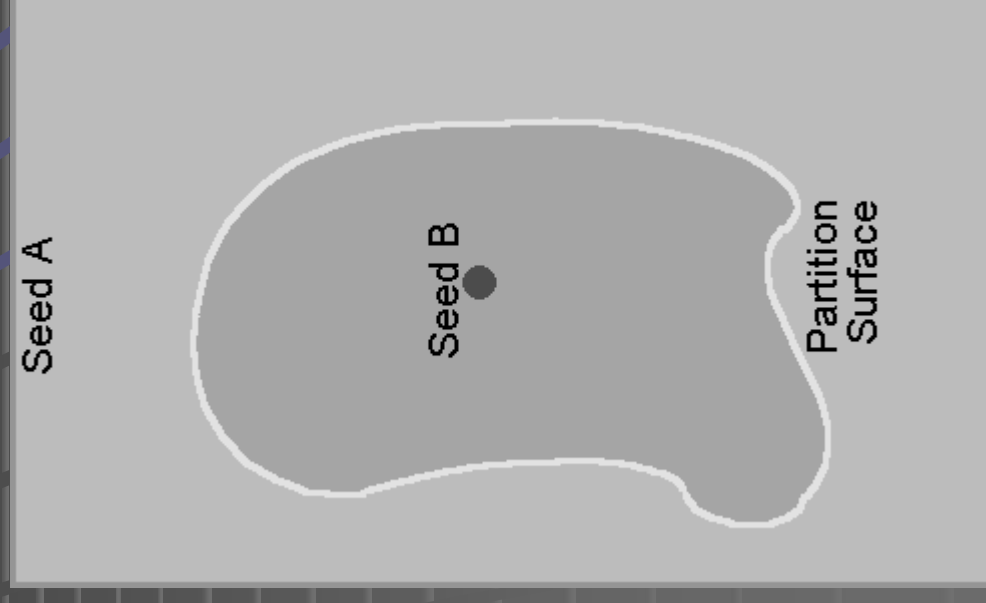


# New Directions - Optimality

- Snakes and level sets are iterative optimisers
  - Not guaranteed to extract the best answer
- Graph theory and its continuous extensions allow optimal solutions to the same problems

# Surfaces and partitionings

- A *partitioning* divides a space into disjoint components
  - Boundaries form simple closed surfaces
- Many Image Analysis (IA) problems seek surfaces or partitionings of a space:
  - Segmentation
  - Object tracking
  - 3D reconstruction



# Geodesic Active Contours and Surfaces

- Proposed for segmentation by Caselles et. al. (PAMI 1997)

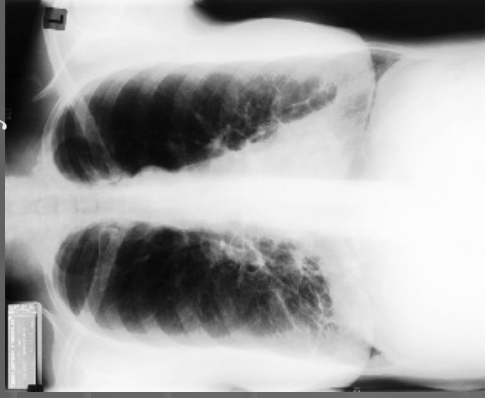
$$E(C) = \int_C g \cdot ds$$

- Weighted minimal surface model

- Variational level set approach

Gradient metric:  $g = \frac{1}{1 + |\nabla G_\sigma * I|}$

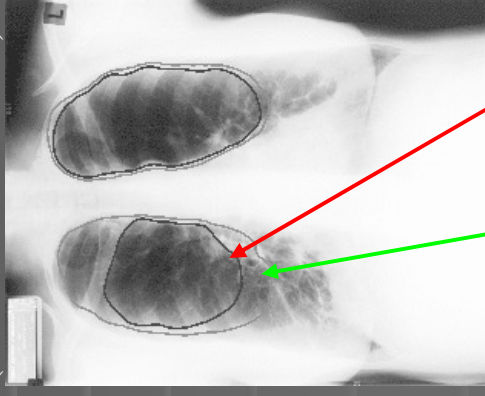
Chest x-ray



Metric

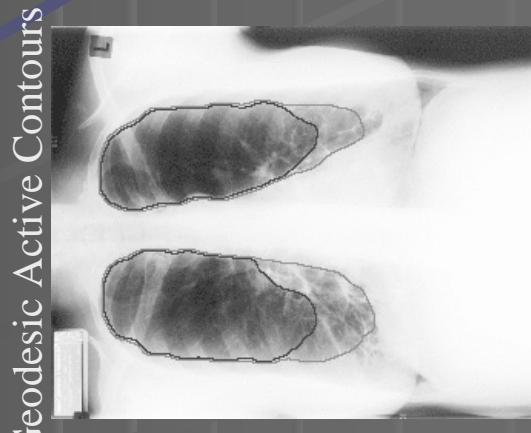


Geodesic Active Contours (Level set evolution)



Lung

Globally Optimal



Healthy lung

# Continuous maximal flows

- Simulate the flow  $\mathbf{F}$  of an incompressible fluid
- Pressure  $P$  is conserved under flow  $\mathbf{F}$
- $\mathbf{F}$  is driven by fluctuations in  $P$
- Boundary conditions:

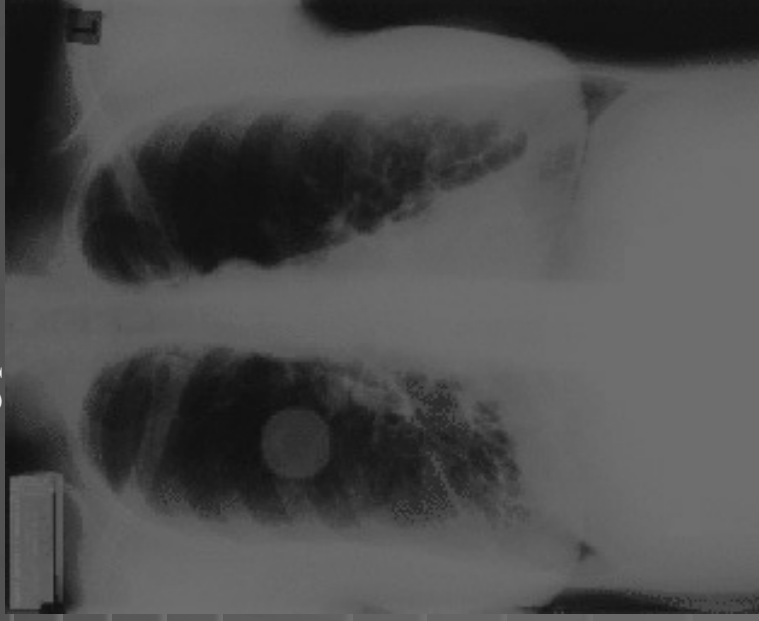
$$P_s = 1$$

$$P_t = 0$$

$$\frac{\partial P}{\partial t} = -\operatorname{div} \mathbf{F}$$
$$\frac{\partial \mathbf{F}}{\partial t} = -\nabla P$$
$$|\mathbf{F}| \leq 8$$

# Example: Lung Segmentation

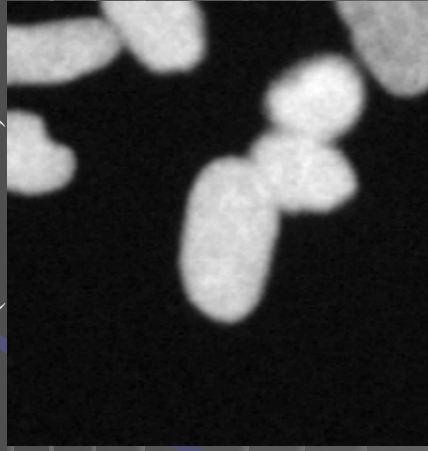
Evolving potential field (P)



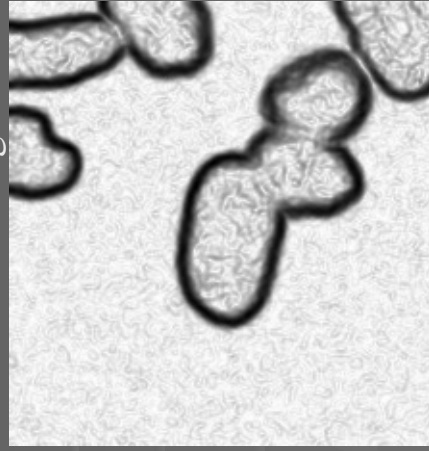
# Results

## 2D segmentation

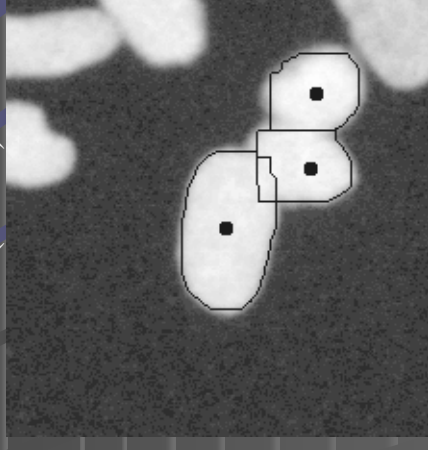
Microscope cell image  
(231 × 221)



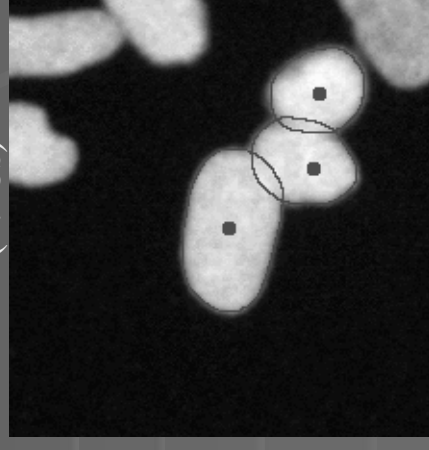
Metric image



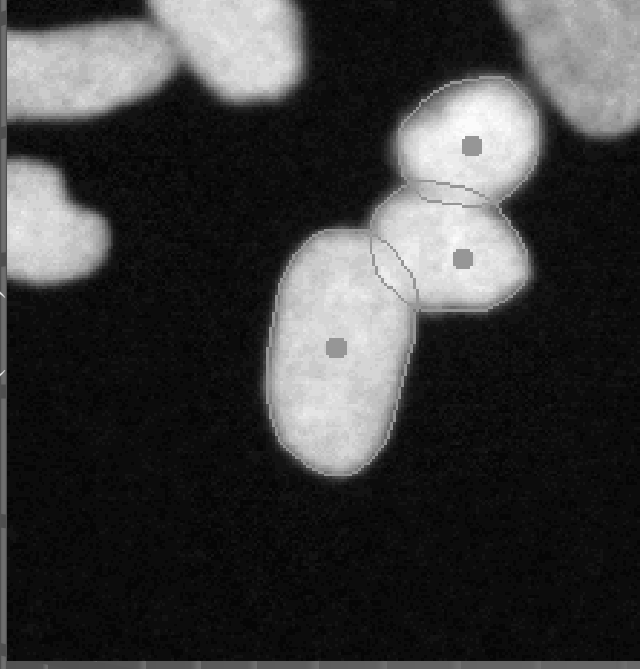
Minimum cuts  
(12.9s)



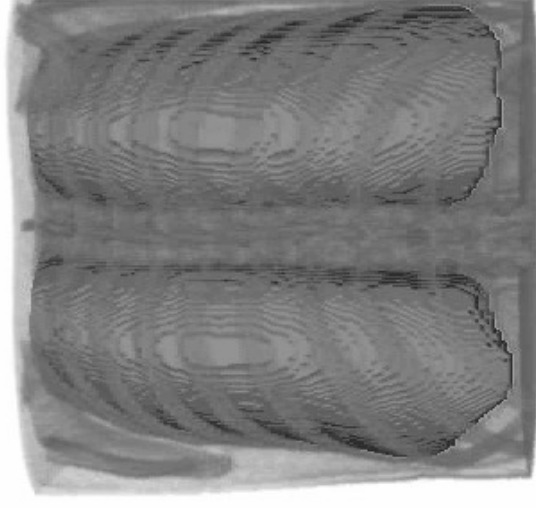
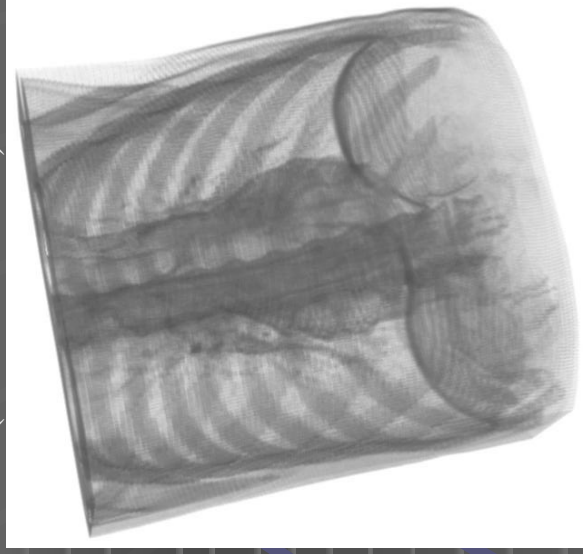
GOGAC  
(1.48s)



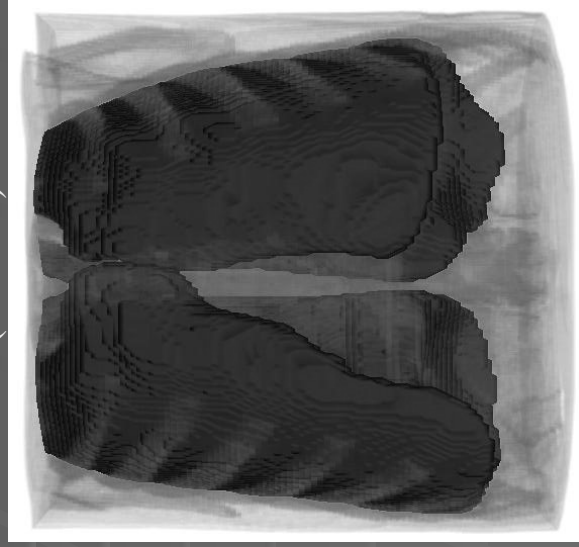
Globally minimal surfaces  
(5.1s)



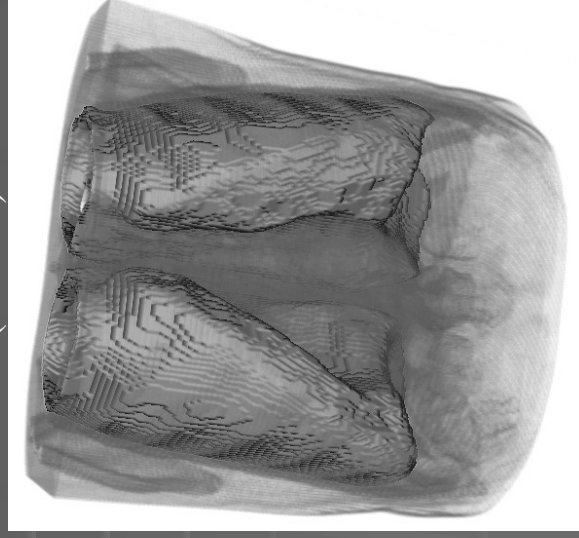
Chest CT  
( $200 \times 160 \times 90$ )



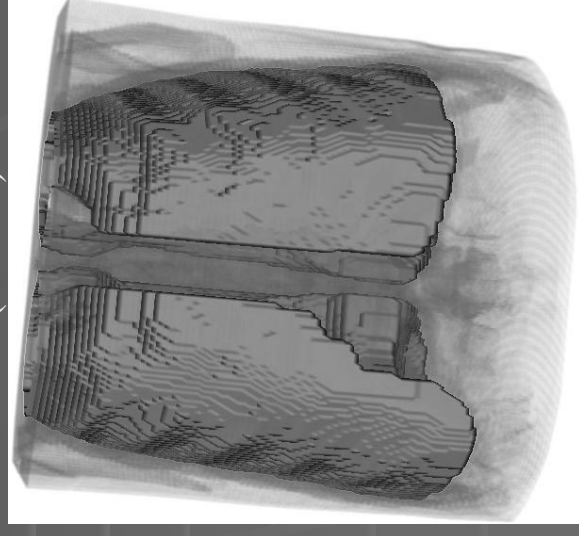
Globally minimal surface  
(8.5 min)



Classical Geodesic Active Surface  
(22 min)



Minimum cut  
(14 min)



# Results

## 3D reconstruction



Minimal surface (140 sec)



Minimum cut (230 sec)



# Summary

- Active contours have emerged as a powerful tool in segmentation
- Great improvements have been made as viewpoints change
  - Snakes → Level Sets (topology)
  - Level Sets → Minimal surfaces (optimality)