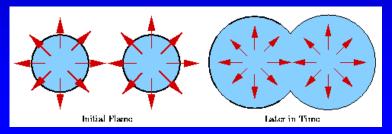
Level Set Methods

- Contour evolution method due to J. Sethian and S. Osher, 1988
- www.math.berkeley.edu/~sethian/level_set.html
- Difficulties with snake-type methods
 - Hard to keep track of contour if it self-intersects during its evolution
 - Hard to deal with changes in topology

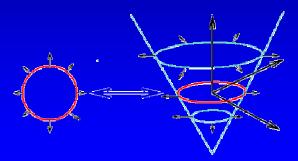


- The level set approach:
 - Define problem in 1 higher dimension
 - Define level set function $z = \phi(x,y,t=0)$ where the (x,y) plane contains the contour, and z = signed Euclidean distance transform value (negative means inside closed contour, positive means outside contour)

How to Move the Contour?

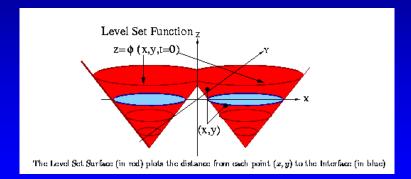
- Move the level set function, $\phi(x,y,t)$, so that it rises, falls, expands, etc.
- Contour = cross section at z = 0, i.e.,

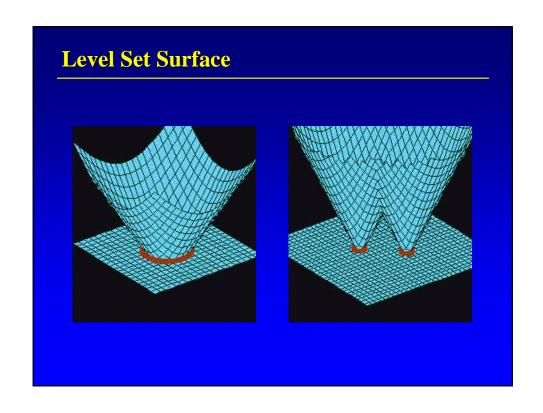
$$\{(x,y) \mid \phi(x,y,t) = 0\}$$



Level Set Surface

• The zero level set (in blue) at one point in time as a slice of the level set surface (in red)





How to Move the Level Set Surface?

- 1. Define a velocity field, F, that specifies how contour points move in time
 - Based on application-specific physics such as time, position, normal, curvature, image gradient magnitude
- 2. Build an initial value for the level set function, $\phi(x,y,t=0)$, based on the initial contour position
- 3. Adjust ϕ over time; contour at time t defined by $\phi(x(t), y(t), t) = 0$

$$\frac{\partial \Phi}{\partial t} + \vec{F} \cdot \nabla \Phi = 0$$
 Hamilton-Jacobi equation

$$\frac{\partial \mathbf{\Phi}}{\partial t} + \mathbf{F} \left(\left(\frac{\partial \mathbf{\Phi}}{\partial \mathbf{x}} \right)^2 + \left(\frac{\partial \mathbf{\Phi}}{\partial \mathbf{y}} \right)^2 \right)^{1/2} = 0$$

Level Set Formulation

• Constraint: level set value of a point on the contour with motion x(t) must always be 0

$$\phi(x(t), t) = 0$$

• By the chain rule

$$\phi_t + \nabla \phi(x(t), t) \cdot x'(t) = 0$$

- Since F supplies the speed in the outward normal direction $\chi'(t) \cdot n = F$, where $n = \nabla \phi / |\nabla \phi|$
- Hence evolution equation for ϕ is

$$\phi_t + F|\nabla \phi| = 0$$

Speed Function

$$F(k) = F_0 + F_I(k) = (1 - \varepsilon k)$$

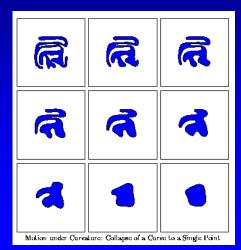
$$F(k) = k_I (x, y) * (1 - \varepsilon k)$$

$$\mathbf{k}_I = \frac{1}{1 + |\nabla \mathbf{G}_{\sigma} * \mathbf{I}(\mathbf{x}, \mathbf{y})|}$$

$$\mathbf{k}_I = \mathbf{e}^{-|\nabla \mathbf{G}_{\sigma} * \mathbf{I}(\mathbf{x}, \mathbf{y})|}$$

Example: Shape Simplification

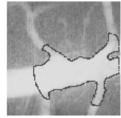
• $F = 1 - 0.1\kappa$ where κ is the curvature at each contour point



Example: Segmentation

- Digital Subtraction Angiogram
- F based on image gradient and contour curvature



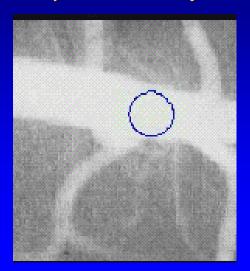


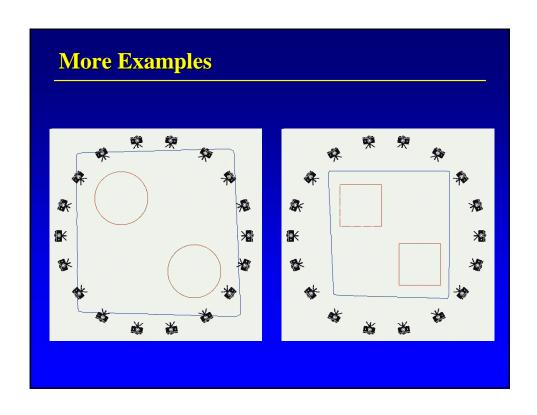


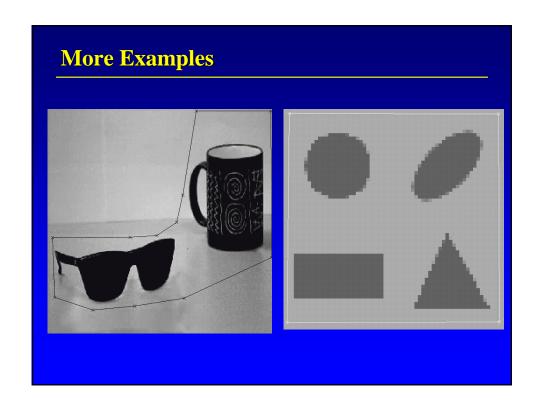
Evolving Front Driven by Function of Image Gradient

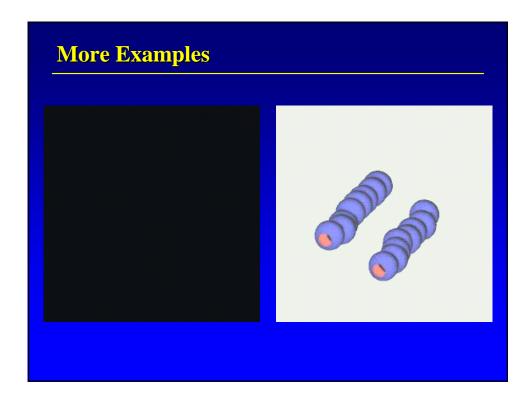
Example (cont.)

• Initial contour specified manually







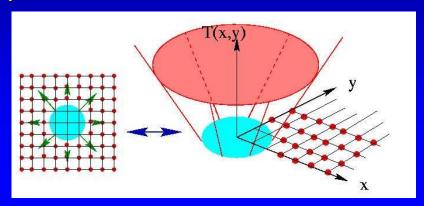


Fast Marching Method

- J. Sethian, 1996
- Special case that assumes the velocity field, F, never changes sign. That is, contour is either always expanding (F>0) or always shrinking (F<0)
- Convert problem to a stationary formulation on a discrete grid where the contour is guaranteed to cross each grid point at most once

Fast Marching Method

- Compute T(x,y) = time at which the contour crosses grid point (x,y)
- At any height, t, the surface gives the set of points reached at time t



Fast Marching Algorithm

- Compute T using the fact that
 - Distance = rate x time

• In 1D: $1 = F \times dT/dx$

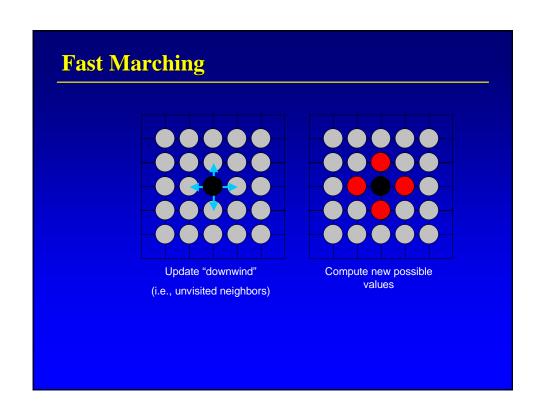
• In 2D: $1 = F \times |\nabla T|$

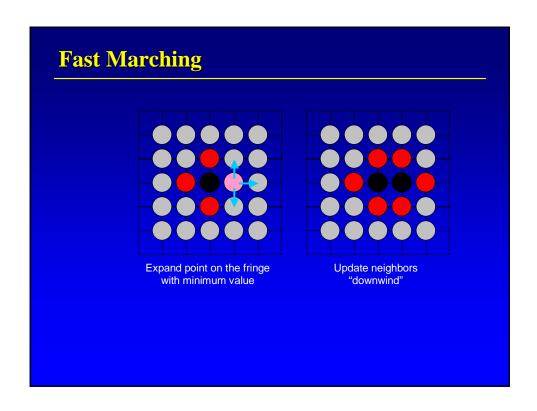
• Contour at time t =

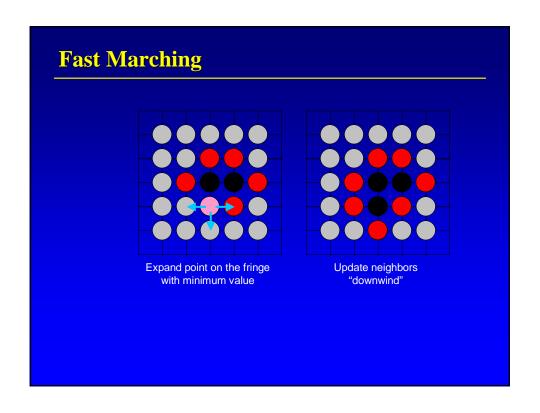
 $\{(x,y)\mid T(x,y)=t\}$

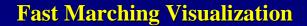
Fast Marching Algorithm

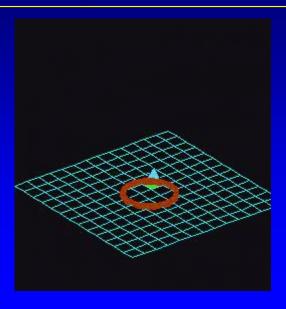
- Construct the arrival time surface T(x,y) incrementally:
 - 1. Build the initial contour
 - 2. Incrementally add on to the existing surface the part that corresponds to the contour moving with speed F (in other words, repeatedly pick a point on the fringe with minimum T value)
 - 3. Iterate until F goes to 0
- Builds level set surface by "scaffolding" the surface patches farther and farther away from the initial contour











Fast Marching + Level Set for Shape Recovery

1. First use the Fast Marching algorithm to obtain "rough" contour

$$|\nabla \mathbf{T}|\mathbf{F} = 1$$
, $\mathbf{F} = e^{-\alpha|\nabla G_{\sigma}*I(\mathbf{x},\mathbf{y})|}$

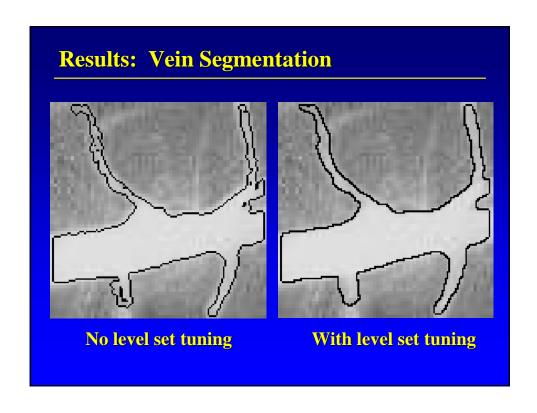
2. Then use the Level Set algorithm to fine tune, using a few iterations, the results from Fast Marching

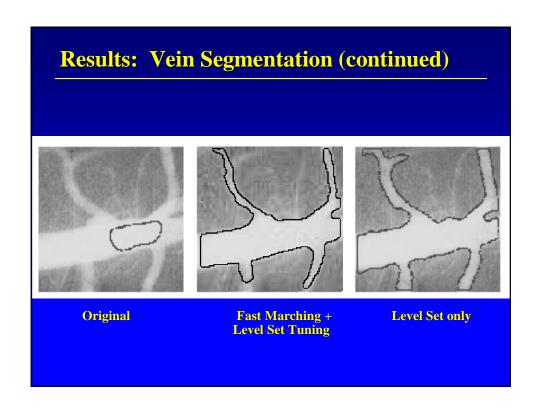
$$\frac{\partial \mathbf{\Phi}}{\partial \mathbf{t}} + \mathbf{k}_{\mathbf{I}} (1 - \varepsilon \mathbf{K}) |\nabla \mathbf{\Phi}| - \beta \nabla \mathbf{P} \cdot \nabla \mathbf{\Phi} = \mathbf{0}$$

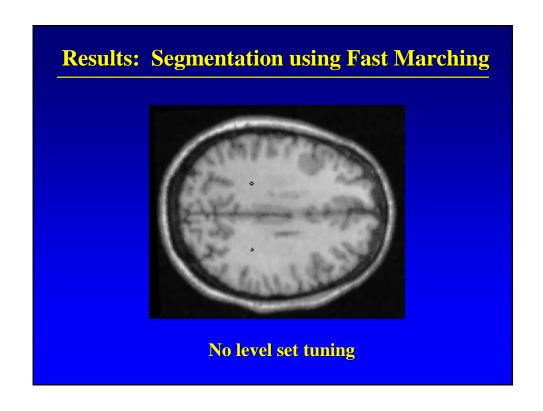
$$k_{I} = \frac{I}{I + |\nabla G_{\sigma} * I(x, y)|}$$

$$P(x, y) = -|\nabla G_{\sigma} * I(x, y)|$$

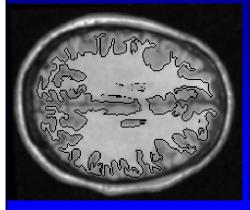


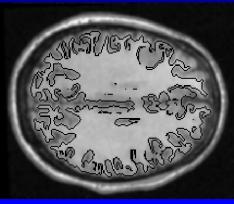






Results: Brain Image Segmentation



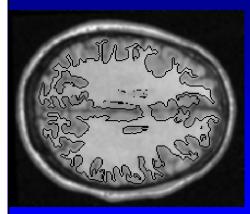


of iterations = 9000

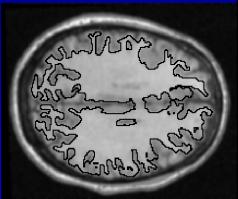
of iterations = 12000

Fast marching only, no level set tuning

Results: Brain Segmentation (continued)



Without level set tuning



With level set tuning

