
Computer Science
Department



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Geometric Active Contours

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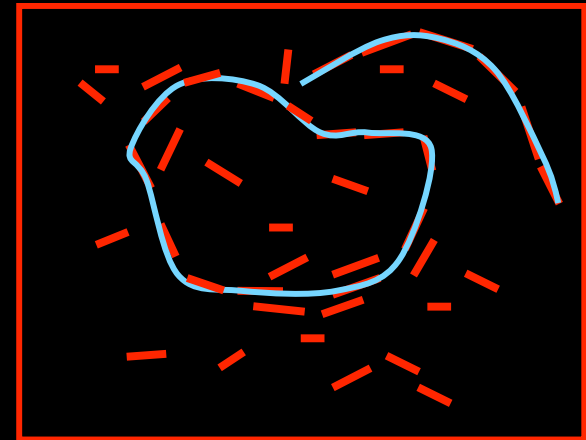
Geometric Image Processing Lab

Edge Detection

q Edge Detection:

- u The process of labeling the locations in the image where the gray level's "rate of change" is high.

n **OUTPUT:** "edgels" locations,
direction, strength



q Edge Integration:

- u The process of combining "local" and perhaps sparse and non-contiguous "edgel"-data into meaningful, long edge curves (or closed contours) for segmentation

n **OUTPUT:** edges/curves consistent with the local data

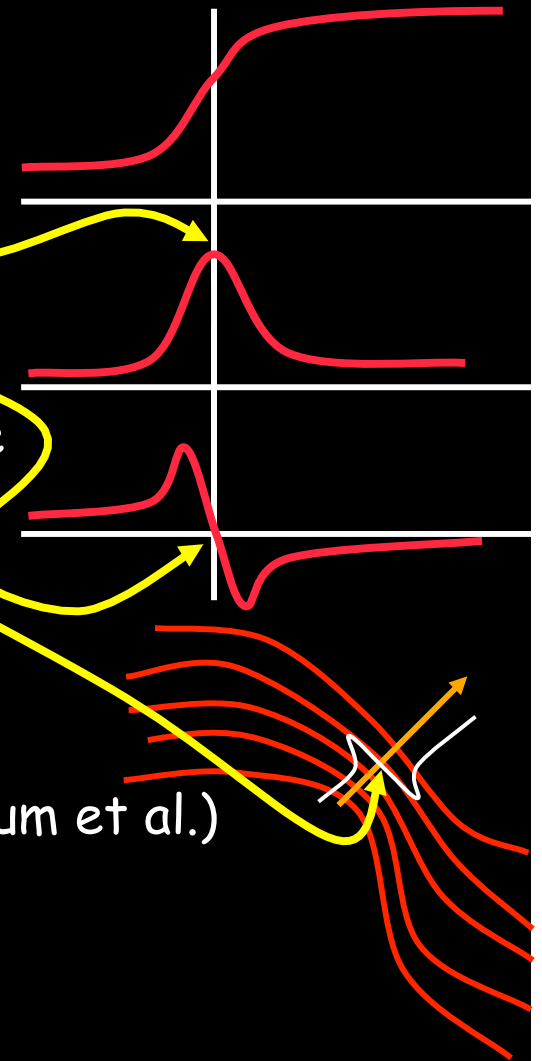
The Classics

q Edge detection:

- u Sobel, Prewitt, Other gradient estimators
- u Marr Hildreth
- u Haralick/Canny/Deriche et al.
- “optimal” directional local max of derivative

q Edge Integration:

- u tensor voting (Rom, Medioni, Williams, ...)
- u dynamic programming (Shashua & Ullman)
- u generalized “grouping” processes (Lindenbaum et al.)



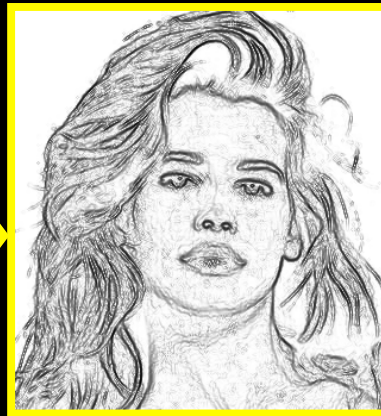
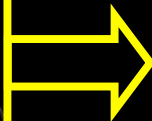


The "New-Wave"

- q Snakes
- q Geodesic Active Contours
- q Model Driven Edge Detection



Image



Edge Indicator
Function



"nice" curves that optimize a
functional of $g(\cdot)$, i.e.

nice: "regularized", smooth,
fit some prior information

Edge Curves

$$\frac{1}{1 + |\nabla(G_\sigma * I)|^2}$$

Geodesic Active Contours

- q Snakes [Terzopoulos-Witkin-Kass 88](#)
 - u Linear functional \Rightarrow efficient implementation
 - u non-geometric \Rightarrow depends on parameterization
- q Open geometric scaling invariant, [Fua-Leclerc 90](#)
- q Non-variational geometric flow
[Caselles et al. 93](#), [Malladi et al. 93](#)
 - u Geometric, yet does not minimize any functional
- q Geodesic active contours [Caselles-Kimmel-Sapiro 95](#)
 - u derived from geometric functional
 - u non-linear \Rightarrow inefficient implementations:
 - n Explicit Euler schemes limit numerical step for stability
- q Level set method [Ohta-Jansow-Karasaki 82](#), [Osher-Sethian 88](#)
 - u automatically handles contour topology
- q Fast geodesic active contours [Goldenberg-Kimmel-Rivlin-Rudzsky 99](#)
 - u no limitation on the time step
 - u efficient computations in a narrow band



Laplacian Active Contours

- q Closed contours on vector fields
 - u Non-variational models Xu-Prince 98, Paragios et al. 01
 - u A variational model Vasilevskiy-Siddiqi 01
- q Laplacian active contours open/closed/robust
Kimmel-Bruckstein 01



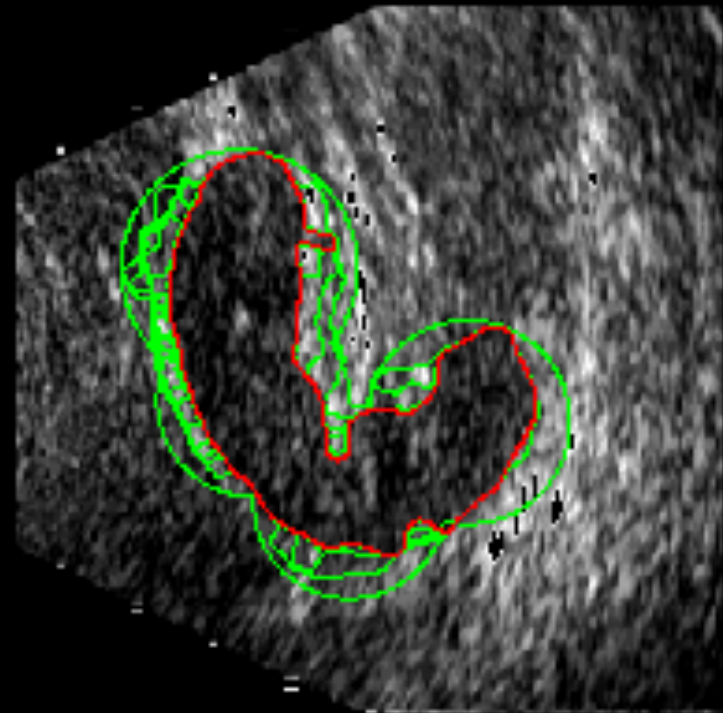
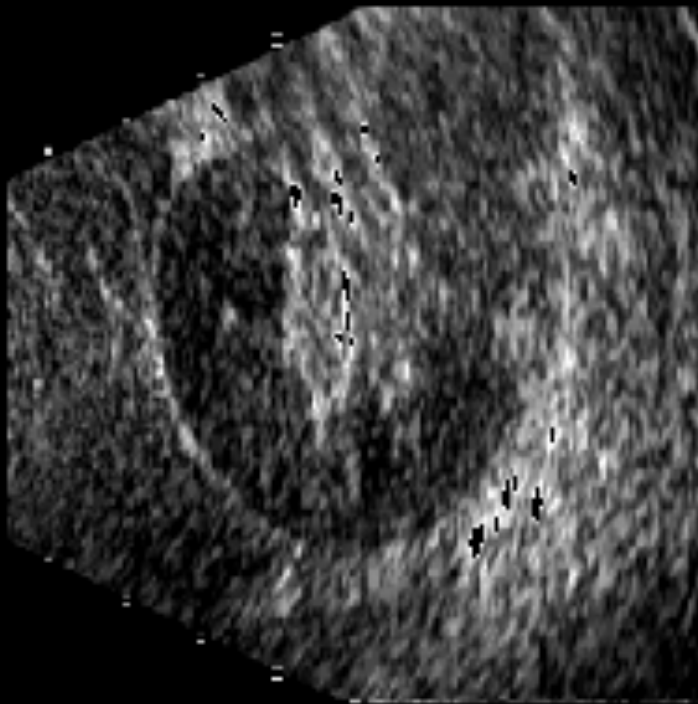
Most recent:
variational measures
for good old operators
Kimmel-Bruckstein 03

Segmentation



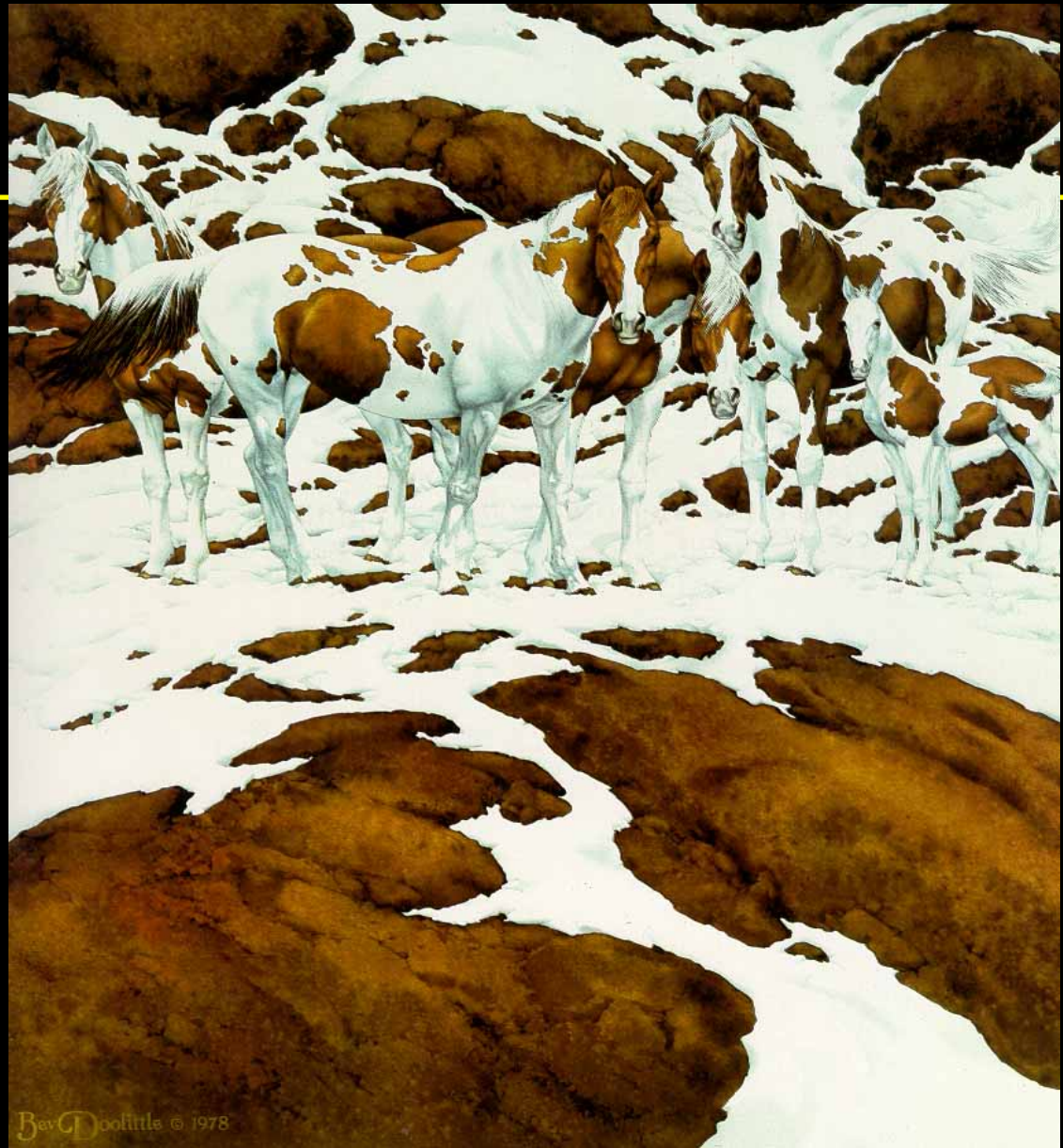
Segmentation

q Ultrasound images



Caselles, Kimmel, Sapiro ICCV'95

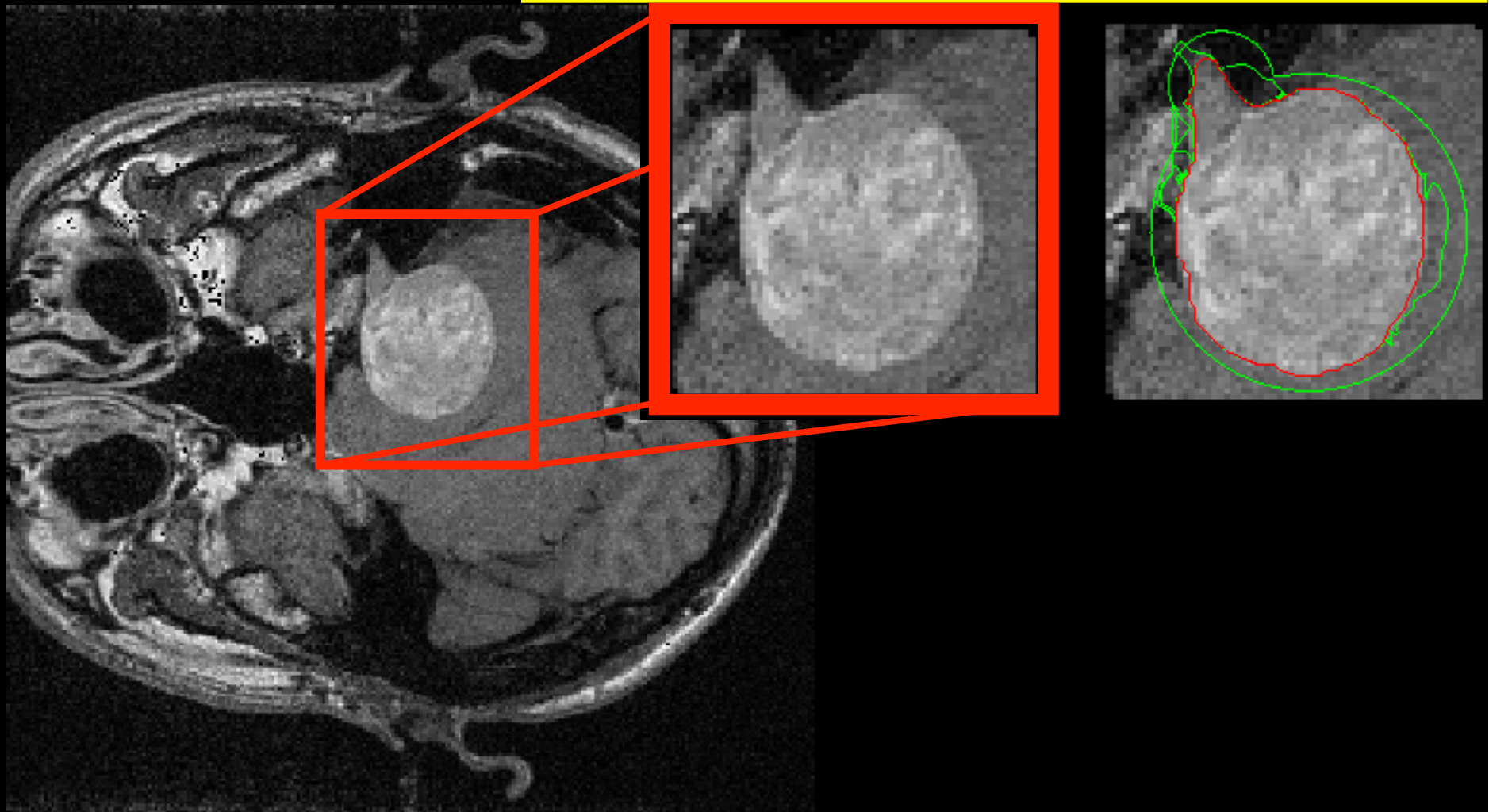
Pintos



Woodland Encounter Bev Doolittle 1985



Segmentation



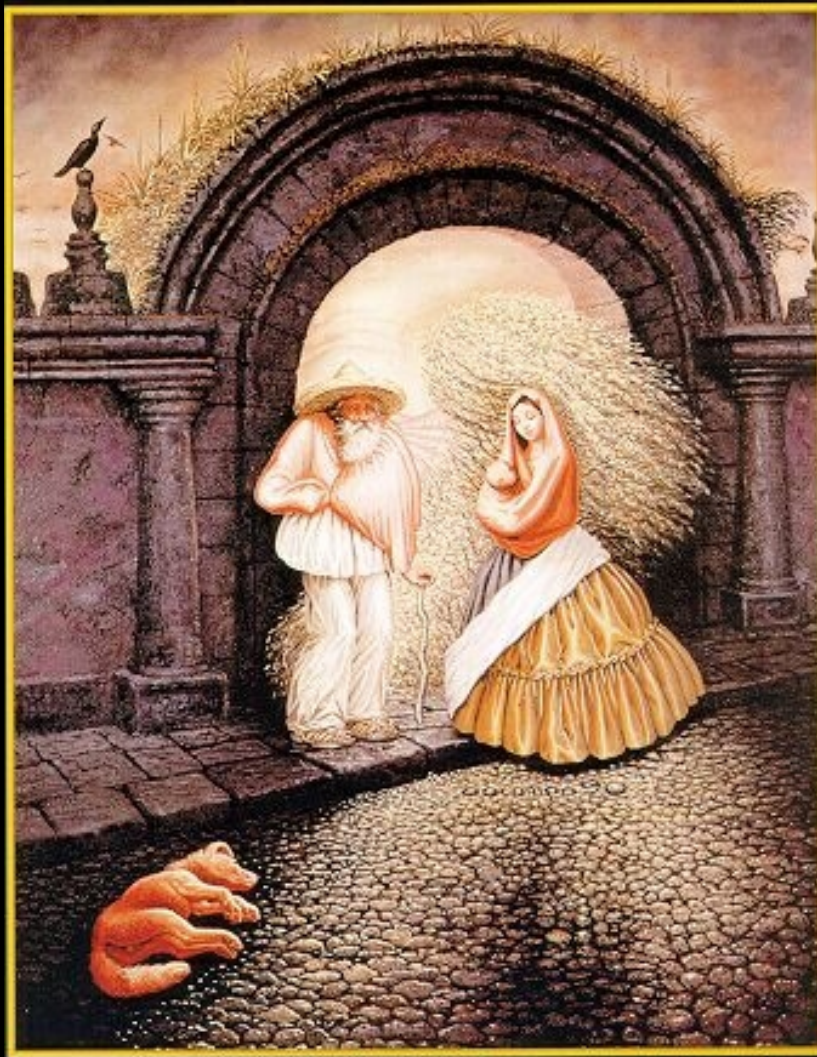
Caselles, Kimmel, Sapiro ICCV'95

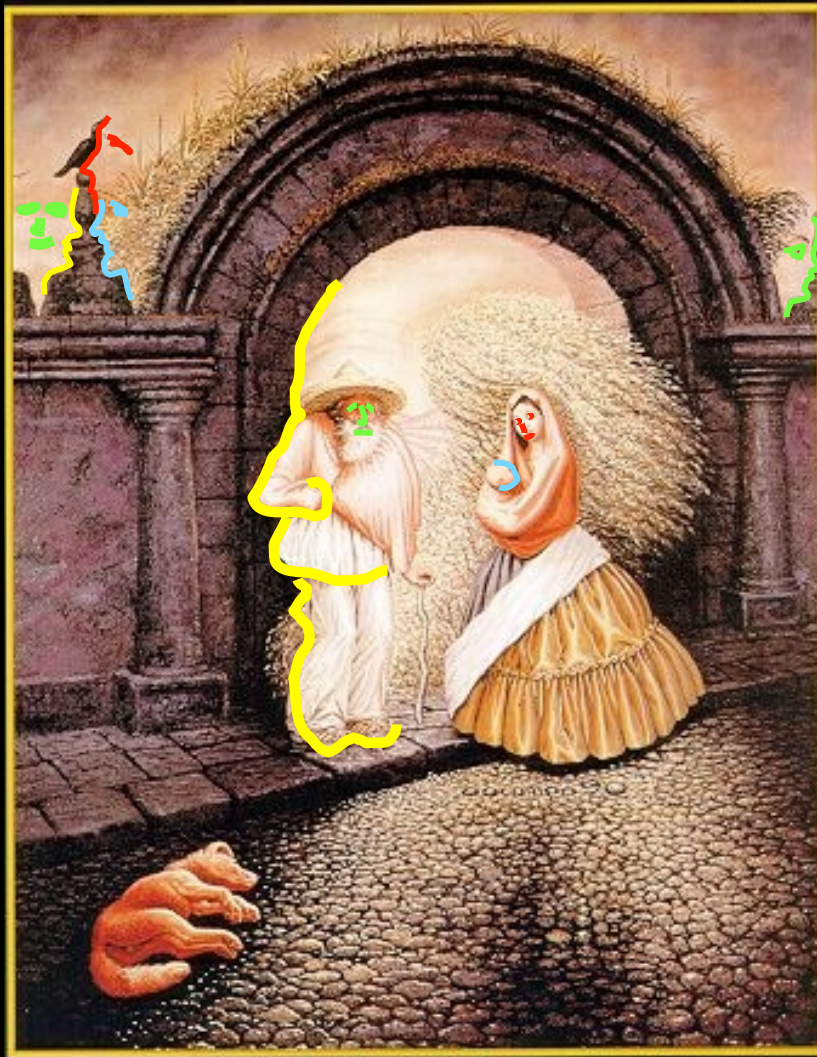
Prior knowledge...



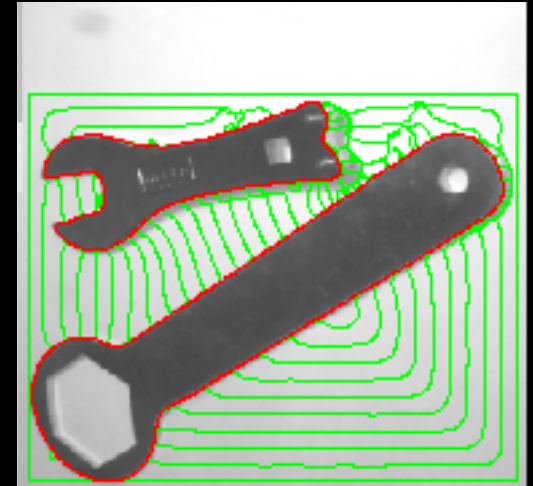
Prior knowledge...







Segmentation



Caselles, Kimmel, Sapiro ICCV'95

Segmentation

q With a good prior who needs the data...

Wrong Prior???

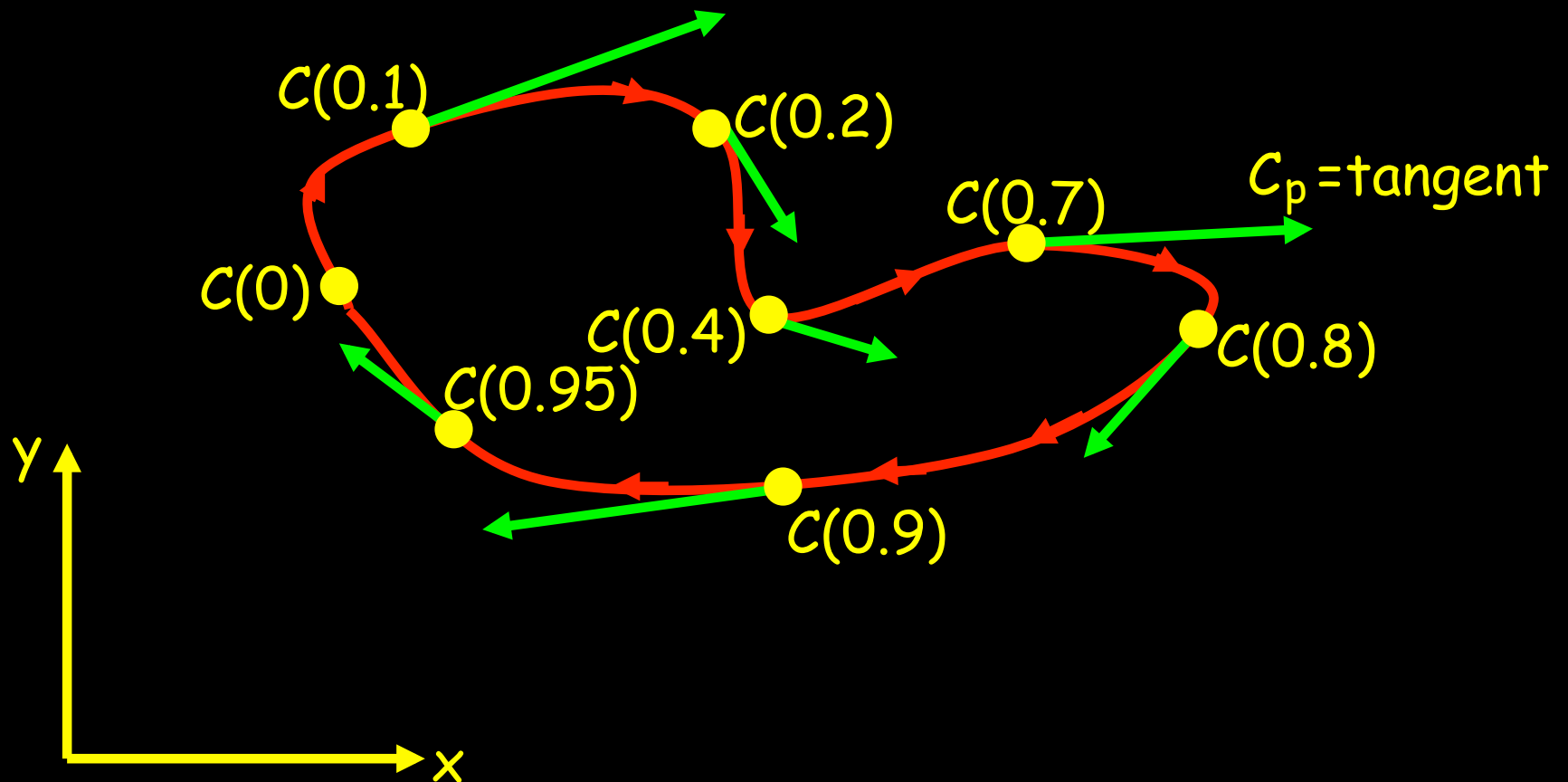


Wrong Prior???



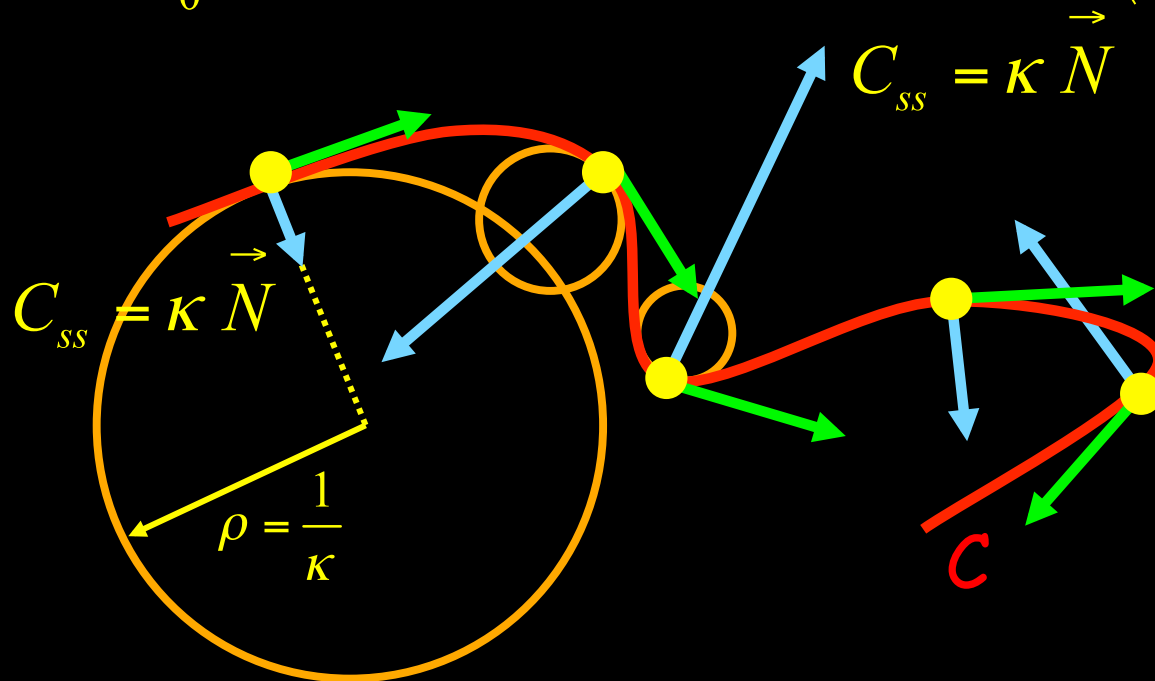
Curves in the Plane

$C(p) = \{x(p), y(p)\}, \quad p \in [0, 1]$



Arc-length and Curvature

$$s(p) = \int_0^p |C_p| dp \Rightarrow |C_s| = 1, \quad \left(C_s = \frac{C_p}{|C_p|} \right)$$



Calculus of Variations

Find C for which $E = \int_0^1 L(C, C_p) dp$ is an extremum

$$E(\varepsilon) = \int L(x + \varepsilon\eta, y, x_p + \varepsilon\eta_p, y_p) dp$$

$$(\delta E)_x = \left(\frac{\partial E}{\partial \varepsilon} \right)_{\varepsilon=0} = \int \left(L_x - \frac{d}{dp} L_{x_p} \right) \eta dp$$

Euler-Lagrange:

$$\left\{ \begin{array}{l} L_x - \frac{d}{dp} L_{x_p} = 0 \\ L_y - \frac{d}{dp} L_{y_p} = 0 \end{array} \right\}$$

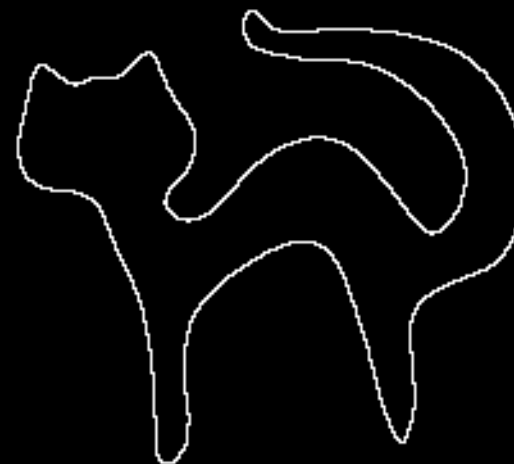
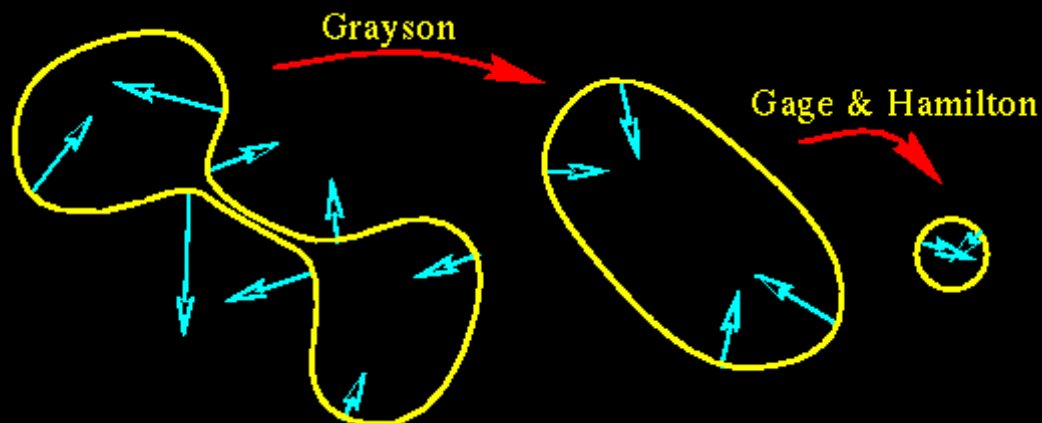
Calculus of Variations

Important Example

⇒ Euler-Lagrange: $\kappa \vec{N} |C_p| = 0$, setting $E = \int_0^1 |C_p| dp$ $ds = |C_p| dp$

⇒

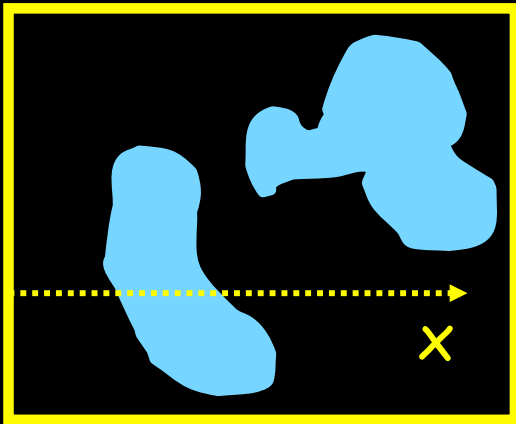
⇒ Curvature flow



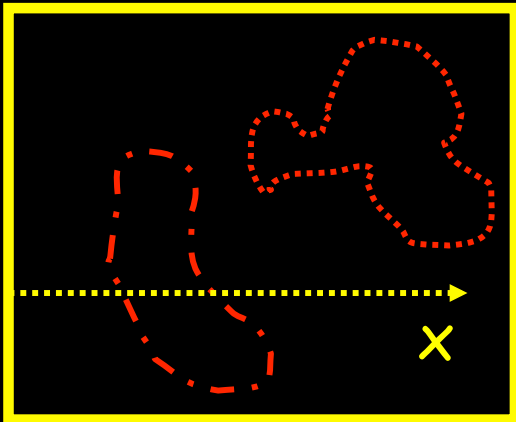
$$C_t = C_{ss} \quad (C_{ss} = \kappa \vec{N})$$

Potential Functions (g)

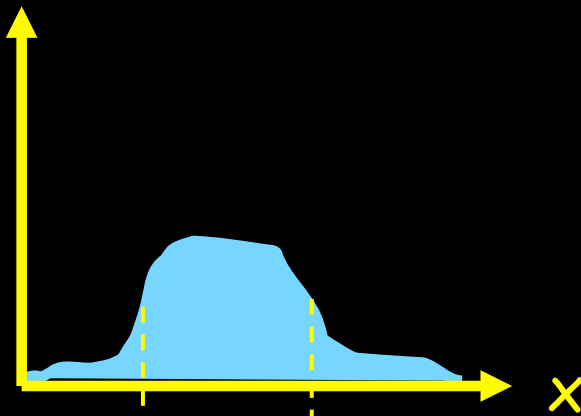
$I(x,y)$
Image



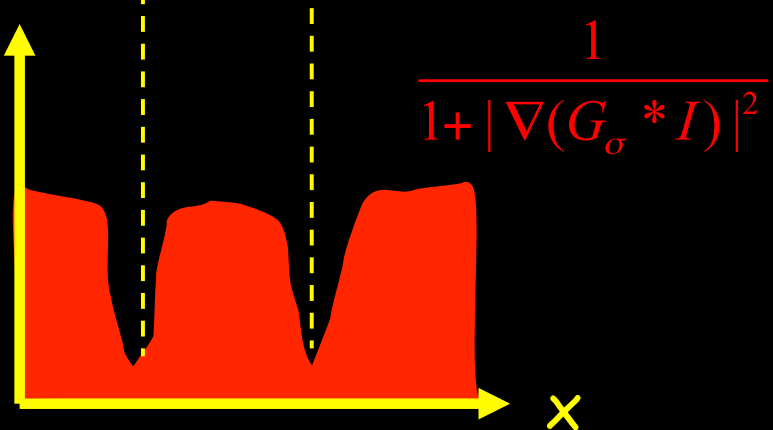
$g(x,y)$
Edges



$I(x)$



$g(x)$





Snakes & Geodesic Active Contours

q Snake model

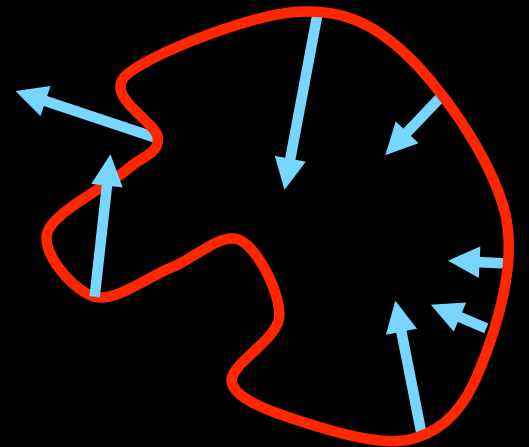
Terzopoulos-Witkin-Kass 88

q Euler Lagrange as a gradient descent

q Geodesic active contour model

Caselles-Kimmel-Sapiro 95

q Euler Lagrange gradient descent

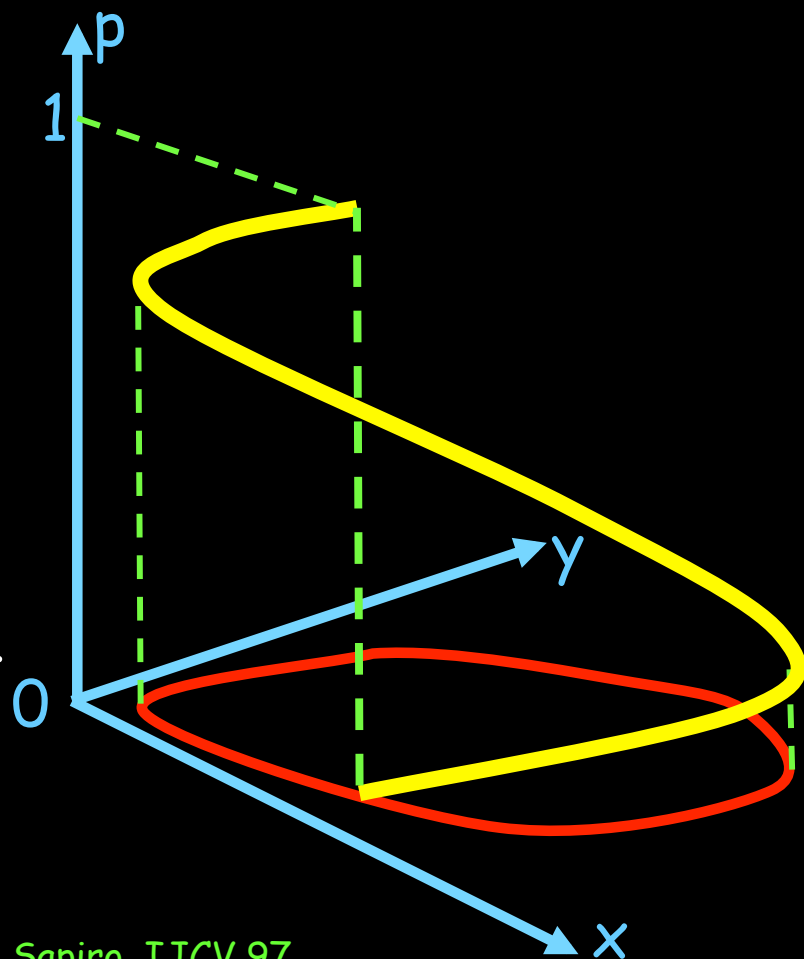


Maupertuis Principle of Least Action

$$\arg \min_C \int_0^1 (g(C(p)) + |C_p|^2) dp =$$
$$\arg \min_C \int_0^L \sqrt{g(C(s)) + E_0} ds$$

Snake = Geodesic active contour
up to some E_0 , i.e. $\tilde{g} = \sqrt{g + E_0}$

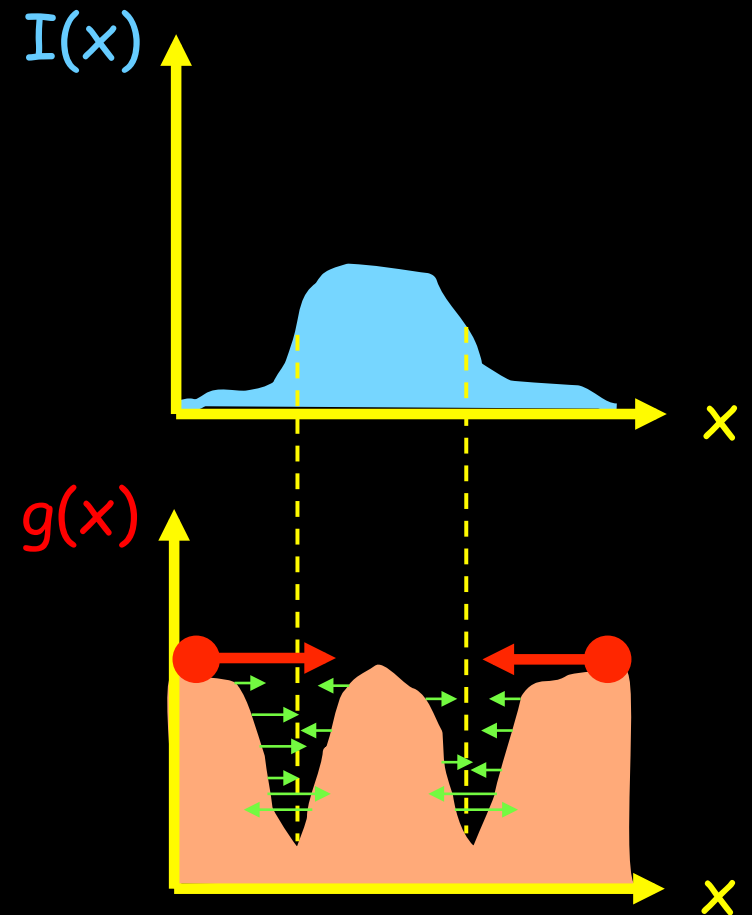
- ⇒ Snakes depend on parameterization.
- ⇒ Different initial parameterizations yield solutions for different geometric functionals



Caselles Kimmel Sapiro, IJCV 97

Geodesic Active Contours in 1D

Geodesic active contours are
reparameterization invariant



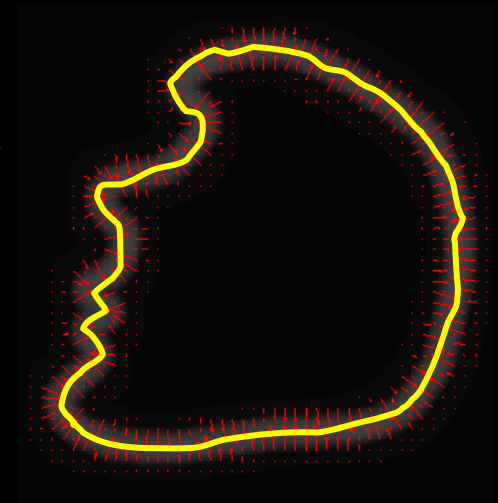
Geodesic Active Contours in 2D



$G_s * I$

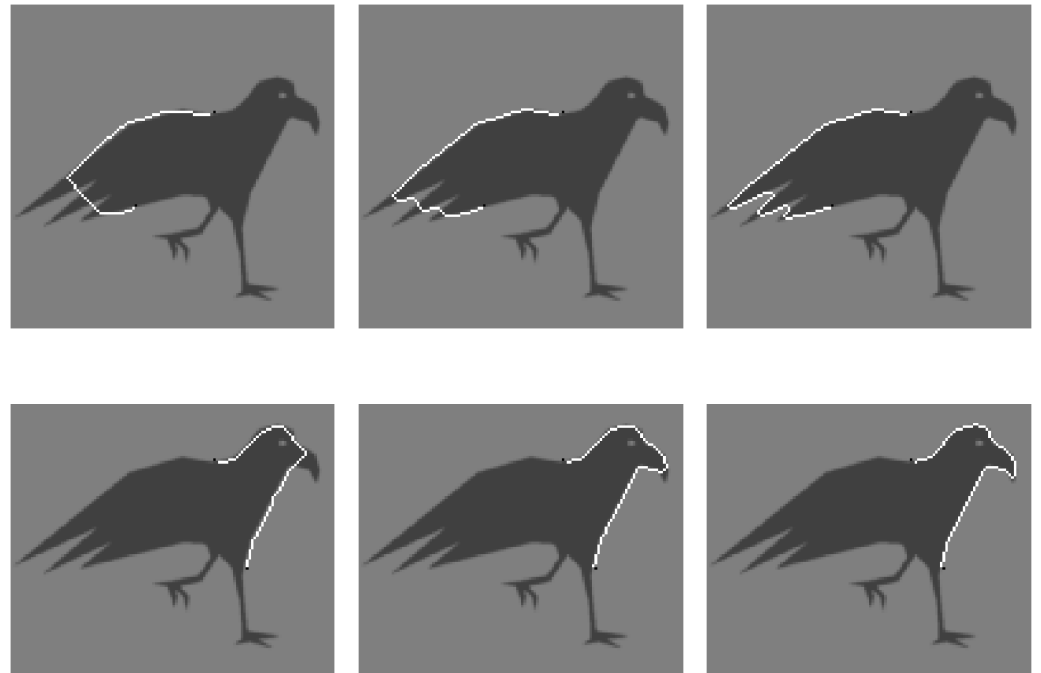
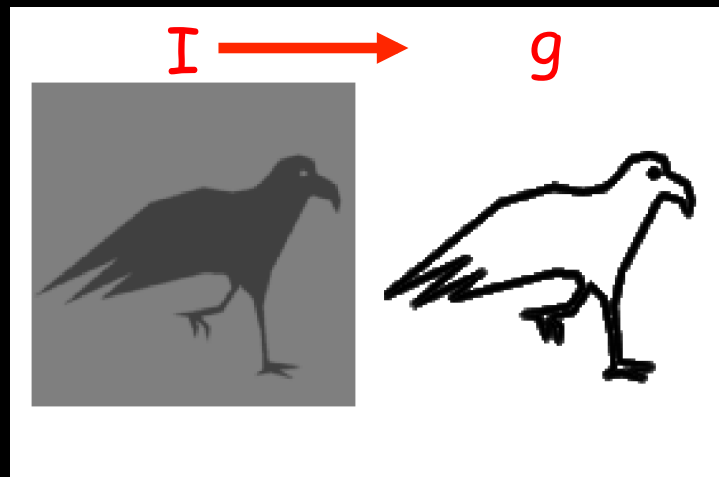


$$g(x) = \frac{1}{1 + |\nabla(G_\sigma * I)|^2}$$



Controlling -max

Smoothness

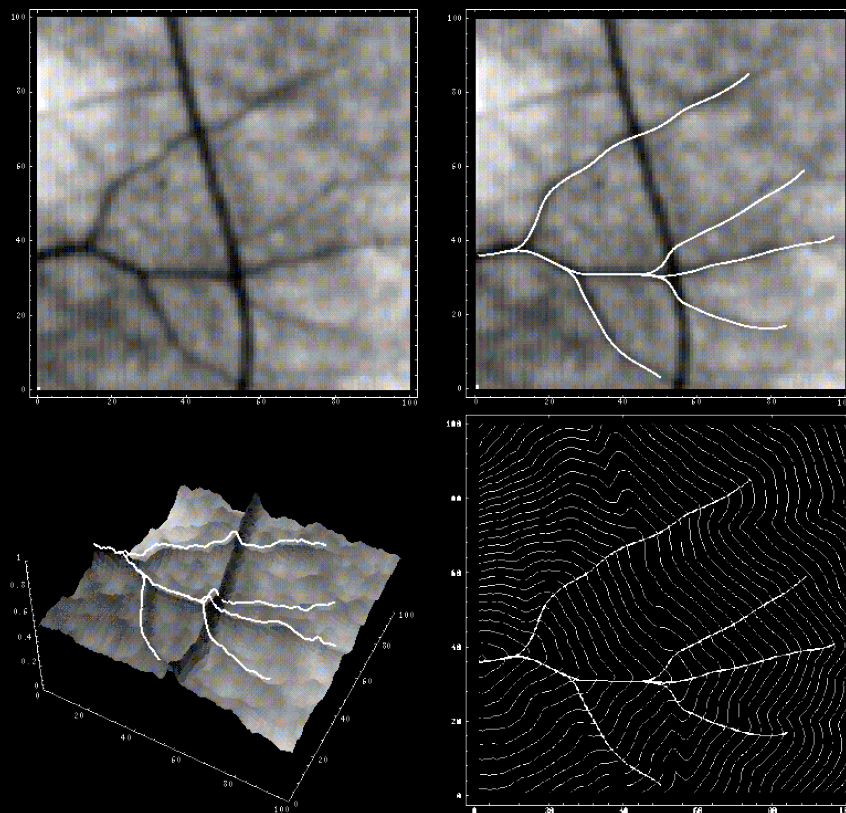


Cohen Kimmel, IJCV 97

Fermat's Principle

In an isotropic medium, the paths taken by light rays are extremal geodesics w.r.t.

i.e.,



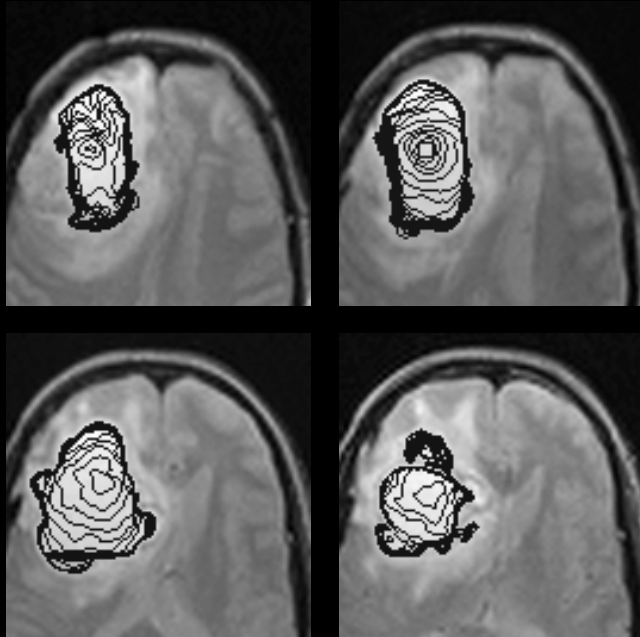
Cohen Kimmel, IJCV 97

Experiments - Color Segmentation



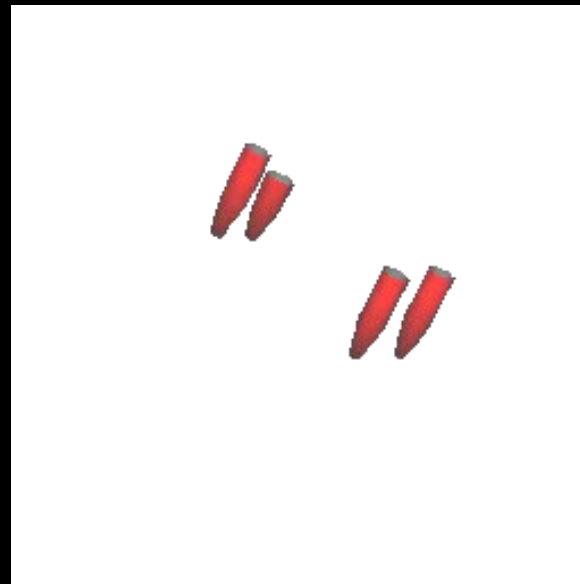
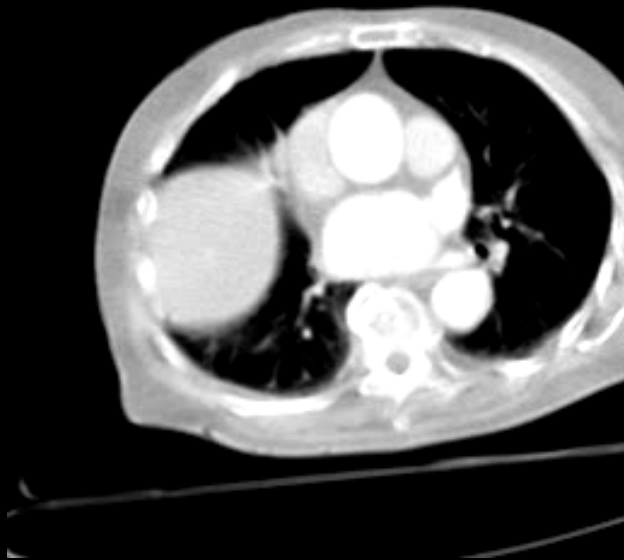
Goldenberg, Kimmel, Rivlin, Rudzsky,
IEEE T-IP 2001

Tumor in 3D MRI



Caselles, Kimmel, Sapiro, Sbert, IEEE T-PAMI 97

Segmentation in 4D



Malladi, Kimmel, Adalsteinsson,
Caselles, Sapiro, Sethian
SIAM Biomedical workshop 96



Tracking in Color Movies



Goldenberg, Kimmel, Rivlin, Rudzsky,
IEEE T-IP 2001



Tracking in Color Movies



Goldenberg, Kimmel, Rivlin, Rudzsky,
IEEE T-IP 2001

