

Computer Vision-IT416

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What is an Active Contour?

Given: Approximate boundary (contour) around the object

Task: Evolve (move) the contour to fit exact object boundary



Image

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Active Contour:

Iteratively “deform” the initial contour so that:

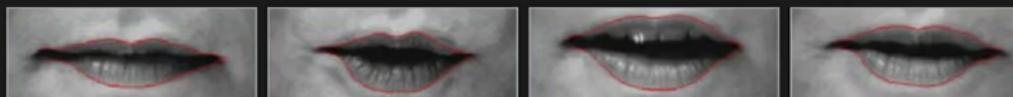
- It is near pixels with high gradient (edges)
- It is smooth



Image

Power of Deformable Contours

Boundaries could deform over time



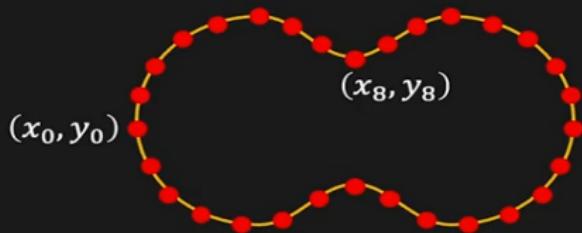
Boundaries could deform with viewpoint



Boundary Tracking: Use the boundary from the current image as initial boundary for the next image.

Representing a Contour

Contour \mathbf{v} : An ordered list of 2D vertices (control points) connected by straight lines of fixed length



$$\mathbf{v} = \{v_i = (x_i, y_i) \mid i = 0, 1, 2, \dots, n - 1\}$$

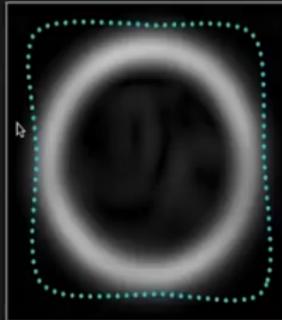
Attracting Contours to Edges



Image with
Initial Contour



Gradient Magnitude
Squared
 $\|\nabla I\|^2$



Blurred Gradient
Magnitude Squared
 $\|\nabla n_\sigma * I\|^2$

Attracting Contours to Edges

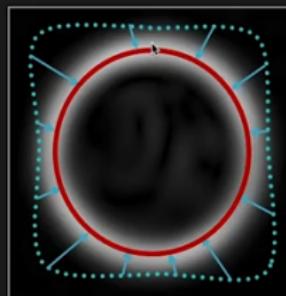


Image with
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Gradient Magnitude
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$$\|\nabla I\|^2$$



Blurred Gradient
Magnitude Squared

$$\|\nabla n_\sigma * I\|^2$$

Maximize Sum of Gradient Magnitude Square

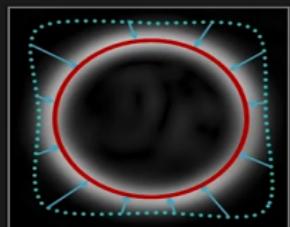
Attracting Contours to Edges



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Maximize Sum of Gradient Magnitude Square

≡ Minimize -ve (Sum of Gradient Magnitude Square)

≡ Minimize $E_{image} = - \sum_{i=0}^{n-1} \|\nabla n_\sigma * I(v_i)\|^2$

Contour Deformation: Greedy Algorithm

1. For each contour point v_i ($i = 0, \dots, n - 1$), move v_i to a position within a window W where the energy function E_{image} for the contour is minimum.



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Greedy solution might be suboptimal and slow.

Sensitivity to Noise and Initialization



Contour fitted to
gradient magnitude

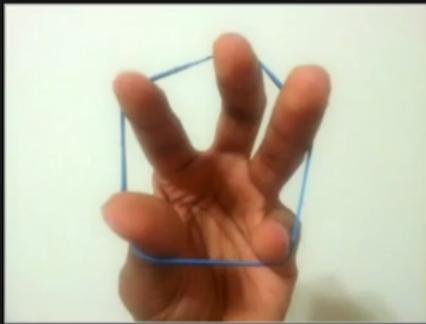
Sensitivity to Noise and Initialization



Contour fitted to
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Solution: Add constraints that make the contour
contract and remain smooth

Making Contours Elastic and Smooth

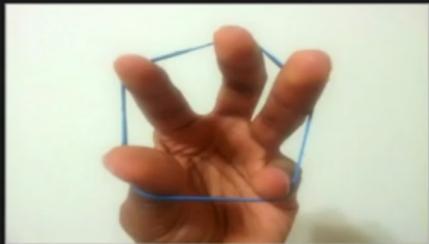


Elastic and contracts
like a rubber band



Smooth
like a metal strip

Making Contours Elastic and Smooth



Elastic and contracts
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Smooth
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Minimize Internal Bending Energy of the Contour:

$$E_{contour} = \alpha E_{elastic} + \beta E_{smooth}$$

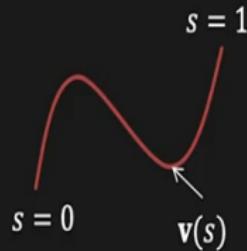
(α, β) : Control the influence of elasticity and smoothness

Elasticity and Smoothness

For point $0 \leq s \leq 1$ on continuous contour $\mathbf{v}(s) = (x(s), y(s))$:

$$E_{elastic} = \left\| \frac{d\mathbf{v}}{ds} \right\|^2$$

$$E_{smooth} = \left\| \frac{d^2\mathbf{v}}{ds^2} \right\|^2$$

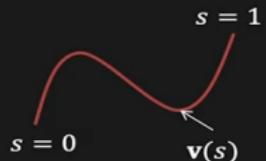


Elasticity and Smoothness

For point $0 \leq s \leq 1$ on continuous contour $\mathbf{v}(s) = (x(s), y(s))$:

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$$E_{smooth} = \left\| \frac{d^2\mathbf{v}}{ds^2} \right\|^2$$



Discrete approximations at control point \mathbf{v}_i :

$$E_{elastic}(\mathbf{v}_i) = \left\| \frac{d\mathbf{v}}{ds} \right\|^2 \approx \|\mathbf{v}_{i+1} - \mathbf{v}_i\|^2 = (x_{i+1} - x_i)^2 + (y_{i+1} - y_i)^2$$

$$E_{smooth}(\mathbf{v}_i) = \left\| \frac{d^2\mathbf{v}}{ds^2} \right\|^2 \approx \|(\mathbf{v}_{i+1} - \mathbf{v}_i) - (\mathbf{v}_i - \mathbf{v}_{i-1})\|^2$$

$$= (x_{i+1} - 2x_i + x_{i-1})^2 + (y_{i+1} - 2y_i + y_{i-1})^2$$

Elasticity and Smoothness

Internal bending energy along the entire contour:

$$E_{contour} = \alpha E_{elastic} + \beta E_{smooth}$$

where:

$$E_{elastic} = \sum_{i=0}^{n-1} [(x_{i+1} - x_i)^2 + (y_{i+1} - y_i)^2]$$

$$E_{smooth} = \sum_{i=0}^{n-1} [(x_{i+1} - 2x_i + x_{i-1})^2 + (y_{i+1} - 2y_i + y_{i-1})^2]$$

Combining the Forces

Image Energy, E_{image} : Measure of how well the contour latches on to edges

Internal Energy, $E_{contour}$: Measure of elasticity and smoothness

Total Energy of Active Contour:

$$E_{total} = E_{image} + E_{contour}$$

Minimize the Total Energy

Contour Deformation: Greedy Algorithm

1. Uniformly sample the contour to get n contour points.
2. For each contour point v_i ($i = 0, \dots, n - 1$), move v_i to a position within a window W where the energy function E_{total} for the entire contour is minimum.

$$E_{total} = E_{image} + E_{contour}$$

3. If the sum of motions of all the contour points is less than a threshold, stop. Else go to Step 1.



Result: Effect of Contour Constraint



Without contour constraint

$$E_{total} = E_{image}$$



With contour constraint

$$E_{total} = E_{image} + E_{contour}$$

Result: Boundary Around Two Objects



Large α

(More like a rubber band)



Small α

(Less like a rubber band)