Snakes: Active Contour Models

CS 6640
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(some material from Zoltan Kato http://www.cab.u-szeged.hu/~kato/variational/)

Materials

- Kass, Witkin, Terzopopoulos, Int. Journal of Computer Vision, 321-331, 1988
- Trucco and Verri, Introductory Techniques for 3-D Computer Vision, Chapter 5
- Bryan Morse, Lecture 21 Image Understanding

Websites

- Andy Witkin's homepage: <u>http://www.cs.cmu.edu/~aw/</u>
- Terzopoulos: http://www.cs.ucla.edu/~dt/vision.html
 - → Snakes: Active Contour Models
- Original snake demo: Kass, Witkin, Terzopoulos 1988: http://www.cs.ucla.edu/~dt/videos/deformable-models/snakes.avi
- Other Demos:
 - Xu/Prince: http://www.iacl.ece.jhu.edu/static/gvf/
 - http://users.ecs.soton.ac.uk/msn/book/new_demo/
 - http://www.markschulze.net/snakes/index.html

Snake's Energy Function

- Position of the snake v(s) = (x(s),y(s))
- $E_{\text{snake}} = \int [E_{\text{int}} v(s) + E_{\text{image}} v(s) + E_{\text{con}} v(s)] ds$
 - Internal: Internal energy due to bending. Serves to impose piecewise smoothness constraint
 - Image: Image forces pushing the snake toward image features (edges, etc...)
 - Constraints: External constraints are responsible for putting the snake near the desired local minimum

Internal Energy

- $E_{int} = [\alpha(s) |v_s(s)|^2 + \beta(s) |v_{ss}(s)|^2]$
 - First order term: membrane, α(s):"elasticity"
 - Second order term: thin plate, β(s): "rigidity, stiffness"
 - If $\alpha(s)=\beta(s)=0$, we allow breaks in the contour

Image Forces

- Edge Functional : negative gradient magnitude: $E_{edge} = |\nabla I(x,y)|^2$
- Better: negative gradient magnitude of Gaussian-smoothed image:

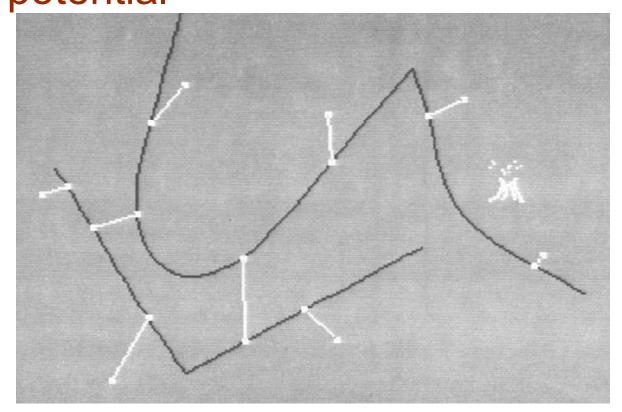
$$E_{edge} = - |\nabla G(\sigma) \otimes I(x,y)|^2$$

→ Attracts the snake to locations of large gradients = strong edges

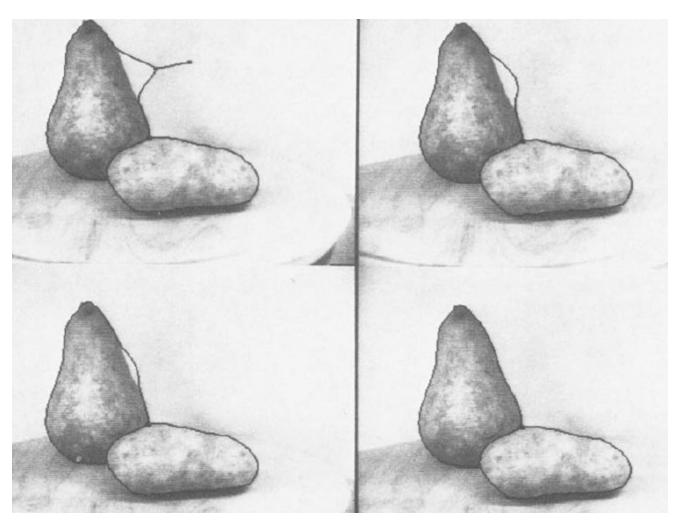
External Constraint Forces

Springs: add -k(x1-x2)2 to E_{con}

 Volcano: 1/r2 repulsion force, combine with image potential



External Constraint Forces: Spring



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Original Demo Kass et al., 1988



http://www.cs.ucla.edu/~dt/videos/deformable-models/snakes.avi

Numerical Solutions

$$-\frac{d^{2}}{ds^{2}}\left(\frac{\partial E}{\partial \left(\frac{d^{2}x}{ds^{2}}\right)} + \frac{\partial E}{\partial \left(\frac{d^{2}y}{ds^{2}}\right)}\right) + \frac{d}{ds}E_{v_{s}} - E_{v} = 0$$

Numerical Solutions

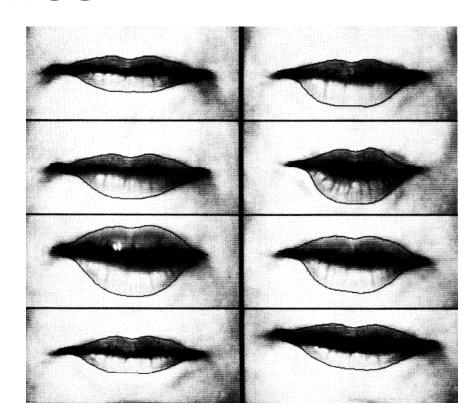
 The spline v(s) which minimizes E*_{snake} must satisfy

$$-\frac{d^{2}}{ds^{2}}\left(\frac{\partial E}{\partial \left(\frac{d^{2}x}{ds^{2}}\right)} + \frac{\partial E}{\partial \left(\frac{d^{2}y}{ds^{2}}\right)}\right) + \frac{d}{ds}E_{v_{s}} - E_{v} = 0$$

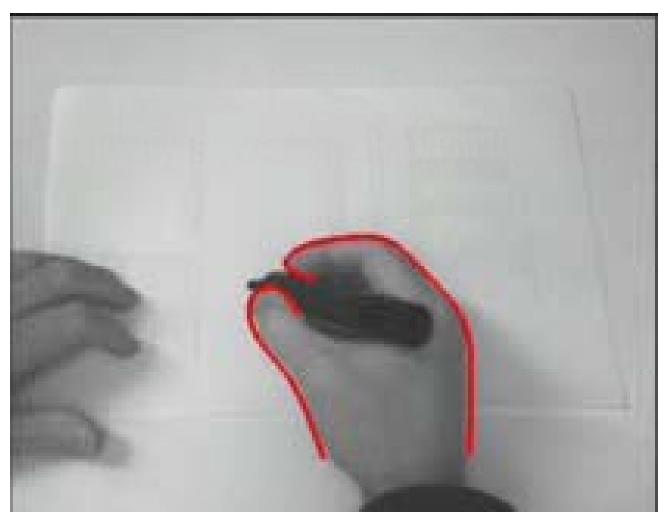
 Solutions: Greedy local updates, Euler Lagrange etc. (see handouts)

Using Snakes for Dynamic Scenes

- Once a snake finds a feature, it "locks on"
- If feature begins to move, the snake will track the same local minimum
- Fast motion could cause the snake to flip into a different minimum



Example Movie Sequences



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