

Boundaries often hard to find

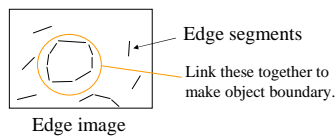


Active Contour Models: Snakes

- Important class of algorithms for finding object boundaries
- Usually assume little about shape other than that it is smooth
- Finding boundaries becomes an optimisation process
- Balance between matching to image and ensuring result is smooth

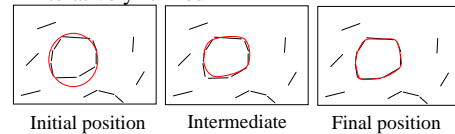
Introduction

- Suppose we wish to find the boundary of an object in an image
- One approach is to find edge segments and link them together



Introduction

- Alternatively use an Active Contour Model (or 'snake')
- A smooth curve which matches to image data
- Curve is initialised near target
- Iteratively refined



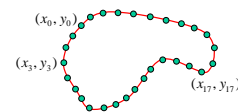
Properties of snakes

- Optimise a combination of *internal* and *external* energy
- Internal energy constrains shape of curve
 - eg to encourage smoothness
- External energy encourages matching to suitable image features
 - Eg strong edges

Simple curve representation

- Represent the curve with a set of n points

$$\mathbf{v}_i = (x_i, y_i) \quad i = 0 \dots n-1$$



Snake Energy

- The total energy of the snake is defined as

$$E_{\text{total}} = E_{\text{internal}} + E_{\text{external}}$$

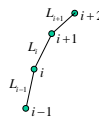
Encourages smoothness or a particular shape

Encourages curve onto image structures (eg edges)

Simple Elastic Curve

- For a curve represented as a set of points a simple elastic energy term is

$$E_{\text{internal}} = \alpha \sum_{i=0}^{n-1} L_i^2$$

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


- This encourages the closed curve to shrink to a point (like a very small elastic band)

External Energy

- The external energy describes how well the curve matches the image data locally
- Numerous forms can be used, attracting the curve toward different image features

Simple Edge Strength

- Suppose we have an image $I(x, y)$
- Generate gradient images $G_x(x, y), G_y(x, y)$
- The absolute edge strength at a point is then
- An external energy term for a snake is thus

$$|G_x(x, y)|^2 + |G_y(x, y)|^2$$

$$E_{\text{external}} = -\sum_{i=0}^{n-1} |G_x(x_i, y_i)|^2 + |G_y(x_i, y_i)|^2$$

(Negative in order that minimising it forces the curve toward large edges)

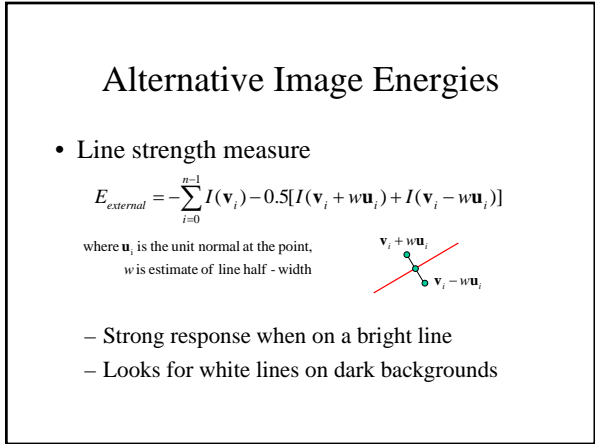
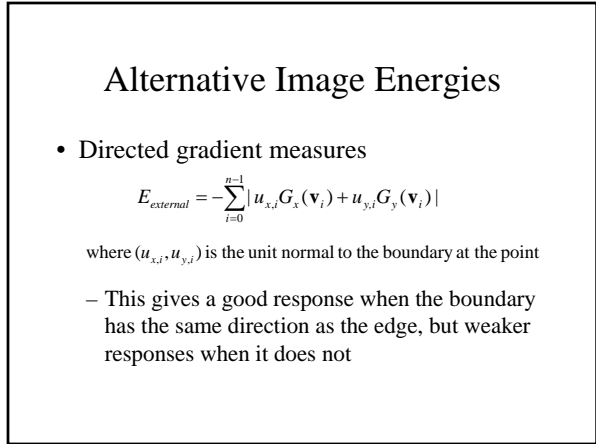
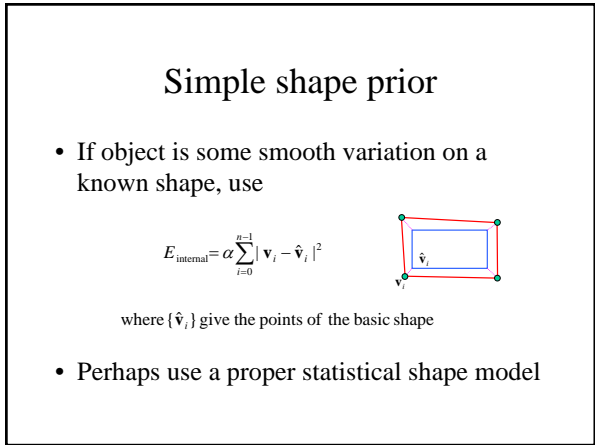
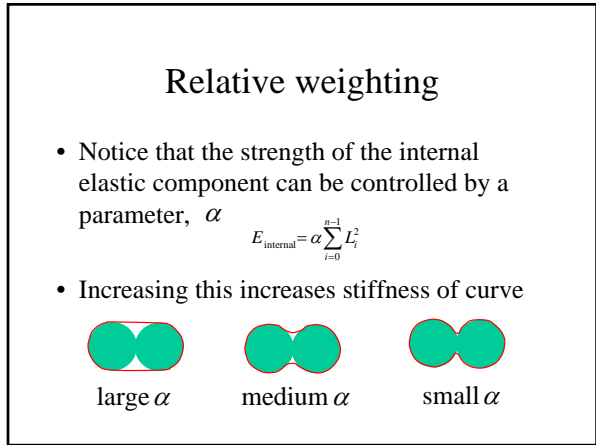
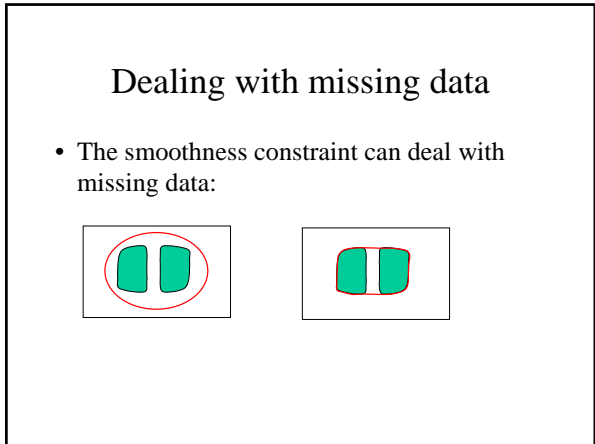
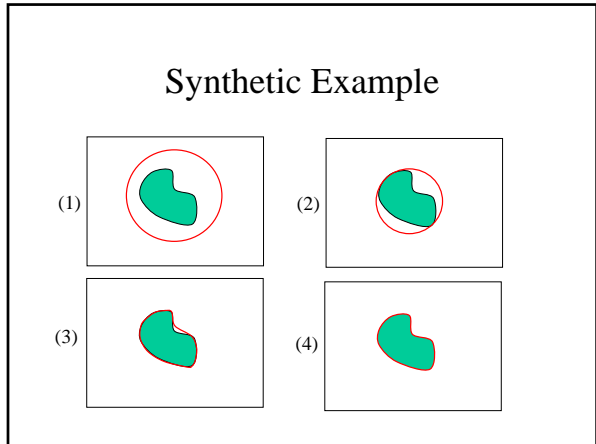
Simple Elastic Snake

- A simple elastic snake is thus defined by
 - a set of n points,
 - An internal elastic energy term
 - An external edge based energy term
- To use this to locate the outline of an object
 - Initialise in the vicinity of the object
 - Modify the points to minimise the total energy

Simple Elastic Snake

- In this case we have

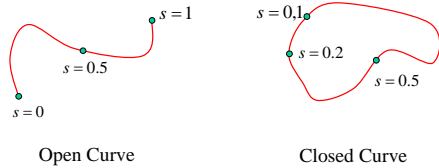
$$E_{\text{total}}(\mathbf{x}) = E_{\text{internal}}(\mathbf{x}) + E_{\text{external}}(\mathbf{x})$$
 where $\mathbf{x} = (x_0, \dots, x_{n-1}, y_0, \dots, y_{n-1})^T$
 - which is a $2n$ -D optimisation problem
 - Use a general optimisation algorithm
 - Note that we can easily compute the derivatives, which allows efficient optimisation



Continuous Curve Representation

- We can represent the curve parametrically

$$\mathbf{v}(s) = (x(s), y(s)) \quad 0 \leq s \leq 1$$



Internal Energy

- The bending energy of a continuous curve is given by

$$E_{\text{int}}(\mathbf{v}(s)) = \alpha(s) \left| \frac{d\mathbf{v}}{ds} \right|^2 + \beta(s) \left| \frac{d^2\mathbf{v}}{ds^2} \right|^2$$

Elasticity
Stiffness

(The more the curve bends, the larger this value)

Discrete representation

- If the curve is represented by n points

$$\mathbf{v}_i = (x_i, y_i) \quad i = 0 \dots n-1$$

$$\frac{d\mathbf{v}}{ds} \approx \frac{\mathbf{v}_{i+1} - \mathbf{v}_{i-1}}{2} \quad \frac{d^2\mathbf{v}}{ds^2} \approx (\mathbf{v}_{i+1} - \mathbf{v}_i) - (\mathbf{v}_i - \mathbf{v}_{i-1}) = \mathbf{v}_{i+1} - 2\mathbf{v}_i + \mathbf{v}_{i-1}$$

$$E_{\text{internal}} = \sum_{i=0}^{n-1} \alpha |\mathbf{v}_{i+1} - \mathbf{v}_{i-1}|^2 + \beta |\mathbf{v}_{i+1} - 2\mathbf{v}_i + \mathbf{v}_{i-1}|^2$$

Elasticity
Stiffness

Snake Energy (continuous form)

- The total energy of the snake is defined as

$$E_{\text{total}} = E_{\text{internal}} + E_{\text{external}}$$

$$E_{\text{internal}} = \int_0^1 E_{\text{int}}(\mathbf{v}(s)) ds \quad \text{Eg Bending energy}$$

$$E_{\text{external}} = \int_0^1 E_{\text{ext}}(\mathbf{v}(s)) ds \quad \text{Eg. Total edge strength under curve}$$

Open Curves



- When using an open curve we must impose constraints on the end points
- Typically fix them

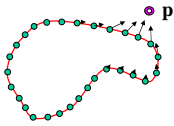
Additional Constraints

- Snakes originally developed for interactive image segmentation
- User could draw initial contour, then nudge it when it goes wrong
- Simply modify the external energy term to
 - Pull nearby points toward cursor, or
 - Push nearby points away from cursor

Interactive forces

- Pull points towards cursor:

$$E_{pull} = \sum_{i=0}^{n-1} \frac{r^2}{|v_i - \mathbf{p}|^2}$$



Nearby points get pulled hardest

Problems with snakes

- Just using smoothness doesn't always capture all useful prior knowledge
- User must define smoothness parameters
- Snake may oversmooth the boundary
- Not trivial to prevent curve self intersecting



Problems with Snakes (II)

- Can't easily deal with relationships between objects
 - Use a statistical shape model instead
- Choice of most suitable external energy sometimes more an art than a science

Suggested Reading

- “Image Processing, Analysis and Machine Vision” (Sonka, Hlavac and Boyle), pp374-380
 - Gives more mathematical detail and more examples
- For the dedicated
 - “Active Contours”, A. Blake and M. Isard (pub. Springer)

Demos

- Snake demo
 - <http://www.isbe.man.ac.uk/~bim/software>
- Matlab Snake Demo
 - By Chris Bregler and Malcolm Slaney, Interval Research Corporation.
 - Download from <http://www.slaney.org/malcolm/pubs.html>