

# Active Contours (Snakes)

Jungah Son

## **Snakes: Active Contour Models**

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### **Abstract**

A snake is an energy-minimizing spline guided by external constraint forces and influenced by image forces that pull it toward features such as lines and edges. Snakes are active contour models: they lock onto nearby edges, localizing them accurately. Scale-space continuation can be used to enlarge the capture region surrounding a feature. Snakes provide a unified account of a number of visual problems, including detection of edges, lines, and subjective contours; motion tracking; and stereo matching. We have used snakes successfully for interactive interpretation, in which user-imposed constraint forces guide the snake near features of interest.

# Initialization

1. Use **ginput** function to pick initial contour points
2. Use **spline** function to interpolate with a spline curve



# Energy formulation

## 1. Internal energy

$$E_{\text{int}} = (\alpha(s)|\mathbf{v}_s(s)|^2 + \beta(s)|\mathbf{v}_{ss}(s)|^2)/2$$

## 2. Image energy

$$E_{\text{image}} = w_{\text{line}}E_{\text{line}} + w_{\text{edge}}E_{\text{edge}}$$

## 3. Constraint energy

# Energy formulation

## 1. Internal energy

$$E_{\text{int}} = (\alpha(s)|\mathbf{v}_s(s)|^2 + \beta(s)|\mathbf{v}_{ss}(s)|^2)/2$$

## 2. Image energy

$$E_{\text{image}} = w_{\text{line}}E_{\text{line}} + w_{\text{edge}}E_{\text{edge}}$$

## 3. Constraint energy

$$\left. \begin{aligned} & E_{\text{ext}} \\ & = E_{\text{image}} + E_{\text{con}} \end{aligned} \right\}$$

# Problem formulation

Minimize

$$\begin{aligned} E_{\text{snake}}^* &= \int_0^1 E_{\text{snake}}(\mathbf{v}(s)) \, ds \\ &= \int_0^1 E_{\text{int}}(\mathbf{v}(s)) + E_{\text{ext}}(\mathbf{v}(s)) \, ds \end{aligned}$$

# Discrete Representation

Minimize

$$\begin{aligned} E_{\text{snake}}^* &= \int_0^1 E_{\text{snake}}(\mathbf{v}(s)) \, ds \\ &= \int_0^1 E_{\text{int}}(\mathbf{v}(s)) + E_{\text{ext}}(\mathbf{v}(s)) \, ds \end{aligned}$$

$$E_{\text{snake}}^* = \sum_{i=1}^n E_{\text{int}}(i) + E_{\text{ext}}(i)$$

# Discrete Representation

Minimize

$$E_{\text{snake}}^* = \sum_{i=1}^n E_{\text{int}}(i) + E_{\text{ext}}(i)$$

$$E_{\text{int}} = (\alpha(s)|\mathbf{v}_s(s)|^2 + \beta(s)|\mathbf{v}_{ss}(s)|^2)/2$$

finite differences



$$\begin{aligned} E_{\text{int}}(i) &= \alpha |\mathbf{v}_i - \mathbf{v}_{i-1}|^2/2 \\ &\quad + \beta |\mathbf{v}_{i-1} - 2\mathbf{v}_i + \mathbf{v}_{i+1}|^2/2 \end{aligned}$$

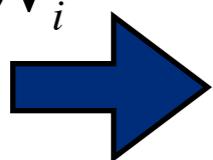
# Discrete Representation

Minimize

$$E_{\text{snake}}^* = \sum_{i=1}^n E_{\text{int}}(i) + E_{\text{ext}}(i)$$

$$= \sum_{i=1}^n \alpha |\mathbf{v}_i - \mathbf{v}_{i-1}|^2/2 + \beta |\mathbf{v}_{i-1} - 2\mathbf{v}_i + \mathbf{v}_{i+1}|^2/2 + E_{\text{ext}}(i)$$

$$\frac{\partial E_{\text{snake}}^*}{\partial \mathbf{v}_i}$$



$$\alpha (\mathbf{v}_i - \mathbf{v}_{i-1}) - \alpha (\mathbf{v}_{i+1} - \mathbf{v}_i)$$

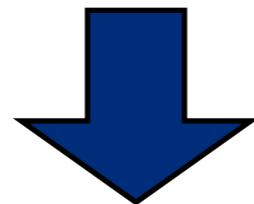
$$+ \beta [\mathbf{v}_{i-2} - 2\mathbf{v}_{i-1} + \mathbf{v}_i] - 2\beta [\mathbf{v}_{i-1} - 2\mathbf{v}_i + \mathbf{v}_{i+1}]$$

$$+ \beta [\mathbf{v}_i - 2\mathbf{v}_{i+1} + \mathbf{v}_{i+2}] + (f_x(i), f_y(i)) = 0$$

$$\underbrace{(f_x(i), f_y(i))}_{\left( \frac{\partial E_{\text{ext}}}{\partial x_i}, \frac{\partial E_{\text{ext}}}{\partial y_i} \right)}$$

# Rearrange terms

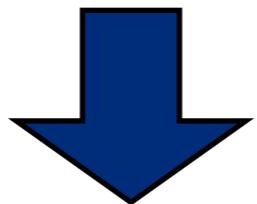
$$\begin{aligned}\alpha (\mathbf{v}_i - \mathbf{v}_{i-1}) - \alpha (\mathbf{v}_{i+1} - \mathbf{v}_i) \\ + \beta [\mathbf{v}_{i-2} - 2\mathbf{v}_{i-1} + \mathbf{v}_i] - 2\beta [\mathbf{v}_{i-1} - 2\mathbf{v}_i + \mathbf{v}_{i+1}] \\ + \beta [\mathbf{v}_i - 2\mathbf{v}_{i+1} + \mathbf{v}_{i+2}] + (f_x(i), f_y(i)) = 0\end{aligned}$$



$$\begin{aligned}\beta \mathbf{v}_{i-2} + (-\alpha - 4\beta) \mathbf{v}_{i-1} \\ + (2\alpha + 6\beta) \mathbf{v}_i \\ + (-\alpha - 4\beta) \mathbf{v}_{i+1} + \beta \mathbf{v}_{i+2} + (f_x(i), f_y(i)) = 0\end{aligned}$$

# Write in matrix form

$$\begin{aligned}\beta \mathbf{v}_{i-2} + (-\alpha - 4\beta) \mathbf{v}_{i-1} \\ + (2\alpha + 6\beta) \mathbf{v}_i \\ + (-\alpha - 4\beta) \mathbf{v}_{i+1} + \beta \mathbf{v}_{i+2} + (f_x(i), f_y(i)) = 0\end{aligned}$$



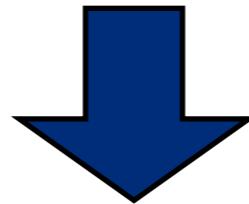
$$\mathbf{Ax} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) = 0$$

$$\mathbf{Ay} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) = 0$$

# Matrix A

$$\begin{aligned}\beta \mathbf{v}_{i-2} + (-\alpha - 4\beta) \mathbf{v}_{i-1} \\ + (2\alpha + 6\beta) \mathbf{v}_i \\ + (-\alpha - 4\beta) \mathbf{v}_{i+1} + \beta \mathbf{v}_{i+2} + (f_x(i), f_y(i)) = 0\end{aligned}$$

$$\boxed{\begin{aligned}\mathbf{Ax} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) &= 0 \\ \mathbf{Ay} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) &= 0\end{aligned}}$$



$$\left[ \begin{array}{ccccccccc} 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & 0 & 0 & \cdots & \beta & -\alpha - 4\beta \\ -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & 0 & \cdots & 0 & \beta \\ \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & \cdots & 0 & 0 \\ 0 & \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & \cdots & 0 & 0 \\ 0 & 0 & \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \beta & 0 & 0 & 0 & 0 & 0 & \cdots & 2\alpha + 6\beta & -\alpha - 4\beta \\ -\alpha - 4\beta & \beta & 0 & 0 & 0 & 0 & \cdots & -\alpha - 4\beta & 2\alpha + 6\beta \end{array} \right]$$

# Matrix A

$$\beta \mathbf{v}_{-1} + (-\alpha - 4\beta) \mathbf{v}_0$$

$$+ (2\alpha + 6\beta) \mathbf{v}_1$$

$$+ \underline{(-\alpha - 4\beta) \mathbf{v}_2} + \beta \mathbf{v}_3 + (f_x(i), f_y(i)) = 0$$

$$\mathbf{A}\mathbf{x} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) = 0$$

$$\mathbf{A}\mathbf{y} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) = 0$$

$i = 1 :$

$$\begin{bmatrix} 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & 0 & 0 & \cdots & \beta & -\alpha - 4\beta \\ -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & 0 & \cdots & 0 & \beta \\ \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & \cdots & 0 & 0 \\ 0 & \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & \cdots & 0 & 0 \\ 0 & 0 & \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \beta & 0 & 0 & 0 & 0 & 0 & \cdots & 2\alpha + 6\beta & -\alpha - 4\beta \\ -\alpha - 4\beta & \beta & 0 & 0 & 0 & 0 & \cdots & -\alpha - 4\beta & 2\alpha + 6\beta \end{bmatrix}$$

# Matrix A

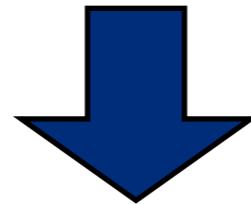
$$\frac{\beta \mathbf{v}_{n-1} + (-\alpha - 4\beta) \mathbf{v}_n}{+ (2\alpha + 6\beta) \mathbf{v}_1}$$

$$+ (-\alpha - 4\beta) \mathbf{v}_2 + \beta \mathbf{v}_3 + (f_x(i), f_y(i)) = 0$$

$$\mathbf{A}\mathbf{x} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) = 0$$

$$\mathbf{A}\mathbf{y} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) = 0$$

1. Snake is circular!



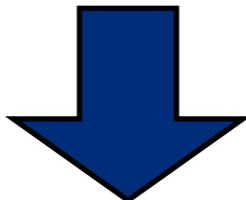
$$\left[ \begin{array}{ccccccccc} 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & 0 & 0 & \cdots & \beta & -\alpha - 4\beta \\ -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & 0 & \cdots & 0 & \beta \\ \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & \cdots & 0 & 0 \\ 0 & \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & \cdots & 0 & 0 \\ 0 & 0 & \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \beta & 0 & 0 & 0 & 0 & 0 & \cdots & 2\alpha + 6\beta & -\alpha - 4\beta \\ -\alpha - 4\beta & \beta & 0 & 0 & 0 & 0 & \cdots & -\alpha - 4\beta & 2\alpha + 6\beta \end{array} \right]$$

# Matrix A

$$\begin{aligned}\beta \mathbf{v}_{i-2} + (-\alpha - 4\beta) \mathbf{v}_{i-1} \\ + (2\alpha + 6\beta) \mathbf{v}_i \\ + (-\alpha - 4\beta) \mathbf{v}_{i+1} + \beta \mathbf{v}_{i+2} + (f_x(i), f_y(i)) = 0\end{aligned}$$

$$\boxed{\begin{aligned}\mathbf{Ax} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) &= 0 \\ \mathbf{Ay} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) &= 0\end{aligned}}$$

2. Use [circshift](#) function!



$$\left[ \begin{array}{ccccccccc} 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & 0 & 0 & \cdots & \beta & -\alpha - 4\beta \\ -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & 0 & \cdots & 0 & \beta \\ \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & 0 & \cdots & 0 & 0 \\ 0 & \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \beta & \cdots & 0 & 0 \\ 0 & 0 & \beta & -\alpha - 4\beta & 2\alpha + 6\beta & -\alpha - 4\beta & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \beta & 0 & 0 & 0 & 0 & 0 & \cdots & 2\alpha + 6\beta & -\alpha - 4\beta \\ -\alpha - 4\beta & \beta & 0 & 0 & 0 & 0 & \cdots & -\alpha - 4\beta & 2\alpha + 6\beta \end{array} \right]$$

# Solve equations

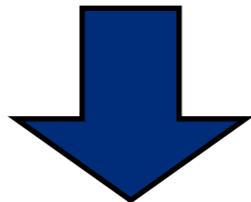
$$\mathbf{A}\mathbf{x} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) = 0$$

$$\mathbf{A}\mathbf{y} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) = 0$$

# Solve equations

$$\mathbf{A}\mathbf{x} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) = 0$$

$$\mathbf{A}\mathbf{y} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) = 0$$



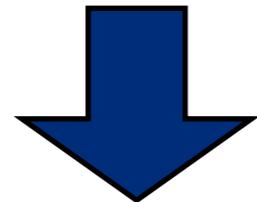
$$\mathbf{A}\mathbf{x}_t + \mathbf{f}_x(\mathbf{x}_{t-1}, \mathbf{y}_{t-1}) = -\gamma(\mathbf{x}_t - \mathbf{x}_{t-1})$$

$$\mathbf{A}\mathbf{y}_t + \mathbf{f}_y(\mathbf{x}_{t-1}, \mathbf{y}_{t-1}) = -\gamma(\mathbf{y}_t - \mathbf{y}_{t-1})$$

# Solve equations

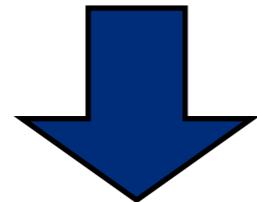
$$\mathbf{A}\mathbf{x} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) = 0$$

$$\mathbf{A}\mathbf{y} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) = 0$$



$$\mathbf{A}\mathbf{x}_t + \mathbf{f}_x(\mathbf{x}_{t-1}, \mathbf{y}_{t-1}) = -\gamma(\mathbf{x}_t - \mathbf{x}_{t-1})$$

$$\mathbf{A}\mathbf{y}_t + \mathbf{f}_y(\mathbf{x}_{t-1}, \mathbf{y}_{t-1}) = -\gamma(\mathbf{y}_t - \mathbf{y}_{t-1})$$



Use LU  
decomposition  
if needed!

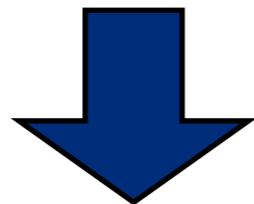
$$\mathbf{x}_t = \underline{\mathbf{(A + \gamma I)^{-1}}(\gamma \mathbf{x}_{t-1} - \mathbf{f}_x(x_{t-1}, y_{t-1}))}$$

$$\mathbf{y}_t = \mathbf{(A + \gamma I)^{-1}}(\gamma \mathbf{y}_{t-1} - \mathbf{f}_y(x_{t-1}, y_{t-1}))$$

# Solve equations

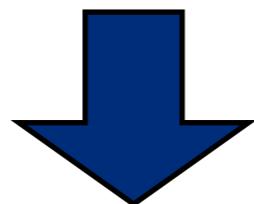
$$\mathbf{Ax} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) = 0$$

$$\mathbf{Ay} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) = 0$$



$$\mathbf{Ax}_t + \mathbf{f}_x(\mathbf{x}_{t-1}, \mathbf{y}_{t-1}) = -\gamma(\mathbf{x}_t - \mathbf{x}_{t-1})$$

$$\mathbf{Ay}_t + \mathbf{f}_y(\mathbf{x}_{t-1}, \mathbf{y}_{t-1}) = -\gamma(\mathbf{y}_t - \mathbf{y}_{t-1})$$



$$\mathbf{x}_t = (\mathbf{A} + \gamma \mathbf{I})^{-1}(\gamma \mathbf{x}_{t-1} - \underline{\mathbf{f}_x(x_{t-1}, y_{t-1})})$$

$$\mathbf{y}_t = (\mathbf{A} + \gamma \mathbf{I})^{-1}(\gamma \mathbf{y}_{t-1} - \underline{\mathbf{f}_y(x_{t-1}, y_{t-1})})$$

Use [interp2](#)  
function

# Last

1. Don't forget to smooth images to enlarge “potential field”
2. Start with a simple case first
3. Snake needs parameter adjustment

# Codes that we used in the discussion session

```
clc;
clear all;
close all;

% Original curve
xys = xy;

img = imread('lena.jpg');
figure;
imshow(img);
hold on;

xs = xys(1,:);
ys = xys(2,:);

hold on;
plot(xs, ys, 'r-');

button = 1;
n = 0;
while button == 1
    [xi,yi,button] = ginput(1);
    plot(xi,yi,'go')
    n = n+1;
    xy(:,n) = [xi;yi];
end

% Interpolate with a spline curve
t = 1:n;
ts = 1: 0.1: n;
xys = spline(t,xy,ts);

xs = xys(1,:);
ys = xys(2,:);

hold on;
plot(xs, ys, 'b-');

n = n+1;
xy(:,n) = [xy(1,1);xy(2,1)];
```

# Codes that we used in the discussion session

```
% Read multiple images from a folder  
  
srcFiles = dir('/Winter 2015/liptracking2/*.jpg');  
for i = 1 : length(srcFiles)  
    filename = strcat('/Winter 2015/liptracking2/',srcFiles(i).name);  
    I = imread(filename);  
end
```