

Lecture 19-Snake Contour

Yuyao Zhang PhD

zhangyy8@shanghaitech.edu.cn

SIST Building 2 302-F

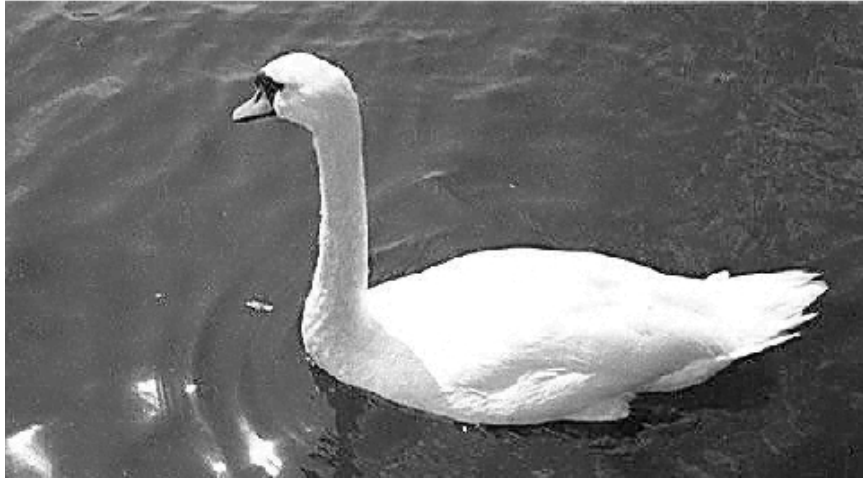
Active Contours (SNAKES)

- Back to boundary detection
 - ✓ This time using perceptual grouping.
- This is non-parametric
 - ✓ We' re not looking for a contour of a specific shape.
 - ✓ Just a good contour.

For Information on SNAKEs

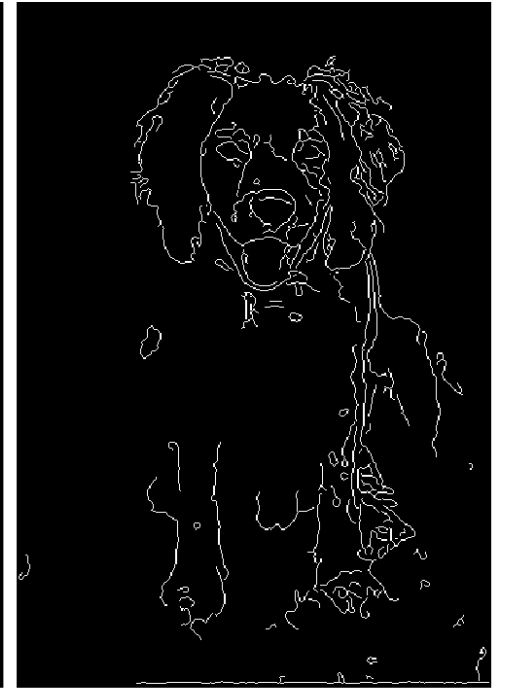
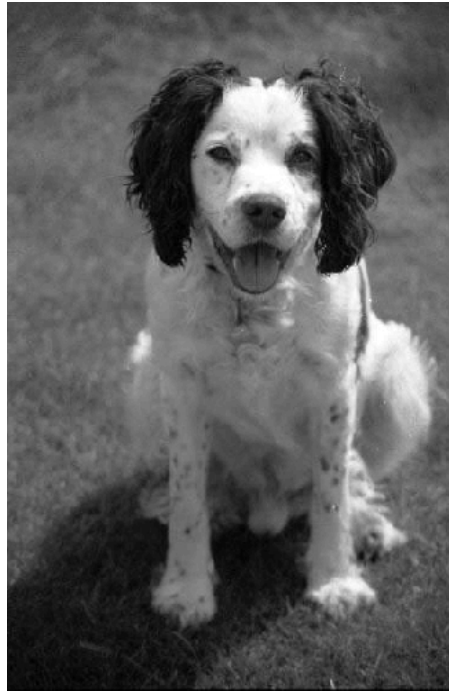
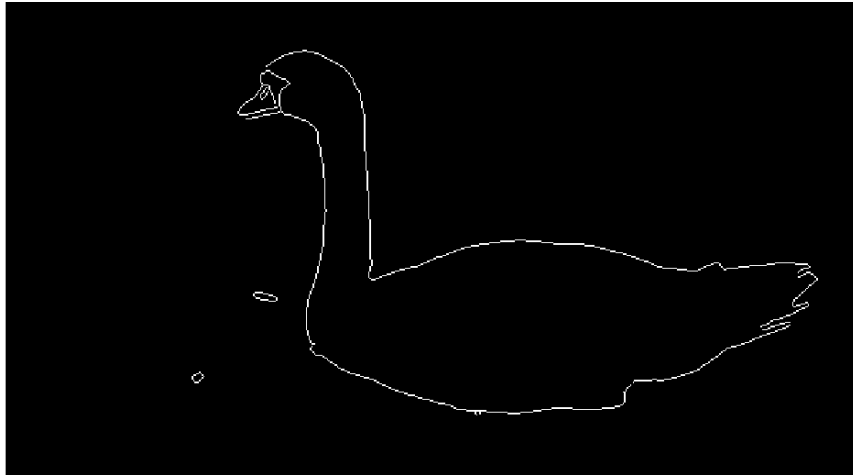
- Kass, Witkin and Terzopoulos, IJCV.
- “Dynamic Programming for Detecting, Tracking, and Matching Deformable Contours” , by Geiger, Gupta, Costa, and Viontzos, IEEE Trans. PAMI 17(3)294-302, 1995
- E. N. Mortensen and W. A. Barrett, Intelligent Scissors for Image Composition, in ACM Computer Graphics (SIGGRAPH `95), pp. 191-198, 1995

Boundary following



Sometimes edge detectors find the boundary pretty well.

Sometimes it's not good enough.



Improve Boundary Detection

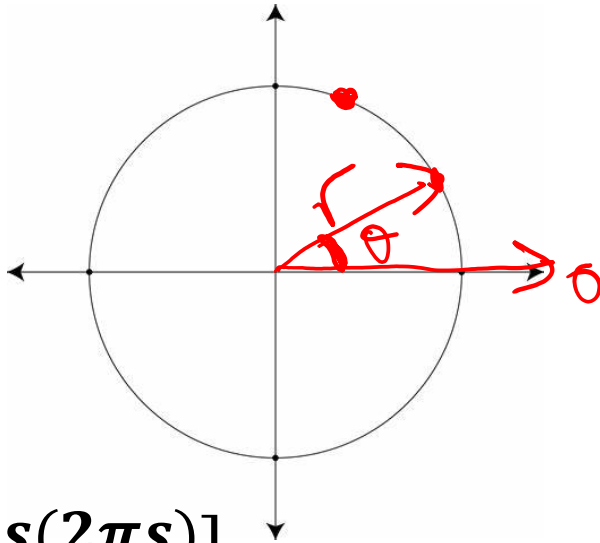
- Idea: segment using curves, not pixels.
- We want a segmentation curve that
 - 1) Conforms to image edges.
 - 2) Generates a smooth and varying curve.

Parametric Curves

➤ Consider $\begin{bmatrix} x(s) \\ y(s) \end{bmatrix}$ $s \in [0, 1]$ continuous.

$$x^2 + y^2 = r^2$$
$$\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$$

➤ E.g.



$$\begin{bmatrix} x(s) \\ y(s) \end{bmatrix} = \begin{bmatrix} r \cos(2\pi s) \\ r \sin(2\pi s) \end{bmatrix}$$

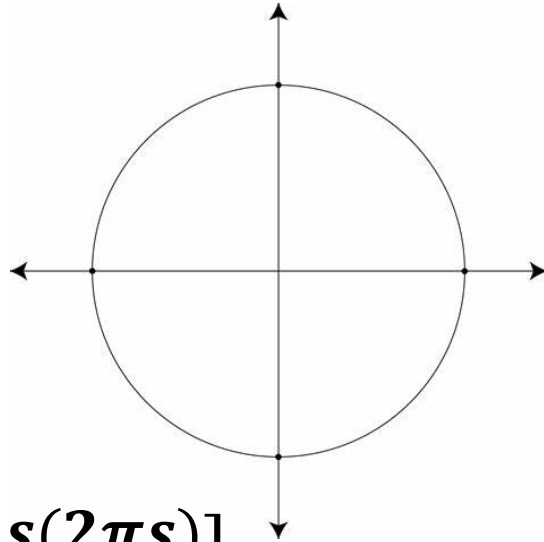


??

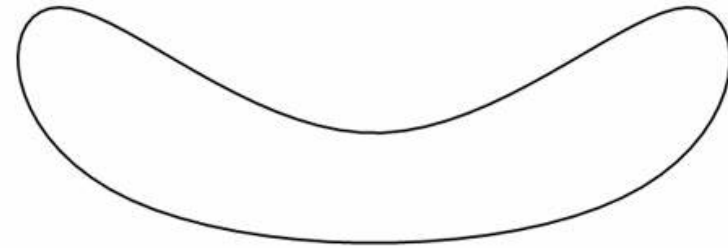
Parametric Curves

➤ Consider $\begin{bmatrix} x(s) \\ y(s) \end{bmatrix}$ $s \in [0, 1]$ continuous.

➤ E.g.



$$\begin{bmatrix} x(s) \\ y(s) \end{bmatrix} = \begin{bmatrix} r \cos(2\pi s) \\ r \sin(2\pi s) \end{bmatrix}$$

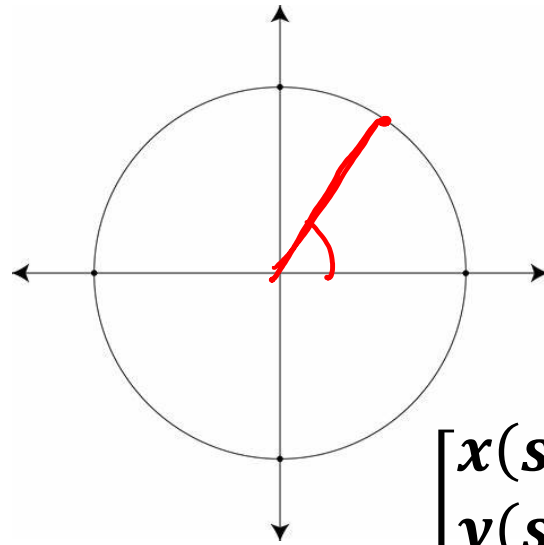


??

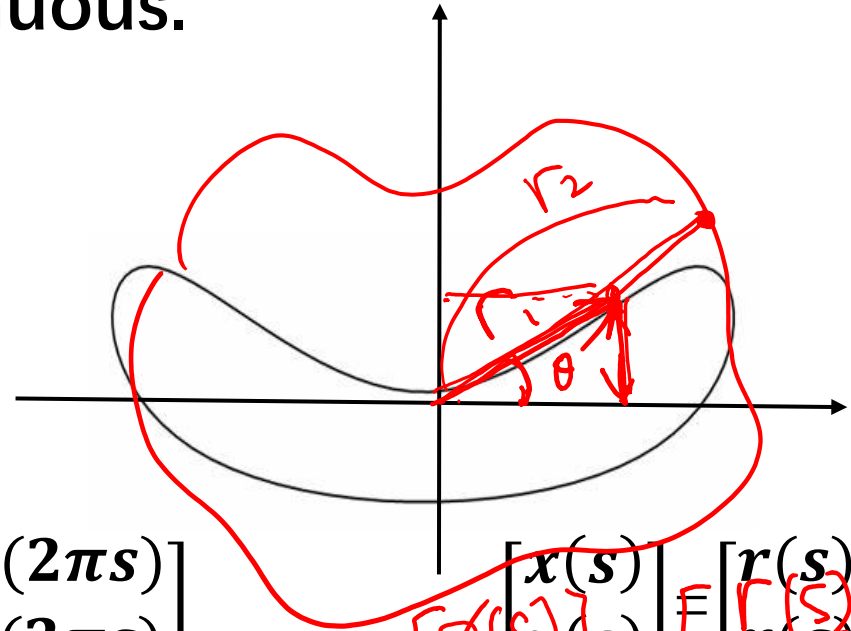
Parametric Curves

➤ Consider $\begin{bmatrix} x(s) \\ y(s) \end{bmatrix}$ $s \in [0, 1]$ continuous.

➤ E.g.



$$\begin{bmatrix} x(s) \\ y(s) \end{bmatrix} = \begin{bmatrix} r \cos(2\pi s) \\ r \sin(2\pi s) \end{bmatrix}$$

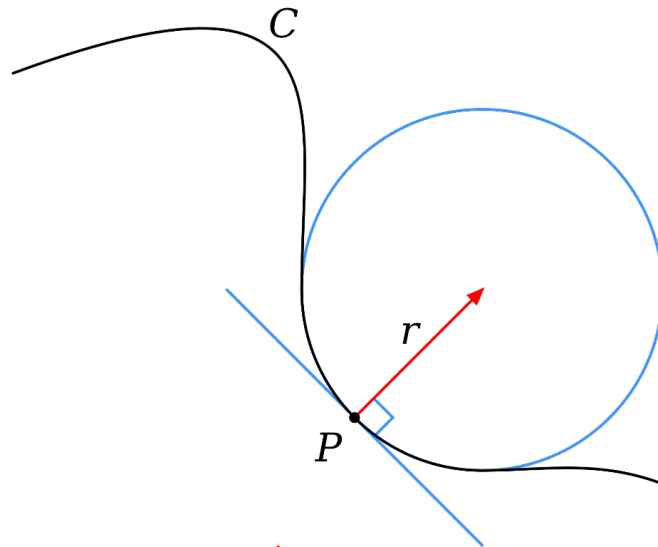


$$\begin{bmatrix} x(s) \\ y(s) \end{bmatrix} = \begin{bmatrix} r(s) \cos(2\pi s) \\ r(s) \sin(2\pi s) \end{bmatrix}$$

➤ We define a curve using $C(s) = [x(s), y(s)]$.

Curvature

$C(s)$



$$K = \lim_{r \rightarrow \infty} \frac{1}{r} = 0$$

- Let C be a plane curve (the precise technical assumptions are given below). The curvature of C at a point is a measure of how sensitive its tangent line is to moving the point to other nearby points.

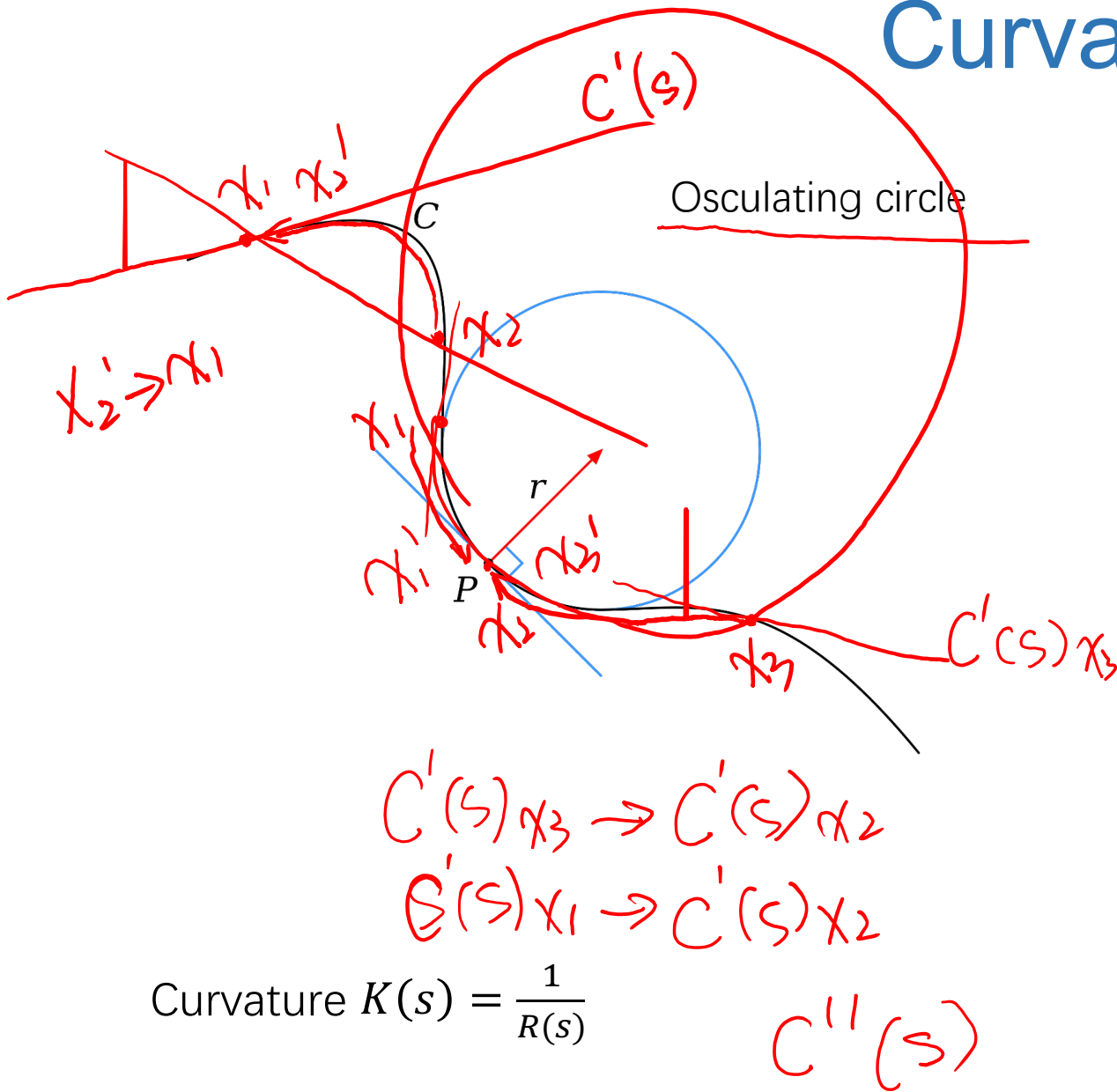
- It is natural to define the curvature of a straight line to be constantly zero. The curvature of a circle of radius r should be large if r is small and small if r is large. Thus the curvature of a circle is defined to be the reciprocal of the radius.

$$K_1 > K_2 \dots > K_6$$



$$\text{Curvature } K(s) = \frac{1}{R(s)}$$

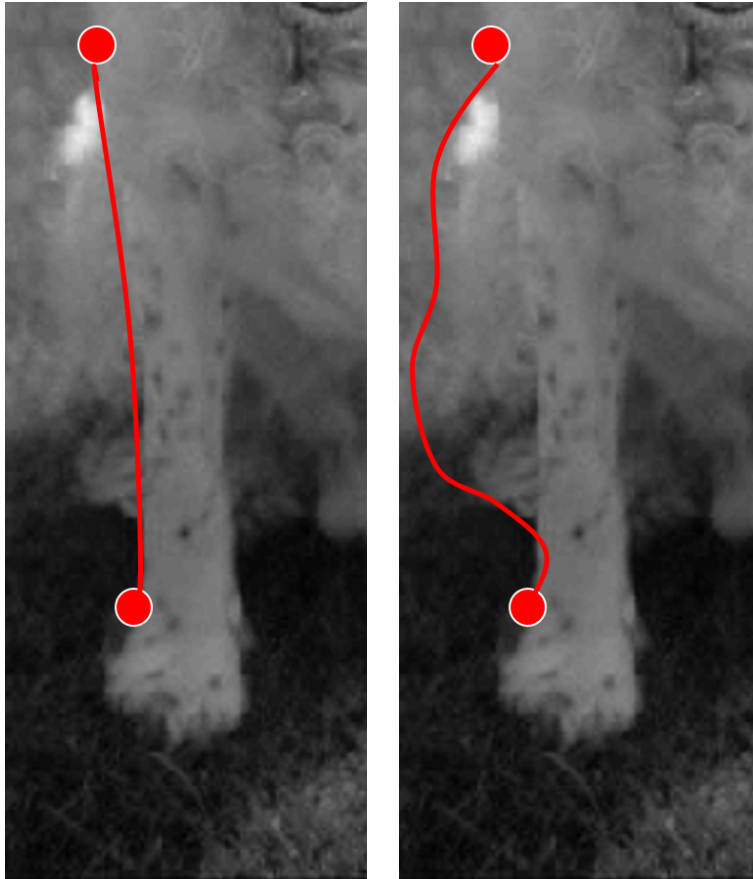
Curvature



- Let C be a plane curve (the precise technical assumptions are given below). The curvature of C at a point is a measure of how sensitive its tangent line is to moving the point to other nearby points.
- It is natural to define the curvature of a straight line to be constantly zero. The curvature of a circle of radius r should be large if r is small and small if r is large. Thus the curvature of a circle is defined to be the reciprocal of the radius.

Curvature of plane curves

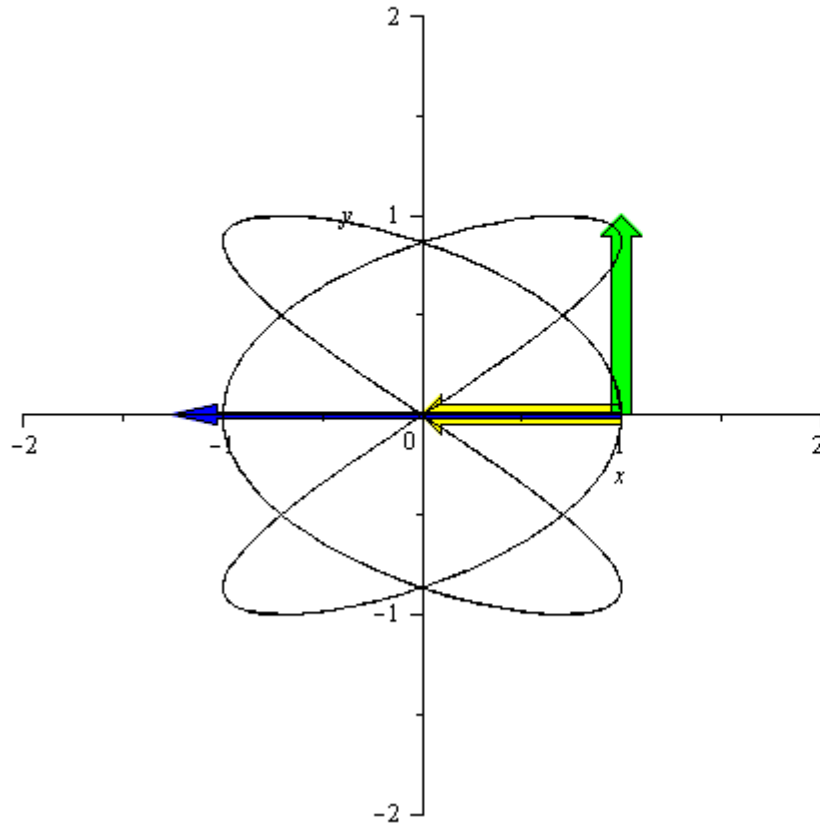
- How do we decide how good a path is? Which of two paths is better?



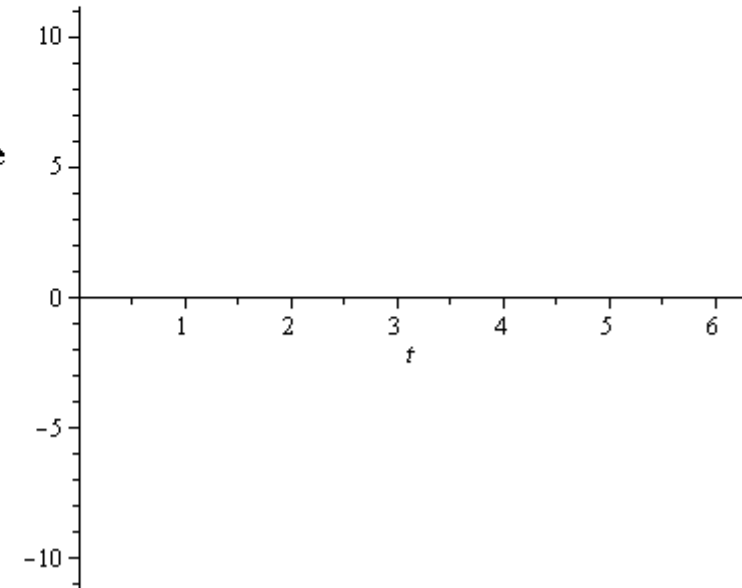
- $T(s) = C'(s)$ is considered as the velocity vector or the unit tangent vector of the curve $C(s)$.
- $\kappa(s) = \frac{1}{R(s)} = C''(s)$ is the curvature of curve $C(s)$.

Curvature of plane curves

Lissajous-Curve with tangent vector (green), normal vector (yellow), and "acceleration vector" (blue)



Curvature



Find energy for the curve

➤ Idea: we want to define an energy function $E(c)$ that matches our intuition about what makes a good segmentation.

➤ Curve will iteratively evolve to reduce/ minimize $E(c)$.

➤ $E(c) = E_{internal}(c) + E_{external}(c)$.

✓ $E_{internal}(c)$ depends only on the shape of the curve.

✓ $E_{external}(c)$ depends on image intensities.

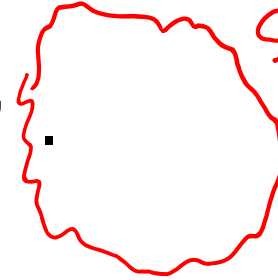
Energy for shape of the curve

➤ $E_{\text{internal}}(c) = \int_0^1 \alpha \|c'(s)\|^2 + \beta \|c''(s)\|^2 ds.$

$\begin{bmatrix} x(s) \\ y(s) \end{bmatrix} = \begin{bmatrix} r(s) \cos 2\pi s \\ r(s) \sin 2\pi s \end{bmatrix}$

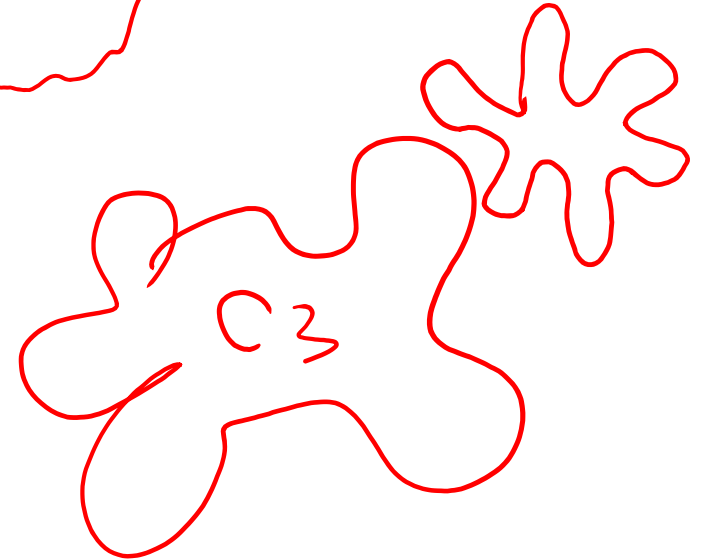
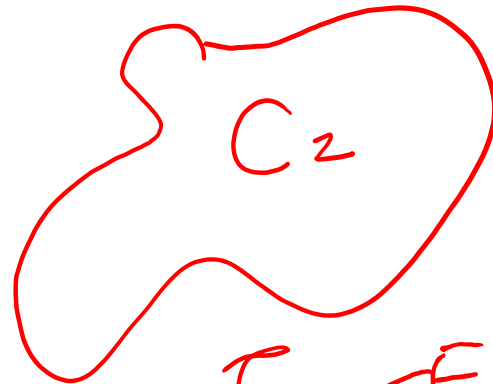
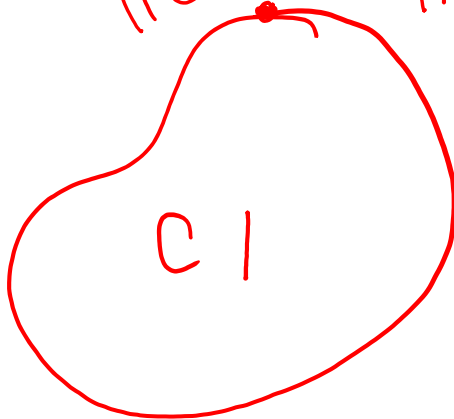
$s \in [0, 1] \quad 2\pi s \in [0, 2\pi]$

✓ Low $c'(s)$ keeps curve not too “stretchy”.



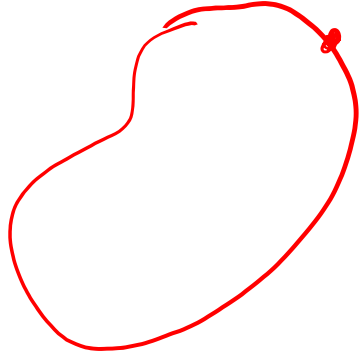
✓ Low $c''(s)$ keeps curve not too “bendy”.

$\|c'(s)\|^2 + \|c''(s)\|^2$



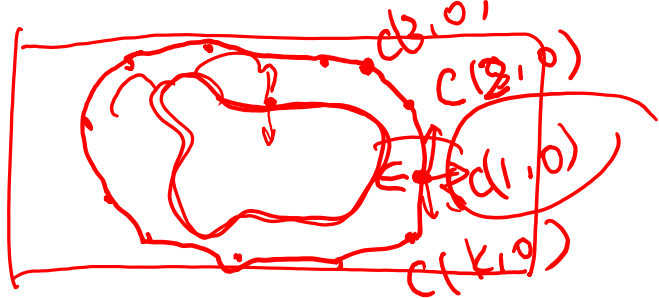
$E_1 < E_2 < E_3$

Energy for image intensities

$$\begin{aligned} \blacktriangleright E_{\text{external}}(\mathbf{c}) &= \int_0^1 - \|\nabla I(\mathbf{c}(s))\|^2 ds \\ &= \int_0^1 - \left\{ \left[\frac{\partial I}{\partial x}(\underline{x}(s), \underline{y}(s)) \right]^2 + \left[\frac{\partial I}{\partial y}(\underline{x}(s), \underline{y}(s)) \right]^2 \right\} ds \end{aligned}$$


✓ No edge, then $\nabla I = 0$, $E_{\text{external}}(\mathbf{c}) = 0$

✓ Big edge, then $\nabla I = \text{big positive value}$, $E_{\text{external}}(\mathbf{c}) = \text{big negative value}$



How to minimize $E(c)$?

➤ Requires: variational calculus. (变分微积分)

1. In practice for digital images, we solve the problem by creating a curve $C(s, t)$.
Where t represents the iteration.
2. Curve approximated by k discrete points (x_i, y_i) .
3. Then we step $C(s, t - 1)$ to $C(s, t)$ by taking a step along gradient of $E(C)$: $\frac{\partial E}{\partial C}$.
4. A snake minimize $E(C)$ must satisfy the Euler equation: $-\nabla E_{external}(c) = 0$.

➤ Result: Curve inches along until points around perimeter stop changing.

Try this

- Launch “snake.m” .
- Load image “circle” . Click the button “Set new points” , initialize starting points.
- Click the button “start” to start.

Problem with basic snake (E_{ext})

- Contour never “sees” strong edges that are far away.
- Small gradient: Snake gets hung up.
- When there is no gradient for external Energy, then only internal Energy working.
- Can not work for outer boundary.

Gradient vector flow (GVF)

- Idea: instead of using exactly the image gradient, create a new vector field over image plane:

- $\vec{V}(x, y) = \begin{bmatrix} V_x(x, y) \\ V_y(x, y) \end{bmatrix}$ vector field of the curve.

- $\vec{e}(x, y) = \begin{bmatrix} e_x(x, y) \\ e_y(x, y) \end{bmatrix}$ vector field of the edge map in an image.

- $\vec{V}(x, y)$ is defined to minimize: 在此处键入公式。

- $\iint \mu \left[\left(\frac{\partial V_x}{\partial x} \right)^2 + \left(\frac{\partial V_x}{\partial y} \right)^2 + \left(\frac{\partial V_y}{\partial x} \right)^2 + \left(\frac{\partial V_y}{\partial y} \right)^2 \right] + \|\nabla e\|^2 \|\vec{V} - \vec{e}\|^2 dx dy$

- Intuition: ∇e is big: gradient is large, \vec{V} follows edge gradient faithfully; ∇e is small: gradient is small, follows along to be as smooth as possible, trades off smooth vs how faithful.

C. Xu and J.L. Prince, "Gradient Vector Flow: A New External Force for Snakes," Proc. IEEE Conf. on Comp. Vis. Patt. Recog. (CVPR), Los Alamitos: Comp. Soc. Press, pp. 66-71, June 1997



Extensions

- Active shape models.
- Active appearance models.
- Level sets.
- FAST: FMRIB's Automated Segmentation Tool.

Discussion

- Try your own image with convolutional snake and GVF snake using the provided implementation or looking for better demon online.
- Find out what snake can do and what snake can not.