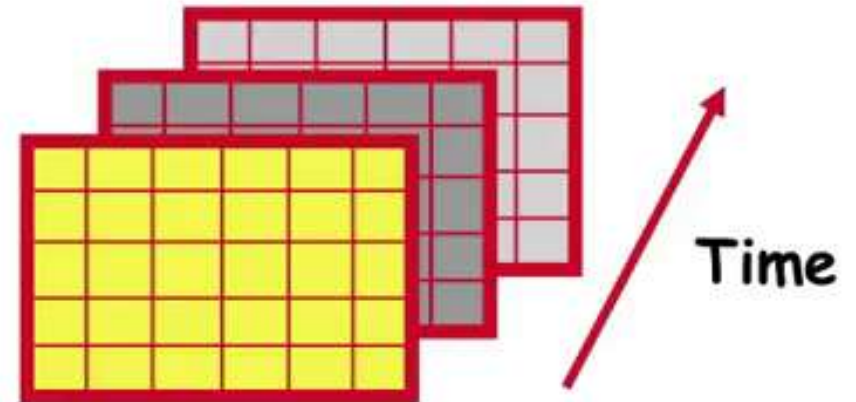


# What and why a discrete computer image?



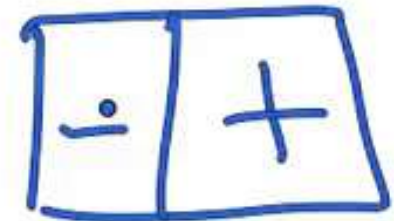
Movie courtesy "Sleepers" by W. Allen

# Discrete image representations



- Classical image processing is based on discrete mathematics (most of it)

- Sums instead of integrals



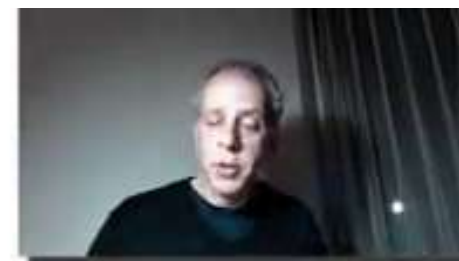
- Re-definition of classical continuous operators such as gradients, Laplacian, etc



# The PDEs approach

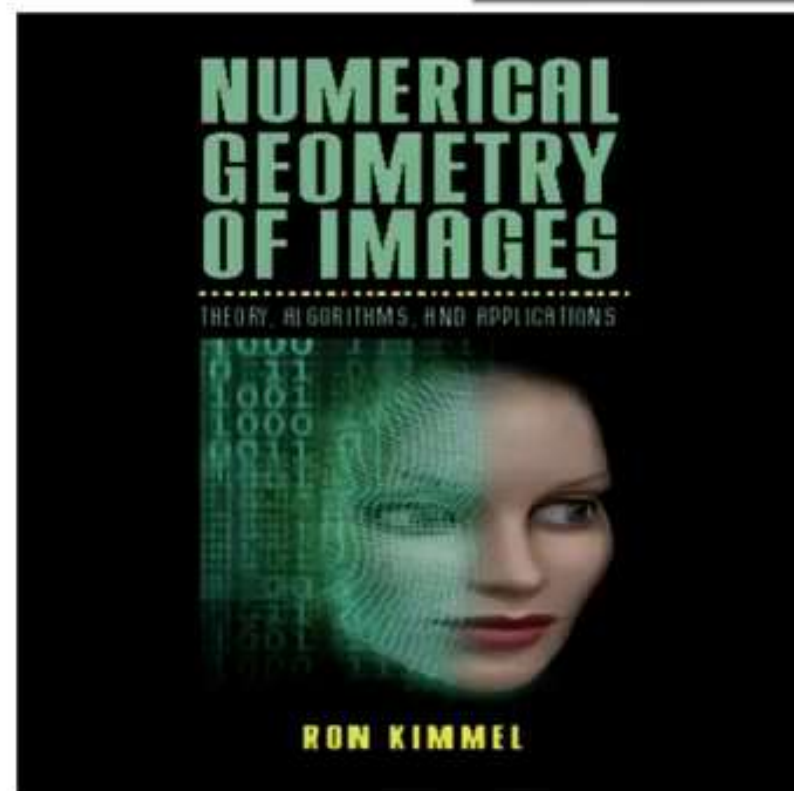
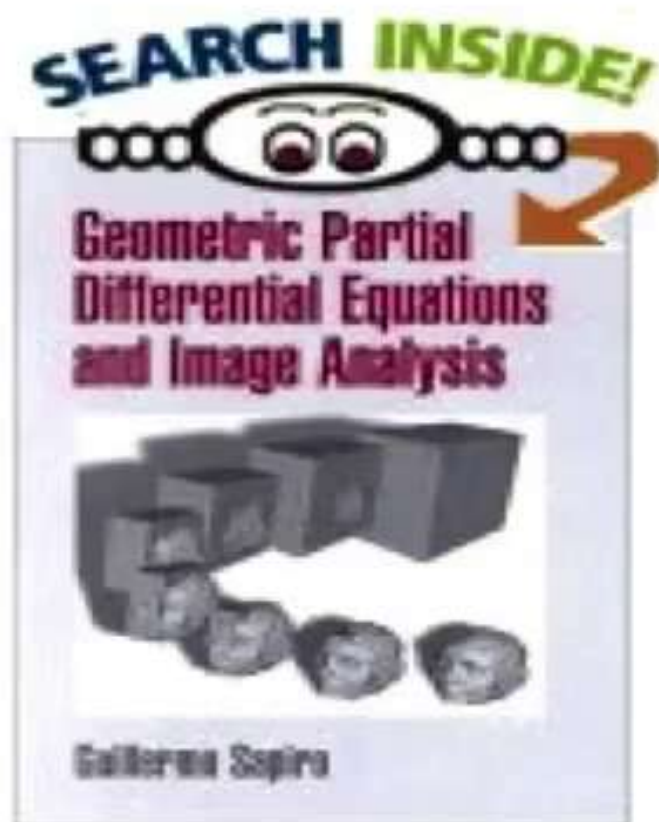
- Images are **continuous** objects
- Image processing is the results of **iteration of infinitesimal operations: PDEs**
- **Differential geometry** on images
- **Computer** image processing is based on **numerical analysis**

# Why? Why Now? Who?



- **Why now:**
  - Computers!!!
  - People
- **Why:**
  - New concepts
  - Accuracy
  - Formal analysis (existence, uniqueness, etc)
- **Consequences:**
  - Many state of the art results
  - New tools in the bookshelf

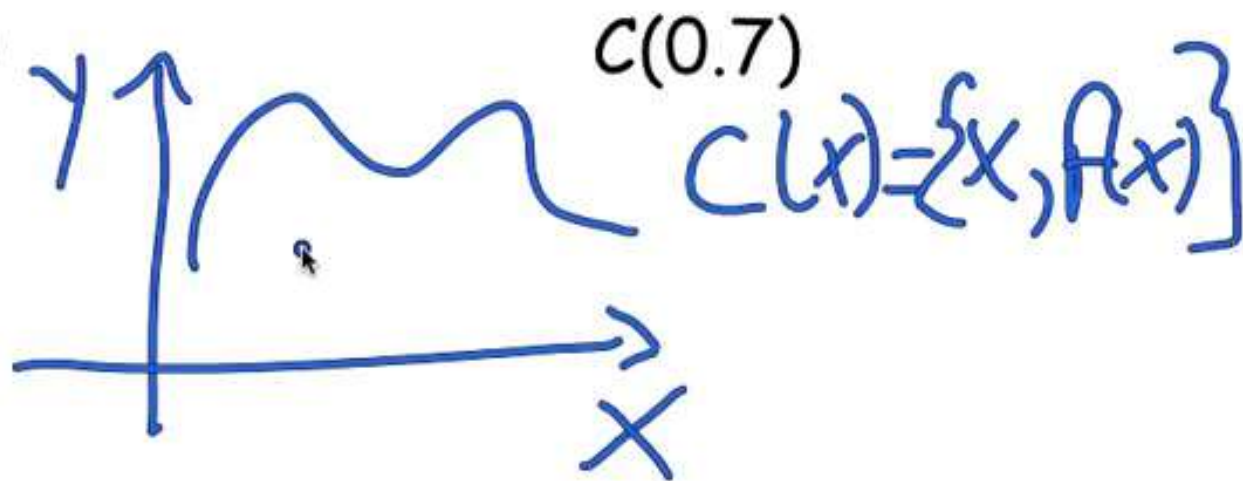






# Planar Curves

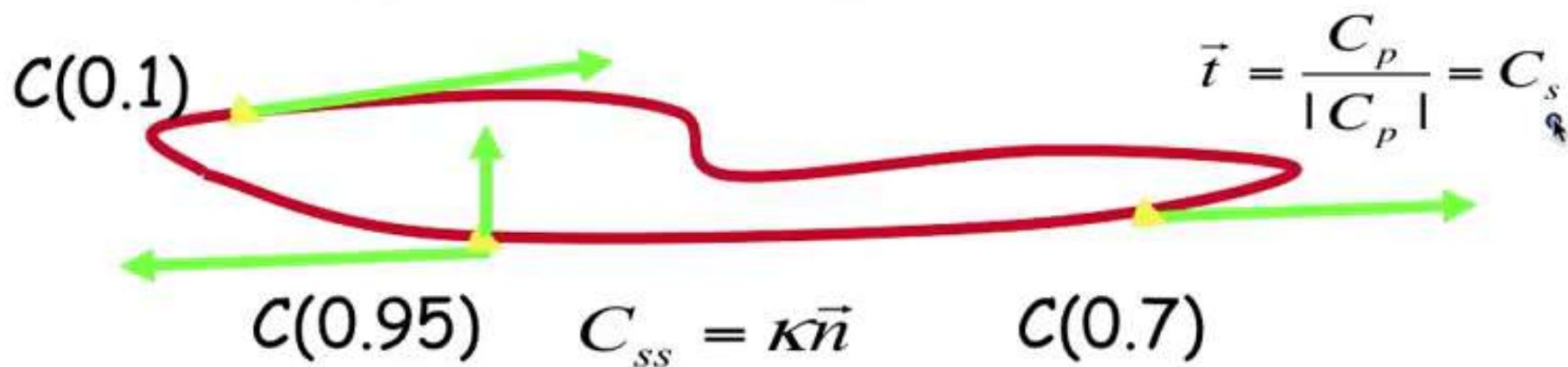
- $C(p) = \{x(p), y(p)\}, \quad p \in [0, 1] \quad C(0) = C(1)$





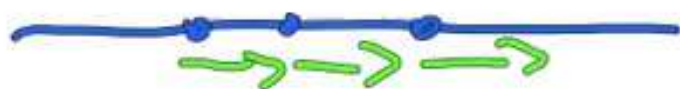
# Planar Curves

- $C(p) = \{x(p), y(p)\}$ ,  $p \in [0, 1]$



$$C_p = \frac{\partial C}{\partial p} = [x_p, y_p]$$

# Planar Curves

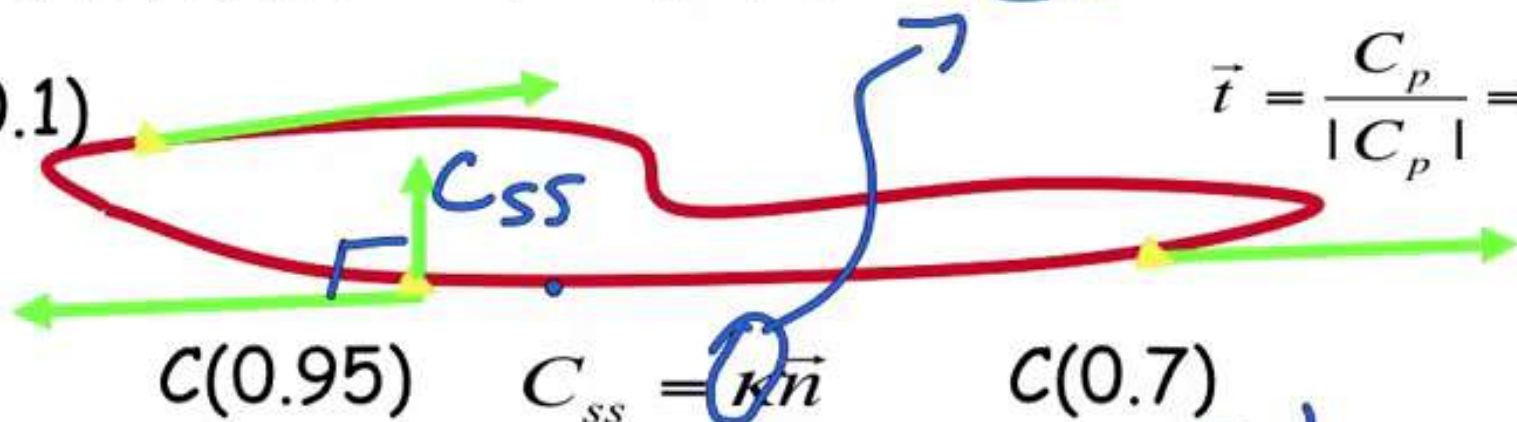


- $C(p) = \{x(p), y(p)\}, \quad p \in [0, 1]$

Curvature



$C(0.1)$



$$\vec{t} = \frac{C_p}{|C_p|} = C_s$$

$C(0.95)$

$$C_{ss} = k\vec{n}$$

$C(0.7)$

$$|C_s| = 1 \quad \frac{\partial}{\partial s} \langle C_s, C_s \rangle = \frac{\partial}{\partial s} 1$$

$$C_s \perp C_{ss} \quad \Rightarrow \quad \langle C_s, C_{ss} \rangle = 0$$

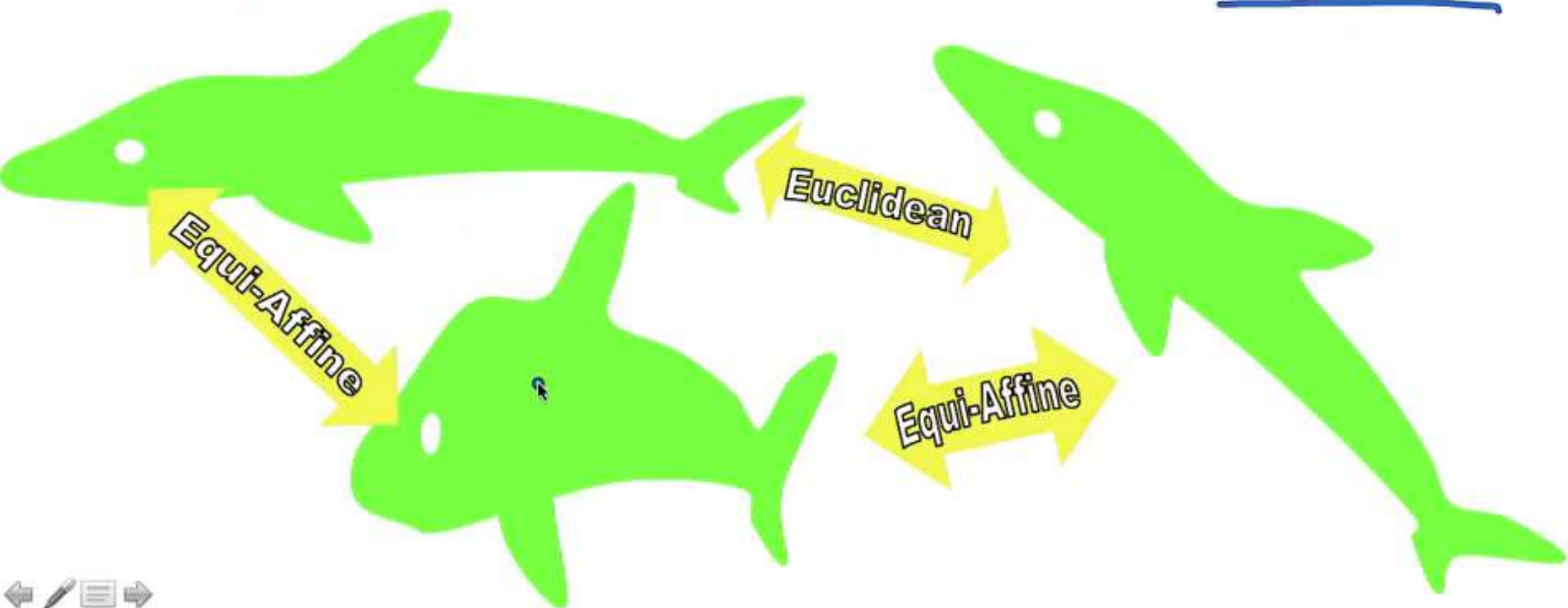






# Linear Transformations

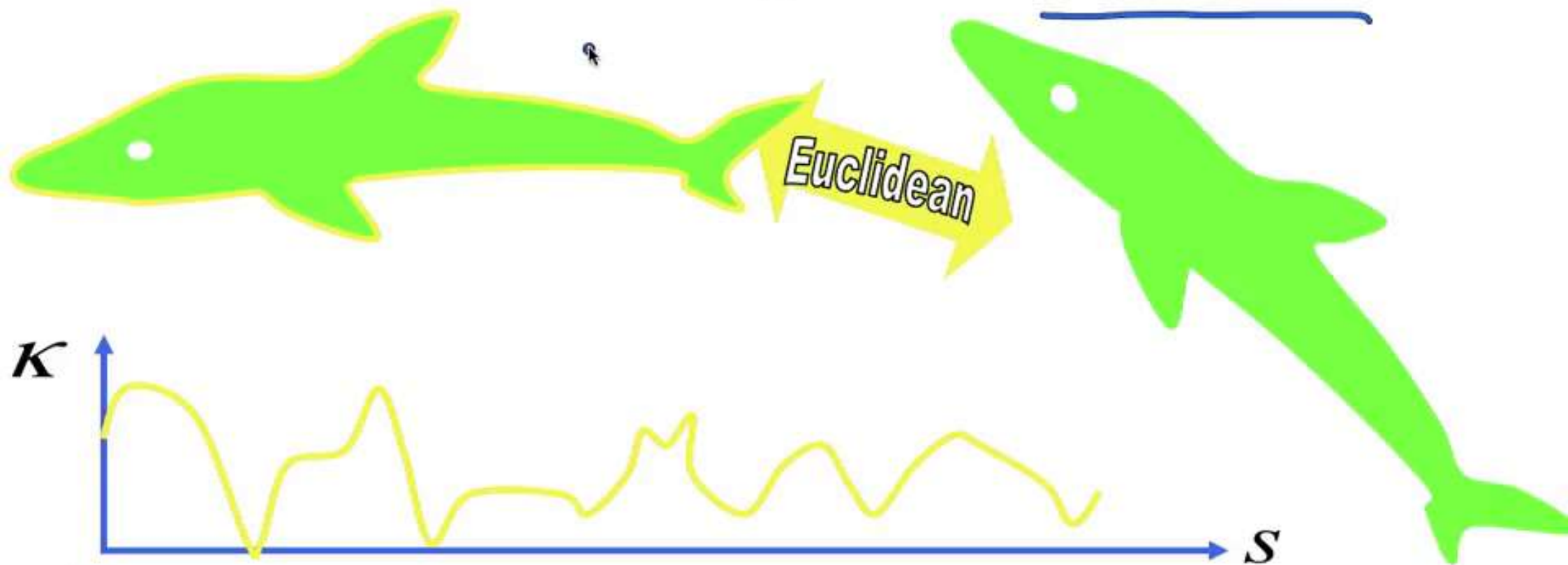
Equi-Affine:  $\{\tilde{x}, \tilde{y}\}^T = A\{x, y\}^T + \bar{b}$ ,  $\det(A) = 1$ .





# Differential Signatures

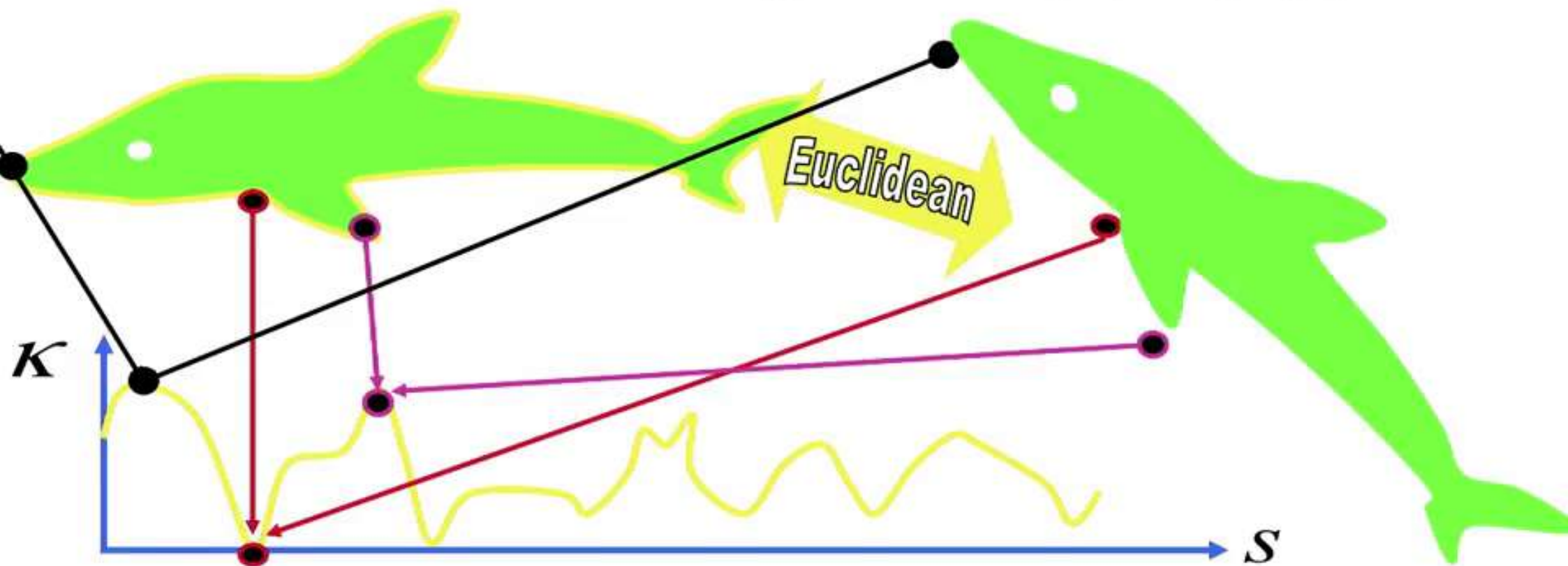
- Euclidean invariant signature  $\{s, \kappa(s)\}$





# Differential Signatures

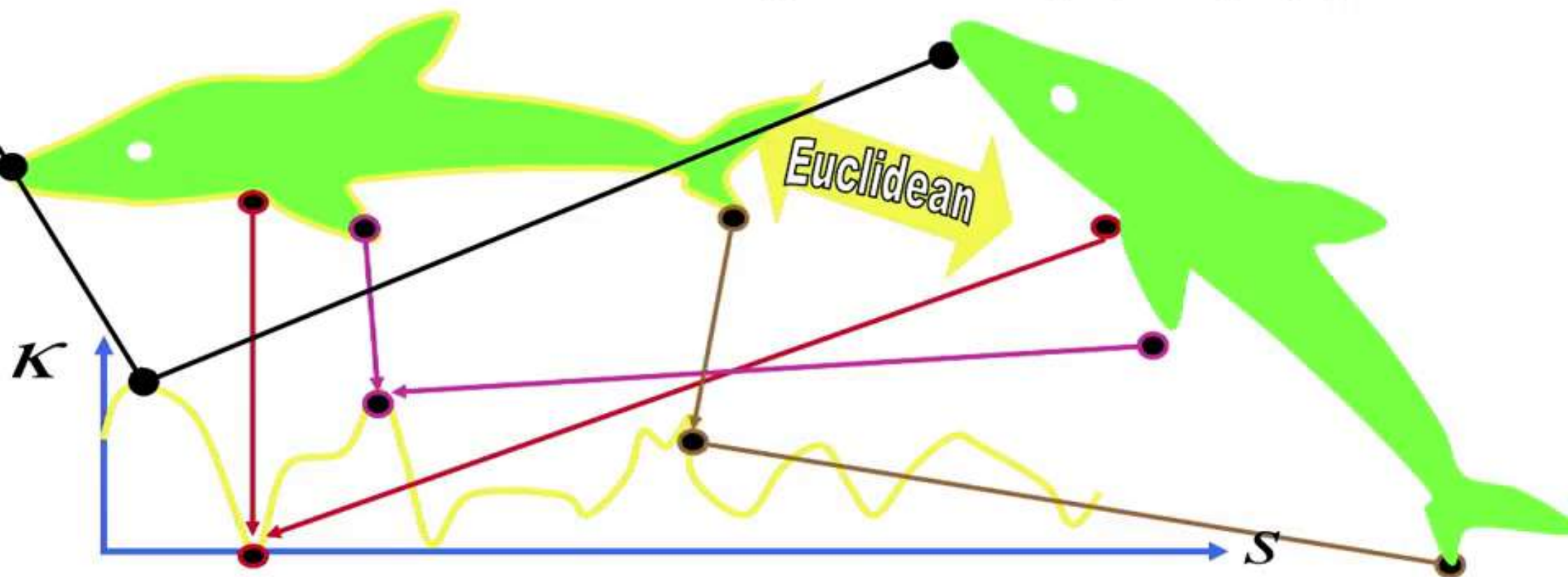
- Euclidean invariant signature  $\{s, \kappa(s)\}$





# Differential Signatures

- Euclidean invariant signature  $\{s, \kappa(s)\}$





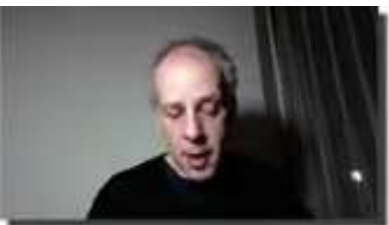


# Differential Signatures

- Euclidean invariant signature

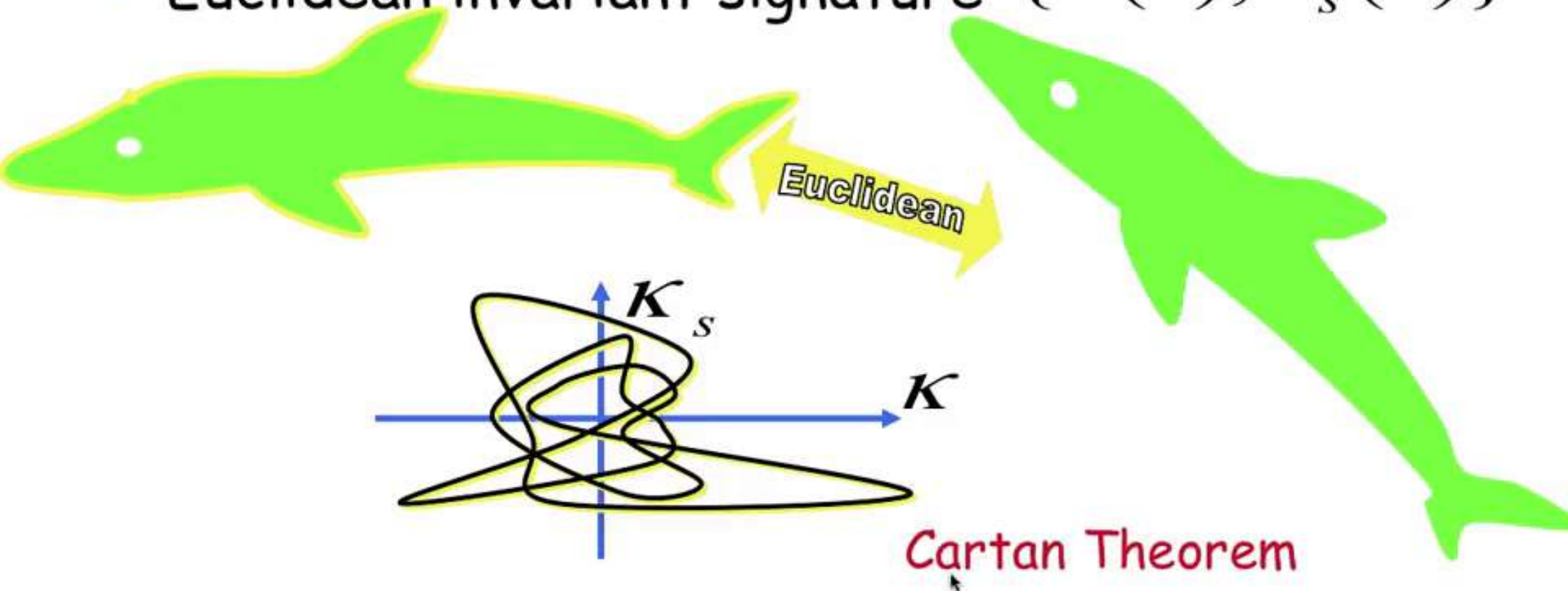
$$\{s, K(s)\}$$





# Differential Signatures

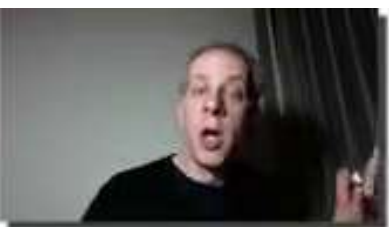
- Euclidean invariant signature  $\{K(s), K_s(s)\}$



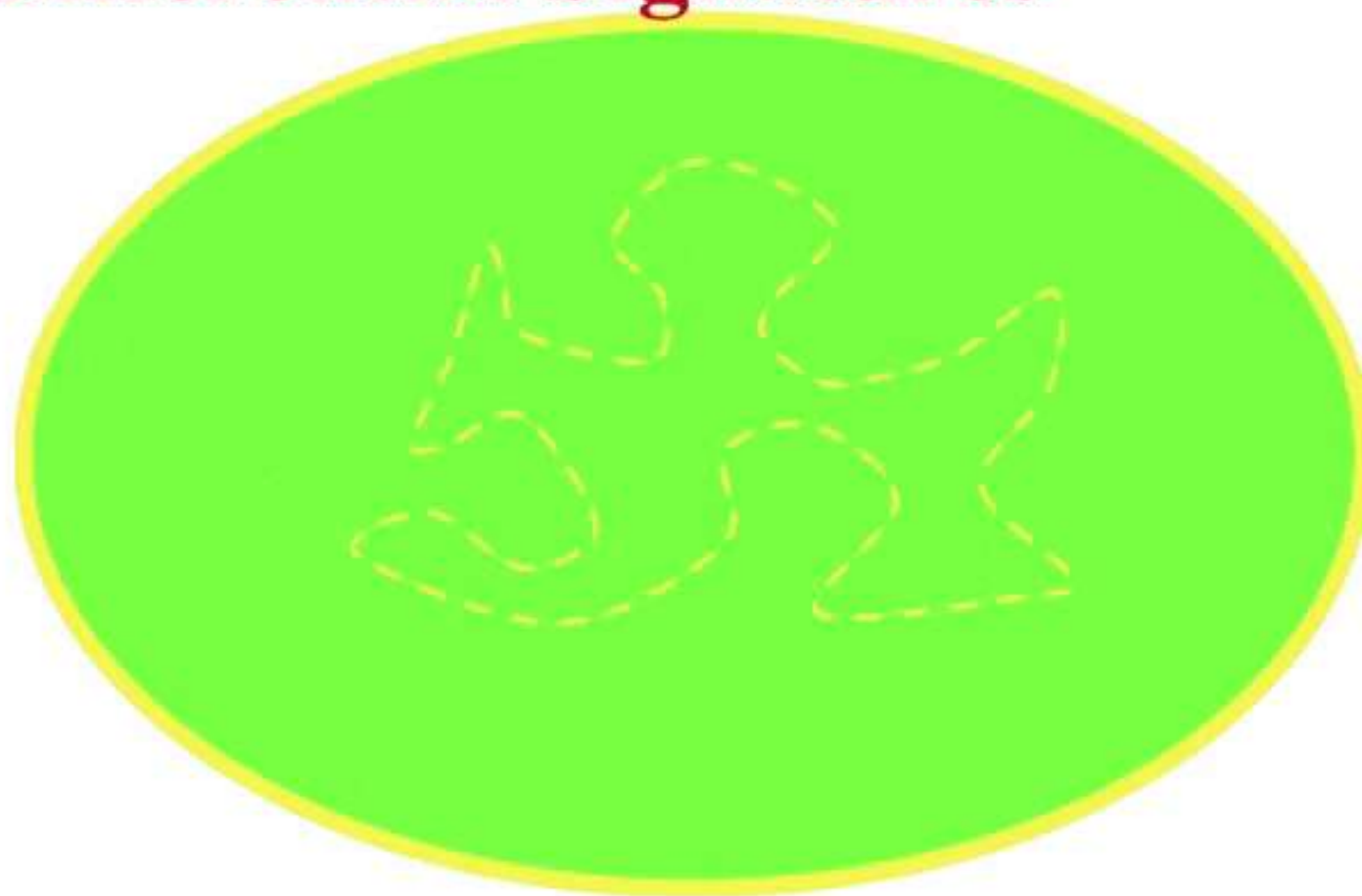


# Differential Signatures



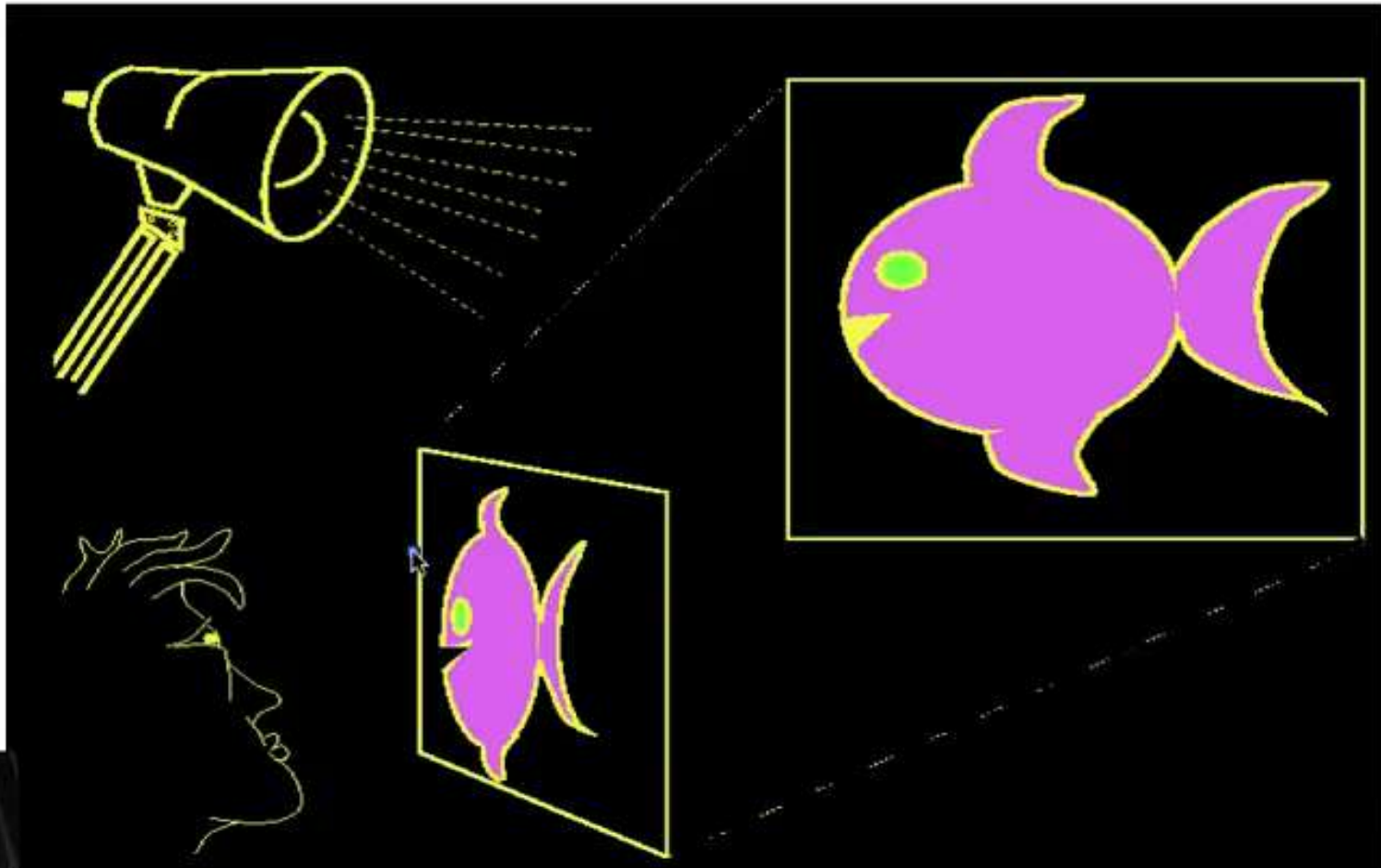


# Differential Signatures

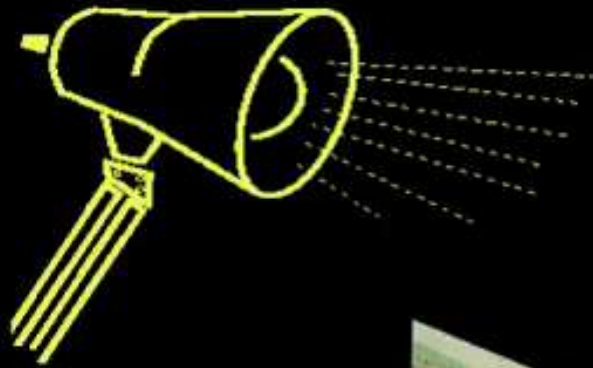




~Affine



~Affine



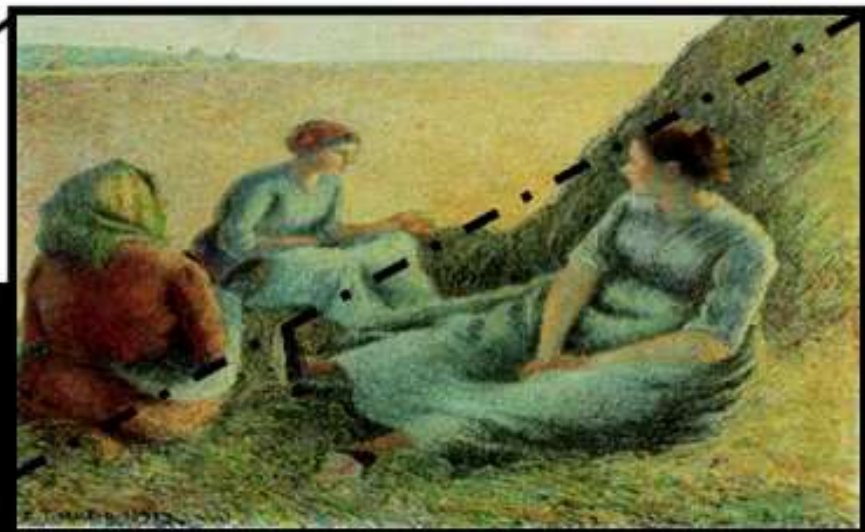
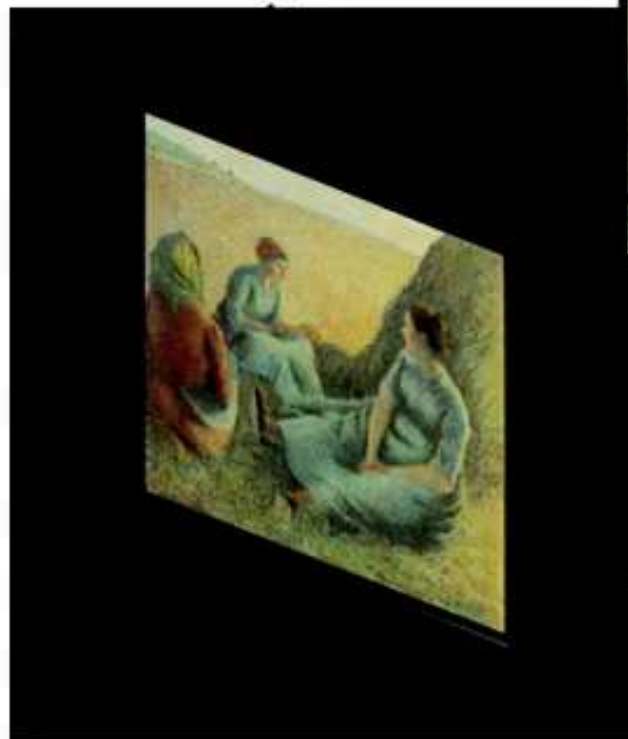
# Image transformation

$$I_2(x, y) = I_1(T_1(x, y), T_2(x, y))$$

$$\begin{pmatrix} T_1(x, y) \\ T_2(x, y) \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} e \\ f \end{pmatrix}$$

- **Equi-affine:**

$$\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = 1$$



$\mathcal{C}(p)$   $\mathcal{C}(r)$  **Invariant arclength should be**

1. Re-parameterization invariant



$$w = \int F(C, C_p, C_{pp}, \dots) dp = \int F(C, C_r, C_{rr}, \dots) dr$$

2. Invariant under the group of transformations





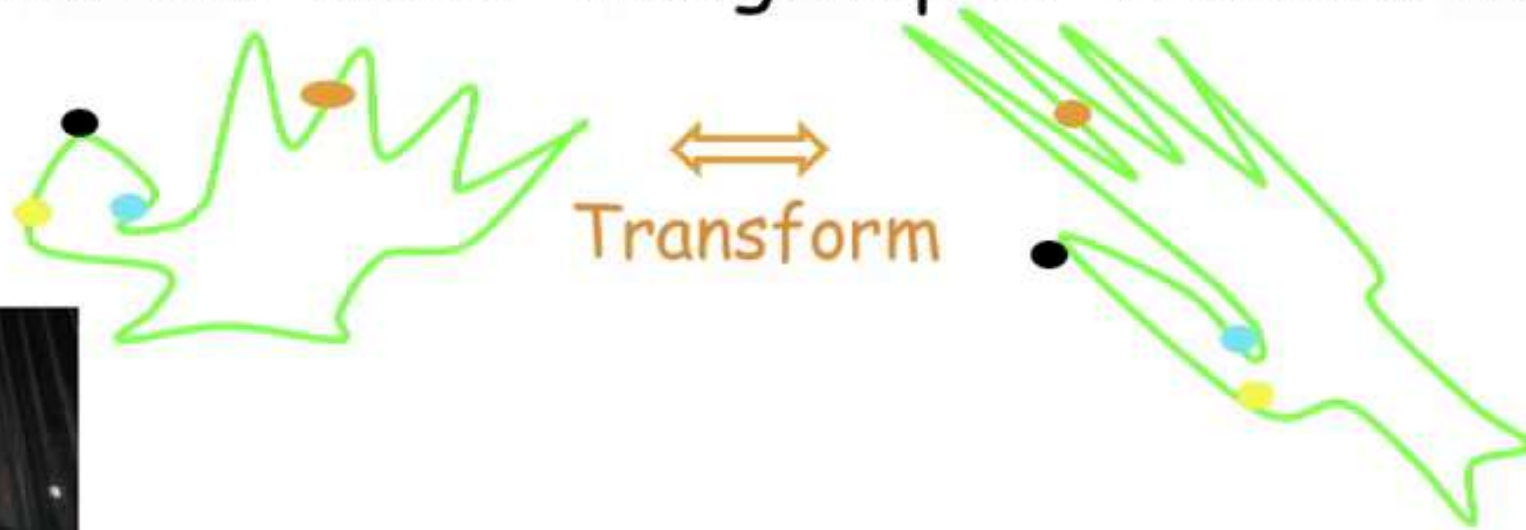
# Invariant arclength should be

## 1. Re-parameterization invariant

$$w = \int F(C, C_p, C_{pp}, \dots) dp = \int \boxed{F(C, C_r, C_{rr}, \dots)} dr$$

Geometric measure

## 2. Invariant under the group of transformations



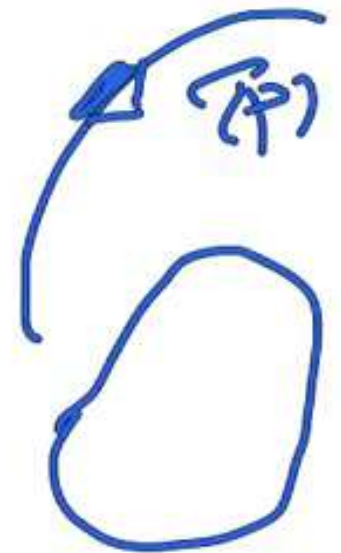
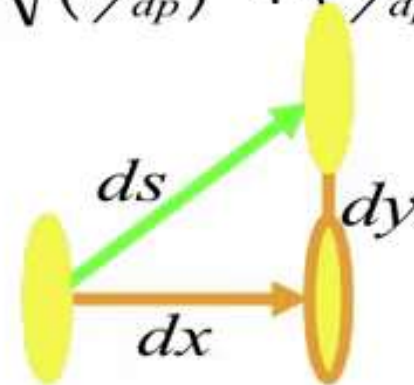
# Euclidean arclength

- Length is preserved, thus

$$ds = \sqrt{dx^2 + dy^2} = \frac{dp}{dp} \sqrt{dx^2 + dy^2} = dp \sqrt{\left(\frac{dx}{dp}\right)^2 + \left(\frac{dy}{dp}\right)^2} = |C_p| dp$$

$$s = \int |C_p| dp$$

$$|C_s| = 1$$



$$\text{Length } L = \int_0^1 |C_p| dp = \int_0^1 \langle C_p, C_p \rangle^{1/2} dp = \int_0^L ds$$



# Equi-affine arclength

- Area is preserved, thus

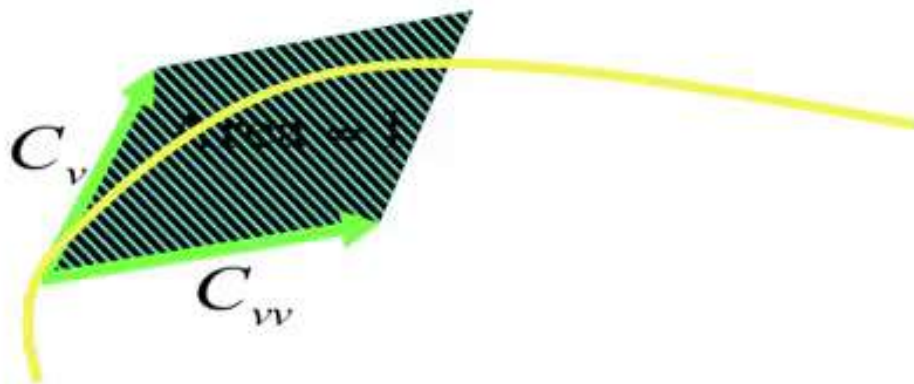
$$\begin{bmatrix} x_v & x_{vv} \\ y_v & y_{vv} \end{bmatrix}$$

$$(C_v, C_{vv}) = 1$$

$$v = \int (C_p, C_{pp})^{1/3} dp$$

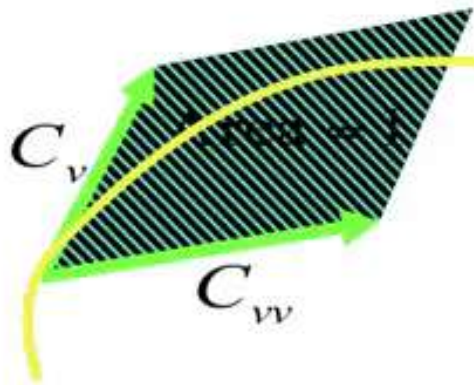
$$v = \int (C_s, C_{ss})^{1/3} ds = \int \kappa^{1/3} ds$$

$$dv = \kappa^{1/3} ds$$



# Equi-affine arclength

- Area is preserved, thus



re-parameterization  
invariance

$$(C_v, C_{vv}) = 1$$

$$v = \int (C_p, C_{pp})^{1/3} dp$$

$$v = \int (C_s, C_{ss})^{1/3} ds = \int \kappa^{1/3} ds$$

$$dv = \kappa^{1/3} ds$$





# $\langle C_s, C_s \rangle = 1$    **Equi-affine curvature**

$$\begin{aligned}(C_v, C_{vv}) &= 1 \Rightarrow \frac{d}{dv} (C_v, C_{vv}) = 0 \\&\Rightarrow \cancel{(C_{vv}, C_{vv})} + (C_v, C_{vvv}) = 0 \\&\Rightarrow (C_v, C_{vvv}) = 0 \\&\Rightarrow C_v \parallel C_{vvv} \Rightarrow C_{vvv} = \mu C_v\end{aligned}$$

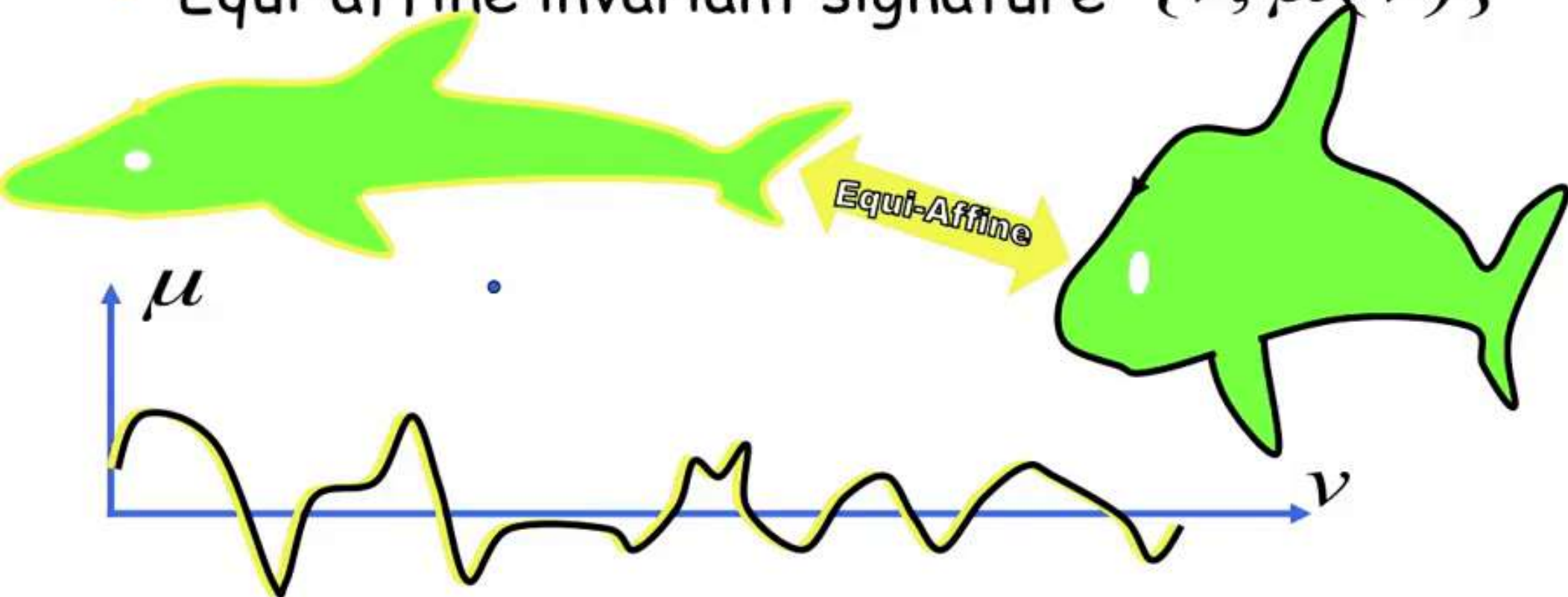
$\mu$  is the affine invariant curvature





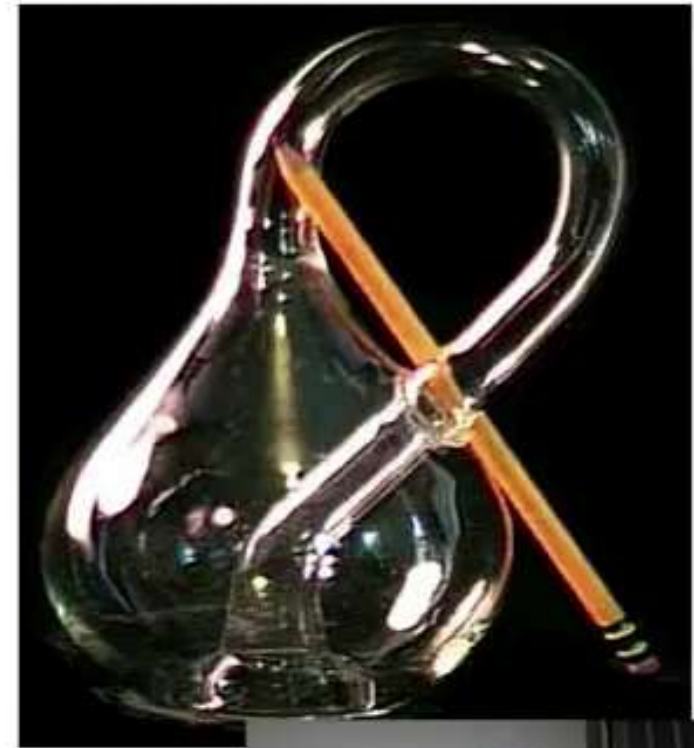
# Differential Signatures

- Equi-affine invariant signature  $\{v, \mu(v)\}$



# Surfaces

- Topology (Klein Bottle)

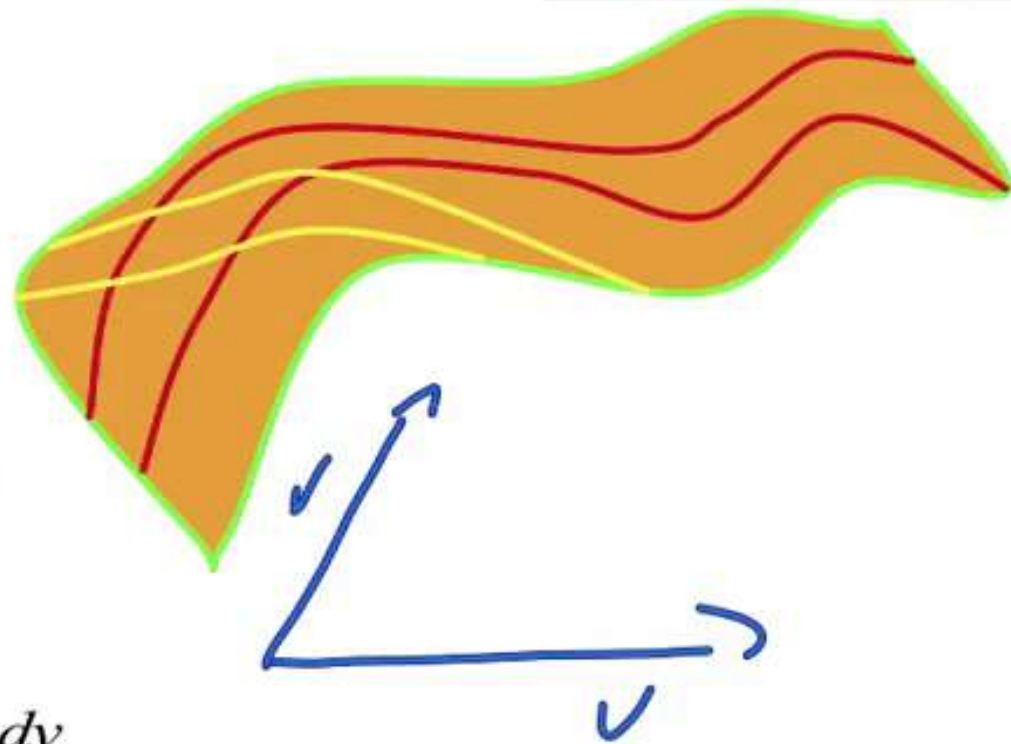


# Surface



$$S(u, v) = \{x(u, v), y(u, v), z(u, v)\}$$

- Normal  $\vec{N} = \frac{S_u \times S_v}{|S_u \times S_v|}$
- Area element  $dA = |S_u \times S_v|$
- Total area  $A = \iint |S_u \times S_v| du dv$

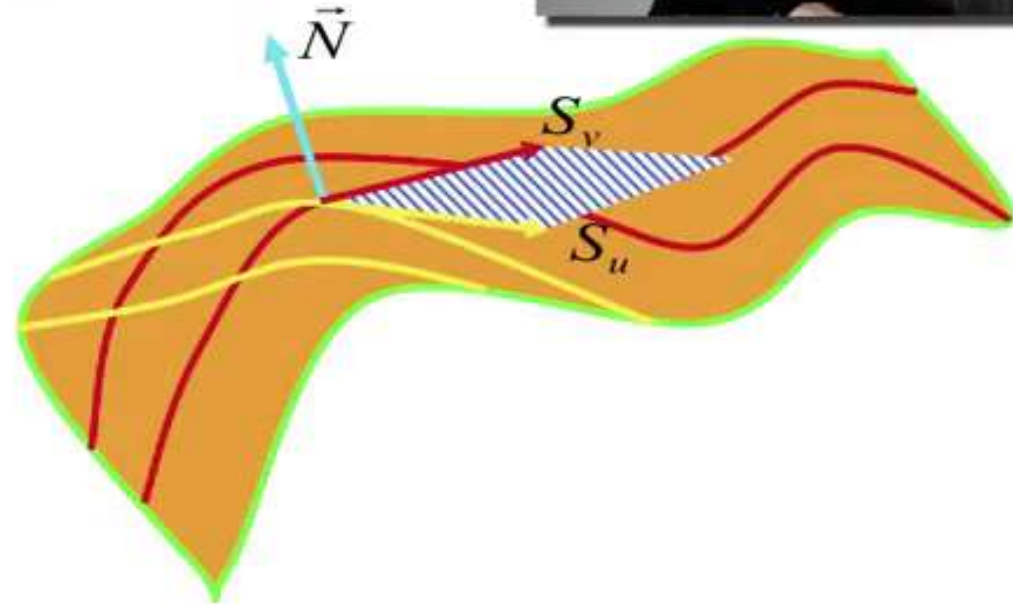




# Surface

$$S(u, v) = \{x(u, v), y(u, v), z(u, v)\}$$

- Normal  $\vec{N} = \frac{S_u \times S_v}{|S_u \times S_v|}$
- Area element  $dA = |S_u \times S_v|$
- Total area  $A = \iint |S_u \times S_v| du dv$

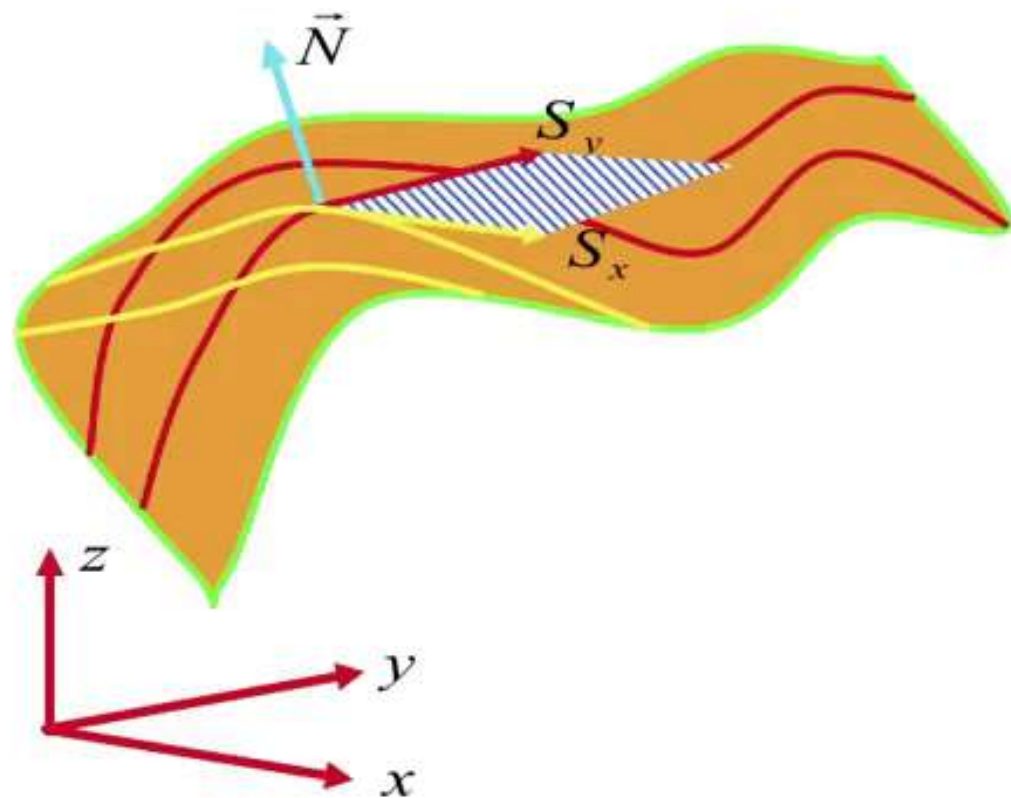




## Example: Surface as graph of function

- A surface,  $S: \mathbf{R}^2 \rightarrow \mathbf{R}^3$

$$S(u, v) = \{x = u, y = v, z(u, v)\}$$





Normal Curvature

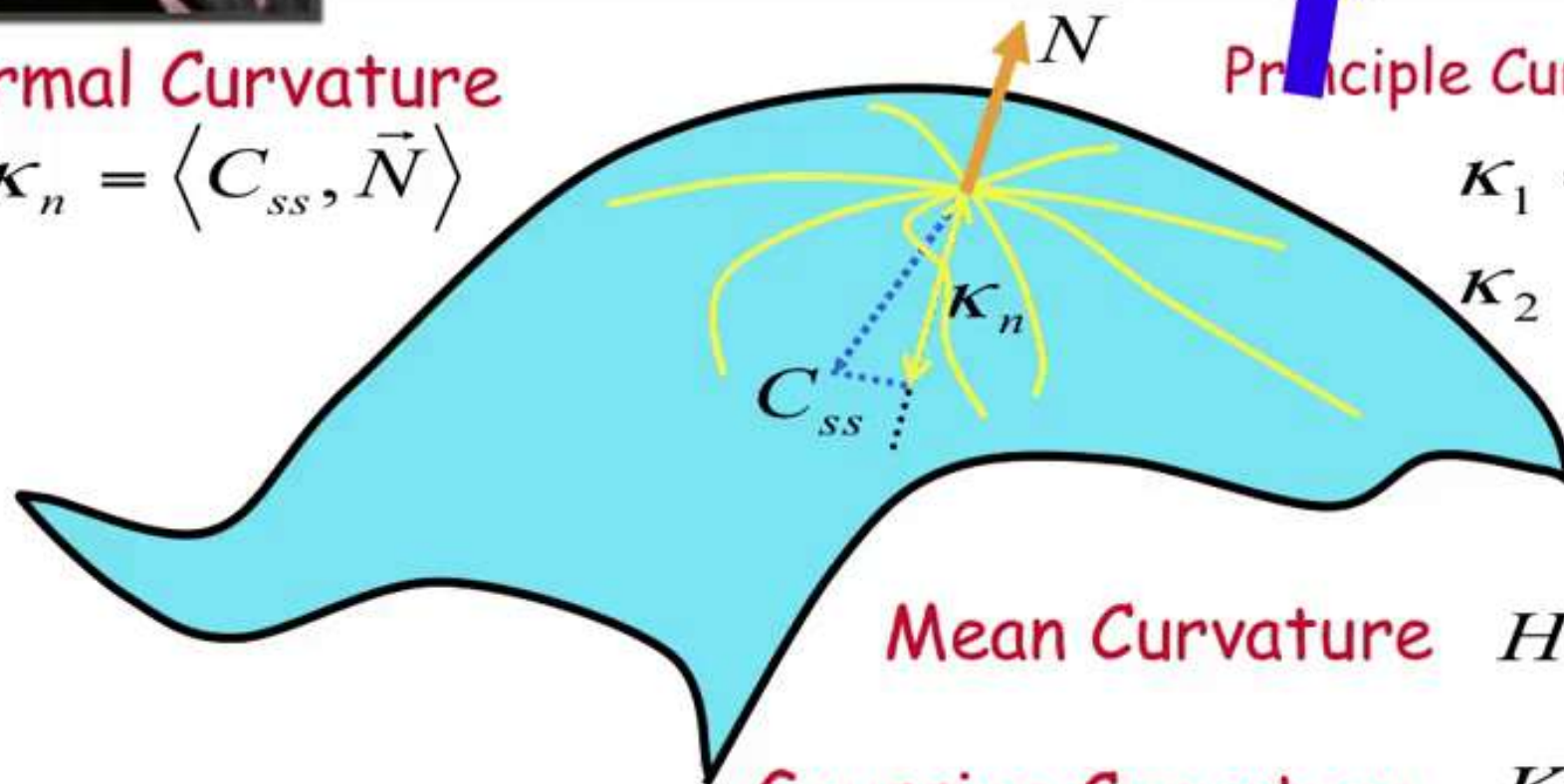
$$\kappa_n = \langle C_{ss}, \vec{N} \rangle$$

Principal

Principle Curvatures

$$\kappa_1 = \max_{\theta}(\kappa)$$

$$\kappa_2 = \min_{\theta}(\kappa)$$



Mean Curvature  $H = \frac{\kappa_1 + \kappa_2}{2}$

Gaussian Curvature  $K = \kappa_1 \kappa_2$

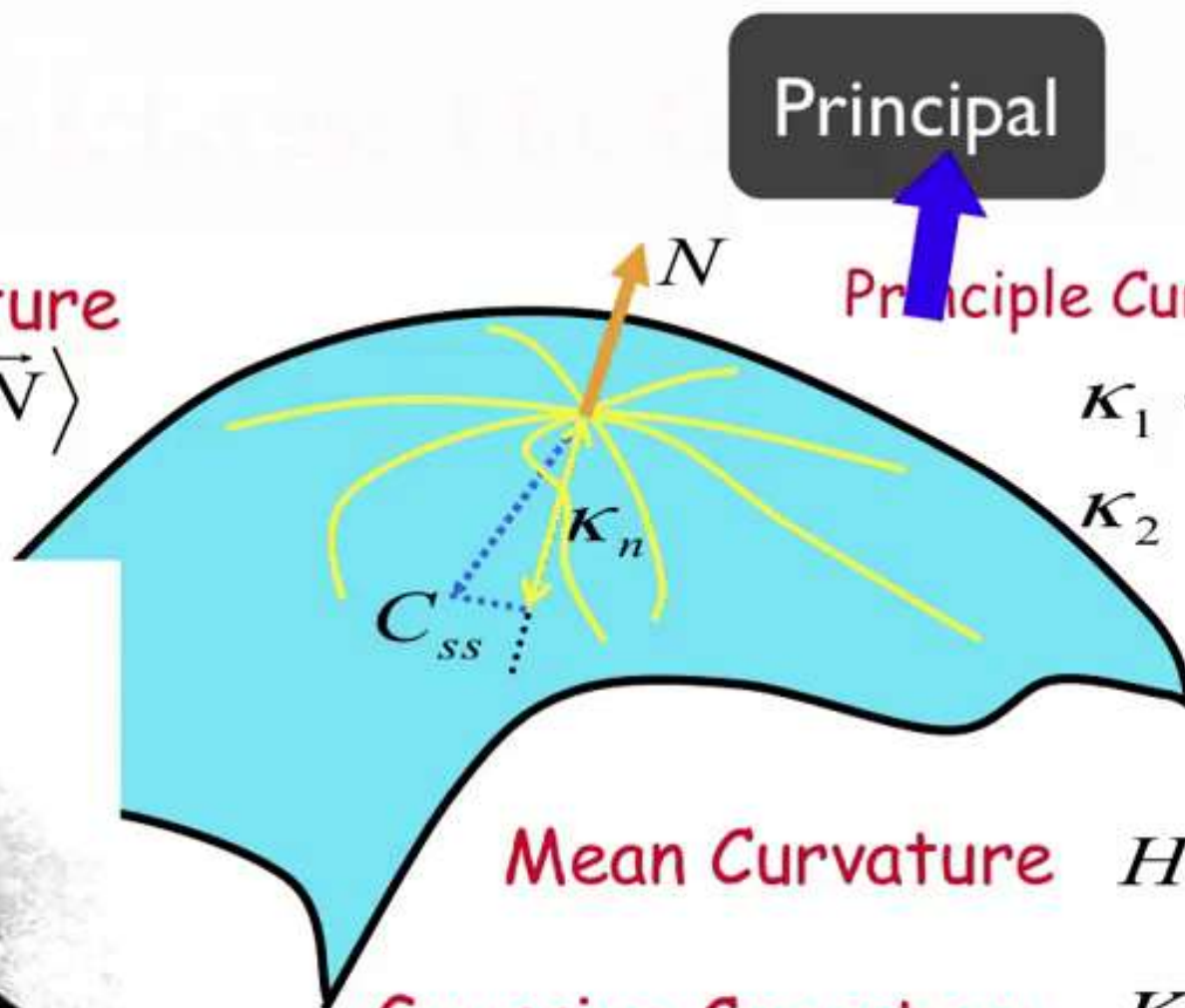


Normal Curvature

$$\kappa_n = \langle C_{ss}, \vec{N} \rangle$$



Gauss



Principal

Principal Curvatures

$$\kappa_1 = \max_{\theta}(\kappa)$$

$$\kappa_2 = \min_{\theta}(\kappa)$$

Mean Curvature

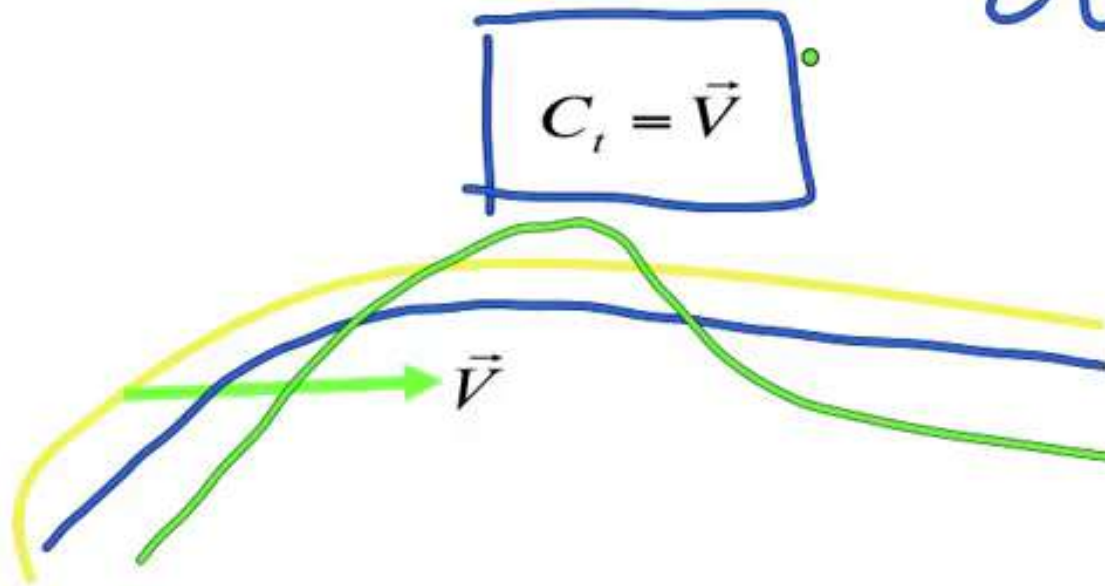
$$H = \frac{\kappa_1 + \kappa_2}{2}$$

Gaussian Curvature

$$K = \kappa_1 \kappa_2$$



$$\frac{\partial C(p)}{\partial t} = \vec{V}(p, t)$$



## Important property

- Tangential components do not affect the geometry of an evolving curve

$$\boxed{C_t = \vec{V}} \Leftrightarrow C_t = \langle \vec{V}, \vec{n} \rangle \vec{n}$$
$$\underline{C_t = \alpha \vec{t}}$$



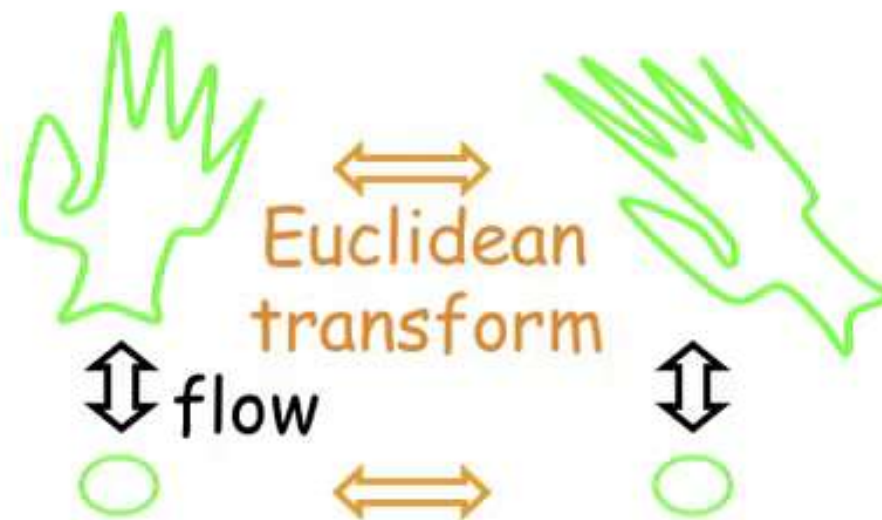
# Curvature flow

- Euclidean geometric heat equation  $C_t = \kappa \vec{n}$

---

$$C_t = C_{ss} \quad \text{where} \quad C_{ss} = \kappa \vec{n}$$

$$\frac{\partial C}{\partial t} = \frac{\partial^2 C}{\partial s^2}$$



# Curvature flow $C_t = \kappa \vec{n}$



Grayson



First becomes convex



Gage-Hamilton



Vanish at a Circular point



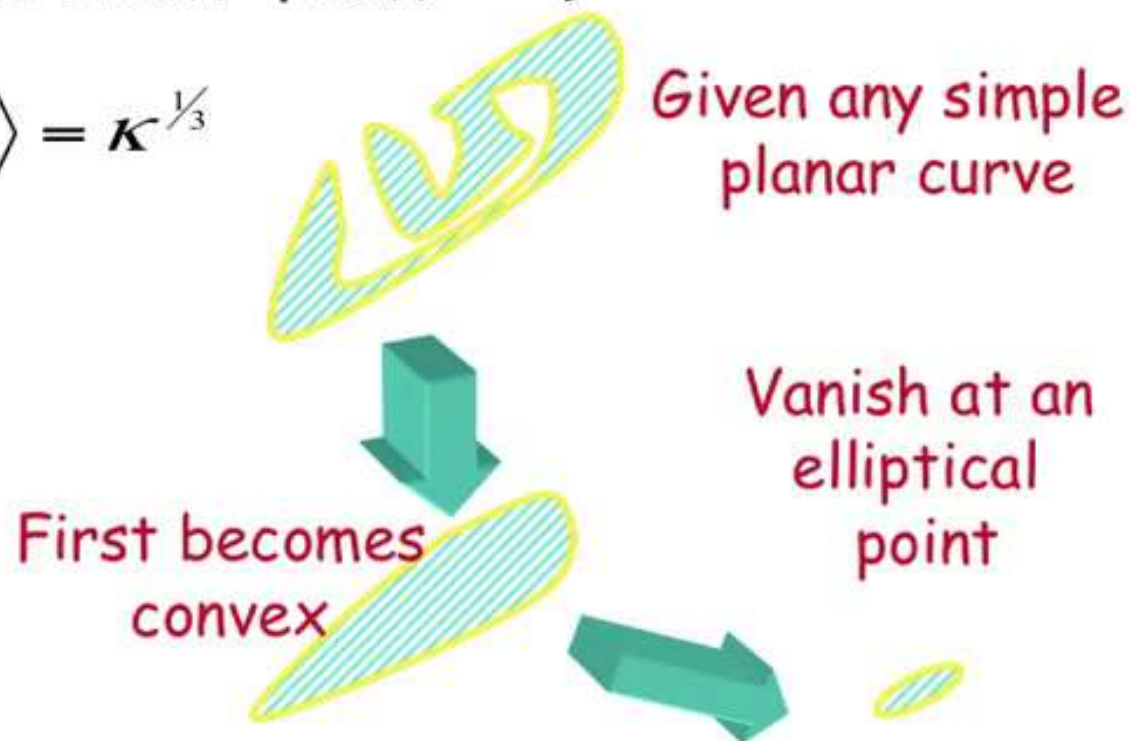




# Affine heat equation $C_t = C_{vv}$

- Special (equi-)affine heat flow  $C_t = \kappa^{1/3} \vec{n}$

$$C_t = \langle C_{vv}, \vec{n} \rangle \vec{n} \quad \text{where} \quad \langle C_{vv}, \vec{n} \rangle = \kappa^{1/3}$$

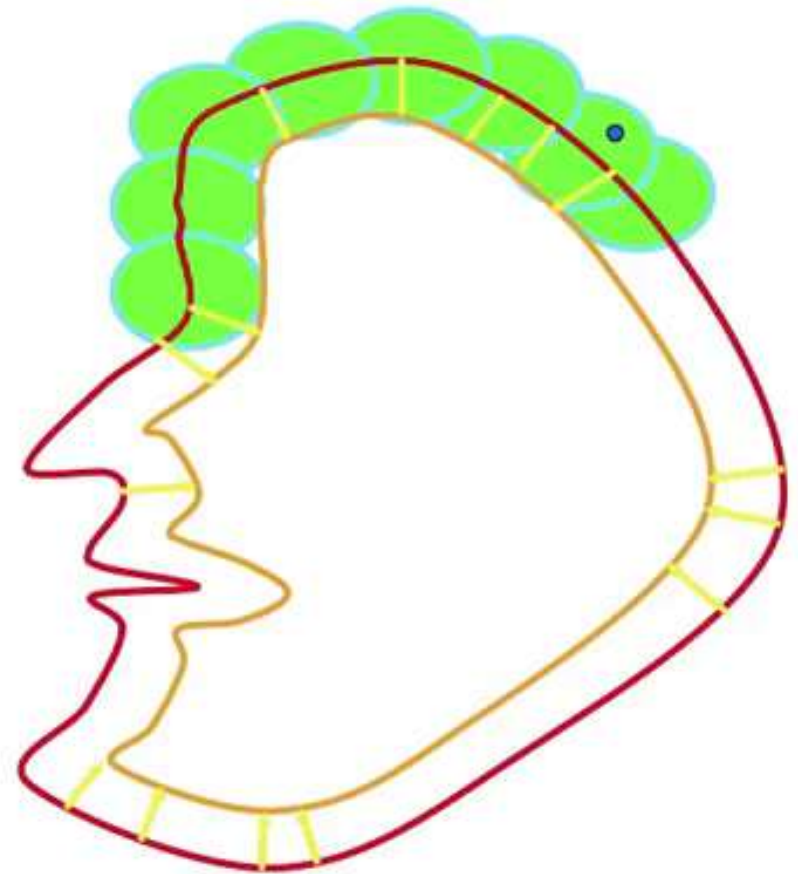




## Constant flow

- Offset curves
- Equal-height contours of the distance transform
- Envelope of all disks of equal radius centered along the curve (Huygens principle)

$$C_t = \vec{n}$$





# Constant flow

$$C_t = \vec{n}$$

- Offset curves

Change in topology





## So far we defined

Constant flow

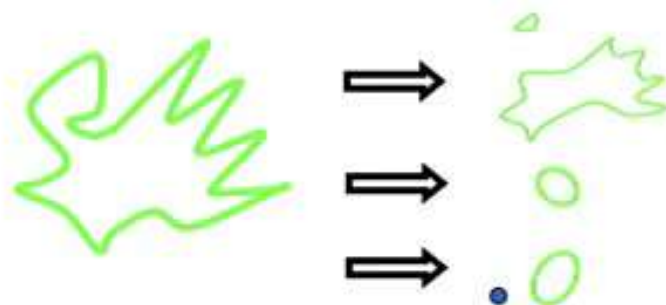
$$C_t = \vec{n}$$

Curvature flow

$$C_t = \kappa \vec{n}$$

Equi-affine flow

$$C_t = \kappa^{1/3} \vec{n}$$







$$C_t = V\vec{n}$$

$$\frac{\partial}{\partial t} L = \frac{\partial}{\partial t} \oint \langle C_p, C_p \rangle^{1/2} dp = 2 \oint \left\langle \frac{\partial}{\partial t} C_p, C_p \right\rangle dp = \dots = - \int_0^L \kappa V ds$$

$$\frac{\partial}{\partial t} A = \frac{1}{2} \frac{\partial}{\partial t} \oint (C, C_p) dp = \oint \left( \frac{\partial}{\partial t} C, C_p \right) dp + \oint \left( C, \frac{\partial}{\partial t} C_p \right) dp = \dots = - \int_0^L V ds$$

$$\frac{\partial}{\partial t} \kappa = \frac{\partial}{\partial t} \left( \frac{(C_p, C_{pp})}{\langle C_p, C_p \rangle^{3/2}} \right) = \dots = V_{ss} + \kappa^2 V$$



Length

Area

Curvature

$$L_t = - \int_0^L \kappa V ds$$

$$A_t = \int_0^L V ds$$

$$\kappa_t = V_{ss} + \kappa^2 V$$



# Constant flow ( $V = 1$ )

Length

$$L_t = -\int_0^L \kappa V ds = -\int_0^L \kappa ds = -2\pi$$

Area

$$A_t = -\int_0^L V ds = -\int_0^L ds = -L$$

Curvature

$$\kappa_t = V_{ss} + \kappa^2 V = \kappa^2$$

The curve vanishes at

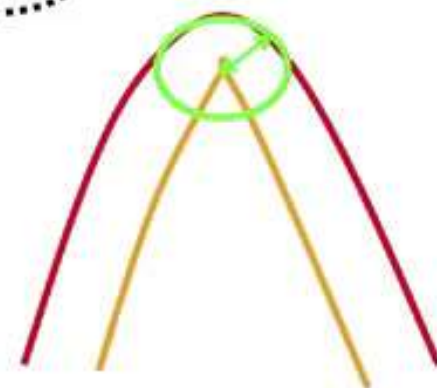
Riccati eq.

Singularity ('shock') at

$$t = \frac{L(0)}{2\pi}$$

$$\kappa(p, t) = \frac{\kappa(p, 0)}{1 - t\kappa(p, 0)}$$

$$t = \rho(p, 0)$$





# Curvature flow ( $V = \kappa$ )

Length

$$L_t = -\int_0^L \kappa V ds = -\int_0^L \kappa^2 ds$$

Area

$$A_t = -\int_0^L V ds = -\int_0^L \kappa ds = -2\pi$$

Curvature

$$\kappa_t = V_{ss} + \kappa^2 V = \kappa_{ss} + \kappa^3$$

The curve vanishes at

$$t = \frac{A(0)}{2\pi}$$





# Equi-Affine flow ( $V = \kappa^{1/3}$ )

Length

$$L_t = -\int_0^L \kappa V ds = -\int_0^L \kappa^{4/3} ds$$



Area

$$A_t = -\int_0^L V ds = -\int_0^L \kappa^{1/3} ds$$

Curvature

$$\kappa_t = V_{ss} + \kappa^2 V = \frac{1}{3} \kappa^{-2/3} \kappa_{ss} - \frac{2}{9} \kappa^{-5/3} \kappa_s^2 + \kappa^{7/3}$$





# Geodesic active contours

✓

$$C_t = \left( g(x, y) \kappa - \langle \nabla g(x, y), \vec{n} \rangle \right) \vec{n}$$

$$C_t = g \quad \kappa \vec{n}$$



$$g \sim \frac{1}{|\nabla I|}$$



# Geodesic active contours

✓

$$C_t = \left( g(x, y) \kappa - \langle \nabla g(x, y), \vec{n} \rangle \right) \vec{n}$$

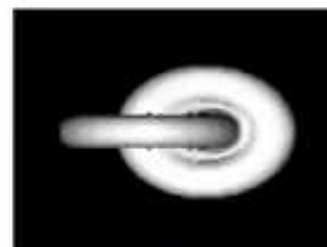
$$C_t = g \kappa \vec{n} - \langle \nabla g, \vec{n} \rangle \vec{n}$$



$$g \sim \frac{1}{|\nabla I|}$$



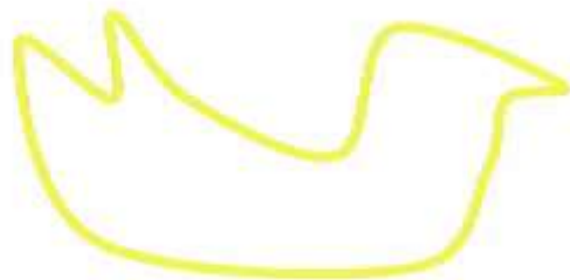
## Surface evolution...



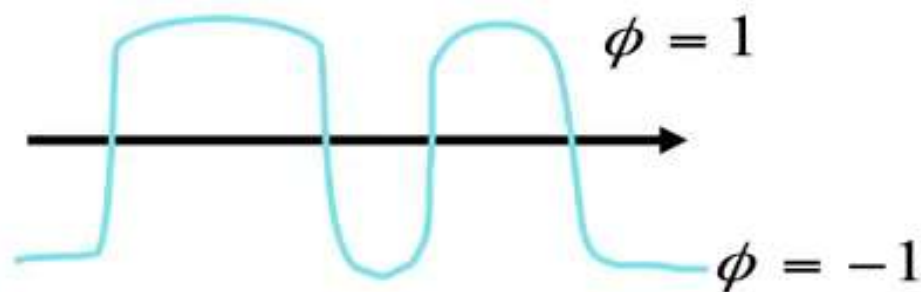
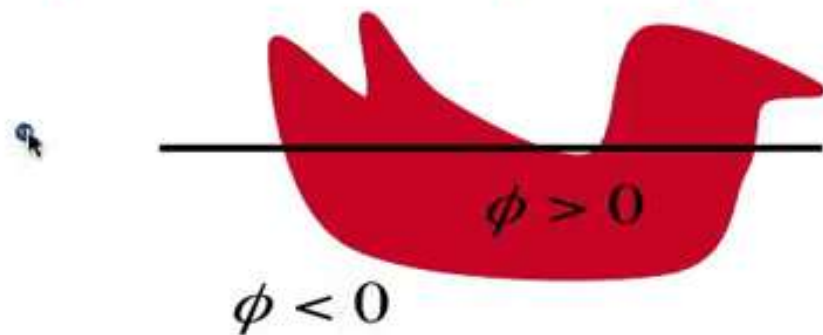
$$\frac{\partial S}{\partial t} = g \kappa \vec{n}$$

# Implicit representation

Consider a closed planar curve  $C(p) : \mathbf{S}^1 \rightarrow \mathbf{R}^2$



The geometric trace of the curve can be alternatively represented implicitly as  $C = \{(x, y) \mid \phi(x, y) = 0\}$



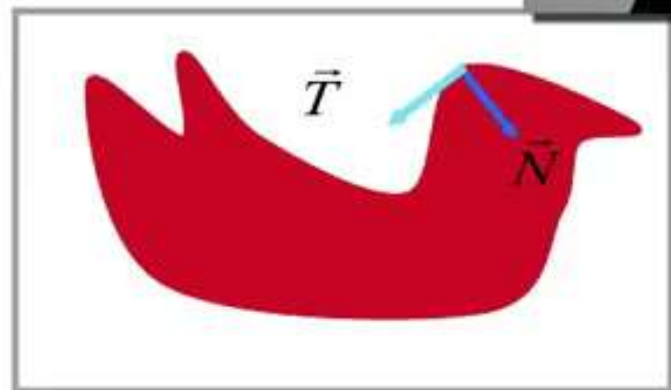


# Properties of level sets



The level set normal  $\phi$

$$\vec{N} = -\frac{\nabla \phi}{|\nabla \phi|} \quad \left( \vec{T} = \frac{\bar{\nabla} \phi}{|\nabla \phi|} \right)$$



*Proof.* Along the level sets we have zero change, that is  $\phi_s = 0$ , but by the chain rule

$$\phi_s(x, y) = \phi_x x_s + \phi_y y_s = \langle \nabla \phi, \vec{T} \rangle$$

$$\begin{bmatrix} a \\ b \end{bmatrix}, \begin{bmatrix} c \\ d \end{bmatrix}$$

So,

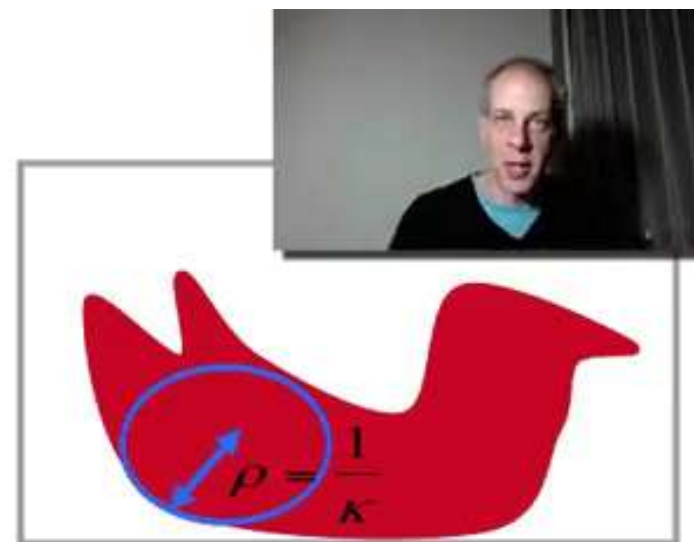
$$\left\langle \frac{\nabla \phi}{|\nabla \phi|}, \vec{T} \right\rangle = 0 \Rightarrow \frac{\nabla \phi}{|\nabla \phi|} \perp \vec{T} \Rightarrow \vec{N} = -\frac{\nabla \phi}{|\nabla \phi|}$$



# Properties of level sets

The level set curvature

$$C_S = k \vec{n} \quad \kappa = \operatorname{div} \left( \frac{\nabla \phi}{|\nabla \phi|} \right) = \alpha_x + \beta_y$$



Proof: zero change along the level sets,  $\phi_{ss} = 0$ , also

$$\phi_{ss}(x, y) = \frac{d}{ds} (\phi_x x_s + \phi_y y_s) = \frac{d}{ds} \langle \nabla \phi, \vec{T} \rangle = \left\langle \frac{d}{ds} \nabla \phi, \vec{T} \right\rangle + \langle \nabla \phi, \kappa \vec{N} \rangle = 0$$

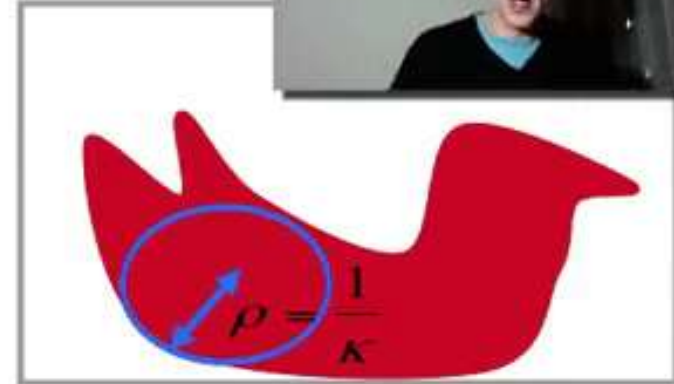
$$\kappa \left\langle \nabla \phi, \frac{\nabla \phi}{|\nabla \phi|} \right\rangle = \kappa |\nabla \phi| = - \left\langle [\phi_{xx} x_s + \phi_{xy} y_s, \phi_{xy} x_s + \phi_{yy} y_s], \frac{\nabla \phi}{|\nabla \phi|} \right\rangle$$

# Properties of level sets

The level set curvature

$$C_S = \kappa \vec{n} \quad \kappa = \operatorname{div} \left( \frac{\nabla \phi}{|\nabla \phi|} \right) = \alpha_x + \beta_y$$

(Handwritten note:  $\operatorname{div}(\alpha, \beta)$ )



Proof: zero change along the level sets,  $\phi_{ss} = 0$ , also

$$\phi_{ss}(x, y) = \frac{d}{ds} (\phi_x x_s + \phi_y y_s) = \frac{d}{ds} \langle \nabla \phi, \vec{T} \rangle = \left\langle \frac{d}{ds} \nabla \phi, \vec{T} \right\rangle + \langle \nabla \phi, \kappa \vec{N} \rangle = 0$$

$$\kappa \left\langle \nabla \phi, \frac{\nabla \phi}{|\nabla \phi|} \right\rangle = \kappa |\nabla \phi| = - \left\langle [\phi_{xx} x_s + \phi_{xy} y_s, \phi_{xy} x_s + \phi_{yy} y_s], \frac{\nabla \phi}{|\nabla \phi|} \right\rangle$$



# Level Set Formulation

(Osher-Sethian)

$$\phi(x, y): \mathbf{R}^2 \rightarrow \mathbf{R} \quad C = \{(x, y): \phi(x, y) = 0\}$$

$$\frac{dC}{dt} = V\vec{N} \Leftrightarrow \frac{d\phi}{dt} = V|\nabla\phi|$$

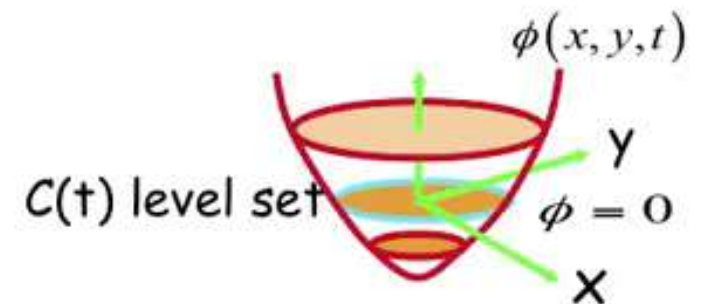
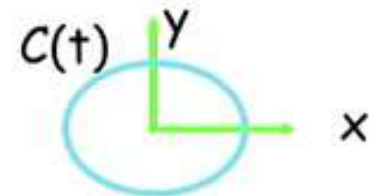
$$0 = \frac{\partial\phi(x, y; t)}{\partial t} = \phi_x x_t + \phi_y y_t + \phi_t$$

$$-\phi_t = \phi_x x_t + \phi_y y_t = \langle \nabla\phi, C_t \rangle = \langle \nabla\phi, V\vec{N} \rangle = V \langle \nabla\phi, \vec{N} \rangle$$

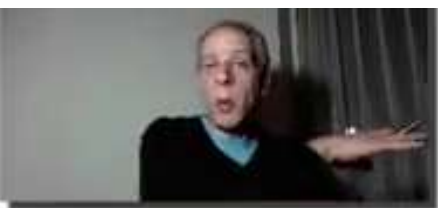
$$\vec{N} = -\frac{\nabla\phi}{|\nabla\phi|}$$

$$-V \langle \nabla\phi, \vec{N} \rangle = V \left\langle \nabla\phi, \frac{\nabla\phi}{|\nabla\phi|} \right\rangle = V |\nabla\phi|$$

$$\phi_t = V |\nabla\phi|$$

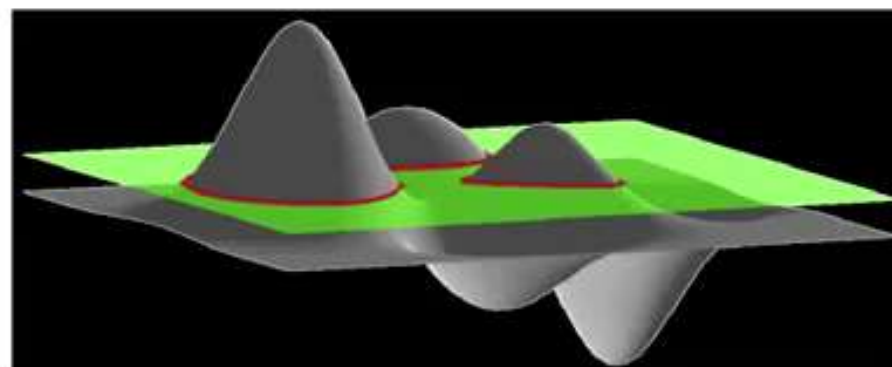
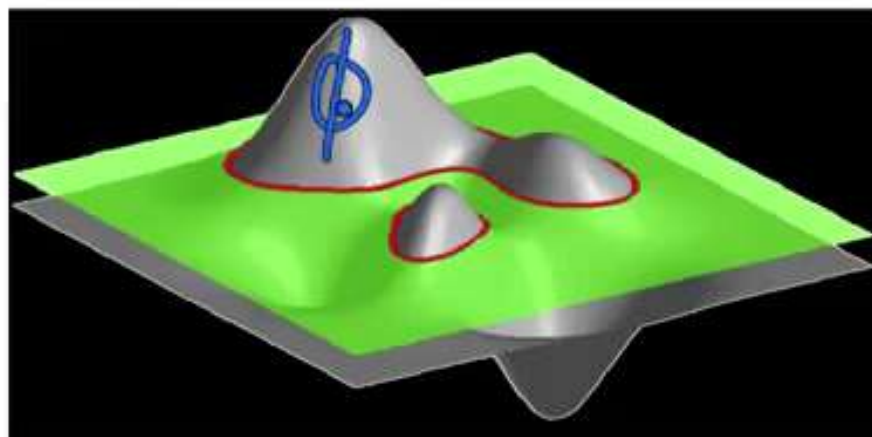
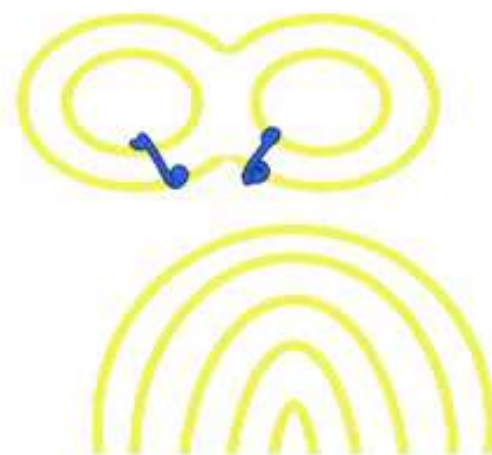






# Level Set Formulation

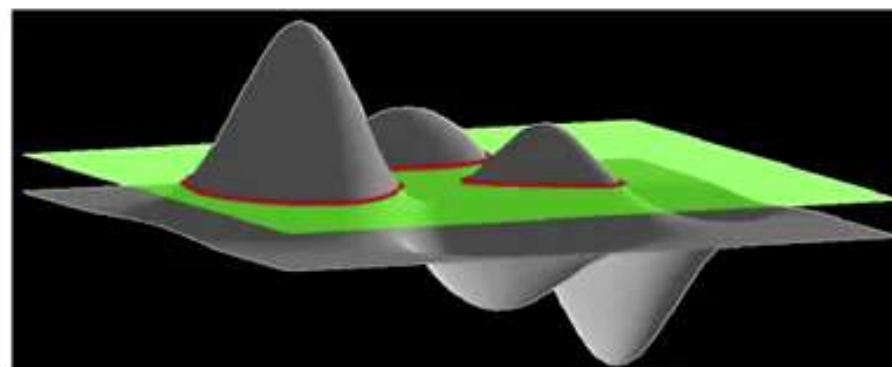
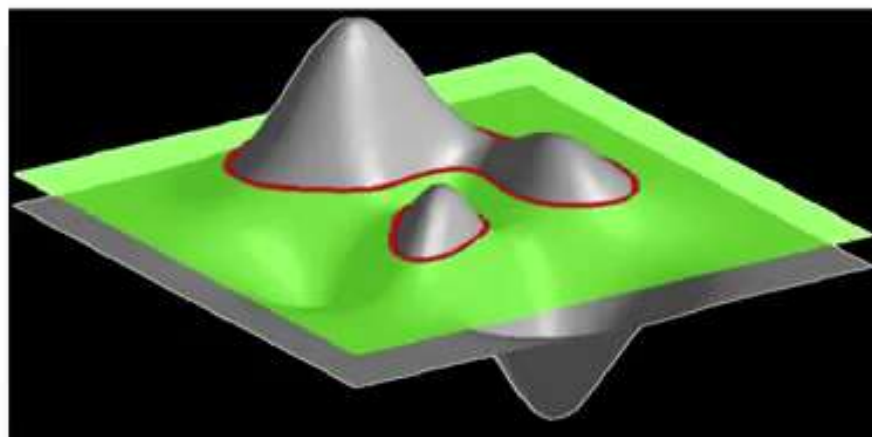
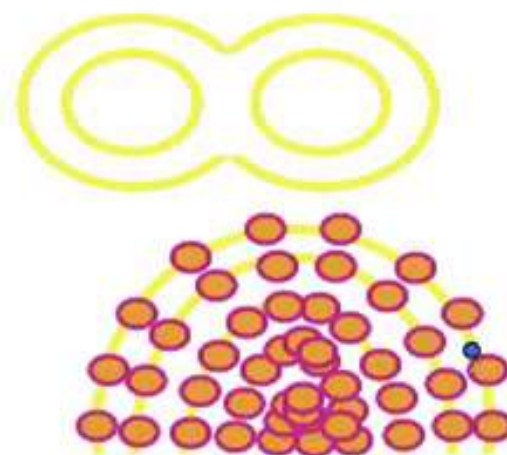
- Handles changes in topology
- Numeric grid points never collide or drift apart.
- Natural philosophy for dealing with gray level images.





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# Level Set Formulation

- Handles changes in topology
- Numeric grid points never collide or drift apart.
- Natural philosophy for dealing with gray level images.

$$V = g \cdot k$$

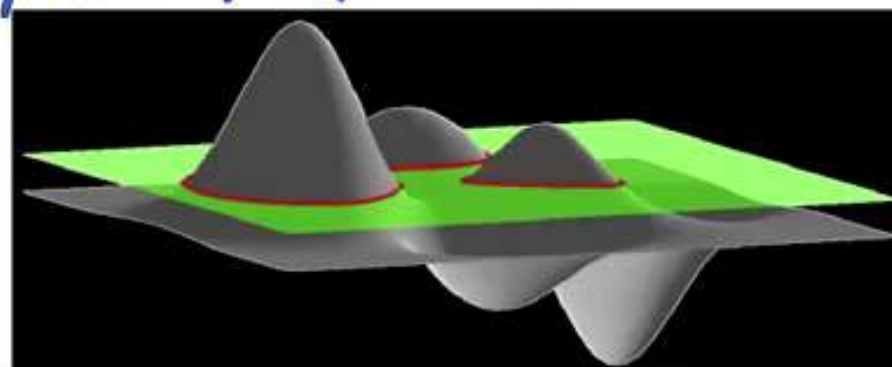
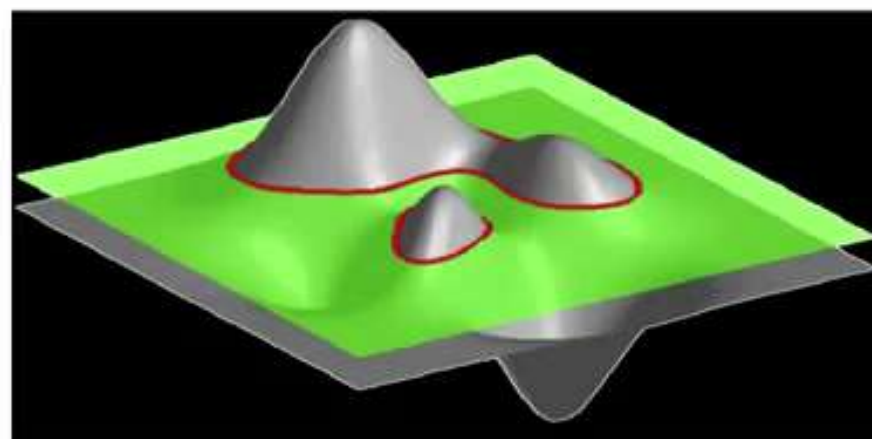
$$\phi_t = g \cdot k$$

$$(\nabla \phi)$$

$$\phi_t = |\nabla \phi|$$

$$\phi_t = \sqrt{|\nabla \phi|^2}$$

$$\phi = 0$$







# Calculus of Variations

Generalization of Calculus that seeks to find the path, curve, surface, etc., for which a given **Functional** has a minimum or maximum.

Goal: find extrema values of integrals of the form

$$\int F(u, u_x) dx$$

It has an extremum only if the **Euler-Lagrange** Differential Equation is satisfied,

$$\left( \frac{\partial}{\partial u} - \frac{d}{dx} \frac{\partial}{\partial u_x} \right) F(u, u_x) = 0$$





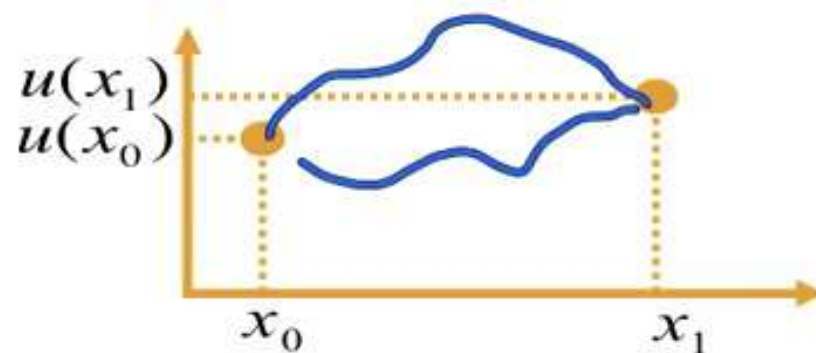


# Calculus of Variations

Example: Find the shape of the curve  $\{x, u(x)\}$  with shortest length:

$$\int_{x_0}^{x_1} \sqrt{1 + u_x^2} dx$$

given  $u(x_0), u(x_1)$



Solution: a differential equation that  $u(x)$  must satisfy,

$$\left( \frac{\partial}{\partial u} - \frac{d}{dx} \frac{\partial}{\partial u_x} \right) F(u, u_x) = 0$$

$$\frac{u_{xx}}{(1 + u_x^2)^{3/2}} = 0 \quad \Rightarrow \quad u_x = a \quad \Rightarrow \quad u(x) = ax + b$$

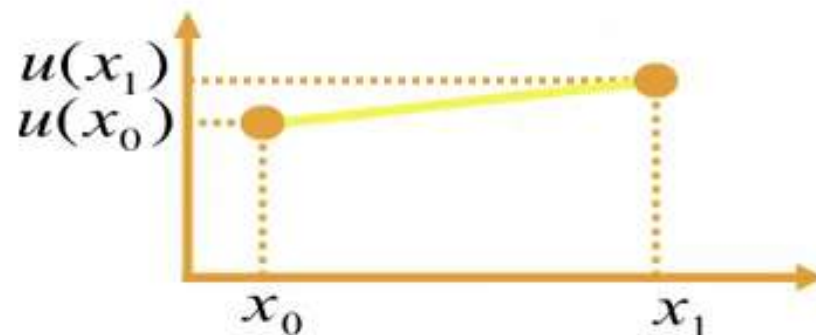


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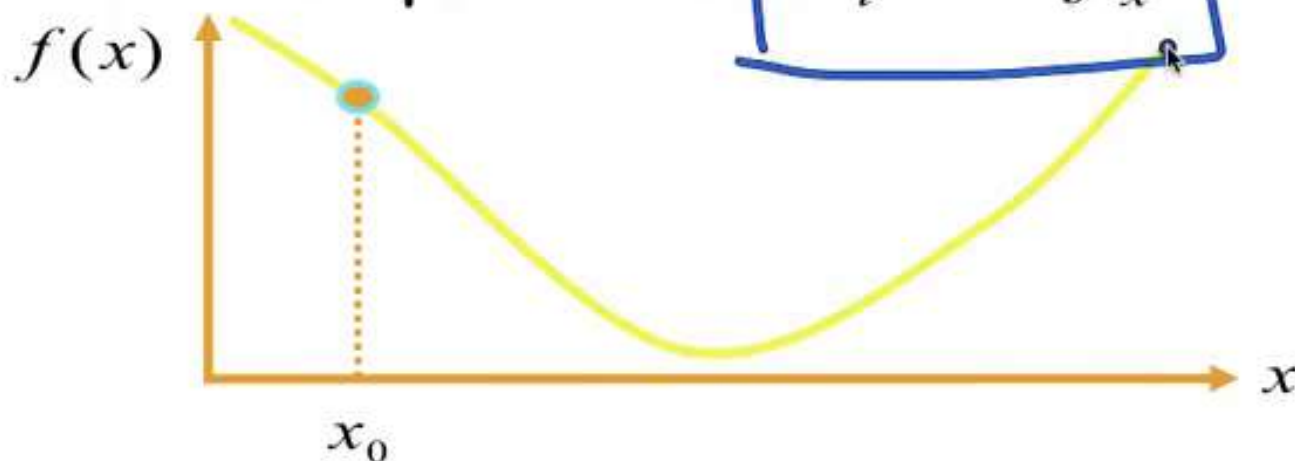


# Extrema points in calculus

$$\forall \eta : \lim_{\varepsilon \rightarrow 0} \left( \frac{df(x + \varepsilon \eta)}{d\varepsilon} \right) = 0 \Leftrightarrow \forall \eta : f_x(x) \eta = 0 \Leftrightarrow \underline{f_x(x) = 0}$$

Gradient descent process

$$x_t = -f_x$$

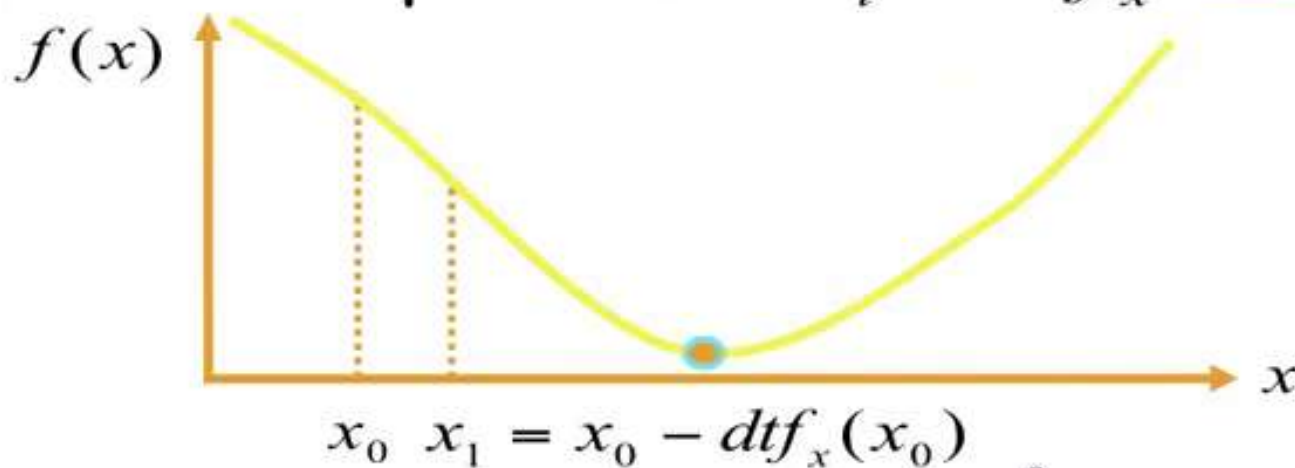




## Extrema points in calculus

$$\forall \eta : \lim_{\varepsilon \rightarrow 0} \left( \frac{df(x + \varepsilon \eta)}{d\varepsilon} \right) = 0 \Leftrightarrow \forall \eta : f_x(x) \eta = 0 \Leftrightarrow f_x(x) = 0$$

Gradient descent process  $x_t = -f_x \frac{x(t+\Delta t) - x(t)}{\Delta t}$







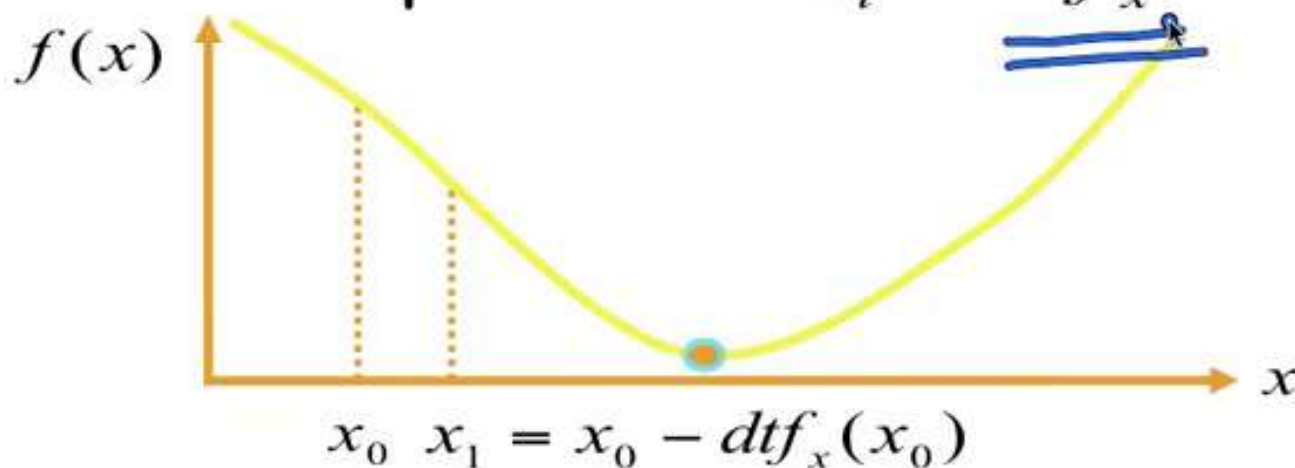
# Extrema points in calculus

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Gradient descent process

$$x_t = -f_x$$

$$x_t = 0$$





# Calculus of variations

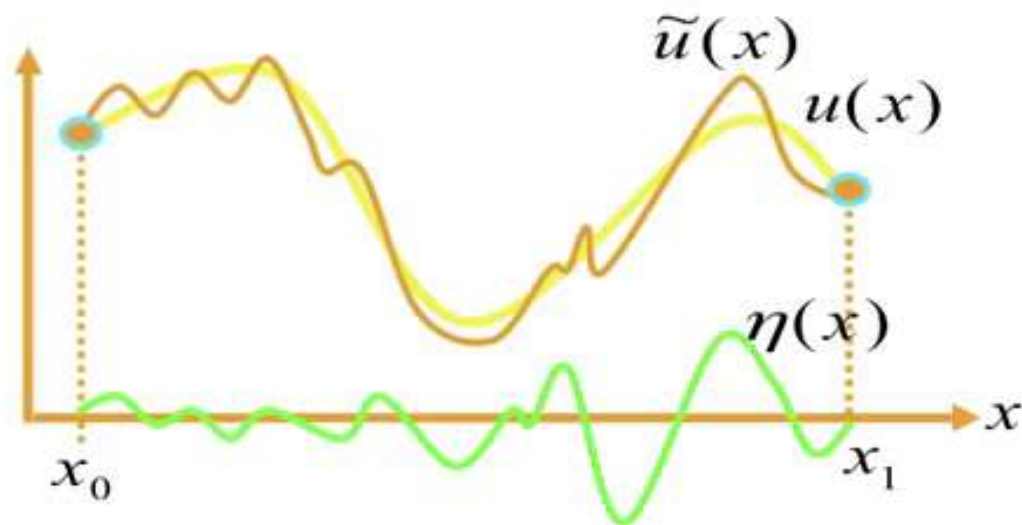
$$E(u(x)) = \int F(u, u_x) dx$$

$$\tilde{u}(x) = u(x) + \varepsilon \eta(x)$$

$$\forall \eta(x) : \lim_{\varepsilon \rightarrow 0} \left( \frac{d}{d\varepsilon} \int F(\tilde{u}, \tilde{u}_x) dx \right) \stackrel{?}{=} 0$$



$$\frac{\delta E(u)}{\delta u} = \left( \frac{\partial}{\partial u} - \frac{d}{dx} \frac{\partial}{\partial u_x} \right) F(u, u_x)$$



Gradient descent process

$$u_t = - \frac{\delta E(u)}{\delta u}$$



# Calculus of variations

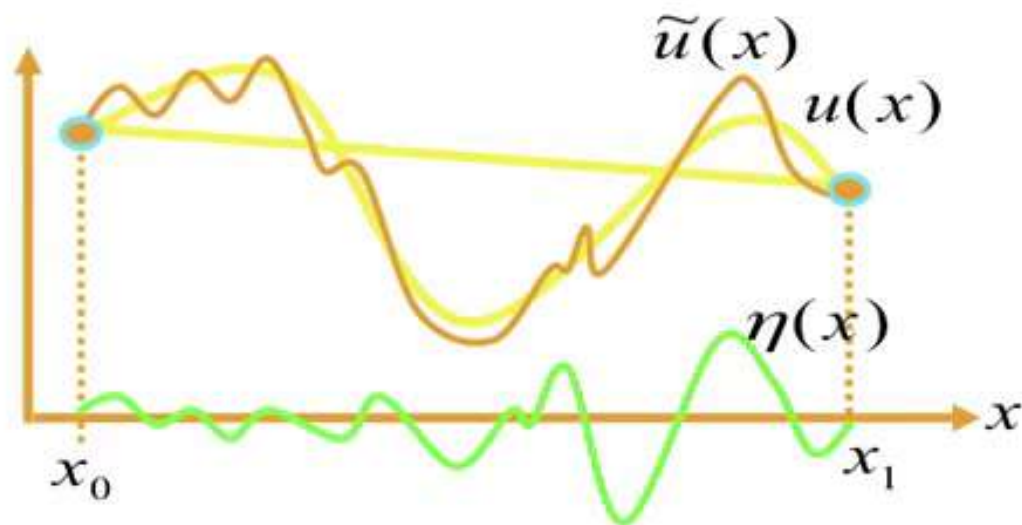
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Gradient descent process

$$u_t = - \frac{\delta E(u)}{\delta u}$$



# Conclusions

- Gradient descent process

Calculus  $\arg \min_x f(x)$

Calculus of variations  $\arg \min_{u(x)}$

$$\Rightarrow x_t = -f_x$$

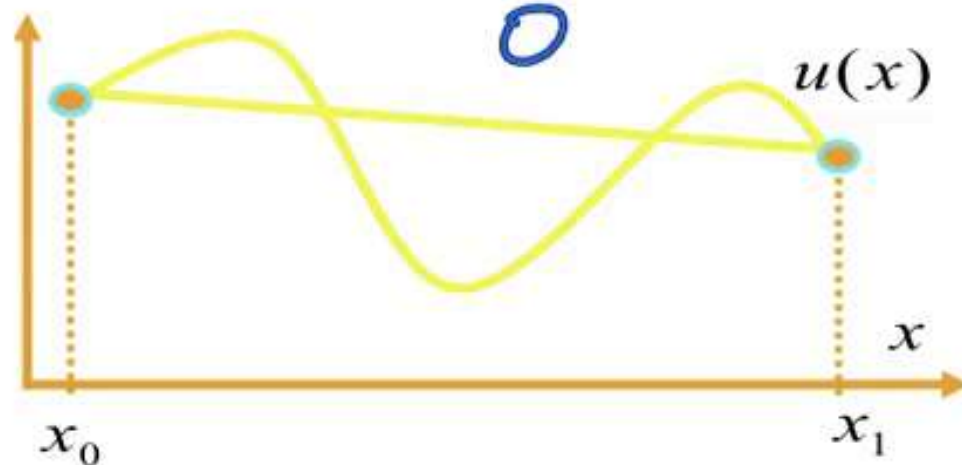
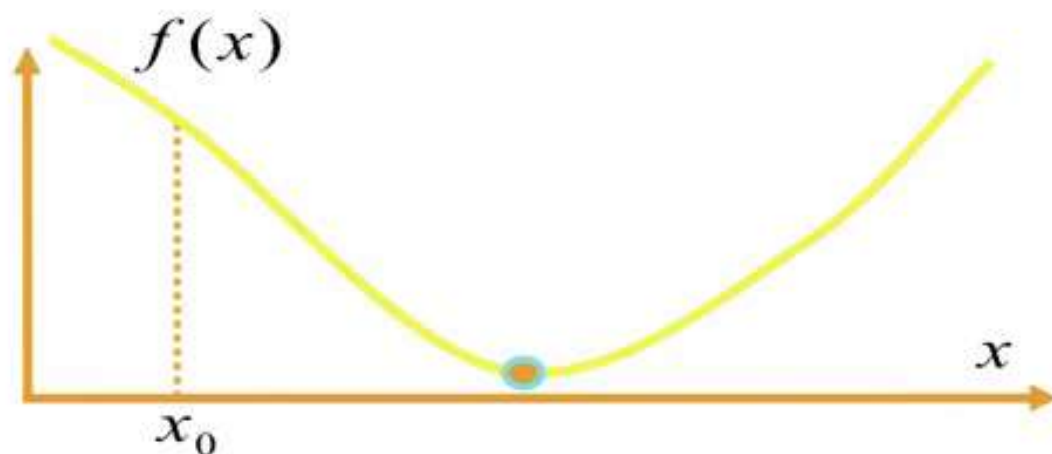
$$\arg \min_{u(x)} \int F(u, u_x) dx$$

$E(u)$

$$\Rightarrow$$

Euler-Lagrange

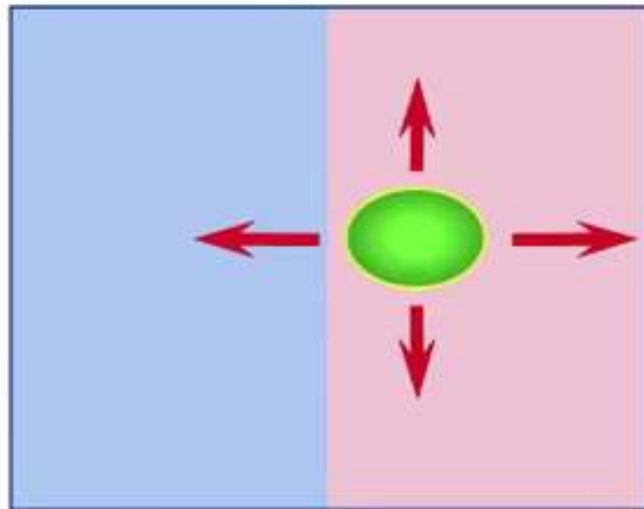
$$u_t = - \frac{\delta E(u)}{\delta u}$$



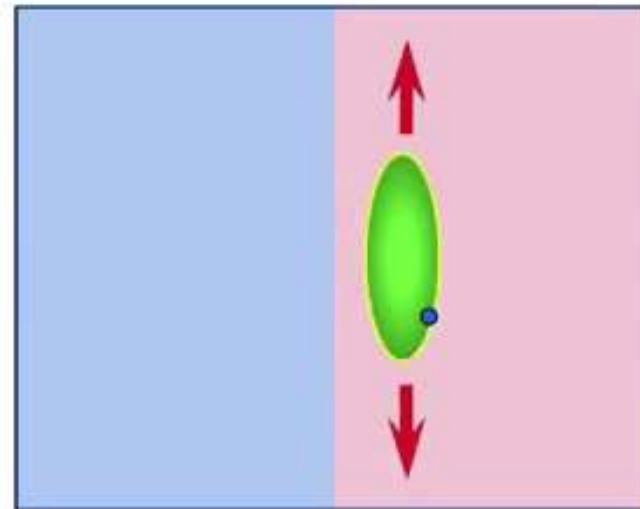


# Anisotropic diffusion

## Isotropic vs. Anisotropic Smoothing

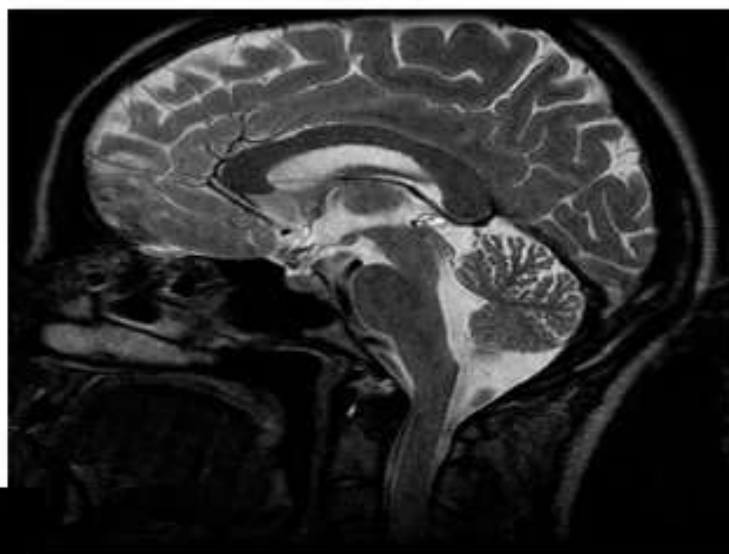


**Isotropic  
smoothing**

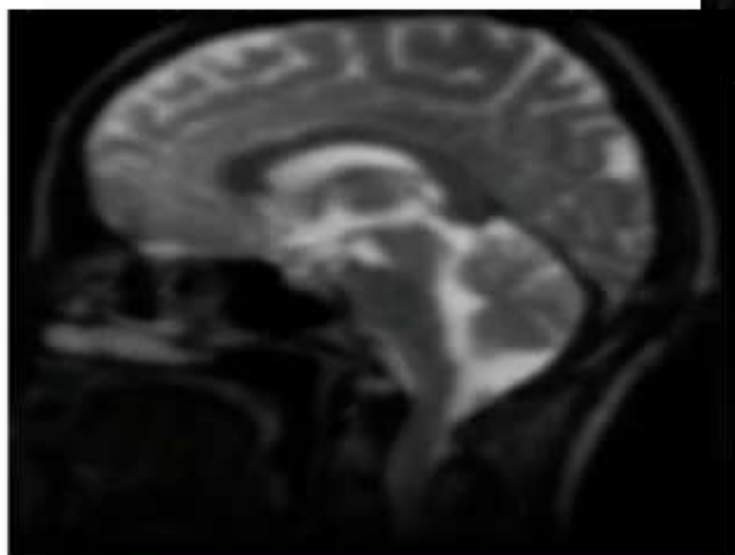


**Anisotropic  
smoothing**

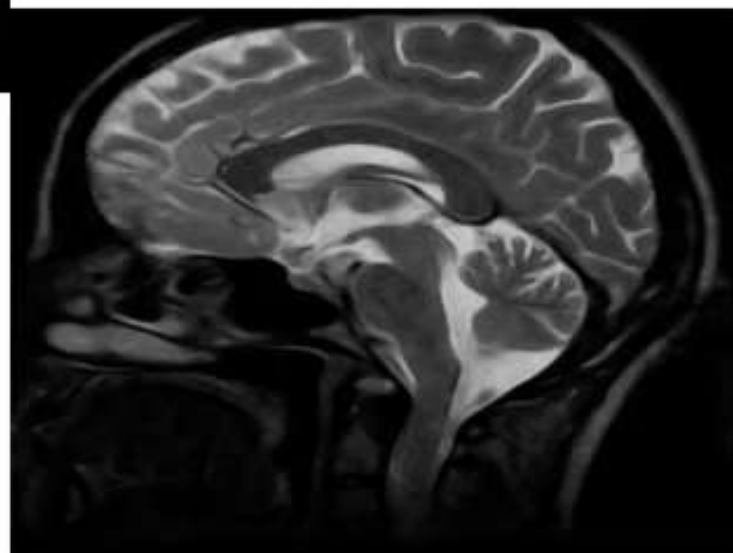
Isotropic  
(Heat equation)



Anisotropic



$$\Delta I = \text{div}(\nabla I)$$



$$\frac{\partial I(x,y,t)}{\partial t} = \Delta I$$

$$\frac{\partial I(x,y,t)}{\partial t} = \text{div}(g(|\nabla I|)\nabla I)$$

$$\min_I \int_{\Omega} \rho(|\nabla I|) d\Omega$$



$$\frac{\partial I(x,y,t)}{\partial t} = \operatorname{div} \left[ \rho' \frac{\nabla I}{|\nabla I|} \right]$$

$$\int F(u, u_x)$$



$$\min_I \int_{\Omega} \rho(|\nabla I|) d\Omega$$

$$\rho(a) = a^2$$

$$\int |\nabla I|^2$$



$$\frac{\partial I(x,y,t)}{\partial t} = \operatorname{div} \left[ \rho' \frac{\nabla I}{|\nabla I|} \right]$$

$$\rho' = 2a$$

$$I_t = \operatorname{div} \left( \overbrace{|\nabla I|}^{\rho'} \underbrace{\frac{\nabla I}{|\nabla I|}} \right)$$





$$\min_I \int_{\Omega} \rho(|\nabla I|) d\Omega$$



$$\frac{\partial I(x,y,t)}{\partial t} = \operatorname{div} \left[ \rho' \frac{\nabla I}{|\nabla I|} \right]$$

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$$\int |\nabla I|^2$$

$$\rho' = 2a$$

$$I_t = \operatorname{div} \left( \overbrace{|\nabla I|}^{\rho'} \frac{\nabla I}{\underbrace{|\nabla I|}} \right)$$

$$= \Delta I$$



$$\min_I \int_{\Omega} \rho(|\nabla I|) d\Omega$$



$$\frac{\partial I(x,y,t)}{\partial t} = \operatorname{div} \left[ \rho' \frac{\nabla I}{|\nabla I|} \right]$$

$\rho(a) = a \Rightarrow \rho' = 1$  Total Variation  
 $\int |\nabla I| \quad \swarrow \quad I_t = \operatorname{div} \left( \frac{\nabla I}{|\nabla I|} \right)$

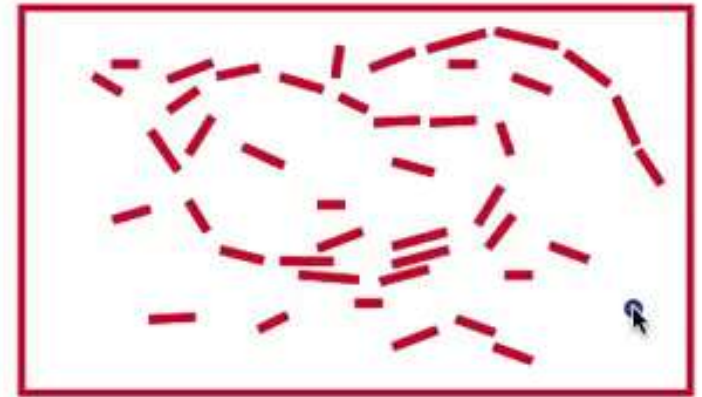


# Edge Detection



## Edge Detection:

- The process of labeling the locations in the image where the gray level's "rate of change" is high.
  - **OUTPUT:** "edgels" locations, direction, strength



## Edge Integration:

- The process of combining "local" and perhaps sparse and non-contiguous "edgel"-data into meaningful, long edge curves (or closed contours) for segmentation
  - **OUTPUT:** edges/curves consistent with the local data

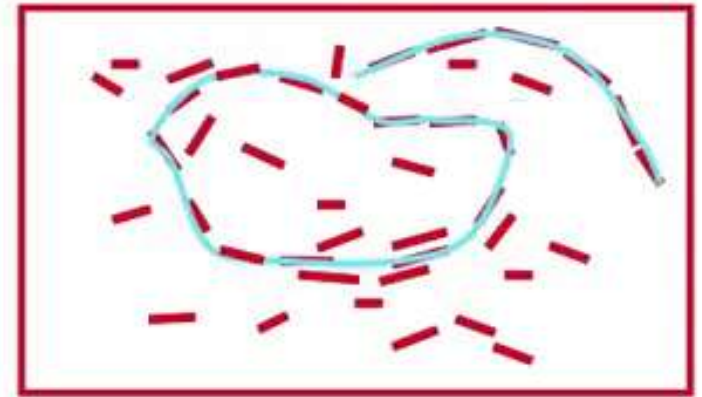


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# Active Contours



Image



Edge Indicator  
Function

$$g(x, y) = \frac{1}{1 + |\nabla(G_o * I)|^2}$$



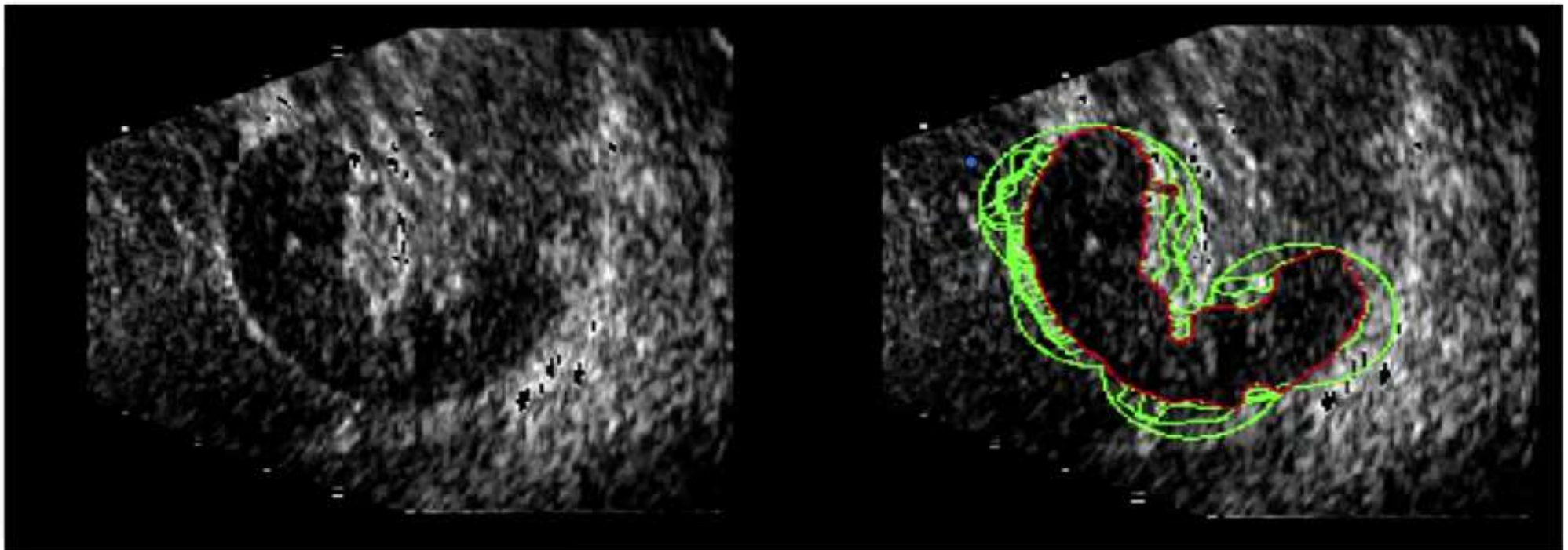
“nice” curves that optimize a functional of  $g()$ , i.e.

$$\int_{\text{curve}} g() ds$$

nice: “regularized”, smooth,  
fit some prior information

Edge Curves

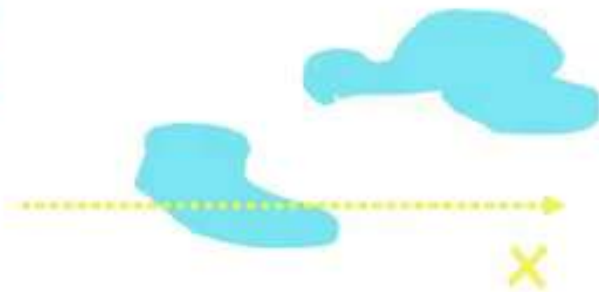
# Segmentation



# Potential Functions



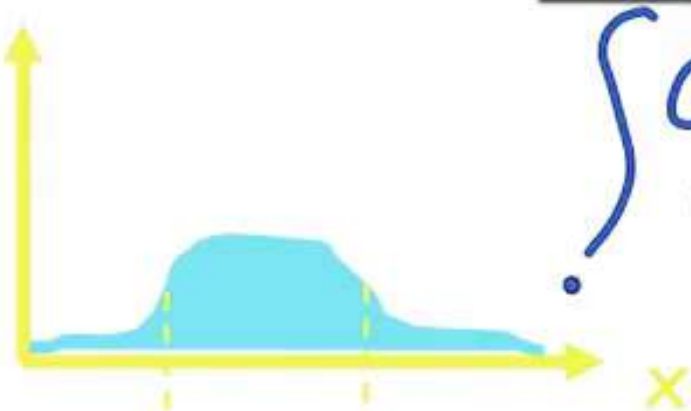
$I(x,y)$   
Image



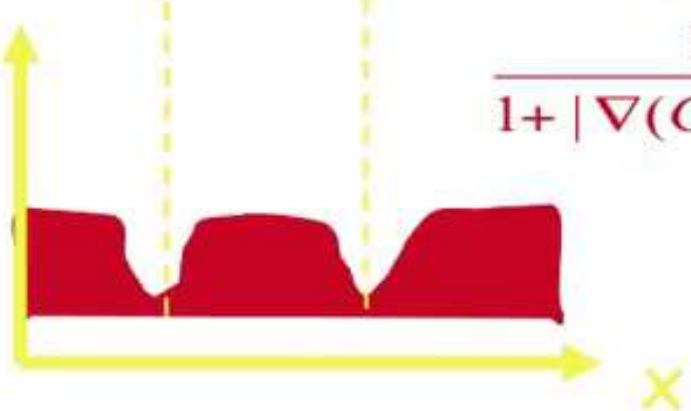
$g(x,y)$   
Edges



$I(x)$



$g(x)$



$\int g$

$$\frac{1}{1 + |\nabla(G_\sigma * I)|^2}$$

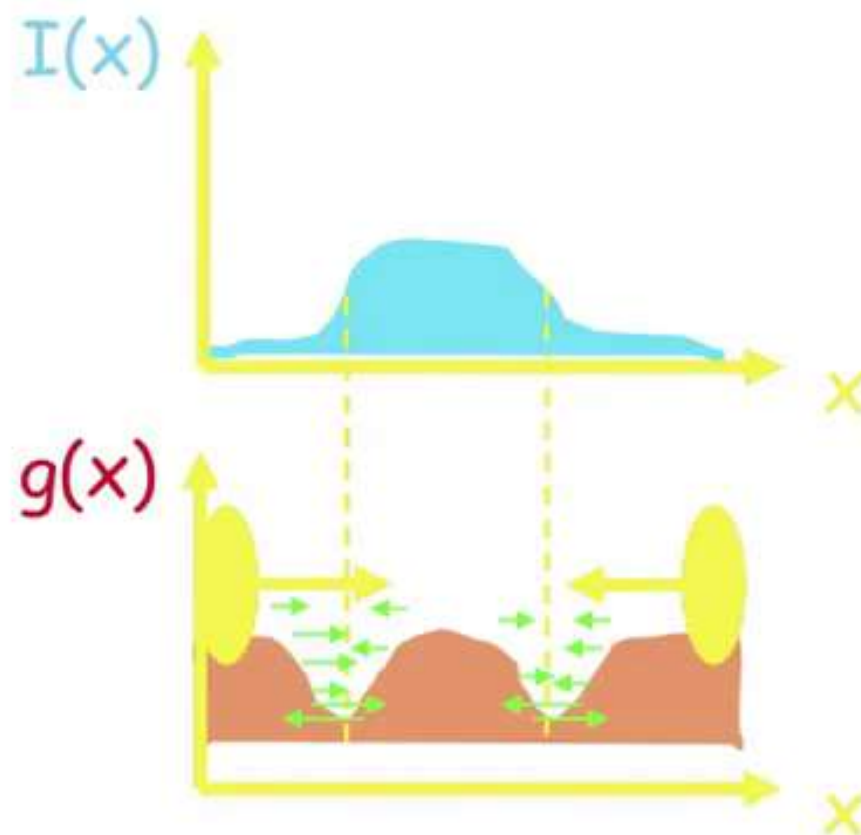


## Geodesic Active Contours in 1D

Geodesic active contours are  
reparameterization invariant

$$\frac{dC}{dt} = \left( g(C)\kappa - \langle \nabla g(C), \vec{N} \rangle \right) \vec{N}$$

$\nearrow$   
 $\int g$





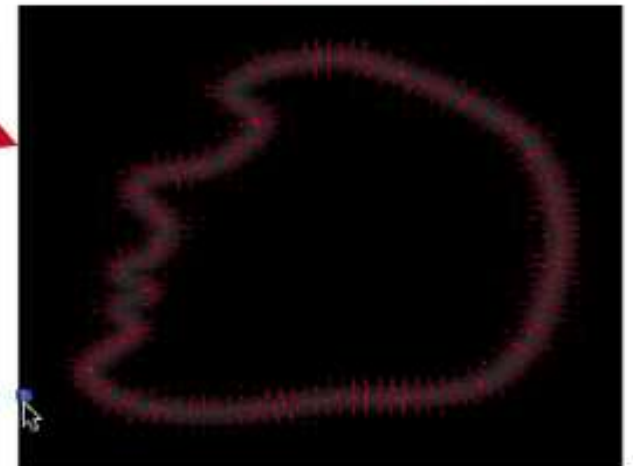
# Geodesic Active Contours in 2D



$G_s * I$



$$g(x) = \frac{1}{1 + |\nabla(G_s * I)|^2}$$



$$\frac{dC}{dt} = (g(C)\kappa - \langle \nabla g(C), \vec{N} \rangle) \vec{N}$$



## Geodesic Active Contours in 2D



$G_s * I$



$$g(x) = \frac{1}{1 + |\nabla(G_s * I)|^2}$$



$$\frac{dC}{dt} = (g(C)\kappa - \langle \nabla g(C), \vec{N} \rangle) \vec{N}$$









# Gray Matter Segmentation

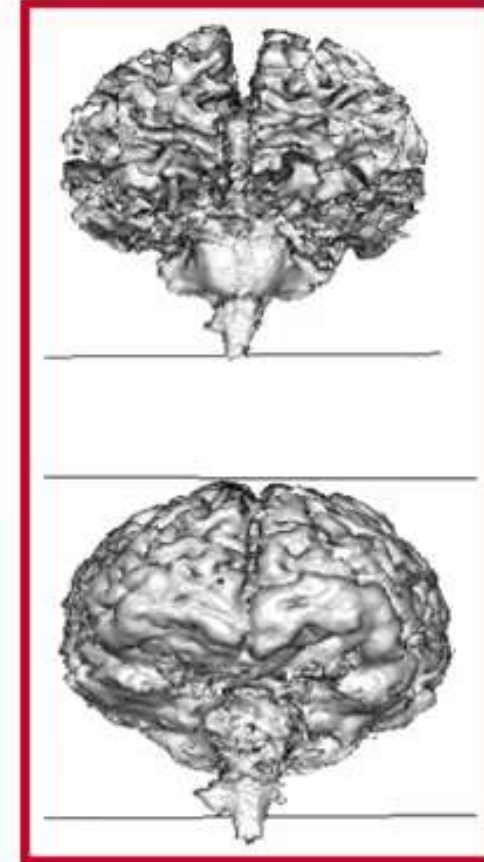
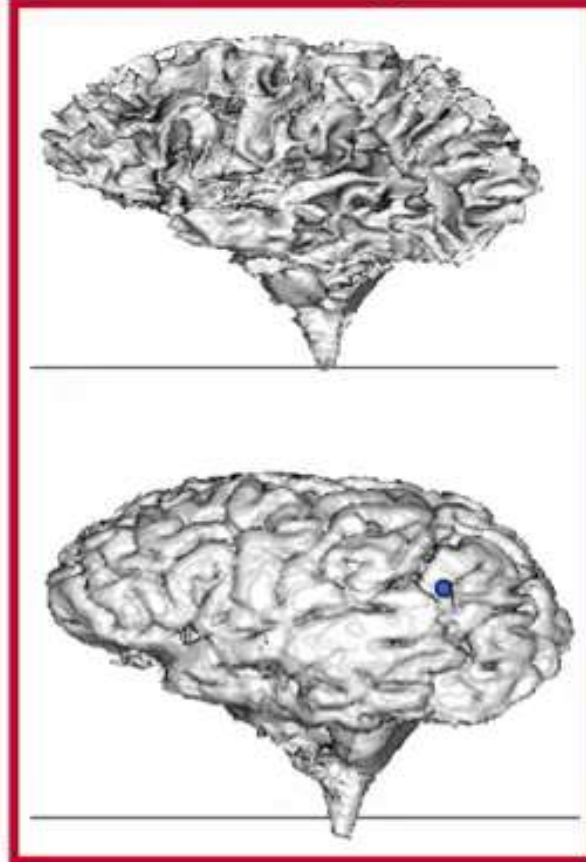
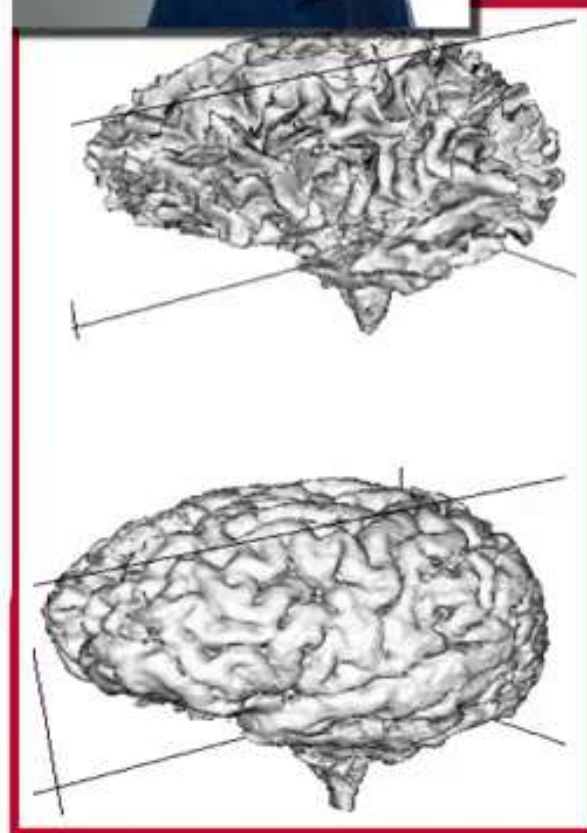
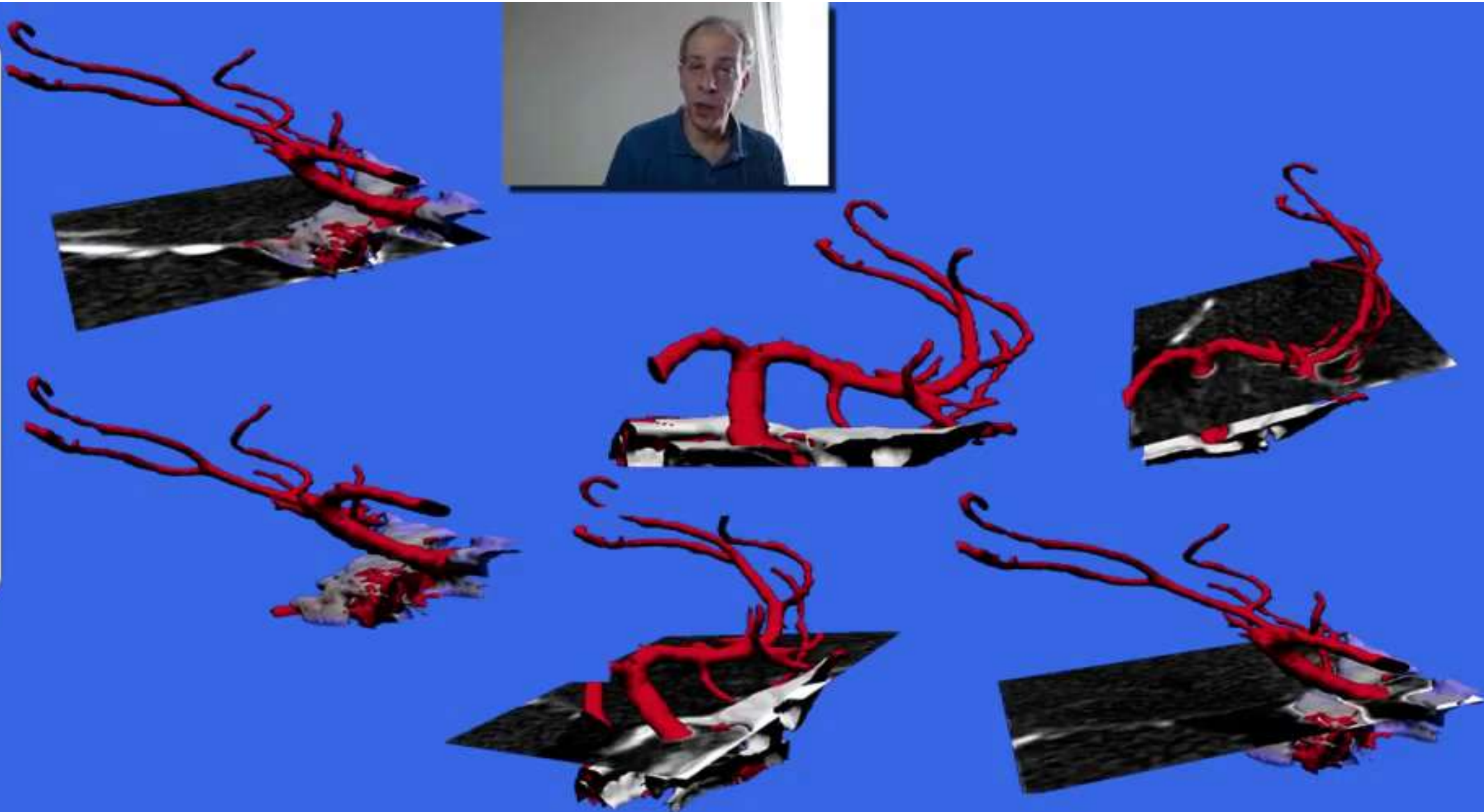


Image courtesy of Goldenberg Kimmel Rivlin Rudzsky,

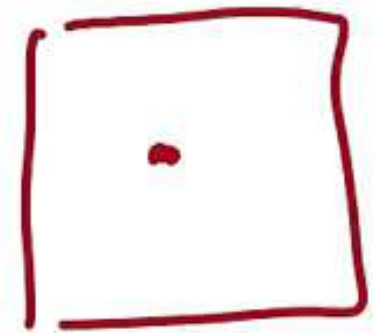
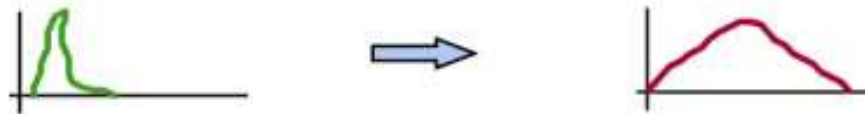


Images courtesy of Holzman-Gazit, Goldshier, Kimmel

# Contrast Enhancement

## □ Contrast enhancement via image deformations

- Approach: Histogram modification



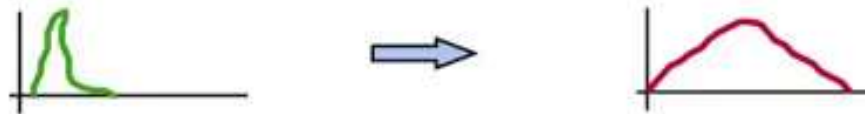
$$\bigcirc \quad \underline{\underline{=}} \quad \frac{\partial I(x,y)}{\partial t} = I(x,y) - (\# \text{pixels of value } \geq I(x,y) )$$

●

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- Approach: Histogram modification

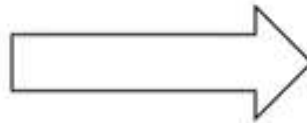
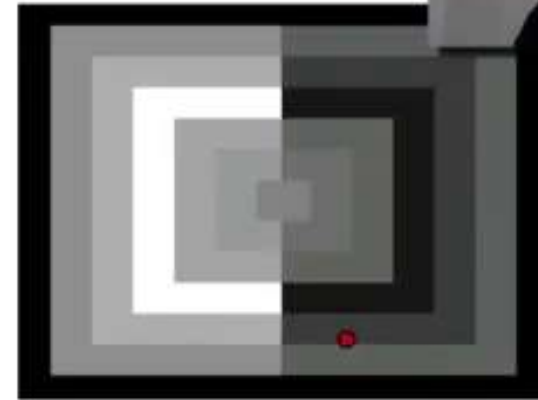
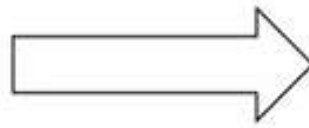
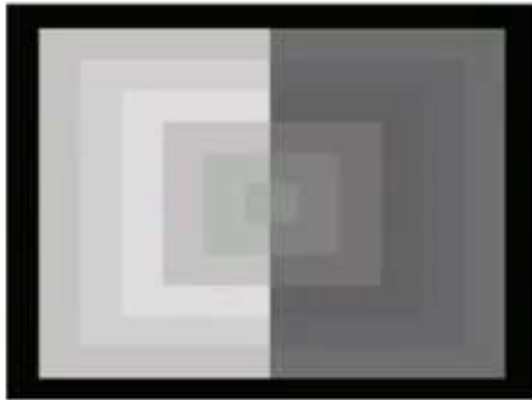


$$\frac{\partial I(x,y)}{\partial t} = I(x,y) - (\# \text{pixels of value } \geq I(x,y) )$$

$$U(I) = \frac{1}{2} \int [I(\vec{x}) - 1/2]^2 d\vec{x} - \frac{1}{4} \iint [I(\vec{x}) - I(\vec{z})] d\vec{x} d\vec{z}$$







• Images courtesy JDE and IEEE