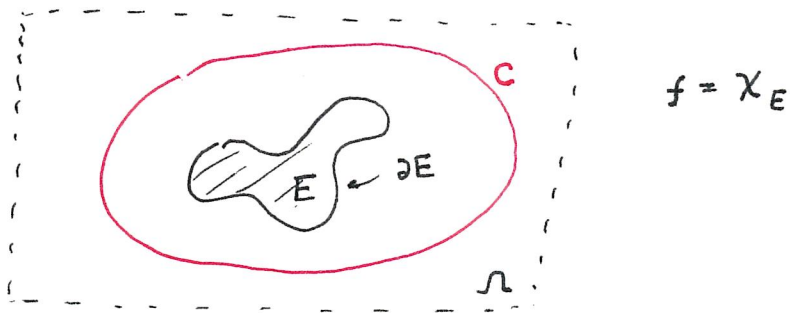


PCMI Lecture # 8

Curves Flows and Level set methods I

In this lecture we begin our exploration of curves and level sets in image data analysis. We will follow Chan & Vese 2001 quite closely.



Task: evolve C so that it coincides with ∂E

Snakes: Kass, Witkin and Terzopoulos 1988

$$C^* = \operatorname{argmin}_C F(C) = \alpha \int_0^1 |C'(s)|^2 ds + \beta \int_0^1 |C''(s)| ds - \lambda \int |\nabla f(C(s))|^2 ds$$

$$\alpha, \beta, \lambda > 0$$

First and second terms: regularity of C
Third term: seeks high gradient regions of f

First term: enforces uniform speed v

$$\int_0^1 v(t) dt = D \quad \int_0^1 (v(t) - D)^2 dt \geq 0$$
$$\Rightarrow \int_0^1 v^2(t) dt \geq \int_0^1 D^2 dt$$

... so choose $v(t) = D$

Second term: this is just $|D^2K|$ so this term encourages low curvature oscillation.

Third term: This encourages us to move the curve c to coincide with ∂E . (This term would, strictly speaking, force the curve to follow any discontinuity but in practice we do not have discontinuities since we would smooth the image a bit.)

Active Contour Method

Ingredient 1 Edge Detectors: Functions which go to 0 as the scale argument goes to infinity which we apply to the norm of the gradient.

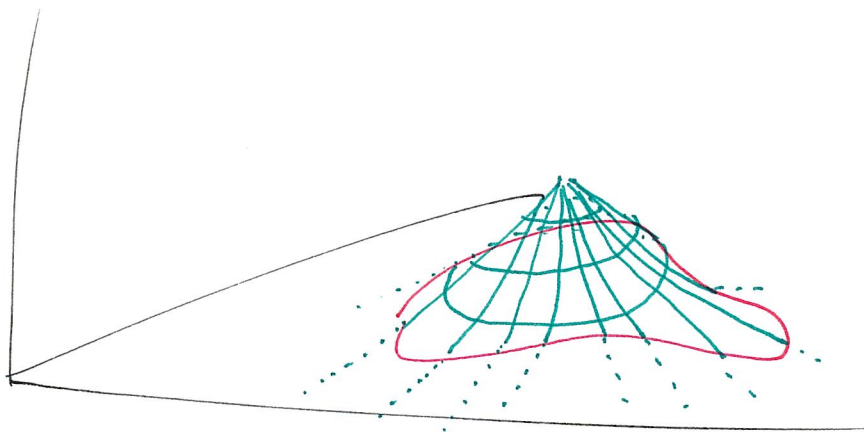
Example:
$$g(|\nabla f|) = \frac{1}{1 + |\nabla K_\sigma * f|^p} \quad p \geq 1$$

note: we have moved the gradient operator to the smoothing kernel K_σ , σ is the width of the smoothing kernel.

(g is of course going to be zero at edges since $\nabla f = \infty$ there.)

Ingredient 2 Level set method:

Idea: represent the curve C implicitly, as the zero level set of a function ϕ .



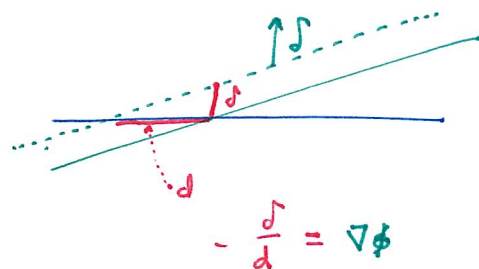
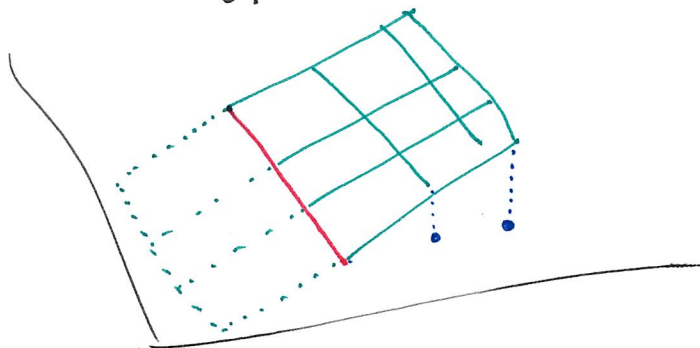
$$C = \{x \mid \phi(x) = 0\}$$

Suppose we want to move C with a vector field $\vec{V}(x)$.
We can do that very simply and implicitly.

$$\frac{\partial \phi}{\partial t} = -\nabla \phi \cdot \vec{V} \quad \text{or} \quad \frac{\partial \phi}{\partial t} + \nabla \phi \cdot \vec{V} = 0$$

note that if \vec{V} is a normal vector field then we can write

$$\frac{\partial \phi}{\partial t} + |\nabla \phi| |\vec{V}| = 0$$

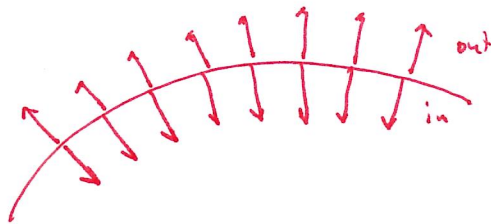


$$\Rightarrow \frac{d}{dt} \phi = 0$$

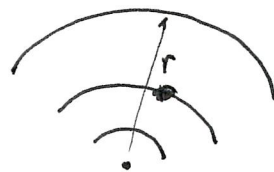
i.e. $\nabla \phi$ is the conversion factor needed to convert vertical motion to horizontal motion.

Ingredient 3

mean curvature



$$H = \nabla \cdot \frac{\nabla \phi}{|\nabla \phi|}$$



$$\phi = \sqrt{x^2 + y^2} = \text{distance to origin}$$

$$\begin{aligned} H &= \nabla \cdot \left(\frac{x}{\sqrt{x^2 + y^2}}, \frac{y}{\sqrt{x^2 + y^2}} \right) \\ &= \frac{\sqrt{x^2 + y^2} - x^2}{(x^2 + y^2)^{3/2}} + \frac{x^2 + y^2 - y^2}{(x^2 + y^2)^{3/2}} \\ &= \frac{1}{\sqrt{x^2 + y^2}} = \frac{1}{r} \end{aligned}$$

Like wise in \mathbb{R}^n we get

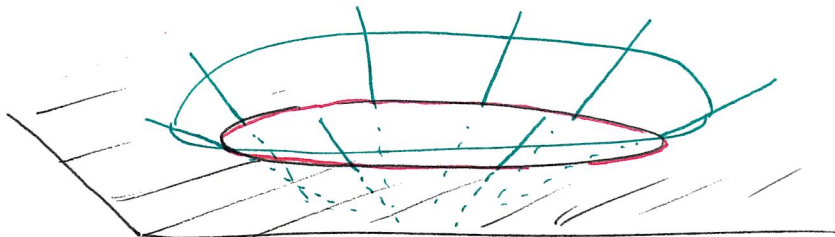
$$H = \nabla \cdot \left(\frac{r}{|r|} \right) = \frac{n-1}{r}$$

For the level set r from the origin.

Note: Some use $\phi < 0$ inside C others use $\phi > 0$ inside C . $\Rightarrow \frac{\nabla \phi}{|\nabla \phi|}$ pointing out, $\frac{\nabla \phi}{|\nabla \phi|}$ pointing in, respectively

Choosing $\phi > 0$ outside:

$$\frac{\partial \phi}{\partial t} = g(|\nabla f|) |\nabla \phi| \left(\nabla \cdot \frac{\nabla \phi}{|\nabla \phi|} + v \right)$$



in this case $\frac{\nabla \phi}{|\nabla \phi|}$ is pointing out and $\nabla \cdot \frac{\nabla \phi}{|\nabla \phi|}$ is positive for this curve C . This makes this particular curve shrink and will continue to shrink even if there is a bit of negative curvature as long as $v > 0$, until $g(|\nabla f|)$ stops it.

It is (I believe) a little more typical to choose $\phi > 0$ inside.

One other model: geodesic mean curvature of Cagelli, Kimmel, and Sapiro

$$F(C) = \int_0^1 |C'(s)| \cdot g(|\nabla f(C(s))|) ds$$



level set formula



$$F(C) = \int g(|\nabla f|) |\nabla \phi| dx$$



EL ...

$$\nabla \cdot \left(g(|\nabla f|) \frac{\nabla \phi}{|\nabla \phi|} \right)$$

(5)

To get propagation speeds invariant to ϕ , we move according to

$$\frac{\partial \phi}{\partial t} = |\nabla \phi| \left(\nabla \cdot \left(g(|\nabla \phi|) \frac{\nabla \phi}{|\nabla \phi|} \right) + v \cdot g(|\nabla \phi|) \right)$$

~~Chen-Vese~~

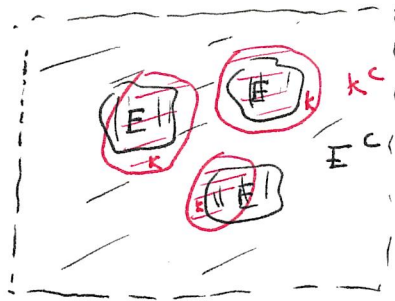
Chan-Vese: Active contours without edges

Inspiration here is the Mumford-Shah model

recall

$$F(u, \Gamma) = \alpha \int_{\Omega} |u - f|^2 + \beta \int_{\Omega \setminus \Gamma} |\nabla u|^2 + \gamma \mathcal{H}^{n-1}(\Gamma)$$

And they consider piecewise constant images...



$$f = \begin{cases} c_1^* & \text{on } E \\ c_2^* & \text{on } E^c \end{cases}$$

$$F_0(K, c_1, c_2) = \int_K (f - c_1)^2 dx + \int_{K^c} (f - c_2)^2$$

$K = E$, $c_1 = c_1^*$, $c_2 = c_2^*$ is the unique minimizing set.

The complete Chan-Vese model is:

$$F_{cv}(K, c_1, c_2) = \alpha H'(c) + \beta (\text{area inside } c) \\ + \gamma_1 \int_{\text{inside}(c)} |f - c_1|^2 dx \\ + \gamma_2 \int_{\text{outside}(c)} |f - c_2|^2 dx$$

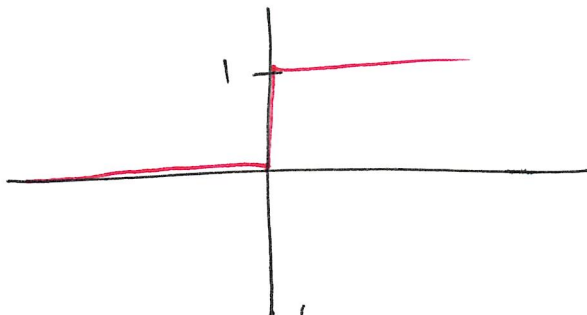
The Mumford-Shah model, specialized to the case of piecewise constant outputs, is

$$\alpha H'(c) + \gamma_1 \int_{\text{in}(c)} |f - \bar{f}_c|^2 + \gamma_2 \int_{\text{out}(c)} |f - \bar{f}_c|^2$$

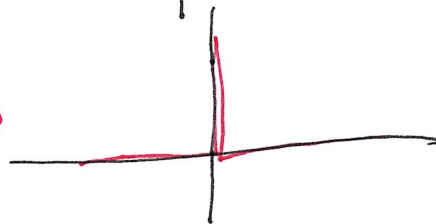
which is just F_{cv} with $\beta = 0$.

Level set formulation of the Chan-Vese model

Heaviside function $H(x)$



$$\frac{d}{dx} H(x) = \delta(x) \text{ (a measure)}$$



(We will, upon implementation, use regularized versions of H and δ , $H_\epsilon, \delta_\epsilon$ $\delta_\epsilon = \frac{d}{dx} H_\epsilon$)

Now we represent C by $\phi = 0$. $\phi > 0$ on the inside is the convention we use.

we note that

$$H'(C) = \int |\nabla \chi_k| dx$$

and $H(\phi) = \chi_k$, so we get

$$\begin{aligned} H'(C) &= \int |\nabla H(\phi)| \\ &= \int \delta(\phi) |\nabla \phi| \end{aligned}$$

$$\text{Area in } C = \int H(\phi) dx$$

and

$$\int (f - c_1)^2 H(\phi)$$

$$\int (f - c_2)^2 (1 - H(\phi))$$

are the last two terms we need.

$$\begin{aligned} \Rightarrow F_{CV}(\phi, c_1, c_2) &= \alpha \int \delta(\phi) |\nabla \phi| \\ &+ \beta \int H(\phi) dx \\ &+ \gamma_1 \int (f - c_1)^2 H(\phi) dx \\ &+ \gamma_2 \int (f - c_2)^2 (1 - H(\phi)) dx \end{aligned}$$

(8)

This makes sense by considering the regularized version, or a change of variables for the δ distribution.

Regularized version:
for smooth δ_ϵ , this is the chain rule.

change of variables

$$\int \delta(kx) f(x) dx$$

$$\begin{aligned} y &= kx \\ dy &= k dx \end{aligned}$$

$$\int \delta(y) f\left(\frac{y}{k}\right) \frac{dy}{k}$$

$$= \frac{f(0)}{k}$$

$$\Rightarrow \delta(kx) k = \delta(x)$$