

Gradient Descent for Variational Problems with Moving Curves and Surfaces

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The Problem

 Define gradient descent procedures to find extremals to variational problems of the form

$$E(\Gamma) = \int_{\Gamma} g(\mathbf{x}, \mathbf{n}(\mathbf{x})) d\sigma$$
,

over the space of sufficiently regular m-surfaces in \mathbf{R}^{m+1} , e.g.

- curves in images (1D in 2D)
- surfaces in space (2D in 3D)



Example 1 - Area

 Find surface with minimal area (Plateau's Problem)

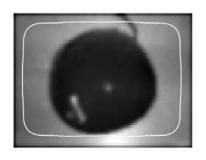
$$E(\Gamma) = \int_{\Gamma} 1 \, d\sigma$$

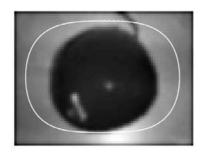
$$g = 1$$

Without boundary conditions:



Example 2 – Curve fitting









 Find a curve aligned to the image edges using Geodesic Active Contours

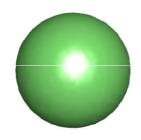
$$E(\Gamma) = \int_{\Gamma} g(|\nabla I|) \, d\sigma$$

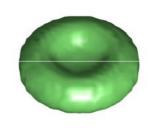


Example 3 – Surface reconstruction















Surface fitting to 3D data using a distance potential

$$E(\Gamma) = \int_{\Gamma} d(\mathbf{x}) \, d\sigma$$

[Zhao et al 2000]

And many, many more examples...



The Level Set Method

- Numerical methods for interface evolution
 [Osher & Sethian 1988]
 [Dervieux & Thomasset 1979]
- Represent a dynamic surface
 Γ implicitly
- Properties such as surface normal and mean curvature can easily be computed
- The surface is moved/deformed by solving a PDE on a fixed grid.

$$\Gamma(t) = \{ \mathbf{x} \in \Omega ; \ \phi(\mathbf{x}, t) = 0 \}$$

$$\phi(\mathbf{x},t) \begin{cases} < 0 & \text{inside } \Gamma \\ = 0 & \text{on } \Gamma \\ > 0 & \text{outside } \Gamma \end{cases}$$

$$\mathbf{n} = \frac{\nabla \phi}{|\nabla \phi|} \qquad \kappa = \nabla \cdot \frac{\nabla \phi}{|\nabla \phi|}$$

$$\frac{\partial \phi}{\partial t} + \mathbf{v} \cdot \nabla \phi = 0 \iff \frac{\partial \phi}{\partial t} + v_n |\nabla \phi| = 0$$



Variational Level Set Method

- Example: (Plateau's Problem) Minimal surfaces
- "Gradient descent" of functional gives surface motion (mean curvature flow)

$$E(\Gamma) = \int_{\Gamma} 1 dS \qquad \longleftrightarrow \qquad \frac{\partial \phi}{\partial t} = \kappa |\nabla \phi|$$



Deriving Gradient Descent

• A "common procedure"

$$E(\Gamma) = \int_{\Gamma} g(\mathbf{x}, \mathbf{n}(\mathbf{x})) d\sigma = \int_{\mathbf{R}^{m+1}} g(\mathbf{x}, \mathbf{n}(\mathbf{x})) |\nabla \phi| \delta(\phi) d\mathbf{x}$$

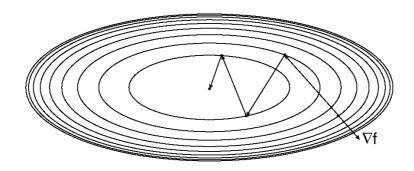
- Euler-Lagrange equation $G(\phi, \mathbf{x})\delta(\phi) = 0$
- Solve PDE until steady state

$$\frac{\partial \phi}{\partial t} = \pm G\delta(\phi)$$

$$\frac{\partial \phi}{\partial t} = \pm G|\nabla \phi|$$

- This is called the gradient descent evolution.
- Where is the gradient?
- What does the replacement of $\delta(\phi)$ mean?
- Can we give a geometric interpretation for this gradient descent procedure?

Gradient Descent in Rm



• Solve $\dot{\mathbf{x}}(t) = -\nabla f(\mathbf{x}(t)), \quad \mathbf{x}(0) = \mathbf{x}_0$ until a stationary point is reached. Here the gradient is

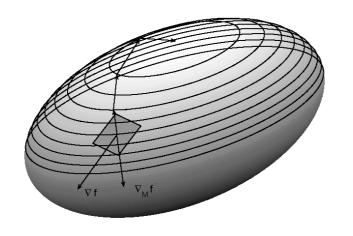
$$\nabla f = (\partial f/\partial x_1, \dots, \partial f/\partial x_m)$$

and

$$f'_{\mathbf{v}}(\mathbf{x}) = \langle \nabla f(\mathbf{x}), \mathbf{v} \rangle$$



Gradient Descent on Manifolds



Example:

$$M = S^{m} \quad \mathbf{n}(\mathbf{x}) = \mathbf{x}$$
$$\nabla_{S^{m}} f(\mathbf{x}) = \nabla \tilde{f} - \langle \mathbf{x}, \nabla \tilde{f} \rangle_{\mathbf{x}} \mathbf{x}$$

- Solve $\dot{\mathbf{x}}(t) = -\nabla_M f(\mathbf{x}(t)), \quad \mathbf{x}(0) = \mathbf{x}_0$ until a stationary point on the manifold is reached.
- The gradient is defined as the unique vector $\nabla_M f(\mathbf{x}) \in T_{\mathbf{x}} M$ such that $df(\mathbf{x})\mathbf{v} = \langle \nabla_M f(\mathbf{x}), \mathbf{v} \rangle_{\mathbf{x}}$

where

$$df(\mathbf{x})\mathbf{v} = \frac{d}{dt}f(\alpha(t))\Big|_{t=0}$$
$$\langle \mathbf{v}, \mathbf{w} \rangle_{\mathbf{x}} = \mathbf{v} \cdot \mathbf{w} \text{ for } \mathbf{v}, \mathbf{w} \in T_{\mathbf{x}}M$$



The Ingredients

- 1. Tangent space at the point x: $T_{\mathbf{x}}M$
- 2. Scalar product on the tangent space
- 3. Differential on M df(x)v

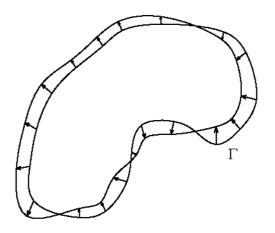
These are the necessary ingredients to define a gradient descent procedure. Can we do this for variational problems like

$$\min_{\Gamma} E(\Gamma) = \int_{\Gamma} g(\mathbf{x}, \mathbf{n}(\mathbf{x})) d\sigma ,$$

???



The Manifold of Admissible m-Surfaces



- Let Γ₀ be an m-dimensional surface in R^{m+1}
- The manifold M is defined as the set of surfaces which can be obtained by a regular evolution $t \mapsto \Gamma(t)$ from Γ_0
- Each surface is a "point" on M
- Need to find the "ingredients": tangent space at the point Γ_0 , and a scalar product.

Normal Velocity

• For a differentiable curve $\alpha: (-\epsilon, \epsilon) \to \mathbf{R}^{m+1}$ such that α belongs to Γ for all t, we have the normal velocity for the evolution $\Gamma(t)$

$$\dot{\Gamma}(t, \mathbf{x}) = \dot{\alpha}(0) \cdot \mathbf{n}(\mathbf{x}_0) = -\frac{\partial \phi(\mathbf{x}_0, 0) / \partial t}{|\nabla \phi(\mathbf{x}_0, 0)|}$$

- This is a continuous function on Γ_0 which can be interpreted as a tangent vector to M.
- The surface evolution is then determined by normal velocities $v \in T_{\Gamma}M$
- In particular: $C^2(\Gamma_0) \subset T_{\Gamma_0}M \subset C(\Gamma_0)$



The Gradient

• The differential at Γ_0 , $dE(\Gamma_0)$: $T_{\Gamma_0}M \to \mathbf{R}$, is defined as

$$dE(\Gamma_0)v = \frac{d}{dt}E(\Gamma(t))\Big|_{t=0} \qquad \dot{\Gamma}(0) = v$$

• Introduce scalar product on $T_{\Gamma}M \subset L^2(\Gamma)$

$$\langle v, w \rangle_{\Gamma} = \int_{\Gamma} v(\mathbf{x}) w(\mathbf{x}) \, d\sigma$$

- By Riesz' lemma $dE(\Gamma)v = \langle w,v \rangle_{\Gamma}$ $w = \nabla_M E(\Gamma)$
- Compare:

$$f'_{\mathbf{v}}(\mathbf{x}) = \langle \nabla f(\mathbf{x}), \mathbf{v} \rangle$$
 $df(\mathbf{x})\mathbf{v} = \langle \nabla_M f(\mathbf{x}), \mathbf{v} \rangle_{\mathbf{x}}$ $dE(\Gamma)v = \langle \nabla_M E(\Gamma), v \rangle_{\Gamma}$

Gradient descent

$$\dot{\Gamma}(t) = -\nabla_M E(\Gamma(t))$$



Back to the Problem

• Functional: $E(\Gamma) = \int_{\Gamma} g(\mathbf{x}, \mathbf{n}(\mathbf{x})) d\sigma$

Theorem: The functional has differential

$$dE(\Gamma)v = \langle \nabla \cdot [g_{\mathbf{n}} + g_{\mathbf{n}}], v \rangle_{\Gamma}$$

$$\nabla_M E = \nabla \cdot [g_{\mathbf{n}} + g_{\mathbf{n}}] \qquad g_{\mathbf{n}} = \nabla_{S^m} g$$

Gradient descent is then

$$\dot{\Gamma} = -\nabla \cdot [g_{\mathbf{n}} + g_{\mathbf{n}}]$$



Examples Again

Area

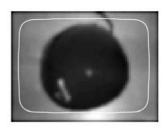
$$E(\Gamma) = \int_{\Gamma} 1 \, d\sigma$$
$$\nabla_M E = \kappa$$



Geodesic Active Contours

$$E(\Gamma) = \int_{\Gamma} f(|\nabla I|) \, d\sigma$$

$$\nabla_M E = f(|\nabla I|) \kappa + \nabla f(|\nabla I|) \cdot \mathbf{n}$$



Surface potential

$$E(\Gamma) = \int_{\Gamma} d(\mathbf{x}) \, d\sigma$$

$$\nabla_M E = g(\mathbf{x})\kappa + \nabla g(\mathbf{x}) \cdot \mathbf{n}$$







Volume functionals

Same analysis for volume (area) functionals

$$E(\Gamma) = \int_{\Omega^{-}} g(\mathbf{x}) d\mathbf{x}$$
 $\Omega^{-} = int(\Gamma)$

The functional has differential

$$dE(\Gamma)v = \langle g(\mathbf{x}), v \rangle_{\Gamma}$$
$$\nabla_M E = g(\mathbf{x})$$

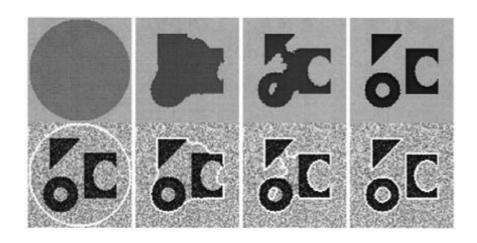


Example - Segmentation

Chan-Vese model:

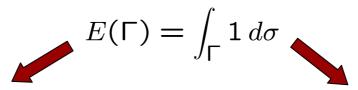
$$E(\Gamma) = \int_{\Omega^{-}} (I - I_0)^2 d\mathbf{x} + \int_{\Omega^{+}} (I - I_1)^2 d\mathbf{x}$$

$$\nabla_M E = (I - I_0)^2 - (I - I_1)^2$$





Connection to Euler-Lagrange



E-L approach

E-L equation

$$\kappa\delta(\phi)=0$$

$$\frac{\partial \phi}{\partial t} = \pm \kappa \delta(\phi)$$

Find sign & extend

Gradient interpretation

- Differential $dE(\Gamma)v = \langle \kappa, v \rangle_{\Gamma}$
- Gradient $\nabla_M E = \kappa$
- Gradient descent

$$\dot{\Gamma} = -\nabla_M E \qquad \dot{\Gamma} = -\frac{\partial \phi/\partial t}{|\nabla \phi|}$$



$$\frac{\partial \phi}{\partial t} = \kappa |\nabla \phi|$$





Gradient Projection for Variational Surface Problems with Constraints

Functional

$$E(\Gamma) = \int_{\Gamma} g(\mathbf{x}, \mathbf{n}(\mathbf{x})) \, d\sigma$$

with constraints in "infinite" form, defined by a functional

$$G(\Gamma) = 0$$

Constraint (sub-manifold)

$$N = \{ \Gamma \in M : G(\Gamma) = 0 \} ,$$

This gives a gradient on the constraint manifold as

$$\nabla_N E(\Gamma) := \nabla E(\Gamma) - \frac{\langle \nabla E(\Gamma), \nabla G(\Gamma) \rangle_{\Gamma}}{\|\nabla G(\Gamma)\|_{\Gamma}^2} \nabla G(\Gamma)$$

Note: this extends naturally to multiple constraints!



Constrained Evolution - Lagrange Functional

• Find extremals (Γ, λ) to the Lagrange functional $E(\Gamma) - \lambda G(\Gamma)$ using gradient descent. This gives

$$\frac{\partial \phi}{\partial t} = (\nabla E - \lambda \nabla G) |\nabla \phi|$$

where the Lagrange multiplier λ is given by

$$\lambda = \frac{\int_{\Gamma} \nabla E \nabla G \, d\sigma}{\int_{\Gamma} (\nabla G)^2 \, d\sigma}$$

Note: this is a orthogonal projection on the constraint manifold.



Example

Problem:

Incorporate visibility constraints in variational surface fitting procedures.

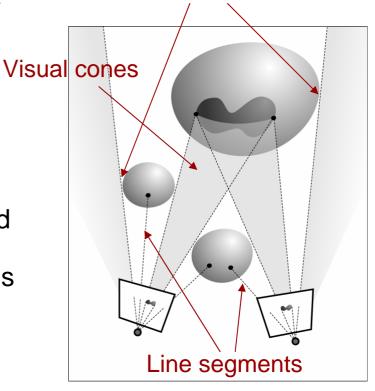
Solution:

- Define forbidden regions using signed distance functions
- Define constraints using these regions
- Use constrained gradient descent to evolve the surface

Why?

- Visibility is an important cue
- Faster convergence, correct topology, avoid oscillatory evolution, always obey visibility during evolution.

Bounding silhouettes





Visibility Constraints

• Group all forbidden regions in a set $W \subset \mathbf{R}^3$ defined using a signed distance function

$$W = \{ \mathbf{x} ; w(\mathbf{x}) \ge 0 \} \ \partial W = \{ w = 0 \}$$

Define the constraint using a functional G

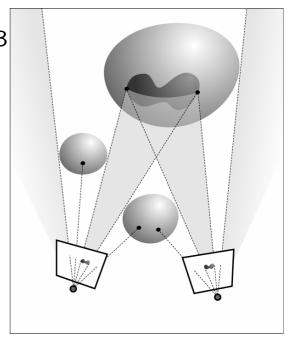
$$G(\Gamma, W) = \int_{\Omega^- \cap W} d\mathbf{x} = 0$$

The gradient of the constraint is

$$\nabla G = H(w)$$

where H(w) is the Heaviside function, and the constrained surface evolution is

$$\frac{\partial \phi}{\partial t} = (\nabla E - \lambda H(w)) |\nabla \phi| \qquad \lambda = \frac{\int_{\Gamma} \nabla E H(w) \, d\sigma}{\int_{\Gamma} (H(w))^2 \, d\sigma}$$





Initialization

- Need initial surface that satisfies $G(\Gamma, W) = 0$
- This is easy to obtain using Boolean operations on the functions

since
$$\Omega^- \setminus W = \Omega^- \cap W^c = \{ \mathbf{x} ; \max(\phi(\mathbf{x}), w(\mathbf{x})) \leq 0 \}$$

• Given any initial function, this can be modified to satisfy the constraint by setting $\phi(\mathbf{x}) = \max(\phi(\mathbf{x}), w(\mathbf{x}))$

Example – pitcher:





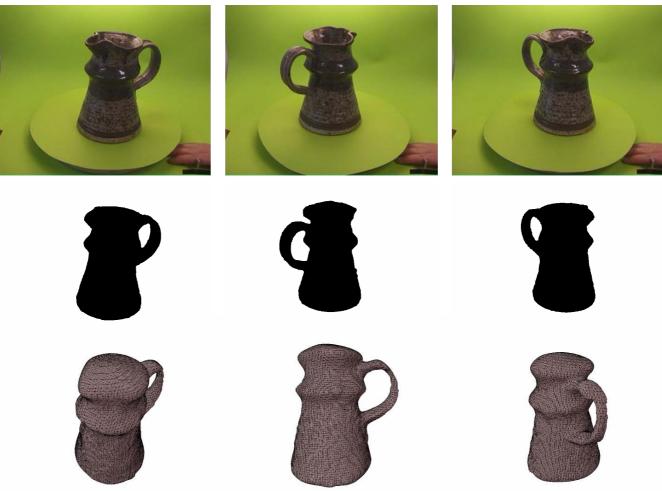




silhouettes

convex hull

lines





Summary

- We introduced:
 - manifold of admissible m-surfaces
 - tangent space
 - scalar product
 - gradient
- Based on geometric quantities such as normal velocity and gradient.
- This means that the theory is valid
 - in any number of dimensions
 - for any surface representation
- References
 - Solem & Overgaard, Scale Space 2005
 - Solem & Overgaard, VLSM 2005



Thank you!

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