

Semi-lagrangian schemes for first and second order Hamilton-Jacobi equations

joint work with M.Falcone, R.Ferretti, G.Russo

E.Carlini

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APPLICATION FOR HJ EQUATION

- **Image Analysis** (contour detection, shape from shading, ...)
- **Front Propagation** (level set method)
- **Optimal Control Process**

EXISTENCE and UNICITY

in the class of **viscosity solution**:

- Y.G. Chen, Y. Giga, S. Goto, *Uniqueness and existence of viscosity solutions of generalized mean curvature flow equations*, Journal Diff. Geom., **33** (1991), 749-786.
- M.G. Crandall, H. Ishii, P.L. Lions, *User's guide to viscosity solutions of second order partial differential equations* Bull. Amer. Math. Soc., **27** (1992), 1-67.
- L.C. Evans, J. Spruck, *Motion of level sets by mean curvature motion I*, Journal Diff. Geom., **33** (1991), 635-681.

OUTLINE

first order: $u_t + H(Du) = 0$

- SL scheme
- High order scheme with Weno reconstruction
- Numerical Test

second order: $u_t + H(D^2u, Du) = 0$

- SL scheme for Mean Curvature Motion codimension 1
- SL scheme for Mean Curvature Motion codimension 2
- Numerical Test

FIRST ORDER EQUATION

$$\begin{cases} u_t(x, t) + H(\nabla u(x, t)) = 0 & x \in \mathbb{R}^N, t \in [0, T] \\ u(x, 0) = u_0(x) \end{cases}$$

Legendre Transform : $H^*(\alpha) = \sup_{p \in \mathbb{R}^N} \{\alpha \cdot p - H(p)\} \quad \alpha \in \mathbb{R}^N.$

$$p \rightarrow H(p) \quad \text{convex} \Rightarrow H^{**} = H$$

$$H(\nabla u) = \sup_{\alpha \in \mathbb{R}^N} \{\alpha \cdot \nabla u - H^*(\alpha)\}$$

Hopf-Lax formula :

$$u(x, t) = \inf_{\alpha \in \mathbb{R}^N} \{t H^*(\alpha) + u_0(x - \alpha t)\}$$

NUMERICAL SOLUTION

Dynamic Programming Principle

$$u(x, t + \Delta t) = \inf_{\alpha \in \mathbb{R}^N} \{ \Delta t H^*(\alpha) + u(x - \alpha \Delta t, t) \}$$

$$\text{SL scheme} \quad \begin{cases} u_i^{n+1} = \min_{a \in \mathbb{R}^N} \{ \Delta t H^*(a) + I[u^n](x_i - \Delta t a) \} \\ u_i^0 = u(x_i, 0). \end{cases}$$

We need to compute:

- $I[u^n](x_i - \Delta t a)$ by an INTERPOLATION procedure \Rightarrow **WENO**
- Legendre transform $H^*(a)$
- minimum for $a \in \mathbb{R}^N$ by PRAXIS algorithm (ftp.netlib.org)

TEST 1: two dimensional circle collapse

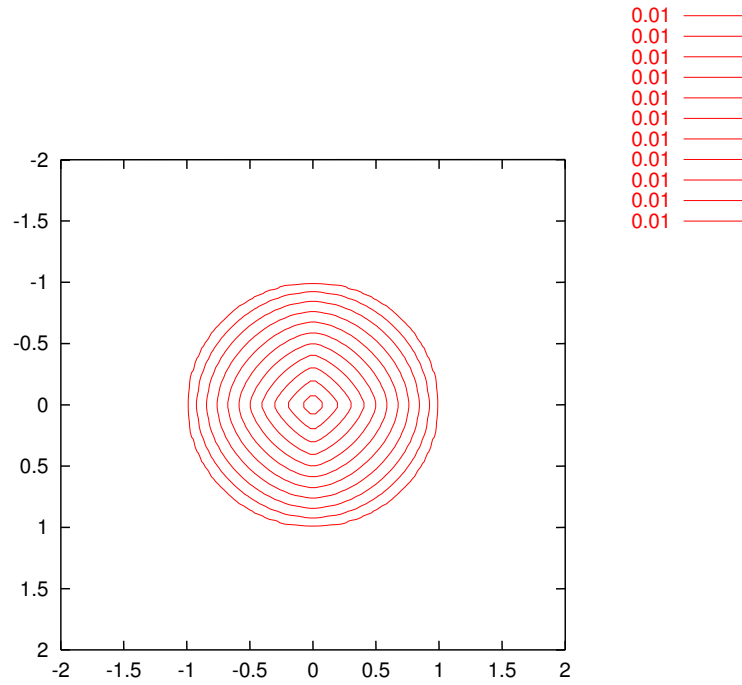
$$\begin{cases} v_t(x, t) + |Dv(x, t)| = 0 \\ v(x, 0) = \max(1 - |x|, 0), \end{cases}$$

OSHER-SETHIAN scheme					
$n.it$	Δx	Δt	relative L^∞ -order	relative L^1 -error	CPU time
10	0.08	0.01	$3.35 \cdot 10^{-2}$	$2.74 \cdot 10^{-2}$	$< 1s$
2	0.08	0.05	$2.14 \cdot 10^{-2}$	$1.63 \cdot 10^{-2}$	$< 1s$
50	0.04	0.02	$1.85 \cdot 10^{-2}$	0.38	

Semi-Lagrangian scheme					
$n.it$	Δx	Δt	relative L^∞ -order	relative L^1 -error	CPU time
10	0.08	0.01	$3.19 \cdot 10^{-2}$	$2.61 \cdot 10^{-2}$	25.69s
2	0.08	0.05	$2.00 \cdot 10^{-2}$	$1.00 \cdot 10^{-2}$	2.48s
10	0.04	0.1	$4.57 \cdot 10^{-3}$	$1.88 \cdot 10^{-2}$	

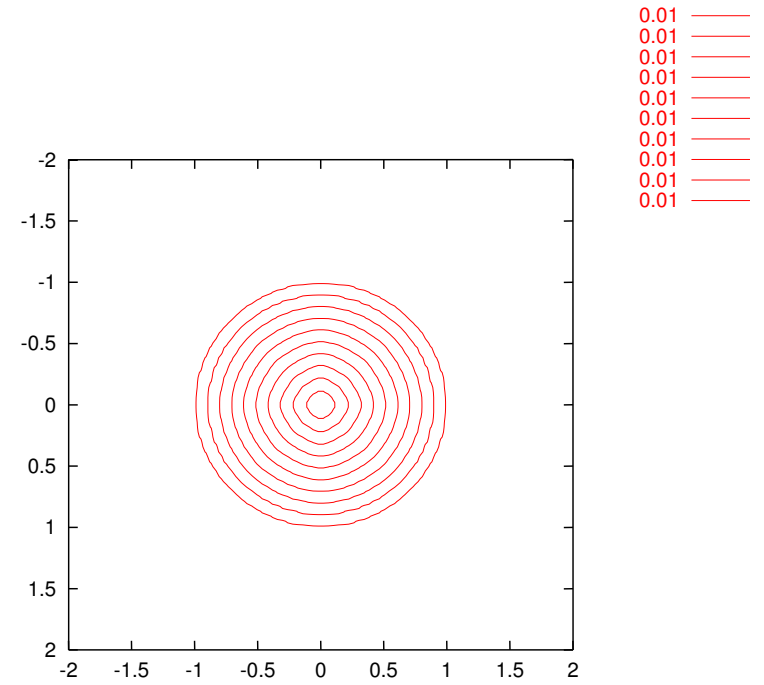
Osher Sethian scheme

Osher Sethian scheme



SL scheme

SL scheme



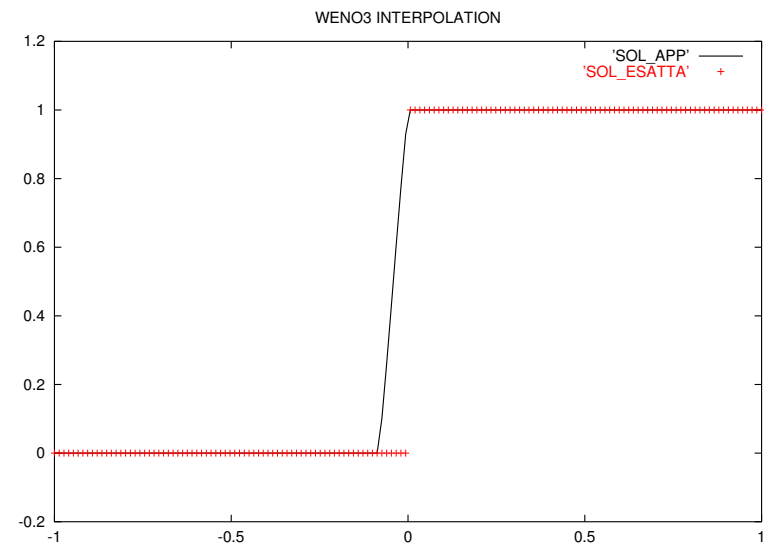
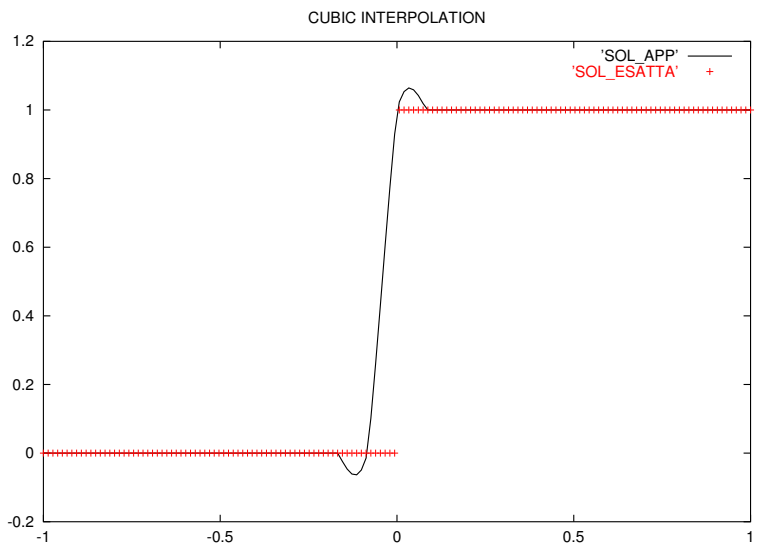
WENO RECONSTRUCTION

Idea: convex combination of lower order reconstruction.

$$I[V](x) = \sum_{k=1}^n \omega_k P_k(x).$$

P_k are lower order Lagrange polynomials interpolating V

w_k are non linear weights, which measure how smooth the functions P_k are in x



ENO-WENO REFERENCES

S.Osher, C.W.Shu *High order essentially non oscillatory schemes for Hamilton Jacobi equation*, SIAM J.Num.Anal., 28, (1991), 907-922.

X.D.Liu, S.Osher, T.Chan *Weighted essentially non oscillatory schemes*, J. of Computational Physics, V115, (1994), 220-212.

C.Jiang, C.W.Shu *Efficient Implementation of WENO Schemes*, J. of Computational Physics, V126, (1996), 202-228.

S.Bryson, D.Levy *Central Schemes for Multi-Dimensional Hamilton Jacobi Equations*, submitted to SICS.

$2n - 1$ WENO RECONSTRUCTION $I[V](x)$

Assume we have an interval 1D with a mesh structure,
we'd like to reconstruct V in x

- V is known on the mesh nodes

- $x \in I_j = [x_j, x_{j+1}]$

1. $S_k = \{x_{j-n+k}, \dots, x_{j+k}\}$ $k = 1, \dots, n$,
 n stencils s.t. x belong to each one

2. $\Gamma = \bigcup_{k=1}^n S_k$

3. $P_k(x)$ polynomials of degree n interpolating V on S_k

4. $Q(x)$ higher order $2n - 1$ polynomial on the larger stencil Γ

5. $C_k(x)$ linear weight s.t. $Q(x) = \sum_{k=1}^n C_k(x)P_k(x)$

6. β_k **SMOOTHNESS INDICATOR** ,

$$\beta_k = \sum_{1 \leq l \leq n} \int_{I_j} h^{2l-1} (D^l P_k(x))^2 dx \quad k = 1, \dots, n \quad h \text{ size of } I_j$$

7. ω_k **NON LINEAR WEIGHT**

$$\omega_k = \frac{\tilde{\omega}_k}{\sum_{k=1}^n \tilde{\omega}_k}, \quad \tilde{\omega}_k = \frac{C_k}{(\epsilon + \beta_k)^2}$$

The final WENO reconstruction is

$$I[V](x) = \sum_{k=1}^n \omega_k P_k(x).$$

POSITIVITY of the LINEAR WEIGHTS $C_k(x)$

$$Q(x) = \sum_{k=1}^n C_k(x) P_k(x) \quad (1)$$

should interpolate V on the larger stencil $\Gamma \Rightarrow$

$$C_k(x) = \gamma_k \prod_{x_l \in \Gamma \setminus S_k} (x - x_l) \quad (2)$$

with γ_k unknowns to be determined.

Using (2) in (1):

$$\begin{aligned} C_1(x_{j-n+1}) &= 1 \\ C_1(x_{j-n+2}) + C_2(x_{j-n+2}) &= 1 \\ &\vdots \\ \sum_{i=1}^k C_i(x_{j-n+k}) &= 1 \quad k = 1, \dots, n \end{aligned}$$

More explicitly we obtain the relation

$$\sum_{i=1}^k \gamma_i \prod_{x_l \in \Gamma \setminus S_k} (x - x_l) = 1 \quad k = 1, \dots, n$$

Using some tools from combinatorics we proved that

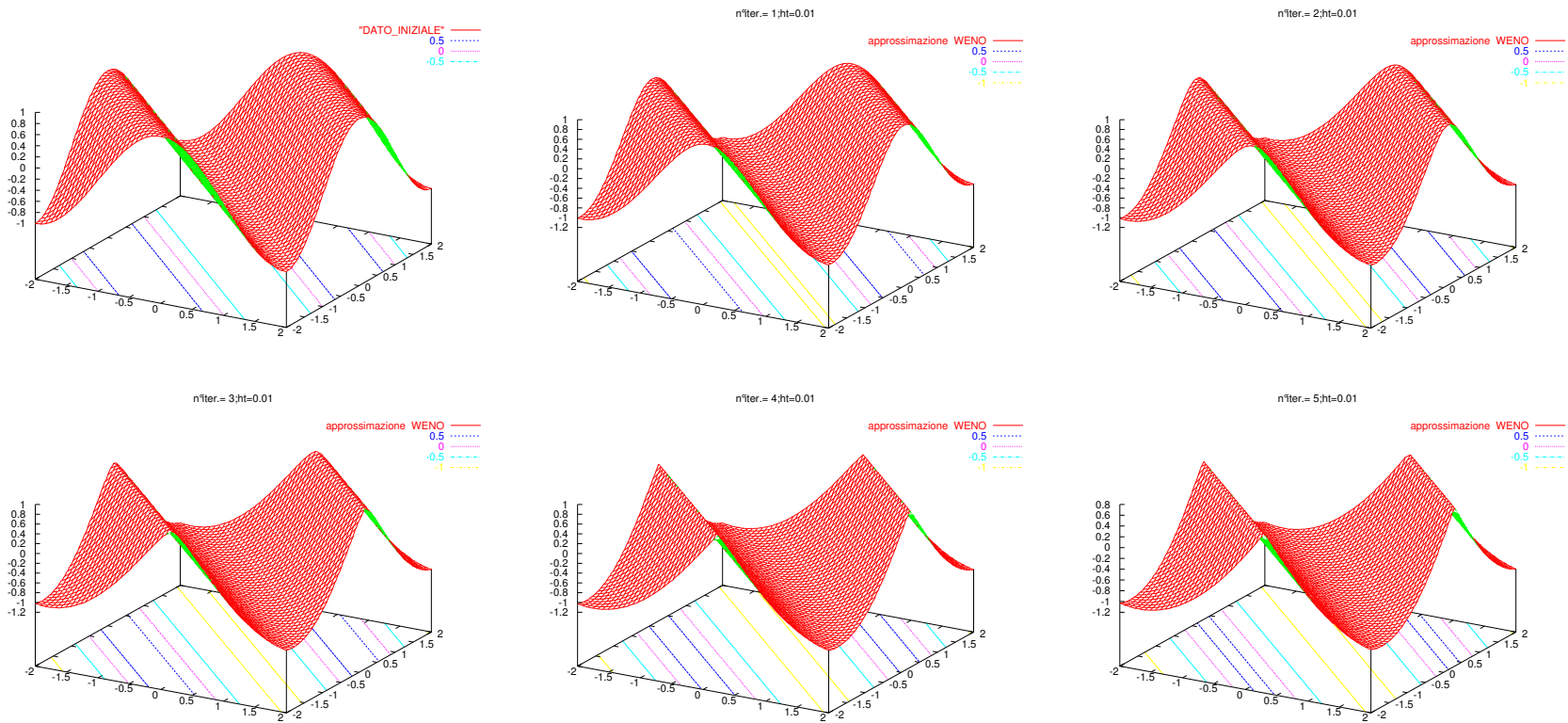
$C_k(x)$ are **positive** for each k

Therefore $I[V](x) = \sum_{k=1}^n \omega_k P_k(x)$ is a **convex** combination and

$$\min_k P_k(x) \leq I[V](x) \leq \max_k P_k(x)$$

TEST 1: two dimensional periodic solution

$$\begin{cases} v_t(x, t) + \frac{(v_{x_1}(x, t) + v_{x_2}(x, t) + 1)^2}{2} = 0 \\ v(x, 0) = v_0(x) = -\cos \frac{\pi(x_1 + x_2)}{2}, \end{cases}$$



WENO 3rd order				
Δx	relative L^∞ -error	L^∞ -order	relative L^1 -error	L^1 -order
0.16	$3.00 \cdot 10^{-3}$		$7.43 \cdot 10^{-4}$	
0.08	$4.66 \cdot 10^{-4}$	2.6	$4.79 \cdot 10^{-5}$	3.9
0.04	$2.68 \cdot 10^{-5}$	4.1	$2.44 \cdot 10^{-6}$	4.2
0.02	$1.48 \cdot 10^{-6}$	4.1	$1.62 \cdot 10^{-7}$	3.9

$$\Delta t = 0.02, T = \frac{0.8}{\pi^2}$$

WENO 5th order				
Δx	relative L^∞ -error	L^∞ -order	relative L^1 -error	L^1 -order
0.16	$1.28 \cdot 10^{-3}$		$1.51 \cdot 10^{-4}$	
0.08	$4.89 \cdot 10^{-5}$	4.3	$3.40 \cdot 10^{-6}$	4.8
0.04	$2.07 \cdot 10^{-6}$	4.5	$1.01 \cdot 10^{-7}$	5
0.02	$2.37 \cdot 10^{-8}$	6.4	$1.09 \cdot 10^{-9}$	6.5

$$\Delta t = 0.02, T = \frac{0.8}{\pi^2}$$

WENO 3rd order				
Δx	relative L^∞ -error	L^∞ -order	relative L^1 -error	L^1 -order
0.16	$1.00 \cdot 10^{-2}$		$1.09 \cdot 10^{-3}$	
0.08	$8.68 \cdot 10^{-5}$	6.8	$3.50 \cdot 10^{-5}$	5
0.04	$9.35 \cdot 10^{-6}$	3.2	$1.97 \cdot 10^{-6}$	4.1
0.02	$2.97 \cdot 10^{-7}$	4.4	$1.09 \cdot 10^{-7}$	4.1

$$\Delta t = 0.03, T = \frac{1.5}{\pi^2}$$

WENO 5th order				
Δx	relative L^∞ -error	L^∞ -order	relative L^1 -error	L^1 -order
0.16	$8.55 \cdot 10^{-3}$		$7.00 \cdot 10^{-4}$	
0.08	$8.53 \cdot 10^{-4}$	3.6	$9.30 \cdot 10^{-5}$	4.3
0.04	$2.08 \cdot 10^{-6}$	8.6	$4.68 \cdot 10^{-8}$	9.4
0.02	$5.34 \cdot 10^{-9}$	8.6	$1.69 \cdot 10^{-10}$	8.1

$$\Delta t = 0.03, T = \frac{1.5}{\pi^2}$$

SL SCHEME for MEAN CURVATURE MOTION

$$\begin{cases} u_t(x, t) = \operatorname{div} \left(\frac{Du(x, t)}{|Du(x, t)|} \right) |Du(x, t)| \\ u(x, 0) = u_0(x) \end{cases}$$

representation formula by Soner and Touzi:

$$u(x, t) = E\{u_0(y(x, t, t))\}, \quad Du \neq 0$$

$$\begin{cases} dy(x, t, s) = \sqrt{2}P(y, t, s)dW(s) \\ y(x, t, 0) = x \end{cases}$$

$$P(y, t, s) = \frac{1}{|Du|^2} \begin{pmatrix} u_{x_2}^2 & -u_{x_1}u_{x_2} \\ -u_{x_1}u_{x_2} & u_{x_1}^2 \end{pmatrix}$$

H.M. Soner, N. Touzi, *A stochastic representation for mean curvature type geometric flows*, Ann. Probab. 31 (2003), no. 3, 1145–1165.

CONSTRUCTION OF THE SCHEME

Soner Touzi formula between t and $t + \Delta t$:

$$u(x, t + \Delta t) = E\{u(y(x, t + \Delta t, \Delta t), t)\}$$

Brownian **dimension reduction** :

$$\sqrt{2}P dW = \frac{\sqrt{2}}{|Du|} \begin{pmatrix} u_{x_2} \\ -u_{x_1} \end{pmatrix} \left(\frac{u_{x_2} dW_1}{|Du|} - \frac{u_{x_1} dW_2}{|Du|} \right) = \sigma d\hat{W}$$

we can replace the Stochastic Initial problem by

$$\begin{cases} dy(x, t, s) = \sigma(y, t, s) d\hat{W}(s) \\ y(x, t, 0) = x. \end{cases}$$

Weak Euler :

$$\begin{cases} y_{k+1} = y_k + \sqrt{2}\sigma(y, t_k, 0)\Delta\hat{W}_k \\ y_0 = x. \end{cases}$$

with

$$P(\Delta\hat{W}_k = \pm\sqrt{\Delta t}) = \frac{1}{2}.$$

Time-discretization

$$u_{\Delta t}(x, t_{n+1}) = \frac{1}{2}u_{\Delta t}(x + \sqrt{2}\sigma(x, t_n, 0)\sqrt{\Delta t}, t_n) + \frac{1}{2}u_{\Delta t}(x - \sqrt{2}\sigma(x, t_n, 0)\sqrt{\Delta t}, t_n).$$

Fully discrete scheme $Du \neq 0$

$$u_j^{n+1} = \frac{1}{2} \left(I[u^n](x_j + \sigma_j^n \sqrt{\Delta t}) + I[u^n](x_j - \sigma_j^n \sqrt{\Delta t}) \right)$$

MODIFIED SCHEME

$$\begin{cases} u_j^{n+1} = \frac{1}{2} \left(I[u^n](x_j + \sigma_j^n \sqrt{\Delta t}) + I[u^n](x_j - \sigma_j^n \sqrt{\Delta t}) \right) & \text{if } |D_j^n| > C \Delta x^s \\ u_j^{n+1} = \frac{1}{4} \sum_{i \in \mathcal{D}(j)} u_i^n & \text{if } |D_j^n| \leq C \Delta x^s \end{cases}$$

CONSISTENCY ERROR

(case $|D_j^n| > C \Delta x^s$)

$$L_{\Delta x, \Delta t} = O\left(\frac{\Delta x^r}{\Delta t}\right) + O\left(\frac{\Delta x^{q-s}}{\Delta t^{\frac{1}{2}}}\right) + O(\Delta t^{\frac{1}{2}}) + O\left(\frac{\Delta t}{\Delta x^s}\right)$$

MONOTONE SCHEME

$$\widehat{H}(u^n; j) = \frac{(1-\lambda)}{2} \left[I_1[u^n](x_j + \sigma_j^n \sqrt{\Delta t}) + I_1[u^n](x_j - \sigma_j^n \sqrt{\Delta t}) \right] \\ + \frac{\lambda}{4} \sum_{i \in \mathcal{D}(j)} u_i^n, \quad \text{with } \lambda \in [0, 1].$$

Theorem Assume $\lambda \geq \frac{1}{1+\delta}$, where $\delta \equiv \frac{C \Delta x^{s+1}}{16 \sqrt{\Delta t} L_{I_1[u^n], \mathcal{G}}}$.

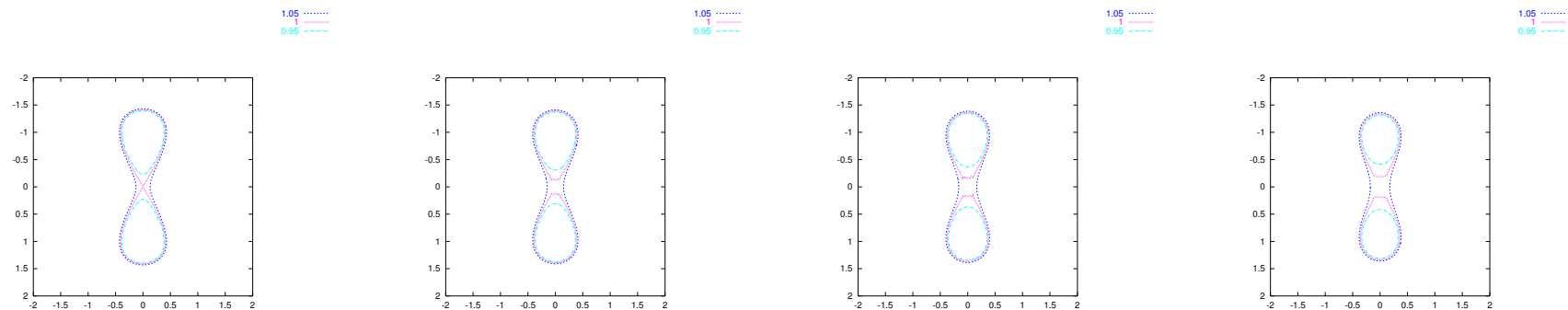
Then $\widehat{H}(u^n; j)$ is monotone.

TEST: two dimensional **circle evolution**

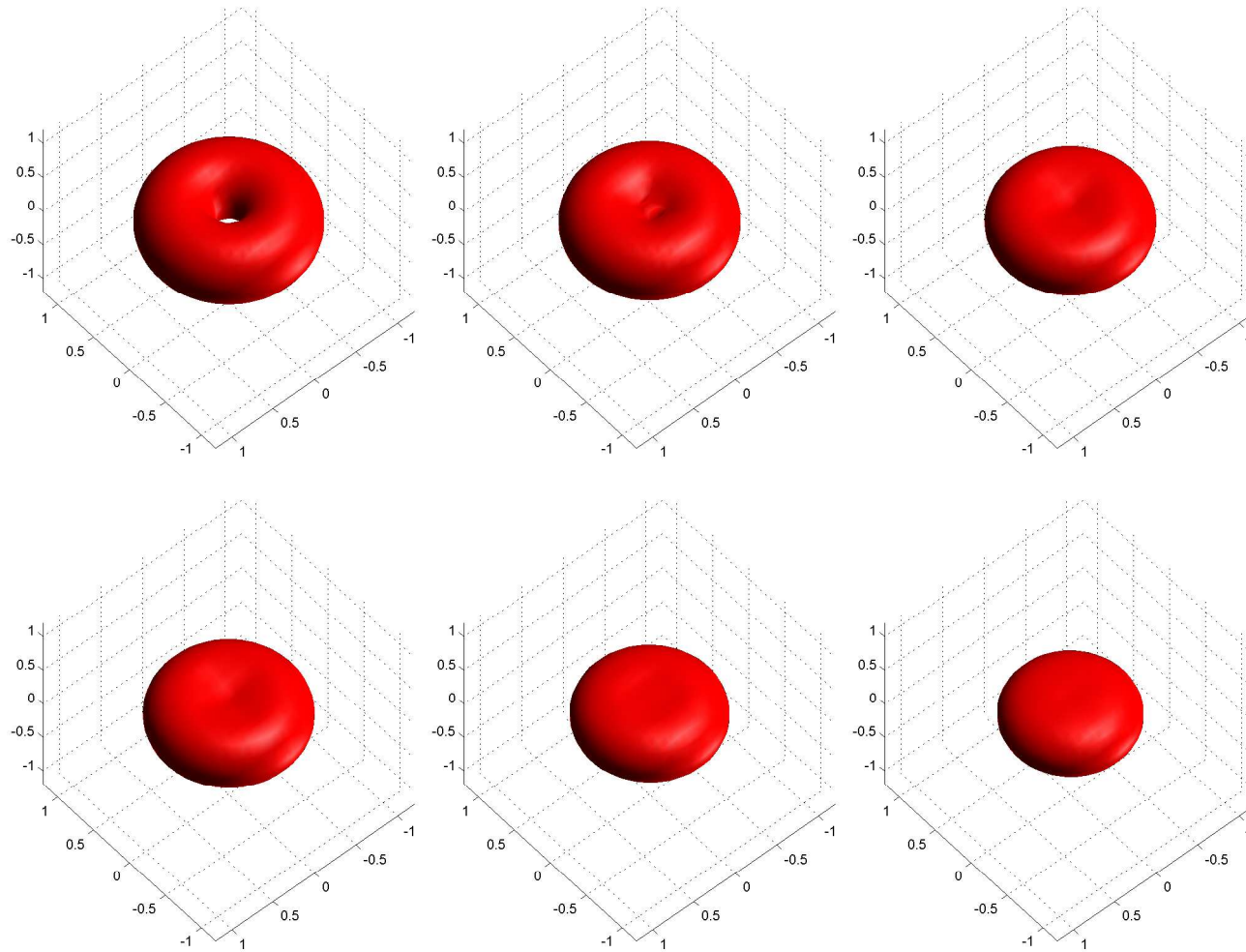
$$\Delta t = O(\Delta x^{\frac{3}{4}}), s = \frac{1}{4}.$$

Δx	Δt	$\ \cdot\ _{\infty}$	$\ \cdot\ _1$	L^{∞} - order	L^1 - order
0.04	0.08	$3.04 \cdot 10^{-4}$	$6.50 \cdot 10^{-6}$		
0.02	0.053	$1.25 \cdot 10^{-4}$	$3.42 \cdot 10^{-6}$	1.2	0.9
0.01	0.032	$5.22 \cdot 10^{-5}$	$1.82 \cdot 10^{-6}$	1.2	1.6
0.005	0.02	$2.09 \cdot 10^{-5}$	$7.75 \cdot 10^{-7}$	1.3	1.2

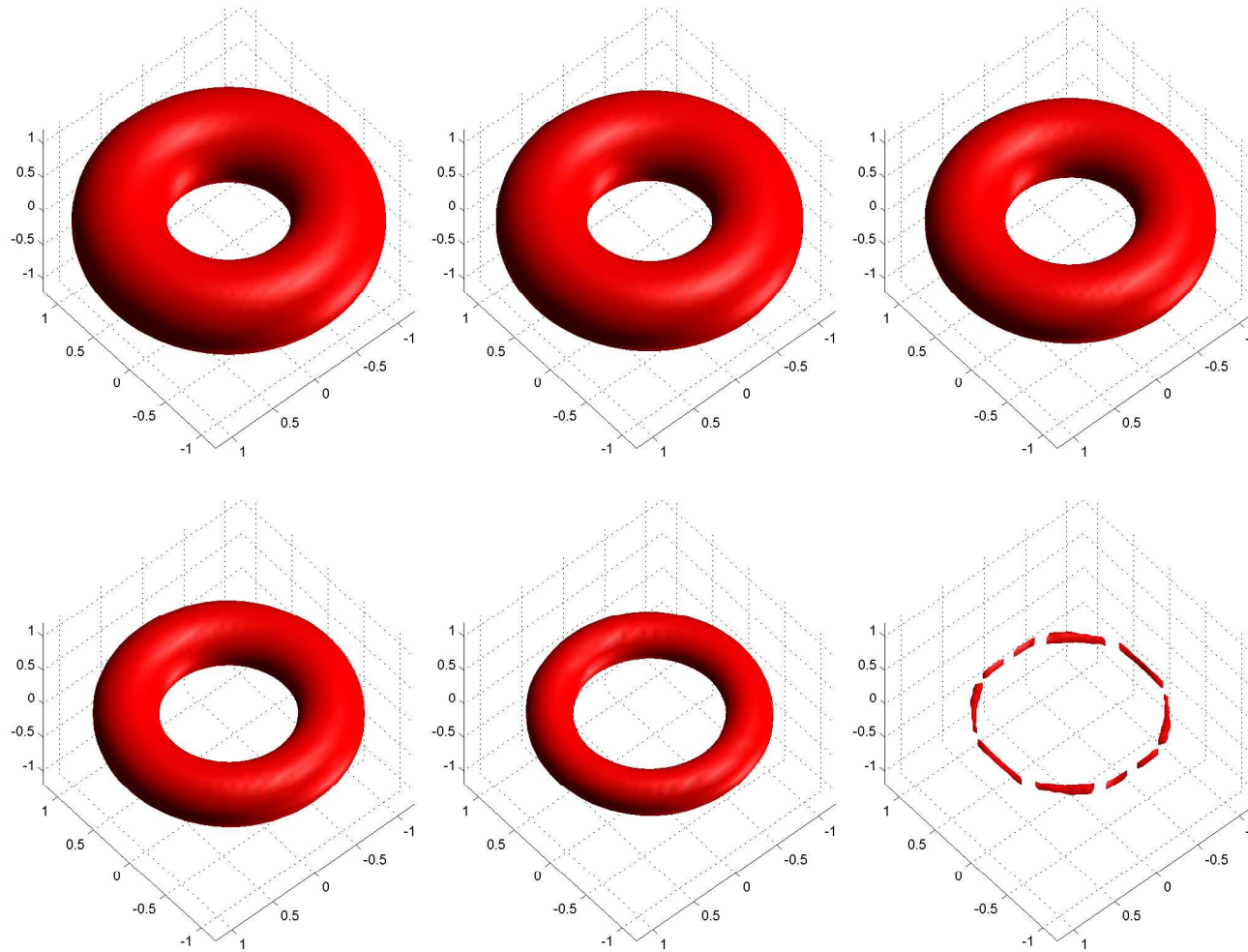
DEVELOPMENT of a NONEMPTY INTERIOR



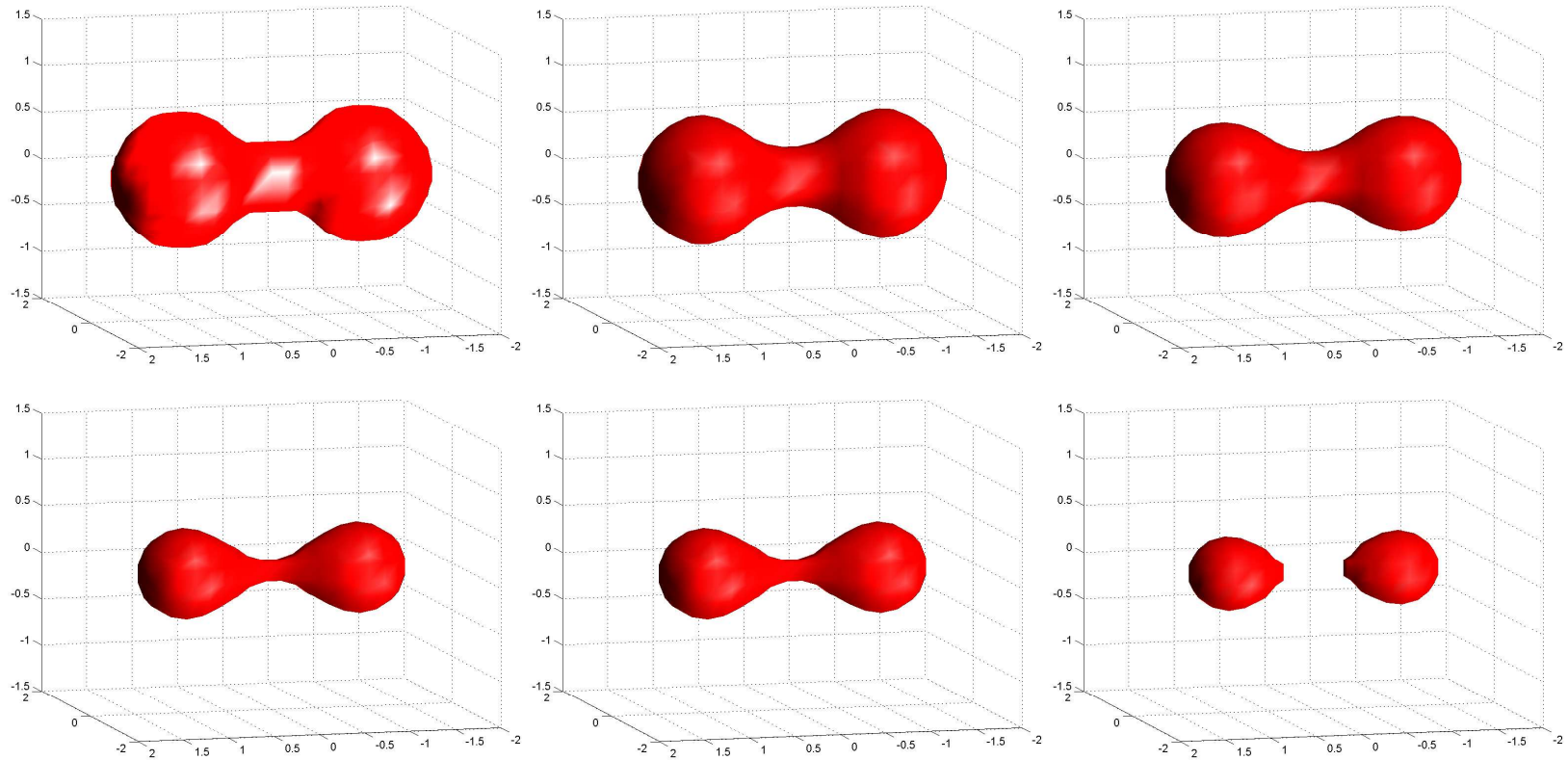
Fattening: evolution of the level curves $u = 0.095, 1, 1.05$



The torus evolving into a sphere



The torus collapsing in a circle



Dumb-bell: topology change in \mathbb{R}^3

MCM codimension 2
$$\begin{cases} u_t = F(D^2u, Du) & \mathbb{R}^3 \times [0, \infty) \\ u(x, 0) = \frac{1}{2}d(x, \mathcal{C})^2 \end{cases}$$

$$F(A, p) = \inf_{\nu \in \mathcal{N}(p)} \{\text{trace}[AP^\nu]\} \quad \mathcal{N}(p) = \{\nu \in S^2 : P^\nu p = 0\}$$

$$P^\nu = \nu\nu^T$$

Soner Touzi formula:

$$u(x, t) = \inf_{\nu \in \mathcal{U}} \{E\{u(y_\nu(x, t, t), 0)\}\}$$

with $\mathcal{U} = \{\nu : [0, T] \rightarrow S^2 \subset \mathbb{R}^3 : \nu \cdot Du(x, t) = 0\}$

$$\begin{cases} dy_\nu(x, t, s) = \sqrt{2}\nu(s)d\widehat{W}(s) \\ y_\nu(x, t, 0) = x \end{cases}$$

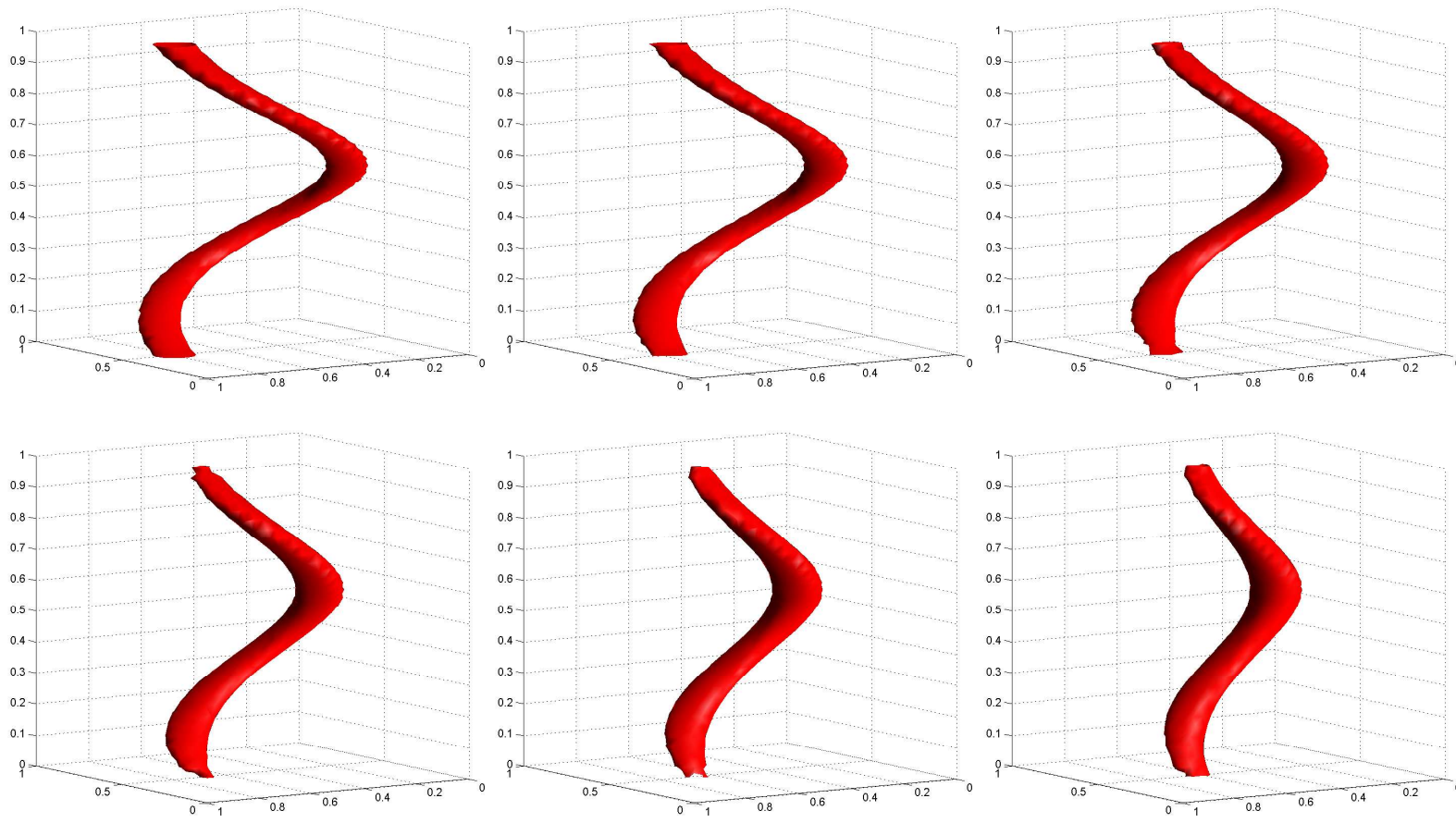
L.Ambrosio, M.Soner, *Level Set Approach to Mean Curvature Flow in Arbitrary Codimension*, J.Differential Geometry, **43** (1996), 693-737.

Time-discretization:

$$u_{\Delta t}(x, t_{n+1}) = \inf_{\nu^n \in \hat{\mathcal{U}}} \left\{ \frac{1}{2} u_{\Delta t}(x + \sqrt{2} \nu^n \sqrt{\Delta t}, t_n) + \frac{1}{2} u_{\Delta t}(x - \sqrt{2} \nu^n \sqrt{\Delta t}, t_n) \right\}.$$

Full-discrete scheme:

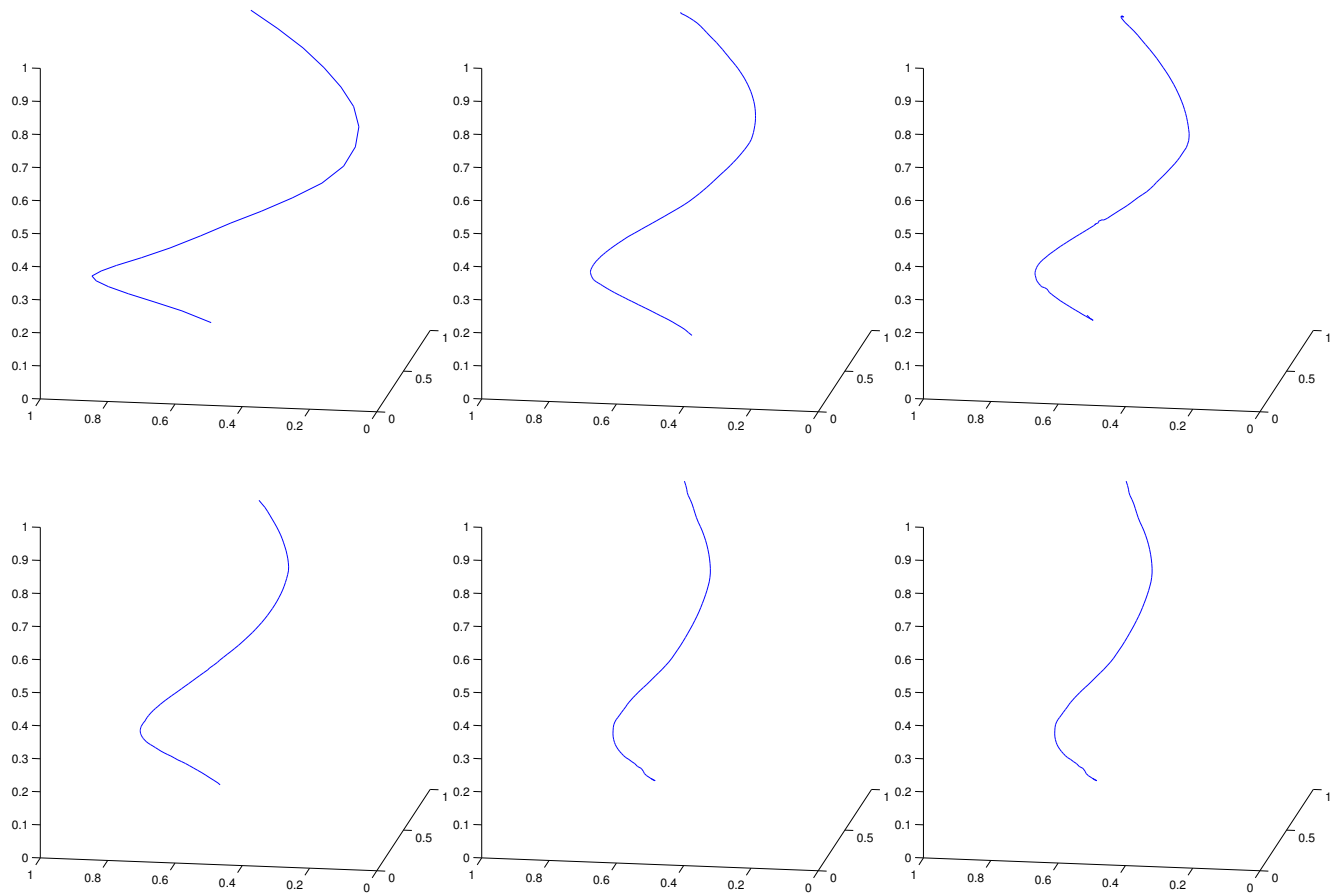
$$u_j^n = \min_{\nu^n \in \mathbb{R}^3} \left\{ \frac{1}{2} I[u^n](x + \sqrt{2} \nu^n \sqrt{\Delta t}) + \frac{1}{2} I[u^n](x - \sqrt{2} \nu^n \sqrt{\Delta t}) + \frac{(|D_j^n \nu^n|)^2}{\epsilon_1} + \frac{(|\nu^n| - 1)^2}{\epsilon_2} \right\}.$$



Evolution of ϵ helical surface, $\epsilon = 0.008$.

OPTIMAL TRAJECTORY ALGORITHM

- $s_0 := x_{jmin}$ point of the curve
- $\nu_j^{n*} = \underset{\nu_j^n \in \hat{U}}{\operatorname{argmin}} \{u^n(s_j + \sqrt{2\Delta s} \nu_j^n)\} \quad j = 0, \dots, \hat{j}$
- $s_{j+1} = s_j + \sqrt{2\Delta s} \nu_j^{n*} \quad j = 0, \dots, \hat{j}$
- $|s_0 - s_{\hat{j}}| \leq \epsilon$



Evolution of a helix in \mathbb{R}^3 .

REFERENCES:

- M. Falcone, R.Ferretti, *Semi-Lagrangian schemes for Hamilton-Jacobi equations discrete representation formulae and Godunov methods*, J. of Computational Physics, 175, (2002), 559-575.
- E.Carlini, M.Falcone, R.Ferretti, *An efficient algorithm for Hamilton-Jacobi equations in High Dimension*, To appear in *Computing and Visualization in Science* .
- E.Carlini, F.Ferretti, G.Russo, *A Weno large time-step for Hamilton-Jacobi equation*, Preprint
- M.Falcone, R.Ferretti, *Consistency of a large time-step scheme for mean curvature motion*, In F.Brezzi, A.Buffa, S.Corsaro and A.Murli (eds), *Numerical Analysis and Advanced Applications- Proceedings on ENUMATH 2001, Ischia*,(2001).