

Spline Element Method for Monge-Ampère equations

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Recent developments in nonlinear analysis and
applications, Cotonou 2010
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Outline

- 1 Fully Nonlinear 2nd Order Elliptic Equations
- 2 Existing Methods
- 3 Features of the Spline Method, Applications, Problems
- 4 Comparative Study of Three Methods
- 5 Numerical results

Fully nonlinear $F(x, u, Du, D^2u) = 0, x \in \Omega.$

We consider $F(x, D^2u(x)) = 0.$

F is uniformly elliptic if, $\exists \lambda, \Lambda > 0$ / for all M in $\mathbb{S}, x \in \Omega$

$$\lambda \|N\| \leq F(x, M + N) - F(x, M) \leq \Lambda \|N\|, \forall N \geq 0,$$

$$\lambda |\xi|^2 \leq \frac{\partial F}{\partial m_{ij}}(x, M) \xi_i \xi_j \leq \Lambda |\xi|^2, \forall \xi \in \mathbb{R}^n.$$

Main examples : Monge-Ampère and Pucci equations

Prescribed Gauss curvature $\det D^2u - K(x)(1 + |Du|^2)^{\frac{n+2}{2}} = 0$

Bellman equation $F(x, D^2u) := \inf_{\alpha \in \mathcal{A}} (L_\alpha u(x) - f_\alpha(x)) = 0$

Isaacs equation $F(x, D^2u) := \sup_\beta \inf_\alpha (L_{\alpha, \beta} u(x) - f_{\alpha, \beta}(x)) = 0$

Monge-Ampère Equation

$$\det D^2 u = f \text{ in } \Omega, \quad u = g \text{ on } \partial\Omega, \quad f > 0$$

Appears in mass transportation problems, kinetic theory, meteorology, fluid mechanics, nonlinear elasticity, antenna design, material sciences and mathematical finance.

Pucci Equations

$$M^+[u] := \alpha \Delta u + (1 - n\alpha) \lambda^+(D^2 u) = f$$

$$M^-[u] := \alpha \Delta u + (1 - n\alpha) \lambda^-(D^2 u) = f, \quad 0 < \alpha \leq \frac{1}{n}.$$

Appear in stochastic control where the control variable is the diffusion coefficient.

Pucci extremal operators play an important role in the theory of fully nonlinear equations.

Monge-Ampère

Concept of generalized solutions and viscosity solutions

Alexandrov-Bakelman solution equivalent to viscosity solution for f continuous

Theoretical convergence of numerical solutions not easy.

Geometric, Oliker and Prussner 1988 for Alexandrov solution

Oberman 2008 for viscosity solution

Some very simple difference methods, Benamou, Froese and Oberman 2010

Partial convergence theory for finite elements, Bohmer 2008

“One of the outstanding challenges to mathematical theory and ingenuity is the construction of algorithms which provide numerical answers to numerical questions . . . The effective use of the digital computer requires a combination of superior mathematical training and mathematical ingenuity”

Richard Bellman, preface to the book

The method of quasi-reversibility, Applications to partial differential equations by Lattès and Lions

Finite element based approaches

Dean and Glowinski, 2003, 2004, 2006. Augmented Lagrangian and Least Squares.

Feng and Neilan, 2007, 2008, 2009. Vanishing moment methodology.

$$-\epsilon \Delta^2 u + \det D^2 u = f, \text{ in } \Omega, \quad u = g, \Delta u = \epsilon^2 \text{ on } \partial\Omega.$$

Benamou, Froese and Oberman, 2009. Fixed point iteration for solving $u = T(u)$ where

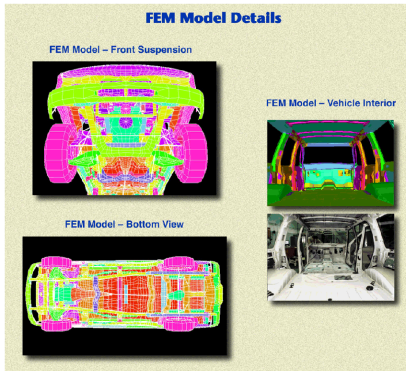
$$T(u) = \Delta^{-1}(\sqrt{(\Delta u)^2 + 2(f - \det D^2 u)})$$

Bohmer 2008, projection.

Finite element implementation with Lagrange multipliers

The finite element method is the most widely used method for solving numerically partial differential equations

A collection of methods falls under the designation f.e.m.



<http://www.epm.ornl.gov/SC98/car.html>

Model problem

$$\begin{cases} -\Delta u = f & \text{in } \Omega \\ u = g & \text{on } \partial\Omega \end{cases}$$

where $\partial\Omega$ will denote the boundary of the bounded domain Ω and Δ denotes the Laplace operator, $\Delta = \sum_{i=1}^n \frac{\partial^2}{\partial x_i^2}$.

Green's identity

$$\int_{\partial\Omega} (-\operatorname{div} \nabla u) v \, dx = \int_{\partial\Omega} \nabla u \cdot \nabla v \, dx - \int_{\partial\Omega} \frac{\partial u}{\partial \nu} v.$$

Find u in $H_0^1(\Omega)$ such that

$$\int_{\Omega} \nabla u \cdot \nabla v = \int_{\Omega} f v, \quad \forall v \in H_0^1(\Omega),$$

Biharmonic equation

$$\begin{cases} \Delta^2 u = f & \text{in } \Omega \\ u = g & \text{on } \partial\Omega \\ \frac{\partial u}{\partial \mathbf{n}} = h & \text{in } \partial\Omega, \end{cases}$$

$$\int_{\Omega} \Delta u \Delta v \, dx = \int_{\Omega} f v, \quad \forall v \in H_0^2(\Omega), \quad (1)$$

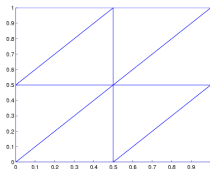
Abstract variational problem Find $u \in V$ such that

$$a(u, v) = \langle l, v \rangle \quad \text{for all } v \in V$$

Lax Milgram lemma

If a is symmetric, u is the unique minimizer in V of the functional $J(v) = 1/2a(v, v) - \langle l, v \rangle$.

Discrete approximations



Find $u_h \in V_h$ such that

$$a(u_h, v) = \langle l, v \rangle \quad \text{for all } v \in V_h$$

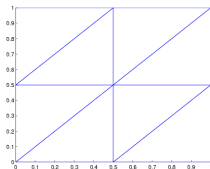
Equivalently, u_h is the unique minimizer in V_h of $J_h(u) = 1/2a(u, u) - \langle l, u \rangle$

Cea's lemma

$$\|u - u_h\|_V \leq C \min_{v \in V_h} \|u - v\|_V$$

for a constant C independent of h .

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Requirements for conforming approximations

Let $k \geq 1$ and suppose Ω is bounded. Then a piecewise infinitely differentiable function $v : \bar{\Omega} \rightarrow \mathbf{R}$ belongs to $H^k(\Omega)$ if and only if $v \in C^{k-1}(\bar{\Omega})$.

Even in two dimensions, conforming finite element spaces can be **very complicated** and there are **no satisfactory answer in three dimensions**.

Nonconforming approximations are not **very popular**.

Stokes equations Find $(\mathbf{u}, p) \in H^1(\Omega)^n \times L_0^2(\Omega)$ such that

$$\begin{cases} -\nu \Delta \mathbf{u} + \nabla p = \mathbf{f} & \text{in } \Omega \\ \operatorname{div} \mathbf{u} = 0 & \text{in } \Omega \\ \mathbf{u} = \mathbf{g} & \text{on } \partial\Omega \end{cases}$$

Equilibrium conditions are very difficult to enforce in the finite element method

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Conforming finite element implementation with Lagrange multipliers

Start with a representation of a piecewise discontinuous polynomial as a vector in \mathbb{R}^N

Express boundary conditions and constraints including global continuity or smoothness conditions as linear relations

Discretize the equation. Then enforce boundary conditions and constraints using Lagrange multipliers

Contrast with other approaches where Lagrange multipliers are introduced before discretization

Leads to saddle point problems which are solved by an augmented Lagrangian algorithm (sequences of linear equations with size $N \times N$)

Main advantages Flexible. No need to implement basis functions. Constraints and smoothness are enforced. Variable degree approximation.

Possible disadvantage Large matrix sizes

“spline” because we use Bernstein basis representations very convenient to express smoothness conditions

Splines on Plane Triangulations

Let $d \geq 1$ and $r \geq 0$

$$S_d^r(\Omega) = \{p \in C^r(\Omega), p|_t \in P_d, \forall t \in \mathcal{T}\}.$$

$$B_{ijk}^d(v) = \frac{d!}{i!j!k!} b_1^i b_2^j b_3^k, \quad i + j + k = d.$$

The set $\mathcal{B}^d = \{B_{ijk}^d(x, y), i + j + k = d\}$ is a basis for P_d .

$$s|_T = \sum_{i+j+k=d} c_{ijk}^T B_{ijk}^d,$$

Interpolation

On the triangle $T = \langle v_1, v_2, v_3 \rangle$ at $\xi_{ijk} = \frac{iv_1 + jv_2 + kv_3}{d}$

On the edge $\langle v_1, v_2 \rangle$ at $\xi_{ij} = \frac{iv_1 + jv_2}{d}$

$$\sum_{i+j=d} \tilde{c}_{ij} \tilde{B}_{ij}^d(v), \quad \tilde{B}_{ij}^d = \frac{d!}{i!j!} b_1^i b_2^j.$$

$$p = \sum_{i+j+k=d} c_{ijk} B_{ijk}^d, \quad q = \sum_{i+j=d} c_{ij0} B_{ij0}^d$$

$$Rc = c_b$$

Derivatives

$D_i c$, $i = 1, 2$ encode respectively the B -net of $\frac{\partial s}{\partial x_i}$.

Integration

$$\int_{\Omega} pq = c^T Id$$

Smoothness conditions

There's a (l, N) matrix H such that s is in $C^r(\Omega)$ if and only if

$$Hc = 0.$$

Features of the Spline Method

Find $u \in W$ such that

$$a(u, v) = \langle l, v \rangle \quad \text{for all } v \in V$$

- Find u in $H_0^1(\Omega)$ such that

$$\int_{\Omega} \nabla u \cdot \nabla v = \int_{\Omega} f v, \quad \forall v \in H_0^1(\Omega),$$

- Find $u \in H^2(\Omega)$, $u = g$ on $\partial\Omega$ such that

$$\int_{\Omega} (\text{cof } D^2 u) Du \cdot Dv \, dx = -n \int_{\Omega} f v \, dx, \quad \forall v \in H_0^1(\Omega).$$

$$V_h = \{c \in \mathbf{R}^N, Rc = 0\},$$

$$W_h = \{c \in \mathbf{R}^N, Rc = G\}$$

The condition $a(u, v) = \langle l, v \rangle$ for all $v \in V_h$ becomes

$$K(c, c)d = L^T d \quad \forall d \in V_h, \text{ that is for all } d \text{ with } Rd = 0$$

$$K(c, c) + \lambda^T R = L^T.$$

$$\begin{bmatrix} K^T & R^T \\ R & 0 \end{bmatrix} \begin{bmatrix} c \\ \lambda \end{bmatrix} = \begin{bmatrix} L \\ G \end{bmatrix}$$

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Recall Cea's lemma $\|u - u_h\|_V \leq C \min_{v \in V_h} \|u - v\|_V$

Assume V_h is a spline space $S_d^r(\mathcal{T}_h)$

Bivariate splines For $d \geq 3r + 2$ and $0 \leq m \leq d$ and for $0 \leq k \leq m$.

$$|f - Qf|_{\rho, k} \leq Ch^{m+1-k} |f|_{\rho, m+1}$$

The constant C depends on the smallest angle in Δ

Trivariate splines For $0 \leq k \leq d$, $f \in W_\rho^{d+1}(\Omega)$

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Advantages of the method

- Can be applied to a wide range of PDEs in science and engineering in both two and three dimensions.
- Constraints and Smoothness are enforced exactly and there is no need to implement basis functions with the required properties. Particularly suitable for higher order PDEs.
- No inf-sup condition
- One gets in a single implementation approximations of variable order.
- The mass and stiffness matrices are assembled easily and this can be done in parallel.

Possible Disadvantages

- Large size matrices for 3D problems and high order approximations
-

$$\begin{bmatrix} K^T & R^T \\ R & 0 \end{bmatrix} \begin{bmatrix} c \\ \lambda \end{bmatrix} = \begin{bmatrix} L \\ G \end{bmatrix} \quad \left(\begin{array}{cc} K^T & R^T \\ R & -\mu I \end{array} \right) \begin{bmatrix} c^{(l+1)} \\ \lambda^{(l+1)} \end{bmatrix} = \begin{bmatrix} L \\ G - \mu \lambda^{(l)} \end{bmatrix}$$

Computing $c^{(1)}$ from $\lambda^{(0)}$, one solves

$$\left(K^T + \frac{1}{\mu} R^T R \right) c^{(l+1)} = K^T c^{(l)} + \frac{1}{\mu} R^T G, \quad l = 1, 2, \dots$$

$$\|c - c^{(l+1)}\| \leq C\mu \|c - c^{(l)}\|$$

Examples of Applications

- G. Awanou, M. J. Lai and P. Wenston, The multivariate spline method for numerical solutions of partial differential equations and scattered data fitting, *Wavelets and Splines* : Athens 2005, pp. 24–74.
- V. Baradmize and M.J. Lai, Spherical spline solution to a PDE on sphere, *Wavelets and Splines* : Athens 2005, pp. 75–92.
- G. Awanou and M. J. Lai, Trivariate spline approximations of 3D Navier-Stokes equations, *Math. Comp.* **74** (2005), pp. 585–601.
- G. Awanou, Robustness of a spline element method with constraints, *Journal of Scientific Computing*, **36**, 3, (2009) pp. 421–432.
- Xian-Liang Hu, Dan-Fu Han, Ming-Jun Lai, Bivariate Splines of Various Degrees for Numerical Solution of Partial Differential Equations, *SIAM Journal of Scientific Computing*, 29(2007) 1338–1354

Test 1 : a smooth solution

Test 2 : solution not in $H^2(\Omega)$

Test 3 : no exact solution. Expects a convex solution.

Method 1 Newton's method.

Dean and Glowinski in Numerical Methods for fully nonlinear equations of the Monge-Ampère type “Newton's and conjugate gradients methods may be well-suited for the solution of ... combines the difficulty of both harmonic and bi-harmonic problems, making the approximation a delicate matter, albeit solvable ... If ... has no solution, we can expect the divergence of the Newton ...”

Method 2 Vanishing moment is analogue of a singular perturbation problem

$$\epsilon \Delta^2 u - \Delta u = f \text{ in } \Omega, u = 0, \frac{\partial u}{\partial n} = 0, \text{ on } \partial\Omega$$

Method 3 Iterative method

Put $F(u) = \det D^2 u - f$. Then $F(u) = 1/n \operatorname{div} ((\operatorname{cof} D^2 u)Du) - f$.
 $F'(u)w = \operatorname{div} ((\operatorname{cof} D^2 u)Dw)$.

Variational formulation Find $u \in H^2(\Omega)$, $u = g$ on $\partial\Omega$ such that

$$-\frac{1}{n} \int_{\Omega} (\operatorname{cof} D^2 u)Du \cdot Dv \, dx = \int_{\Omega} fv \, dx, \quad \forall v \in H^2(\Omega) \cap H_0^1(\Omega).$$

Algorithm 1 Newton's method

$$F'(u_k)(u_{k+1} - u_k) = -F(u_k)$$

$$\operatorname{div} ((\operatorname{cof} D^2 u_k)(Du_{k+1} - Du_k)) = -\frac{1}{n} \operatorname{div} ((\operatorname{cof} D^2 u_k)Du_k) + f,$$

$$\int_{\Omega} (\operatorname{cof} D^2 u_k)Du_{k+1} \cdot Dv \, dx = \frac{n-1}{n} \int_{\Omega} (\operatorname{cof} D^2 u_k)Du_k \cdot Dv \, dx - \int_{\Omega} fv \, dx, \quad \forall v \in V_0.$$

Algorithm 2 (Newton's method of vanishing moment formulation Feng-Neilan)

$$\begin{aligned} \epsilon \int_{\Omega} \Delta u_{k+1} \Delta v \, dx + \int_{\Omega} (\operatorname{cof} D^2 u_k) D u_{k+1} \cdot D v \, dx &= \frac{n-1}{n} \int_{\Omega} (\operatorname{cof} D^2 u_k) D u_k \cdot D v \, dx \\ &+ \epsilon^3 \int_{\partial\Omega} \frac{\partial v}{\partial n} \, ds - \int_{\Omega} f v \, dx, \quad \forall v \in H^2(\Omega) \cap H_0^1(\Omega) \end{aligned}$$

with $u_{k+1} = g$ on $\partial\Omega$

Algorithm 3 (Iterative method of Benamou-Froese-Oberman)

$$\Delta u_{k+1} = \sqrt{(\Delta u_k)^2 + 2(f - \det D^2 u_k)}$$

with $u_{k+1} = g$ on $\partial\Omega$

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Analysis of algorithms

Based on the work of Loeper and Rapetti Newton's method can be shown to converge based on global Schauder estimates due to Wang. For a C^3 domain, $g \in C^3(\overline{\Omega})$, $\inf f > 0$, and $f \in C^\alpha$ for some $\alpha \in (0, 1)$

2D Newton's method

d	L^2 norm	H^1 norm	H^2 norm
3	$1.0610 \cdot 10^{-3}$	$1.1101 \cdot 10^{-2}$	$1.6383 \cdot 10^{-1}$
4	$3.5127 \cdot 10^{-5}$	$4.8553 \cdot 10^{-4}$	$9.0596 \cdot 10^{-3}$
5	$4.1572 \cdot 10^{-6}$	$6.5142 \cdot 10^{-5}$	$1.9364 \cdot 10^{-3}$
6	$1.9685 \cdot 10^{-7}$	$3.6401 \cdot 10^{-6}$	$1.4774 \cdot 10^{-4}$
7	$2.2699 \cdot 10^{-8}$	$4.1498 \cdot 10^{-7}$	$2.2424 \cdot 10^{-5}$
8	$1.2430 \cdot 10^{-9}$	$2.2586 \cdot 10^{-8}$	$1.5479 \cdot 10^{-6}$
Rate	$10^{3.7312} d^{-13.5794}$	$10^{3.7312} d^{-12.9970}$	$10^{4.8498} d^{-11.3673}$

Test 1 : $u(x, y) = e^{(x^2+y^2)/2}$ $h = 1/2$

d	L^2 norm	H^1 norm	H^2 norm
3	$1.2809 \cdot 10^{-4}$	$2.6554 \cdot 10^{-3}$	$8.9587 \cdot 10^{-2}$
4	$1.6278 \cdot 10^{-6}$	$4.5619 \cdot 10^{-5}$	$1.7395 \cdot 10^{-3}$
5	$1.1531 \cdot 10^{-7}$	$2.3916 \cdot 10^{-6}$	$1.3444 \cdot 10^{-4}$
6	$1.7705 \cdot 10^{-9}$	$6.8506 \cdot 10^{-8}$	$5.5403 \cdot 10^{-6}$
7	$1.4548 \cdot 10^{-10}$	$3.7545 \cdot 10^{-9}$	$3.9490 \cdot 10^{-7}$
8	$8.1014 \cdot 10^{-12}$	$5.3353 \cdot 10^{-10}$	$7.2159 \cdot 10^{-8}$
Rate	$10^{4.3451} d^{-16.8127}$	$10^{5.2200} d^{-15.9649}$	$10^{5.9692} d^{-14.4722}$

Test 1 : $u(x, y) = e^{(x^2+y^2)/2}$ $h = 1/4$

ϵ	L^2 norm	H^1 norm	H^2 norm
10^{-3}	1.727110^{-3}	2.091010^{-2}	8.633810^{-1}
10^{-4}	1.856310^{-4}	3.558110^{-3}	2.005010^{-1}
10^{-5}	1.891710^{-5}	4.070010^{-4}	2.411910^{-2}
10^{-6}	1.818210^{-6}	4.077510^{-5}	2.438810^{-3}
10^{-7}	1.244110^{-7}	4.095110^{-6}	2.494910^{-4}
10^{-8}	1.011910^{-7}	2.296210^{-6}	1.302910^{-4}
10^{-9}	1.138410^{-7}	2.379010^{-6}	1.338210^{-4}
10^{-10}	1.151610^{-7}	2.390310^{-6}	1.343810^{-4}
10^{-11}	1.153010^{-7}	2.391410^{-6}	1.344310^{-4}
10^{-14}	1.153110^{-7}	2.391610^{-6}	1.344410^{-4}
Reduced	$1.1531 \cdot 10^{-7}$	$2.3916 \cdot 10^{-6}$	$1.3444 \cdot 10^{-4}$

Comparison of Newton's method with (Newton) vanishing moment method Test 1,
 $h = 1/4$, $d = 5$. No boundary layers issue with spline element method.

h	n_{it}	L^2 norm	H^1 norm	H^2 norm
$1/2^1$	41	$2.8275 \cdot 10^{-6}$	$6.1372 \cdot 10^{-5}$	$1.8845 \cdot 10^{-3}$
$1/2^2$	37	$5.4642 \cdot 10^{-8}$	$2.1971 \cdot 10^{-6}$	$1.2972 \cdot 10^{-4}$
$1/2^3$	38	$8.3164 \cdot 10^{-10}$	$7.2252 \cdot 10^{-8}$	$8.4790 \cdot 10^{-6}$
$1/2^4$	37	$2.7871 \cdot 10^{-9}$	$1.4089 \cdot 10^{-8}$	$1.0809 \cdot 10^{-6}$

BFO iterative method for Test 1, $d = 5$

d	n_{it}	L^2 norm	H^1 norm	H^2 norm
3	38	$2.1905 \cdot 10^{-6}$	$1.3368 \cdot 10^{-4}$	$1.9600 \cdot 10^{-2}$
4	39	$5.1317 \cdot 10^{-9}$	$5.9766 \cdot 10^{-7}$	$7.7660 \cdot 10^{-5}$
5	35	$2.2563 \cdot 10^{-9}$	$1.2616 \cdot 10^{-8}$	$1.0166 \cdot 10^{-6}$
6	34	$1.2775 \cdot 10^{-9}$	$1.3206 \cdot 10^{-8}$	$2.9168 \cdot 10^{-6}$
7	31	$2.8787 \cdot 10^{-9}$	$3.5309 \cdot 10^{-8}$	$8.2791 \cdot 10^{-6}$
8	26	$2.1192 \cdot 10^{-8}$	$2.6174 \cdot 10^{-7}$	$1.7105 \cdot 10^{-5}$

BFO iterative method for Test 1, $h = 1/2^4$

Newton's method diverges for this test $d = 3$

h	L^2 norm	H^1 norm
$1/2^1$	2.195410^{-2}	1.640910^{-1}
$1/2^2$	3.609710^{-3}	6.140510^{-2}
$1/2^3$	1.068510^{-3}	4.097810^{-2}
$1/2^4$	5.083810^{-3}	2.804810^{-1}
$1/2^5$	2.579710^{+3}	2.268810^{+5}
$1/2^6$	1.845210^{+4}	3.592210^{+6}

Test 2 : $u(x, y) = -\sqrt{2 - x^2 - y^2} \notin H^2(\Omega)$ $d = 3$

Damped Newton's method, $d = 3, \tau = 50$.
Results here better than Glowinski's

$$F'(u_k)(u_{k+1} - u_k) = -\frac{1}{\tau}F(u_k), \tau \geq 1$$

Convergence for smooth solutions on smooth domains for some $\tau \geq 1$

h	L^2 norm	H^1 norm
$1/2^1$	$2.1173 \cdot 10^{-2}$	$1.5928 \cdot 10^{-1}$
$1/2^2$	$2.6709 \cdot 10^{-3}$	$3.4032 \cdot 10^{-2}$
$1/2^3$	$2.6400 \cdot 10^{-4}$	$7.7037 \cdot 10^{-3}$
$1/2^4$	$2.5420 \cdot 10^{-5}$	$2.1618 \cdot 10^{-3}$

Vanishing moment Feng-Neilan $\epsilon = 10^{-3}$, $d = 5$

h	L^2 norm	H^1 norm
$1/2^1$	7.668010^{-3}	7.449110^{-2}
$1/2^2$	1.453610^{-3}	3.924410^{-2}
$1/2^3$	9.872710^{-3}	2.511210^{-1}
$1/2^4$	5.681910^{-3}	2.492710^{-1}
$1/2^5$	$1.9830 \cdot 10^{+4}$	$1.1812 \cdot 10^{+6}$

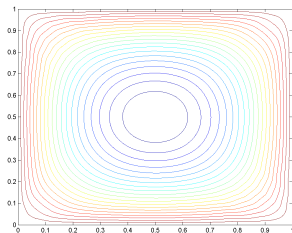
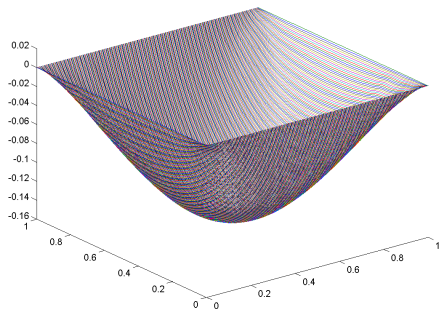
Vanishing moment Feng-Neilan $\epsilon = 10^{-2}$, $d = 5$

h	L^2 norm	H^1 norm
$1/2^1$	7.825410^{-3}	9.318410^{-2}
$1/2^2$	1.064610^{-2}	9.520110^{-2}
$1/2^3$	1.130610^{-2}	9.615410^{-2}
$1/2^4$	1.150010^{-2}	9.133610^{-2}
$1/2^5$	1.162510^{-2}	8.778510^{-2}
$1/2^6$	1.168110^{-2}	8.563210^{-2}

BFO iterative method $d = 3$, $r = 0$ and $r = 1$

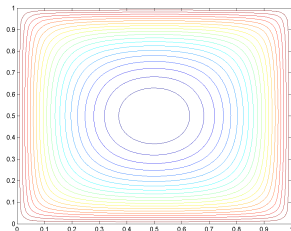
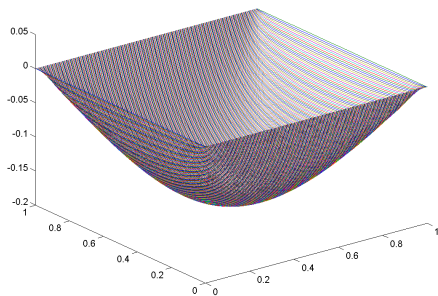
h	n_{it}	L^2 norm	H^1 norm
$1/2^1$	50	$2.3921 \cdot 10^{-1}$	1.1900
$1/2^2$	159	$1.2585 \cdot 10^{-1}$	$7.1292 \cdot 10^{-1}$
$1/2^3$	151	$1.0341 \cdot 10^{-1}$	$6.4299 \cdot 10^{-1}$
$1/2^4$	160	$9.6031 \cdot 10^{-2}$	$6.2088 \cdot 10^{-1}$
$1/2^5$	199	$9.4551 \cdot 10^{-2}$	$6.2453 \cdot 10^{-1}$
$1/2^6$	8	$1.6977 \cdot 10^{-2}$	$2.2925 \cdot 10^{-1}$

No smooth solution Damped Newton's method



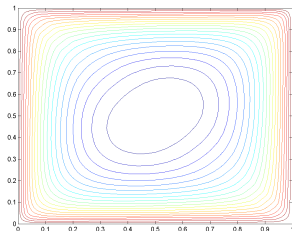
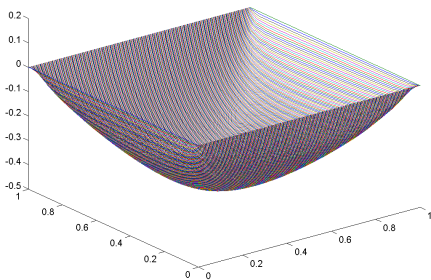
Test 3 : $f(x, y) = 1$ and $g(x, y) = 0$ on
 $h = 1/2^4$, $d = 3$, $r = 1$, $\tau = 100$.

No smooth solution Vanishing moment Feng-Neilan



Test 3 : $f(x, y) = 1$ and $g(x, y) = 0$ on $h = 1/2^4, d = 3$.

No smooth solution Benamou-Froese-Oberman



Test 3 : $f(x, y) = 1$ and $g(x, y) = 0$ on $h = 1/2^4, d = 3$.

3D Newton's method

d	L^2 norm	H^1 norm	H^2 norm
3	$5.7401 \cdot 10^{-3}$	$3.9308 \cdot 10^{-2}$	$3.8805 \cdot 10^{-1}$
4	$4.9653 \cdot 10^{-4}$	$4.7672 \cdot 10^{-3}$	$7.6089 \cdot 10^{-2}$
5	$6.9025 \cdot 10^{-5}$	$7.7162 \cdot 10^{-4}$	$1.5425 \cdot 10^{-2}$
6	$1.0005 \cdot 10^{-5}$	$1.0310 \cdot 10^{-4}$	$2.5992 \cdot 10^{-3}$
7	$1.8956 \cdot 10^{-6}$	$1.7983 \cdot 10^{-5}$	$4.8870 \cdot 10^{-4}$
Rate	$10^{2.3278} d^{-9.4276}$	$10^{3.0318} d^{-9.0347}$	$10^{3.4794} d^{-7.8305}$

$$u(x, y, z) = e^{(x^2+y^2+z^2)/3} \text{ on } \mathcal{I}_1$$

d	L^2 norm	H^1 norm	H^2 norm
3	$4.7273 \cdot 10^{-3}$	$2.6606 \cdot 10^{-2}$	$1.7829 \cdot 10^{-1}$
4	$6.5001 \cdot 10^{-5}$	$9.6609 \cdot 10^{-4}$	$2.2183 \cdot 10^{-2}$
5	$2.1399 \cdot 10^{-6}$	$4.5600 \cdot 10^{-5}$	$1.5825 \cdot 10^{-3}$
6	$1.3815 \cdot 10^{-6}$	$2.2368 \cdot 10^{-5}$	$6.0920 \cdot 10^{-4}$
7	$1.5283 \cdot 10^{-7}$	$1.6753 \cdot 10^{-6}$	$5.6189 \cdot 10^{-5}$
Rate	$10^{3.1038} d^{-11.8706}$	$10^{3.6366} d^{-11.0367}$	$10^{3.7955} d^{-9.3039}$

$$u(x, y, z) = e^{(x^2+y^2+z^2)/3} \text{ on } \mathcal{T}_2$$

3D version of BFO

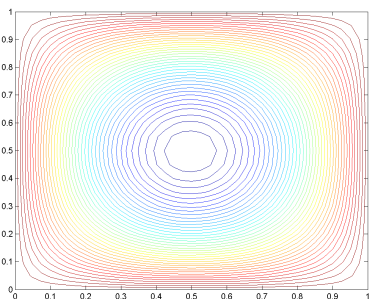
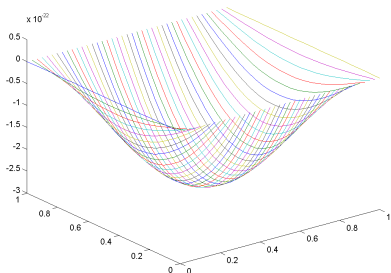
$$\Delta u_{k+1} = ((\Delta u_k)^3 + 9(f - \det D^2 u_k))^{\frac{1}{3}},$$

with $u_{k+1} = g$ on $\partial\Omega$.

$$\det D^2 u \leq \frac{1}{n^n} (\Delta u)^n,$$

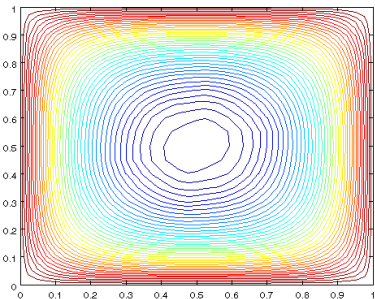
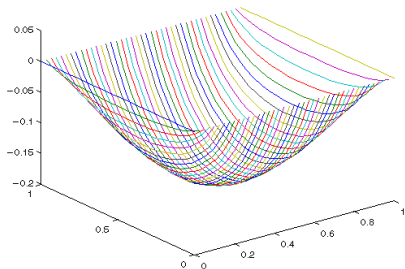
Enforces partial convexity since $\Delta u_k \geq 0$

No smooth solution for $f=1$, $g=0$

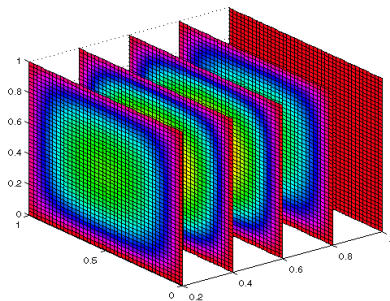
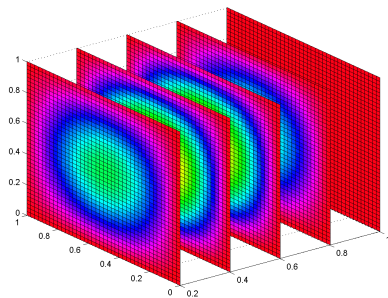


Test 2 : $f(x, y, z) = 1$ and $g(x, y, z) = 0$ on \mathcal{I}_2 , $d = 3$, $r = 2$, $\epsilon = 10^{-3}$. Vanishing moment

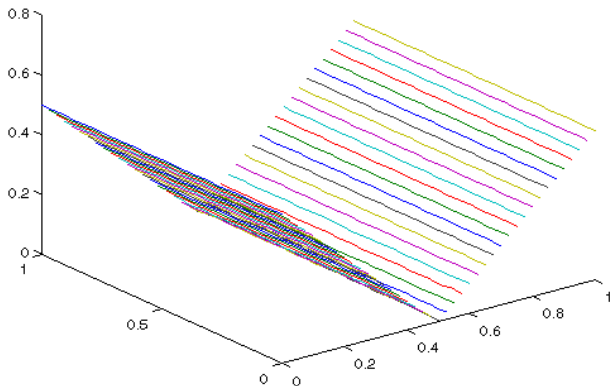
No smooth solution for $f=1$, $g=0$



Test 2 : $f(x, y, z) = 1$ and $g(x, y, z) = 0$ on \mathcal{I}_2 , $d = 5$, $r = 1$. 3D BFO



Slices in the x -direction on Domain 2 and \mathcal{I}_3 , $d = 3$, Vanishing moment $r = 2$, $\epsilon = 10^{-5}$ and BFO $d = 5$, $r = 1$

$f=0, g(x, y, z) = |x - 1/2|$ 3D BFO

Concluding remarks

For homogeneous boundary conditions, i.e. for $u = 0$ on $\partial\Omega$,
i.e. for $u(x, y) = \pm x(1 - x)y(1 - y)\exp(x^2 + y^2)$, both concave
and convex solutions

Newton's method : initial guess $\pm u_0$

Vanishing moment (Feng-Neilan) : $\pm\epsilon$

Benamou-Froese-Oberman : $u_{k+1} = \pm T(u_k)$

Last two perform (to some extent) when there is no smooth
solution

Vanishing moment (with Newton) gives Newton's method result
for $\epsilon = 10^{-9}$

Future directions

Proven convergence methods

Benamou-Brenier “ It follows from this theoretical result that a natural computational solution of the L^2 MKP is the numerical resolution of the Monge-Ampère equation”
...”Unfortunately, this fully nonlinear second-order elliptic equation has not received much attention from numerical analysts and, to the best of our knowledge, there is no efficient finite-difference or finite-element methods, comparable to those developed for linear second-order elliptic equations (such as fast Poisson solvers, multigrid methods, preconditioned conjugate gradient methods,...). ”

Connections with optimal transportation in particular work of Barrett and Prigozhin

Other engineering applications

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