Adaptive Grid Methods for Q-tensor Theory of Liquid Crystals

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Motivation

- model: *Q*-tensor model of nematic liquid crystal cell
- aim: model dynamics of defect movement
- problem: characteristic lengths with large scale differences
- uniform grid: many grid points needed to capture defect behaviour
- idea: use adaptive grid methods to ensure there is no waste of computational effort

Adaptive Grid Methods I

- local mesh refinement
 - extra nodes added locally in regions of high error
 - h refinement, p refinement
 - often requires complicated data structures which need updating frequently
- moving mesh methods
 - existing node points are moved to regions of high error
 - same grid connectivity maintained
 - r refinement
 - comparatively easy extension of existing software

Adaptive Grid Methods II

- velocity-based methods
 - mesh point velocities are calculated directly
 - moving finite element methods
 - geometric conservation laws
- location-based methods
 - mesh points are calculated directly
 - equidistribution methods
 - harmonic mapping
- adaptive grid on physical domain is image of uniform grid on computational domain under a suitable mapping

Grid Mapping in 1D

- coordinates: physical $x \in [0, 1]$, computational $\xi \in [0, 1]$
- coordinate transformation:

$$x = x(\xi, t), \ \xi \in [0, 1], \qquad x(0, t) = 0, \ x(1, t) = 1$$

• uniform mesh on computational domain:

$$\xi_i = \frac{i}{N}, \qquad i = 0, 1, \dots, N, \qquad N \in \mathbb{Z}^+$$

• corresponding physical mesh:



Equidistribution Principle in 1D

- choose (positive) monitor function M(x,t)
- equidistribution principle (EP):

$$\int_0^{x(\xi,t)} M(s,t) \, ds = \xi \int_0^1 M(s,t) \, ds$$

• discrete forms:

$$\int_{x_i}^{x_{i+1}} M(s,t) \, ds = \int_{x_{i-1}}^{x_i} M(s,t) \, ds, \qquad i = 1, \dots, N-1$$

or

$$\int_{x_{i-1}}^{x_i} M(s,t) \, ds = \frac{1}{N} \int_0^1 M(s,t) \, ds, \qquad i = 1, \dots, N$$

Monitor Functions

• (scaled) arc-length monitor function

$$M(u(x,t)) = \sqrt{\hat{\alpha} + \left(\frac{\partial u(x,t)}{\partial x}\right)^2}$$

user-prescribed parameter $\hat{\alpha} > 0$

• various other ideas e.g. Beckett and Mackenzie (2000)

$$M(u(x,t)) = \delta + \left|\frac{\partial u(x,t)}{\partial x}\right|^{\frac{1}{m}}$$

$$\delta$$
, *m* positive constants, $\delta = \int_0^1 \left| \frac{\partial u(x,t)}{\partial x} \right|^{\frac{1}{m}} dx$

Aside on MMPDEs

- equidistribution principle differentiate EP twice with respect to ξ
- variational principle

find Euler-Lagrange equation associated with

$$I[x] = \frac{1}{2} \int_0^1 x_{\xi}^2(\xi) M^2(x(\xi)) \, d\xi$$

$$x_{\xi\xi} + \frac{M_{\xi}}{M}x_{\xi} = 0$$

elliptic equidistribution generator

Adaptive Grid Methods III

- dynamic methods:
 - moving mesh partial differential equation (MMPDE)
 - mesh equation and underlying PDE solved together

- static methods: at a fixed time,
 - equation is discretised and solved on an initial mesh
 - adaptive mesh is constructed based on EP for monitor function
 - solution is interpolated onto new mesh
 - solution mesh generation loop

Practical Algorithm

- Sanz-Serna and Christie, JCP 67, 1986
- 1D example:

$$u_t = F(u, u_x, u_{xx}, x, t)$$

plus boundary and initial conditions

- time level t_n , grid x_i^n , approximation U_i^n to $u(x_i^n, t_n)$
- equidistribute solution arc-length:
- (i) Use numerical scheme on grid x_i^n to find \bar{U}_i^{n+1} .
- (ii) Join points $(x_j^n, \overline{U}_j^{n+1})$ by straight lines. Find the points on the polygon which divide its length into equal parts and project onto the *x*-axis. Compute U_j^{n+1} via interpolation.

Q-tensor Theory

• aim: minimise free energy density

$$\mathcal{F} = \int_V F(\theta, \phi, \nabla \theta, \nabla \phi) \, dV$$

- problems with multivalued angles/singularities
- tensor order parameter

$$Q = \begin{bmatrix} q_1 & q_2 & q_3 \\ q_2 & q_4 & q_5 \\ q_3 & q_5 & -q_1 - q_4 \end{bmatrix}$$

express free energy density as

$$\mathcal{F} = \int_V F(q_i, \nabla q_i) \, dV, \qquad i = 1, 2, 3, 4, 5$$

1D Model Problems

- homogeneous uniaxial alignment in $\Omega \equiv z \in [0, d]$
- z-axis aligned with n

$$Q = \sqrt{\frac{3}{2}}S \begin{bmatrix} -\frac{1}{3} & 0 & 0\\ 0 & -\frac{1}{3} & 0\\ 0 & 0 & \frac{2}{3} \end{bmatrix}$$

 ${\cal Q}$ depends only on scalar order parameter ${\cal S}$

• bulk energy densities $(A_s, B_s, C_s, L_{1s}, S_{eq}, F_{eq} + ve)$

$$\frac{1}{2}A_sS^2 - \frac{1}{3}B_sS^3 + \frac{1}{4}C_sS^4 + \left(\frac{2L_{1s}+1}{6}\right)\left(\frac{\partial S}{\partial z}\right)^2 (1)$$
$$\frac{F_{eq}}{S_{eq}}S\left(2 - \frac{1}{S_{eq}}S\right) + \left(\frac{2L_{1s}+1}{6}\right)\left(\frac{\partial S}{\partial z}\right)^2 (2)$$

Analytic Solutions



Theoretical Accuracy

• measure of error: using linear interpolant S_I

 $||e||_{L_{\infty}(0,d)} = \max_{z \in [0,d]} |S_{exact}(z) - S_{I}(z)|$

• for green problem, it can be shown that

$$\|e\|_{L_{\infty}(0,d)} \le \frac{C}{N^2}$$

with both uniform and adaptive grids

• for blue problem, using practical measure

$$l_{\infty} = \max_{j=0,...,N/2} |S_f(z_j) - S_N(z_j)|$$

adaptive grid error is $O(N^{-2})$

Efficiency

• CPU times (in seconds) required to solve blue problem

	Adaptive Grid				
accuracy	N	solve	grid	total	
1×10^{-4}	115	1.9491e-1	7.6075e-4	1.9590e-1	
1×10^{-5}	258	2.2016e-1	9.7614e-4	2.2137e-1	
1×10^{-6}	476	2.4822e-1	1.3882e-3	2.4344e-1	
1×10^{-7}	817	2.7932e-1	1.9825e-3	2.8150e-1	

	Unif	orm Grid	
accuracy	N	total	% speedup
1×10^{-4}	174	1.9371e-1	-1.13
1×10^{-5}	338	2.3071e-1	4.05
1×10^{-6}	568	2.5882e-1	5.94
1×10^{-7}	1051	3.2753e-1	14.05

Order Reconstruction Problem I Barberi et al., Eur. J. Phys. E (2004)

- cell surface treated at boundaries to induce alignments uniformly tilted by a specified tilt angle but oppositely directed
- two topologically different equilibrium states: mostly horizontal alignment with a slight splay, mostly vertical alignment with a bend of almost π radians



Order Reconstruction Problem II

- aim: model order reconstruction which takes place when an electric field is applied
- no longer purely uniaxial: need full Q-tensor
- 5 coupled PDEs for q_i s, plus PDE for electric potential U
- 1D domain $z \in [0, d]$, monitor based on $T(z) = tr(Q^2)$

$$M(T(z)) = \sqrt{\hat{\alpha} + \left(\frac{dT}{dz}\right)^2}$$

• quantify order reconstruction via measure of biaxiality

$$b = \sqrt{1 - \frac{6 \operatorname{tr}(Q^3)^2}{\operatorname{tr}(Q^2)^3}}$$

coded using COMSOL Multiphysics

Numerical Results



- solutions for electric field strength V just below and above the critical voltage at which switching occurs
- adaptive grid with 256 quadratic elements

Grid Trajectories



 approx. 25% fewer points (less CPU time) needed for adaptive grid

Exchange of Eigenvalues



Detail of Biaxial Transition



Solution Accuracy: Uniform Grid



(a) N = 32.





(b) N = 64.



(c) N = 128.

(d) N = 256.

Solution Accuracy: Switching Times

Ν	uniform	adaptive
128	no switching occurs	2.064×10^{-3}
256	2.100×10^{-3}	2.064×10^{-3}
512	2.065×10^{-3}	2.064×10^{-3}
1024	2.064×10^{-3}	2.064×10^{-3}

• when the uniform grid is not fine enough, switching either does not occur at all or occurs at a later time

Equidistribution in 2D

- physical $\mathbf{x} = [x, y]^T$, computational $\xi = [\xi, \eta]^T$
- minimise

$$I[\xi] = \frac{1}{2} \int_D \left[(\nabla \xi)^T G_1^{-1} \nabla \xi + (\nabla \eta)^T G_2^{-1} \nabla \eta \right] d\mathbf{x}$$

 G_1 , G_2 symmetric positive definite monitor matrices

• Euler-Lagrange equations: modified gradient flow

$$\frac{\partial \xi}{\partial t} = \frac{P}{\tau} \nabla \cdot (G_1^{-1} \nabla \xi), \qquad \frac{\partial \eta}{\partial t} = \frac{P}{\tau} \nabla \cdot (G_2^{-1} \nabla \eta) = 0$$

spatial balance operator P temporal smoothing parameter τ

Equidistribution in 2D

- interchange roles of dependent/independent variables
- Winslow-type monitor matrices

$$G_1 = G_2 = \begin{bmatrix} w & 0 \\ 0 & w \end{bmatrix}, \qquad w(\mathbf{x}, t) = \sqrt{\hat{\alpha} + |\nabla[\operatorname{tr}(Q^2)]|^2}$$

• MMPDE

$$\frac{\partial \mathbf{x}}{\partial t} = P(a\mathbf{x}_{\xi\xi} + b\mathbf{x}_{\xi\eta} + c\mathbf{x}_{\eta\eta} + d\mathbf{x}_{\xi} + e\mathbf{x}_{\eta})$$

a, b, c, d, e depend on ω , x_{ξ} , x_{η} , y_{ξ} , y_{η}

coded using COMSOL Multiphysics

2D Test Problem

Zhang et al., Liquid Crystals (2004)

- 2D test problem
 - square cell $[0,d] \times [0,d]$
 - variable pretilt on x = 0, fixed pretilt on x = d
 - periodic boundary conditions on y = 0, y = d

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Summary

- 1D model problem analysis shows
 - adaptive grid error is $O(N^{-2})$ for a simple model problem
 - adaptive grid error appears to be $O(N^{-2})$ for a more realistic model problem
- arc-length monitor function based on $tr(Q^2)$ works well
- to obtain a specified level of accuracy, adaptive grid requires fewer points: inaccurate solutions/switching times can be obtained if uniform grid is not fine enough
- 3-year EPSRC project (from June 1st 2007) with Ainsworth, Mottram (Strathclyde) and Newton (Hewlett-Packard)

Adaptive Numerical Methods for Optoelectronic Devices

- Adaptive Grid Methods for Q-Tensor Theory of Liquid 720 Crystals: A One-Dimensional Feasibility Study Strathclyde Mathematics Research Report No. 13, 2006.
- Adaptive Solution of a One-dimensional Order 724 Reconstruction Problem in Q-Tensor Theory of Liquid Crystals
 Liquid Crystals, 34 (4), pp. 479 - 487, 2007.