

# Efficient preconditioned iterative solvers for director-based models of nematic liquid crystals

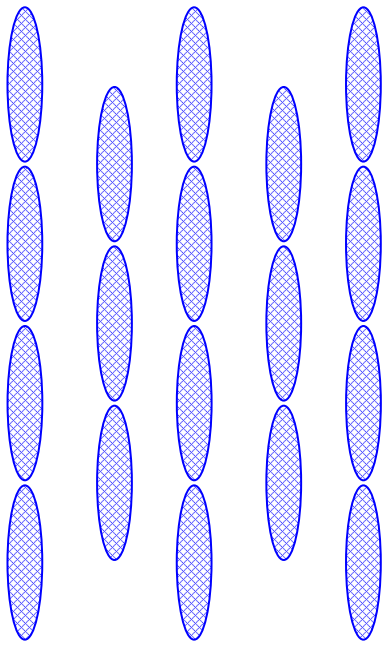
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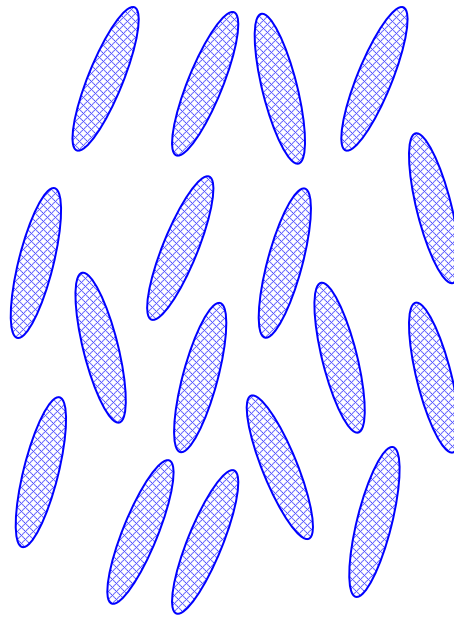
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# Liquid Crystals

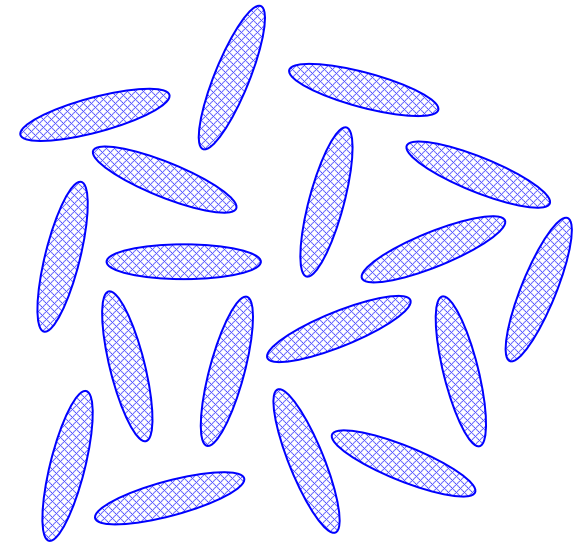
- occur between solid crystal and isotropic liquid states



solid



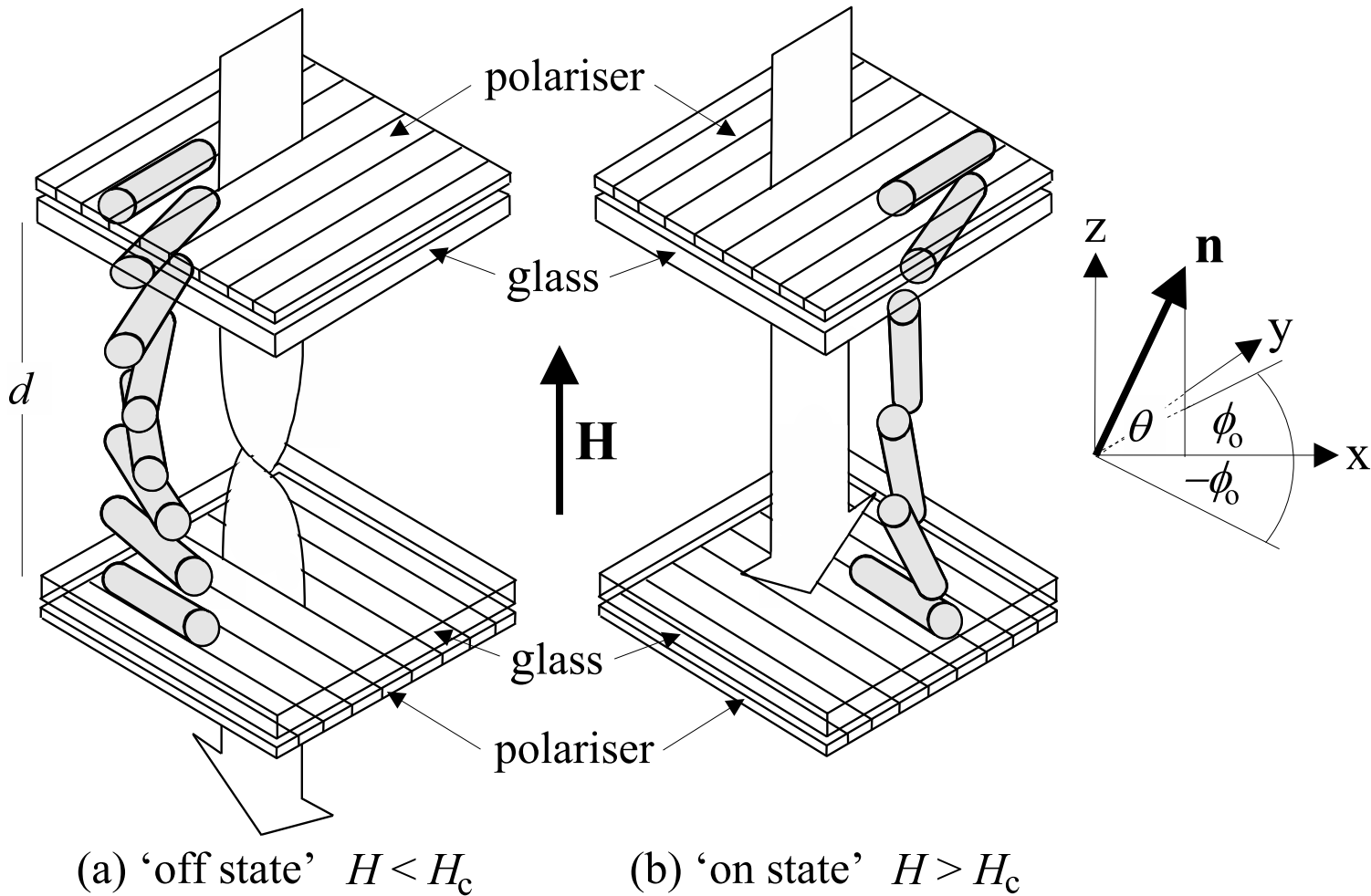
liquid crystal



liquid

- may have different **equilibrium** configurations
- **switch** between stable states by altering applied voltage, magnetic field, boundary conditions, . . .

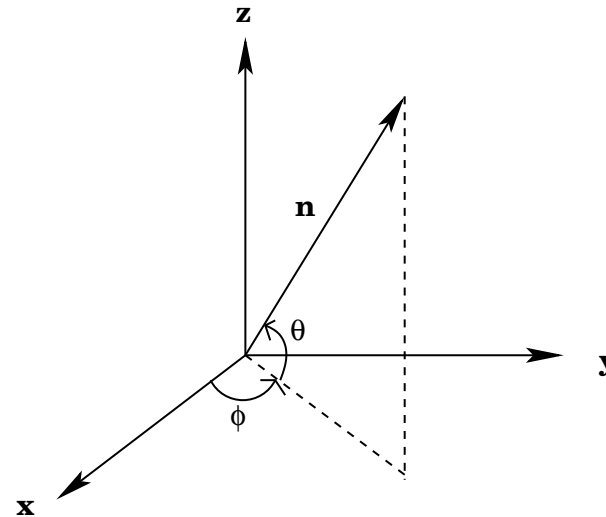
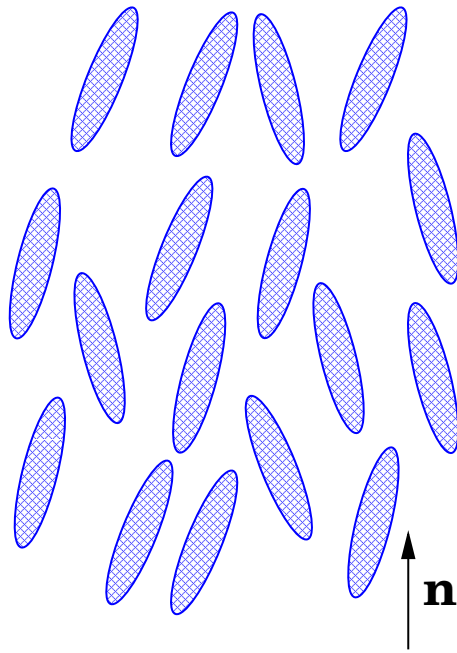
# Liquid Crystal Displays



twisted nematic device

*Static and Dynamic Continuum Theory of Liquid Crystals,*  
*Iain W. Stewart (2004)*

# Modelling: Director-based Models



- **director**: average direction of molecular alignment

unit vector

$$\mathbf{n} = (\cos \theta \cos \phi, \cos \theta \sin \phi, \sin \theta)$$

- $\mathbf{n}$  varies with **space** and **time**

$$\theta \equiv \theta(x, y, z, t), \quad \phi \equiv \phi(x, y, z, t)$$

# Finding Equilibrium Configurations

- minimise the **free energy**

$$\mathcal{F} = \int_V F_{bulk}(\theta, \phi, \nabla\theta, \nabla\phi) + \int_S F_{surface}(\theta, \phi) dS$$

$$F_{bulk} = F_{elastic} + F_{electrostatic}$$

- elastic energy

$$F_{elastic} = \frac{1}{2}K \|\nabla \mathbf{n}\|^2$$

- electrostatic energy

$$F_{electrostatic} = -\frac{1}{2}\epsilon_0\epsilon_{\perp}E^2 - \frac{1}{2}\epsilon_0\epsilon_a(\mathbf{n} \cdot \mathbf{E})^2$$

# Equilibrium Equations

- nondimensionalised equilibrium equations on  $z \in [0, 1]$

$$F = \frac{1}{2} \int_0^1 [(u_z^2 + v_z^2 + w_z^2) - \alpha^2 \pi^2 (\beta + w^2) U_z^2] dz$$

- dimensionless parameters:

$$\alpha^2 = \frac{\epsilon_0 \epsilon_a V^2}{K \pi^2}, \quad \beta = \frac{\epsilon_{\perp}}{\epsilon_a}$$

- director  $\mathbf{n} = (u, v, w)$ , electric potential  $U$ :  $E = -\frac{dU}{dz}$

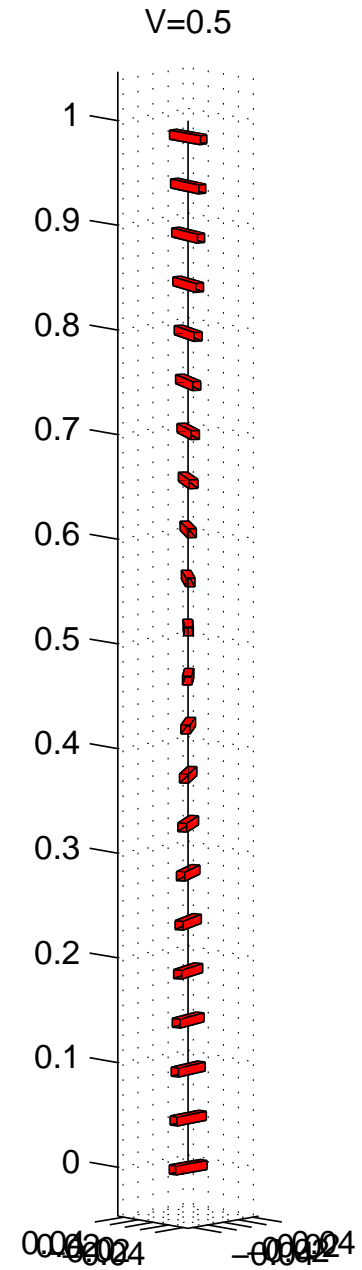
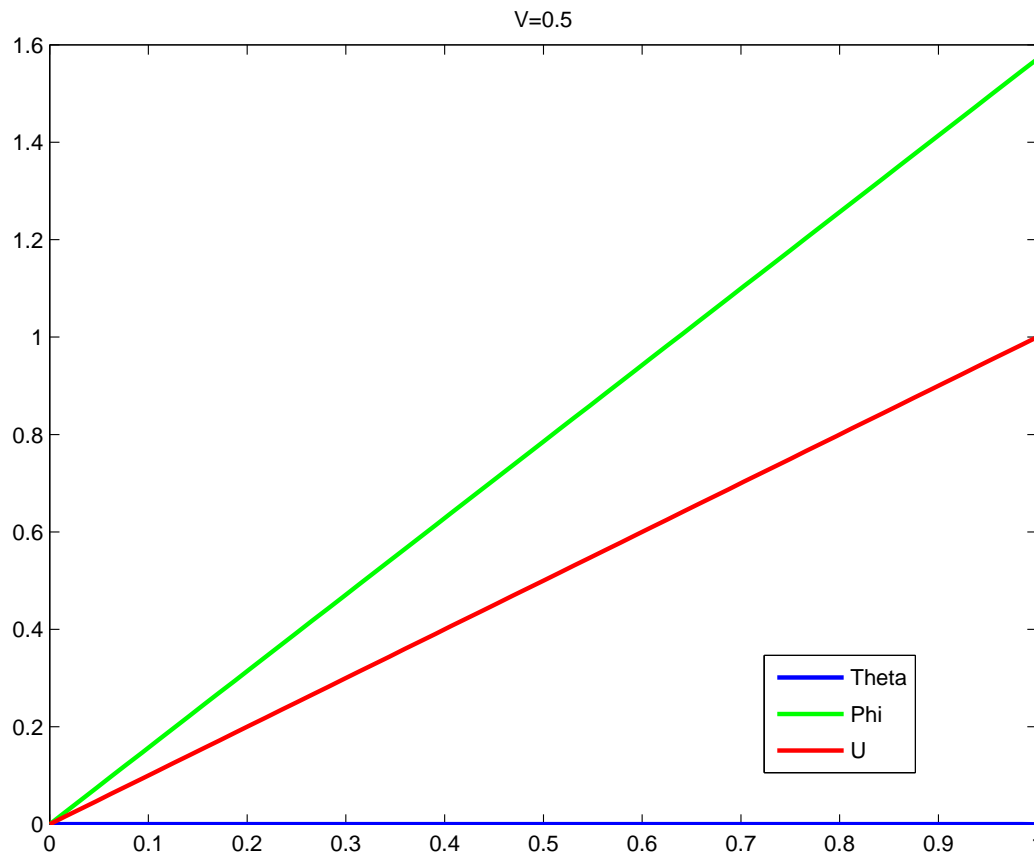
- boundary conditions:

$$\text{at } z = 0: \mathbf{n} = (1, 0, 0), \quad \text{at } z = 1: \mathbf{n} = (0, 1, 0)$$

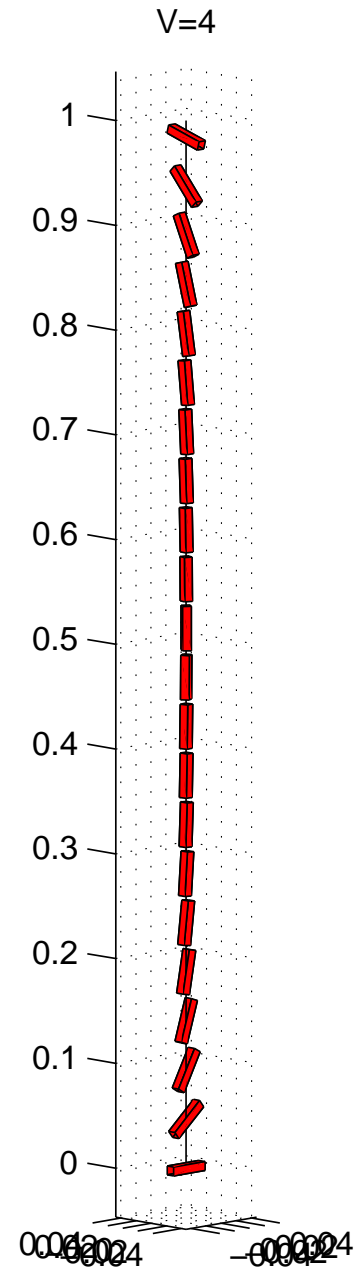
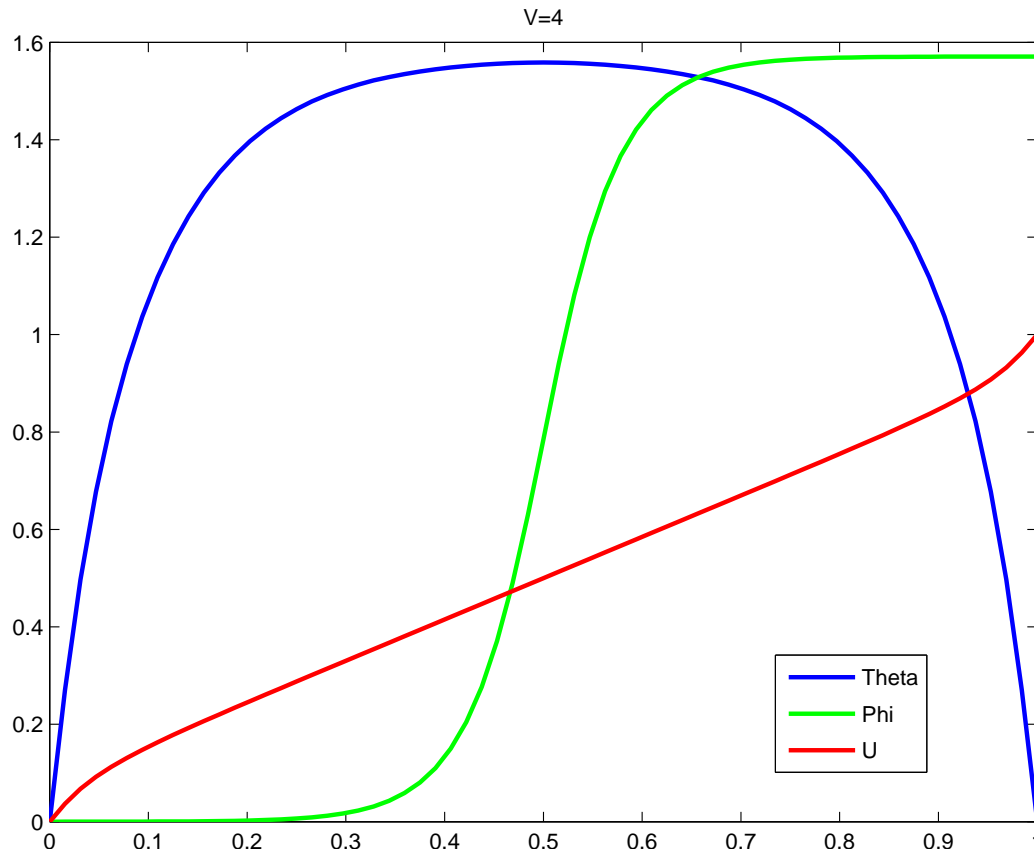
- unknowns  $u, v, w, U$

# Off State

$$\theta(z) \equiv 0, \quad \phi(z) = \frac{\pi}{2}z, \quad U(z) = z$$



# On State

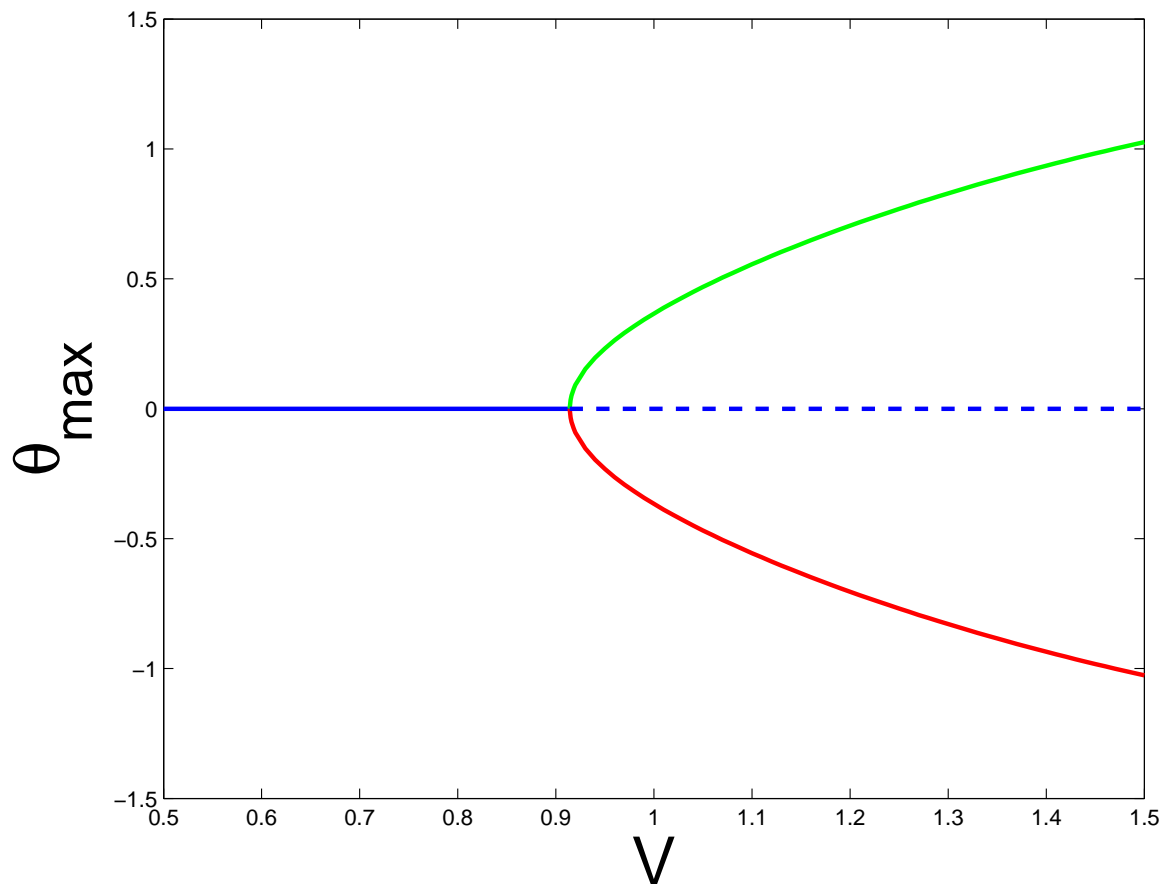




# Critical Voltage

- switching occurs at

$$V_c = \frac{\pi}{2} \sqrt{\frac{3K}{\epsilon_0 \epsilon_a}}$$



# Discrete Free Energy

- grid of  $N + 1$  points  $z_k$  a distance  $\Delta z$  apart,  $n = N - 1$  unknowns for each variable
- **piecewise linear** approximation, weighted average

$$F \approx \frac{\Delta z}{2} \sum_{k=0}^{N-1} \left\{ \left[ \frac{u_{k+1} - u_k}{\Delta z} \right]^2 + \left[ \frac{v_{k+1} - v_k}{\Delta z} \right]^2 + \left[ \frac{w_{k+1} - w_k}{\Delta z} \right]^2 - \alpha^2 \pi^2 \left( \beta + \left[ \frac{w_k^2 + w_{k+1}^2}{2} \right] \right) \left[ \frac{U_{k+1} - U_k}{\Delta z} \right]^2 \right\}$$

- equivalent to **mid-point** finite differences, **linear** finite elements

# Constrained Minimisation

- minimise  $F$  subject to pointwise constraint

$$u_j^2 + v_j^2 + w_j^2 = 1, \quad j = 1, \dots, n$$

- constraints applied via **Lagrange multipliers**: minimise

$$G = \frac{\Delta z}{2} [f - \dots - \lambda_j(u_j^2 + v_j^2 + w_j^2 - 1) - \dots]$$

- solve  $\nabla \mathbf{G}(\mathbf{x}) = \mathbf{0}$  for  $\mathbf{x} = [\mathbf{u}, \mathbf{v}, \mathbf{w}, \lambda, \mathbf{U}]$

$$N + 1 \text{ gridpoints} \Rightarrow n = N - 1 \text{ unknowns}$$

- use Newton's method:  $5n \times 5n$  **Hessian**  $\nabla^2 \mathbf{G}(\mathbf{x})$

$$\nabla^2 \mathbf{G} = \begin{bmatrix} \nabla_{\mathbf{nn}}^2 \mathbf{G} & \nabla_{\mathbf{n}\lambda}^2 \mathbf{G} & \nabla_{\mathbf{nU}}^2 \mathbf{G} \\ \nabla_{\lambda\mathbf{n}}^2 \mathbf{G} & \nabla_{\lambda\lambda}^2 \mathbf{G} & \nabla_{\mathbf{U}\lambda}^2 \mathbf{G} \\ \nabla_{\mathbf{U}\mathbf{n}}^2 \mathbf{G} & \nabla_{\lambda\mathbf{U}}^2 \mathbf{G} & \nabla_{\mathbf{UU}}^2 \mathbf{G} \end{bmatrix}$$

# Hessian Components

- matrix notation:  $\nabla_{\mathbf{nn}}^2 \mathbf{G} = A$   $A \in \mathbb{R}^{3n \times 3n}$
- $A$  is block diagonal with **symmetric tridiagonal** blocks
- $A$  can be **indefinite**
  
- matrix notation:  $\nabla_{\mathbf{n}\lambda}^2 \mathbf{G} = B$   $B \in \mathbb{R}^{3n \times n}$
- $B = -\Delta z [B_u, B_v, B_w]^T$ ,  $B_u, B_v, B_w$  are **diagonal**
- $B^T B = \Delta z^2 I_n$  when constraints are satisfied
  
- matrix notation:  $\nabla_{\mathbf{UU}}^2 \mathbf{G} = -C$   $C \in \mathbb{R}^{n \times n}$
- $C$  is **symmetric positive definite** and **tridiagonal**
  
- matrix notation:  $\nabla_{\mathbf{nU}}^2 \mathbf{G} = D$   $D \in \mathbb{R}^{3n \times n}$
- $D = k[0, 0, D_w]^T$ ,  $D_w$  is **tridiagonal**,  $\text{rank}(D) = n - 1$

# Iterative solution

- full Hessian matrix:

$$H = \begin{bmatrix} A & B & D \\ B^T & 0 & 0 \\ D^T & 0 & -C \end{bmatrix}$$

saddle-point problem

- outer iteration: **Newton's method**  $\text{tol}=1e-4$
- inner iteration: **MINRES**  $\text{tol}=1e-4$
- Hessian is poorly conditioned:

$$\kappa(H) = O(N^4)$$

# Nullspace Method

- **Idea:** use information about nullspace of  $B$  to eliminate constraint blocks

$$A\delta\mathbf{n} + B\delta\lambda + D\delta\mathbf{U} = -\nabla_{\mathbf{n}}G \quad (1)$$

$$B^T\delta\mathbf{n} = -\nabla_{\lambda}G \quad (2)$$

$$D^T\delta\mathbf{n} - C\delta\mathbf{U} = -\nabla_{\mathbf{U}}G \quad (3)$$

- write solution of (2) as  $\delta\mathbf{n} = \widehat{\delta\mathbf{n}} + Z\mathbf{z}$  where columns of  $Z \in \mathbb{R}^{3n \times 2n}$  form a basis for  $\mathcal{N}(B^T)$
- reduced  $3n \times 3n$  system:

$$\begin{bmatrix} Z^T A Z & Z^T D \\ D^T Z & -C \end{bmatrix} \begin{bmatrix} \mathbf{z} \\ \delta\mathbf{U} \end{bmatrix} = \begin{bmatrix} -Z^T (\nabla_{\mathbf{n}}G + A\widehat{\delta\mathbf{n}}) \\ -\nabla_{\mathbf{U}}G - D^T\widehat{\delta\mathbf{n}} \end{bmatrix}$$

- $B^T B$  is **diagonal** so solves with  $(B^T B)^{-1}$  are cheap

# Nullspace of $B^T$

- permute entries of B:

$$B = -\Delta z \operatorname{diag}(\mathbf{n}_1, \mathbf{n}_2, \dots, \mathbf{n}_n), \quad \mathbf{n}_j = [u_j, v_j, w_j]^T$$

- eigenvectors of **orthogonal projection**  $I - \mathbf{n}_j \otimes \mathbf{n}_j$  are orthogonal to  $\mathbf{n}_j$ , e.g.

$$\mathbf{l}_j = \frac{1}{\sqrt{u_j^2 + v_j^2}} \begin{bmatrix} -v_j \\ u_j \\ 0 \end{bmatrix}, \quad \mathbf{m}_j = \frac{1}{\sqrt{u_j^2 + v_j^2}} \begin{bmatrix} -u_j w_j \\ -v_j w_j \\ u_j^2 + v_j^2 \end{bmatrix}$$

- at least one of  $u_j, v_j, w_j$  nonzero as  $|\mathbf{n}_j| = 1$
- $3n \times 2n$  nullspace matrix

$$Z = \operatorname{diag}([\mathbf{l}_1, \mathbf{m}_1], [\mathbf{l}_2, \mathbf{m}_2], \dots, [\mathbf{l}_n, \mathbf{m}_n])$$

# Solving the Reduced System

- write  $\bar{A} = Z^T A Z$  and  $\bar{D} = Z^T D$ :

$$\mathcal{H} = \begin{bmatrix} \bar{A} & \bar{D} \\ \bar{D}^T & -C \end{bmatrix}$$

- block preconditioner:  $\mathcal{P} = \begin{bmatrix} \bar{A} & 0 \\ 0 & C \end{bmatrix}$
- preconditioned matrix:

$$\tilde{\mathcal{H}} = \mathcal{P}^{-1/2} \mathcal{H} \mathcal{P}^{-1/2} = \begin{bmatrix} I & M^T \\ M & -I \end{bmatrix}$$

$$M = C^{-1/2} \bar{D} \bar{A}^{-1/2}$$

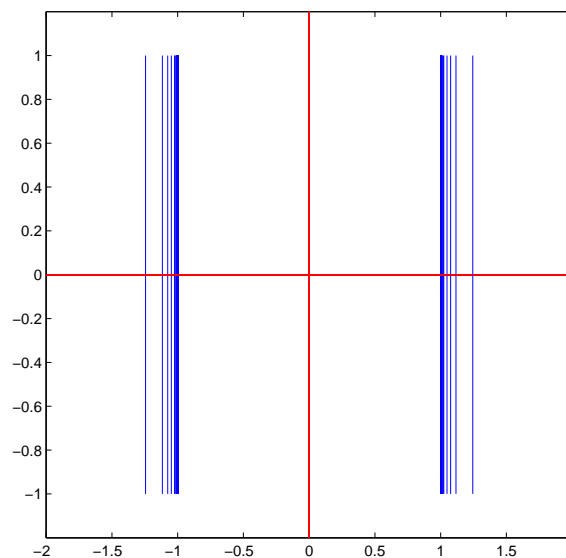
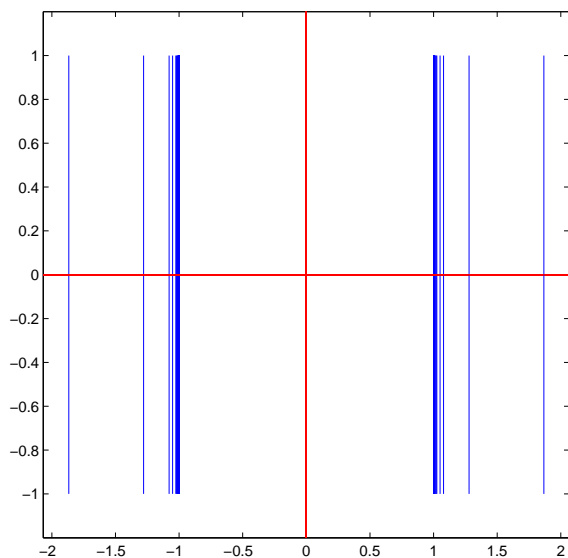
- off state:  $M = 0$



# Preconditioned Spectrum

$$M = C^{-1/2} Z^T D (Z^T A Z)^{-1/2}$$

- non-zero singular values  $\sigma_k$ ,  $k = 1, \dots, n - 1$
- eigenvalues of  $\tilde{\mathcal{H}}$  are
  - (i) **1** with multiplicity  $n + 1$
  - (ii) **-1** with multiplicity  $1$
  - (iii)  $\pm \sqrt{1 + \sigma_k^2}$  for  $k = 1, \dots, n - 1$



sample eigenvalue plots for  $N = 64$

# Estimate of MINRES convergence

- eigenvalues in two symmetric intervals

$$[-\beta, -1] \cup [1, \beta], \quad \beta = \sqrt{1 + \sigma_{\max}^2}$$

$N$	first Newton step	last Newton step
32	1.3563e+00	1.5472e+00
64	1.3579e+00	1.5479e+00
128	1.3583e+00	1.5481e+00
256	1.3584e+00	1.5482e+00

- to achieve  $\|\mathbf{r}_k\| \leq \epsilon \|\mathbf{r}_0\|$  need

$$k \simeq \frac{1}{2} \sqrt{1 + \sigma_{\max}^2} \ln \left( \frac{2}{\epsilon} \right)$$

- for  $\epsilon = 1e - 4$ ,  $k \simeq 6$

# Iteration Counts

- diagonal scaling

$N$	8	16	32	64	128	256
first Newton step	15	40	117	382	1293	5126
last Newton step	37	134	414	1617	7466	34755

# Iteration Counts

- diagonal scaling

$N$	8	16	32	64	128	256
first Newton step	15	40	117	382	1293	5126
last Newton step	37	134	414	1617	7466	34755

- reduced block preconditioning

$N$	8	16	32	64	128	256
first Newton step	5	5	5	5	5	5
last Newton step	5	5	5	5	5	5

- independent of problem size and Newton iteration

# Computing Time

- elapsed time (tic/toc)
- A: **full** direct, B: **reduced** direct, C: **reduced** block

$N$	A	B	C
8	7.54e-02	7.17e-02	2.85e-03
16	7.67e-03	7.37e-03	2.60e-03
32	1.11e-02	1.06e-02	3.51e-03
64	1.67e-02	1.56e-02	4.95e-03
128	3.55e-02	3.30e-02	8.62e-03
256	1.18e-01	1.26e-01	1.26e-02
512	4.89e-01	4.40e-01	2.26e-02
1024	1.40e+00	1.37e+00	4.64e-02
2048	5.25e+00	5.15e+00	1.12e-01
4096	2.11e+01	2.12e+01	1.78e-01

# Summary and Conclusions

- The **reduced block preconditioner** is very efficient for this problem.
- The **nullspace method** seems to be ideal for dealing with the unit vector constraint.
- Although this example features a 1D cell, the same method should work well for more complicated liquid crystal cells as the **orientation space** and **unit vector constraint** is exactly the same.
- The reduced block preconditioner is cheap to invert here: hopefully **approximate solves** (e.g. multigrid) can be used for other problems with less structure.

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THANKS!

# Diagonal Preconditioning

$$H = \begin{bmatrix} A & B & D \\ B^T & 0 & 0 \\ D^T & 0 & -C \end{bmatrix}$$

$$\mathcal{D} = \begin{bmatrix} D_A & 0 & 0 \\ 0 & \Delta z^3 I & 0 \\ 0 & 0 & D_C \end{bmatrix} \quad \begin{array}{l} D_A = \text{diag}(A) \\ D_C = \text{diag}(C) \end{array}$$



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- estimated condition of  $\mathcal{D}^{-1}H$  is  $O(N^2)$

$$\lambda_{\min} = -2, \quad \lambda_s = O(N^{-2}), \quad \lambda_{s+1} = O(N^{-2}), \quad \lambda_{\max} = 2$$