

*Two Finite Difference Methods
for Poisson-Boltzmann
Equation*

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Outline of this talk:

- I-L. Chern, J-L. Liu and W-C Wang, "Accurate Evaluation for Poisson-Boltzmann Equations with Interfaces," *Methods and Applications of Analysis*, Vol. 10, No. 2, pp. 309-328 (2003).
- Chern, I-Liang and Yu-Chen Shu, "Coupling interface method for elliptic interface problems," *Journal of Computational Physics*, Vol. 225, No. 2, pp.2138-2174 (2007).

Outline of First Part:

- The problem: study the electrostatics for macromolecule in mobile ionic solution
- Numerical issue: less accuracy on electronic force due to
 - singular point charges
 - interface problems
 - open domain problem
- Treatments of singularities
 - singular charges: multipole method
 - interface problems: jump condition capturing method
 - open domain problem: coordinate patching
- Numerical examples

Outline of Second Part:

- Elliptic Interface Problems
- Coupling interface method
 - 1d: CIM1, CIM2
 - 2d: CIM2
 - d dimension
 - Hybrid CIM
 - Numerical validation
- Application to the Poisson-Boltzmann equation:
 - Treatment of complex interfaces

The problem

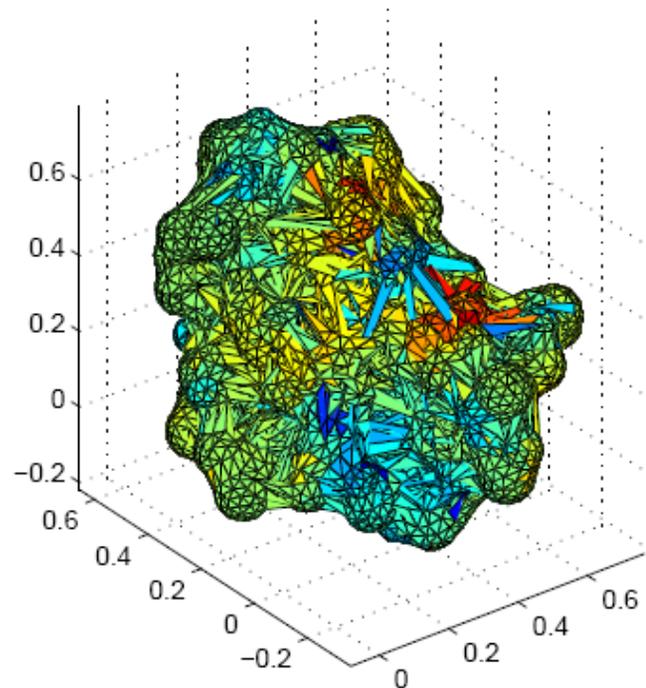
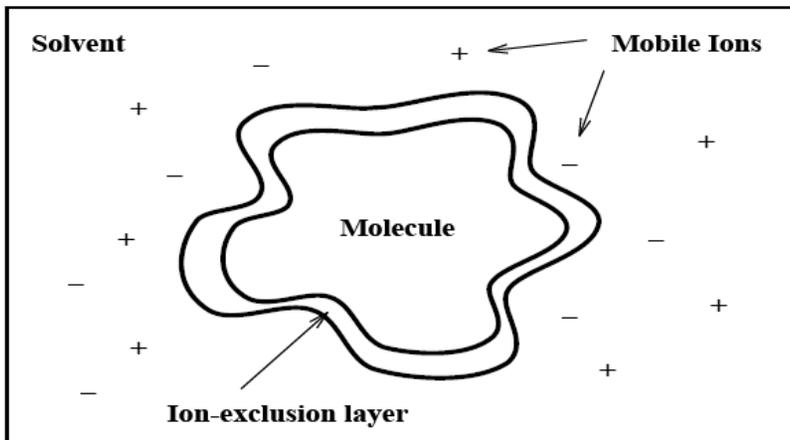
- Motivation:

- understanding functions of macromolecules in solutions
- drug design

- Goal : study corresponding electrostatics

The model

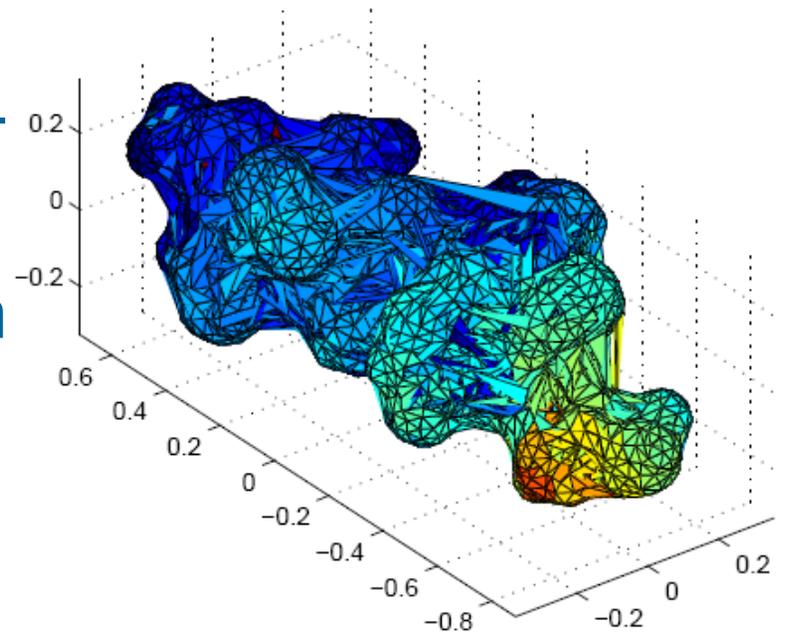
- macromolecule: a structured and polarized clusters of atoms
- ionic solvent: a continuum (Debye and Huckel 1924)



Biomolecule in solvent

Poisson-Boltzmann model

- Macromolecule: 50 Å
- Hydrogen layer: 1.5 – 3 Å
- Molecule surface: thin
- Dielectric constants:
 - 2 inside molecule
 - 80 in water



a hydrophilic protein (PDB ID:1DNG)

Equations

- **The Poisson Equation**

$$-\nabla \cdot [\epsilon(\mathbf{x})\nabla\Phi(\mathbf{x})] = K(\mathbf{x})(\rho_+ + \rho_-) + 4\pi Q(x)$$

- **Boltzmann distribution**

$$\rho_+ = \left(\frac{k_B T}{e_c}\right) \exp\left(-\frac{e_c \Phi(\mathbf{x})}{k_B T}\right)$$

$$\rho_- = \left(\frac{k_B T}{-e_c}\right) \exp\left(\frac{e_c \Phi(\mathbf{x})}{k_B T}\right)$$

where

$$Q(x) = \sum_{i=1}^{N_m} q_i \delta(\mathbf{x} - \mathbf{x}_i)$$

$$\epsilon(\mathbf{x}) = \begin{cases} \epsilon_1, & \mathbf{x} \in \Omega_1, \\ \epsilon_2, & \mathbf{x} \in \Omega_2 \cup \Omega_3. \end{cases}$$

$$K(\mathbf{x}) = \begin{cases} 0, & \mathbf{x} \in \Omega_1 \cup \Omega_2, \\ \bar{\kappa}^2, & \mathbf{x} \in \Omega_3, \end{cases}$$

* $\epsilon(\mathbf{x})$ the dielectric parameter,

* e_c the charge of an electron,

* k_B the Boltzmann constant,

* T the temperature,

* \mathbf{x}_i the atomic location,

* q_i the atomic partial charge,

[Brief Review]

Previous numerical methods (from late 80 to now) for PBE can be classified into

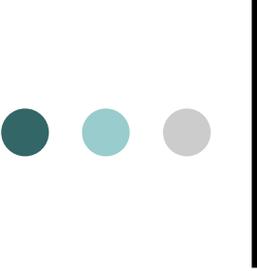
- Finite difference methods
- Finite element methods
- Boundary element methods

■ Review article: B.Z. Lu, Y.C. Zhou, M.J. Holst, J.A. McCammon, CiCP 2008

[Numerical issues]

less accuracy on electronic force due to

- singular point charges
- interface problems (discontinuity of dielectric function)
- open domain problem



Treatment of point charge singularities

separate singular part

$$\phi = \bar{\phi} + \tilde{\phi}.$$

where

$$\bar{\phi}(\mathbf{x}) = \begin{cases} \phi^*(\mathbf{x}) + \phi^0(\mathbf{x}) & \mathbf{x} \in \Omega_1 \\ 0 & \mathbf{x} \in \Omega_2 \cup \Omega_3 \end{cases}.$$

and ϕ^* is the potential in the free space induced by Q , i.e.

$$\phi^*(\mathbf{x}) = \begin{cases} C \sum_{i=1}^m \frac{1}{\epsilon_1} \frac{z_i}{4\pi |\mathbf{x} - \mathbf{x}_i|}, & \mathbf{x} \in R^3 \\ C \sum_{i=1}^m -\frac{1}{\epsilon_1} \frac{z_i}{2\pi} \log(|\mathbf{x} - \mathbf{x}_i|), & \mathbf{x} \in R^2 \end{cases}.$$

ϕ^0 is a harmonic function in Ω_1 satisfying

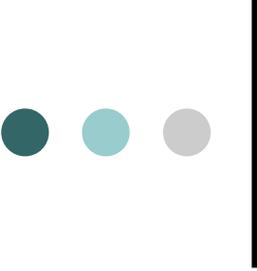
$$\begin{cases} \Delta \phi^0 = 0 & \text{in } \Omega_1 \\ \phi^0 = -\phi^* & \text{on } \Gamma_1. \end{cases}$$

The introduction of ϕ^0 is to force $[\bar{\phi}] = 0$ across Γ_1 .

The correction potential satisfies

$$-\nabla \cdot \left(\epsilon(\mathbf{x}) \nabla \tilde{\phi}(\mathbf{x}) \right) + K(\mathbf{x}) \sinh(\tilde{\phi}(\mathbf{x})) = [\epsilon \bar{\phi}_n]_{\Gamma_1} \delta_{\Gamma_1}.$$

Thus, the point charge singularity is transferred into surface singularity.



Damped Newton's Method

$$-\nabla \cdot (\epsilon(\mathbf{x}) \nabla v^l) + K(\mathbf{x}) \cosh(\phi^l) v^l = \nabla \cdot (\epsilon(\mathbf{x}) \nabla \tilde{\phi}^l) - K(\mathbf{x}) \sinh(\phi^l) + [\epsilon \bar{\phi}_n]_{\Gamma_1}$$
$$\tilde{\phi}^{l+1} = \tilde{\phi}^l + v^l \quad (1)$$

Since the direction v^n is indeed a descent direction for the functional $E(\phi)$,

$$E(\phi^l + \lambda^l v^l) < E(\phi^l) \text{ for small } \lambda^l > 0,$$

we can accelerate the convergence of the Newton's method globally by performing a line search to find a suitable damping parameter λ^l that minimizes $E(\phi^l + \lambda^l v^l)$ and replace (1) by

$$\tilde{\phi}^{l+1} = \tilde{\phi}^l + \lambda^l v^l.$$



Treatment of surface singularities

In nonlinear iteration, we need to solve the following linear Poisson-Boltzmann equation:

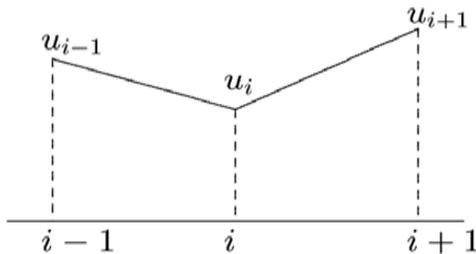
$$-\nabla \cdot \left(\epsilon(\mathbf{x}) \nabla \tilde{\phi}(\mathbf{x}) \right) + H(\mathbf{x}) \tilde{\phi}(\mathbf{x}) = f(\mathbf{x}) + k\delta|_{\Gamma_1}, \quad . \quad (2)$$

where $\epsilon(\mathbf{x})$ is discontinuous across Γ_1 , and $H(\mathbf{x})$ and $f(\mathbf{x})$ are discontinuous across Γ_1 .

- **the jump condition capturing scheme**
- **skew variable**

Jump condition capturing scheme

1-D Case



- uniform grid
- interface is on the grid
- The 1-d equation with singular source

$$(\epsilon(x)u')' = f + k\delta(x - x_i)$$

is discretized by

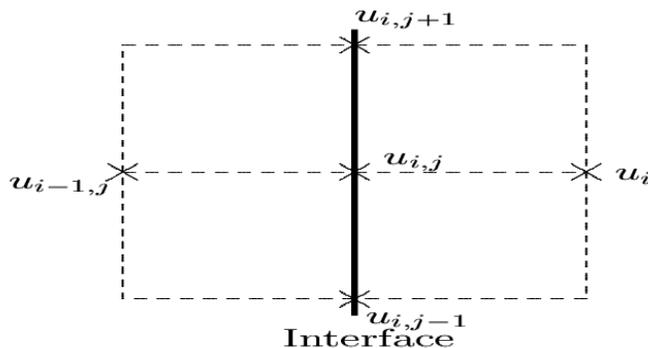
$$\begin{aligned} & \frac{1}{\Delta x} \left(\epsilon_{i+1/2} \left(\frac{u_{i+1} - u_i}{\Delta x} \right) - \epsilon_{i-1/2} \left(\frac{u_i - u_{i-1}}{\Delta x} \right) \right) \\ &= \frac{1}{2} (f_{i,+} + f_{i,-}) + \frac{k}{\Delta x} \end{aligned}$$

- truncation error
 - * $O(\Delta x^2)$ off the interface
 - * $O(\Delta x)$ on the interface
- global error $O(\Delta x^2)$

Jump condition capturing scheme

- 2-D Case: (A wrong approach)

$$\begin{aligned} & \frac{1}{\Delta x} \left(\epsilon_{i+1/2,j} \left(\frac{u_{i+1,j} - u_{i,j}}{\Delta x} \right) - \epsilon_{i-1/2,j} \left(\frac{u_{i,j} - u_{i-1,j}}{\Delta x} \right) \right) \\ & + \frac{1}{\Delta y} \left(\bar{\epsilon}_{i,j+1/2} \left(\frac{u_{i,j+1} - u_{i,j}}{\Delta y} \right) - \bar{\epsilon}_{i,j-1/2} \left(\frac{u_{i,j} - u_{i,j-1}}{\Delta y} \right) \right) \\ & = \frac{1}{2} (f_{i+,j} + f_{i-,j}) + \frac{k}{\Delta x} \end{aligned}$$

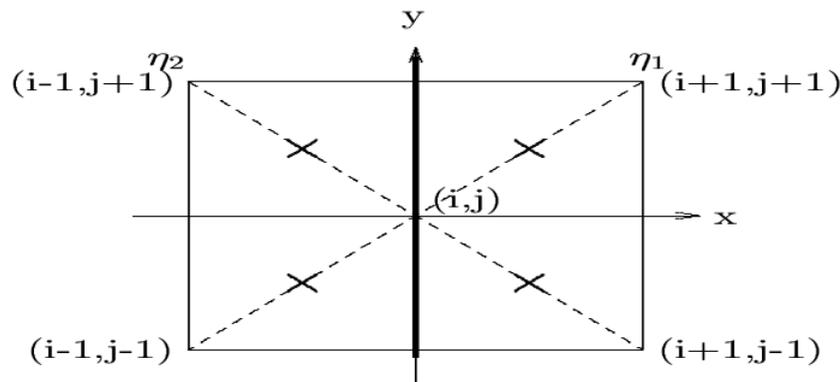


- $\bar{\epsilon}_{i,j+1/2}$ is not defined.
- The local truncation error is $O(1)$.

Jump condition capturing scheme

- 2-D Case: (A right approach) --- Use skew variable

$$\begin{aligned} & \frac{1}{\Delta\eta_1} \left(\epsilon_{i+1/2, j+1/2} \left(\frac{u_{i+1, j+1} - u_{i, j}}{\Delta\eta_1} \right) - \epsilon_{i-1/2, j-1/2} \left(\frac{u_{i, j} - u_{i-1, j-1}}{\Delta\eta_1} \right) \right) \\ & + \frac{1}{\Delta\eta_2} \left(\epsilon_{i-1/2, j+1/2} \left(\frac{u_{i-1, j+1} - u_{i, j}}{\Delta\eta_2} \right) - \epsilon_{i+1/2, j-1/2} \left(\frac{u_{i, j} - u_{i+1, j-1}}{\Delta\eta_2} \right) \right) \\ & = \frac{1}{2} (f_{i+, j} + f_{i-, j}) + k \left(\frac{1}{\Delta\eta_1} + \frac{1}{\Delta\eta_2} \right) \end{aligned}$$

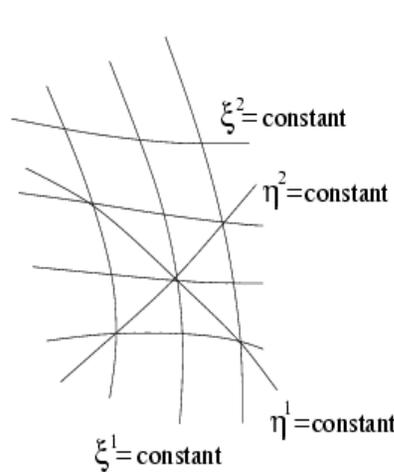
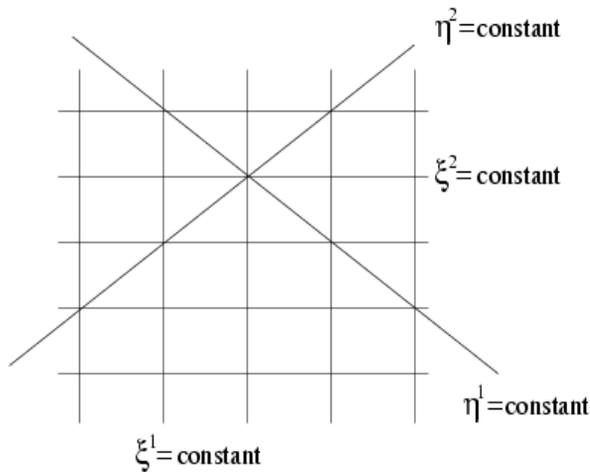


$$\frac{\partial}{\partial\eta_1} \left(\epsilon \frac{\partial u}{\partial\eta_1} \right) + \frac{\partial}{\partial\eta_2} \left(\epsilon \frac{\partial u}{\partial\eta_2} \right) = f + k\delta|_{\Gamma_1}$$

General Case

Let us denote by (ξ^1, ξ^2) the variables in the computational domain, $\mathbf{X}(\xi^1, \xi^2) \in R^2$ the position vector in the physical space.

we define



$$\eta^1 = \frac{\xi^1 \Delta \xi^2 + \xi^2 \Delta \xi^1}{\sqrt{(\Delta \xi^1)^2 + (\Delta \xi^2)^2}}$$

$$\eta^2 = \frac{\xi^2 \Delta \xi^1 - \xi^1 \Delta \xi^2}{\sqrt{(\Delta \xi^1)^2 + (\Delta \xi^2)^2}}$$

$$\Delta \eta^1 = \Delta \eta^2 = \frac{2 \Delta \xi^1 \Delta \xi^2}{\sqrt{(\Delta \xi^1)^2 + (\Delta \xi^2)^2}}$$

finite difference discretization in the skewed variable

$$-\partial_\mu(\epsilon\sqrt{\hat{g}}\hat{g}^{\mu\nu}\partial_\nu\tilde{\phi}) + \sqrt{\hat{g}}H(x)\tilde{\phi} = \sqrt{\hat{g}}f$$

$$\hat{g}^{\mu\nu} = \langle \nabla\eta^\mu, \nabla\eta^\nu \rangle$$

$$\hat{g}_{\mu\nu} = \left\langle \frac{\partial\mathbf{X}}{\partial\eta^\mu}, \frac{\partial\mathbf{X}}{\partial\eta^\nu} \right\rangle$$

$$\hat{g} = \det(\hat{g}_{\mu\nu})$$

$$\sqrt{\hat{g}} = \det\left(\frac{\partial\mathbf{X}}{\partial\boldsymbol{\eta}}\right) = \frac{1}{2} \begin{pmatrix} \Delta\xi^1 & \Delta\xi^2 \\ \Delta\xi^2 & \Delta\xi^1 \end{pmatrix} \det\left(\frac{\partial\mathbf{X}}{\partial\xi}\right)$$

$$\hat{g}^{\mu\gamma}\hat{g}_{\gamma\nu} = \delta_\nu^\mu$$

finite difference discretization in the skewed variable

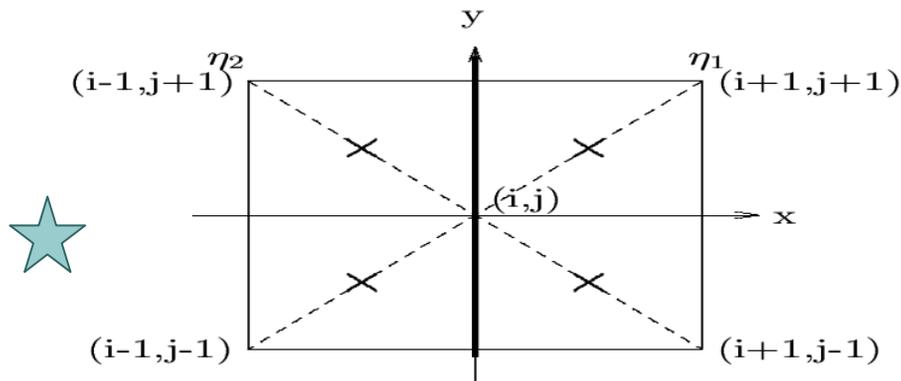
the jump conditions can be incorporated into the following finite difference discretization

$$\begin{aligned}
 & -\frac{1}{(\Delta\eta)^2} \left(\begin{aligned}
 & (\epsilon\sqrt{\hat{g}_h}\hat{g}_h^{11})_{i+\frac{1}{2},j+\frac{1}{2}}(\tilde{\phi}_{i+1,j+1} - \tilde{\phi}_{i,j}) - (\epsilon\sqrt{\hat{g}_h}\hat{g}_h^{11})_{i-\frac{1}{2},j-\frac{1}{2}}(\tilde{\phi}_{i,j} - \tilde{\phi}_{i-1,j-1}) \\
 & + (\epsilon\sqrt{\hat{g}_h}\hat{g}_h^{12})_{i+\frac{1}{2},j+\frac{1}{2}}(\tilde{\phi}_{i,j+1} - \tilde{\phi}_{i+1,j}) - (\epsilon\sqrt{\hat{g}_h}\hat{g}_h^{12})_{i-\frac{1}{2},j-\frac{1}{2}}(\tilde{\phi}_{i-1,j} - \tilde{\phi}_{i,j-1}) \\
 & + (\epsilon\sqrt{\hat{g}_h}\hat{g}_h^{21})_{i-\frac{1}{2},j+\frac{1}{2}}(\tilde{\phi}_{i,j+1} - \tilde{\phi}_{i-1,j}) - (\epsilon\sqrt{\hat{g}_h}\hat{g}_h^{21})_{i+\frac{1}{2},j-\frac{1}{2}}(\tilde{\phi}_{i+1,j} - \tilde{\phi}_{i,j-1}) \\
 & + (\epsilon\sqrt{\hat{g}_h}\hat{g}_h^{22})_{i-\frac{1}{2},j+\frac{1}{2}}(\tilde{\phi}_{i-1,j+1} - \tilde{\phi}_{i,j}) - (\epsilon\sqrt{\hat{g}_h}\hat{g}_h^{22})_{i+\frac{1}{2},j-\frac{1}{2}}(\tilde{\phi}_{i,j} - \tilde{\phi}_{i+1,j-1})
 \end{aligned} \right) \\
 & = \sqrt{(\hat{g}_h)_{i,j}}(f_{i,j} - H_{i,j}\tilde{\phi}_{i,j})
 \end{aligned}$$

At the interface, the right-hand side is replaced by

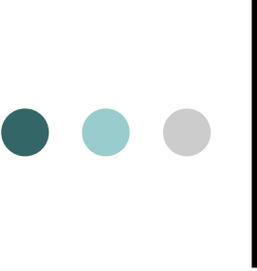
$$\begin{aligned}
 & \sqrt{(\hat{g}_h)_{i,j}} \left(\frac{1}{2}(f_{i^+,j} + f_{i^-,j}) - \frac{1}{2}(H_{i^+,j}\tilde{\phi}_{i^+,j} + H_{i^-,j}\tilde{\phi}_{i^-,j}) \right) \\
 & + \text{terms involving } [\tilde{\phi}] \text{ and } [\epsilon\tilde{\phi}_n]
 \end{aligned}$$

finite difference discretization in the skewed variable



$$\begin{aligned}
 & ((\hat{g}_h)_{11})_{i+\frac{1}{2}, j+\frac{1}{2}} \\
 & \langle D_1 \mathbf{X}_{i+\frac{1}{2}, j+\frac{1}{2}}, D_1 \mathbf{X}_{i+\frac{1}{2}, j+\frac{1}{2}} \rangle \\
 & \left\langle \frac{\mathbf{X}_{i+1, j+1} - \mathbf{X}_{i, j}}{\Delta \eta}, \frac{\mathbf{X}_{i+1, j+1} - \mathbf{X}_{i, j}}{\Delta \eta} \right\rangle
 \end{aligned}$$

- ★ The local truncation error is $O(\Delta \eta)$ on the interface and $O(\Delta \eta^2)$ elsewhere.
- ★ The resulting flux function is 2nd order accurate, even on the interface.
- ★ The symmetry and positivity of this discretization is essentially unconditional.



Summary of the Algorithm

- **Step 1** Evaluate the free space Poisson kernel $\phi^*(\mathbf{x})$ and $\phi_n^*(\mathbf{x})$ for $\mathbf{x} \in \Gamma_1$ (by multipole method).
- **Step 2** Evaluate $\phi_n^0(x)$ for $x \in \Gamma_1$ by solving the Laplace equation in Ω_1 so that $\tilde{\phi} = 0$ on Γ_1 .
- **Step 3** Compute the correction potential $\tilde{\phi}$ by the damped Newton's method. The resulting linearized Poisson Boltzmann equation is discretized using the jump condition capturing scheme. The initial trial of $\tilde{\phi}$ of the iteration is set to be zero.

Numerical experiments

Parameters: $K^+ = 2$, $\varepsilon^- = 2$, $\varepsilon^+ = 80$, $n_r = 32$, $n_\theta = 96$

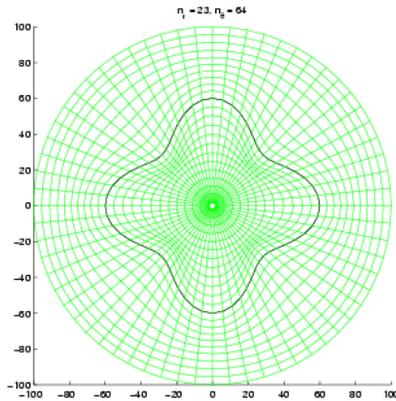


Figure 5: The shape of the macromolecule for Example 1

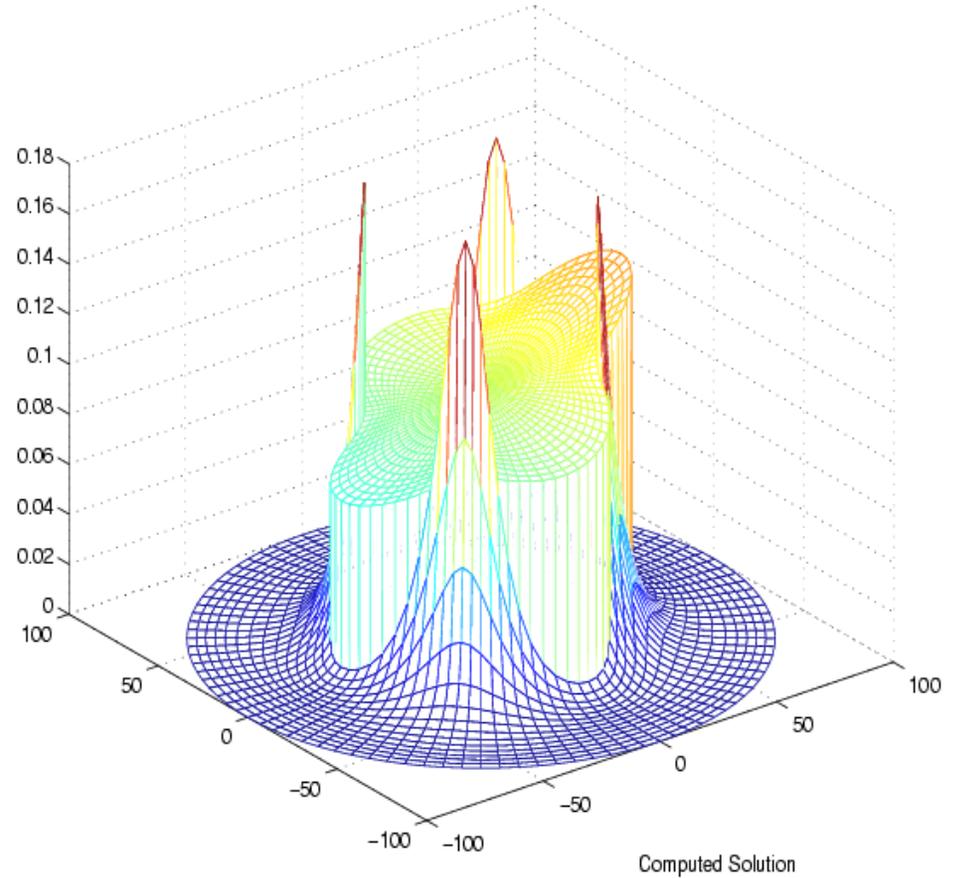


Figure 6: Computed solution for Example 1

Numerical experiments

- **Example 1**

In this test example, Ω_1 is given by a star-shaped region (Fig 2-3) $r < 50 * (1 + 0.2 \cos(4\theta))$ with the exact solution:

$$u(x, y) = \begin{cases} 0.1 \exp(x/2) \cos(y/2) & \text{inside } \Gamma_1 \\ 100 \exp(-\bar{\kappa}/\sqrt{\epsilon_2}r) & \text{outside } \Gamma_1 \end{cases} \quad (15)$$

We take $\bar{\kappa}^2 = 2.0 \text{\AA}^{-2}$, which corresponds to the ionic strength $I_s = 0.2357$. The result is listed in Table 1.

$n_r \times n_\theta$	L^∞ error in ϕ	order	L^∞ error in $\epsilon^- \phi_n^-$	order
14×32	6.066E-03	—	1.214E-01	—
23×64	2.253E-03	1.429	6.372E-02	0.930
41×128	6.315E-04	1.835	1.747E-02	1.867
77×256	1.810E-04	1.803	4.097E-03	2.092
149×512	4.841E-05	1.903	9.481E-04	2.112

Table 1: Error and order of accuracy in ϕ and the flux for Example 1.

Numerical experiments

- **Example 4**

In this example, we perform an actual simulation on the Poisson-Boltzmann equation. The interface is the same as in Example 3. Here we take $\bar{\kappa}^2 = 1.27 \text{ \AA}^{-2}$ and $C = 15,000$. In case (a), we put six charges with alternating sign corresponding to $z_i = \pm 1$. In case (b), z_i 's are randomly chosen between ± 1 and sum to zero. The results are plotted in Fig 9-11.

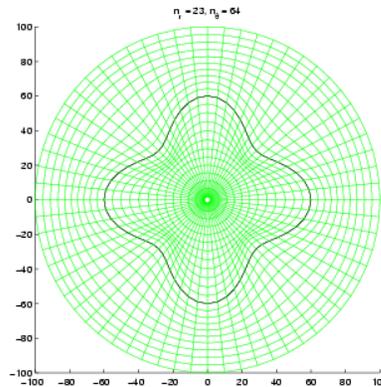


Figure 5: The shape of the macromolecule for Example 1

Parameters: $K^+ = 1.27$, $C = 15000$, $\epsilon^- = 2$, $\epsilon^+ = 80$, $n_r = 32$, $n_\theta = 96$

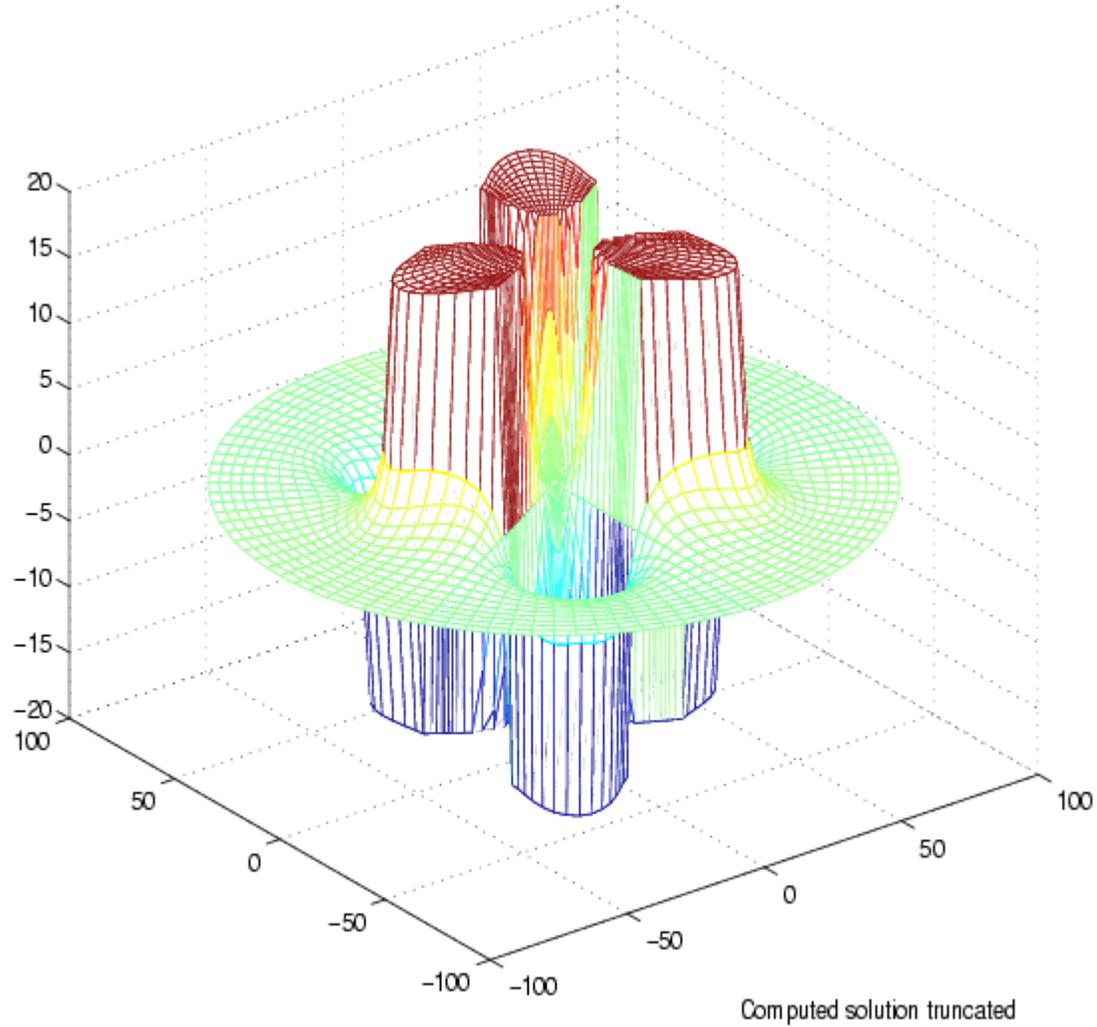


Figure 12: Computed solution for Example 4a

Computed solution outside of (including) the interface

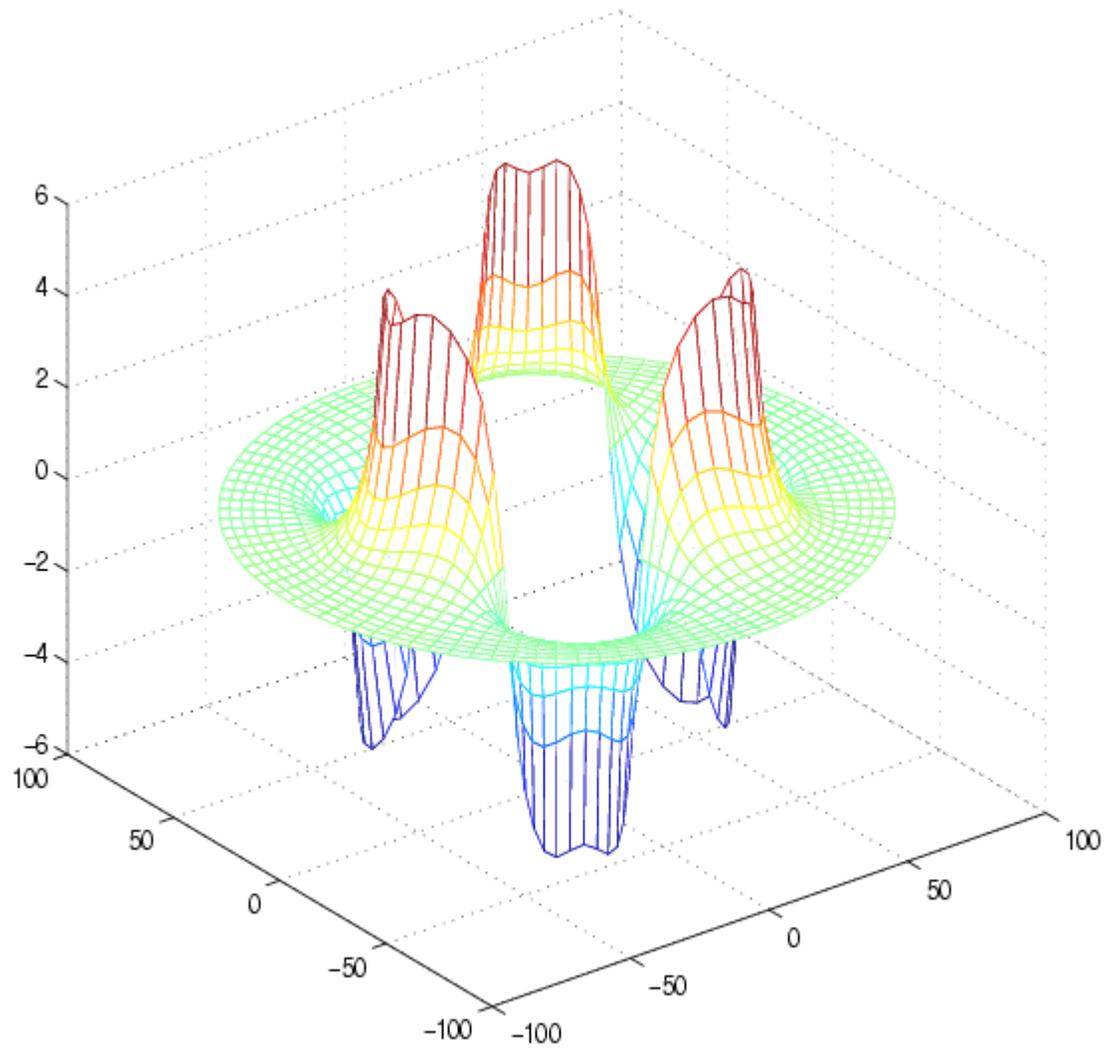


Figure 13: Computed solution outside of Γ_1 for Example 4a

Computed solution outside of (including) the interface

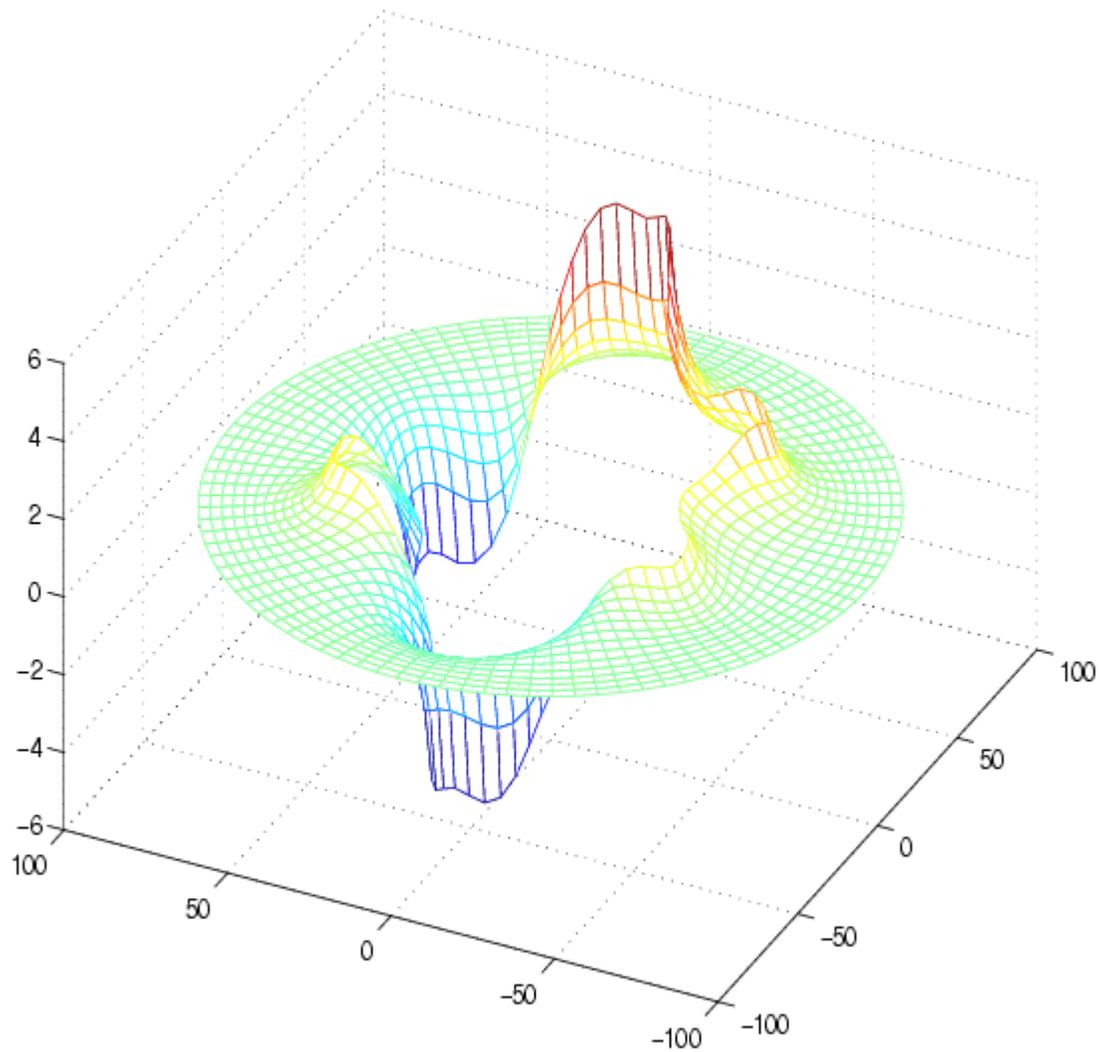


Figure 15: Computed solution outside of Γ_1 for Example 4b

Part 1-Conclusion

1. Point charge singularity is resolved by Green's function and a harmonic function
2. Surface singularity is solved by jump condition capturing method with a skew variable
3. The resulting linear system is symmetric and positive definite, standard fast solver can be adopted
4. Second order accurate for electric field

Part 2: Coupling Interface Method for Elliptic Interface Problems

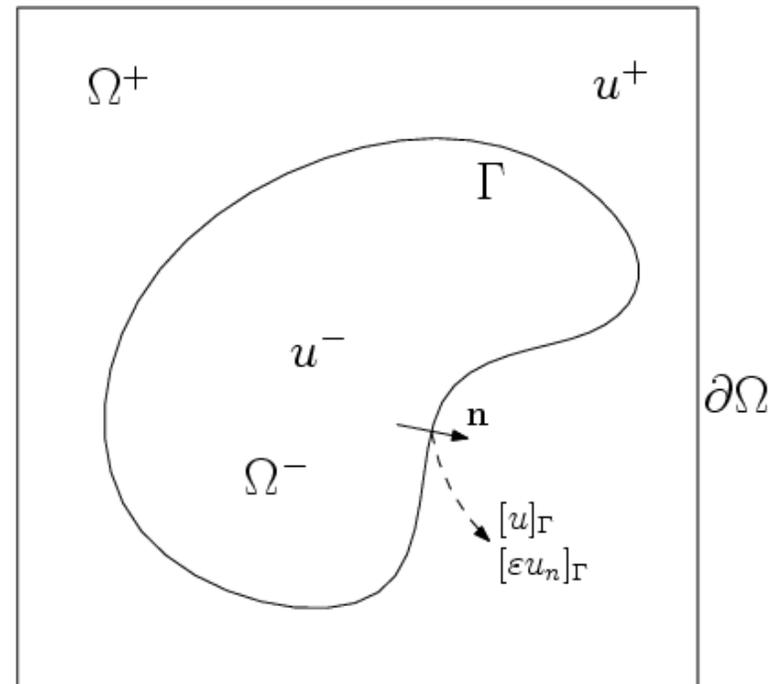
$$-\nabla \cdot (\varepsilon(\mathbf{x})\nabla u(\mathbf{x})) = f(\mathbf{x}), \quad \mathbf{x} \in \Omega \setminus \Gamma,$$

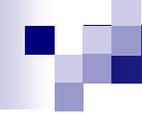
$$[u] = \tau, \quad [\varepsilon u_n] = \sigma \quad \text{on } \Gamma,$$

$$u = g \quad \text{on } \partial\Omega.$$

ε and u are discontinuous,

f is singular across Γ





Three classes of approaches

- Boundary integral approach
- Finite element approach:
- Finite Difference approach:
 - Body-fitting approach
 - Fixed underlying grid: more flexible for moving interface problems

Regular Grid Methods for Solving Elliptic Interface Problems

- **Regularization approach** (Tornberg-Engquist, 2003)
 - Harmonic Averaging (Tikhonov-Samarskii, 1962)
 - Immersed Boundary Method (IB Method) (Peskin, 1974)
 - Phase field method
- **Dimension un-splitting approach**
 - Immersed Interface Method (IIM) (LeVeque-Li, 1994)
 - Maximum Principle Preserving IIM (MIIM) (Li-Ito, 2001)
 - Fast iterative IIM (FIIM) (Li, 1998)
- **Dimension splitting approach**
 - Ghost Fluid Method (Fedkiw et al., 1999, Liu et al. 2000)
 - Decomposed Immersed Interface Method (DIIM) (Berthelsen, 2004)
 - Matched Interface and Boundary Method (MIB) (YC Zhou et al., 2006)
 - **Coupling interface method (CIM) (Chern and Shu 2007)**

Coupling Interface Method (CIM)

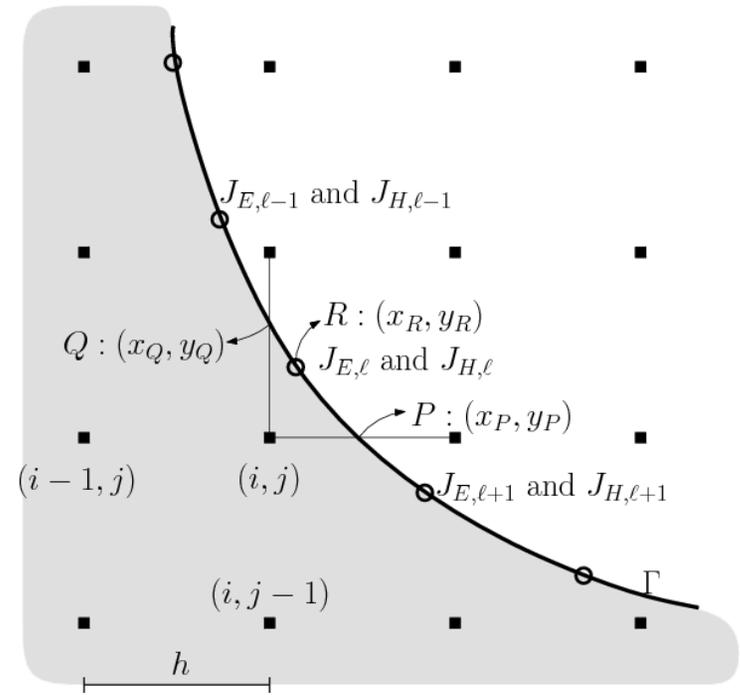
■ CIM

- CIM1 (first order)
- CIM2 (2nd order)
- Hybrid CIM (CIM1 + CIM2)

for complex interface problems

■ Augmented CIM

- Auxiliary variables on interfaces



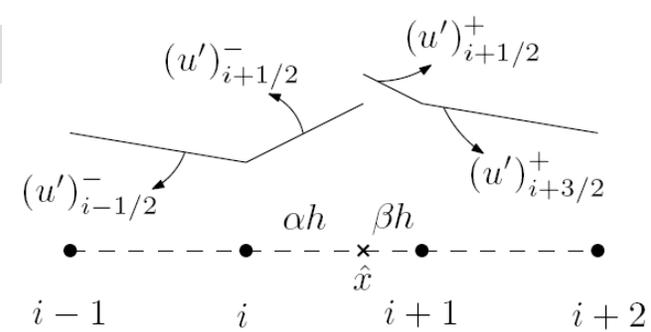
Numerical Issues for dealing with interface problems

- Accuracy: second-order in maximum norm.
- Simplicity: easy to derive and program.
- Stability: nice stencil coefficients for linear solvers.
- Robustness: capable to handle complex interfaces.

CIM outline

- 1d: CIM1, CIM2
- 2d: CIM2
- d dimension
- Hybrid CIM
- Numerical validation
- Application to the Poisson-Boltzmann equation

CIM1: one dimension



$$\begin{cases} u^-(x) := u_i + (u')_{i+1/2}^-(x - x_i) & \text{for } x_i \leq x < \hat{x} \\ u^+(x) := u_{i+1} + (u')_{i+1/2}^+(x - x_{i+1}) & \text{for } \hat{x} < x < x_{i+1}. \end{cases}$$

$$(u_{i+1} - \beta h (u')_{i+1/2}^+) - (u_i + \alpha h (u')_{i+1/2}^-) \approx \tau$$

$$\varepsilon^+ (u')_{i+1/2}^+ - \varepsilon^- (u')_{i+1/2}^- \approx \sigma.$$

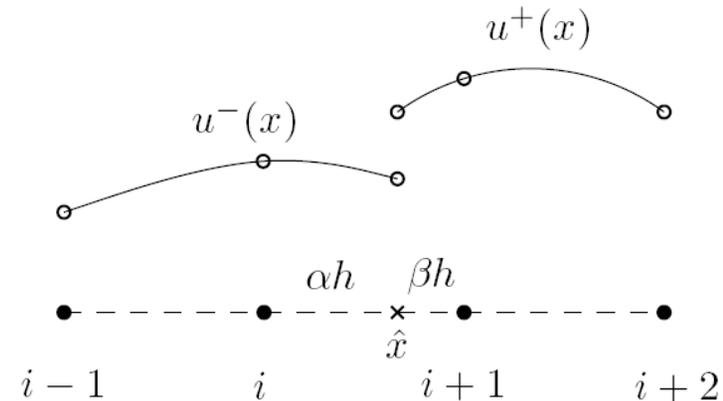
$$(u')_{i+1/2}^- = \frac{1}{h} \left(\bar{\rho}^+ (u_{i+1} - u_i) - \bar{\rho}^+ \tau - \beta h \frac{\sigma}{\bar{\varepsilon}} \right) + O(h)$$

$$(u')_{i+1/2}^+ = \frac{1}{h} \left(\bar{\rho}^- (u_{i+1} - u_i) - \bar{\rho}^- \tau + \alpha h \frac{\sigma}{\bar{\varepsilon}} \right) + O(h)$$

$$\bar{\varepsilon} = \alpha \varepsilon^+ + \beta \varepsilon^-, \quad \bar{\rho}^\pm = \varepsilon^\pm / \bar{\varepsilon}.$$

$$- (\varepsilon u')'(x_i) = - \frac{1}{h} \varepsilon_i \left((u')_{i+1/2}^- - (u')_{i-1/2}^- \right) + O(1).$$

CIM2: One dimension



- Quadratic approximation and match two grid

$$u_\ell(x) = u_i + \left(\frac{u_i - u_{i-1}}{h} + \frac{1}{2} h \underline{u_i''} \right) (x - x_i) + \frac{1}{2} \underline{u_i''} (x - x_i)^2 + O(h^3),$$

$$u_r(x) = u_{i+1} + \left(\frac{u_{i+2} - u_{i+1}}{h} - \frac{1}{2} h \underline{u_{i+1}''} \right) (x - x_{i+1}) + \frac{1}{2} \underline{u_{i+1}''} (x - x_{i+1})^2 + O(h^3)$$

- Match two jump conditions

$$u_r(\hat{x}) - u_\ell(\hat{x}) = \tau, \quad \hat{\varepsilon}^+ u_r'(\hat{x}) - \hat{\varepsilon}^- u_\ell'(\hat{x}) = \sigma,$$

CIM2: One dimension

$$u_i'' = \frac{1}{h^2} \left(L^{(\ell)} u_i + J_i^{(\ell)} \right) + O(h)$$

$$u_{i+1}'' = \frac{1}{h^2} \left(L^{(r)} u_{i+1} + J_{i+1}^{(r)} \right) + O(h),$$

$$L^{(\ell)} u_i := a_{i,-1} u_{i-1} + a_{i,0} u_i + a_{i,1} u_{i+1} + a_{i,2} u_{i+2}$$

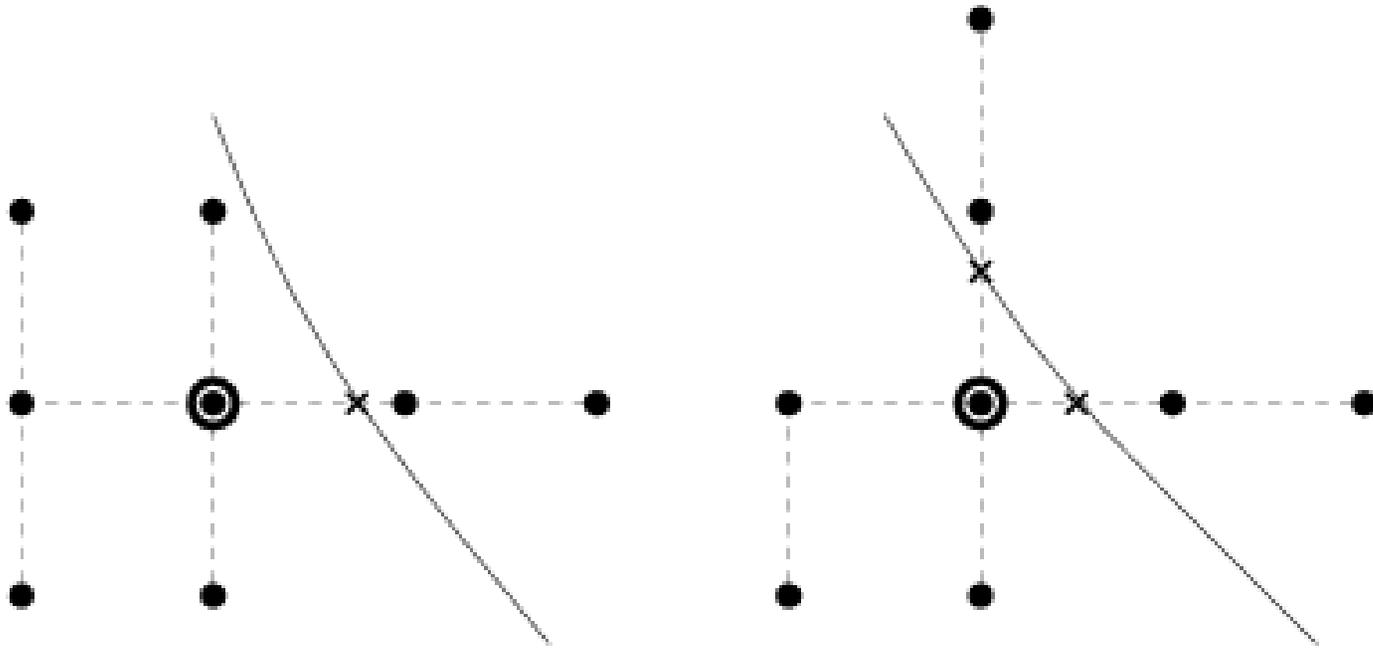
$$L^{(r)} u_{i+1} := a_{i+1,-2} u_{i-1} + a_{i+1,-1} u_i + a_{i+1,0} u_{i+1} + a_{i+1,1} u_{i+2}$$

$$J_i^{(\ell)} := a_{i,\tau} \frac{\tau}{h^2} + a_{i,\sigma} \frac{\sigma}{\hat{\varepsilon} h}$$

$$J_{i+1}^{(r)} := -a_{i+1,\tau} \frac{\tau}{h^2} + a_{i+1,\sigma} \frac{\sigma}{\hat{\varepsilon} h}.$$

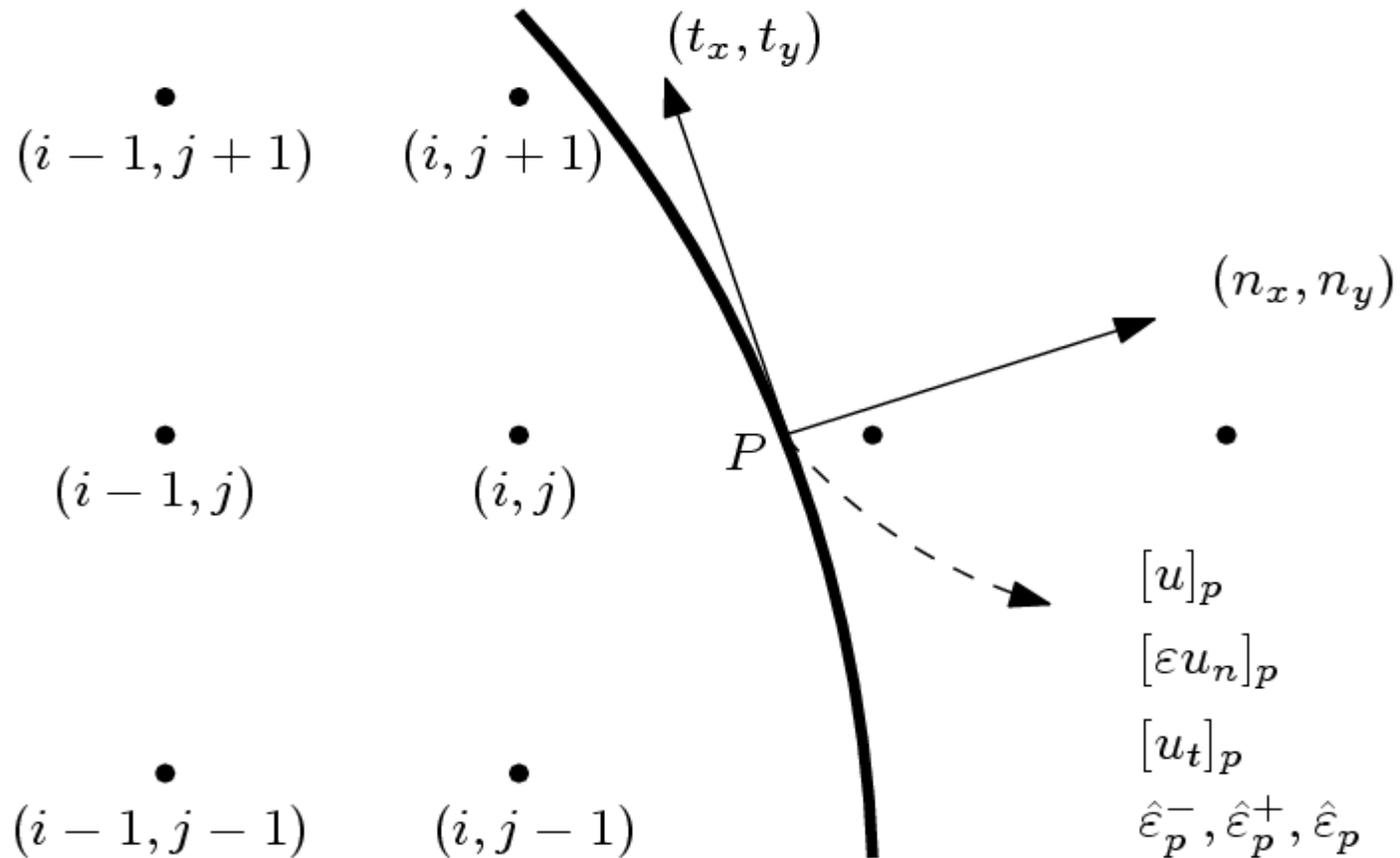
CIM2: 2 dimensions

Stencil at a normal on-front points (bullet) (8 points stencil)

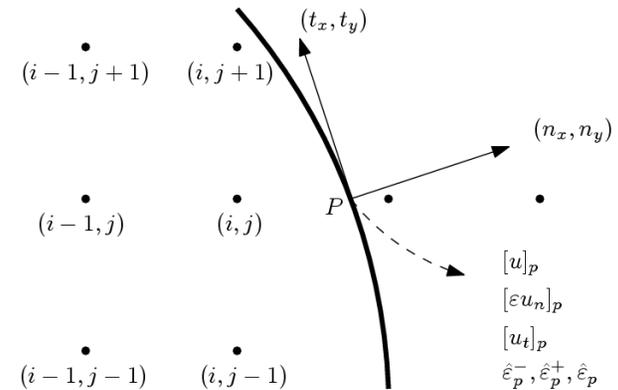


(a) Two dimension: 2 cases

CIM2 Case 1:



CIM2 (Case 1):



- Dimension splitting approach

$$\underline{u_{xx}} = \frac{1}{h^2} \left(Lu + a_\tau [u]_p + a_\sigma h \frac{[\varepsilon u_x]_p}{\hat{\varepsilon}_p} \right) + O(h)$$

- Decomposition of jump condition

$$[\varepsilon u_x]_p = [\varepsilon u_n]_p n_x + (\hat{\varepsilon}_p^+ [u_t]_p + (\hat{\varepsilon}_p^+ - \hat{\varepsilon}_p^-) (u_t^-)_p) t_x$$

- One side interpolation

$$\begin{aligned} (u_t^-)_p &\leftarrow \left(\frac{u_{i,j} - u_{i-1,j}}{h} + \left(\frac{1}{2} + \alpha \right) h \underline{u_{xx}} \right) t_x \\ &+ \left((1 + \alpha) \frac{u_{i,j+1} - u_{i,j-1}}{2h} - \alpha \frac{u_{i-1,j+1} - u_{i-1,j-1}}{2h} \right) t_y + O(h^2) \\ &= \frac{1}{h} T u + h \left(\frac{1}{2} + \alpha \right) t_x u_{xx} + O(h^2) \end{aligned}$$

CIM2 (Case 1):

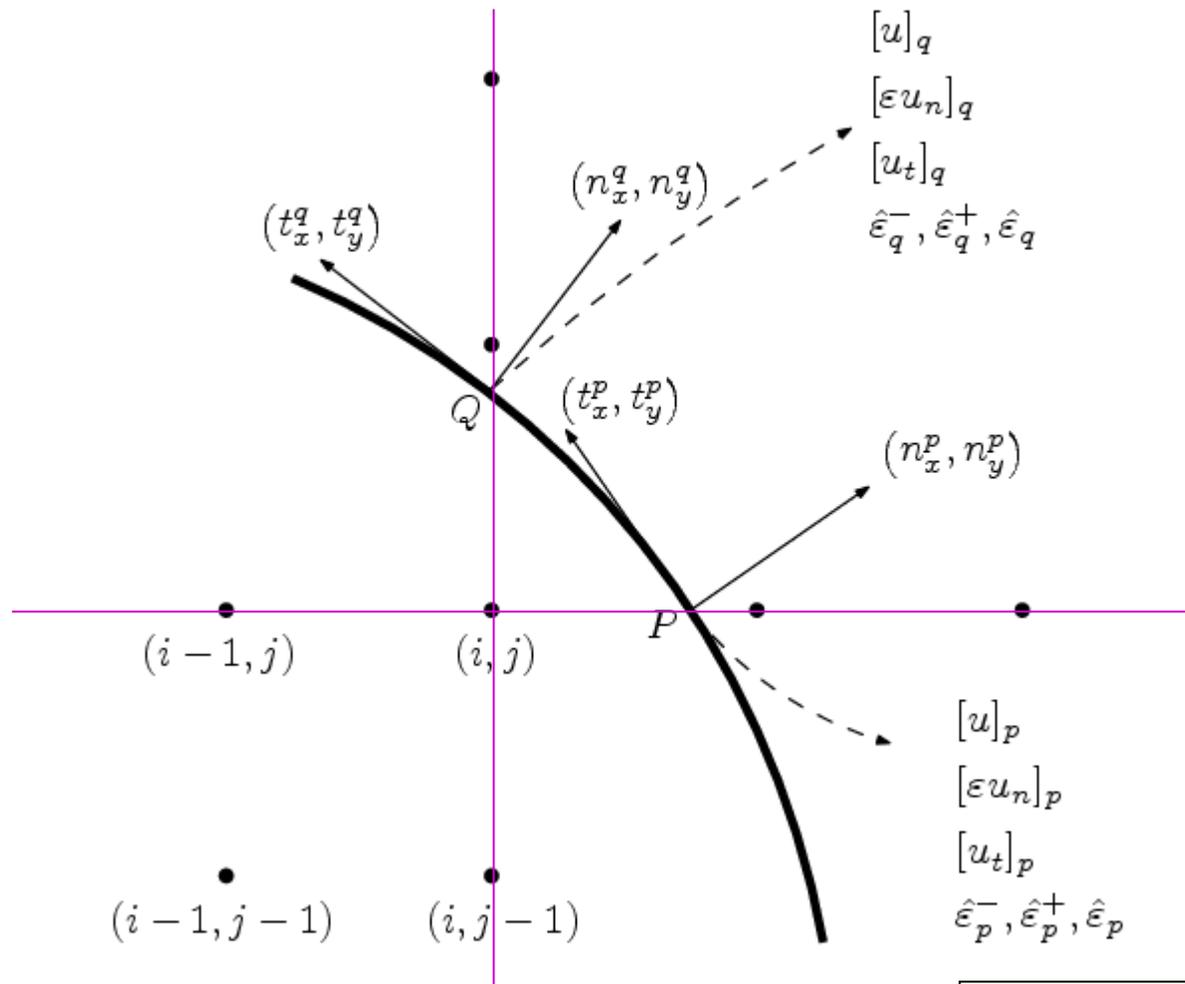
Bounded by 1 and $\varepsilon^+/\varepsilon^-$.

$$\left(1 - \left(\frac{1}{2} + \alpha\right)a_t t_x\right) u_{xx} = \frac{1}{h^2} (Lu + a_t T u + J)$$

$$a_t = a_\sigma (\rho^+ - \rho^-) t_x, \rho^\pm = \hat{\varepsilon}_p^\pm / \hat{\varepsilon}_p,$$

$$J = a_\tau [u]_p + a_\sigma h \left(\frac{[\varepsilon u_n]}{\varepsilon} n_x + \rho^+ [u_t] t_x \right)$$

CIM2 (Case 2):

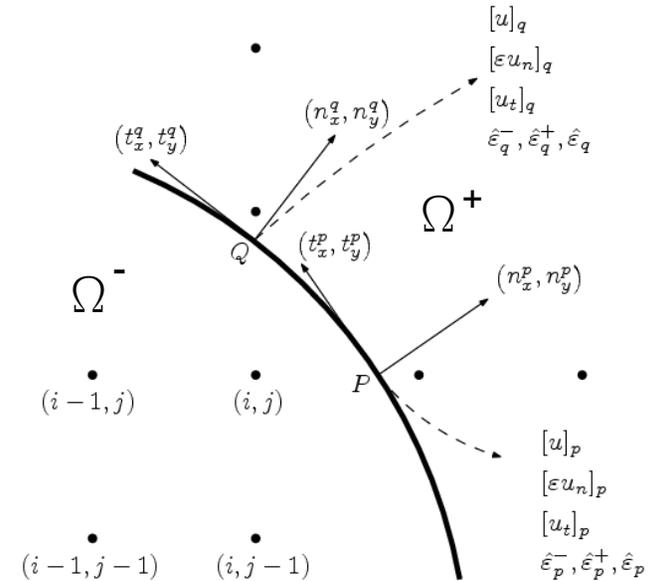


CIM2 (Case 2):

■ Dimension splitting approach

$$u_{xx} = \frac{1}{h^2} \left(L_x u + a_{\tau,p} [u]_p + a_{\sigma,p} h \frac{[\varepsilon u_x]_p}{\hat{\varepsilon}_p} \right) + O(h)$$

$$u_{yy} = \frac{1}{h^2} \left(L_y u + a_{\tau,q} [u]_q + a_{\sigma,q} h \frac{[\varepsilon u_y]_q}{\hat{\varepsilon}_q} \right) + O(h)$$



■ Decomposition of jump conditions

$$[\varepsilon u_x]_p = [\varepsilon u_n]_p n_x^p + (\hat{\varepsilon}_p^+ [u_t]_p + (\hat{\varepsilon}_p^+ - \hat{\varepsilon}_p^-) (u_t^-)_p) (t_x^p)$$

$$[\varepsilon u_y]_q = [\varepsilon u_n]_q n_y^q + (\hat{\varepsilon}_q^+ [u_t]_q + (\hat{\varepsilon}_q^+ - \hat{\varepsilon}_q^-) (u_t^-)_q) (t_y^q)$$

■ One-side interpolation

$$(u_t^-)_p \approx \left(\frac{u_{i,j} - u_{i-1,j}}{h} + \left(\frac{1}{2} + \alpha_p \right) h u_{xx} \right) t_x^p + \left((1 + \alpha_p) \frac{u_{i,j} - u_{i,j-1}}{h} - \alpha_p \frac{u_{i-1,j} - u_{i-1,j-1}}{h} + \frac{1}{2} h u_{yy} \right) t_y^p$$

$$(u_t^-)_q \approx \left(\frac{u_{i,j} - u_{i,j-1}}{h} + \left(\frac{1}{2} + \alpha_q \right) h u_{yy} \right) t_y^q + \left((1 + \alpha_q) \frac{u_{i,j} - u_{i-1,j}}{h} - \alpha_q \frac{u_{i,j-1} - u_{i-1,j-1}}{h} + \frac{1}{2} h u_{xx} \right) t_x^q$$

The second order derivatives are coupled by jump conditions

CIM2 (Case 2): results a coupling matrix

$$\mathbf{M} \begin{bmatrix} u_{xx} \\ u_{yy} \end{bmatrix} = \begin{bmatrix} L_x u + a_{t,p} T_p u + J_p \\ L_y u + a_{t,q} T_q u + J_q \end{bmatrix}$$

$$\mathbf{M} = \begin{bmatrix} 1 - (\frac{1}{2} + \alpha_p) a_{t,p} t_x^p & -\frac{1}{2} a_{t,p} t_y^p \\ -\frac{1}{2} a_{t,q} t_x^q & 1 - (\frac{1}{2} + \alpha_q) t_y^q \end{bmatrix}$$

$$a_{t,p} = a_{\sigma,p} (\rho_p^+ - \rho_q^-) t_x^p$$

$$a_{t,q} = a_{\sigma,q} (\rho_q^+ - \rho_p^-) t_y^q$$

$$\rho_p^\pm = \hat{\varepsilon}_p^\pm / \hat{\varepsilon}_p$$

$$\rho_q^\pm = \hat{\varepsilon}_q^\pm / \hat{\varepsilon}_q$$

$$J_p = a_{\tau,p} [u]_p + a_{\sigma,p} h \left(\frac{[\varepsilon u_n]_p}{\hat{\varepsilon}_p} n_x^p + \rho_p^+ [u_t]_p t_x^p \right)$$

$$J_q = a_{\tau,q} [u]_q + a_{\sigma,q} h \left(\frac{[\varepsilon u_n]_q}{\hat{\varepsilon}_q} n_y^q + \rho_q^+ [u_t]_q t_y^q \right)$$

Theorem: $\det(\mathbf{M})$ is positive when local curvature is zero or h is small

CIM1: d dimensions

■ Dimension splitting approach

$$\frac{\partial}{\partial x_k} u(\mathbf{x} + \frac{1}{2} h \mathbf{e}_k) \approx \frac{1}{h} \left(\bar{\rho}_{k+}^+ (u(\mathbf{x} + h \mathbf{e}_k) - u(\mathbf{x})) - \bar{\rho}_{k+}^+ [u]_{\hat{\mathbf{x}}_{k+}} - \beta_{k+} h \frac{[\varepsilon \nabla u \cdot \mathbf{e}_k]_{\hat{\mathbf{x}}_{k+}}}{\bar{\varepsilon}_{k+}} \right)$$

$$\frac{\partial}{\partial x_k} u(\mathbf{x} - \frac{1}{2} h \mathbf{e}_k) \approx \frac{1}{h} \left(\bar{\rho}_{k-}^+ (u(\mathbf{x}) - u(\mathbf{x} - h \mathbf{e}_k)) + \bar{\rho}_{k-}^+ [u]_{\hat{\mathbf{x}}_{k-}} - \beta_{k-} h \frac{[\varepsilon \nabla u \cdot \mathbf{e}_k]_{\hat{\mathbf{x}}_{k-}}}{\bar{\varepsilon}_{k-}} \right)$$

■ Decomposition of jump conditions

$$[\varepsilon \nabla u \cdot \mathbf{e}_k]_{\hat{\mathbf{x}}_k} = [\varepsilon \nabla u \cdot \mathbf{n}_k]_{\hat{\mathbf{x}}_k} (\mathbf{n}_k \cdot \mathbf{e}_k) + [\varepsilon \nabla u \cdot \mathbf{t}_k]_{\hat{\mathbf{x}}_k} (\mathbf{t}_k \cdot \mathbf{e}_k)$$

$$= \sigma_k (\mathbf{n}_k \cdot \mathbf{e}_k) + (\hat{\varepsilon}_k^+ [\nabla u \cdot \mathbf{t}_k]_{\hat{\mathbf{x}}_k} + (\hat{\varepsilon}_k^+ - \hat{\varepsilon}_k^-) \nabla u^-(\hat{\mathbf{x}}_k) \cdot \mathbf{t}_k) (\mathbf{t}_k \cdot \mathbf{e}_k)$$

■ One-side interpolation

- $j = k$: $\frac{\partial}{\partial x_k} u^-(\hat{\mathbf{x}}_{k\pm}) \approx \frac{\partial}{\partial x_k} u(\mathbf{x} \pm \frac{1}{2} h \mathbf{e}_k)$
- $j \perp k$: $\frac{\partial}{\partial x_j} u^-(\hat{\mathbf{x}}_{k\pm}) = \begin{cases} D_j^{(s_j)} u(\mathbf{x}) & \text{if } \gamma_{j+\frac{1}{2}} + \gamma_{j-\frac{1}{2}} < 2 \\ \frac{\partial}{\partial x_j} u^-(\mathbf{x} \pm \frac{1}{2} h \mathbf{e}_j) & \text{if } \gamma_{j+\frac{1}{2}} + \gamma_{j-\frac{1}{2}} = 2 \end{cases}$

CIM2: d dimensions

- Dimension splitting approach

$$\frac{\partial^2}{\partial x_k^2} u(\mathbf{x}) = \frac{1}{h^2} \left(L_k^{(s_k)} u(\mathbf{x}) + a_{\tau,k} [u]_{\hat{\mathbf{x}}_k} + s_k a_{\sigma,k} h \frac{[\varepsilon \nabla u \cdot \mathbf{e}_k]_{\hat{\mathbf{x}}_k}}{\hat{\varepsilon}_k} \right) + O(h)$$

- Decomposition of jump conditions

$$[\varepsilon \nabla u \cdot \mathbf{e}_k]_{\hat{\mathbf{x}}_k} = [\varepsilon \nabla u \cdot \mathbf{n}_k]_{\hat{\mathbf{x}}_k} (\mathbf{n}_k \cdot \mathbf{e}_k) + [\varepsilon \nabla u \cdot \mathbf{t}_k]_{\hat{\mathbf{x}}_k} (\mathbf{t}_k \cdot \mathbf{e}_k)$$

$$= \sigma_k (\mathbf{n}_k \cdot \mathbf{e}_k) + (\hat{\varepsilon}_k^+ [\nabla u \cdot \mathbf{t}_k]_{\hat{\mathbf{x}}_k} + (\hat{\varepsilon}_k^+ - \hat{\varepsilon}_k^-) \nabla u^-(\hat{\mathbf{x}}_k) \cdot \mathbf{t}_k) (\mathbf{t}_k \cdot \mathbf{e}_k)$$

- One-side interpolation

$$\begin{aligned} & \nabla u^-(\hat{\mathbf{x}}_k) \cdot \mathbf{t}_k \\ &= \frac{1}{h} T_k u(\mathbf{x}) + h \left(s_k \left(\frac{1}{2} + \alpha_k \right) (\mathbf{t}_k \cdot \mathbf{e}_k) \frac{\partial^2}{\partial x_k^2} u(\mathbf{x}) + \frac{1}{2} \sum_{j=1, j \neq k}^d s_j (\mathbf{t}_k \cdot \mathbf{e}_j) \frac{\partial^2}{\partial x_j^2} u(\mathbf{x}) \right) \end{aligned}$$

CIM2: d dimensions, coupling matrix

$$\mathbf{M} \left(\frac{\partial^2}{\partial x_k^2} u(\mathbf{x}) \right)_{k=1}^d = \frac{1}{h^2} (L u(\mathbf{x}) + T u(\mathbf{x}) + J),$$

$$m_{k,j} = \begin{cases} 1 - |s_k| \left(\frac{1}{2} + \alpha_k \right) a_{t,k} (\mathbf{t}_k \cdot \mathbf{e}_k) & j = k \\ -\frac{1}{2} s_j s_k a_{t,k} (\mathbf{t}_k \cdot \mathbf{e}_j) & j \neq k, \end{cases}$$

$$L = (L_1, \dots, L_d)^T,$$

$$T = (s_1 a_{t,1} T_1, \dots, s_d a_{t,d} T_d)^T$$

$$J = (J_1, \dots, J_d)^T.$$

Classification of grids for complex interface (number of grids)

- Interior grids: $O(h^{-d})$
- Normal on-fronts (CIM2): $O(h^{1-d})$
- Exceptional (CIM1): $O(1)$

- The resulting scheme is still **2nd order**

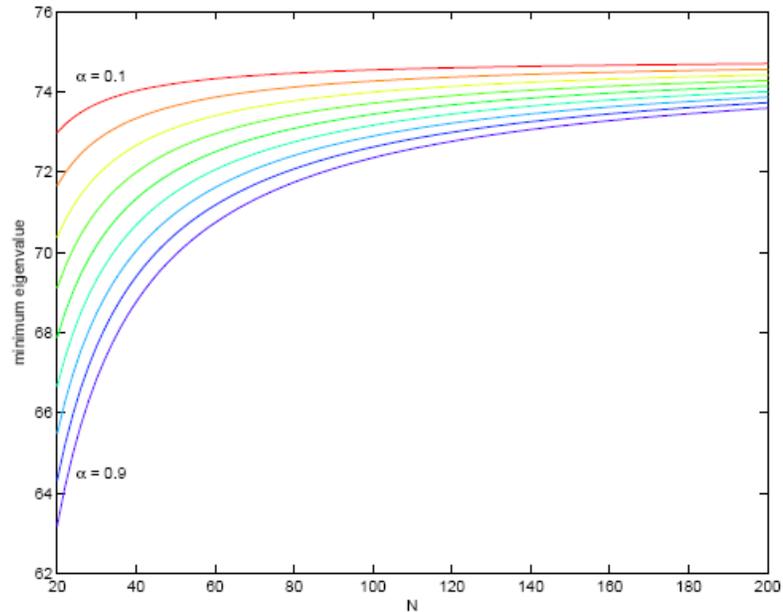


Numerical Validation

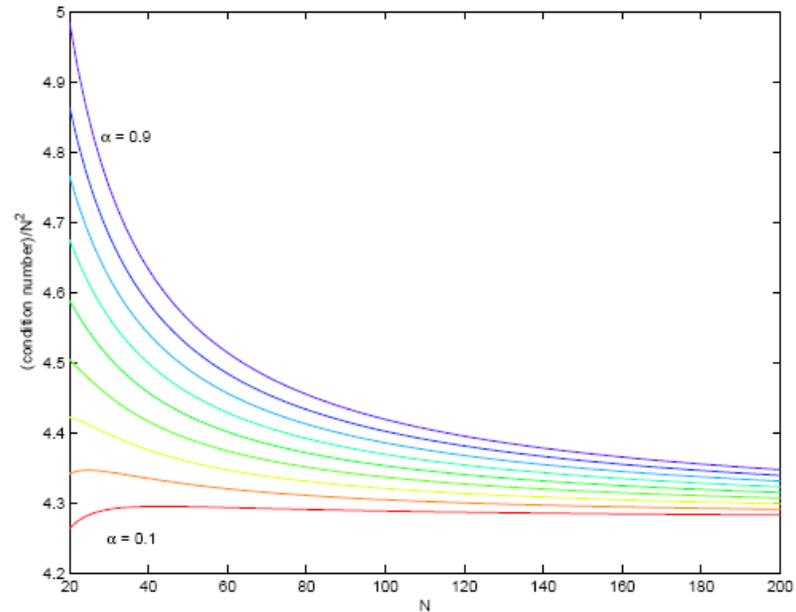
- Stability of CIM2 in 1d
- Orientation error of CIM2 in 2d
- Convergence tests of CIM1
- Comparison results (CIM2)
- Complex interfaces results (Hybrid CIM)

Stability Issue of CIM2 in 1-d

Let $A(\alpha, N)$ be the resulting matrix.



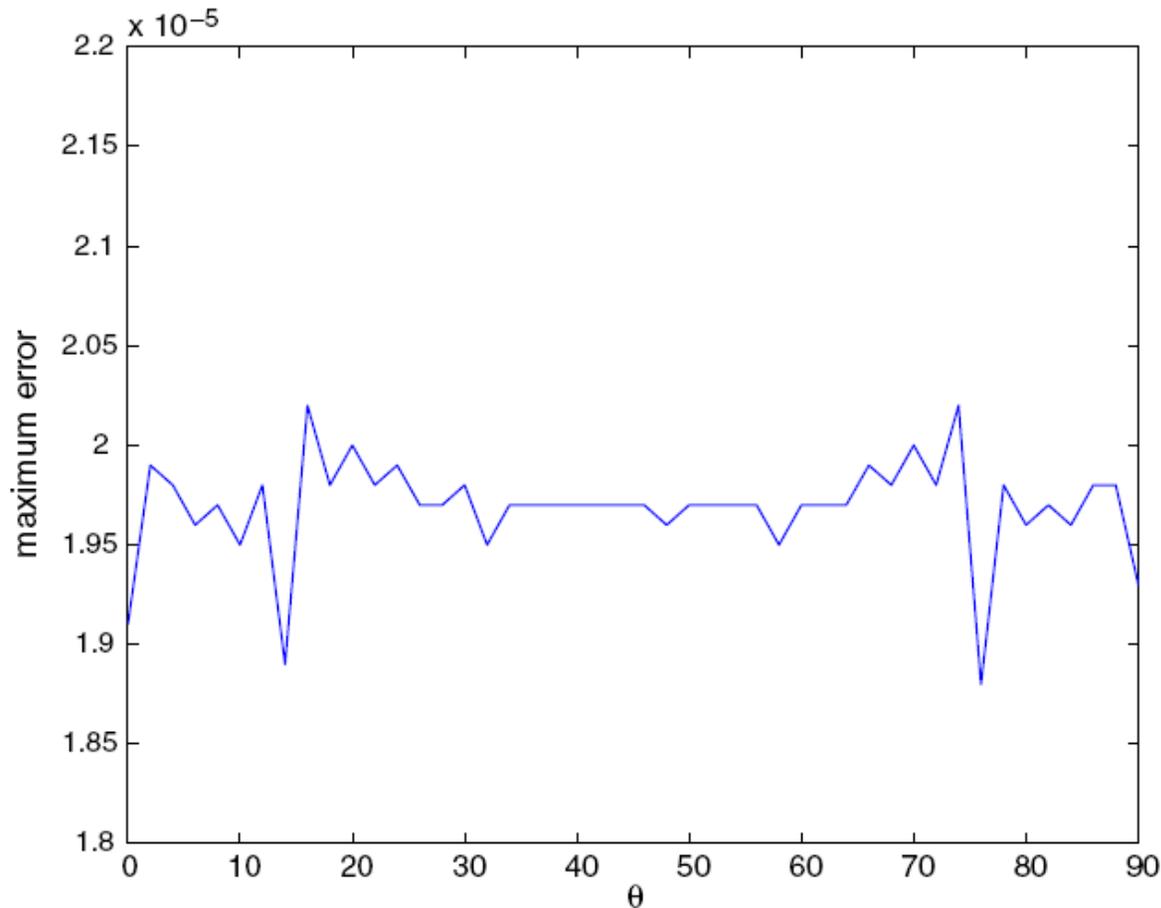
(a) Minimal eigenvalue $\lambda_{min}(\alpha, N)$



(b) Scaled condition number $condA(\alpha, N)/N^2$

Insensitive to the location of the interface in a cell.

Orientation error from CIM2 is small



Insensitive to the orientation of the interface.

Comparison Table (for CIM2)

Method	EJIIM [49]	MIIM [30, 8]	DIIM [5]	JCCS [48]	MIB [53]
Year	2000	2001,2003	2004	2004	2006
2D Example 3	✓	✓	✓		
2D Example 4				✓	✓
3D Example 5		✓			

Example 3 (for CIM2)

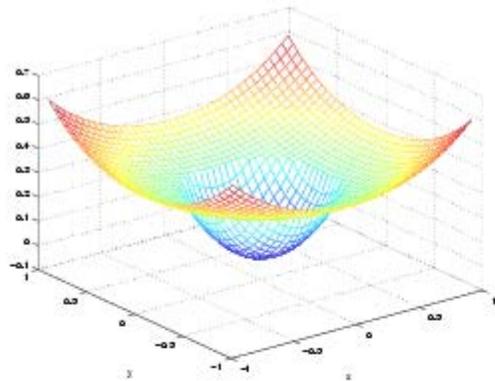
$$\phi(x, y) = r - 0.5, \quad \Omega^- = \{(x, y) | \phi(x, y) < 0\}, \quad \Omega^+ = \{(x, y) | \phi(x, y) > 0\},$$

$$\varepsilon(x, y) = \begin{cases} 1 + r^2 & (x, y) \in \Omega^- \\ b & (x, y) \in \Omega^+ \end{cases}$$

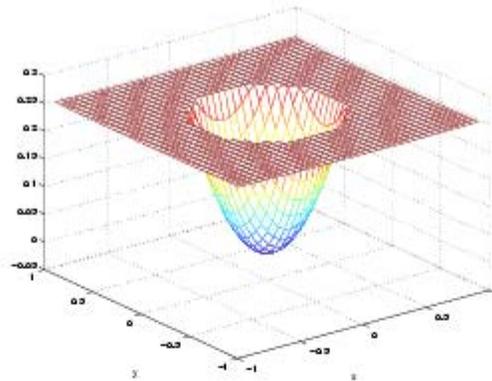
$$u_e(x, y) = \begin{cases} r^2 & (x, y) \in \Omega^- \\ (r^4/2 + r^2 + 0.1 \log(2r))/b - (0.5^4/2 + 0.5^2)/b + 0.5^2 & (x, y) \in \Omega^+ \end{cases}$$

$$f(x, y) = -8r^2 - 4,$$

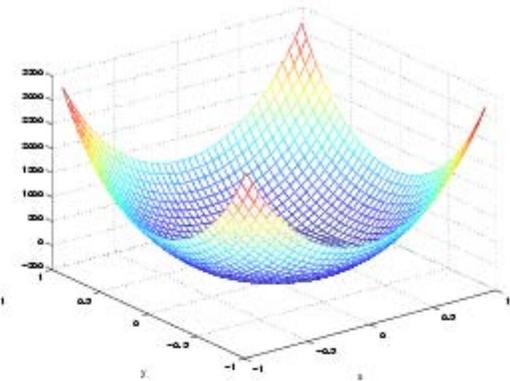
Example 3, figures



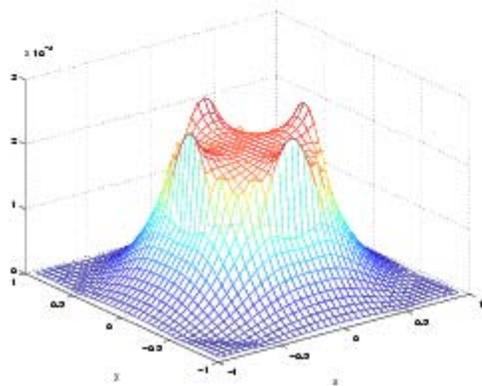
(a) Exact Solution: $b = 10$



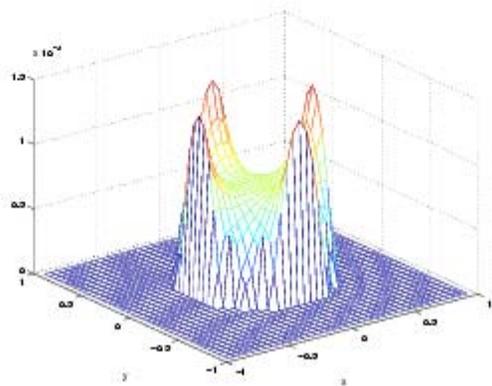
(b) Exact Solution: $b = 1000$



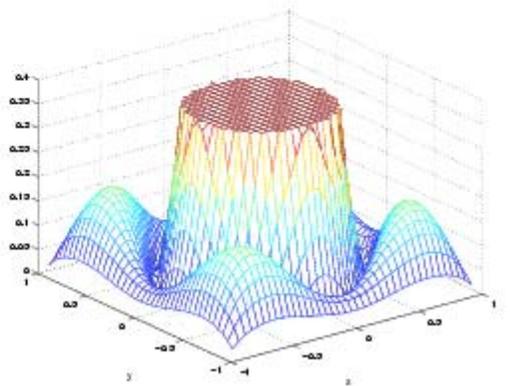
(c) Exact Solution: $b = 0.001$



(d) $|u - u_e|$: $b = 10$



(e) $|u - u_e|$: $b = 1000$



(f) $|u - u_e|$: $b = 0.001$

Example 3, Comparison table 1

n	CIM			DIIM	EJIIM	MIIM
	CPU	$\ \nabla u - \nabla u_e\ _{\infty, \Gamma}$	$\ u - u_e\ _{\infty}$			
20	0.00	1.557×10^{-2}	1.259×10^{-3}	5.378×10^{-4}	7.6×10^{-4}	–
40	0.02	4.714×10^{-3}	2.565×10^{-4}	1.378×10^{-4}	2.4×10^{-4}	4.864×10^{-4}
80	0.17	1.305×10^{-3}	5.215×10^{-5}	3.470×10^{-5}	7.9×10^{-5}	1.448×10^{-4}
160	0.74	3.462×10^{-4}	1.142×10^{-5}	8.704×10^{-6}	2.2×10^{-5}	3.012×10^{-5}
320	3.65	8.948×10^{-5}	2.725×10^{-6}	2.177×10^{-6}	5.3×10^{-6}	8.226×10^{-6}
640	15.86	2.276×10^{-5}	6.740×10^{-7}	–	–	2.060×10^{-6}

Table 2: Example 1: $b = 10$

Example 3, Comparison table 2

n	CPU	CIM		DIIM	MIIM
		$\ \nabla u - \nabla u_e\ _{\infty, \Gamma}$	$\ u - u_e\ _{\infty}$	$\ u - u_e\ _{\infty}$	$\ u - u_e\ _{\infty}$
32	0.04	6.841×10^{-3}	2.732×10^{-4}	2.083×10^{-4}	5.136×10^{-4}
64	0.19	1.920×10^{-3}	3.875×10^{-5}	5.296×10^{-5}	8.235×10^{-5}
128	1.03	5.156×10^{-4}	5.337×10^{-6}	1.330×10^{-5}	1.869×10^{-5}
256	4.84	1.345×10^{-4}	7.241×10^{-7}	3.330×10^{-6}	4.026×10^{-6}
512	22.52	3.463×10^{-5}	9.891×10^{-8}	—	9.430×10^{-7}

Table 3: Example 1: $b = 1000$

Example 3 for CIM2

n	CIM			DIIM	MIIM
	CPU	$\ \nabla u - \nabla u_e\ _{\infty, \Gamma}$	$\ u - u_e\ _{\infty}$	$\ u - u_e\ _{\infty}$	
32	0.03	8.030×10^0	4.278×10^{-1}	4.971×10^0	9.346×10^0
64	0.18	1.829×10^0	1.260×10^{-1}	1.176×10^0	2.006×10^0
128	1.03	4.658×10^{-1}	3.773×10^{-2}	2.900×10^{-1}	5.808×10^{-1}
256	5.3	1.254×10^{-1}	1.365×10^{-2}	7.086×10^{-2}	1.374×10^{-1}
512	23.48	4.141×10^{-2}	2.446×10^{-3}	—	3.580×10^{-2}

Table 4: Example 1: $b = 0.001$

Example 4 (for CIM2)

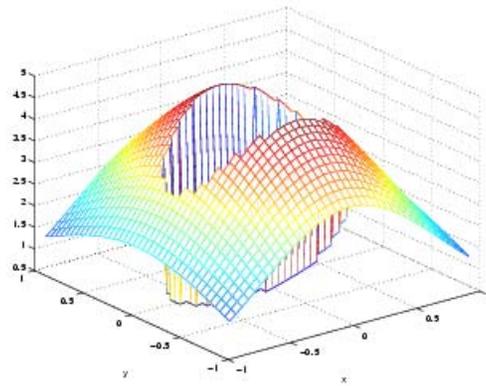
$$\phi(x, y) = \left(\frac{x^2}{18/27}\right)^2 + \left(\frac{y}{10/27}\right)^2 - 1, \quad \Omega^- = \{(x, y) | \phi(x, y) < 0\}, \quad \Omega^+ = \{(x, y) | \phi(x, y) > 0\}$$

$$\varepsilon(x, y) = \begin{cases} \varepsilon^- & (x, y) \in \Omega^- \\ \varepsilon^+ & (x, y) \in \Omega^+ \end{cases}$$

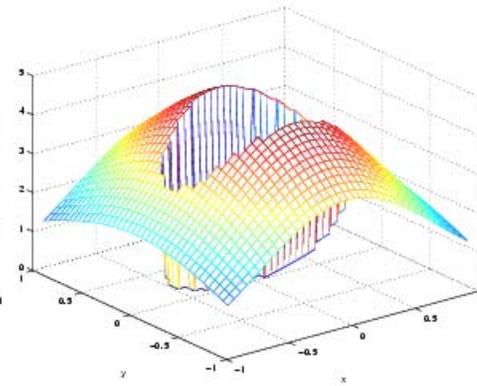
$$u_e(x, y) = \begin{cases} e^x \cos y & (x, y) \in \Omega^- \\ 5e^{-x^2-y^2/2} & (x, y) \in \Omega^+ \end{cases}$$

$$f(\mathbf{x}) = \begin{cases} 0 & (x, y) \in \Omega^- \\ -5e^{-x^2-y^2/2}(-3 + 4x^2 + y^2); & (x, y) \in \Omega^+ \end{cases}$$

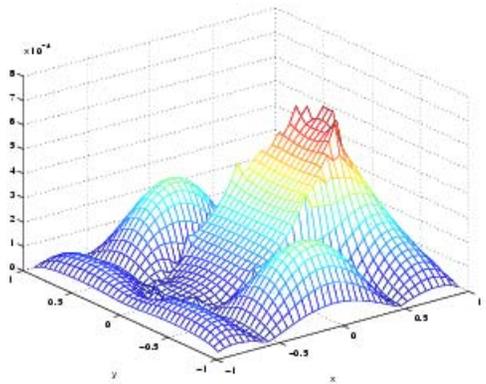
Example 4, figures



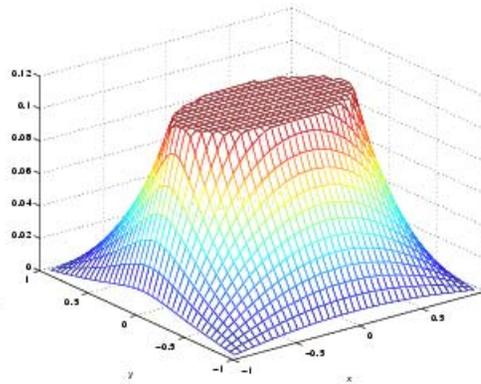
(a) Exact Solution: $\varepsilon^- = 10, \varepsilon^+ = 1$



(b) Exact Solution: $\varepsilon^- = 1000, \varepsilon^+ = 1$



(c) $|u - u_\varepsilon|$: $\varepsilon^- = 10, \varepsilon^+ = 1$



(d) $|u - u_\varepsilon|$: $\varepsilon^- = 1000, \varepsilon^+ = 1$

Example 4 (for CIM2)

n	Total time	CIM		MIB	JCCS
		$\ \nabla u - \nabla u_e\ _{\infty, \Gamma}$	$\ u - u_e\ _{\infty}$	$\ u - u_e\ _{\infty}$	$\ u - u_e\ _{\infty}$
20	0.01	2.289×10^{-2}	4.067×10^{-3}	2.659×10^{-2}	1.755×10^{-2}
40	0.03	8.068×10^{-3}	6.171×10^{-4}	5.206×10^{-3}	4.961×10^{-3}
80	0.19	3.164×10^{-3}	1.682×10^{-4}	1.487×10^{-3}	1.352×10^{-3}
160	1.13	9.935×10^{-4}	3.975×10^{-5}	3.746×10^{-4}	3.548×10^{-4}
320	6.20	2.293×10^{-4}	7.390×10^{-6}	7.803×10^{-5}	9.096×10^{-5}

Table 6: Example 2: $\varepsilon^- = 10$, $\varepsilon^+ = 1$

Example 4, Comparison table 2

n	Total time	CIM		MIB	JCCS
		$\ \nabla u - \nabla u_e\ _{\infty, \Gamma}$	$\ u - u_e\ _{\infty}$	$\ u - u_e\ _{\infty}$	$\ u - u_e\ _{\infty}$
20	0.01	1.551×10^0	3.539×10^{-1}	9.130×10^{-2}	2.803×10^0
40	0.08	4.682×10^{-1}	1.100×10^{-1}	2.764×10^{-2}	7.543×10^{-1}
80	0.41	8.966×10^{-2}	2.028×10^{-2}	7.524×10^{-3}	1.940×10^{-1}
160	2.26	2.799×10^{-2}	6.462×10^{-3}	2.169×10^{-3}	4.906×10^{-2}
320	7.29	6.343×10^{-3}	1.437×10^{-3}	4.841×10^{-4}	1.232×10^{-2}

Table 7: Example 2: $\varepsilon^- = 1000$, $\varepsilon^+ = 1$.

Example 5 (for CIM2)

■ 3 dimensions

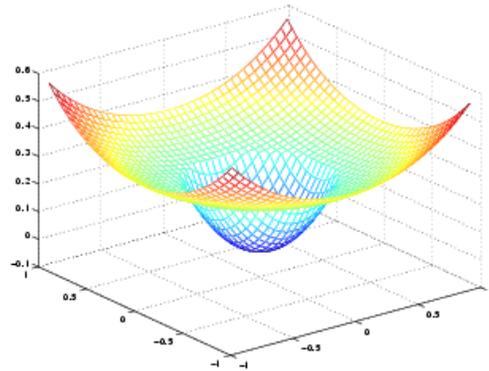
$$\phi(x, y, z) = r - 0.5, \quad \Omega^- = \{(x, y, z) | \phi(x, y, z) < 0\}, \quad \Omega^+ = \{(x, y, z) | \phi(x, y, z) > 0\}$$

$$\varepsilon(x, y, z) = \begin{cases} 1 + r^2 & (x, y, z) \in \Omega^- \\ b & (x, y, z) \in \Omega^+ \end{cases}$$

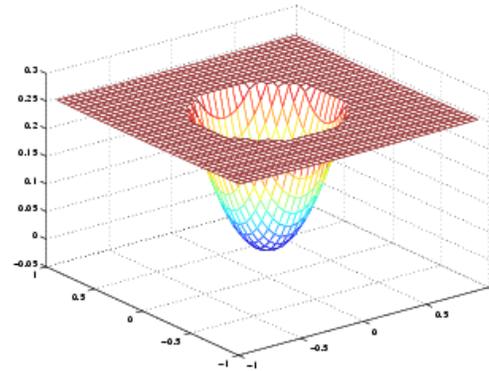
$$u_e(x, y, z) = \begin{cases} r^2 & (x, y, z) \in \Omega^- \\ (r^4/2 + r^2)/b - (0.5^4/2 + 0.5^2)/b + 0.5^2 & (x, y, z) \in \Omega^+ \end{cases}$$

$$f(x, y, z) = -(10r^2 + 6),$$

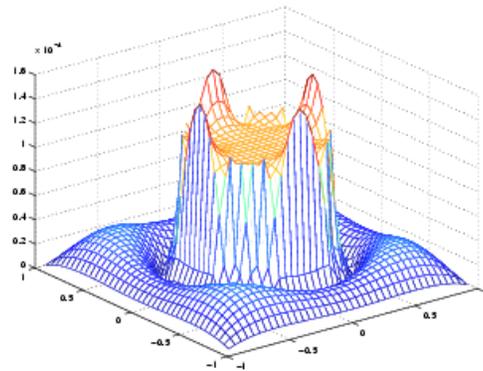
Example 5, figures



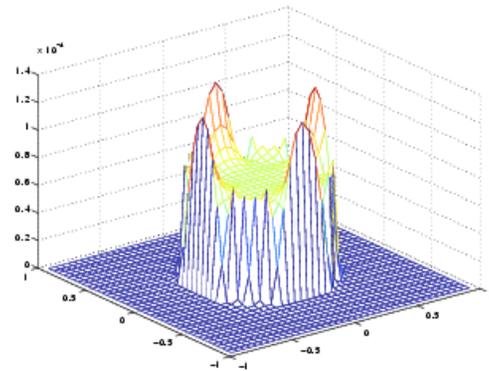
(a) Exact Solution: $b = 10$



(b) Exact Solution: $b = 1000$



(c) $|u - u_e|$: $b = 10$



(d) $|u - u_e|$: $b = 1000$

Example 5 (for CIM2)

n	CPU	CIM			MIIM, 27 points	
		$\ \nabla u - \nabla u_e\ _{\infty, \Gamma}$	$\ u_a - u_e\ _{\infty} / \ u_e\ _{\infty}$	Order	$\ u_a - u_e\ _{\infty} / \ u_e\ _{\infty}$	Order
26	1.52	1.005×10^{-2}	1.822×10^{-4}		1.247×10^{-3}	
52	20.5	3.685×10^{-3}	4.153×10^{-5}	2.133	3.979×10^{-3}	1.648
104	212	9.729×10^{-4}	9.529×10^{-6}	2.124	9.592×10^{-4}	2.052
208	2355	2.540×10^{-4}	2.230×10^{-6}	2.095	–	–

Table 10: Example 4: $b = 1$

Example 5 (CIM2)

n	CIM				MIIM, 27 points	
	CPU	$\ \nabla u - \nabla u_e\ _{\infty, \Gamma}$	$\ u_a - u_e\ _{\infty} / \ u_e\ _{\infty}$	Order	$\ u_a - u_e\ _{\infty} / \ u_e\ _{\infty}$	Order
26	1.45	7.174×10^{-3}	4.332×10^{-4}		1.525×10^{-3}	
52	19.14	2.693×10^{-3}	9.240×10^{-5}	2.229	5.240×10^{-4}	1.541
104	161	7.401×10^{-4}	1.636×10^{-5}	2.498	1.010×10^{-4}	2.375
208	1867	1.979×10^{-4}	3.330×10^{-6}	2.297	–	–

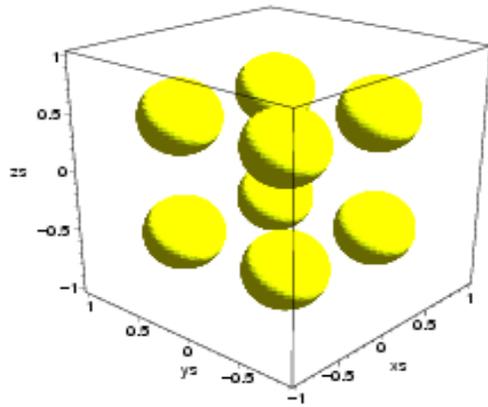
Table 11: Example 4: $b = 10$

Example 5 (CIM2)

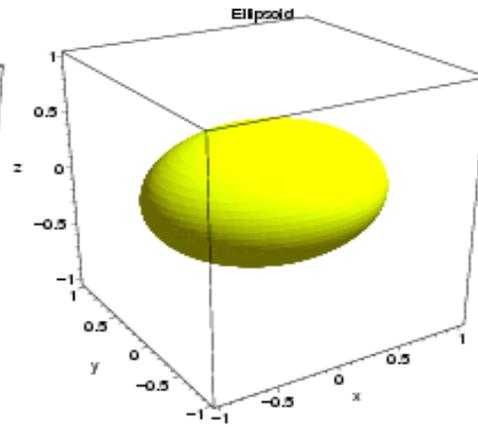
n	CPU	CIM			MIIM, 27 points	
		$\ \nabla u - \nabla u_e\ _{\infty, \Gamma}$	$\ u_a - u_e\ _{\infty} / \ u_e\ _{\infty}$	Order	$\ u_a - u_e\ _{\infty} / \ u_e\ _{\infty}$	Order
26	1.48	6.825×10^{-3}	9.133×10^{-4}		3.845×10^{-3}	
52	24.54	2.594×10^{-3}	2.466×10^{-4}	1.889	1.111×10^{-3}	1.649
104	209	7.183×10^{-4}	3.447×10^{-5}	2.839	1.605×10^{-4}	2.791
208	3299	1.925×10^{-4}	4.727×10^{-6}	2.866	—	—

Table 12: Example 4: $b = 1000$

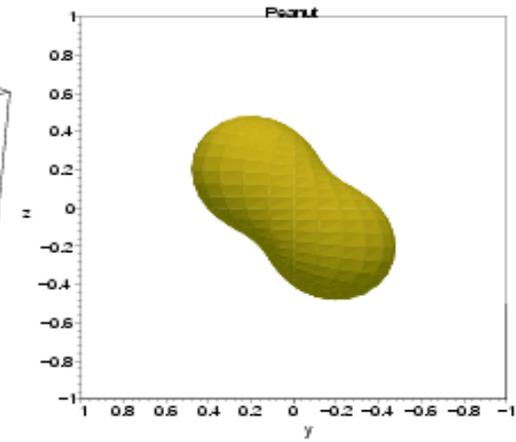
Convergence tests for CIM1: interfaces



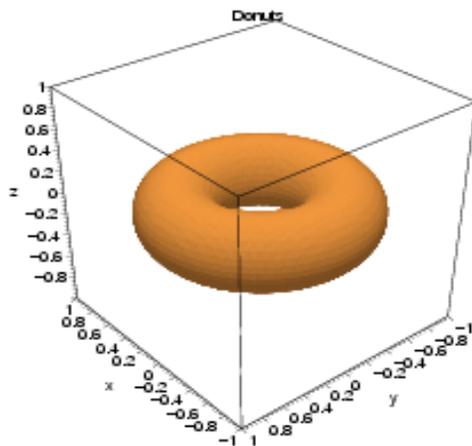
(a) 8 balls



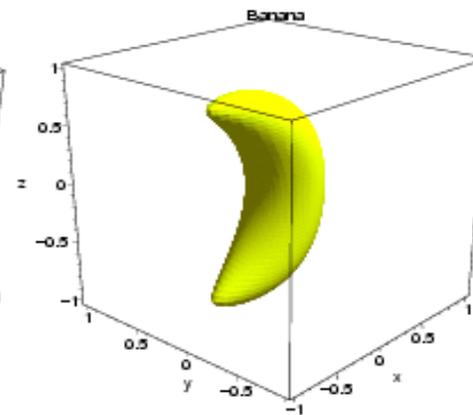
(b) Ellipsoid



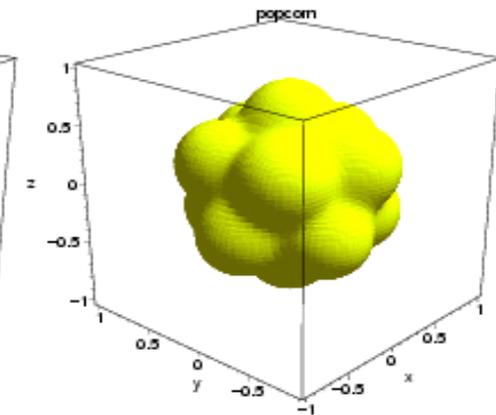
(c) Peanut



(d) Donut

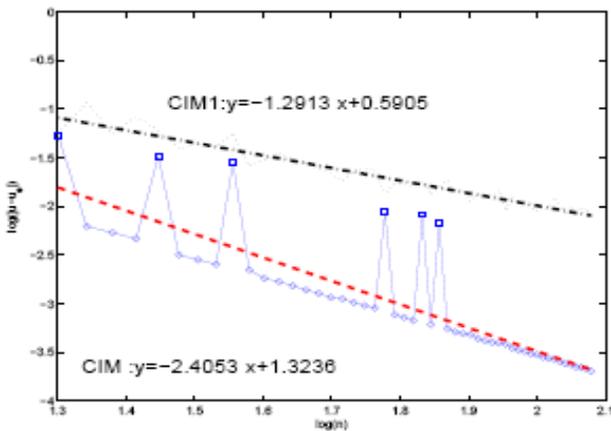


(e) Banana

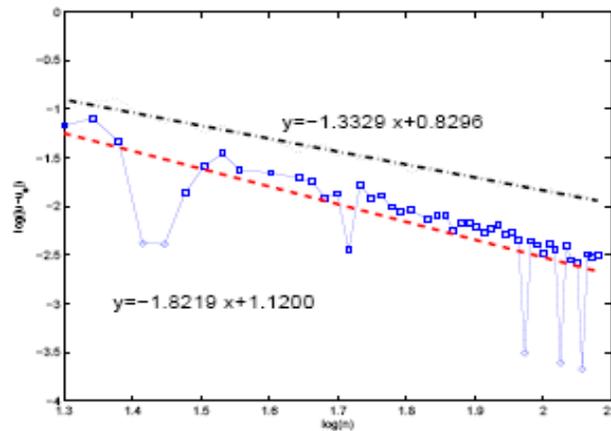


(f) Popcorn

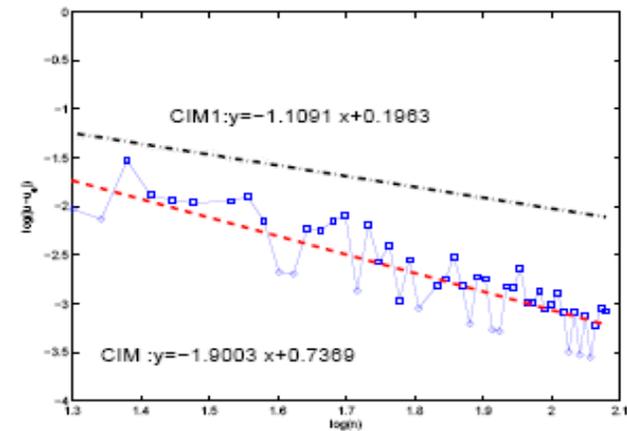
Convergence of hybrid CIM (order 1.8)



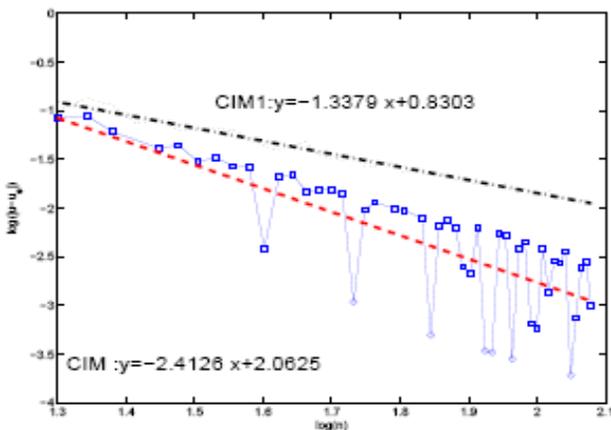
(a) 8 balls



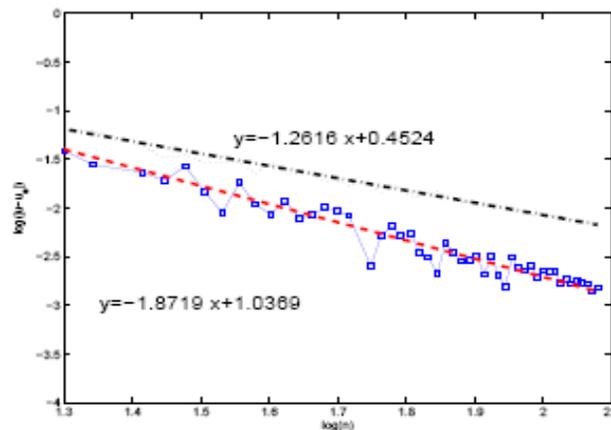
(b) Ellipsoid



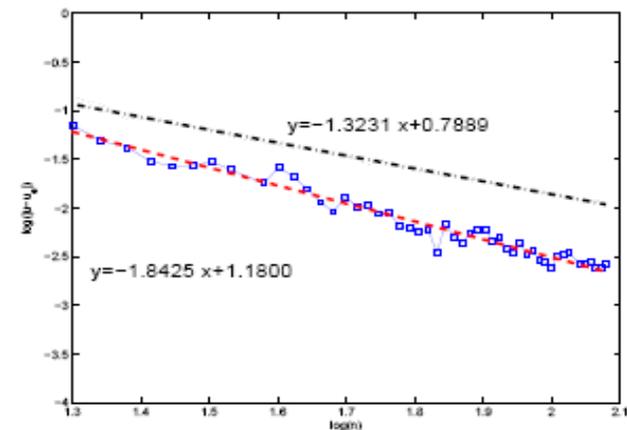
(c) Peanut



(d) Donut



(e) Banana



(f) Popcorn

Comparison results

- Second order for u and its gradients in maximum norm for CIM2
- Insensitive to the contrast of epsilon
- Less absolute error despite of using smaller size of stencil
- Linear computational complexity

CIM: Ingredients

- Dimension splitting approach.
- Decomposition of jump conditions
- **Coupling**: express tangential derivative in terms of principal second derivative to **reduce number of interpolation points**.
- Solvability of stencil coefficients

CIM:merits

- Accuracy: **second-order** for u and its gradients in maximum norm with smaller absolute errors than other existing methods.
- Simplicity: **smaller size stencil**, easy to program.
- Stability: nice stencil coefficients for linear solvers.
- Robustness: **capable to handle complex interfaces**.
- Speed: linear computational complexity

Poisson-Boltzmann equation

$$-\nabla[\epsilon(r)\nabla\phi(r)] + K(r) \sinh(\phi(r)) = Q(r)$$

$$Q(r) = C \sum_{i=1}^{N_m} q_i \delta(r - r_i), C = \frac{4\pi e_c^2}{k_B T}$$

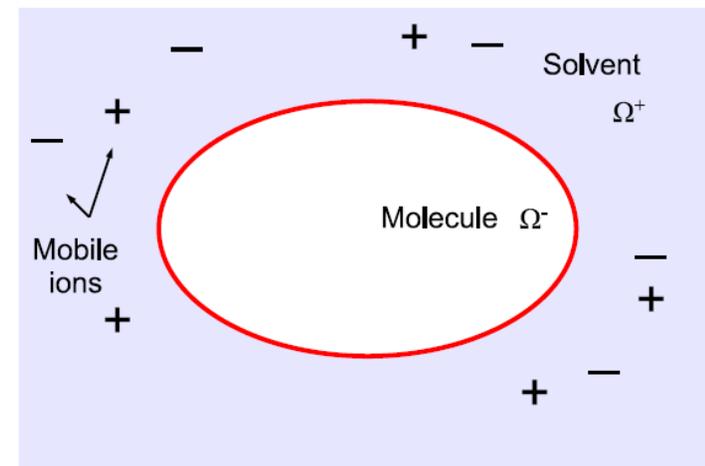
$$\epsilon_1 \approx 1 \sim 2, \epsilon_2 \approx 80,$$

$$5249.0 \leq C \leq 10500.0,$$

$$-1 \leq q_i \leq 1,$$

$K =$

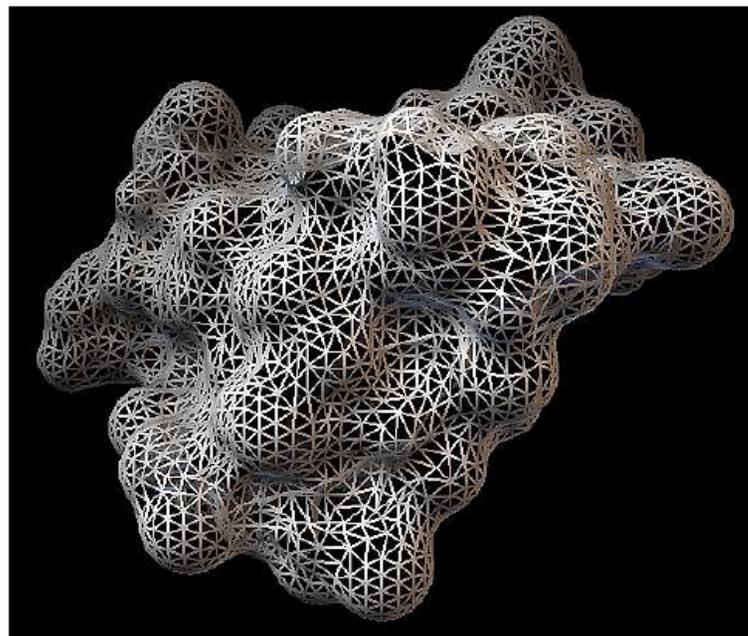
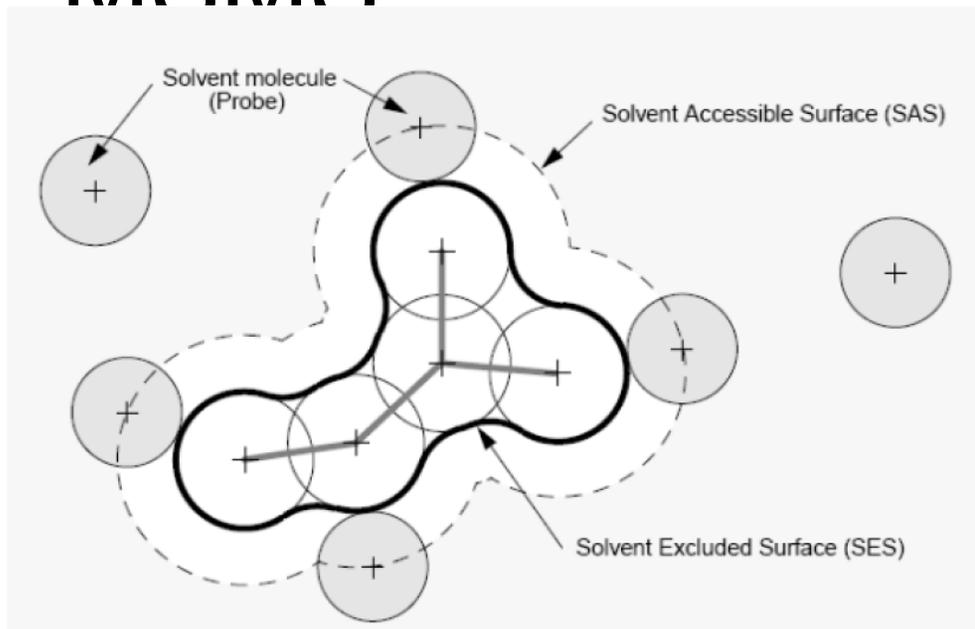
$$\bar{\kappa}^2 = 8.486902807 \text{ \AA}^{-2} I_s,$$



Numerical procedure

- Construction of molecular surface (by MSMS)
- Treatment of singular charges $C \sum q_i (x - x_i)$
- Nonlinear iteration by damped Newton's method for the perturbed equation
- Coupling interface method to solve elliptic interface problem
- Algebraic multigrid for solving linear systems

Construction of molecular surface: MSMS



The interface calculated by computer software MSMS of molecule 1crn with probe radius 1.4 and triangulation density 3.0.

Treatment of point charge singularities

separate singular part

$$\phi = \bar{\phi} + \tilde{\phi}.$$

where

$$\bar{\phi}(\mathbf{x}) = \begin{cases} \phi^*(\mathbf{x}) + \phi^0(\mathbf{x}) & \mathbf{x} \in \Omega_1 \\ 0 & \mathbf{x} \in \Omega_2 \cup \Omega_3 \end{cases}.$$

and ϕ^* is the potential in the free space induced by Q , i.e.

$$\phi^*(\mathbf{x}) = \begin{cases} C \sum_{i=1}^m \frac{1}{\epsilon_1} \frac{z_i}{4\pi |\mathbf{x} - \mathbf{x}_i|}, & \mathbf{x} \in R^3 \\ C \sum_{i=1}^m -\frac{1}{\epsilon_1} \frac{z_i}{2\pi} \log(|\mathbf{x} - \mathbf{x}_i|), & \mathbf{x} \in R^2 \end{cases}.$$

ϕ^0 is a harmonic function in Ω_1 satisfying

$$\begin{cases} \Delta \phi^0 = 0 & \text{in } \Omega_1 \\ \phi^0 = -\phi^* & \text{on } \Gamma_1. \end{cases}$$

The introduction of ϕ^0 is to force $[\bar{\phi}] = 0$ across Γ_1 .

The correction potential satisfies

$$-\nabla \cdot \left(\epsilon(\mathbf{x}) \nabla \tilde{\phi}(\mathbf{x}) \right) + K(\mathbf{x}) \sinh(\tilde{\phi}(\mathbf{x})) = [\epsilon \bar{\phi}_n]_{\Gamma_1} \delta_{\Gamma_1}.$$

Thus, the point charge singularity is transferred into surface singularity.

Damped Newton's Method

$$-\nabla \cdot (\epsilon(\mathbf{x}) \nabla v^l) + K(\mathbf{x}) \cosh(\phi^l) v^l = \nabla \cdot (\epsilon(\mathbf{x}) \nabla \tilde{\phi}^l) - K(\mathbf{x}) \sinh(\phi^l) + [\epsilon \bar{\phi}_n]_{\Gamma_1}$$
$$\tilde{\phi}^{l+1} = \tilde{\phi}^l + v^l \quad (1)$$

Since the direction v^n is indeed a descent direction for the functional $E(\phi)$,

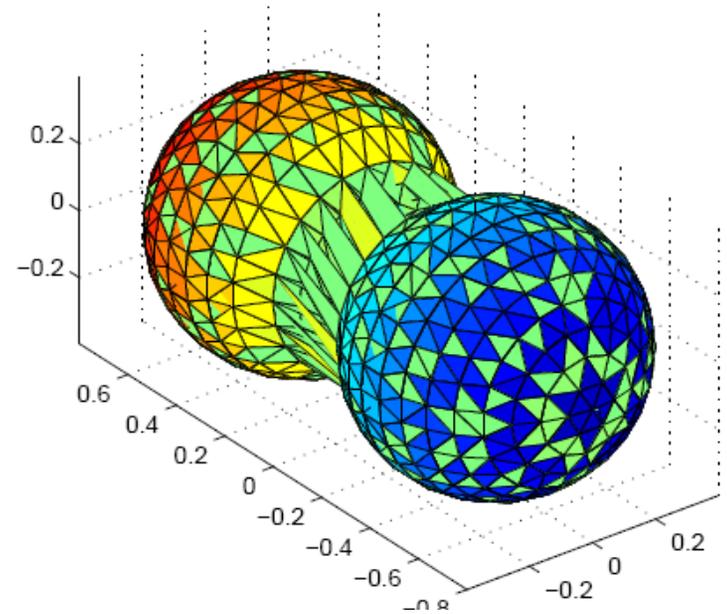
$$E(\phi^l + \lambda^l v^l) < E(\phi^l) \text{ for small } \lambda^l > 0,$$

we can accelerate the convergence of the Newton's method globally by performing a line search to find a suitable damping parameter λ^l that minimizes $E(\phi^l + \lambda^l v^l)$ and replace (1) by

$$\tilde{\phi}^{l+1} = \tilde{\phi}^l + \lambda^l v^l.$$

Numerical Validation—Artificial molecule

$$u_e(r) = \begin{cases} e^{-(x^2+y^2+z^2)} & r \in \Omega^- \\ 0 & r \in \Omega^+ \end{cases}$$



N	Newton iteration	$\ \nabla u - \nabla u_e\ _{\infty, \Gamma}$	order	$\ u - u_e\ _{\infty}$	order
10	4	6.572e-002	—	8.136e-003	—
20	3	1.378e-002	2.2538	2.025e-003	2.0064
40	3	3.115e-003	2.1292	4.901e-004	2.0467



Summary of computing Poisson-Boltzmann equation

- Ingredients: CIM + AMG + damped Newton's iteration
- Capable to handle complex interfaces
- Second order accuracy for potential and electric field for molecules with smooth surfaces