

# Modeling and Simulation of Charged Molecules with Legendre-Transformed Poisson-Boltzmann Electrostatic Free Energy Functional

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# Outline

- 1 Legendre-Transformed (LT) Poisson-Boltzmann (PB) functional
- 2 Applications to interface problem
- 3 Numerical methods and results
- 4 Conclusions

# Legendre-Transformed Poisson-Boltzmann functional

## Classical Poisson-Boltzmann energy functional

$$I[\phi] = \int_{\Omega} \left[ -\frac{\epsilon}{2} |\nabla\phi|^2 + f\phi - B(\phi) \right] dx$$

- $\Omega \subseteq \mathbb{R}^3$  : bounded region
- $\phi : \Omega \rightarrow \mathbb{R}$  : electrostatic potential
- $\epsilon : \Omega \rightarrow \mathbb{R}$  : dielectric coefficient
- $f : \Omega \rightarrow \mathbb{R}$  : fixed charge density
- $B : \mathbb{R} \rightarrow \mathbb{R}$  : strictly convex,  $B(0) = 0$  and  $B(\infty) = \infty$ .  
E.g.  $B(\phi) = \cosh(\phi)$

The Euler-Lagrange equation of  $I[\phi]$  is

$$\text{PBE : } \quad \nabla \cdot \epsilon \nabla \phi - B'(\phi) = -f$$

# Legendre-Transformed Poisson-Boltzmann functional

Legendre Transform: For function  $B$ ,  $\forall \xi \in \mathbb{R}$ ,  $B^*(\xi) = \sup_{s \in \mathbb{R}} (s\xi - B(s))$

## LT PB energy functional (Maggs, 2012)

$$J[D] = \int_{\Omega} \left[ \frac{1}{2\epsilon} |D|^2 + B^*(f - \nabla \cdot D) \right] dx + \int_{\partial\Omega} g D \cdot n dS$$

$$D : \Omega \rightarrow \mathbb{R}^3, D = -\epsilon \nabla \phi$$

## Equivalence between two functionals (Ciotti & Li, 2018)

$\forall \phi \in H_g^1(\Omega)$ ,  $\forall D \in H(\text{div}, \Omega)$ , we have  $I[\phi] \leq J[D]$

- $H_g^1(\Omega) = \{u \in H^1(\Omega) : u = g \text{ on } \partial\Omega\}$
- $H(\text{div}, \Omega) = \{D \in [L^2(\Omega)]^3 : \nabla \cdot D \in L^2(\Omega)\}$

Moreover, we have  $D_B = -\epsilon \nabla \phi_B$  and

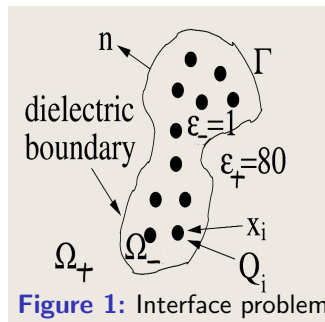
$$I[\phi_B] = \max_{\phi \in H_g^1(\Omega)} I[\phi] = \min_{D \in H(\text{div}, \Omega)} J[D] = J[D_B]$$

# Applications to interface problem

Classical PB energy functional:

$$I_{\Gamma}[\phi] = \int_{\Omega} \left[ -\frac{\epsilon_{\Gamma}}{2} |\nabla \phi|^2 + f\phi - \chi_{+} B(\phi) \right] dx$$

$$\epsilon_{\Gamma} = \begin{cases} \epsilon_{+} & \text{in } \Omega_{+} \\ \epsilon_{-} & \text{in } \Omega_{-} \end{cases}$$



LT PB energy functional:

$$J_{\Gamma}[D] = \int_{\Omega} \left[ \frac{1}{2\epsilon_{\Gamma}} |D|^2 + \chi_{+} B^{*}(f - \nabla \cdot D) \right] dx + \int_{\partial\Omega} g D \cdot n dS, \forall D \in V_{\Gamma}$$

$$V_{\Gamma} = \{ D \in H(\text{div}, \Omega) : \nabla \cdot D = f \text{ in } \Omega_{-} \}$$

## Equivalence between two functionals (Ciotti & Li, 2018)

$\forall \phi \in H_g^1(\Omega), \forall D \in V_\Gamma$ , we have

$$I_\Gamma[\phi] \leq J_\Gamma[D]$$

Moreover, we have

$$I_\Gamma[\phi_\Gamma] = \max_{\phi \in H_g^1(\Omega)} I_\Gamma[\phi] = \min_{D \in V_\Gamma} J_\Gamma[D] = J_\Gamma[D_\Gamma]$$

Here,  $D_\Gamma = -\epsilon_\Gamma \nabla \phi_\Gamma$ .

# Numerical methods and results

## Problem

Convex optimization problem with constraint:

$$\begin{cases} \min J_{\Gamma}[D] \\ \text{s.t. } \nabla \cdot D = f \text{ in } \Omega_- \end{cases}$$

## Penalty method

Given a penalty coefficient  $\mu$ , we need to minimize the following:

$$J_{\Gamma,\mu}[D] = J_{\Gamma}[D] + \frac{1}{\mu} \int_{\Omega_-} (\nabla \cdot D - f)^2 dx$$

$$= \int_{\Omega} \left[ \frac{1}{2\epsilon_{\Gamma}} |D|^2 + \chi_+ B^*(f - \nabla \cdot D) + \frac{1}{\mu} \chi_- (\nabla \cdot D - f)^2 \right] dx + \int_{\partial\Omega} gD \cdot n dS$$

One good optimization algorithm is the limited-memory BFGS method.

$$J_{\Gamma,\mu}[D] = J_{\Gamma}[D] + \frac{1}{\mu} \int_{\Omega_-} (\nabla \cdot D - f)^2 dx$$

$$= \int_{\Omega} \left[ \frac{1}{2\epsilon_{\Gamma}} |D|^2 + \chi_+ B^*(f - \nabla \cdot D) + \frac{1}{\mu} \chi_- (\nabla \cdot D - f)^2 \right] dx + \int_{\partial\Omega} g D \cdot n dS$$

## Convergence theorem of penalty method

- 1 For each  $\mu$ , there exists a unique  $D_{\Gamma,\mu}$  which minimizes  $J_{\Gamma,\mu}[D]$
- 2  $\min J_{\Gamma,\mu}[D] \rightarrow \min_{D \in V_{\Gamma}} J_{\Gamma}[D]$  as  $\mu \rightarrow 0$
- 3  $D_{\Gamma,\mu} = \operatorname{argmin} J_{\Gamma,\mu}[D] \rightarrow \operatorname{argmin}_{D \in V_{\Gamma}} J_{\Gamma}[D] = D_{\Gamma}$  as  $\mu \rightarrow 0$



Computational region  $\Omega = [-1, 1]^3$ .

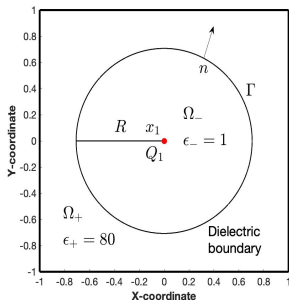
## Particular solution

For  $\Gamma$  which is a sphere in  $\mathbb{R}^3$ , one particular solution  $D_{\Gamma, \mu}$  is given by

$$D_{\Gamma, \mu}(r) = \begin{cases} \frac{\epsilon_+ \exp(-\lambda r)(\lambda r + 1)}{r^3} \mathbf{v} & \text{in } \Omega_+ \\ -2C\epsilon_- \mathbf{v} & \text{in } \Omega_- \end{cases}$$

Here  $\mathbf{v} = (x, y, z)$ ,  $r$  is the distance to the origin,  $C$  is a constant to be decided in terms of  $\mu$ ,  $\lambda = \sqrt{\frac{1}{\epsilon_+}}$  is also a constant.

# Numerical methods and results



**Figure 2:** Computational domain

In numerical test, constants include

- Radius  $R = 1/\sqrt{2}$
- Tolerance for the norm of the gradient  $= 3 * 10^{-6}$

Parameters include

- Number of intervals on each direction  $N$
- Penalty coefficient  $\mu$

## Result for $\mu = 1$ , change $N$

N	L1 rel. error	L2 rel. error	Time	Epoch	Order
10	0.6250	0.6286	5s	1131	-
20	0.2944	0.4386	1m	1910	0.6119
40	0.1296	0.3046	20m	3867	0.5858
80	0.0657	0.2110	2h	5895	0.5564
120	0.0445	0.1716	14h	8439	0.5222

**Table 1:** Result for  $\mu = 1$

The minimal value of the functional is approximately  $1.23 * 10^4$ . The order of convergence is approximately 0.5.

# Result for $\mu = 1$ , change $N$

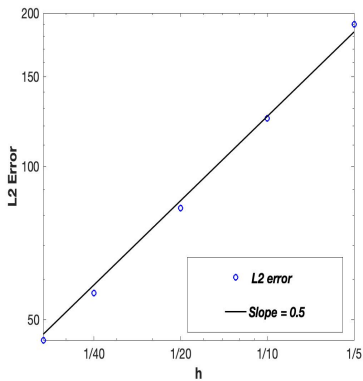
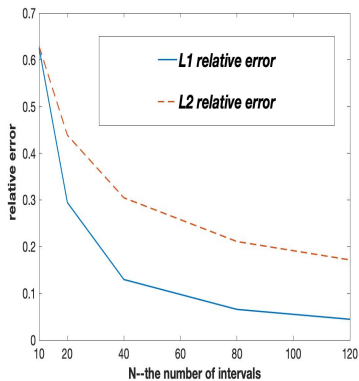


Figure 3: Result for  $\mu = 1$

## Result for $\mu = 10^{-2}$ , change $N$

N	L1 rel. error	L2 rel. error	Time	Epoch	Order
10	0.6209	0.6250	30s	10983	-
20	0.2938	0.4371	13m	23582	0.6068
40	0.1298	0.3043	3.5h	56498	0.5818
80	0.0657	0.2109	3d8h	128958	0.5551

**Table 2:** Result for  $\mu = 10^{-2}$

The minimal value of the functional is approximately  $1.19 * 10^4$ . The order of convergence is approximately 0.5.

# Result for $\mu = 10^{-2}$ , change $N$

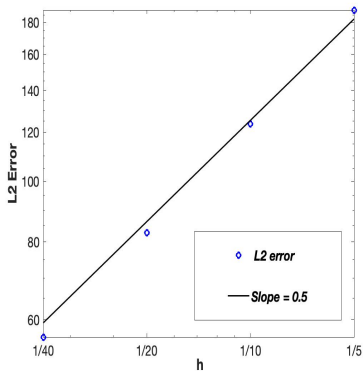
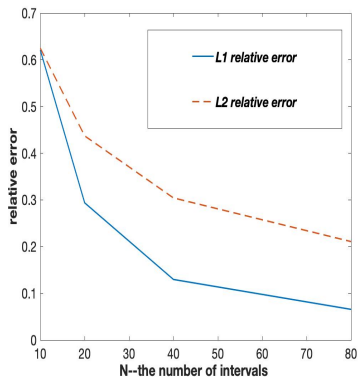


Figure 4: Result for  $\mu = 10^{-2}$

## Result for $\mu = 10^{-5}$ , change $N$

N	L1 rel. error	L2 rel. error	Time	Epoch	Order
10	0.6209	0.6250	2m	5576	-
20	0.2938	0.4371	40m	80235	0.6067
40	0.1298	0.3043	5d	928830	0.5817

**Table 3:** Result for  $\mu = 10^{-5}$

The minimal value of the functional is approximately  $1.15 * 10^4$ . The order of convergence is approximately 0.5.

# Result for $\mu = 10^{-5}$ , change $N$

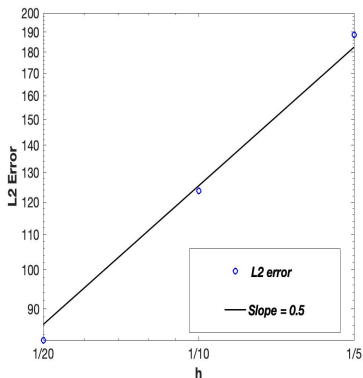
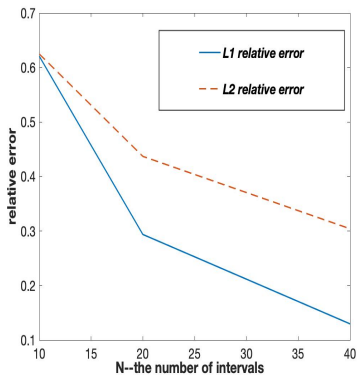


Figure 5: Result for  $\mu = 10^{-5}$



## Result for $N = 40$ , change $\mu$

$\mu$	L1 rel. error	L2 rel. error	Time	Epoch	Value
$10^5$	1.4923	2.4879	30m	2150	$2.92 \cdot 10^7$
$10^4$	0.9509	1.5560	10m	3018	$2.92 \cdot 10^6$
$10^3$	0.3119	0.5732	10m	3003	$3.02 \cdot 10^5$
$10^2$	0.1456	0.3360	5m	2588	$4.05 \cdot 10^4$
10	0.1296	0.3077	10m	2520	$1.44 \cdot 10^4$
1	0.1296	0.3046	20m	3867	$1.18 \cdot 10^4$
$10^{-1}$	0.1297	0.3043	1h	11960	$1.15 \cdot 10^4$
$10^{-2}$	0.1298	0.3043	3h30m	56498	$1.15 \cdot 10^4$
$10^{-5}$	0.1298	0.3043	5d	928830	$1.15 \cdot 10^4$

**Table 4:** Result for  $N = 40$

# Test for constraint

## Numerical quadrature in $\Omega_-$

As  $\mu \rightarrow 0$ ,

$$E(D) = \int_{\Omega_-} (\nabla \cdot D - f)^2 dx \rightarrow 0$$

N	$\mu$	E(D)	Min Value of LT PB functional
20	1	262.2015	$1.1317 \cdot 10^4$
20	$10^{-2}$	0.0262	$1.1039 \cdot 10^4$
20	$10^{-5}$	$2.6250 \cdot 10^{-8}$	$1.1036 \cdot 10^4$

**Table 5:** Result for numerical quadrature in  $\Omega_-$

# Conclusions

- ① LT PB functional: A novel convex functional for electrostatic energy.
- ② Penalty method converges: Theoretical proof and numerical test.
- ③ Further improvement: Combine the model with the level-set method for molecular dynamics.