

An Efficient and Accurate Model for Water with an Improved Non-Bonded Potential

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A molecular mechanical model for liquid water is developed that uses a physically-motivated potential to represent Pauli repulsion and dispersion instead of the standard Lennard-Jones potential. The model has three-atomic sites and a virtual site located on the \angle HOH bisector (i.e., a TIP4P-type model). Pauli-repulsive interactions are represented using a Buckingham-type exponential decay potential. Dispersion interactions are represented by both C_6/r^6 and C_8/r^8 terms. This higher order C_8 dispersion term has been neglected by most force fields. The ForceBalance code was used to define parameters that optimally reproduce the experimental physical properties of liquid water. The resulting model is in good agreement with the experimental density, dielectric constant, enthalpy of vaporization, isothermal compressibility, thermal expansion coefficient, diffusion coefficient, and radial distribution function. A GPU-accelerated implementation of this improved non-bonded potential can be employed in OpenMM without modification by using the CustomNonBondedForce feature. Efficient and automated parameterization of these non-bonded potentials provides a rational strategy to define a new molecular mechanical force field that treats repulsion and dispersion interactions more rigorously without major modifications to existing simulation codes or a substantially larger computational cost.

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I. INTRODUCTION

Molecular mechanical force fields underlie many simulations of materials and biomolecules. These models must effectively capture the significant intermolecular interactions present in a system using computationally-efficient functions. Although these interactions all originate from complex electron–electron, proton–proton, and electron–proton Coulombic interactions, they can be effectively simplified into pairwise electrostatic, London dispersion, and Pauli repulsion interactions.

In most popular molecular mechanical models, electrostatic interactions are described by Coulombic interactions between a set of point charges (q) at atomic centers or positions defined with respect to those atomic centers. Some more elaborate models extend this to include the effects of induced polarization and charge transfer,^{1–6} but force field developers have been largely successful in identifying static charges that capture these complex electrostatic interactions in an effective way.

London dispersion is an ubiquitous, attractive intermolecular force arising through interaction between instantaneous electric moments in neighboring atoms. To a reasonable approximation, this interaction can be approximated as a pairwise sum,

$$\mathcal{V}_{disp}(r) = -\frac{C_6}{r^6} - \frac{C_8}{r^8} \tag{1}$$

The magnitude coefficients C_6 and C_8 depend on the strength of the dispersion interaction between the interacting atoms. The C_6 term is the strongest and longest-range term, although quantum chemical analysis and equations of state have found the C_8 term can also make a substantial contribution to dispersion interactions. Odd-numbered terms, like C_7 and C_9 , are less significant for water molecules because of the small magnitude of these coefficients.⁷

Pauli repulsion originates from the overlap of electron density clouds of atoms at close range. As the electron density of atoms follows an exponential dependence, the interaction potential can be described accurately as an exponential decay (i.e., $A \cdot \exp(-br)$), but polynomial terms are often used instead (e.g., A/r^{12}). In most popular molecular mechanical models, the dispersion and repulsive terms are represented using the Lennard-Jones 12-6 potential⁸,

$$\mathcal{V}_{nb}(r) = \frac{A_{LJ}}{r^{12}} - \frac{C_6}{r^6}. \quad (2)$$

In this potential, the A_{LJ}/r^{12} term is intended to represent Pauli repulsion, while the $-C_6/r^6$ term represents London dispersion interactions. Higher order dispersion terms are neglected.

These models require the definition of atomic charges and the A_{LJ} and C_6 Lennard-Jones parameters. Generally, the parameters are fit such that the properties predicted by the model are as close as possible to the physical properties of the liquid. Because these intermolecular potentials and simulation algorithms are inexact, the parameters may take on “effective” values that yield accurate predictions of the targeted physical properties but are inconsistent with the true nature of the intermolecular interactions between molecules in the liquids.

The correct strength of dispersion interactions has been a contentious subject in biomolecular simulation.^{9–12} One proposal is that short-range dispersion interactions are overestimated because of the use of long-range cutoffs of dispersion interactions.¹³ In a previous studies by our group, force field dispersion parameters were compared to *ab initio* values calculated using the eXchange-hole Dipole Moment (XDM) model.^{14,15} This analysis showed that molecular C_6 coefficients vary widely between force fields, but they were systematically higher than the *ab initio* values. This trend was attributed to the neglect of higher order dispersion terms from the Lennard-Jones potential.

There are several alternatives to the Lennard-Jones potential that could resolve some of these issues. Buckingham proposed an intermolecular function to describe the interactions of noble gases¹⁶,

$$\mathcal{V}_{nb}(r) = A \cdot \exp(-b \cdot r) - \frac{C_6}{r^6} - \frac{C_8}{r^8}. \quad (3)$$

Here, the repulsive interaction is described by the $A \cdot \exp(-b \cdot r)$ term, where the A and b parameters define the strength of the repulsion. Dispersion interactions are represented by the C_6/r^6 and C_8/r^8 terms. This potential has several advantages over the Lennard-Jones potential. The exponential term is a more realistic description of Pauli repulsion than the polynomial A/r^{12} term, which is advantageous in simulations where there are frequent repulsive contacts because of strong attractive interactions, high temperatures, or high pressures.

Wang et al. successfully employed a modified Buckingham repulsive potential to develop an improved water model, but this model only included a C_6/r^6 dispersion term.¹⁷ Explicit inclusion of the C_8/r^8 term describes the dispersion interaction more realistically as the sum of the C_6/r^6 term and shorter-range C_8/r^8 term, instead of the current practice where all dispersion interactions are effectively included in the C_6/r^6 term.

One issue with the basic form of the Buckingham potential is that the potential becomes infinitely negative as $r \rightarrow 0$. This is a consequence of the exponential term being finite at $r = 0$ but the dispersion terms becoming infinitely negative. To resolve this issue, Tang and Toennis proposed that the dispersion terms be damped by an incomplete gamma function¹⁸,

$$f_{damp,n}(r) = 1 - \exp(-\zeta r) \sum_{k=0}^n \frac{(\zeta r)^k}{k!} \quad (4)$$

Multiplying the dispersion terms of Eqn. 3 by the corresponding damping terms (Eqn. 4) yields

$$\mathcal{V}_{nb}(r) = A \cdot \exp(-b \cdot r) - f_{damp,6}(r) \frac{C_6}{r^6} - f_{damp,8}(r) \frac{C_8}{r^8}. \quad (5)$$

The degree of the damping is determined by the coefficient, ζ , a parameter that corresponds to the strength of the damping. Based on the work of Tang and Toennis, a value of $\zeta = 35.8967 \text{ nm}^{-1}$ was used in the simulations presented here.

This non-bonded potential is much more amenable for calculating the long-range component of the C_6 interaction energy using lattice summation methods.¹⁹ As the C_8/r^8 term is shorter range than the C_6/r^6 term, a non-bonded cutoff can be applied to this term without neglecting a large component of the dispersion energy.

Although this type of intermolecular potential has found use in models related to chemical physics, molecular simulation of condensed matter still overwhelmingly employ the Lennard-Jones potential to describe these interactions. The foremost barrier to adopting these non-bonded potentials is that it would be necessary to define a complete set of parameters to describe the interaction of each pair of atoms. Determination of optimal parameters has generally been a slow process, where a large set of parameter combinations must be tested. In 2014, Leeping, Martinez, and Pande released the ForceBalance code²⁰, which allows optimal parameters for force fields to be determined using a gradient-directed optimization of the parameters.

The ForceBalance code was successfully used to develop new parameters for the TIP3P and TIP4P water models that best described the physical properties of water. A target function was defined based on the enthalpy of vaporization (ΔH_{vap}), density (ρ), isothermal compressibility (κ_T), heat capacity (C_p), dielectric constant (ε_0), and the thermal expansion coefficient (α) of liquid water at standard conditions (298.15 K, 101.325 kPa).

$$\begin{aligned}
\delta = & b_1 \cdot (\rho_{ref} - \rho_{calc})^2 + \\
& b_2 \cdot (\Delta H_{vap,ref} - \Delta H_{vap,calc})^2 + \\
& b_3 \cdot (\kappa_{T,ref} - \kappa_{T,calc})^2 + \\
& b_4 \cdot (C_{p,ref} - C_{p,calc})^2 + \\
& b_5 \cdot (\varepsilon_{0,ref} - \varepsilon_{0,calc})^2 + \\
& b_6 \cdot (\alpha_{ref} - \alpha_{calc})^2
\end{aligned} \tag{6}$$

The models derived from this optimization, termed TIP3P-FB and TIP4P-FB, provided a significantly improved description of the physical, transport, and structure properties of water. This code provides a viable path to determining appropriate parameters for simulations using the potential defined in Eqn. 5.

The cornerstone of any explicit atomistic biomolecular force field is a model for liquid water. Dozens of these models have been developed.^{21,22} Non-polarizable models with a Lennard-Jones non-bonded potential are the most widely supported in popular molecular simulation codes. Although the TIP3P²³, SPC,²⁴ and SPC/E²⁵ water models are still widely used, more recently, researchers have developed new parameter sets that provide more accurate descriptions of liquid water. The TIP4P/2005,²⁶ TIP3P-FB,²⁰ TIP4P-FB,²⁰ and OPC²⁷ water models are notably successful models, although they all employ the Lennard-Jones potential to represent Pauli repulsion and dispersion interactions.

Although these established Lennard-Jones based water models are effective in describing many of the physical, structural, and transport properties of liquid water, it would be difficult to improve their performance in describing the solvation of molecules, parameterize their interactions based on quantum mechanical data directly, or to assign physical meaning to the Lennard-Jones interactions on an atomic basis. In this work, we present the development of a new model for liquid water using ForceBalance. This water model is intended to serve as the cornerstone of a new force field based on this improved potential.

II. COMPUTATIONAL METHODS

The simulation cell used in the optimization of the model was cubic with an initial side-length of 18.64 Å, containing 215 water molecules. Electrostatic interactions were calculated using the Particle Mesh Ewald method.²⁸ A correction for the long range component of the C_6 term of the dispersion interaction was included using the method of Shirts et al.²⁹

All simulations were performed using OpenMM 7.2.³⁰ Eqn. 5 was implemented using OpenMM’s CustomNonbondedForce feature that allows the Lennard-Jones interaction potential to be replaced with an alternative equation. For better expression code readability, a Python package, OpenMM Transformer³¹, was used to generate the custom force expressions for Eqn. 5. The ForceBalance code was used to generate optimal parameters using a gradient-directed optimization of the target function defined in Eqn. 7.

$$\begin{aligned}
 \delta(A, b, C_6, C_8, q_{LP}, l_{LP}) = & b_1 \cdot (\rho_{ref} - \rho_{calc})^2 + \\
 & b_2 \cdot (\Delta H_{vap,ref} - \Delta H_{vap,calc})^2 + \\
 & b_3 \cdot (\kappa_{T,ref} - \kappa_{T,calc})^2 + \\
 & b_4 \cdot (C_{p,ref} - C_{p,calc})^2 + \\
 & b_5 \cdot (\varepsilon_{0,ref} - \varepsilon_{0,calc})^2 + \\
 & b_6 \cdot (\alpha_{ref} - \alpha_{calc})^2
 \end{aligned}
 \tag{7}$$

The enthalpy of vaporization was calculated from the average potential energy of the simulation,

$$\Delta H_{vap} = RT - \langle \mathcal{V} \rangle_{liq} / N_{mol}
 \tag{8}$$

where N_{mol} is the number of molecules in the simulation and \mathcal{V} is the potential energy.

The thermal expansion coefficient (α) was calculated from,

$$\alpha = \frac{[\langle V \cdot \mathcal{V} \rangle - \langle U \cdot \mathcal{V} \rangle + P(\langle V^2 \rangle - \langle V \rangle^2)]}{k_B T^2 \langle V \rangle}
 \tag{9}$$

where V is the volume of the system.

The isothermal compressibility was calculated from,

$$\kappa_T = \frac{\langle V^2 \rangle - \langle V \rangle^2}{k_B T \langle V \rangle}
 \tag{10}$$

The dielectric constant was calculated using the relation

$$\varepsilon_0 = 1 + \frac{4\pi(\langle M^2 \rangle - \langle M \rangle^2)}{3\langle V \rangle k_B T} \quad (11)$$

where M is the net dipole moment of the simulation cell.

The diffusion coefficient was calculated from an NVE trajectory of the system using the Einstein relation with the correction for finite-size effects by Yeh and Hummer³²,

$$D = \frac{1}{6t} \langle |r_i(t) - r_i(0)|^2 \rangle + 2.837297 \frac{k_B T}{6\pi\eta L}. \quad (12)$$

III. RESULTS AND DISCUSSION

TABLE I. Optimal parameters for a 4-point water model with the Buckingham 6-8 potential.

parameter	value
A (kJ mol ⁻¹)	1.598577×10^6
b (nm ⁻¹)	20.88390
C_6 (kJ mol ⁻¹ nm ⁶)	2.853362×10^{-3}
C_8 (kJ mol ⁻¹ nm ⁸)	3.196540×10^{-5}
ζ (nm ⁻¹)	35.8967
q_O (e)	-1.05519892369
l_{LP} (nm)	0.0106

The ForceBalance parameter optimization method was used to determine the optimal parameters for the Buckingham-type potential with 6th and 8th order dispersion (B68) water model. The values of the A , b , C_6 , q_O , and C_8 were allowed to change freely in this optimization. The optimization was repeated for values of l_{LP} between 0.009 nm and 0.12 nm, with the final value of 0.0106 nm yielding the minimum value of the target function. For this optimization, 34 iterations were needed to converge to an optimal set of parameters. These parameters are presented in Table I.

This model is in generally good agreement for the targeted physical properties (Table II). The dielectric constants are considerably improved over the TIP3P water model. The self-diffusion coefficient predicted by the B68 model is 0.3×10^{-5} cm²/s faster than the experimental value. The O–O radial distribution function of this model is presented in

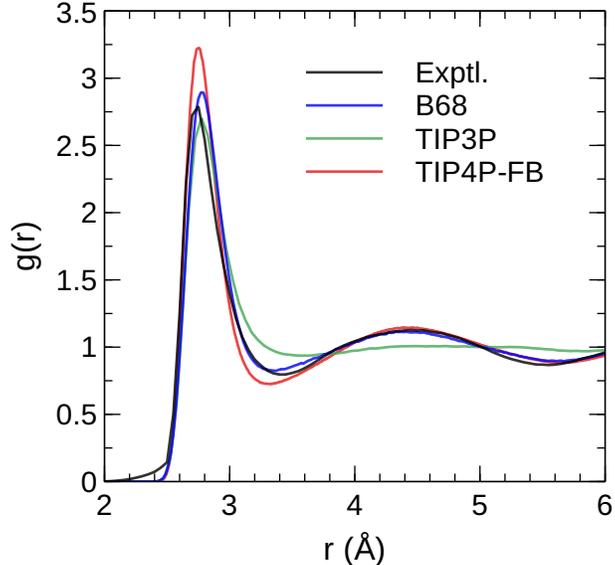


FIG. 1. Radial distribution functions of the TIP3P, TIP4P-FB, and B68 water models compared to the experimental profile from Ref. 33.

TABLE II. Properties of liquid water (298.15 K and 101.32 kPa) predicted by the optimized B68 water model. Density, dielectric constant, enthalpy of vaporization, thermal expansion coefficient, isothermal compressibility, and heat capacity. The properties predicted by the TIP3P and TIP4P-FB models from Ref. 20 are included for comparison. The experimental values are taken from Refs. 34, 35, 36.

Property	TIP3P	TIP4P-FB	B68	Exptl.
ρ (kg m ⁻³)	983.6	995.7	996.9 \pm 0.9	997.075 \pm 0.001
ϵ_0	95.8	77.3	78.4 \pm 6.7	78.4
ΔH_{vap} (kJ \cdot mol ⁻¹)	37.11	45.21	44.04 \pm 0.07	43.99
α (10 ⁻⁴ K ⁻¹)	9.05	2.44	2.53 \pm 1.42	2.5721 \pm 0.0005
κ_T (10 ⁻⁶ bar ⁻⁶)	57.8	45.3	44.2 \pm 3.4	45.2472 \pm 0.0003
C_p (J \cdot mol ⁻¹ \cdot K ⁻¹)	69.58	79.33	73.51 \pm 3.21	75.33
D (10 ⁻⁵ cm ² /s)	6.05	2.21	2.68 \pm 0.07	2.29

Figure 1, along with those for Lennard-Jones based TIP3P and TIP4P-FB models and an experimental profile determined by X-ray scattering experiments.³³ The first peak of the B68 radial distribution function is less steep than TIP4P-FB model, which is in better agreement

with the experimental data. This is likely due to the softer exponential repulsive potential in the B68 model. The region corresponding to the first minimum of the rdf ($r = 2.5 - 3.8 \text{ \AA}$) is in incrementally better agreement with the experimental data than the TIP4P-FB model.

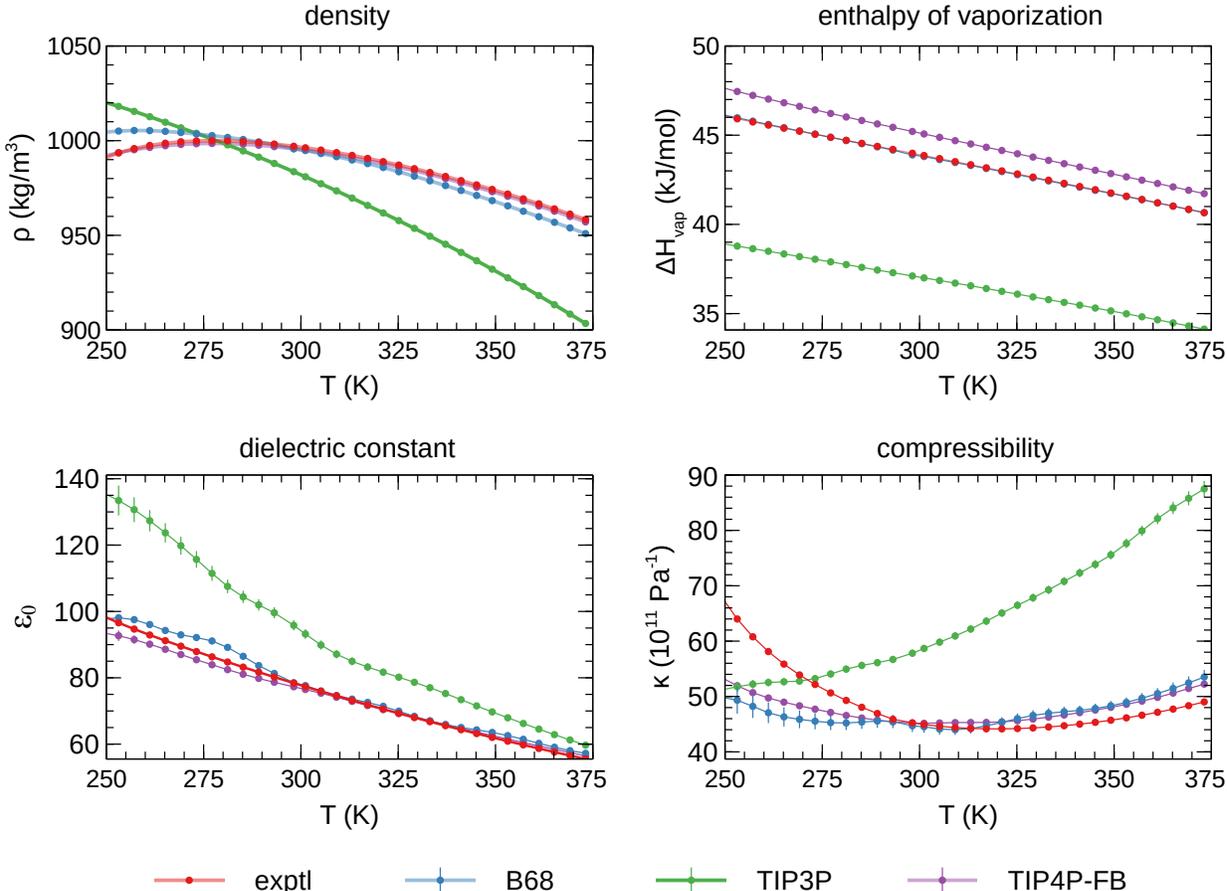


FIG. 2. Temperature dependence of liquid water properties over the temperature range of 270 K to 350 K calculated using the TIP3P, TIP4P-FB, and B68 water models compared to the experimental values. Data for TIP3P and TIP4P-FB are taken from Ref. 20. The curve for the enthalpy of vaporization for the B68 model overlaps with the experimental curve.

The temperature dependence of the dielectric constant, thermal expansion coefficient, compressibility, and density of liquid water calculated using these models is presented in Figure 2. The TIP3P potential is a poor model for the temperature dependence of all four properties. The B68 model is in generally good agreement with the experimental values over the full temperature range, with similar performance to the TIP4P-FB water model for most properties, although the B68 model performs somewhat less accurately than the

TABLE III. The simulation rate of MD simulations using the B68 model and a Lennard-Jones model (TIP4P-FB) and for a $64 \times 64 \times 64 \text{ \AA}^3$ simulation cell of liquid water. All values are in ns/day using a 2 fs time step. A 3.2 GHz Intel(R) i7-8700 CPU with a Titan Xp NVIDIA GPU was used in these benchmarks.

	TIP4P-FB	B68
CPU	6.17	0.61
CUDA	170.0	113.1
OpenCL	143.0	90.7

TIP4P-FB model at temperatures below 298 K.

Historically, part of the justification of the use of Lennard-Jones potentials was that they could be calculated using a small number of multiplication operations, while more complex potentials like Eqn. 5 require the evaluation of additional terms, like the exponential functions. To assess the additional computational cost of the more complex B68 potential in comparison to traditional Lennard-Jones models, benchmark simulations were performed for a $64 \text{ \AA} \times 64 \text{ \AA} \times 64 \text{ \AA}$ simulation cell containing 8673 water molecules. For comparison, simulations were also performed using the TIP4P-FB water model, which employs a conventional Lennard-Jones non-bonded potential. These benchmarks are presented in Table III. When run on the CPU only, the B68 simulations are 10 times slower. This can be attributed to the calculations of exponential terms in the repulsive term and damping functions in the B68 potential, which are computationally-slower operations than the purely polynomial terms in the Lennard-Jones potential. The difference in performance is much smaller for the GPU implementations; the B68 CUDA simulation runs at 113 ns/day vs 170 ns/day for the CPU implementation. The OpenCL implementation shows a similar trend but the performance for both models is slower than the CUDA implementation. We note that this code is generated automatically and more efficient implementations may be possible if these potentials are explicitly added to the simulation codes. Significantly, the CustomNonBondForce code performs part of the energy and force evaluation using slower double precision operations. Nevertheless, these results show that the Lennard-Jones potential can be replaced with a more physically-motivated potential at an acceptable computational cost.

The C_6 coefficient of this model is similar to the XDM-calculated value (B68 C_6 coefficient:

$2.86 \times 10^{-3} \text{ kJ mol}^{-1} \text{ nm}^6$, XDM: $2.57 \times 10^{-3} \text{ kJ mol}^{-1} \text{ nm}^6$) but the C_8 coefficient is 6 times smaller (B68: $3.20 \times 10^{-5} \text{ kJ mol}^{-1} \text{ nm}^8$, XDM: $1.83 \times 10^{-4} \text{ kJ mol}^{-1} \text{ nm}^8$). This may reflect the small effect of dispersion interactions on the intermolecular interactions of water relative to electrostatic interactions, so these parameters may not hold truly physical values when there are other significant approximations in the model (i.e., induced polarization). Secondly, the model is sensitive to the damping coefficients and the selection of different values for these terms may yield different dispersion coefficients. Given the small magnitude of the C_8 coefficients for this model, pair-specific dispersion coefficients may be needed to describe dispersion interactions between water and other species. For liquid water, the use of a Buckingham-type repulsive term appears to be more significant than the inclusion of C_8 dispersion.

IV. CONCLUSIONS

Using the ForceBalance code, a molecular mechanical model for liquid water was developed. This model differs from established water models because it replaces the Lennard-Jones non-bonded potential with a more sophisticated potential that describes pairwise interatomic Pauli repulsion using a Buckingham-type exponential function and includes damped C_6 and C_8 dispersion interactions. The physical properties and radial distribution function are in excellent agreement with experimental data. The CustomNonBondForce feature of OpenMM was used to generate a GPU-accelerated implementation of this potential automatically. Benchmark simulations show this potential is only modestly more computationally-intensive than conventional Lennard-Jones-based potentials. The structural and physical properties of liquid water predicted by this model is comparable to models like TIP4P-FB, which employ the Lennard-Jones potential, although the more realistic description of the intermolecular potential of this model could open new opportunities to develop improved molecular mechanical force fields.

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AIP PUBLISHING DATA SHARING POLICY

The Python scripts and sample input files to reproduce the results of this study are openly available on our GitHub repository, <https://github.com/mohebifar/openmm-transformer/>.³¹

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