Accelerated kinetic over-relaxation for dynamic energy minimization in lattice many-body problems

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A kinetic over-relaxation minimizer, based on a lattice version of the Boltzmann equation is presented. The method is validated against standard Metropolis Monte Carlo, and proves very effective in attaining (global) minima of classical pair potentials, involving solid body rotations. Linear scaling of the computational time to minimum with the system size is demonstrated.

Introduction.— Finding the minimum energy configuration in classical and quantum many-body systems is crucial for many applications in science and engineering [1-3]. The highly-dimensional configuration-space of many-body systems confines analytical results to very special cases [4], and also commands major extensions of standard minimization tools from linear and non-linear algebra. To this regard, dynamic minimization procedures in real or fictitious time, such as Monte Carlo (MC) [5, 6] or Langevin Molecular Dynamics (LMD) [7, 8] and their modern variants, have proven very effective for the last decades. Here, we present a new kind of dynamic minimizer, which is based on a lattice version of Boltzmann’s kinetic equation, Lattice Boltzmann (LB) for short. The method is applicable to a class of lattice potentials involving solid body rotations. The LB propagation-relaxation dynamics proves very effective in reaching the (global) minimum energy configuration of lattice many-body potentials. The time to minimum is found to be a fraction of a standard Metropolis Monte Carlo, and, perhaps more importantly, it is found to scale linearly with the system size.

The physical problem.— We represent water molecules by solid bodies, sitting on the nodes of a regular lattice, consisting of a finite number of point particles carrying either mass and/or charge, and moving along the links of the lattice at discrete times. Here, we consider the standard so-called D3Q27 lattice, consisting of the 26 neighbors of the center node of a cubic cell (6 face centers, 12 edge centers and 8 vertices). In fig. 1 (a) a typical configuration of water molecules on the lattice is shown. The centers of mass (COM) of the water molecules are fixed at the lattice nodes. The molecules are free to rotate about their COM under the action of a torque arising from the intermolecular interaction. Each molecule interacts with its 26 neighbors and its orientation is characterized by the quaternion technique [9]. In this way, in order to determine the rotation matrix one needs the time derivative of the components of the quaternion which are functions of the angular velocity Ω.

The angular velocity evolves in time under the effect of the torque:

\[ \mathbf{T}(\mathbf{x}) = -\partial_q \sum_y V(q(\mathbf{x}), q(\mathbf{y})) \] (1)

where the angular potential \( V(q(\mathbf{x}), q(\mathbf{y})) \) is a function of the orientations \( q(\mathbf{x}) \) and \( q(\mathbf{y}) \) of the interacting molecules located at \( \mathbf{x} \) and \( \mathbf{y} \). In this paper, we shall explore two specific lattice potentials for the water based on the TIP3P [10, 11] and the three-dimensional Ben-Naim (BN3d) [13] potential, their respective geometry being given in fig. 1 (b) and (c), respectively.

Lattice TIP3P (LTIP3P) model.— In the LTIP3P model the potential of two neighboring molecules located at lattice nodes \( i \) and \( j \) is given by

\[ V(\mathbf{x}_i, \mathbf{x}_j) = \frac{1}{4\pi\epsilon} \sum_{k=1}^3 \sum_{l=1}^3 \frac{Q_k Q_l}{|\mathbf{x}_i - \mathbf{x}_k|} \] (2)

in which \( \epsilon \) is the relative permittivity. The indices \( k \) and \( l \) refer to the charges of the water molecules associated with two hydrogens and one oxygen atoms for each molecule. Both hydrogens carry the same positive charge (+Q/2) while the oxygen carries a negative charge (−Q). The Lennard-Jones potential is neglected since the COM of the water molecules are fixed at the lattice nodes. We use the standard parameters for TIP3P water model given in [10]. Besides, the distance between the COM of a molecule with its nearest neighbor is fixed to 2.77 Å, as observed from regular molecular dynamics simulation of the bulk liquid [12]. In order to determine the full potential of a single molecule we sum over the first shell of neighboring molecules only (26 molecules).

Lattice BN3d (LBN3d) model.— In the LBN3d model the potential of two neighboring water molecules located at lattice nodes \( i \) and \( j \) (with the tetrahedral structure carrying two donors and two acceptor) is given by the BN3d potential [13] as
Tetrahedral geometry of two neighbor H atoms

energy configuration is reached. To this purpose, the Eqs. (4) and (5) are evolved until the (global) minimum is monitored in time. It has been shown that the decay of the total energy is associated to the dynamic formation of hydrogen bonds between the water-like molecules. Hence, the number of hydrogen bonds per molecule (HBs) serves as a representative order parameter to describe the energy landscape of this classical many-body system.

**The computational method.** — The Langevin equation is solved in the over-damped approximation, to deliver \( \Omega = (I^{-1} \cdot T)/\gamma \). The quaternion equation is solved by means of a Lattice Boltzmann technique. In this formulation, each lattice node hosts a set of discrete distributions \( q_{i,\mu} \), \( i = 0, n \), connecting each lattice site to its \( n \) neighbors (here \( n = 26 \)). The lattice is chosen with sufficient symmetry (D3Q27) to support macroscopic advection and diffusion phenomena. The LB dynamics reads as follows

\[
q_{i,\mu}(\mathbf{x} + \mathbf{c}_i \Delta t, t + \Delta t) = (1 - \omega)q_{i,\mu}(\mathbf{x}, t) + \omega q_{i,\mu}^eq(\mathbf{x}, t) + \dot{q}_{i,\mu}(\mathbf{x}, t) \Delta t
\]

where \( \omega = \Delta t/\tau \) is a dimensionless relaxation parameter, \( \tau \) being the typical relaxation time due to “particle” collisions and \( \mathbf{c}_i \) is the particle velocity in the direction of the \( i \)-th lattice link. Such collisions drive the distribution to the following local equilibrium

\[
q_{i,\mu}^eq(\mathbf{x}, t) = w_i q_{i,\mu}(1 + u_i)
\]

where \( q_{i,\mu} = \sum_q q_{i,\mu} \) is the quaternion “density”, \( w_i \) are standard lattice weights normalized to unity and \( u_i = \mathbf{u} \cdot \mathbf{c}_i/c_s^2 \), is the projection of the local flow velocity upon the \( i \)-th lattice link, \( c_s = c/\sqrt{3} \) being the lattice speed of sound given by \( c_s^2 = \sum_i w_i c_i^2 \). It should be noted that

\[
V(\mathbf{x}_i, \mathbf{x}_j) = W(\rho) e^{-\frac{1}{2} \sum_{k=1}^{4} \frac{(\mathbf{r}_{ij,k} - R_{HB})^2}{\sigma_R^2}} \sum_{k=1}^{4} \frac{c_{kl}^{HB} e^{-\frac{(\mathbf{r}_{ij,k} - 1)^2 + (\mathbf{r}_{ij,k} + 1)^2}{2 \sigma_R^2}}}{\sigma_R^2}.
\]
where \( p \) (\( 0 < p < 1 \)) is a parameter fixing the fraction of the distance traveled in a single sub-step. In our simulations, \( p = 1/3 \) proved adequate to avoid numerical instabilities.

The LB method as a kinetic over-relaxation scheme. —

It is well known that relaxation methods can be cast in the form of a diffusive process in fictitious time. Numerical stability of the discretized version imposes the following diffusive Courant-Friedrichs-Lewy (CFL) constraint [20] on the relaxation parameter \( C_d \)

\[
C_d \equiv \frac{D\Delta t}{\Delta x^2} < 1/d
\]

in \( d \) spatial dimensions. The LB scheme is based on a streaming-relaxation dynamics in the extended kinetic space, which features some advantageous properties from the computational point of view. First, since the LB dynamics is hyperbolic (first order in both space and time), the stability constraint comes in advection rather than diffusion form, namely \( C_a \equiv c\Delta t / \Delta x \leq 1 \). In fact, LB operates strictly at \( C_a = 1 \), which means that the timestep scales linearly with the space resolution \( \Delta x \), instead of quadratically as in a diffusion process [16]. This is a significant advantage as the size of the system is increased. The stability condition for the relaxation step reads as \( \Delta t < 2\tau \). This is independent of the lattice spacing, because relaxation is fully local. The above condition is a statement of positivity of the LB mean free path \( \lambda = c_s(\tau - \Delta t/2) \). Note that while the LB dynamics recovers diffusive behavior in the limit of low Knudsen number [16], \( \lambda \Delta t \Delta x / q \ll 1 \) (\( q \) is a generic quantity, say any component of the quaternion), it is by no means restricted to this regime. In the opposite limit, \( Kn \to \infty \), the LB dynamics describes the free-ballistic regime, which is still a perfectly meaningful dynamics for the purpose of minimization. Remarkably, it is precisely in this quasi-ballistic regime, \( Kn \gg 1 \), that the LB dynamics provides its best as a dynamic many-body minimizer. This stands in sharp contrast with the standard use of LB as a hydrodynamic solver in kinetic disguise. As detailed shortly, in the case of LTIP3P, LB shows optimal performance around \( \tau / \Delta t \sim 20 \), which, in lattice units, corresponds to \( C_d \sim 20 \), an order of magnitude above the CFL limit!

Numerical Results. — We have run several simulations based on both LTIP3P and LBN3d, with different so-called “fractional time stepping” [18, 19]. The Eq. (6) then modifies into

\[
q_{i,\mu}(x + c_i \Delta t, t + \Delta t) = p[(1 - \omega)q_{i,\mu}(x, t) + \omega q_{i,\mu}^eq(x, t) + \dot{q}_{i,\mu}(x, t)\Delta t] + (1 - p)[(1 - \omega)q_{i,\mu}(x + c_i \Delta t, t) + \omega q_{i,\mu}(x + c_i \Delta t, t)\Delta t],
\]

where \( p \) (\( 0 < p < 1 \)) is a parameter fixing the fraction of the distance traveled in a single sub-step. In our simulations, \( p = 1/3 \) proved adequate to avoid numerical instabilities.

LTIP3P simulations. — In fig. 2 (upper row) we have shown the total potential energy of the system in units of KJ/mol vs the LB steps in panel (a) and their corresponding HBs in panel (b), with \( N = 10^3 \) and \( \gamma = 50\Delta t^{-1} \). The figures in the lower row report the same information as in the upper row, for the case of LBN3d with \( N = 6^3 \) and \( \gamma = 10^2\Delta t^{-1} \).

Parameters. For the sake of validation, the same minimization problem has also been solved by an annealed Metropolis MC method.
The figure shows the energy vs MC sweeps (in blue) and LB steps (in red) for the LTIP3P (a) and the LBN3d (b) potentials. In the case of LTIP3P, the main LB parameters are \( N = 10^3 \), \( \gamma = 50\Delta t^{-1} \) and \( \tau / \Delta t = 20 \) and in the case of LBN3d \( N = 6^3 \), \( \gamma = 10^3 \Delta t^{-1} \) and \( \tau / \Delta t = 1000 \). The insets show the very early LB steps/MC sweeps.

**Comparison with simulated annealing Monte Carlo.**

In order to validate the model, we have compared our LB results with simulated annealing MC. In each MC sweep we randomly rotate the molecules, and if the total potential energy of the system \( E \) is lower than the previous distribution, the move is accepted. If not, the move is accepted with probability \( \propto e^{-\beta E} \), where \( \beta \) is proportional to the inverse of the temperature \( T^{-1} \). In the annealed version, \( \beta \) is progressively increased during the MC simulation. In the present work, the MC scheme is characterized by three types of moves: I. global rotations; II. local rotation; III. global refinement. In I each molecule rotates about a random axis by a random angle, chosen in the range \([0^\circ, 360^\circ]\). In II, we randomly pick one water molecule and rotate it about a random axis by a random angle in the range \([0^\circ, 10^\circ]\). III is similar to I, but the random angles are chosen in a narrower range \([0^\circ, 10^\circ]\). In fig. 3 we compare both cases, the LTIP3P and the LBN3d, with the simulated annealing MC discussed above. As it can be seen from the figure, in both cases LB attains the same minimum as in MC, which provides a validation of the LB model. Although our MC scheme can certainly be improved, the computational performance of the LB minimizer still appears to be pretty remarkable.

**Time to minimum versus system size.**

Two important quality factors of dynamic minimizers are their robustness towards changes in the initial conditions and the scaling of the time to solution with the size of the problem. The LB model is a deterministic minimizer, hence exposed to a sensitivity to the initial conditions for the minimization of highly corrugated potentials. However, this sensitivity can be strongly mitigate by a proper choice of LB parameters. On the other hand, owing to its strong locality, it features a remarkable linear scaling with the system size. As a result, the power of the LB minimizer is probably best displayed for large systems, a statement which is only accrued by the excellent amenability of LB scheme to parallel computing. As it can be seen from fig. 3 (a), the LB model minimizes water energy based on the LTIP3P at different system sizes within nearly the same number of LB steps. Since the computational cost of the LB time-step scales linearly with the system size, we conclude that CPU time-to-minimum is also linear with the system size. This stands in notable contrast with annealed MC, where the number of sweeps to reach the global minimum grows rapidly with the system size. Since each MC sweep takes about the same CPU time as a LB step, the end result is that LB becomes increasingly advantageous at larger system sizes. It should be further noted that even when LB does not get to the global minimum, due to a poor choice of the initial condition, it often attains a close-by local minimum. As a result, one may envisage a synergistic coupling between LB as a fast quasi-global minimizer, to be combined with a subsequent MC mini-
mization taking the system to its global minimum.

Summary.— Summarizing, from the sheer viewpoint of dynamic minimization, the LB dynamics can be seen as a real-time, extreme over-relaxation dynamics in kinetic space. While standard use of the LB equation is in the hydrodynamic regime, where it provides a macroscopic solver in “kinetic disguise”, for the purpose of dynamic minimization, it is here found and documented, that LB provides its best when operated in the non-hydrodynamic regime. This unexpected and welcome result might open up a new field of applications for “kinetic over-relaxation”, as a complement/alternative to Monte Carlo and Langevin-Molecular-Dynamics to a class of lattice potentials.

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