

Ab initio Molecular Dynamics

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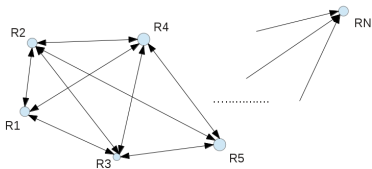
August 25, 2015

Overview

- Equations of motion
- Integrators, Sources of error
- General Lagrangian for AIMD
- Car–Parrinello MD
- Born–Oppenheimer MD
- *ab initio* Langevin MD
- Stability and efficiency

Equations of Motion (EOM)

Newton's EOM for a set of classical point particles in a potential.



$$M_I \ddot{\mathbf{R}}_I = - \frac{dV(\mathbf{R})}{d\mathbf{R}_I}$$

These EOM generate for a given number of particles N in a volume V the micro canonical ensemble (NVE ensemble).

The total energy E is a constant of motion!

Total Energy

Total Energy = Kinetic energy + Potential energy

$$\text{Kinetic energy} = T(\dot{\mathbf{R}}) = \sum_{I=1}^N \frac{M_I}{2} \dot{\mathbf{R}}^2$$

$$\text{Potential energy} = V(\mathbf{R})$$

We will use the total energy as an indicator for the numerical accuracy of simulations.

Lagrange Equation

$$\frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{\mathbf{R}}} = \frac{\partial \mathcal{L}}{\partial \mathbf{R}}$$

with

$$\mathcal{L}(\mathbf{R}, \dot{\mathbf{R}}) = T(\dot{\mathbf{R}}) - V(\mathbf{R})$$

Equivalent to Newton's EOM in Cartesian coordinates, but is more general and flexible.

- Extended systems
- Constraints

Integration of EOM

Discretization of time

$$\begin{aligned} \mathbf{R}(t) &\rightarrow \mathbf{R}(t + \tau) \rightarrow \mathbf{R}(t + 2\tau) \rightarrow \dots \rightarrow \mathbf{R}(t + m\tau) \\ \mathbf{V}(t) &\rightarrow \mathbf{V}(t + \tau) \rightarrow \mathbf{V}(t + 2\tau) \rightarrow \dots \rightarrow \mathbf{V}(t + m\tau) \end{aligned}$$

- **Efficiency:** minimal number of force evaluations, minimal number of stored quantities
- **Stability:** minimal drift in constant of motion (energy)
- **Accuracy:** minimal distance to exact trajectory

Sources of Errors

- **Type of integrator**
predictor-corrector, time-reversible, symplectic
- **Time step τ**
short time accuracy measured as $\mathcal{O}(\tau^n)$
- **Consistency of forces and energy**
e.g. cutoffs leading to non-smooth energy surfaces
- **Accuracy of forces**
e.g. convergence of iterative force calculations (SCF, constraints)

Velocity Verlet Integrator

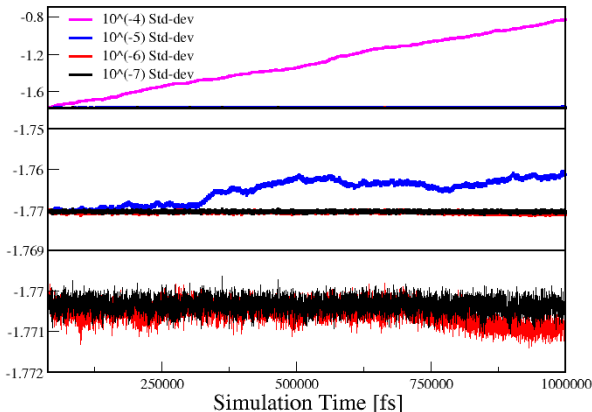
$$\mathbf{R}(t + \tau) = \mathbf{R}(t) + \tau \mathbf{V}(t) + \frac{\tau^2}{2M} \mathbf{f}(t)$$

$$\mathbf{V}(t + \tau) = \mathbf{V}(t) + \frac{\tau}{2M} [\mathbf{f}(t) + \mathbf{f}(t + \tau)]$$

- Efficiency: 1 force evaluation, 3 storage vectors
- Stability: time reversible
- Accuracy: $\mathcal{O}(\tau^2)$
- Simple adaptation for constraints (shake, rattle, roll)
- Simple adaptation for multiple time steps and thermostats

Test on Required Accuracy of Forces

Classical Force Field Calculations, 64 molecules, 330 K
TIP3P (flexible), SPME ($\alpha = 0.44$, GMAX = 25),



Stability: Accuracy of Forces

Stdev. Δf Hartree/Bohr	Stdev. Energy μ Hartree	Drift μ Hartree/ns	Drift Kelvin/ns
—	170.35	35.9	0.06
10^{-10}	179.55	-85.7	-0.14
10^{-08}	173.68	6.5	0.01
10^{-07}	177.83	-58.2	-0.10
10^{-06}	—	-385.4	-0.63
10^{-05}	—	9255.8	15.21
10^{-04}	—	972810.0	1599.31

Born–Oppenheimer MD: The Easy Way

$$M_I \ddot{\mathbf{R}}_I = - \frac{dV(\mathbf{R})}{d\mathbf{R}_I} \quad \text{EOM}$$

$$V(\mathbf{R}) = \min_{\Phi} [E_{\text{KS}}(\{\Phi(\mathbf{r})\}; \mathbf{R}) + \text{const.}] \quad \text{Kohn–Sham BO potential}$$

Forces

$$\begin{aligned} f_{\text{KS}}(\mathbf{R}) &= \frac{d \min_{\Phi} E_{\text{KS}}(\{\Phi(\mathbf{r})\}; \mathbf{R})}{d\mathbf{R}} \\ &= \frac{\partial E_{\text{KS}}}{\partial \mathbf{R}} + \frac{\partial \text{const.}}{\partial \mathbf{R}} + \sum_i \underbrace{\frac{\partial (E_{\text{KS}} + \text{const.})}{\partial \Phi_i}}_{=0} \frac{\partial \Phi_i}{\partial \mathbf{R}} \end{aligned}$$

Computational Details

- System

- 64 water molecules
- density 1 gcm^{-3}
- Temperature $\approx 330 \text{ K}$
- Timestep 0.5 fs

- DFT Calculations

- GPW, TZV2P basis (2560 bsf), PBE functional
- Cutoff 280 Rydberg, $\epsilon_{\text{default}} = 10^{-12}$
- OT-DIIS, Preconditioner FULL_SINGLE_INVERSE
- Reference trajectory (1 ps), $\epsilon_{\text{SCF}} = 10^{-10}$

Stability in BOMD

Unbiased initial guess; $\Phi(t) = \Phi_0(\mathbf{R}(t))$

ϵ_{SCF}	MAE E_{KS} Hartree	MAE f Hartree/Bohr	Drift Kelvin/ns
10^{-08}	$1.2 \cdot 10^{-11}$	$5.1 \cdot 10^{-09}$	0.0
10^{-07}	$9.5 \cdot 10^{-10}$	$5.6 \cdot 10^{-08}$	0.1
10^{-06}	$6.9 \cdot 10^{-08}$	$4.8 \cdot 10^{-07}$	0.4
10^{-05}	$7.4 \cdot 10^{-06}$	$5.6 \cdot 10^{-06}$	2.3
10^{-04}	$3.3 \cdot 10^{-04}$	$5.9 \cdot 10^{-05}$	50

Consistent with results from classical MD
Note accuracy of forces!

Efficiency: Initial Guess of Wavefunction

4th order Gear predictor (PS extrapolation in CP2K)

Method	ϵ_{SCF}	Iterations	Drift (Kelvin/ns)
Guess	10^{-06}	14.38	0.4
Gear(4)	10^{-07}	6.47	5.7
Gear(4)	10^{-06}	5.22	11.8
Gear(4)	10^{-05}	4.60	86.8

What is the problem?

Time reversibility has been broken!

Generalized Lagrangian

$$\mathcal{L}(\mathbf{q}, \dot{\mathbf{q}}, \mathbf{x}, \dot{\mathbf{x}}) = \frac{1}{2}M\dot{\mathbf{q}}^2 + \frac{1}{2}\mu\dot{\mathbf{x}}^2 - E(\mathbf{q}, \mathbf{y}) + k\mu G(\|\mathbf{x} - \mathbf{y}\|)$$

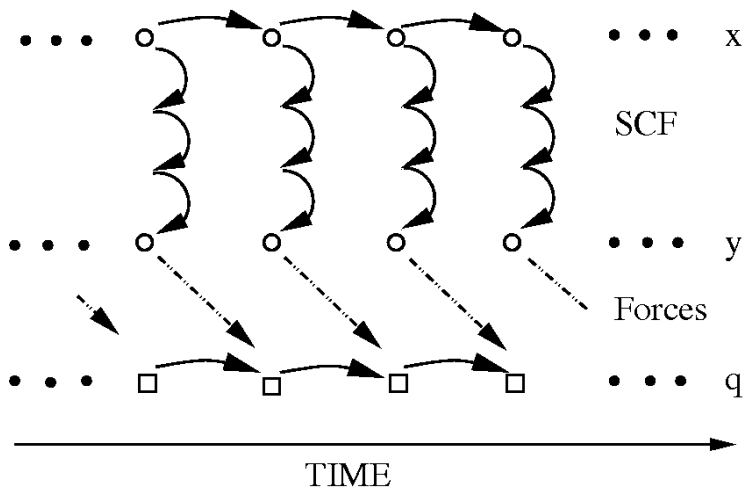
$\mathbf{y} = F(\mathbf{q}, \mathbf{x})$ wavefunction optimization

$G(\|\mathbf{x} - \mathbf{y}\|)$ wavefunction retention potential

Equations of motion

$$M\ddot{\mathbf{q}} = -\frac{\partial E}{\partial \mathbf{q}} - \frac{\partial E}{\partial \mathbf{y}} \frac{\partial F}{\partial \mathbf{q}} + k\mu \frac{\partial G}{\partial \mathbf{y}} \frac{\partial F}{\partial \mathbf{q}}$$
$$\mu\ddot{\mathbf{x}} = -\frac{\partial E}{\partial \mathbf{y}} \frac{\partial F}{\partial \mathbf{x}} + k\mu \left[\frac{\partial G}{\partial \mathbf{x}} + \frac{\partial G}{\partial \mathbf{y}} \frac{\partial F}{\partial \mathbf{x}} \right]$$

Dynamical System



Car–Parrinello Molecular Dynamics

$$\mathbf{y} = \mathbf{x} \quad \Rightarrow \quad G(\|\mathbf{x} - \mathbf{y}\|) = 0$$

Lagrangian

$$\mathcal{L}(\mathbf{q}, \dot{\mathbf{q}}, \mathbf{x}, \dot{\mathbf{x}}) = \frac{1}{2}M\dot{\mathbf{q}}^2 + \frac{1}{2}\mu\dot{\mathbf{x}}^2 - E(\mathbf{q}, \mathbf{x})$$

Equations of motion

$$M\ddot{\mathbf{q}} = -\frac{\partial E}{\partial \mathbf{q}}$$
$$\mu\ddot{\mathbf{x}} = -\frac{\partial E}{\partial \mathbf{x}}$$

Properties of CPMD

- **Accuracy:** Medium
Distance from BO surface controlled by mass μ
Requires renormalization of dynamic quantities (e.g. vibrational spectra)
- **Stability:** Excellent
All forces can be calculated to machine precision easily
- **Efficiency:** Good
Efficiency is strongly system dependent (electronic gap)
Requires many nuclear gradient calculations

Not implemented in CP2K!

BOMD

$$\mathbf{y} = \text{Min}_{\mathbf{x}} E(\mathbf{q}, \mathbf{x}) \quad \text{and} \quad \mu = 0$$

Lagrangian

$$\mathcal{L}(\mathbf{q}, \dot{\mathbf{q}}) = \frac{1}{2} M \dot{\mathbf{q}}^2 + E(\mathbf{q}, \mathbf{y})$$

$$\mathcal{L}(\mathbf{x}, \dot{\mathbf{x}}) = \frac{1}{2} \dot{\mathbf{x}}^2 + kG(\|\mathbf{x} - \mathbf{y}\|)$$

Equations of motion

$$\left. \begin{aligned} M\ddot{\mathbf{q}} &= -\frac{\partial E}{\partial \mathbf{q}} \\ \ddot{\mathbf{x}} &= -k\frac{\partial G}{\partial \mathbf{x}} \end{aligned} \right\} \text{decoupled equations}$$

BOMD with Incomplete SCF Convergence

$$\mathbf{y} = F(\mathbf{q}, \mathbf{x}) \approx \text{Min}_{\mathbf{x}} E(\mathbf{q}, \mathbf{x}) \quad \text{and} \quad \mu = 0$$

Equations of motion

$$M\ddot{\mathbf{q}} = -\frac{\partial E}{\partial \mathbf{q}} - \frac{\partial E}{\partial \mathbf{y}} \frac{\partial F}{\partial \mathbf{q}}$$
$$\ddot{\mathbf{x}} = -k \frac{\partial G}{\partial \mathbf{x}} - \frac{\partial G}{\partial \mathbf{y}} \frac{\partial F}{\partial \mathbf{x}}$$

SCF Error: Neglect of force terms $\frac{\partial E}{\partial \mathbf{y}} \frac{\partial F}{\partial \mathbf{q}}$ and $\frac{\partial G}{\partial \mathbf{y}} \frac{\partial F}{\partial \mathbf{x}}$.

EOM are coupled through terms neglected!

ASPC Integrator

Integration of electronic DOF (\mathbf{x}) has to be

- **accurate**: good wavefunction guess gives improved efficiency
- **stable**: do not destroy time-reversibility of nuclear trajectory

ASPC: Always Stable Predictor Corrector

- J. Kolafa, J. Comput Chem. **25**: 335–342 (2004)
- ASPC(k): time-reversible to order $2k + 1$

Orthogonality Constraint

Wavefunction extrapolation for non-orthogonal basis sets

$$\mathbf{C}_{\text{init}} = \sum_{j=0}^K B_{j+1} \underbrace{\mathbf{C}_{t-j\tau} \mathbf{C}_{t-j\tau}^T \mathbf{S}_{t-j\tau}}_{\text{PS}} \mathbf{C}_{t-\tau}$$

Coefficients B_j are given by ASPC algorithm

$$\begin{aligned} \mathbf{x} &= \mathbf{C}_{\text{init}} \\ \mathbf{y}[t] &= \mathbf{C}_t \\ \mathbf{S} & \text{ overlap matrix} \end{aligned}$$

J. VandeVondele et al. *Comp. Phys. Comm.* **167**: 103–128 (2005)

Importance of Time-Reversibility

Method	ϵ_{SCF}	Iterations	Drift (Kelvin/ns)
Guess	10^{-06}	14.38	0.4
ASPC(3)	10^{-06}	5.01	0.2
ASPC(3)	10^{-05}	3.02	4.5
Gear(4)	10^{-07}	6.47	5.7
Gear(4)	10^{-06}	5.22	11.8
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Efficiency and Drift

Method	ϵ_{SCF}	Iterations	Drift (Kelvin/ns)
Guess	10^{-06}	14.38	0.4
ASPC(4)	10^{-06}	5.01	0.2
ASPC(4)	10^{-05}	3.02	4.5
ASPC(4)	10^{-04}	1.62	1742.4
ASPC(4)	10^{-02}	1.03	21733.2

Efficiency and Drift

Method	ϵ_{SCF}	Iterations	Drift (Kelvin/ns)
ASPC(4)	10^{-04}	1.62	1742.4
ASPC(5)	10^{-04}	1.63	1094.0
ASPC(6)	10^{-04}	1.79	397.4
ASPC(7)	10^{-04}	1.97	445.8
ASPC(8)	10^{-04}	2.06	24.1

Langevin BOMD

Starting point: BOMD with ASPC(k) extrapolation

Analysis of forces

$$f_{\text{BO}}(\mathbf{R}) = \underbrace{f_{\text{HF}}(\mathbf{R}) + f_{\text{Pulay}}(\mathbf{R})}_{f(\mathbf{R})} + f_{\text{nsc}}(\mathbf{R})$$

- f_{BO} : correct BO force
- f_{HF} : Hellmann–Feynman force
- f_{Pulay} : Pulay force
- f_{nsc} : non-self consistency error force
- f : approximate BO force

Forces in Approximate BOMD

Approximate f_{nsc} by

$$\tilde{f}_{\text{nsc}} = - \int d\mathbf{r} \left[\left(\frac{\partial V_{\text{xc}}(\rho^i)}{\partial \rho^i} \right) \Delta\rho + V_{\text{H}}(\Delta\rho) \right] \nabla_I \rho^i$$

with $\Delta\rho = \rho^o - \rho^i$, ρ^o final (output) density, ρ^i initial (predicted) density.

Now assume

$$f(\mathbf{R}) + \tilde{f}_{\text{nsc}}(\mathbf{R}) = f_{\text{BO}}(\mathbf{R}) - \gamma_D \dot{\mathbf{R}}$$

where γ_D is a constant friction parameter.

Langevin EOM

$$M\ddot{\mathbf{R}} = f_{\text{BO}}(\mathbf{R}) - (\gamma_D + \gamma_L)\dot{\mathbf{R}} + \Theta$$

with Θ a Gaussian random noise term and

$$\langle \Theta(0)\Theta(t) \rangle = 6(\gamma_D + \gamma_L)Mk_B T \delta(t)$$

Given temperature $T = \frac{k_B}{3} \langle M\dot{\mathbf{R}}^2 \rangle$ and an arbitrary γ_L this determines γ_D

Langevin BOMD

T. D. Kühne et al., Phys. Rev. Lett. **98** 066441 (2007)

- single SCF step plus force correction needed
extremely efficient for systems with slow SCF convergence
- γ_D is small: correct statistics and dynamics
- difficult to stabilize in complex systems



Car–Parrinello molecular dynamics

Jürg Hutter*

The Car–Parrinello (CP) method made molecular dynamics simulation with on-the-fly computation of interaction potentials from electronic structure theory computationally feasible. The method reformulates *ab initio* molecular dynamics (AIMD) as a two-component classical dynamical system. This approach proved to be valuable far beyond the original CP molecular dynamics method. The modern formulation of Born–Oppenheimer (BO) dynamics is based on the same basic principles and can be derived from the same Lagrange function as the CP method. These time-reversible BO molecular dynamics methods allow higher accuracy and efficiency while providing similar longtime stability as the CP method. AIMD is used in many fields of computational physics and chemistry. Its applications are instrumental in fields as diverse as enzymatic catalysis and the study of the interior of planets. With its versatility and predictive power, AIMD has become a major approach in atomistic simulations. © 2011 John Wiley & Sons, Ltd.

How to cite this article:

WIREs Comput Mol Sci 2012, 2: 604–612 doi: 10.1002/wcms.90

Summary: BOMD in CP2K

- Born–Oppenheimer MD with ASPC(3) is default in CP2K
- SCF convergence criteria depends on system
 10^{-5} – 10^{-6} is a reasonable starting guess
- Best used together with OT and the
FULL_SINGLE_INVERSE preconditioner
- Langevin dynamics is an option
needs some special care, mostly for uniform systems