# *Integration Algorithm*



2024 Winter Seminar January  $12<sup>th</sup>$ , 2024

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# *Content*

*Introduction*

*Stability*

*Efficiency*

*Accuracy*



But there are much more things that we should understand here!

# *The list of integrator implemented in OpenMM*

#### In OpenMM,



#### In OpenMMtools (https://openmmtools.readthedocs.io/en/stable/ integrators.html),







#### *Basics of MD simulation*

In classical mechanics, the equation of motion is integrated to generate the trajectory.

$$
\frac{\mathrm{d}}{\mathrm{d}t}\left(\frac{\partial \mathcal{L}}{\partial \dot{\mathbf{r}}_i}\right) - \frac{\partial \mathcal{L}}{\partial \mathbf{r}_i} = 0, \qquad \dot{q}_\alpha = \frac{\partial \mathcal{H}}{\partial p_\alpha}, \ \dot{p}_\alpha = -\frac{\partial \mathcal{H}}{\partial q_\alpha}
$$

Energy is conserved in classical mechanics!

By solving F=ma, we sample the **microcanonical**ensemble (constant E) for ensemble averages.

$$
\langle a \rangle = \frac{\int dx \ a(x) \delta(\mathcal{H}(x) - E)}{\int dx \ \delta(\mathcal{H}(x) - E)} = \lim_{T \to \infty} \frac{1}{T} \int_0^T dt \ a(x_t) \equiv \bar{a}.
$$

We need to sample the x\_t in microcanonical ensemble, which we call it **trajectory**.(.dcd)

In here, we introduce the time discretization parameter dt, known as the **time step.** Starting with the initial cond x\_0, x\_dt, x\_2dt, x\_3dt are generated by applying the integrator iteratively.

$$
A = \langle a \rangle = \frac{1}{M} \sum_{n=1}^{M} a(\mathbf{x}_{n \Delta t}) \equiv \bar{a}.
$$

A discretization of the equations of motion can be obtained by Taylor expansion:

$$
\mathbf{r}_i(t + \Delta t) = \mathbf{r}_i(t) + \Delta t \mathbf{v}_i(t) + \frac{\Delta t^2}{2m_i} \mathbf{f}_i(t) + \frac{\Delta t^3}{3!} \ddot{\mathbf{r}}_i(t) + \mathcal{O}(\Delta t^4)
$$
  

$$
\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \frac{\Delta t}{m_i} \mathbf{f}_i(t) + \frac{\Delta t^2}{2} \ddot{\mathbf{v}}_i(t) + \frac{\Delta t^3}{3!} \ddot{\mathbf{v}}_i(t) + \mathcal{O}(\Delta t^4).
$$

By the **Euler algorithm**, the trajectory is calculated according to:

$$
\mathbf{r}_i(t + \Delta t) = \mathbf{r}_i(t) + \Delta t \mathbf{v}_i(t) + \frac{\Delta t^2}{2m_i} \mathbf{f}_i(t) + \mathcal{O}(\Delta t^3)
$$

$$
\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \frac{\Delta t}{m_i} \mathbf{f}_i(t) + \mathcal{O}(\Delta t^2)
$$

WE DO NOT USE THIS, since this is not stable (neither time-reversible nor **symplectic**)

A discretization of the equations of motion can be obtained by Taylor expansion:

$$
\mathbf{r}_i(t + \Delta t) = \mathbf{r}_i(t) + \Delta t \mathbf{v}_i(t) + \frac{\Delta t^2}{2m_i} \mathbf{f}_i(t) + \frac{\Delta t^3}{3!} \ddot{\mathbf{r}}_i(t) + \mathcal{O}(\Delta t^4)
$$
  

$$
\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \frac{\Delta t}{m_i} \mathbf{f}_i(t) + \frac{\Delta t^2}{2} \ddot{\mathbf{v}}_i(t) + \frac{\Delta t^3}{3!} \ddot{\mathbf{v}}_i(t) + \mathcal{O}(\Delta t^4).
$$

By the **Verlet algorithm** (1967), the trajectory is calculated according to:

$$
\mathbf{r}_i(t - \Delta t) = \mathbf{r}_i(t) - \Delta t \mathbf{v}_i(t) + \frac{\Delta t^2}{2m_i} \mathbf{f}_i(t) - \frac{\Delta t^3}{3!} \ddot{\mathbf{r}}_i(t) + \mathcal{O}(\Delta t^4).
$$

The updating equation for the positions and velocities are:

$$
\mathbf{r}_i(t + \Delta t) = 2\mathbf{r}_i(t) - \mathbf{r}_i(t - \Delta t) + \frac{\Delta t^2}{m_i} \mathbf{f}_i(t) + \mathcal{O}(\Delta t^4),
$$

$$
\mathbf{v}_i(t) = \frac{\mathbf{r}_i(t + \Delta t) - \mathbf{r}_i(t - \Delta t)}{2\Delta t} + \mathcal{O}(\Delta t^3).
$$

A discretization of the equations of motion can be obtained by Taylor expansion:

$$
\mathbf{r}_i(t + \Delta t) = \mathbf{r}_i(t) + \Delta t \mathbf{v}_i(t) + \frac{\Delta t^2}{2m_i} \mathbf{f}_i(t) + \frac{\Delta t^3}{3!} \ddot{\mathbf{r}}_i(t) + \mathcal{O}(\Delta t^4)
$$
  

$$
\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \frac{\Delta t}{m_i} \mathbf{f}_i(t) + \frac{\Delta t^2}{2} \ddot{\mathbf{v}}_i(t) + \frac{\Delta t^3}{3!} \ddot{\mathbf{v}}_i(t) + \mathcal{O}(\Delta t^4).
$$

By the **Leap-frog algorithm**, the trajectory is calculated according to:

$$
\mathbf{v}_i(t + \frac{\Delta t}{2}) = \mathbf{v}_i(t - \frac{\Delta t}{2}) + \frac{\Delta t}{m_i} \mathbf{f}_i(t),
$$

$$
\mathbf{r}_i(t + \Delta t) = \mathbf{r}_i(t) + \Delta t \mathbf{v}_i(t + \frac{\Delta t}{2}).
$$

The velocities are updated at half time steps and 'leap' ahead the positions as:

$$
\mathbf{v}_i(t) = \frac{\mathbf{v}_i(t - \frac{\Delta t}{2}) + \mathbf{v}_i(\Delta t + \frac{\Delta t}{2})}{2}
$$

A discretization of the equations of motion can be obtained by Taylor expansion:

$$
\mathbf{r}_i(t + \Delta t) = \mathbf{r}_i(t) + \Delta t \mathbf{v}_i(t) + \frac{\Delta t^2}{2m_i} \mathbf{f}_i(t) + \frac{\Delta t^3}{3!} \ddot{\mathbf{r}}_i(t) + \mathcal{O}(\Delta t^4)
$$
  

$$
\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \frac{\Delta t}{m_i} \mathbf{f}_i(t) + \frac{\Delta t^2}{2} \ddot{\mathbf{v}}_i(t) + \frac{\Delta t^3}{3!} \ddot{\mathbf{v}}_i(t) + \mathcal{O}(\Delta t^4).
$$

By the **Velocity-Verletalgorithm** (1982), the trajectory is calculated according to:

$$
\mathbf{r}_i(t + \Delta t) = \mathbf{r}_i(t) + \Delta t \mathbf{v}_i(t) + \frac{\Delta t^2}{m_i} \mathbf{f}_i(t) + \mathcal{O}(\Delta t^3),
$$
  

$$
\mathbf{v}_i(t + \Delta t) = \mathbf{v}(t) + \frac{\Delta t}{2m_i} (\mathbf{f}_i(t) + \mathbf{f}_i(t + \Delta t)) + \mathcal{O}(\Delta t^3).
$$

This is very stable and has become the perhaps most widely used integration algorithm. The Velocity-Verlet scheme is a **symplectic** integrator.

# *What properties should integration algorithms satisfy?*

# **1. Stability**

- **Energy conservation**
- Symplecticity
- Time reversibility

# **2. Efficiency**

- Maximum permissible time step
- Constraint algorithm
- Hydrogen mass repartitioning
- **Multi-step integration**

# **3. Accuracy**

- Configurational sampling
- **Dynamical properties**

# *What properties should integration algorithms satisfy?*

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It would be nice if we can derive the Verlet integration algorithm directly from the classical mechanics such as Hamilton's equation. Then '**symplecticity**' (the phase-space volume preserving property) is guaranteed.

For the time evolution of any function  $a(x)$  of the phase space vector,

 $\sim$   $\sim$ 

$$
\frac{\mathrm{d}a}{\mathrm{d}t} = \sum_{\alpha=1}^{3N} \left[ \frac{\partial a}{\partial q_{\alpha}} \dot{q}_{\alpha} + \frac{\partial a}{\partial p_{\alpha}} \dot{p}_{\alpha} \right].
$$

$$
\dot{q}_{\alpha} = \frac{\partial \mathcal{H}}{\partial p_{\alpha}}, \qquad \qquad \dot{p}_{\alpha} = -\frac{\partial \mathcal{H}}{\partial q_{\alpha}}
$$

$$
\frac{da}{dt} = \sum_{\alpha=1}^{3N} \left[ \frac{\partial a}{\partial q_{\alpha}} \frac{\partial \mathcal{H}}{\partial p_{\alpha}} - \frac{\partial a}{\partial p_{\alpha}} \frac{\partial \mathcal{H}}{\partial q_{\alpha}} \right]
$$

$$
= \{a, \mathcal{H}\}.
$$

Let's define the Liouville operator L as:  $iLa = \{a, \mathcal{H}\}\$ 

$$
iL = \sum_{\alpha=1}^{3N} \left[ \frac{\partial \mathcal{H}}{\partial p_{\alpha}} \frac{\partial}{\partial q_{\alpha}} - \frac{\partial \mathcal{H}}{\partial q_{\alpha}} \frac{\partial}{\partial p_{\alpha}} \right] \qquad a(\mathbf{x}_{t}) = e^{iLt} a(\mathbf{x}_{0}).
$$

$$
\mathbf{x}_{t} = e^{iLt} \mathbf{x}_{0}.
$$

We call the operator exp(iLt) as the classical propagator.

From now on, let's split the Liouville operator into two:

$$
iL = iL_1 + iL_2,
$$
  

$$
\mathcal{H} = \frac{p^2}{2m} + U(x).
$$

 $iL_1 = \sum_{\alpha=1}^{3N} \frac{\partial \mathcal{H}}{\partial p_{\alpha}} \frac{\partial}{\partial q_{\alpha}}$  $iL_1 = \frac{p}{m} \frac{\partial}{\partial x}, \qquad iL_2 = F(x) \frac{\partial}{\partial x},$  $iL_2 = -\sum^{3N} \frac{\partial \mathcal{H}}{\partial q_{\alpha}} \frac{\partial}{\partial p_{\alpha}}.$ 

Mark T. Tuckerman, *Statistical mechanics: Theory and Molecular Simulation 2thed*, **2023**

Let's define the Liouville operator L as:  $iLa = \{a, \mathcal{H}\}\$ 

$$
iL = \sum_{\alpha=1}^{3N} \left[ \frac{\partial \mathcal{H}}{\partial p_{\alpha}} \frac{\partial}{\partial q_{\alpha}} - \frac{\partial \mathcal{H}}{\partial q_{\alpha}} \frac{\partial}{\partial p_{\alpha}} \right] \qquad a(\mathbf{x}_{t}) = e^{iLt} a(\mathbf{x}_{0}).
$$

$$
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Mark T. Tuckerman, *Statistical mechanics: Theory and Molecular Simulation 2thed*, **2023**

By the Trotter theorem,  $e^{A+B} = \lim_{P \to \infty} \left[ e^{B/2P} e^{A/P} e^{B/2P} \right]^P$ 

$$
e^{iLt} \approx \left[ e^{iL_2 \Delta t/2} e^{iL_1 \Delta t} e^{iL_2 \Delta t/2} \right]^P + \mathcal{O} \left( P \Delta t^3 \right)
$$

For the Hamiltonian  $\mathcal{H} = p^2/2m + U(x)$ 

$$
\exp(iL\Delta t) \approx \exp\left(\frac{\Delta t}{2}F(x)\frac{\partial}{\partial p}\right)\exp\left(\Delta t \frac{p}{m}\frac{\partial}{\partial x}\right)\exp\left(\frac{\Delta t}{2}F(x)\frac{\partial}{\partial p}\right)
$$

$$
\begin{pmatrix} x(\Delta t) \\ p(\Delta t) \end{pmatrix} \approx \exp\left(\frac{\Delta t}{2}F(x)\frac{\partial}{\partial p}\right) \exp\left(\Delta t \frac{p}{m}\frac{\partial}{\partial x}\right) \exp\left(\frac{\Delta t}{2}F(x)\frac{\partial}{\partial p}\right) \begin{pmatrix} x \\ p \end{pmatrix}.
$$

$$
\begin{pmatrix} x(\Delta t) \\ p(\Delta t) \end{pmatrix} \approx \exp\left(\frac{\Delta t}{2}F(x)\frac{\partial}{\partial p}\right) \exp\left(\Delta t \frac{p}{m}\frac{\partial}{\partial x}\right) \exp\left(\frac{\Delta t}{2}F(x)\frac{\partial}{\partial p}\right) \begin{pmatrix} x \\ p \end{pmatrix}.
$$

$$
\exp\left(\frac{\Delta t}{2}F(x)\frac{\partial}{\partial p}\right) \begin{pmatrix} x \\ p \end{pmatrix} = \begin{pmatrix} x \\ p + \frac{\Delta t}{2}F(x) \end{pmatrix}
$$

$$
\exp\left(\Delta t \frac{p}{m} \frac{\partial}{\partial x}\right) \begin{pmatrix} x \\ p + \frac{\Delta t}{2} F(x) \end{pmatrix} = \begin{pmatrix} x + \Delta t \frac{p}{m} \\ p + \frac{\Delta t}{2} F\left(x + \Delta t \frac{p}{m}\right) \end{pmatrix}
$$

$$
\exp\left(\frac{\Delta t}{2}F(x)\frac{\partial}{\partial p}\right)\left(\frac{x+\Delta t\frac{p}{m}}{p+\frac{\Delta t}{2}F\left(x+\Delta t\frac{p}{m}\right)}\right)
$$

$$
= \left(\begin{matrix} x + \frac{\Delta t}{m} \left(p + \frac{\Delta t}{2} F(x)\right) \\ p + \frac{\Delta t}{2} F(x) + \frac{\Delta t}{2} F\left(x + \frac{\Delta t}{m} \left(p + \frac{\Delta t}{2} F(x)\right)\right)\end{matrix}\right).
$$

$$
\exp\left(\frac{\Delta t}{2}F(x)\frac{\partial}{\partial p}\right)\begin{pmatrix} x + \Delta t \frac{p}{m} \\ p + \frac{\Delta t}{2}F\left(x + \Delta t \frac{p}{m}\right) \end{pmatrix}
$$

$$
= \left(\begin{matrix} x + \frac{\Delta t}{m} \left(p + \frac{\Delta t}{2} F(x)\right) \\ p + \frac{\Delta t}{2} F(x) + \frac{\Delta t}{2} F\left(x + \frac{\Delta t}{m} \left(p + \frac{\Delta t}{2} F(x)\right)\right) \end{matrix}\right)
$$

$$
x(\Delta t) = x(0) + \Delta tv(0) + \frac{\Delta t^2}{2m}F(x(0))
$$

$$
v(\Delta t) = v(0) + \frac{\Delta t}{2m} \left[ F(x(0)) + F(x(\Delta t)) \right]
$$

This is just the update step in the Velocity Verletalgorithm!

#### **Symplecticity is now guaranteed!**

For the N-particle systems:  $\mathcal{H} = \sum_{i=1}^{N} \frac{\mathbf{p}_i^2}{2m_i} + U(\mathbf{r}_1, ..., \mathbf{r}_N),$ 

$$
iL = \sum_{i=1}^{N} \frac{\mathbf{p}_i}{m_i} \cdot \frac{\partial}{\partial \mathbf{r}_i} + \sum_{i=1}^{N} \mathbf{F}_i \cdot \frac{\partial}{\partial \mathbf{p}_i}.
$$

Again, we can split the Liouville operator into two:  $iL = iL_1 + iL_2$ ,

$$
p(\Delta t/2) = p(0) + \frac{\Delta t}{2} F(x(0))
$$
  
\n
$$
p = p + 0.5 * \Delta t * F
$$
  
\n
$$
x(\Delta t) = x(0) + \frac{\Delta t}{m} p(\Delta t/2)
$$
  
\n
$$
p = p + 0.5 * \Delta t * F
$$
  
\n
$$
x = x + \Delta t * p/m
$$
  
\nRecalculate the force  
\n
$$
p = p + 0.5 * \Delta t * F.
$$
  
\n
$$
p = p + 0.5 * \Delta t * F.
$$

### *Symplecticity*

Q. Why symplecticity is good to be guaranteed?

A. Sympletic integrator has the important property that there exists a shadow Hamiltonian that remains close to the true Hamiltonian and is exactly **conserved**by the algorithm.



Mark T. Tuckerman, *Statistical mechanics: Theory and Molecular Simulation 2thed*, **2023**

# *What properties should integration algorithms satisfy?*

# 1. Stability

#### **Energy conservation**  $\mathcal{L}^{\text{max}}$

#### **Symplecticity**  $\mathcal{L}^{\text{max}}$

**Time reversibility** P.

# **2. Efficiency**

- Maximum permissible time step
- **Constraint algorithm**
- Hydrogen mass repartitioning
- **Multi-step integration**

# 3. Accuracy

- **Configurational sampling**
- Dynamical properties

# *The maximum permissible time step*



#### TABLE I. Characteristic Oscillation Periods of Atomic Motions in MD Simulations.<sup>a</sup>

 ${}^{\text{a}}f_c$ : force constant; I: moment of inertia, or atomic mass for bond stretching; calc.: calculated from eq. (1); sim.: highest frequency significant peak in spectrum of angle respectively dihedral motion from simulation. An entry of "-" means not applicable, or not determinable.

The maximum time step in simulation is limited by the fastest motions which invariably involve **hydrogen** atoms, which is usually set to be **1 fs** for usual simulation.

Some simulation techniques are further required to increase the maximum time step, but how?

#### *The idea of constraint*

Rigid distance (or angle, proper angle, improper angle) **constraints** are used to increase the integration step size from **1 fs to 2 fs.**

The idea is to use **the method of Lagrange multipliers**.

$$
\sigma_i(\{\mathbf{r}_k\}) = |\mathbf{r}_m - \mathbf{r}_n| - d_i
$$

We apply a constraint force to atoms m and n, which produces a combined displacements δ along the constraint algorithm such that  $\sigma = 0$  at the end of the time step.

This requires **iteratively** solving a system of nonlinear equations (i.e. Newton iteration)

$$
\delta^{N+1} = \delta^N - \mathbf{J}^{-1} \boldsymbol{\sigma}^N \qquad \mathbf{J}_{ij} = \frac{\partial \sigma_i}{\partial \delta_j}
$$

There are many constraint algorithms in MD simulation package which differentiates with the way to construct the jacobian matrix J and the way to perform the iteration.

List of algorithms: SHAKE, M-SHAKE, SHAPE, RATTLE, SETTLE, LINCS, P-LINCS, CCMA....

# *Constraint algorithm implemented in MD simulation package*

In OpenMM,

**SETTLE** for water molecules only.

**SHAKE** for isolated clusters of one heavy atom with up to three hydrogens bonded to it.

**CCMA**for anything not handled by one of the above algorithms.

(LINCS is not implemented in OpenMM)

$$
\delta^{N+1} = \delta^N - \mathbf{J}^{-1} \boldsymbol{\sigma}^N \qquad \mathbf{J}_{ij} = \frac{\partial \sigma_i}{\partial \delta_j}
$$

### **Remark.**

**M-SHAKE** constructs the jacobian matrix and then invert it, which is quite stable but high-cost.

LINCS circumvents the invertion of jacobian matrix by representing J<sup>^</sup>(-1) as a power series, but for strongly connected system, this series converges very slowly or even fail to converge.

**SHAKE** approximates J<sup> $\land$ </sup>(-1) using its upper triangle form for improved convergence at very little extra cost. This method is very popular but hard to be implemented efficiently on parallel architectures.

**SETTLE** uses an analytical solution for rigid water molecules rather than solving it iteratively. But only applicable to water molecules.

**CCMA** approximates J $\sqrt{-1}$  with a different matrix K $\sqrt{-1}$  that is easier to calculate on parallel architectures.

### *Convergence issues on constraint algorithm*

Cholesterol in MARTINI force field is a notorious molecule due to highly coupled constraints.

Applying **LINCS algorithm** with small LINCS Order and Iterations fails to be converged, which causes a negative energy drift and subsequently to cooling.



Such issue can be circumvented by additional incorporating the virtual sites in cholesterol.



*J. Phys. Chem. B 2021*, 125, 9537-9546. *J. Chem. Theory Comput. 2023,* 19, 1592-1601.

#### *Shot time step is sometimes inevitable*

dt=2fs is often used in simulating rigid models of water with constraint (SHAKE, SETTLE) "Fast internal vibrations are usually **decoupled** from rotational and translational motions?"



*J. Chem. Theory Comput.* **2024**, 20, 368-374 **25** 

#### *Hydrogen Mass Repartitioning (HMR)*

The idea is to increase the mass of the hydrogen atoms while decreasing the mass of the oxygen atoms to increase the integration step size from **2 fs to 4 fs.**





<sup>a</sup> /: corresponding smallest moments of inertia; resulting dynamical properties:  $\eta$ : viscosity; D: diffusion constant. Values of H<sub>2</sub>O and D<sub>2</sub>O from Lide et al.<sup>33</sup> and hydrogen-bond lifetime (T<sub>H bond</sub>) value of H<sub>2</sub>O from Montrose<sup>31</sup>; RMS drift of the total energy over 12 runs at a time step of 4 fs; maximum time step  $(\Delta t_{\text{max}})$  at a maximum order of 10 of the drift as a function of time step.

> *J. Comp. Chem., 1999*, 20, 8, 786-798. *J. Chem. Theory Comput. 2015*, 11, 1864-1874.

$$
U(\mathbf{r}_1, ..., \mathbf{r}_N) = \sum_{\text{bonds}} \frac{1}{2} K_{\text{bond}} (r - r_0)^2 + \sum_{\text{bends}} \frac{1}{2} K_{\text{bend}} (\theta - \theta_0)^2
$$
  
+ 
$$
\sum_{\text{tors}} \sum_{n=0}^{6} A_n [1 + \cos(C_n \phi + \delta_n)]
$$
  
+ 
$$
\sum_{i,j \in nb} \left\{ 4\epsilon_{ij} \left[ \left( \frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left( \frac{\sigma_{ij}}{r_{ij}} \right)^6 \right] + \frac{q_i q_j}{r_{ij}} \right\}.
$$

**Intramolecular potentials (fast force)** have large and rapidly varying components while **nonbonded potentials (slow force)** have slowly varying components due to long-range nature.

But we should choose integrator time step dt for the fast force.

Why don't we use a integrator capable of separating the time scales for a gain in computational efficiency, such as allowing the slow forces to be recomputed less frequently than the fast forces?

$$
\dot{x} = \frac{p}{m}
$$
  

$$
\dot{p} = F_{\text{fast}}(x) + F_{\text{slow}}(x).
$$

$$
\dot{x} = \frac{p}{m}
$$
  

$$
\dot{p} = F_{\text{fast}}(x) + F_{\text{slow}}(x).
$$

Liouville operator for this system is given by

$$
iL = \frac{p}{m}\frac{\partial}{\partial x} + [F_{\text{fast}}(x) + F_{\text{slow}}(x)]\frac{\partial}{\partial p}
$$

This separation (kinetic/force) leads to the standard velocity Verlet algorithm:

$$
iL = iL_1 + iL_2
$$
  
\n
$$
iL_1 = \frac{p}{m} \frac{\partial}{\partial x}
$$
  
\n
$$
iL_2 = [F_{\text{fast}}(x) + F_{\text{slow}}(x)] \frac{\partial}{\partial p}
$$

$$
\dot{x} = \frac{p}{m}
$$
  

$$
\dot{p} = F_{\text{fast}}(x) + F_{\text{slow}}(x).
$$

Liouville operator for this system is given by

$$
iL = \frac{p}{m}\frac{\partial}{\partial x} + [F_{\text{fast}}(x) + F_{\text{slow}}(x)]\frac{\partial}{\partial p}
$$

This separation (fast/slow) leads to the **reference system propagator (RESPA) algorithm**:

$$
iL = iL_{\text{fast}} + iL_{\text{slow}}
$$

$$
iL_{\text{fast}} = \frac{p}{m} \frac{\partial}{\partial x} + F_{\text{fast}}(x) \frac{\partial}{\partial p}
$$

$$
iL_{\text{slow}} = F_{\text{slow}}(x) \frac{\partial}{\partial p}.
$$

$$
\exp(iL\Delta t) = \exp\left(iL_{\text{slow}}\frac{\Delta t}{2}\right) \exp(iL_{\text{fast}}\Delta t) \exp\left(iL_{\text{slow}}\frac{\Delta t}{2}\right)
$$

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$$
\exp(iL\Delta t) = \exp\left(iL_{\text{slow}}\frac{\Delta t}{2}\right)\exp(iL_{\text{fast}}\Delta t)\exp\left(iL_{\text{slow}}\frac{\Delta t}{2}\right)
$$

In here, Δt is the time step of the slow force.

Introducing the time step of the fast force  $\delta t = \Delta t/n$ ,

$$
\exp(iL_{\text{fast}}\Delta t) = \left[\exp\left(\frac{\delta t}{2}F_{\text{fast}}\frac{\partial}{\partial p}\right)\exp\left(\delta t \frac{p}{m}\frac{\partial}{\partial x}\right)\exp\left(\frac{\delta t}{2}F_{\text{fast}}\frac{\partial}{\partial p}\right)\right]^{n}.
$$

$$
\exp(iL\Delta t) = \exp\left(\frac{\Delta t}{2}F_{\text{slow}}\frac{\partial}{\partial p}\right)
$$
  
\$\times \left[\exp\left(\frac{\delta t}{2}F\_{\text{fast}}\frac{\partial}{\partial p}\right)\exp\left(\delta t\frac{p}{m}\frac{\partial}{\partial x}\right)\exp\left(\frac{\delta t}{2}F\_{\text{fast}}\frac{\partial}{\partial p}\right)\right]^{n}\$  
\$\times \exp\left(\frac{\Delta t}{2}F\_{\text{slow}}\frac{\partial}{\partial p}\right).

$$
\exp(iL\Delta t) = \exp\left(iL_{\text{slow}}\frac{\Delta t}{2}\right)\exp(iL_{\text{fast}}\Delta t)\exp\left(iL_{\text{slow}}\frac{\Delta t}{2}\right)
$$

In here, Δt is the time step of the slow force.

Introducing the time step of the fast force  $\delta t = \Delta t/n$ ,

$$
p = p + 0.5 * \Delta t * F_{slow}
$$
  
for i = 1 to n  

$$
p = p + 0.5 * \delta t * F_{fast}
$$

$$
x = x + \delta t * p/m
$$
Recalculate fast force  

$$
p = p + 0.5 * \delta t * F_{fast}
$$
endfor  
Recalculate slow force  

$$
p = p + 0.5 * \Delta t * F_{slow}.
$$

#### *Resonance Instability*

There is a limit on the size of the time step for the slow forces by that of the fast forces

So called the **resonant time step** is set by the fast frequency w.

The higher the value of w, the smaller time step should be chosen, which limits the computational savings afforded by the algorithm independent of how slow the slow force is.



# *What properties should integration algorithms satisfy?*

# 1. Stability

**Energy conservation** 

#### **Symplecticity**

**Time reversibility** P.

# 2. Efficiency

- Maximum permissible time step
- **Constraint algorithm**
- Hydrogen mass repartitioning
- Multi-step integration

# **3. Accuracy**

- Configurational sampling
- **Dynamical properties**

*Operator splitting order is related to the configurational sampling*

$$
\begin{bmatrix}\n\dot{\mathbf{x}} \\
\dot{\mathbf{v}}\n\end{bmatrix} = \underbrace{\begin{bmatrix}\n\mathbf{v} \\
0\n\end{bmatrix}}_{R} + \underbrace{\begin{bmatrix}\n0 \\
-M^{-1}\nabla U(\mathbf{x})\n\end{bmatrix}}_{V} + \underbrace{\begin{bmatrix}\n0 \\
-\gamma \mathbf{v} + \sqrt{2\gamma} (\beta M)^{-1/2} \dot{\mathbf{W}}\n\end{bmatrix}}_{Q}
$$
\n
$$
R : e^{\mathcal{L}_{R}\tau} : \begin{bmatrix}\n\Delta \mathbf{x} \\
\Delta \mathbf{v}\n\end{bmatrix} = \begin{bmatrix}\n\mathbf{v} \\
0\n\end{bmatrix} \tau
$$
\n
$$
V : e^{\mathcal{L}_{V}\tau} : \begin{bmatrix}\n\Delta \mathbf{x} \\
\Delta \mathbf{v}\n\end{bmatrix} = \begin{bmatrix}\n0 \\
-M^{-1}\nabla U(\mathbf{x})\n\end{bmatrix} \tau
$$
\n
$$
O : e^{\mathcal{L}_{O}\tau} : \begin{bmatrix}\n\Delta \mathbf{x} \\
\Delta \mathbf{v}\n\end{bmatrix} = \begin{bmatrix}\n0 \\
(a(\tau) - 1)\mathbf{v} + \sqrt{1 - a(\tau)^{2}} (\beta M)^{-1/2} \xi\n\end{bmatrix}
$$

Once a splitting is defined, the propagator exp(Ldt) can be approximated as a Trotter factorization.

$$
e^{[A+B+C+D]\Delta t}
$$
  
=  $e^{A\Delta t/2}e^{[B+C+D]\Delta t}e^{A\Delta t/2} + O(\Delta t^3)$   
=  $e^{A\Delta t/2}e^{B\Delta t/2}e^{[C+D]\Delta t}e^{B\Delta t/2}e^{A\Delta t/2} + O(\Delta t^3)$   
=  $e^{A\Delta t/2}e^{B\Delta t/2}e^{C\Delta t/2}e^{D\Delta t}e^{C\Delta t/2}e^{B\Delta t/2}e^{A\Delta t/2} + O(\Delta t^3)$ 

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# *Operator splitting order is related to the configurational sampling*

ORVRO  
\n
$$
e^{[\mathcal{L}_o + \mathcal{L}_r + \mathcal{L}_r + \mathcal{L}_h]\Delta t}
$$
\n
$$
\simeq e^{\mathcal{L}_o \Delta t/2} e^{\mathcal{L}_r \Delta t/2} e^{\mathcal{L}_h \Delta t/2} e^{\mathcal{L}_v \Delta t} e^{\mathcal{L}_h \Delta t/2} e^{\mathcal{L}_r \Delta t/2} e^{\mathcal{L}_o \Delta t/2}
$$

RVOVR (ABOBA) 
$$
e^{[L_0 + L_v + L_r + L_h]\Delta t}
$$
  

$$
\simeq e^{L_r \Delta t/2} e^{L_h \Delta t/2} e^{L_v \Delta t/2} e^{L_o \Delta t} e^{L_v \Delta t/2} e^{L_h \Delta t/2} e^{L_r \Delta t/2}
$$

VRORV (BAOAB) 
$$
e^{[\mathcal{L}_0 + \mathcal{L}_r + \mathcal{L}_r + \mathcal{L}_h]\Delta t}
$$
  
 $\simeq e^{\mathcal{L}_v \Delta t/2} e^{\mathcal{L}_r \Delta t/2} e^{\mathcal{L}_h \Delta t/2} e^{\mathcal{L}_0 \Delta t} e^{\mathcal{L}_h \Delta t/2} e^{\mathcal{L}_r \Delta t/2} e^{\mathcal{L}_r \Delta t/2}$ 

$$
\begin{array}{lll}\n\text{ROVOR} & \qquad & e^{[\mathcal{L}_o + \mathcal{L}_v + \mathcal{L}_r + \mathcal{L}_h] \Delta t} \\
& \simeq & e^{\mathcal{L}_r \Delta t/2} e^{\mathcal{L}_o \Delta t/2} e^{\mathcal{L}_h \Delta t/2} e^{\mathcal{L}_v \Delta t} e^{\mathcal{L}_h \Delta t/2} e^{\mathcal{L}_o \Delta t/2} e^{\mathcal{L}_o \Delta t/2}.\n\end{array}
$$

VOROV  
\n
$$
e^{[\mathcal{L}_o + \mathcal{L}_r + \mathcal{L}_r + \mathcal{L}_h]\Delta t}
$$
\n
$$
\simeq e^{\mathcal{L}_v \Delta t/2} e^{\mathcal{L}_o \Delta t/2} e^{\mathcal{L}_r \Delta t/2} e^{\mathcal{L}_h \Delta t} e^{\mathcal{L}_r \Delta t/2} e^{\mathcal{L}_o \Delta t/2} e^{\mathcal{L}_v \Delta t/2}
$$

### *Operator splitting order is related to the configurational sampling*



### *Operator splitting order is related to the dynamic properties*



#### Table 1. Definition of Dynamical Properties<sup>a</sup>

#### Table 2. Comparison of Properties for Different Splittings<sup>a</sup>



### *Takeaways*

■ Liouville operator splitting is important to ensure the stability of the integration algorithm, and to provide the direct translation of algorithms.

■ Efficiency of the integration algorithm can be gained by increasing the maximum permissible time step.

■ Several strategies are the constraint algorithm, hydrogen mass repartitioning, and multistep integration algorithms.

■ Changing the splitting orders of the integration procedure can impact on the configurational sampling and the dynamic properties of the system.



*Q&A*