Multiple Time Stepping

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Objectives: Space-time multiresolution algorithms

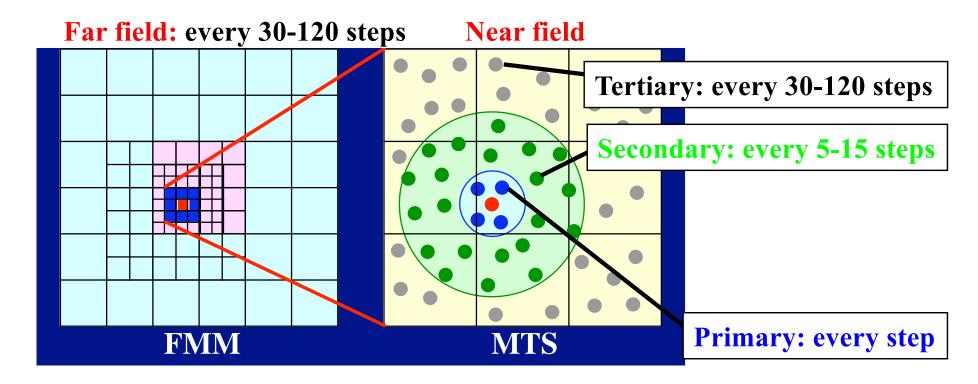
- > Tree codes: fast multipole method
- > Multiple time stepping





Temporal Locality: Multiple Time Stepping

- Different force-update schedules for different force components
 - → i) Reduced computation
 - ii) Enhanced data locality & parallel efficiency



A. Nakano et al., Comput. Phys. Commun. 83, 197 ('94)

https://aiichironakano.github.io/cs653/Nakano-MRMD-CPC94.pdf





Loop Invariant for Long-time Stability

Reversible symplectic integrator *via* split-operator method

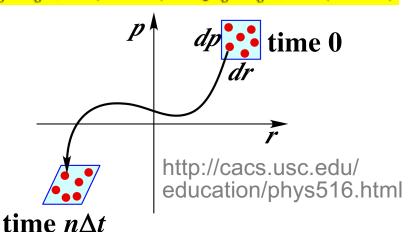
$$\Gamma(t + n\Delta t) = e^{iL_{\text{long}}n\Delta t/2} \left(e^{iL_{\text{short}}\Delta t}\right)^n e^{iL_{\text{long}}n\Delta t/2} \Gamma(t)$$

SYMPLECTIC-MTS(positions \mathbf{r}^N , velocities \mathbf{v}^N) initialize long-range accelerations, $\mathbf{a}_{long}^{N}(\mathbf{r}^{N})$ for outer step $\leftarrow 1$ to Max outer $\mathbf{v}^N \leftarrow \mathbf{v}^N + \mathbf{a}_{\text{long}}^N \times \text{Max_inner} \times \Delta t/2$ initialize short-range accelerations, $\mathbf{a}_{\text{short}}^{N}(\mathbf{r}^{N})$ for inner step $\leftarrow 1$ to Max inner $\mathbf{v}^N \leftarrow \mathbf{v}^N + \mathbf{a}_{\text{short}}^N \Delta t/2$ $\mathbf{r}^N \leftarrow \mathbf{r}^N + \mathbf{v}^N \Lambda t$ update $\mathbf{a}_{\text{short}}^{N}(\mathbf{r}^{N})$ $\mathbf{v}^N \leftarrow \mathbf{v}^N + \mathbf{a}_{\text{short}}^N \Delta t/2$ update $\mathbf{a}_{long}^{N}(\mathbf{r}^{N})$ $\mathbf{v}^N \leftarrow \mathbf{v}^N + \mathbf{a}_{long}^N \times Max_inner \times \Delta t/2$

Phase-space volume is a simulation-loop invariant

Long-time stability

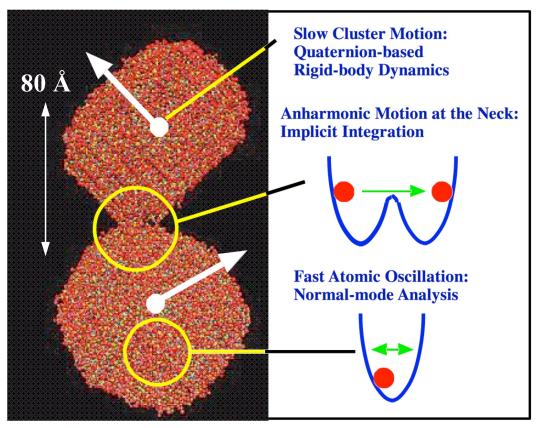
$$\frac{\partial (p_{n\Delta t}^{N}, r_{n\Delta t}^{N})}{\partial (p_{0}^{N}, r_{0}^{N})}^{T} \begin{pmatrix} 0 & \mathbf{I} \\ -\mathbf{I} & 0 \end{pmatrix} \frac{\partial (p_{n\Delta t}^{N}, r_{n\Delta t}^{N})}{\partial (p_{0}^{N}, r_{0}^{N})} = \begin{pmatrix} 0 & \mathbf{I} \\ -\mathbf{I} & 0 \end{pmatrix}$$

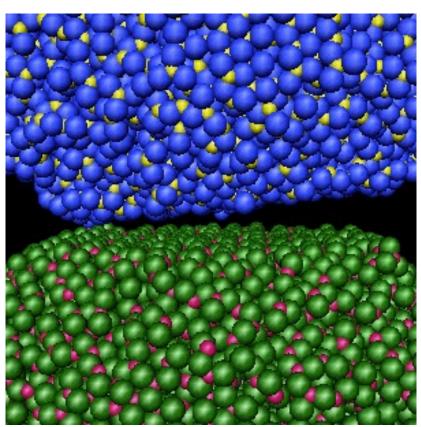


M. Tuckerman, B.J. Berne & G.J. Martyna, *J. Chem. Phys.* **97**, 1990 ('92) https://aiichironakano.github.io/cs653/Tuckerman-RESPA-JCP92.pdf

Clustering-based Hierarchical Dynamics

10⁻⁶ sec simulation requires 10⁹ iterations ($\Delta t = 10^{-15}$ sec): 1,000-fold increase of Δt ?



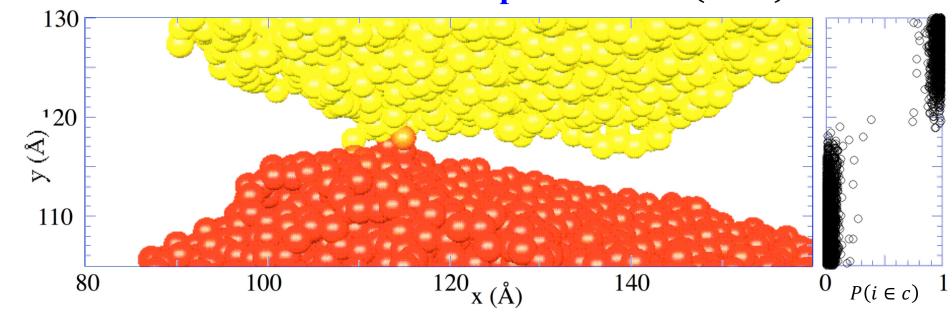


Rigid-body/implicit-integration/normal-mode approach achieves 28-fold speedup over a conventional MD

$$m_i \frac{d^2 \mathbf{z}_i}{dt^2} = \mathbf{F}_i \left(\left\{ \mathbf{z}_i + \mathbf{r}_i^{\text{RigidBody}} \right\} \right) - \mathbf{F}_i \left(\left\{ \mathbf{r}_i^{\text{RigidBody}} \right\} \right) + \frac{\partial^2 V}{\partial \mathbf{r}_{\min,i}^2} \left(\mathbf{r}_i^{\text{NormalMode}} - \mathbf{r}_{\min,i} \right)$$

Fuzzy Clustering Facilitates Seamless Integration of Hierarchical Abstraction

Fractional membership function: $P(i \in c)$



Clustering based on chemical cohesion, v_{ij} cf. fuzzy c-means algorithm, Bezdek

$$E_c(i) = \frac{1}{2} \sum_{j(\neq i)} P(j \in c) v_{ij} (|\vec{r}_i - \vec{r}_j|)$$

A. Nakano, Comput. Phys. Commun. 105, 139 ('97)

https://aiichironakano.github.io/cs653/Nakano-fuzzy-CPC97.pdf

Fuzzy Clustering Improves the Numerical Accuracy of Hierarchical Dynamics

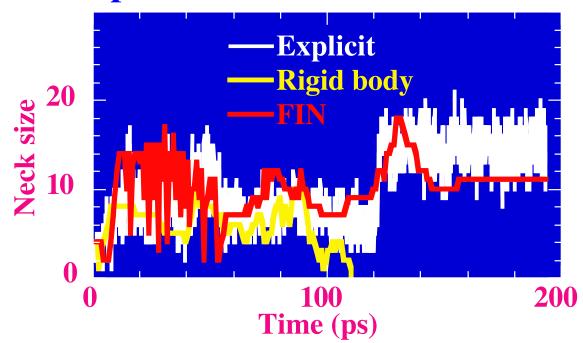
Maximum entropy principle

Constrained maximization:
$$S_i = -\sum_c P(i \in c) \log P(i \in c)$$

$$\sum_c P(i \in c) = 1; \sum_c E_c(i) P(i \in c) = \text{const.}$$

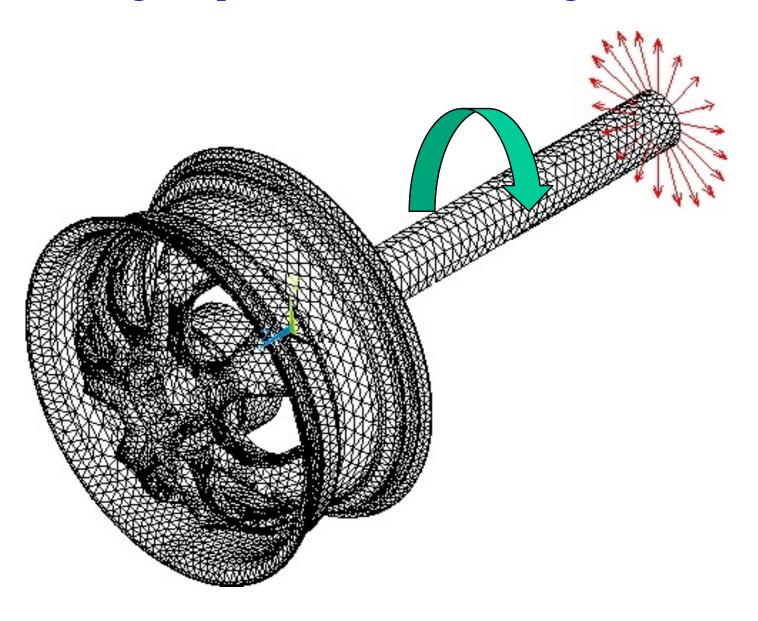
$$P(i \in c) = \exp[-E_c(i)/k_{\rm B}T]/\sum_{c'} \exp[-E_{c'}(i)/k_{\rm B}T]$$

Fixed-point iteration to determine P



Lesson

Use the right representation at each length/time scale



Multiscale MD/QD/FD Simulation

- Hybrid atoms (molecular dynamics, MD)-electrons (quantum dynamics, QD)-electromagnetic field (field dynamics, FD) simulations
- Multiple time-scales: atoms, $\Delta t_{\rm MD}$ (10⁻¹⁵ s) > electrons, $\Delta t_{\rm QD}$ (10⁻¹⁸ s) > electromagnetic field ($e^2/\hbar c \times \Delta t_{\rm QD} = \Delta t_{\rm QD}/136$)
- Split-operator formulation:

$$\exp\left(\frac{iL_{\text{MD}}\Delta t_{\text{MD}}}{2}\right) \times \left[\exp\left(\frac{iH_{\text{QD}}\Delta t_{\text{QD}}}{2}\right)\exp(iL_{\text{FD}}\Delta t_{\text{FD}})^{N_{\text{FD}}}\exp\left(\frac{iH_{\text{QD}}\Delta t_{\text{QD}}}{2}\right)\right]^{N_{\text{QD}}} \times \exp\left(\frac{iL_{\text{MD}}\Delta t_{\text{MD}}}{2}\right)$$

Divide-&-conquer Maxwell-Ehrenfest-surface hopping (DC-MESH) code implemented on heterogeneous CPU (central processing unit)-GPU (graphics processing unit) parallel computers

cf. T. Linker et al., Science Adv. 8, eabk2625 ('22) T. M. Razack et al., PDSEC ('24); SC ('25)

What We Have Learned So Far

- Molecular dynamics (MD) represents the dynamic, irregular dwarf (i.e., interaction among spatially-distributed entities)
- Data locality (e.g., finite interaction range) is essential to achieve high scalability, which in turn should be expressed using appropriate data structures (e.g., linked-list cells)
- If there is no obvious locality, consider divide-conquer-"recombine (e.g., interactive cells in fast multipole method)" —multiresolution in space
- Different subtasks may require different update schedules; consider divide-&-conquer or multiresolution in time
- Q: Any spatiotemporal multiresolution in "your" application? Any interesting papers?
- Tip: Learn a new concept by applying it to what you know well