

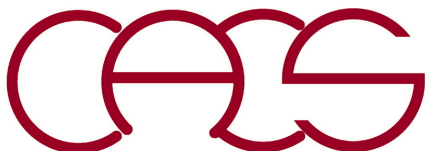
Multiple Time Stepping

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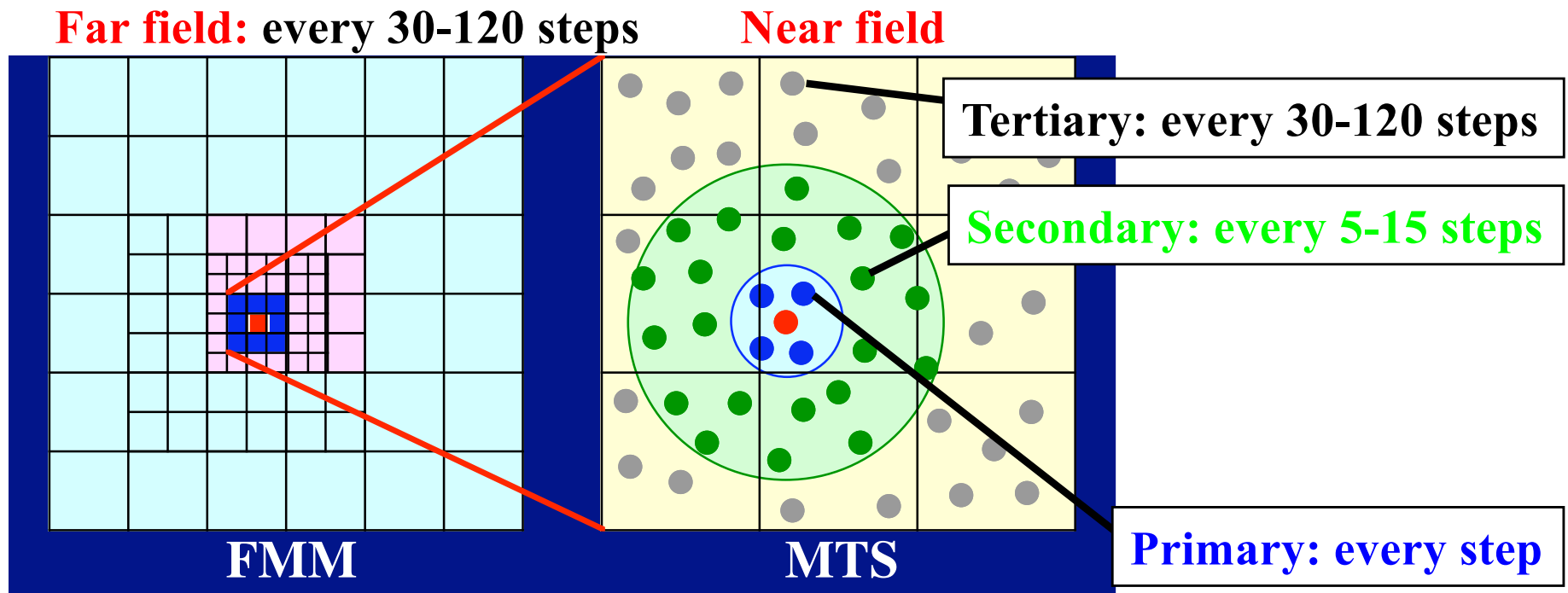
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Objectives: Space-time multiresolution algorithms
> Tree codes: fast multipole method
> Multiple time stepping



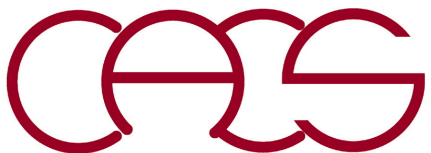
Temporal Locality: Multiple Time Stepping

- 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

$L_{\text{short/long}}$: Short/long-range Liouville operator

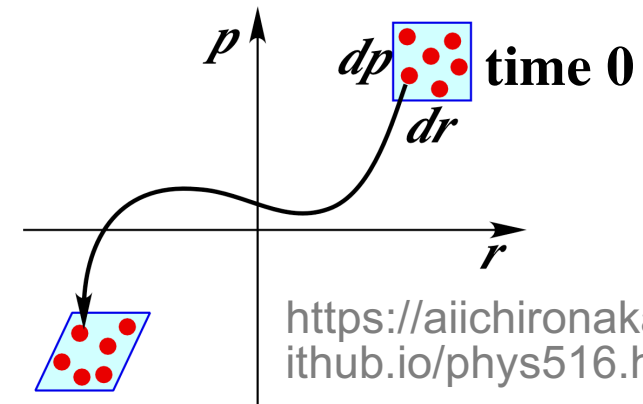
$$\Gamma(t + n\Delta t) = e^{iL_{\text{long}}n\Delta t/2} \left(e^{iL_{\text{short}}\Delta t} \right)^n e^{iL_{\text{long}}n\Delta t/2} \Gamma(t)$$

**Phase-space volume is a
simulation-loop invariant**



Long-time stability

$$\frac{\partial(p_{n\Delta t}^N, r_{n\Delta t}^N)^T}{\partial(p_0^N, r_0^N)} \begin{pmatrix} 0 & \mathbf{I} \\ -\mathbf{I} & 0 \end{pmatrix} \frac{\partial(p_{n\Delta t}^N, r_{n\Delta t}^N)}{\partial(p_0^N, r_0^N)} = \begin{pmatrix} 0 & \mathbf{I} \\ -\mathbf{I} & 0 \end{pmatrix}$$



<https://aiichironakano.github.io/phys516.html>

time $n\Delta t$

SYMPLECTIC-MTS(positions \mathbf{r}^N , velocities \mathbf{v}^N)

initialize long-range accelerations, $\mathbf{a}_{\text{long}}^N(\mathbf{r}^N)$

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

$\mathbf{v}^N \leftarrow \mathbf{v}^N + \mathbf{a}_{\text{short}}^N \Delta t / 2$

$\mathbf{r}^N \leftarrow \mathbf{r}^N + \mathbf{v}^N \Delta t$

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 $\mathbf{a}_{\text{long}}^N(\mathbf{r}^N)$

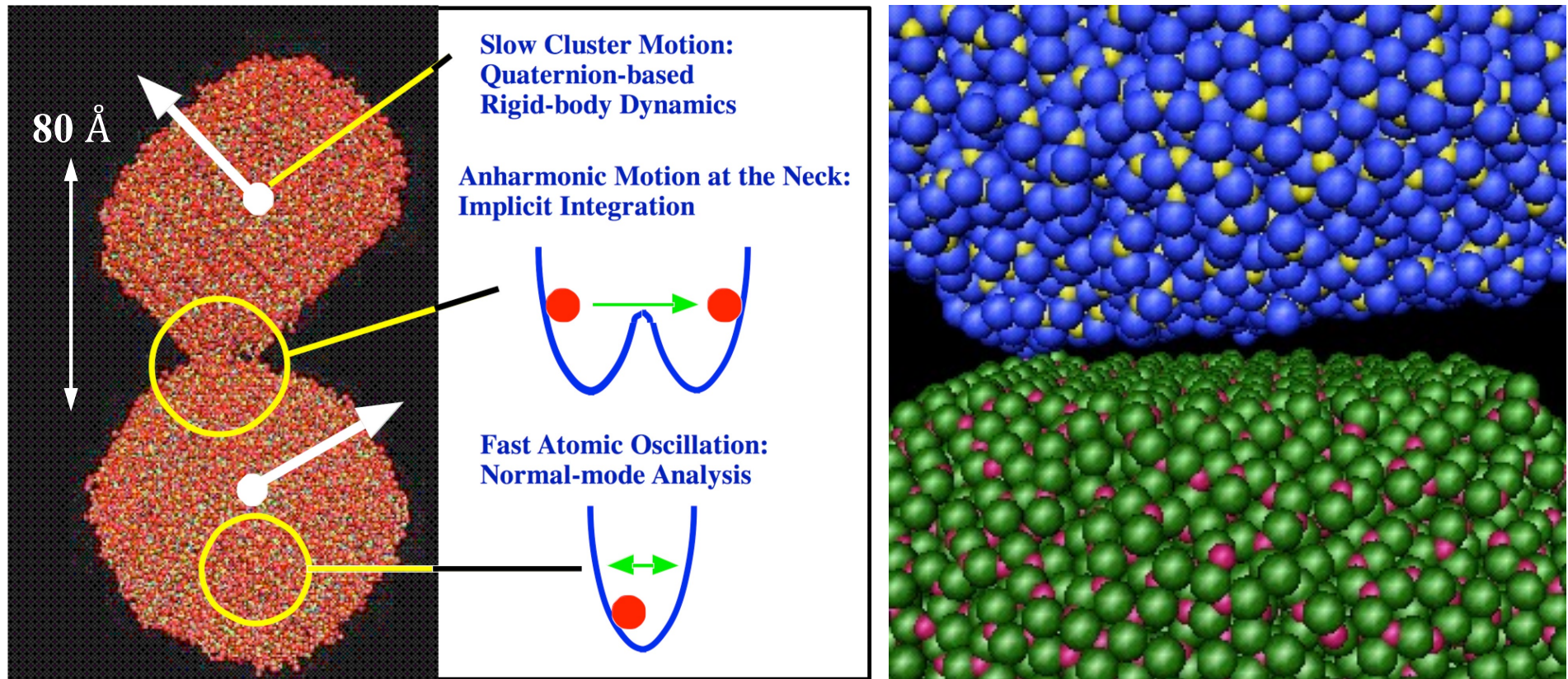
$\mathbf{v}^N \leftarrow \mathbf{v}^N + \mathbf{a}_{\text{long}}^N \times \text{Max_inner} \times \Delta t / 2$

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 10^9 iterations ($\Delta t = 10^{-15}$ sec):
1,000-fold increase of Δt ?

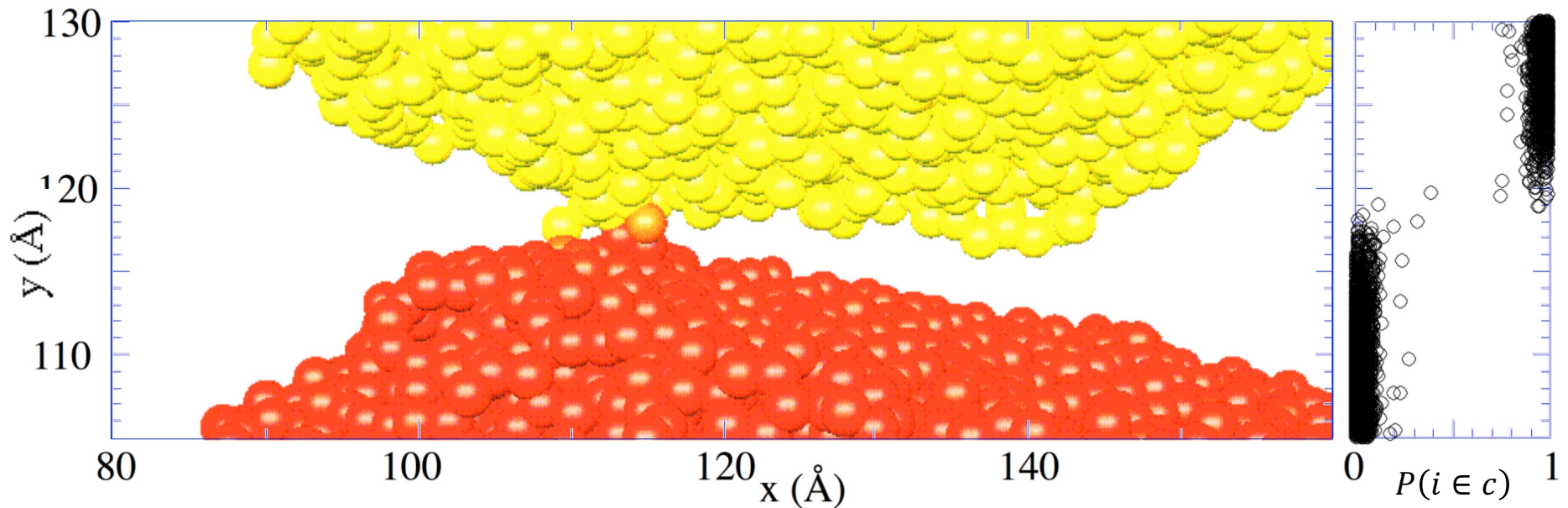


**Rigid-body/implicit-integration/normal-mode approach
achieves 28-fold speedup over a conventional MD**

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

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

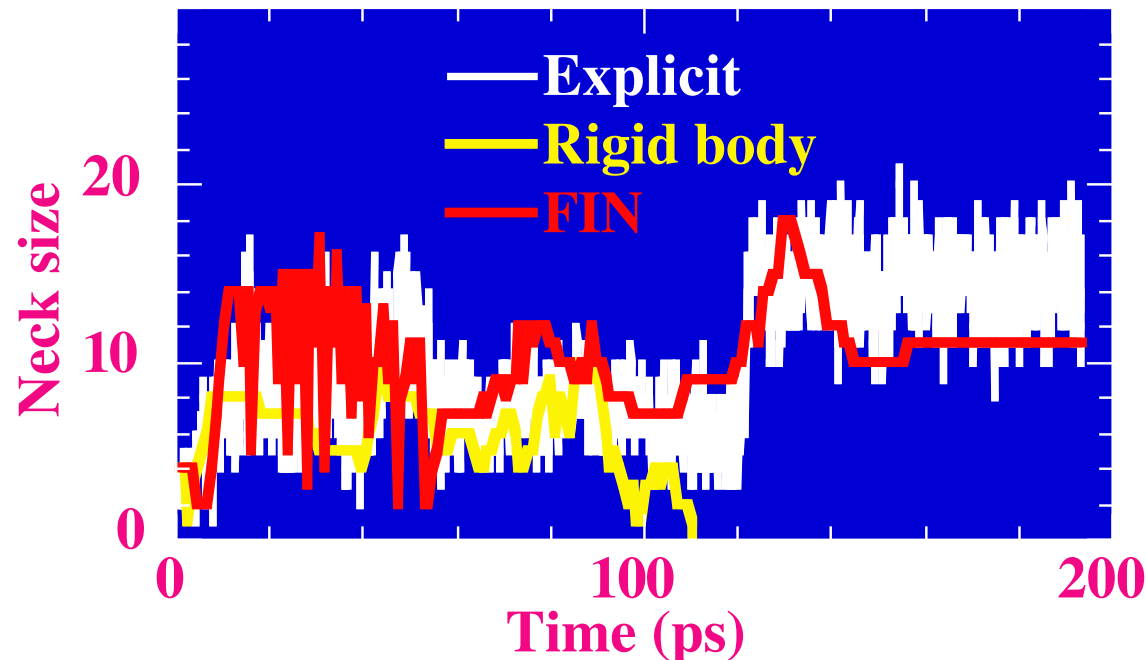
Maximum entropy principle

Constrained maximization: $S_i = -\sum_c P(i \in c) \log P(i \in c)$
 $\sum_c P(i \in c) = 1; \sum_c E_c(i) P(i \in c) = \text{const.}$



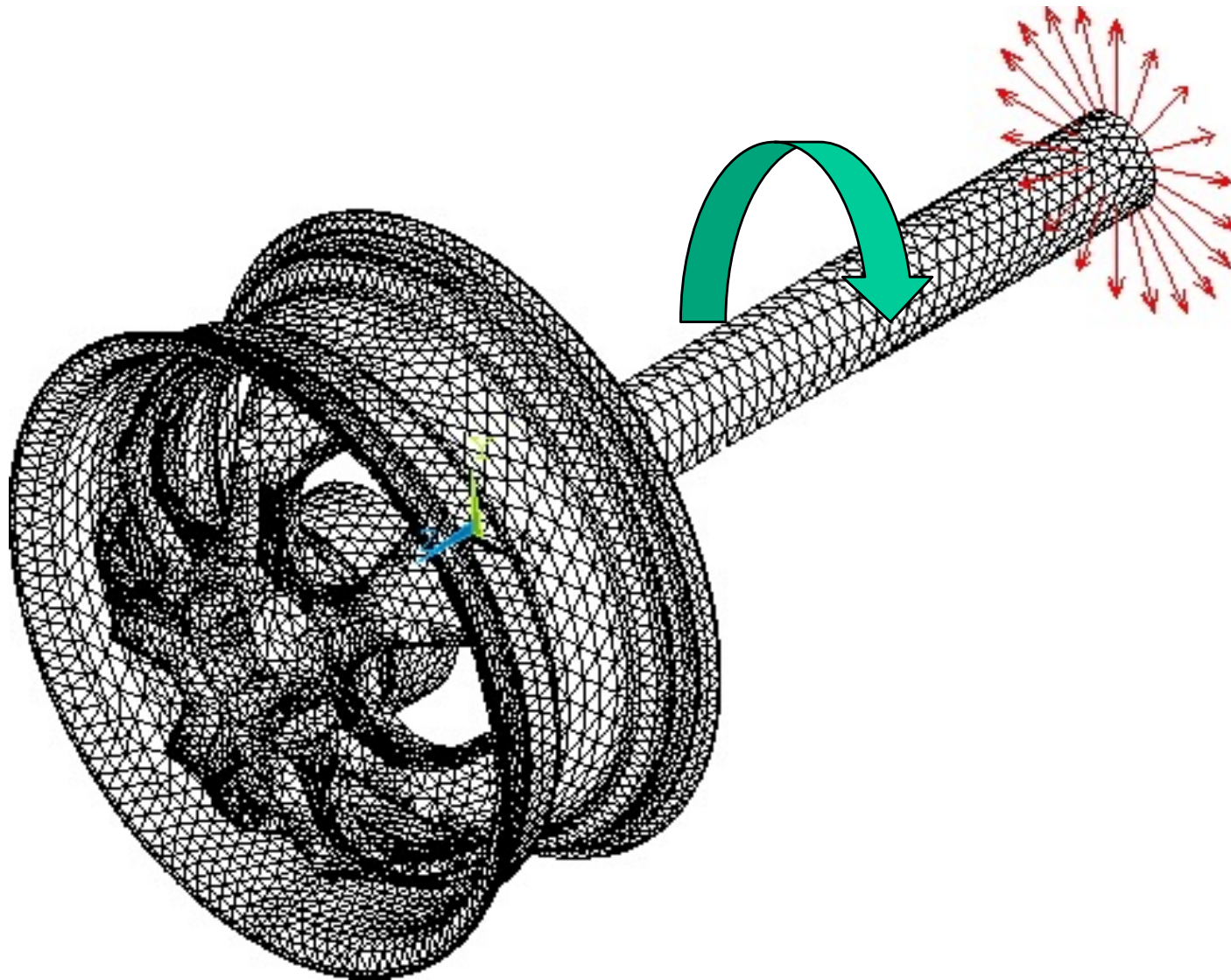
$$P(i \in c) = \exp[-E_c(i)/k_B T] / \sum_{c'} \exp[-E_{c'}(i)/k_B T]$$

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, Δt_{MD} (10^{-15} s) > electrons, Δt_{QD} (10^{-18} s) > electromagnetic field ($e^2/\hbar c \times \Delta t_{\text{QD}} = \Delta t_{\text{QD}}/136$)
- Split-operator formulation:

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

Hamiltonian operator

- 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

cf. <https://aiichironakano.github.io/cs653-papers.html>