

Quantum Materials Dynamics at the Nexus of Exascale Computing, Artificial Intelligence, and Quantum Computing

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Anikeya Aditya, Lindsay Bassman, Hikaru Ibayashi, Thomas Linker,
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Materials Science, and Quantitative & Computational Biology
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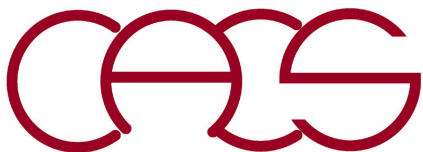
Collaborators:

F. Shimojo, K. Shimamura (*Kumamoto University, Japan*)

*UCLA-IPAM Workshop on “Increasing the Length, Time, and Accuracy of
Materials Modeling Using Exascale Computing”*

Organizers: Dr. V. Ehlacher, Dr. V. Gavini, Dr. D. Perez, Dr. S. Plimpton

March 27, 2023, Los Angeles, CA



Current & Future Supercomputing

- Won two DOE supercomputing awards to develop & deploy metascalable (“design once, scale on future platforms”) simulation algorithms

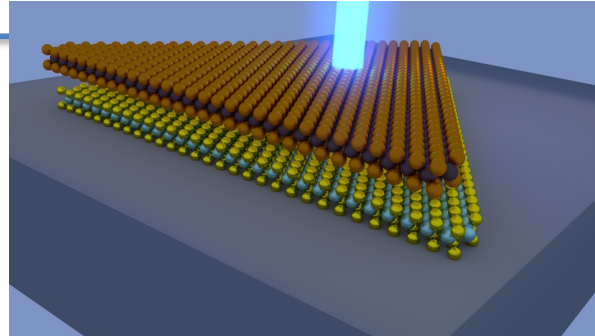
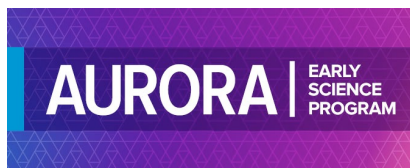


- Atomistic simulations on million cores (pre-exascale)

Title: AI-Guided Exascale Simulations of Quantum Materials Manufacturing and Control
PI and Co-PIs: Aiichiro Nakano–PI, Rajiv K. Kalia, Ken-ichi Nomura, Priya Vasishta



786,432-core IBM Blue Gene/Q
281,088-core Intel Xeon Phi
560-node (2,240-GPU) AMD/NVIDIA Polaris



Early Science Projects for Aurora
Supercomputer Announced
Metascalable layered materials genome
Investigator: Aiichiro Nakano, University of Southern California



2 exaflop/s
Intel Aurora (forthcoming)

exaflop/s = 10^{18} mathematical operations per second

- One of the initial simulation users of the next-generation DOE supercomputer

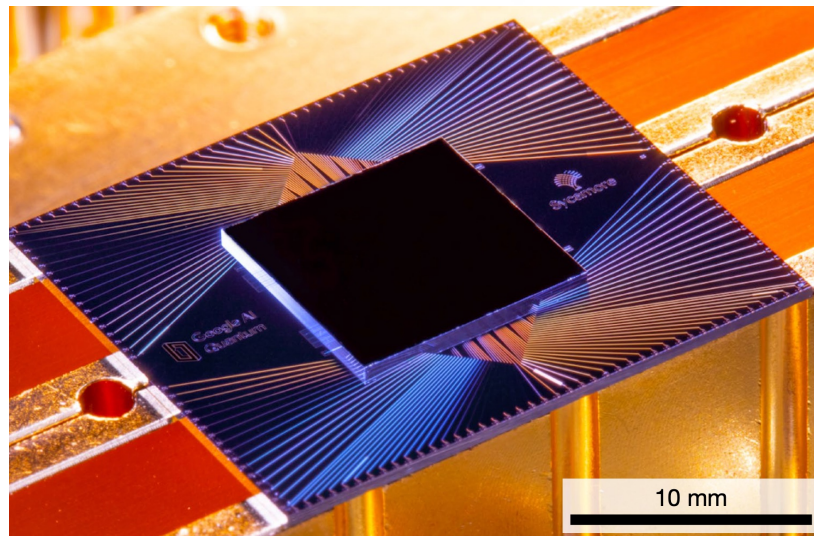
Changing Computing Landscape for Science

Post-exascale Computing for Science



Compute Cambrian explosion

Quantum Computing for Science



AI for Science

DOE readies multibillion- dollar AI push

U.S. supercomputing leader
is the latest big backer
in a globally crowded field

By **Robert F. Service**, in Washington, D.C.

Science 366, 559 (Nov. 1, '19)



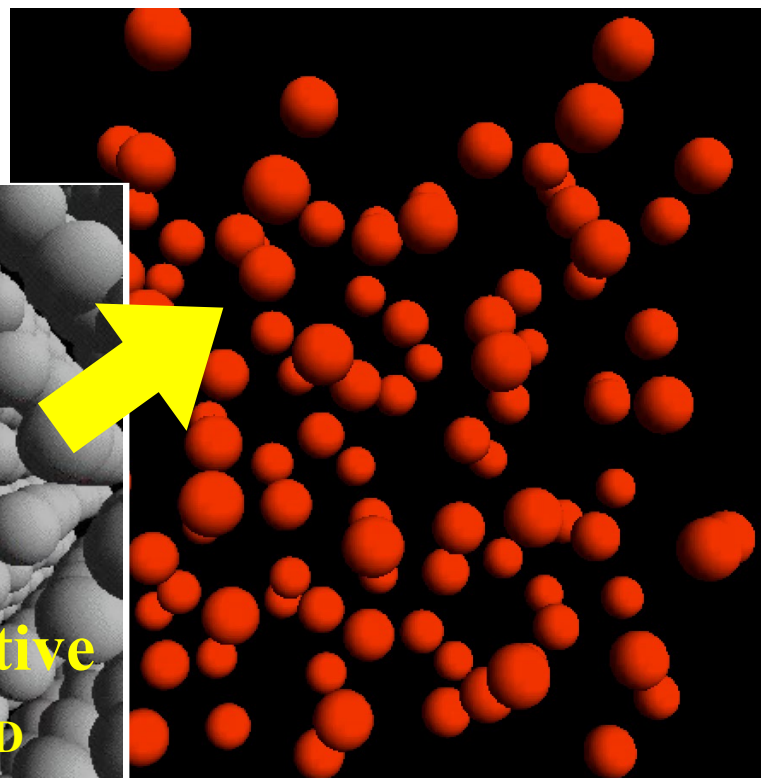
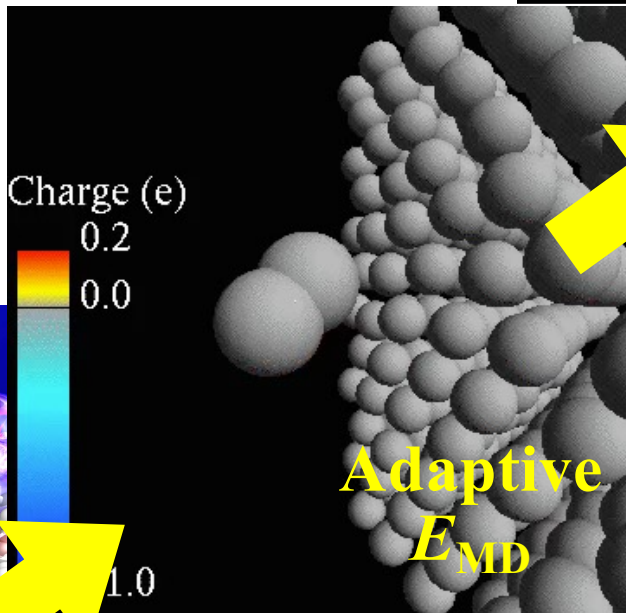
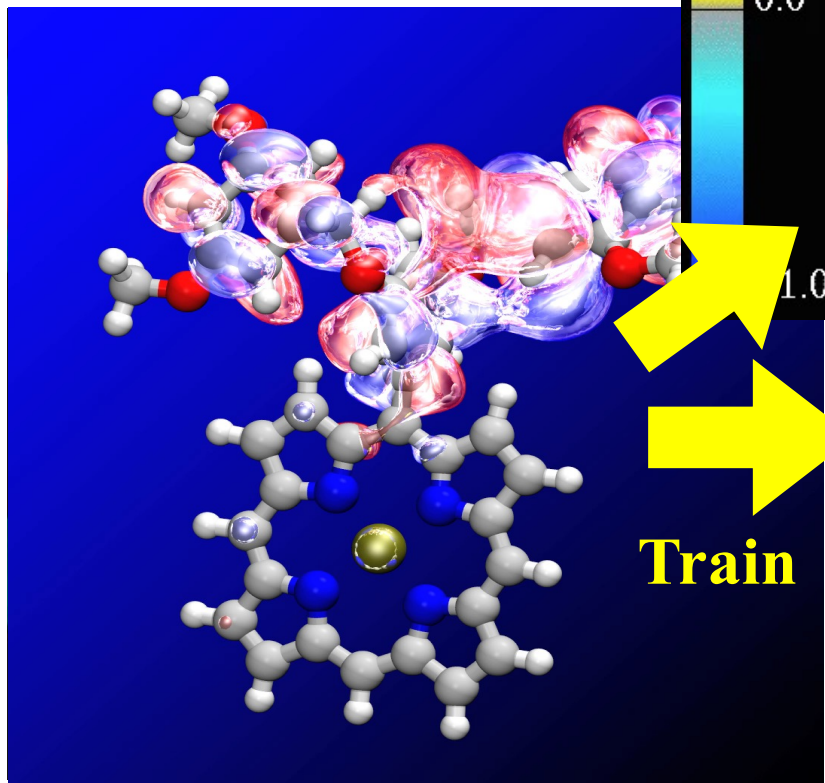
Use all to advance science!

Molecular Dynamics & Machine Learning

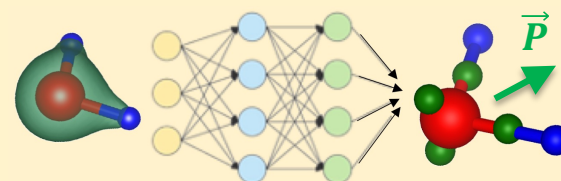
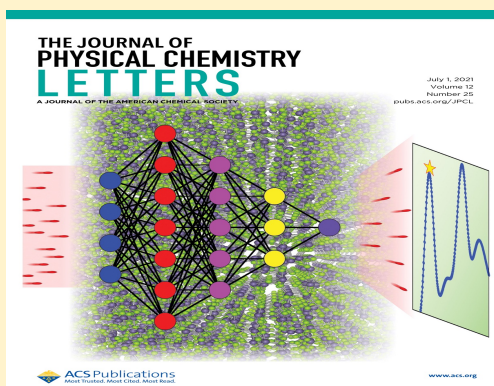
Molecular Dynamics (*MD*)

Reactive MD (*RMD*)

Nonadiabatic quantum
MD (*NAQMD*)



First principles-based neural-network
quantum molecular dynamics (*NNQMD*)



Physical Review Letters
Editor's choice
(May 25, 2021)

BES

Exa-scale

BASIC ENERGY SCIENCES

EXASCALE REQUIREMENTS REVIEW

An Office of Science review sponsored jointly by
Advanced Scientific Computing Research and Basic Energy Sciences

16,661-atom QMD

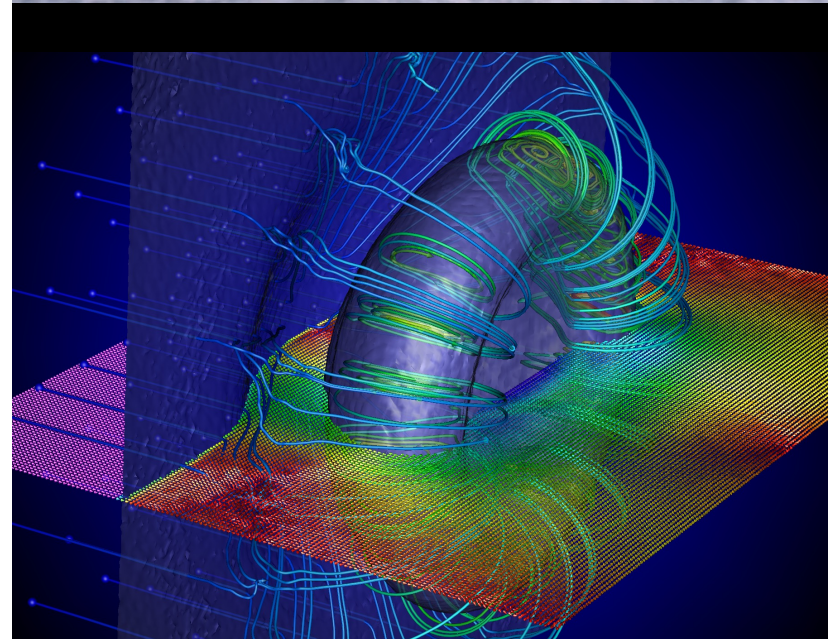
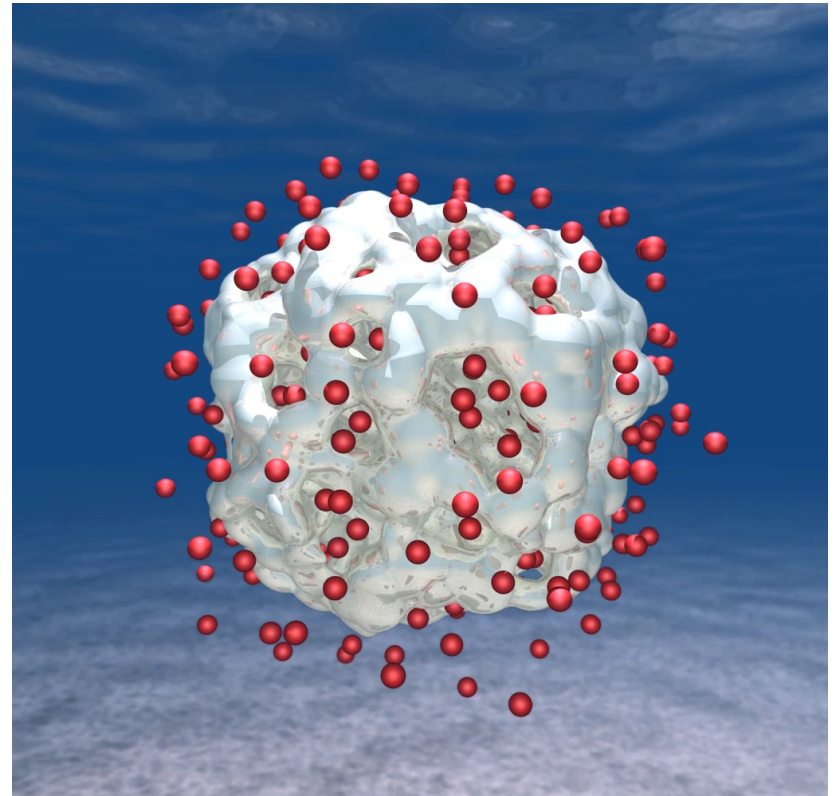
Shimamura *et al.*,
Nano Lett.
14, 4090 ('14)

10⁹-atom RMD

Shekhar *et al.*,
Phys. Rev. Lett.
111, 184503 ('13)

NOVEMBER 3-5, 2015

ROCKVILLE, MARYLAND



BES

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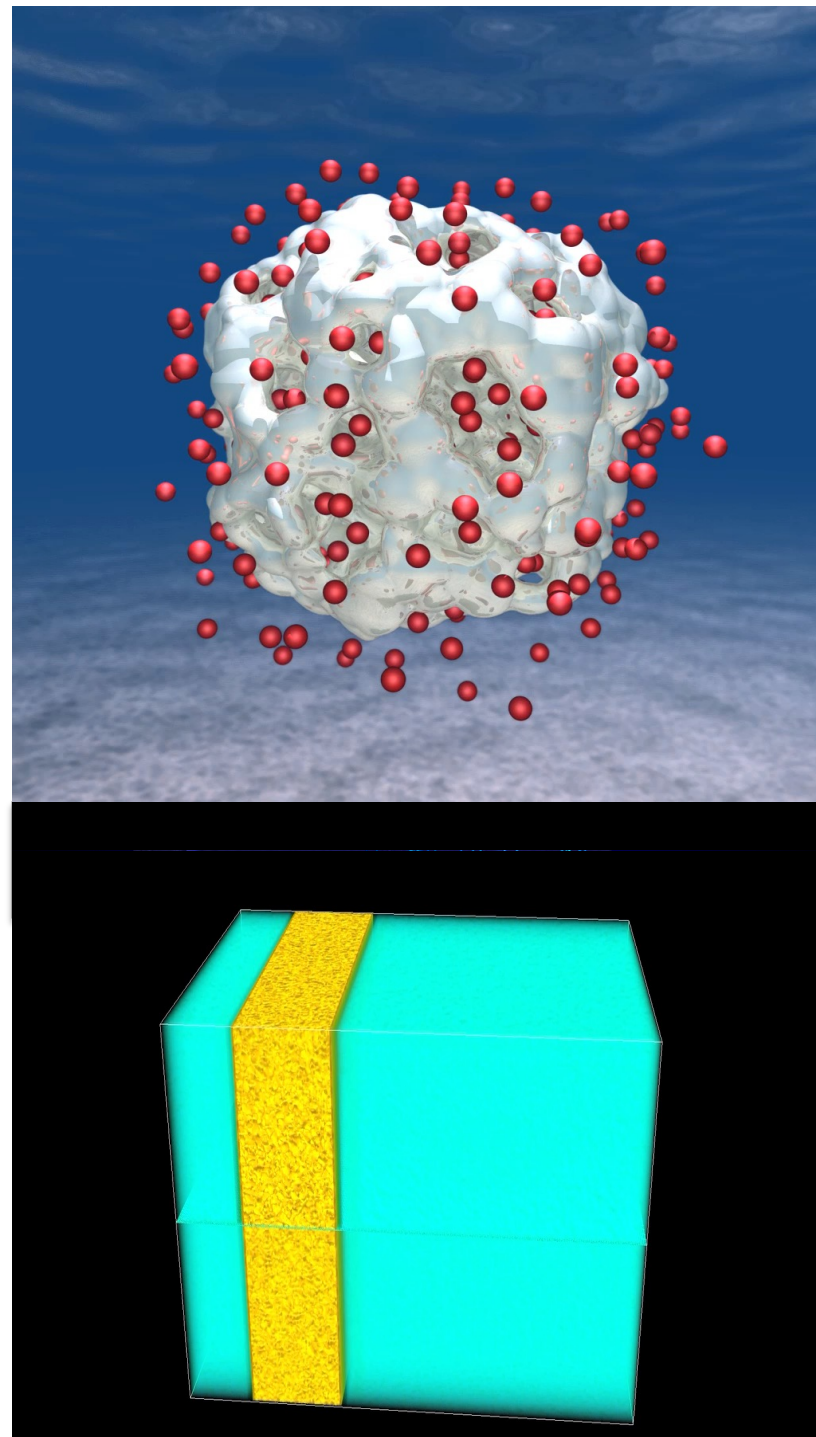
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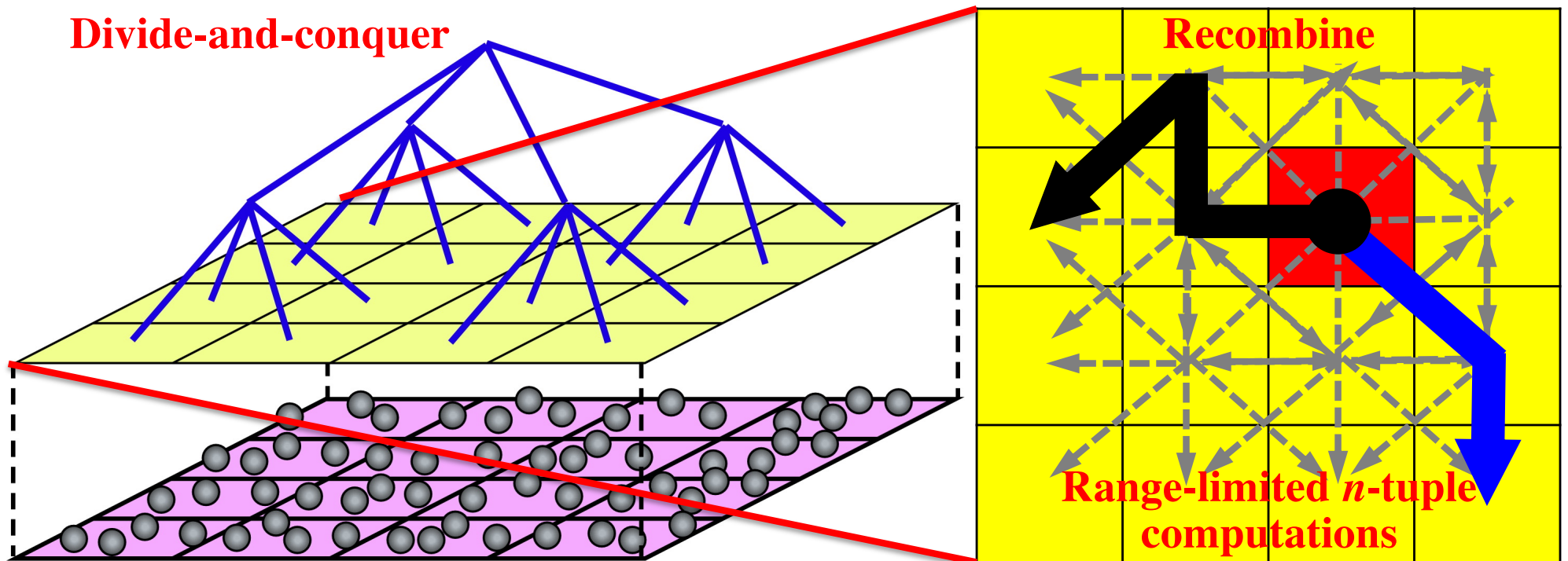
111, 184503 ('13)

NOVEMBER 3-5, 2015

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Divide-Conquer-Recombine (DCR) Engines



M. Kunaseth et al., ACM/IEEE SC13

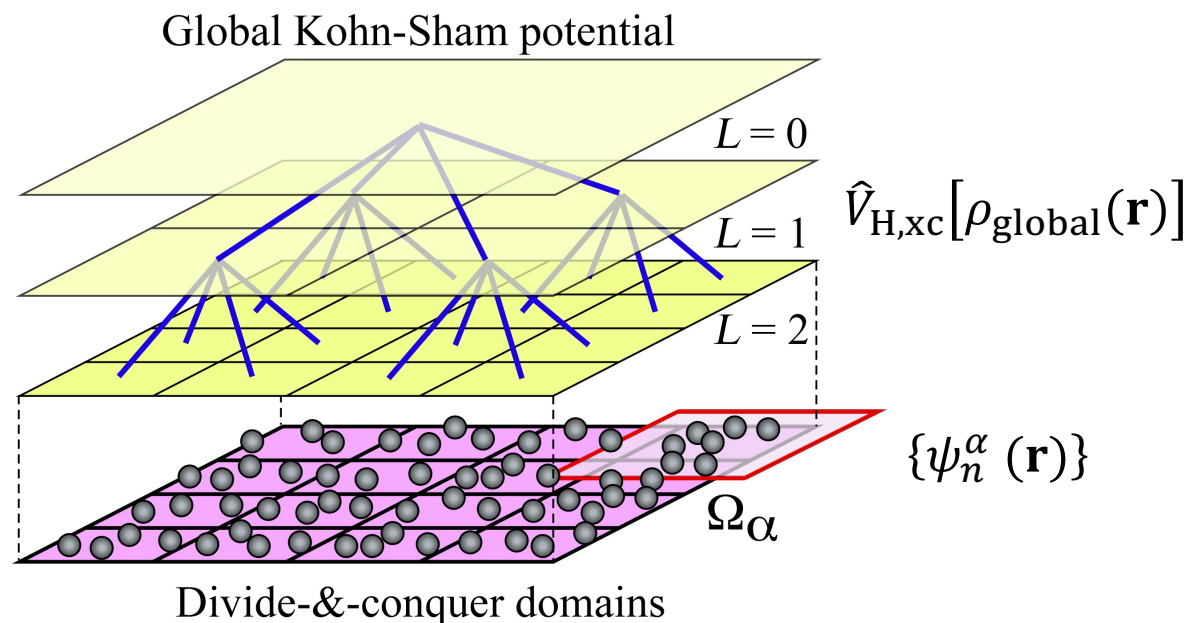
- **Variable N -charge problem: $O(N^3) \rightarrow O(N)$: Extended-Lagrangian reactive molecular dynamics algorithm eliminates speed-limiting charge iteration**
Nomura et al., *Comput. Phys. Commun.* **192**, 91 ('15); Liu et al., *IEEE/ACM ScalA18*
cf. Niklasson, *Phys. Rev. Lett.* **100**, 123004 ('08)
- **Quantum N -body problem: $O(C^N) \rightarrow O(N)$: Lean divide-&-conquer density functional theory algorithm for QMD minimizes $O(N)$ prefactor**
Shimojo et al., *J. Chem. Phys.* **140**, 18A529 ('14); Nomura et al., *IEEE/ACM SC14*;
Tiwari et al., *ACM HPCAsia20*, best paper award

Divide-&-Conquer Density Functional Theory

Yang, *Phys. Rev. Lett.* **66**, 1438 ('91)

Shimojo *et al.*, *Comput. Phys. Commun.* **167**, 151 ('05)

Shimojo *et al.*, *J. Chem. Phys.* **140**, 18A529 ('14)



- **Overlapping spatial domains:** $\Omega = \cup_\alpha \Omega_\alpha$
- **Domain Kohn-Sham equations**

Global-local
self-consistent
field (SCF)
iteration

$$\left(-\frac{1}{2} \nabla^2 + \hat{V}_{\text{ion}} + \hat{V}_{H,xc}[\rho_{\text{global}}(\mathbf{r})] \right) \psi_n^\alpha(\mathbf{r}) = \epsilon_n^\alpha \psi_n^\alpha(\mathbf{r})$$

- **Global & domain electron densities**

$$\rho_{\text{global}}(\mathbf{r}) = \sum_\alpha p_\alpha(\mathbf{r}) \rho_\alpha(\mathbf{r}) \quad \leftarrow \quad \rho_\alpha(\mathbf{r}) = \sum_n [\psi_n^\alpha]^2 \Theta(\mu - \epsilon_n^\alpha)$$

Domain support function $\sum_\alpha p_\alpha(\mathbf{r}) = 1$

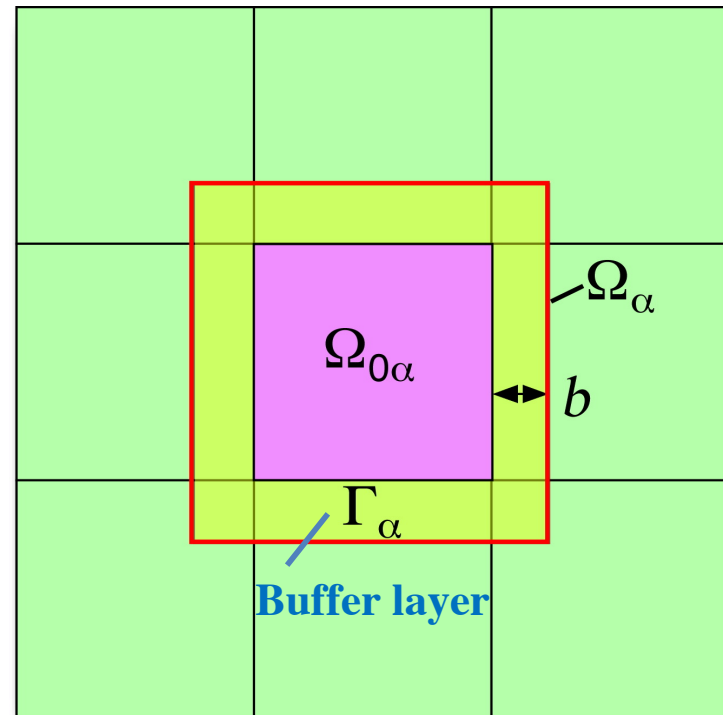
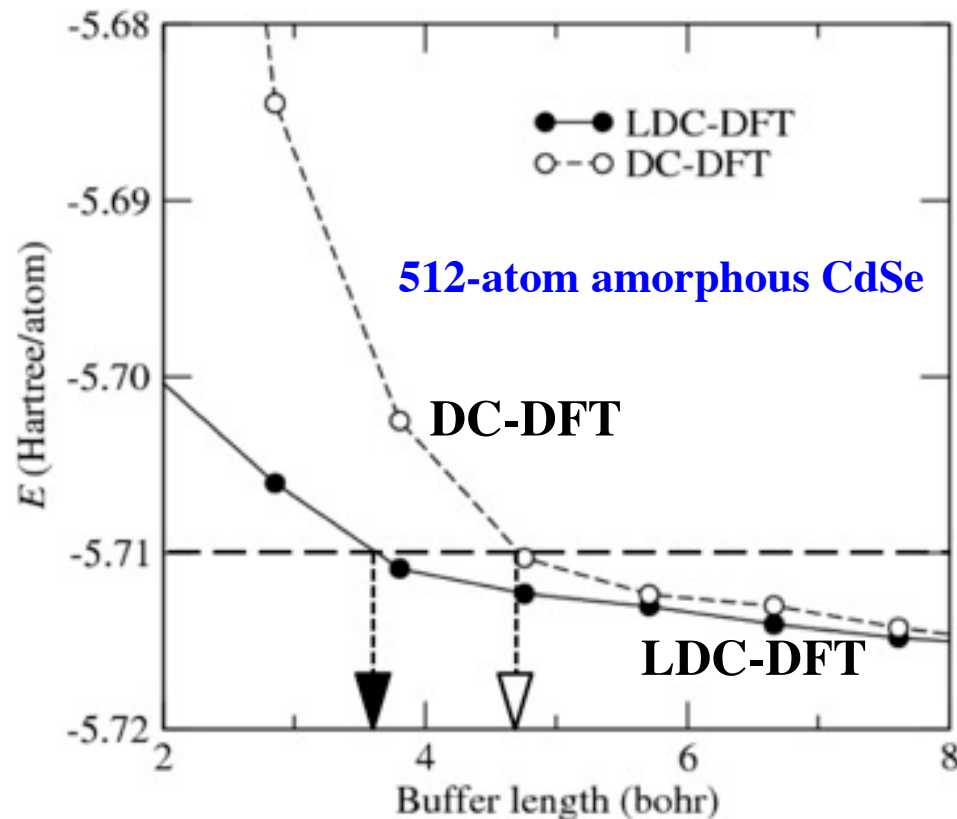
Global chemical potential $N = \int d\mathbf{r} \rho_{\text{global}}(\mathbf{r})$

Lean Divide-&-Conquer (LDC) DFT

- Density-adaptive boundary potential to reduce the $O(N)$ prefactor

$$v_{\alpha}^{bc}(\mathbf{r}) = \int d\mathbf{r}' \frac{\partial v(\mathbf{r})}{\partial \rho(\mathbf{r}')} \left(\rho_{\alpha}(\mathbf{r}) - \rho_{\text{global}}(\mathbf{r}) \right) \cong \frac{\rho_{\alpha}(\mathbf{r}) - \rho_{\text{global}}(\mathbf{r})}{\xi}$$

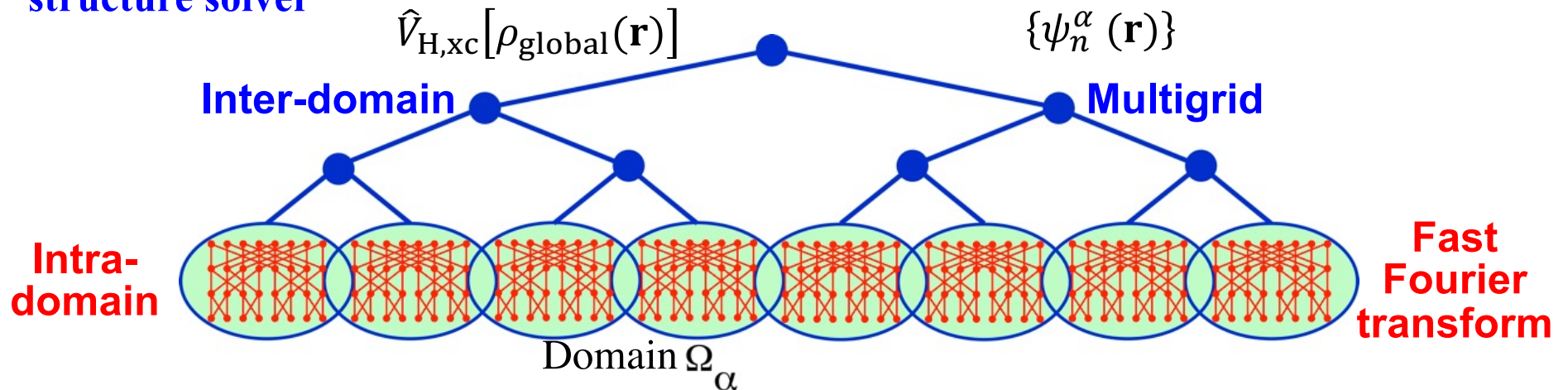
- More rapid energy convergence of LDC-DFT compared with nonadaptive DC-DFT



- Factor 2.03 (for $\nu = 2$) \sim 2.89 (for $\nu = 3$) reduction of the computational cost with an error tolerance of 5×10^{-3} a.u. (per-domain complexity: n^{ν})

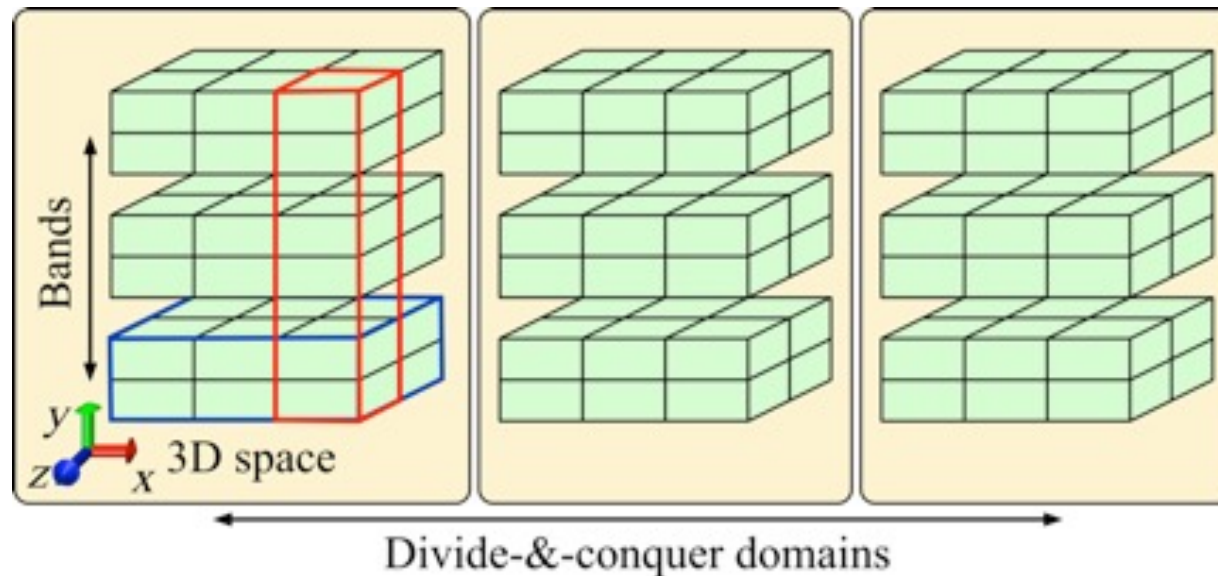
Hierarchical Computing: QMD

- **GSLF: Globally scalable (real-space multigrid) + locally fast (plane wave) electronic-structure solver**



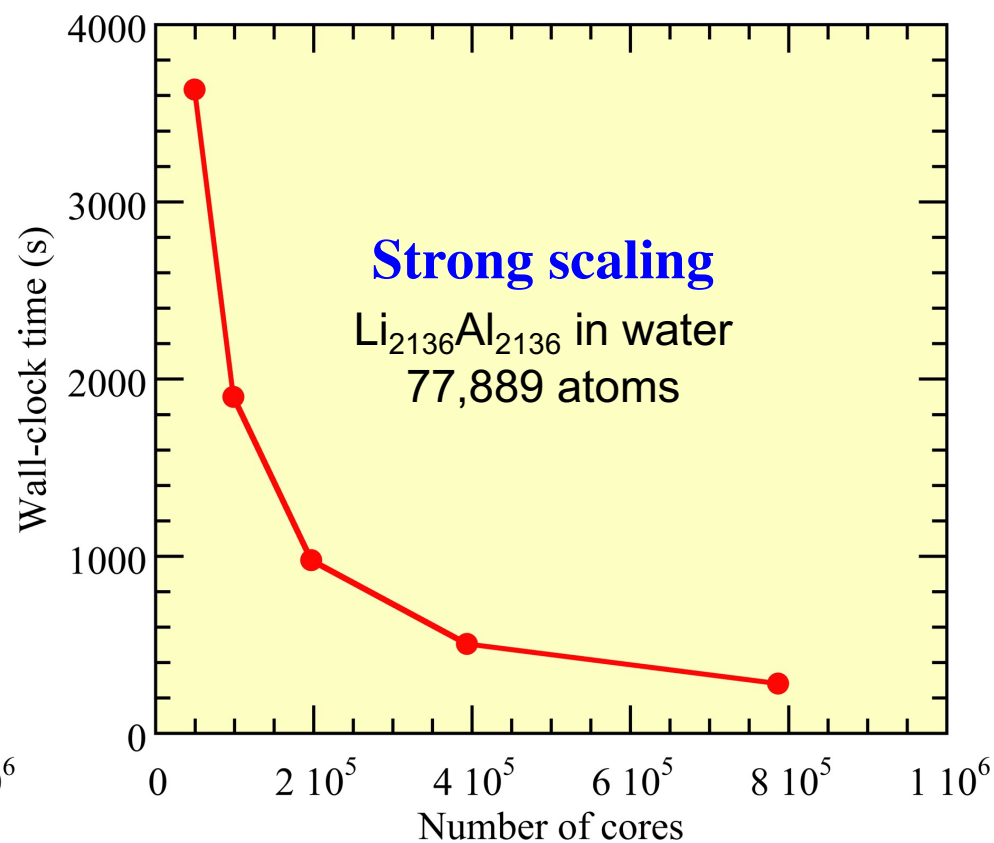
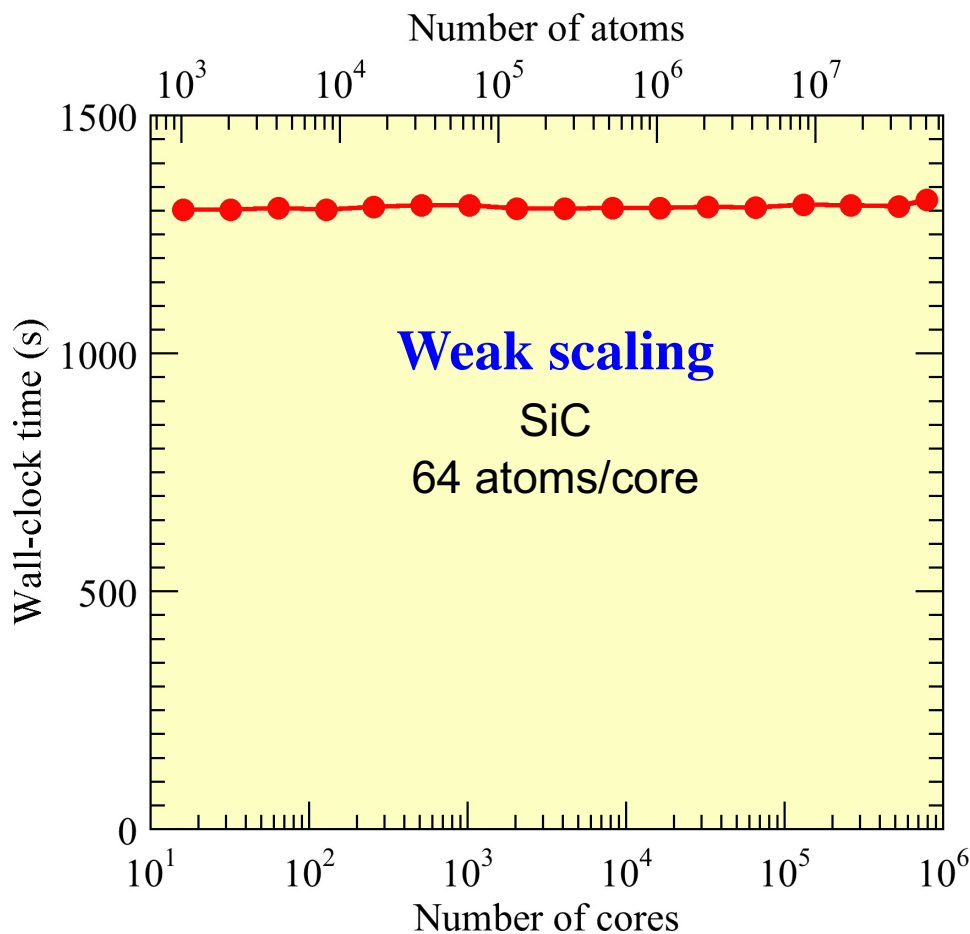
GSLF maximally exposes **data locality** → scalability

- **Hierarchical band (i.e., Kohn-Sham orbital) + space + domain (BSD) decomposition**

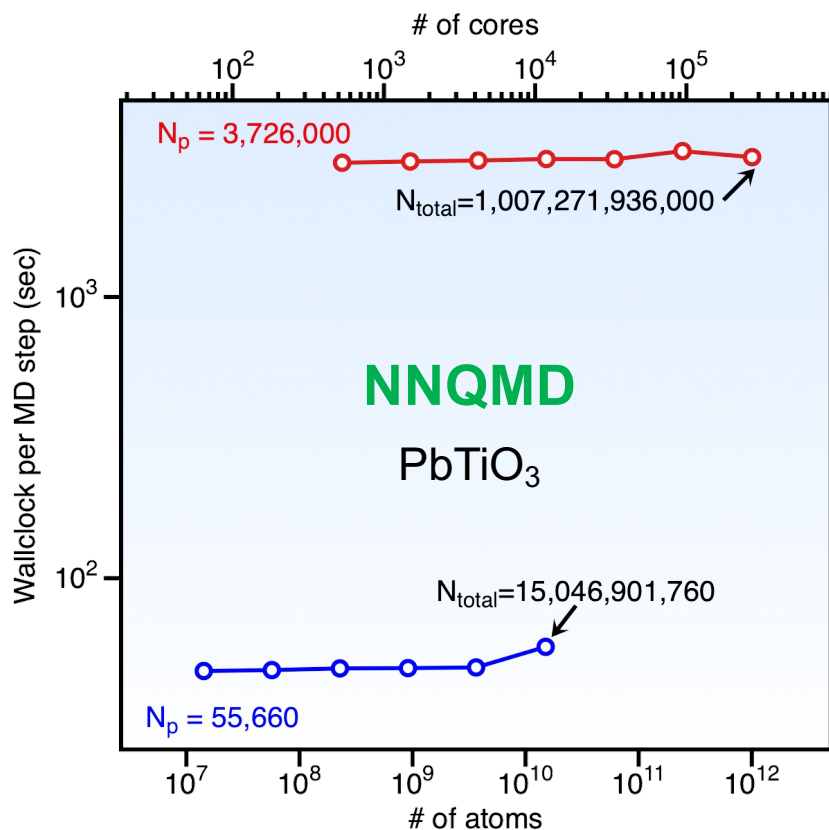


Parallel Performance: QMD

- **Weak-scaling parallel efficiency is 0.984 on 786,432 cores for a 50,331,648-atom SiC system**
- **Strong-scale parallel efficiency is 0.803 on 786,432 cores**



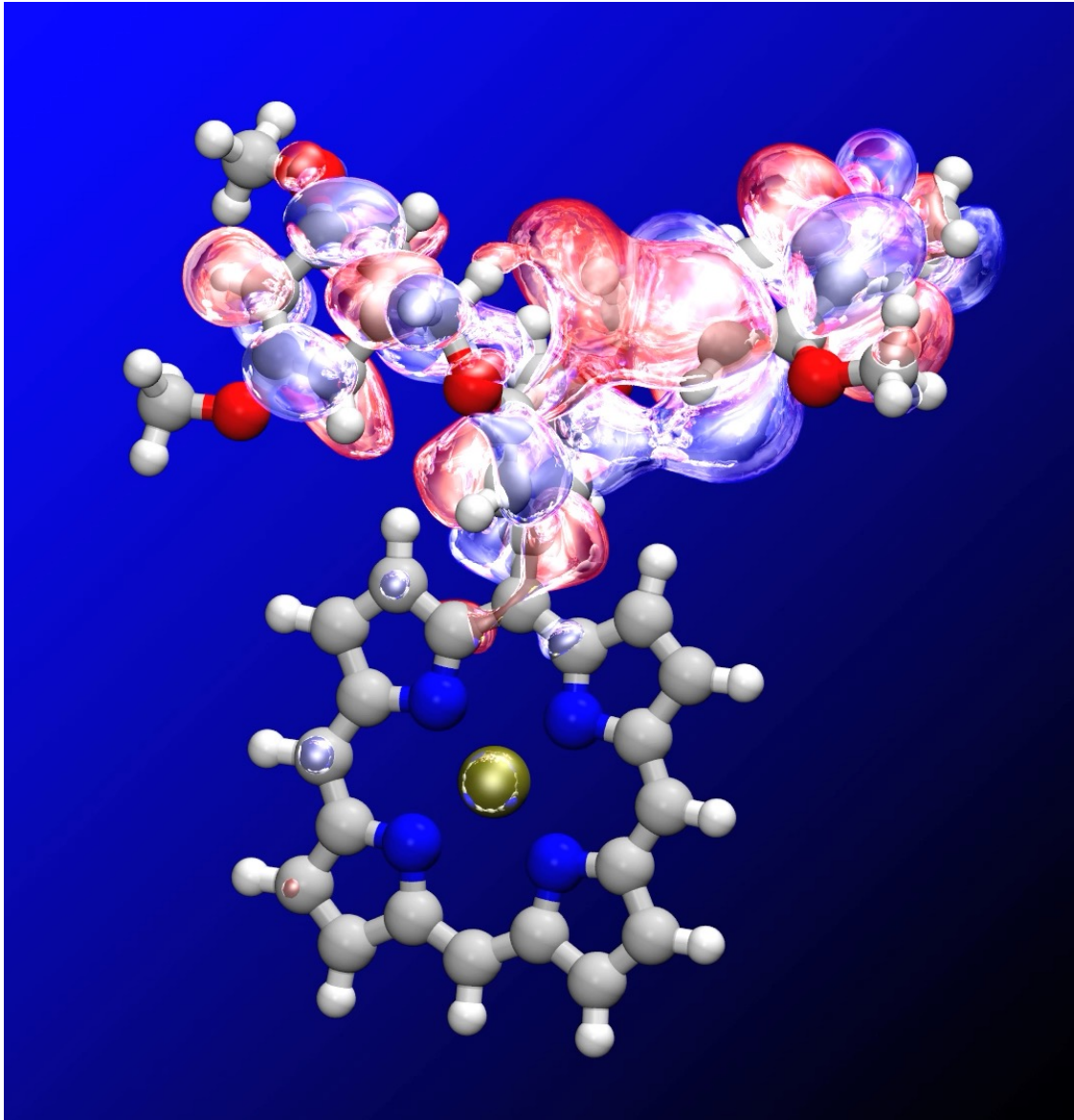
Parallel Performance: NNQMD



- **Trillion-atom scalability:** Weak-scaling parallel efficiency is **0.984** on 4,224 compute nodes (270,336 cores) of Theta supercomputer at Argonne National Laboratory for a trillion-atom neural-network quantum molecular dynamics (NNQMD) of PbTiO3
- **GPU acceleration:** **70× speedup** on GPU (NVIDIA A100) over CPU (AMD EPYC 7513) of Polaris supercomputer at Argonne

Number of atoms	1,000	10,000	100,000
Speedup (A100 GPU over EPYC CPU)	4.37	24.7	68.9

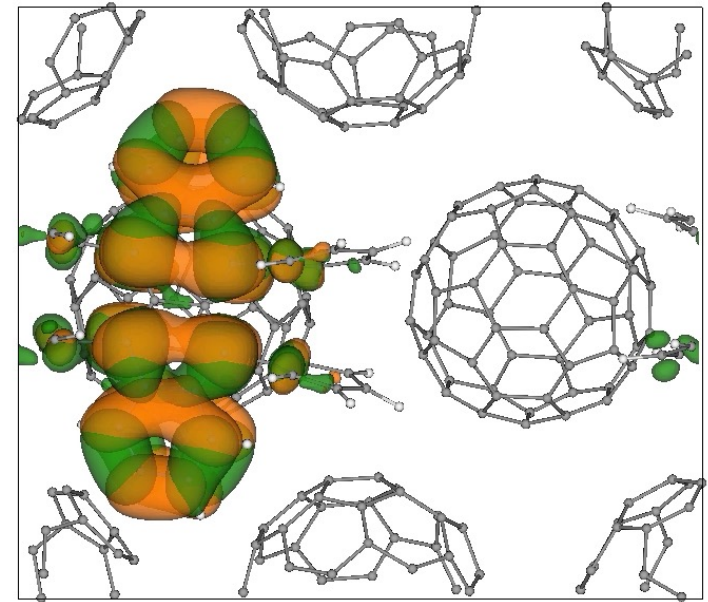
Nonadiabatic Quantum Molecular Dynamics



Appl. Phys. Lett. **98**, 113301 ('11); *ibid.* **100**, 203306 ('12); *J. Chem. Phys.* **136**, 184705 ('12); *Comput. Phys. Commun.* **184**, 1 ('13); *Appl. Phys. Lett.* **102**, 093302 ('13); *ibid.* **102**, 173301 ('13); *J. Chem. Phys.* **140**, 18A529 ('14); *IEEE Computer* **48(11)**, 33 ('15); *Sci. Rep.* **5**, 19599 ('16); *Nature Commun.* **8**, 1745 ('17); *Nano Lett.* **18**, 4653 ('18); *Nature Photon.* **13**, 425 ('19)

Zn porphyrin

Rubrene/C₆₀

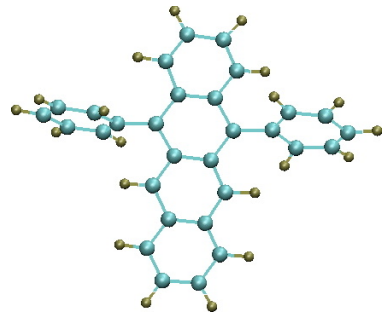


quasi-electron; quasi-hole

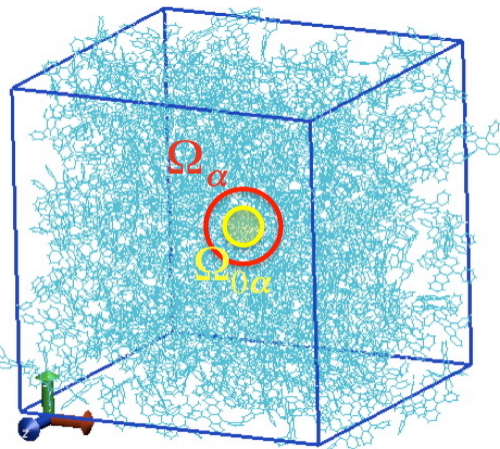
- **Excited states:** Linear-response time-dependent density functional theory [Casida, '95]
- **Interstate transitions:** Surface hopping [Tully, '90; Jaeger, Fisher & Prezhdoo, '12]

Singlet Fission in Amorphous DPT

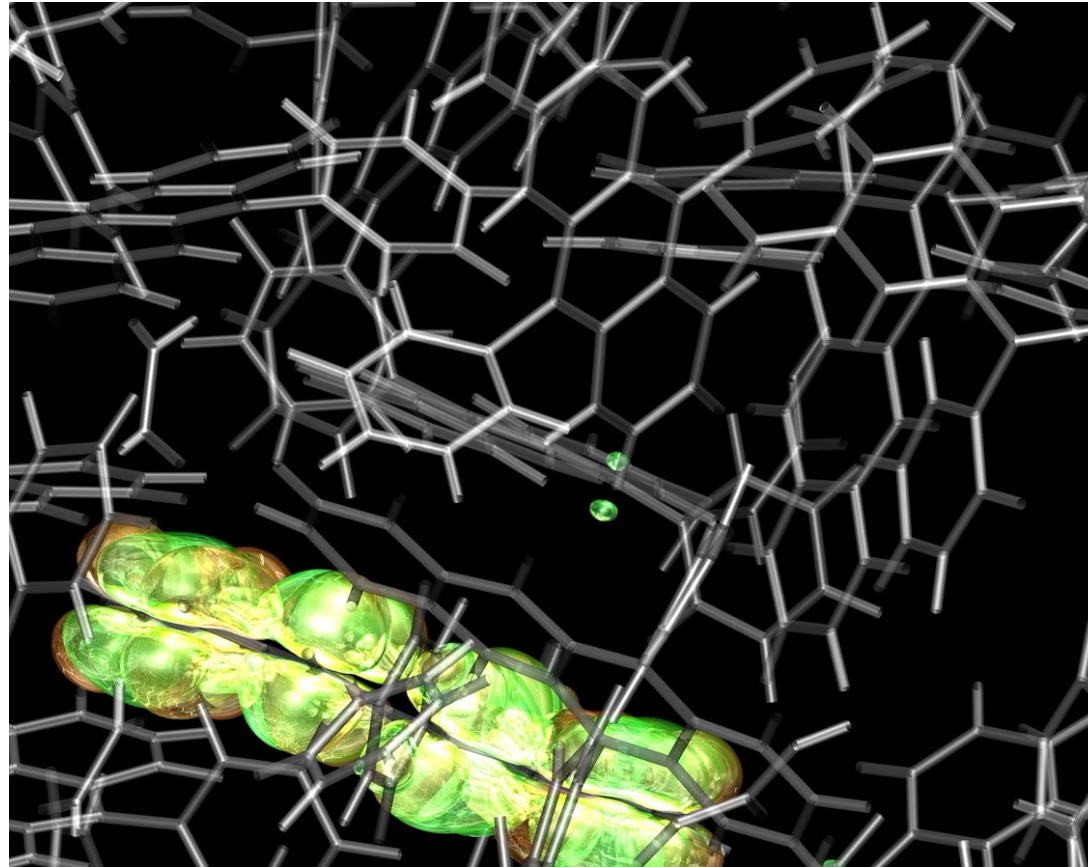
- Photo-current doubling by splitting a singlet exciton into 2 triplet excitons
- Singlet fission in mass-produced disordered organic solid → efficient low-cost solar cells
- **Experimental breakthrough:** SF found in amorphous diphenyl tetracene (DPT)



DPT molecule



Amorphous DPT



Quasi-electron

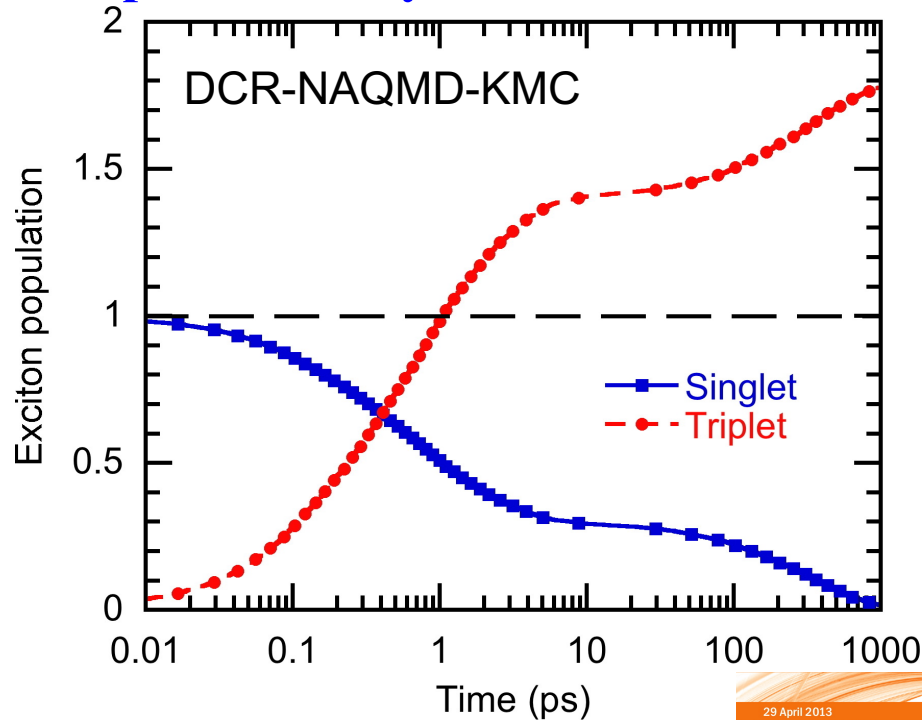
Quasi-hole

W. Mou *et al.*, *Appl. Phys. Lett.* **102**, 173301 ('13)

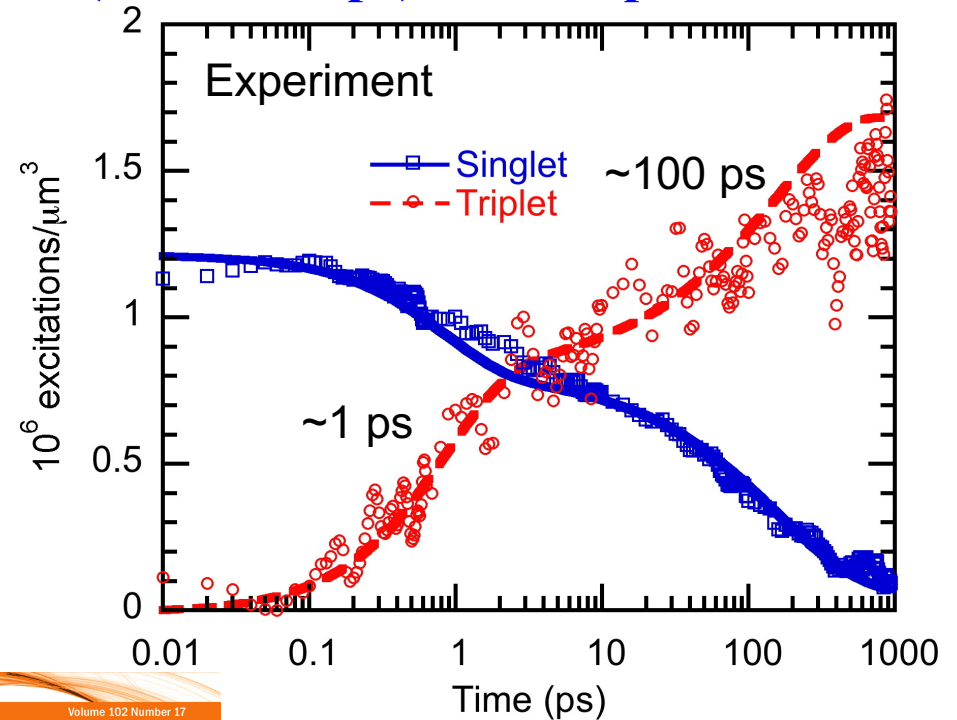
- **Divide-conquer-recombine nonadiabatic QMD** (phonon-assisted exciton dynamics) + time-dependent perturbation theory (singlet-fission rate) + kinetic Monte Carlo calculations of exciton population dynamics in **6,400-atom** amorphous DPT

DCR-NAQMD Informed Kinetic Monte Carlo

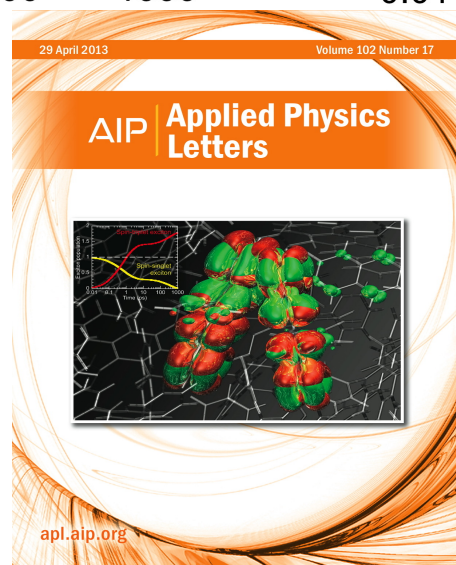
- DCR-NAQMD-KMC exciton population dynamics reproduces the experimentally observed two time-scales (~ 1 & 100 ps) in amorphous DPT



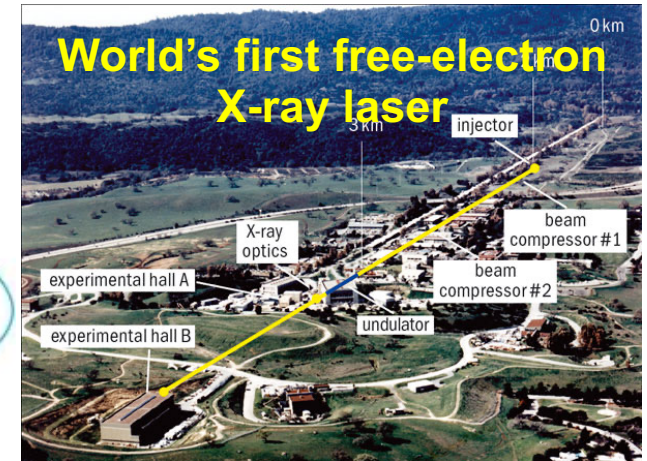
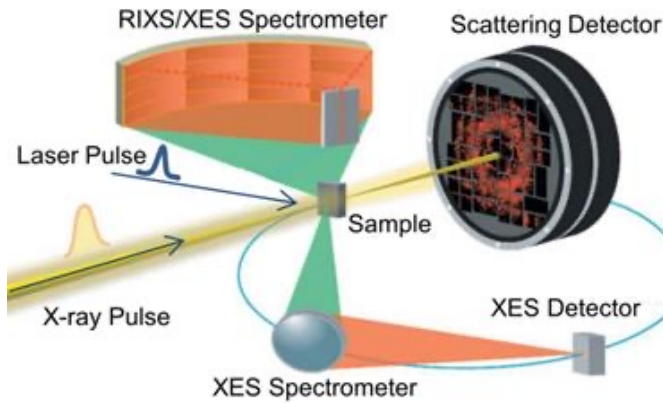
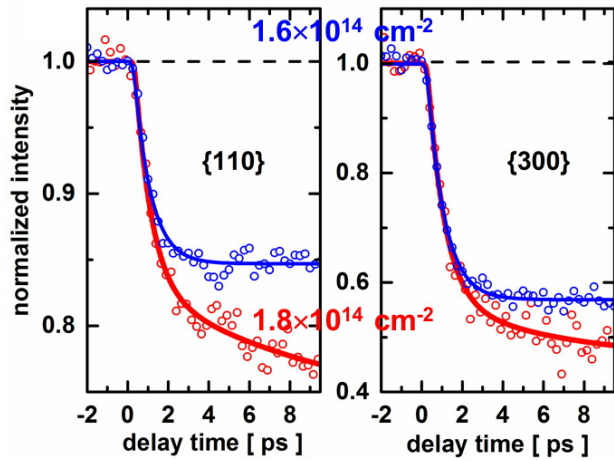
Mou *et al.*,
Appl. Phys. Lett.
100, 173301 ('13)



Roberts *et al.*,
J. Am. Chem. Soc.
134, 6388 ('12)



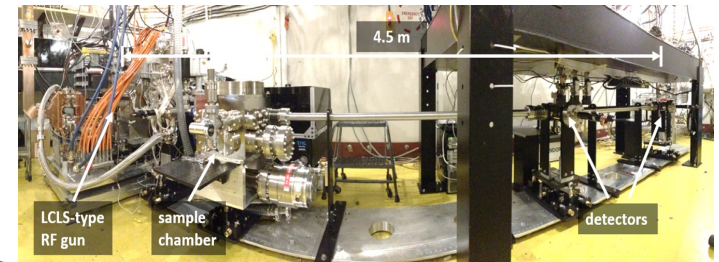
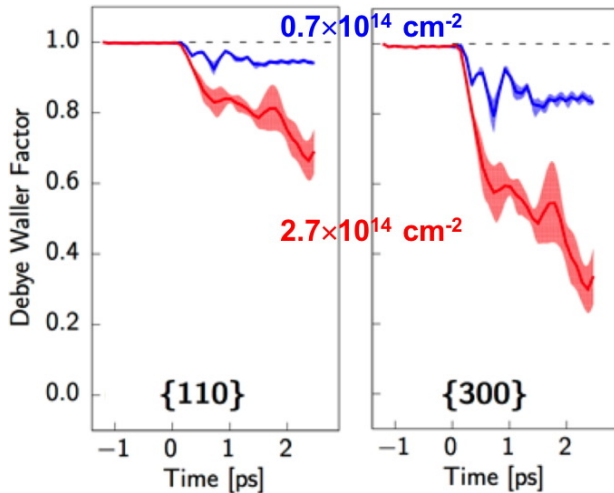
INCITE/AURORA-MAGICS-LCLS Synergy



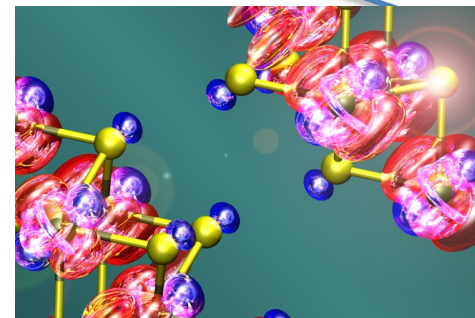
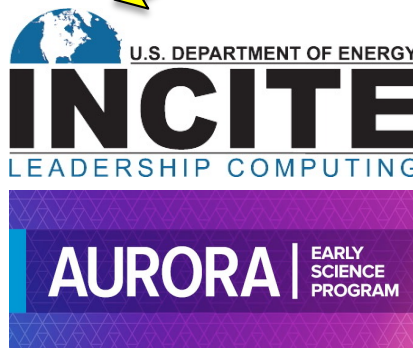
Linac Coherent Light Source

LCLS

DOE INCITE & Aurora ESP Awards

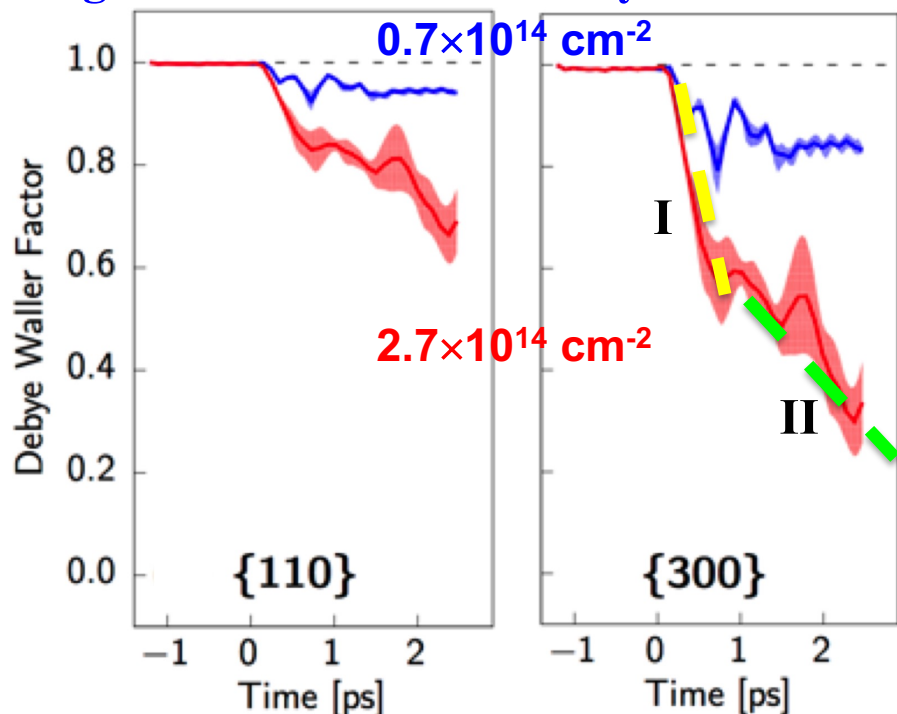


Ultrafast electron diffraction (UED) at SLAC



Strong Electron-Lattice Coupling in MoSe₂

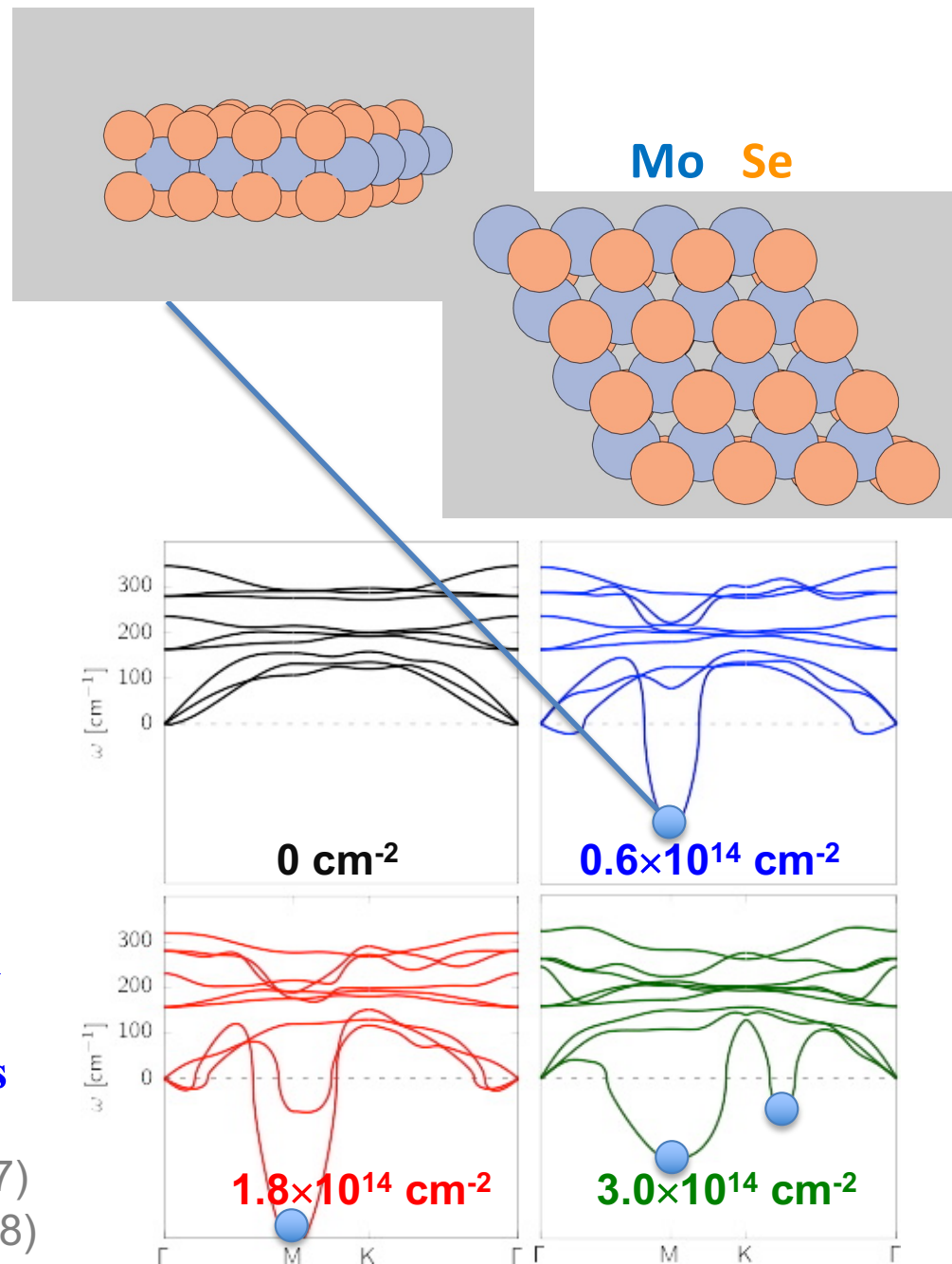
- NAQMD simulations reproduce (1) rapid photo-induced lattice dynamics & (2) mono- to bi-exponential transition at higher electron-hole density



- Rapid lattice dynamics is explained by the softening of M-point (1/2 0 0) phonon
- Bi-exponential transition is explained by the softening of additional phonon modes at higher electron-hole densities

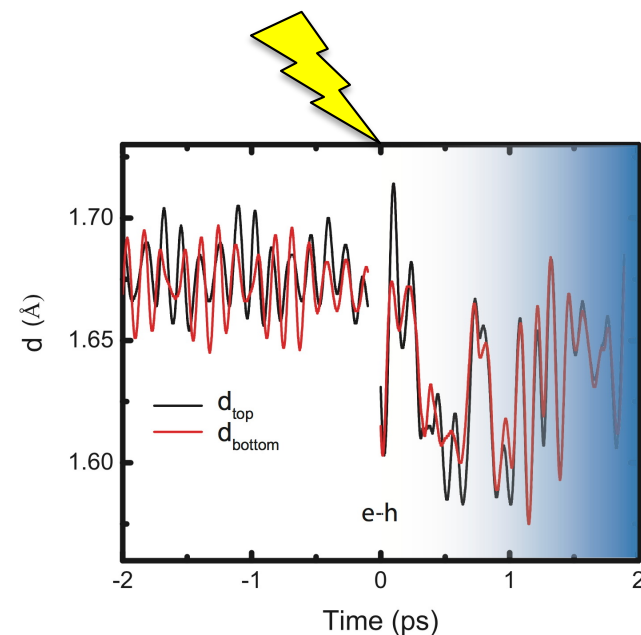
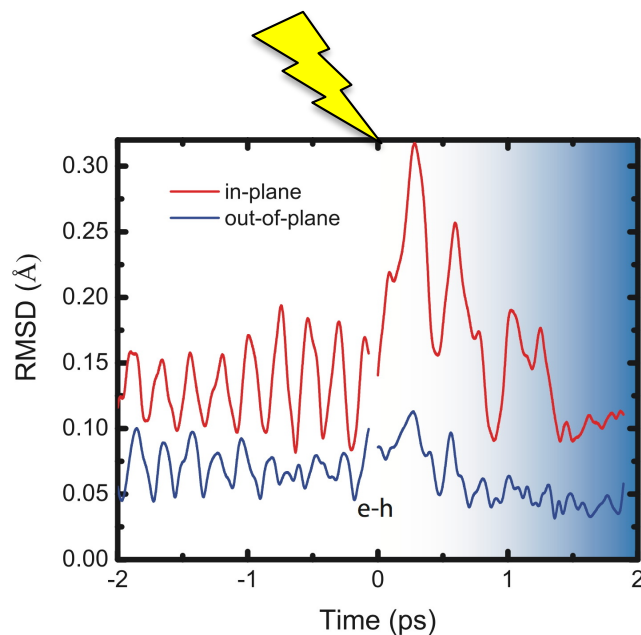
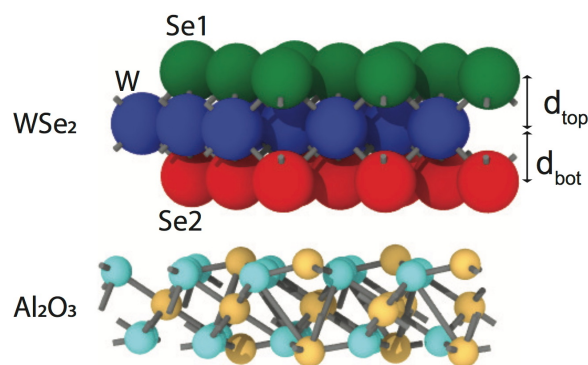
Lin *et al.*, *Nature Commun.* **8**, 1745 ('17)

Bassman *et al.*, *Nano Lett.* **18**, 4653 ('18)

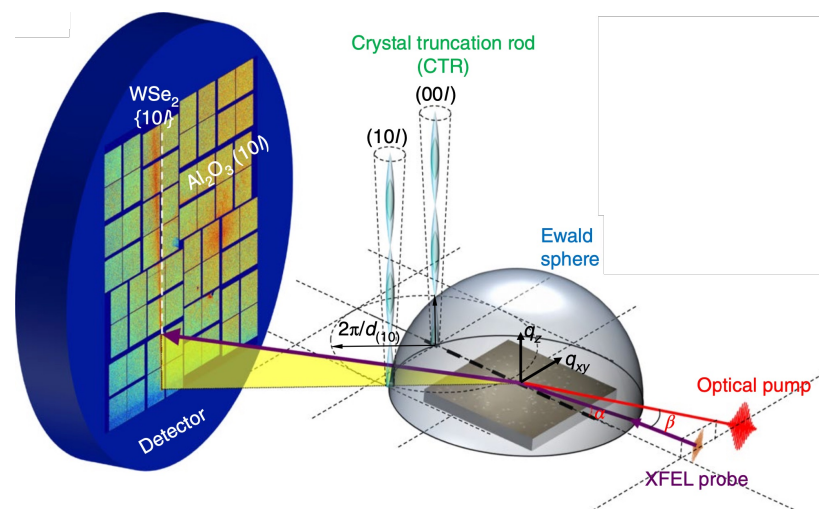


WSe₂ Monolayer on Al₂O₃ Substrate

- NAQMD simulation to study photoexcitation dynamics of WSe₂ monolayer on Al₂O₃ substrate



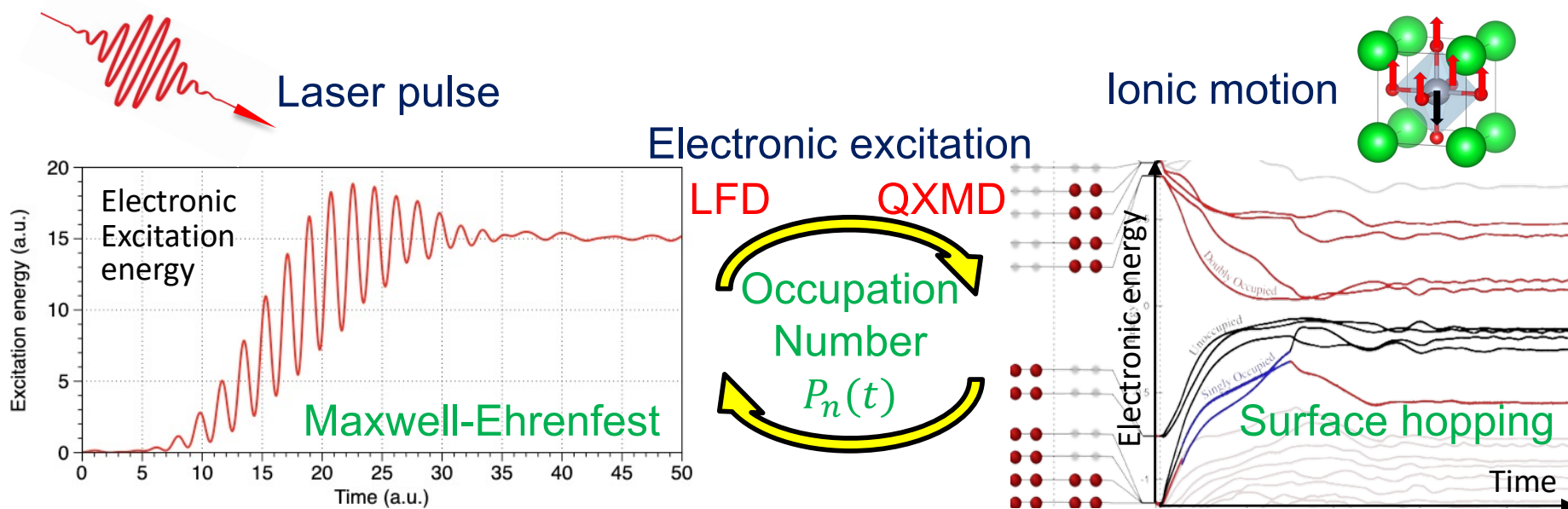
- Enhanced in-plane atomic displacements upon photoexcitation
- Photo-induced intralayer contraction of W-Se distances
- Good agreement with femtosecond surface X-ray scattering experiments at LCLS



Tung, et al., *Nature Photonics* **13**, 425 ('19)

Light-Matter Interaction: DC-MESH

- **DC-MESH (divide-&-conquer Maxwell + Ehrenfest + surface-hopping):** $O(N)$ algorithm to simulate photo-induced quantum materials dynamics
- **LFD (local field dynamics)** solves Maxwell equations for light & real-time time-dependent density functional theory (RT-TDDFT) equations for electrons to describe light-matter interaction
- **QXMD (quantum molecular dynamics with excitation)** describes nonadiabatic coupling of excited electrons & ionic motions based on surface-hopping approach [Nature Commun. 8, 1745 ('17); Nature Photon., 13, 425 ('19)]
- **LFD-QXMD handshaking** via electronic occupation numbers
- **GSLD: Globally sparse (interdomain Hartree coupling via multigrid) & locally dense (intradomain nonlocal exchange-correlation computation via BLAS) solver**



LFD Algorithm

- **Hamiltonian in the α -th domain** [Yabana, *Phys. Rev. B* **85**, 045134 ('12)]

$$\hat{h}(t, \mathbf{R}(t)) = \overbrace{\frac{1}{2} \left(\frac{\nabla}{i} + \frac{1}{c} \mathbf{A}(\mathbf{r}_\alpha, t) \right)^2 - \phi(\mathbf{r}_\alpha, t) + \hat{v}_{\text{xc}} + v_{\text{ion}}(\mathbf{r}, \mathbf{R})}^{\hat{h}_{\text{el}}(t)} + \overbrace{\Delta \dot{\mathbf{R}} \cdot \frac{\partial}{\partial \mathbf{R}} v_{\text{ion}}}^{\hat{h}_{\text{el-ion}}}$$

Electromagnetic vector & scalar potentials at the α -th domain Nonadiabatic coupling

- **Trotter expansion of time propagator**

$$\exp(-i\hat{h}\Delta_{\text{MD}}) \cong \underbrace{\exp(-i\hat{h}_{\text{el-ion}}\Delta_{\text{MD}}/2)}_{\text{QXMD}} \underbrace{\mathcal{T} \exp \left(-i \int_t^{t+\Delta_{\text{MD}}} dt \hat{h}_{\text{el}}(t) \right)}_{\text{LFD}} \underbrace{\exp(-i\hat{h}_{\text{el-ion}}\Delta_{\text{MD}}/2)}_{\text{QXMD}}$$

- **Self-consistent propagator** [Sato, *J. Chem. Phys.* **143**, 224116 ('15); Lian, *Adv. Theo. Sim.* **1**, 1800055 ('18)]

$$\mathcal{T} \exp \left(-i \int_t^{t+\Delta_{\text{MD}}} dt \hat{h}_{\text{el}}(t) \right) \cong \prod_{n=1}^{N_{\text{QD}}=\Delta_{\text{MD}}/\Delta_{\text{QD}}} \exp \left(-i\Delta_{\text{QD}} \hat{h}_{\text{el}} \left(t + \left(n - \frac{1}{2} \right) \Delta_{\text{QD}} \right) \right)$$

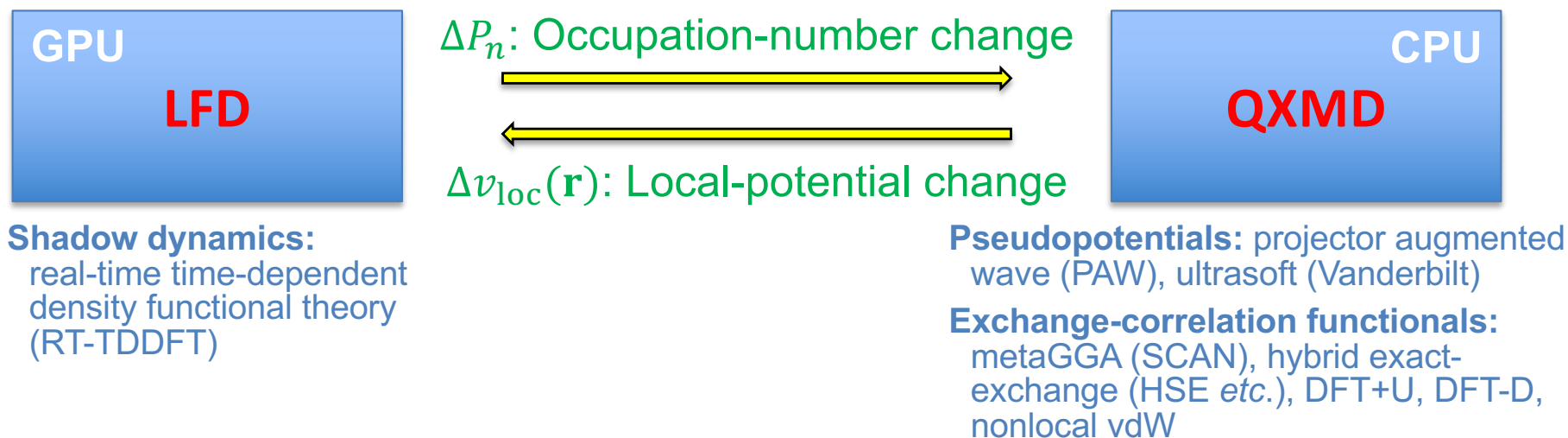
- **Nonlocal exchange-correlation propagator** [Vlcek, *J. Chem. Phys.* **150**, 184118 ('19)]

$$\exp(-i\Delta_{\text{QD}}\hat{h}_{\text{el}}) \cong \frac{1 - i\hat{v}_{\text{nl}}\Delta_{\text{QD}}/2}{\| (1 - i\hat{v}_{\text{nl}}\Delta_{\text{QD}}/2) |\psi_n(t)\rangle \|} \exp(-i\Delta_{\text{QD}}\hat{h}_{\text{loc}}) \frac{1 - i\hat{v}_{\text{nl}}\Delta_{\text{QD}}/2}{\| (1 - i\hat{v}_{\text{nl}}\Delta_{\text{QD}}/2) |\psi_n(t)\rangle \|}$$

$$\hat{v}_{\text{xc}} = \underbrace{\hat{v}_{\text{loc}}}_{\text{local}} + \underbrace{\hat{v}_{\text{nl}}}_{\text{nonlocal}}; \quad \hat{h}_{\text{el}} = \hat{h}_{\text{loc}} + \hat{v}_{\text{nl}}$$

Reduced-Communication Shadow Dynamics

- At each molecular-dynamics step, LFD informs QXMD of occupation-number change due to light-electron & electron-electron interactions
- QXMD performs excited-state quantum molecular dynamics & informs LFD of local-potential change for the next $N_{\text{QD}} (= \Delta_{\text{MD}}/\Delta_{\text{QD}})$ quantum-dynamics steps
- “Shadow” electronic wave functions in LFD are resident on GPU, while QXMD wave functions on CPU, to minimize CPU-GPU data transfers

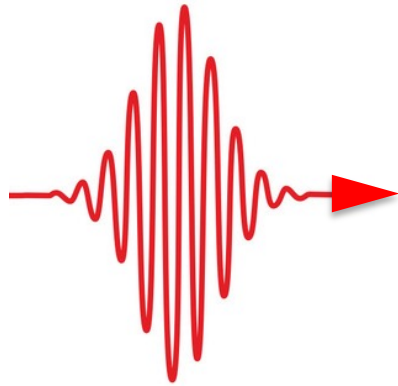


- **BLASified nonlocal propagator in LFD on GPU**

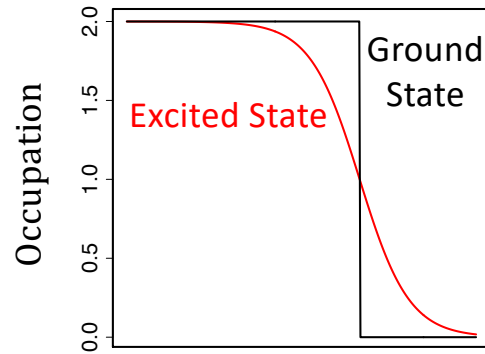
Real-time scissor approximation [Wang, *J. Phys. Condens. Mat.* **31**, 214002 ('19)]

$$\hat{v}_{\text{nl}}|\psi_n(t)\rangle \cong \Delta_{\text{sci}} \sum_{m \geq \text{LUMO}} |\psi_m\rangle \langle \psi_m | \psi_n(t)\rangle$$

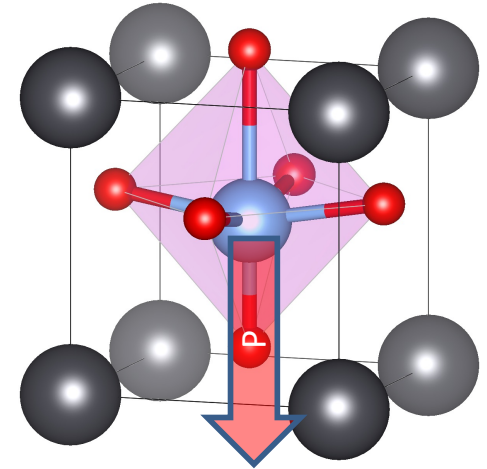
Multiscaling from DC-MESH to XS-NNQMD



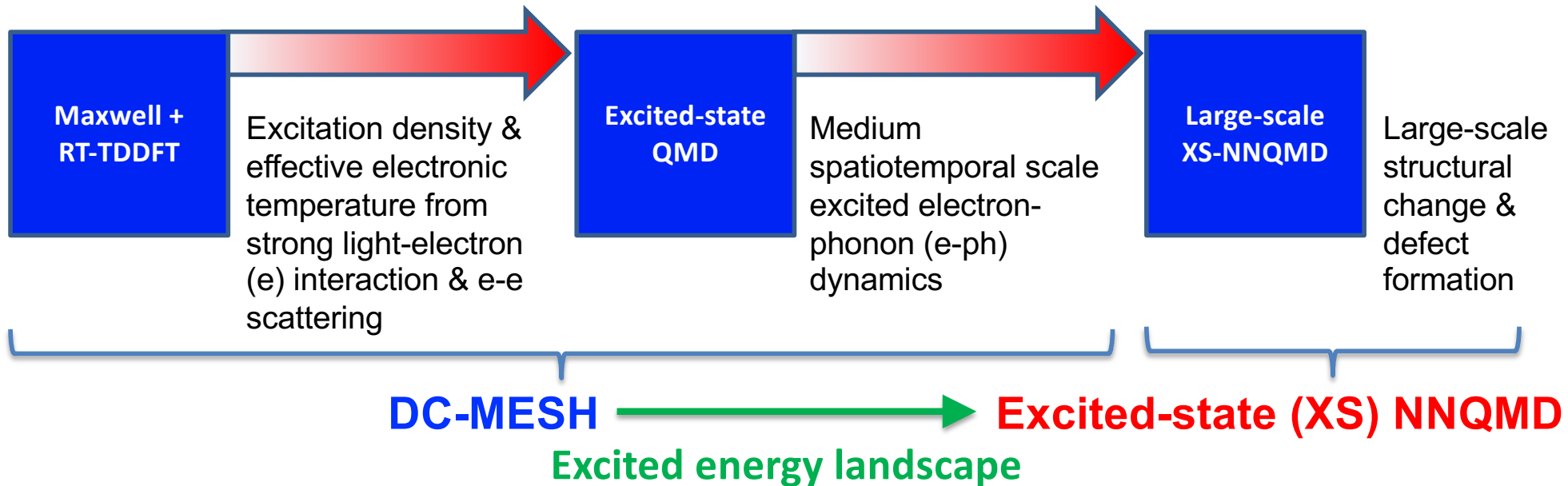
Electronic excitation by ultrafast laser pulse



Change in electronic occupation due to electron-electron interaction



Change in polarization dynamics due to electron-ion interaction

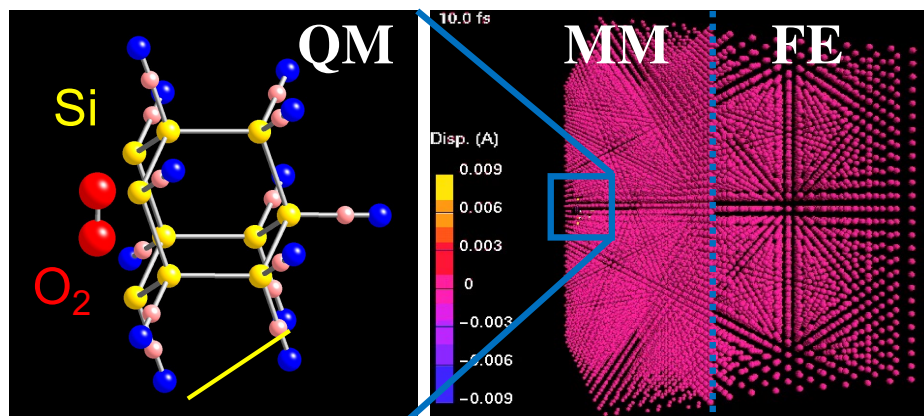
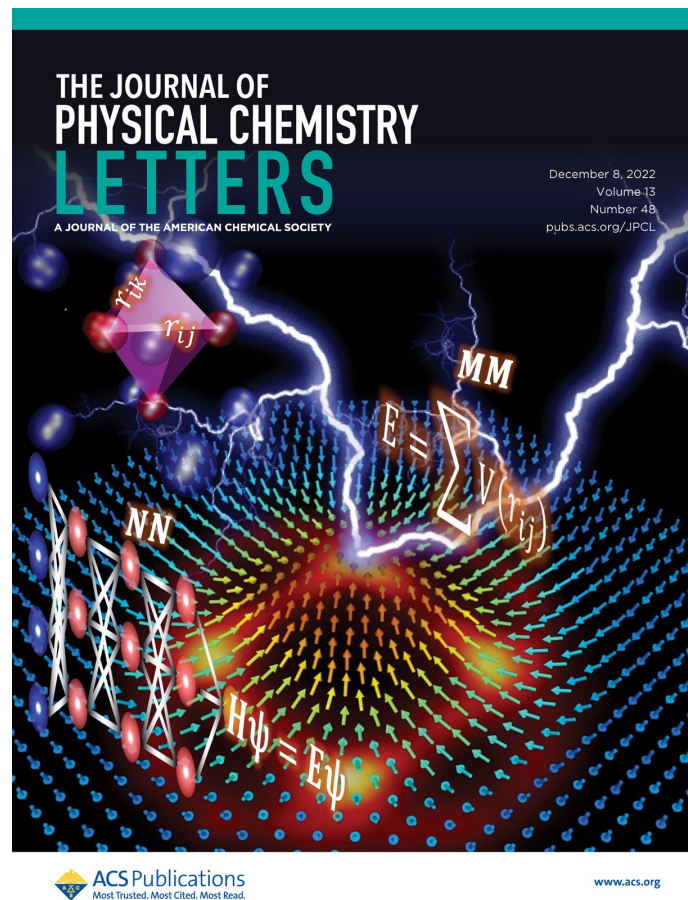


Multiscale QM/MM → NN/MM

- **Multiscale quantum challenge:** Complex response of ferroelectric topological defects to external stimuli encompasses picosecond-to-nanosecond time & nanometer-to-micrometer length scales
- **QM/MM:** Overcame the challenge taking cue from multiscale quantum-mechanics (QM)/molecular mechanics (MM) approach (2013 Nobel chemistry prize)

Warshel, *Angew. Chem.* **53**, 10020 ('14)

- **NN/MM:** NNQMD for ferroelectric (PbTiO₃: PTO) embedded in MM for paraelectric (SrTiO₃: STO) to apply appropriate strain boundary condition

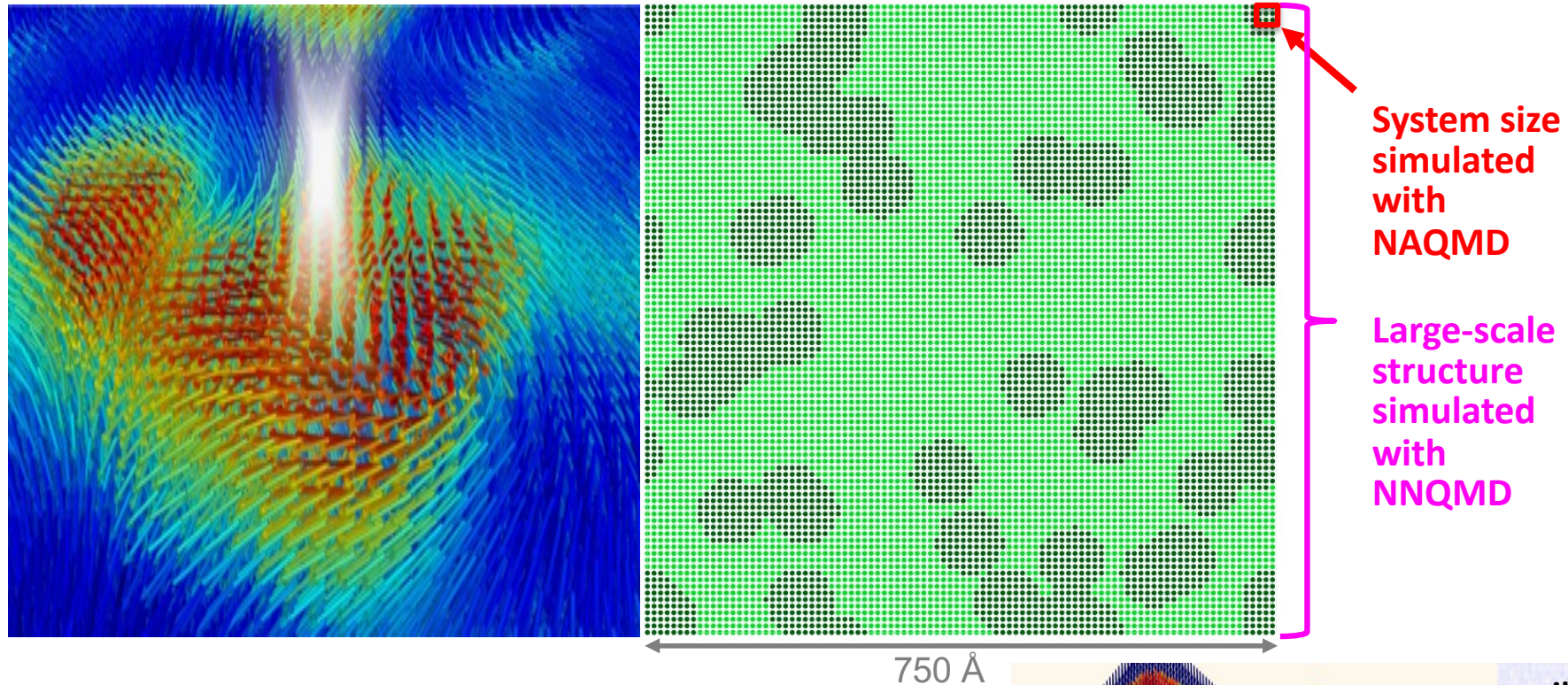


QM/MM/FE (finite-element method)

Ogata *et al*, *Comput. Phys. Commun.* **138**, 143 ('01)

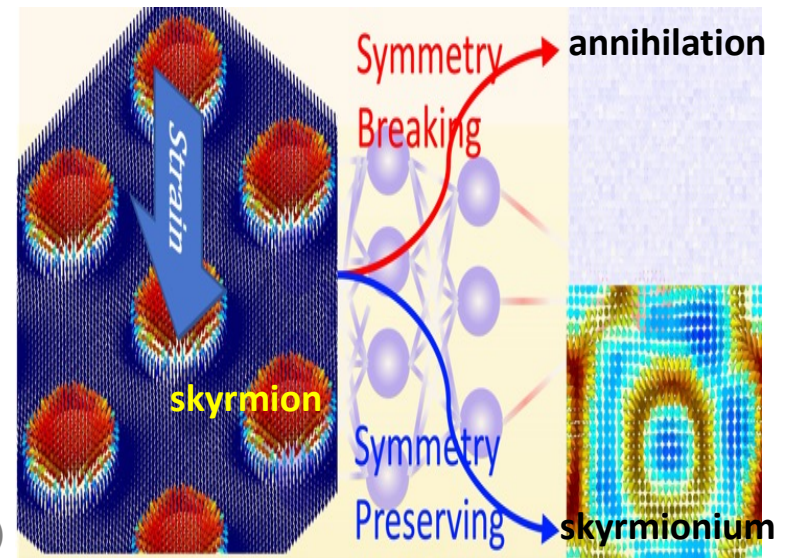
Linker *et al.*, *J. Phys. Chem. Lett.*
13, 11335 ('22)

Application: Ferroelectric Opto-Topotronics



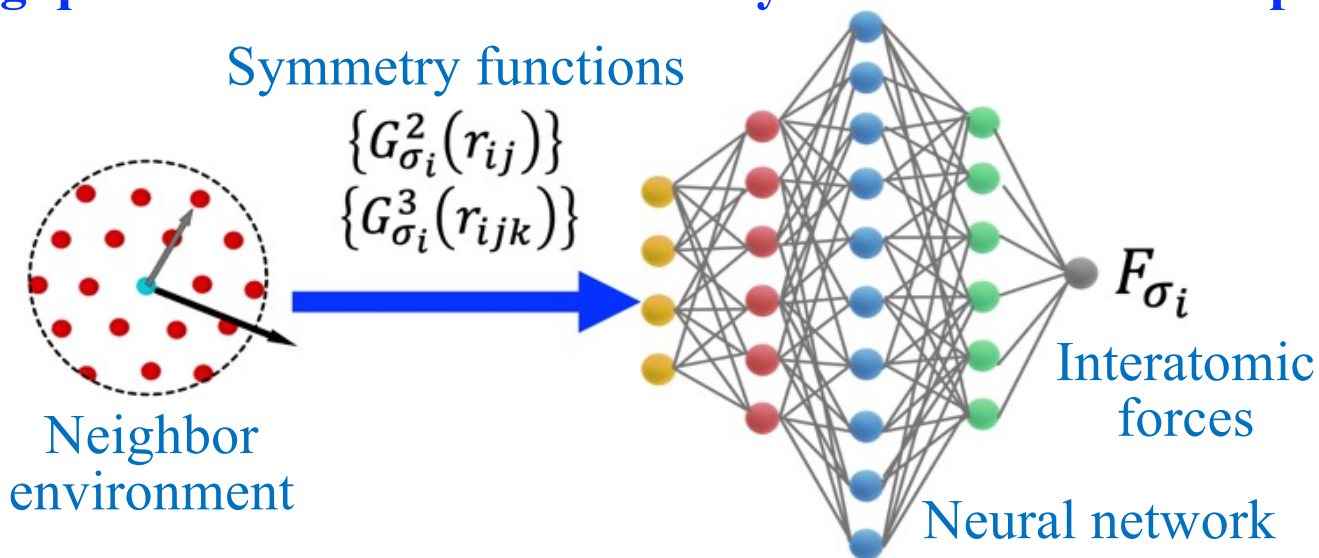
- Quantized ferroelectric topology is protected against thermal noise → future ultralow-power opto-electronics applications
- Billion-atom NNQMD revealed photo-induced topological phase-transition dynamics (*cf.* Kibble-Zurek mechanism in cosmology)
- Symmetry-controlled skyrmion-to-skyrmionium* switching *Composite of skyrmions with opposite topological charges

Linker *et al.*, *Science Adv.* **8**, eabk2625 ('22); *JPCL* **13**, 11335 ('22)

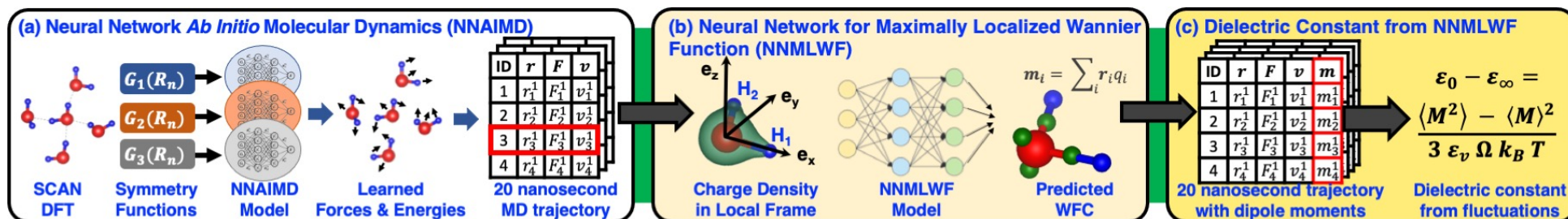


Neural-Network Quantum Molecular Dynamics

- NNQMD@scale could revolutionize atomistic modeling of materials, providing quantum-mechanical accuracy at a fraction of computational cost

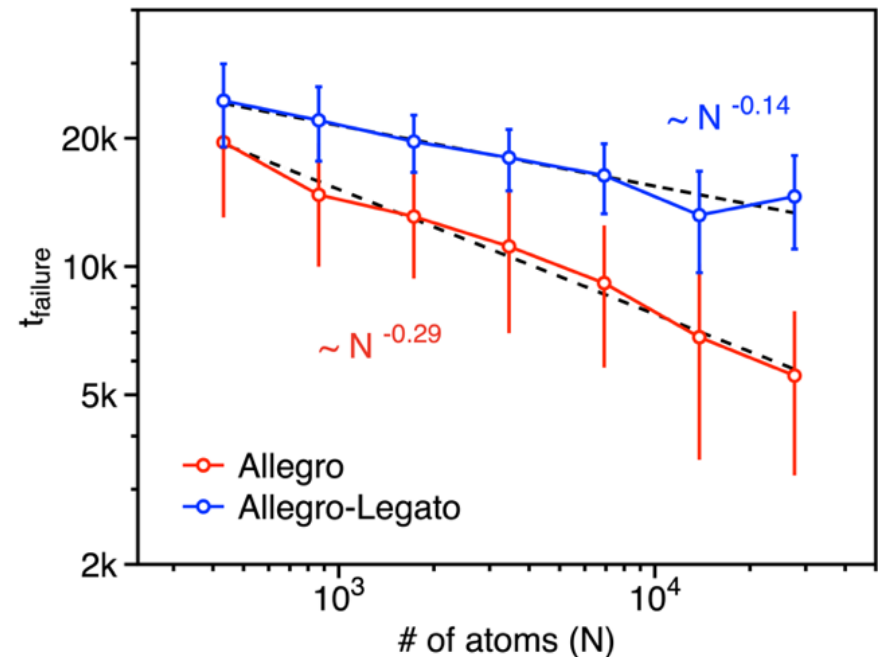
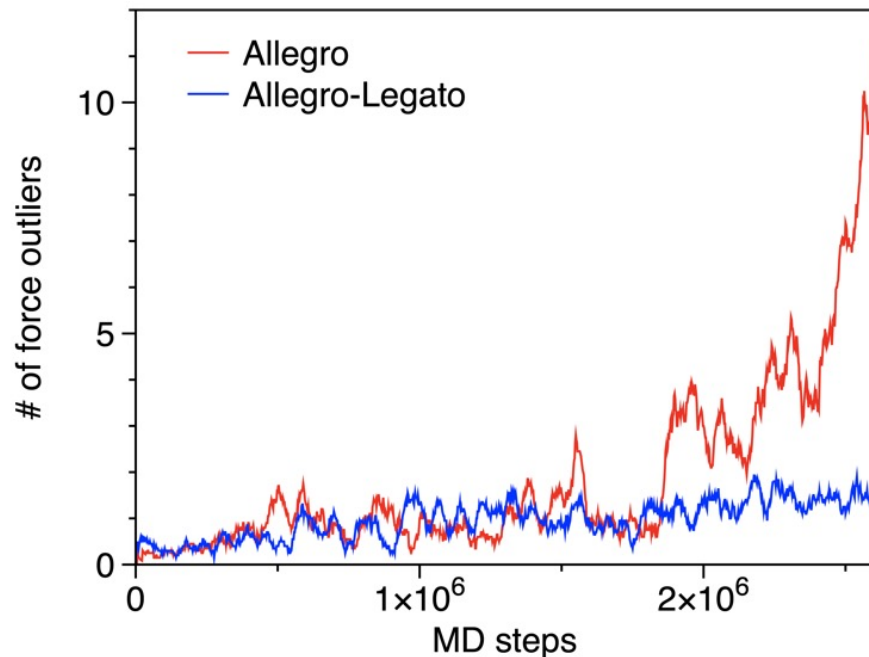


- Neural networks predict: (1) atomic forces for performing MD simulations; & (2) maximally-localized Wannier-function (MLWF) centers for computing quantum properties like electronic dipoles



Fast & Robust NNQMD: Allegro-Legato

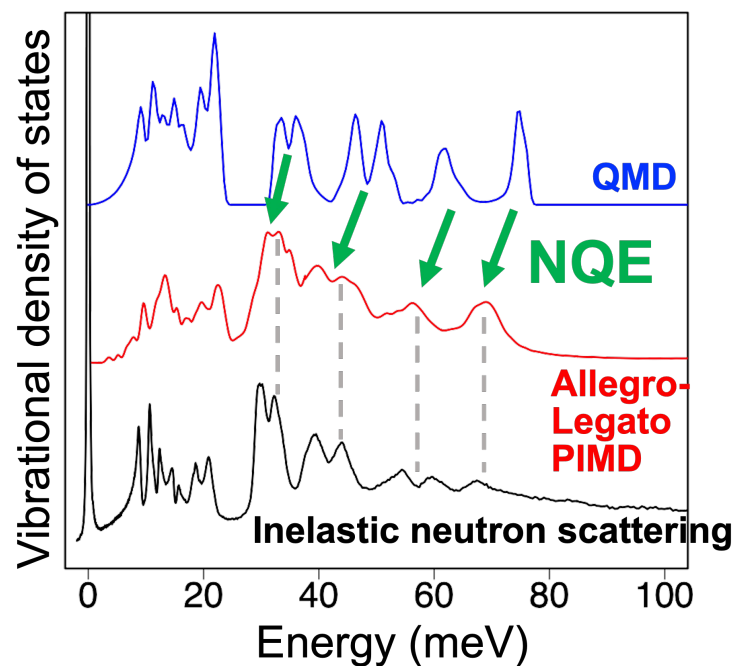
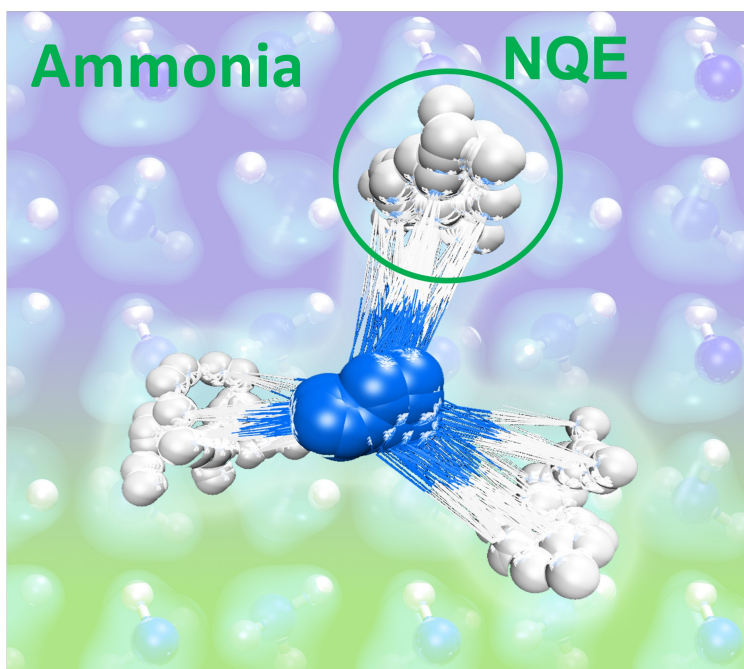
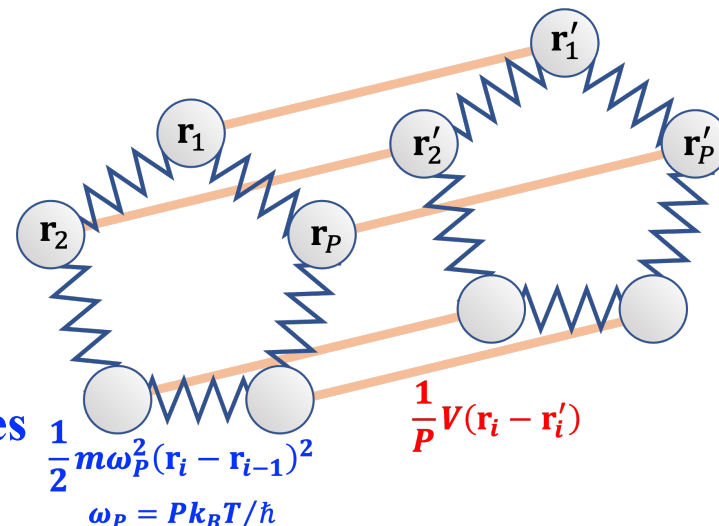
- **Allegro (fast) NNQMD: State-of-the-art *accuracy & speed* founded on group-theoretical equivariance & local descriptors** [Musaelian *et al.*, *Nat. Commun.* **14**, 579 ('23)]
- **Fidelity-scaling problem: On massively parallel computers, growing number of unphysical (adversarial) force predictions prohibits simulations involving larger numbers of atoms for longer times**
- **Allegro-Legato (fast and “smooth”): *Sharpness aware minimization (SAM)* enhances the *robustness* of Allegro through improved smoothness of loss landscape**
 $\mathbf{w}_* = \operatorname{argmin}_{\mathbf{w}} [L(\mathbf{w}) + \max_{\|\epsilon\|_2 \leq \rho} \{L(\mathbf{w} + \epsilon) - L(\mathbf{w})\}]$ (L : loss; \mathbf{w} : model parameters)
- **Elongated time-to-failure scaling, $t_{\text{failure}} = O(N^{-\beta})$, without sacrificing accuracy or speed, thereby achieving spectroscopically stable long-time Hamiltonian trajectory**



H. Ibayashi *et al.*, *ISC 2023* (arXiv: 2303.08169)

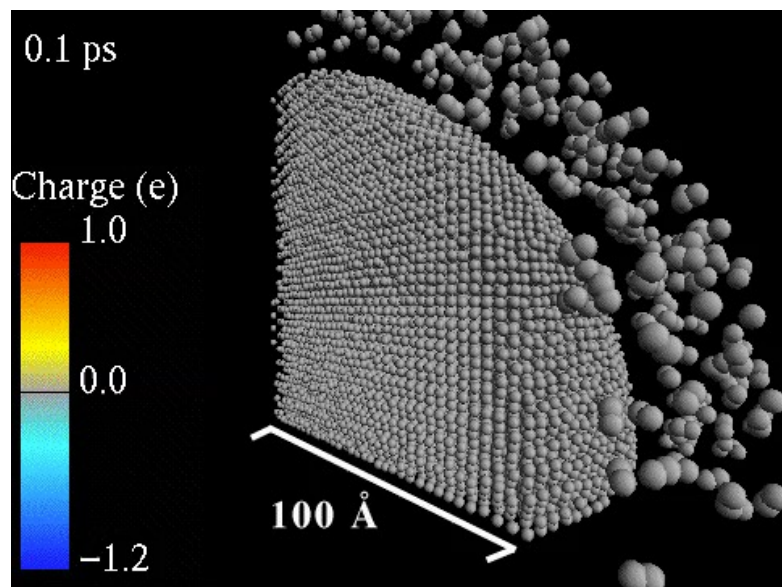
Nuclear-Quantum NNQMD

- **Allegro-Legato-PIMD**: Incorporate nuclear quantum effect (NQE) through path-integral molecular dynamics (PIMD)
- NNQMD trained by QMD achieves the required large number (P) of replicas at low temperature & long-time Hamiltonian dynamics to resolve fine vibrational structures
- NQE down-shifts inter-molecular vibrational modes in ammonia to explain high-resolution inelastic neutron scattering experiments



Charge-Transfer NNQMD

- Incorporated charge transfer for accurately describing chemical reactions through charge equilibration (QEq) in reactive molecular dynamics (RMD)



Campbell, *Phys. Rev. Lett.* **82**, 4866 ('99)

$$q_*^N = \operatorname{argmin}_{q^N} E_{\text{Coulomb}}(\mathbf{r}^N, q^N) \text{ s.t. } \sum_i q_i = 0$$

$$E_{\text{Coulomb}}(\mathbf{r}^N, q^N) = \sum_i \chi_i q_i + \frac{1}{2} \sum_{i,j} q_i H(r_{ij}) q_j$$

Electronegativity

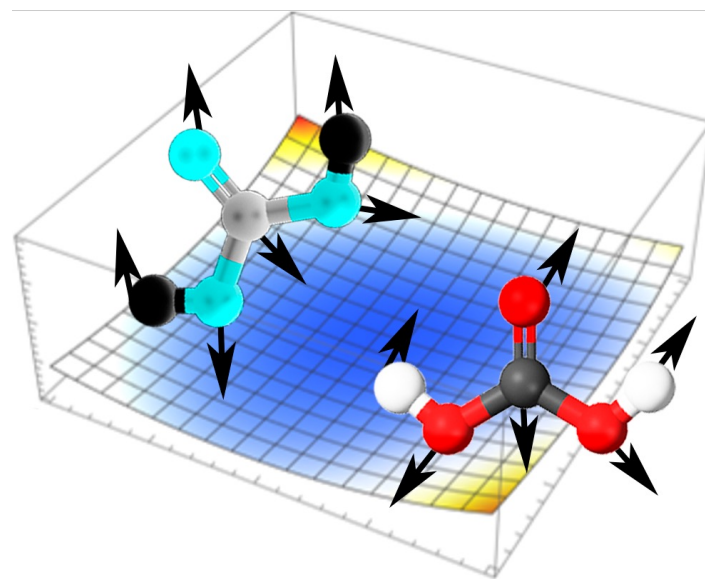
Coulombic interaction

Rappe, *J. Phys. Chem.* **95**, 3358 ('91); van Duin, *J. Phys. Chem. A* **105**, 9396 ('01)
Nakano, *Comput. Phys. Commun.* **104**, 59 ('97); Nomura, *ibid.* **192**, 91 ('15)

- 4G NNQMD: Separate neural network to predict electronegativity**

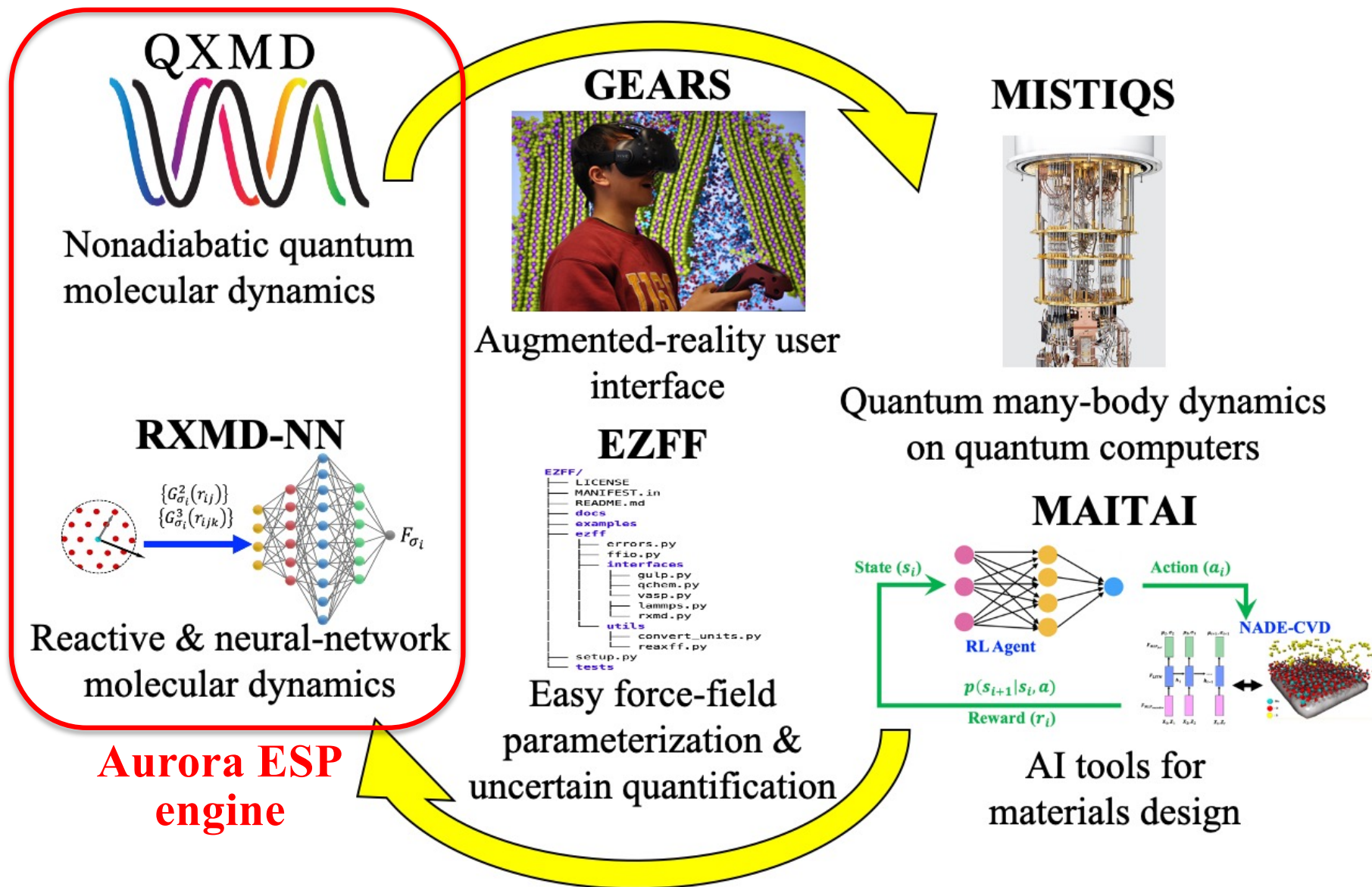
Ko, *Nat. Commun.* **12**, 398 ('21)

- Allegro-Legato-4G: Allegro-Legato to achieve smooth loss landscape & robust long-time dynamics in 4G-NNQMD**



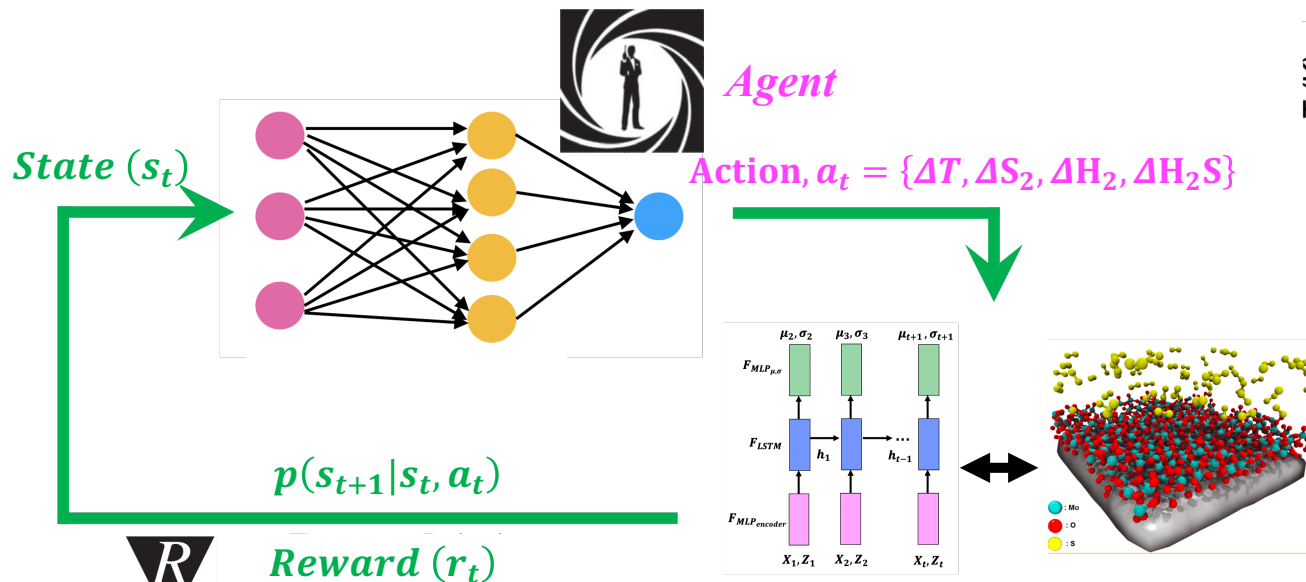
AIQ-XMaS Software Suite

AI & Quantum-Computing Enabled Exa Quantum Materials Simulator



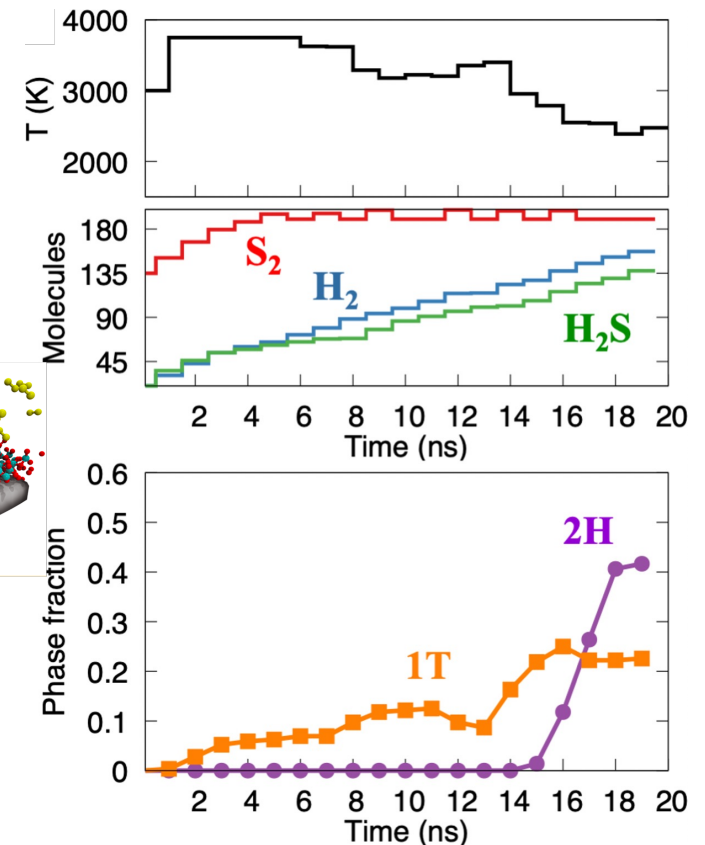
Reinforcement Learning for Growth

- In a manner AI plays a board game of Go, use reinforcement learning (RL) to design optimal growth conditions (e.g., temperature & gas-pressure control) to achieve desired properties such as minimal defect density
- AI model combines:
 - RL agent to design actions
 - Neural network-based dynamic model trained by reactive molecular-dynamics (RMD) to predict new states

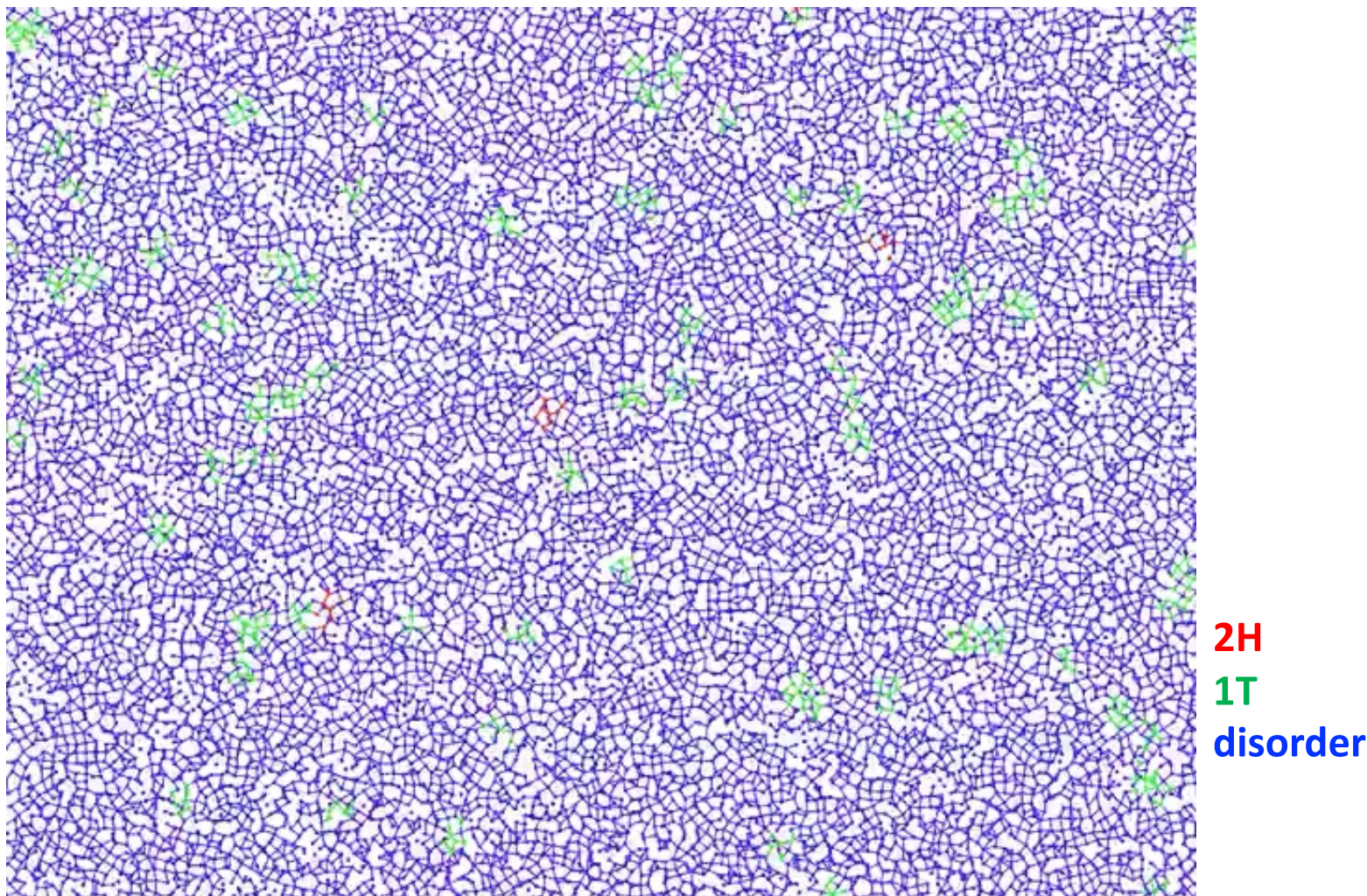


Rajak et al.,
npj Comput. Mater. 7, 108 ('21)

cf. Sgroi et al., *Phys. Rev. Lett.* 126, 020601 ('21)



Computational Synthesis of MoS₂



- Found novel transformation pathways to the **stable 2H phase** via the **metastable 1T phase** during chemical vapor deposition (CVD) growth of MoS₂

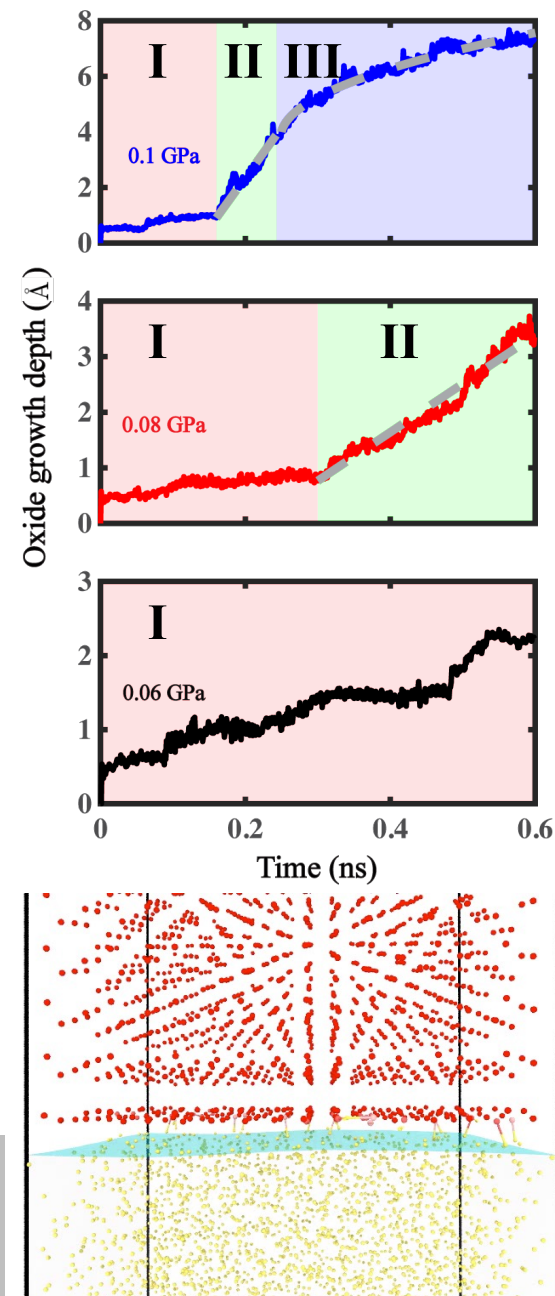
S. Hong *et al.*, *J. Phys. Chem. Lett.* **10**, 2739 ('19)

Active Control of TMDC Oxidation

- Transition-metal dichalcogenide (TMDC) semiconductors form the basis of future low-power two-dimensional (2D) electronics
- Oxidation of TMDC was proposed as a scalable synthetic pathway to critical semiconductor-insulator interfaces [Jo, Yang *et al.*, *Nano Lett.* **20**, 8592 ('20); Illarionov, *Nat. Commun.* **11**, 3385 ('20); Liu, *IEEE T. Electron. Dev.*, early edition ('23)]
- Active oxidation (enhanced by plasma, ultraviolet, laser & pressure) is being explored for controlled oxidation of TMDC [Lai, *Nanoscale* **10**, 18758 ('18); Reidy, *arXiv:2211.16789* ('23)]
- Reactive molecular dynamics simulations reveal pressure control of oxidation stages during oxidation of ZrS₂: (I) layer-by-layer to (II) reaction-controlled to (III) diffusion-controlled

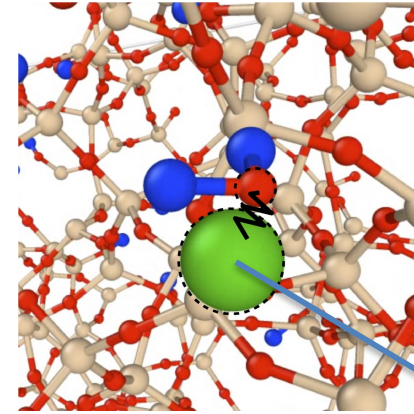
Jo, Yang *et al.*, *Nano Lett.* **20**, 8592 ('20)
Yang, *ACS Nano*, accepted ('23)

○
Zr (non-bonded)
Zr (bonded)



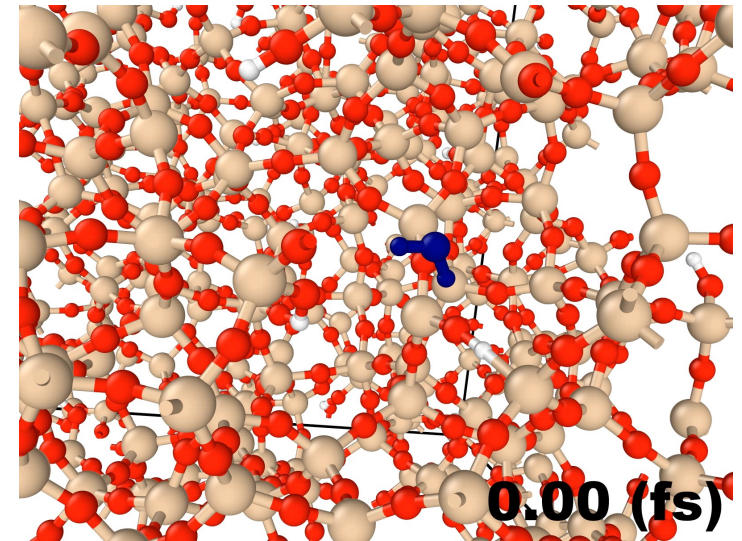
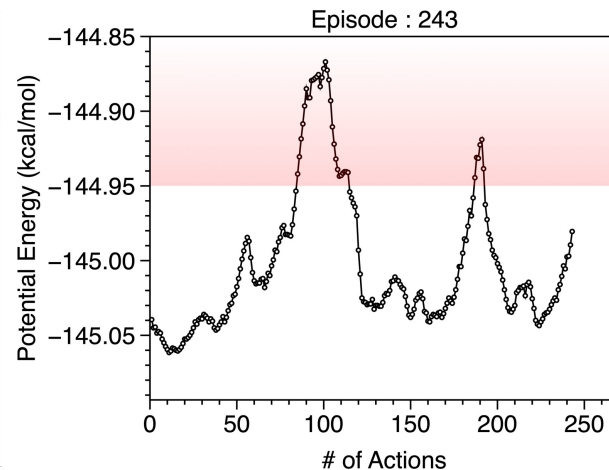
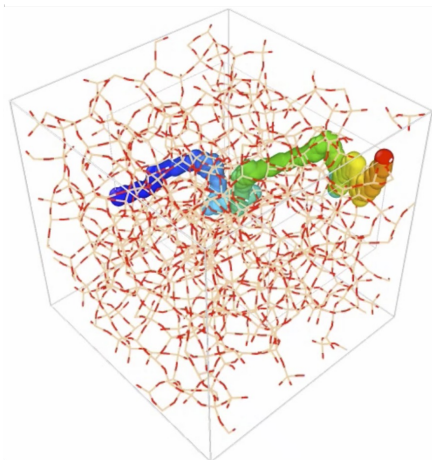
Reinforcement Learning for Long-Time Dynamics

- Reinforcement learning agents autonomously discover low-activation-barrier migration pathways to study long-time dynamics
- Multiple agents share experience using an asynchronously updated replay buffer [Silver *et al.*, *Nature* **529**, 484 ('16)]
- Estimate migration time based on the transition-state theory:



Agent

$$t_{\text{migration}} = \sum_{i \in \{\text{activation events}\}} \frac{\hbar}{k_B T} \exp\left(\frac{E_i^{\text{activation}}}{k_B T}\right)$$



2-seconds trajectory

Low-activation-barrier H₂O migration pathway in SiO₂
Application to CO₂ storage through mineralization in wet silicates

Quantum Computing (QC) for Science

Quantum computing utilizes quantum properties such as superposition & entanglement for computation

- U.S. Congress (Dec. 21, '18) signed National Quantum Initiative Act to ensure leadership in quantum computing & its applications

- Quantum supremacy demonstrated by Google

F. Arute, *Nature* **574**, 505 ('19)

- Quantum computing for science:
Universal simulator of quantum many-body systems

R. P. Feynman, *Int. J. Theo. Phys.* **21**, 467 ('82);

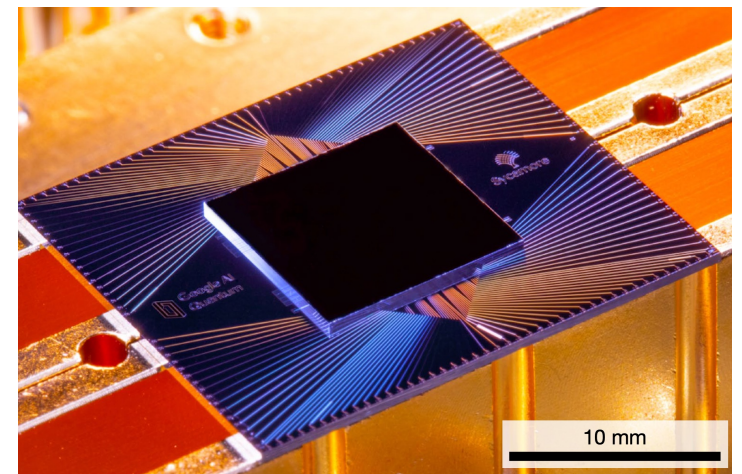
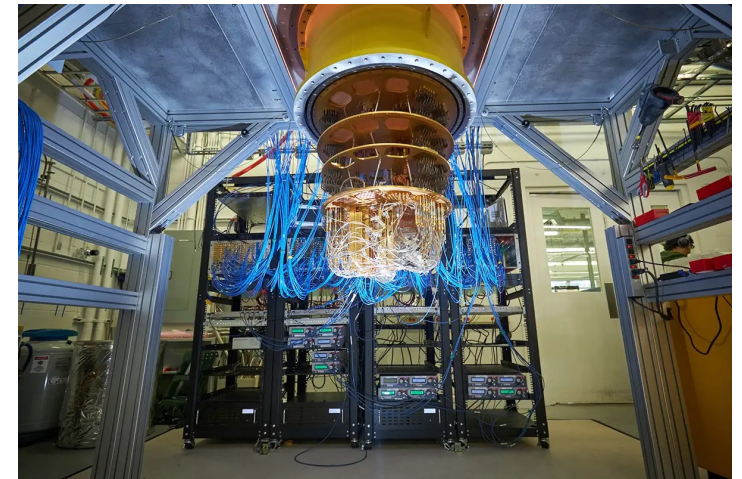
S. Lloyd, *Science* **273**, 1073 ('96)

- Success in simulating *static* properties of quantum systems (*i.e.*, ground-state energy of small molecules)

A. Aspuru-Guzik *et al.*, *Science* **309**, 1704 ('05)

- Challenge: Simulate quantum many-body *dynamics* on current-to-near-future noisy intermediate-scale quantum (NISQ) computers

J. Preskill, *Quantum* **2**, 79 ('18)

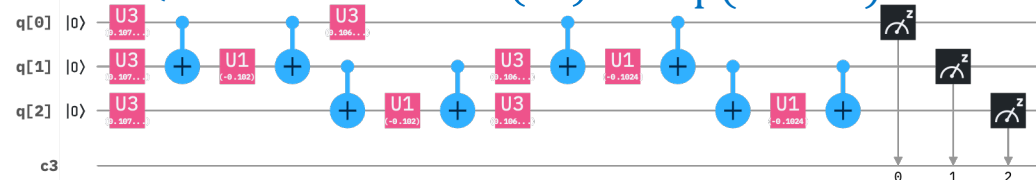
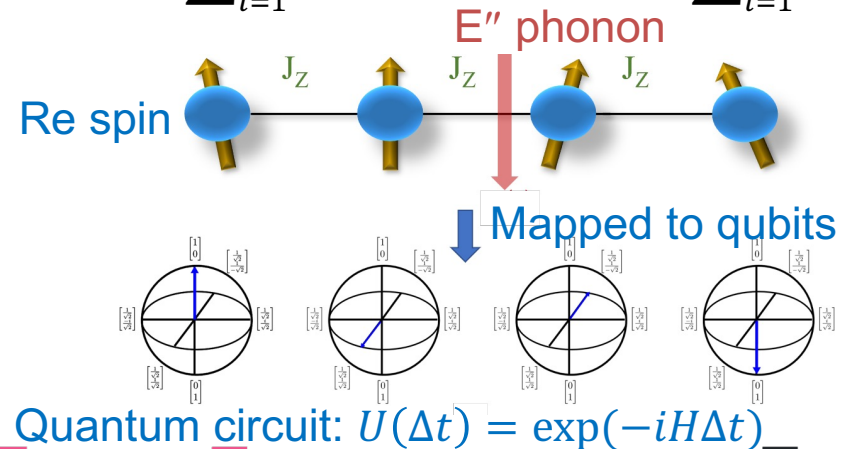


54-qubit Google Sycamore

Quantum Computing of Magnetism

- Simulated quantum many-body dynamics on IBM's Q16 Melbourne & Rigetti's Aspen quantum processors
- Electromagnetic-field control of quantum states in a chain of rhenium-magnets in MoSe₂ monolayer to realize desired material properties on demand, thereby pushing the envelope of “quantum materials science”

$$H(t) = -J_Z \sum_{i=1}^{N-1} \sigma_Z^i \sigma_Z^{i+1} - \epsilon_{ph} \sin(\omega_{ph} t) \sum_{i=1}^N \sigma_X^i$$

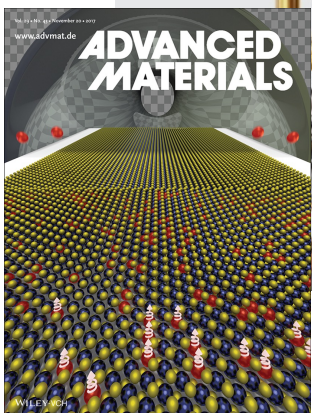
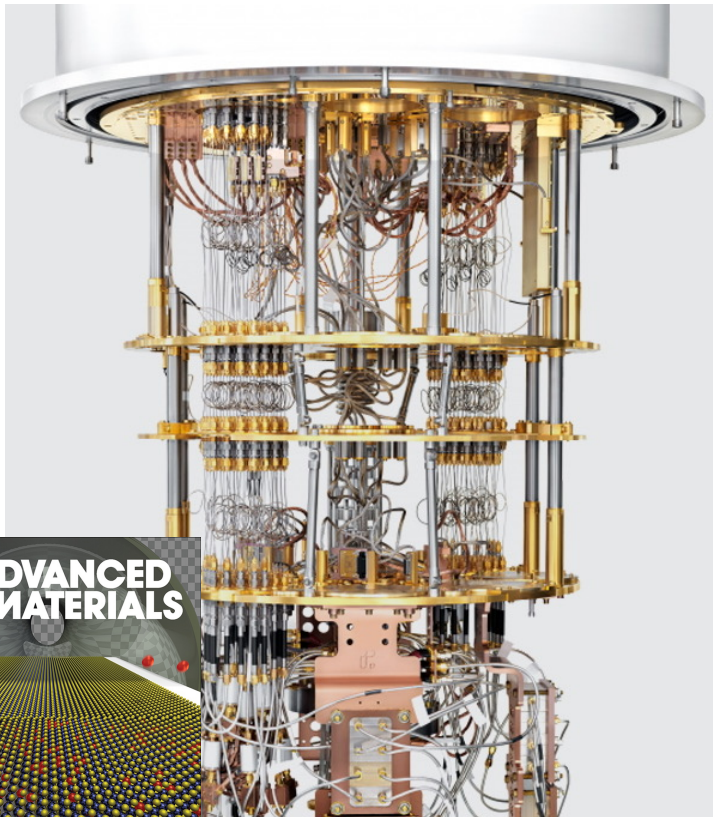


Quantum program

```

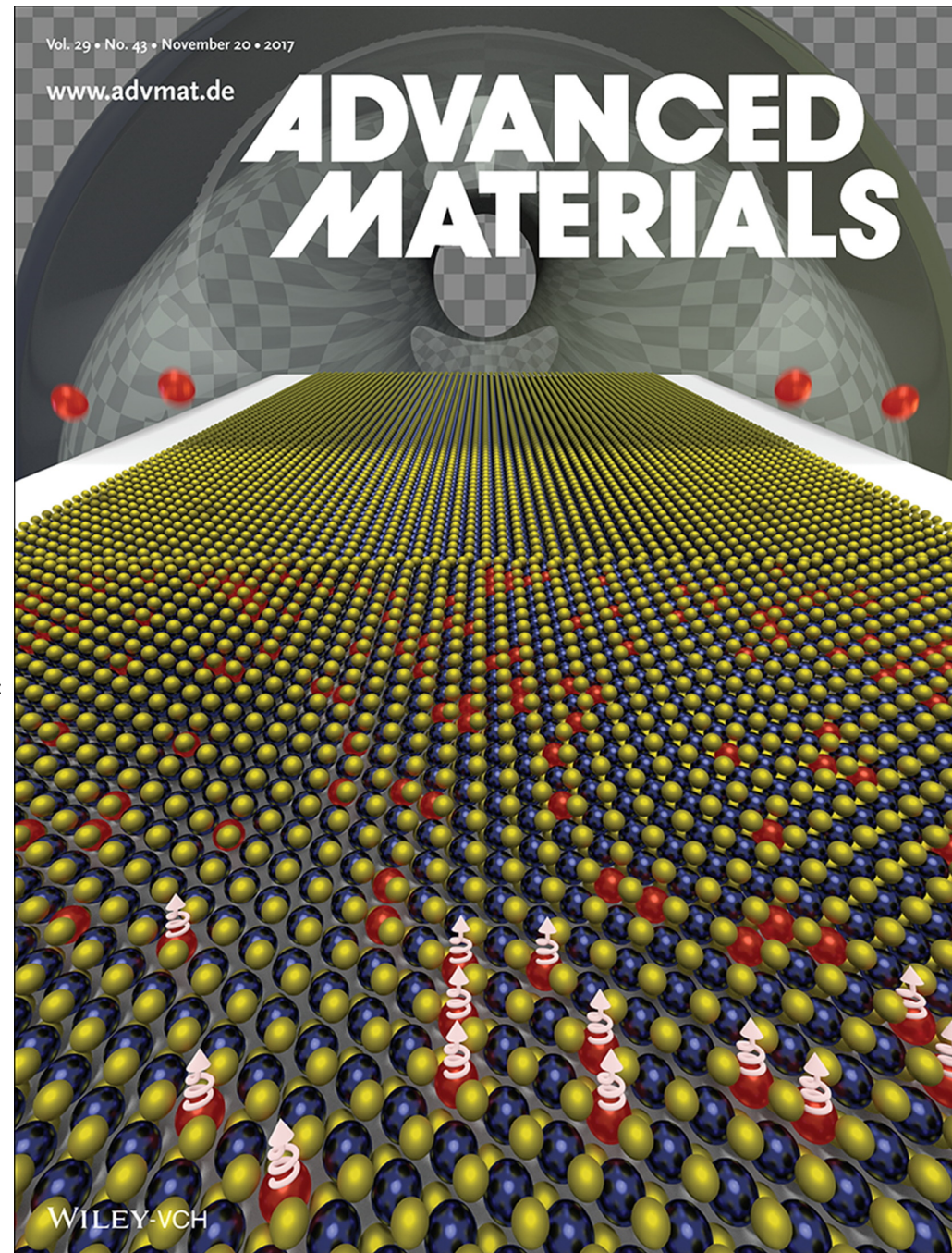
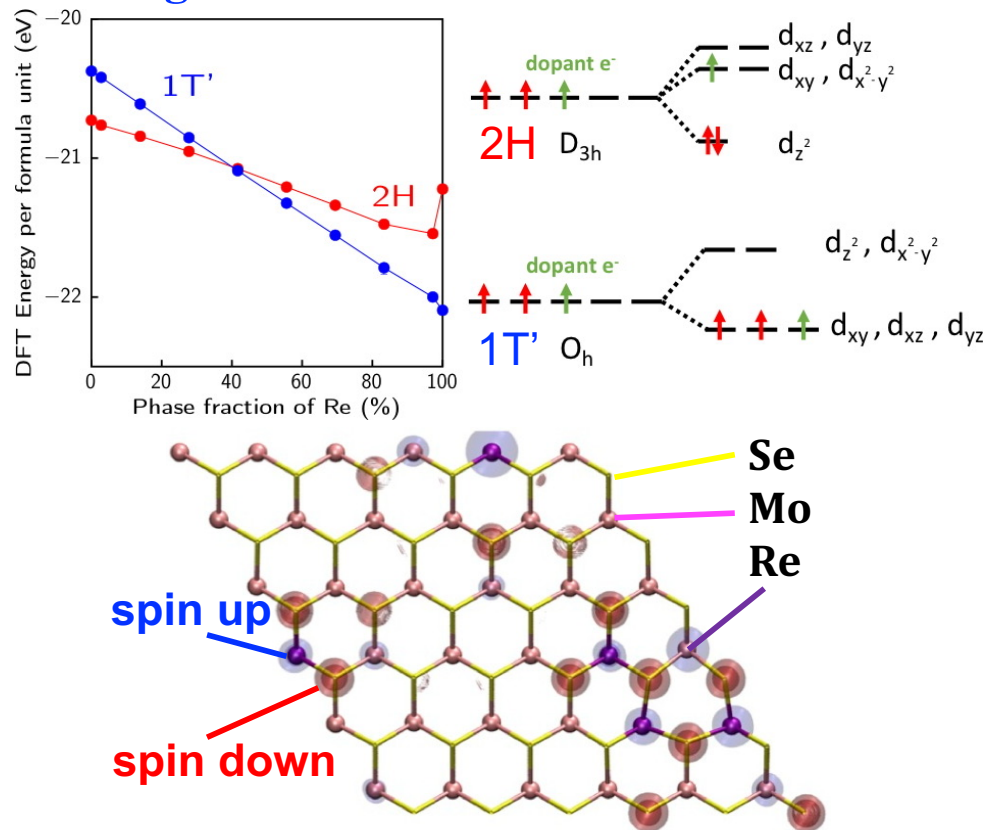
32 #define the two non-commuting terms that comprise the Hamiltonian
33 Hz = PauliTerm("Z", 0, epsilon_0)
34 Hy = PauliTerm("Y", 0, epsilon_ph*np.sin(w_ph*t))
35 #exponentiate the terms of the Hamiltonian for use in Trotter approx
36 exp_Hz = exponential_map(Hz)(delta_t/(2.0*hbar))
37 exp_Hy = exponential_map(Hy)(delta_t/hbar)
    
```

Do it yourself at <https://quantum-computing.ibm.com>



Emergent Magnetism: Structural Transition *via* Doping

- Experiment at Rice shows 2H-to-1T' phase transformation by alloying MoSe₂ with Re
- QMD simulations at USC elucidate its electronic origin
- Simulation & experiment show novel magnetism centered at Re atoms



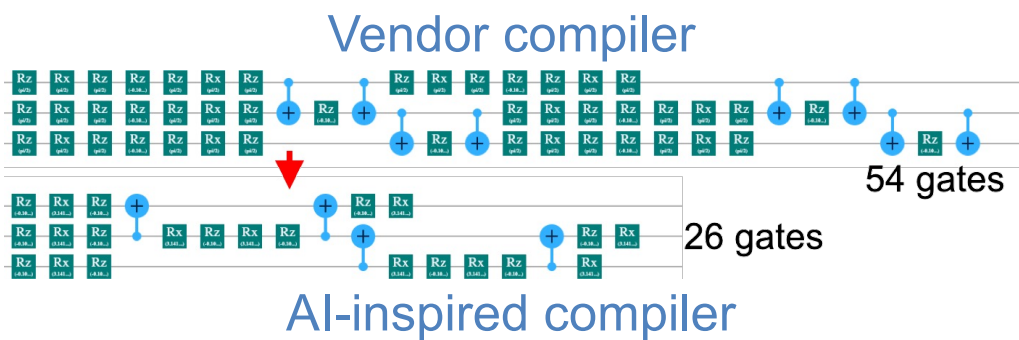
Kochat *et al.*, *Adv. Mater.* 29, 1703754 ('17)

Quantum Dynamics on Quantum Computers

- Quantum-dynamics simulations on quantum computers show dynamic suppression of magnetization by THz radiation

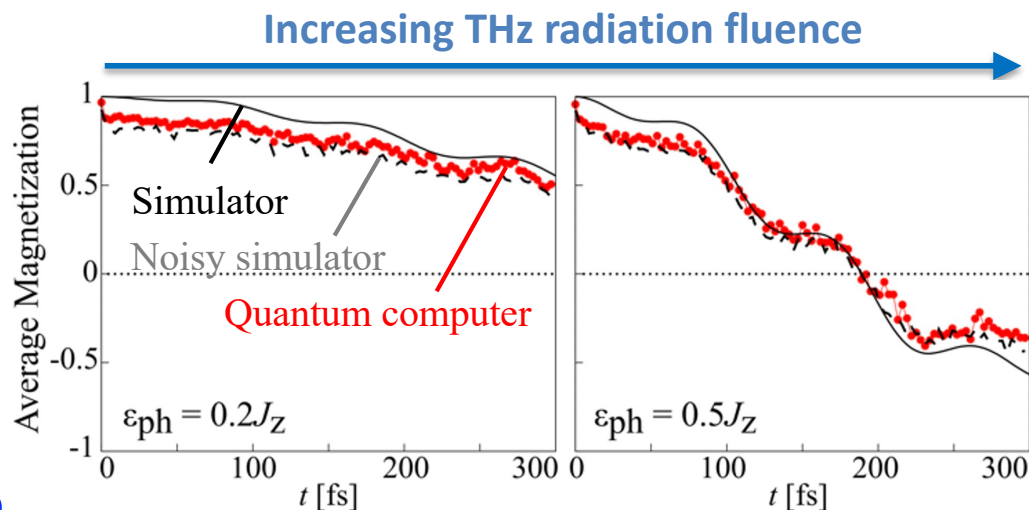
Bassman *et al.*,
Phys. Rev. **101**, 184305 ('20)

- AI-inspired quantum compiler reduced the circuit size by 30% to mitigate environmental noise



Bassman *et al.*,
Quantum Sci. Tech. **6**, 014007 ('21)

Lindsay Bassman: *Maria Curie Fellow* ('22-);
Science, She Says Award ('23)



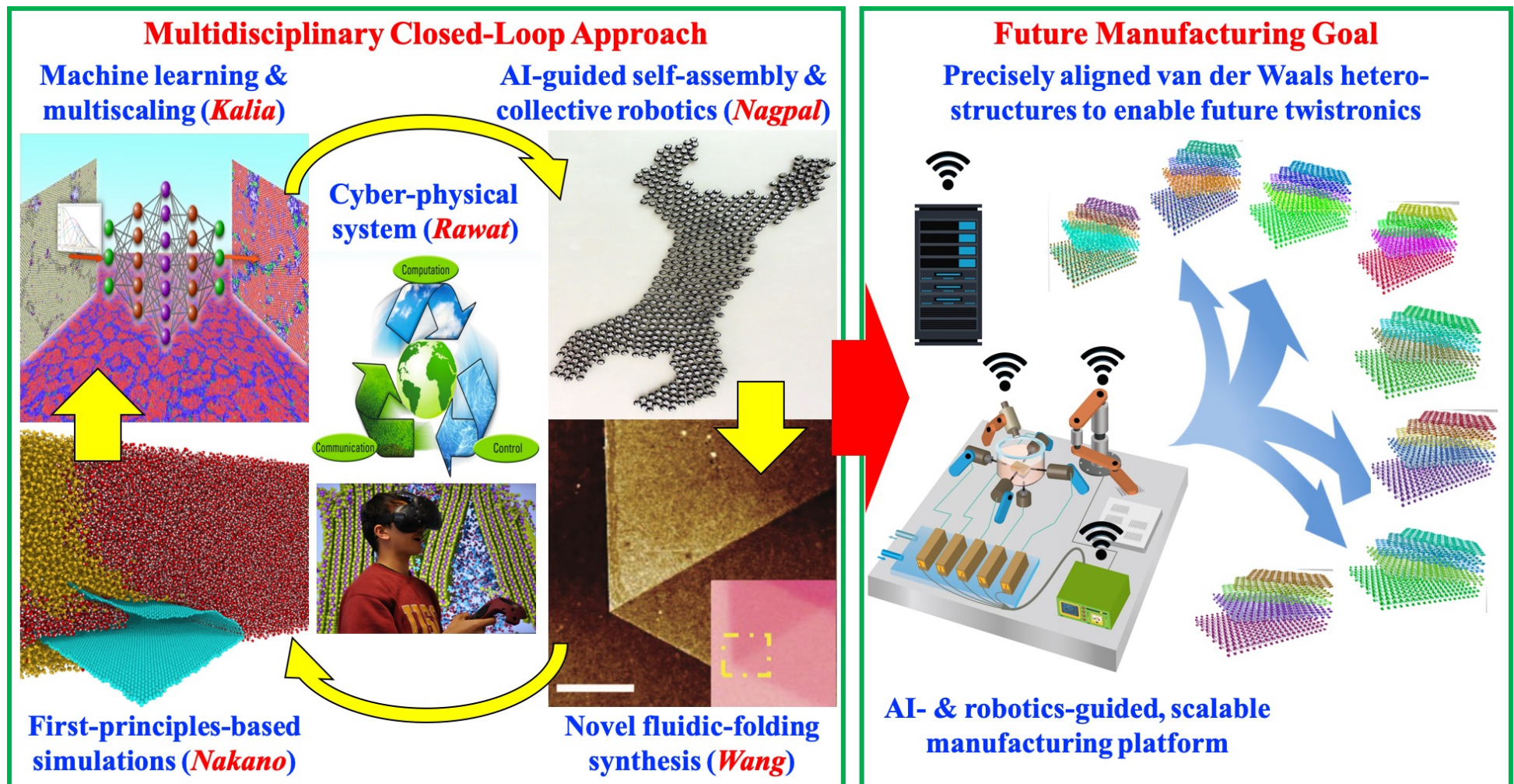
- Full-stack, cross-platform software for quantum dynamics simulations on NISQ computers

MISTIQS
Multiplatform Software
for Time-dependent
Quantum Simulation

Powers *et al.*, *SoftwareX* **14**, 100696 ('21)
<https://github.com/USCCACS/MISTIQS>

Princeton-USC-Howard Future Manufacturing

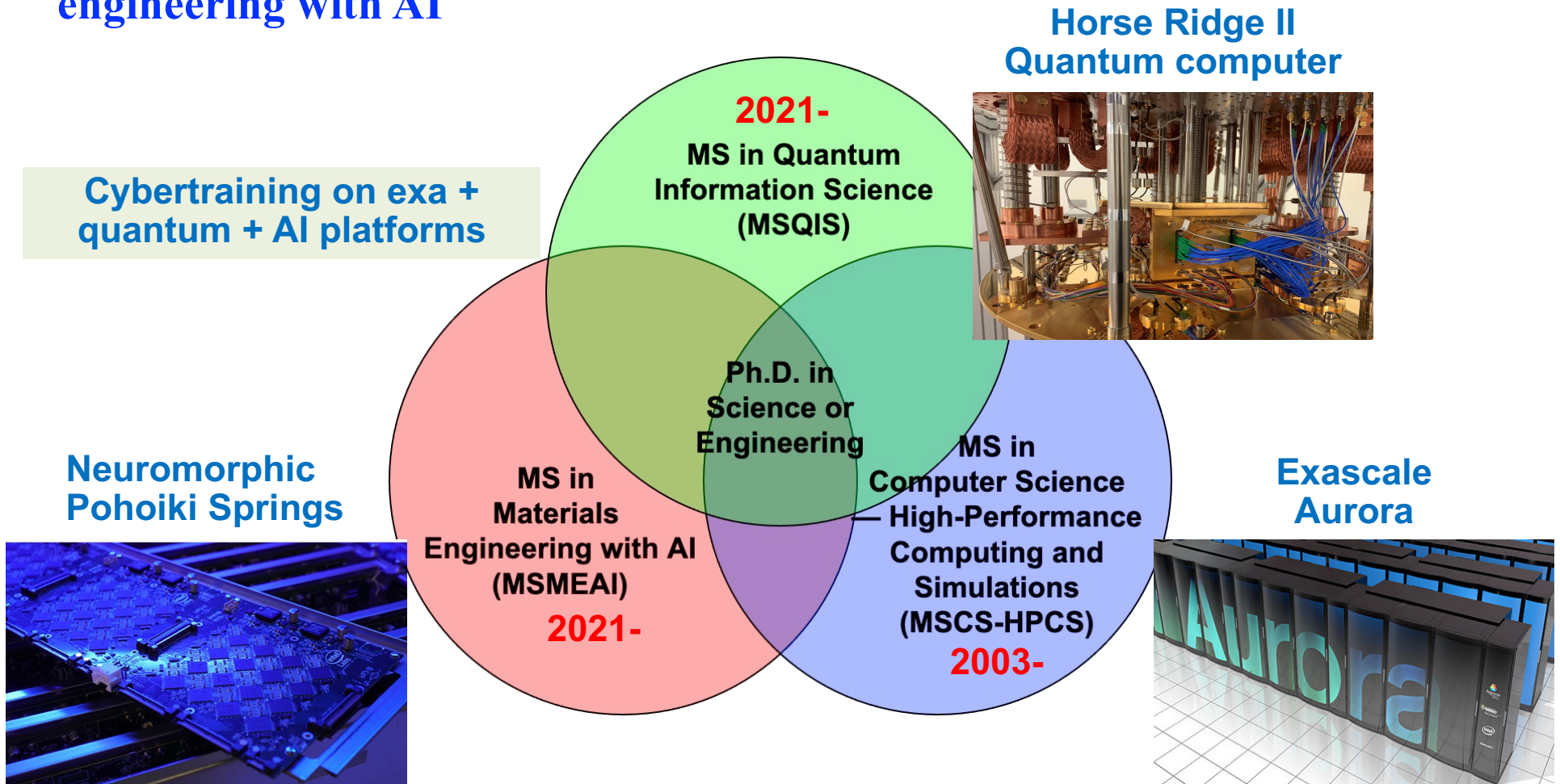
FMRG: Artificial Intelligence Driven Cybermanufacturing of Quantum Material Architectures
\$3.75M NSF project (2020-2025)



Nagpal (Princeton); Kalia, Nakano, Wang (USC); Rawat (Howard)

Training Cyber Science Workforce

- New generation of computational scientists at the **nexus of exascale computing, quantum computing & AI**
- **Unique dual-degree program:** Ph.D. in materials science or physics, along with MS in computer science specialized in high-performance computing & simulations, MS in quantum information science or MS in materials engineering with AI



USC-Howard Cybertraining

CyberMAGICS: Cyber Training on Materials Genome Innovation for Computational Software

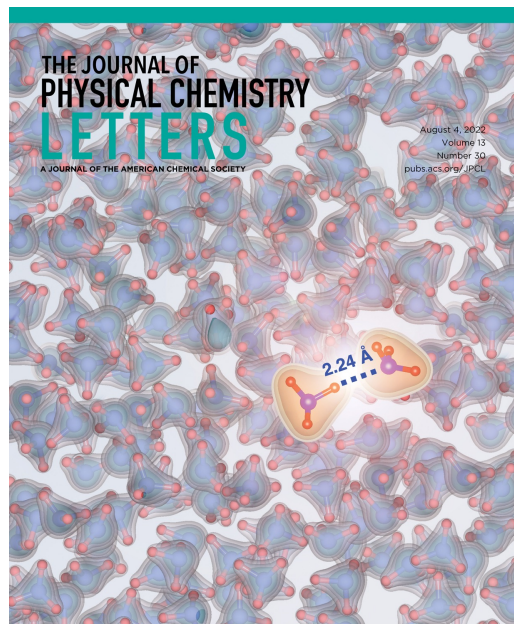
- Train a new generation of materials cyberworkforce, who will solve challenging materials genome problems through innovative use of advanced cyberinfrastructure at the exa-quantum-AI nexus



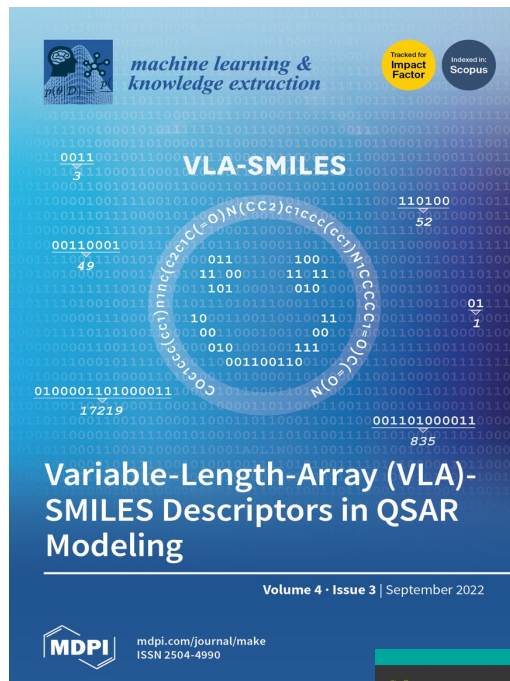
\$1M NSF CyberTraining (2021-25) project

Nakano, Nomura, Vashishta (USC); Dev, Wei (Howard)
External advisors: T. Germann, S. Plimpton, *et al.*

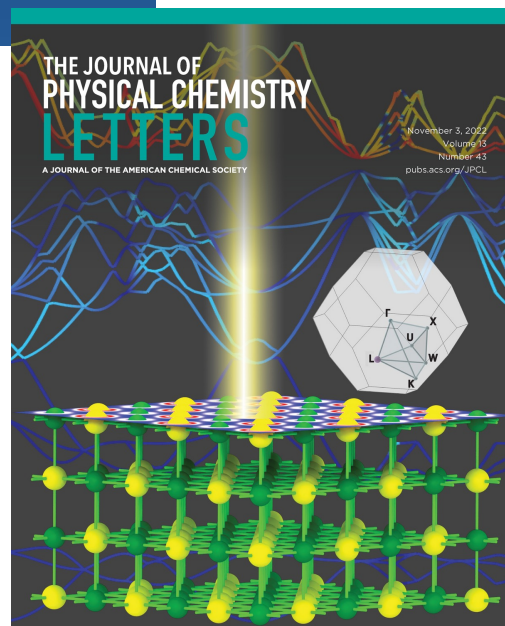
Other Applications



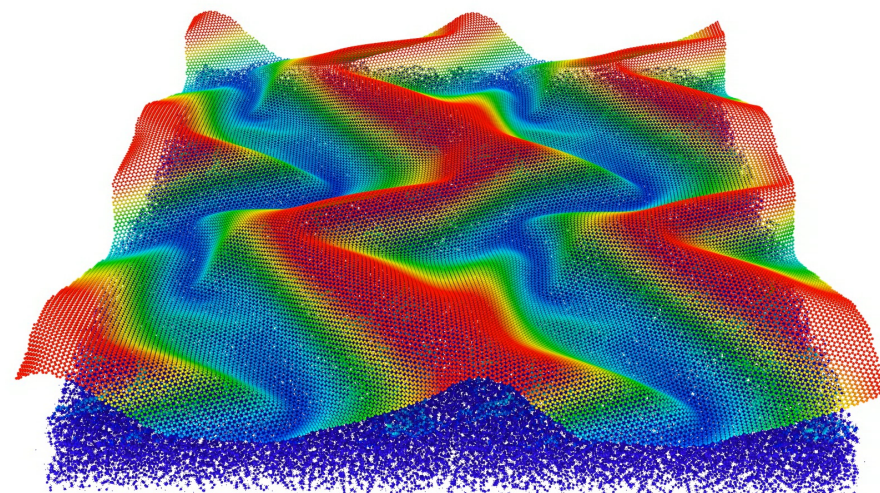
Green ammonia for renewable power (Aug. 4, '22)



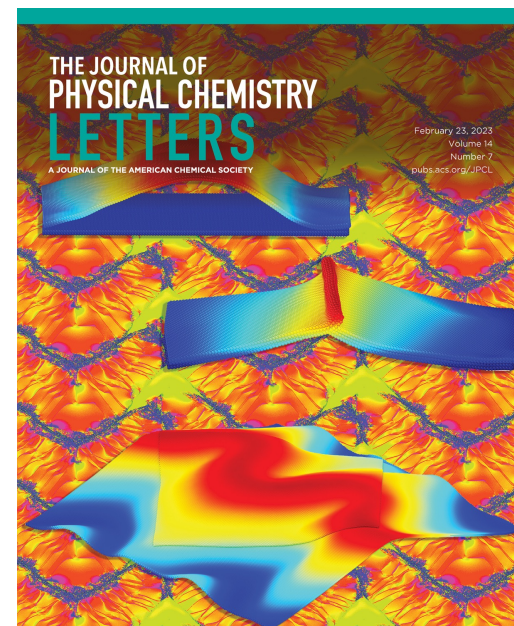
Molecular design (Sep. '22)



Phase-change memory (Nov. 3, '22)



Strain self-assembly of 2D metasurface (Feb. 23, '23)

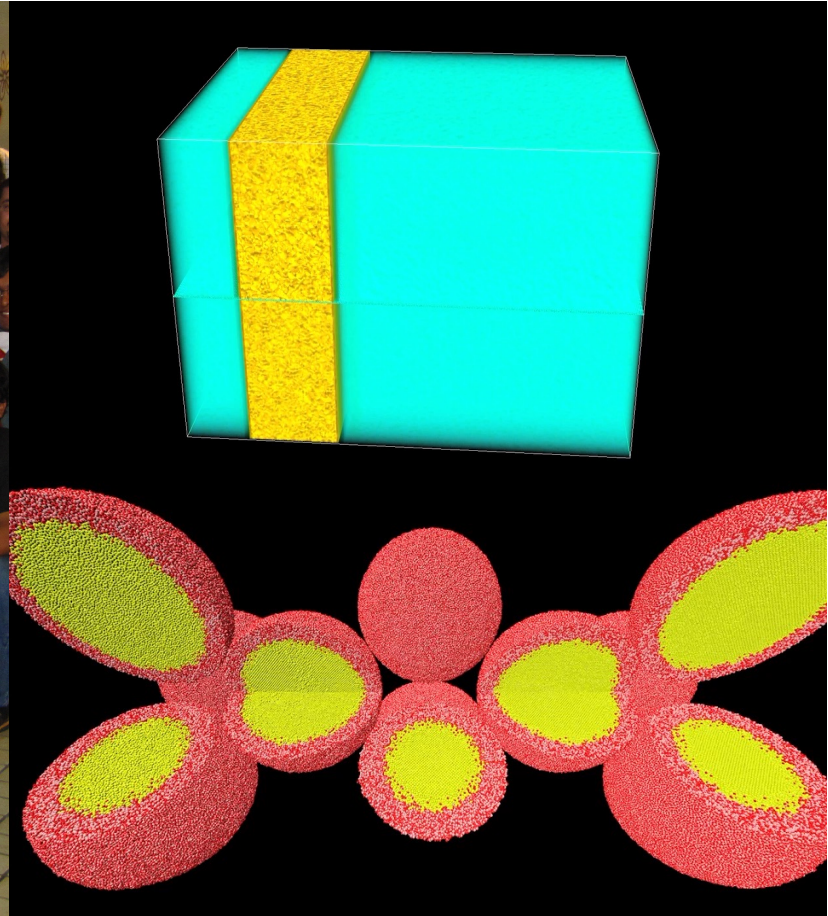


Conclusion

1. Large spatiotemporal-scale quantum, reactive & neural molecular dynamics simulations based on a common algorithmic framework
2. Broad applications of (exa+AI+quantum)4Science



Thank You



Research supported by
DOE-CMS/Neutron/INCITE/Aurora-ESP,
NSF-FM/CyberTraining, ONR-MURI×2, Sony

