

ATPESC Track 7: Scalable Molecular Visualization and Analysis Tools in VMD

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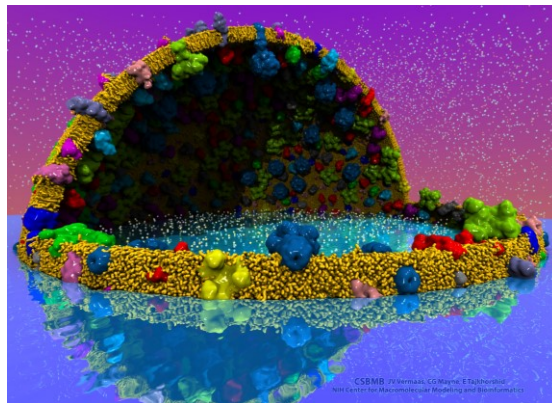
<http://www.ks.uiuc.edu/Research/vmd/>

Argonne Training Program on Exascale Computing (ATPESC)

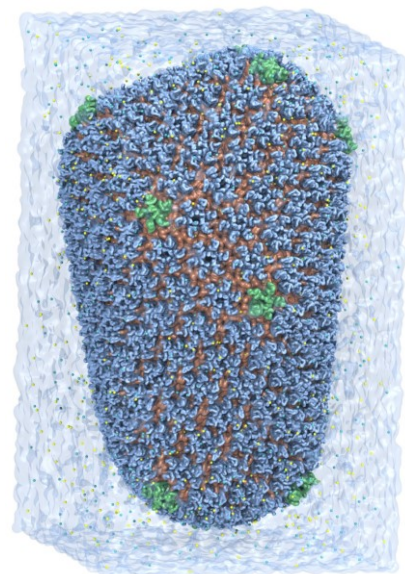
3:30pm-4:30pm, St. Charles Amphitheater, Q Center,
St. Charles, IL, Thursday August 9th, 2018

VMD – “Visual Molecular Dynamics”

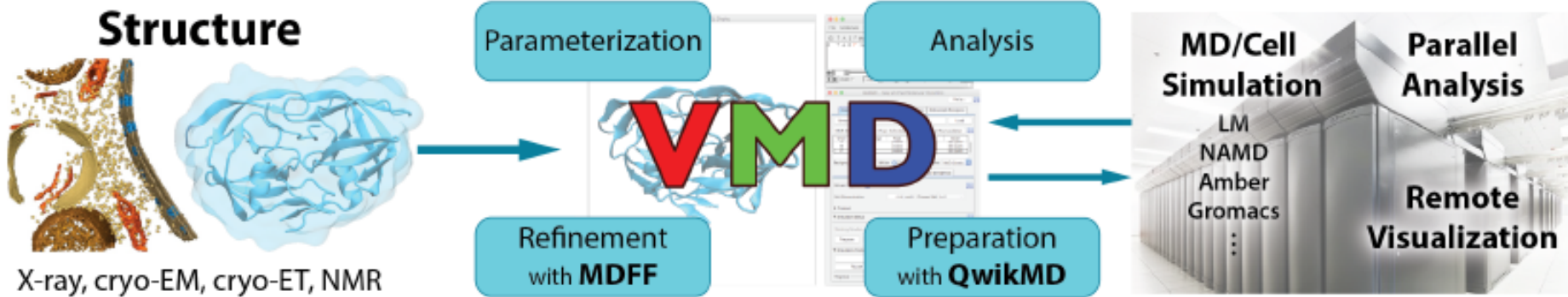
- 100,000 active users worldwide
- Visualization and analysis of:
 - Molecular dynamics simulations
 - Lattice cell simulations
 - Quantum chemistry calculations
 - Cryo-EM densities, volumetric data
- User extensible scripting and plugins
- <http://www.ks.uiuc.edu/Research/vmd/>



Cell-Scale Modeling



MD Simulation

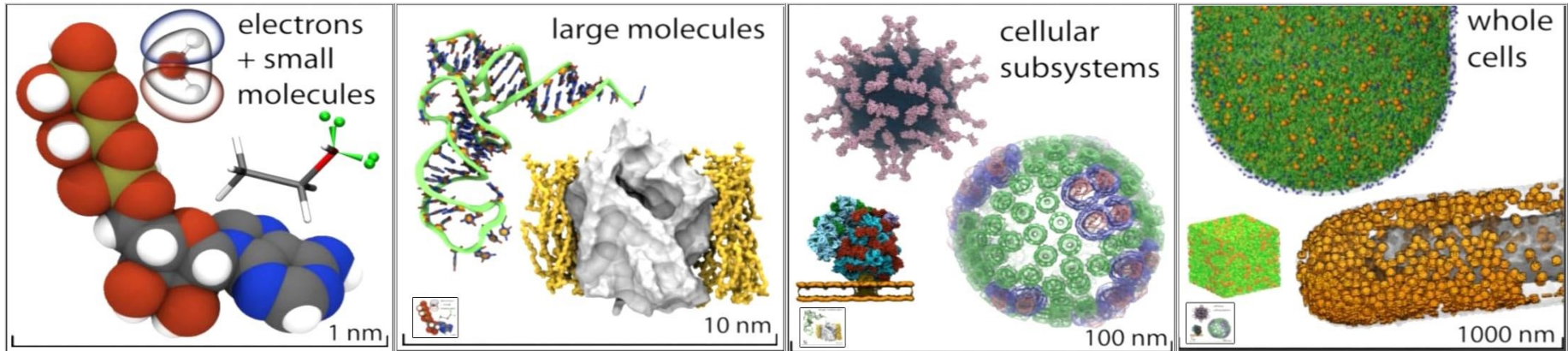


VMD Hands-On Tutorials

- <http://www.ks.uiuc.edu/Training/Tutorials/#vmd>
 - Main VMD tutorial
 - QwikMD simulation preparation and analysis plugin
 - VMD images and movies tutorial
 - Structure check
 - VMD quantum chemistry visualization tutorial
 - Visualization and analysis of CPMD data with VMD
 - Parameterizing small molecules using ffTK

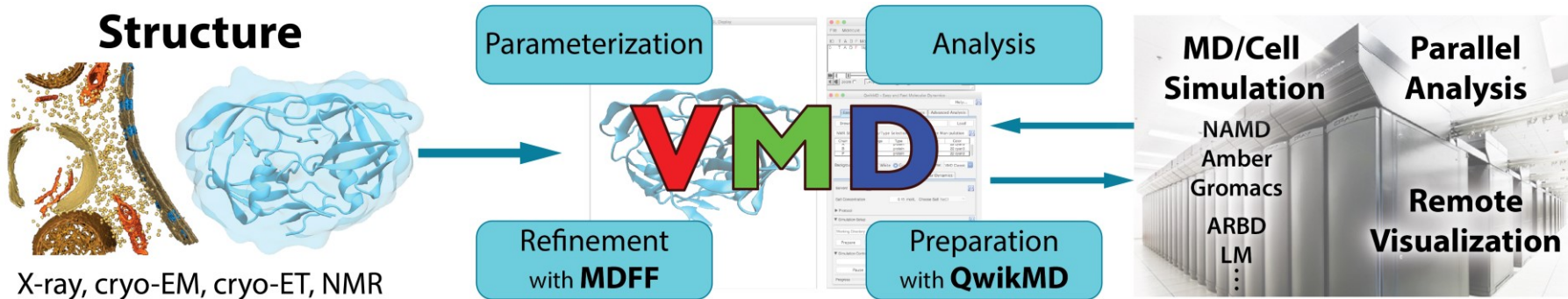
VMD Interoperability Serves Many Communities

- Uniquely interoperable with a broad range of tools:
 - AMBER, CHARMM, CPMD, DL_POLY, GAMESS, GROMACS, HOOMD, LAMMPS, NAMD, and many more ...
- Supports key data types, file formats, and databases
- Incorporates tools for simulation preparation, visualization, and analysis



VMD Interoperates with Mainstream Research Tools

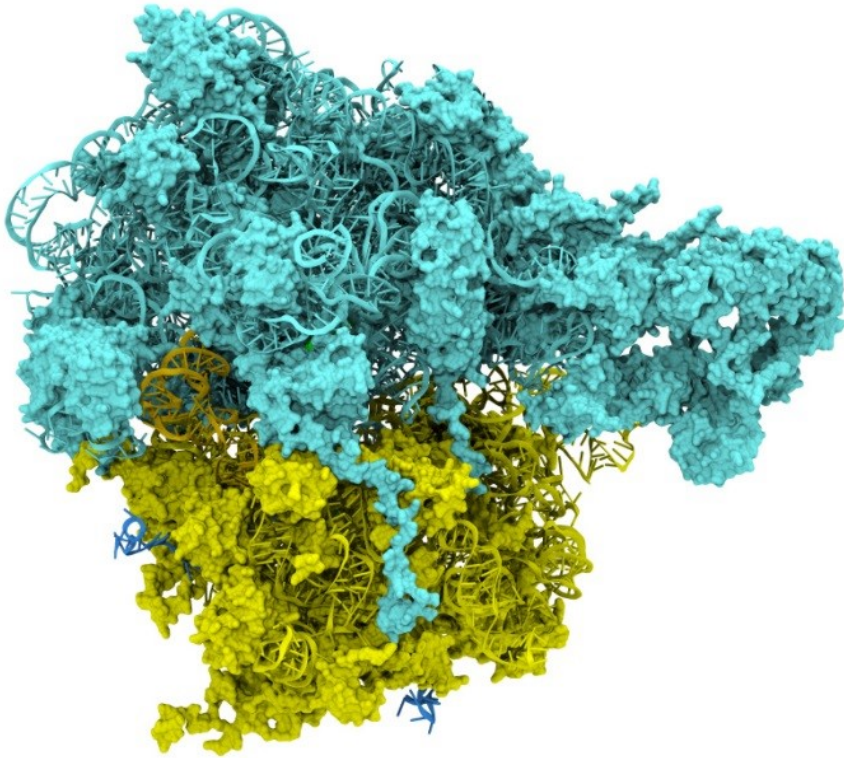
- Provides tools for simulation preparation, visualization, and analysis
- Interpret and process multi-modal structural information
- Connects with key software tools to enable state-of-the-art simulations
- Openness, extensibility, and interoperability are VMD hallmarks
- Uses advanced algorithms and hardware technologies to address data size challenges posed by cutting-edge experimental imaging and simulation



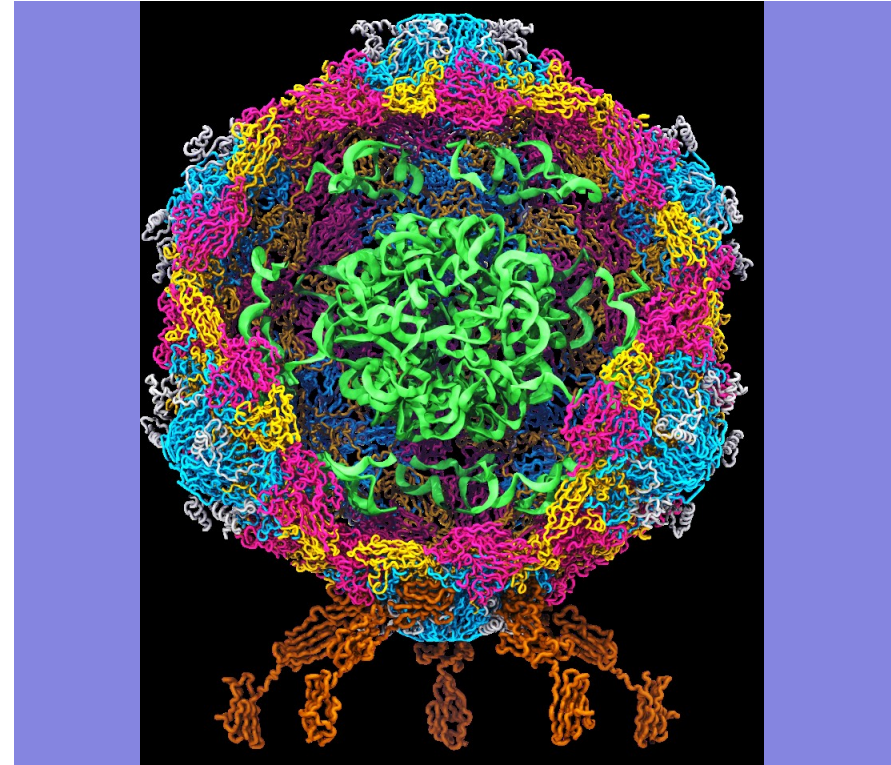
Goal: A Computational Microscope

Study the molecular machines in living cells

Ribosome: target for antibiotics



Poliovirus

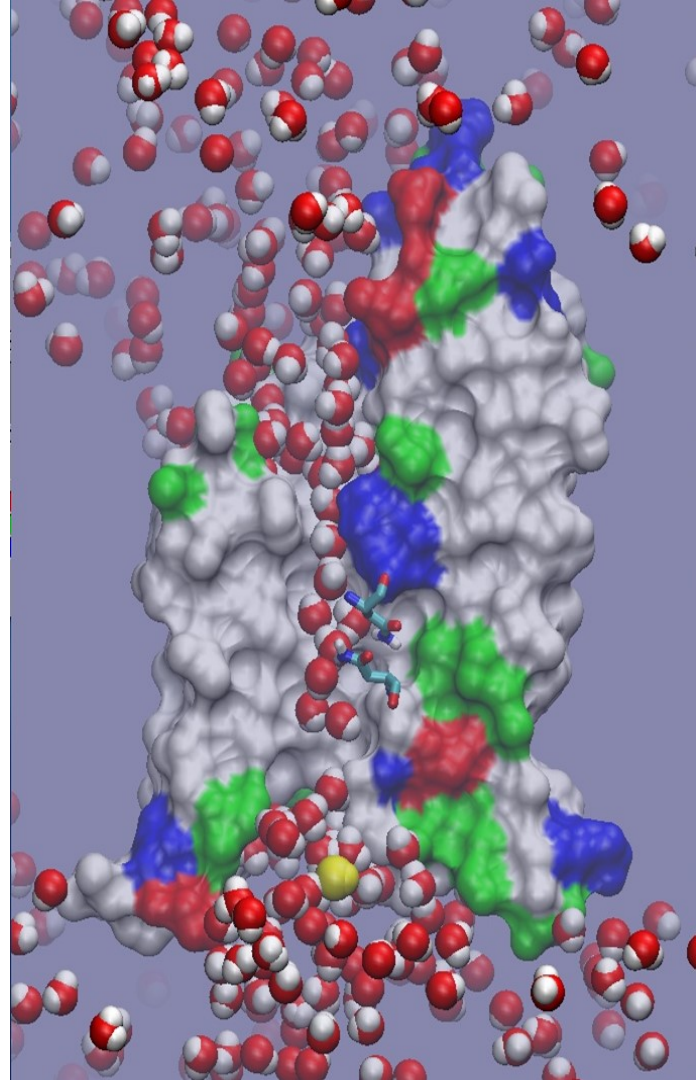


VMD Approach to Visualization

- Molecular scene is composed of “graphical representations”
- Each representation encapsulates a group of selected atoms, a drawing style, coloring style, and other parameters
- Representations are independent of each other, can be toggled on/off easily, allowing molecular scenes to be built-up incrementally
- VMD atom selection language is shared with its analytical and scripting interfaces

Selection, Filtering

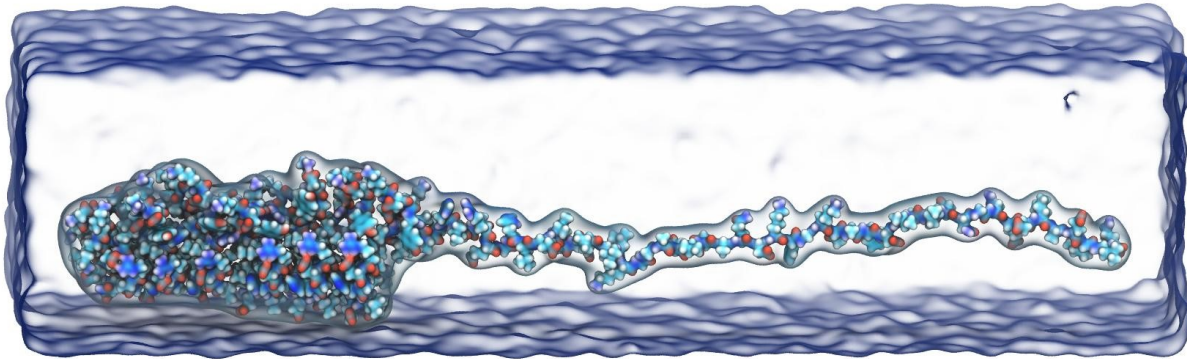
- Most viz tools allow interactive visual picking, menu-driven selections of structure components to display or operate on
- VMD also extensively uses a text-based selection language (think google):
 - water within 10 of protein and $z > 0$
 - nucleic or protein or ions
 - segname BR
 - name "C.*"
- Allows selection on user-defined data fields
- ***Promotes synergy between interactive and scripting interfaces, visualization and quantitative analysis tasks***
- Works well with huge time-varying structures



Structure Visualization

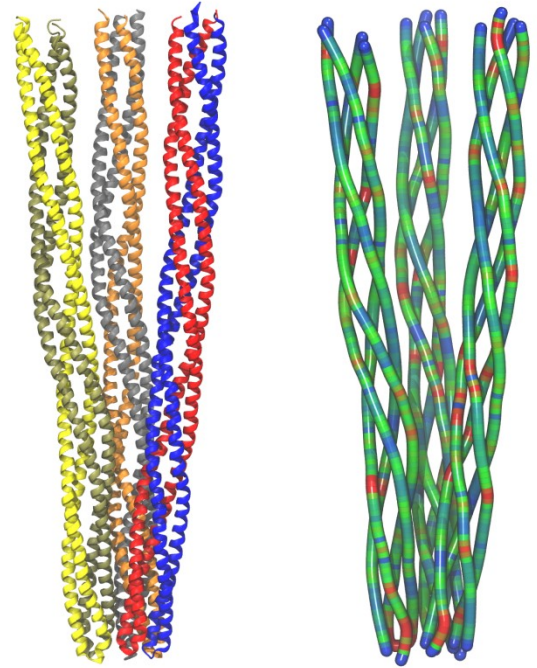
Molecular representations provide different levels of abstraction, atomic detail vs. higher level organizational information

- Atoms, VdW spheres, bonds, ball-stick, ...
- Coarse-grained “beads”
- Ribbons, secondary structure, “cartoon” reps, RNA/DNA
- Molecular surfaces
- Molecular orbitals (quantum chemistry)



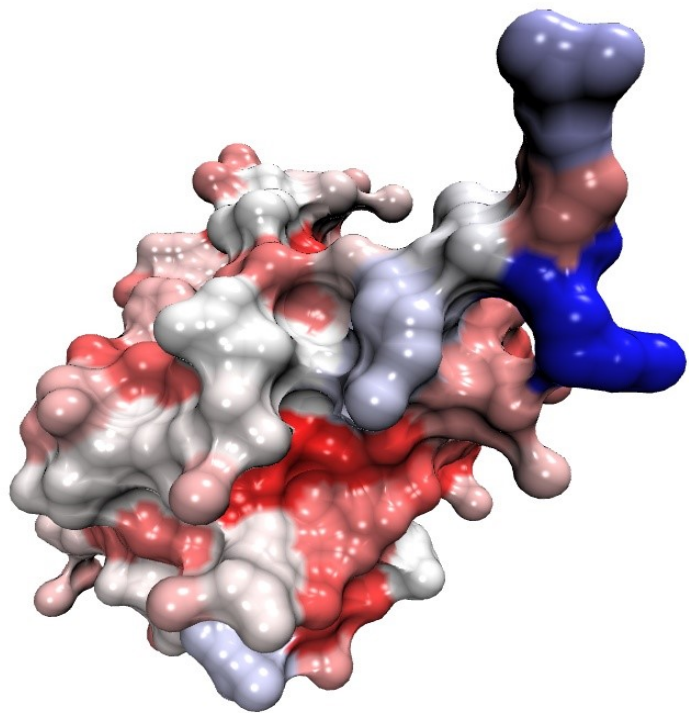
Computed Properties

- Smoothing of thermal noise
- Secondary structure
- Hydrogen bonds, salt bridges
- Forces, energies, stress, strain
- Time averaging of electrostatic fields, occupancy maps
- Quality-of-fit cross correlation with cryo-EM density maps
- Normal modes, principal component analysis, essential dynamics
- Cluster simulation trajectory timesteps by structural similarity

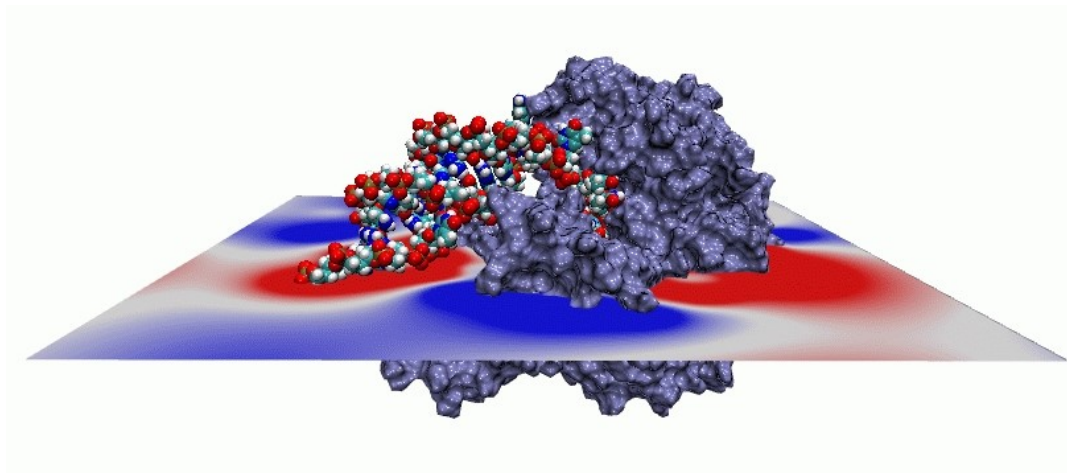


Chemoreceptor trimer-of-dimers analysis with Bendix plugin in VMD

Display of Computed Properties on Structures

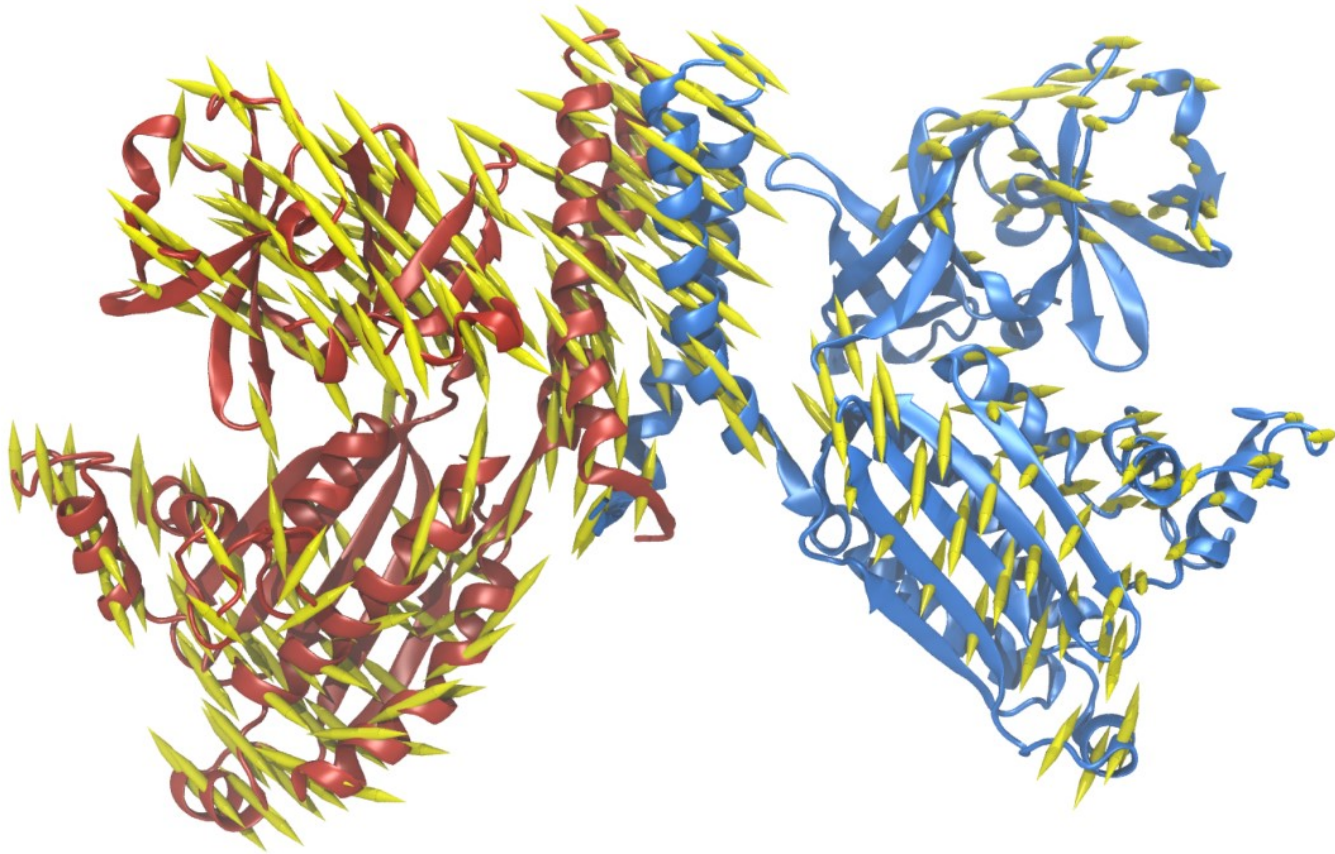


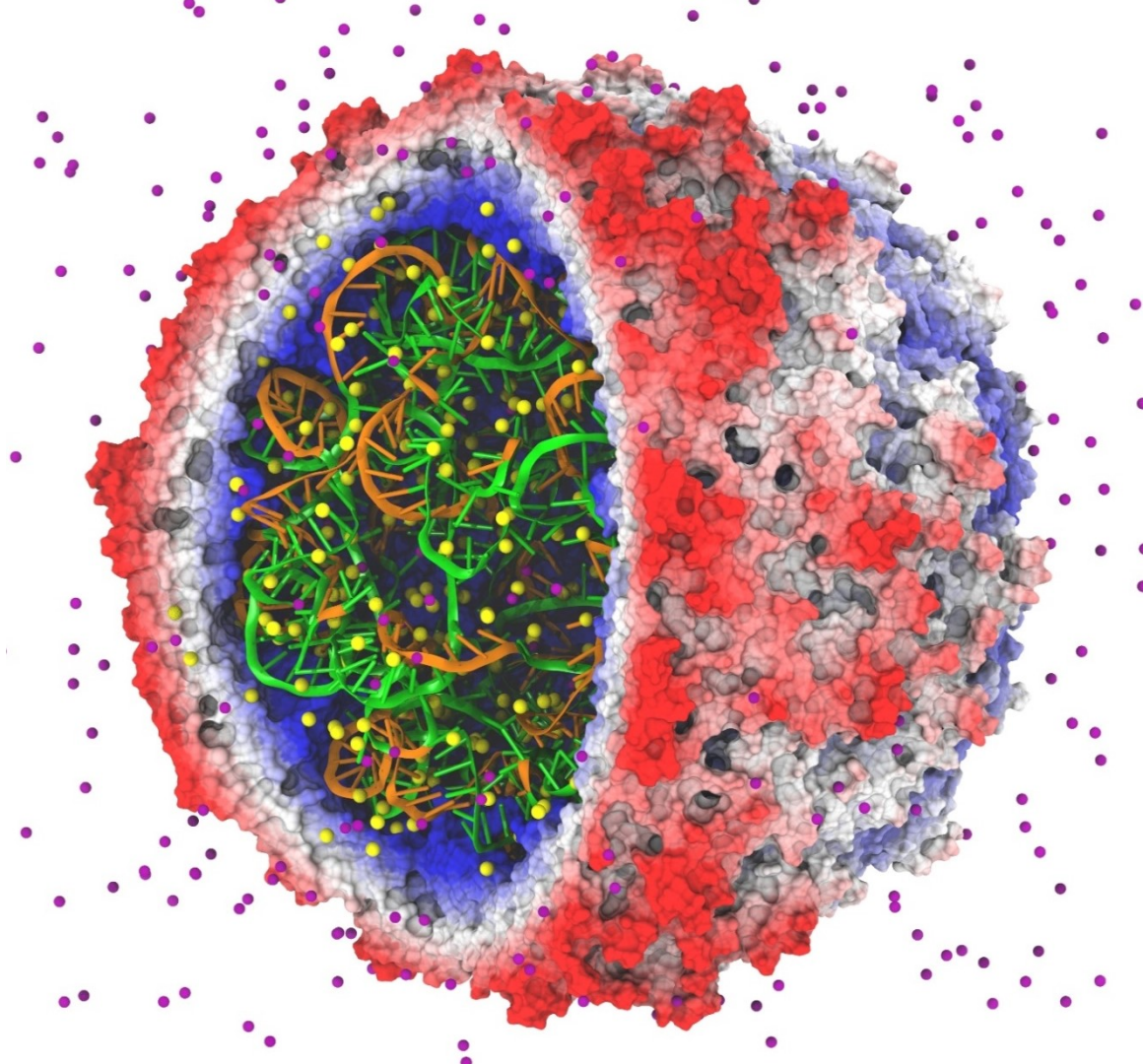
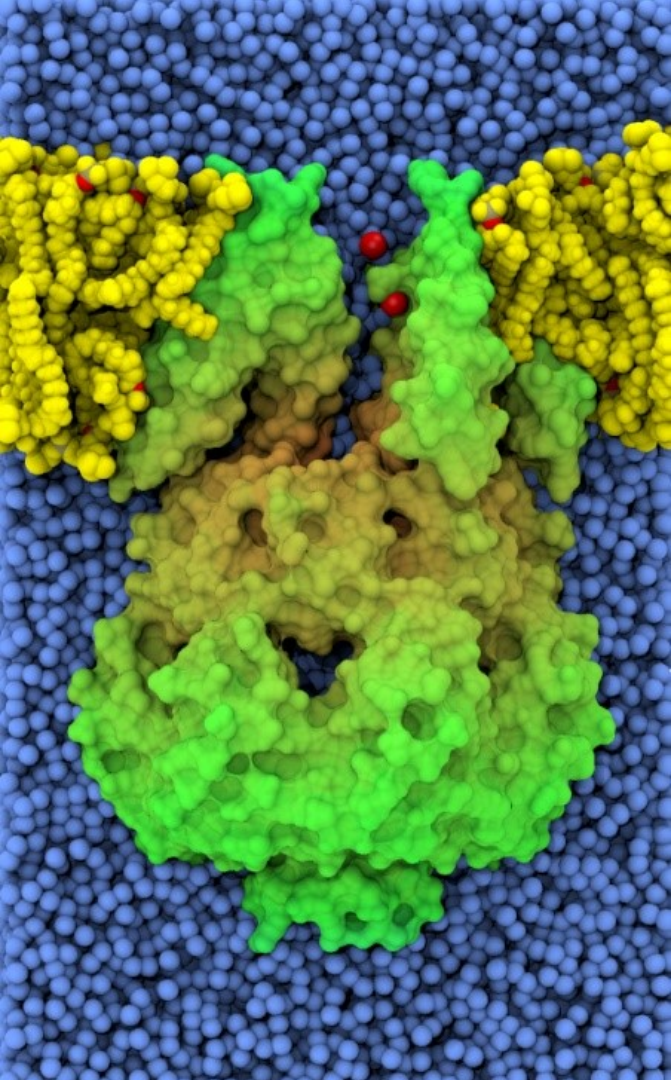
Per-residue solvent-accessible surface area of Ubiquitin



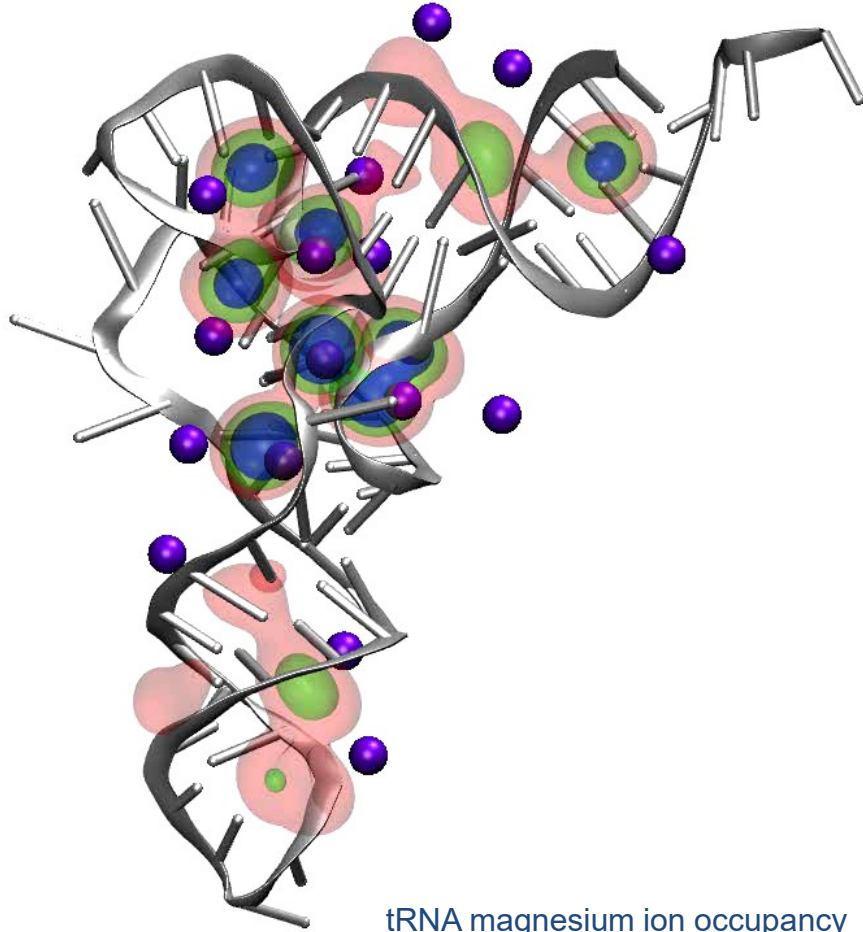
PME electrostatic potential contour for a helicase on a volumetric slice plane

CheA kinase PCA: first principal component porcupine plot





Computing Molecular Properties



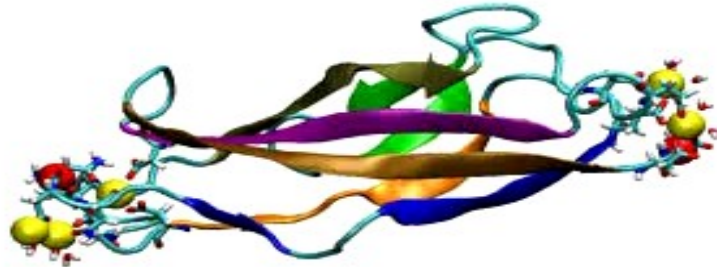
tRNA magnesium ion occupancy

Compute properties, e.g., density, distance, occupancy, electrostatic potential over thousands of MD simulation trajectory frames

Example: display binding sites for diffusively bound ions as probability density isosurfaces

Visualization of Molecular Dynamics

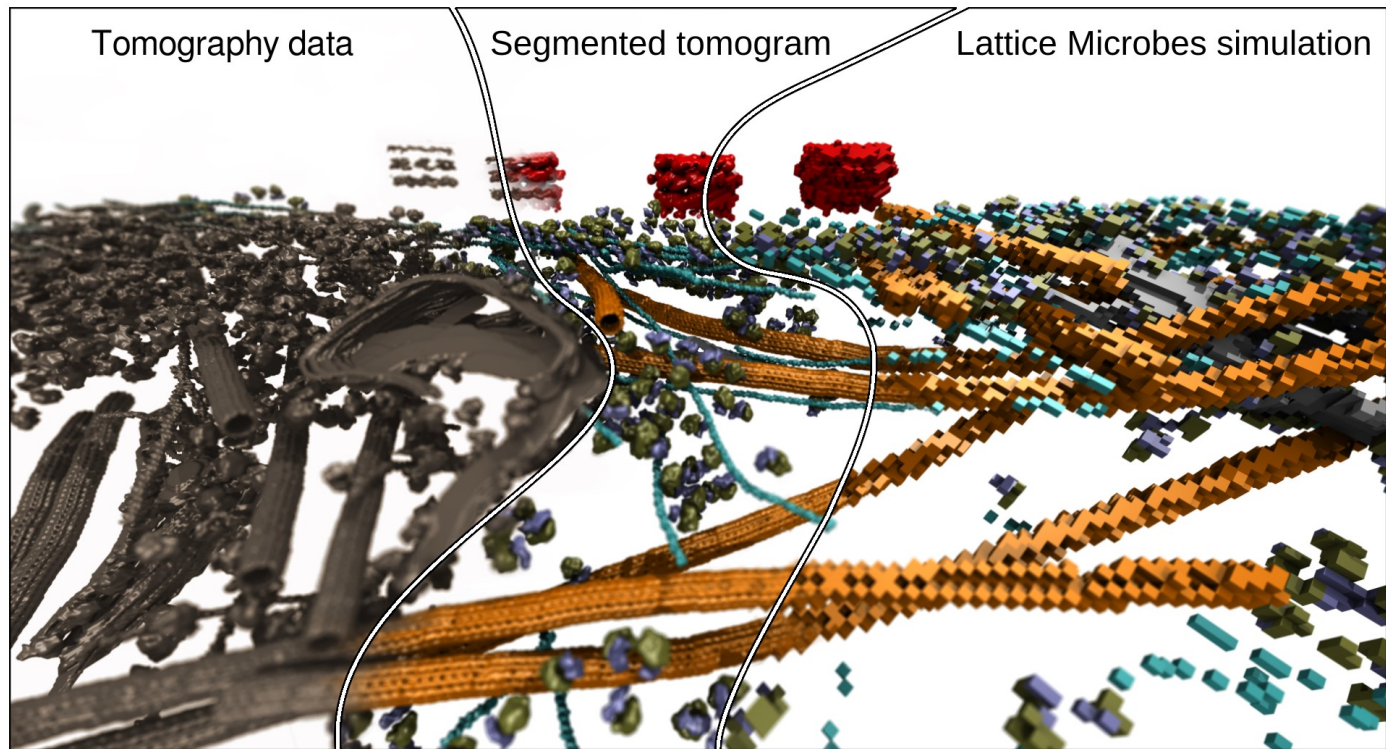
- Molecular dynamics simulations save trajectories of atomic coordinates as simulated time progresses
- Researchers study trajectories by analyzing force profiles, energies, structural changes, etc.
- **Visualization selections, graphics, structure properties recomputed for each trajectory timestep!**



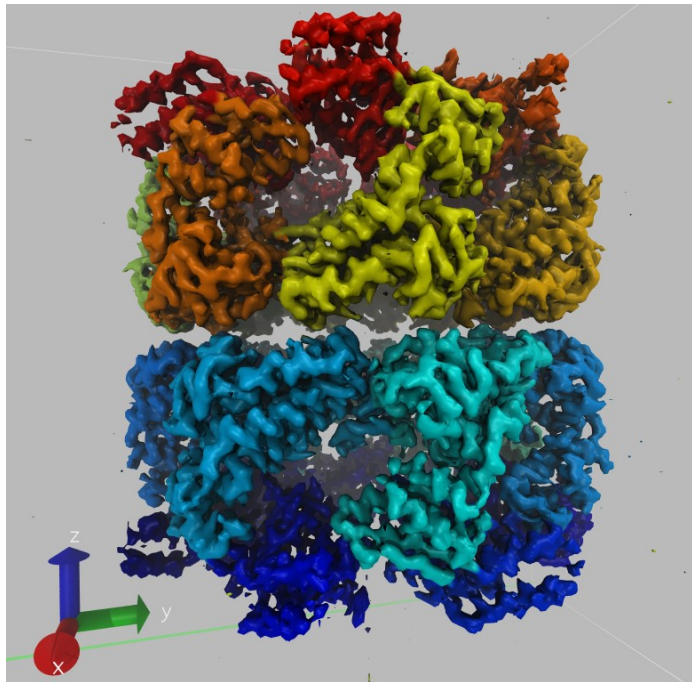
Cryo-EM / Cryo-ET Density Map Segmentation

Evaluate 3-D volumetric electron density maps and segment them, to identify key structural components

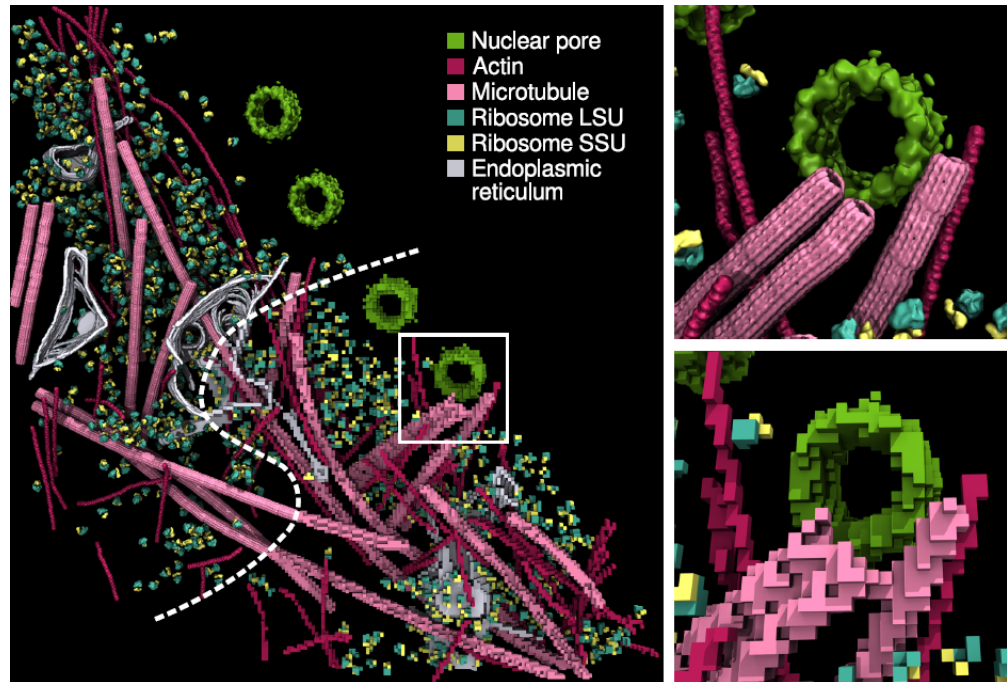
Index/label components so they can be referred to, colored, analyzed, and simulated...



Density Map Segmentation

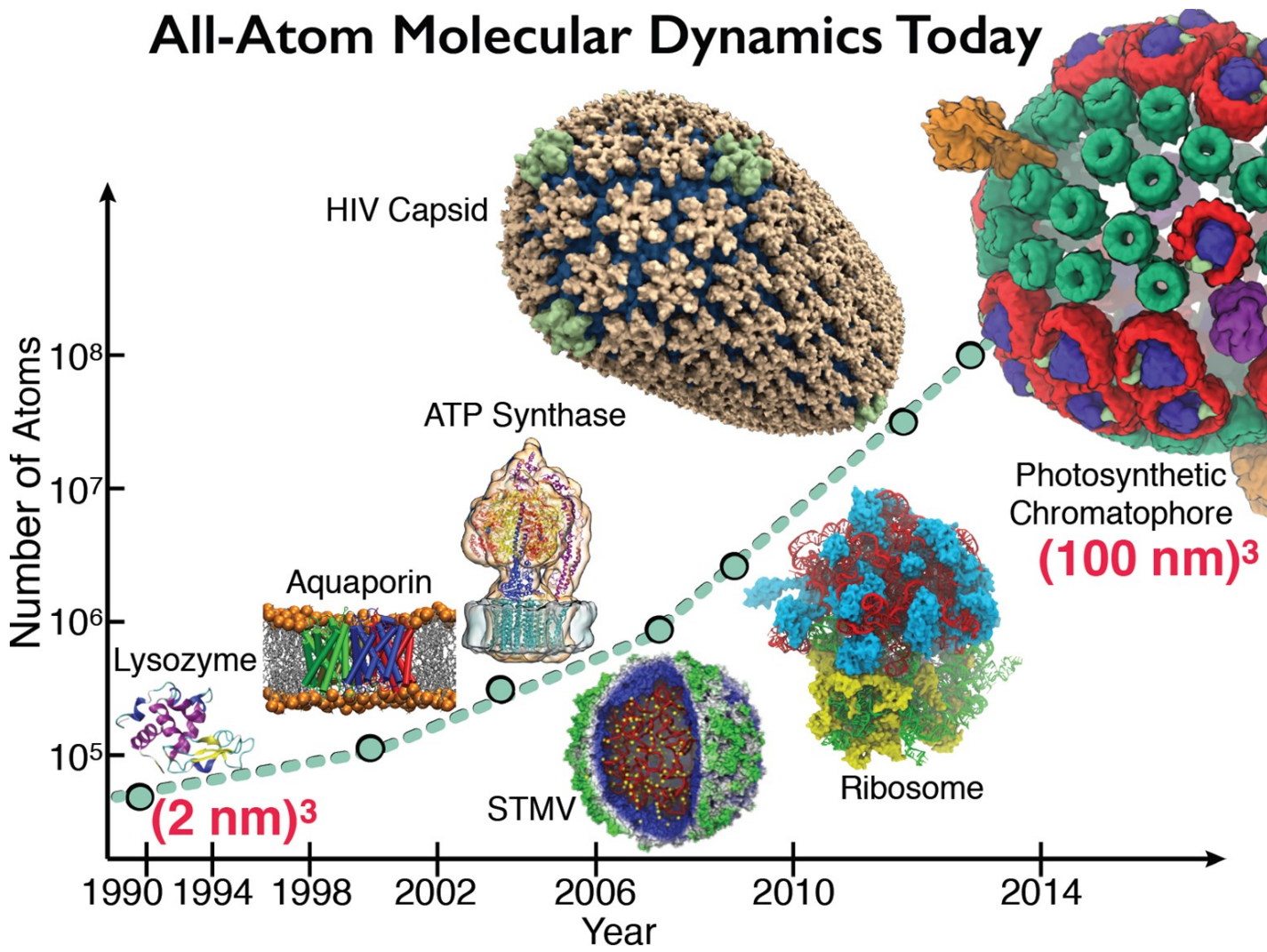


VMD GPU-accelerated density map segmentation of GroEL

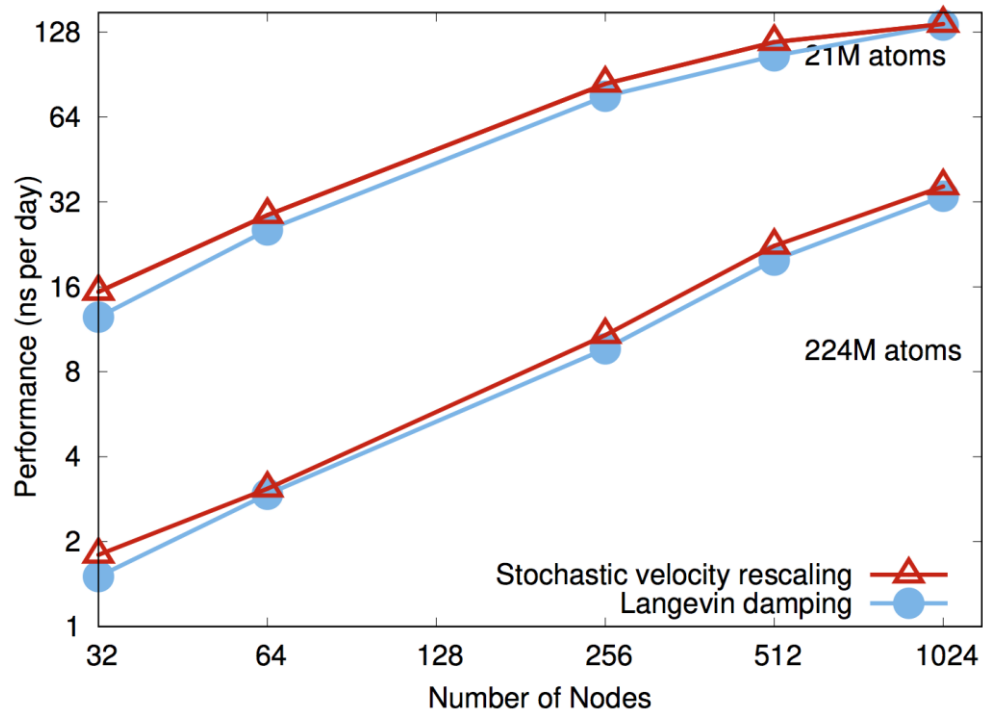


Earnest, et al. *J. Physical Chemistry B*, 121(15): 3871-3881, 2017.

All-Atom Molecular Dynamics Today



NAMD on Summit, May 2018



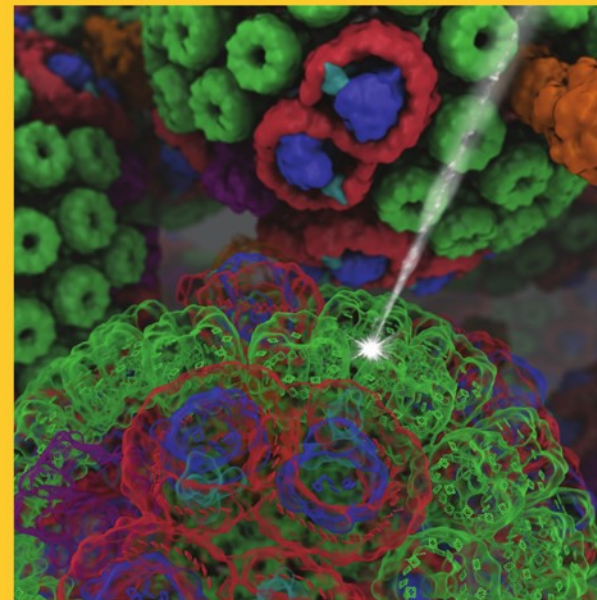
NAMD simulations can generate up to 10TB of output per day on 20% of Summit

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Molecular Dynamics Trajectory Analysis

- MD simulations sample femtosecond timescales
- Millions of timesteps stored per trajectory
- Dynamics of biomolecular complexes are main interest, but **solvent often accounts for half or more of the simulation content**

Skip I/O for regions of bulk solvent where possible [1]

- Modern MD tools, e.g., VMD, NAMD, LAMMPS, HOOMD, employ extensive **embedded scripting (Python, Tcl, etc)** to permit simulation preparation, custom simulation protocols, **analysis, and visualization**
- Unified collective variables module allows identical analytical computations to be performed within LAMMPS, NAMD, and VMD, during **pre-simulation modeling, in-situ, and post-hoc [2]**

[1] Immersive Out-of-Core Visualization of Large-Size and Long-Timescale Molecular Dynamics Trajectories.

J. Stone, K. L. Vandivort, and K. Schulten. G. Bebis et al. (Eds.): *7th International Symposium on Visual Computing (ISVC 2011)*, LNCS 6939, pp. 1-12, 2011.

[2] Using collective variables to drive molecular dynamics simulations. G. Fiorin, M. L. Klein, and J. Hénin. *Molecular Physics*, 111:22-23, 3345-3362, 2013.

Petascale Molecular Dynamics I/O and Storage Challenges

- NAMD simulations can produce up to **10TB/day @ 1024 nodes (~20%) of ORNL Summit**, more as optimizations raise NAMD performance further
- Petascale science campaigns require **months of simulation runs**
- **Long-term storage of large-fractional petabytes impractical**
- **Historical “download output files for analysis and visualization” approach is a non-starter at this scale**
- Demands visualization and analysis operate on the data **in-place** on the HPC system, **whether *post-hoc*, *in-transit*, or *in-situ***
- Analyses must **identify salient features of structure, dynamics**, cull data that don't contribute to biomolecular processes of interest

VMD Petascale Visualization and Analysis

- Analyze/visualize large trajectories too large to transfer off-site:
 - User-defined parallel analysis operations, data types
 - Parallel rendering, movie making
- Supports GPU-accelerated Cray XK7 nodes for both visualization and analysis:
 - **GPU accelerated trajectory analysis w/ CUDA**
 - **OpenGL and GPU ray tracing for visualization and movie rendering**
- Parallel I/O rates up to **275 GB/sec** on 8192 Cray XE6 nodes – can read in **231 TB in 15 minutes!**

Parallel VMD currently available on:

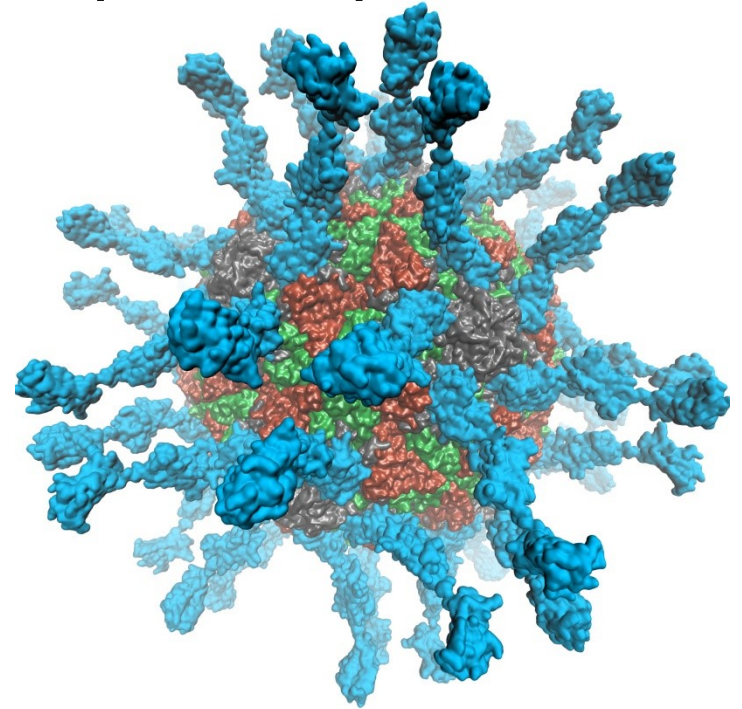
**ORNL Titan, NCSA Blue Waters, Indiana Big Red II,
CSCS Piz Daint, and similar systems**



NCSA Blue Waters Hybrid Cray XE6 / XK7
22,640 XE6 dual-Opteron CPU nodes
4,224 XK7 nodes w/ Telsa K20X GPUs

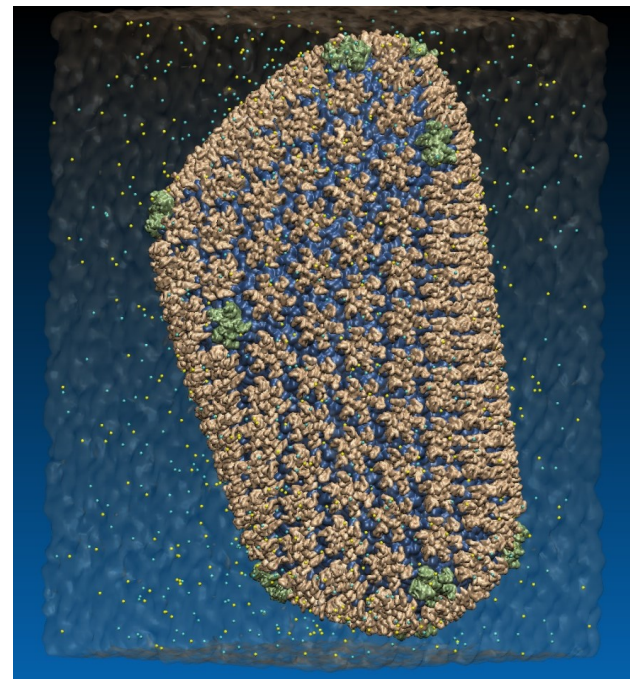
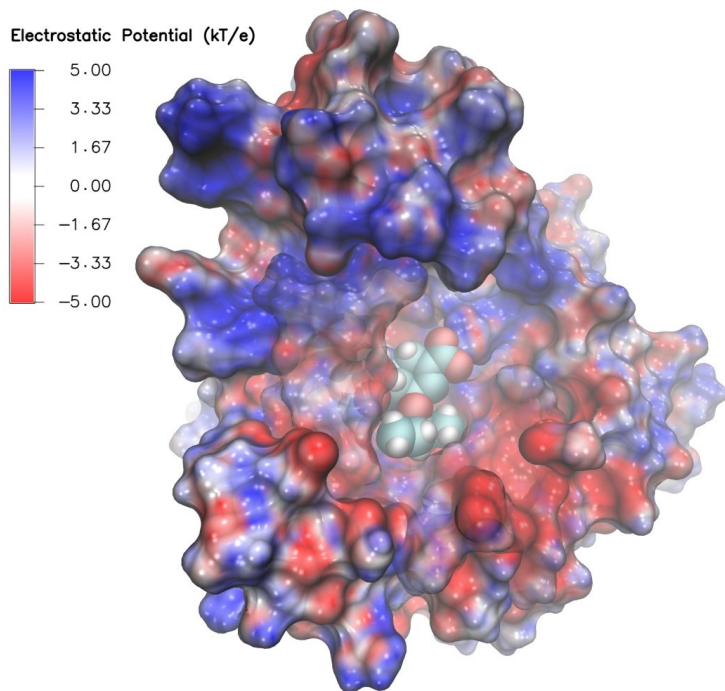
VMD supports EGL for in-situ and parallel rendering on clouds, clusters, and supercomputers

- Eliminate dependency on windowing systems
- Simplified deployment of parallel VMD builds supporting off-screen rendering
- Maintains 100% of VMD OpenGL shaders and rendering features
- Support high-quality vendor-supported commercial OpenGL implementations in HPC systems that were previously limited to Mesa



Poliovirus

VMD EGL Rendering: Supports full VMD GLSL shading features Vulkan support coming soon...



Swine Flu A/H1N1 neuraminidase bound to Tamiflu

64M atom HIV-1 capsid simulation

High Performance Molecular Visualization: In-Situ and Parallel Rendering with EGL.

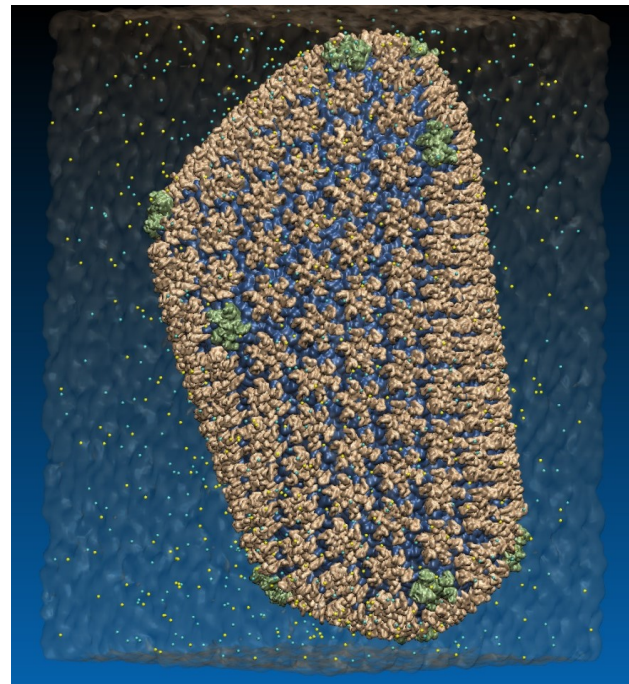
J. E. Stone, P. Messmer, R. Sisneros, and K. Schulten. High Performance Data Analysis and Visualization Workshop, IEEE IPDPSW, pp. 1014-1023, 2016.

VMD EGL Performance on Amazon EC2 Cloud

MPI Ranks	EC2 “G2.8xlarge” GPU Instances	HIV-1 movie rendering time (sec), (I/O %) 3840x2160 resolution
1	1	626s (10% I/O)
2	1	347s (19% I/O)
4	1	221s (31% I/O)
8	2	141s (46% I/O)
16	4	107s (64% I/O)
32	8	90s (76% I/O)

Performance at 32 nodes reaches ~48 frames per second

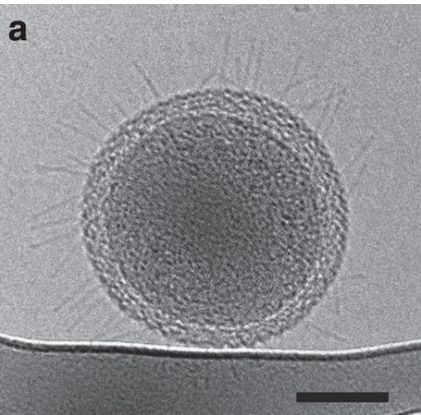
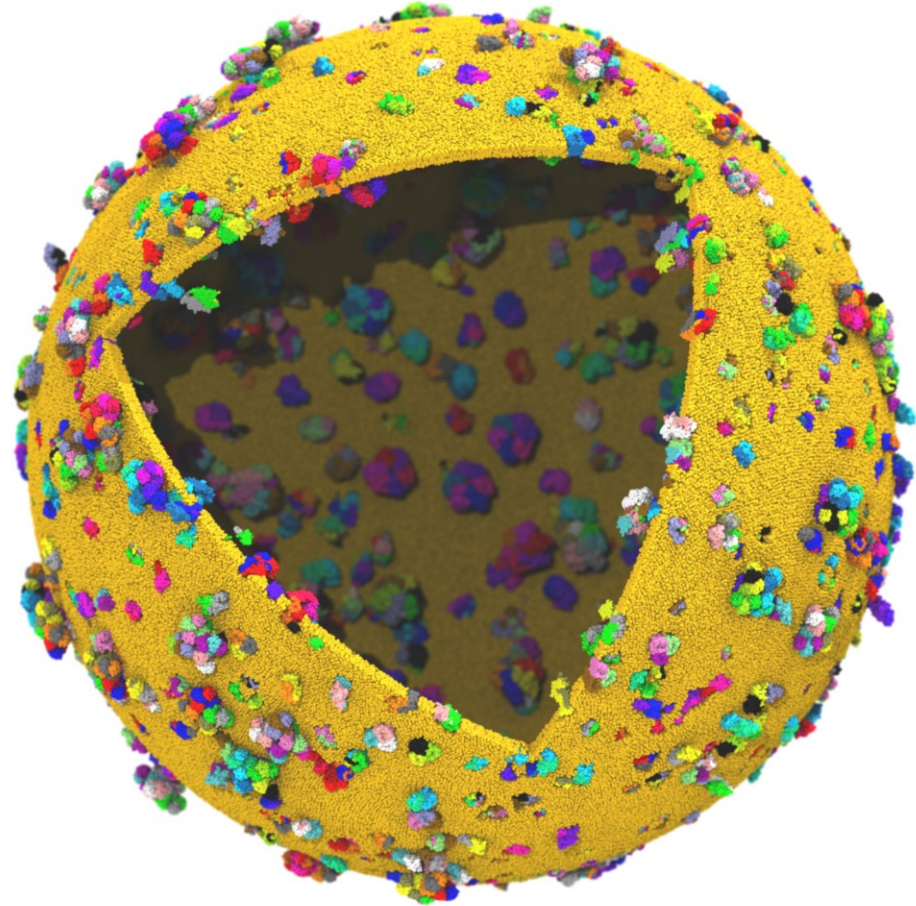
High Performance Molecular Visualization: In-Situ and Parallel Rendering with EGL.
J. E. Stone, P. Messmer, R. Sisneros, and K. Schulten. High Performance Data Analysis and Visualization Workshop, IEEE IPDPSW, pp. 1014-1023, 2016.



64M atom HIV-1 capsid simulation rendered via EGL

Next Generation: Simulating a Proto-Cell

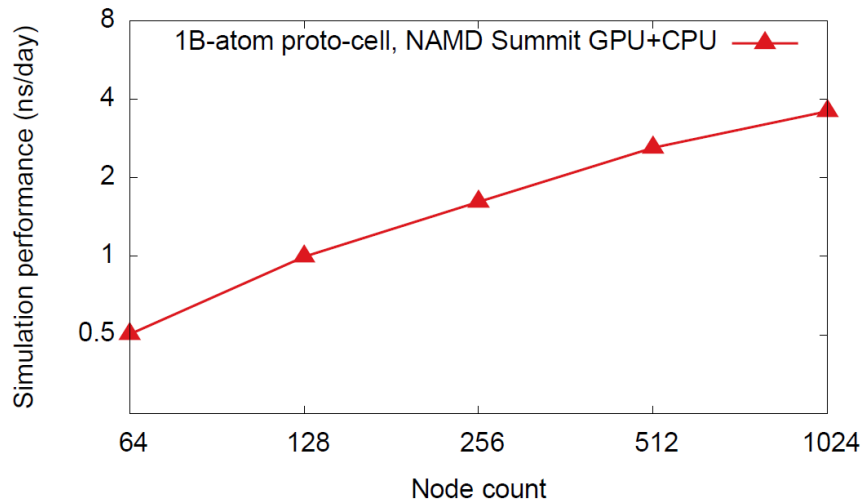
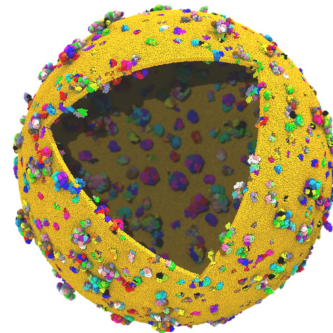
- Emulate aspects of the *Mycoplasma mycoides* bacterium
- 200nm diameter
- ~1 billion atoms w/ solvent
- ~1400 proteins in membrane



Cryo-ET image of ultra-small bacteria (scale bar 100nm)
Luef et al. Nature Comm., 6:6372, 2015.

Proto-Cell Data Challenges

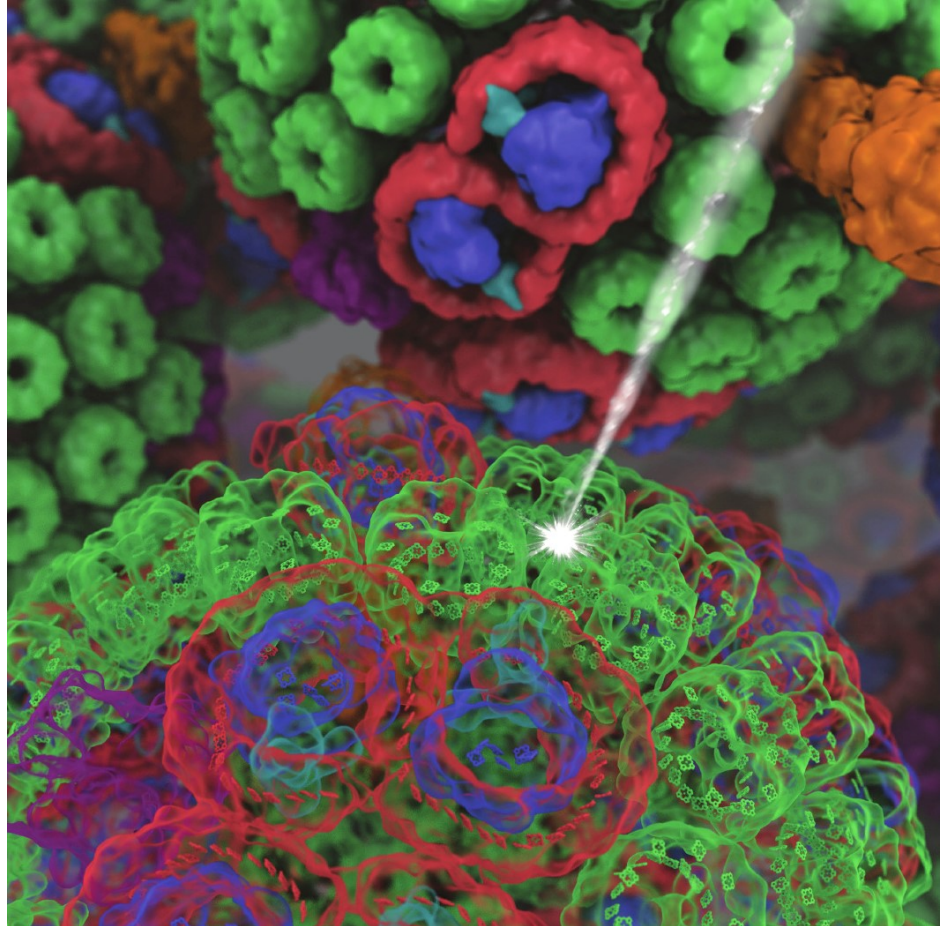
- **1B-atom proto-cell requires nodes with more than TB RAM to build complete model...**
- **1B-atom proto-cell binary structure file: 63GB**
- **Trajectory frame atomic coordinates: 12GB, 1.2TB/ns of simulation (1 frame per 10ps)**
- **Routine modeling and visualization tasks are a big challenge at this scale**
 - **Models contain thousands of atomic-detail components** that must work together in harmony
 - **Exploit persistent memory technologies** to enable “instant on” operation on massive cell-scale models – eliminate several minutes of startup during analysis/visualization of known structure
 - **Sparse output of results at multiple timescales** will help ameliorate visualization and analysis I/O
 - **Data quantization, compression, APIs like ZFP**



High Fidelity Ray Tracing

- **Interactive RT** on laptops, desk, cloud
- Large-scale parallel rendering: in situ or post hoc visualization tasks
- AO, DoF, instancing,
- Stereoscopic panorama and full-dome projections
- Omnidirectional VR: YouTube, HMDs

- Built-in ray tracing engines:
 - **Tachyon**: cross-platform RT
 - **NVIDIA OptiX**: GPU-accelerated and remote RT on VCA clusters
 - **Intel OSPRay**: CPU x86/Phi-optimized parallel rendering w/ MPI

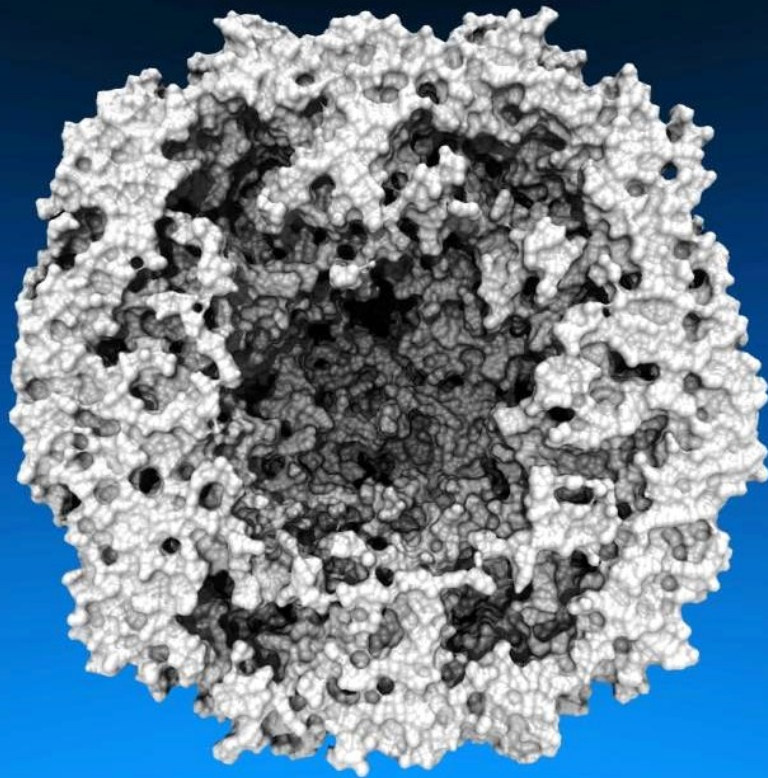
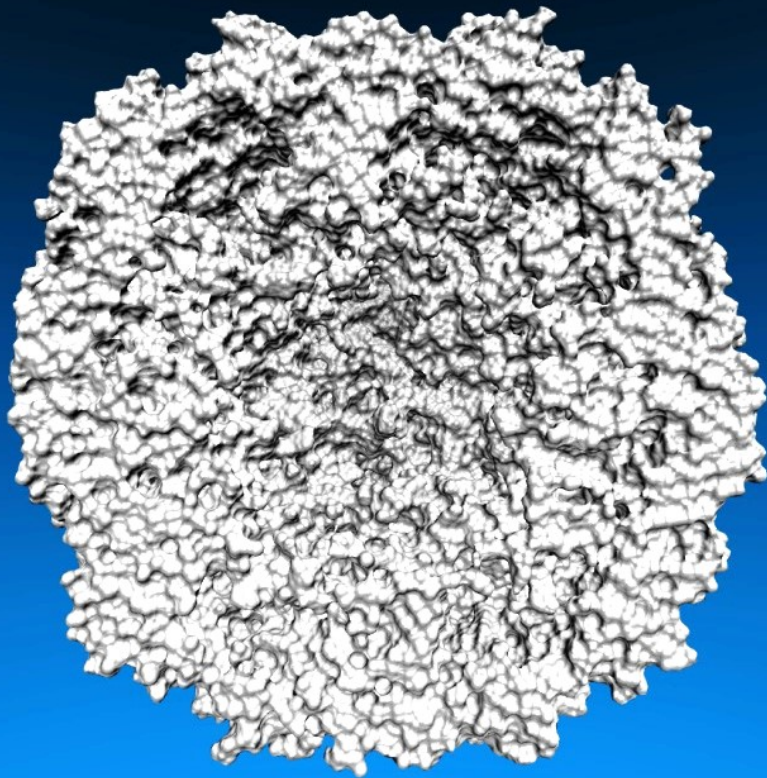


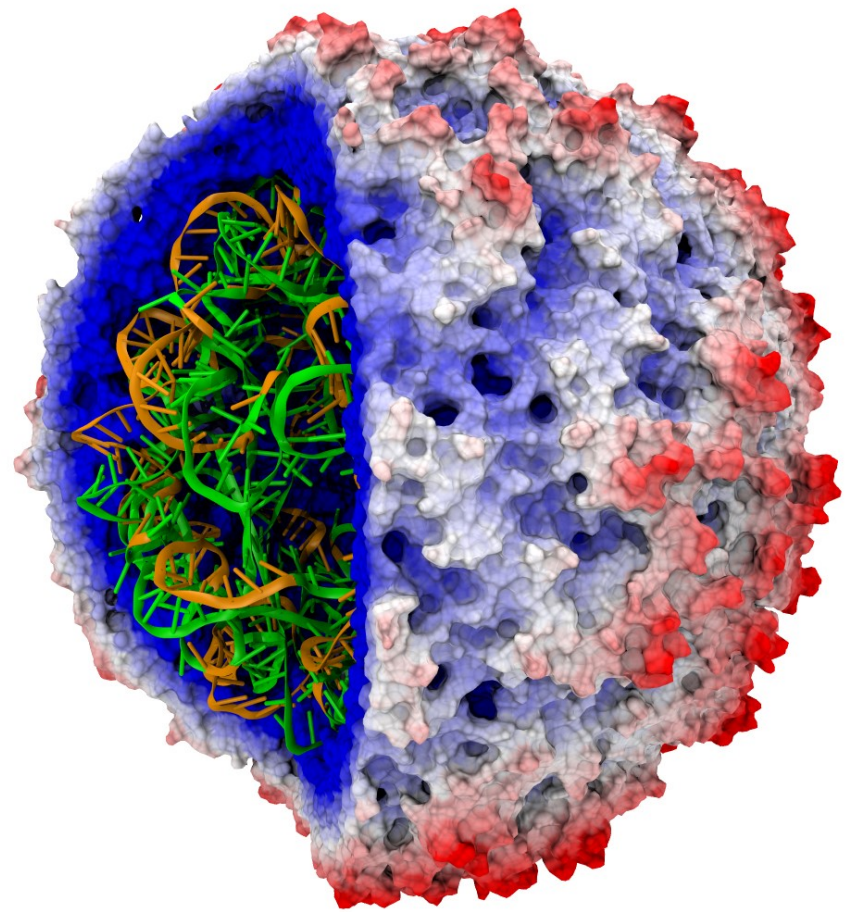
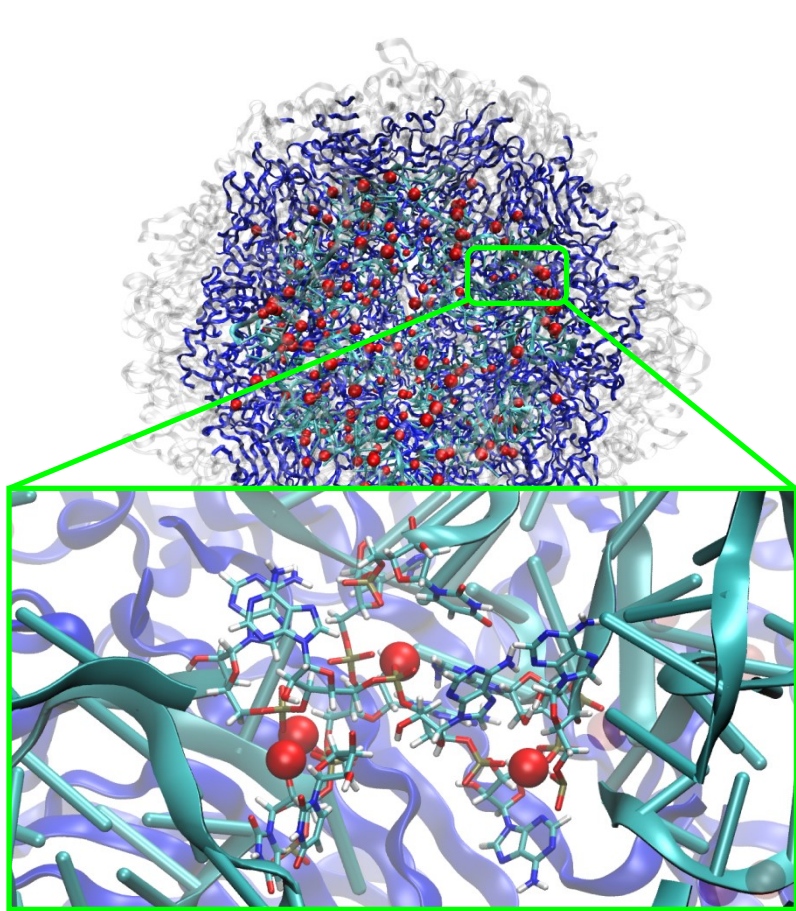
VMD/OptiX all-atom Chromatophore

Lighting Comparison, STMV Capsid

Two lights, no shadows

Ambient occlusion + two lights, 144 AO rays/hit





Satellite Tobacco Mosaic Virus

Biomedical Technology Research Center for Macromolecular Modeling and Bioinformatics
Beckman Institute, University of Illinois at Urbana-Champaign - www.ks.uiuc.edu

VMD w/ OptiX

- Interactive RT on laptops, desktops, and cloud
- Large-scale parallel rendering: in situ or post hoc visualization tasks
- Remote RT on NVIDIA VCA clusters
- Stereoscopic panoramic and full-dome projections
- **Omnidirectional VR for YouTube, VR HMDs**
- **GPU memory sharing via NVLink**

GPU-Accelerated Molecular Visualization on Petascale Supercomputing Platforms.

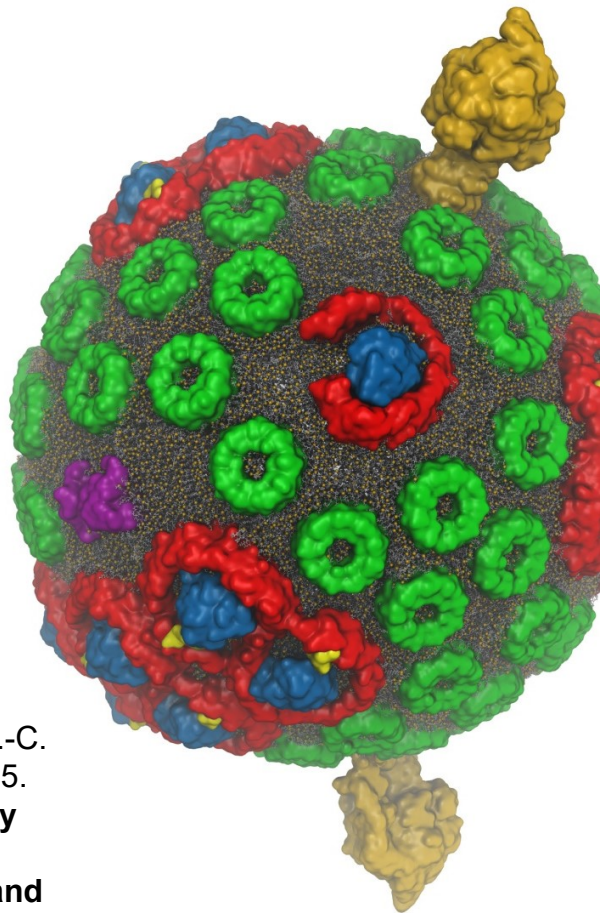
J. E. Stone, K. L. Vandivort, and K. Schulten. UltraVis'13, pp. 6:1-6:8, 2013.

Visualization of Energy Conversion Processes in a Light Harvesting Organelle at Atomic Detail. M. Sener, et al. SC'14 Visualization and Data Analytics Showcase, 2014.

Chemical Visualization of Human Pathogens: the Retroviral Capsids. J. R. Perilla, B.-C. Goh, J. E. Stone, and K. Schulten. SC'15 Visualization and Data Analytics Showcase, 2015.

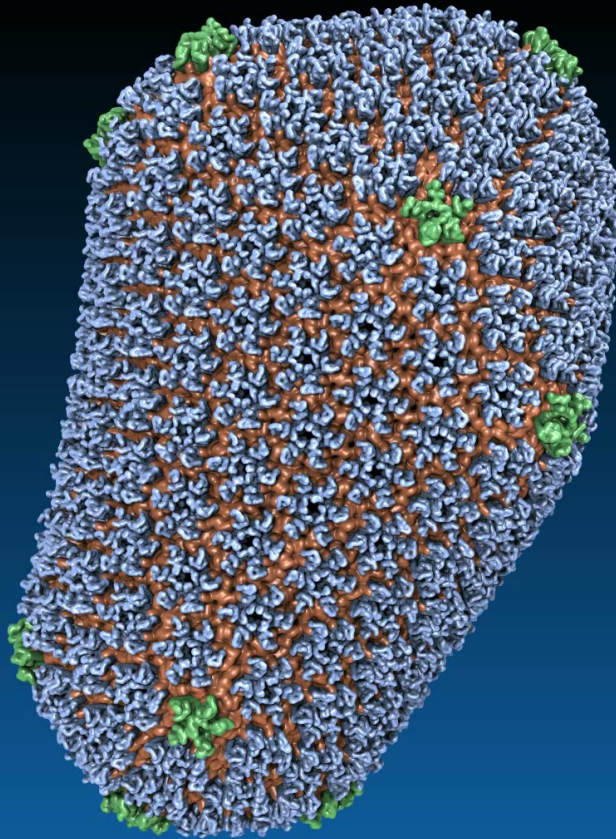
Atomic Detail Visualization of Photosynthetic Membranes with GPU-Accelerated Ray Tracing. J. E. Stone et al., J. Parallel Computing, 55:17-27, 2016.

Immersive Molecular Visualization with Omnidirectional Stereoscopic Ray Tracing and Remote Rendering J. E. Stone, W. R. Sherman, and K. Schulten. HPDAV, IPDPSW, pp. 1048-1057, 2016.

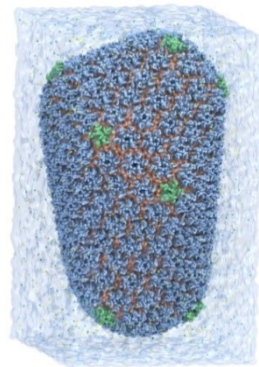


VMD/OptiX GPU Ray Tracing of
all-atom Chromatophore w/ lipids.

HIV-1 Capsid



HIV-1 Parallel HD Movie Rendering on Blue Waters Cray XE6/XK7

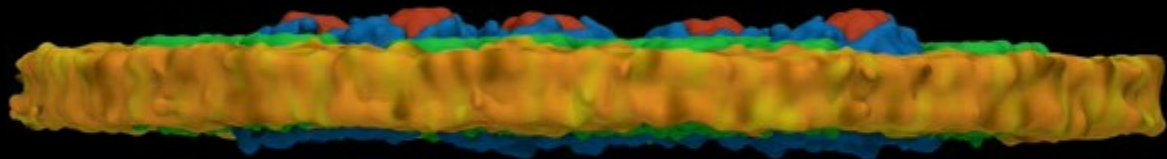


New VMD TachyonL-OptiX on XK7 vs. Tachyon on XE6:
K20X GPUs yield **up to twelve times** geom+ray tracing speedup

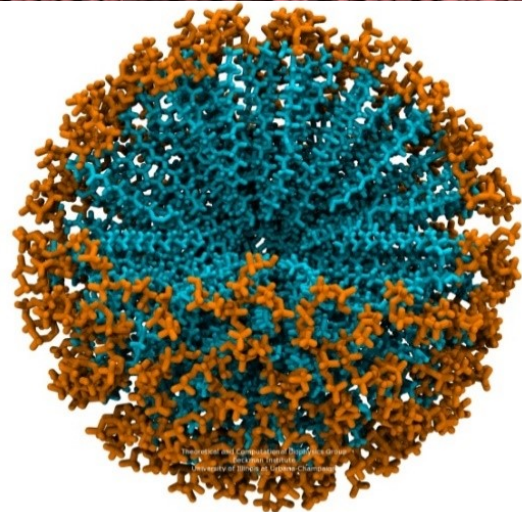
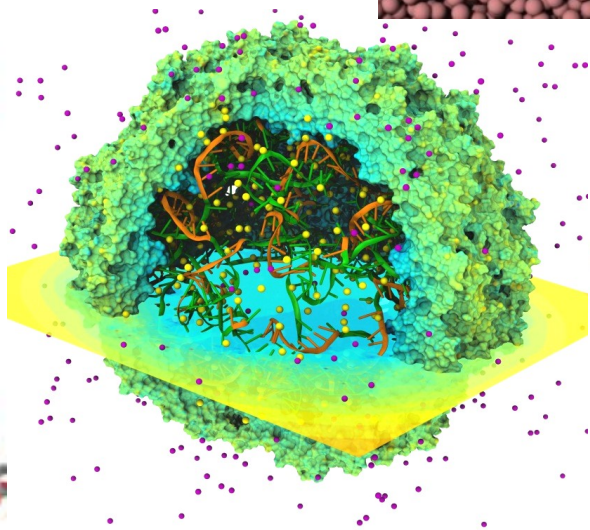
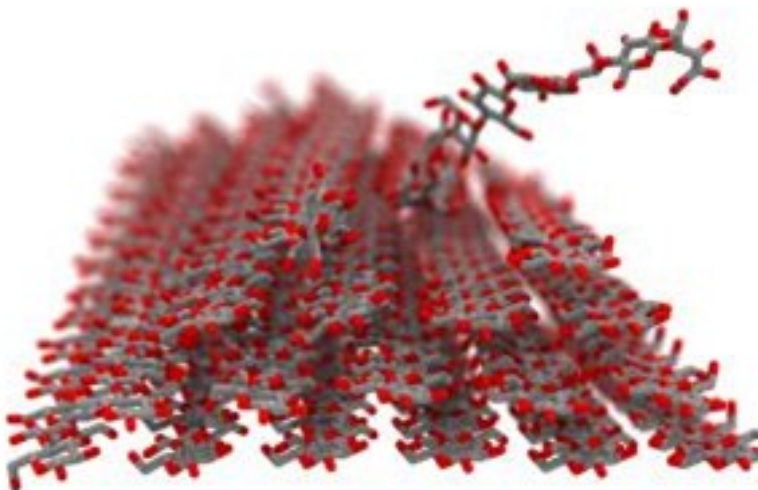
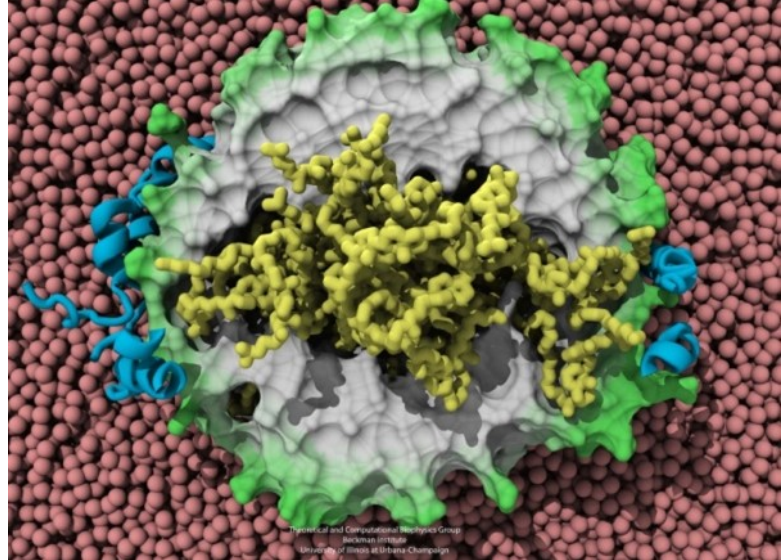
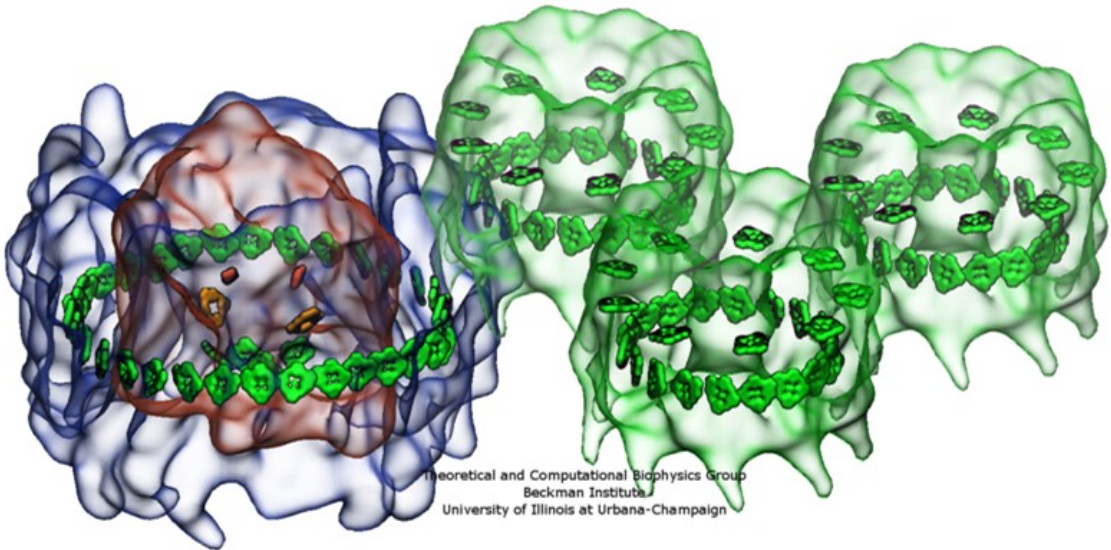
Ray Tracer Version	Node Type and Count	Script Load	State Load	Geometry + Ray Tracing	Total Time
New TachyonL-OptiX	64 XK7 Tesla K20X GPUs	2 s	39 s	435 s	476 s
New TachyonL-OptiX	128 XK7 Tesla K20X GPUs	3 s	62 s	230 s	295 s
TachyonL-OptiX [1]	64 XK7 Tesla K20X GPUs	2 s	38 s	655 s	695 s
TachyonL-OptiX [1]	128 XK7 Tesla K20X GPUs	4 s	74 s	331 s	410 s
TachyonL-OptiX [1]	256 XK7 Tesla K20X GPUs	7 s	110 s	171 s	288 s
Tachyon [1]	256 XE6 CPUs	7 s	160 s	1,374 s	1,541 s
Tachyon [1]	512 XE6 CPUs	13 s	211 s	808 s	1,032 s

[1] GPU-Accelerated Molecular Visualization on Petascale Supercomputing Platforms.

J. E. Stone, K. L. Vandivort, and K. Schulten. UltraVis'13: Proceedings of the 8th International Workshop on Ultrascale Visualization, pp. 6:1-6:8, 2013.



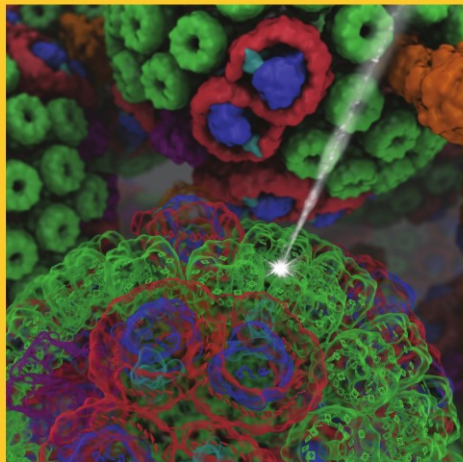
20 M atom chromatophore patch



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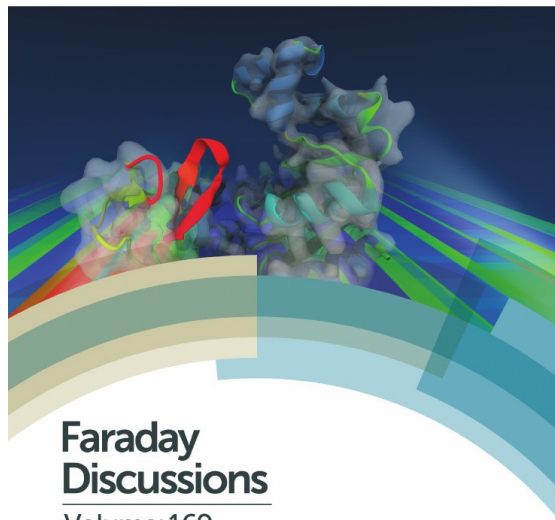
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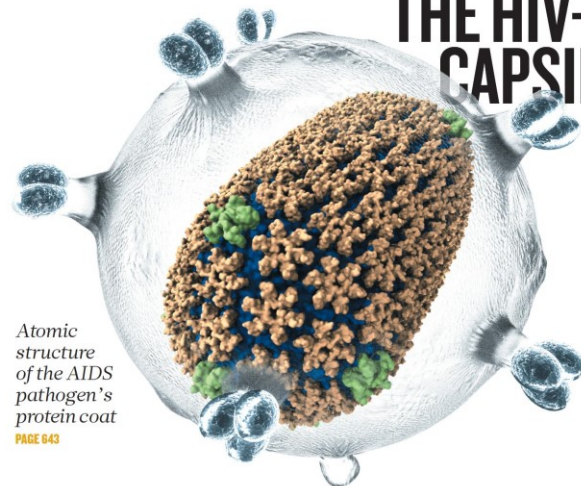
Molecular Simulations and Visualization



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THE HIV-1 CAPSID



Atomic
structure
of the AIDS
pathogen's
protein coat

PAGE 643

COSMOLOGY
**THE FIRST
LIGHT**
*In pursuit of the most
distant galaxies*
PAGE 554

CITATION
**CROSSING THE
BORDERS**
*International collaborations
make the most impact*
PAGE 557

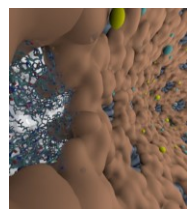
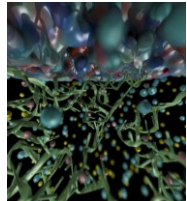
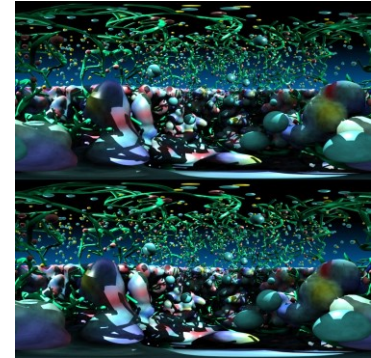
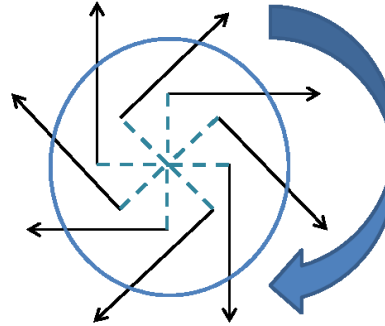
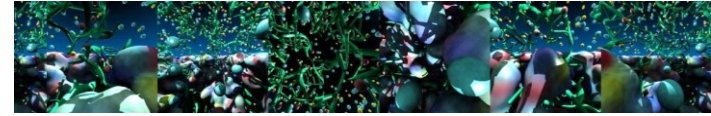
ANTICANCER DRUGS
**A SITTING
TARGET**
*An indirect hit on
'undruggable' KRAS protein*
PAGES 577 & 638

NATURE.COM/NATURE
30 May 2013



Omnidirectional Stereoscopic Ray Tracing

- **Ray trace 360° images and movies for Desk and VR HMDs: Oculus, Vive, Cardboard**
- Stereo spheremaps or cubemaps allow very high-frame-rate interactive OpenGL display
- **AO lighting, depth of field, shadows, transparency, curved geometry, ...**
- **Summit 6x Tesla V100 GPU nodes:**
 - Render many omni-stereo viewpoints, no acceleration structure rebuilds, tens of frames/sec per-node!
 - OptiX multi-GPU rendering, NVLink compositing and data distribution, etc...
 - Future: AI for warping between views

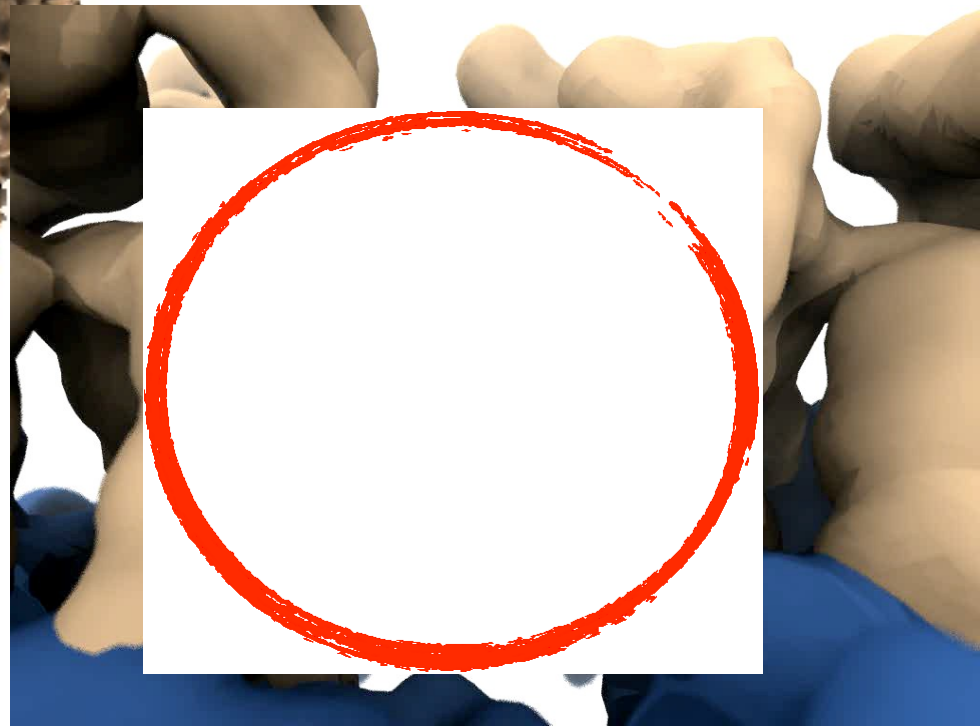
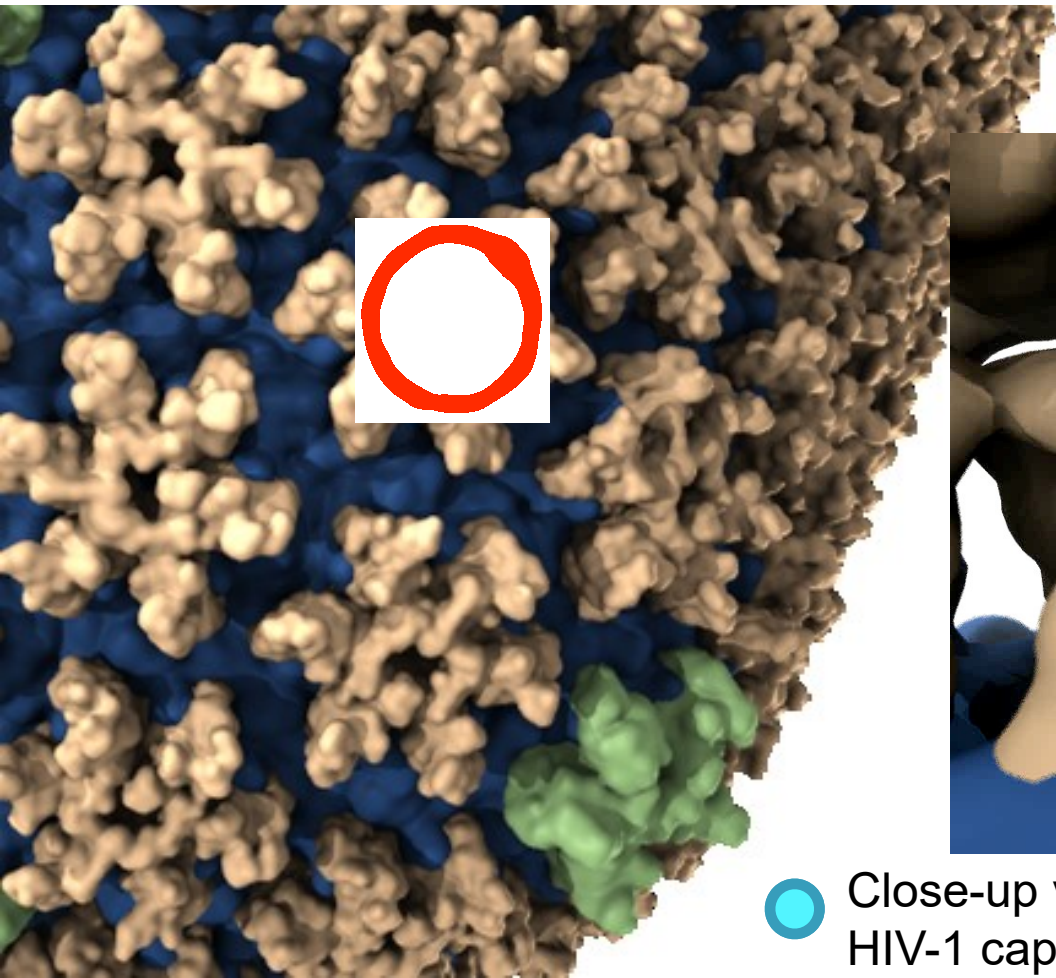


Atomic Detail Visualization of Photosynthetic Membranes with GPU-Accelerated Ray Tracing. J. E. Stone, et al. J. Parallel Computing, 55:17-27, 2016.

Immersive Molecular Visualization with Omnidirectional Stereoscopic Ray Tracing and Remote Rendering. J. E. Stone, W. R. Sherman, and K. Schulten. High Performance Data Analysis and Visualization Workshop, IEEE International Parallel and Distributed Processing Symposium Workshops (IPDPSW), pp. 1048-1057, 2016.

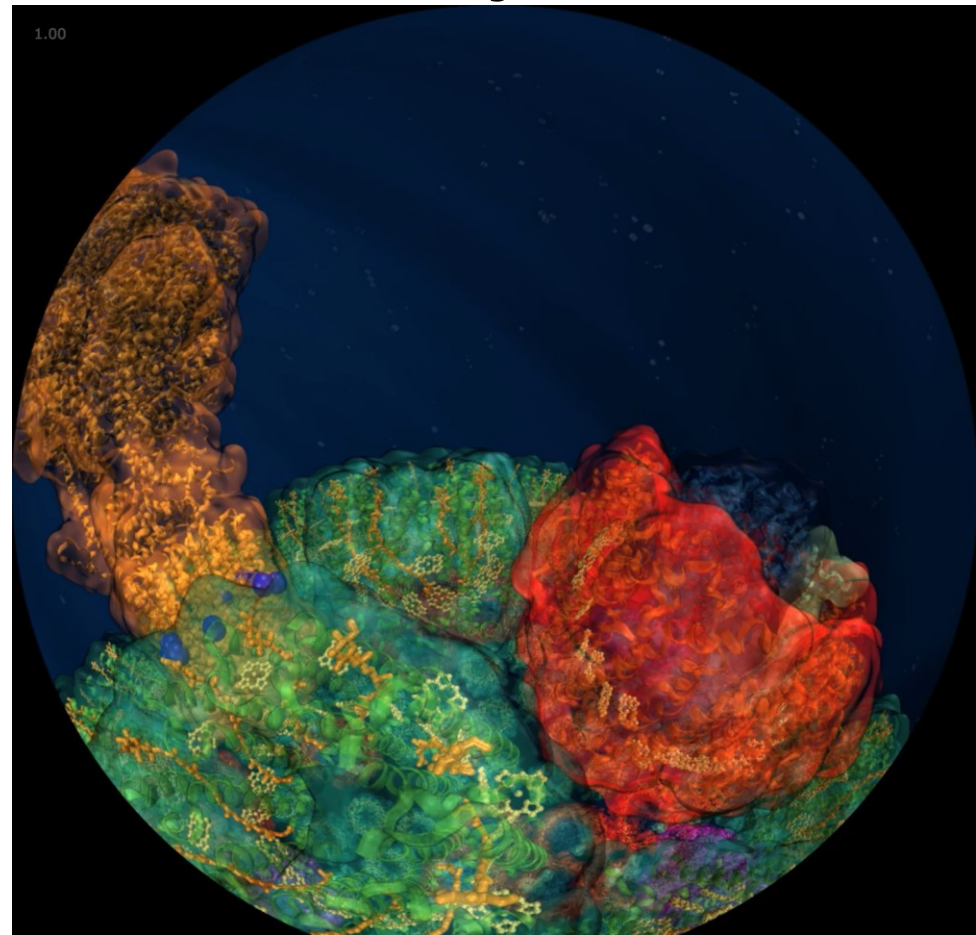
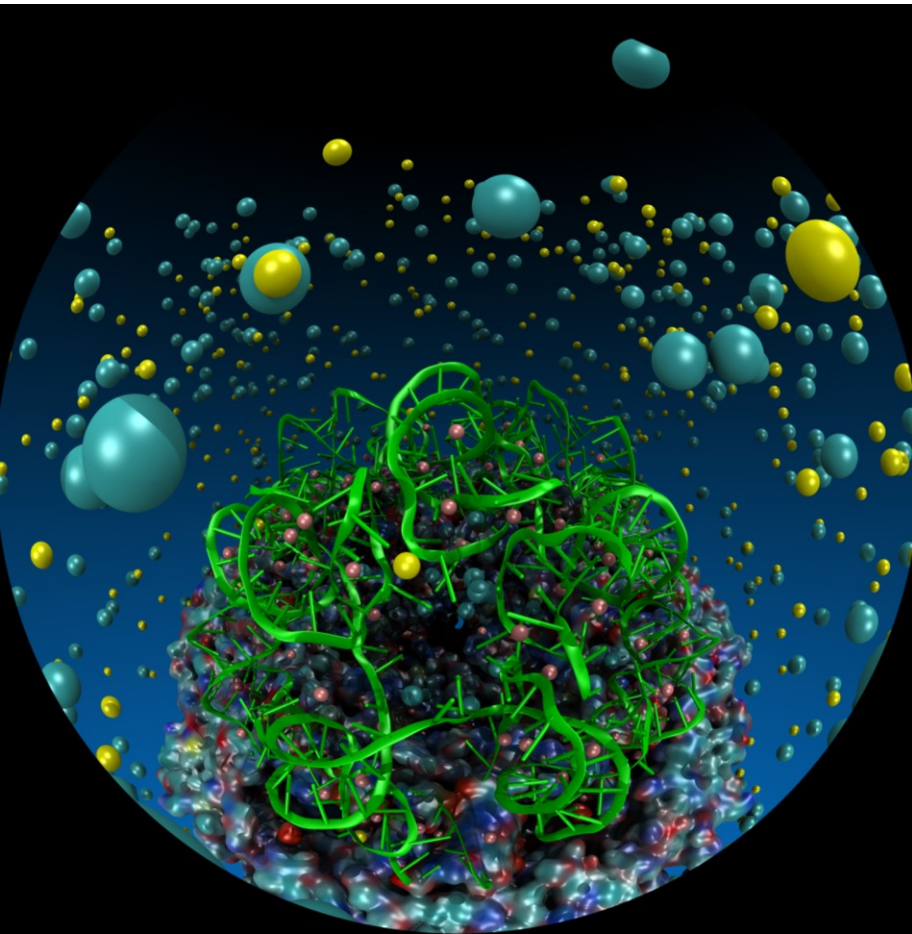
Goal: Intuitive interactive viz. in crowded molecular complexes

Results from 64 M atom, 1 μ s sim!



Close-up view of chloride ions permeating through HIV-1 capsid hexameric centers

Planetarium Dome Master Projections



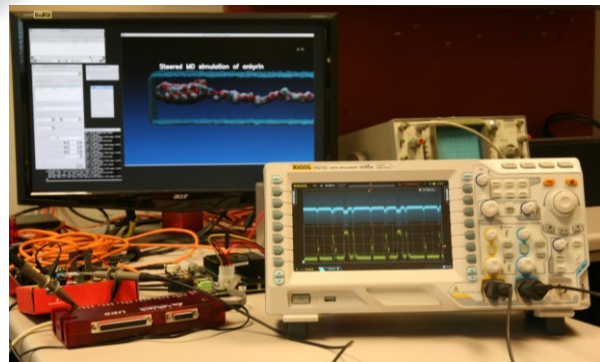
Technology Opportunities and Collaborations

- Supercomputer Centers, Cray, IBM
 - Remote visualization
 - Performance, power profiling and optimization
- NVIDIA
 - **GPU computing**
 - **Ray tracing**
 - **Remote visualization**
 - ARM, Tablets, power profiling and optimization
- Intel
 - **x86, Xeon Phi optimization**
 - **Ray tracing**
- Amazon
 - Cloud deployment of VMD/NAMD, related tools
 - Remote visualization
- Universities:
 - **G. Fiorin, J. Henin, Toni Giorgino, collective variables**
 - **T. Ertl, U. Stuttgart: visualization algorithms**
 - M. Kuttel, U. Cape Town: visualization and analysis
 - W. Sherman, Indiana U.: VR HMDs, visualization

GPU computing,
Ray tracing,
Remote viz.



VR HMDs, 6DoF input devices

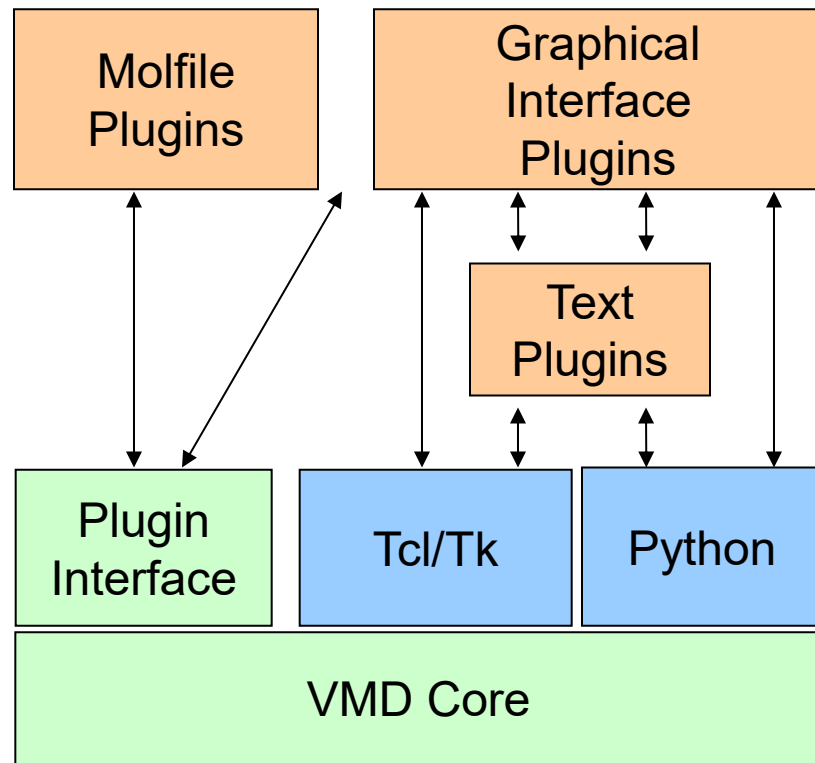


Energy efficiency: ARM+GPU

VMD is a Platform for Developing Research Tools

Over 110 VMD Plugins, Half Developed by Users

- VMD user-extensible scripting w/ Tcl/Tk, Python
- User-developed plugins:
 - Alanine Scanning
 - Collective Variable Analyzer
 - Clustering Tool
 - Carbon Nanostructure Builder
 - TorsionPlot
 - RMSD Trajectory Tool
 - Many others...



QwikMD: Guided MD Simulation and Training

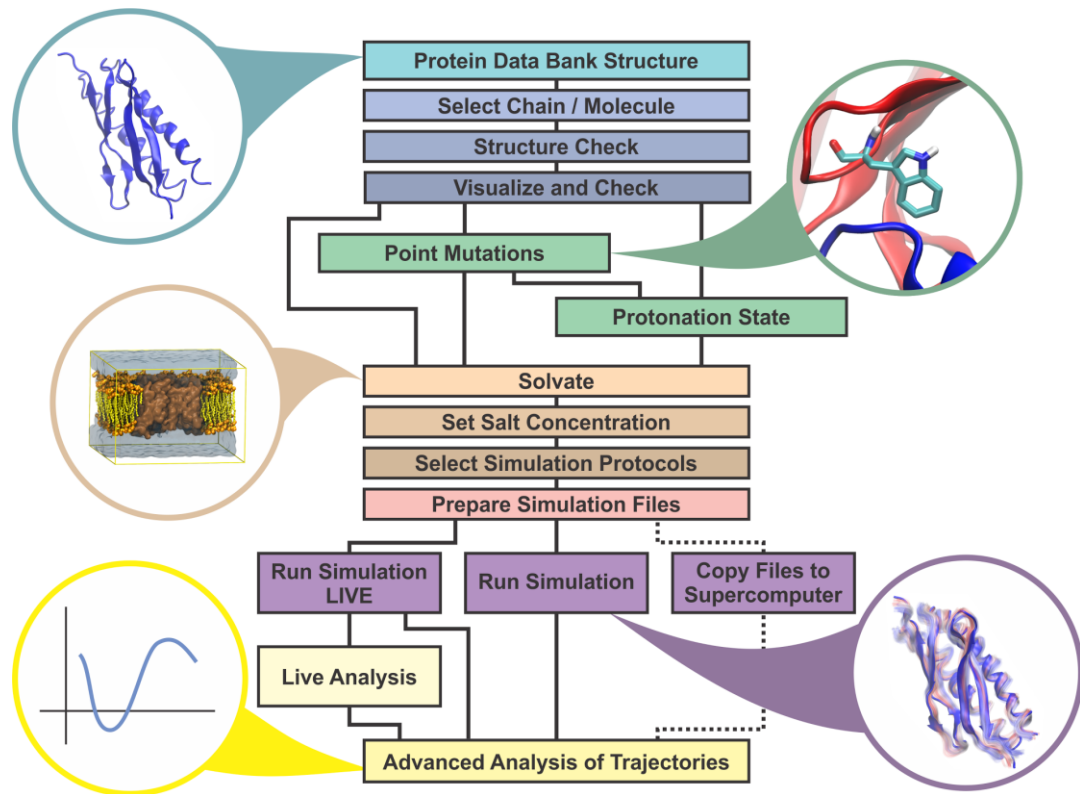
Smooths initial learning curve (non-expert users)

Training: used in 4 Center workshops to-date

Speed up tedious simulation preparation tasks (expert users)

Reproducibility:
detailed log of all steps

Interactive preparation, simulation, and analysis



Selected VMD Plugins: Center Developed, and User Developed

Analysis

APBSRun
CatDCD
Contact Map
[GofRGUI](#)
[HeatMapper](#)
ILSTools
[IRSpecGUI](#)
MultiSeq
NAMD Energy
NAMD Plot
NetworkView
[NMWiz](#)
[ParseFEP](#)
PBCTools
PMEpot
[PropKa GUI](#)
RamaPlot
RMSD Tool
[RMSD Trajectory Tool](#)
[RMSD Visualizer Tool](#)
Salt Bridges
Sequence Viewer
Symmetry Tool
Timeline
[TorsionPlot](#)
VolMap

Modeling

AutoIonize
AutoPSF
Chirality
Cionize
Cispeptide
CGTools
Dowser
ffTK
Inorganic Builder
MDFF
Membrane
Merge Structs
Molefacture
Mutator
[Nanotube](#)
Psfgen
[RESPTool](#)
RNAView
Solvate
SSRestrains
Topotools

Visualization

Clipping Plane Tool
[Clone Rep](#)
DemoMaster
[Dipole Watcher](#)
[Intersurf](#)
[Navigate](#)
NavFly
[MultiMolAnim](#)
Color Scale Bar
Remote
Palette Tool
ViewChangeRender
ViewMaster
[Virtual DNA Viewer](#)
VMD Movie Maker

Simulation

AlaScan
AutoIMD
IMDMenu
NAMD GUI
NAMD Server
QMTTool

Collaboration

Remote Control

Data Import and Plotting

Data Import
Multiplot
PDBTool
MultiText

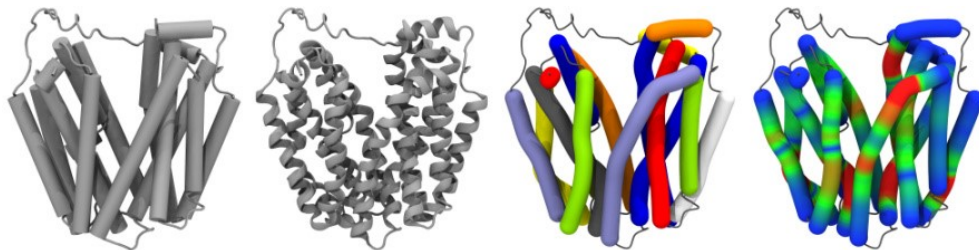
Externally Hosted Plugins and Extensions

[Check sidechains](#)
[MultiMSMS](#)
[Interactive Essential Dynamics](#)
[Mead Ionize](#)
[Clustering Tool](#)
[iTrajComp](#)
[Swap RMSD](#)
[Intervor](#)
[SurfVol](#)
[vmdICE](#)

75 MolFile I/O Plugins:
structure, trajectory, sequence,
and density map

<http://www.ks.uiuc.edu/Research/vmd/plugins/>

Example VMD Visualization and Analysis Plugins



Bendix

Dahl ACE, Chavent M and Sansom MSP Bendix: intuitive helix geometry analysis and abstraction. *Bioinformatics* 2012 28(16): 2193-2194.



Normal Mode Wizard

Bakan A, Meireles LM, Bahar I ProDy: Protein Dynamics Inferred from Theory and Experiments. *Bioinformatics* 2011 27(11):1575-1577.

Making Our Research Tools Easily Accessible

- Cloud based deployment
 - Full virtual machines (known as “AMI” in Amazon terminology)
 - Amazon AWS EC2 GPU-accelerated instances:
<http://www.ks.uiuc.edu/Research/cloud/>
- Docker “container” images available in NVIDIA NGC registry
 - Users obtain Docker images via registry, download and run on the laptop, workstation, cloud, or supercomputer of their choosing
 - <https://ngc.nvidia.com/registry/>
 - <https://ngc.nvidia.com/registry/hpc-vmd>



Clusters, Supercomputers

Workstations,
Servers,
Cloud



Molecular dynamics-based refinement and validation for sub-5 Å cryo-electron microscopy maps. Abhishek Singharoy, Ivan Teo, Ryan McGreevy, John E. Stone, Jianhua Zhao, and Klaus Schulten. *eLife*, 10.7554/eLife.16105, 2016. (66 pages).

QwikMD-integrative molecular dynamics toolkit for novices and experts. Joao V. Ribeiro, Rafael C. Bernardi, Till Rudack, John E. Stone, James C. Phillips, Peter L. Freddolino, and Klaus Schulten. *Scientific Reports*, 6:26536, 2016.

High performance molecular visualization: In-situ and parallel rendering with EGL. John E. Stone, Peter Messmer, Robert Sisneros, and Klaus Schulten. *2016 IEEE International Parallel and Distributed Processing Symposium Workshop (IPDPSW)*, pp. 1014-1023, 2016.

Molecular Dynamics Flexible Fitting (MDFF)

X-ray crystallography



APS at Argonne

MDFF



Electron microscopy

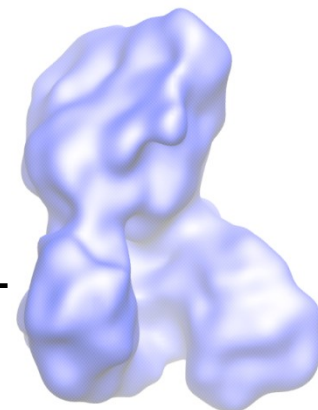


FEI microscope

ORNL Titan

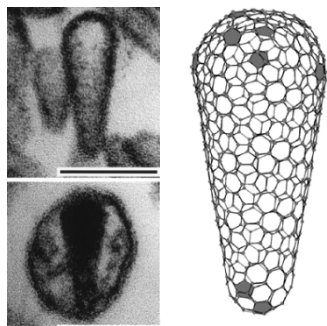


Molecular dynamics-based refinement and validation for sub-5Å cryo-electron microscopy maps. A. Singharoy, I. Teo, R. McGreevy, J. E. Stone, J. Zhao, and K. Schulten. *eLife* 2016;10.7554/eLife.16105

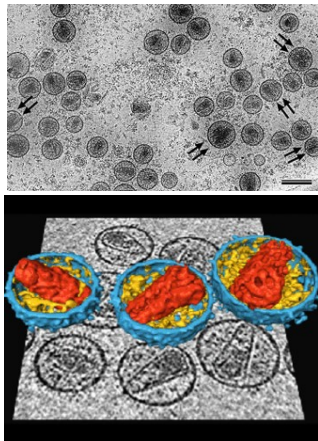


Structural Route to the all-atom HIV-1 Capsid

1st TEM (1999) 1st tomography (2003)

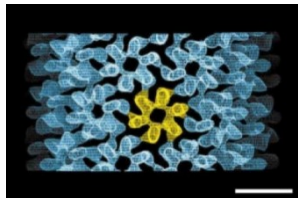


Ganser et al. *Science*, 1999
Briggs et al. *EMBO J*, 2003
Briggs et al. *Structure*, 2006

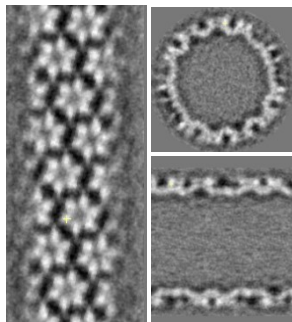


cryo-ET (2006)

hexameric tubule

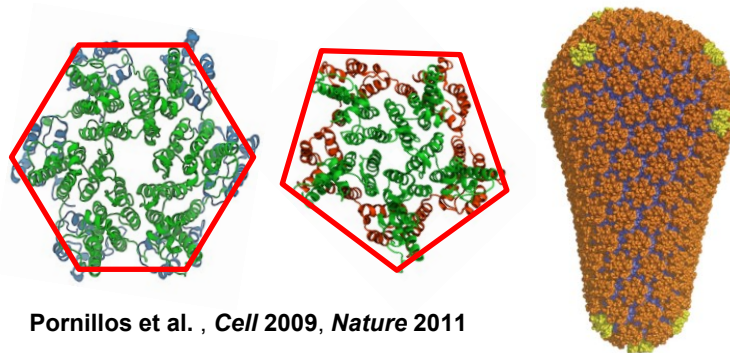


Li et al., *Nature*, 2000



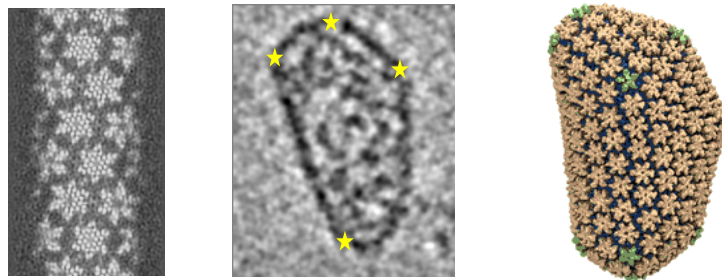
Byeon et al., *Cell* 2009

Crystal structures of separated hexamer and pentamer



Pornillos et al. , *Cell* 2009, *Nature* 2011

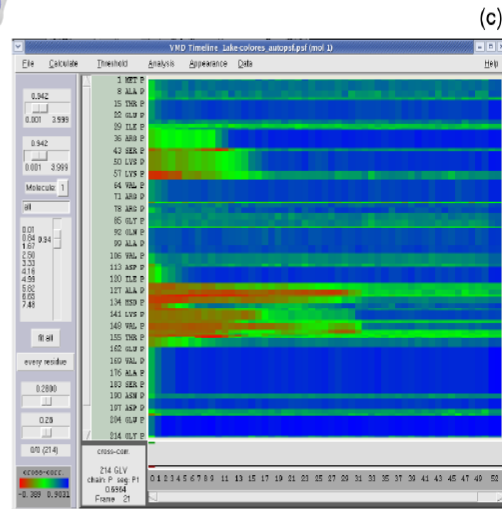
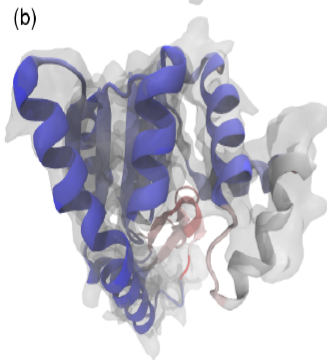
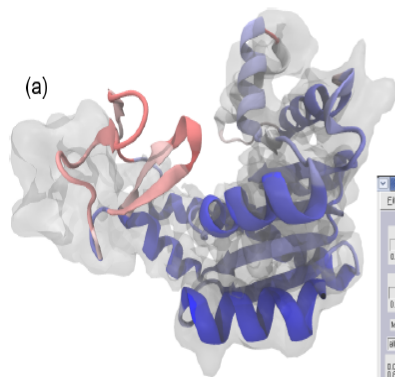
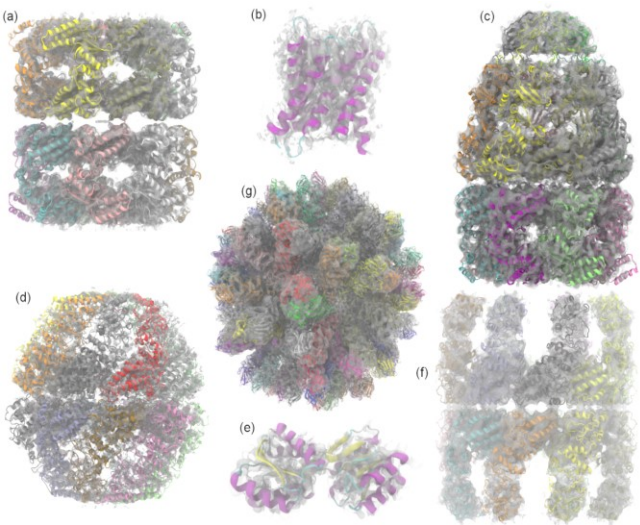
High res. EM of hexameric tubule, tomography of capsid,
all-atom model of capsid by MDFF w/ NAMD & VMD,
NSF/NCSA Blue Waters computer at Illinois



Zhao et al. , *Nature* 497: 643-646 (2013)

Evaluating Quality-of-Fit for Structures Solved by Hybrid Fitting Methods

Compute Pearson correlation to evaluate quality-of-fit between a reference cryo-EM density map and a **simulated density map** from an **all-atom structure**.



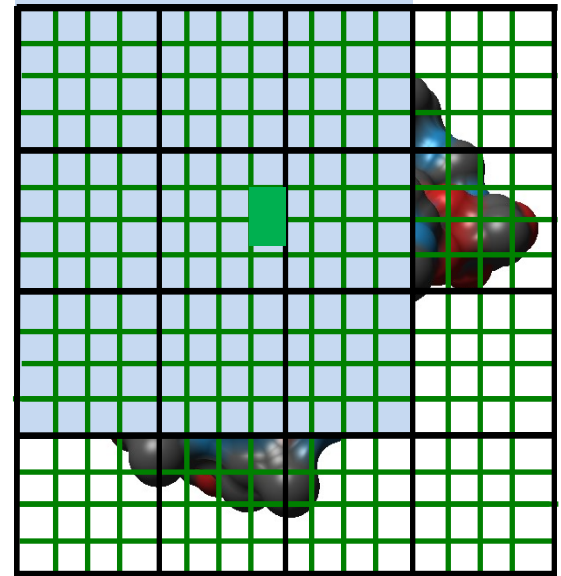
MDFF Cross Correlation Timeline
Regions with poor fit **Regions with good fit**

MDFFF Density Map Algorithm

- Build spatial acceleration data structures, optimize data for GPU
- Compute 3-D density map:

$$\rho(\vec{r}; \vec{r}_1, \vec{r}_2, \dots, \vec{r}_N) = \sum_{i=1}^N e^{-\frac{|\vec{r}-\vec{r}_i|^2}{2\alpha^2}}$$

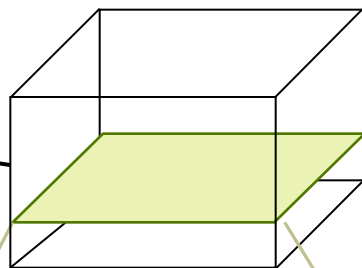
- Truncated Gaussian and spatial acceleration grid ensure linear time-complexity



3-D density map lattice point and the neighboring spatial acceleration cells it references

Single-Pass MDFF GPU Cross-Correlation

3-D density map decomposes into 3-D grid of 8x8x8 tiles containing CC partial sums and local CC values

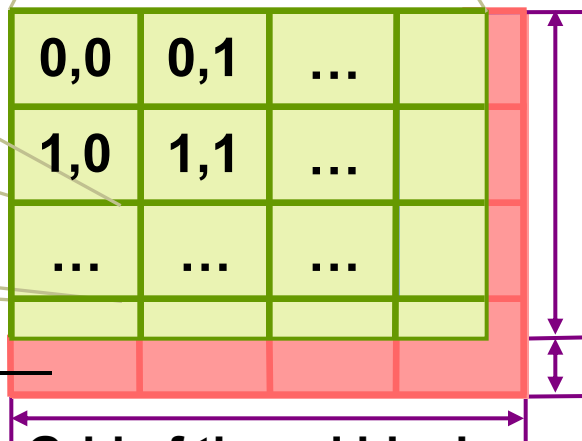
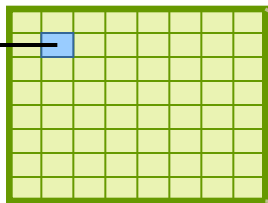


Fusion of density and CC calculations into a single CUDA kernel!!!

Spatial CC map and overall CC value computed in a single pass

Small 8x8x2 CUDA thread blocks afford large per-thread register count, shared memory

Each thread computes 4 z-axis density map lattice points and associated CC partial sums



Threads producing results that are used

Inactive threads, region of discarded output

Padding optimizes global memory performance, guaranteeing coalesced global memory accesses

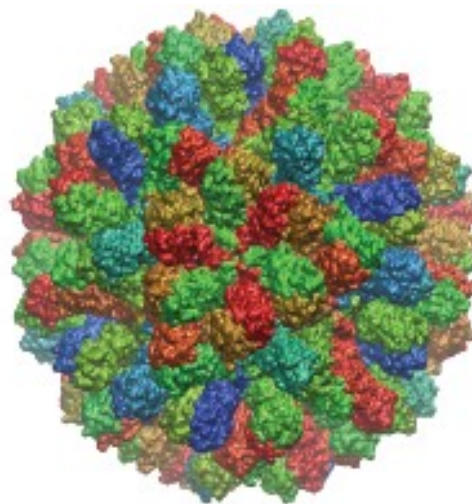
Grid of thread blocks

Parallel MDFF Cross Correlation Analysis on Cray XK7

Rabbit Hemorrhagic Disease Virus (RHDV)

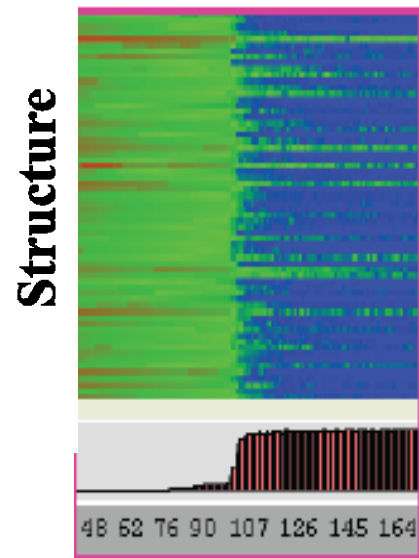
Traj. frames	10,000
Structure component selections	720
Single-node XK7 (projected)	336 hours (14 days)
128-node XK7	3.2 hours 105x speedup
2048-node XK7	19.5 minutes 1035x speedup

Calculation of 7M CCs would take **5 years** using serial CPU algorithm!



**RHDV colored
by relative CC**

Relative CC
-0.0032 0.02



Time

VMD Tesla V100 Cross Correlation Performance

Rabbit Hemorrhagic Disease Virus: 702K atoms, 6.5Å resolution

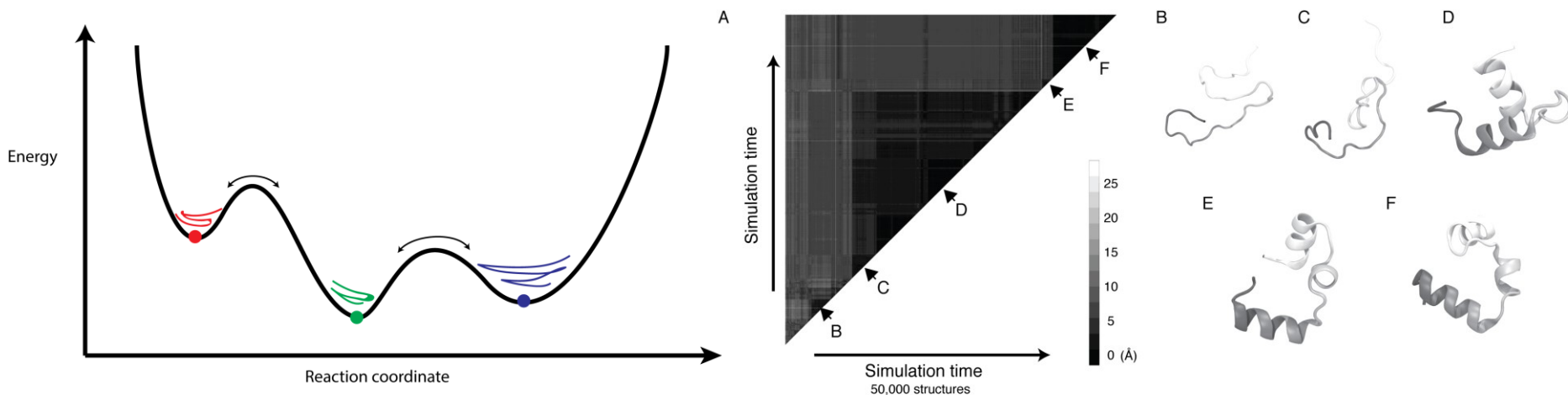
Volta GPU architecture almost 2x faster than previous gen Pascal:

Application and Hardware platform	Runtime, Speedup vs. Chimera, VMD+GPU
Chimera Xeon E5-2687W (2 socket) [1]	15.860s, 1x
VMD-CUDA IBM Power8 + 1x Tesla K40 [2]	0.488s, 32x 0.9x
VMD-CUDA Intel Xeon E5-2687W + 1x Quadro K6000 [1,2]	0.458s, 35x 1.0x
VMD-CUDA Intel Xeon E5-2698v3 + 1x Tesla P100	0.090s, 176x 5.1x
VMD-CUDA IBM Power8 “Minsky” + 1x Tesla P100	0.080s, 198x 5.7x
VMD-CUDA Intel Xeon E5-2697Av4 + 1x Tesla V100	0.050s, 317x 9.2x
VMD-CUDA IBM Power9 “Newell” + 1x Tesla V100	0.049s, 323x 9.3x

[1] GPU-Accelerated Analysis and Visualization of Large Structures Solved by Molecular Dynamics Flexible Fitting. J. E. Stone, R. McGreevy, B. Isralewitz, and K. Schulten. Faraday Discussions 169:265-283, 2014.

[2] Early Experiences Porting the NAMD and VMD Molecular Simulation and Analysis Software to GPU-Accelerated OpenPOWER Platforms. J. E. Stone, A.-P. Hynninen, J. C. Phillips, K. Schulten. International Workshop on OpenPOWER for HPC (IWOPH'16), LNCS 9945, pp. 188-206, 2016.

Clustering Analysis of Molecular Dynamics Trajectories: Requires I/O+Memory for All-Pairs of Trajectory Frames



GPU-Accelerated Molecular Dynamics Clustering Analysis with OpenACC. J.E. Stone, J.R. Perilla, C. K. Cassidy, and K. Schulten. In, Robert Farber, ed., *Parallel Programming with OpenACC*, Morgan Kaufmann, Chapter 11, pp. 215-240, 2016.

Use of Node-Local Burst Buffers and Non-Volatile Memory DIMMs

- **Perform viz+analysis in-transit in node-local SSDs, persistent memory NVDIMMs**
- ORNL Summit I/O:
 - Parallel FS: 2.5 TB/s
 - **Node-local PCIe “burst buffer” SSDs: 10+ TB/sec, 7PB capacity**
- Plenty of capacity for full-detail MD trajectories, **could enable ~100x increase in temporal resolution** in cases where it would be valuable to the science
- **Enable all-pairs trajectory clustering analyses and resulting visualizations**
- Future systems with NVDIMMs (3D Xpoint, phase change memory) could eventually provide bandwidths approaching DRAM
- Use NVDIMMs w/ **mmap()**, **APIs like PMDK** to perform formerly-out-of-core calculations using persistent memory:
<https://github.com/pmem/pmdk>
- **Imagine future Summit-like machines w/ NVLink-connected GPUs w/ access to high-bandwidth persistent memory on each node**

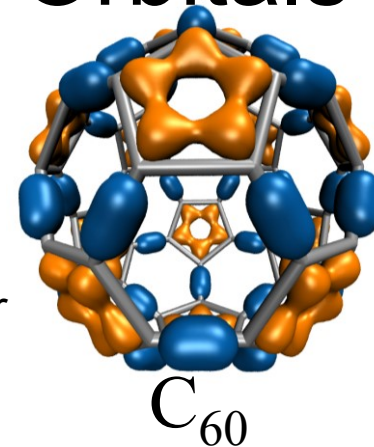
Trade FLOPS for Reduced I/O

ORNL Summit compute node:

- 6x Tesla V100 GPUs, 2x POWER9 CPUs
- GPUs Peak: ~46 DP TFLOPS, ~96 SP TFLOPS
- Peak IB rate per node: ~23GB/sec
- **Ratio of FLOPS vs. I/O:**
 - ~2,000 DP FLOPS/byte, ~4000 SP FLOPS/byte
 - ~16K FLOPS per FP word**

Unconventional approach: Recompute to avoid I/O

Computing+Visualizing Molecular Orbitals



- Movies of simulation trajectories provides insight into results
- QM, and hybrid (QM/MM) MO visualizations historically done from huge “cube” files, impractical
- Store QM wavefunctions + Gaussian basis set, only 10s of KB per stored timestep compared to 100s of MB
- **Recompute MO grid on-the-fly from QM basis set, huge decrease in RAM+I/O in exchange for heavy FP arithmetic**

NAMD goes quantum: An integrative suite for hybrid simulations. Melo, M. C. R.; Bernardi, R. C.; Rudack T.; Scheurer, M.; Riplinger, C.; Phillips, J. C.; Maia, J. D. C.; Rocha, G. D.; Ribeiro, J. V.; Stone, J. E.; Neese, F.; Schulten, K.; Luthey-Schulten, Z.; **Nature Methods**, 2018.

<http://dx.doi.org/10.1038/nmeth.4638>

High Performance Computation and Interactive Display of Molecular Orbitals on GPUs and Multi-core CPUs. J. E. Stone, J. Saam, D. Hardy, K. Vandivort, W. Hwu, K. Schulten, *2nd Workshop on General-Purpose Computation on Graphics Processing Units (GPGPU-2)*, ACM International Conference Proceeding Series, volume 383, pp. 9-18, 2009.

MO Kernel for One Grid Point (Naive C)

```
...
for (at=0; at<numatoms; at++) {
    int prim_counter = atom_basis[at];
    calc_distances_to_atom(&atompos[at], &xdist, &ydist, &zdist, &dist2, &xdiv);
    for (contracted_gto=0.0f, shell=0; shell < num_shells_per_atom[at]; shell++) {
        int shell_type = shell_symmetry[shell_counter];
        for (prim=0; prim < num_prim_per_shell[shell_counter]; prim++) {
            float exponent      = basis_array[prim_counter    ];
            float contract_coeff = basis_array[prim_counter + 1];
            contracted_gto += contract_coeff * expf(-exponent*dist2);
            prim_counter += 2;
        }
        for (tmpshell=0.0f, j=0, zdp=1.0f; j<=shell_type; j++, zdp*=zdist) {
            int imax = shell_type - j;
            for (i=0, ydp=1.0f, xdp=pow(xdist, imax); i<=imax; i++, ydp*=ydist, xdp*=xdiv)
                tmpshell += wave_f[ifunc++] * xdp * ydp * zdp;
        }
        value += tmpshell * contracted_gto;
        shell_counter++;
    }
}
}.....
```

Loop over atoms

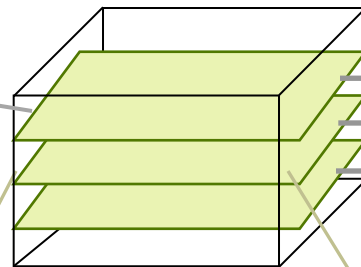
Loop over shells

Loop over primitives:
largest component of
runtime, due to **expf()**

Loop over angular
momenta
(unrolled in real code)

MO GPU Parallel Decomposition

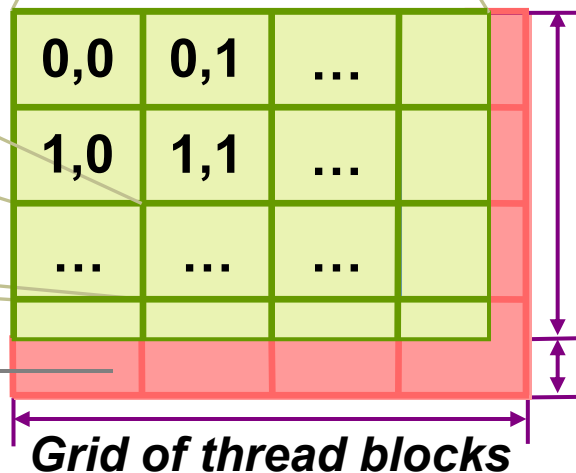
MO 3-D lattice decomposes into 2-D slices (CUDA grids)



...
GPU 2
GPU 1
GPU 0

Lattice computed using multiple GPUs

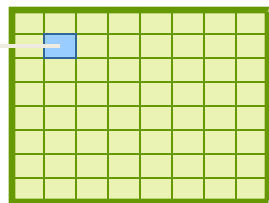
Small 8x8 thread blocks afford large per-thread register count, shared memory



Threads producing results that are used

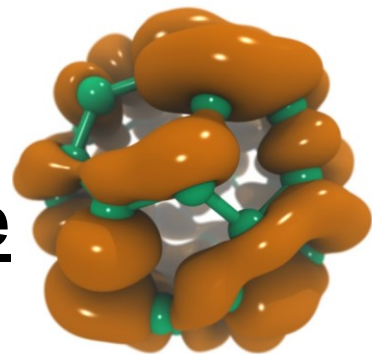
Threads producing results that are discarded

Each thread computes one MO lattice point.



Padding optimizes global memory performance, guaranteeing coalesced global memory accesses

VMD C₆₀ MO Viz. Perf, 516x519x507 Grid: @ .13s/frame, avoids 3.8GB/s I/O per-node



Hardware platform	Runtime,	Speedup
IBM Power8 (ORNL 'crest') + 1x Tesla K40 [1]	3.49s,	1.0x
Intel Xeon E5-2697Av4 + 1x Tesla V100	0.610s,	5.7x
Intel Xeon E5-2697Av4 + 2x Tesla V100	0.294s,	11.8x
Intel Xeon E5-2697Av4 + 3x Tesla V100	0.220s,	15.9x
IBM Power9 "Newell" + 1x Tesla V100	0.394s,	8.8x
IBM Power9 "Newell" + 2x Tesla V100	0.207s,	16.8x
IBM Power9 "Newell" + 3x Tesla V100	0.151s,	23.1x
IBM Power9 "Newell" + 4x Tesla V100	0.130s,	26.8x

NVLink perf.
boost w/ no
code tuning
(YET)

[1] Early Experiences Porting the NAMD and VMD Molecular Simulation and Analysis Software to GPU-Accelerated OpenPOWER Platforms. J. E. Stone, A.-P. Hynninen, J. C. Phillips, K. Schulten. International Workshop on OpenPOWER for HPC (IWOPH'16), LNCS 9945, pp. 188-206, 2016.

MO Kernel Structure, Opportunity for NVRTC JIT...

Data-driven execution, but representative loop trip counts in (...)

Loop over atoms (1 to ~200) {

Loop over electron shells for this atom type (1 to ~6) {

Loop over primitive functions for this shell type (1 to ~6) {

Small loop trip counts result in significant loop overhead. **Runtime kernel generation and NVRTC JIT compilation can achieve in a large (1.8x!) speed boost via loop unrolling, constant folding, elimination of array accesses, ...**

Loop over angular momenta for this shell type (1 to ~15) {}

}

}

Molecular Orbital Computation and Display Process

Runtime Kernel Generation, NVRTC Just-In-Time (JIT) Compilation

**One-time
initialization**

**Initialize Pool of GPU
Worker Threads**

Read QM simulation log file, trajectory

Preprocess MO coefficient data
eliminate duplicates, sort by type, etc...

Generate/compile basis set-specific CUDA kernel

For current frame and MO index,
retrieve MO wavefunction coefficients

**Compute 3-D grid of MO wavefunction amplitudes
using basis set-specific CUDA kernel**

Extract isosurface mesh from 3-D MO grid

Render the resulting surface

**For each trj frame,
for each MO shown**

```
for (shell=0; shell < maxshell; shell++) {  
    float contracted_gto = 0.0f;  
  
    // Loop over the Gaussian primitives of CGTO  
    int maxprim = const_num_prim_per_shell[shell_counter];  
    int shell_type = const_shell_symmetry[shell_counter];  
    for (prim=0; prim < maxprim; prim++) {  
        float exponent = const_basis_array[prim_counter];  
        float contract_coeff = const_basis_array[prim_counter + 1];  
        contracted_gto += contract_coeff * expf(-exponent*dist2);  
        prim_counter += 2;  
    }  
  
    contracted_gto = 1.832937 * expf(-7.868272*dist2);  
    contracted_gto += 1.405380 * expf(-1.881289*dist2);  
    contracted_gto += 0.701383 * expf(-0.544249*dist2);  
}
```

General loop-based
data-dependent MO
CUDA kernel



Runtime-generated data-
specific MO CUDA kernel
compiled via **CUDA**
NVRTC JIT...



1.8x Faster

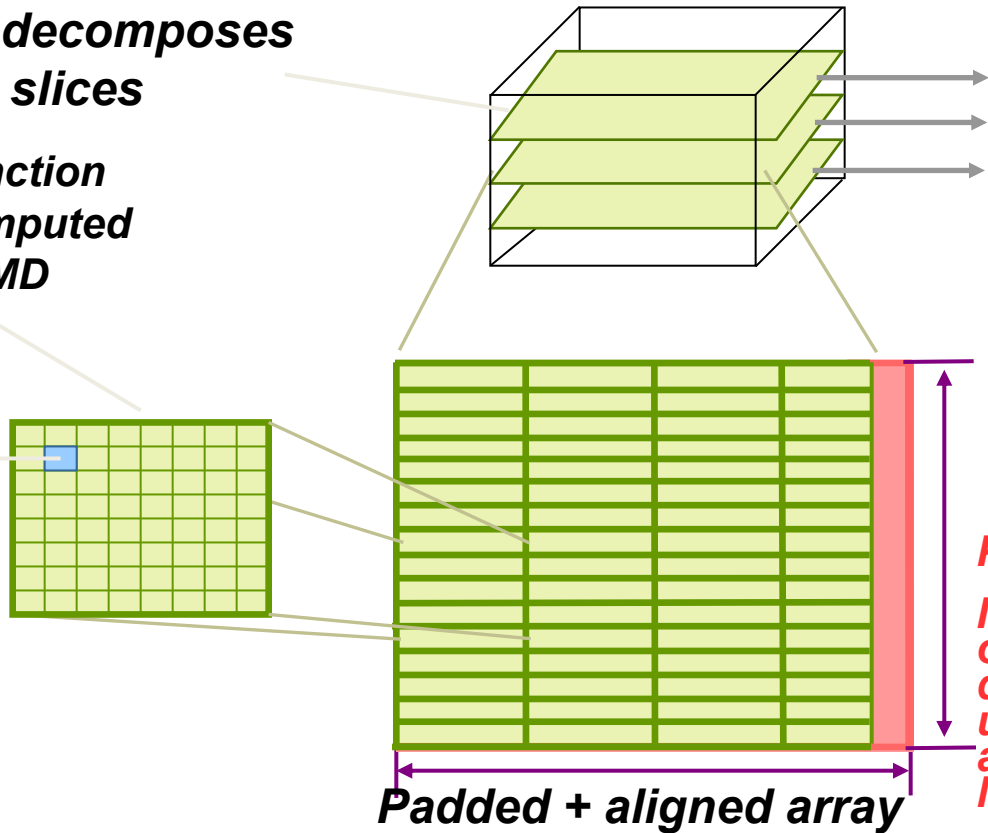
MO CPU Parallel Decomposition

AVX-512 Kernels on KNL

MO 3-D lattice decomposes into 2-D slices

Vectors of wavefunction amplitudes are computed using hardware SIMD instructions

Each CPU thread computes 1, 4, 8, 16 MO lattice points per loop iteration: C, SSE, AVX2 or AVX-512ER



...
Thread 2
Thread 1
Thread 0

Lattice decomposed across many CPU threads

SIMD lanes producing results that are used

Padding:
Inactive SIMD lanes or region of discarded output used to guarantee aligned vector loads+stores

Padded + aligned array

AVX-512ER MO CGTO Loop

```
int maxprim = num_prim_per_shell[shell_counter];
int shelltype = shell_types[shell_counter];
for (prim=0; prim<maxprim; prim++) {
    float exponent      = basis_array[prim_counter    ];
    float contract_coeff = basis_array[prim_counter + 1];

    // contracted_gto += contract_coeff * exp(exponent*dist2);
    __m512 expval = _mm512_mul_ps(_mm512_set1_ps(exponent * MLOG2EF), dist2);

    // expf() approximation required, use (base-2) AVX-512ER instructions...
    __m512 retval = _mm512_exp2a23_ps(expval);
}
```

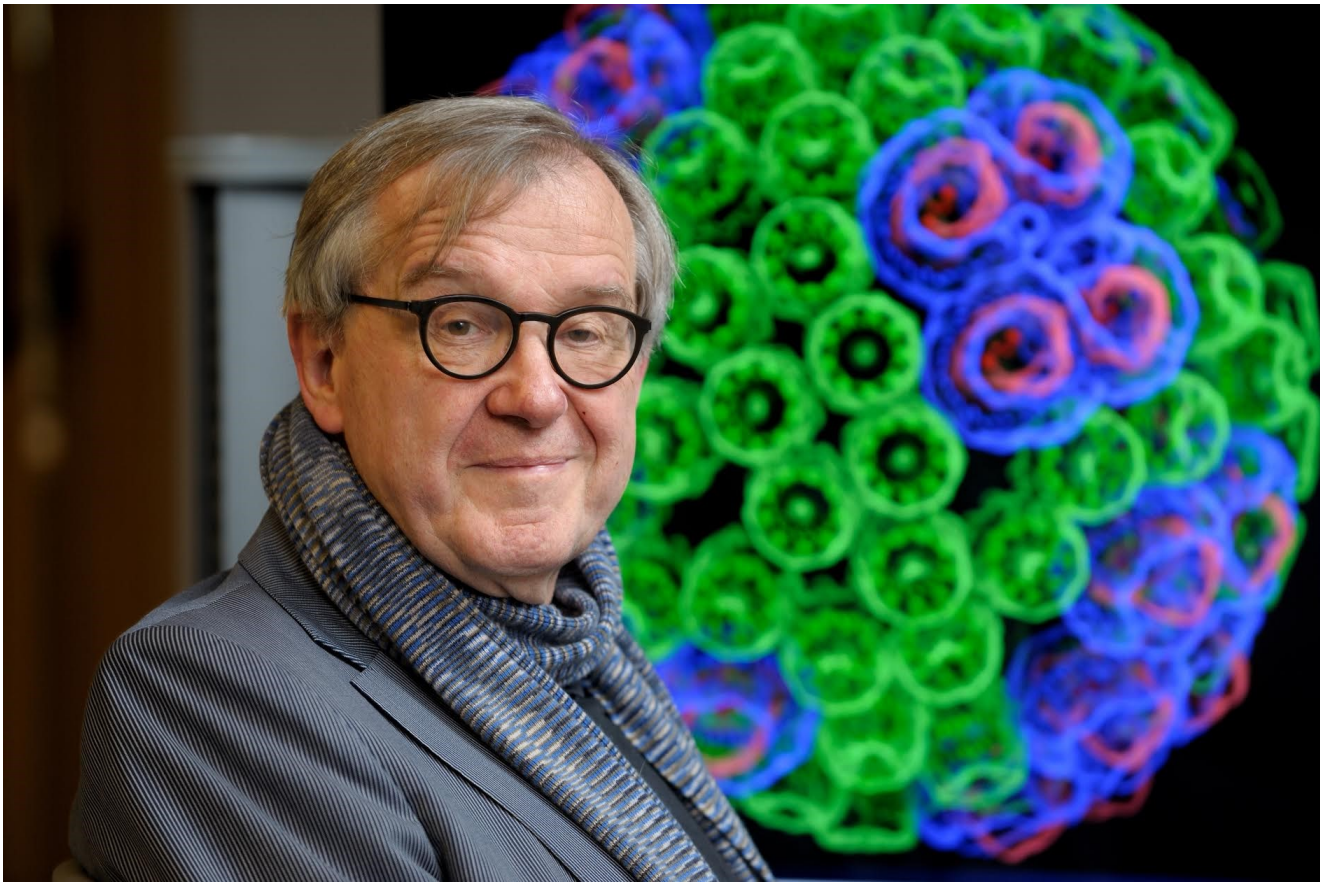

Performance of AVX-512ER Intrinsics vs. Autovectorization on KNL: Small 172x173x169 Grid

- Intel C++ '15 autovectorization (**fail**): **220+ sec**
- Hand-coded SSE2 w/ existing thread scheme: **48.5 sec**
- Hand-coded AVX-512ER w/ existing thread scheme: **6.3 sec**
- Hand-coded AVX-512ER, refactoring thread pool: **0.2 sec**
- **Hand-coded AVX-512ER tuned thread pool: 0.131 sec**
- **Hand-coded AVX-512ER+FMA tweaks: 0.107 sec**

Further improvement will require attention to details of cache behaviour and further tuning of low-level threading constructs for Xeon Phi/KNL

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“When I was a young man, my goal was to look with mathematical and computational means at the inside of cells, one atom at a time, to decipher how living systems work. That is what I strived for and I never deflected from this goal.” – Klaus Schulten

Related Publications

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- **NAMD goes quantum: An integrative suite for hybrid simulations.** Melo, M. C. R.; Bernardi, R. C.; Rudack T.; Scheurer, M.; Riplinger, C.; Phillips, J. C.; Maia, J. D. C.; Rocha, G. D.; Ribeiro, J. V.; Stone, J. E.; Neese, F.; Schulten, K.; Luthey-Schulten, Z.; Nature Methods, 2018. (In press)
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- **Probing Biomolecular Machines with Graphics Processors.** J. Phillips, J. Stone. *Communications of the ACM*, 52(10):34-41, 2009.
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