Using Accelerator Directives to Adapt Science Applications for State-of-the-Art HPC Architectures

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Overview

- My perspective about directive-based accelerator programming today and in the near-term ramp up to exascale computing
- Based on our ongoing work developing VMD and NAMD molecular modeling tools supported by our NIH-funded center since the mid-90's
- What is a person like me doing using directives? I'm the same guy that likes to give talks about CUDA and OpenCL, x86 intrinsics, and similarly lower level programming techniques. Why am I here?





Spoilers:

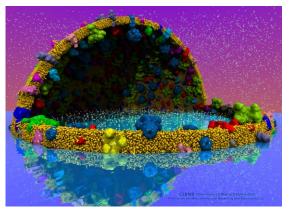
- Directives are a key solution in the "all options on the table" type of approach that I believe is required as we work toward exascale computing
- There aren't enough HPC developers in the world to write everything entirely in low level APIs fast enough to keep pace
- Science is an ever changing landscape significant methodological developments come every few years in active fields like biomolecular modeling...
- Code gets (re)written for new science methodologies before you've finished optimizing the old code for the previous science method!?!?!?!
- Hardware is still changing very rapidly, and more disruptively than during the blissful heyday of "Peak Moore's Law"



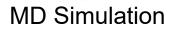


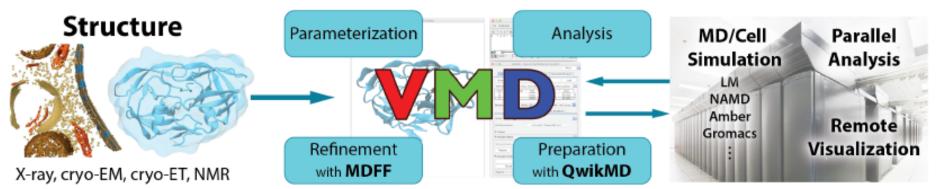
VMD – "Visual Molecular Dynamics"

- Visualization and analysis of:
 - Molecular dynamics simulations
 - Lattice cell simulations
 - Quantum chemistry calculations
 - Sequence information
- User extensible scripting and plugins
- http://www.ks.uiuc.edu/Research/vmd/



Cell-Scale Modeling





Major Approaches For Programming Hybrid Architectures

- Use drop-in libraries in place of CPU-only libraries
 - Little or no code development
 - Examples: MAGMA, BLAS-variants, FFT libraries, etc.
 - Speedups limited by Amdahl's Law and overheads associated with data movement between CPUs and GPU accelerators
- Generate accelerator code as a variant of CPU source, e.g. using OpenMP and OpenACC directives, and similar
- Write lower-level accelerator-specific code, e.g. using CUDA, OpenCL, other approaches





Challenges Adapting Large Software Systems for State-of-the-Art Hardware Platforms

- Initial focus on key computational kernels eventually gives way to the need to optimize an ocean of less critical routines, due to observance of Amdahl's Law
- Even though these less critical routines might be easily ported to CUDA or similar, the sheer number of routines often poses a challenge
- Need a low-cost approach for getting "some" speedup out of these second-tier routines
- In many cases, it is completely sufficient to achieve memorybandwidth-bound GPU performance with an existing algorithm





Amdahl's Law and Role of Directives

- Initial partitioning of algorithm(s) between host CPUs and accelerators is typically based on initial performance balance point
- Time passes and accelerators get MUCH faster...
- Formerly harmless CPU code ends up limiting overall performance!
- Need to address bottlenecks in increasing fraction of code
- **Directives** provide **low cost, low burden**, approach to **improve incrementally** vs. status quo
- **Directives are complementary to lower level approaches** such as CPU intrinsics, CUDA, OpenCL, and they all need to coexist and interoperate very gracefully alongside each other





How Do Directives Fit In?

- Single code base is typically maintained
- Almost "deceptively" simple to use
- Easy route for incremental, "gradual buy in"
- Rapid development cycle, but success often follows minor refactoring and/or changes to data structure layout
- **Profiling during development** provides critical feedback on incremental directive efficacy
- Higher abstraction level than other techniques for programming accelerators
- In many cases, **performance can be "good enough" due to memory-bandwidth limits**, or based on return on developer time or some other metric





Why Not Use Directives Exclusively?

- Some projects do...but:
 - Back-end runtimes for compiler directives sometimes have unexpected extra overheads that could be a showstopper in critical algorithm steps
 - High abstraction level may mean lack of access to hardware features exposed only via CUDA or other lower level APIs
 - Fortunately, interoperability APIs enable directive-based approaches to be used side-by-side with hand-coded kernels, libraries, etc.
 - Presently, sometimes-important capabilities like JIT compilation of runtime-generated kernels only exist within lower level APIs such as CUDA and OpenCL



What Do Existing Accelerated Applications Look Like?

I'll provide examples from digging into modern versions of VMD that have already incorporated acceleration in a deep way.

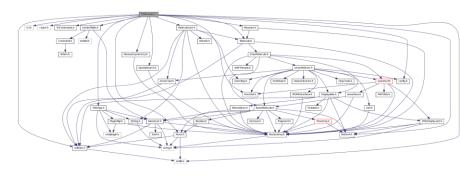
Questions:

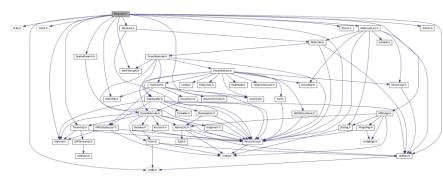
- How much code needs to be "fast", or "faster"
- What fraction runs on accelerator now?
- Using directives, how much more coverage can be achieved, and with what speedup?
- Do I lose access to any points of execution or resource control that are critical for the application's performance?



Example of VMD Module Connectivity

- Early progress focused acceleration efforts on handful of high level analysis routines that were the most computationally demanding
- Future hardware requires pervasive acceleration
- Top image shows script interface links to top level analytical routines
- Bottom image shows links among subset of data analytics algorithms to leaf-node functions



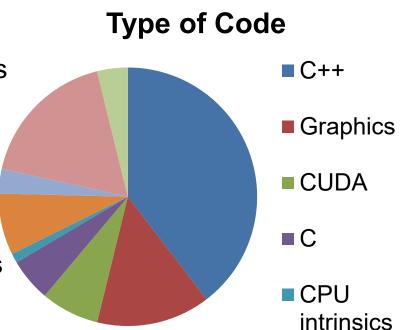




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VMD Software Decomposition

- Computational code is 50% of VMD core
- Hand-written accelerator + vectorized code (CUDA + CPU intrinsics) represents only 14% of core computational code
 - 20,000 lines of CUDA
 - 3,100 lines of intrinics
- Percent coverage of leaf-node analytical functions is lower yet
- Need to evolve VMD toward high coverage of performance-critical analysis code with fine-grained parallelism on accelerators and vectorization

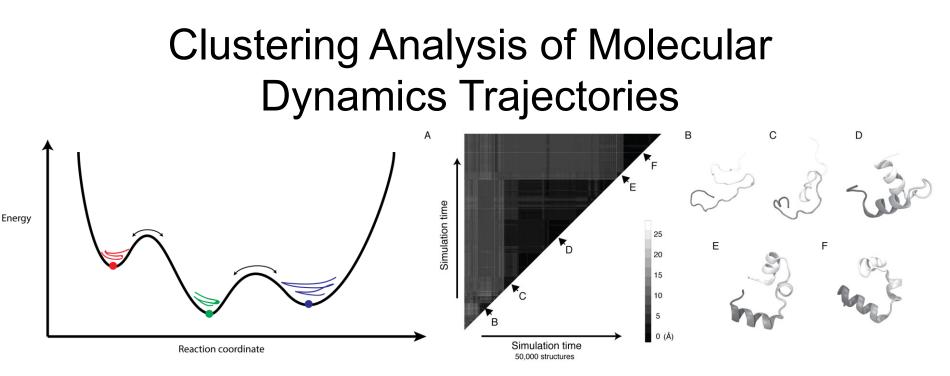




Directive-Based Parallel Programming with OpenACC

- Annotate loop nests in existing code with #pragma compiler directives:
 - Annotate opportunities for parallelism
 - Annotate points where host-GPU memory transfers are best performed, indicate propagation of data
- Evolve original code structure to improve efficacy of parallelization
 - Eliminate false dependencies between loop iterations
 - Revise algorithms or constructs that create excess data movement
- How well does this work if we stick with "low cost, low burden" philosophy I claim to support?





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.



Serial QCP RMSD Inner Product Loop

- Simple example where directive based parallelism can be applied easily and effectively
- Such a loop is inherently a memory-bandwidth-bound algorithm, so that's the goal for acceleration

for (int I=0; I<cnt; I++) { double x1, x2, y1, y2, z1, z2; x1 = crdx1[l];y1 = crdy1[l];z1 = crdz1[1]: G1 += x1*x1 + y1*y1 + z1*z1;x2 = crdx2[l]; $y^{2} = crdy^{2}[1];$ z2 = crdz2[1]; $G2 += x2^{*}x2 + y2^{*}y2 + z2^{*}z2;$ a0 += x1 * x2: a1 += x1 * y2; a2 += x1 * z2; a3 += v1 * x2: a4 += y1 * y2; a5 += y1 * z2; a6 += z1 * x2; a7 += z1 * v2: a8 += z1 * z2:



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OpenACC QCP RMSD Inner Product Loop

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- Such a loop is inherently a memory-bandwidth-bound algorithm, so that's the goal for acceleration

long i, j, k;

#pragma acc kernels copyin(crds[0:tsz]), copy(rmsdmat[0:msz])
for (k=0; k<(framecount*(framecount-1))/2; k++) {
 // compute triangular matrix index 'k' in a helper function
 // to ensure that the compiler doesn't think that we have
 // conflicts or dependencies between loop iterations
 acc_idx2sub_tril(long(framecount-1), k, &i, &j);
 long x1addr = j * 3L * framecrdsz;
 long x2addr = i * 3L * framecrdsz;</pre>

#pragma acc loop vector(256) for (long l=0; l<cnt; l++) { // abridged for brevity ...</pre>

```
rmsdmat[k]=rmsd; // store linearized triangular matrix }
```



OpenACC QCP RMSD Inner Product Loop Performance Results

- Xeon 2867W v3, w/ hand-coded AVX and FMA intrinsics: 20.7s
- Tesla K80 w/ OpenACC: 6.5s (3.2x speedup)
- OpenACC on K80 achieved 65% of theoretical peak memory bandwidth, with 2016 compiler and just a few lines of #pragma directives. Excellent speedup for minimal changes to code.
- Future OpenACC compiler revs should provide higher performance yet



Caveat Emptor

- Compilers are not all equal...
- ...sometimes they make me want to scream...
- ...but they all improve with time
- If we begin using directives now to close the gap on impending doom arising from Amdahl's Law, the compilers should be robust and performant when it really counts





Directives and Hardware Evolution

- Ongoing hardware advancements are addressing ease-of-use gaps that remained a problem for both directives and hand-coded kernels
- Unified memory: eliminate many cases where programmer used to have to hand-code memory transfers explicitly, blurs CPU/GPU boundary
- What about distributing data structures across multiple NVLink-connected GPUs?





Performance Tuning, Profiling Wish List

- Some simple examples on my wish list:
 - Make directive runtimes more composable with external resource management, tasking frameworks, and runtime systems, interop APIs are already a start, to build more commonality there.
 - Help profiling tools to clearly identify functions, call chains, and resources associated with code produced by compiler directives and their runtime system(s), to clearly differentiate from hand-coded kernels, and resources used by other runtimes
 - Allow directive-based programming systems support things like application-determined hardware scheduling priorities that encompass both hand-coded and directive-generated kernels
 - Allow programmer oversight about what resources directive kernels are allowed to use, CPU affinity, etc



Directives and Potential Hardware Evolution

Think of ORNL Summit node as an "entry point" to potential future possibilities... Questions:

- Would the need for ongoing growth in memory bandwidth among tightly connected accelerators w/ HBM predict even denser nodes?
 - Leadership systems use 6-GPU nodes now, how many in 2022 or thereafter?
- As accelerated systems advance, will directives encompass peer-to-peer accelerator operations better?
- What if future accelerators can directly RDMA to remote accelerators (over a communication fabric) via memory accesses?
- In the future, will directives make it easier to program potentially complex collective operations, reductions, fine-grained distributed-shared-memory data structures among multiple accelerators?



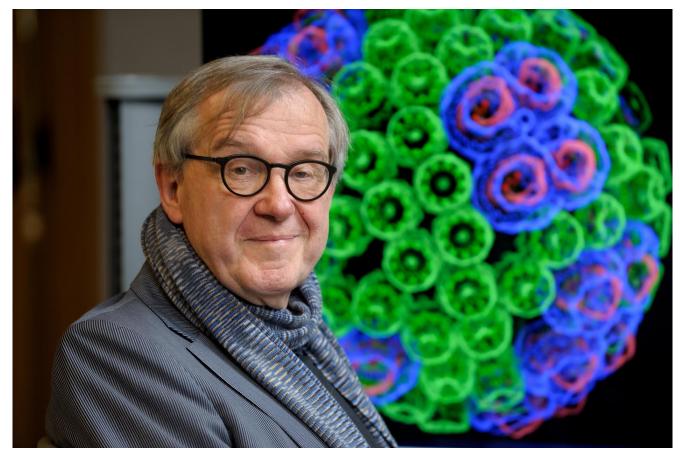




Please See These Other Talks:

- S8727 Improving NAMD Performance on Volta GPUs
- S8718 Optimizing HPC Simulation and Visualization Codes
 using Nsight Systems
- S8709 Accelerating Molecular Modeling Tasks on Desktop and Pre-Exascale Supercomputers
- S8747 ORNL Summit: Petascale Molecular Dynamics Simulations on the Summit POWER9/Volta Supercomputer
- S8665: VMD: Biomolecular Visualization from Atoms to Cells
 Using Ray Tracing, Rasterization, and VR



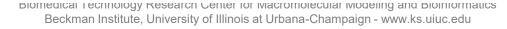


"When I was a young man, my goal was to look with mathematical and computational means at the inside of cells, one atom at a time, to decipher how living systems work. That is what I strived for and I never deflected from this goal." – Klaus Schulten

Related Publications http://www.ks.uiuc.edu/Research/gpu/

- 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)
- Challenges of Integrating Stochastic Dynamics and Cryo-electron Tomograms in Whole-cell Simulations.
 T. M. Earnest, R. Watanabe, J. E. Stone, J. Mahamid, W. Baumeister, E. Villa, and Z. Luthey-Schulten.
 J. Physical Chemistry B, 121(15): 3871-3881, 2017.
- 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, and K. Schulten. International Workshop on OpenPOWER for HPC (IWOPH'16), LNCS 9945, pp. 188-206, 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 Workshop (IPDPSW), pp. 1048-1057, 2016.
- **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 International Parallel and Distributed Processing Symposium Workshop (IPDPSW), pp. 1014-1023, 2016.
- Evaluation of Emerging Energy-Efficient Heterogeneous Computing Platforms for Biomolecular and Cellular Simulation Workloads. J. E. Stone, M. J. Hallock, J. C. Phillips, J. R. Peterson, Z. Luthey-Schulten, and K. Schulten.25th International Heterogeneity in Computing Workshop, IEEE International Parallel and Distributed Processing Symposium Workshop (IPDPSW), pp. 89-100, 2016.







Related Publications

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- Atomic Detail Visualization of Photosynthetic Membranes with GPU-Accelerated Ray Tracing. J. E. Stone, M. Sener, K. L. Vandivort, A. Barragan, A. Singharoy, I. Teo, J. V. Ribeiro, B. Isralewitz, B. Liu, B.-C. Goh, J. C. Phillips, C. MacGregor-Chatwin, M. P. Johnson, L. F. Kourkoutis, C. Neil Hunter, and K. Schulten. J. Parallel Computing, 55:17-27, 2016.
- Chemical Visualization of Human Pathogens: the Retroviral Capsids. Juan R. Perilla, Boon Chong Goh, John E. Stone, and Klaus Schulten. SC'15 Visualization and Data Analytics Showcase, 2015.
- Visualization of Energy Conversion Processes in a Light Harvesting Organelle at Atomic Detail. M. Sener, J. E. Stone, A. Barragan, A. Singharoy, I. Teo, K. L. Vandivort, B. Isralewitz, B. Liu, B. Goh, J. C. Phillips, L. F. Kourkoutis, C. N. Hunter, and K. Schulten. Visualization and Data Analytics Showcase, 2014.
 ***Winner of the SC'14 Visualization and Data Analytics Showcase
- Runtime and Architecture Support for Efficient Data Exchange in Multi-Accelerator Applications. J. Cabezas, I. Gelado, J. E. Stone, N. Navarro, D. B. Kirk, and W. Hwu. IEEE Transactions on Parallel and Distributed Systems, 26(5):1405-1418, 2015.
- Unlocking the Full Potential of the Cray XK7 Accelerator. M. D. Klein and J. E. Stone. Cray Users Group, Lugano Switzerland, May 2014.
- 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.
- Simulation of reaction diffusion processes over biologically relevant size and time scales using multi-GPU workstations. M. J. Hallock, J. E. Stone, E. Roberts, C. Fry, and Z. Luthey-Schulten. Journal of Parallel Computing, 40:86-99, 2014.



Related Publications

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- **GPU-Accelerated Molecular Visualization on Petascale Supercomputing Platforms.** J. Stone, K. L. Vandivort, and K. Schulten. UltraVis'13: Proceedings of the 8th International Workshop on Ultrascale Visualization, pp. 6:1-6:8, 2013.
- Early Experiences Scaling VMD Molecular Visualization and Analysis Jobs on Blue Waters. J. Stone, B. Isralewitz, and K. Schulten. In proceedings, Extreme Scaling Workshop, 2013.
- Lattice Microbes: High-performance stochastic simulation method for the reaction-diffusion master equation. E. Roberts, J. Stone, and Z. Luthey-Schulten. J. Computational Chemistry 34 (3), 245-255, 2013.
- Fast Visualization of Gaussian Density Surfaces for Molecular Dynamics and Particle System Trajectories. M. Krone, J. Stone, T. Ertl, and K. Schulten. *EuroVis Short Papers,* pp. 67-71, 2012.
- 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.
- Fast Analysis of Molecular Dynamics Trajectories with Graphics Processing Units Radial Distribution Functions. B. Levine, J. Stone, and A. Kohlmeyer. *J. Comp. Physics*, 230(9):3556-3569, 2011.





Related Publications http://www.ks.uiuc.edu/Research/gpu/

- Quantifying the Impact of GPUs on Performance and Energy Efficiency in HPC Clusters. J. Enos, C. Steffen, J. Fullop, M. Showerman, G. Shi, K. Esler, V. Kindratenko, J. Stone, J Phillips. *International Conference on Green Computing*, pp. 317-324, 2010.
- **GPU-accelerated molecular modeling coming of age.** J. Stone, D. Hardy, I. Ufimtsev, K. Schulten. *J. Molecular Graphics and Modeling*, 29:116-125, 2010.
- **OpenCL: A Parallel Programming Standard for Heterogeneous Computing.** J. Stone, D. Gohara, G. Shi. *Computing in Science and Engineering,* 12(3):66-73, 2010.
- An Asymmetric Distributed Shared Memory Model for Heterogeneous Computing Systems. I. Gelado, J. Stone, J. Cabezas, S. Patel, N. Navarro, W. Hwu. ASPLOS '10: Proceedings of the 15th International Conference on Architectural Support for Programming Languages and Operating Systems, pp. 347-358, 2010.



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- **GPU Clusters for High Performance Computing**. V. Kindratenko, J. Enos, G. Shi, M. Showerman, G. Arnold, J. Stone, J. Phillips, W. Hwu. *Workshop on Parallel Programming on Accelerator Clusters (PPAC),* In Proceedings IEEE Cluster 2009, pp. 1-8, Aug. 2009.
- Long time-scale simulations of in vivo diffusion using GPU hardware. E. Roberts, J. Stone, L. Sepulveda, W. Hwu, Z. Luthey-Schulten. In *IPDPS'09: Proceedings of the 2009 IEEE International Symposium on Parallel & Distributed Computing*, pp. 1-8, 2009.
- 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 Pricessing Units (GPGPU-2), ACM International Conference Proceeding Series, volume 383, pp. 9-18, 2009.
- **Probing Biomolecular Machines with Graphics Processors**. J. Phillips, J. Stone. *Communications of the ACM*, 52(10):34-41, 2009.
- **Multilevel summation of electrostatic potentials using graphics processing units**. D. Hardy, J. Stone, K. Schulten. *J. Parallel Computing*, 35:164-177, 2009.





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- Adapting a message-driven parallel application to GPU-accelerated clusters.
 J. Phillips, J. Stone, K. Schulten. *Proceedings of the 2008 ACM/IEEE Conference on Supercomputing*, IEEE Press, 2008.
- **GPU acceleration of cutoff pair potentials for molecular modeling applications**. C. Rodrigues, D. Hardy, J. Stone, K. Schulten, and W. Hwu. *Proceedings of the 2008 Conference On Computing Frontiers*, pp. 273-282, 2008.
- **GPU computing**. J. Owens, M. Houston, D. Luebke, S. Green, J. Stone, J. Phillips. *Proceedings of the IEEE*, 96:879-899, 2008.
- Accelerating molecular modeling applications with graphics processors. J. Stone, J. Phillips, P. Freddolino, D. Hardy, L. Trabuco, K. Schulten. *J. Comp. Chem.*, 28:2618-2640, 2007.
- Continuous fluorescence microphotolysis and correlation spectroscopy. A. Arkhipov, J. Hüve, M. Kahms, R. Peters, K. Schulten. *Biophysical Journal*, 93:4006-4017, 2007.



