

Transforming Additive Manufacturing through Exascale Computing (ExaAM)

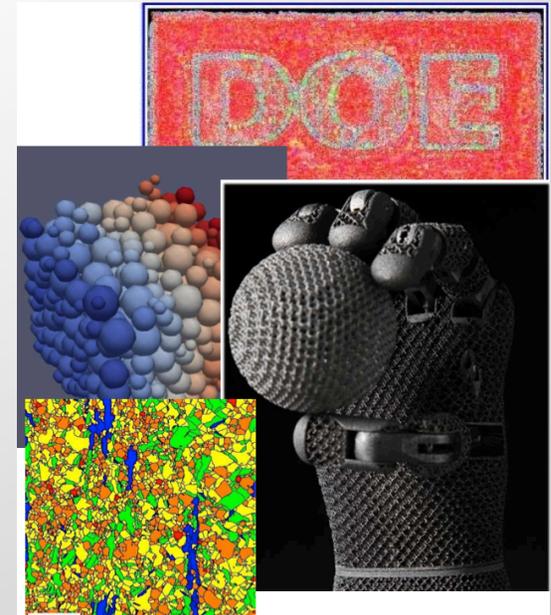
PI: John Turner (turnerja@ornl.gov)

Co-PI: **Jim Belak** (belak@llnl.gov)

www.ExaAM.org (no content yet)

Plus too many to name at ORNL(Babu), LLNL (King, Matthews), LANL(Henson, Carlson) and NIST (Levine)

SOS21, Davos Switzerland, 20-23 March, 02017

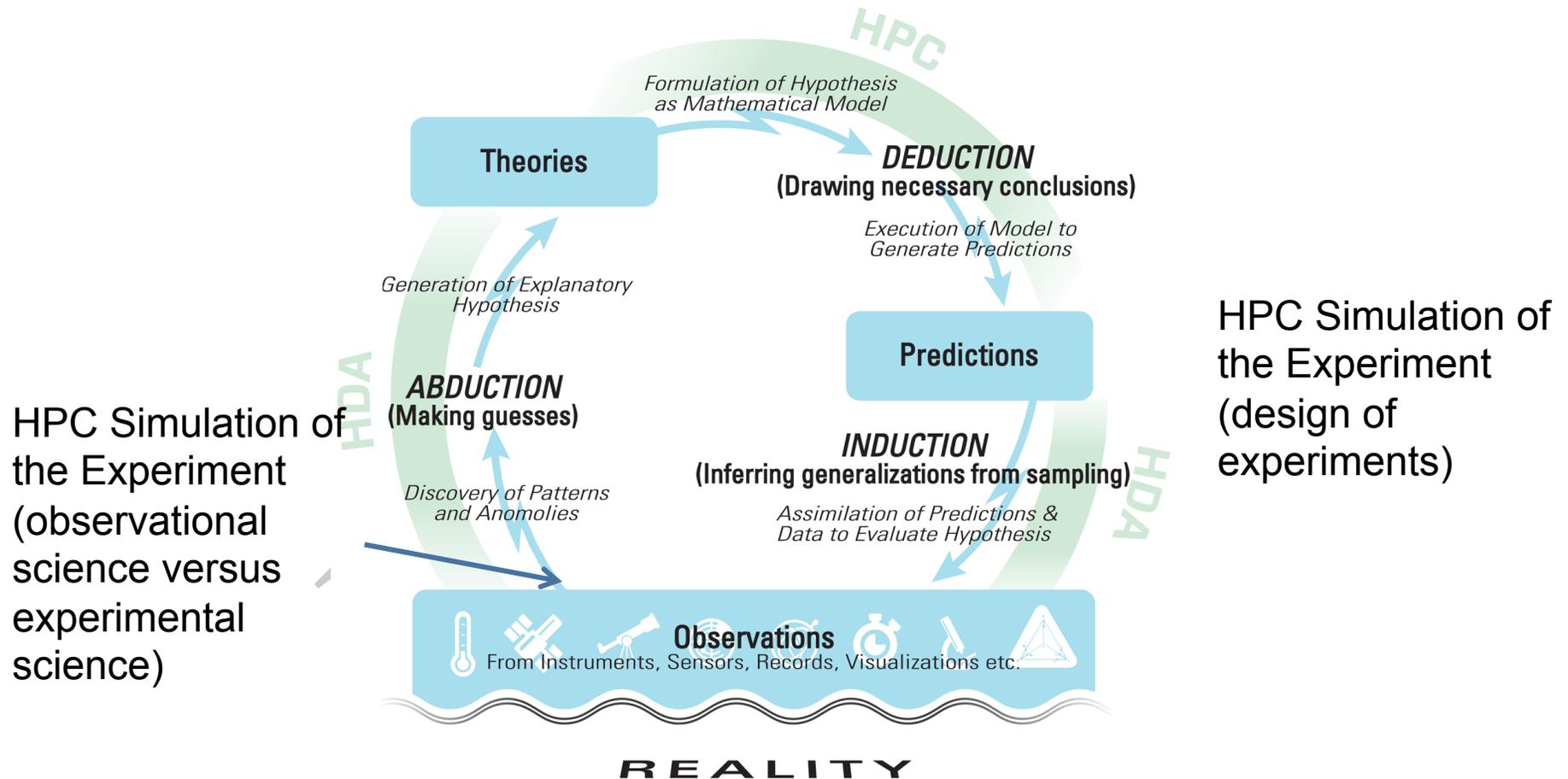


LLNL-PRES-XXXXXX

This work was performed under the auspices of the U.S. Department of Energy by Lawrence Livermore National Laboratory under contract DE-AC52-07NA27344. Lawrence Livermore National Security, LLC



(from Dan Reed's talk) Something is missing. The first thing we do after making an observation is ask if it is consistent with our framework of understanding.



HPC Simulation of the Experiment (observational science versus experimental science)

HPC Simulation of the Experiment (design of experiments)

BDEC doc

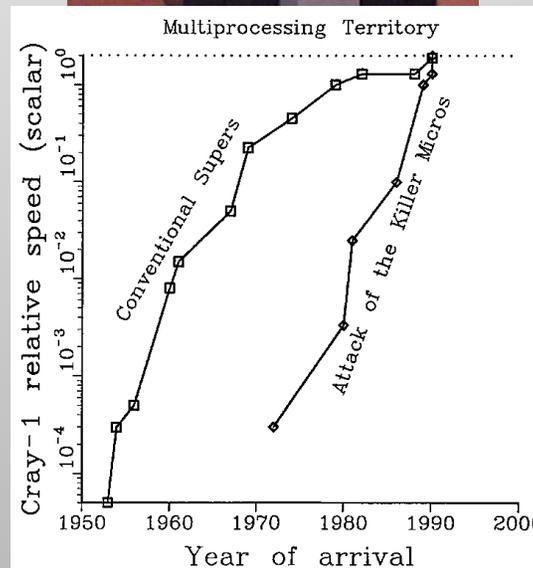
Eugene Brooks (1987) "Attack of the Killer Micros" (LLNL Massively Parallel Computing Initiative – 1990) Fastforward 2010+ "Attack of the Killer Cell Phones"



Game Processors

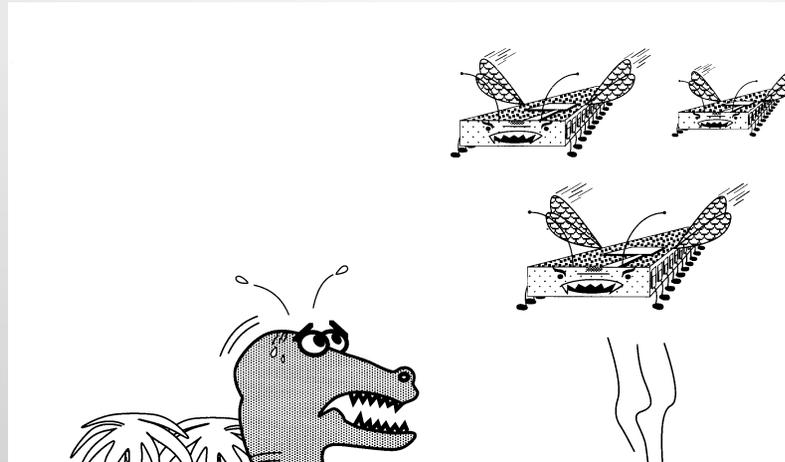


Billions and Billions of Cell Phone Processors

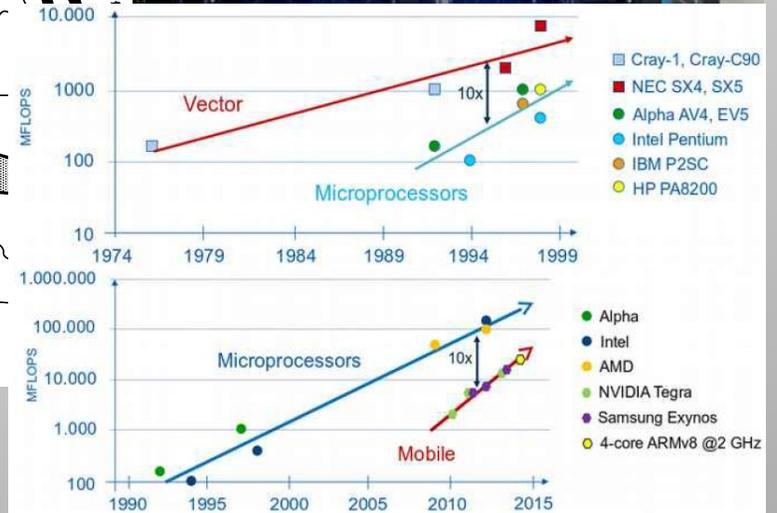


Apology to Mateo Valero (Director Barcelona Supercomputer Center)

ARM-Based Mont Blanc (BSC)



Mare Nostrum (BSC)



(Crisis) Commonly asked questions in exascale: What is the problem?

Power, energy, and heat dissipation are the central issues. Imagine a computer with **billions and billions** of cell phone processors (14MW) or **millions and millions** of throughput optimized cores, GPGPUs (20MW)

- How do you program it to work on one science problem?
- The architecture will be **heterogeneous** and **hierarchical**, with very high **flop/byte ratios**.
- Single program multiple data bulk synchronous parallelism will no longer be viable.

Data Movement will be expensive and computation will be cheap

- Need to present the physics so the computation occurs where the data is!
- Traditional global checkpoint/restart will be impractical: need local / micro checkpoint (flash memory?)

Simulation codes will need to become **fault tolerant and resilient**

- Recover from soft and hard errors, and anticipating faults
- Ability to drop or replace nodes and keep on running
- The curse of silent errors

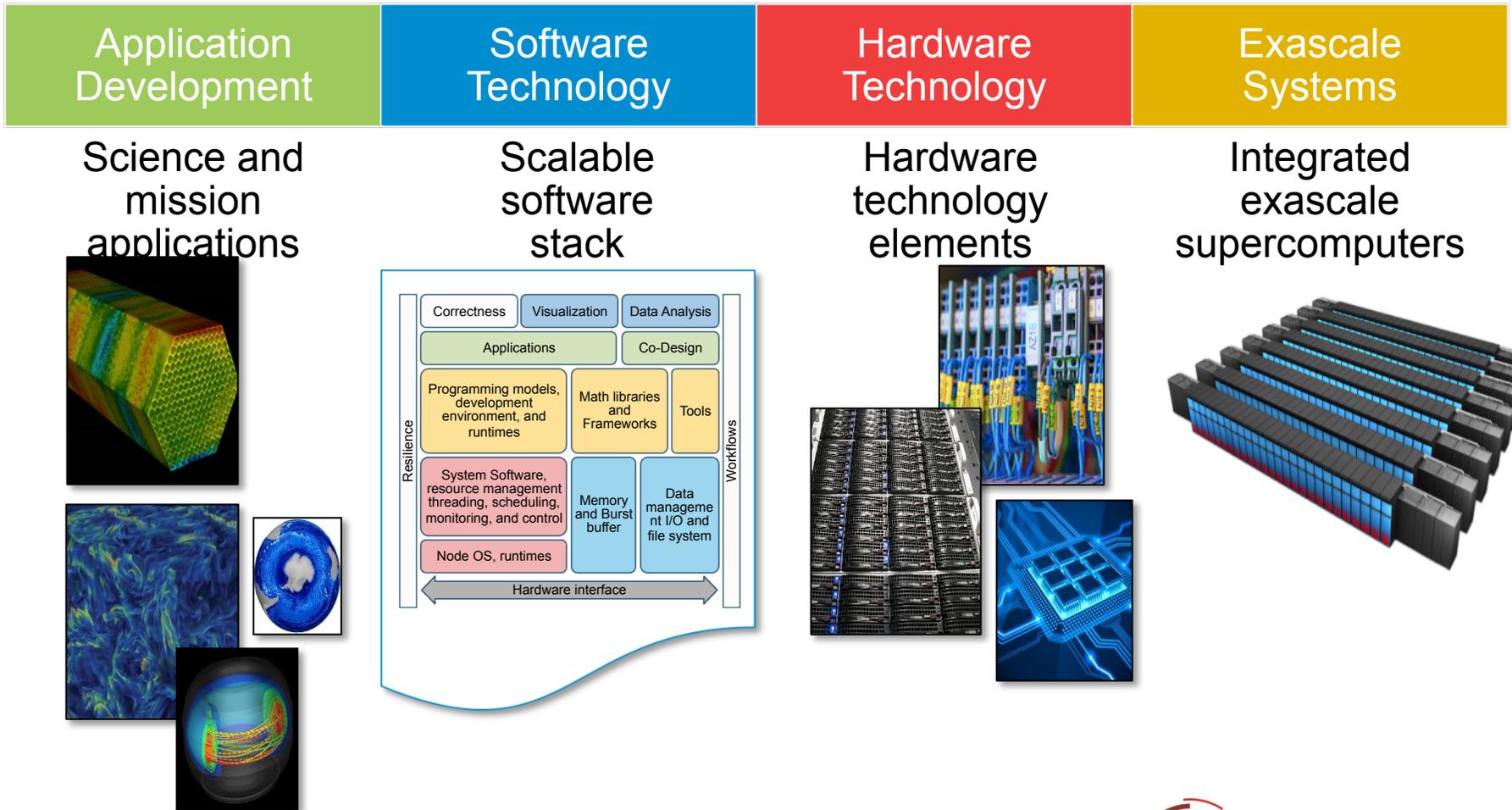
What is the Exascale Computing Project (ECP)?

- Created in support of President Obama's National Strategic Computing initiative (NSCI)
- A collaborative effort of two US Dept of Energy (DOE) offices:
 - Office of Science (DOE-SC)
 - National Nuclear Security Administration (NNSA)
- A 10-year project to accelerate the development of a **capable** exascale ecosystem
 - 50x the performance of today's 20 PF/s systems
 - Operates in a power envelope of 20–30 MW
 - Is sufficiently resilient (average fault rate: $\leq 1/\text{week}$)
 - Includes a software stack that meets the needs of a broad spectrum of applications and workloads
 - Led by DOE laboratories
 - Executed in collaboration with academia and industry



A **capable** exascale computing system will have a well-balanced ecosystem (software, hardware, applications)

ECP has formulated a holistic approach that uses co-design and integration to achieve capable exascale

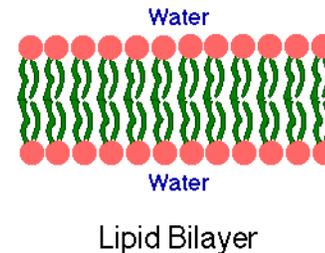


Creating Community Codes (1994-7) LAMMPS

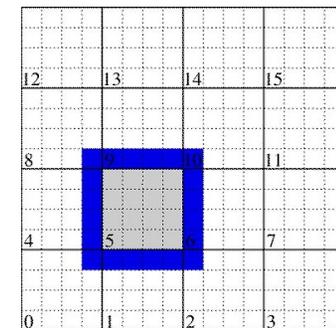
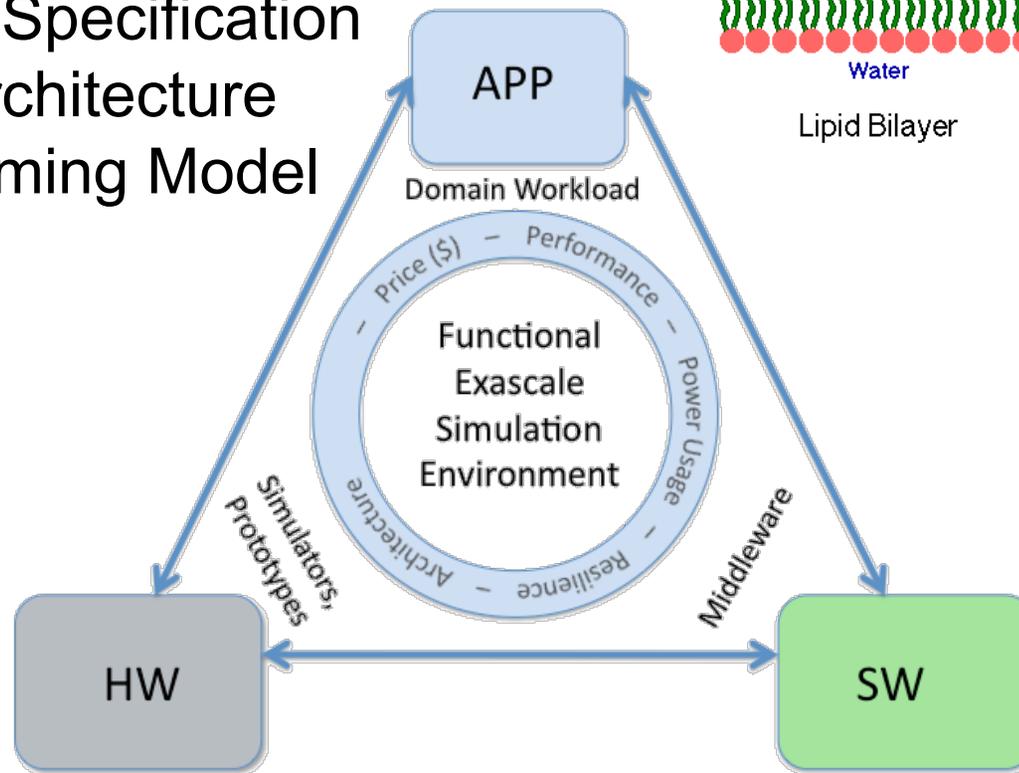
CRADA: Cray (Carpenter), Bristol-Myers (Stouch), Dupont (Lustig), LLNL (Belak), Sandia (Plimpton)

Three Essential Elements:

- Problem Specification
- Target Architecture
- Programming Model



Diffusion of a small drug molecule through a bio-membrane

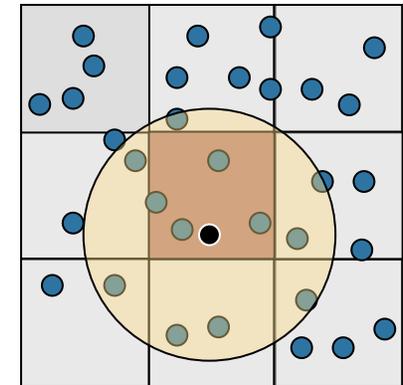


Cray T3D, Clock Speed: 0.15GHz, Date: 1994
 Microprocessor Peak Teraflops: 0.02, Memory: 0.01TB
 Processors: 128. Power Consumption: 41.40 kW

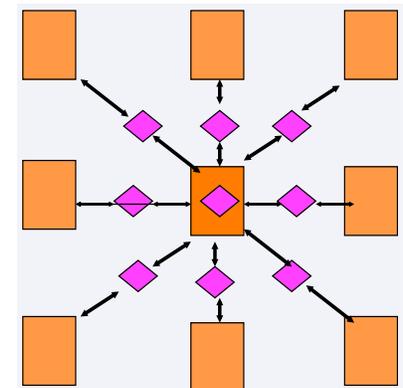
SPMD, MPI,
 Geometric domain
 decomposition

1997: LAMMPS became a successful community code because of Steve's idea to release it as open source, the ubiquitous Beowolf cluster and the stability of MPI.

- 20 Years Ago
 - Hardware: Linux Beowolf Cluster
 - Software Programming Model: MPI/SPMD
 - Application Code: LAMMPS
- ~5 Years in the Future
 - Hardware: Exascale? (Petascale in a Rack) Challenge: a scaled exascale computer with today's technology?
 - Software Programming Model: MPI+X, Q? Task-based with "Control" of Data Locality (User and Runtime)
 - Application Codes???



Worker



Work

Opportunity: Ubiquitous Peta-scale computing, in Industry, in Labs, in Academia.

Our Exascale AM Project is one of 15 initial ECP application development projects (plus 7 seeds)

Exascale Application Driver



Advanced Manufacturing

Gaps and Opportunities

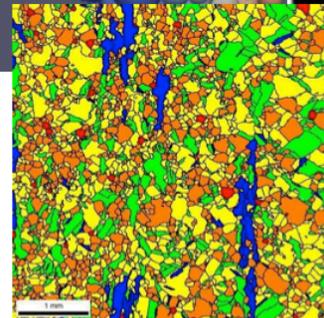
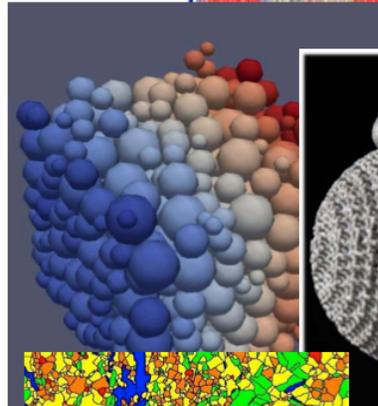
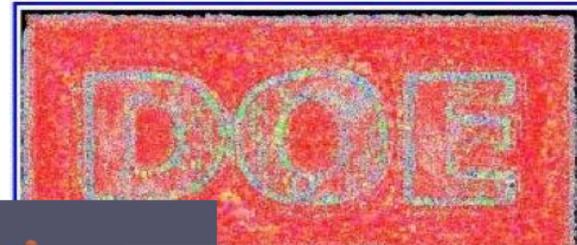
- ✓ Advance quality, reliability, and application breadth of additive manufacturing (AM)
- ✓ Accelerate innovation in clean energy manufacturing institutes (NNMIs)
- ✓ Capture emerging manufacturing markets

Simulation Challenge Problems

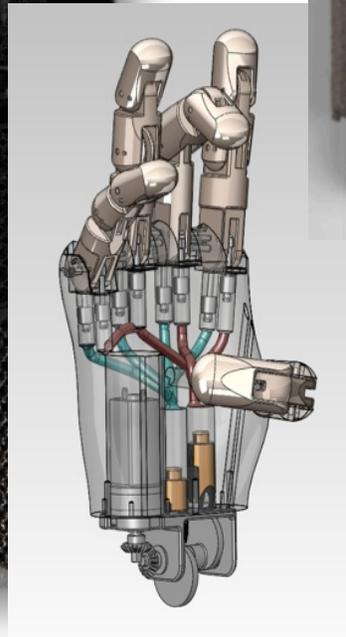
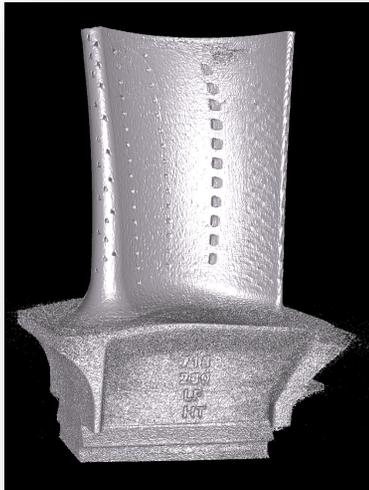
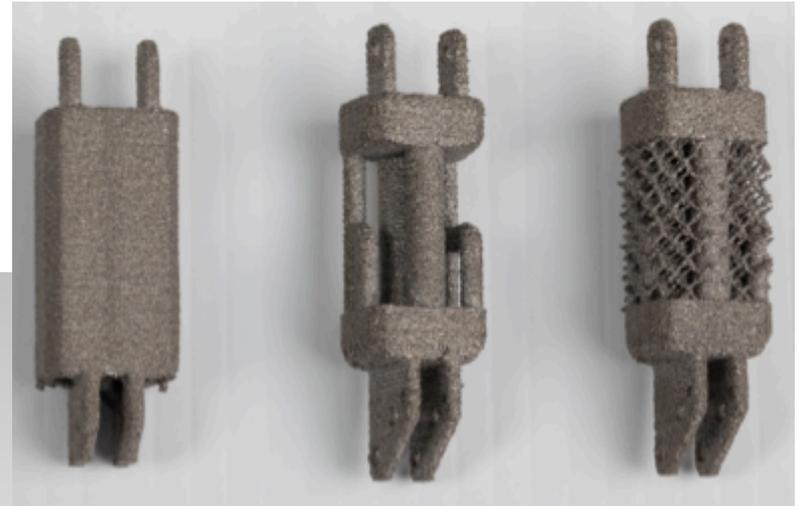
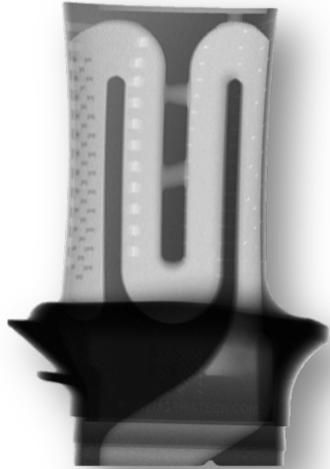
- ✓ Continuum level predictions of non-uniform microstructure and its relationship to process parameters
- ✓ Predictive mesoscale models for dendritic solidification then scale-bridged to continuum

Prospective Outcomes and Impact

- ✓ Routine qualification of AM parts via process-aware design specs and reproducibility through process control
- ✓ Fabrication of metal parts with unique properties such as light weight strength and failure-proof joints and welds



Additive Manufacturing, a.k.a. 3D printing, is being used for metal as well as polymers

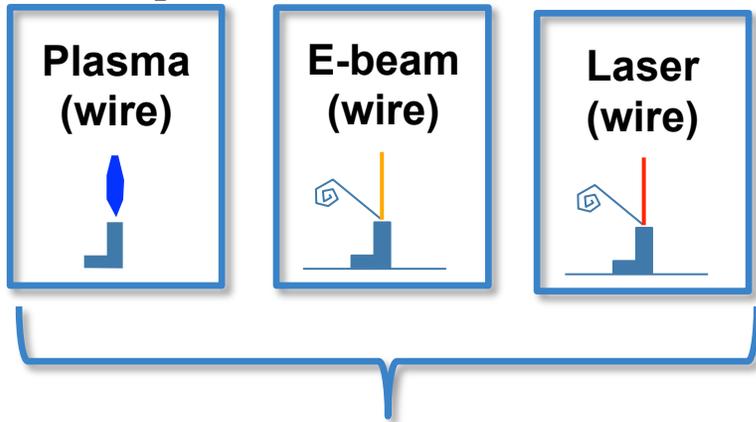


21.1 g

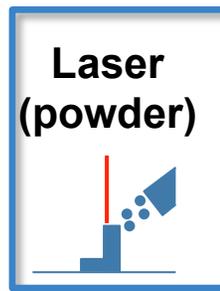
12.1 g

14.4 g

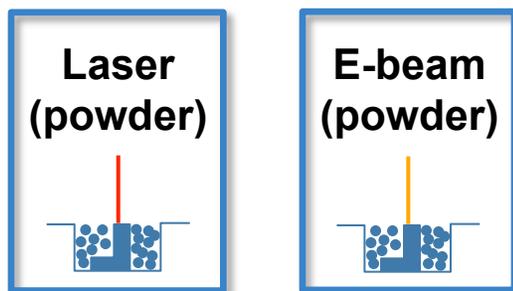
Multiple metal AM technologies



Large Melt Pool Technologies

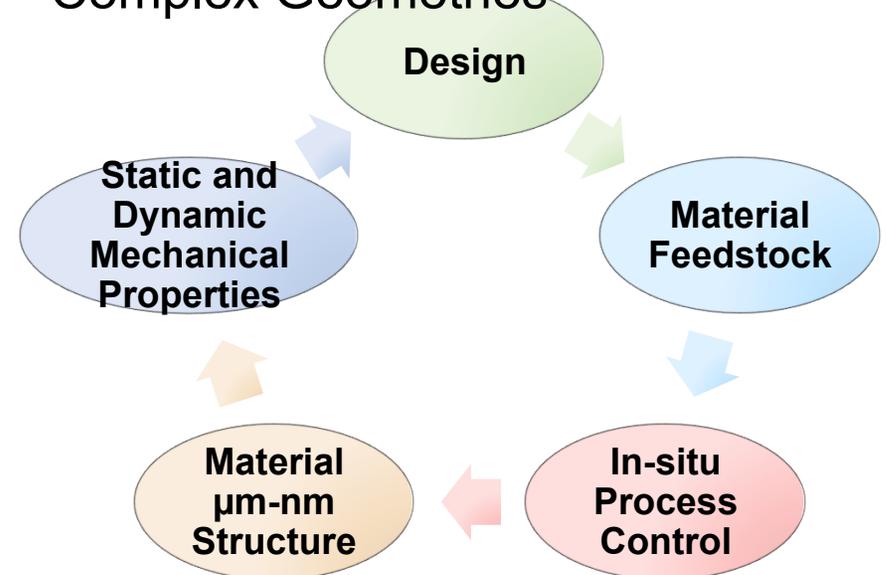


Direct Metal Deposition



Physical processes are similar

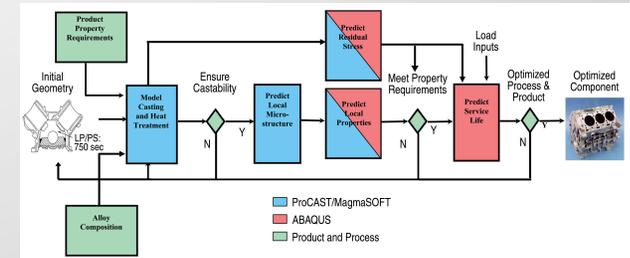
- Energy Deposition
- Melting & Powder Addition
- Evaporation & Condensation
- Heat & Mass Transfer
- Solidification
- Solid-State Phase Transformation
- Repeated Heating and Cooling
- Complex Geometries



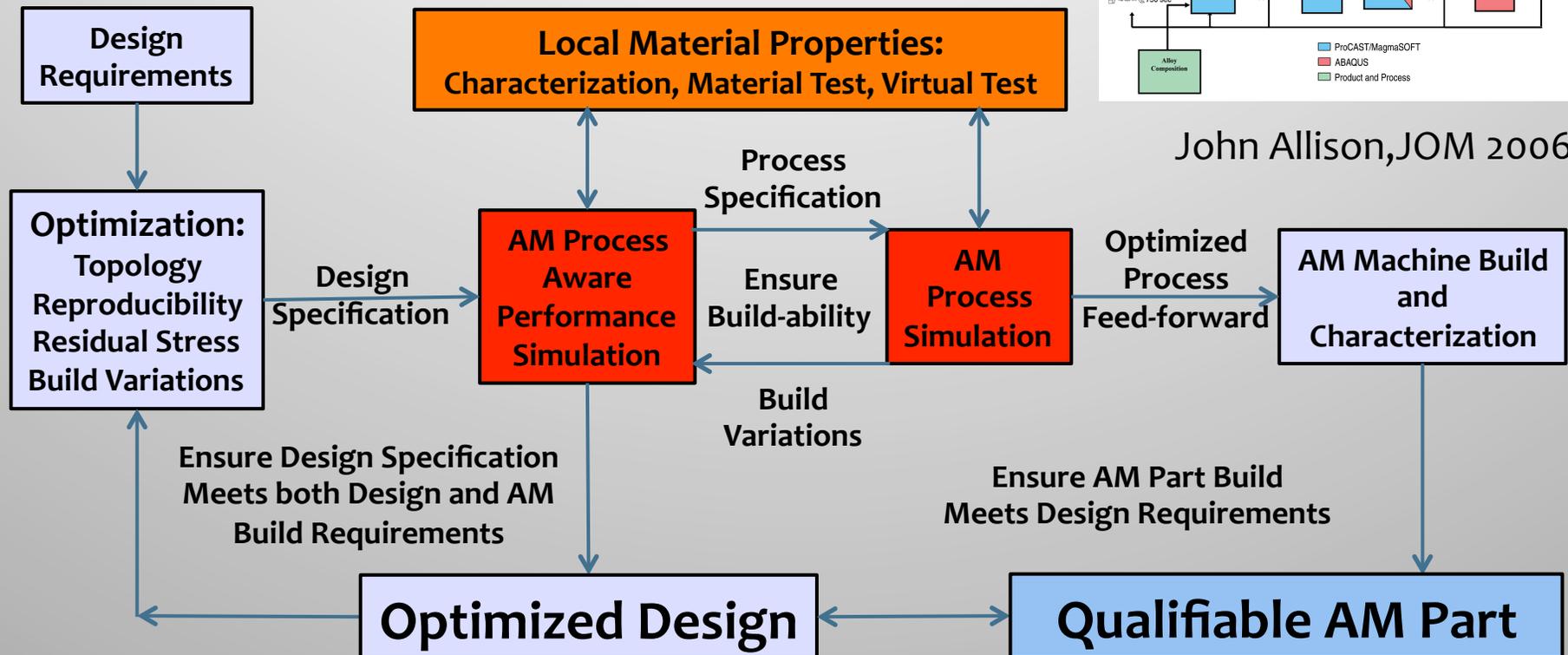
Multiple computational challenges must be addressed for AM

- $1 \text{ m}^3 \sim 10^{12}$ particles $\sim 10^9 \text{ m}$ of “weld” line (assuming $50\mu\text{m}$ particles) and build times of hours
- Large temperature gradients, rapid heating and cooling
 - necessary / sufficient coupling between thermomechanics and melt/solidification
- Heterogeneous and multi-scale
 - resolution of energy sources and effective properties of powder for continuum simulations
- Path optimization
- Large number of parameters and missing understanding
 - key uncertainties and propagation of those uncertainties
- Validation is difficult as characterization is limited

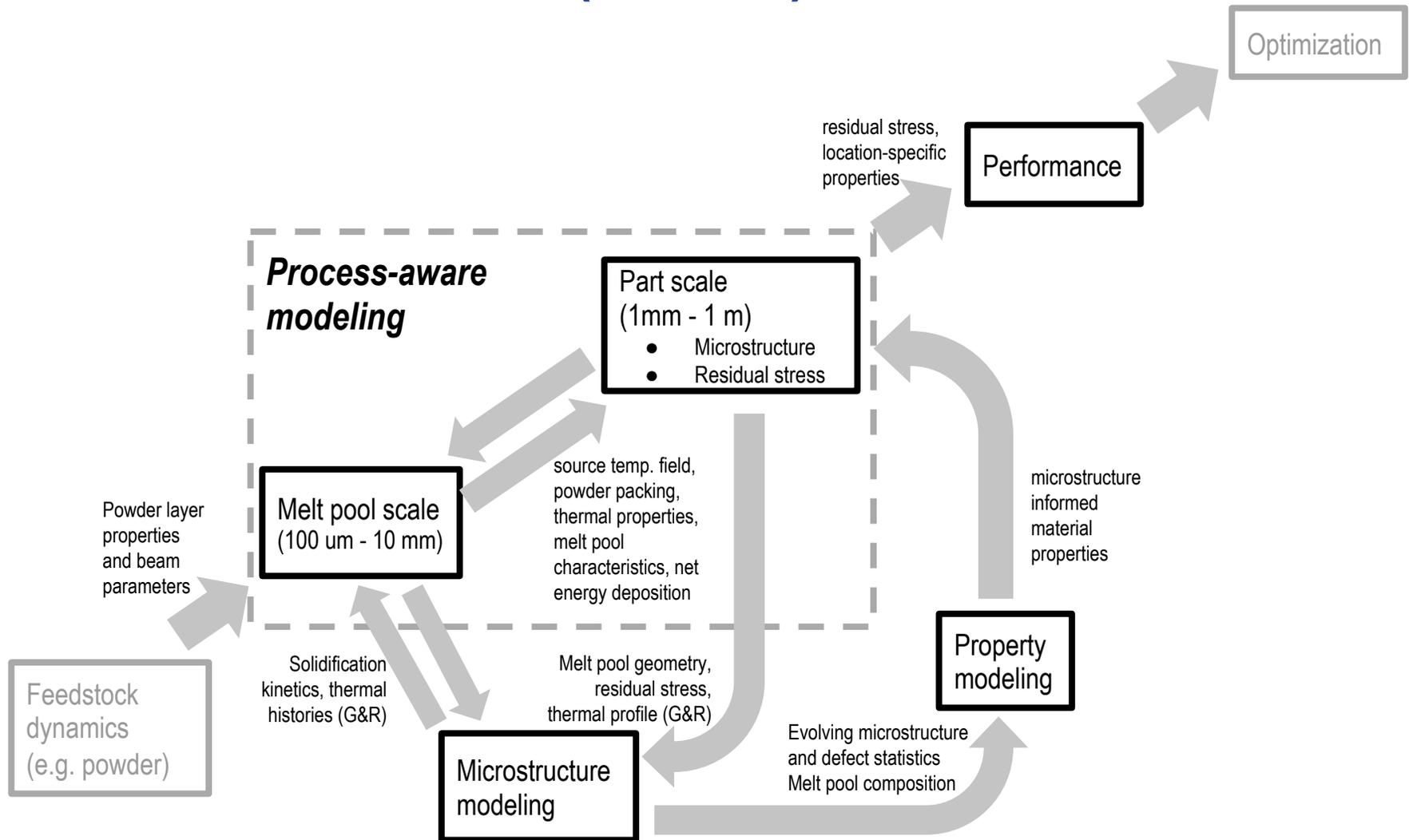
How are we going to address the problem? Workflow for Design Optimization of Additively Manufactured Parts (Integrated Computational Engineering)



John Allison, JOM 2006



Intra-HPC Workflow (ExaAM)

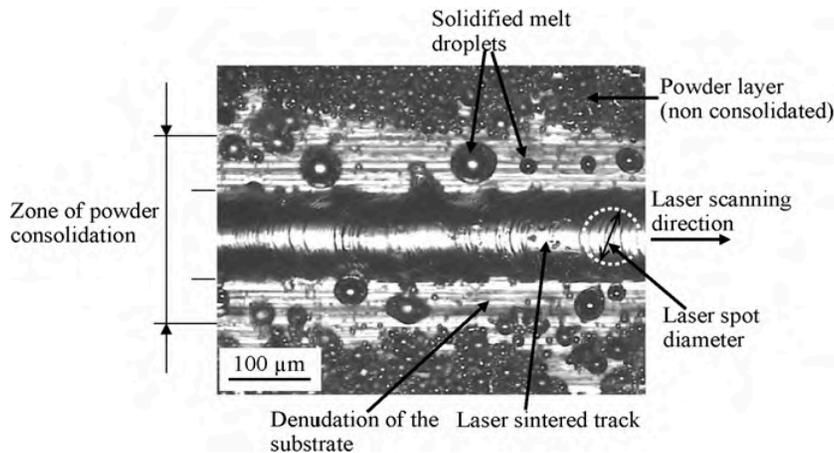


Mesoscopic 3D simulations provide insight into possible surface finish improvement options

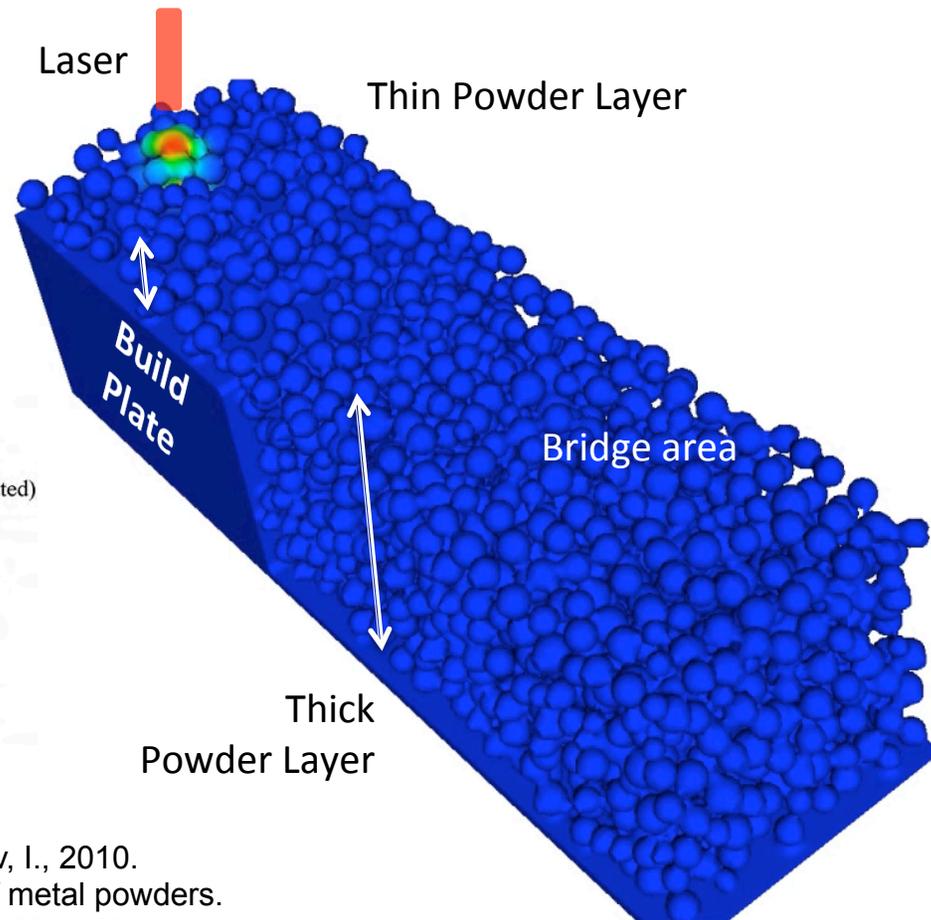
Khairallah, S.A., Anderson, A., 2014.

Mesoscopic Simulation Model of Selective Laser Melting of Stainless Steel Powder.

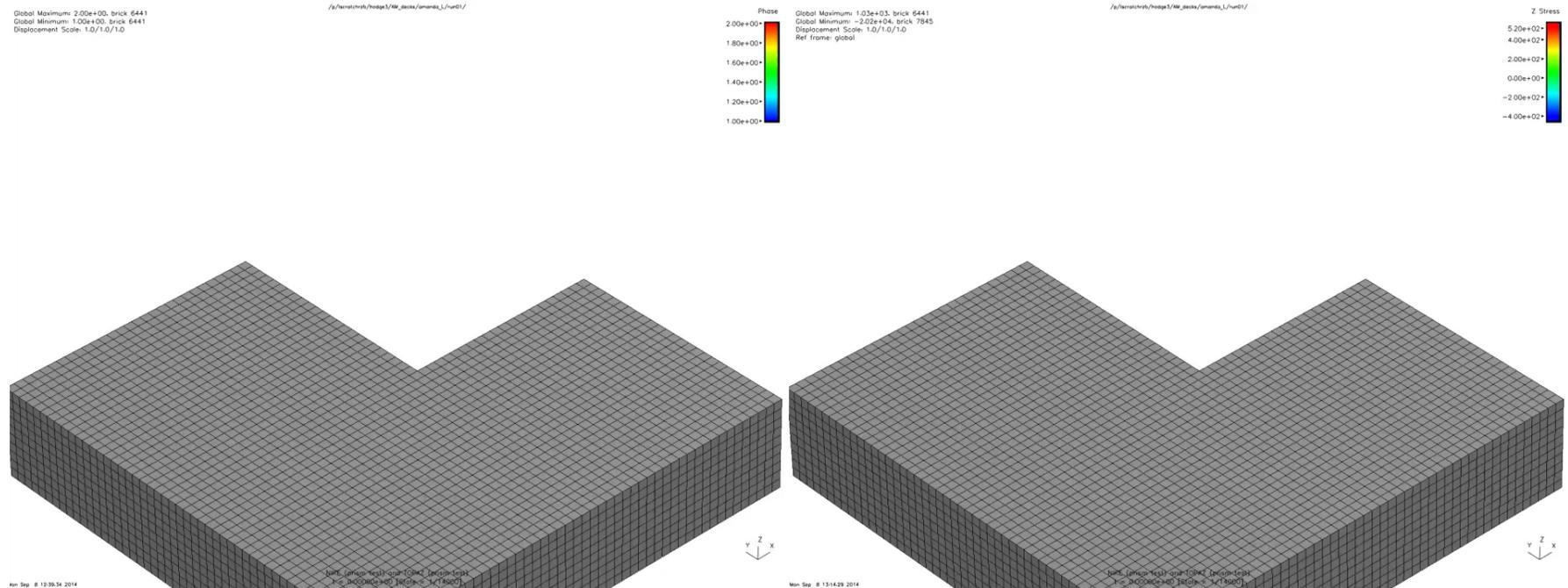
Journal of Materials Processing Technology 214, 2627-2636 DOI. 10.1016/j.jmatprotec. 2014.06.001.



Yadroitsev, I., Gusarov, A., Yadroitsava, I., Smurov, I., 2010.
Single track formation in selective laser melting of metal powders.
Journal of Materials Processing Technology 210, 1624-1631.



The effective medium model has been used to predict the development of residual stresses



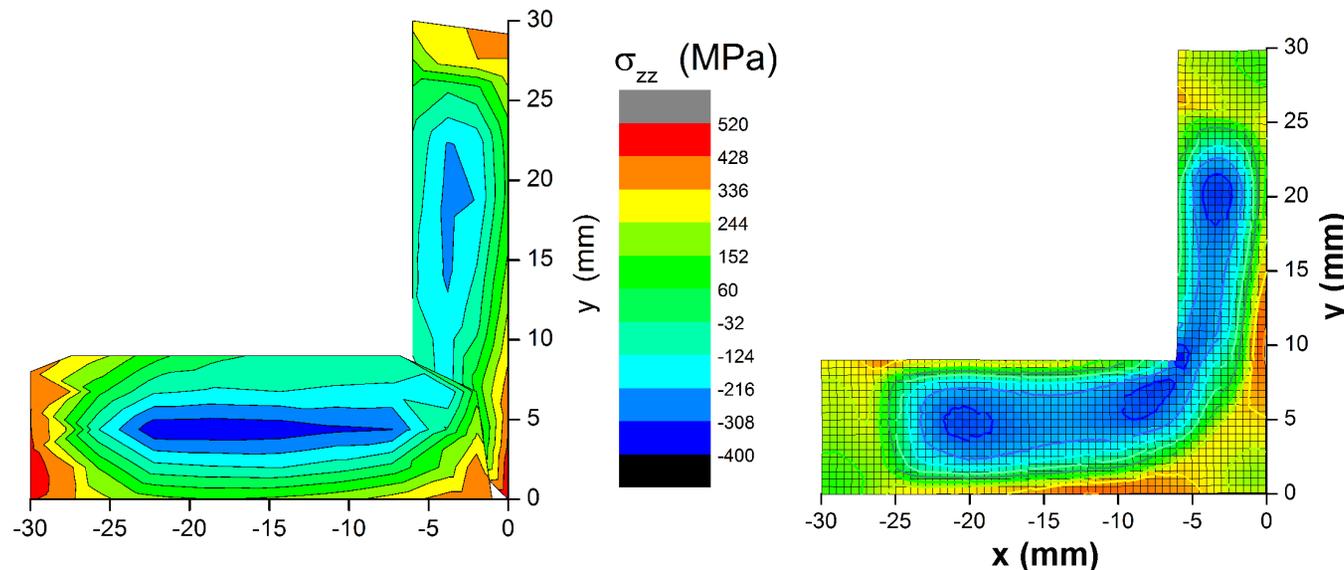
Hodge, N.E., Ferencz, R.M., Vignes, R.M., 2016. Experimental Comparison of Residual Stresses for a Thermomechanical Model for the Simulation of Selective Laser Melting. Additive Manufacturing DOI. <http://dx.doi.org/10.1016/j.addma.2016.05.011>.

Simulation compares well with digital image correlation combined with neutron diffraction

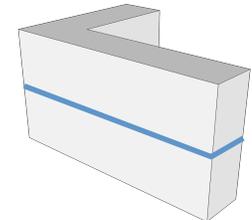
- Vertical stress assessed while on the build plate
- Perimeter values assessed from digital image correlation at horizontal cut

Experiments

Diablo



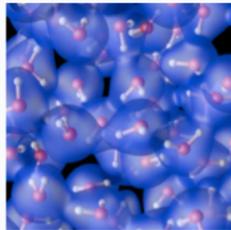
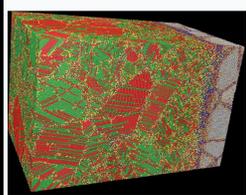
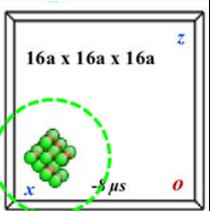
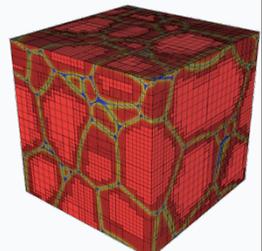
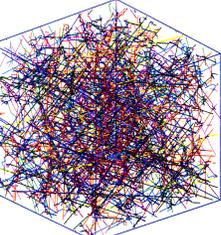
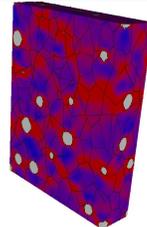
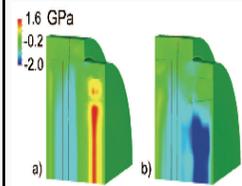
σ_{zz} normal
to plane 15mm
above base



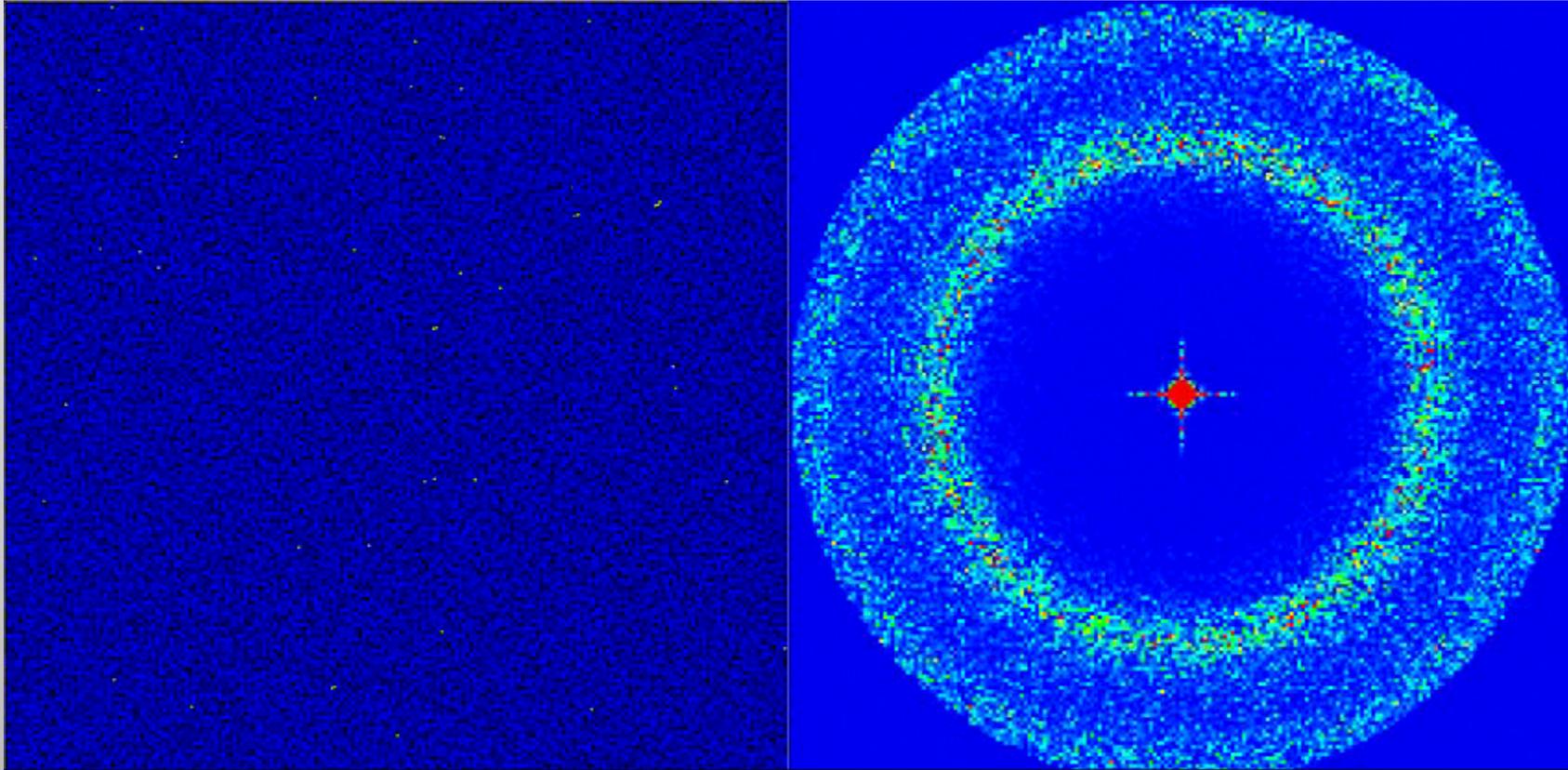
Hodge, N.E., Ferencz, R.M., Vignes, R.M., 2016. Experimental Comparison of Residual Stresses for a Thermomechanical Model for the Simulation of Selective Laser Melting. Additive Manufacturing DOI. <http://dx.doi.org/10.1016/j.addma.2016.05.011>.

We can model residual stresses and design for them

Exascale is about better Physics Fidelity: Coupling Atomistic with Microstructural Scales through Overlapping Simulations

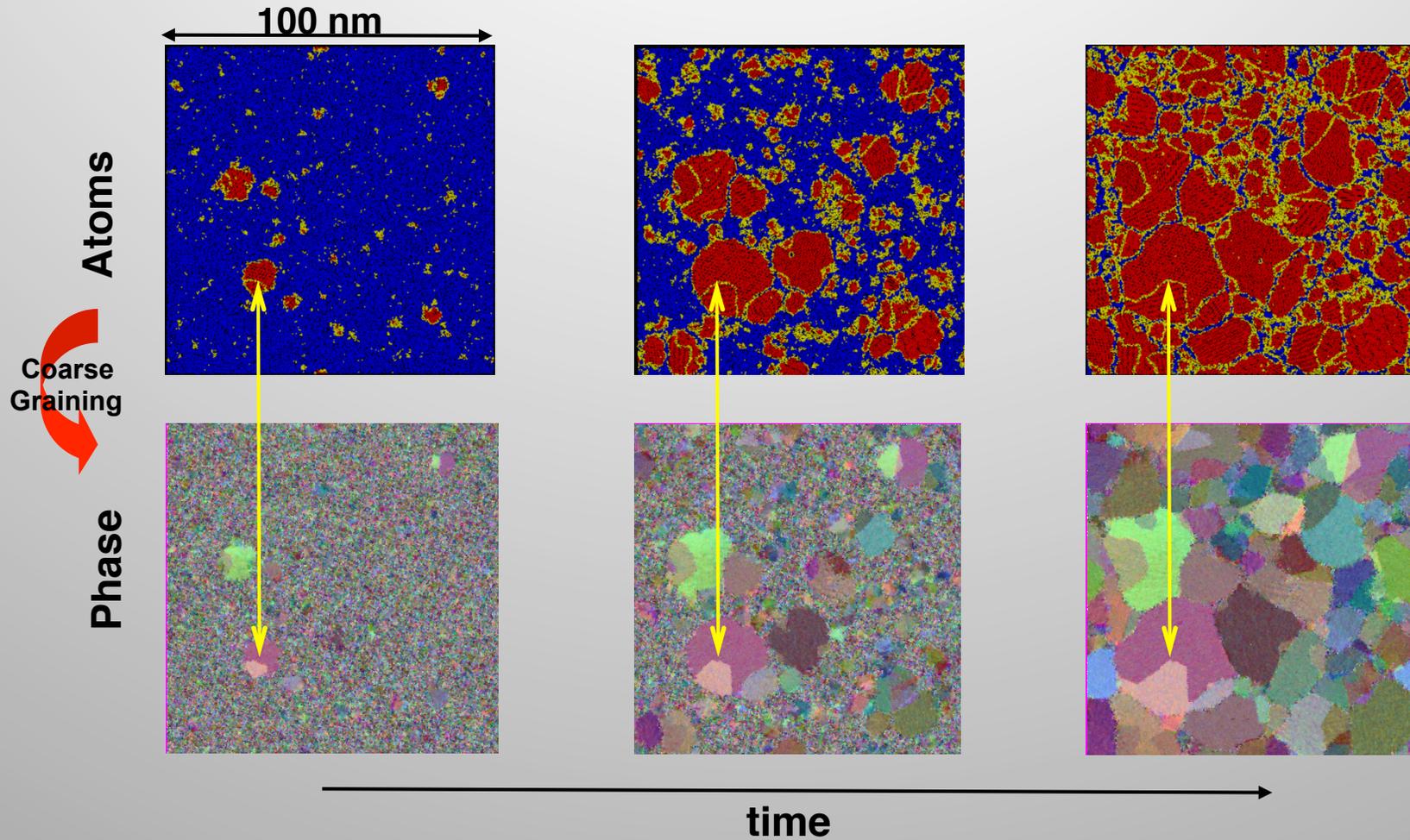
Ab-initio	Atoms	Long-time	Microstructure	Dislocation	Crystal	Continuum
Inter-atomic forces, EOS, excited states	Defects and interfaces, nucleation	Defects and defect structures	Meso-scale multi-phase, multi-grain evolution	Meso-scale strength	Meso-scale material response	Macro-scale material response
						
Code: Qbox/ LATTE Motif: Particles and wavefunctions, plane wave DFT, ScaLAPACK, BLACS, and custom parallel 3D FFTs Prog. Model: MPI + CUBLAS/CUDA	Code:SPaSM/ ddcMD/CoMD Motif: Particles, explicit time integration, neighbor and linked lists, dynamic load balancing, parity error recovery, and <i>in situ</i> visualization Prog. Model: MPI + Threads	Code: SEAKMC Motif: Particles and defects, explicit time integration, neighbor and linked lists, and <i>in situ</i> visualization Prog. Model: MPI + Threads	Code: AMPE/GL Motif: Regular and adaptive grids, implicit time integration, real-space and spectral methods, complex order parameter Prog. Model: MPI	Code: ParaDiS Motif: “segments” Regular mesh, implicit time integration, fast multipole method Prog. Model: MPI	Code: VP-FFT Motif: Regular grids, tensor arithmetic, meshless image processing, implicit time integration, 3D FFTs. Prog. Model: MPI + Threads	Code: ALE3D/ LULESH Motif: Regular and irregular grids, explicit and implicit time integration. Prog. Model: MPI + Threads

Rapid Solidification: Molecular Dynamics (MD) are now large enough to model the initiation of realistic microstructure

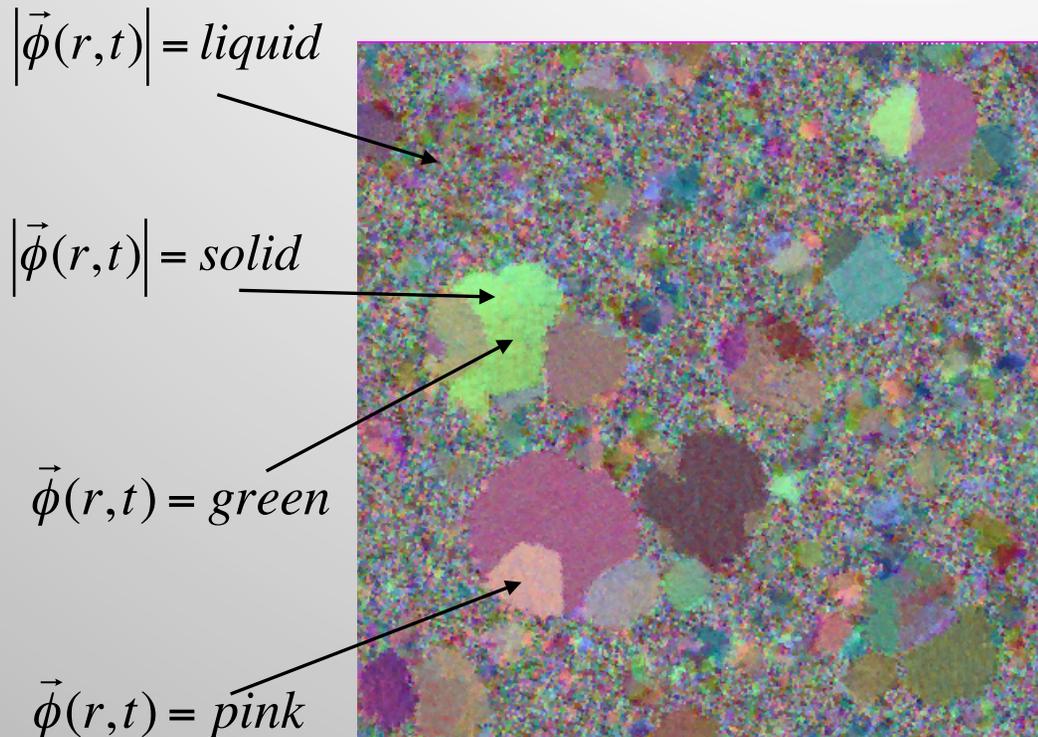


Simulations suggest novel in situ x-ray scattering experiments using emerging sources such as LCLS

Multi-scale paradigm: Phase-field model and MD simulations that overlap in space and time



What is Phase Field Modeling (PFM)?



Thermodynamic representation of phase (or “color”) everywhere

- Each color represents a different value of the phase field $\vec{\phi}$ (solid orientation)
- Free energy describes how colors interact and evolve
- Accuracy depends on fidelity of physics in the equations

Evolution Equations

$$F(P,T) = \int dx \left\{ |\nabla \vec{\phi}|^2 + f(\vec{\phi}, P, T) + \dots \right\}$$

$$\frac{\partial \vec{\phi}}{\partial t} = -\Gamma \frac{\delta F}{\delta \vec{\phi}} + \text{noise}$$

What does a crystallographic-aware phase-field model of polycrystal solidification look like?

Pusztai et al., have proposed a 3D quaternion-based phase-field model

- Represents crystal orientation with quaternion order parameter
- Quaternions are widely used to analyze crystallography of polycrystal interfaces
- Quaternion algebra is fast, efficient, avoids singularities, ...

Free Energy

$$F = \int \left[\frac{\varepsilon_\varphi^2}{2} |\nabla \varphi|^2 + f(\varphi, c, T) + HT[1 - p(\varphi)] \left(\sum_i (\nabla q_i)^2 \right)^{1/2} \right] d^3r$$

Evolution

$$\frac{\partial q_i}{\partial t} = -M_q \frac{\delta F}{\delta q_i} + \zeta_i = M_q \left[\nabla \cdot \left(D \frac{\nabla q_i}{|\nabla q_i|} \right) - 2\lambda q_i \right] + \zeta_i$$

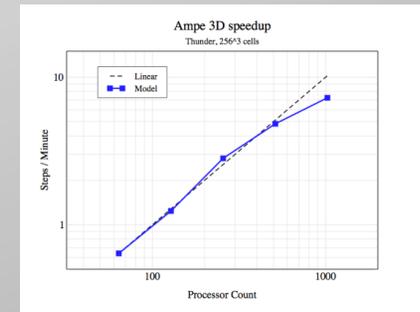
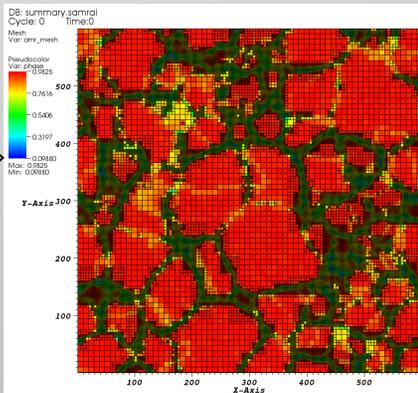
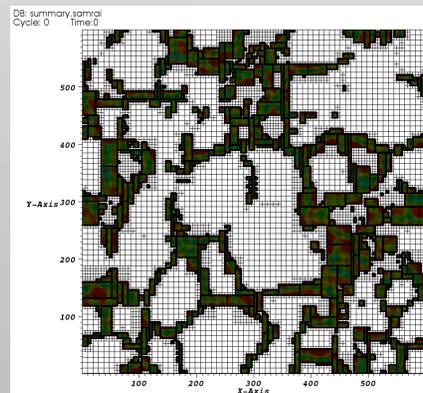
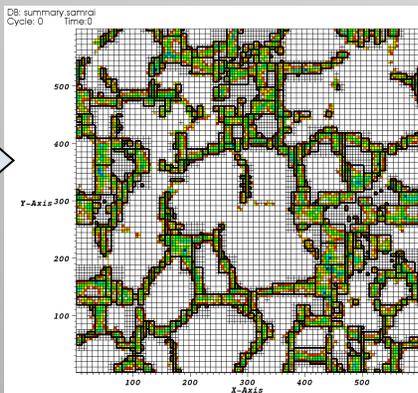
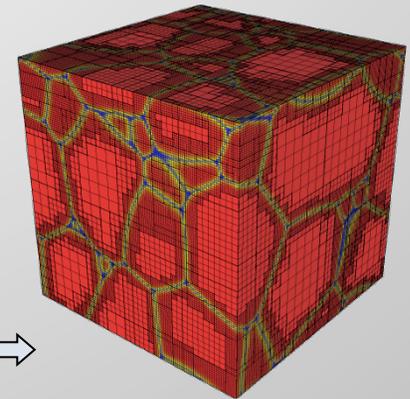
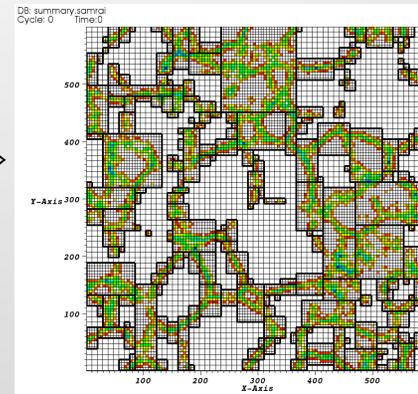
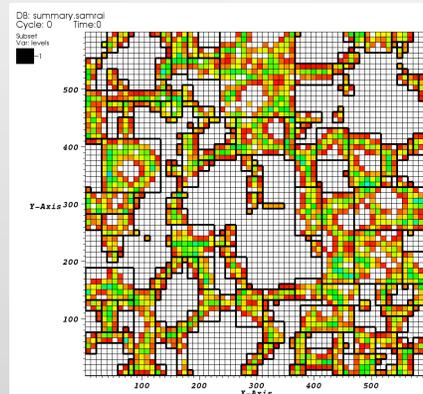
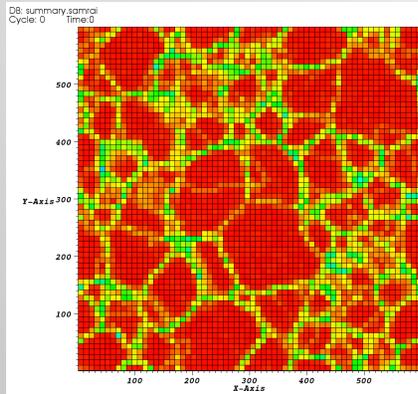
Where q_i is the quaternion order parameter, M_q is the associated mobility and ζ is the fluctuation in q .

We have implemented the Pusztai model in our 3D AMR code

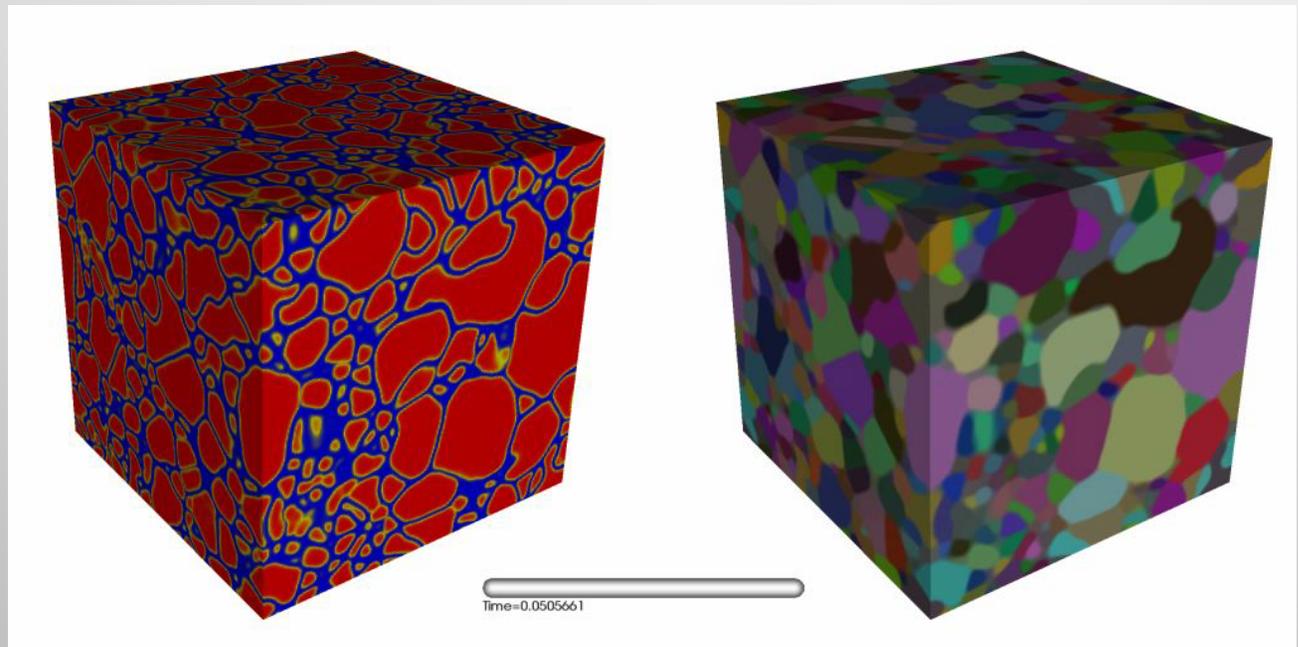
- Enhance energy functional to represent energetics of grain boundaries
- Crystal symmetry aware quaternion mathematics
- Extend energy functional to include elasticity and alloy concentration

Refs: T. Pusztai, G. Bortel, and L. Granasy, "Phase field theory of polycrystalline solidification in three dimensions," Europhys. Lett, 71 (2005) 131-137; Dorr, M.R., Fattebert, J.-L., Wickett, M.E., Belak, J.F., and Turchi, P.E.A., "A Numerical Algorithm for the Solution of a Phase-field Model of Polycrystalline Materials," J. Comp. Phys. Vol. 229, 626 (2010).

Representation of MD Data onto the AMR Grid Hierarchy using the SAMRAI AMR Library



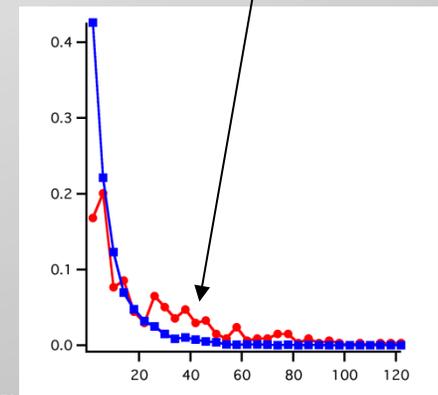
MD nucleated microstructure onto the micro-second hydro time-scale with the crystallographic quaternion model



Phase Order Parameter

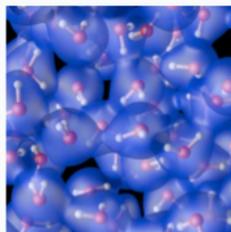
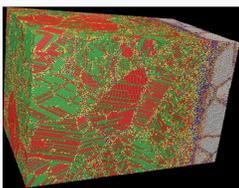
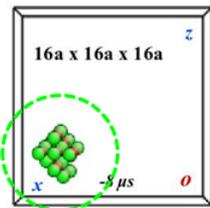
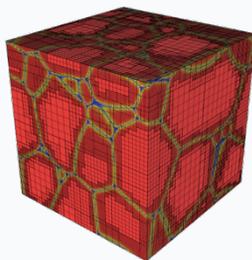
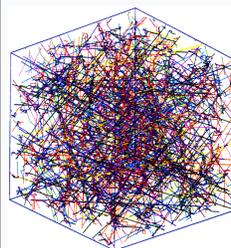
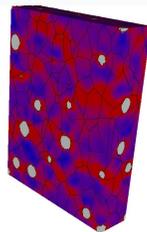
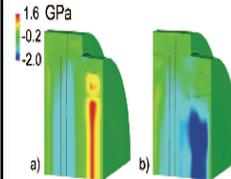
Quaternion Order Parameter

Growth of large grains
Blue: MD nucleation
Red: Phase-field evolution



While significant grain coarsening has occurred on the microsecond scale, the microstructure is far from log-normal

ExMatEx and Materials Science Workflow: The 7 Pillars of Materials Science and Adaptive Physics Refinement

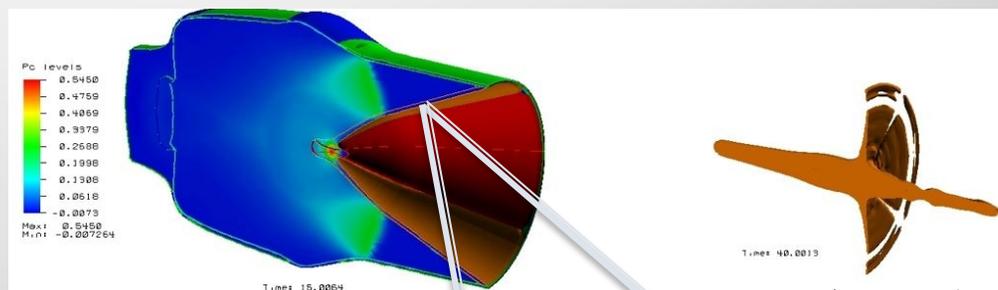
Ab-initio	Atoms	Long-time	Microstructure	Dislocation	Crystal	Continuum
Inter-atomic forces, EOS, excited states	Defects and interfaces, nucleation	Defects and defect structures	Meso-scale multi-phase, multi-grain evolution	Meso-scale strength	Meso-scale material response	Macro-scale material response
						
Code: Qbox/ LATTE	Code:SPaSM/ ddcMD/CoMD	Code: SEAKMC	Code: AMPE/GL	Code: ParaDiS	Code: VP-FFT	Code: ALE3D/ LULESH
Motif: Particles and wavefunctions, plane wave DFT, ScaLAPACK, BLACS, and custom parallel 3D FFTs	Motif: Particles, explicit time integration, neighbor and linked lists, dynamic load balancing, parity error recovery, and <i>in situ</i> visualization	Motif: Particles and defects, explicit time integration, neighbor and linked lists, and <i>in situ</i> visualization	Motif: Regular and adaptive grids, implicit time integration, real-space and spectral methods, complex order parameter	Motif: “segments” Regular mesh, implicit time integration, fast multipole method	Motif: Regular grids, tensor arithmetic, meshless image processing, implicit time integration, 3D FFTs.	Motif: Regular and irregular grids, explicit and implicit time integration.
Prog. Model: MPI + CUBLAS/CUDA	Prog. Model: MPI + Threads	Prog. Model: MPI + Threads	Prog. Model: MPI	Prog. Model: MPI	Prog. Model: MPI + Threads	Prog. Model: MPI + Threads

Use Case: Shaped-charge jets, breakup and 3D effects (e.g. spinning) require crystal plasticity and anisotropy

What is required:

Resolution: 10^{12} zones (10 cm cube)
Simulation time: 100 μ sec (10^5 steps)
Strain rate: 10^6 /sec
Strain: 1-3
Using Small Strain Crystal Plasticity Model:
 $\sim 10^4$ sec (~ 3 h) wall clock on 10^9 cores
Large Strain Crystal Plasticity Model: 10x
Twinning / Scale Bridging Model: 100x

ALE3D simulation of shaped-charge jet
(Rose McCallen, LLNL)



$$\Delta \varepsilon \geq 1$$

What we can do today:

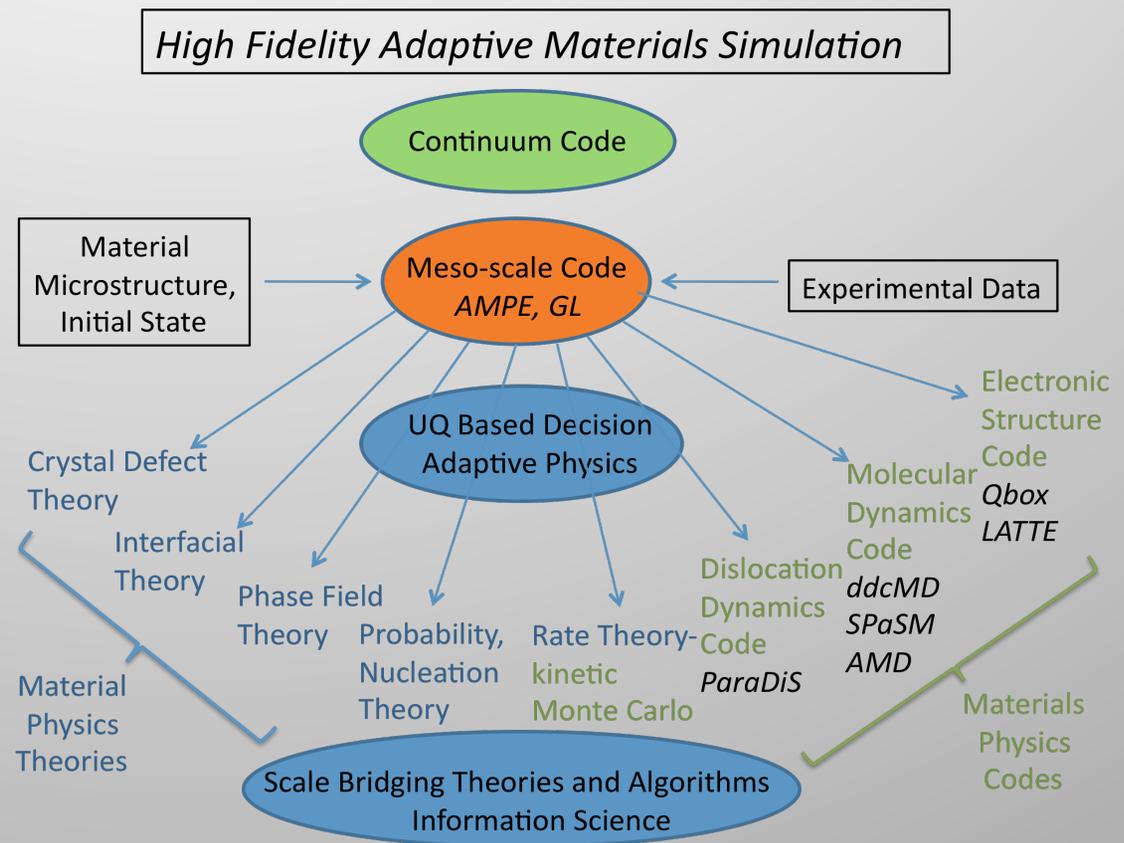
Crystal plasticity simulation of high rate deformation (Nathan Barton, LLNL)
Model: Small Strain Crystal Plasticity
Number Zones: 10^7 (100 micron cube)
Simulation time: 10 μ sec (10^4 steps)
Strain rate: 10^6 /sec
Strain: 0.15
Wall Clock: 1 day on 1/10 Cielo



$$\Delta \varepsilon = 0.15$$

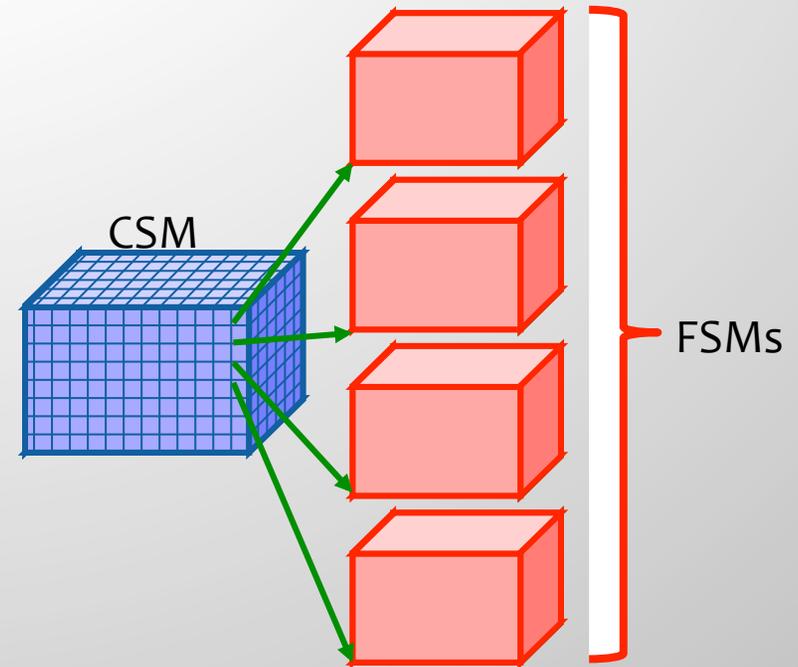
Objective: Full utilization of exascale concurrency and locality

- Task-based embedded Scale-Bridging escapes the traditional synchronous SPMD paradigm and exploits the heterogeneity expected in exascale hardware.
- To achieve this, we are developing a UQ-driven *adaptive physics refinement* approach.
- Coarse-scale simulations dynamically spawn tightly coupled and self-consistent fine-scale simulations as needed.
- This *task-based* approach naturally maps to exascale heterogeneity, concurrency, and resiliency issues.



Direct multi-scale embedding requires full utilization of exascale concurrency and locality

- *Brute force multi-scale coupling*: Full fine scale model (FSM, e.g. a crystal plasticity model) run for every zone & time step of coarse scale mode (CSM, e.g. an ALE code)
- *Adaptive Sampling*:
 - Save FSM results in database
 - Before running another FSM, check database for FSM results similar enough to those needed that interpolation or extrapolation suffices
 - Only run full FSM when results in database not close enough

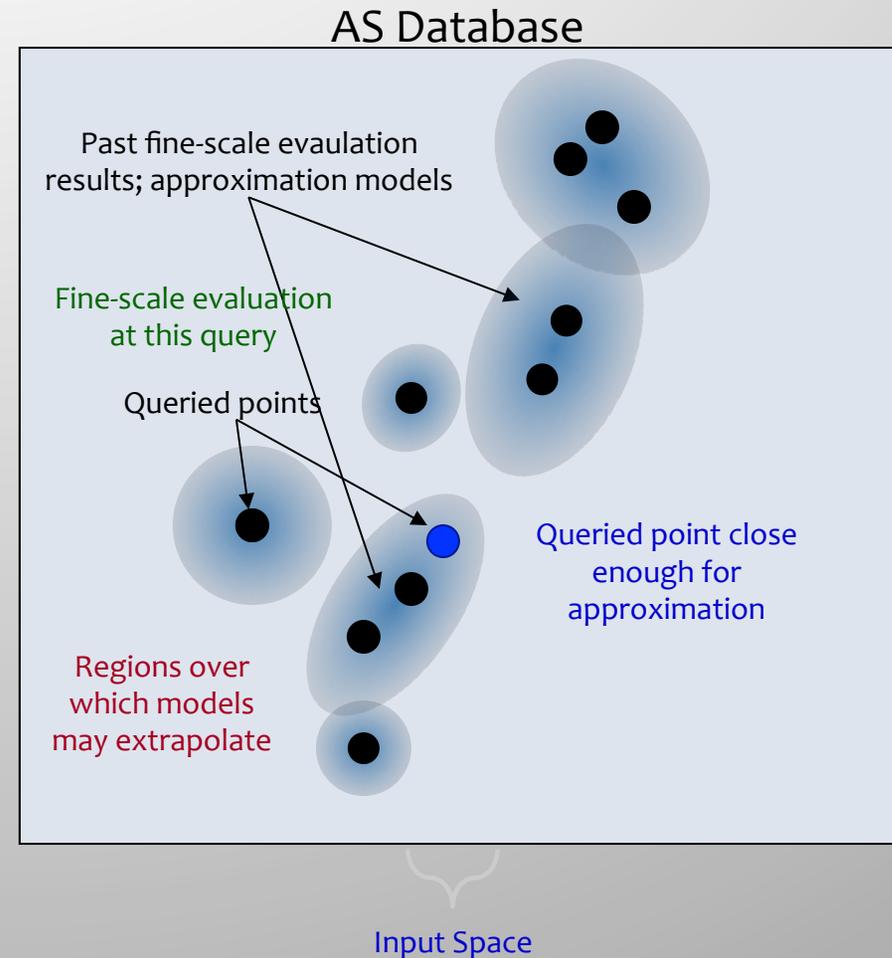


- Heterogeneous, hierarchical MPMD algorithms map naturally to anticipated heterogeneous, hierarchical architectures
- Escape the traditional bulk synchronous SPMD paradigm, improve scalability and reduce scheduling
- Task-based MPMD approach leverages concurrency and heterogeneity at exascale while enabling novel data models, power management, and fault tolerance strategies

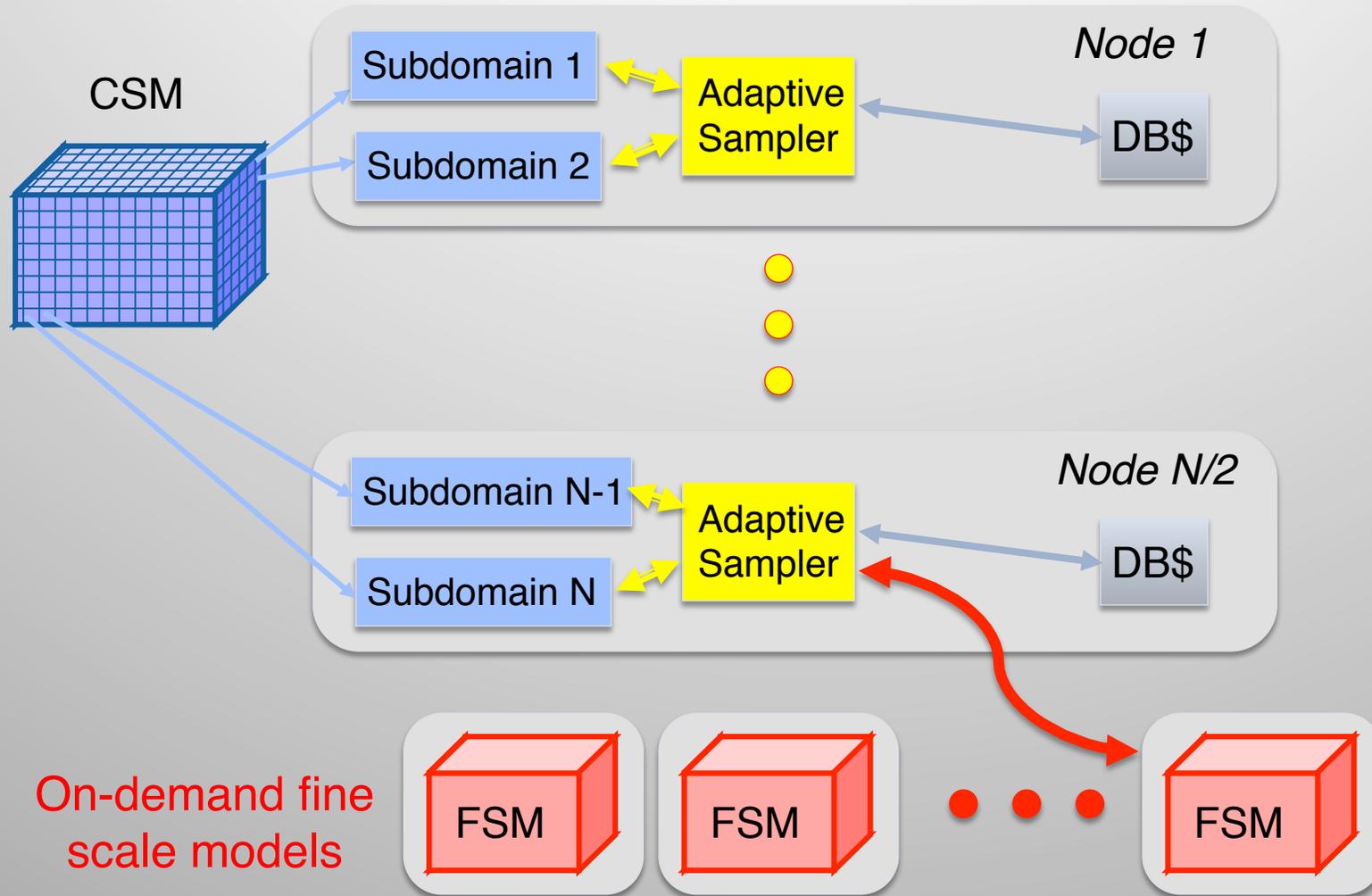
Ref: Barton *et.al*, 'A call to arms for task parallelism in multi-scale materials modeling,' *Int. J. Numer. Meth. Engng* 2011; 86:744–764

Adaptive Sampling builds response on the fly

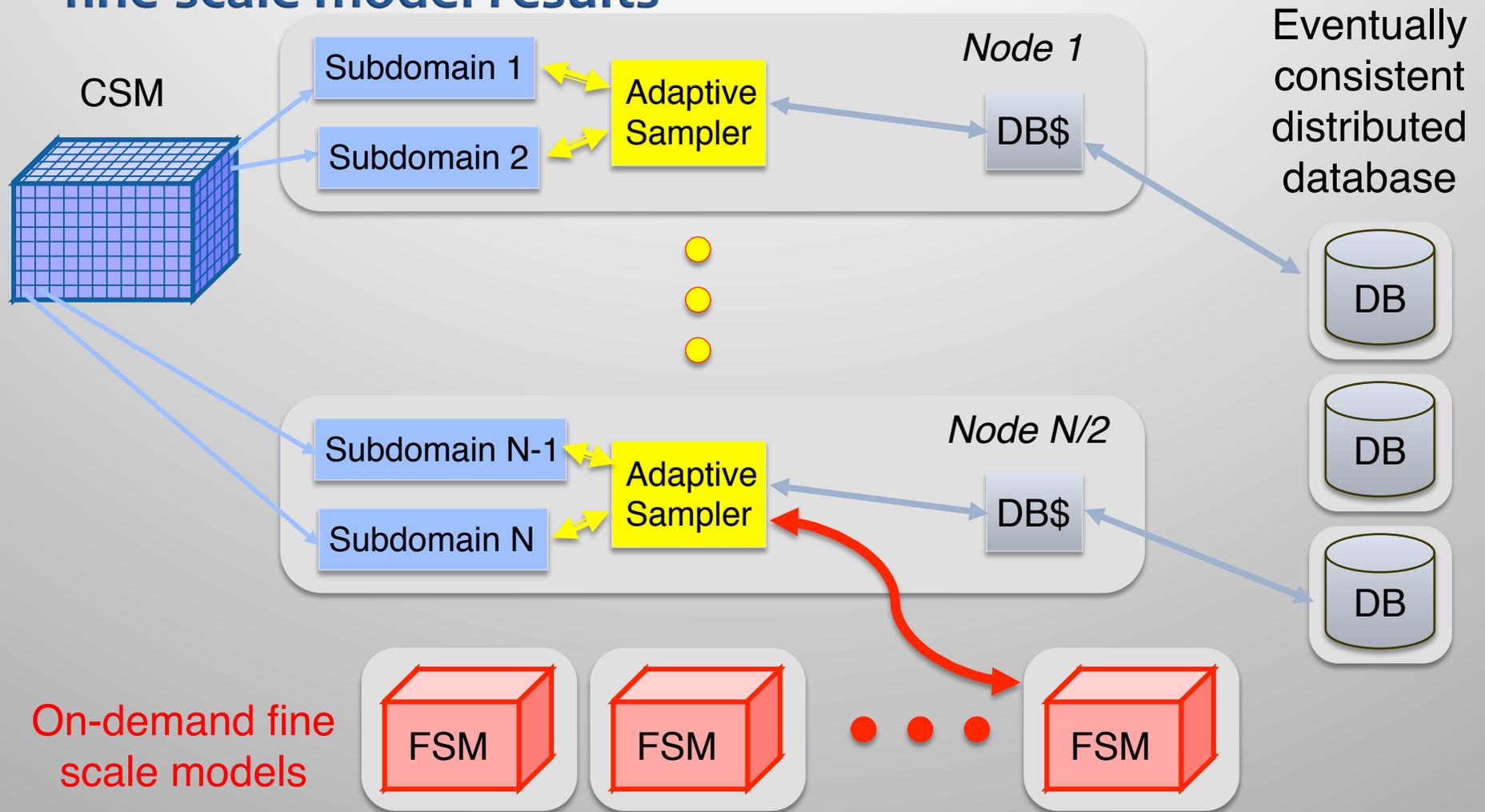
- Coarse scale model queries database for fine-scale material response
- If possible, approximate response from past evaluations
- Otherwise perform fine scale evaluation
- Fine-scale evaluations grow database



Tradeoff: re-use vs. re-computation of expensive fine-scale model results



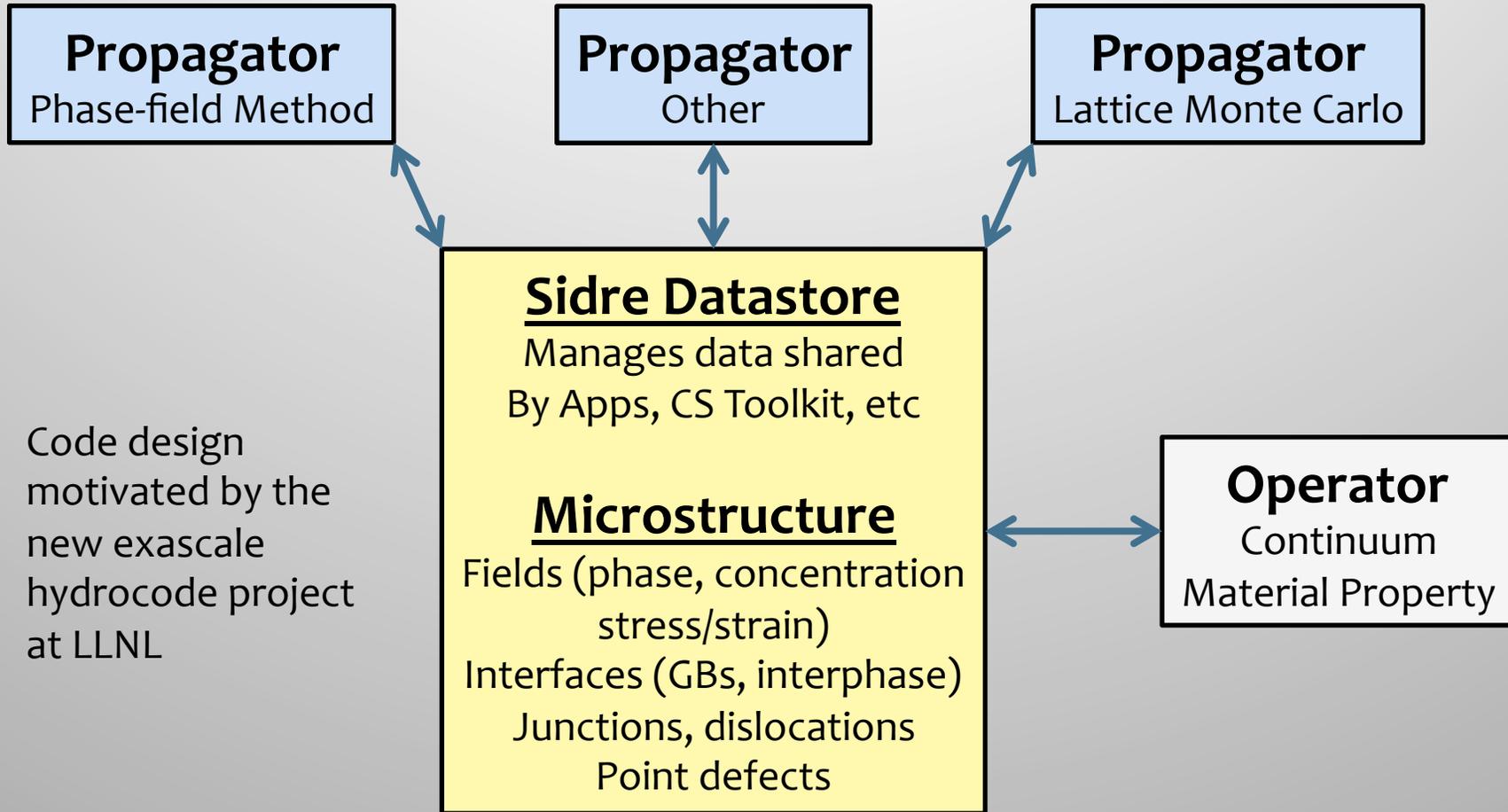
Tradeoff: re-use vs. re-computation of expensive fine-scale model results



Several practical challenges must be addressed

- Interoperability between (often) legacy coarse and fine-scale models, and backing database
 - Scale-bridging implementations to date often “hacked” to work within MPI
 - “Monolithic” programming models (e.g. Charm++, Chapel) may require impractical rewrites of entire code base
 - Web/cloud technologies are often service-based, and emphasize programmer productivity, code agility, and code maintainability over performance
 - Component-based frameworks (e.g. Uintah, Pathos) have demonstrated success for multiphysics HPC applications
- Distributed database
 - Most implementations depend on conventional TCP/IP, which is often not supported on HPC software stacks (Infiniband)
 - Stanford’s RAMCloud an interesting option

Seeds for Contemplation: Promote State (Data) be be the first class citizen.



More Seeds for Contemplation

- **Workflow Manager:** Imagine imbedded chips (Neuromorphic? Vickie from I,Robot) that monitor and predict work flow during the simulation and moves tasks (both work and data) to optimal locations.
- **Machine Learning Constitutive Models:** the constitutive model is a response relation (e.g. stress-strain) that depends on the constitution of the material. It is the closure relation in the continuum conservation equations. Imagine using an ML model that learns from experimental data and direct numerical simulation.
- **Programming:** What does a scientific application look like when expressed in a programming framework designed for AI? e.g. TensorFlow code for $O(N)$ electronic structure.

