Possible Stack for Parallel Programming Models for Scientific Computing

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Contents

- 1. Challenges for parallel programming
- Possible domain-specific programming model for scientific computing
- 3. An instance : JASMIN framework
- 4. Conclusion

1. Challenges for parallel programming

In the past decade: Parallel Programming

18 years of legacy codes for various applications such as fusion energy, high energy physics, climate forecasting, facility and experimental design, materials, chemistry, etc.

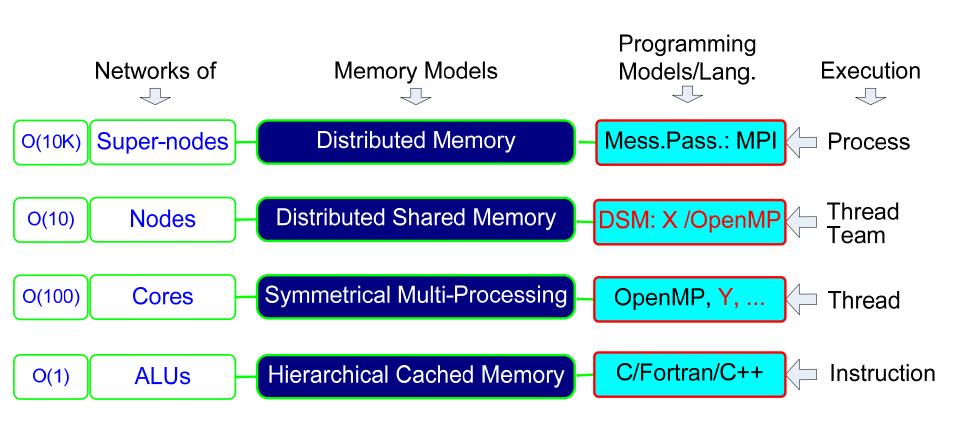
MPI for O(10K) parallelism, OpenMP for O(1) parallelism.

Terascale to Petascale Machines

In the next decade: Three parallel computing Points

Terascale Laptop: Uni-Supernode — few-node — many-core Petascale Desktop: Multi-Supernode — multi-node — many-core Exascale Center: Many-Supernode — many-node — many-core 2018 Goal: Make Common elements Petascale = Terascale + more; Exascale = Petascale + more.

Machine Parallelism: Nested/Hybrid Programming Models



Emerging models: Accelerators, Resilience, Energy.

In the next ten years: Parallel Programming

Petascale to Exascale Applications

Reconstruction of 15 years of MPI legacy codes, New generation codes for multi-physics complex systems

Big and increasing gaps for realization

Nested/Hybrid Parallel Programming Models

MPI O(10K), DSM-X O(10), OpenMP O(100), ILP O(1);

Accelerator: OpenCL/CUDA; Resilience, Energy, ...

Petascale to Exascale machines

In the next ten years: Parallel Programming

Big and increasing gaps for realization

- 1. Evolving and nested/hybrid parallel programming
- 2. Complex management for data structure, hierarchical memory, irregular communication, etc.;
- 3. Multilevel/Hybrid load imbalance arising from physics, numerical stencils, communication, run time status of machine, etc.;
- 4. Implementation and integration for fast numerical algorithms;
- 5. Code extensibility for more and more complex applications: O(10K~1M) lines;
- 6. Data management and visualization.

Great Challenges:

Fussy Parallel Programming
Load imbalance
Fast algorithms implementation
Code complexity
Visualization Interfaces

Good solutions

Scientific
Computing and
Engineering
Applications

common

Frameworks

Software

Applicaiton A: parallel code A

.....

Application N: parallel code N

Frameworks enable:

Applications, algorithms, parallel experts, computer experts can cooperate tightly in the development of complex codes.

Encapsulates and seperates fussy works of parallel computing from applications (e.g. data structure, parallel programming, numerical libraries);

Encapsulates code complexity and applies software engineering for code extensibility and maintenance;

Accelerates the developments of codes towards petascale/exascale.

Meet the expectation of application/physics experts

Think Parallel, Write Sequential

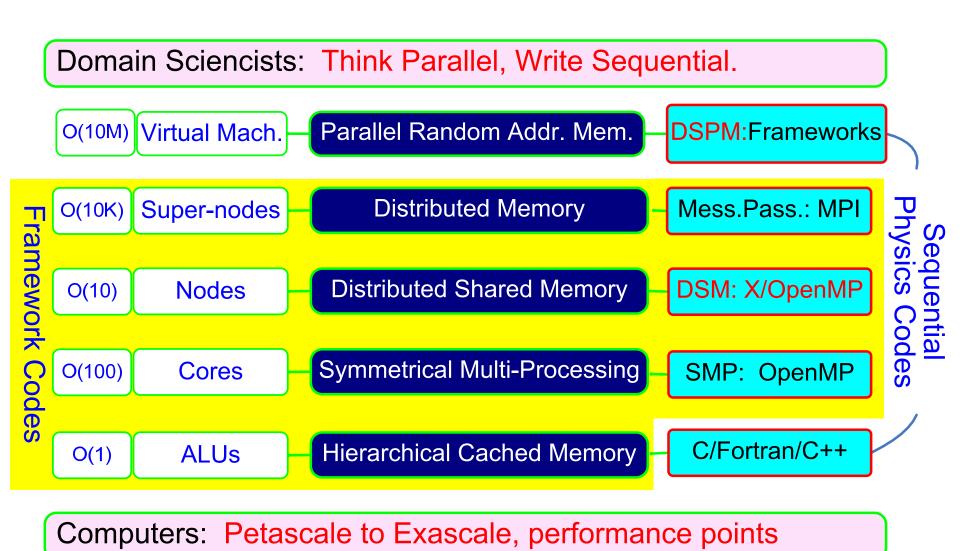
- 1. Significantly simplify or reuse the parallelism patterns using the emerging programming models;
- 2. The return on their rewrite efforts can be leveraged for multiple years even the machine is rapidly changing (e.g., 20 years? ----- 20 or more years for MPI);
- 3. Once infrastructure in place, ratio of science experts vs. parallel experts: 10:1, physics added as serial code, now and in the future.

--M.Heroux, LANL, July. 2011, DOE Workshop on Exascale Programming Challenges

Possible Domain Specific Programming Model for Scientific Computing

----- Think Parallel, Write Sequential

Possible Stack of Programming Models: Frameworks



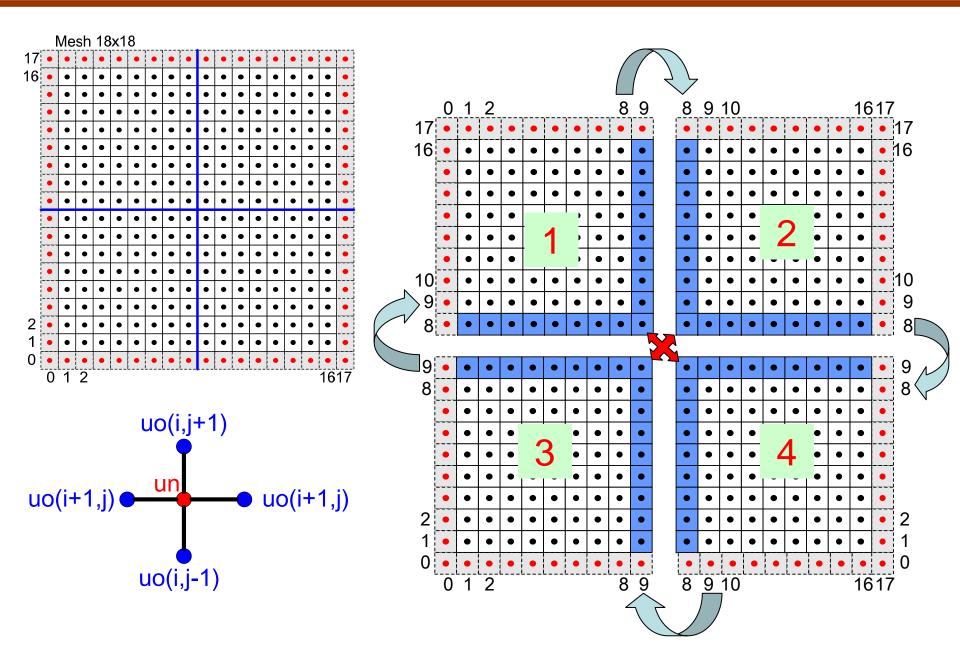
Framework-based DSPM is possible for scientific computing?

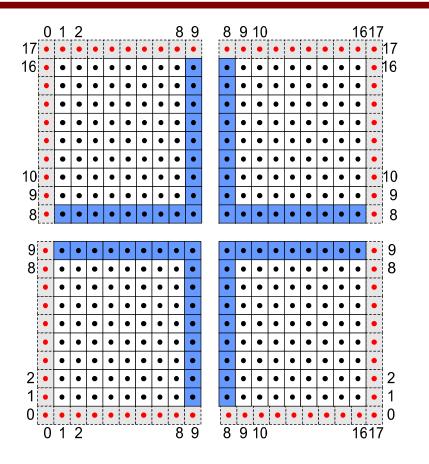
DDM Graph Based Patterns: Halo exchanges, Collectives

separate parallel programming from serial codes:

Computational Patterns

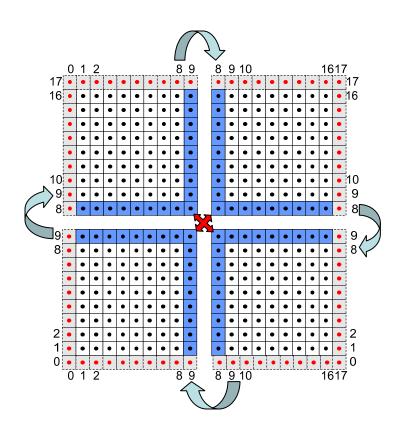
Digraph Based Patterns: data driven, dynamic tasks





```
ist(1)=ist(3)=1, iend(1)=iend(3)=8;
jst(1)=jst(2)=9, jend(1)=jend(2)=16;
ist(2)=ist(4)=9, iend(2)=iend(4)=8;
jst(3)=jst(4)=1, jend(3)=jend(4)=16.
```

ENDDO for i,j,b



Do MPI message passing fill ghost cells for uo;

DO b= 1, 4 in parallel

DO i = ist(b), iend(b)

DO j = jst(b), jend(b)

uo(i,j+1)

uo(i+1,j)

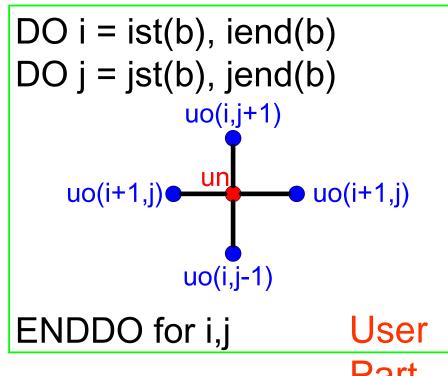
uo(i,j-1)

ENDDO for i,j,b

```
DO MPI message passing
    fill ghost cells for uo;
DO b= 1, 4 in parallel
```

serial computation for block b;

ENDDO for b



Part

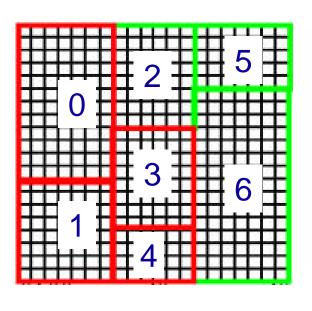
DO b=1, NB fill ghost cells for uo;

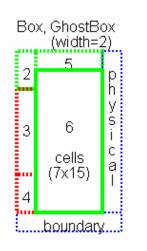
DO p=1, NP in parallel

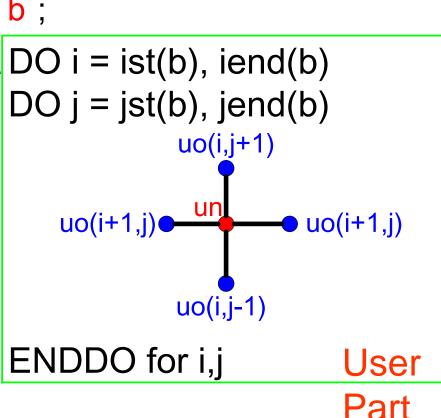
DO b=bst(p), bend(p)

serial computation for block b;

ENDDO for b ENDDO for p in parallel







3. An instance: JASMIN framework

----- Think Parallel, Write Sequential

JASMIN: Parallel Patterns + Libarary → DSPM

DDM Graph Based Patterns: Halo exchanges, Collectives

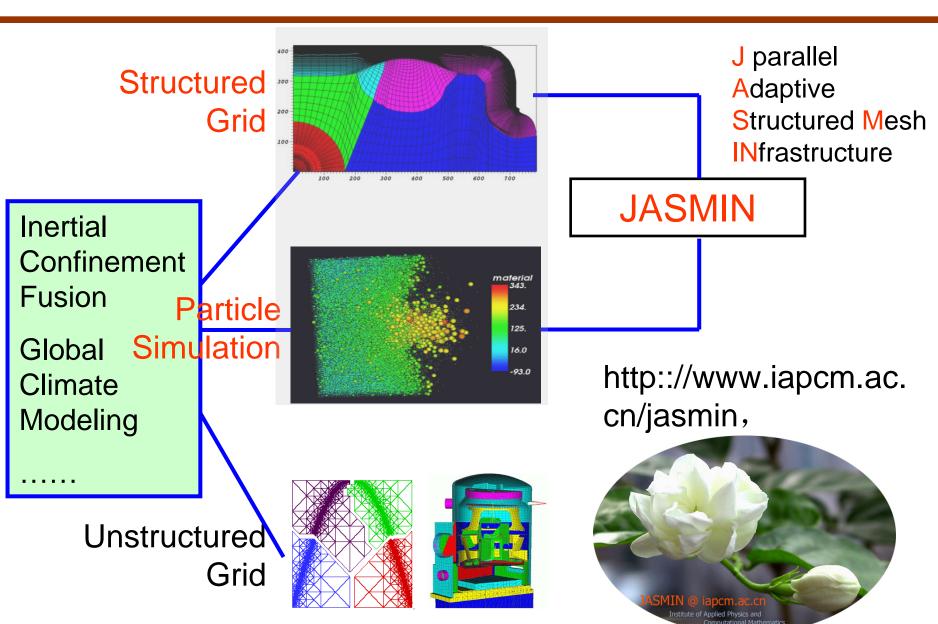


C++ Components = parallelism/libs + interfaces -> serial

numerical subroutines

Digraph Based Patterns: data driven, dynamic tasks, barriers

3.1 JASMIN



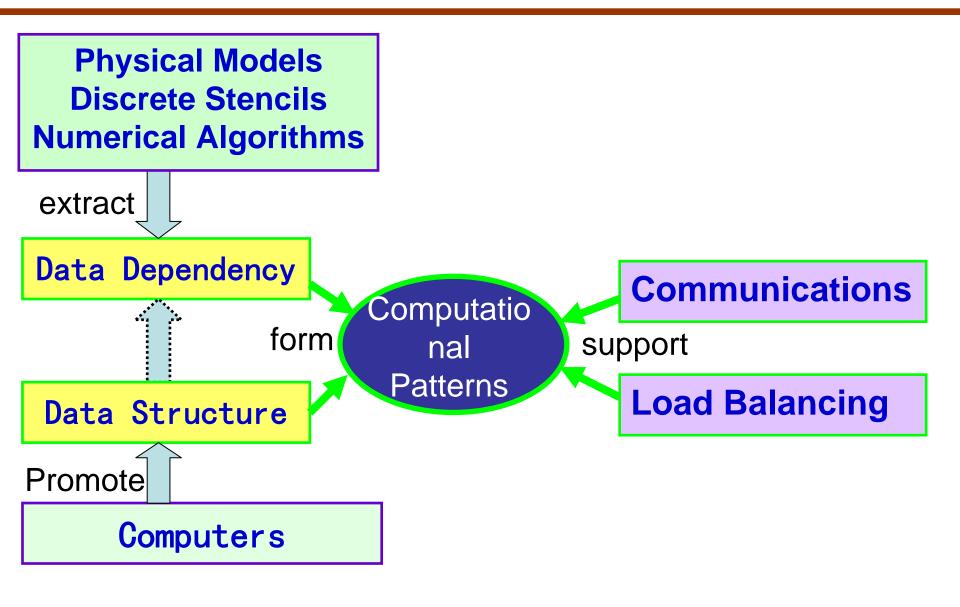
3.1 JASMIN

Motivations:

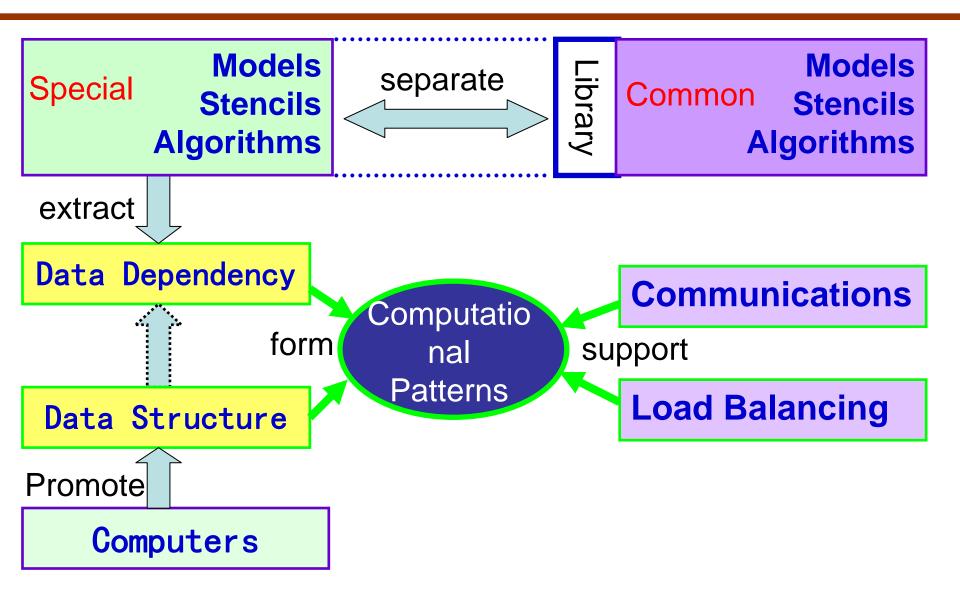
Supports the developments of parallel codes for large scale scientific computing on personal computers.

- Hides parallel programming using millons of cores and the hierarchy of parallel computers;
- Integrates the efficient implementations of parallel fast algorithms;
- Provides efficient data structures and solver libraries;
- Supports software engineering for code extensibility.

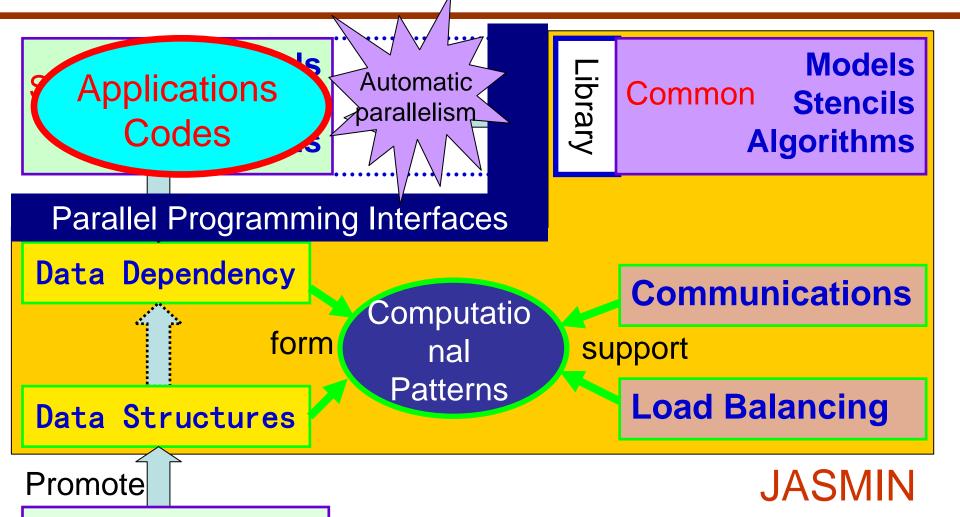
3.2 Basic Ideas



3.2 Basic Ideas



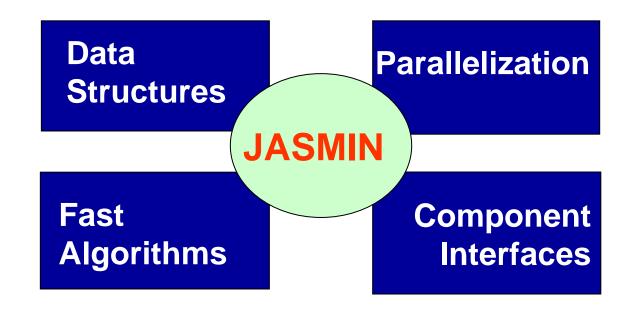
3.2 Basic Ideas



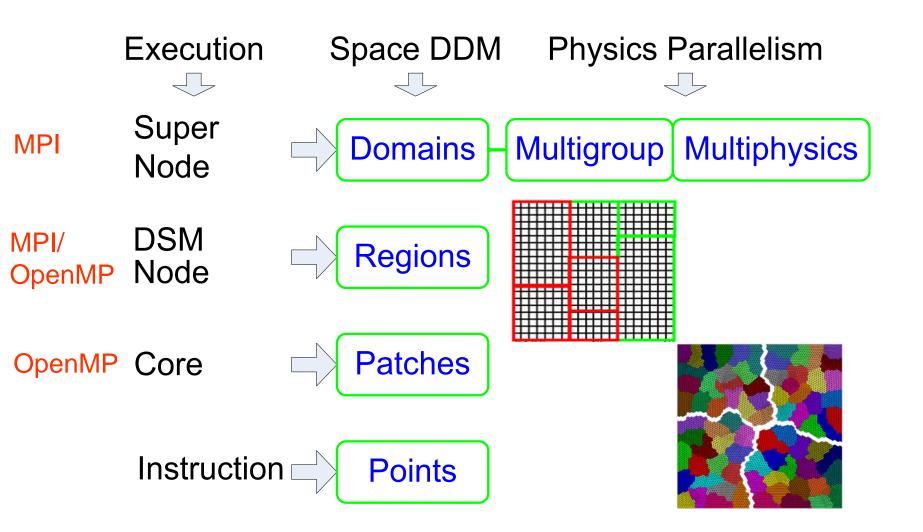
Computers

parallel middlewares for scientific computing on structured meshes

3.3 key factors

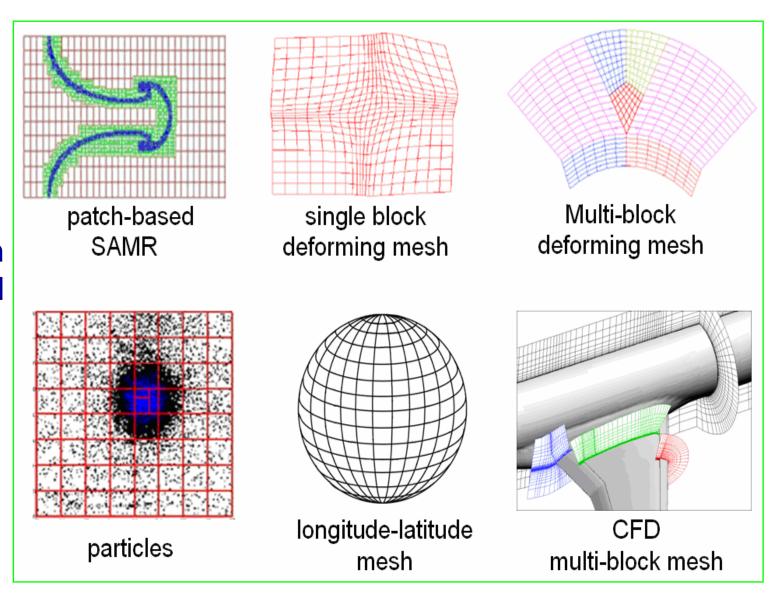


3.3.1 Data Structures



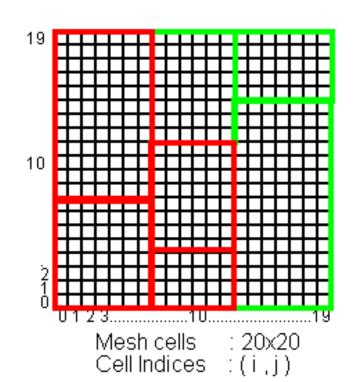
Neighboring Graph Based Patterns: Halo exchanges, Collectives

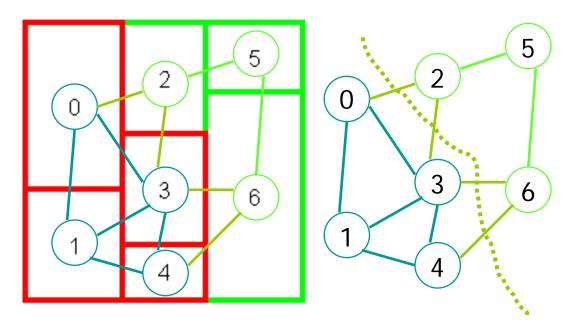
3.3.1 Data Structures



Mesh supported

3.3.2 Communications

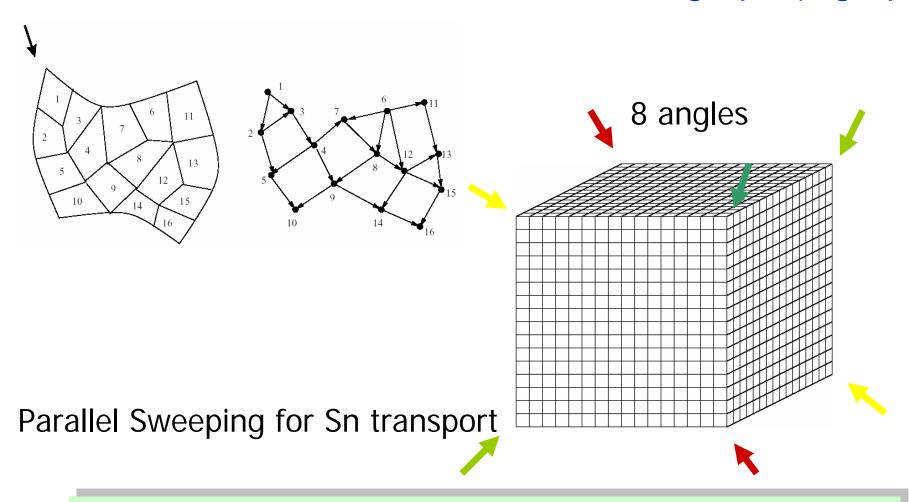




Distributed Undirected Graph (2 processors)

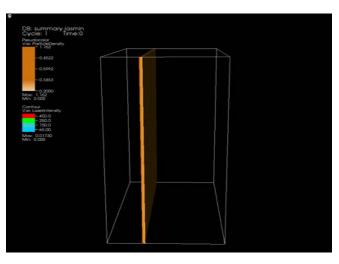
3.3.2 Communications

distributed directed graph (digraph)

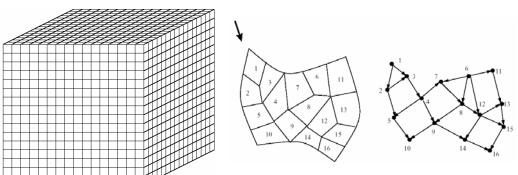


Digraph Based Patterns: data driven, dynamic tasks

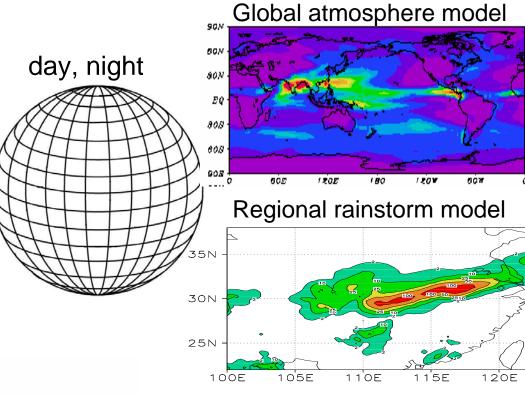
3.3.3 dynamic load balancing

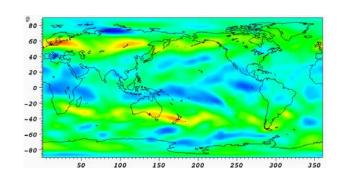


extreme load imbalance

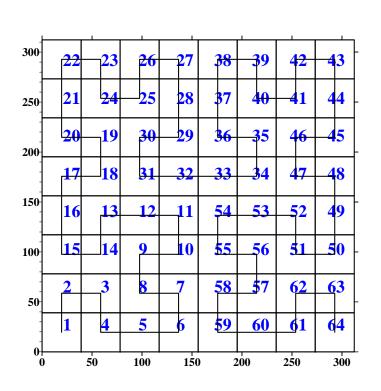


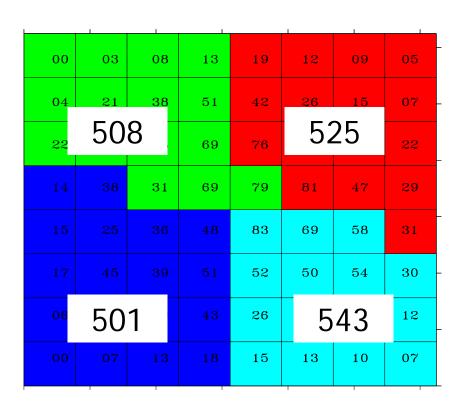
Radiation and neutron transport





3.3.3 dynamic load balancing

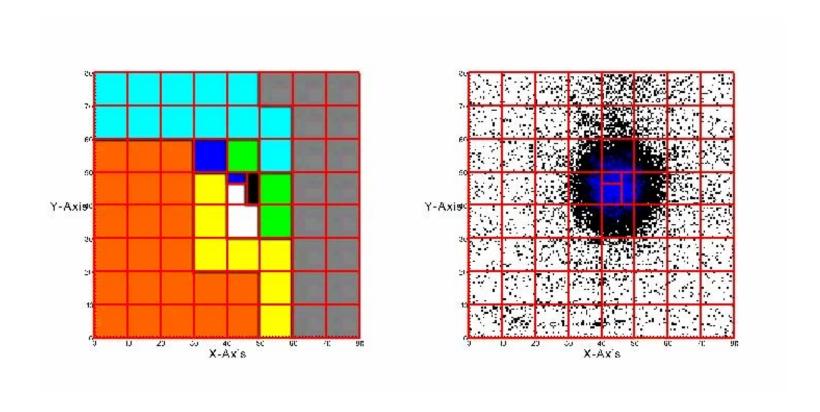




space filling curves : 2D,3D->1D

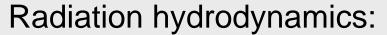
MAW/Cycle methods: 1D balancing

3.3.3 dynamic load balancing



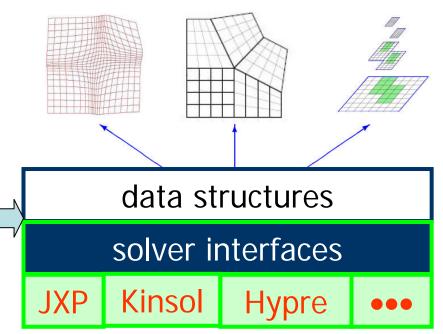
Redistributes particles among 8 Processes

3.3.4 Solvers for linear systems



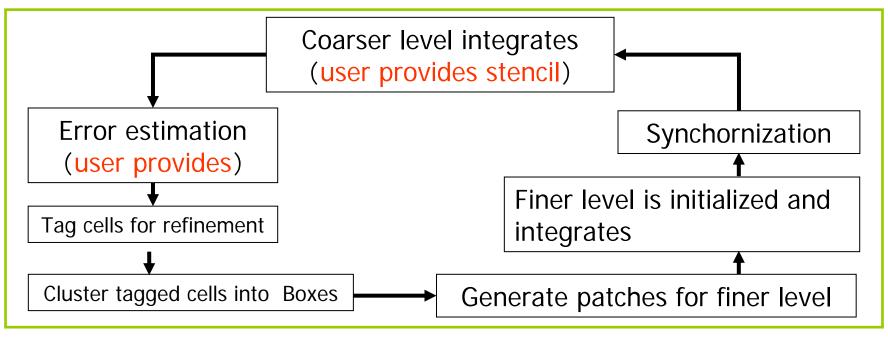
Usual algorithms : $O(N^{1.5})$

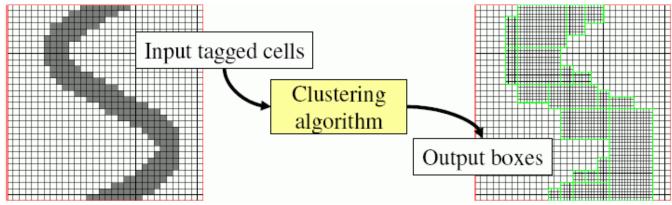
PAMG+DDM: O(NlogN)



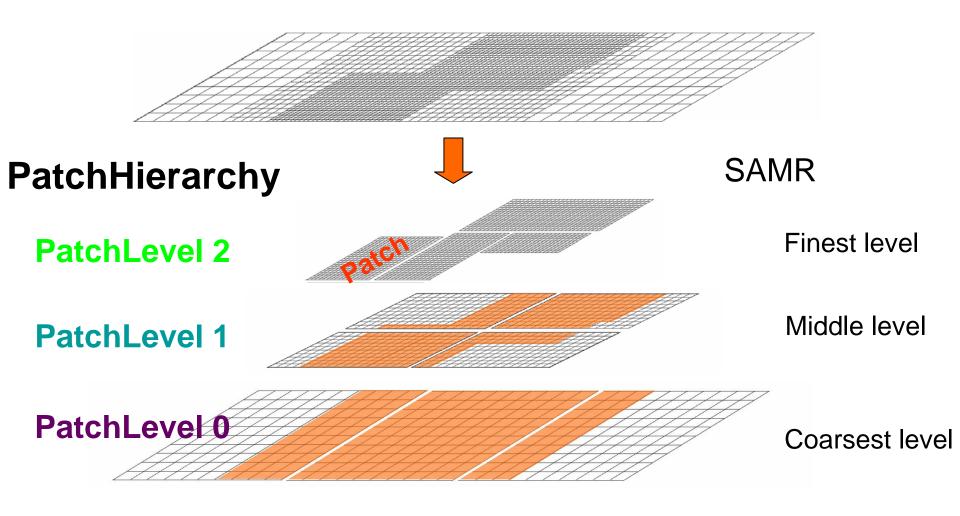
3.3.5 **SAMR**

h-adaptivity: Advance-Estimate-Tag-Refine-Synchronize

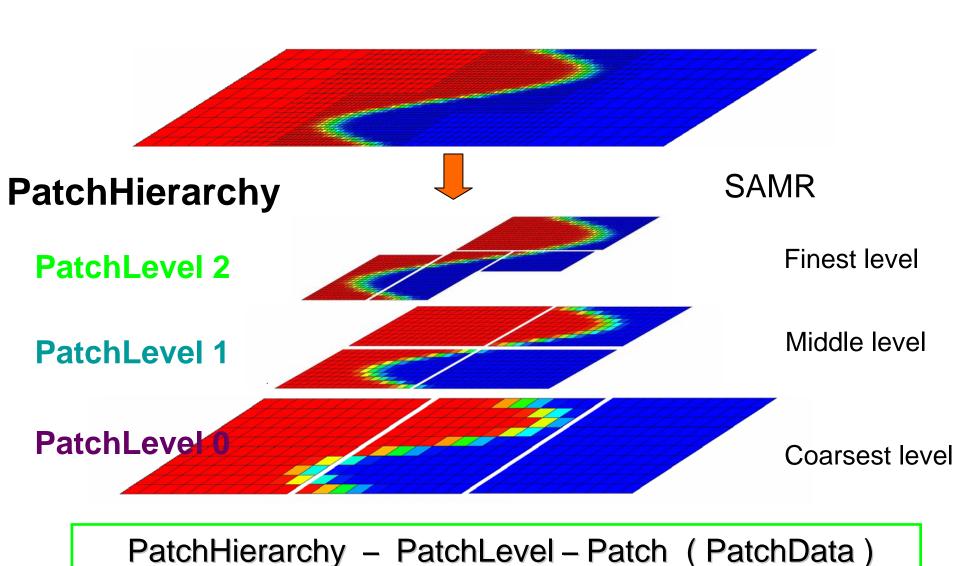




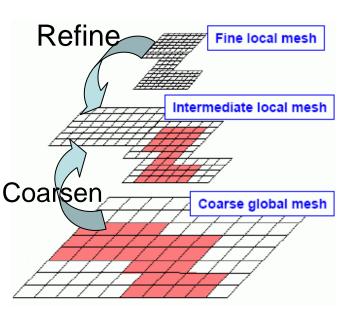
3.3.5 SAMR: data structure

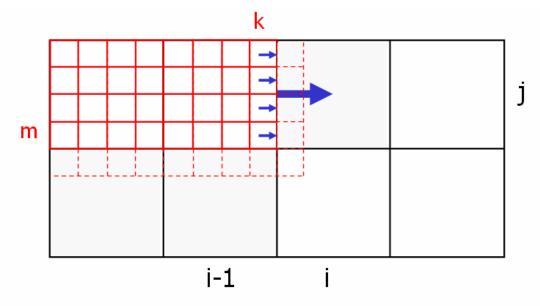


3.3.5 SAMR: data structure



3.3.5 SAMR: Communication



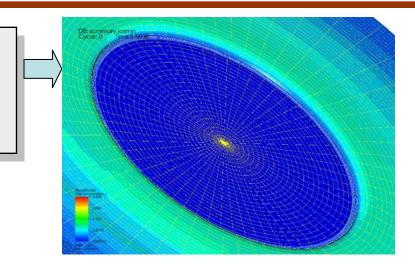


SAMR: using flux conservation interpolation schemes of high order precision

$$F_{i-\frac{1}{2},j}^{coarse} = \frac{1}{r^2} \sum_{q=0}^{r-1} \sum_{p=0}^{r-1} F_{k+\frac{1}{2},m+p}^{fine} \left(t + q \Delta t^{fine} \right)$$

3.3.5 **SAMR**

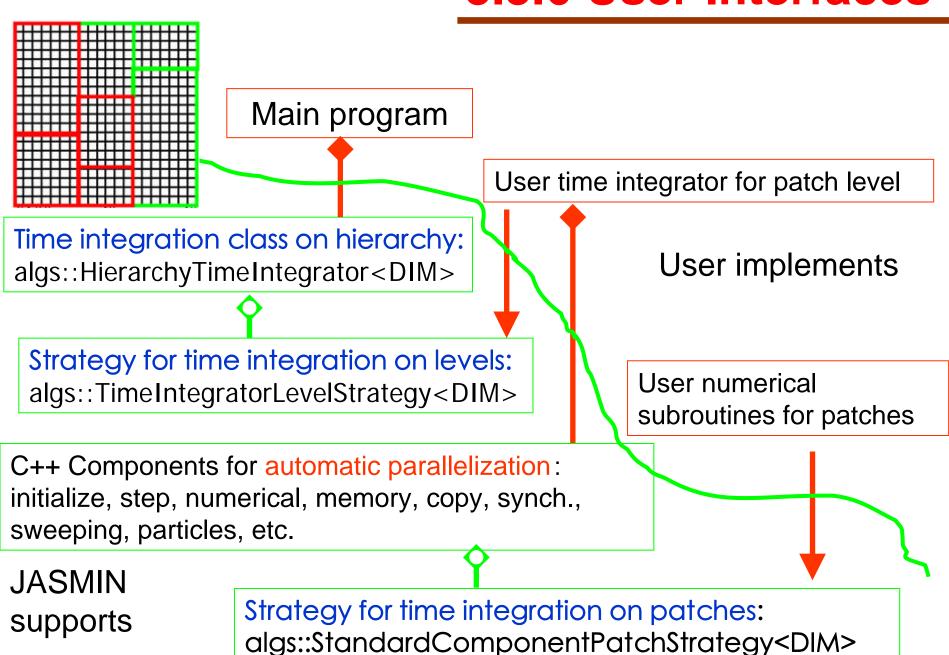
ICF 2-D radiation hydrodynamics simulation using LARED-S using three levels of SAMR meshes.



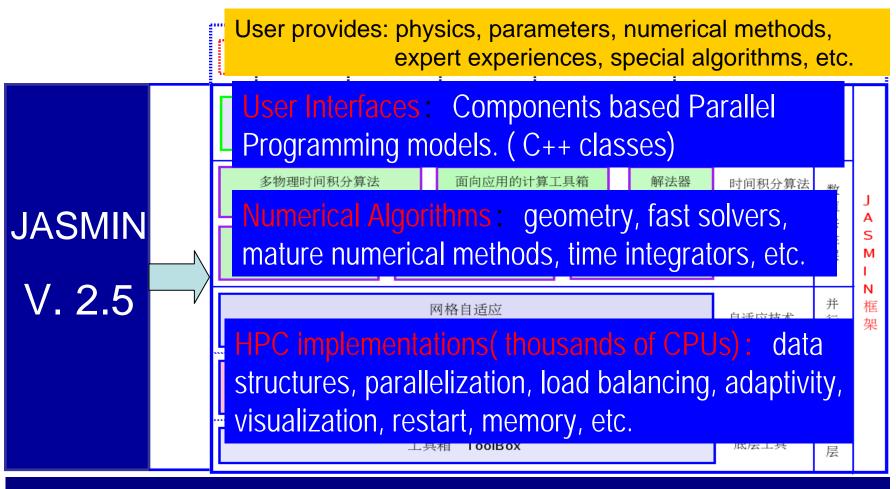
Speedup = 160.

Resolution of		# cells	#	# CPU	Time	
coarsest level			steps	cores	(hours)	
Resolution of finest level: 10240x2048. Physical time =1.324ns						
10240x2048	1	2097	1024 cores require 20 days			
640x128	3	38 ~ 124	28700	64	24.6	

3.3.6 User Interfaces



3.4 Current Version



Architecture: Multilayered, Modularized, Object-oriented;

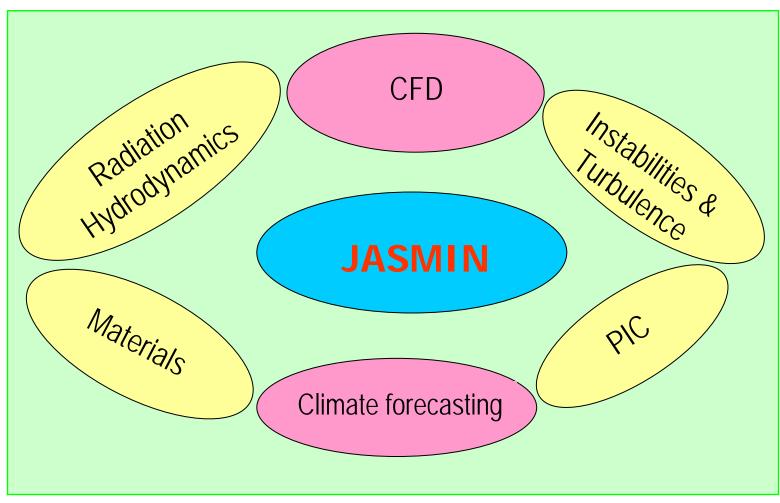
Codes: C++/C/F90/F77, MPI+MPI+MPI+OpenMP, 660K LOC;

Installation: Personal computers, Cluster, MPP.

3.4 Current Version

Applications

currently



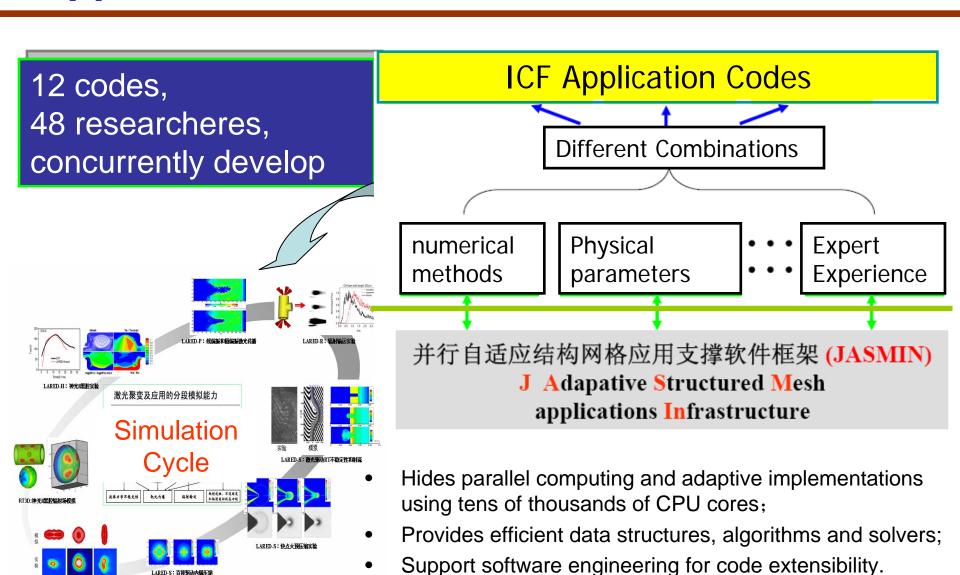
25 codes

3.5 Some Applications on PetaFlops System TianHe-1A

Codes	# CPU cores	Codes	# CPU cores
LARED-S	32,768	RH2D	1,024
LARED-P	72,000	HIME3D	3,600
LAP3D	16,384	PDD3D	4,096
MEPH3D	38,400	LARED-R	512
MD3D	80,000	LARED Integration	128
		RT3D	1,000
JMES-FDTD	60,000	NEPTUNE	1,024

Simulation times: several hours to tens of hours.

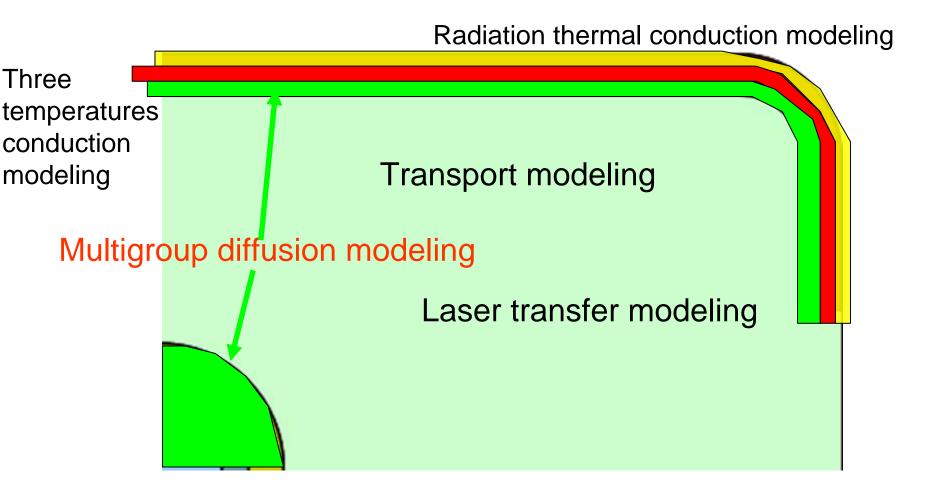
Applications-1: Inertial Confinement Fusion



Applications-1: Inertial Confinement Fusion

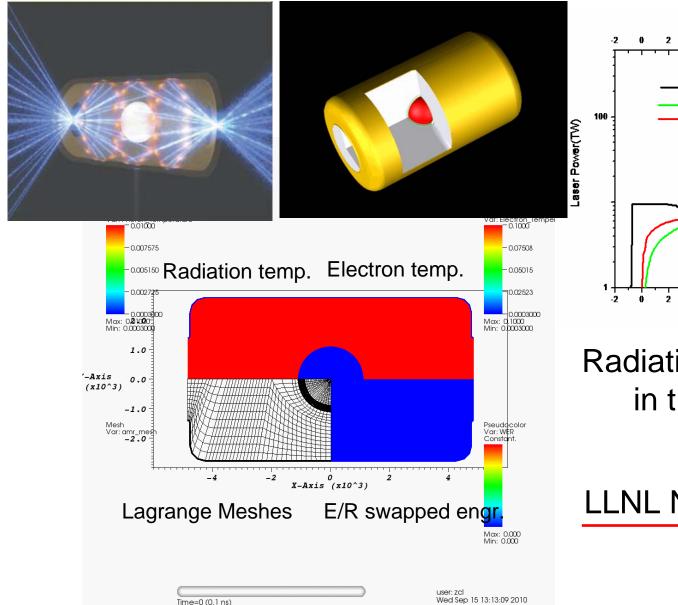
Codes	Year 2004	Year 2010				
LARED-H	serial	Parallel				
2-D radiation hydrodynamics Lagrange	Single bolck	Multiblock				
code	Without capsule	NIF ignition target				
LARED-R	Serial	Parallel (2048 cores)				
2-D radiation transport code						
Scale up a factor of 1000 ores)						
LARED-S	Single level	SAMR				
3-D radiation hydrodynamics Euler code	2-D: single group	Multi-group diffusion				
Try drody framinos Euror codo	3-D: no radiation	3-D: radiaiton multigroup diffusion				
LARED-P	serial	Parallel (36000 cores)				
3-D laser plasma interaction code		Terascale of particles				

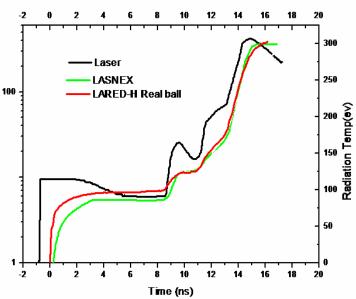
ICF-1: Integration simulations for ignition targets



Multi-physical modeling for radiation hydrodynamcs simulations

ICF-1: Integration simulations for ignition targets





Radiation temperatures in the hohlraum.

LLNL NIF base target

ICF-2: 3-D simulations for laser plasma interactions

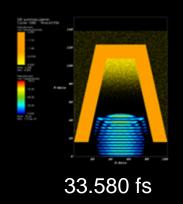
LARED-P: super-strong lasers transfer and focus over the plasma cone and generate high energy electrons.

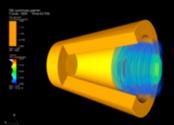
#mesh: 768M #partiles: 40G

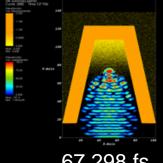
CPU Cores: 72,000 Para.Effi.: 44%

Phys. time: 320 fs Sim. time: 4.5 hours #steps: 9500

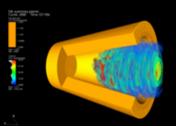
> Output data 640GB/20

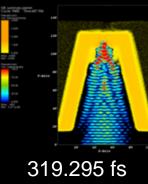


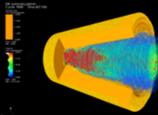




67.298 fs







ICF-3: 3-D simulations for laser plasma hydrodynamics

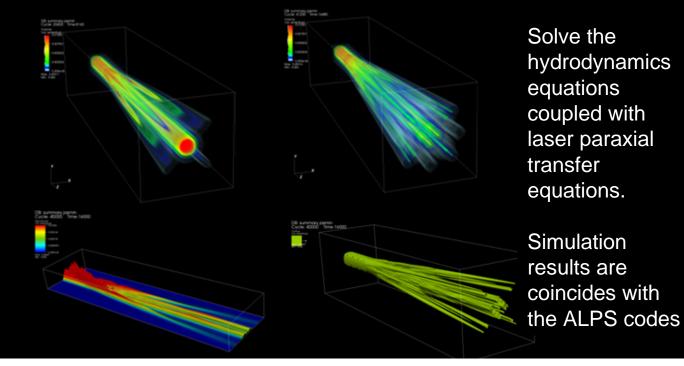
LAP3D: lasers filament and self-focus while transfer over a long distance in the lower density plasma environments.

#mesh: 2.1G

CPU Cores: 16,384 Para.Effi.: 50%

Phys. time: 56.5 ps Sim. time: 12.8 hours #steps: 41,200

> Output data 1.74TB/ 104 Parallel rendering using 72 cores



ICF-4: 3-D simulations for radiation hydrodynamics instabilities

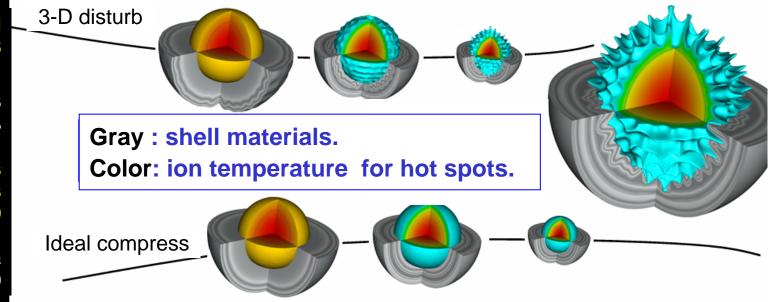
LARED-S: radiation hydrodynamics instabilities occur over the capsule interfaces while the capsule shell rapidly slows down.

#mesh: 160M 256x256x256

Cores: 32,768 Para.Effi.: 52%

P. time: 0.1 ps S. time: 39 hours #steps: 778,580

Output data 4.4TB/ 670



Applications-2: Material simulations

PDD3D: 3-D discrete dislocation simulations for the locally plastic deformations of metal materials while the shock is enforced.

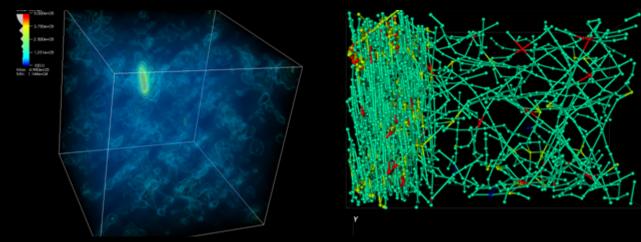
#dislocation: 3 M

CPU Cores: 4096 Para.Effi.: 47%

Phys. time: 5.5 ns Sim. time: 12 hours #steps: 1,800

> Output data 196 GB/ 196

The details of dislocation structures are discovered.



The stretch simulations of Cu crystal (0.12 mm³, r=10⁷/s). Left: dislocation density; Right: local structures.

Applications-2: Material simulations

MD3D: 3-D molecule dynamics simulations for the structures and dynamics shock responses of metal materials with nano-scale defects.

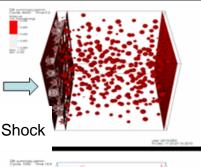
Dislocation holes release and collapse.

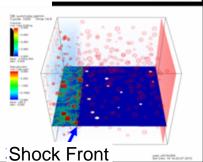
#Molecules: 50G

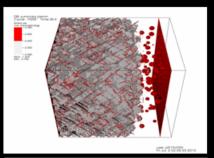
#Cores: 80,000 Para.Effi.: 62%

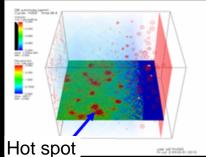
Ph.time: 5.5 ps Sim. time: 3.0 hours #steps: 30,000

> Output data 162 GB/ 150

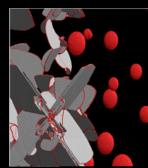












Discation holes collapse and interact with each other, hot spots are formed to generate various dynamics responses.

Left: local spots; Right: LLNL's results in 2005.

Applications-3: Electromagnetic Simulations

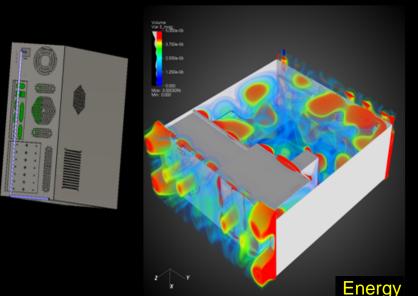
JMES3D: 3-D FDTD simulations for destroy mechanism of electromagnetic waves with different frequencies and directions.

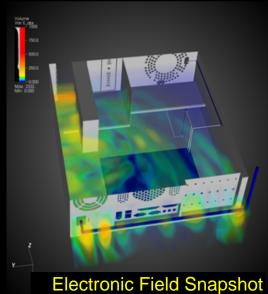
mesh: 1.2 G

#Cores: 60,000 Para.Effi.: 70%

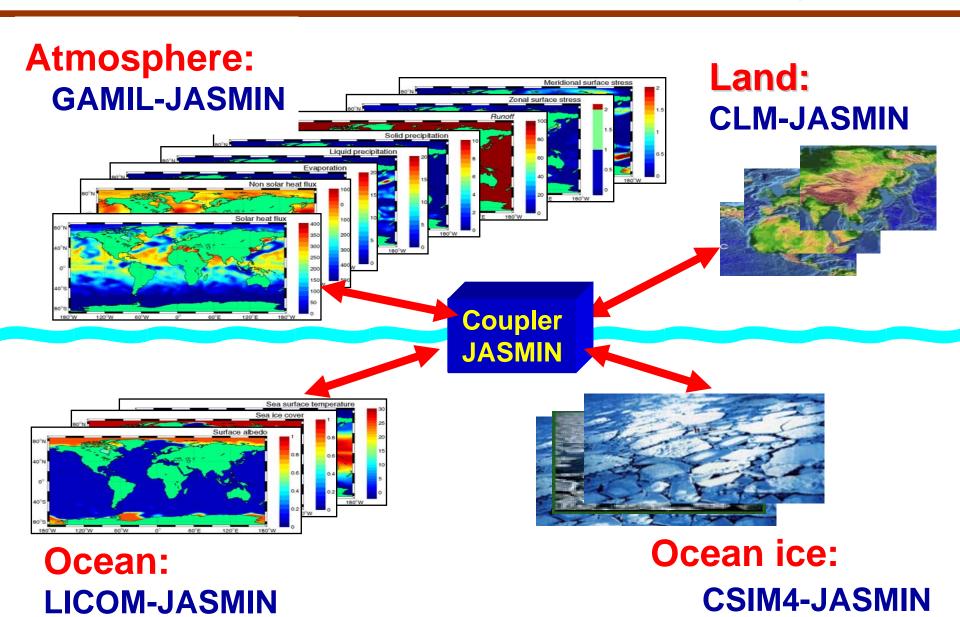
Phy.ime: 250 ns S.time: 8.1 hours #steps: 254,000

Output data 736 GB/ 172

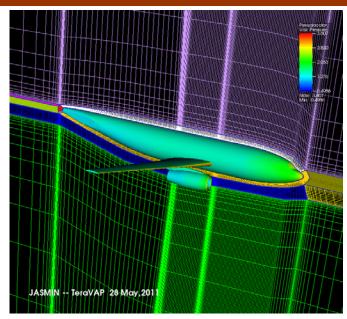




Applications-4: Climate Forecasting

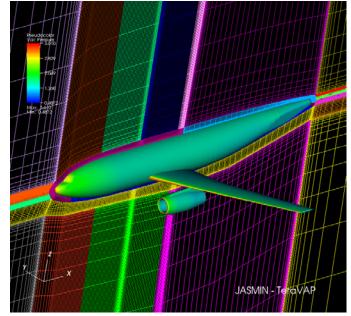


Applications-5: CFD



mesh blocks distribution

DDM 64 cores



cells: 9,574,784 # blocks: 194 # cores: 2048 # s/10K steps 800 # speedup 580

4. Conclusion

- Domain specific programming models are possible to enable domain scientists "think parallel ,write sequential".
- A possible stack of programming model for scientific computing is "Framework-based DSPM —MPI—DSM-X— OpenMP—ILP".
- Numerical fast algorithms are essential components for the implementation of DSPM.
- JASMIN shows the possibility of DSMP on structured meshes.