

# Lawrence Livermore National Laboratory

## Computationally Efficient Modeling of High-Efficiency Clean Combustion Engines

Daniel Flowers (PI), Salvador Aceves, Nicholas Killingsworth,  
Matthew McNenly, Thomas Piggott, Mark Havstad, Russell Whitesides,  
Randy Hessel (U Wisc), J.Y. Chen (UCB)



Project ID # ACE012

2011 DOE Hydrogen Program and Vehicle Technologies Program Annual Merit  
Review and Peer Evaluation Meeting

May 10, 2011 - Washington, DC

This presentation does not contain any proprietary or confidential information

This work performed under the auspices of the U.S. Department of Energy by  
Lawrence Livermore National Laboratory under Contract DE-AC52-07NA27344

# Overview

## Timeline

- Ongoing project with yearly direction from DOE

## Budget

- FY09 funding: \$1M
- FY10 funding: \$1M

### *Split among 3 projects:*

- *Combustion Numerics*
- *HECC Simulation*
- *Exhaust Speciation*

## Barriers

- Inadequate understanding of the fundamentals of HECC
- Inadequate understanding of the fundamentals of mixed mode operation
- Computational expense of HECC simulations

## Partners

- Sandia, Oak Ridge, Los Alamos
- Ford
- UC Berkeley, University of Wisconsin, University of Michigan, Lund Institute of Technology, Chalmers University, Tianjin University
- FACE working group, AEC MOU, SAE

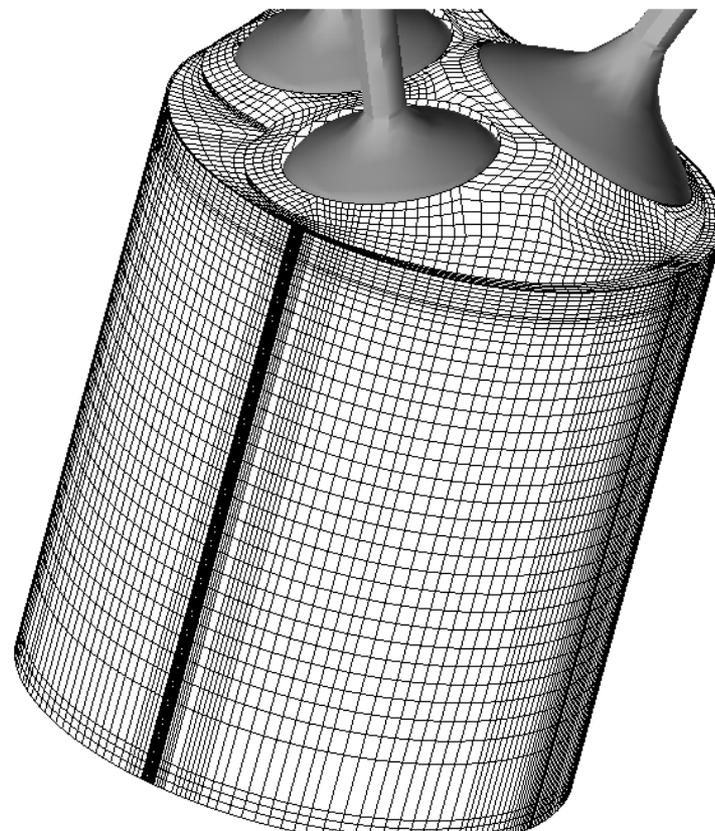
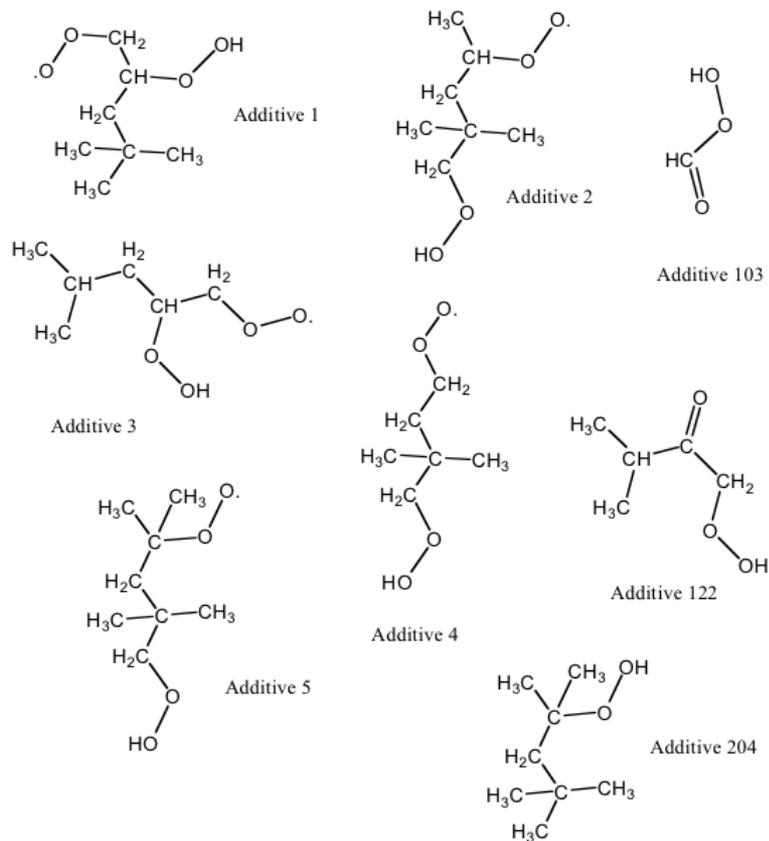


# Relevance to DOE objectives

- By 2015, improve the fuel economy of light-duty gasoline vehicles by 25 percent and of light-duty diesel vehicles by 40 percent, compared to the baseline 2009 gasoline vehicle.
  - **Light-duty research focuses on reducing fuel consumption through investigating HCCI and PCCI part load, and transition to SI or CIDI for full load operation**
- By 2015, improve heavy truck efficiency to 50 percent with demonstration in commercial vehicle platforms. This represents about a 20 percent improvement over current engine efficiency.
  - **Heavy-engine research directed towards high efficiency strategies, such as Partially Premixed Combustion and Low-temperature Diesel Combustion**
- By 2018, further increase the thermal efficiency of a heavy truck engine to 55 percent which represents about a 30 percent improvement over current engines.
  - **We continue to provide the engine research community with insight and simulation tools for advanced combustion concepts**



# Objective: Enhance understanding of clean and efficient engine operation through detailed numerical modeling

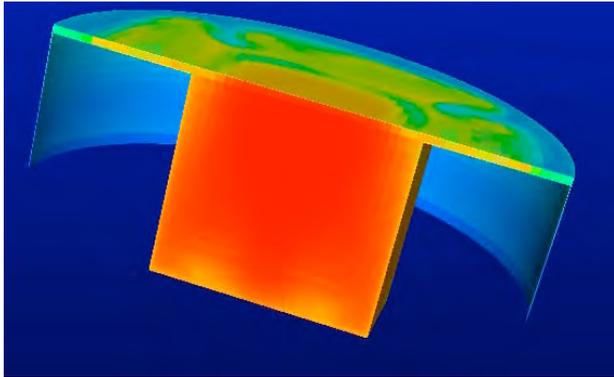


## Chemical kinetics

## Fluid mechanics

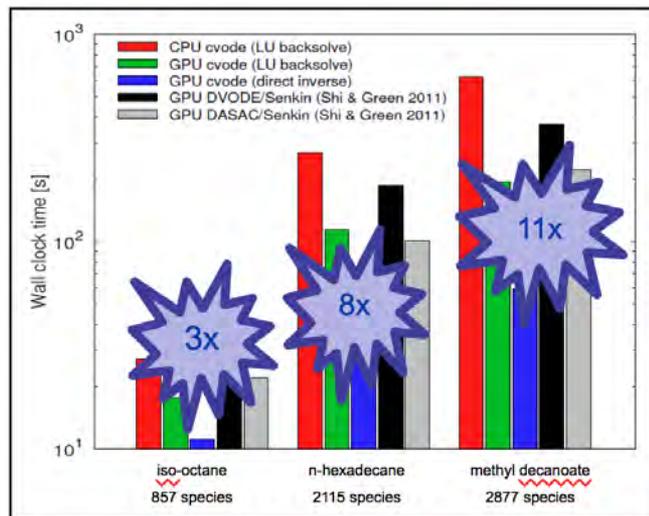


# Milestones: We have developed and experimentally validated detailed engine modeling tools



- *11x speedup demonstrated for GPU zero-dimensional ignition delay calculations (Feb. 2011)*

- *Developed and Deployed Parallel Multi-zone Combustion Model in Kiva4-mpi (Dec. 2010)*



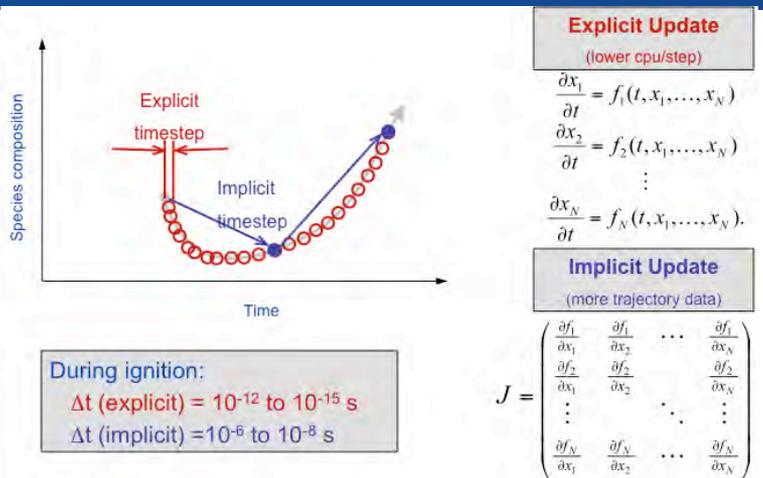
- *Completed License agreement for Multi-zone Combustion Model with Convergent Science Inc. (Oct. 2010)*
- *Up to 300x speedup demonstrated for GPU based thermodynamic property calculations (Aug. 2010)*

# Approach: Collaborate with industry, academia and national labs in the development of analysis tools leading to clean, efficient engines

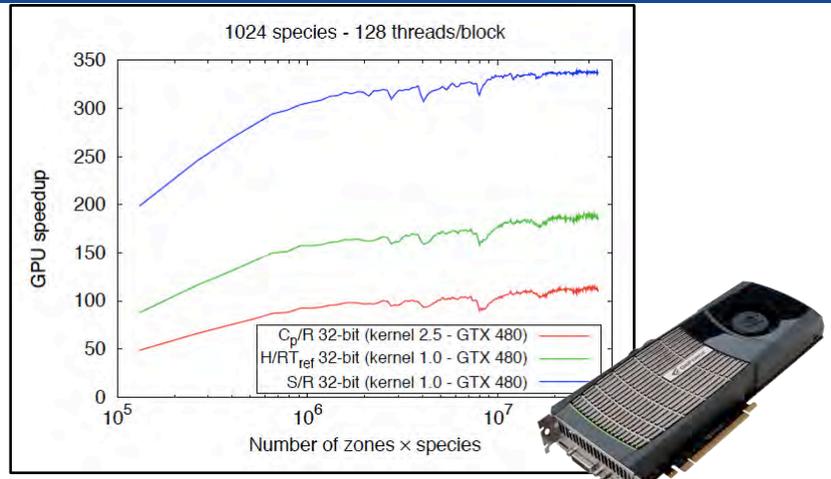
- Gain fundamental and practical insight into HECC regimes through numerical simulations and experiments
- Develop and apply numerical tools to simulate HECC by combining multidimensional fluid mechanics with chemical kinetics
- Reduce computational expense for HECC simulations
- Democratize simulation: bring computational tools to the Desktop PC



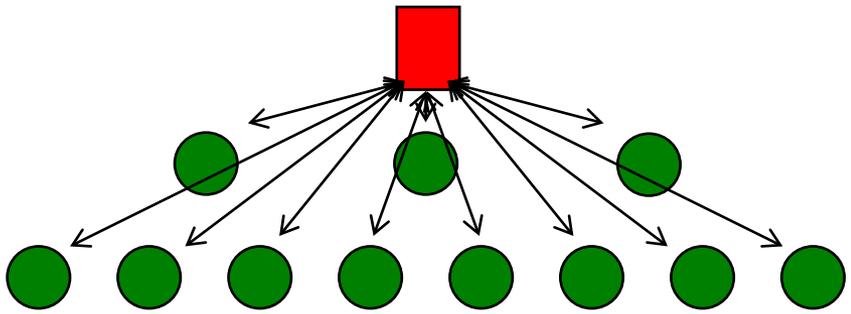
# Technical Accomplishments: We have made significant progress in improving and applying our advanced simulation tools to HECC



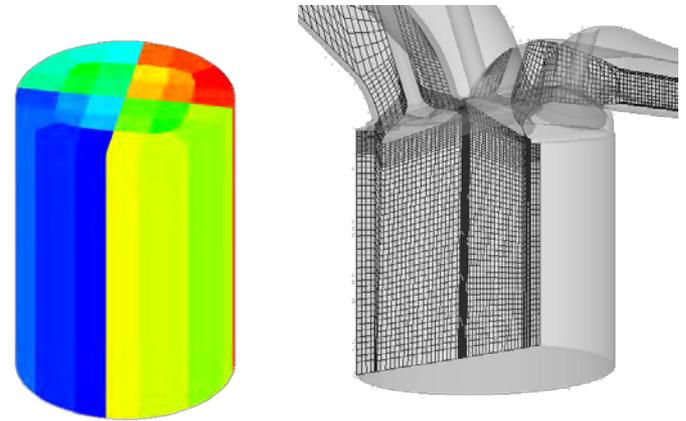
## Improved Numerics



## New Computing Architectures



## KIVA4-mpi multi-zone development

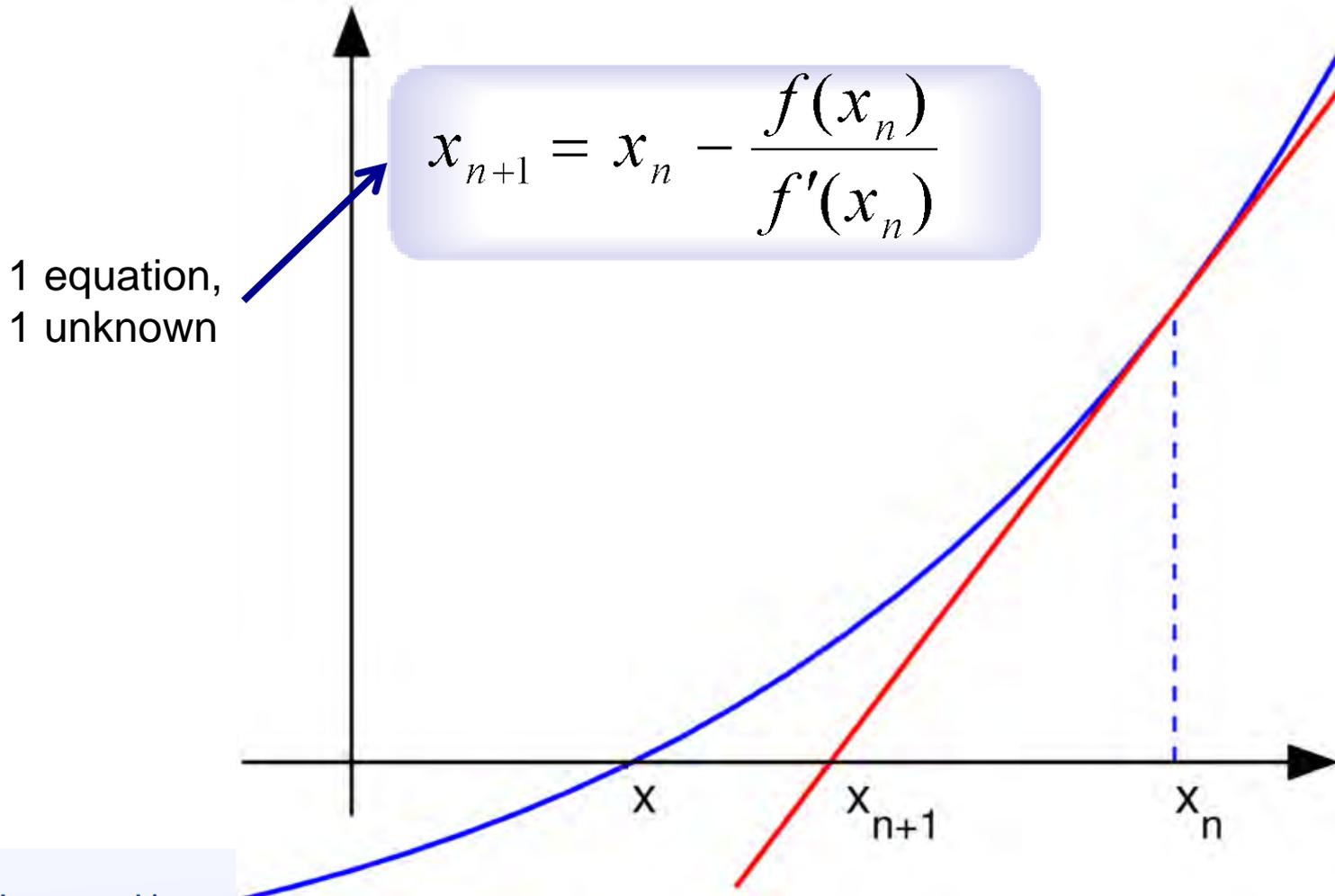


## Parallel CFD and MZ-chemistry



# Chemical kinetics involves solving systems of non-linear differential equations

The Newton-Raphson method efficiently solves nonlinear equations



When solving a system of differential equations, the **Jacobian matrix**  $J = \frac{\partial f_i}{\partial y_j}$  plays the role of the derivative

N equations,  
N unknowns

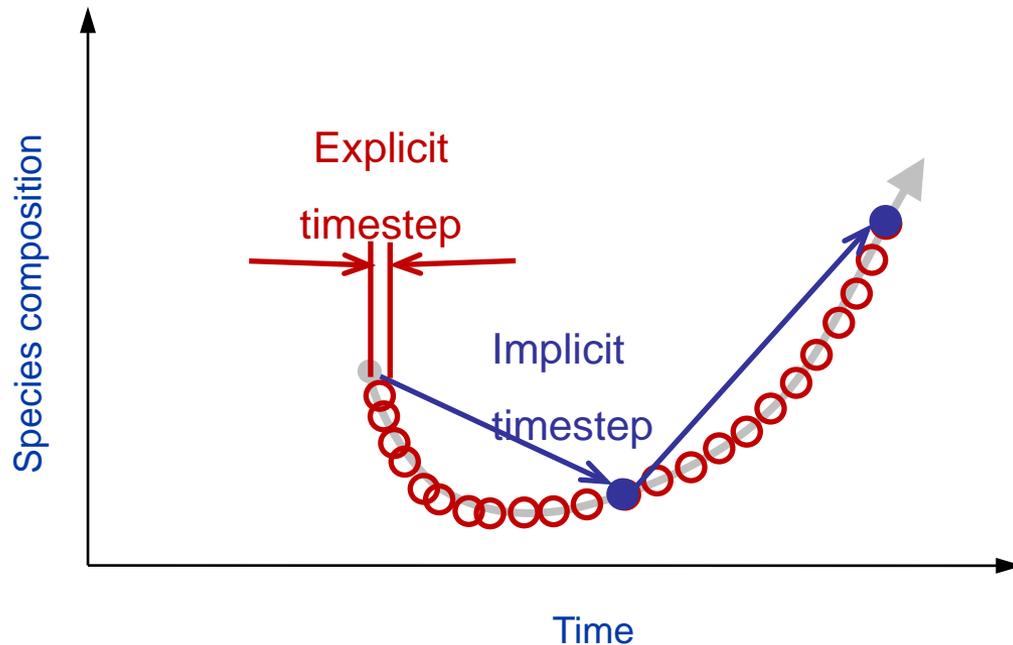
$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

$$\left( I - \Delta t \frac{\partial f_i}{\partial y_j} \right) (\hat{y}_j^{k+1} - \hat{y}_j^k) = -\hat{y}_i^k + y_i + \Delta t f_i(\hat{y}_1^k, \dots, \hat{y}_N^k)$$

$$J = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_N} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & & \frac{\partial f_2}{\partial x_N} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_N}{\partial x_1} & \frac{\partial f_N}{\partial x_2} & \dots & \frac{\partial f_N}{\partial x_N} \end{pmatrix}$$

The Jacobian is the key player in solving chemical kinetics ODEs

# >95% of the chemistry solution CPU cost is spent constructing and solving the Jacobian system



## Explicit Update

(lower cpu/step)

$$\begin{aligned} \frac{\partial x_1}{\partial t} &= f_1(t, x_1, \dots, x_N) \\ \frac{\partial x_2}{\partial t} &= f_2(t, x_1, \dots, x_N) \\ &\vdots \\ \frac{\partial x_N}{\partial t} &= f_N(t, x_1, \dots, x_N). \end{aligned}$$

## Implicit Update

(more trajectory data)

$$J = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_N} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & & \frac{\partial f_2}{\partial x_N} \\ \vdots & & \ddots & \vdots \\ \frac{\partial f_N}{\partial x_1} & \frac{\partial f_N}{\partial x_2} & \dots & \frac{\partial f_N}{\partial x_N} \end{pmatrix}$$

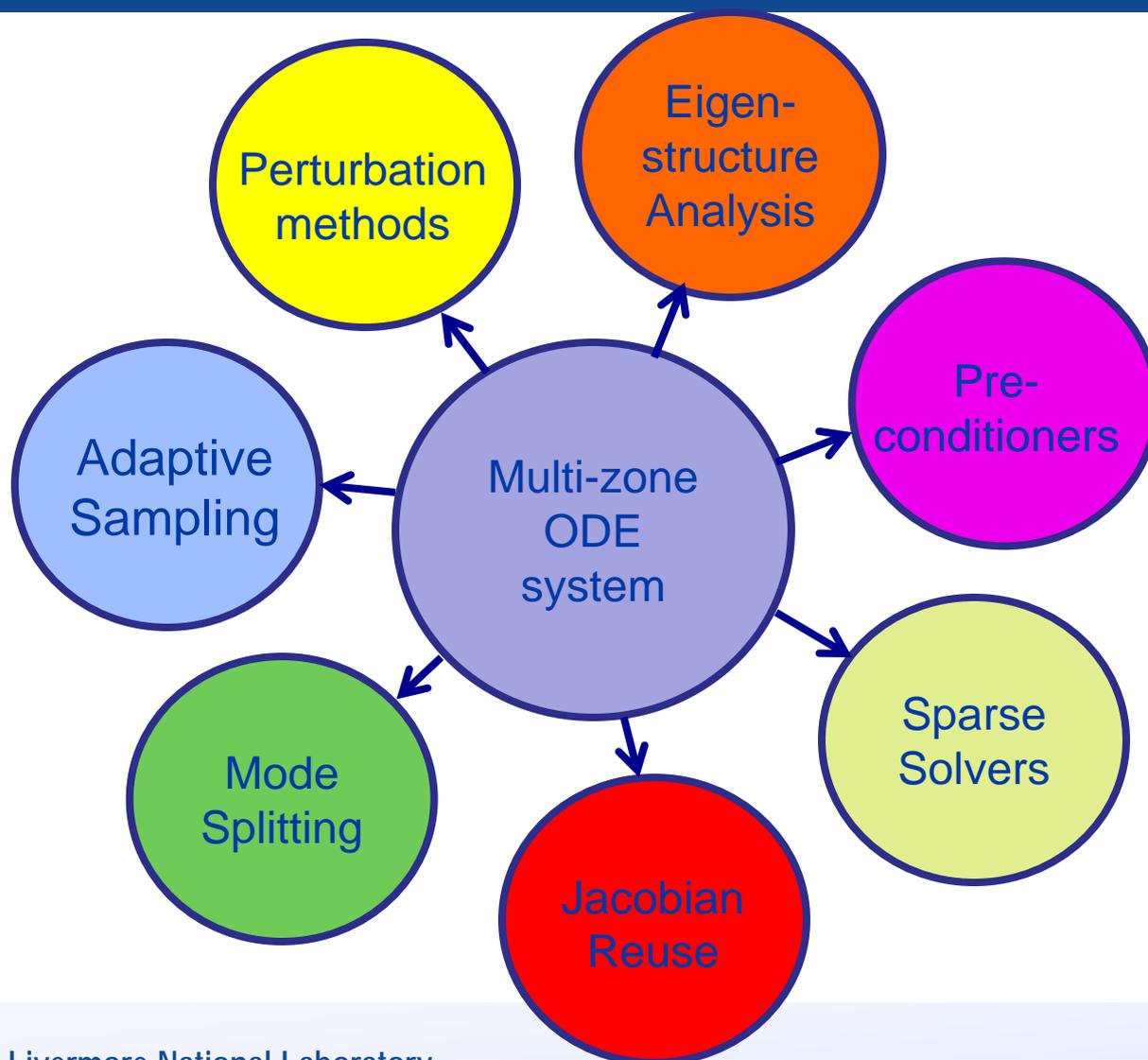
During ignition:

$$\Delta t \text{ (explicit)} = 10^{-12} \text{ to } 10^{-15} \text{ s}$$

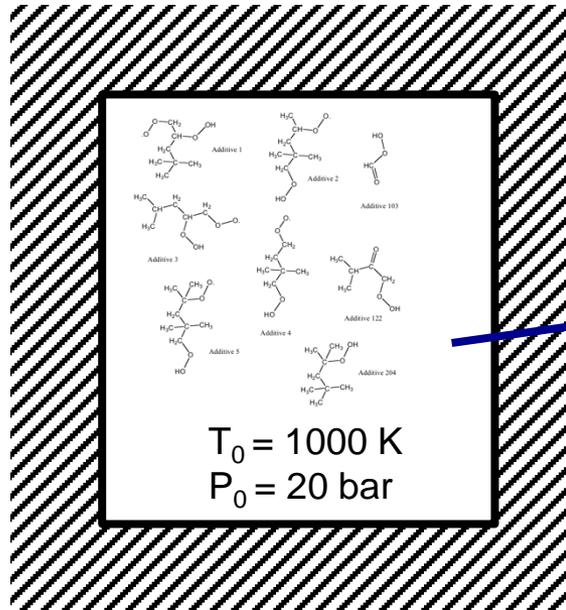
$$\Delta t \text{ (implicit)} = 10^{-6} \text{ to } 10^{-8} \text{ s}$$



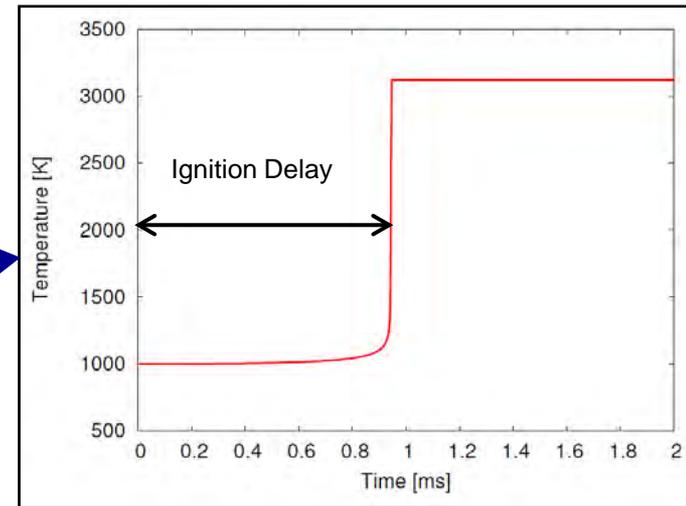
# Opportunities for 1000x speedup in computational chemistry cost through applied mathematics



# Constant volume ignition delay is the basis for numerical chemistry development



**Constant Volume Reactor**



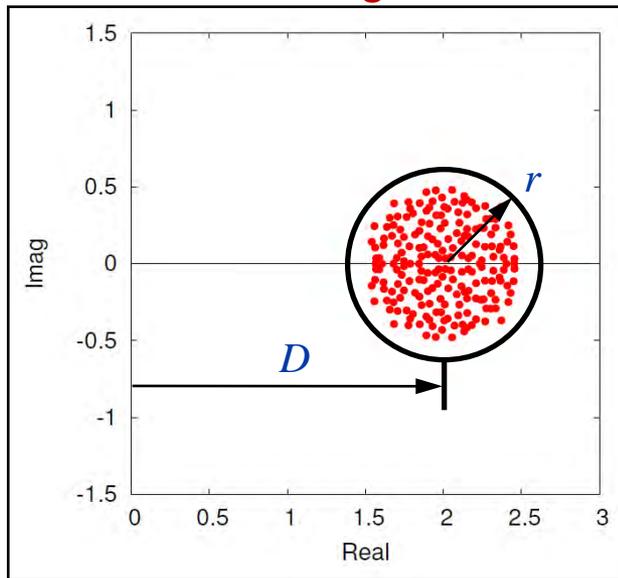
**Temperature Time History**

The Constant Volume Reactor is the basic unit for chemistry in multidimensional CFD codes

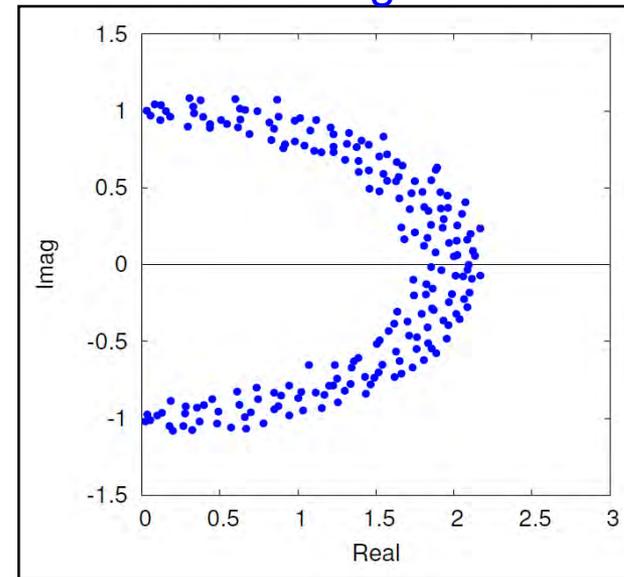
# Applied mathematics techniques identify opportunities for improved solver convergence

Eigenvalues of the Jacobian determine how quickly an ODE solution will converge

Fast convergence



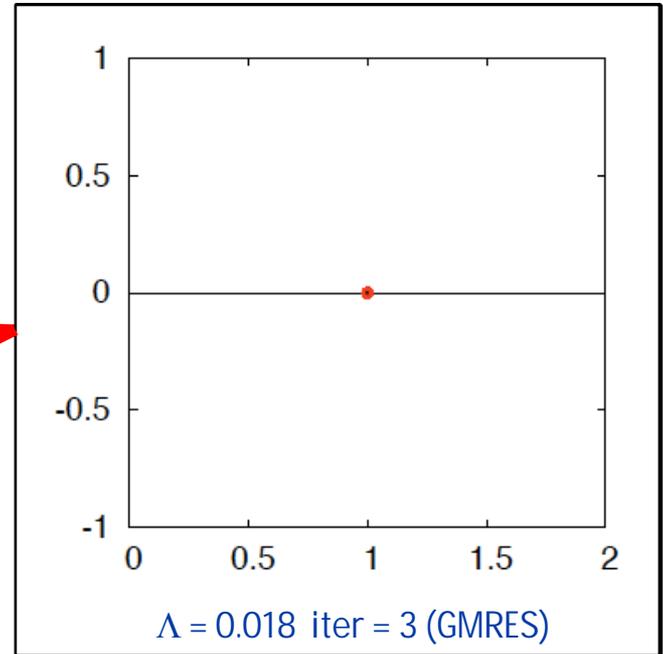
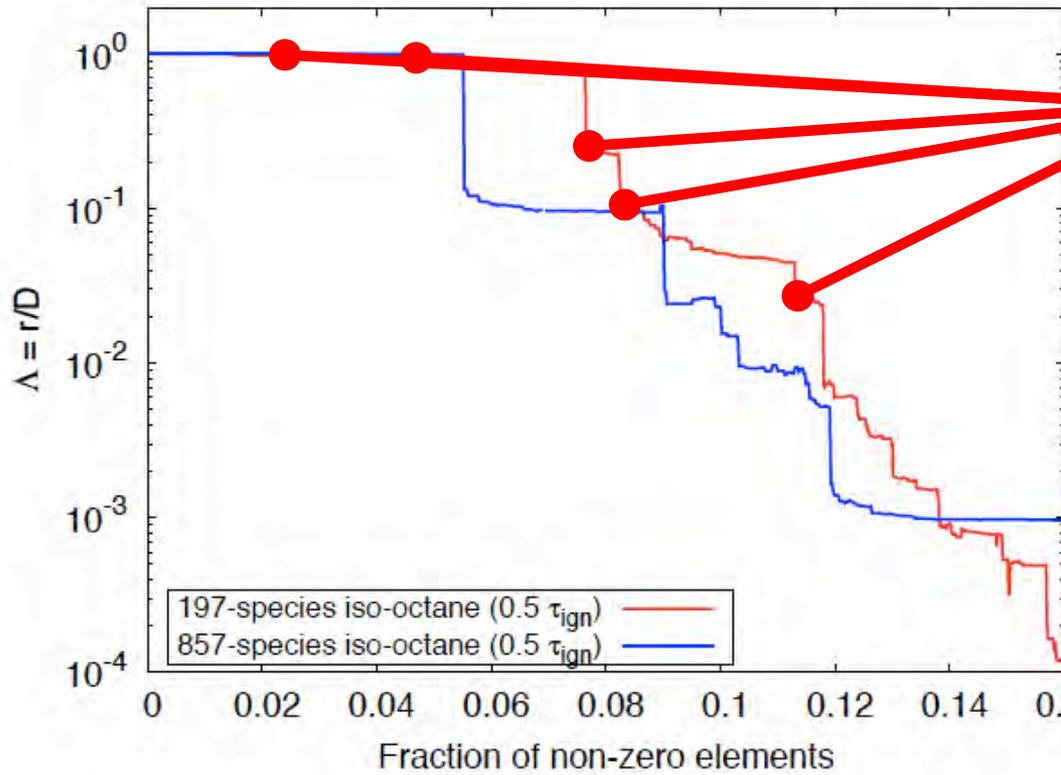
Slow convergence



ODEs with tightly clustered eigenvalues far from the origin converge faster.

# Adjusting the Jacobian (“preconditioning”) with low-cost operations enables rapid solver convergence

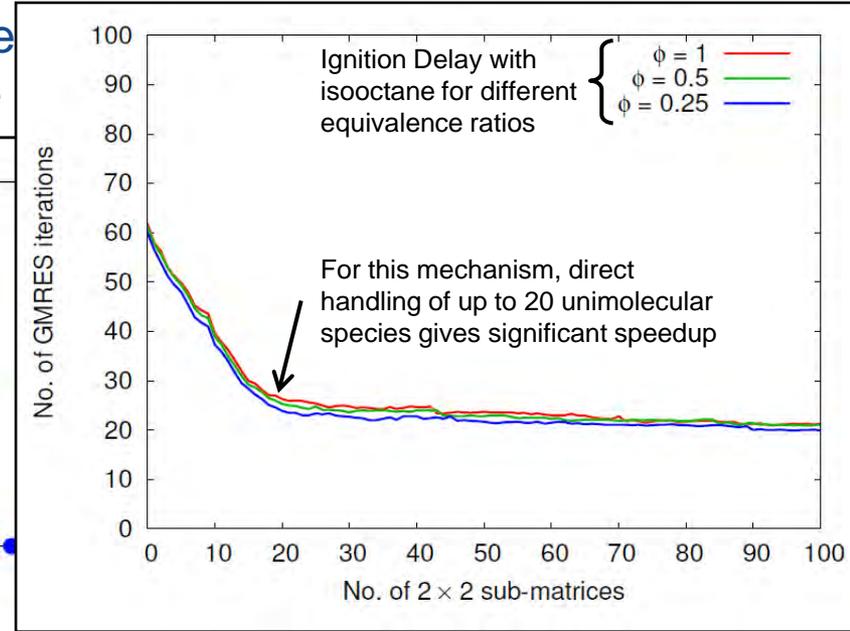
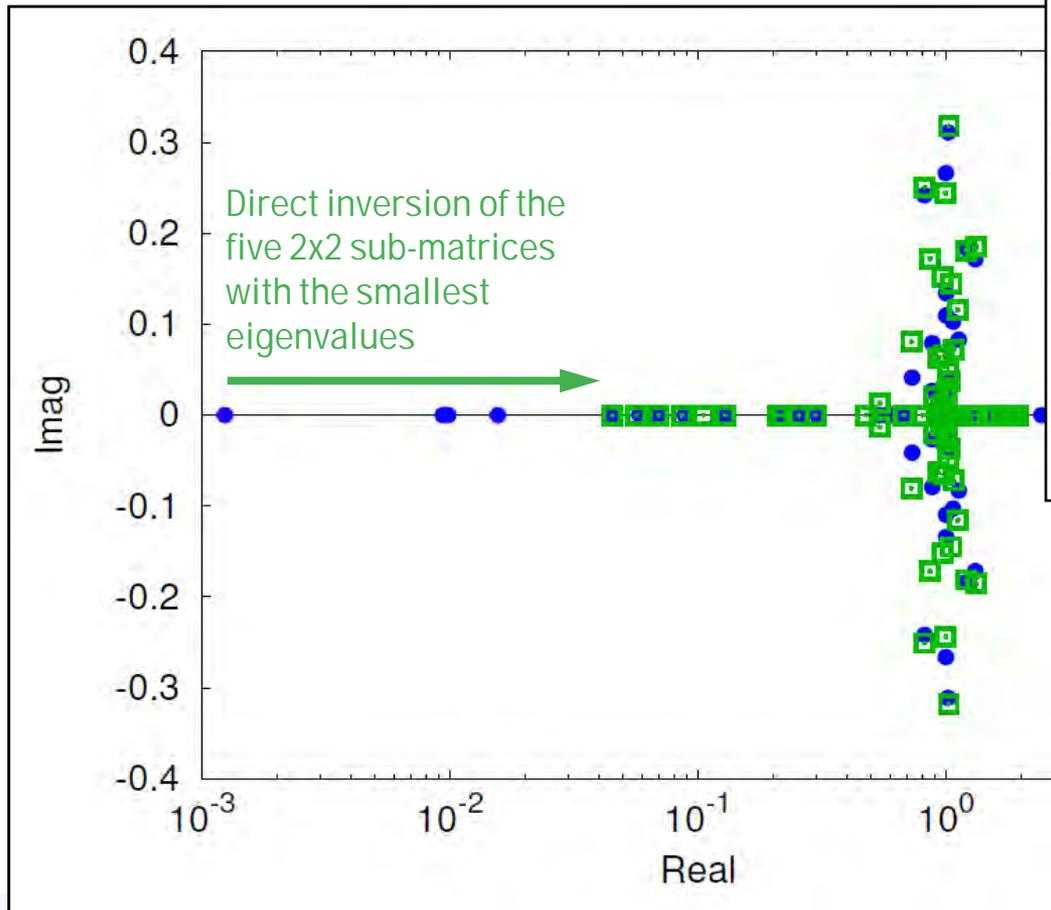
Heuristic 1: only include the off-diagonal Jacobian terms corresponding to the fastest reaction rates



Rapid convergence achieved with less than 10% of the Jacobian information.

# Direct handling of certain unimolecular reactions gives much faster convergence

Unimolecular reactions can produce eigenvalue near zero (after scaling) that slow convergence

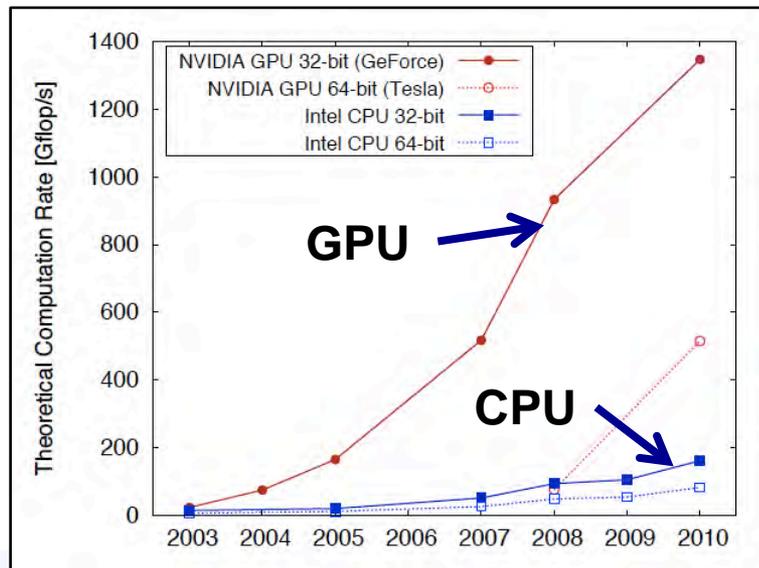


Number of iterations are reduced by a factor of 3 with less than 1% of the Jacobian information.

# Graphical Processing Units (GPUs) can bring supercomputing to the desktop workstation



Nvidia GeForce 480



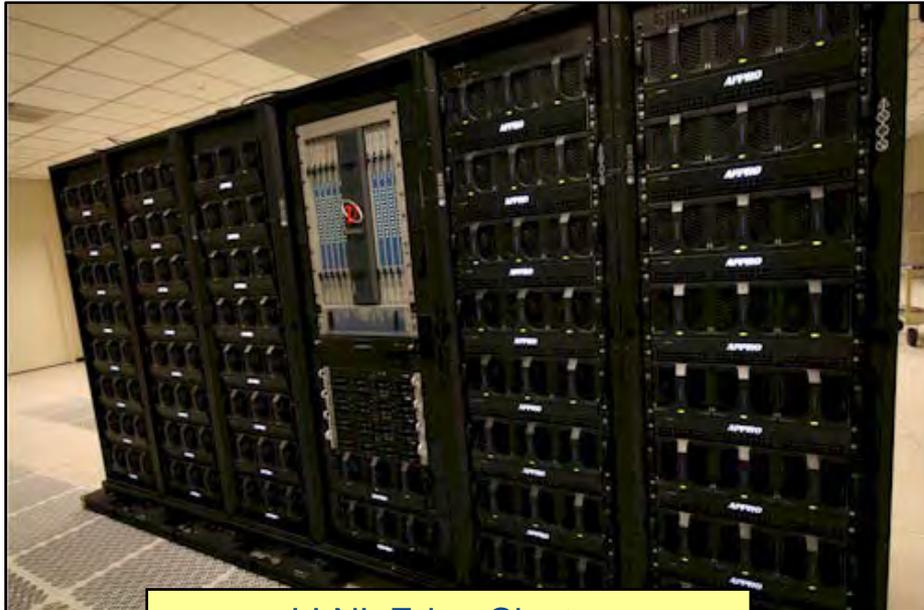
Data from NVIDIA's, "CUDA C Programming Guide Version 3.1," 2010.

Lawrence Livermore National Laboratory

- ½ Teraflop for \$500
- 480 parallel processors
- Codes must be redesigned to take advantage of architecture
- Massively parallel computing on the desktop
- Fortran/C++ Compilers designed for GPUs now available



# LLNL is a center for research on using GPU architectures for large-scale scientific simulations



LLNL Edge Cluster

GPUs:	412	
Type:	Tesla M2050	
Cores:	185,000	GPU
	2,500	CPU
Tflop/s:	212 (64-bit)	
Price:	4 - 5 M\$	

**AMAX GPU WORKSTATIONS**



Tesla 20-Series Ready

Off-the-shelf Desktop

GPUs:	1	
Type:	Tesla C2050	
Cores:	448	GPU
	8	CPU
Tflop/s:	0.5 (64-bit)	
Price:	5.7 K\$	

# Getting the most out of the GPU involves designing suitable algorithms

- Thermodynamic property evaluation illustrates algorithmic design
- Specific heat
  - Two polynomials: Low and high temperature

$$C_p = \begin{cases} a_1 + a_2T + a_3T^2 + a_4T^3 + a_5T^4, T < 1000 K \\ b_1 + b_2T + b_3T^2 + b_4T^3 + b_5T^4, T > 1000 K \end{cases}$$



# Thermodynamic property evaluation illustrates GPU algorithm design

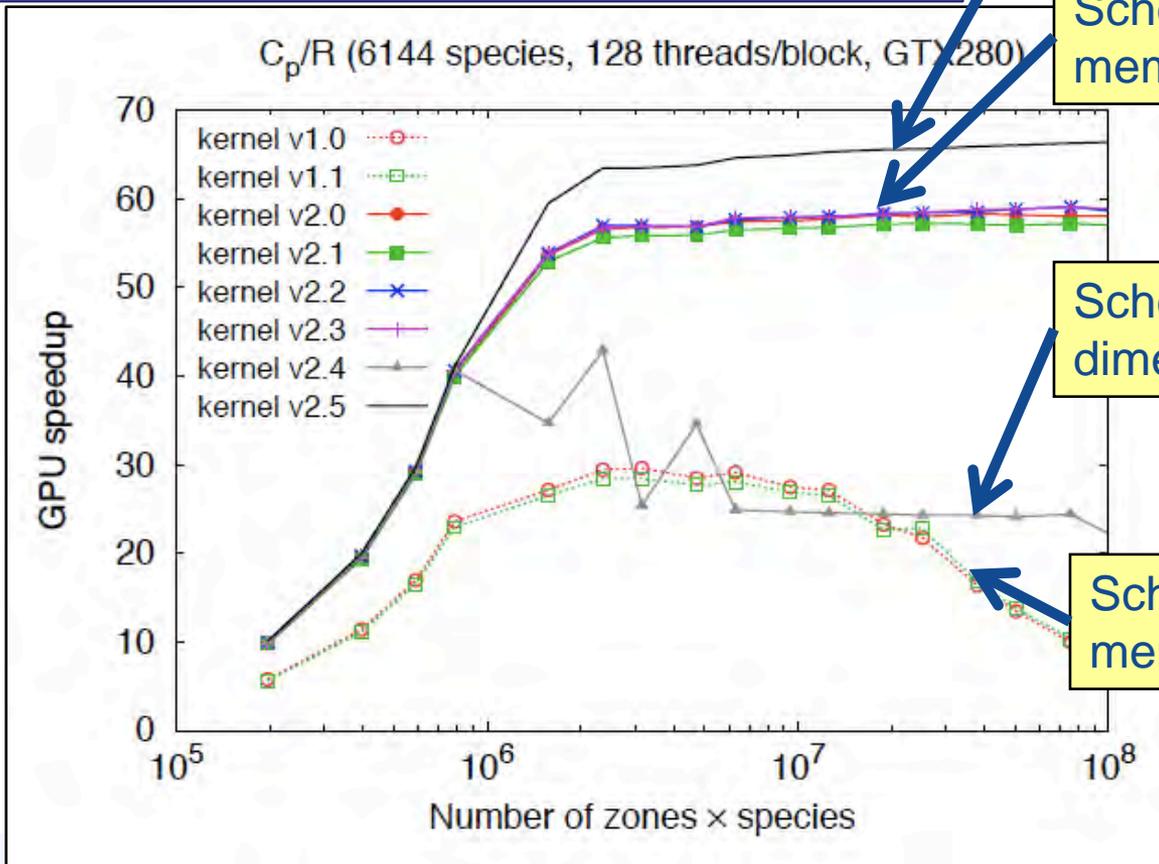
$$C_p = \begin{cases} a_1 + a_2T + a_3T^2 + a_4T^3 + a_5T^4, & T < 1000K \\ b_1 + b_2T + b_3T^2 + b_4T^3 + b_5T^4, & T > 1000K \end{cases}$$

Scheme v2.5 (shared memory, both high and low T calculated)

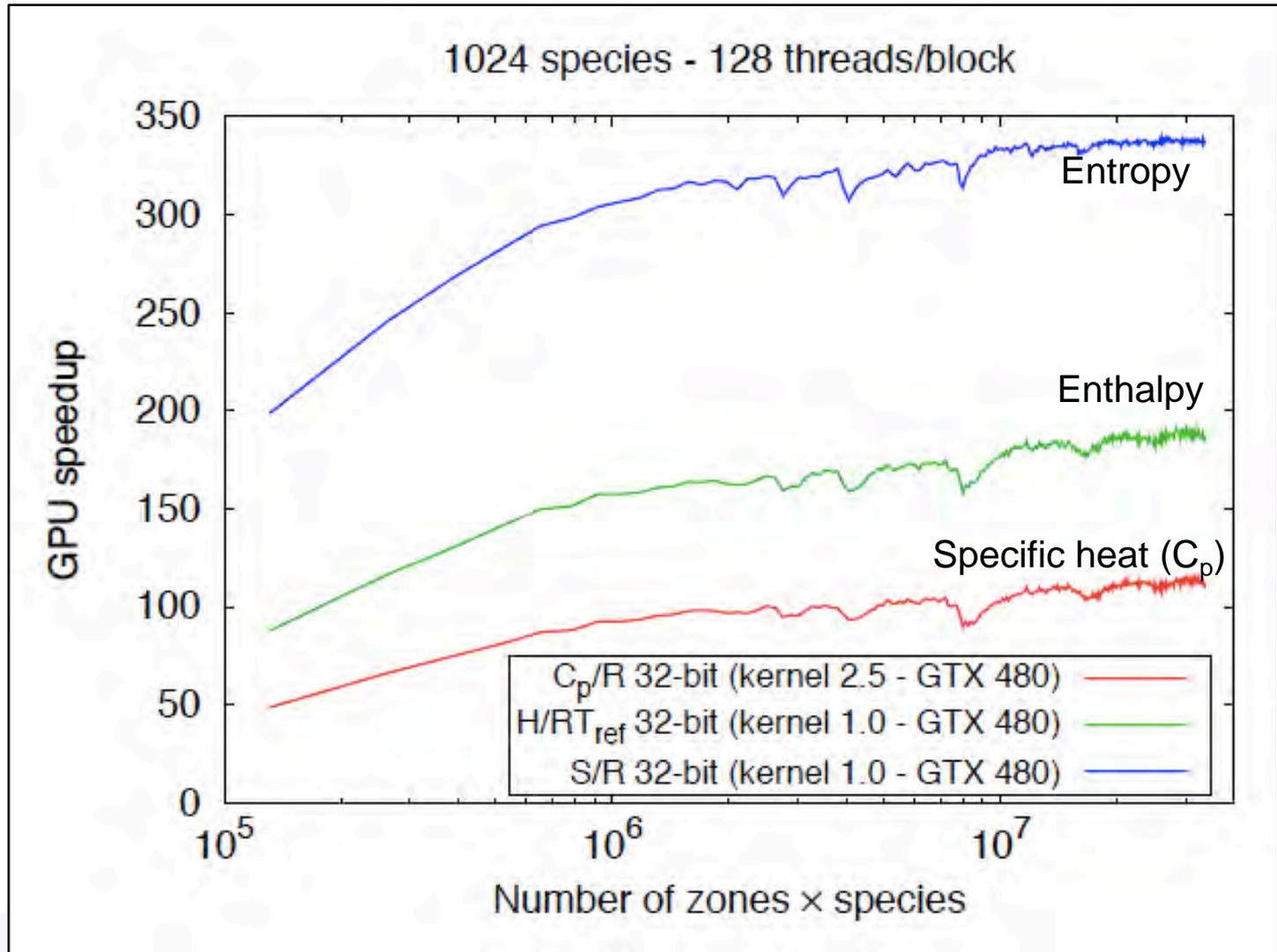
Schemes v2.0 – v2.3 (shared memory)

Scheme v2.4 (switched dimensions)

Schemes v1.0, v1.1 (no shared memory, brute force)

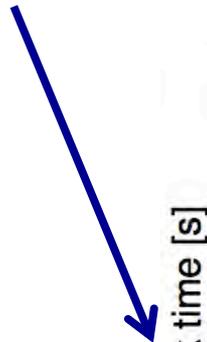


# Enthalpy and entropy calculations have greater speedup with the GPU; GPUs perform best with more arithmetic operations per memory access



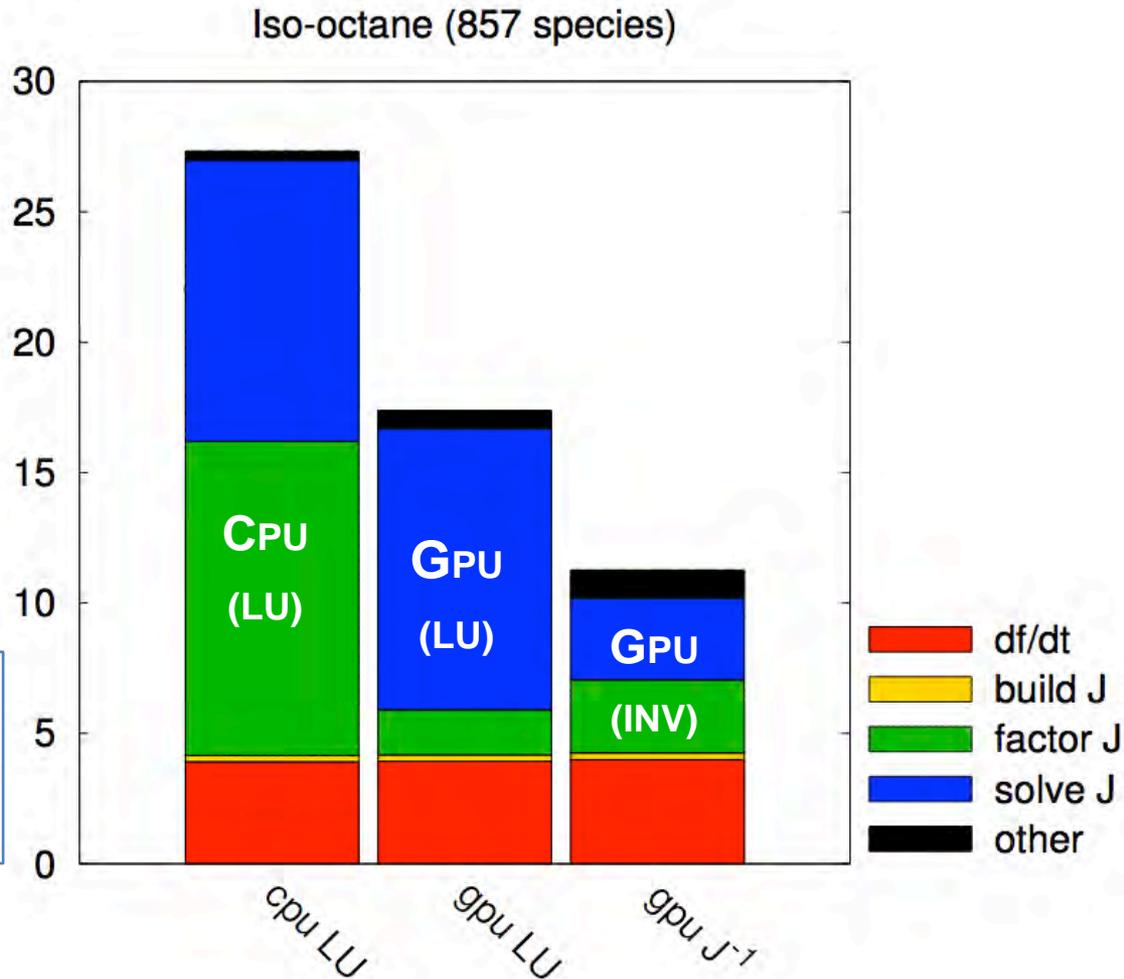
# We are doing ignition delay calculations with large mechanisms to determine best practices for GPUs

Simulation Time

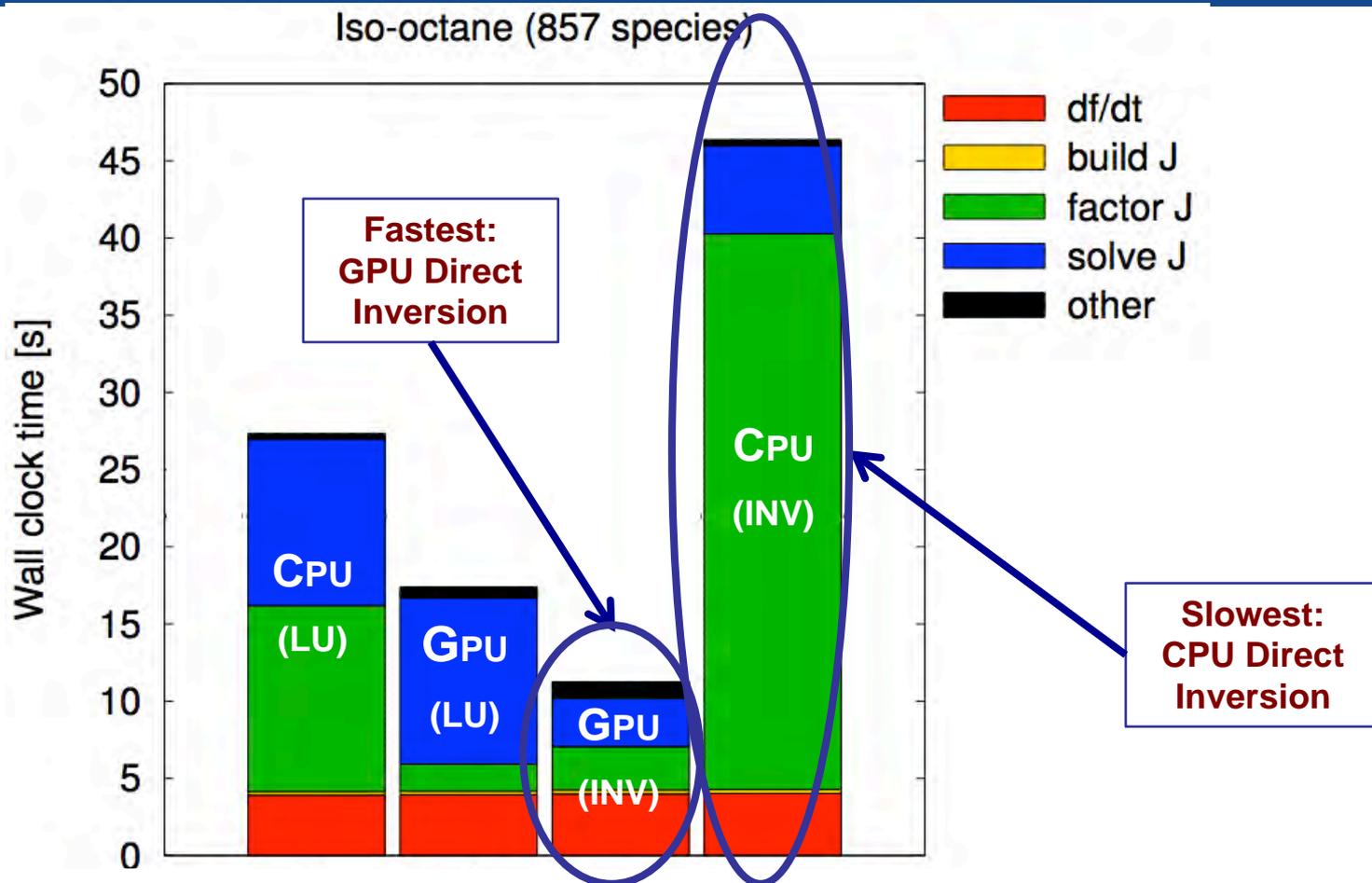


Wall clock time [s]

Shorter bars mean faster simulation



# Direct matrix inversion is the most effective solver strategy on the GPU, but performs poorly on the CPU

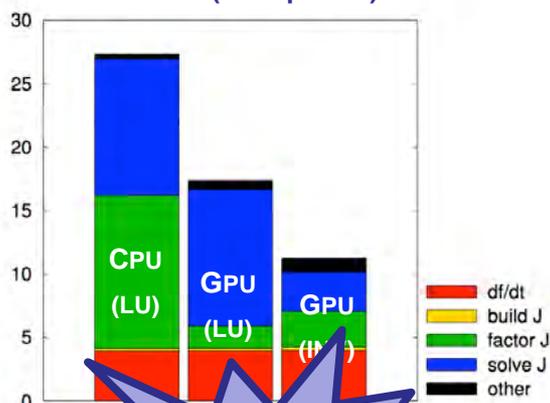


**GPUs can do many high repetitive calculations with little computational effort**

**CPUs get bogged down by these kinds of calculations**

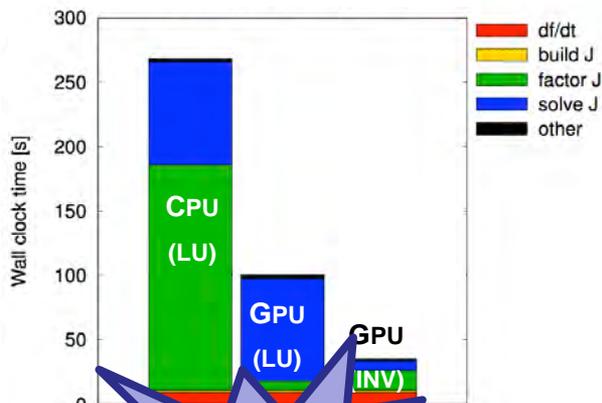
# We have achieved up to 11x speedup with GPU for ignition delay with large mechanisms

Iso-octane (857 species)



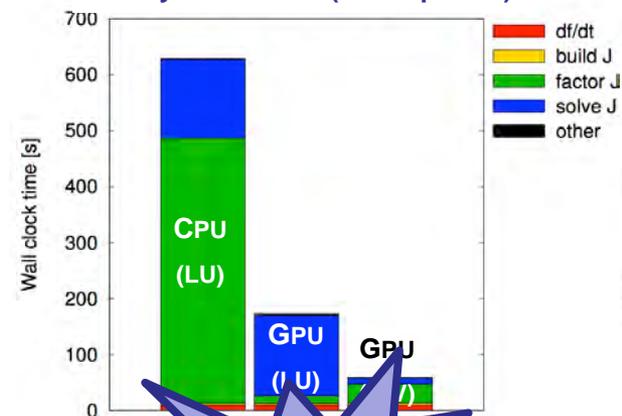
3x  
Speedup

n-hexadecane (2115 species)



8x  
Speedup

methyl-decanoate (2887 species)

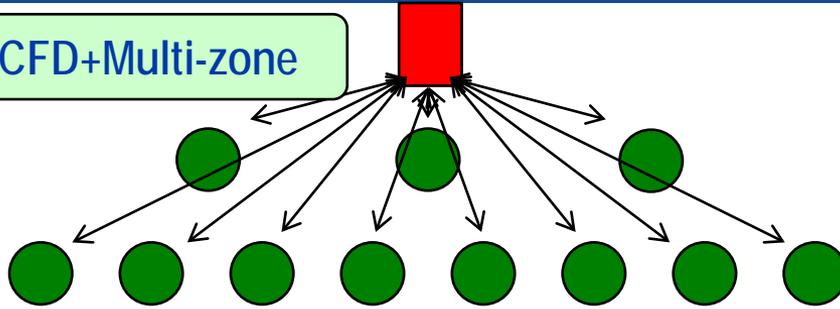


11x  
Speedup

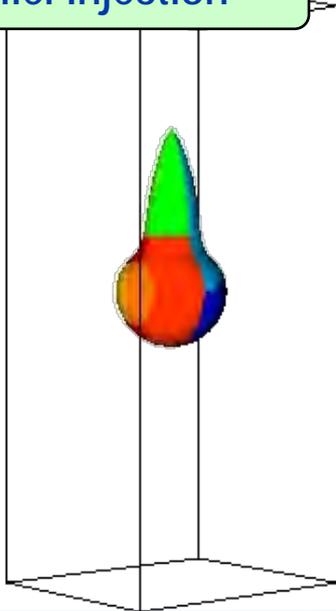


# We have done extensive KIVA-4mpi submodel development and testing to enable large-scale parallel CFD for engine simulation

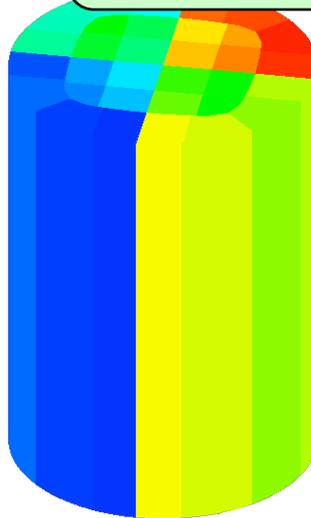
Parallel CFD+Multi-zone



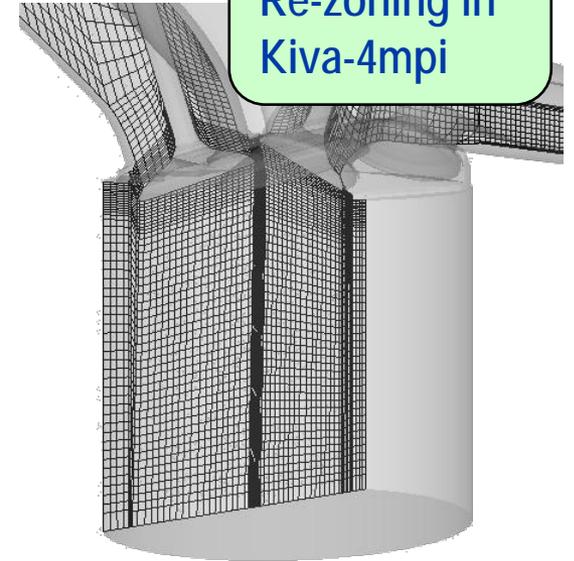
Parallel Injection



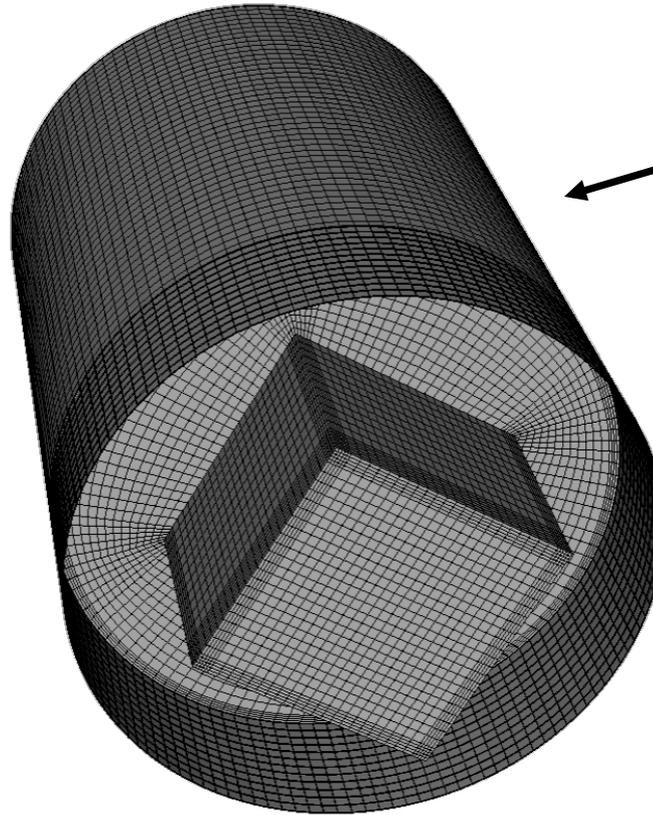
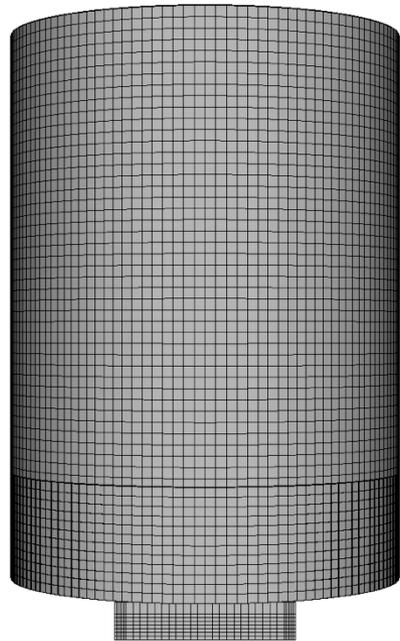
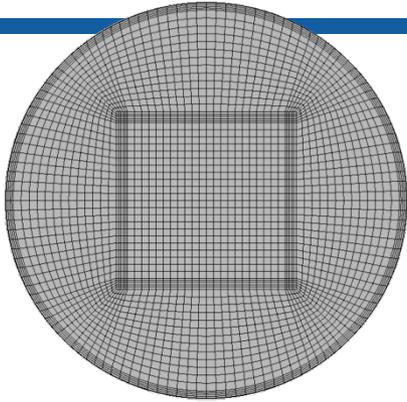
Mesh Partitioning Strategies



Robust Grid Re-zoning in Kiva-4mpi



# KIVA-4mpi-Multizone is tested with a well-characterized 3D benchmark case, square-bowl HCCI

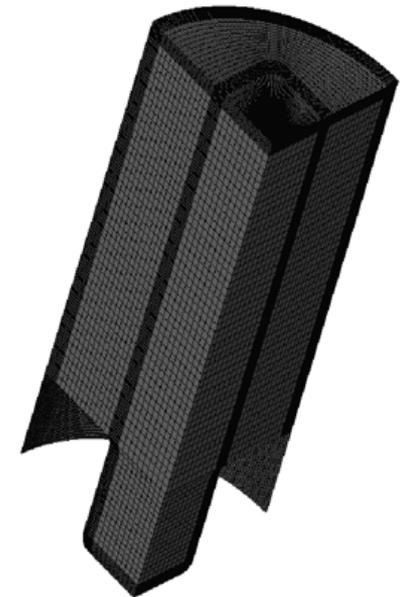


Grid 1: ~25,000 cells

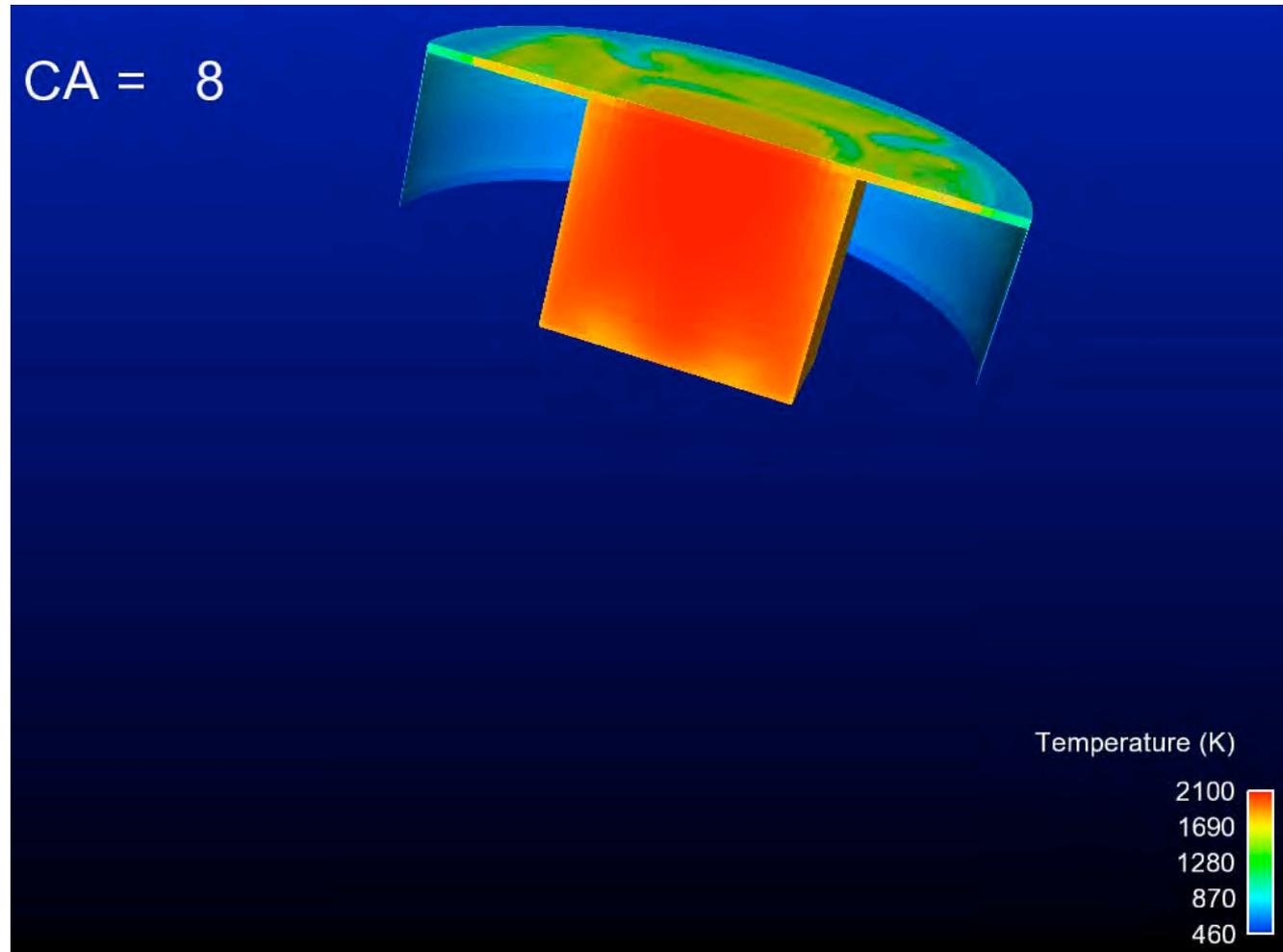
Grid 2: ~70,000 cells

Grid 3: ~200,000 cells

Kiva-3V Grid: ~400,000 cells



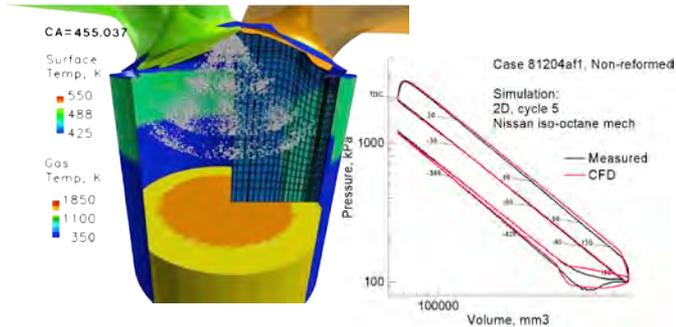
# Kiva-4mpi-MZ allows us to investigate more complex geometries with large detailed mechanisms



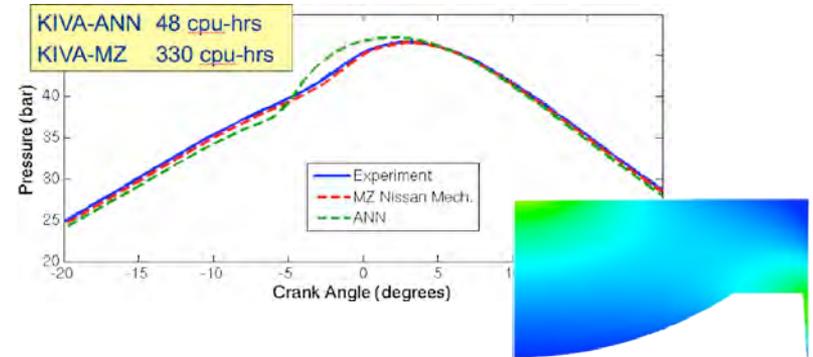
## Kiva4-mpi-multi-zone simulation



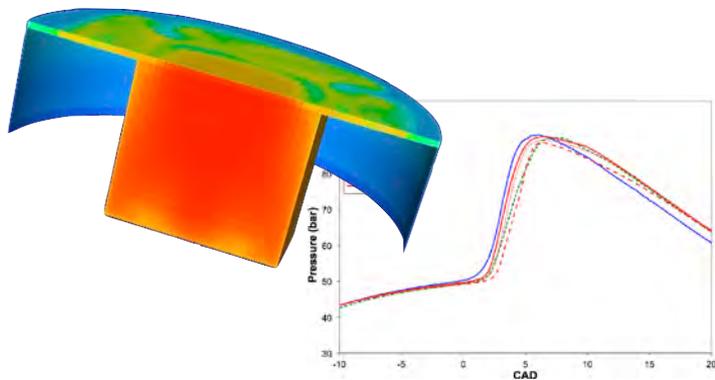
# We consider it vital to conduct simulations with close coupling to experimental data



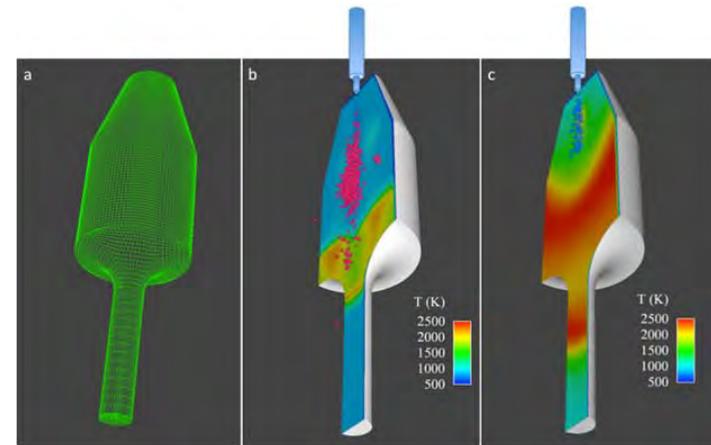
**Sandia NVO PCCI**



**Sandia DI PCCI**

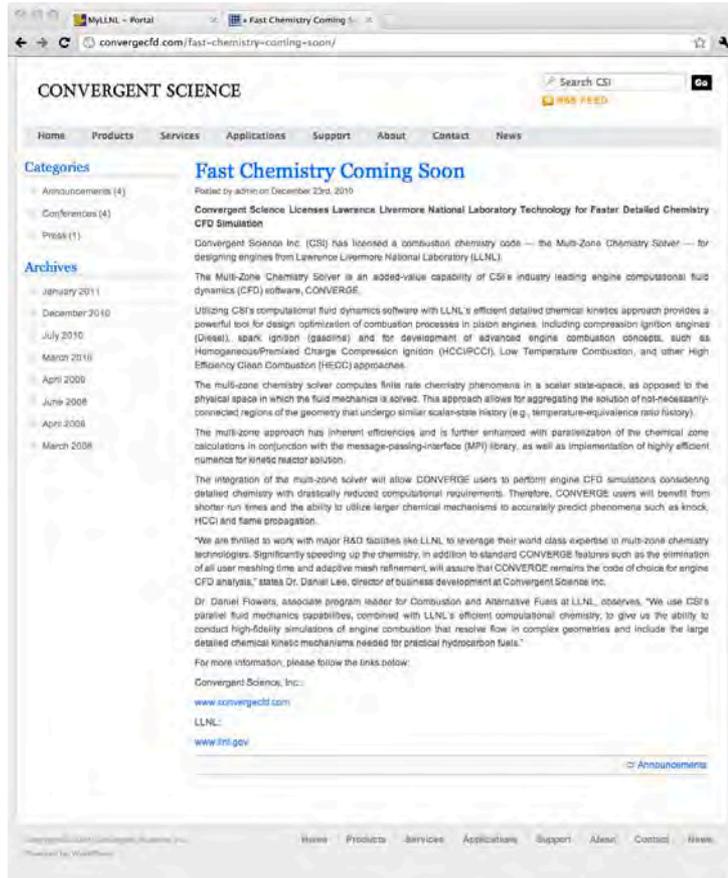


**Lund HCCI**



**NREL IQT (DI Diesel)**

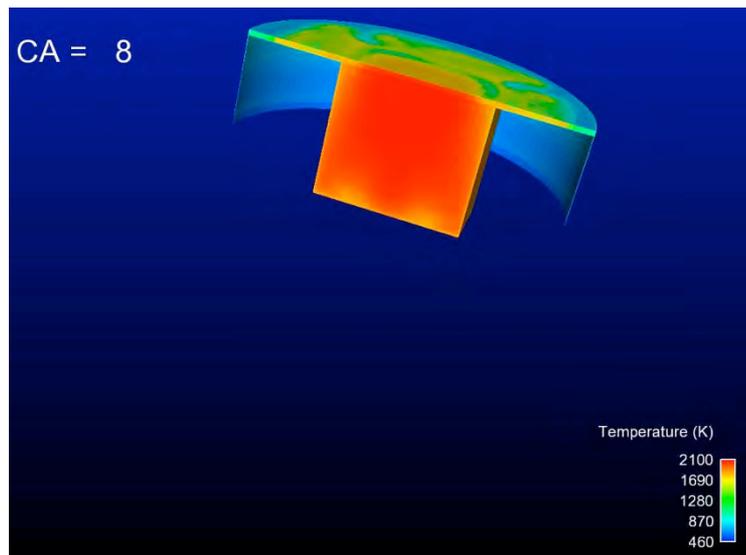
# We completed a 5 year licensing agreement for the LLNL Multi-zone Model with Convergent Science Inc. (CSI)



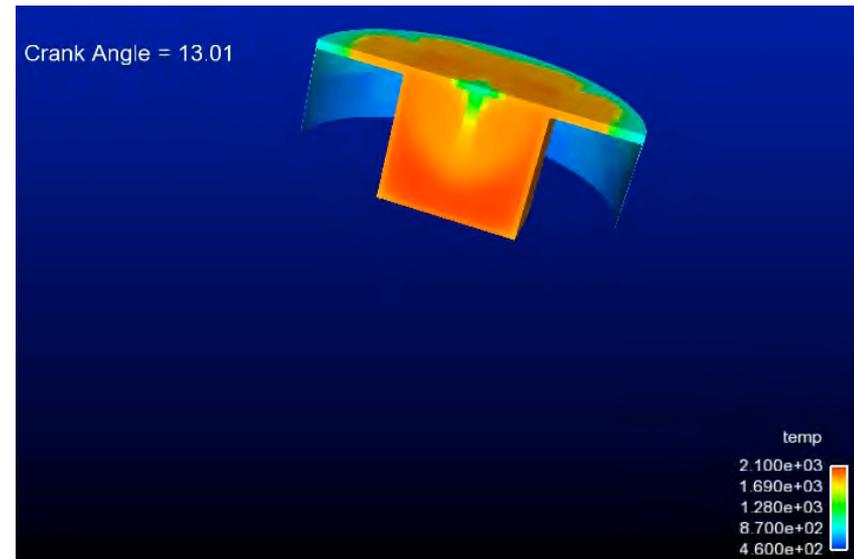
- CONVERGE from CSI is high performance parallel CFD solver widely used in US industry
- LLNL has CONVERGE-MZ licenses for complex 3D problems on our large-scale parallel computers
- LLNL is working with CSI to implement and test the CONVERGE multi-zone model



# We are benchmarking KIVA-4mpi-multi-zone and CONVERGE-multi-zone



**Kiva4-mpi-multi-zone**



**CONVERGE-multi-zone**

# Collaboration: We have ongoing interactions Industry, National Labs, and Universities

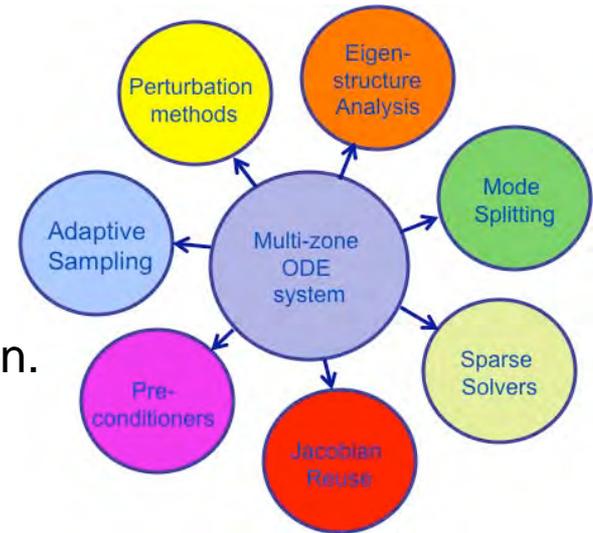
- **Convergent Science Inc** - Multi-zone license agreement
- **Advanced Engine Combustion (AEC) working group** (Industry, National labs, Univ. of Wisc.) - semiannual presentations
- **Fuels for Advanced Combustion Engines (FACE)** - working group
- **Sandia National Laboratory** - HCCI and PCCI, gaseous injection
- **Oak Ridge National Laboratory** - SI-HCCI transition and  $^{14}\text{C}$  exhaust analysis for HCCI and Diesel engines
- **Los Alamos National Laboratory** - Kiva4 development
- **Other Universities** - UC Berkeley, University of Wisconsin, University of Michigan, Lund Institute, Chalmers University, Tianjin University
- **Ford** - gaseous direct injection
- **Delphi** - direct injection



# Future Work: We will explore strategies for improving efficiency of CFD and chemistry simulations

## ■ **Improved computational chemistry solvers**

- Sparse solvers (CPU & GPU)
- More efficient data structures
- Hybrid solver solutions
- Solver parallelization compatibility
- Reaction sort with submatrix direct inversion.
- New integration error control logic
- Increase GPU shared memory reuse
- GPU particle motion/collision algorithms



## ■ **Improved parallel CFD with chemistry**

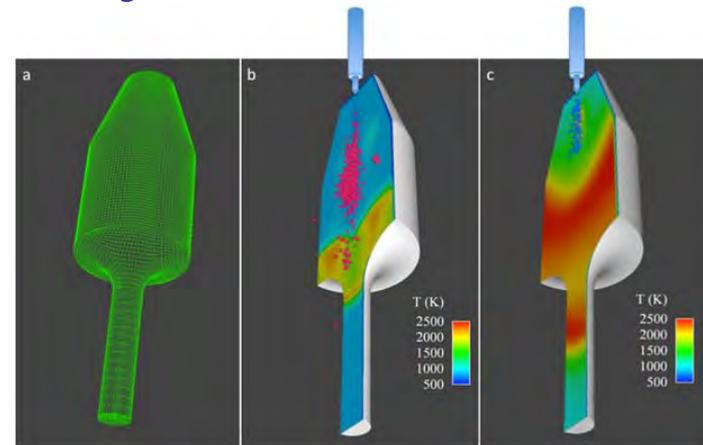
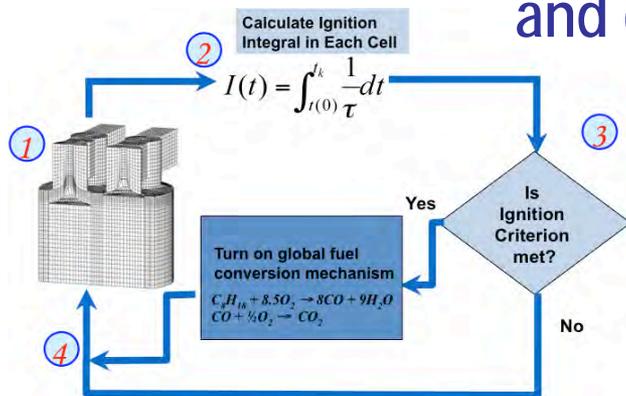
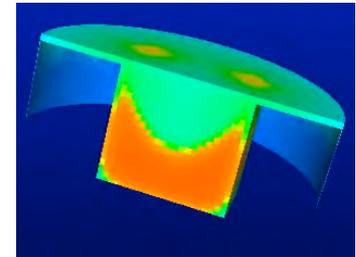
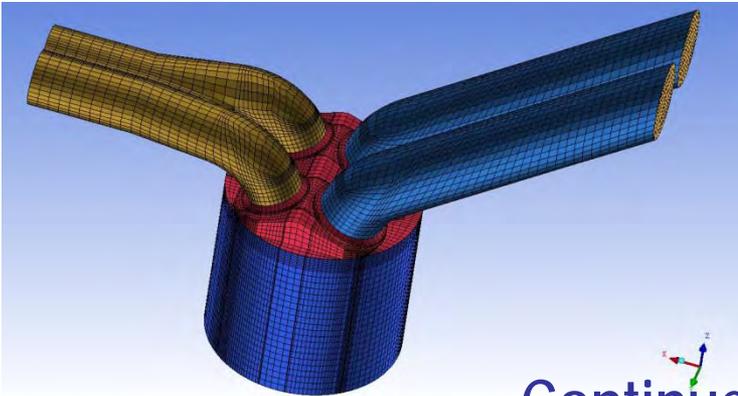
- Multi-criteria multi-zone
- Spray parcel models
- Spray initialization



# Future work: extend applicability and computational efficiency of analysis tools

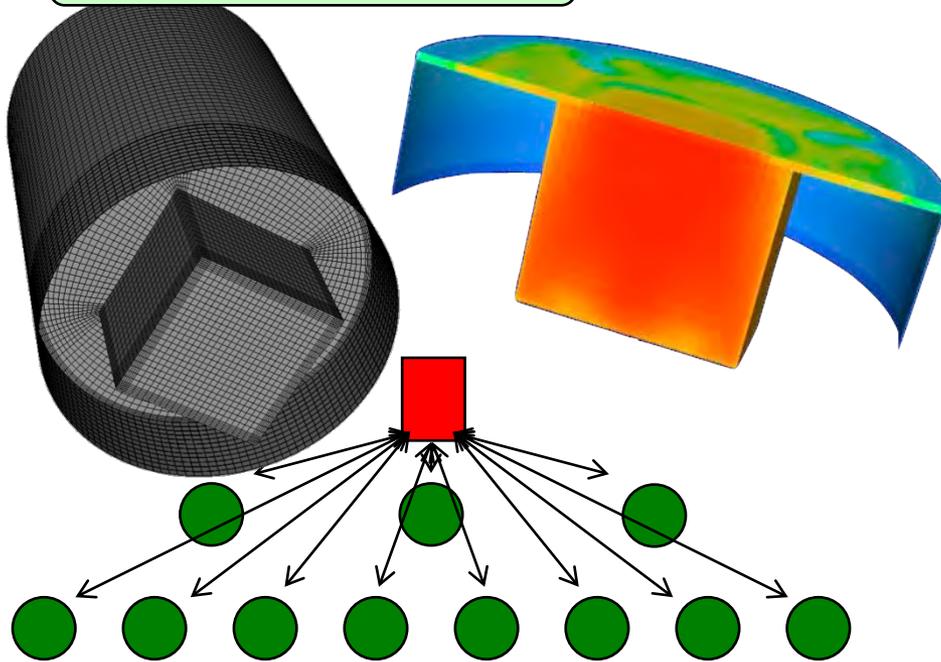
Enable 3-D fluid mechanics and detailed kinetics in today's desktop PCs

Continue to validate and develop parallel CFD, multi-zone, and ANN for highest fidelity in fluids and chemistry

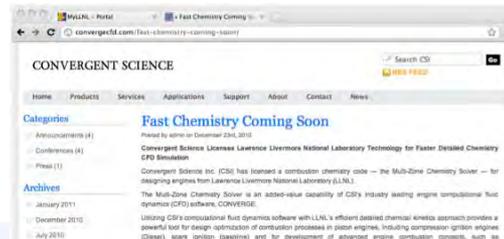
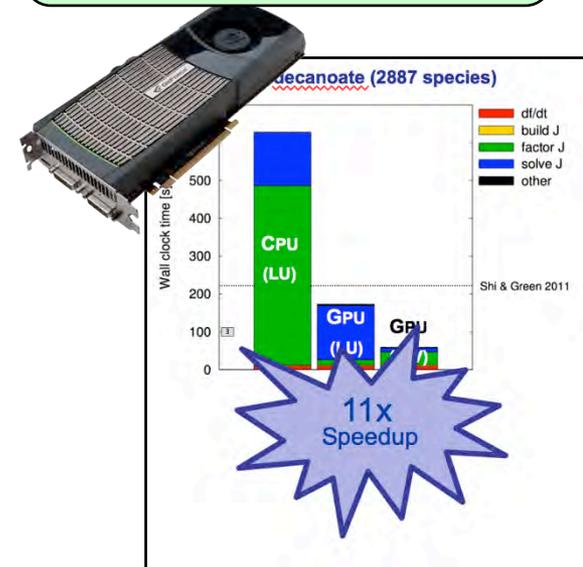


# Summary: we are enhancing our analysis capabilities and improving computational performance

## Parallel CFD+Multi-zone



## Accelerated Combustion Chemistry Simulation



## Model Commercialization

