

Advancing parabolic operators in thermodynamic MHD models: Explicit super time-stepping vs. implicit methods with Krylov solvers

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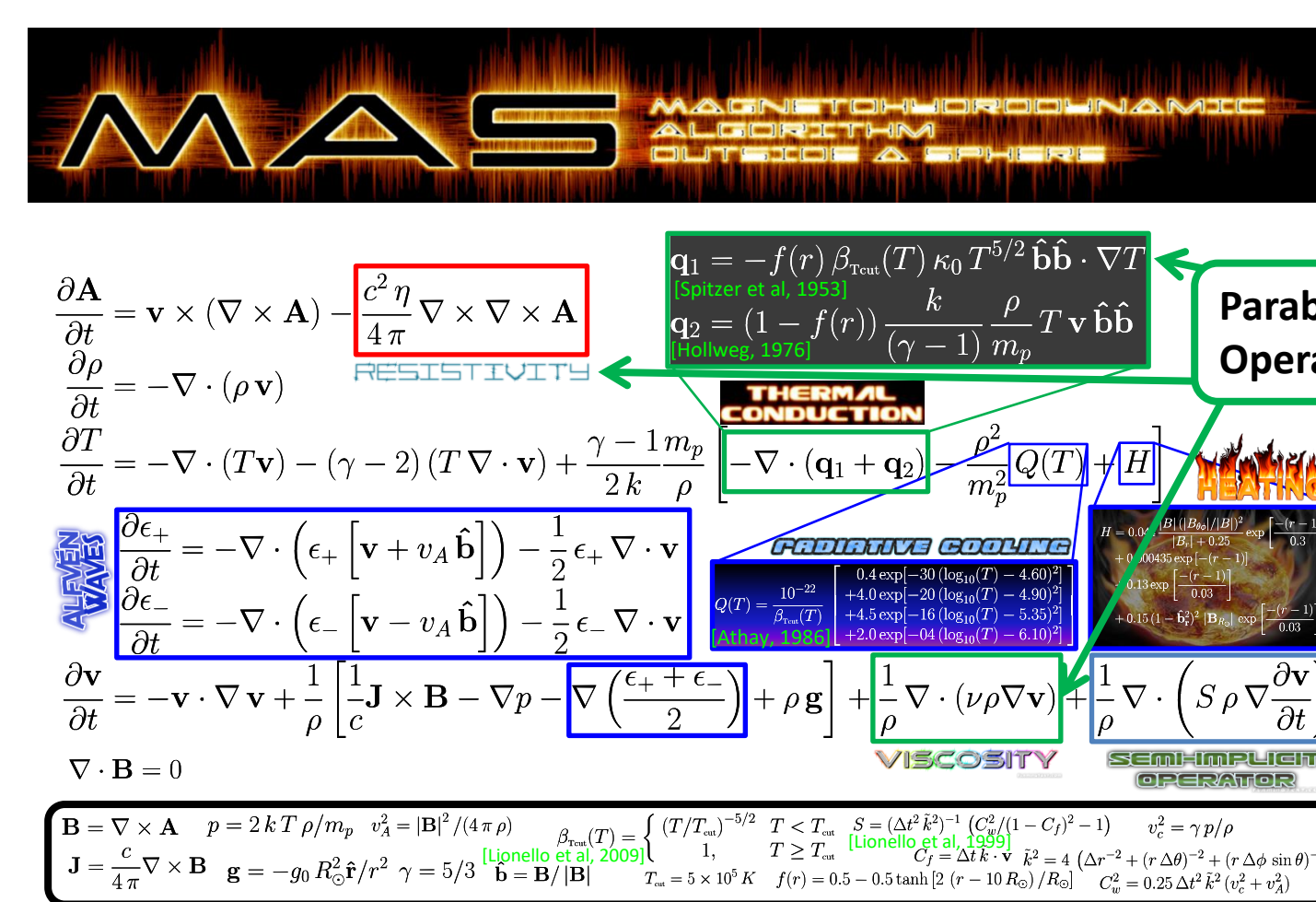
INTRODUCTION

We explore the performance and scaling of using explicit super time-stepping (STS) algorithms versus implicit schemes with Krylov solvers for integrating parabolic operators in thermodynamic MHD models of the solar corona. Specifically, we compare the second-order Runge-Kutta Legendre (RKL2) STS method with implicit backward Euler computed using the preconditioned conjugate gradient (PCG) solver with both a point-Jacobi and a non-overlapping domain decomposition ILU0 preconditioner.

The algorithms are used to integrate a scalar operator (anisotropic Spitzer thermal conduction) and a vector operator (artificial kinematic viscosity) at time-steps much larger than the explicit Euler limit. A key component of the comparison is the use of a real-world simulation on large state-of-the-art HPC systems.

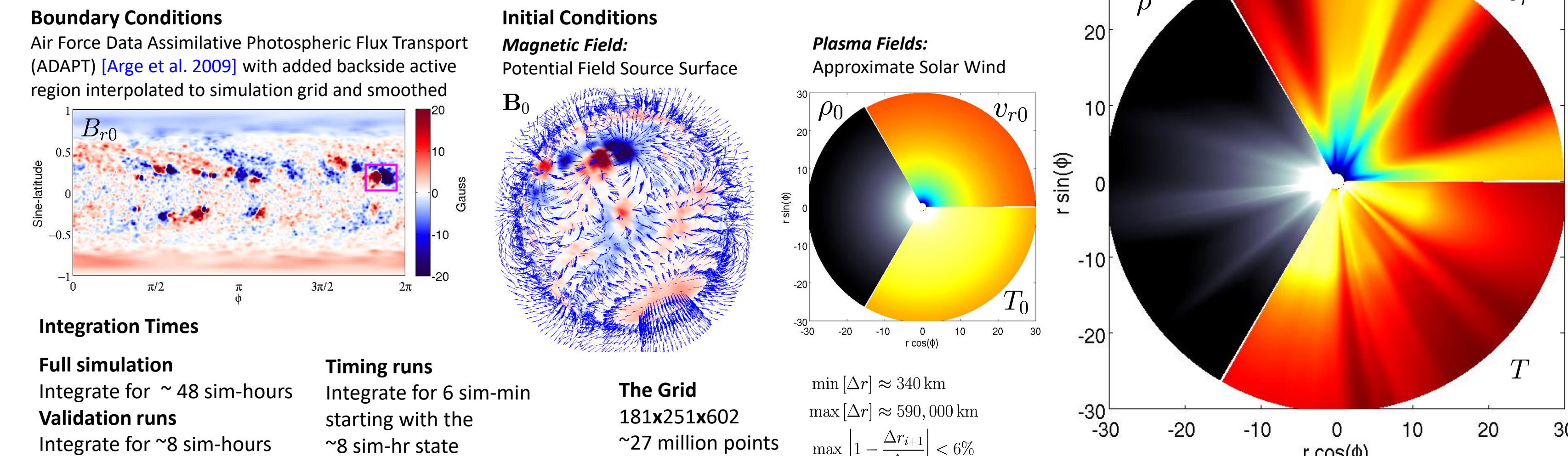
A paper detailing this study has been submitted and is made available [arXiv:1610.01265].

MODEL AND NUMERICAL METHODS



- Established MHD Solar coronal model [Lionello, et al Ap.J. 690 902 (2008)]
- FORTRAN 90 + MPI-2
- Used extensively in solar physics research
- Parabolic operators tested with algorithms:
 - Finite difference on logically rectangular, staggered, non-uniform spherical grid
 - Advective terms are upwinded
 - Wave terms use predictor-corrector
 - Semi-implicit term solved using BE+PCG for both predictor and corrector
 - Parabolic terms are operator split and use second-order central differences
 - Matrix operators stored in modified DIA format, PC2 preconditioner stored in CSR format

REAL-WORLD 3D TEST CASE



Implicit Backward Euler with Preconditioned Conjugate Gradient

Backward Euler (BE): simplest L-stable method

$$\frac{\partial u}{\partial t} = F(u, r) \Rightarrow \frac{u^{n+1} - u^n}{\Delta t} = M u^{n+1}$$

Applying BE to PDE yields a system of equations to solve $(1 - \Delta t M) u^{n+1} = u^n \Rightarrow A x = y$

To solve system, use Krylov subspace iterative method (e.g. Conjugate Gradient)

Preconditioners (helps solver converge)

We use two communication-free PCs:

PC1: Point-Jacobi / Diagonal scaling (DIAG)

PC2: Non-overlapping domain decomposition with zero-fill incomplete LU factorization (NDD+ILU0)

PC1 (DIAG):

```
do j = 1 : N
  P_jj = A_jj
enddo
do i = 1 : N
  z_i = P_ii r_i
enddo
```

PC2 (NDD+ILU0):

```
LU = A
do i = 2 : N
  do k = 1 : i - 1 (LU_ik ≠ 0)
    LU_ik = LU_ik / LU_kk
  enddo
  do j = k + 1 : N (LU_ij ≠ 0)
    LU_ij = LU_ij - LU_ik LU_kj
  enddo
enddo
enddo
P = LU
```

Point-2-Point comm+sync

Global comm+sync

$$x_0 = u^n \quad z_0 = P^{-1} r_0$$

$$r_0 = b - A x_0 \quad p_0 = z_0$$

$$P \approx A \quad r_r = r_0 \cdot z_0$$

```
do k = 0 : k_max
  y_k = A p_k
  alpha_k = (p_k · y_k)
  x_{k+1} = x_k + alpha_k p_k
  r_{k+1} = r_k - alpha_k y_k
  z_{k+1} = P^{-1} r_{k+1}
  r_r = r_r + alpha_k z_{k+1}
  Check r_r for convergence
  beta_k = r_r / r_{old}
  p_{k+1} = beta_k p_k + z_{k+1}
enddo
u^{n+1} = x_{k+1}
```



- Unconditionally stable, but explicit!
- Main idea: Runge-Kutta method with stages added for more stability, rather than for more accuracy
- Two main flavors are RKC (Chebyshev-based) and RKL (Legendre-based) and can be recursive or factored
- RKL2 [Meyers et al, J. Comp. Phys. 257, 594 (2014)]

Very easy to implement (ideal for directive parallelization e.g. OpenMP/OpenACC)

No global synchronization points

Can include nonlinearities and/or flux limiting

New method - low circulation

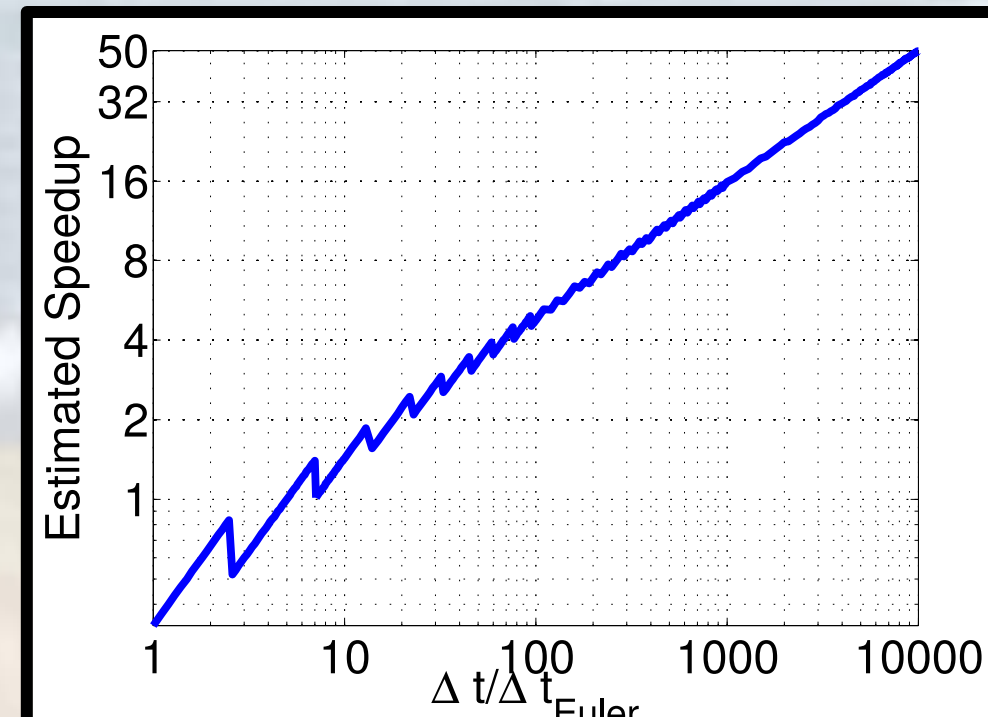
Number of sub-cycles can be large depending on the problem and grid

Amplification factor not monotonically decreasing for increasing wave modes - can lead to undesirable results when using large time-steps and/or short simulation times

Number of sub-steps

$$s \geq \left\lceil \frac{1}{2} \left(\sqrt{9 + 16 \frac{\Delta t}{\Delta t_{Euler}}} - 1 \right) \right\rceil$$

mod(s, 2) ≠ 0



Point-2-Point comm+sync

$$y_0 = u^n$$

$$F_0 = M y_0$$

$$y_1 = y_0 + \tilde{\mu}_1 \Delta t F_0$$

```
do j = 2 : s
  y_j = mu_j y_{j-1} + nu_j y_{j-2} + (1 - mu_j - nu_j) y_0 + mu_j Delta t M y_{j-1} + gamma_j Delta t F_0
enddo
```

$$u^{n+1} = y_s$$

$$b_0 = b_1 = b_2 = \frac{1}{3} \quad b_j = \frac{j^2 + j - 2}{2j(j+1)}$$

$$\tilde{\mu}_1 = \frac{4}{3(s^2 + s - 2)} \quad \tilde{\mu}_j = \frac{4(2j-1)}{j(s^2 + s - 2)} \frac{b_j}{b_{j-1}}$$

$$\mu_j = \frac{2j-1}{j} \frac{b_j}{b_{j-1}} \quad \nu_j = -\frac{j-1}{j} \frac{b_j}{b_{j-2}}$$

$$\gamma_j = (b_j - 1) \tilde{\mu}_j$$

HPC ENVIRONMENT

San Diego Supercomputing Center

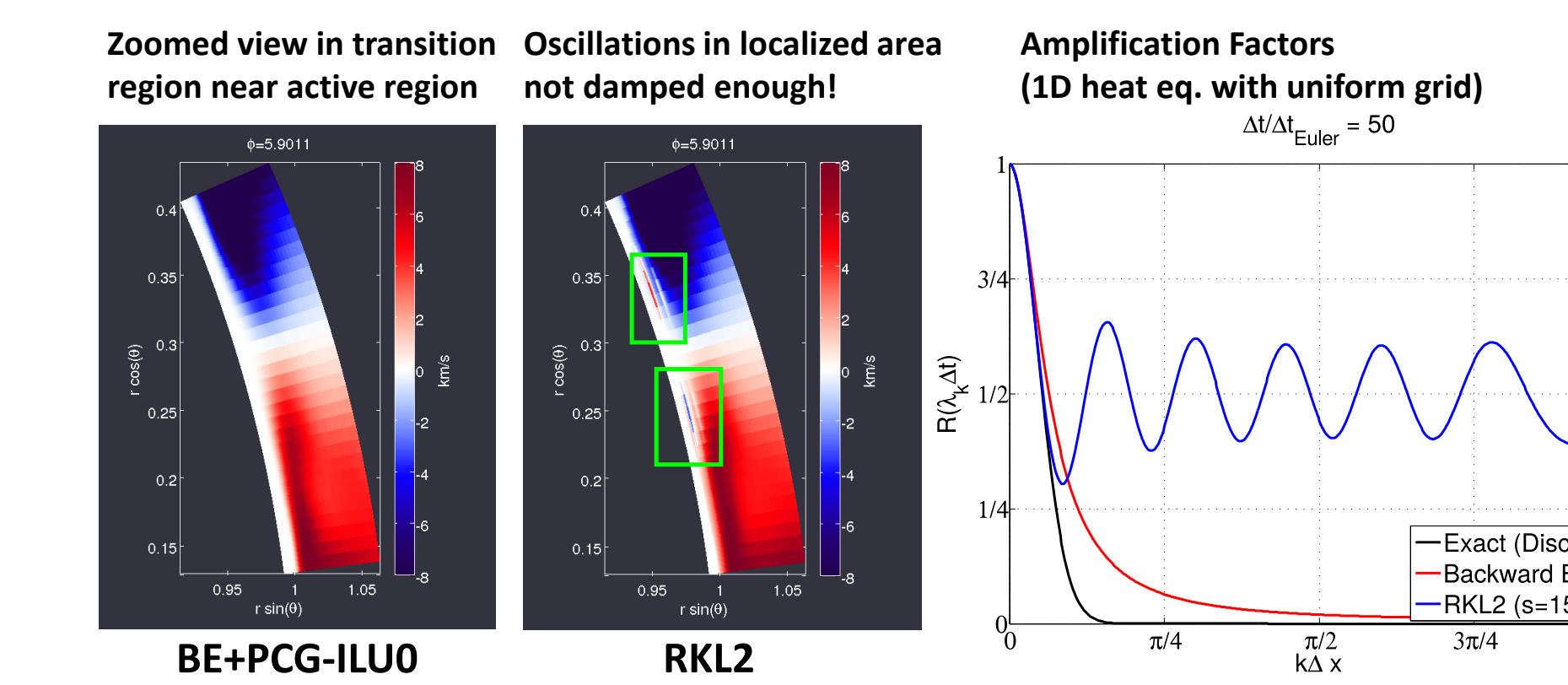
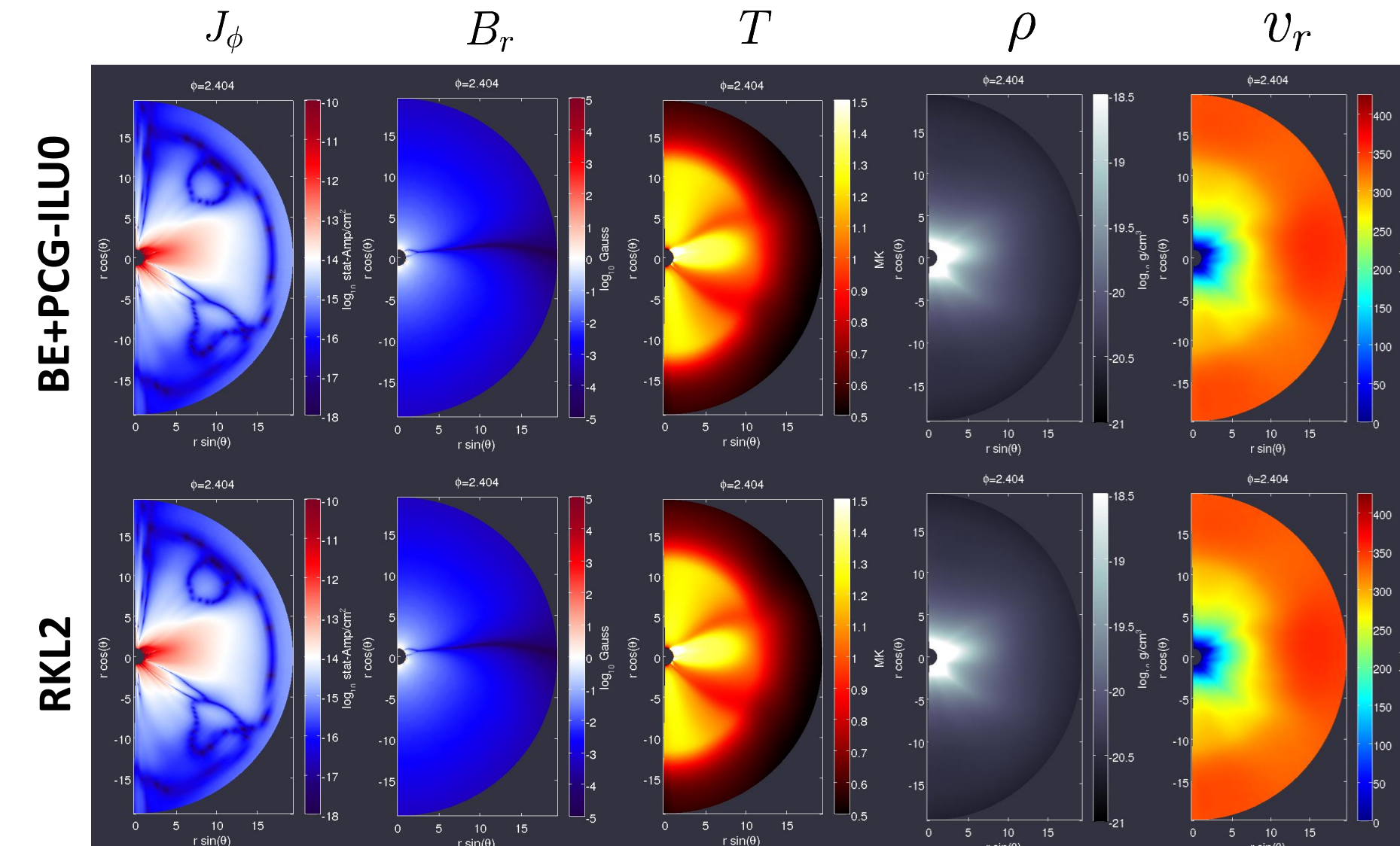
Texas Advanced Computing Center

	SDSC Comet	TACC Stampede
Processor Model	Intel Xeon E5-2660v3	Intel Xeon E5-2680v2
Clock Speed	2.5 GHz	2.7 GHz
CPU Cores/node	24	16
DRAM/node (ECC)	128 GB DDR4-2133MHz	32 GB DDR3-1600MHz
Max # of cores/job	1728	4096
Max Flops/node	960 GFlops	346 GFlops
Network	FDR InfiniBand Hybrid Fat-Tree	FDR InfiniBand 2-level Fat-Tree
MPI Library	MVAPICH 2.1	Intel 2015.2.164
Compiler	Intel 2015.2.164	Intel 2015.2.164
OS	CentOS 2.6.32-573	CentOS 2.6.32-431

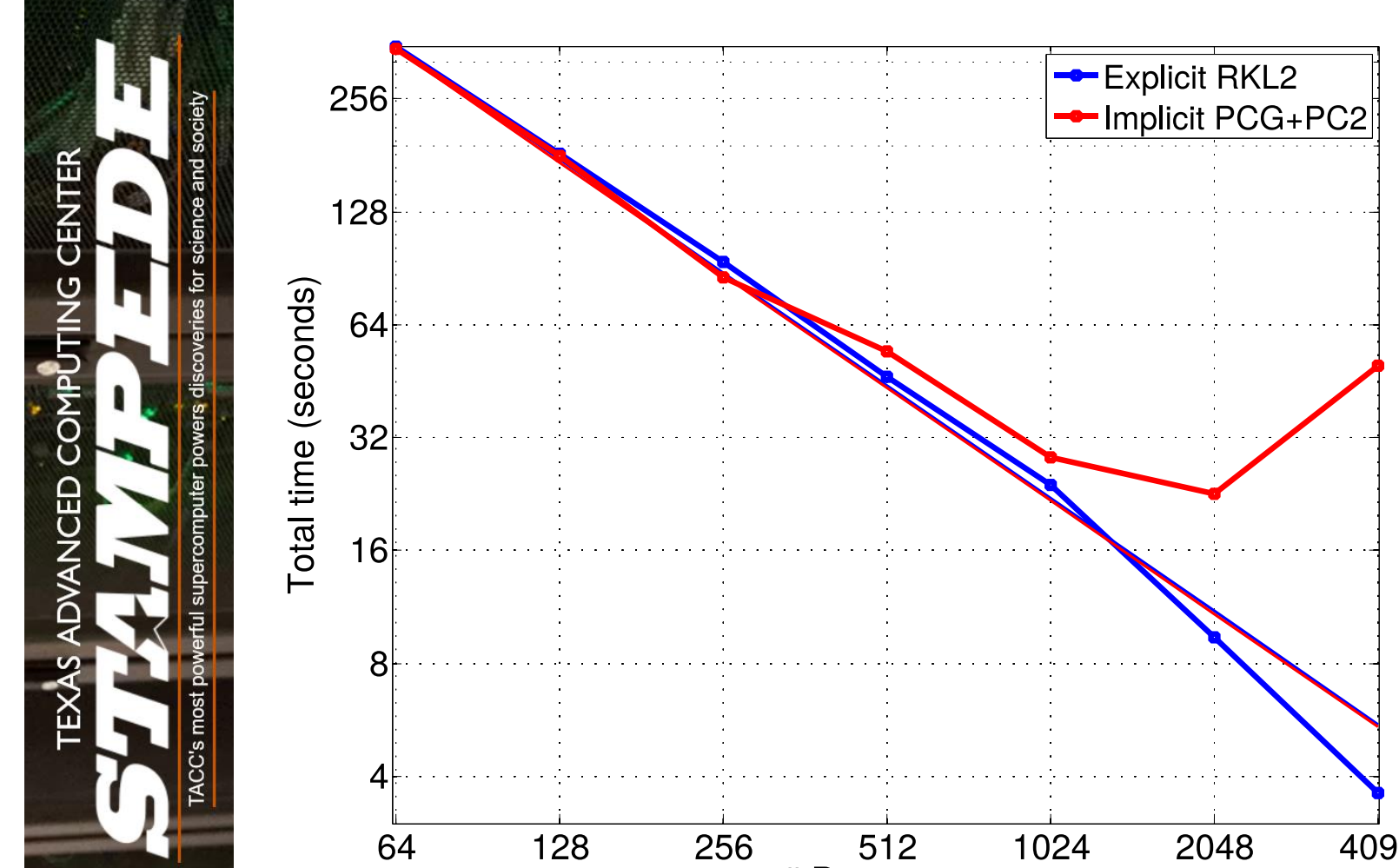
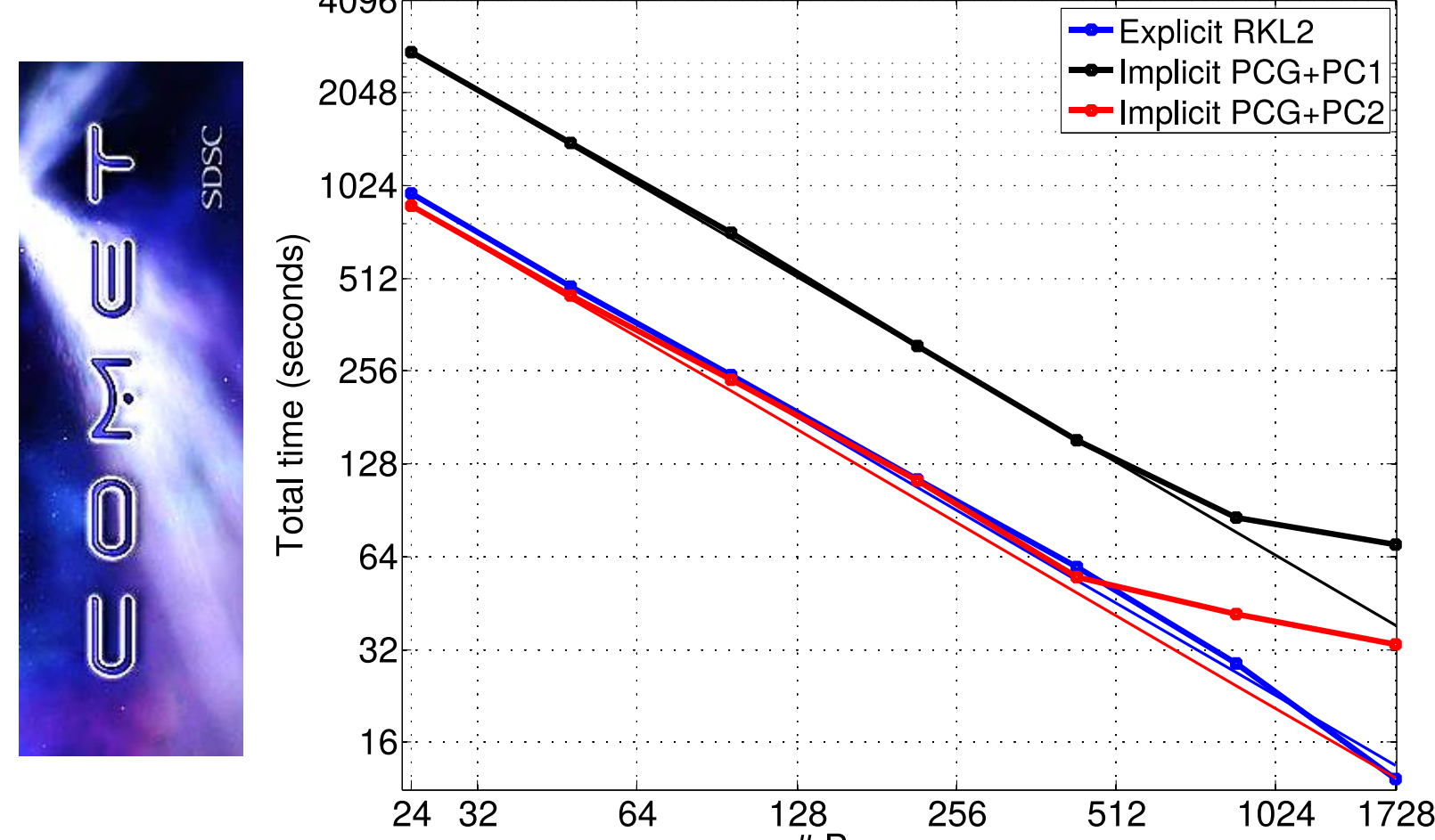
Allocations provided by: XSEDE

VALIDATION

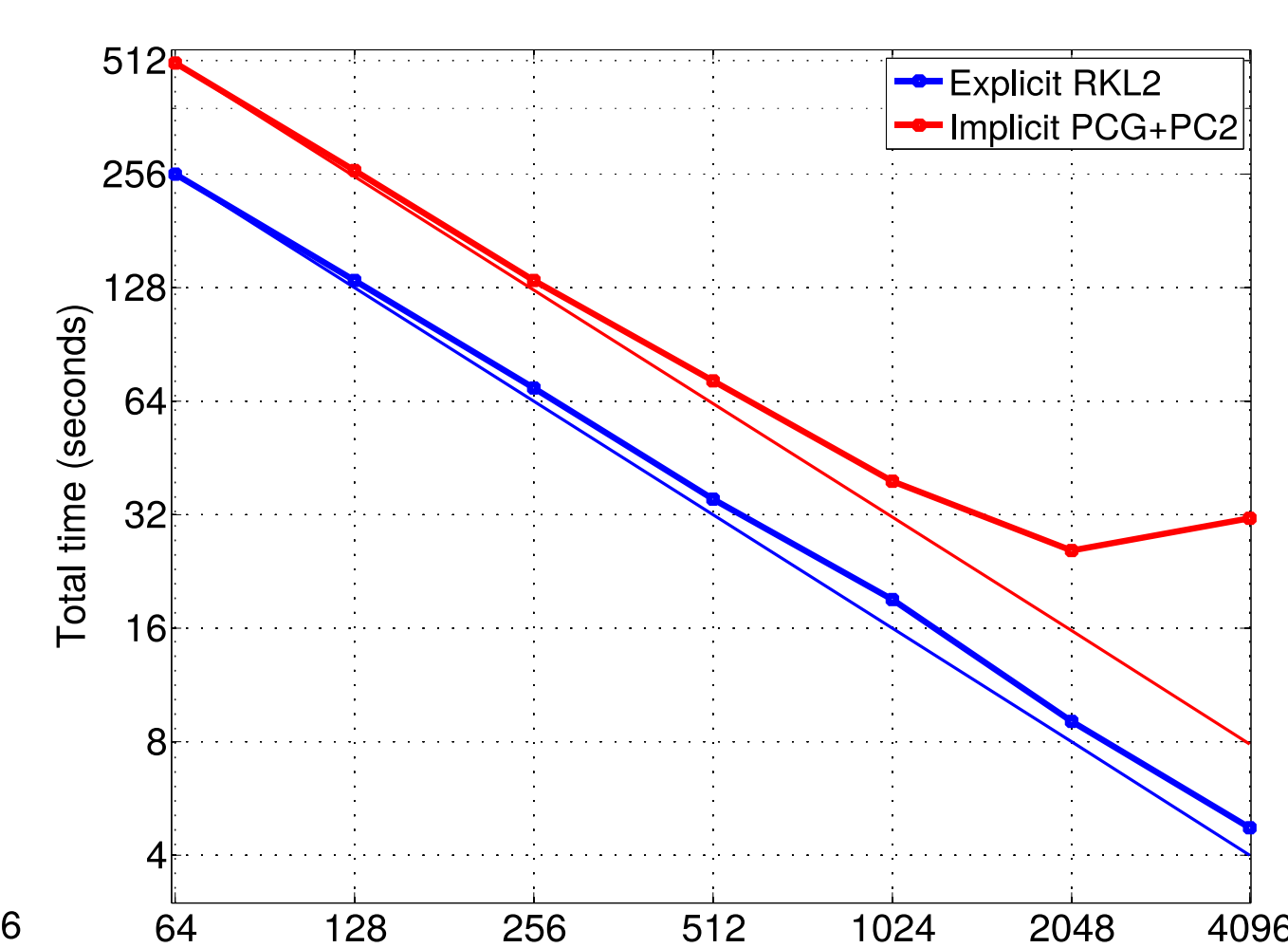
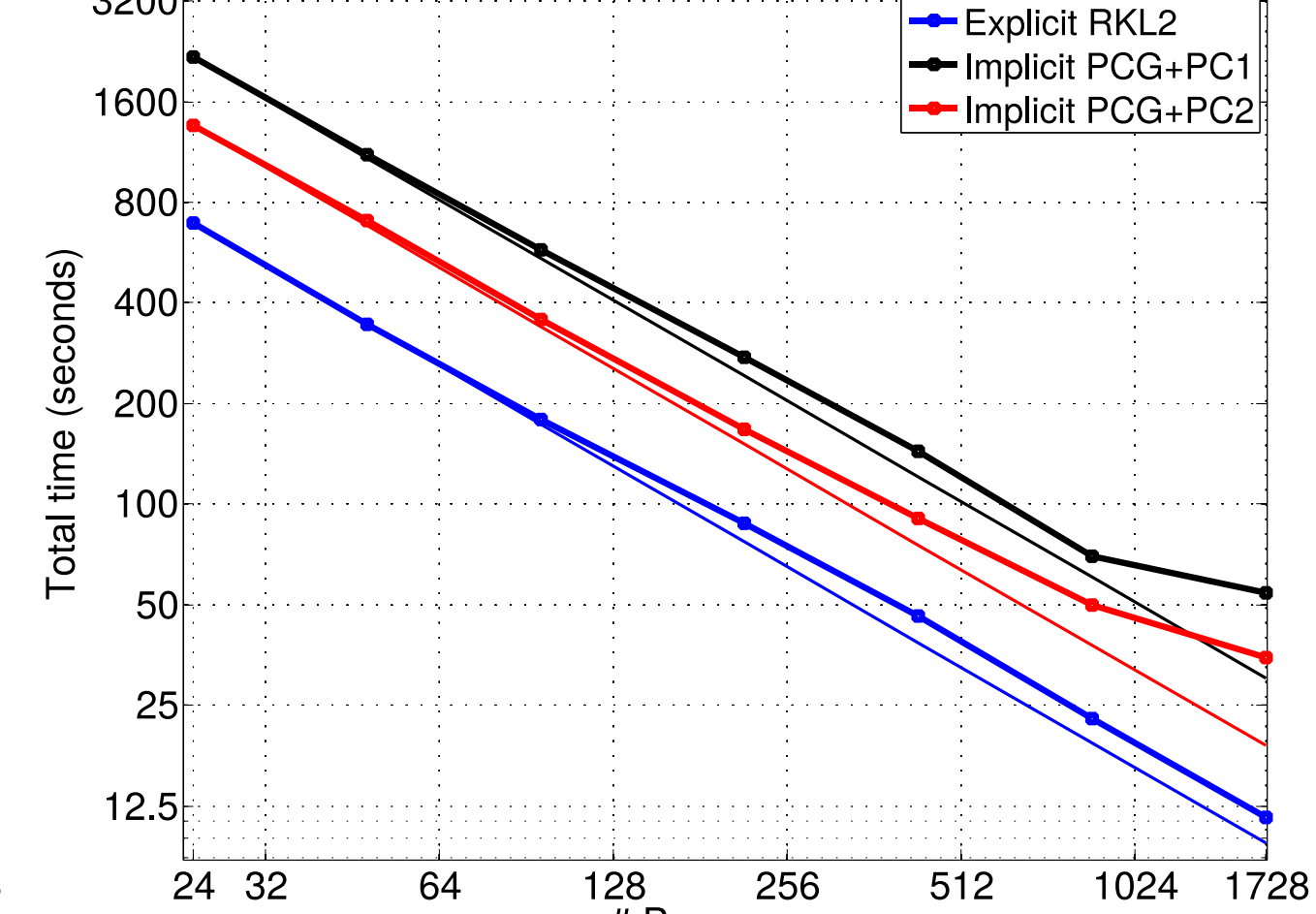
Solution cuts taken after relaxation:



Thermal Conduction

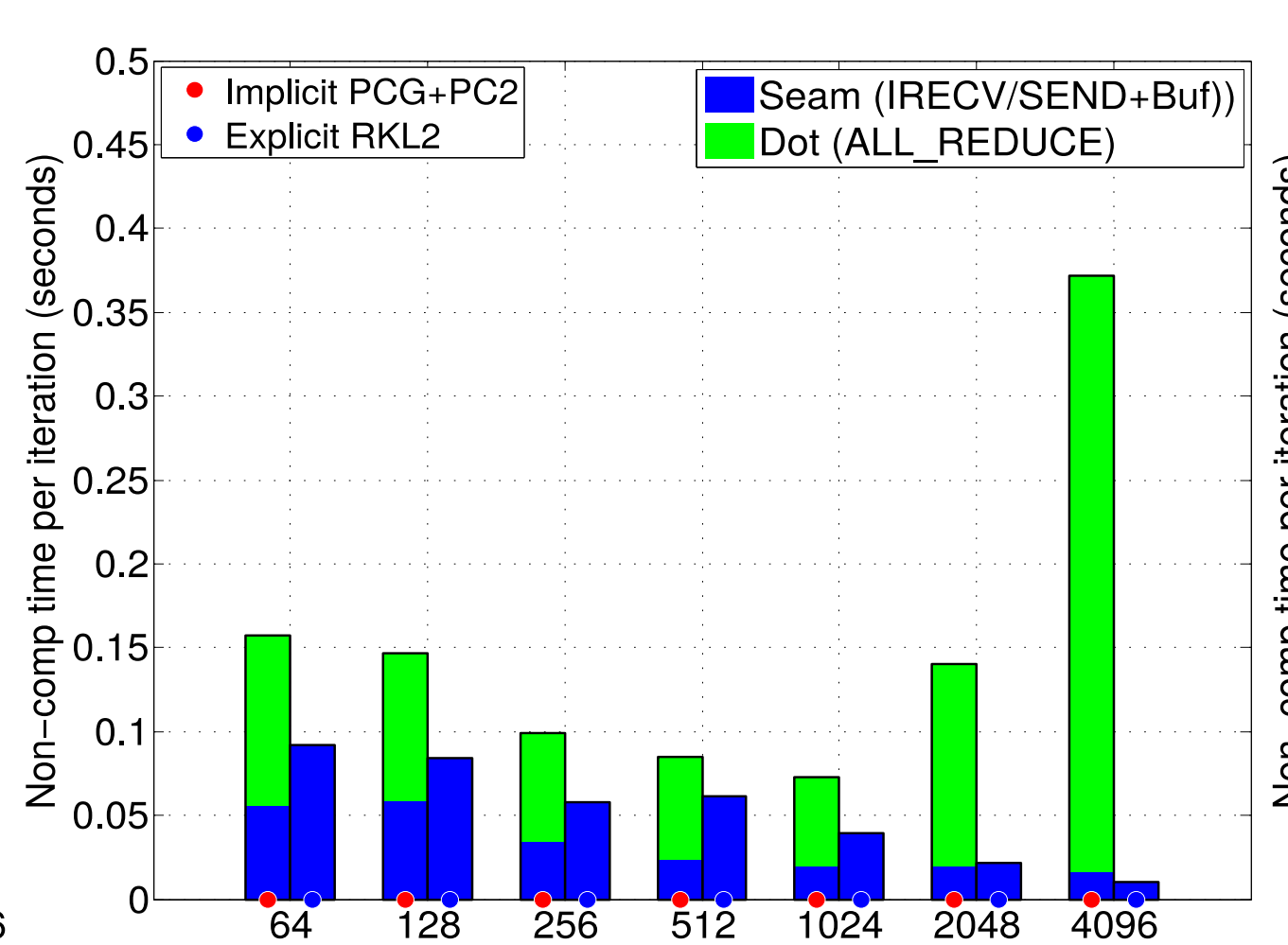
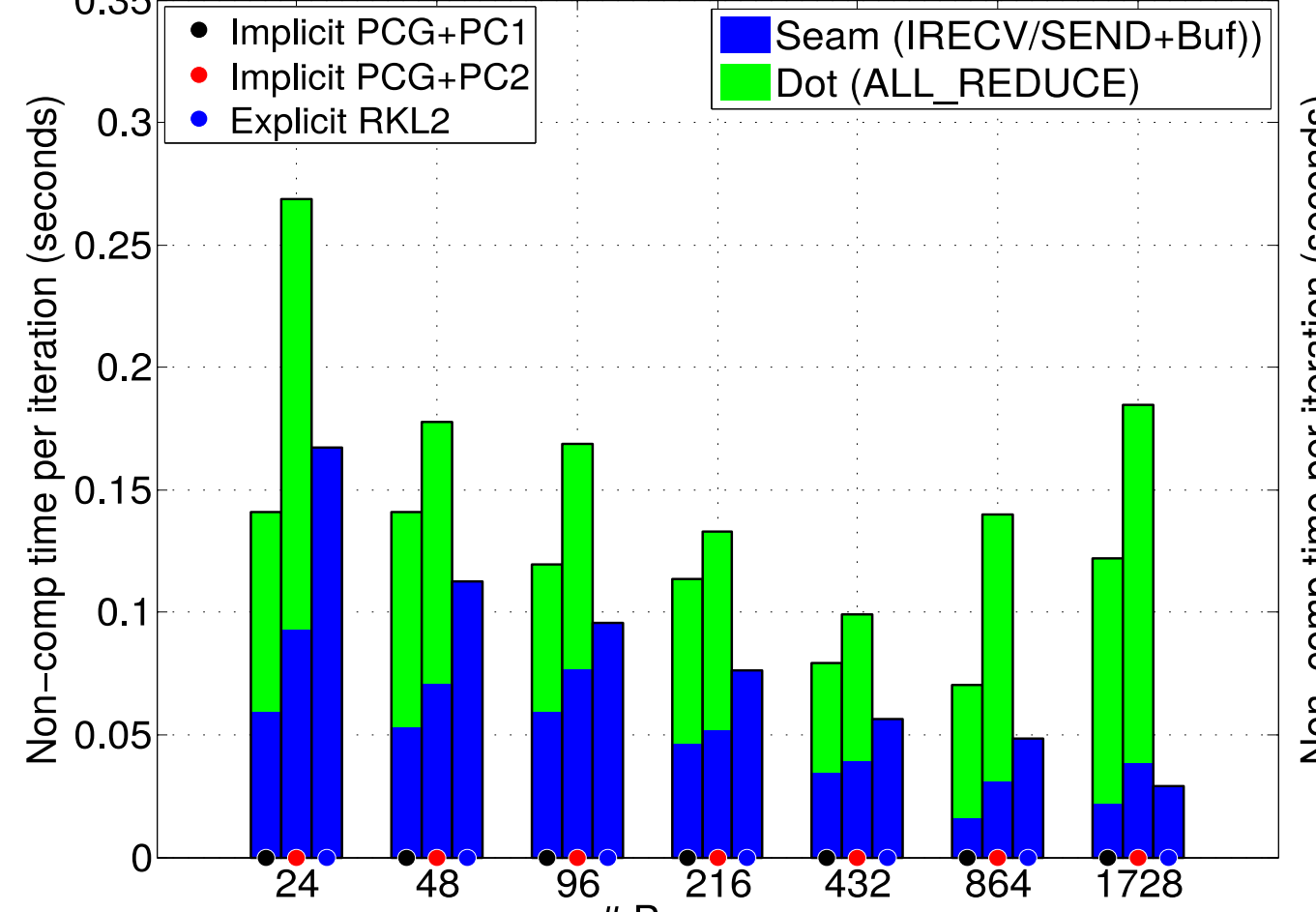


Viscosity

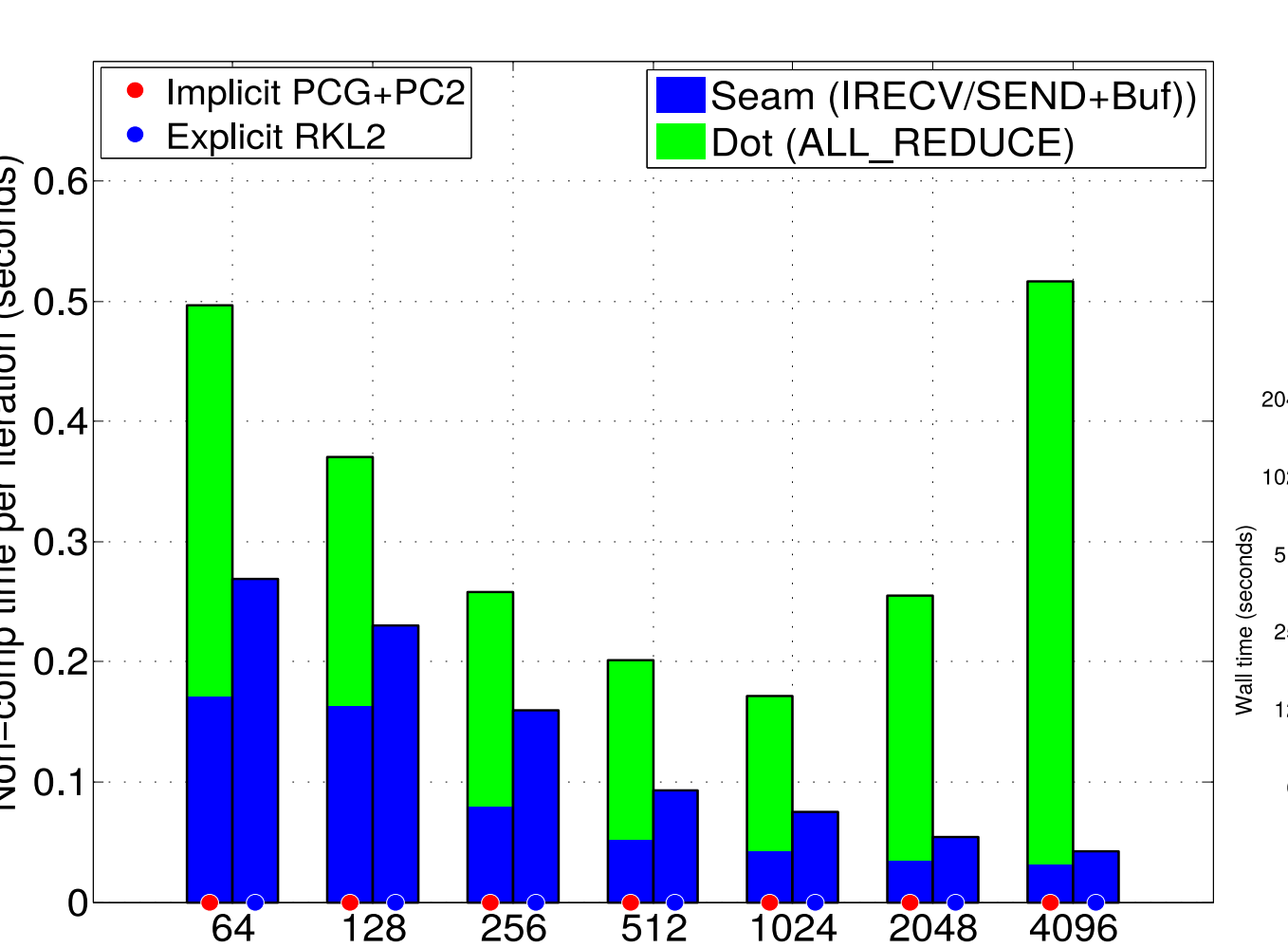
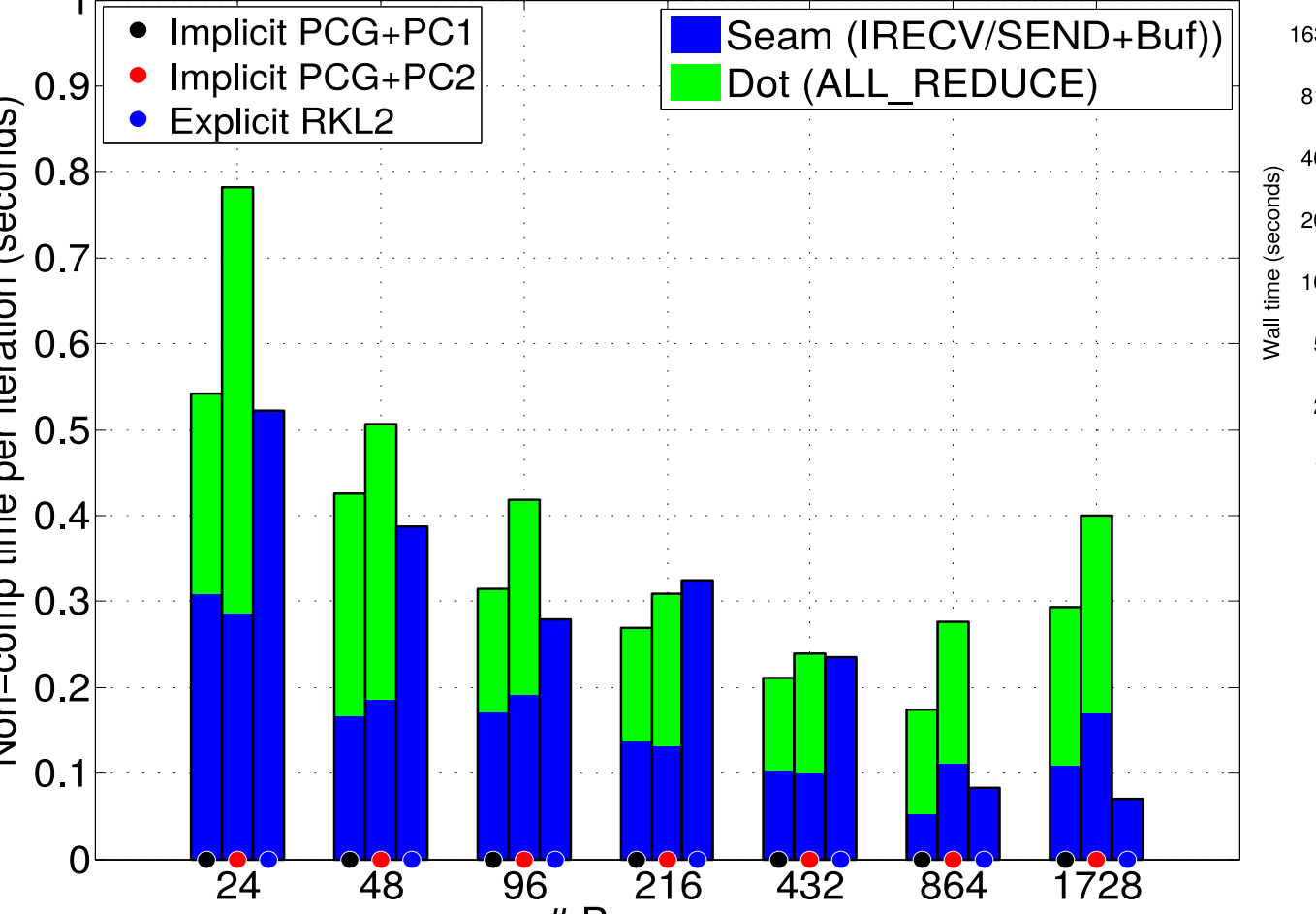


RESULTS

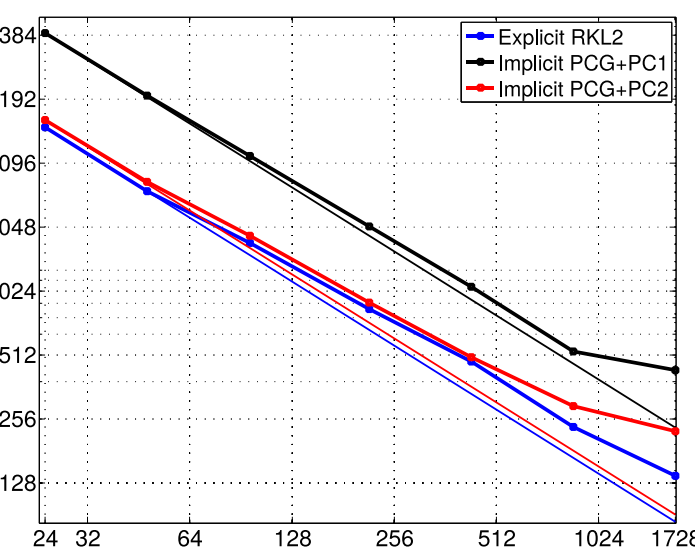
Thermal Conduction



Viscosity



Wall time (full code)



Wall time (full code)

