

Parallel Computation of Axisymmetric Jets

SciPy.in 2010

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Outline

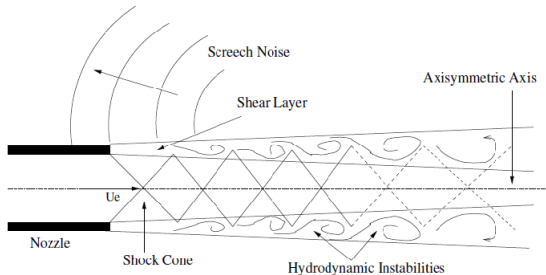
- 1 Introduction
- 2 Python and Cython Implementation
- 3 Parallelization of Code
- 4 Conclusion

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Flow Physics

- Imperfectly expanded jet
- Resonant screech loop
 - Instability wave growth
 - Instability-shock interaction
 - Acoustic feedback
 - Receptivity processes



In-house C code

- Favre-filtered Axi-symmetric Navier Stokes Solver
- Spatail discretization - 5th order WENO scheme
- Advancement in time by 2nd order TVD Runge-Kutta time stepping
- Smagorinsky's eddy viscosity model for subgrid scale modeling
- Constants for subgrid models -> based on DNS results of Erlebacher et al
- Accepts input parameters in text file
- Uses procedural programming approach

Parallel computing

- Shared memory
 - All processors access all memory as global address space
 - Changes in a memory location effected by one processor are visible to all other processors
 - Advantage: Data sharing between tasks is fast and uniform
 - Disadvantage: Lack of scalability between memory and CPUs
- Distributed memory
 - Processors have their own local memory
 - Memory addresses in one processor do not map to another processor
 - Advantage: Memory is scalable with number of processors
 - Disadvantage: Synchronization between tasks is programmer's responsibility

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Optimization with Cython

Implementation	Average CPU time for a iteration	Speedup
Python	193.34 s	
In-house C code	0.61 s	316.9x

- Compiling in Cython
- Declaration of variables
- Declaration of numpy arrays
- Switching off boundscheck and wraparound
- Multi-threading

Software	Version
Python	2.6
Cython	0.12.1
matplotlib	0.99.1.1

Performance

Implementation	¹ Average CPU time for an iteration	Speedup
Python	193.34 s	
Compilation in Cython	174.97 s	1.11x
Variable type declaration	155.79 s	1.24x
Numpy array declaration	1.19 s	162.5x
boundscheck and wraparound	0.94 s	205.7x
In-house C code	0.61 s	

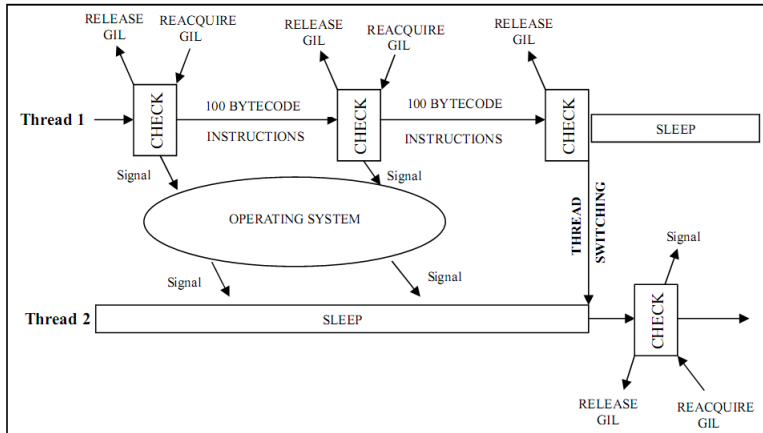
Processor : Intel(R) Core(TM)2 Duo CPU E7500@ 2.93GHz with 3072 KB cache size

¹Average CPU time computed by averaging the time for 3 iterations on a 350 x 300 uniform mesh

Multi-threading

- No OpenMP support
- Threading module of Python employed
- Global Interpreter Lock (GIL)
 - Controls thread execution
 - Ensures exclusive access to interpreter internals
 - Necessary for a Python operation
- Python does not have a thread scheduler
- *with nogil* statement was used to get rid of GIL

Thread Scheduling



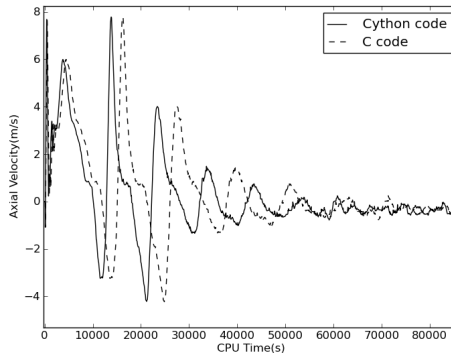
Final Performance

- Spatial derivative for axial and radial direction
 - independent of each other
 - computed on 2 threads

Implementation	Average CPU time for an iteration	Speedup
Python	193.34 s	
Compilation in Cython	174.97 s	1.11x
Variable type declaration	155.79 s	1.24x
Numpy array declaration	1.19 s	162.5x
boundscheck and wraparound	0.94 s	205.7x
Multi-threading	0.52 s	371.8x
In-house C code	0.61 s	

Significance

- LES computations \rightarrow 100 thousand iterations or more for developed flow
- Improvement of 0.09 s/iteration saves 2.5 hrs of CPU time



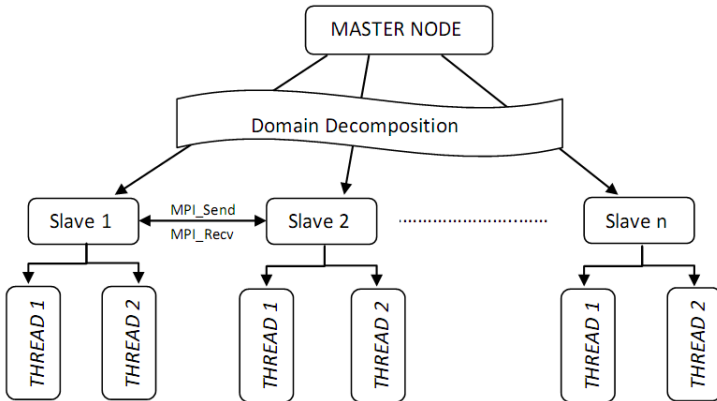
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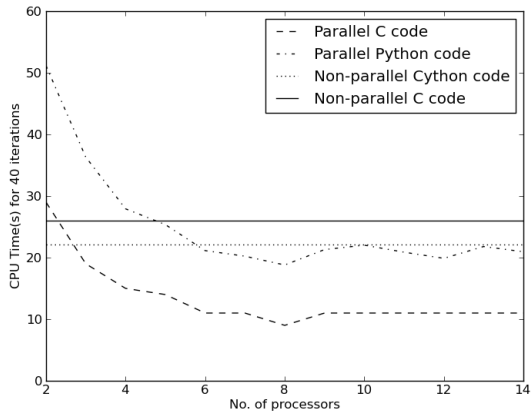
Parallel programming models

- Thread based model
 - Single process creates a number of sub-processes (threads)
 - Threads communicate with each other through global memory
 - OpenMP is used on the subdomain created by MPI based parallelization
- Message Passing model
 - Different sets of task running on different machines
 - Data exchange is through communication
 - Domain decomposition by radially splitting the domain
 - Three grid point interface

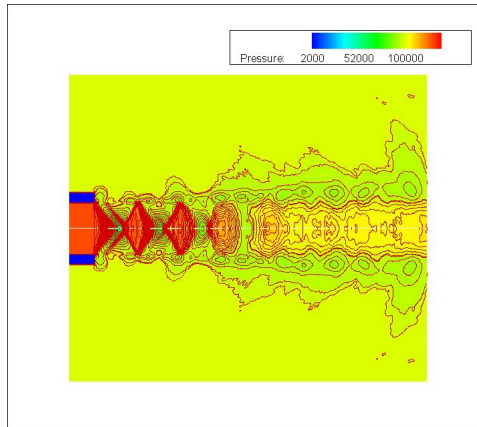
Parallel scheme



Performance results



Pressure contour



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Conclusion

- Python, on optimization with Cython, provides performance equivalent to C code
- Shared and Distributed memory model successfully implemented
- Optimization of inter-processor communication could further enhance performance
- Computational results successfully matched with the experimental results in literature