

An Orthogonal Recursive Bisection (ORB) Based Time Advancement Algorithm for CFD-DEM Solvers

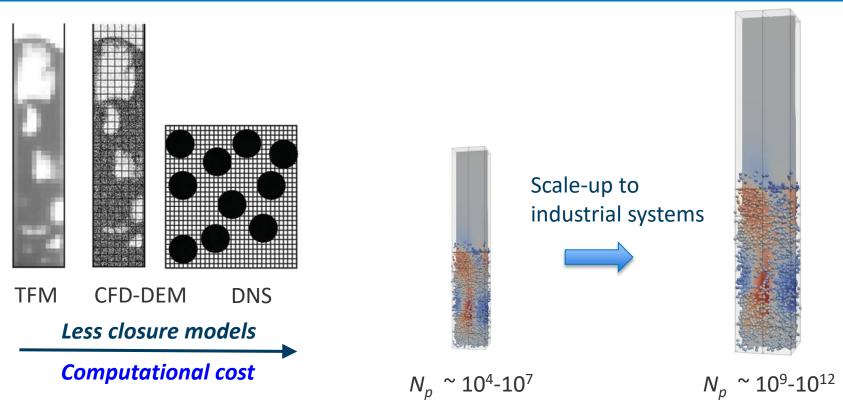
Hari Sitaraman, Ray Grout

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Motivation



- CFD-DEM balance between computational cost and modeling closures
- Goal: Achieve simulations of industry relevant granular flows
- Specifically, increase performance of well-established CFD-DEM solver, MFIX
- MFIX Multiphase Flow with Interface Exchanges
 - Developed at National Energy Technology Laboratory
 - MFiX-Exa being developed as part of DOE Exascale Computing Project

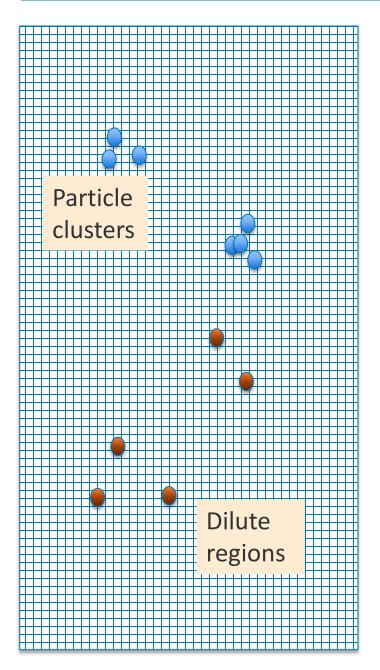
Current approaches for particle advance

• 2nd order explicit-Verlet scheme

•
$$\vec{v}_i^{n+1/2} = \vec{v}_i^{n-1/2} + \Delta t \frac{\vec{F}_i^n}{m_i}$$

- $\vec{x}_i^{n+1} = \vec{x}_i^n + \Delta t \ \vec{v}_i^{n+1/2}$
- Constant time step size for all particles
 - Collisional time scale determined by solid phase properties
- Challenge with constant time step
 - Fluid residence time scales with system dimensions
 - Particle time scale is intrinsic to phase properties
- For large-scale systems computational cost increases
 - More number of particles
 - More number of time steps

Adaptive time stepping - idea



- Particle clustering is common in industrial systems
- Reduce overall computational cost with localized time stepping method
- Adaptive time stepping approach
 - Identify particle clusters
 - Advance particle subsets with local timescale
 - Lower costs for dilute regions
 - Identify dilute/clustered regions?
 - synchronization in a global fluid time step?
 - Collision misses?

Mathematical model and numerical methods

Computational model

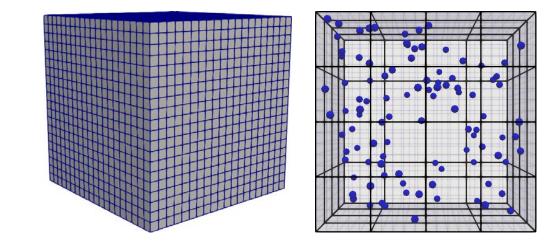
Continuous phase equations solved by SIMPLE/projection schemes

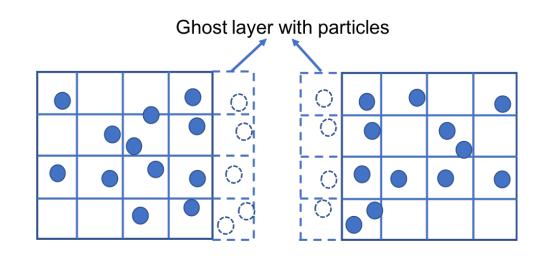
Discrete phase equations solved using 2nd order velocity-verlet scheme

- Continuous to discrete coupling through void fraction and momentum interaction term
- Discrete to continuous coupling through viscous and pressure drag

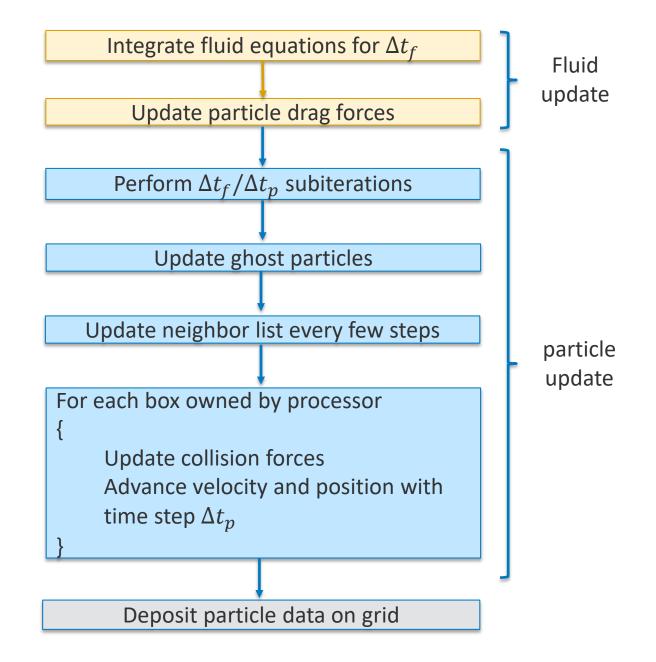
Code base – MFIX-Exa

- Slimmed down version of multiphase code – MFIX
 - developed at LBNL, NETL, NREL and CU, Boulder
 - Test bed for performance optimizations
 - uses adaptive-mesh refinement library, AMReX
 - Decomposition of domain into boxes
 - Boxes distributed among processors
 - load balancing
 - Space filling curve
 - knapsack

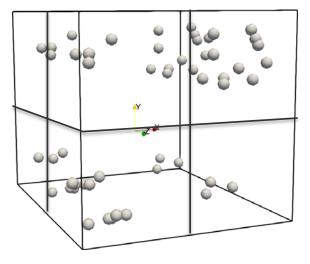




Tasks in a coupled time step

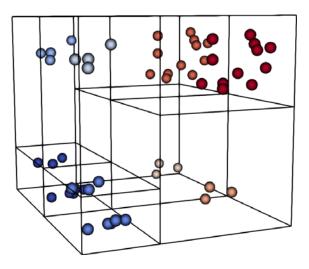


Orthogonal Recursive Bisection (ORB) method



Classic explicit scheme

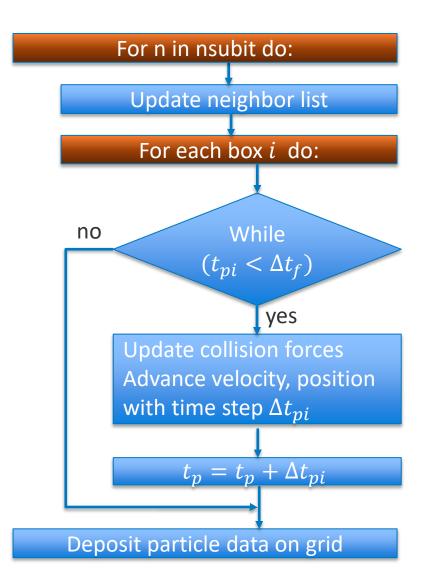
- Find global minimum time step, dt_{min}
- N_subit = dt/dt_{min}
- Update neighbor data
- Perform N_subit sub iterations
 - Loop over each box
 - Compute forces
 - Advance using explicit scheme



Explicit ORB scheme

- Build the ORB tree
- Update neighbor data
- Loop over each box
 - Find minimum time step, dt_{min}
 - N_subit=dt/dt_{min} (can be 1!!)
 - Perform N_subit sub iterations
 - Compute forces
 - Advance using explicit scheme

Reduce neighbor exchange errors – strategy 1



- Advance in smaller timestep chunks
 - Particle update to fluid time level happens in a few iterations of smaller time steps
 - This will increase number of neighbor updates
 - More number of updates will increase accuracy
 - Algorithm tends to constant global timestep method for large number of sub-iterations

Reduce neighbor exchange errors – strategy 2

 Δt_{ps} = user defined sub time step

While $(t_{pi} < \Delta t_f)$

First pass

- Redistribute, Update neighbor list
- Do adaptive time stepping for time Δt_{ps}
- Store particle position and velocity

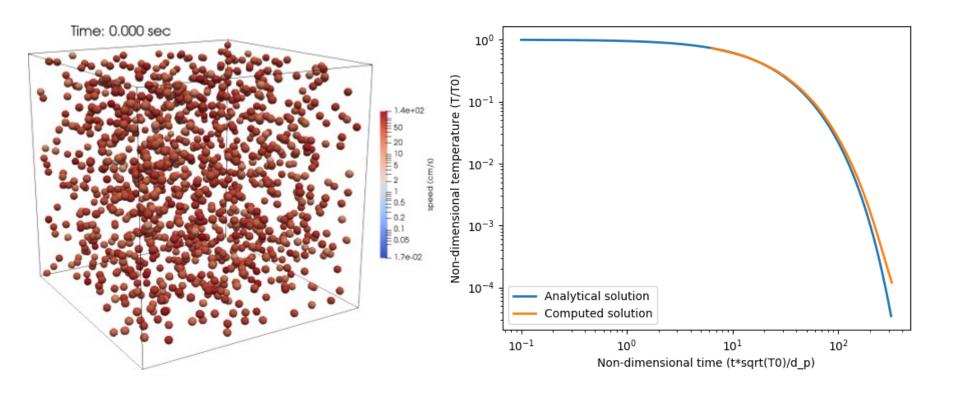
second pass

- Repeat 2 times
 - Redistribute, Update neighbor list
 - Do adaptive time stepping for time $\Delta t_{ps}/2$
- Compute error between first and second pass
- $t_{pi} = t_{pi} + \Delta t_{ps}$
- If error is large $\Delta t_{ps} = \Delta t_{ps}/2$

- Two-pass error correction method
 - Adaptively change particle update time step based on an error metric
- Advantages
 - Detect time stepping failure
 - Reduces neighbor update MPI communication events
 - Temporal locality in Cache
 - Repeated usage of memory
- Disadvantage
 - More floating point operations

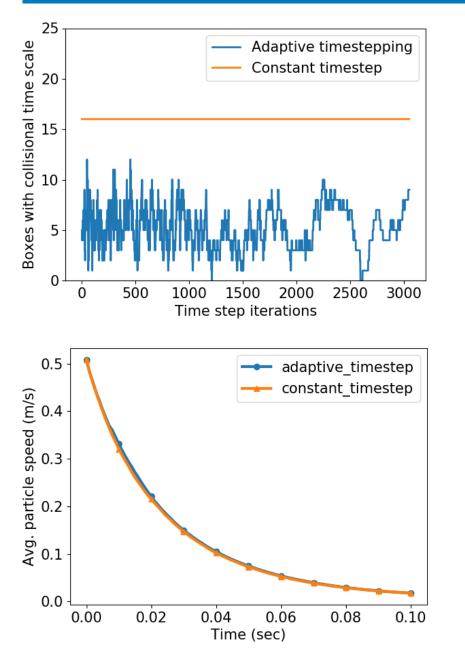
Results

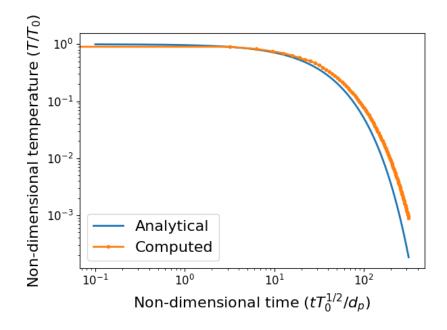
Test case 1 – Homogenous cooling system (HCS)



- Decay of total particle energy with time
 - Viscous and collisional losses
- Initial conditions random velocity and position distribution
- Analytic solution Haff's law¹ predicts decay of non-dimensional temperature

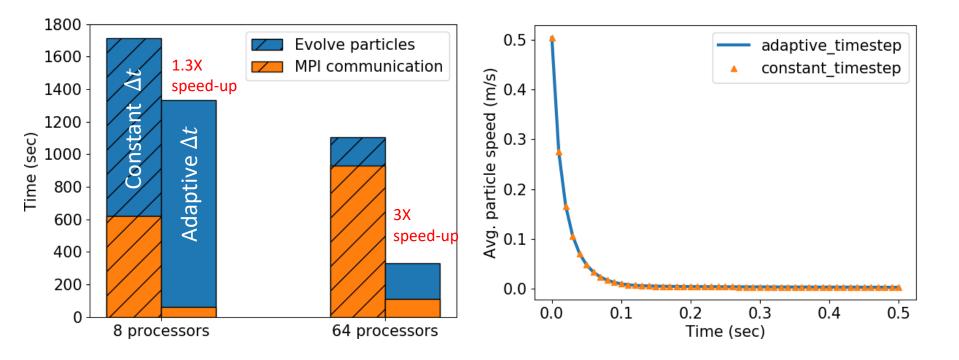
Test case 1 – Homogenous cooling system (HCS)





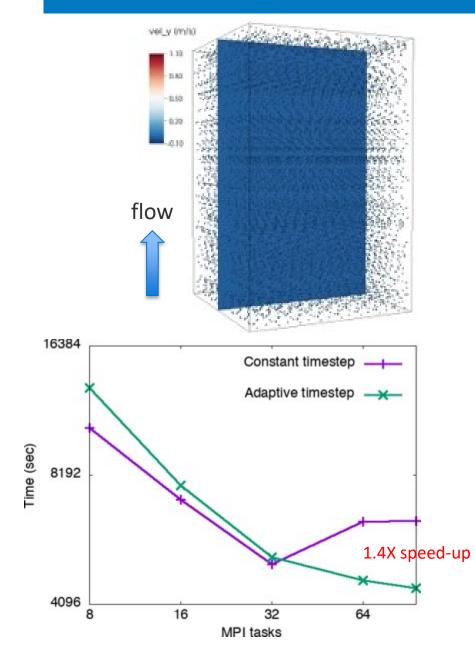
- 300 particles, 16 ORB leaves, 8000 cells
- Single processor run
- Initial guess for adaptive timestep = $10 \ \mu s$
- Collisional time step = 0.3 μs
- Speed-up of 1.6x obtained with adaptive time stepping

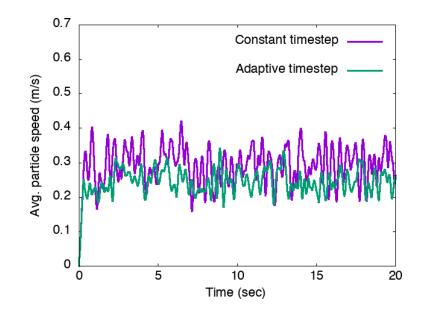
Test case 1 – Homogenous cooling system (HCS)



- Multi-processor HCS run with 40,000 particles with 128 ORB leaves
- Improvements are more pronounced at larger processor counts
 - Less MPI communications with respect to neighbor communication
 - ~ 3X improvement seen with 64 processors
- Retrieves identical solution with respect to constant timestep case

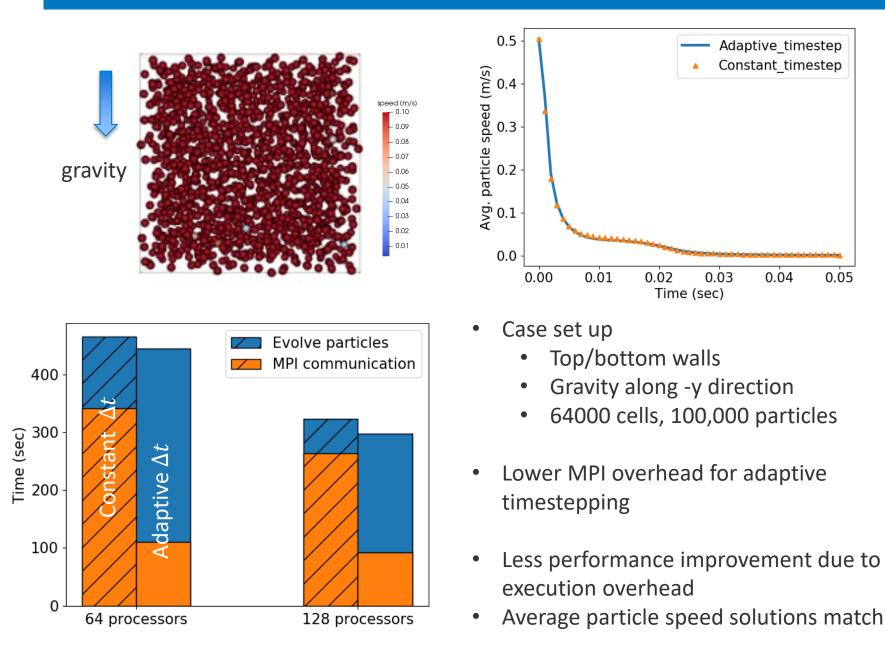
Test case 2 – Riser flow





- Case set up
 - Lateral wall boundaries
 - Constant y pressure gradient
 - 51200 cells, 14000 particles
- Better strong scaling observed with adaptive timestep
 - Less MPI communications
- Average speed solutions are very similar

Test case 3 – Settling due to gravity



Conclusions and future work

- Conclusions
 - Developed an adaptive time stepping algorithm
 - Using Orthogonal recursive bisection
 - Local time steps for subsets of particles at the ORB leaves
 - Two-pass error correction method to reduce collision misses
 - Performance improvement
 - Significant performance improvement for parallel cases
 - Reduces MPI communication overheads
- Future work
 - Other decomposition methods
 - K-means clustering
 - Currently studied 3 canonical DEM cases
 - Application to realistic systems
 - What is the correct error tolerance for different DEM systems?

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