

Development of Scalable Parallel Implicit SPH using LAMMPS and Trilinos

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LAMMPS/Trilinos Integration

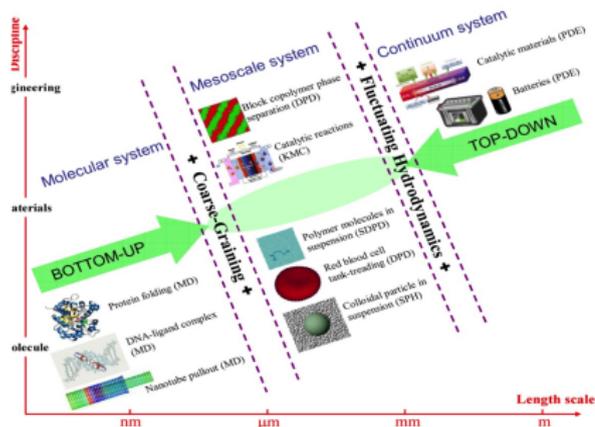
Smoothed Particle Hydrodynamics

Implementation of Navier Stokes Equations

Numerical Examples

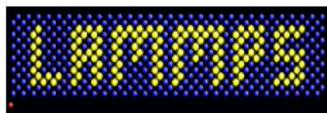
Conclusion

LAMMPS/Trilinos Integration



Research focuses of CM4

- Funded by ASCR MMICC, DOE.
- Developing particle- and grid-based methods for mesoscale material processes.
- Concurrent coupling of these methods.
- Exploring fast solution techniques for exascale computing.
- Integrating mathematical and computational models for applications relevant to synthesis of new materials.



Goal

- Develop large scale parallel 3D implicit simulation capability.
- Use LAMMPS, Sandia's massively parallel molecular dynamics code.
 - LAMMPS is a classical molecular dynamics code.
 - LAMMPS can simulate any particle system *e.g.*, MD, SPH, DPD, *etc.*
 - Provides modular framework easy to add new capabilities.
 - Demonstrated massively parallel scalability via MPI and spatial domain decomposition.

Problem

- LAMMPS has no capability for implicit time integration.
 - Only explicit time integration used in MD.
 - Need distributed memory parallel linear algebra infrastructure: *e.g.*, vectors, matrices, linear solvers, preconditioners, *etc.*

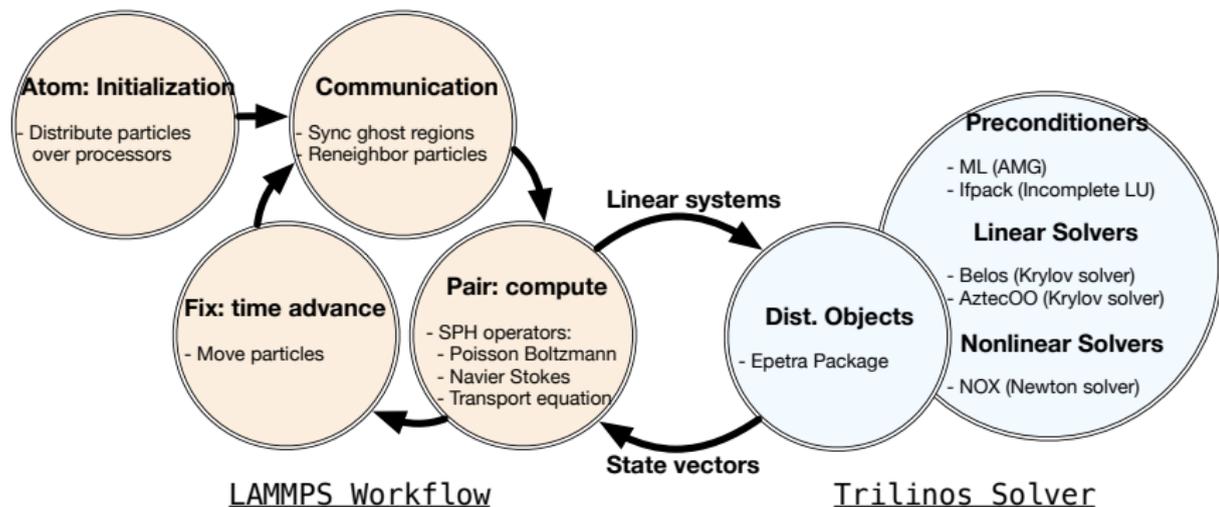
Solution

- Integrate LAMMPS with Trilinos solver packages.

- Open source C++ software framework for solving large scale multi-physics scientific and engineering problems: <https://trilinos.org>.
- Developed and maintained by Sandia National Labs.
- Trilinos is made of packages:
 - The current Trilinos library consists of more than 50 packages.
 - Each package is an independent piece of software but inter-operates with other packages.
 - Use a set of packages as needed, like LEGO blocks.



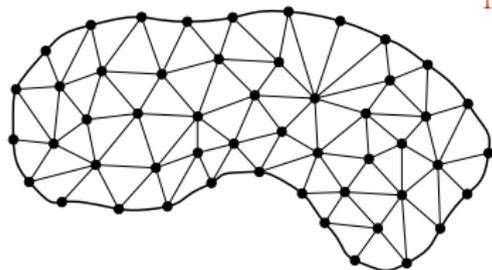
By Alan Chia (Lego Color Bricks) CC BY-SA 2.0 via Wikimedia Commons.



Let each code handle what it was designed to do well

- LAMMPS handles particle data, parallel data distribution, ghosting.
- Trilinos handles distributed memory linear solvers, preconditioners, *etc.*
- Developed implicit solver and time integration framework can be applied to any particle based models in LAMMPS.

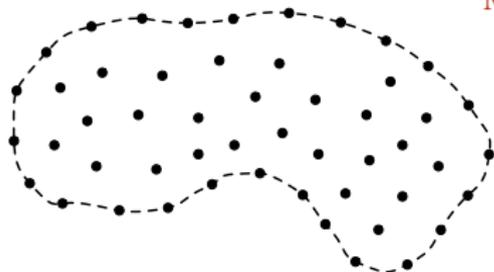
Smoothed Particle Hydrodynamics



Mesh: a list of points with their connectivities.

Motivation of meshfree methods

- Generating a suitable mesh is a challenging task.
- Easier to handle large deformation, moving boundary and fluid structure interaction problems than grid-based approaches.
- By advecting points in Lagrangian form, the non-linear advection term in Navier Stokes equations can be removed.

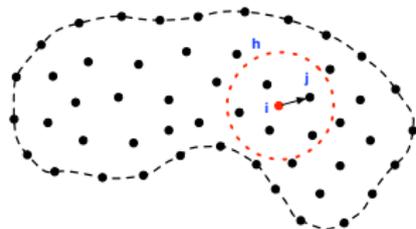
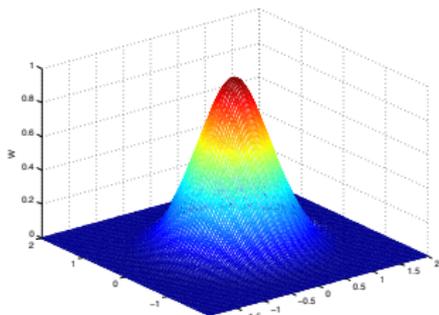


Meshfree: points are scattered on the domain.

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Smoothed Particle Hydrodynamics (SPH) Interpolation



Begin with a trivial identity

$$f(x) = \int f(x')\delta(x-x')dx'$$

Consider integral interpolants with a compact support characterized by h :

$$f(x) = \int f(x')W(x-x',h)dx' \quad \rightarrow \quad \langle f(x_i) \rangle = \sum_j^N f(x_j)W(x_i-x_j,h)V_j.$$

W is an interpolating kernel with these properties:

$$\int W(u,h)du = 1 \quad \text{and} \quad \lim_{h \rightarrow 0} W(u,h) = \delta(u).$$

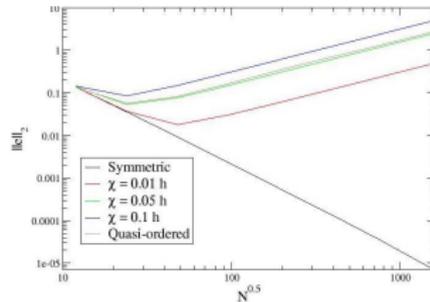
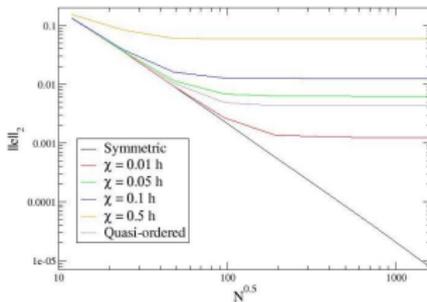
Second Order SPH Discretization

Standard SPH operators are defined as:

$$\nabla_0 f_i = \sum_j^N (f_j - f_i) \nabla_{x_i} W_{ij} V_j$$

$$\nabla_0^2 f_i = 2 \sum_j^N \frac{f_i - f_j}{r_{ij}} \mathbf{e}_{ij} \cdot \nabla_{x_i} W_{ij} V_j.$$

These operators lack 1st order consistency.



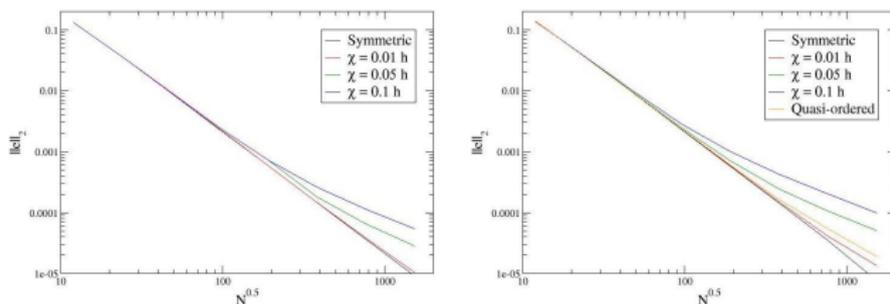
L^2 error for uncorrected gradient and Laplacian operators; χ is random perturbation applied to particles.

The “corrected” SPH scheme uses correction tensors to obtain 1st order consistency:

$$\nabla_1 f_i = \sum_j^N (f_j - f_i) \mathbf{G}_i \nabla_{x_i} W_{ij} V_j,$$

$$\nabla_1^2 f_i = 2 \sum_j^N (\mathbf{L}_i : \mathbf{e}_{ij} \otimes \nabla_{x_i} W_{ij}) \left(\frac{f_i - f_j}{r_{ij}} \mathbf{e}_{ij} \cdot \nabla_1 f_i \right) V_j,$$

where the correction tensors \mathbf{G} and \mathbf{L} are derived from a Taylor expansion².



L^2 error for “corrected” gradient and Laplacian operators; χ is random perturbation applied to particles.

Implementation of Navier Stokes Equations

Consider an incompressible flow governed by the Navier-Stokes (NS) equations:

$$\frac{d\mathbf{u}}{dt} = -\frac{1}{\rho}\nabla p + \nu\nabla^2\mathbf{u} + \mathbf{g},$$

$$\nabla \cdot \mathbf{u} = 0,$$

where \mathbf{g} is a body force. Splitting the equations into prediction/correction steps, we get:

$$\text{Helmholtz} \begin{cases} \frac{\mathbf{u}^* - \mathbf{u}^n}{\Delta t} = -\frac{1}{\rho}\nabla p^n + \nu\nabla^2\left(\frac{\mathbf{u}^* + \mathbf{u}^n}{2}\right) + \mathbf{g} & \mathbf{x} \in \Omega, \\ \mathbf{u}^* = \mathbf{u}_{\partial\Omega} & \mathbf{x} \in \partial\Omega, \end{cases}$$

$$\text{Corrector} \begin{cases} \frac{\mathbf{u}^{n+1} - \mathbf{u}^*}{\Delta t} = -\frac{1}{\rho}\nabla(p^{n+1} - p^n) & \mathbf{x} \in \Omega, \\ \nabla \cdot \mathbf{u}^{n+1} = 0 & \mathbf{x} \in \Omega, \\ \mathbf{u}^{n+1} \cdot \mathbf{n} = \mathbf{u}_{\partial\Omega} \cdot \mathbf{n} & \mathbf{x} \in \partial\Omega. \end{cases}$$

Splitting the equations into prediction/correction steps, we get:

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 \end{aligned}$$

By taking the divergence of the second set of equations, we obtain the Poisson problem for the pressure difference:

$$\text{Poisson} \begin{cases} -\frac{1}{\rho} \nabla^2 (p^{n+1} - p^n) = -\frac{\nabla \cdot \mathbf{u}^*}{\Delta t} & \mathbf{x} \in \Omega, \\ \nabla (p^{n+1} - p^n) \cdot \mathbf{n} = 0 & \mathbf{x} \in \partial\Omega. \end{cases}$$

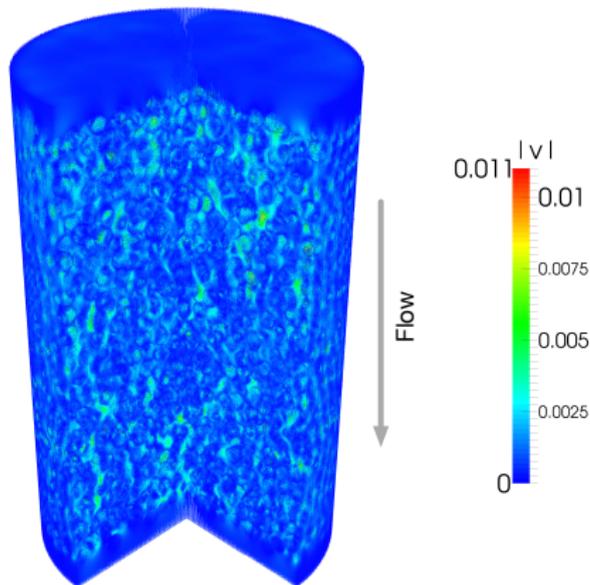
Resulting systems of equations are solved by a preconditioned (algebraic multigrid) GMRES solver.

Numerical Examples

- 3D Complex geometry: Pore-scale Flow in Bead Pack.

Benchmark: 3D Pore-scale Flow in Bead Pack³

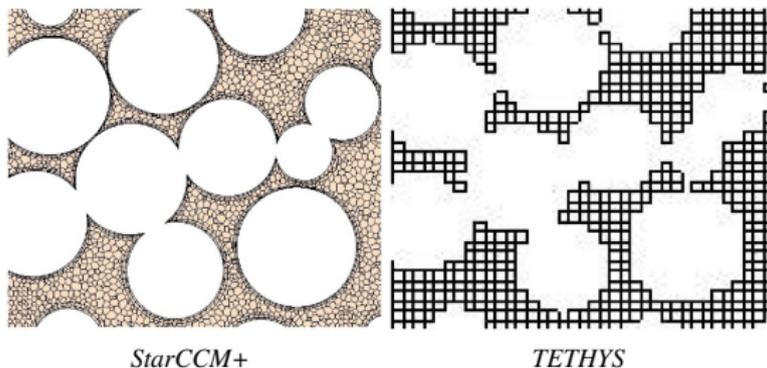
The pore geometry is constructed from voxel data provided by MRI measurements.



Steady-state solution of the flow in a bead pack.

Parameter	Symbol	Value
Bead diameter	d (mm)	0.5
# of Beads	-	6864
Column diameter	D (mm)	8.8
Column length	L (mm)	12.8
Porosity	ϵ	0.4267
Volumetric flow rate	Q (kg/s)	$2.771e-5$
Fluid density	ρ (kg/m ³)	997.561
Fluid dynamic viscosity	μ (pa-s)	$8.887e-4$

[2] Yang, et. al., *Intercomparison of 3D Pore-scale Flow and Solute Transport Simulation Methods*, *Advances in Water Resources*, in review.

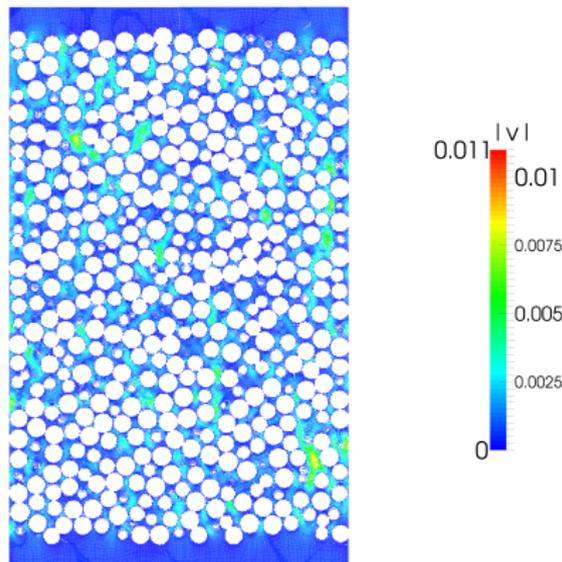


Computational cost of different methods

Code	Mesh	40 [μm]		20 [μm]		Description
		Time	Machine	Time	Machine	
StarCCM+	Tet	15 hrs	4 cpus	-	-	Finite Volume, CD-adapco
TETHYS	Hex	4 hrs	480 cpus	9 hrs	1600 cpus	Finite Volume, PNNL
iRMB-LBM	Hex	4.5 hrs	1 gpu K40c	61.07 hrs	2 gpus K40c	Lattice Boltzmann Tech. Univ. Braunschweig
ISPH	-	0.17 hrs	960 cpus	0.21 hrs	7680 cpus	SPH, SNL

Pressure drop along the axial direction

Code	Resolution	ΔP [Pa]	Diff [%]
Reference ⁴	-	14.29	-
StarCCM+	40 μm	13.61	4.48
ISPH	40 μm	13.26	4.76
TETHYS	40 μm	13.32	6.79
TETHYS	20 μm	13.19	7.70
iRMB-LBM	40 μm	15.20	6.37
iRMB-LBM	20 μm	16.26	13.79

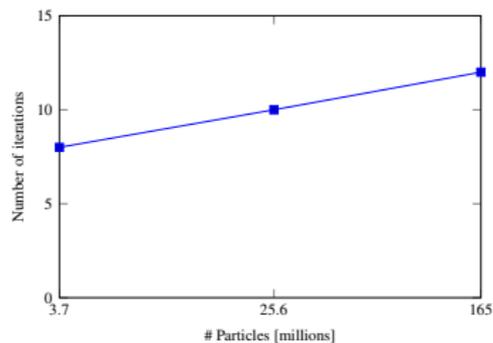
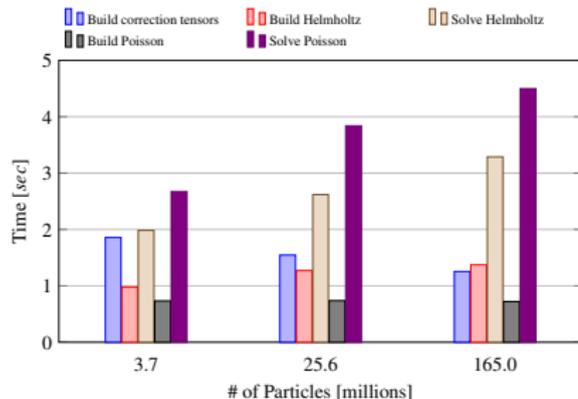


Velocity profile of a vertical cross-section.

[3] B.Eisfeld and K.Schnitzlein, *The influence of confining walls on the pressure drop in packed beds*, *Chemical Engineering Science*, 2001.

$\approx 30k$ Particles per processor

- In theory, AMG convergence factor is independent of the problem size.
- Here, we observe the # of iterations grows moderately with respect to the # of DOFs.



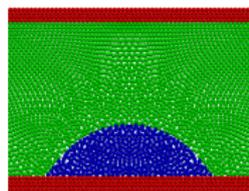
Conclusion

- Demonstrated scalable parallel Implicit SPH method.
- With local correction operators, our ISPH method delivers efficient and accurate solutions that are comparable to other numerical methods.
- Implicit time integration allows to use a large time step.
- Trilinos interface can be applied to problems arising from any particle-based models in LAMMPS.

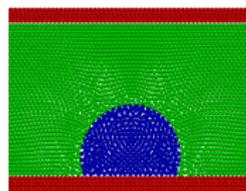
- **Discretizations:**
 - implemented second order SPH;
 - implemented MLS with arbitrary order of approximation and ALE scheme.
- **Highly scalable parallel code:**
 - demonstrated the weak scalability up to 134 million particles with 32k cores;
 - applied the implicit SPH method to solve a real problem which demands highly intensive computation.
- **Multi-physics capabilities:**
 - provide capability to solve electro-kinetic flows coupled with the Poisson Boltzmann equation.
- **Boundary conditions:**
 - Morris mirroring technique with Holmes modification for Dirichlet BCs;
 - continuous boundary force method proposed by *Pan et al.* for Robin BCs;
 - partial slip boundary (Robin) condition with no-penetration (Dirichlet) on normal directions;

On-going and Future work

- **Multi-phase flow:** Continuum Surface Force (CSF) and Pairwise Force (PF) model.

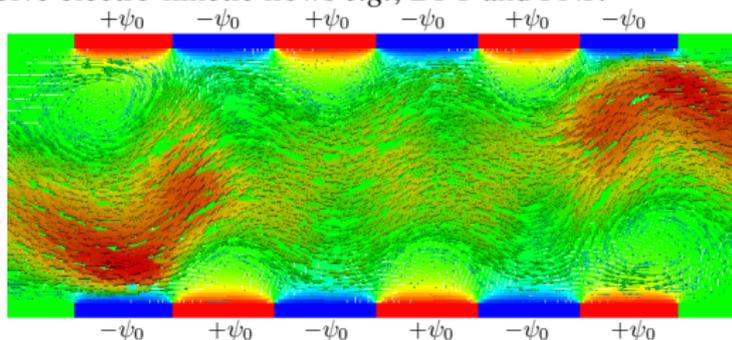


$\theta=60$



$\theta=120$

- **Multi-physics capabilities:** adding improved physics description and coupling strategy to solve electro-kinetic flows *e.g.*, DFT and PNP.



Microfluidic mixing channel with alternating electric potentials on walls.

This code is a research code and we look for more collaborations for interesting application problems.

Acknowledgement

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