



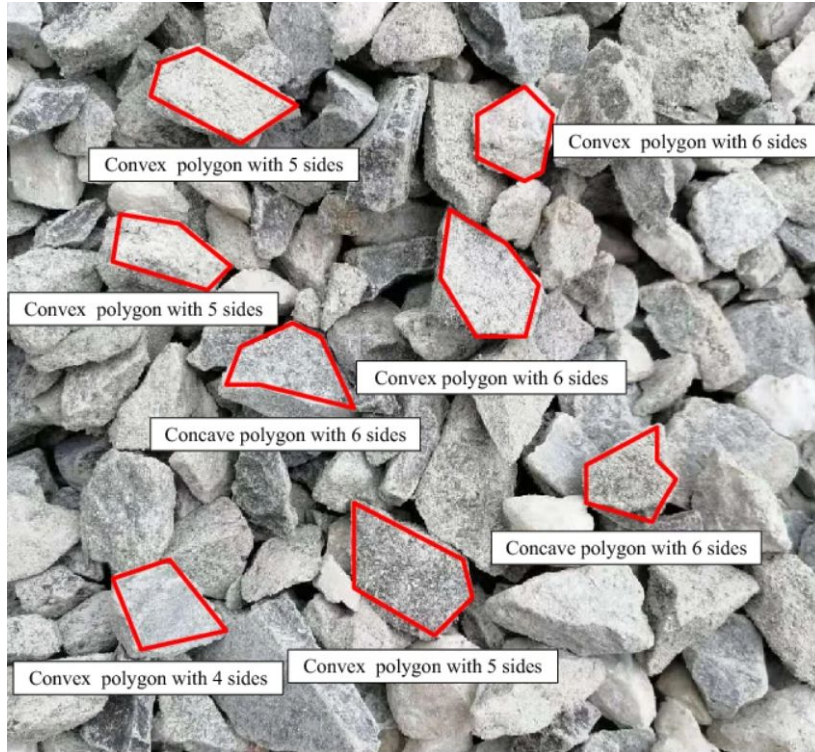
Direct numerical simulation of flow past randomly distributed Platonic polyhedrons

Aashish Goyal, Guodong Gai, Zihao Cheng, and Anthony Wachs

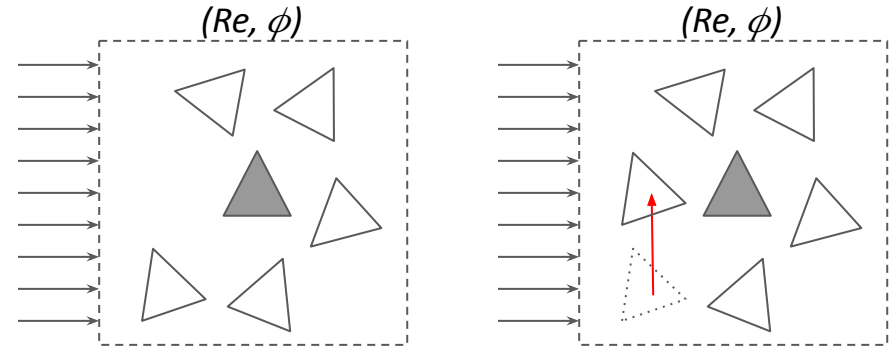
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Vancouver, BC, Canada

*2024 Multiphase Flow Science Workshop, Morgantown, WV, United States
August 13–14, 2024*

Background



Chen et al., Nature Scientific Reports, 2023



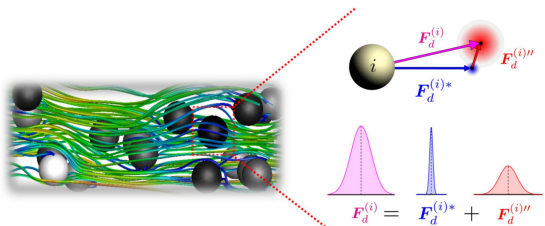
$$\mathbf{F}_i = \langle \mathbf{F}_i \rangle (Re, \phi, \kappa) + \Delta \mathbf{F}_i (Re, \phi, \kappa, \{\mathbf{r}_1, \dots, \mathbf{r}_M\})$$

Tenneti (2011)
Beetstra (2007)
Tang (2015)
Bogner (2015)

Sanjeevi and
Padding 2020

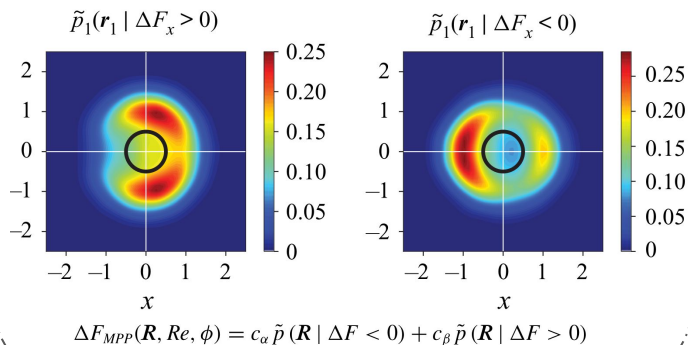
Fluctuating term models

Stochastic model

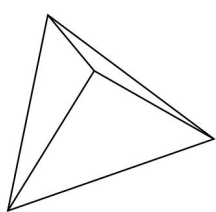
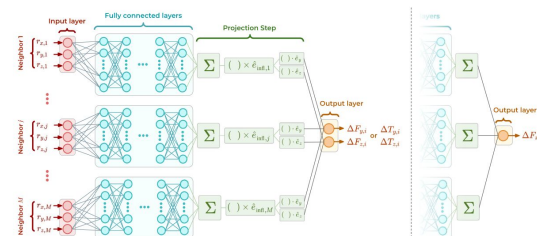


Lattanzi et al. (2022)

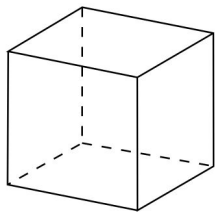
Probability driven model



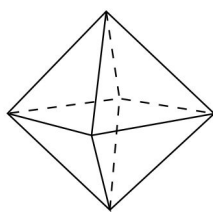
NN (FNN, PINN, CNN, GNN)



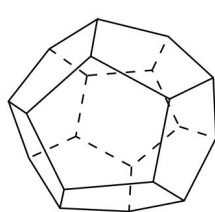
(a)



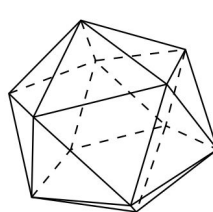
(b)



(c)



(d)



(e)

Capture particle DOFs

Accurate

Computationally expensive

Non-scalable on $O(1000)$ cores

Five Platonic polyhedrons with increasing sphericity κ . (a) Tetrahedron, $\kappa = 0.670$, (b) Cube, $\kappa = 0.806$, (c) Octahedron, $\kappa = 0.846$, (d) Dodecahedron, $\kappa = 0.910$, and (e) Icosahedron, $\kappa = 0.940$.

Particle-laden flow solver

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

$$\rho \frac{\partial \mathbf{u}}{\partial t} + \rho \mathbf{u} \cdot \nabla \mathbf{u} = -\nabla p + \mu \nabla^2 \mathbf{u} + \mathbf{f}_b$$

+

$$\nabla^2 \approx -(1 - \partial_{xx})(1 - \partial_{yy})(1 - \partial_{zz})$$

Converts one 3D problem into three 1D problems !!!

Step 1

Step 2

Step 3

Step 4

$$p^{*,n+\frac{1}{2}} = p^{n-\frac{1}{2}} + \phi^{n-\frac{1}{2}}$$

Pressure prediction

$$\rho \frac{\xi^{n+1} - \mathbf{u}^n}{\Delta t} - \mu \nabla^2 \mathbf{u}^n + \nabla p^{*,n+\frac{1}{2}} + \rho \mathbf{u}^n \cdot \nabla \mathbf{u}^n = \mathbf{f}_b$$

$$\rho \frac{\eta^{n+1} - \xi^{n+1}}{\Delta t} - \frac{\mu}{2} \partial_{xx}(\eta^{n+1} - \eta^n) = 0$$

$$\rho \frac{\zeta^{n+1} - \eta^{n+1}}{\Delta t} - \frac{\mu}{2} \partial_{yy}(\zeta^{n+1} - \zeta^n) = 0$$

$$\rho \frac{\mathbf{u}^{n+1} - \zeta^{n+1}}{\Delta t} - \frac{\mu}{2} \partial_{zz}(\mathbf{u}^{n+1} - \mathbf{u}^n) = 0$$

Velocity prediction

$$\theta - \partial_{xx} \theta = -\frac{1}{\Delta t} \nabla \cdot \mathbf{u}^{n+1}$$

$$\psi - \partial_{yy} \psi = \theta$$

$$\phi^{n+\frac{1}{2}} - \partial_{zz} \phi^{n+\frac{1}{2}} = \psi$$

Penalty step

$$p^{n+\frac{1}{2}} = p^{n-\frac{1}{2}} + \phi^{n+\frac{1}{2}} - \frac{\mu}{2} \nabla \cdot (\mathbf{u}^{n+1} + \mathbf{u}^n)$$

Pressure correction

Fixed rigid bodies in solver

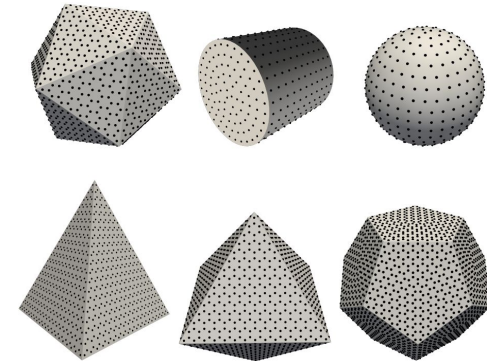
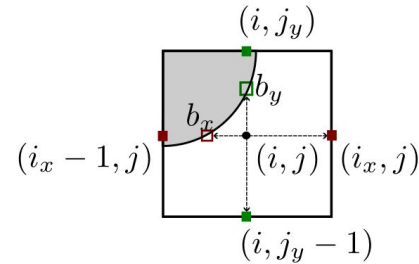
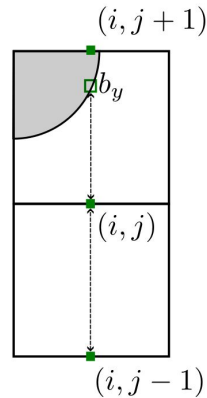
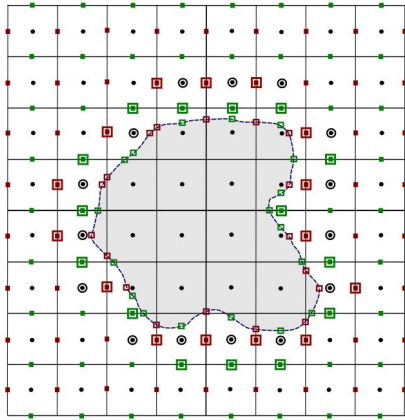
Modifications for
particle-laden simulations

Nodes presence in
particles or fluid

Diffusion stencil

Divergence stencil

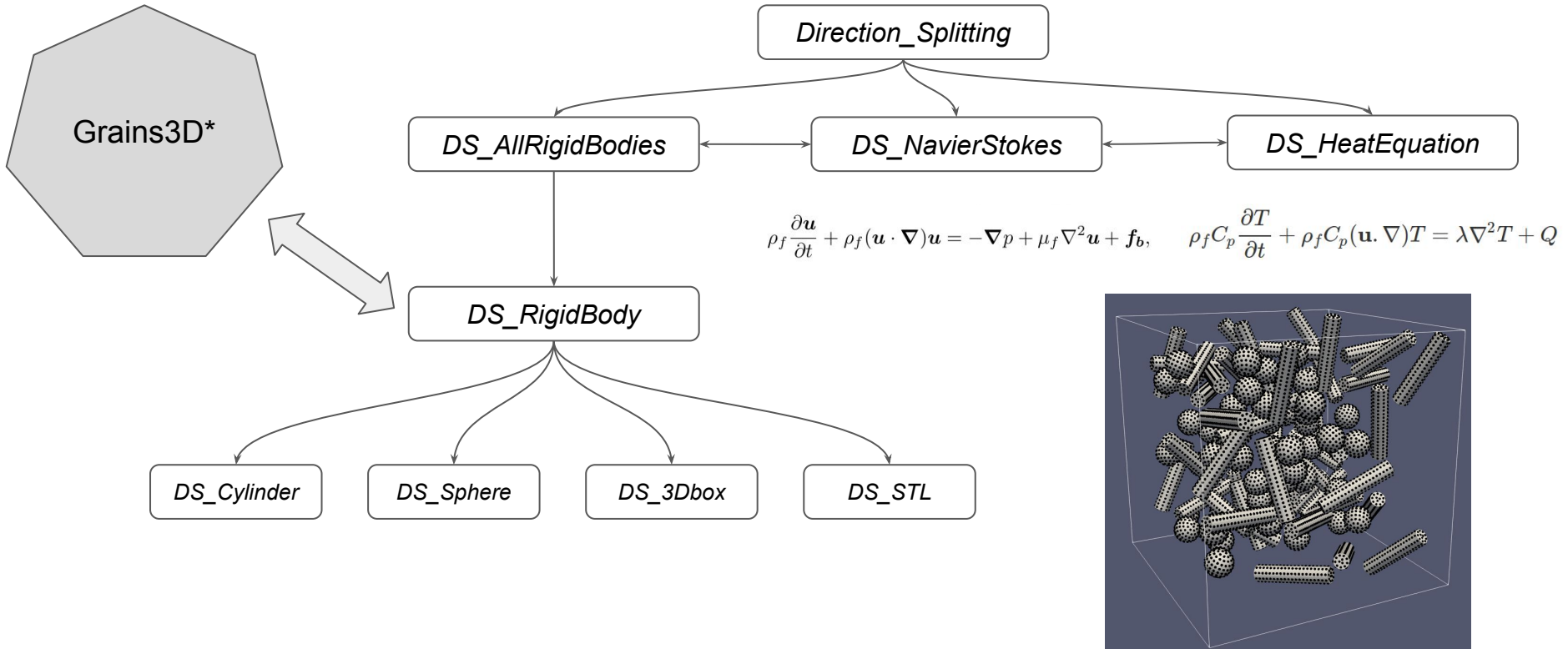
Surface stress



These methods can work for any particle shape.

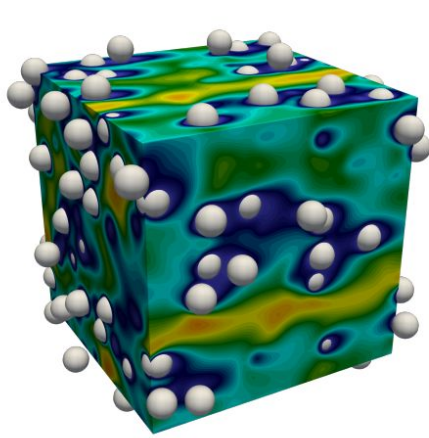
Goyal and Wachs (2023, 2024)

Overall Framework in PacFiC

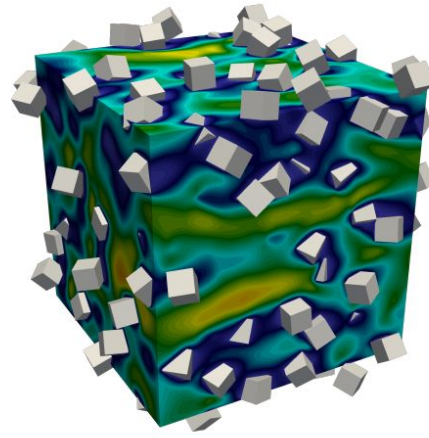
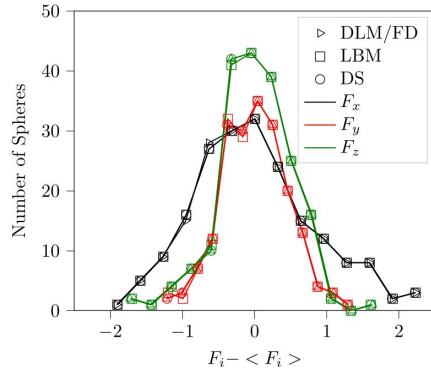


* A. Wachs et al. (2012)

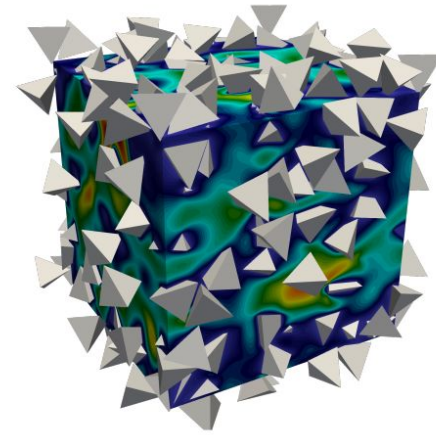
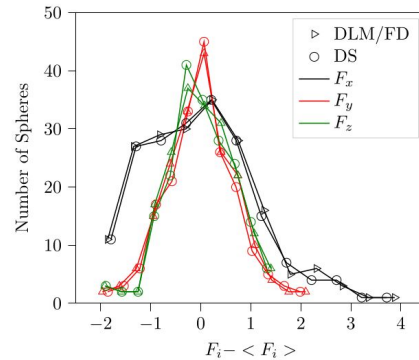
Validation: Random non-spherical rigid bodies



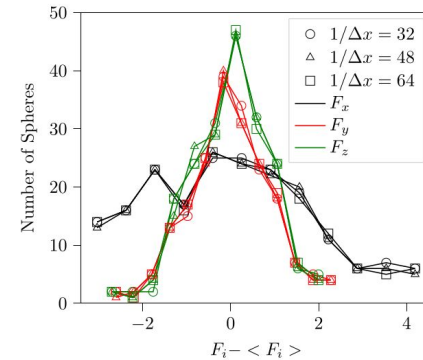
(a) Spheres



(b) Cubes



(c) Tetrahedrons



Data generation

Summary of parameters used for PR-DNS of a random array of each Platonic polyhedron at $Re = 1, 10, 100$.

ϕ	L	N_{RB}	$1/\Delta\tilde{x}$	FV grid cells
0.05	$23D_{eq}$	1146	32	$\sim 399 \times 10^6$
0.10	$23D_{eq}$	2292	32	$\sim 399 \times 10^6$
0.20	$18D_{eq}$	2228	40	$\sim 373 \times 10^6$



(a) Tetrahedrons



(b) Cubes



(c) Octahedrons



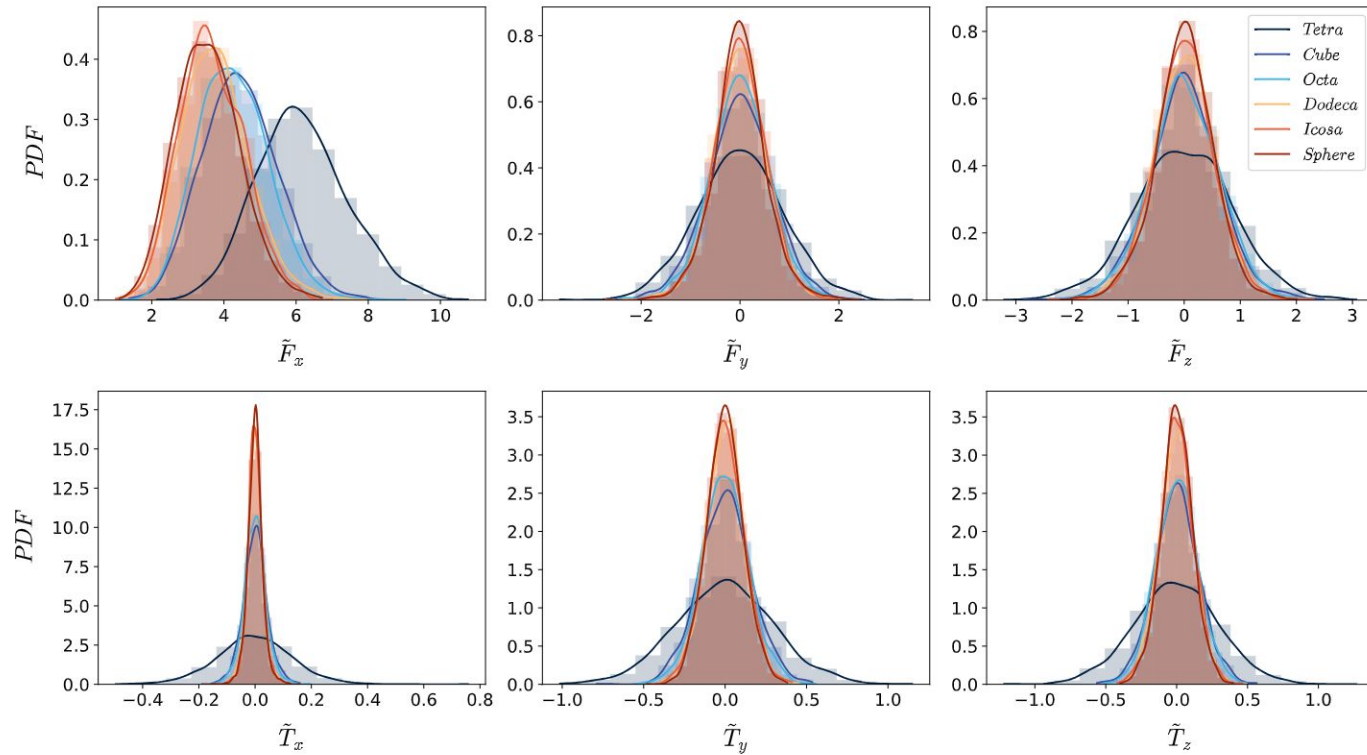
(d) Dodecahedrons



(e) Icosahedrons

Spatial distribution of Platonic polyhedrons in a tri-periodic box at $\phi = 0.2$.

Force and torque distributions

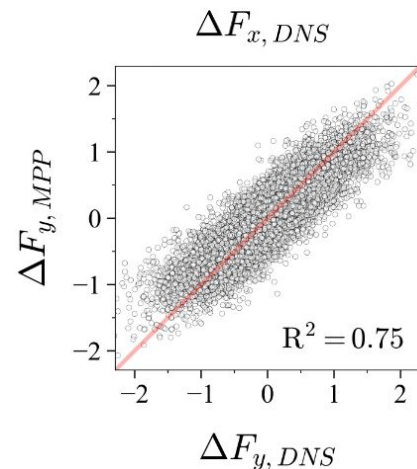
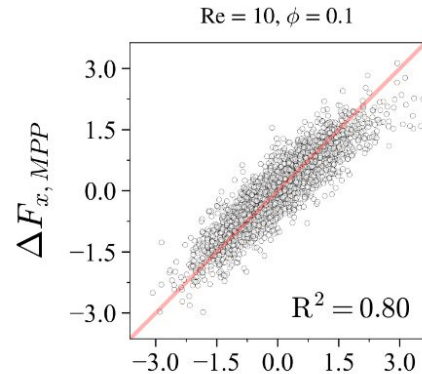
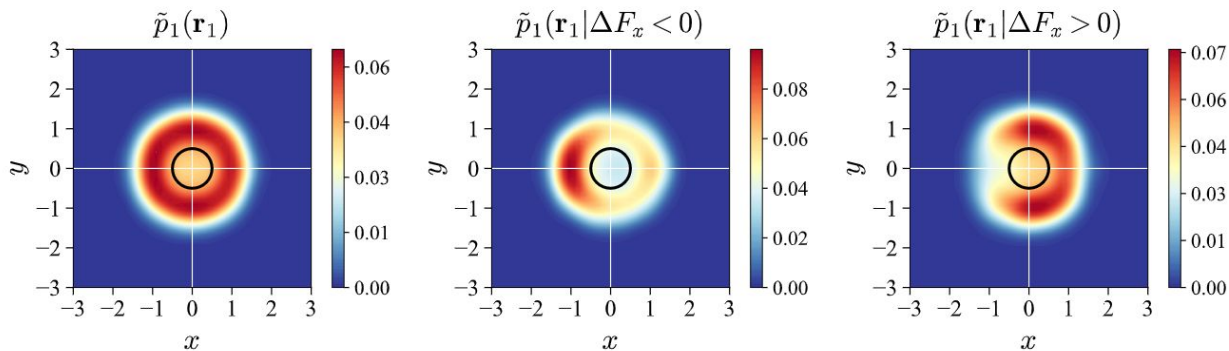


Distribution of normalized force and torque for flow past Platonic polyhedrons at $Re = 10$ and $\phi = 0.1$.

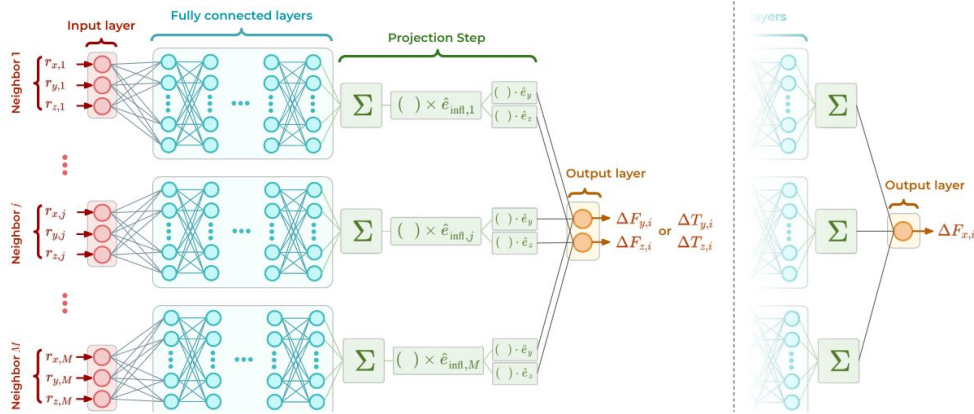
Microstructure-based Probability-driven Point-particle model

$$\Delta F_{MPP}(\mathbf{R}, Re, \phi) = c_\alpha \tilde{p}(\mathbf{R} | \Delta F < 0) + c_\beta \tilde{p}(\mathbf{R} | \Delta F > 0)$$

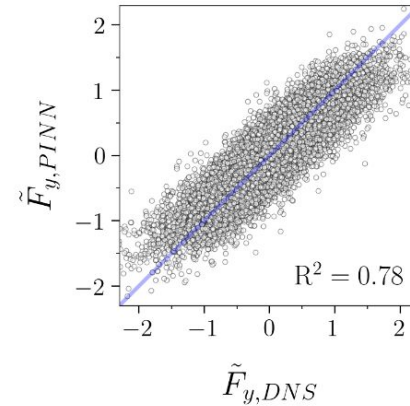
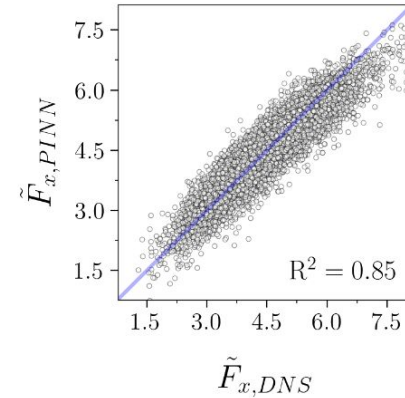
$$\Delta F_{MPP}(\mathbf{R}, Re, \phi) = \sum_{j=1}^M c_{\alpha,j} \tilde{p}_j(\mathbf{r}_j | \Delta F < 0) + \sum_{j=1}^M c_{\beta,j} \tilde{p}_j(\mathbf{r}_j | \Delta F > 0)$$



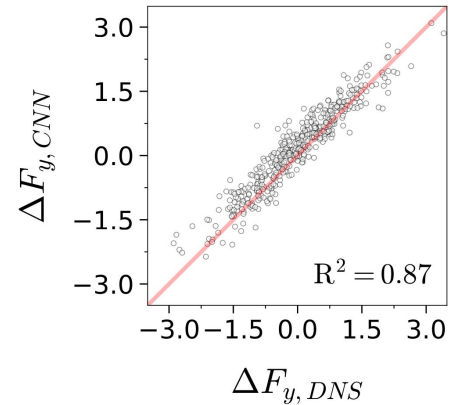
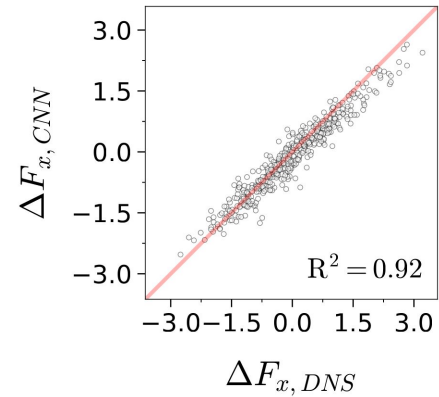
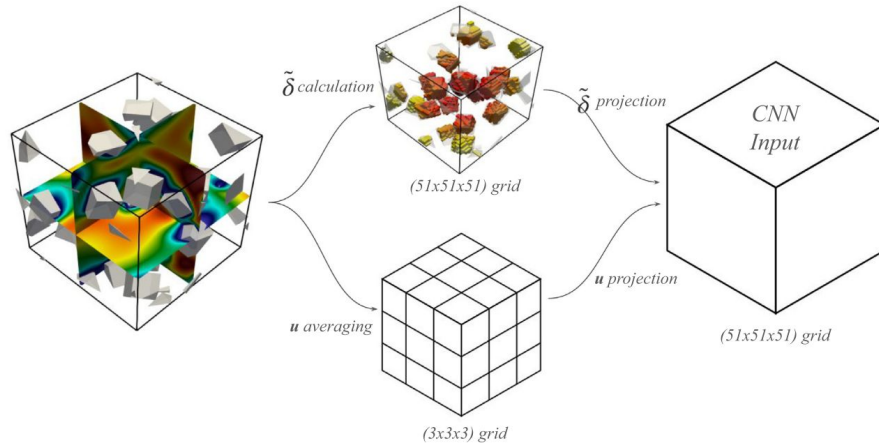
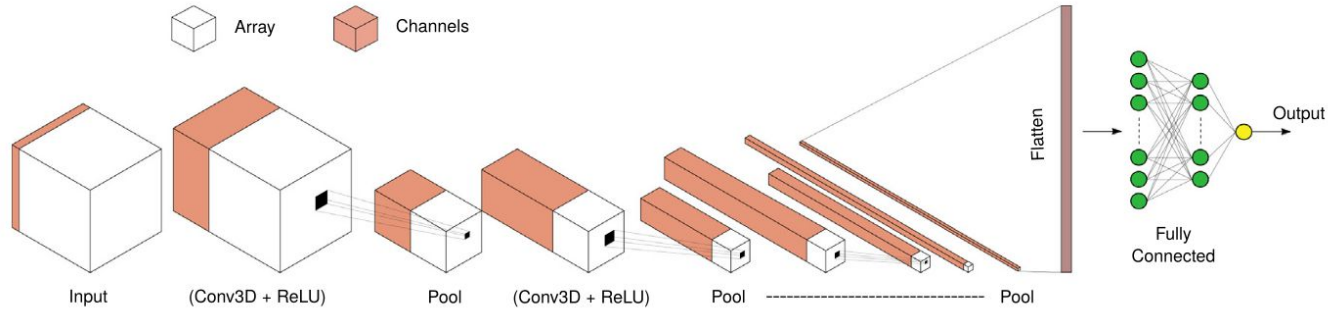
Physics Informed Neural Network (PINN)



- Each NN block handles one neighbor with three input parameters.
- All NN blocks share the same set of parameters.



Convolutional Neural Network



PINN vs. CNN

PINN

ϕ	Re	Tetrahedron			Cube			Octahedron			Dodecahedron			Icosahedron		
		\tilde{F}_x	\tilde{F}_L	\tilde{T}_L	\tilde{F}_x	\tilde{F}_L	\tilde{T}_L	\tilde{F}_x	\tilde{F}_L	\tilde{T}_L	\tilde{F}_x	\tilde{F}_L	\tilde{T}_L	\tilde{F}_x	\tilde{F}_L	\tilde{T}_L
0.05	1	0.83	0.79	0.81	0.91	0.83	0.89	0.89	0.83	0.90	0.92	0.86	0.91	0.93	0.86	0.91
	10	0.79	0.69	0.61	0.87	0.77	0.81	0.83	0.78	0.82	0.88	0.82	0.84	0.87	0.81	0.84
	100	0.47	0.01	0.03	0.61	0.43	0.07	0.56	0.40	0.15	0.64	0.70	0.41	0.63	0.73	0.60
0.1	1	0.80	0.72	0.76	0.87	0.82	0.89	0.87	0.82	0.89	0.89	0.84	0.90	0.88	0.84	0.91
	10	0.76	0.63	0.60	0.85	0.78	0.85	0.82	0.76	0.84	0.86	0.79	0.86	0.85	0.78	0.88
	100	0.47	0.13	0.08	0.60	0.47	0.16	0.60	0.50	0.19	0.70	0.71	0.57	0.65	0.72	0.67
0.2	1	0.66	0.56	0.60	0.77	0.73	0.86	0.80	0.74	0.87	0.82	0.80	0.89	0.82	0.78	0.90
	10	0.65	0.50	0.47	0.77	0.70	0.80	0.80	0.71	0.81	0.81	0.76	0.86	0.81	0.75	0.87
	100	0.45	0.20	0.08	0.60	0.42	0.21	0.64	0.46	0.20	0.68	0.65	0.52	0.71	0.65	0.60

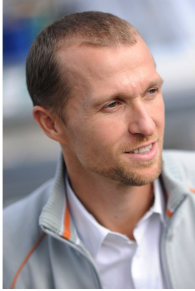
CNN

0.05	1	0.91	0.88	0.85	0.88	0.96	0.95	0.94	0.96	0.94	0.96	0.97	0.97	0.97	0.97	0.93
	10	0.92	0.73	0.66	0.95	0.93	0.87	0.94	0.92	0.87	0.97	0.94	0.90	0.97	0.94	0.82
	100	0.70	0.58	0.40	0.76	0.51	0.34	0.78	0.65	0.14	0.82	0.76	0.34	0.83	0.77	0.34
0.1	1	0.84	0.81	0.77	0.94	0.95	0.92	0.92	0.95	0.94	0.95	0.96	0.96	0.95	0.95	0.96
	10	0.84	0.75	0.63	0.92	0.87	0.86	0.93	0.90	0.90	0.95	0.93	0.91	0.95	0.95	0.92
	100	0.70	0.58	0.25	0.77	0.56	0.08	0.85	0.55	0.10	0.84	0.75	0.44	0.80	0.73	0.57
0.2	1	0.72	0.69	0.59	0.81	0.83	0.87	0.83	0.86	0.84	0.85	0.91	0.93	0.87	0.93	0.94
	10	0.72	0.61	0.52	0.80	0.81	0.81	0.84	0.83	0.77	0.85	0.89	0.88	0.86	0.87	0.90
	100	0.53	0.42	0.25	0.62	0.49	0.24	0.68	0.49	0.14	0.71	0.67	0.43	0.73	0.66	0.55

Relevant Publications

- **Aashish Goyal** and Anthony Wachs, “An accurate and scalable direction-splitting solver for flows laden with non-spherical rigid bodies - Part 1: fixed rigid bodies”, **Communication in Computational Physics** 2023
- Antoine Morente, **Aashish Goyal** and Anthony Wachs, “A Highly Scalable Direction Splitting Solver on Regular Cartesian Grid to Compute Flow in Complex Geometries Described by STL files”, **Fluids** 2023
- **Aashish Goyal** and Anthony Wachs, “An accurate and scalable direction-splitting solver for flows laden with non-spherical rigid bodies - Part 2: moving rigid bodies”, **Computers and Fluids** 2024
- **Aashish Goyal**, Gai Guodong, Zihao Cheng, and Anthony Wachs, “Flow past a random array of statistically homogeneously distributed stationary Platonic polyhedrons: Data analysis, Probability maps, and PINN model”, **International Journal of Multiphase Flows** 2024

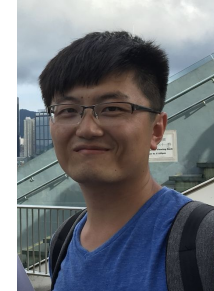
Special Thanks !!!



Dr. Anthony Wachs
UBC Chemical & Biological Engineering
UBC Mathematics



Dr. Zihao Cheng
UBC Mechanical



Dr. Guodong Gai
UBC Mathematics



Thank you !!!

