

A FILTERED COARSE-GRAIN CFD-DEM APPROACH FOR SIMULATING FLUIDIZED PARTICLES

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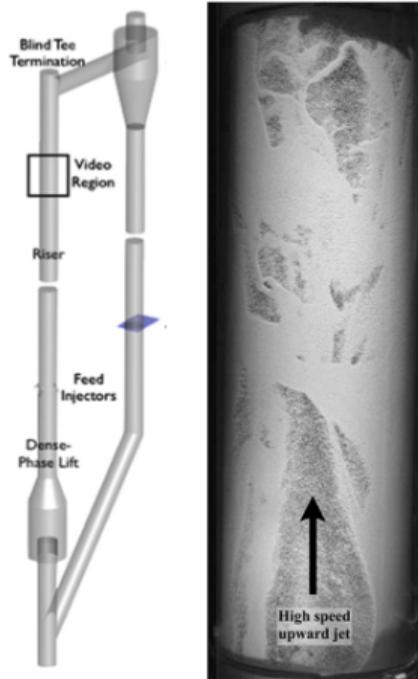
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Introduction

- Circulating fluidized beds (CFB) are one of the prominent gas-solid reactors used in the industry
- Contact between the fluid and particles promotes heat and mass transfer
- Efficiency of CFB reactors heavily influenced by riser hydrodynamics
- **Focus on** developing scalable coarse-grain DEM models for simulation of risers with polydisperse Geldart group-A particles



Schematic representation of CFB¹

¹Shaffer, F, et al., *Powder Technology* (2013)

CFD-DEM gas-phase equations

- Euler–Lagrange (CFD-DEM) approach adopted for many industrial applications
- Volume-filtered Navier–Stokes equations:

$$\frac{\partial}{\partial t} (\varepsilon_f \mathbf{u}_f) + \nabla \cdot (\varepsilon_f \rho_f \mathbf{u}_f) = 0$$

$$\frac{\partial}{\partial t} (\varepsilon_f \rho_f \mathbf{u}_f) + \nabla \cdot (\varepsilon_f \rho_f \mathbf{u}_f \otimes \mathbf{u}_f) = \nabla \cdot (\boldsymbol{\tau} - \mathbf{R}_u) + \varepsilon_f \rho_f \mathbf{g} + \mathbf{F}^{\text{inter}}$$

- Main unclosed terms:
 - Pseudo-turbulent stress tensor, \mathbf{R}_u
 - Interphase momentum exchange, $\mathbf{F}^{\text{inter}}$

Modeling pseudo-turbulent Reynolds stress

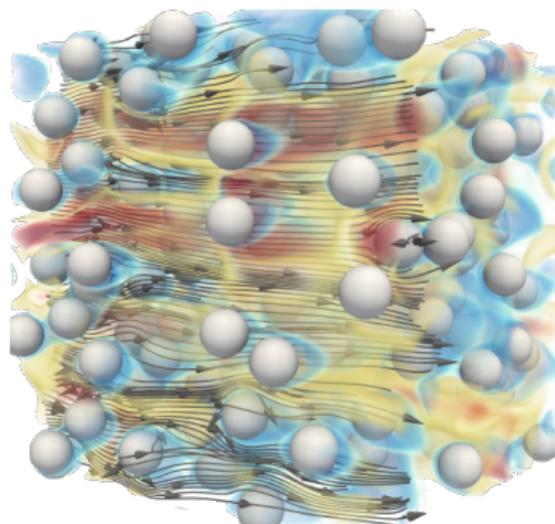
- Subgrid-scale model accounting for pseudo-turbulence (Mehrabadi, et al.²)
- Pseudo-turbulent kinetic energy:

$$k_f = \frac{1}{2} \langle \mathbf{u}_f'' \cdot \mathbf{u}_f'' \rangle$$

$$\approx E_f \left[2\varepsilon_p + 2.5\varepsilon_p (1 - \varepsilon_p)^3 \exp(-\varepsilon_p \text{Re}_p^{1/2}) \right]$$

- Pseudo-turbulent stress tensor:

$$\mathbf{R}_u = \langle \mathbf{u}_f'' \otimes \mathbf{u}_f'' \rangle \approx 2k_f \left(\mathbf{b} + \frac{1}{3} \mathbf{I} \right)$$



Pseudo-turbulence in particle-laden flows³

²Mehrabadi, et al. *Journal of Fluid Mechanics*, 2015

³Lattanzi, et al. *Journal of Fluid Mechanics*, 2022

Particle equation of motion

- Particle transport equation in traditional CFD-DEM:

$$\frac{d\mathbf{u}_p}{dt} = \mathbf{f}_p^{\text{inter}} + \mathbf{f}_p^{\text{col}} + \mathbf{g}$$

- Major drawback: requires tracking each individual particle

- Alternatively, n_{pp} particles lumped into 'parcels'

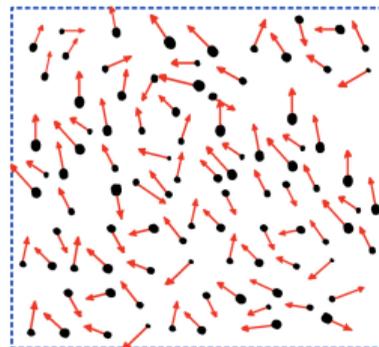
- Common approaches:

- CG-DEM

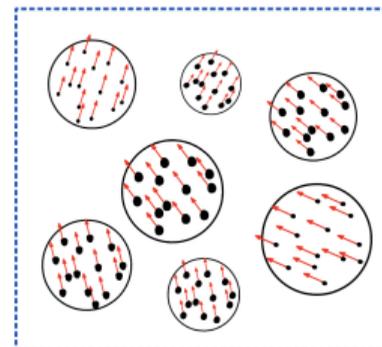
- Collisions modeled using soft-sphere approach with modified restitution coefficient

- Assume uniform particle properties within parcels

- MP-PIC



Particles



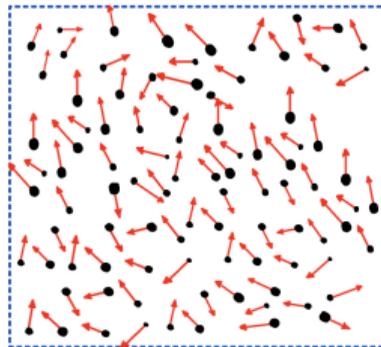
Parcels

Particle equation of motion

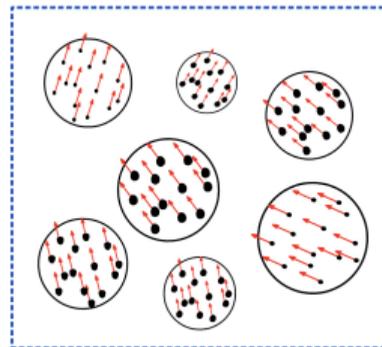
- Particle transport equation in traditional CFD-DEM:

$$\frac{d\mathbf{u}_p}{dt} = \mathbf{f}_p^{\text{inter}} + \mathbf{f}_p^{\text{col}} + \mathbf{g}$$

- Major drawback: requires tracking each individual particle
- Alternatively, n_{pp} particles lumped into 'parcels'
- Common approaches:
 - CG-DEM
 - MP-PIC
 - Collisions modeled using a stochastic approach based on solid stress
 - Assume uniform particle properties within parcels
 - Does not converge to deterministic equations in the limit $n_{pp} \rightarrow 1$



Particles



Parcels

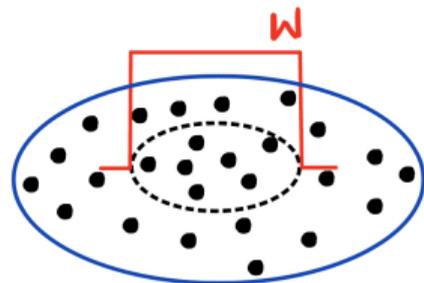
Objective of current work

To develop a rigorous coarse-grain CFD-DEM framework that converges to underlying deterministic equations in the limit $n_{pp} = 1$, with closures that account for variations within the parcel

Coarse-grain modeling using local particle filtering

- Particles are lumped into parcels with n_{pp} particles per parcel
- Parcel properties determined via local averaging using a particle-based filter
- Filtering kernel, W , defined for parcel i :

$$\sum_{j=1}^{n_{pp}} \mathcal{V}_p^{(j)} W(|\hat{\mathbf{x}}_p^{(i)} - \mathbf{x}_p^{(j)}|) = 1$$



Box filter over a parcel

- Notation:
 - $j \rightarrow$ particle index
 - $\mathcal{V}_p \rightarrow$ volume of the particle
 - $\mathbf{x}_p \rightarrow$ particle position
 - $W \rightarrow$ kernel function

Parcel properties obtained using local particle filtering

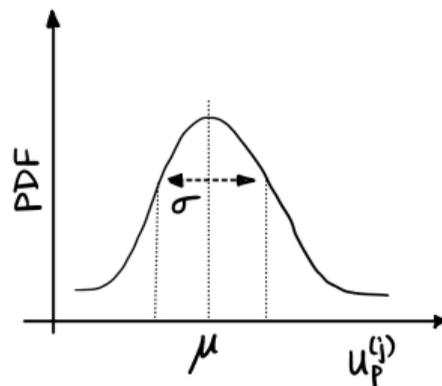
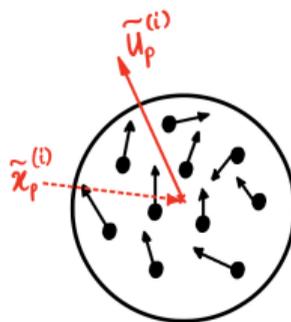
- Position and velocity of parcel i :

$$\hat{\mathbf{x}}_p^{(i)} = \sum_{j=1}^{n_{pp}} \mathbf{x}_p^{(j)} \mathcal{V}_p^{(j)} W_{ij}$$

$$\frac{d\hat{\mathbf{x}}_p^{(i)}}{dt} = \hat{\mathbf{u}}_p^{(i)} = \sum_{j=1}^{n_{pp}} \mathbf{u}_p^{(j)} \mathcal{V}_p^{(j)} W_{ij}$$

- Velocity distribution within parcels is characterized by a mean velocity and granular temperature

$$\theta_p = \frac{1}{3} \overline{\mathbf{u}_p'' \cdot \mathbf{u}_p''}$$



PDF of velocity of particles in a parcel

$$\mu = \hat{\mathbf{u}}_p^{(i)} \quad \sigma = \sqrt{3\theta_p}$$

Parcel equation of motion

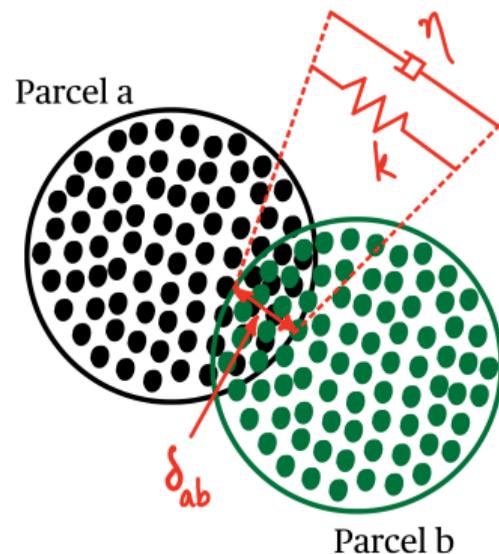
Parcel transport equation:

$$\frac{d\hat{\mathbf{u}}_p}{dt} = \hat{\mathbf{f}}_p^{\text{inter}} + \hat{\mathbf{f}}_p^{\text{col}} + \mathbf{g}$$

- Collisions modeled as a mass-spring-dashpot system using soft-sphere approach
- Coefficient of restitution modified to account for the dissipation of granular energy⁵

$$\frac{\ln e_{CG}}{\ln e} = \sqrt{n_{pp}} \frac{\sqrt{1 - \frac{(\ln e)^2}{(\ln e)^2 + \pi^2}}}{\sqrt{1 - \frac{n_{pp}(\ln e)^2}{(\ln e)^2 + \pi^2}}}$$

- Upper limit on n_{pp} imposed ($e = 0.8 \Rightarrow n_{pp} \leq 199$)



⁵Benyahia, S. and Galvin, J.E., Industrial & Engineering Chemistry Research (2010)

Filtered drag force acting on parcels

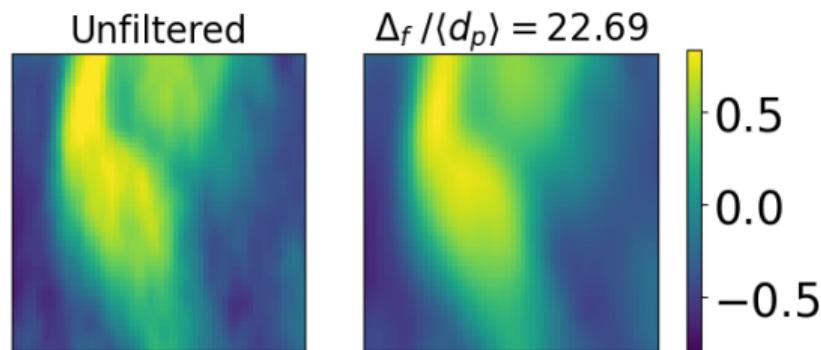
➤ Filtered drag force:

$$\begin{aligned}\widehat{\mathbf{f}}_{drag}^{(i)} &= (\beta[\mathbf{u}_f^{(i)} - \mathbf{u}_p^{(i)}]) = \sum_{j=1}^{n_{pp}} \beta(\varepsilon_f^{(j)}, \text{Re}_p^{(j)}) [\mathbf{u}_f^{(j)} - \mathbf{u}_p^{(j)}] \mathcal{V}_p^{(j)} W_{ij} \\ &= \beta(\bar{\varepsilon}_f^{(i)}, \text{Re}_{CG}^{(i)}) [\tilde{\mathbf{u}}_f^{(i)} - \hat{\mathbf{u}}_p^{(i)}] (1 + H)\end{aligned}$$

➤ Eulerian-based filtering:

$$\tilde{\mathbf{u}}_f = \overline{\varepsilon_f \mathbf{u}_f} / \bar{\varepsilon}_f$$

$$\bar{\varepsilon}_f(\mathbf{x}) = \int_{\Omega} \mathcal{G}(\mathbf{y}) \varepsilon_f(\mathbf{x} - \mathbf{y}) d\mathbf{y}$$



Filtered velocity field

Drag correction factor

➤ Drag correction factor:

$$H = \frac{\widehat{\beta[\mathbf{u}_f^{(i)} - \mathbf{u}_p^{(i)}]}}{\beta(\bar{\varepsilon}_f^{(i)}, \text{Re}_{CG}^{(i)})[\tilde{\mathbf{u}}_f^{(i)} - \hat{\mathbf{u}}_p^{(i)}]} - 1$$

- $H = 0$: drag obtained using resolved quantities require no correction
- $H > 0$: drag overestimated when computed using resolved quantities
- $H < 0$: drag underestimated when computed using resolved quantities

➤ Similar closure appears in filtered two-fluid method^{5,6,7}

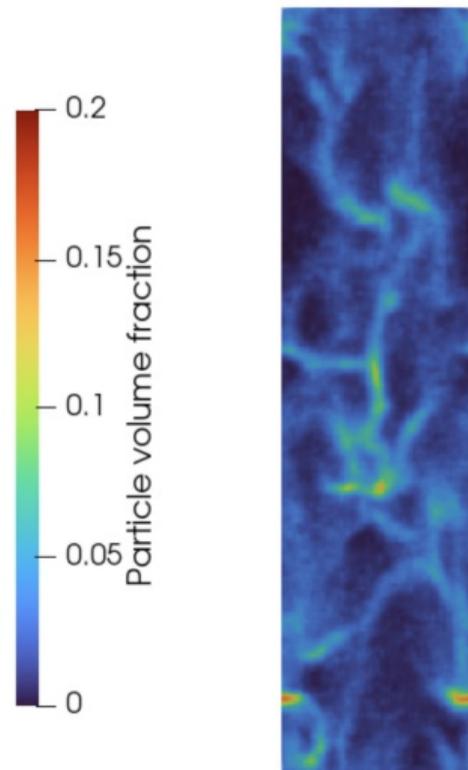
⁵Igci, Y, et al., *Industrial & Engineering Chemistry Research* (2008)

⁶Parmentier, J, et al., *AIChE Journal* (2012)

⁷Milioli, C, et al., *AIChE Journal* (2013)

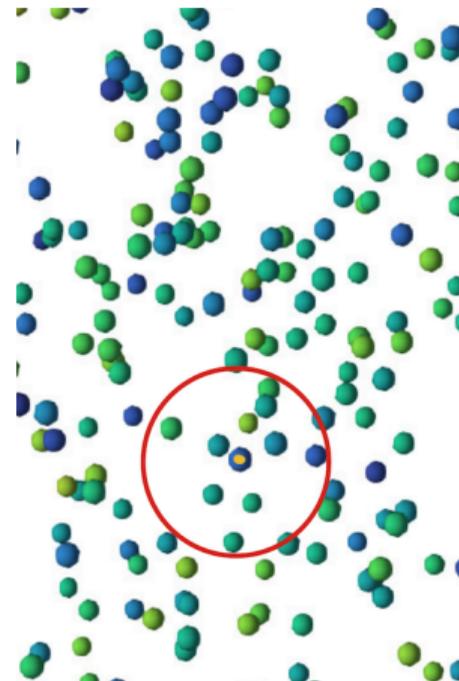
Modeling drag correction factor

- Simulations of “cluster-induced turbulent” flow in a triply periodic domain with monodisperse and polydisperse distribution of Geldart group-A particles
 - Particle mean diameter : $75 \mu m$
 - Particle density : $2250 kg/m^3$
 - Particle volume fraction : 0.02
 - No. of particles $\sim 2.1M$
 - Domain : $960d_p \times 240d_p \times 240d_p$
 - $dx = dy = dz = 3d_p$



Modeling drag correction factor

- Simulations of “cluster-induced turbulent” flow in a triply periodic domain with Geldart group-A particles
- Efficient, embarrassingly parallel, detection of particles within parcels using KDTree algorithm

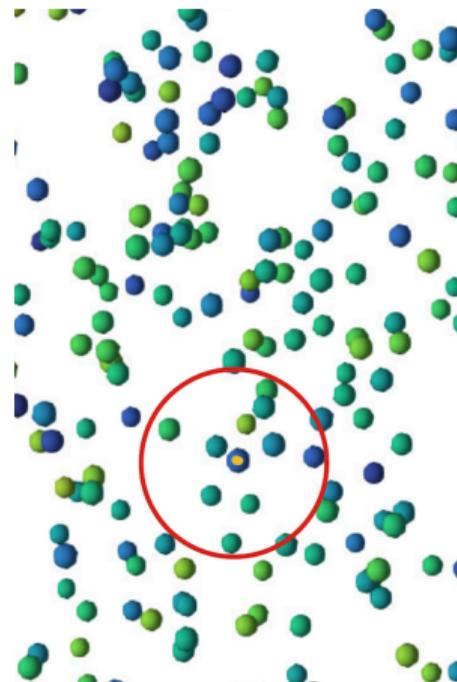


Zoomed-in view of particles in flow

Modeling drag correction factor

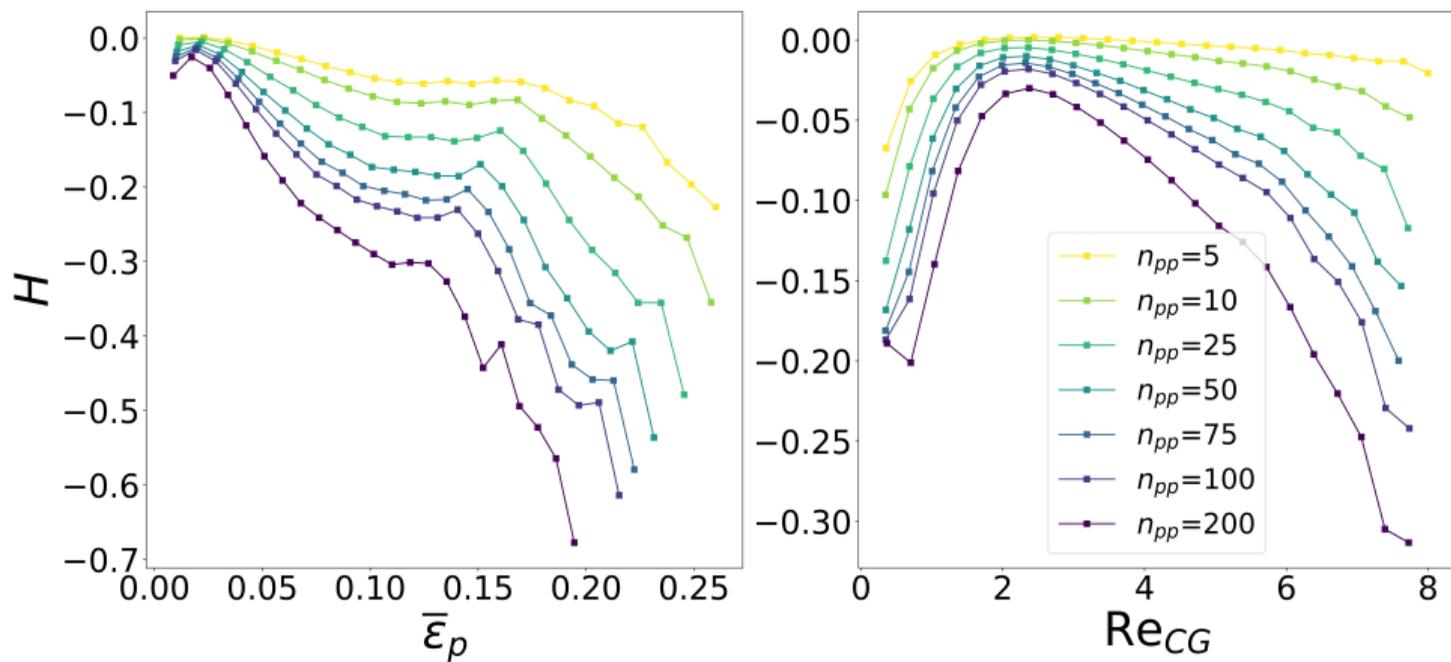
- Simulations of “cluster-induced turbulent” flow in a triply periodic domain with Geldart group-A particles
- Efficient, embarrassingly parallel, detection of particles within parcels using KDTree algorithm
- Filtered quantities obtained using a volume-weighted box filter

$$\hat{\mathbf{u}}_p^{(i)} = \sum_{j=1}^{n_{pp}} \mathbf{u}_p^{(j)} \mathcal{V}_p^{(j)} W_{ij} = \frac{\sum_{j=1}^{n_{pp}} \mathbf{u}_p^{(j)} \mathcal{V}_p^{(j)}}{\sum_{j=1}^{n_{pp}} \mathcal{V}_p^{(j)}}$$



Zoomed-in view of particles in flow

Variation of drag correction factor with monodisperse particles



Parcels with polydisperse particles

- Solid mass and volume are conserved within each parcel

$$m_{CG} = \sum_{j=1}^{n_{pp}} m_p^{(j)}, \quad \mathcal{V}_{CG} = \sum_{j=1}^{n_{pp}} \mathcal{V}_p^{(j)}$$

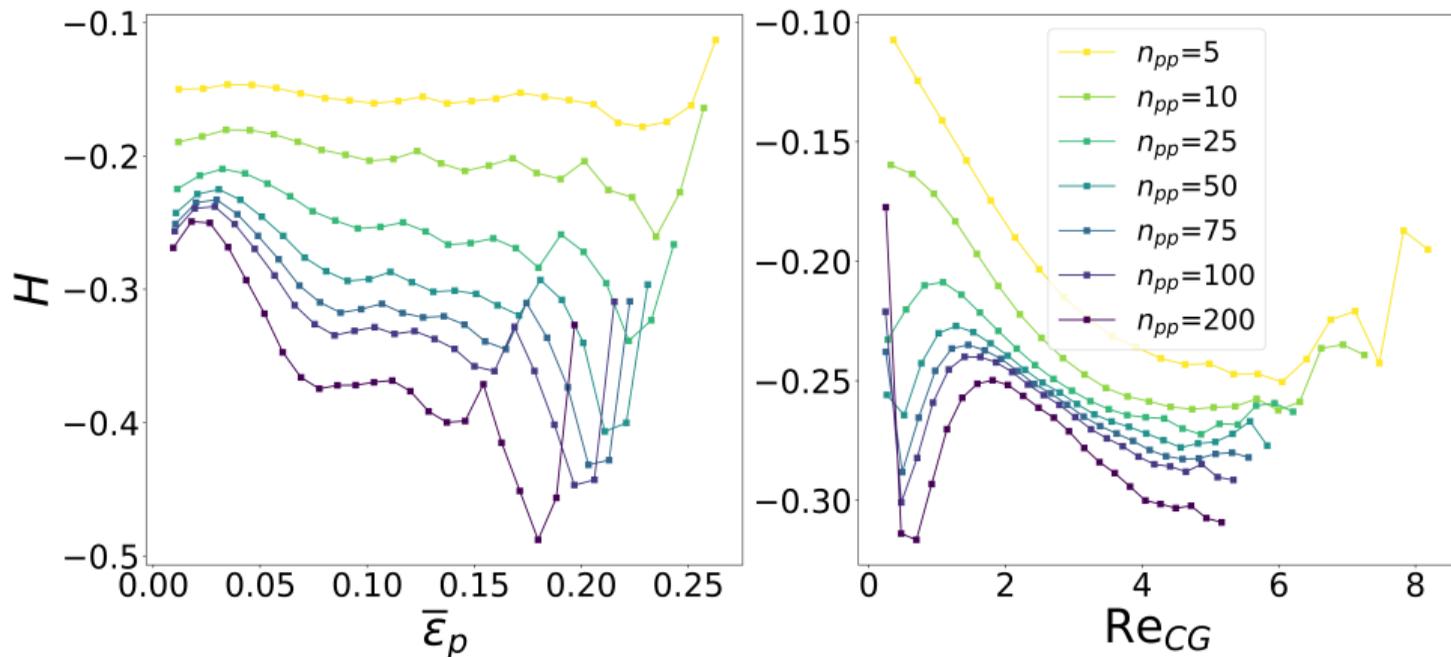
- Equivalent diameter of the particles in the parcels computed using volume constraint

$$d_{p,\text{eff}} = \left(\frac{\sum_{j=1}^{n_{pp}} d_p^3}{n_{pp}} \right)^{1/3}$$

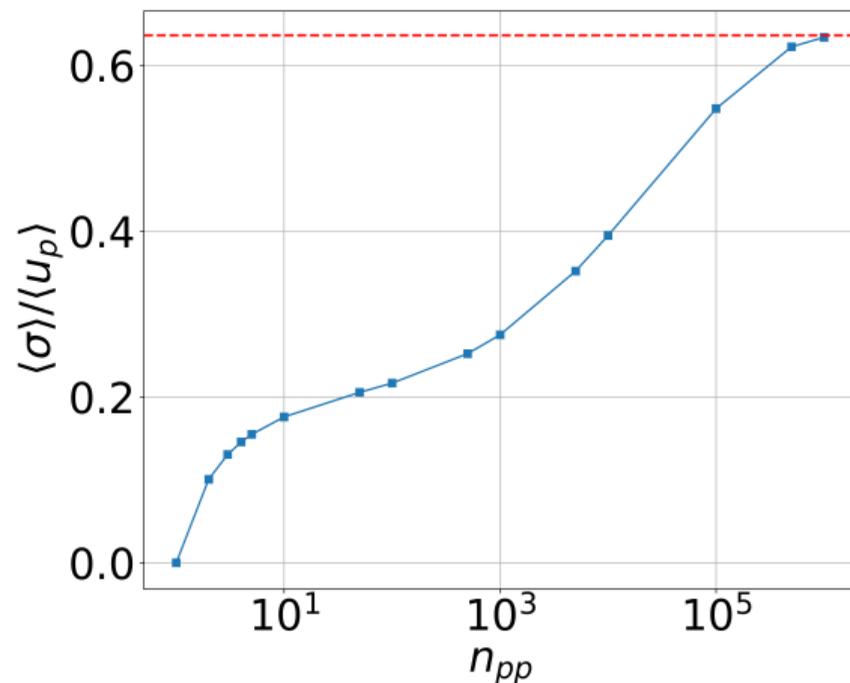
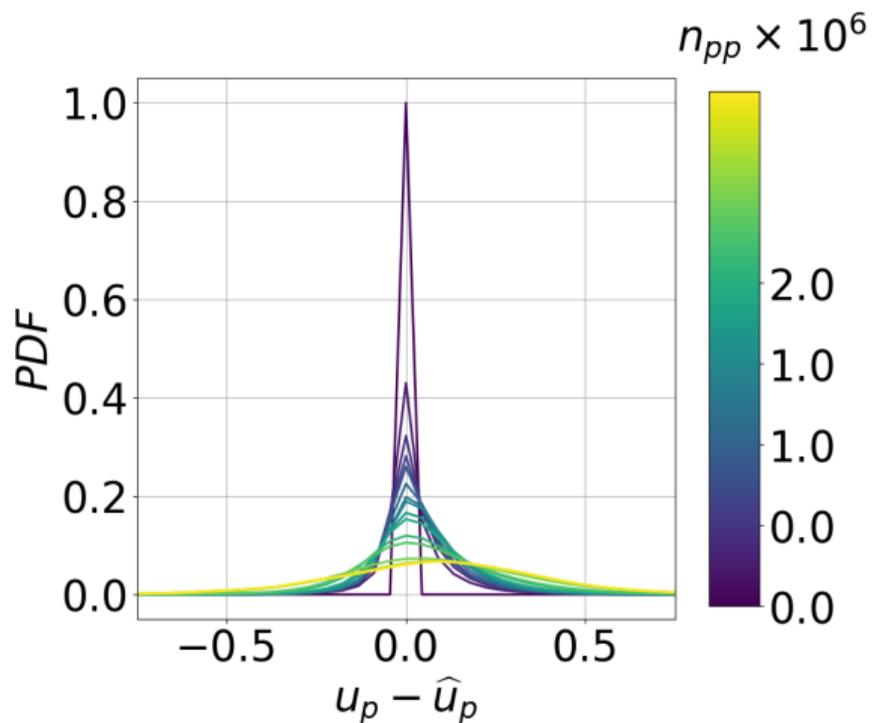
- Reynolds number of parcels computed using the effective diameter

$$\text{Re}_{CG} = \frac{\rho_f |\widetilde{\mathbf{u}}_f - \widehat{\mathbf{u}}_p| d_{p,\text{eff}}}{\nu}$$

Variation of drag correction factor with polydisperse particles



Velocity variation within parcels



Conclusion

Summary

- A rigorous formulation of a scalable filtered coarse-grain CFD-DEM is presented
- Unclosed sub-filter terms arise and require modeling
- Sub-filter drag force quantified with simulations of “cluster induced turbulent” flows

Future work

- Symbolic regression will be employed to obtain closed-form algebraic models
- Formulation will be extended to capture high-order particle statistics within parcels like granular temperature, in addition to exchanging particles between parcels
- Borrow ideas from moment methods for the formulation



Thank you!