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Consistent closures for Euler-Lagrange models of bi-disperse gas-particle suspensions derived from particle-resolved direct numerical simulations

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ABSTRACT

Particle-Resolved Direct Numerical Simulation (PR-DNS) is employed to simulate momentum and energy transport in bi-disperse gas-solid suspensions by means of a novel hybrid immersed-boundary/fictitious domain (HFD-IB) method. First, we demonstrate the accuracy of the new HFD-IB method against several verification tests. Subsequently, we simulate momentum and energy transfer in bi-disperse suspensions in the limit of high Stokes number, and the predicted flow and temperature fields are used, in conjunction with the open-source parallel data processing library CPPPO (Municchi et al., 2016), to assess the validity of existing closures for momentum and heat transfer in the frame of Particle-Unresolved Euler-Lagrange (PU-EL) models. We propose a correction to the drag force model proposed by Beetstra et al. (2009) which consistently takes into account the pressure contribution to the total fluid-particle interaction force in PU-EL models. Also, we propose a stochastic closure model for the per-particle drag coefficient based on a modified log-normal distribution. Finally, we assess the existence of an analogy between the particle-based drag coefficient and the conditionally-averaged Nusselt number. Indeed, our PR-DNS data indicates that a stochastic closure similar to that for the drag can be used to close the particle-based Nusselt number in dense bidisperse suspensions.

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1. Introduction

Numerical simulations of large scale particle flows, which are widely encountered in industrial applications, are normally performed using averaged equations of motion. In these descriptions the solid and fluid phases are modeled as interpenetrating continua [3]. These models are normally based on the kinetic theory of granular flows [4] and contain unclosed terms that have to be modeled somehow. In a multi-scale approach [5,6], these models can be derived from more detailed simulations where particles are described as a discrete phase. In particular, in case the trajectory of each particle is tracked and collisions are resolved, one obtains the so-called Computational Fluid Dynamics-Discrete Elements Method (CFD-DEM) that, in case fluid cells are larger than particle diameters, can also be referred to as the Particle-Unresolved Euler-Lagrange approach (PU-EL) [7–13]. However, even PU-EL equations have several unclosed terms, like the interphase transport coefficients, that account for, for example, fluidparticle heat and momentum transfer. Following the multi-scale paradigm in our present contribution, we seek to obtain certain closures from fully resolved simulations, i.e., where the detailed flow and temperature (or concentration) fields are resolved on a sub-particle level. This latter approach can be denoted as Particle-Resolved Euler-Lagrange (PR-EL), or Particle-Resolved Direct Numerical Simulation (PR-DNS) if turbulence models are used or not, respectively. PR-DNS has already been extensively used to derive closures for the drag coefficient in mono- and bi-disperse suspensions [14,2,15], or for the Nusselt/Sherwood number in mono-disperse suspensions [16–20]. However, almost the totality of this previous work focused on closures for Euler-Euler-based simulations in a coarse scale. Naturally, the question arises of the same closures can be used for PU-EL simulations, and we will demonstrate that indeed this is not the case.

1.1. Upscaling and closure development strategies

A major difference between closures for Euler-Euler and PU-EL models is that the latter require a particle-based description of the interphase transfer processes, while the former (i.e., EE models) require average exchange coefficients. Thus, particle-based models are affected by per-particle fluctuations that arise simply due to the random arrangement of individual particles, and occur even in low-Reynolds number flows. When developing closures for continuum formulations, these quantities, are obtained from averages

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Nomenclature		р	pressure [-]
Abbrouid	ations	Pe	Peclet number [–]
CED	computational fluid dynamics	Per	longitudinal Peclet number [–]
DEM	discrete element method	Pr	Prandtl number [_]
DEIVI	discrete element method	0	interface beat transfer rate $[-]$
DINS	direct numerical simulation	Q r	radial coordinate [m]
EL	Euler-Lagrange	I D-	
HFD-IB	hybrid fictitious-domain/immersed-boundary	Re	Reynolds number [-]
PR	particle-resolved	S	HFD-IB interpolation distance [m]
PU	particle-unresolved	T	temperature [K]
		t	temporal coordinate [–]
Creek cl	naracters	U_s	superficial velocity [m/s]
N CICCR CI	log_normal standard deviation [_]	v	particle velocity [m/s]
ß	log normal lower bound []	V	CFD cell volume [m ³]
p	nog-normal lower bound [-]	x, y, z	cartesian coordinates [-]
η	particle diameter [-]		
γ		Subscrip	nte/superscripts
ĸ	tortuosity [–]	Subscrip	consistent
Λ	interphase saturation coefficient [-]	(cons)	consistent
λο	characteristic fluid-particle system length [–]	() ()	evaluated from the HFD-IB algorithm
v	fluid kinematic viscosity [m²/s]	$\Phi_{m,j}$	evaluated using discrete marker field at j
Ω	computational domain	Deen, G	unn, Sunn refers to one of these correlations
ω	relaxation factor [–]	F	relative to the drag force
ϕ	volume fraction [–]	G	global quantity
Φ_m	marker field	mix	relative to the homogeneous mixture
Φ.	sampled field	Nu	relative to the Nusselt number
0 [°]	phase density $[kg/m^3]$	*	dimensional quantity
σ	scaled standard deviation [–]	0	reference value
Â	saturation margin [_]	32	mean Sauter quantity
Α	fluid temperature [_]	∇p	contribution from pressure gradient
0	filter size []	Φ.,	evaluated using marker field
ϱ		h	hulk quantity
		c	property related to CFD cell c
Latin ch	aracters	d	contribution from drag
f	total interphase force [–]	f	property related to fluid phase
u _.	velocity field [–]	J	guantity related to the whole domain (global)
C'	ensemble conditional averaging Kernel [–]	g	imposed value
М	coefficient matrix [–]	1	ninposed value
U	discrete binning set [–]	1	property related to particle number i
\mathcal{P}	modified log-normal distribution	ĸ	property related to particles specie k
Α	surface [m ²]	M	property related to particles with the largest radii
A_n	specific exchange surface [1/m]	т	property related to particles with the smallest radii
Cn	fluid thermal capacity [I/kg K]	n	quantity at time step <i>n</i>
d	particle diameter [m]	р	property related to particle phase
d _{cr}	distance between cell c and particle surface [m]	P1, P2	property interpolated at point P1 or P2
F	interphase drag coefficient [_]	S	property evaluated at particle surface
F-	Beetstra interphase drag coefficient [_]	sat	value at saturation
ГВ	corrected interphase drag coefficient []		
г _{согг} Ь	CED cells per particle diameter []	Averaoi	ng/filtering operators
11	CFD cens per particle diameter [-]	/()\	ensemble average
I V	indicator function [-]	\(*)/ /()\	ensemble average conditional on Φ
K	kernel function	<u>((*)</u>)Φ	chischible average conditional off Ψ
k	heat conductivity [W/mK]	(*)	volume average
N_p	number of particles [-]	$\widehat{(*)}$	flux average
Nu	Nusselt number [–]	$\widetilde{\mathbf{i}}$	
Nu _{bi}	bi-disperse Nusselt number [–]	(*)	ravie average

within each realization, so that the fluctuations in the particle population are lost and the final standard deviation is calculated based on the ensemble of realizations. On the contrary, closures for PU-EL models are based on the whole studied population and thus, they may require stochastic models to take into account the single particle variability [14,21].

The process of upscaling of fluid quantities that we adopted is known as (spatial) filtering or coarse-graining [22]. The local domain where this operation is performed (in PR-DNS) can be identified with a fluid cell used in PU-EL. In general, closure models derived using this approach have a functional dependence on

the filter size, i.e., the support of the filtering kernel or, in other words, the size of the coarse-grained cell [8]. Moreover, while the velocity field is statistically homogeneous in homogeneous particle configurations, the temperature field is generally inhomogeneous [19,20]. This is in contrast with the assumption of separation of scales required for the development of continuum formulations, e.g., Euler-Euler models, and poses a challenge also in the development of particle-based models such as PU-EL-based models. In the present work we refer to this issue as *saturation*, since this term does reflect the physical process that is behind.

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