

Modeling Particle-Fluid Momentum Transfer in Polydisperse Gas-Solid Flows Through Direct Numerical Simulations Based on the Immersed Boundary Method

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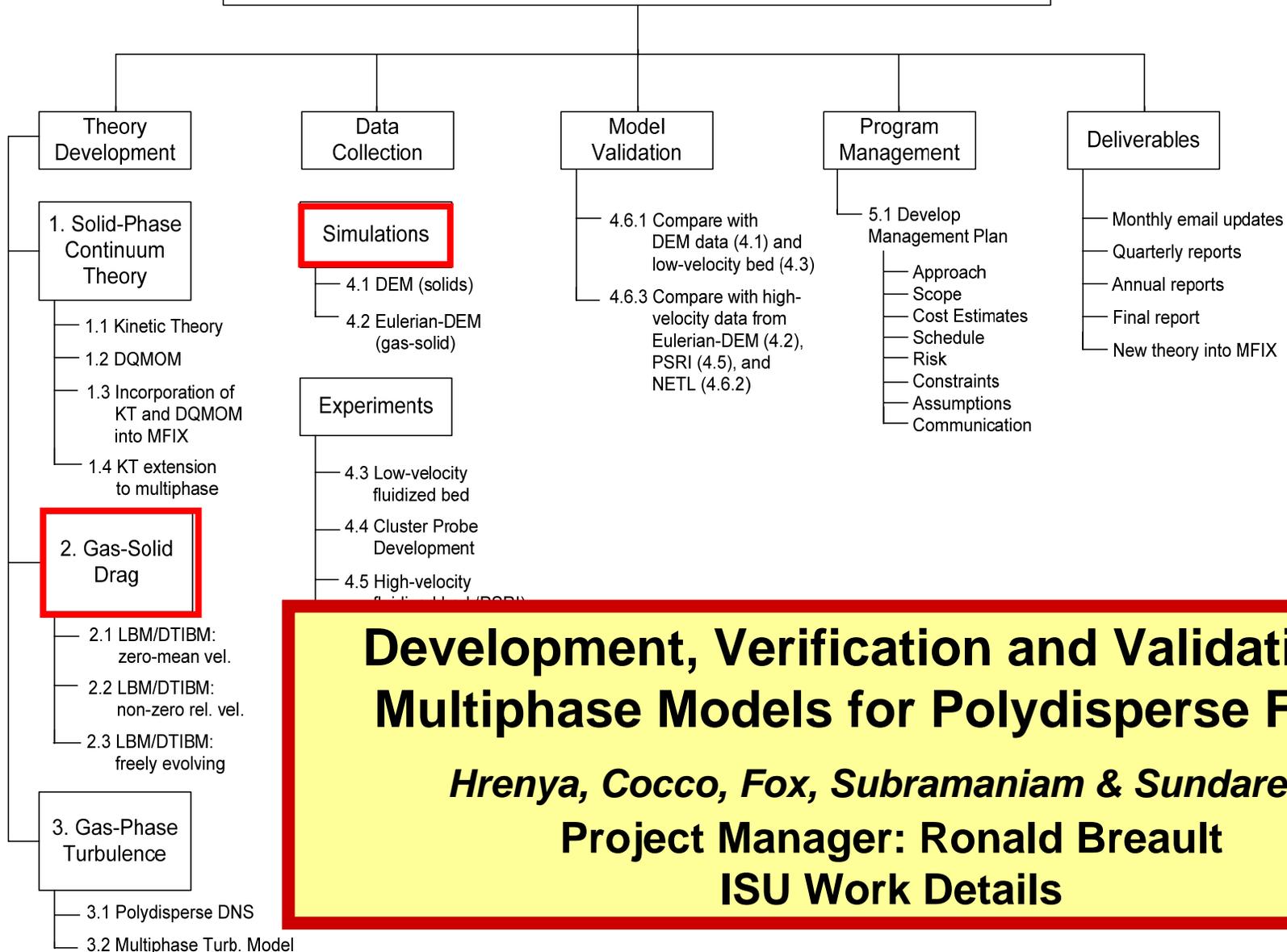
Department of Mechanical Engineering

Iowa State University



Project scope: Work breakdown structure

Development, Verification, and Validation of Multiphase Models for Polydisperse Flows



Development, Verification and Validation of Multiphase Models for Polydisperse Flows

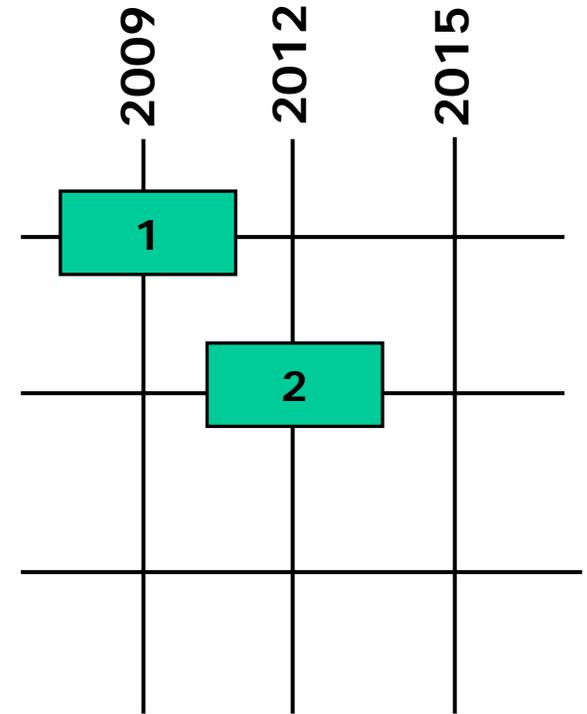
Hrenya, Cocco, Fox, Subramaniam & Sundaresan

Project Manager: Ronald Breault

ISU Work Details

Connections to Multiphase Flow Roadmap

1. Develop drag relations that can handle particle size and density distributions
2. Development of constitutive relations for continuum models from high fidelity simulations



MFIX Two-Fluid Model

MFIX two-fluid model

ϵ_g, ϵ_s

$\mathbf{u}_g, \mathbf{v}_{sm}$

Gas-phase

\mathbf{u}_g

\mathbf{l}_{gp}

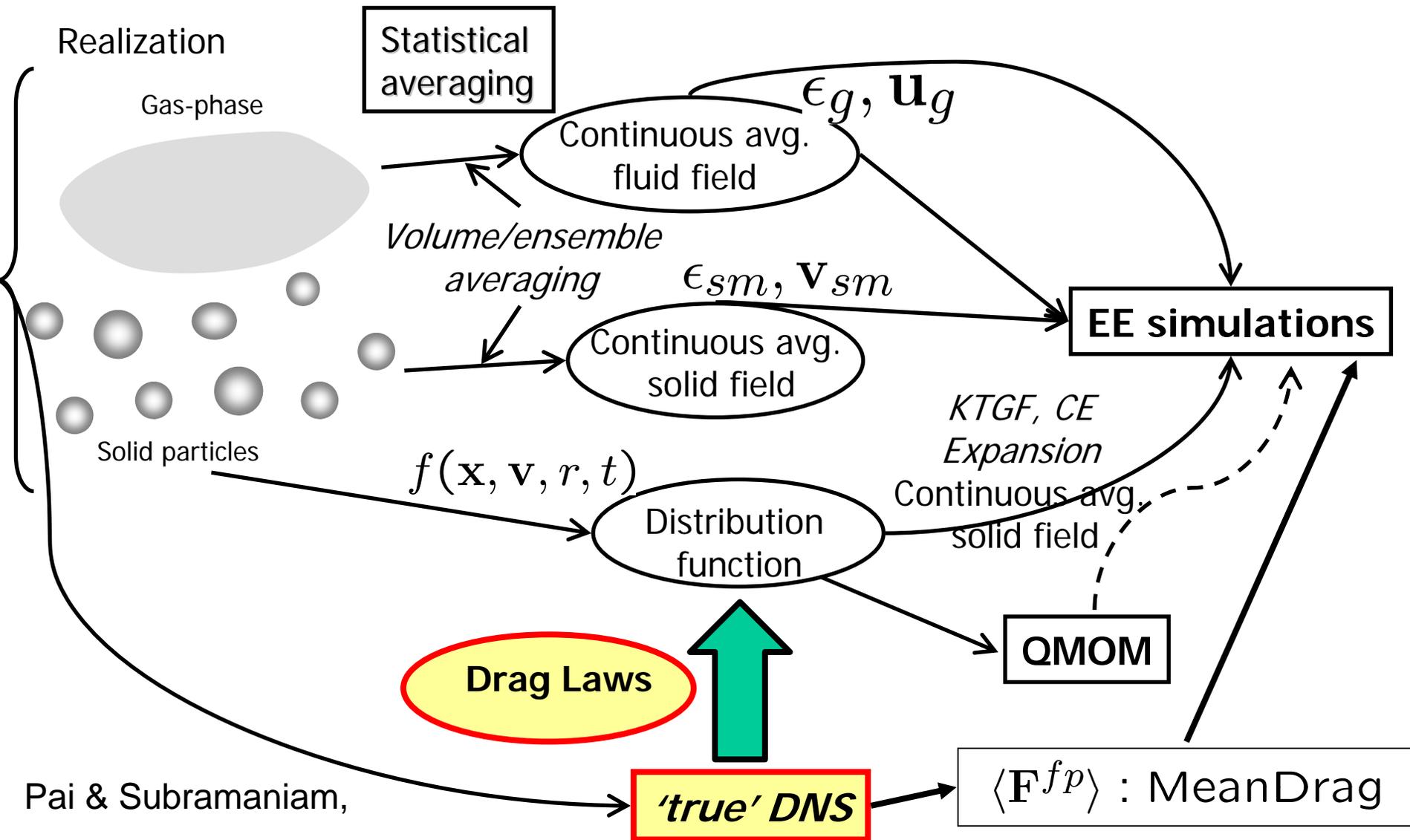
\mathbf{v}_{sm}

Solid particles

Gas-particle interaction

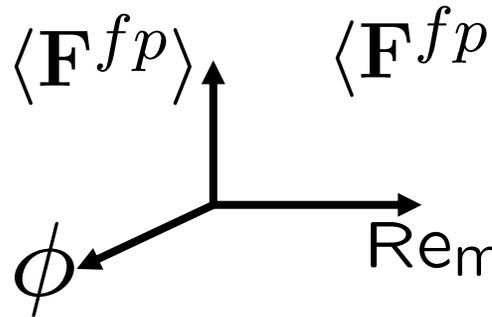
Drag Laws: Correlations for *average* force on particles

Introduction



Pai & Subramaniam,
JFM (2009) (to appear)

Drag Laws (Computational)

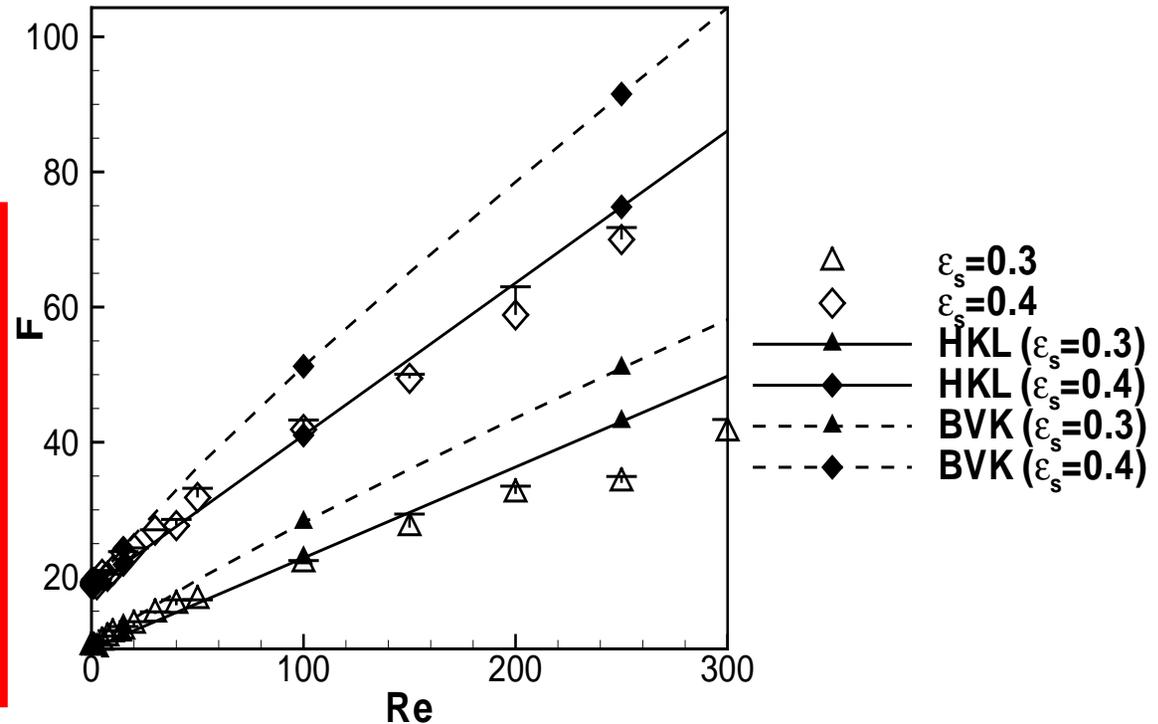
$$\langle \mathbf{F}^{fp} \rangle$$

$$\langle \mathbf{F}^{fp} \rangle = \langle \mathbf{F}^{fp} \rangle (\phi, Re_m)$$

Monodisperse drag laws

➤ Hill et al (JFM 2001)

➤ Beetstra et al (AiChE, 2007)

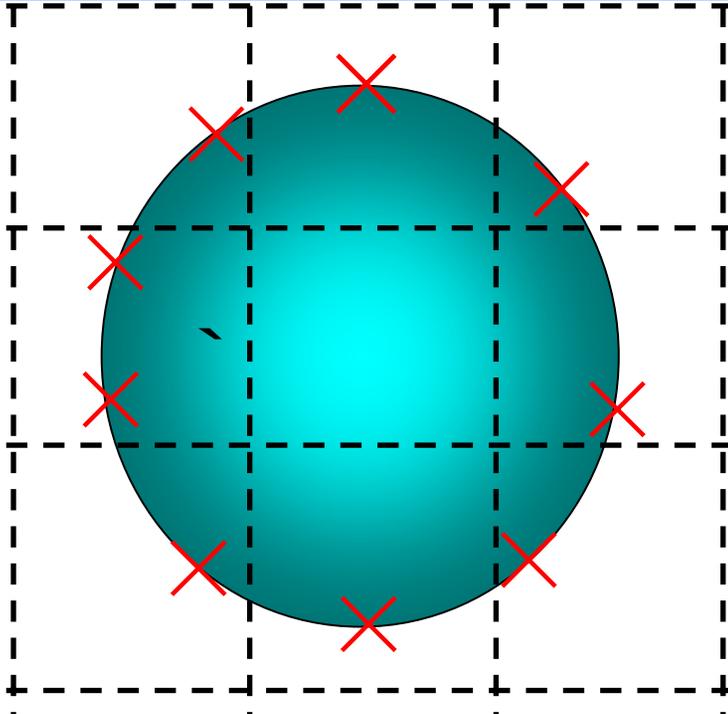
Using Lattice Boltzmann Method (LBM)



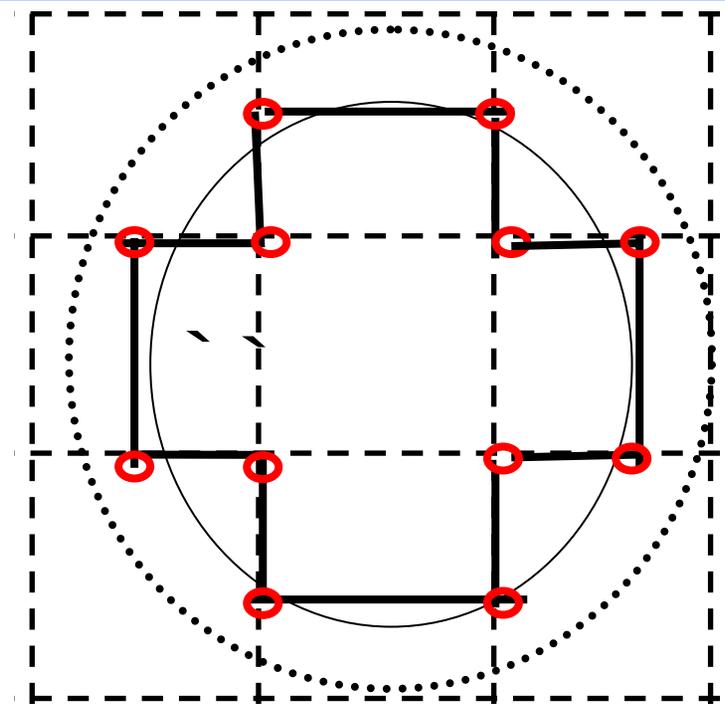
Present Approach: Immersed Boundary Method (IBM)

"Direct Numerical Simulation of Gas-Solids Flow based on the Immersed Boundary Method", Garg et al. in *Computational Gas-Solids Flows and Reacting Systems: Theory, Methods and Practice*, eds S. Pannala, M. Syamlal and T. J. O'Brien (in review)

IBM- LBM Comparison



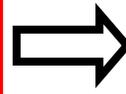
- IBM : Continuum Navier-Stokes Solver
- Incompressible flow solution: elliptic pressure solve
- Sphere: impose BC on boundary
- Drag : integrating stress tensor at sphere boundary (crosses)



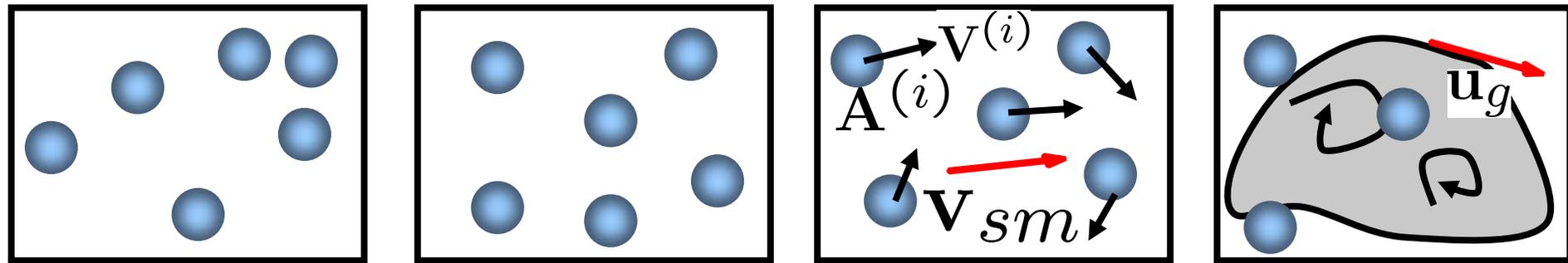
- LBM : Discretized distribution function of fluid molecules
- Parallel local operations, always compressible; $Ma = Kn Re$
- Sphere : Stair-step function
- Drag: reported for hydrodynamic radius (dotted circle)

Mean Drag

Drag Laws: Correlations for *average* force on particles



What is the averaging performed over?



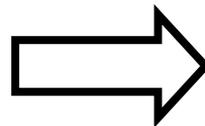
Particle Configurations

Particle velocity distribution

Particle acceleration distribution: from DNS

Fluid velocity fluctuations (not necessarily turbulence)

Modeling approach



$f(\mathbf{x}, \mathbf{v}, r, t)$

One-particle distribution function

Drag Law Model

$$f(\mathbf{x}, \mathbf{v}, r, t)$$

□ One-particle distribution function
(Mean number density)

Modified KTGF (Hrenya)

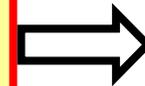
QMOM (Fox)

Accounts for:

- dependence on particle size
- dependence on particle velocity

Does not explicitly account for:

- neighbouring particle effects
- effect of fluid



Need to be modeled

Mean Drag

Conditional average of acceleration (drag)

$$\langle \mathbf{F}^{fp} \rangle(\mathbf{x}, t) = \int_{[\mathbf{v}, r]} m(r) \langle \mathbf{A} | \dots, \mathbf{x}, \mathbf{v}, r; t \rangle f(\mathbf{x}, \mathbf{v}, r, t) d\mathbf{v} dr$$

Instantaneous Particle Acceleration Models

Drag Law: Mean Acceleration

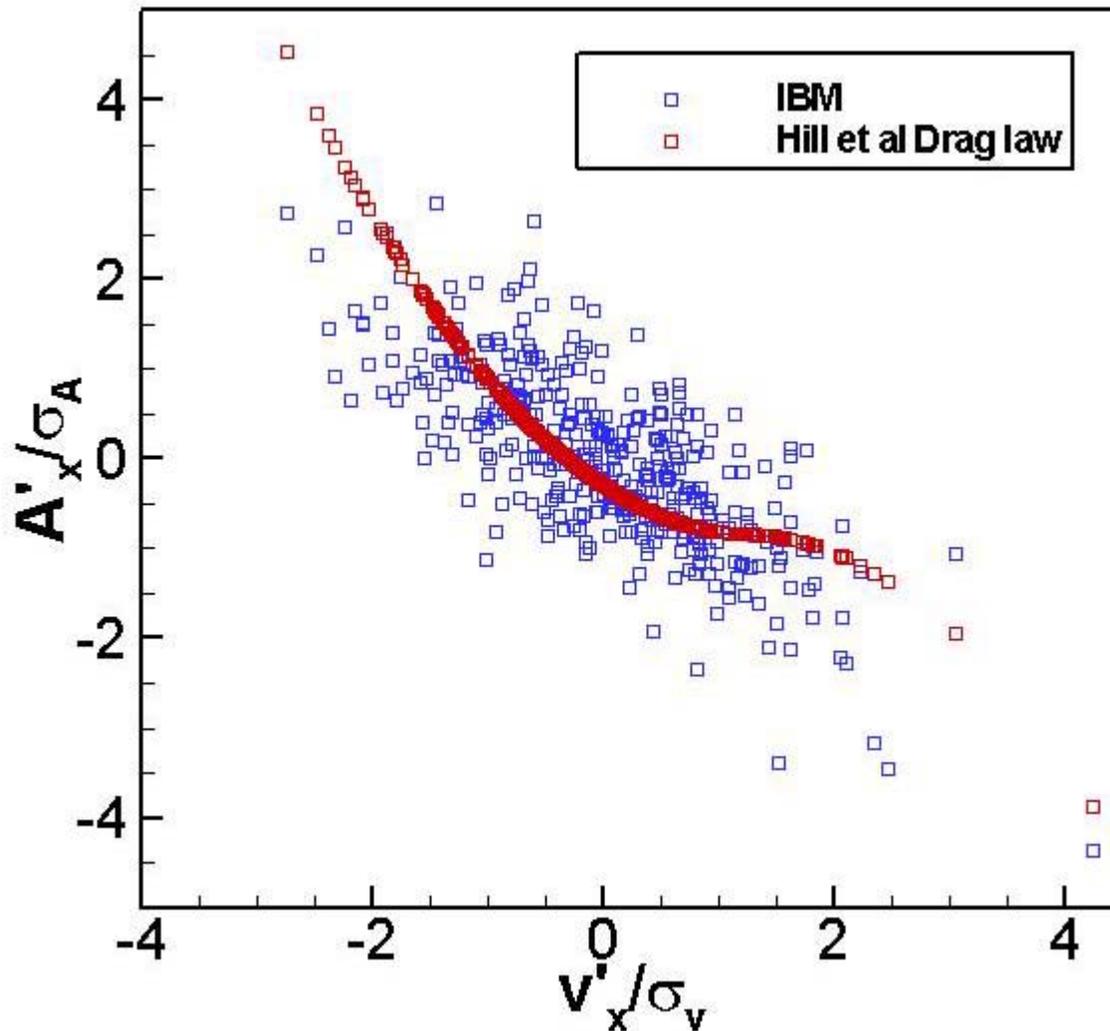
$$\langle A_i \rangle = \beta \left(\langle v_i \rangle - \langle u_i^{(f)} \rangle \right)$$

Simple extension of mean acceleration model to instantaneous particle acceleration

$$A_i = -\beta W_i$$

$$W_i = v_i - \langle u_i^{(f)} \rangle$$

Fluctuating Particle Acceleration-velocity Scatter



$$Re_m = 20$$

$$Re_T = 16$$

$$\phi = 0.2$$

$$A_i = -\beta W_i$$

β : Hill et al.

(JFM 2001)

Drag law applied to velocity distribution does not recover the acceleration distribution

Role of Particle Acceleration Fluctuations

Mean particle acceleration

Fluctuations in particle acceleration:

➤ correlate with fluctuations in particle velocity

$$\langle A_i'' v_j'' \rangle$$

$$S + \Gamma_{vis}$$

Koch (Phys. Fluids 1990)

Fluctuations in particle acceleration:

➤ generate a hierarchy of moments

provide closures for **ALL** moments

To account for effect of fluid phase, neighbour particle interactions etc

Continuum model (MFIx-EE)

Affect granular temperature equation

Need a model for:

$$\langle \mathbf{A} \mid \dots \mathbf{x}, \mathbf{v}, r; t \rangle$$

Instantaneous Particle Acceleration Model

$$dv_i = A_i^{(d)} dt + B_{ij} dW_j \quad \longrightarrow \quad \text{Langevin Model}$$

Drift term

$$A_i^{(d)} = \beta_{(i)} \langle W_i \rangle dt - \gamma_{ij} v_j'' dt + \Sigma_{ij} dW_j$$

Mean slip

Velocity fluctuations (particle velocity distribution)

Wiener process (effect of neighbouring particle + fluid phase fluctuations)

γ_{ij}

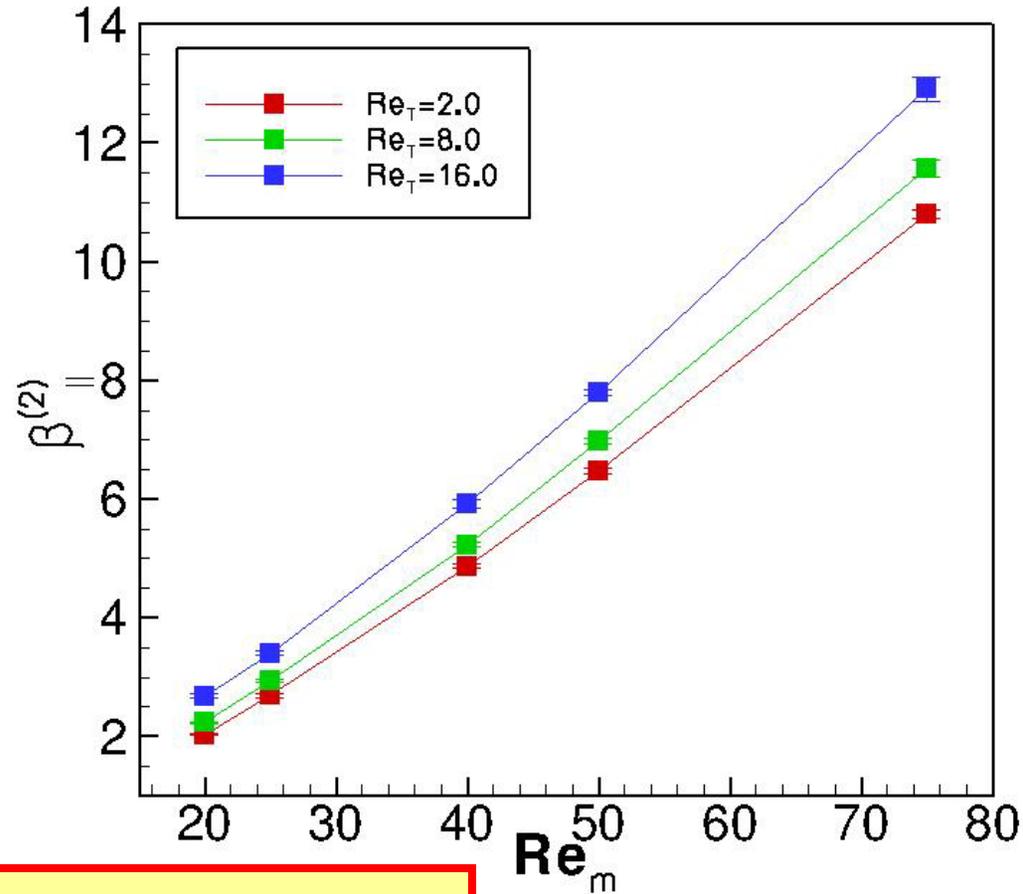
- inverse of Lagrangian particle velocity autocorrelation time
- how long particle retains memory of initial velocity
- function of Stokes number

Langevin Model: Coefficients

$$\beta_{(i)} = \beta_{(i)}(\phi, Re_m, Re_T)$$

- Depends on volume fraction
- Reynolds number based on mean slip velocity
- Reynolds number based on particle granular temperature

Volume fraction = 0.2



γ, Σ

Freely evolving suspensions

Langevin Model: Coefficients

$$dv_i'' = -\gamma v_i'' dt + \Sigma d\mathcal{W}_j$$

Use Lagrangian structure function

$$D_L(s) = \left\langle [v_i''(t+s) - v_i''(t)]^2 \right\rangle$$

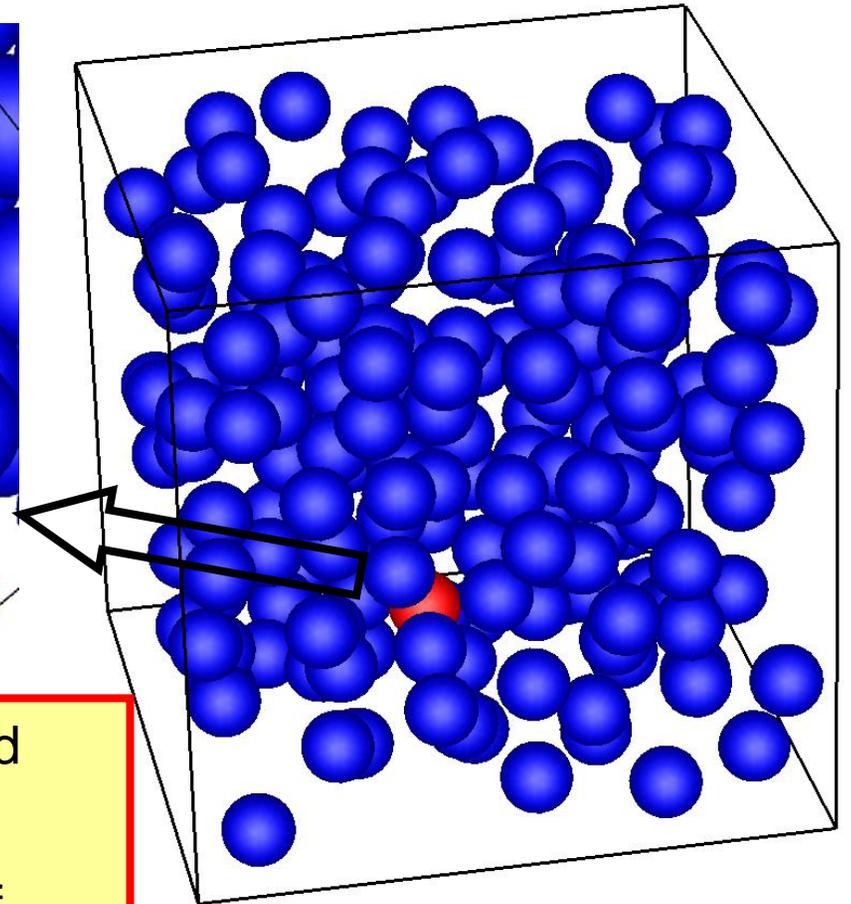
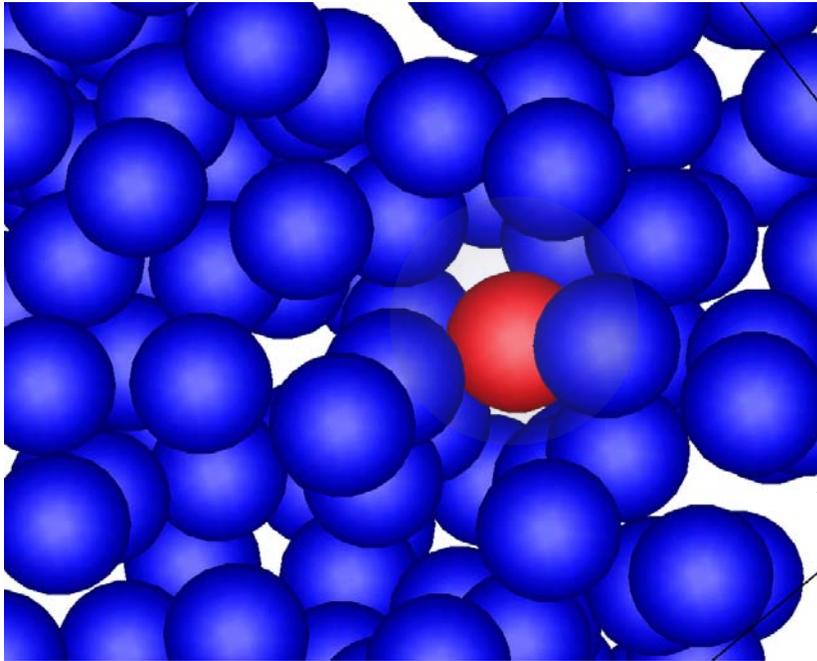
Lagrangian structure function for Langevin model

$$D_L^*(s) = \Sigma^2 s$$

$$\frac{\Sigma^2}{2\gamma} = \frac{1}{3}T$$

Extract the structure function from the DNS of freely evolving suspensions (particles feel the fluid force)

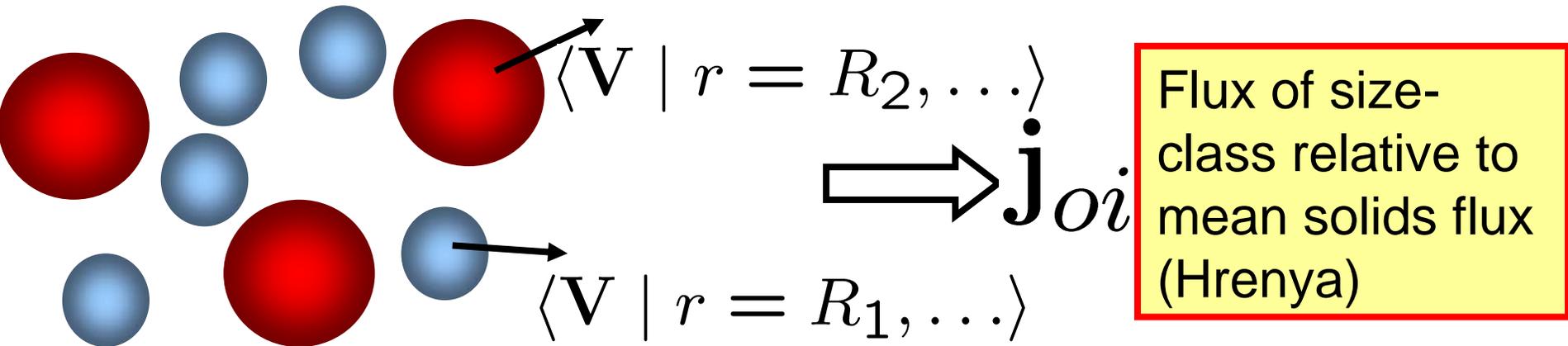
Test Particle in a Homogeneous Assembly



- Assembly of fixed particles: initialized with a velocity distribution
- Particles within radius of influence of moving test particle are allowed to move
- Velocity autocorrelation and Lagrangian structure function will be extracted to determine coefficients

Drag laws: Effect of Particle Size Distribution

Bidisperse Example



Segregation due to drag manifests as flux of size-class relative to mean solids flux

$$\langle \mathbf{A} \mid r = R_\alpha, \dots \rangle - \langle \mathbf{A} \mid \dots \rangle$$

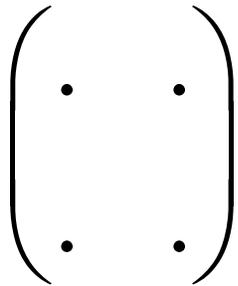
Driving force is drag conditional on size

Instantaneous Particle Acceleration Model

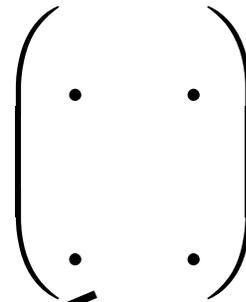
Extension to bidisperse case

$$\langle W_i | r = R_\alpha \rangle = \langle v_i | r = R_\alpha \rangle - \langle u_i^{(f)} \rangle$$

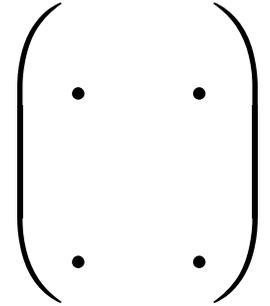
$$m^\alpha dv_i^\alpha = -\beta_{(i)}^{\alpha\eta} \langle W_i | r = R_\eta \rangle dt - \gamma_{ij}^{\alpha\eta} v_j''^{(\eta)} dt + \Sigma_{ij}^{\alpha\eta} dW_j^\eta$$



Effect of species diffusion velocity on mean drag of a size class



Effect of particle velocity distribution in a size-class



Effect of neighbor particles

All terms include effect of the presence of other size-class

Mean drag: Bidisperse (Equal Velocities)

$$m^\alpha dv_i^\alpha = -\beta_{(i)}^{\alpha\eta} \langle \mathbf{W}_i | r = R_\eta \rangle dt - \gamma_{ij}^{\alpha\eta} v_j''^{(\eta)} dt + \sum_{ij}^{\alpha\eta} d\mathcal{W}_j^\eta$$

no relative velocity
between size classes

$$F_{D-\alpha}^* = \frac{|\langle \mathbf{F} \rangle_{D-\alpha}|}{3\pi\mu D_\alpha (1-\phi) |\langle \mathbf{W} | r = R_\alpha; t \rangle|}$$

Average
drag force
per particle

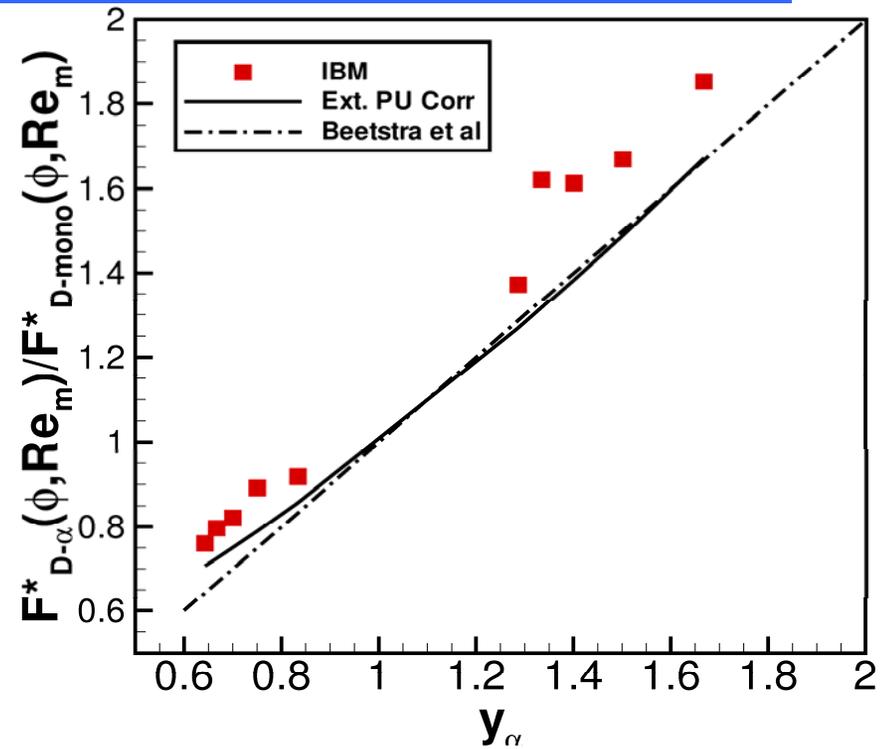
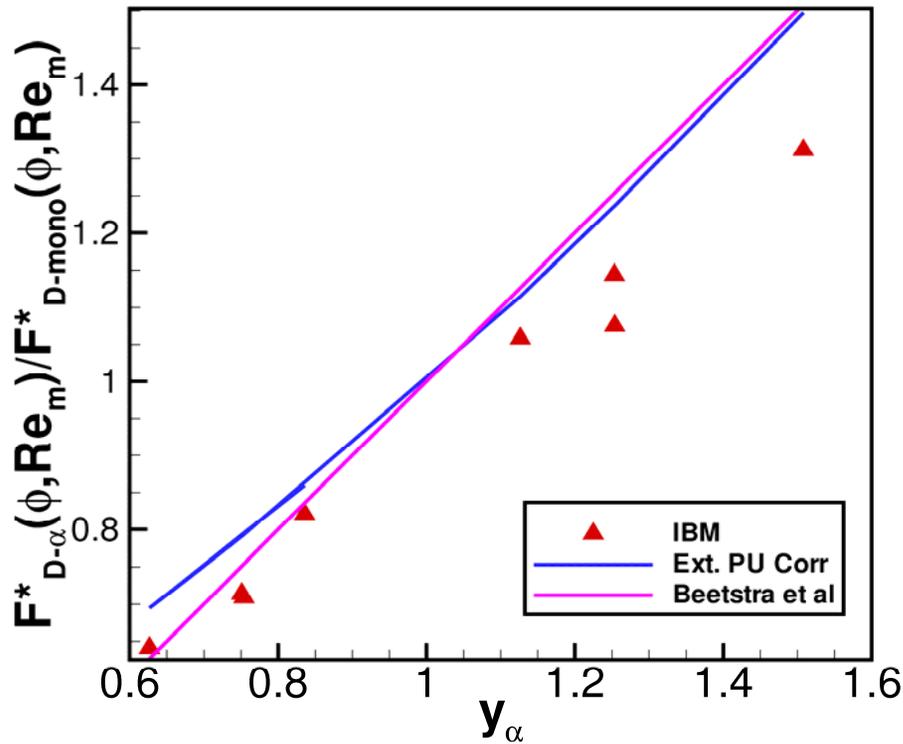
Express the mean drag in terms of an equivalent
monodisperse suspension (Sauter mean diameter)

$$y_\alpha = \frac{D_\alpha}{\langle D \rangle}$$

$$\frac{F_{D-\alpha}^*(\phi, \text{Re}_m)}{F_{D-\text{mono}}^*(\phi, \text{Re}_m)} = y_\alpha$$

No dependence on
Reynolds number
(Beetstra et al. AiChE
J. 2007)

IBM Bi-disperse Simulations : Normalized force



- $Re_m = 20$
- $D_2/D_1 = 1.5, 2$
- $\phi_2/\phi_1 = 1, 2, 3, 4$

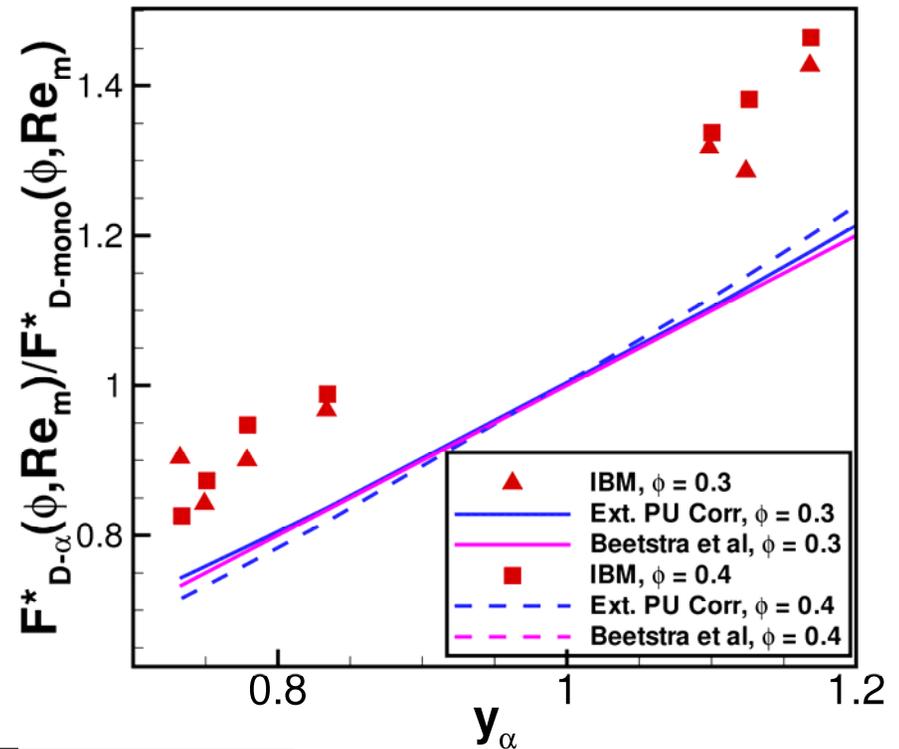
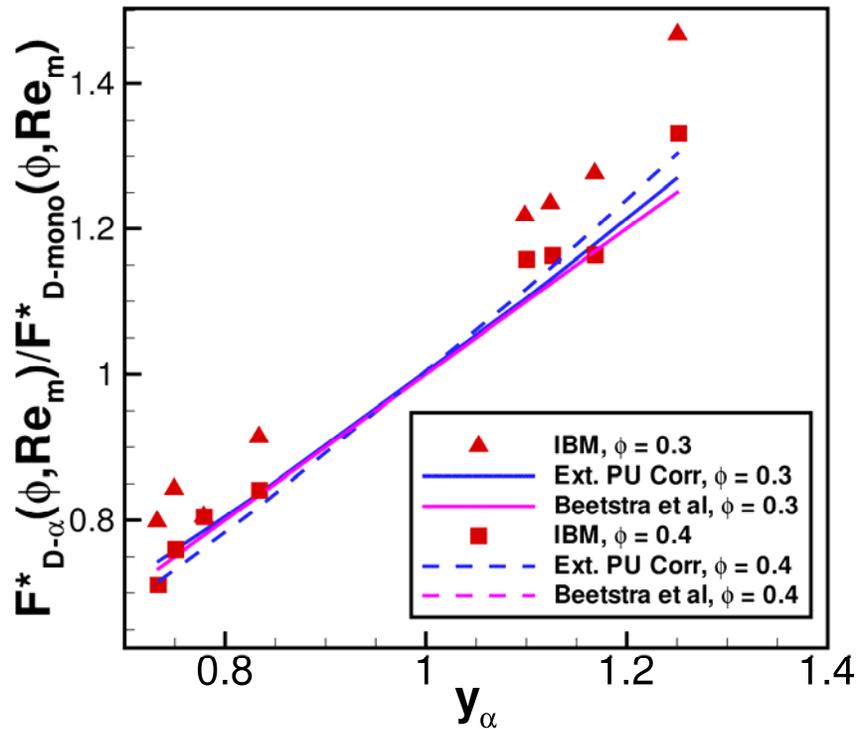
Vol. fraction = 0.2

$D_1/\Delta x : 20$
 $L/D_2 : 6$
 $MIS : 4$

- $Re_m = 50$
- $D_2/D_1 = 1.5$
- $\phi_2/\phi_1 = 0.5, 1, 2, 3, 4$

IBM simulations indicate a dependence on Reynolds number

IBM Bi-disperse Simulations : Normalized force



- $Re_m = 50$
- $D_2/D_1 = 1.5$
- $\phi_2/\phi_1 = 1, 2, 3, 4$

Vol. fraction = 0.3, 0.4

$D_1 / \Delta x : 30$
 $L/D_2 : 4$
 $MIS : 4$

- $Re_m = 65$
- $D_2/D_1 = 1.5$
- $\phi_2/\phi_1 = 1, 2, 3, 4$

Magnitude of drag also is different at higher Reynolds numbers

Current Efforts

1. Development of test particle simulations to extract coefficients of the Langevin model
2. Propose a new bi-disperse drag law at moderate Reynolds numbers
3. Data-driven exploration of the parameter space for DNS of polydisperse systems
4. Publication in preparation

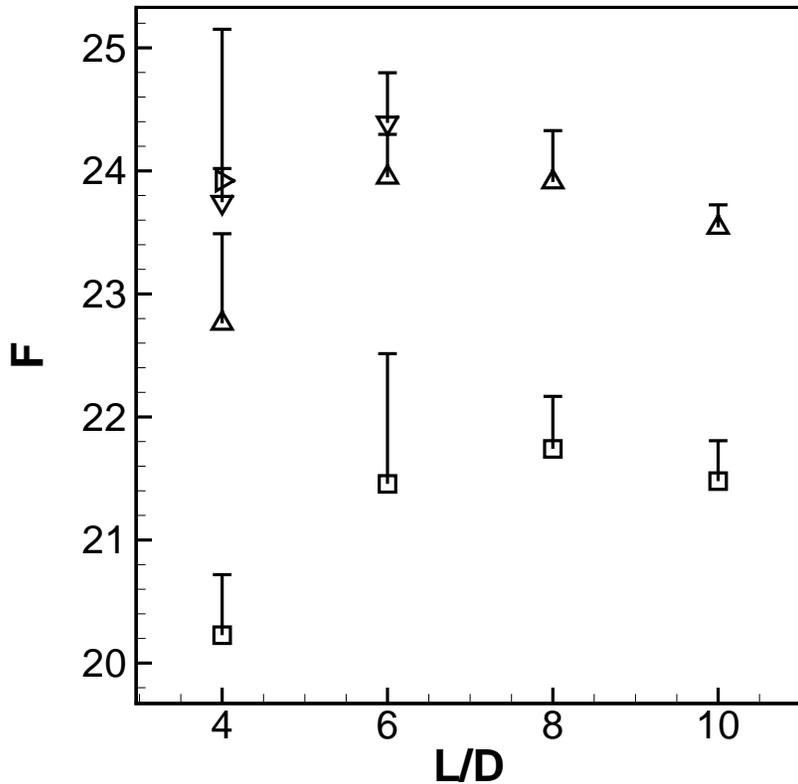
“Effect of hydrodynamic forces on particle velocity fluctuations in suspensions at moderate Reynolds numbers”. S. Tenneti, R. Garg, S. Subramaniam, R.O. Fox, C.M. Hrenya. *In preparation, to be submitted to NETL special issue journal (2009)*

Acknowledgements

- This work is supported by Department of Energy grant DE-FC26-07NT43098 through the National Energy Technology Laboratory
- Rahul Garg, Iowa State University

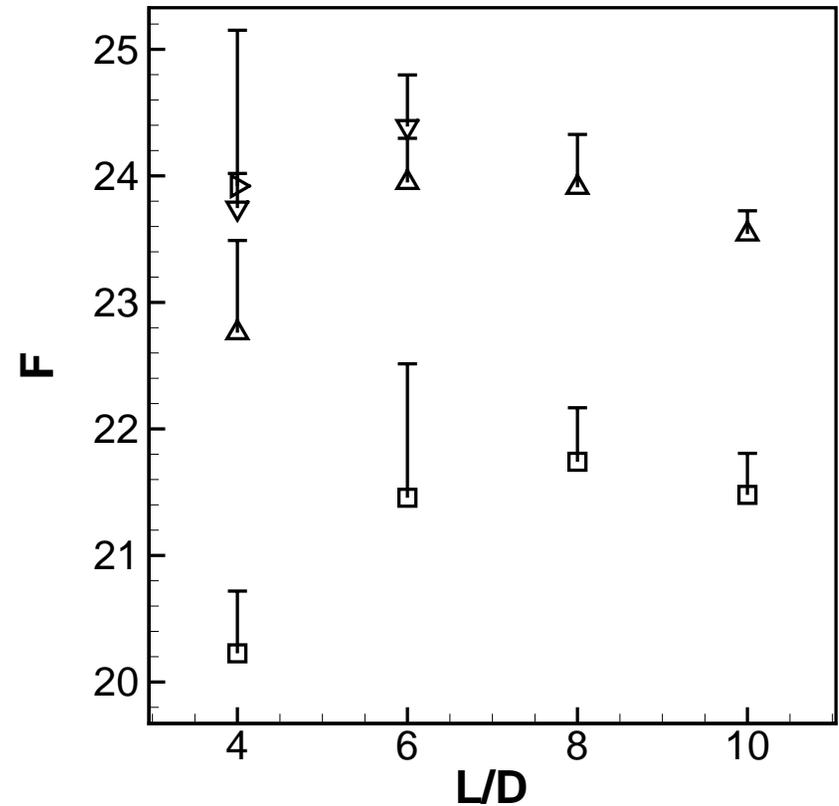
BACKUP

IBM Numerical convergence



□ $Re_m = 20$

□ $\phi = 0.3$



□ $Re_m = 20$

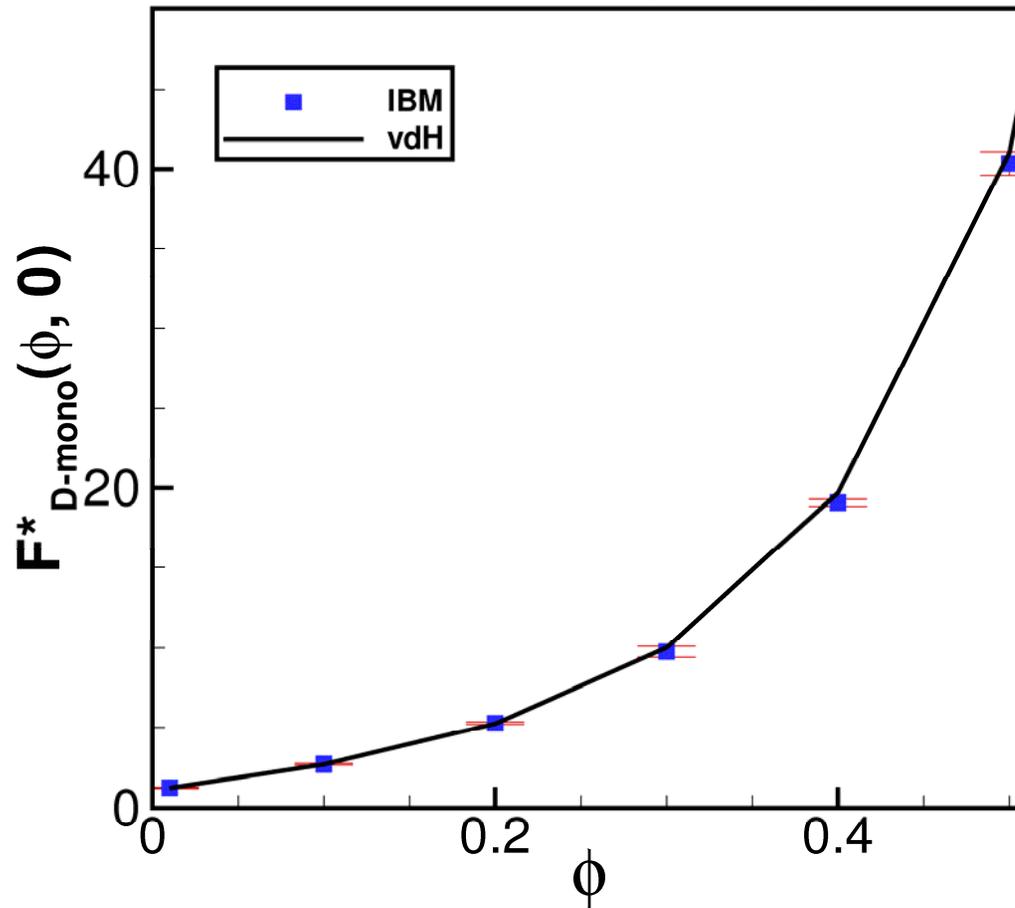
□ $\phi = 0.4$

$D / \Delta x : 10, 20, 30$

MIS : 5

IBM : Monodisperse Stokes Flow Regime

$$F_{D-\text{mono}}^*(\phi, 0) = \frac{10\phi}{(1-\phi)^2} + (1-\phi)^2 (1 + 1.5\sqrt{\phi})$$



Random arrays

Re=0.01

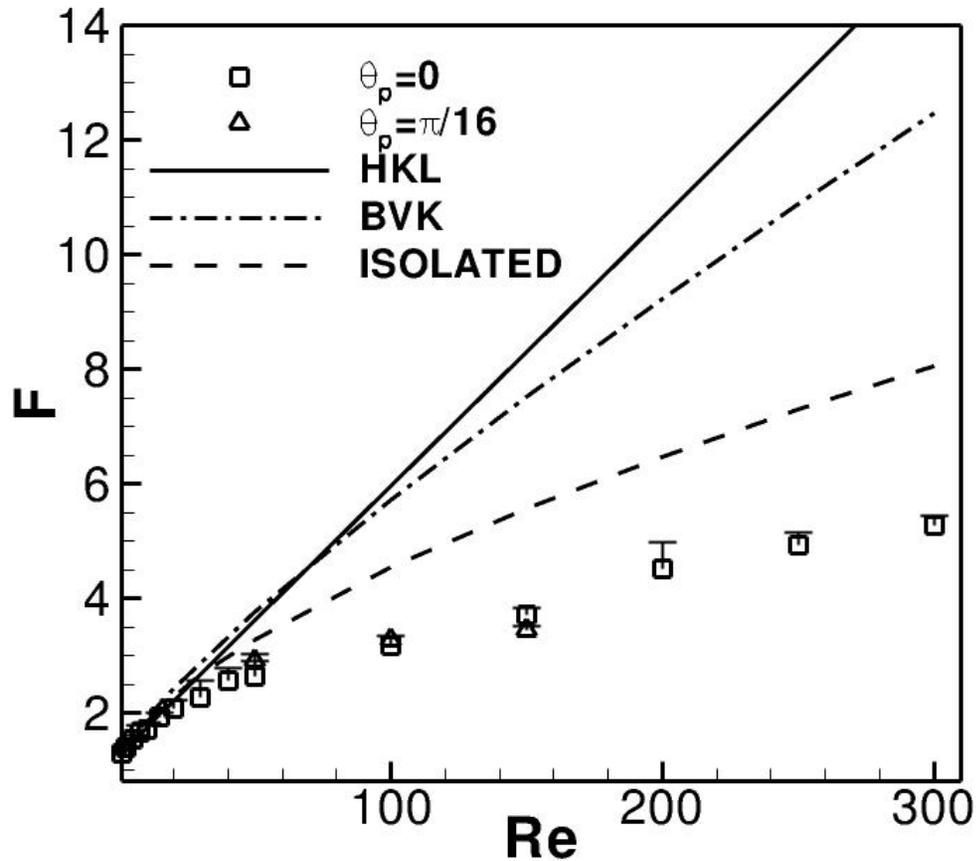
D/ Δx : 10-40

L/D : 4-15

MIS : 5

Excellent agreement of IBM with LBM drag law in Stokes regime

IBM : Monodisperse Dilute Arrays

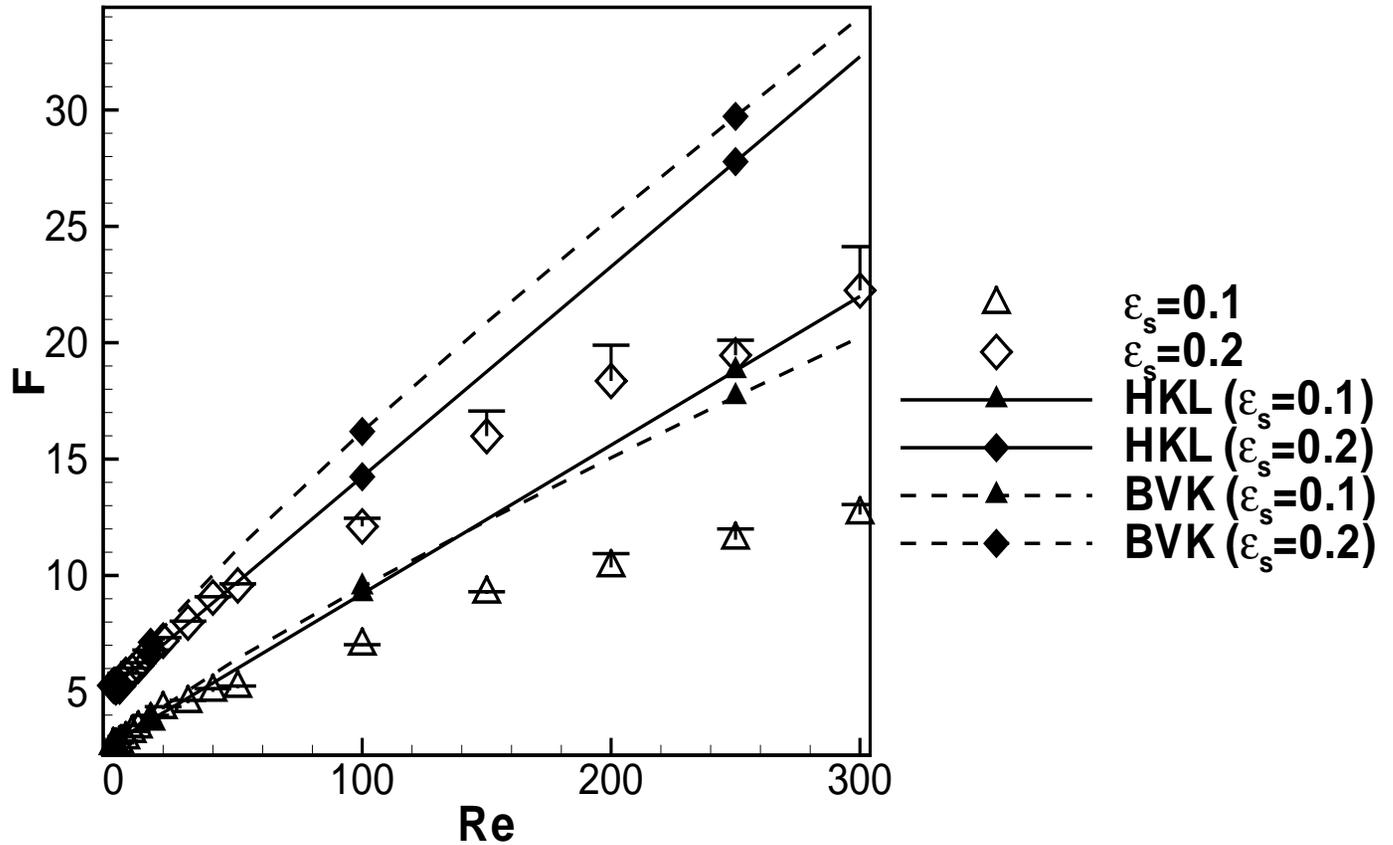


Volume Fraction = 0.01

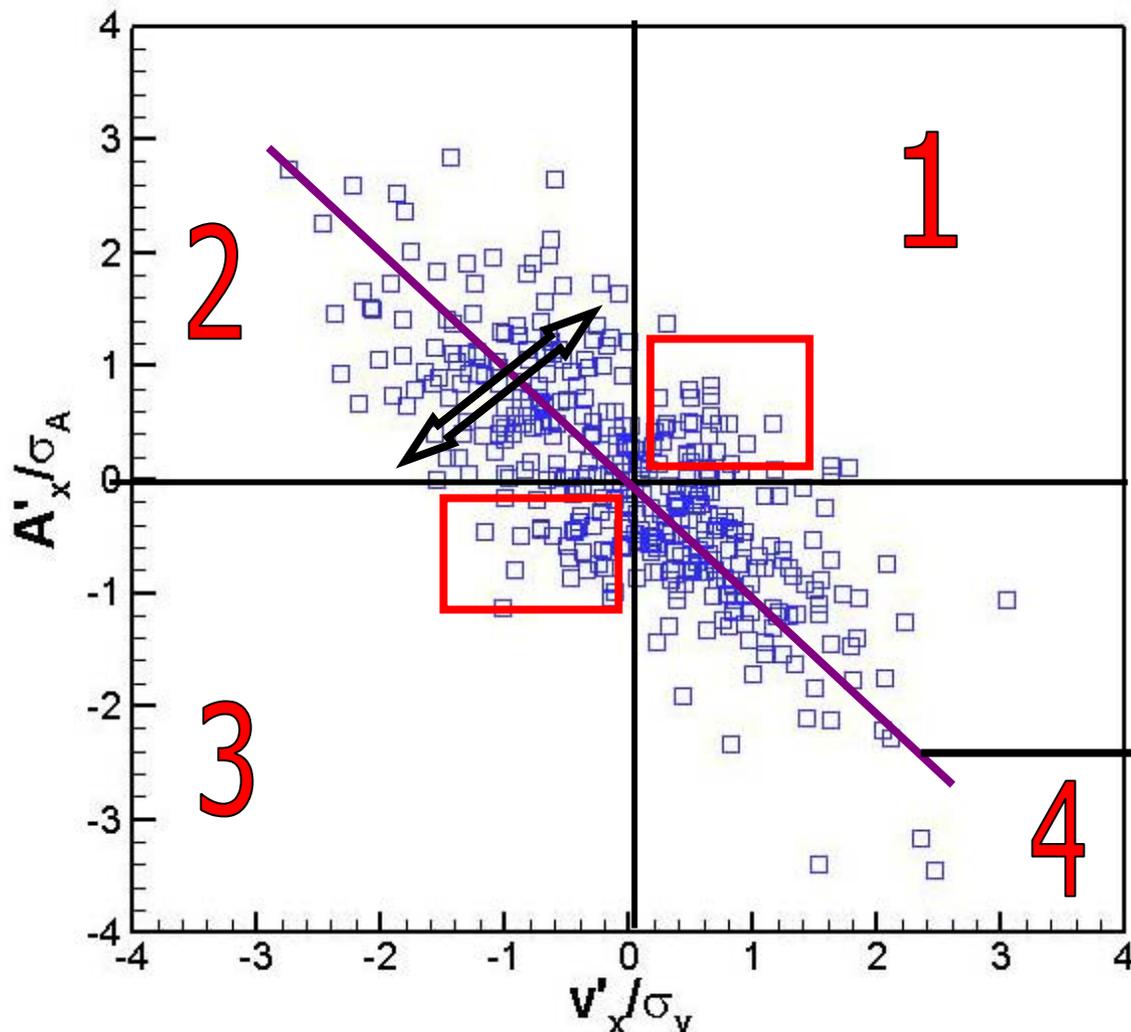
$$D / \Delta x = 10$$

$$L/D = 15$$

IBM : Monodisperse Moderately Dense Arrays



Fluctuating particle acceleration-velocity scatter



$$\begin{aligned} \text{Re}_m &= 20 \\ \text{Re}_T &= 16 \\ \phi &= 0.2 \end{aligned}$$

$$\text{Re}_m = \frac{\rho_f(1-\phi)|\langle \mathbf{W} \rangle D}{\mu_f}$$

$$\text{Re}_T = \frac{DT^{1/2}}{\nu_f}$$

Dilute Stokes flow

Note: some positive velocity fluctuations (less slip) result in *positive* acceleration fluctuations (more drag)

Drag law forms: Bi-disperse (Equal Velocities)

van der Hoef et al

$$F_{D-\alpha}^*(\phi, 0) = y_\alpha F_{D-\text{mono}}^*(\phi, 0)$$

$$y_\alpha = \frac{D_\alpha}{\langle D \rangle} \quad \langle D \rangle = \frac{\sum_{\alpha=1}^2 N_\alpha D_\alpha^3}{\sum_{\alpha=1}^2 N_\alpha D_\alpha^2}$$

Yin et al

$$F_{D-\alpha}^*(\phi, 0) = \frac{1}{1-\phi} + \left(F_{D-\text{mono}}^*(\phi, 0) - \frac{1}{1-\phi} \right) [ay_\alpha + (1-a)y_\alpha^2].$$

Drag law forms: Bi-disperse (Equal Velocities)

$$F_{D-\alpha}^* = F_{D-\alpha}^* (\phi, \text{Re}_m)$$

Mixture Reynolds number

$$\text{Re}_m = \frac{\rho_f (1-\phi) |\langle \tilde{\mathbf{V}} \rangle - \langle \mathbf{u}^{(f)} \rangle| \langle D \rangle}{\mu_f}$$

Mass-weighted mean
particle velocity

$$\langle \tilde{\mathbf{V}} \rangle = \frac{\sum_{\alpha=1}^2 \rho_{\alpha} \phi_{\alpha} \langle \mathbf{v} | r = R_{\alpha}; t \rangle}{\sum_{\alpha=1}^2 \phi_{\alpha} \rho_{\alpha}}$$

Beetstra et al

$$F_{D-\alpha}^* (\phi, \text{Re}_m) = y_{\alpha} F_{D-\text{mono}}^* (\phi, \text{Re}_m)$$

Drag law forms: Bi-disperse (Equal Velocities)

$$F_{D-\alpha}^* = F_{D-\alpha}^* (\phi, \text{Re}_m)$$

Mixture Reynolds number

$$\text{Re}_m = \frac{\rho_f (1-\phi) |\langle \tilde{\mathbf{V}} \rangle - \langle \mathbf{u}^{(f)} \rangle| \langle D \rangle}{\mu_f}$$

Mass-weighted mean
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$$\langle \tilde{\mathbf{V}} \rangle = \frac{\sum_{\alpha=1}^2 \rho_{\alpha} \phi_{\alpha} \langle \mathbf{v} | r = R_{\alpha}; t \rangle}{\sum_{\alpha=1}^2 \phi_{\alpha} \rho_{\alpha}}$$

Extension of Yin et al's drag law

$$F_{D-\alpha}^* (\phi, \text{Re}_m) = F_{D-\alpha}^* (\phi, 0) [1 + \alpha' (\phi, \text{Re}_m) \text{Re}_m]$$

Beetstra et al: Monodisperse drag law

Instantaneous particle acceleration model

$$dv_i = A_i^{(d)} dt + B_{ij} dW_j \quad \longrightarrow \quad \text{Langevin Model}$$

$$dv_i = -\beta_{(i)} \langle W_i \rangle dt - \gamma_{ij} v_j'' dt + \Sigma_{ij} dW_j$$

Second moment of particle velocity

$$\frac{d}{dt} \langle v_i'' v_j'' \rangle = \Sigma_{ik} \Sigma_{jk} - \gamma_{il} \langle v_j'' v_l'' \rangle - \gamma_{jk} \langle v_i'' v_k'' \rangle$$

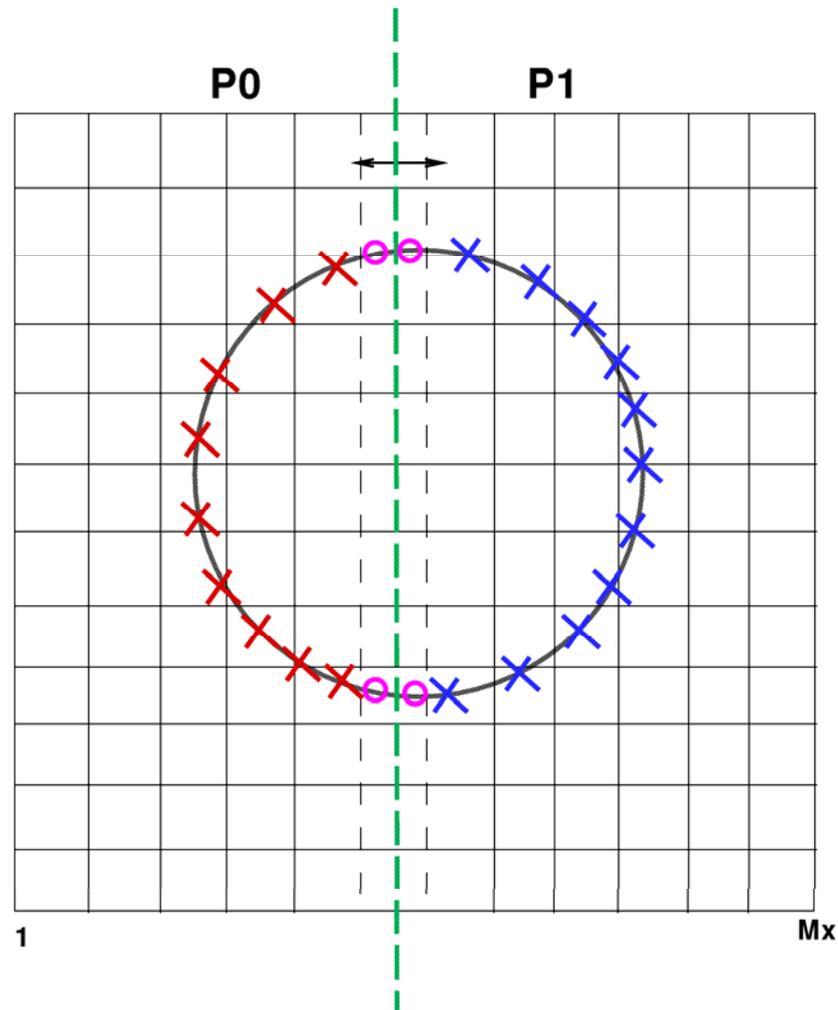
$$m \frac{d}{dt} \langle v_i'' v_j'' \rangle = S_{ij,h} - \Gamma_{ij,h}$$

Sangani &
Koch (1999)

Bidisperse Simulations: Resolution Requirements

- Smaller particles must be well resolved
 - Vol. fraction = 0.2, $Re = 100$, $D/\Delta x = 30$
- Box length should be large compared to larger diameter
 - $L/D_2 = 6$
- Number of larger particles must be large enough

Required box size: 360^3 : Need for parallel IBM



Schematic of domain decomposition

- Cyclic tridiagonal system of equations
- Previously used Gauss-Siedel : iterative solver
 - Bad for parallelization (Need to communicate in every iteration)
- Implemented direct solver based on Sherman-Morrison formula
 - Results in solving two tridiagonal systems
- Parallelized tridiagonal solver
 - Parallel partition algorithm (LANL)

Parallel IBM: Validation Study

$$\epsilon_l = \frac{1}{M_x M_y M_z} \sum_{i=1}^{M_x} \sum_{j=1}^{M_y} \sum_{k=1}^{M_z} \left| Q_l^{(p)}(i, j, k) - Q_l^{(s)}(i, j, k) \right|$$

Configuration	$\frac{D}{\Delta x}$	ϵ_1	ϵ_2	ϵ_3	ϵ_4
Simple	46.4	0.63271616E-14	0.86949156E-16	0.88190940E-16	0.41440159E-14
FCC	29.24	0.23248653E-14	0.62514764E-15	0.62624656E-15	0.13055491E-14
Monodisperse	20	0.12112644E-11	0.79061513E-12	0.86040804E-12	0.57968924E-12
Bi-disperse	15	0.30301074E-15	0.89068307E-16	0.11269121E-15	0.65079639E-16

Validity of parallel IBM established: good for production runs