

**U. S. Department of Energy – National Energy Technology Laboratory (NETL)
2023 Virtual Workshop on Multiphase Flow Science – Morgantown-WV-USA – August 1-2, 2023**

**Scale sensitive sub-grid models for effective drag,
filtered and residual stresses in fluidized gas-particle flows**

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At the 2021 NETL Multiphase Flow Science workshop we presented:

On the effect of particle Froude number in sub-grid modeling of gas-solid fluidized flows

Among the conclusions of that work we stated that:

Before new sub-grid models could be derived accounting for particle Froude number, further work would be required to account for:

higher domain average gas Reynolds numbers

a variety of domain average solid volume fractions

Those goals have been achieved, and related outcomes are now exposed in two presentations:

- 1) Results accounting for ranges of domain average gas Reynolds numbers and solid volume fractions, for a range of particle Froude numbers**
- 2) New sub-grid models for effective drag, filtered and residual stresses**

Presentation (2) follows next.

Filtered two-fluid modeling

We ultimately aim to provide sub-grid models for filtered two-fluid modeling.

$$\begin{aligned} \frac{\partial}{\partial t}(\rho_g \bar{\phi}_g) + \nabla \cdot (\rho_g \bar{\phi}_g \tilde{\mathbf{v}}_g) &= 0 & \frac{\partial}{\partial t}(\rho_s \bar{\phi}_s) + \nabla \cdot (\rho_s \bar{\phi}_s \tilde{\mathbf{v}}_s) &= 0 \\ \frac{\partial}{\partial t}(\rho_g \bar{\phi}_g \tilde{\mathbf{v}}_g) + \nabla \cdot (\rho_g \bar{\phi}_g \tilde{\mathbf{v}}_g \tilde{\mathbf{v}}_g) &= -\bar{\phi}_g \nabla \cdot \tilde{\boldsymbol{\sigma}}_g - \nabla \cdot \mathbf{r}'_g - \bar{\mathbf{M}}_I + \rho_g \bar{\phi}_g \mathbf{g} \\ \frac{\partial}{\partial t}(\rho_s \bar{\phi}_s \tilde{\mathbf{v}}_s) + \nabla \cdot (\rho_s \bar{\phi}_s \tilde{\mathbf{v}}_s \tilde{\mathbf{v}}_s) &= -\nabla \cdot \bar{\boldsymbol{\sigma}}_s - \nabla \cdot \mathbf{r}'_s - \bar{\phi}_s \nabla \cdot \tilde{\boldsymbol{\sigma}}_g + \mathbf{B}'_{gs} + \bar{\mathbf{M}}_I + \rho_s \bar{\phi}_s \mathbf{g} \\ \bar{\mathbf{M}}_I &= (1-H)\bar{\beta}(\tilde{\mathbf{v}}_g - \tilde{\mathbf{v}}_s) \\ \tilde{\boldsymbol{\sigma}}_g &= \left[\tilde{\mathbf{P}}_g - (\lambda_g + \frac{2}{3}\mu_g)(\nabla \cdot \tilde{\mathbf{v}}_g) \right] \mathbf{I} - 2\mu_g \tilde{\boldsymbol{\mathbf{s}}}_g \\ \bar{\boldsymbol{\sigma}}_s &= \left[\bar{\mathbf{P}}_s - (\lambda_s + \frac{2}{3}\mu_s)(\nabla \cdot \mathbf{v}_s) \right] \mathbf{I} - 2\bar{\mu}_s \tilde{\boldsymbol{\mathbf{s}}}_s = \mathbf{P}_{\text{fil},s} \mathbf{I} - 2\mu_{\text{fil},s} \tilde{\boldsymbol{\mathbf{s}}}_s \\ \mathbf{r}'_l &= \rho_l \bar{\phi}_l \tilde{\mathbf{v}}_l \tilde{\mathbf{v}}_l - \rho_l \bar{\phi}_l \tilde{\mathbf{v}}_l \tilde{\mathbf{v}}_l = \mathbf{P}_{\text{res},l} \mathbf{I} - 2\mu_{\text{res},l} \tilde{\boldsymbol{\mathbf{s}}}_l \\ \tilde{\boldsymbol{\mathbf{s}}}_l &= \frac{1}{2} \left[\nabla \tilde{\mathbf{v}}_l + (\nabla \tilde{\mathbf{v}}_l)^T \right] - \frac{1}{3} (\nabla \cdot \tilde{\mathbf{v}}_l) \mathbf{I} \end{aligned}$$

Effective, filtered and residual closures

$$H = 1 - \frac{\beta_{\text{eff}}}{\bar{\beta}} \quad \beta_{\text{eff}} = \frac{\overline{\beta(\mathbf{v}_g - \mathbf{v}_s)}}{(\tilde{\mathbf{v}}_g - \tilde{\mathbf{v}}_s)}$$

$$\mathbf{P}_{\text{fil},s} = \frac{1}{3} \text{tr} \left[\bar{\mathbf{P}}_s - (\lambda_s + \frac{2}{3}\mu_s)(\nabla \cdot \mathbf{v}_s) \right]$$

$$\mu_{\text{fil},s} = \bar{\mu}_s$$

$$\mathbf{P}_{\text{res},l} = \frac{1}{3} \text{tr}(\mathbf{r}'_l)$$

$$\mu_{\text{res},l} = \frac{|\mathbf{r}'_{\text{shear},l}|}{2|\tilde{\boldsymbol{\mathbf{s}}}_{\text{shear},l}|}$$

We go for effective, filtered and residual parameters by filtering over predictions from highly resolved simulations (HRS) with microscopic two-fluid modeling.



Microscopic two-fluid modeling

On the basis of Anderson and Jackson' formulation, with microscopic closures as implemented into the MFI code by Agrawal et al. (2001).

$$\frac{\partial}{\partial t}(\rho_g \phi_g) + \nabla \cdot (\rho_g \phi_g \mathbf{v}_g) = 0$$

$$\frac{\partial}{\partial t}(\rho_s \phi_s) + \nabla \cdot (\rho_s \phi_s \mathbf{v}_s) = 0$$

$$\frac{\partial}{\partial t}(\rho_g \phi_g \mathbf{v}_g) + \nabla \cdot (\rho_g \phi_g \mathbf{v}_g \mathbf{v}_g) = -\phi_g \nabla \cdot \boldsymbol{\sigma}_g - \mathbf{M}_I + \rho_g \phi_g \mathbf{g}$$

$$\frac{\partial}{\partial t}(\rho_s \phi_s \mathbf{v}_s) + \nabla \cdot (\rho_s \phi_s \mathbf{v}_s \mathbf{v}_s) = -\nabla \cdot \boldsymbol{\sigma}_s - \phi_s \nabla \cdot \boldsymbol{\sigma}_g + \mathbf{M}_I + \rho_s \phi_s \mathbf{g}$$

$$\mathbf{M}_I = \beta(\mathbf{v}_g - \mathbf{v}_s)$$

$$\boldsymbol{\sigma}_\ell = \left[P_\ell - \left(\lambda_\ell + \frac{2}{3} \mu_\ell \right) (\nabla \cdot \mathbf{v}_\ell) \right] \mathbf{I} - 2\mu_\ell \mathbf{s}_\ell$$

$$\mathbf{s}_\ell = \frac{1}{2} \left[\nabla \mathbf{v}_\ell + (\nabla \mathbf{v}_\ell)^T \right] - \frac{1}{3} (\nabla \cdot \mathbf{v}_\ell) \mathbf{I}$$

Microscopic closures

Drag

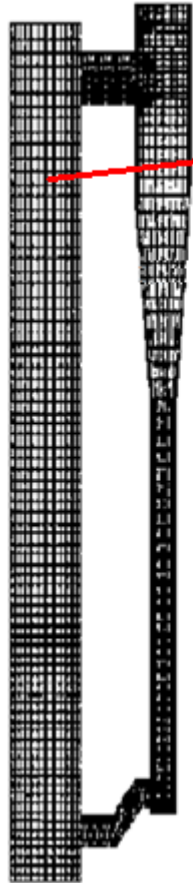
Wen and Yu (1966)

Solid phase pressure and viscous stresses

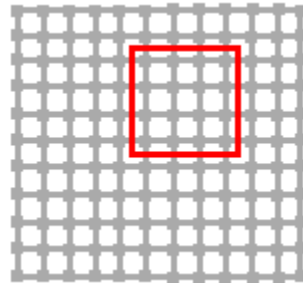
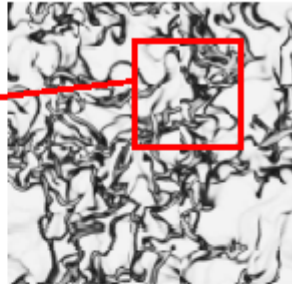
Lun et al. (1984),

as adapted by Agrawal et al. (2001)

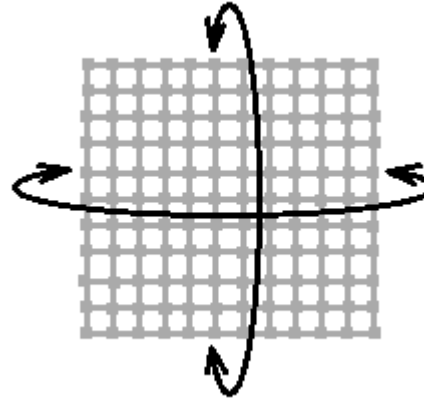
Highly resolved simulations (MFIx) / filtering



LSS



HRS



- All periodic boundaries
- 16 x 16 cm domain
- 128 x 128 grids
- 1.25 x 1.25 mm grid cells
- up to 4 x 4 cm filter sizes

d_p	40 / 75 / 150 / 300 μm
$Fr_p = v_t^2 / (gd_p)$	12.21 / 64.85 / 286.69 / 799.22
ρ_s	1500 kg/m^3
e	0.9
ρ_g	1.3 kg/m^3
μ_g	1.8×10^{-5} kg/(m s)
$\langle \phi_s \rangle$	0.05 / 0.15 / 0.25 / 0.35 / 0.45 / 0.55
$\langle Re_g \rangle / \langle Re_g \rangle_{\text{susp}}$	1 / 8.15 / 16.30 / 24.45

Filtered data classified
by narrow ranges of

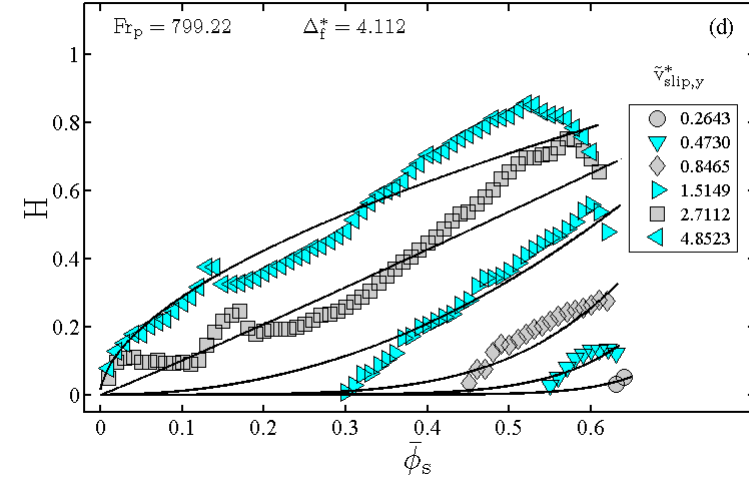
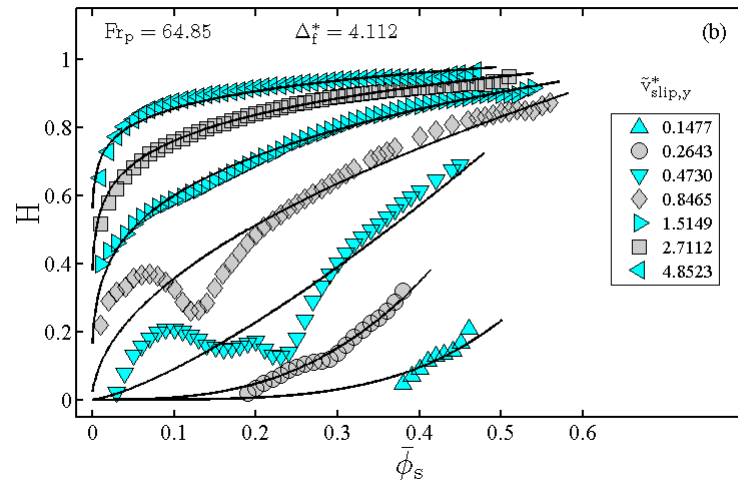
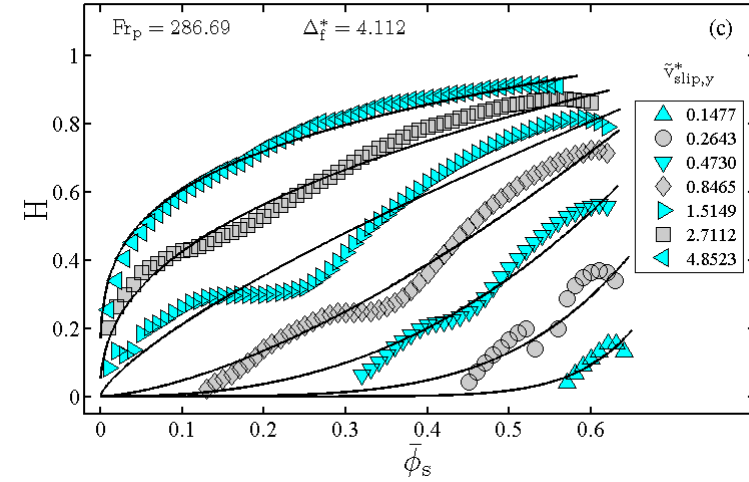
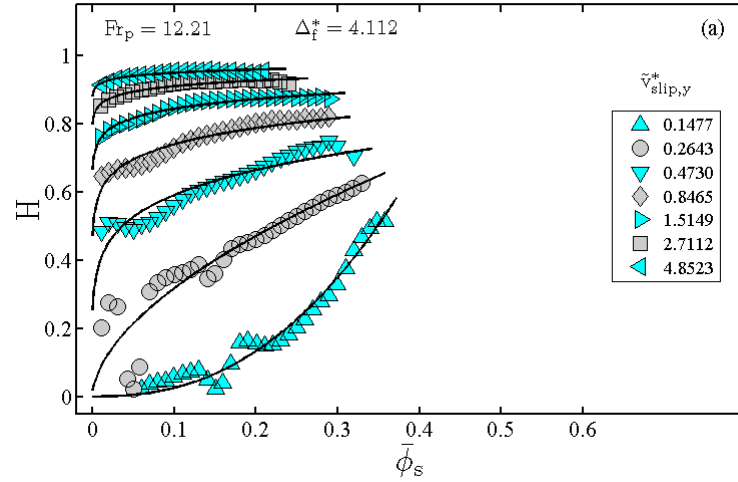
$$\bar{\phi}_s, \tilde{V}_{\text{slip},y}$$

Results

$$\tilde{v}_{slip,y}^* = \left| \tilde{v}_{slip,y} / v_{t75} \right|$$

$$\Delta_f^* = \Delta_f / (v_{t75}^2 / g)$$

$$Fr_p = v_t^2 / (gd_p)$$





$$H = \min[H_1, H_2]$$

$$H_1 = \left(a_1 + \frac{a_2}{(\tilde{v}_{slip,y}^*)} + \frac{a_3}{(\tilde{v}_{slip,y}^*)^2} \right) \bar{\phi}_s \left(b_1 + \frac{b_2}{(\tilde{v}_{slip,y}^*)} + \frac{b_3}{(\tilde{v}_{slip,y}^*)^2} \right)$$

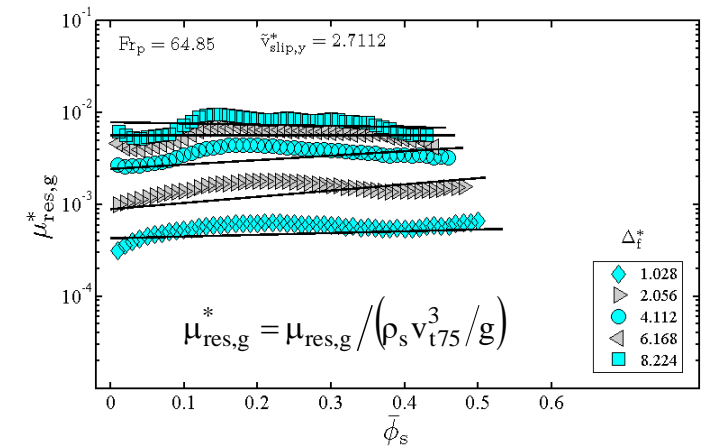
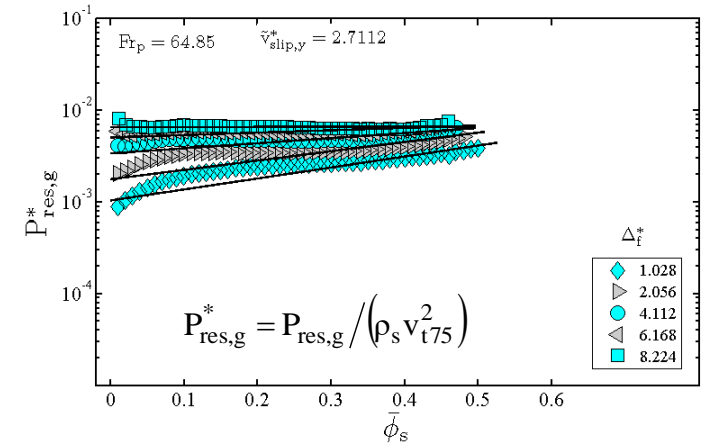
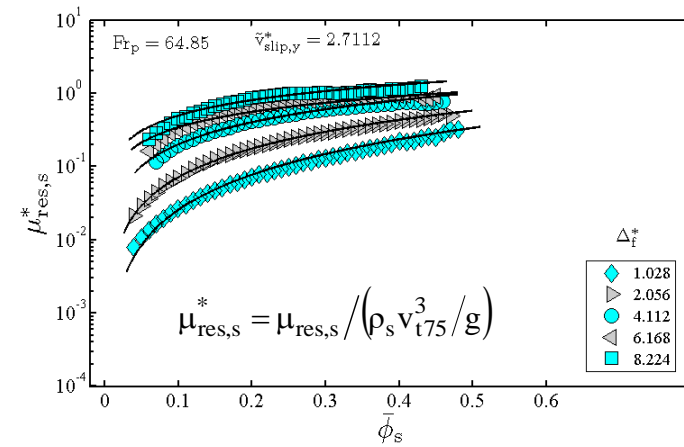
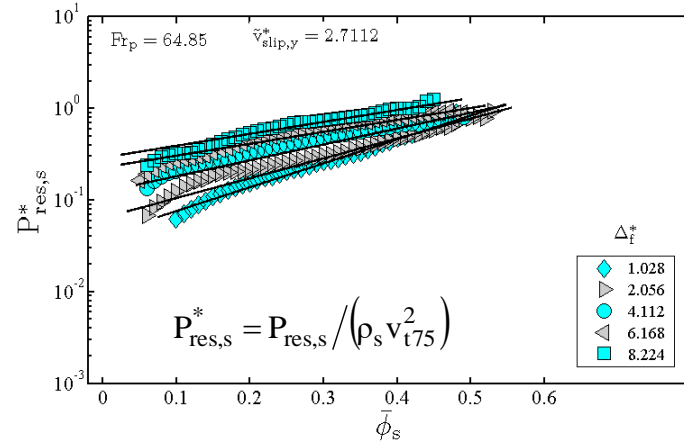
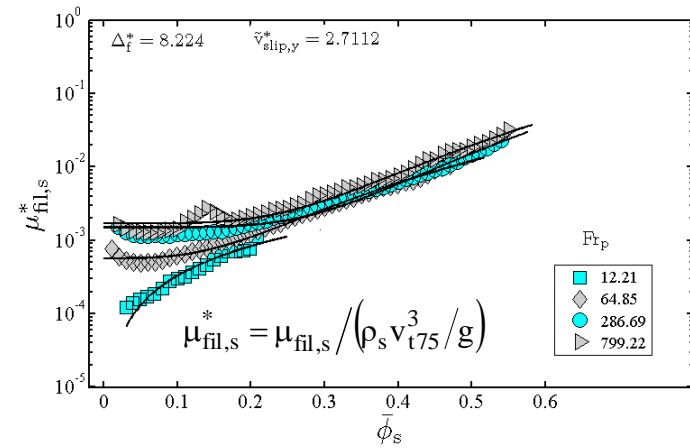
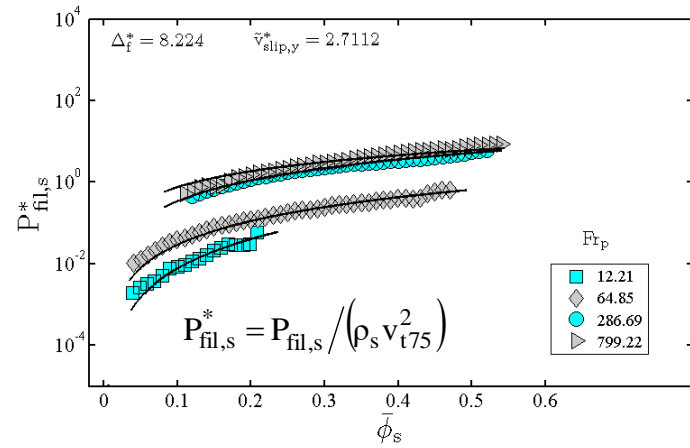
$$H_2 = c_1 + c_2 \bar{\phi}_s$$

The various coefficients were presented in numerical form since no precise enough mathematical rules of correlation could be found.

$\tilde{v}_{slip,y}^* < 0.8$

$\tilde{v}_{slip,y}^* \geq 0.8$

Fr _p	Δ _f *	a ₁	a ₂	a ₃	b ₁	b ₂	b ₃	c ₁	c ₂	Fr _p	Δ _f *	a ₁	a ₂	a ₃	b ₁	b ₂	b ₃	c ₁	c ₂
12.21	1.028	0.9854	-0.2496	0.02155	0.2483	-0.04851	0.01608	1.6250	-3.1250	12.21	1.028	0.9936	0.01384	-0.1084	0.00261	0.2264	-0.03508	1.6250	-3.1250
	2.056	1.3470	-0.3945	0.07045	0.2037	-0.08908	0.04547	3.0769	-7.6923		2.056	1.0030	-0.06721	0.001259	0.0006804	0.1089	0.005454	3.0769	-7.6923
	4.112	3.0220	-1.6890	0.3178	0.2862	-0.2219	0.07850	1.9583	-4.1667		4.112	1.0050	-0.1141	0.02910	0.0007523	0.06413	0.009367	1.9583	-4.1667
	6.168	3.2150	-1.7820	0.3271	0.3906	-0.2859	0.08462	1.7143	-3.5714		6.168	0.9815	-0.04778	-0.005081	-0.008357	0.07968	-0.004212	1.7143	-3.5714
	8.224	3.3330	-1.8660	0.3409	0.4024	-0.2989	0.08604	1.5313	-3.1250		8.224	1.0260	-0.2034	0.1048	0.005166	0.02557	0.0320	1.5313	-3.1250
64.85	1.028	1.7040	-0.5144	0.05743	1.5330	-0.3026	0.0701	5.0833	-8.3333	64.85	1.028	1.0890	-0.05508	0.02728	0.05604	0.7911	-0.07715	5.0833	-8.3333
	2.056	5.9680	-3.2330	0.5615	1.4590	-0.3163	0.1287	14.2500	-25.0000		2.056	1.0690	-0.06079	0.1826	0.03451	0.4012	0.1231	14.2500	-25.0000
	4.112	-5.2010	4.3050	-0.4414	-1.5430	1.6020	-0.1158	17.6667	-33.3333		4.112	1.0380	-0.03234	0.1445	0.02561	0.2478	0.1500	17.6667	-33.3333
	6.168	-8.9100	5.5750	-0.2332	-2.2630	1.8720	-0.1006	10.0000	-20.0000		6.168	1.0360	-0.08281	0.1976	0.02734	0.1685	0.1924	10.0000	-20.0000
	8.224	45.1000	-34.5600	6.6860	-1.2540	1.1070	0.03978	16.3333	-33.3333		8.224	1.0320	-0.1124	0.2470	0.02838	0.1181	0.2365	16.3333	-33.3333
286.69	1.028	0.8370	-0.2250	0.01917	1.6980	-0.1343	0.01282	4.6154	-7.6923	286.69	1.028	1.3570	-1.6210	0.6979	0.6291	0.1795	-0.02427	4.6154	-7.6923
	2.056	2.5740	-0.5549	0.1240	1.0660	0.7769	0.01280	8.8571	-14.2857		2.056	1.1690	-0.1748	0.4089	0.1745	1.0320	0.2343	8.8571	-14.2857
	4.112	3.6660	-1.6480	0.3834	0.4496	0.7892	0.07359	12.8000	-20.0000		4.112	1.0970	-0.1998	0.5041	0.07632	0.7575	0.3945	12.8000	-20.0000
	6.168	10.3600	-6.9670	1.4820	-0.3457	1.2150	0.08068	21.3333	-33.3333		6.168	1.0510	-0.02845	0.4537	0.02634	0.8111	0.3702	21.3333	-33.3333
	8.224	21.9600	-16.1200	3.2460	-1.0910	1.6000	0.07133	16.0000	-25.0000		8.224	1.0170	0.1761	0.2892	-0.009139	0.9301	0.2604	16.0000	-25.0000
799.22	1.028	40.4200	-22.9900	3.5350	11.5700	-3.2870	0.5080	1.1636	-1.8182	799.22	1.028	1.1400	-2.9500	2.3110	2.0100	-5.8210	5.7340	1.1636	-1.8182
	2.056	0.1136	0.4505	-0.01229	0.08879	1.8980	-0.1001	4.5714	-7.1429		2.056	0.8188	1.9540	-0.8404	-0.02071	4.4780	-0.9635	4.5714	-7.1429
	4.112	1.6550	0.9124	0.04843	1.0920	3.2500	-0.1753	12.4000	-20.0000		4.112	1.1010	-0.6132	1.7310	0.07201	2.0530	1.5660	12.4000	-20.0000
	6.168	24.1600	-51.7000	28.4400	3.3310	-3.3320	3.6430	8.8571	-14.2857		6.168	0.5055	3.9500	-1.7140	-0.4612	5.4510	-1.0690	8.8571	-14.2857
	8.224	0.2778	3.1140	0.2333	-0.4787	4.8970	-0.04863	10.3333	-16.6667		8.224	0.8726	1.6800	1.0220	-0.3450	4.5040	0.1909	10.3333	-16.6667



$$P_{fil,s}^* ; \mu_{fil,s}^* ; \mu_{res,s}^* = \left[a_1 (\tilde{v}_{slip,y}^*)^{a_2} + a_3 \right] \bar{\phi}_s \left[b_1 (\tilde{v}_{slip,y}^*)^{b_2} + b_3 \right] + c_1$$

$$P_{res,s}^* = \left[a_1 (\tilde{v}_{slip,y}^*)^{a_2} + a_3 \right] \exp \left\{ \left[b_1 (\tilde{v}_{slip,y}^*)^{b_2} + b_3 \right] \bar{\phi}_s \right\}$$

$$P_{res,g}^* ; \mu_{res,g}^* = \left[a_1 (\tilde{v}_{slip,y}^*)^2 + a_2 (\tilde{v}_{slip,y}^*) + a_3 \right] \exp \left\{ \left[b_1 (\tilde{v}_{slip,y}^*)^{b_2} + b_3 \right] \bar{\phi}_s \right\}$$

Aiming for a smaller set of equations, we correlate only to a most influential third parameter (in addition to the meso-scale markers):

$$\begin{aligned} Fr_p & \text{ for } P_{fil,s}^* \mu_{fil,s}^* \\ \Delta_f^* & \text{ for } P_{res,s}^* \mu_{res,s}^* P_{res,g}^* \mu_{res,g}^* \end{aligned}$$

The various coefficients were presented in numerical form since no precise enough mathematical rules of correlation could be found.

	Fr_p	a_1	a_2	a_3	b_1	b_2	b_3	c_1	Δ_f^*	a_1	a_2	a_3	b_1	b_2	b_3	c_1
$P_{fil,s}^*$	12.21	5.805e-2	3.505e0	5.811e-4	-2.963e1	-3.012e-2	3.115e1	0.0	1.028	1.070e-2	1.321e0	1.742e-3	0.0	0.0	6.000e0	-
	64.85	2.648e-1	2.251e0	0.0	-2.887e0	-2.709e-1	4.135e0	0.0	2.056	3.422e-1	6.362e-2	-3.008e-1	0.0	0.0	5.000e0	-
	286.69	2.302e-6	1.567e1	2.846e0	2.626e-1	1.477e0	5.498e-1	0.0	4.112	3.682e-1	8.971e-2	-2.806e-1	0.0	0.0	4.000e0	-
	799.22	0.0	0.0	1.490e1	0.0	0.0	1.2957e0	0.0	6.168	2.261e-1	2.381e-1	-6.050e-2	0.0	0.0	3.000e0	-
	12.21	2.831e-3	1.000e0	-3.427e-4	-7.160e-2	-1.440e0	1.366e0	0.0	8.224	9.972e-2	6.553e-1	9.870e-2	0.0	0.0	3.000e0	-
$\mu_{fil,s}^*$	64.85	3.987e-2	1.000e0	8.183e-3	-1.004e-1	1.211e0	3.669e0	5.700e-4	1.028	3.712e-1	1.000e0	1.508e-2	5.193e0	5.000e-2	-3.856e0	0.0
	286.69	2.295e-1	8.817e-1	-6.087e-2	-1.049e-1	-2.600e0	5.162e0	1.500e-3	2.056	1.225e0	3.900e-1	-3.836e-1	-1.282e-1	-9.438e-1	1.350e0	0.0
	799.22	2.408e-3	4.990e0	6.539e-2	4.940e-1	1.517e0	2.280e0	1.700e-3	4.112	1.131e0	3.725e-1	3.709e-1	0.0	0.0	1.000e0	0.0
	12.21	2.831e-3	1.000e0	-3.427e-4	-7.160e-2	-1.440e0	1.366e0	0.0	6.168	4.255e-1	4.626e-1	1.041e0	0.0	0.0	7.000e-1	0.0
	64.85	3.987e-2	1.000e0	8.183e-3	-1.004e-1	1.211e0	3.669e0	5.700e-4	8.224	1.903e-1	1.326e0	1.769e0	0.0	0.0	7.000e-1	0.0
$P_{res,s}^*$	286.69	2.295e-1	8.817e-1	-6.087e-2	-1.049e-1	-2.600e0	5.162e0	1.500e-3	1.028	1.847e-4	-1.943e-4	2.113e-4	-4.935e-1	-1.565e0	2.865e0	-
	799.22	2.408e-3	4.990e0	6.539e-2	4.940e-1	1.517e0	2.280e0	1.700e-3	2.056	2.423e-4	-2.688e-4	7.239e-4	-5.492e0	-3.458e-1	6.235e0	-
	12.21	2.831e-3	1.000e0	-3.427e-4	-7.160e-2	-1.440e0	1.366e0	0.0	4.112	1.263e-3	-3.373e-3	3.259e-3	-2.817e0	-7.000e-1	2.681e0	-
	64.85	3.987e-2	1.000e0	8.183e-3	-1.004e-1	1.211e0	3.669e0	5.700e-4	6.168	1.778e-3	-4.722e-3	4.818e-3	-2.825e0	-7.000e-1	2.003e0	-
	286.69	2.295e-1	8.817e-1	-6.087e-2	-1.049e-1	-2.600e0	5.162e0	1.500e-3	8.224	2.106e-3	-5.653e-3	6.413e-3	-4.537e0	-4.500e-1	2.935e0	-
$\mu_{res,g}^*$	799.22	2.408e-3	4.990e0	6.539e-2	4.940e-1	1.517e0	2.280e0	1.700e-3	1.028	-6.672e-6	1.416e-4	9.345e-5	-6.489e1	-1.000e-2	6.469e1	-
	12.21	2.831e-3	1.000e0	-3.427e-4	-7.160e-2	-1.440e0	1.366e0	0.0	2.056	7.645e-5	-6.980e-5	5.178e-4	-2.988e0	-4.388e-1	3.485e0	-
	64.85	3.987e-2	1.000e0	8.183e-3	-1.004e-1	1.211e0	3.669e0	5.700e-4	4.112	6.027e-4	-1.523e-3	2.137e-3	-1.132e0	-1.037e0	1.519e0	-
	286.69	2.295e-1	8.817e-1	-6.087e-2	-1.049e-1	-2.600e0	5.162e0	1.500e-3	6.168	1.405e-3	-3.069e-3	3.685e-3	-1.379e-2	-4.000e0	-1.336e-3	-
	799.22	2.408e-3	4.990e0	6.539e-2	4.940e-1	1.517e0	2.280e0	1.700e-3	8.224	1.823e-3	-3.886e-3	5.017e-3	-1.536e-1	-2.078e0	-2.829e-1	-

Conclusions

- Accounting of the proposed sub-grid models:
 - ✓ At the meso-scale: $\bar{\phi}_s$, $\tilde{v}_{slip,y}^*$ and Δ_f^*
 - ✓ At the micro-scale: Fr_p
 - ✓ At the macro-scale: $\langle \phi_s \rangle$ and $\langle Re_g \rangle$ (in average)
- Possible improvements:
 - ✓ Account for alternative or additional micro-scale markers
 - ✓ Further assess number and type of meso-scale markers
 - ✓ Correlate to macro-scale markers

(refer to companion presentation for behavior analysis)

Acknowledgements

This work was supported by
CNPq FAPESP

Please direct any questions to milioli@sc.usp.br

Thank you very much!

