

Uncertainty Quantification Tools for Multiphase Flow Simulations using MFIX

X. Hu¹, A. Passalacqua², R. O. Fox¹

¹Iowa State University, Department of Chemical and Biological Engineering, Ames, IA

²Iowa State University, Department of Mechanical Engineering, Ames, IA

Project Manager: Charles Miller

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Outline

- 1 Introduction and background
- 2 Project objectives and milestones
- 3 Technical progress
- 4 Future work

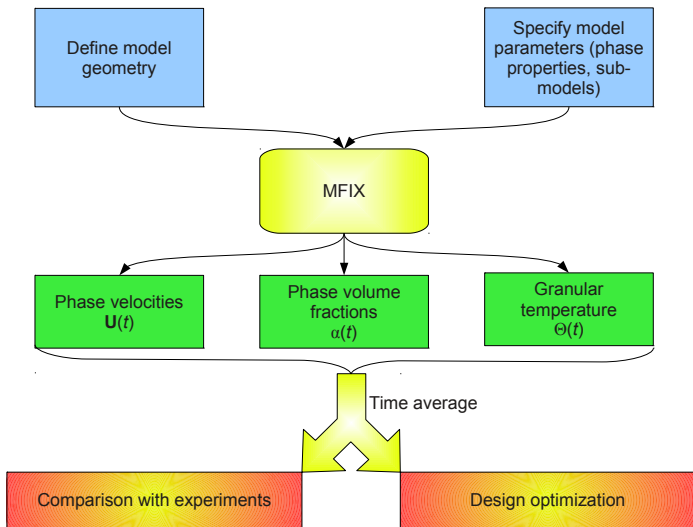
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Eulerian multiphase models for gas-particle flows

- Widely used in both academia and industry
- Computationally efficient
- Applicable to real-world cases (gasifiers, combustors, ...)
- Directly provide averaged quantities of interest in design and optimization studies
- Implemented into many computational codes
 - Commercial codes, eg. FLUENT
 - Open-source codes, eg. OpenFoam®[®], MFIX

Typical steps in a simulation project with MFIx



Motivations and objectives

Models and uncertainty

- Models present a strongly non-linear relation between inputs and outputs
- Input parameters are affected by uncertainty
 - Experimental inputs
 - Theoretical assumptions
- Need to quantify the effect of uncertainty on the simulation results

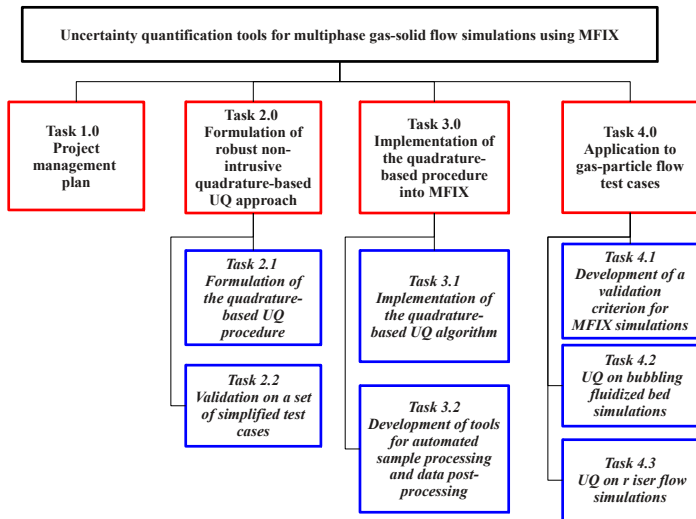
Main objectives

- Develop an efficient quadrature-based uncertainty quantification (UQ) procedure
- Develop a reconstruction procedure for the probability distribution function (PDF) of the system response
- Apply such a procedure to multiphase gas-particle flow simulations considering parameters of interest in applications

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Project tasks



Project milestones and current status

Milestone n.	Description	Due on	Status
1	Submission of project management plan	Dec. 30, 2011	Completed
2	Formulation of the quadrature-based UQ procedure	Jul. 1, 2012	Completed
3	Validation of the quadrature-based UQ procedure on simplified test-cases	Oct. 1, 2012	Completed
4	Implementation of the quadrature-based UQ algorithm into MFIX	May 31, 2013	Completed
5	Development of automated tools for processing input/output data	Oct. 1, 2013	Completed
6	Development of a Validation Criterion for MFIX Simulations	Jan. 3, 2014	Completed
7	UQ on bubbling fluidized bed simulations	Mar. 31, 2014	Completed
8	UQ on riser flow simulations	Sept. 1, 2014	In progress, on time
9	Preparation of final report	Sept 31, 2014	Starts on Sept. 1, 2014

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Basic concepts

- We study **propagation** of uncertainty from inputs to outputs
- Sources of uncertainty
 - Domain geometry
 - Initial/boundary conditions
 - Model inputs and parameters
 - Model assumptions
- **Non-intrusive** approach
 - Sample the space of the uncertain input parameters of the model
 - Obtain simulation results for each sample
 - The moments (statistics) of the model results are the quantity of interest
 - Low-order statistics for practical purposes (mean, variance, ...)
 - Reconstructed PDF of the response

Quadrature-based UQ procedure – univariate case

- Consider a probability space $\mathcal{P}(\Omega, \mathfrak{F}, P)$
 - A sample space Ω , a σ -algebra \mathfrak{F} , and a probability measure P
 - **One random variable** (uncertain parameter) ξ
 - A random process $\kappa(\xi, x)$
- The objective is to compute the moments of the random process:

$$m_n = \int_{\Omega} \kappa(\xi, x)^n p(\xi) d\xi$$

Direct quadrature approach

- Sample Ω using Gaussian quadrature formulae.
- Evaluate the model for each sample to obtain abscissas.
- Approximate moments directly using quadrature weights and abscissas.

Quadrature-based UQ procedure – univariate case

- Use Gaussian quadrature formula, the moments about the origin of the response can be approximated as

$$m_n = \int_{\Omega} \kappa(\xi, x)^n p(\xi) d\xi \approx \sum_{i=1}^M w_i(x) [\kappa(\xi_i, x)]^n$$

where

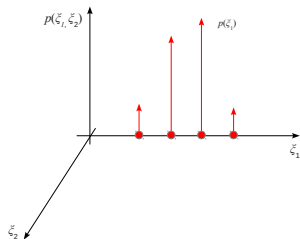
- M is the number of nodes
- $w_i(x)$ are the quadrature weights
- ξ_i are the quadrature nodes

Weight functions

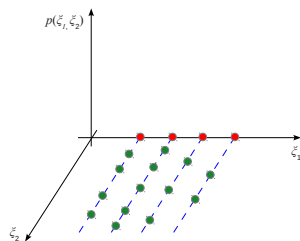
The form of $p(\xi)$ depends on the assumed PDF of the uncertain parameter (uniform, Gaussian, ...)

Quadrature-based UQ procedure – multivariate case

- Sampling procedure for a bivariate case ($\xi = \xi_1, \xi_2$) using CQMOM



Find weights w_{l_1} and nodes ξ_{1,l_1}



Use conditional moments $\langle \xi_2^j \rangle_{l_1}$ to find weights w_{l_1, l_2} and nodes ξ_{2, l_1, l_2}

Moments of the system response

$$\langle \kappa^n(\xi) \rangle = \int_{\mathbb{R}^2} [\kappa(\xi)]^n p(\xi) d\xi \approx \sum_{l_1=1}^{M_1} \sum_{l_2=1}^{M_2} w_{l_1} w_{l_1, l_2} [\kappa(\xi_{1, l_1}, \xi_{2, l_1, l_2})]^n$$

Low-order statistics

- Central moments

$$\mu_n = \sum_{i=0}^n \binom{n}{j} (-1)^{n-i} m_i \mu^{n-j}$$

- Mean

$$\mu = m_1/m_0$$

- Variance (distance from the mean)

$$\sigma^2 = \frac{m_2}{m_0} - \mu^2$$

- Skewness (symmetry of the distribution)

$$\gamma_1 = \frac{\mu_3}{\sigma^3} = \frac{m_3/m_0 - 3\mu m_2/m_0 + 2\mu^3}{\sigma^3}$$

- Kurtosis (importance of tails)

$$\gamma_2 = \frac{\mu_4}{\sigma^4} = \frac{m_4/m_0 - 4\mu m_3/m_0 + 6\mu^2 m_2/m_0 - 3\mu^4}{\sigma^4}$$

Reconstruction of the response PDF

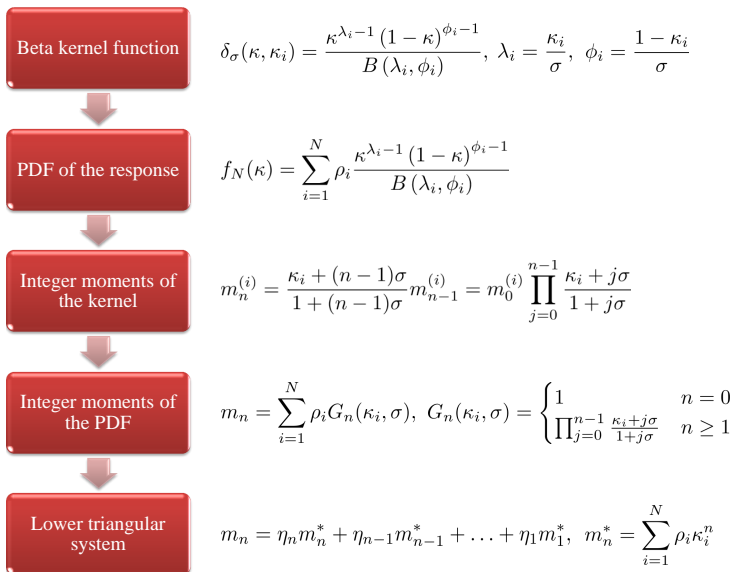
- The foundation of extended quadrature method of moments (EQMOM):

$$f_N(\kappa) = \sum_{i=1}^N \rho_i \delta_\sigma(\kappa, \kappa_i)$$

where

- N is the number of non-negative kernel functions
- ρ_i is the i -th quadrature weight used in the PDF reconstruction
- $\delta_\sigma(\kappa, \kappa_i)$ is the kernel density function
- The choice of the kernel density function $\delta_\sigma(\kappa, \kappa_i)$
 - **Beta** distribution: κ on bounded interval $[a, b]$
 - **Gamma** distribution: semi-infinite κ on $[a, +\infty)$
 - **Gaussian** distribution: κ on the whole real set

Beta EQMOM



Beta EQMOM

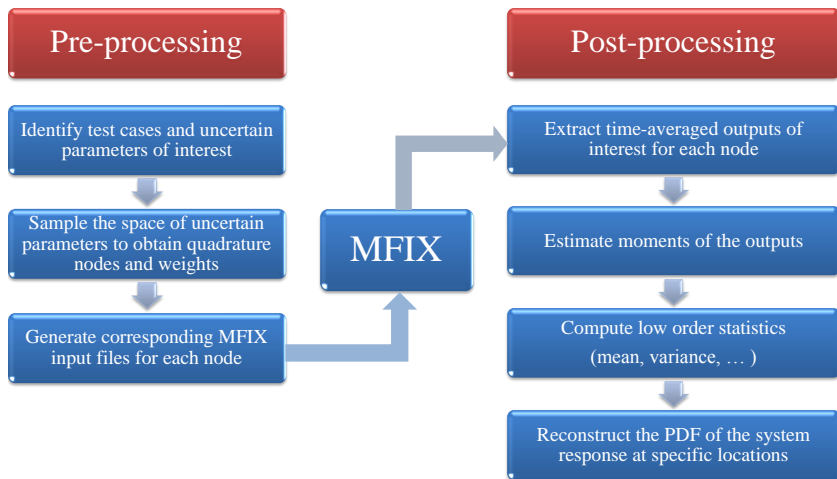
- The algorithm to solve for σ is:
 - ① Guess σ
 - ② Compute the moments m_n^* from the system $\mathbf{A}(\sigma)\mathbf{m}^* = \mathbf{m}$
 - ③ Use the Wheeler algorithm to find weights and abscissas from \mathbf{m}^*
 - ④ Compute m_{2N}^* using weights and abscissas
 - ⑤ Compute

$$J_N(\sigma) = m_{2N} - \eta_{2N}m_{2N}^* - \eta_{2N-1}m_{2N-1}^* - \dots - \eta_1m_1^*$$

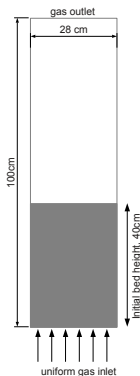
- ⑥ If $J_N(\sigma) \neq 0$, compute a new guess for σ and iterate from step 1 until convergence
- The normalized distribution for κ on bounded interval $[a, b]$ is

$$f_N(\kappa) = \frac{1}{b-a} \sum_{i=1}^N \rho_i \frac{\left(\frac{\kappa-a}{b-a}\right)^{\lambda_i-1} \left(\frac{b-\kappa}{b-a}\right)^{\phi_i-1}}{B(\lambda_i, \phi_i)}$$

Implementation into MFI



Bubbling fluidized bed



- Two fluid model (MFIx)
- Kinetic theory closures

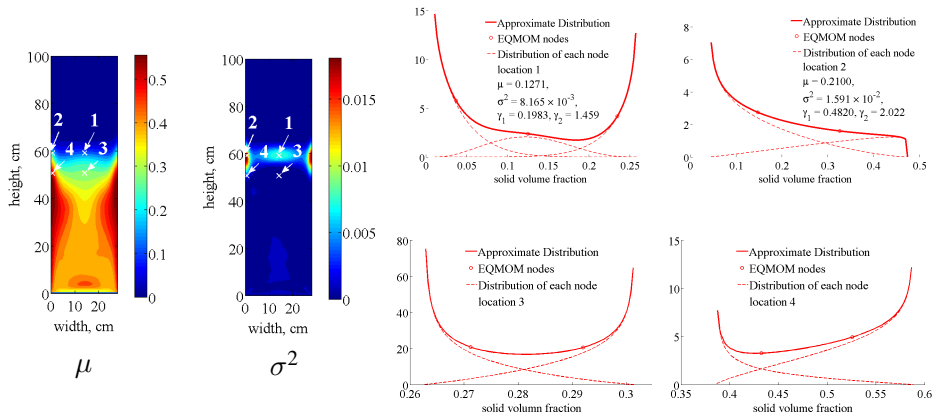
Properties

- Gas density $\rho_g = 1.225 \text{ kg/m}^3$
- Solid density $\rho_s = 2500 \text{ kg/m}^3$
- Uncertain particle size
 - Mean diameter $d_s = 275 \text{ }\mu\text{m}$
 - Uniform distribution
 - Standard deviation $\sigma = 0.3 d_s$

Initial and boundary conditions

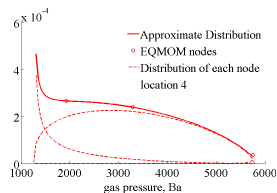
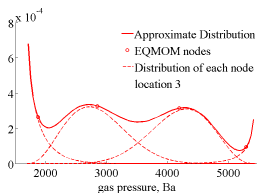
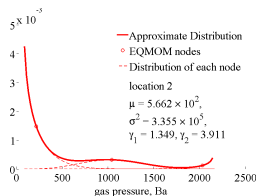
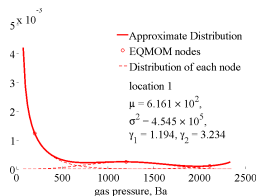
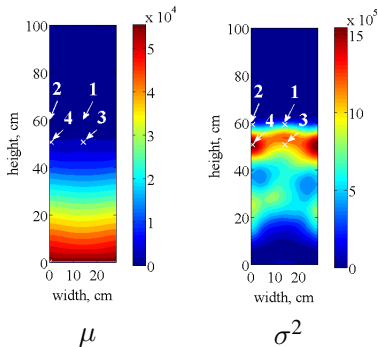
- Bed height: 0.40 m, void fraction: 0.4
- Inlet gas velocity: 0.38 m/s
- Inlet: constant gas inflow
- Outlet: zero gas pressure
- Wall: no-slip boundary condition

Solid volume fraction



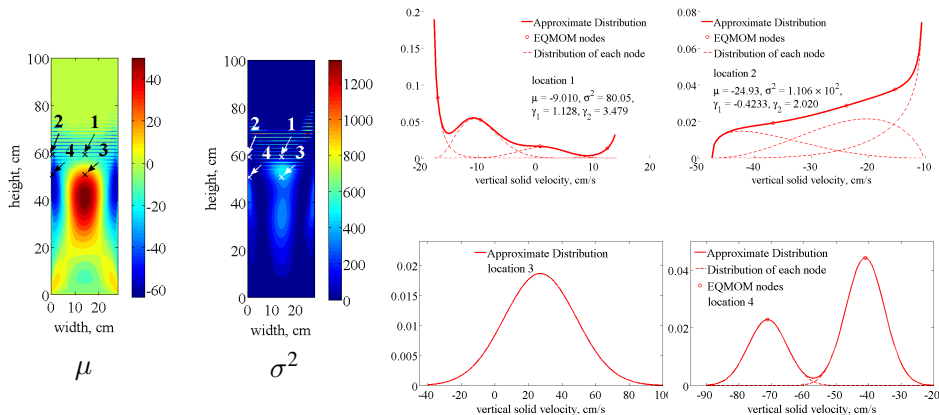
- The effect of uncertain particle diameter focuses on the interface of the bed, in particular near the wall.

Gas pressure



- The effect of uncertain particle diameter focuses on the interface of the bed, especially near the wall.

Vertical solid velocity



- The effect of uncertain particle diameter focuses on the interface of the bed and also near the wall.

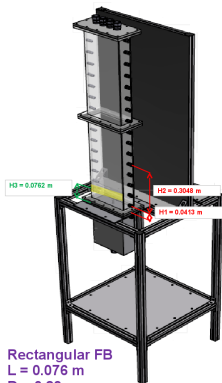
Descriptions of Small Scale Challenge Problem SSCP-I

Properties

- Particle diameter $3256 \mu\text{m}$
- Particle density 1131 kg/m^3
- Sphericity 0.94
- Gas density 1.204 kg/m^3
- Gas viscosity $1.79 \times 10^{-5} \text{ Pa} \cdot \text{s}$
- Minimum fluidization velocity 1.05 m/s

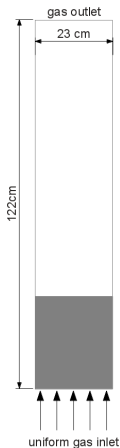
Uncertain parameters

- Particle-particle restitution coefficient 0.84
- Particle-wall restitution coefficient 0.92
- Independent variables



Rectangular FB
 $L = 0.076 \text{ m}$
 $D = 0.23 \text{ m}$
 $H = 1.22 \text{ m}$

Descriptions of Small Scale Challenge Problem SSCP-I



Initial conditions

- Bed height 0.163 m
- Void fraction 0.40
- Superficial gas velocity 2.19 m/s

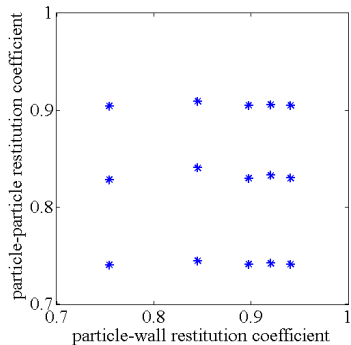
Boundary conditions

- Inlet: constant inlet velocity
- Outlet: zero gas pressure
- Wall: no-slip (gas)
Johnson-Jackson (particles)

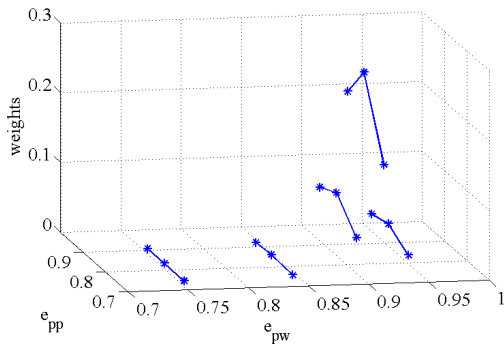
- Two fluid model (MFIx)
- Kinetic theory closures

- Simulation time 90 s
- Convergence criterion 10^{-4}

Sampling method: CQMOM

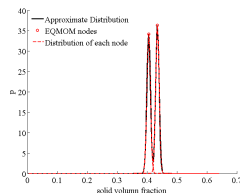
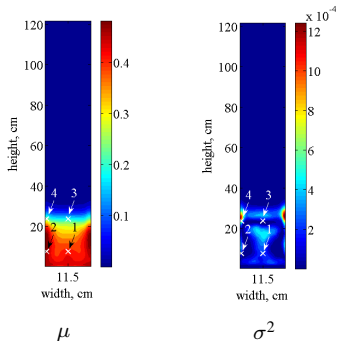


Samples generated by CQMOM

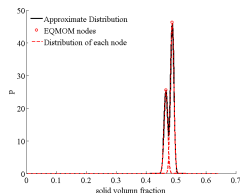


Weights for each sample

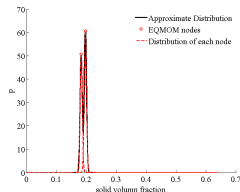
Solid volume fraction



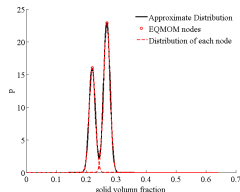
Location 1



Location 2



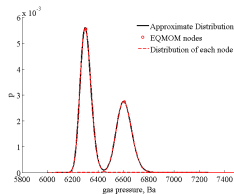
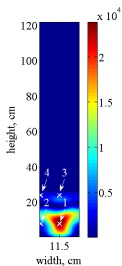
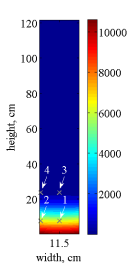
Location 3



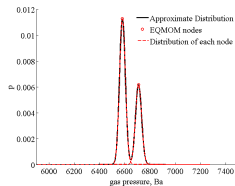
Location 4

- The effect of uncertain parameters focuses on the interface of the bed, in particular near the wall, and also on the center of the bed.

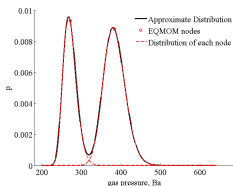
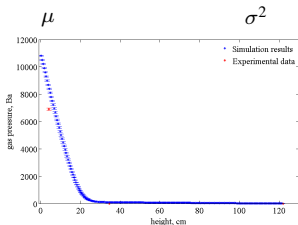
Gas pressure



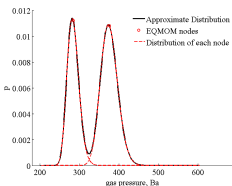
Location 1



Location 2



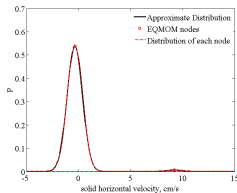
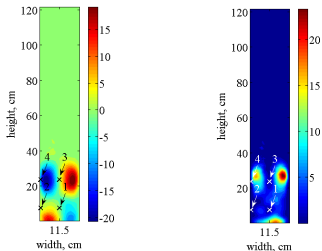
Location 3



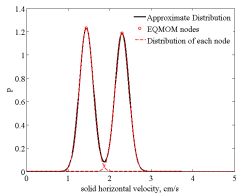
Location 4

- The effect of uncertain parameters focuses on the center of the bed.

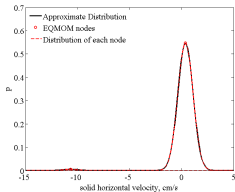
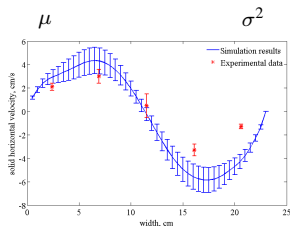
Solid horizontal velocity



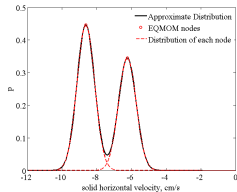
Location 1



Location 2



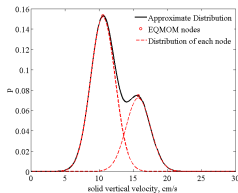
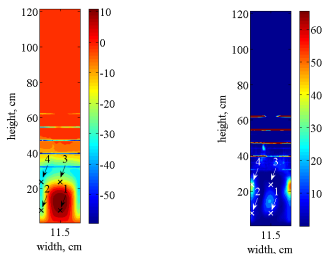
Location 3



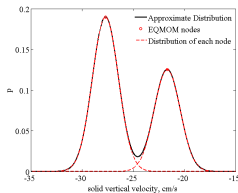
Location 4

- The effect of uncertain parameters focuses on the interface and bottom of the bed.

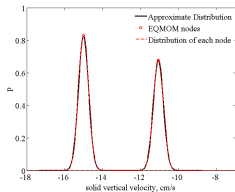
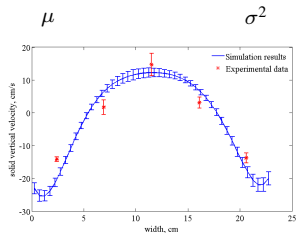
Solid vertical velocity



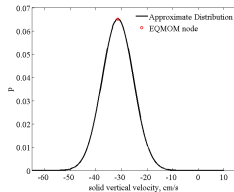
Location 1



Location 2



Location 3



Location 4

- The effect of uncertain parameters focuses on locations near the wall.

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Future work

- Apply the developed procedure to gas-particle flow simulations in risers
- Prepare for the final report

Budget

DOE UCR – FE0006946								
Cost plan status								
Baseline Reporting Quarter	Budget Period 2				Budget Period 3			
	Q3		Q4		Q1		Q2	
	Apr. 1, 2013 - Jun. 30, 2013		Jul. 1, 2013 - Sept. 30, 2013		Oct 1, 2013 - Dec. 31, 2013		Jan. 1, 2014 - Mar. 31, 2014	
	Q3	Cumulative total	Q4	Cumulative total	Q1	Cumulative total	Q2	Cumulative total
Baseline cost plan								
Federal share	26603.00	174123.00	22905.00	197028.00	28347.00	225375.00	23592.00	248967.00
Non-federal share	1850.00	12950.00	1850.00	14800.00	1850.00	16650.00	1850.00	18500.00
Total planned	28453.00	187073.00	24755.00	211828.00	30197.00	242025.00	25442.00	267467.00
Actual incurred cost								
Federal share	31646.29	134567.13	33777.04	168344.17	33456.69	201800.86	29371.87	231172.73
Non-federal share	1850.00	12950.00	1850.00	14800.00	1850.00	12950.00	1850.00	14800.00
Total incurred costs	33496.29	147517.13	35627.04	183144.17	35306.69	218450.56	31221.87	249672.73
Variance								
Federal share	5043.29	-39555.87	10872.04	-28683.83	5109.69	-23574.14	5779.87	-17794.27
Non-federal share	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Total variance	5043.29	-39555.87	10872.04	-28683.83	5109.69	-23574.14	5779.87	-17794.27

Personnel and publications

Personnel

- 1 Assistant professor (Alberto Passalacqua) from October 2011
- 1 Ph.D. student (Xiaofei Hu) from June 2012

Publications

- X. Hu, A. Passalacqua, R. O. Fox, P. Vedula, A quadrature-based uncertainty quantification approach with reconstruction of the probability distribution function of the system response, in preparation.
- X. Hu, A. Passalacqua, R. O. Fox, A quadrature-based uncertainty quantification approach with reconstruction of the probability distribution function of the system response in bubbling fluidized beds, 2013 AIChE Annual Meeting, San Francisco, CA, November 3rd – 8th, 2013.
- X. Hu, A. Passalacqua, R. O. Fox, Validation of a quadrature-based uncertainty quantification approach using NETL small scale challenge problem SSCP-I, ASME 2014 Verification and Validation Symposium, Las Vegas, NV, May 7th – 9th, 2014.
- X. Hu, A. Passalacqua, R. O. Fox, A quadrature-based uncertainty quantification approach in a multiphase gas-particle flow simulation in a riser, 2014 AIChE Annual Meeting, Atlanta, GA, November 16th – 21st, 2014, submitted.

Thanks for your attention!

Questions?