

Computational & Multiscale Mechanics of Materials University of Liège, Belgium



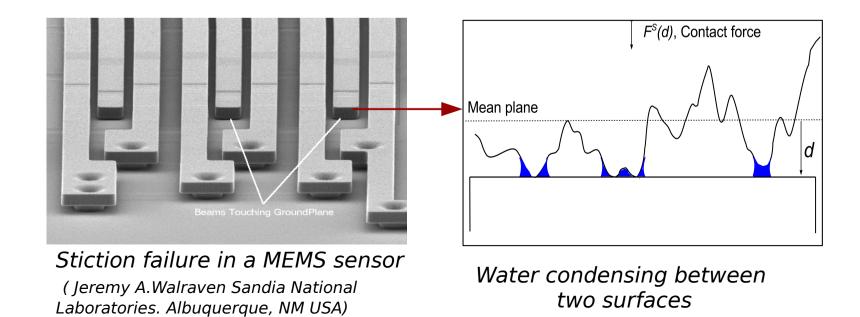
A stochastic multi-scale analysis for MEMS stiction failure

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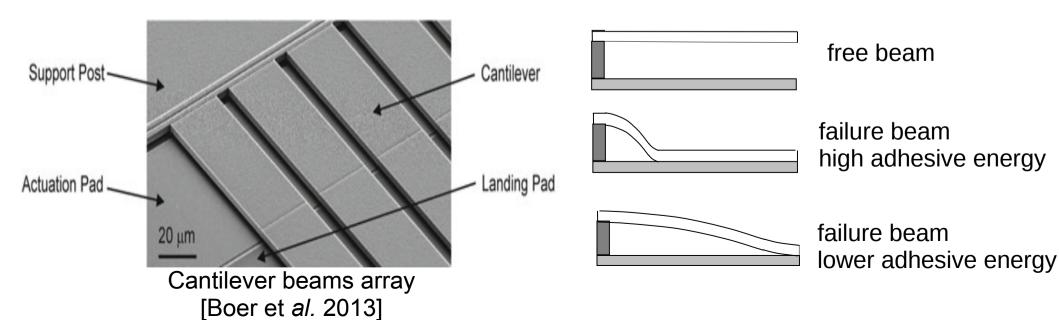
Maarten Arnst, Ludovic Noels

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- (*) PhD candidate at the Belgian National Fund for Education at the Research in Industry and Farming (FNRS -FRIA).

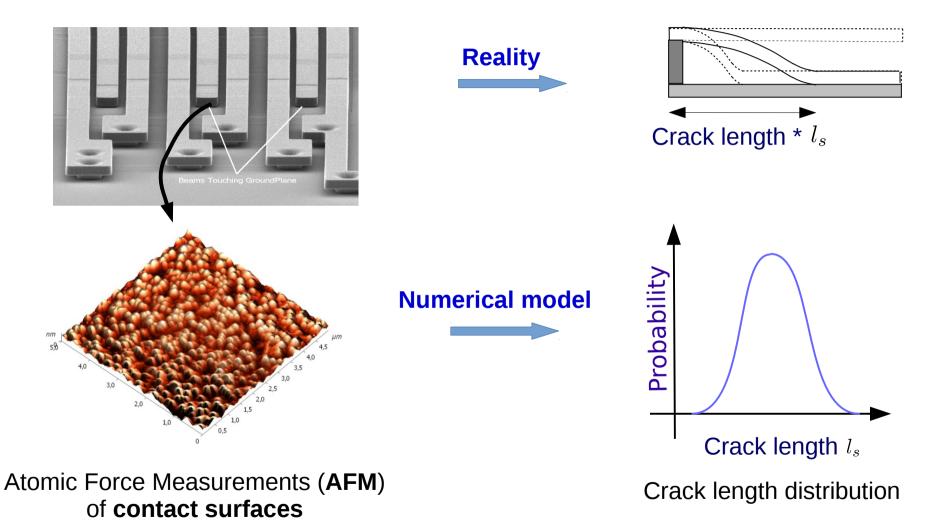
- MEMS devices, e.g. accelerometers, digital mirrors, pressure sensors, resonators
 - Successfully applied in the industry. E.g. accelerometers for air bag systems
- Stiction: a common failure of MEMS
 - Due to dominance of surface adhesive forces at the micrometer scale
 - E. g., van der Waals forces and capillary forces
 - In humid conditions, the capillary forces are dominant
 - Depends on the surface topologies
 - The physical contact happens at the high asperities



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 - Depends on the surface topologies
 - The physical contact happens at the high asperities
 - An uncertain phenomenon: [Boer et al. 2007, 2013], [W. M. van Spengen 2003, 2015]
 - e.g. Stiction test: Some beams are stuck with different configurations, other beams are released



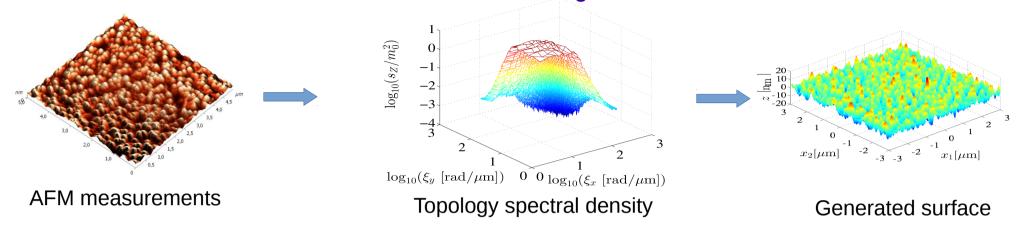
- Construct a numerical model for micro cantilever beam structures
 - To predict the crack length l_s and its uncertainties from the surface topology
 - At an acceptable computational cost



^{*} The crack length l_s characterizes the required energy to release the cantilever beam out of the failure configuration

- Multiscale model
- Probabilistic multiscale model
 - Direct Monte-Carlo multiscale method (high computational cost)
 - Stochastic model-based multiscale method (acceptable computational cost)
- Numerical results: comparison, evaluation
- Conclusions and perspectives

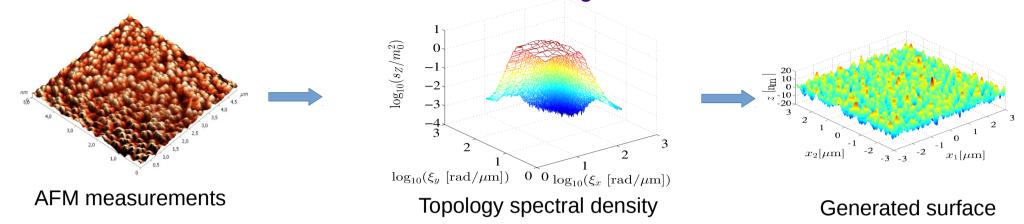
• Lower scale: surface characterization and surface generator



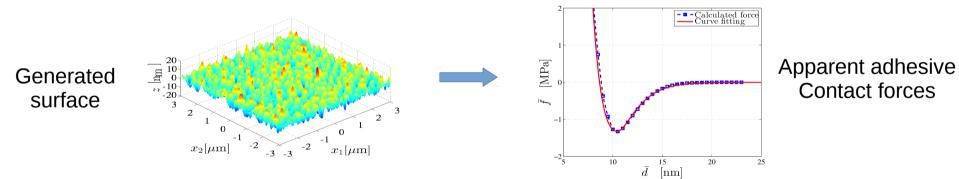
Meso-scale: evaluate the contact forces by a semi-analytical contact model

• Upper-sale: Integrate the contact forces as contact law into FE model

Lower scale: surface characterization and surface generator



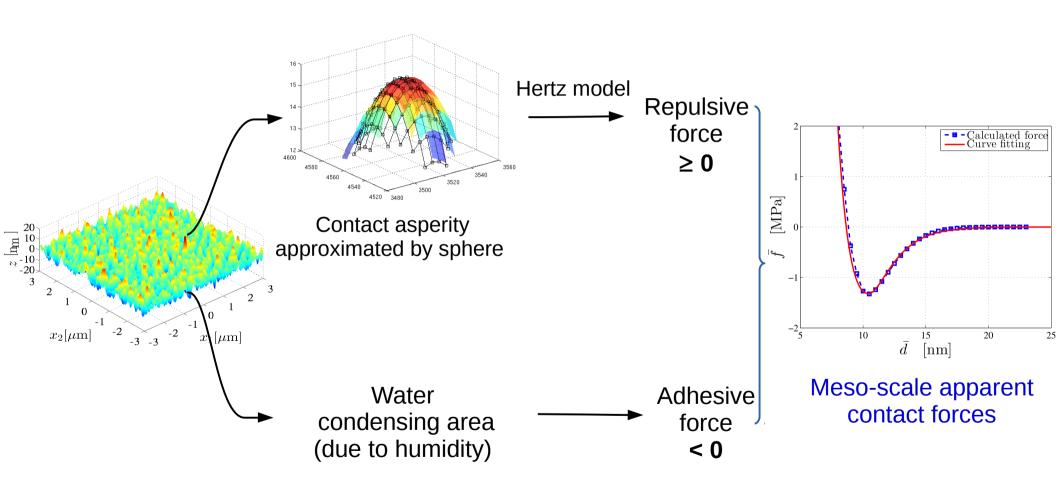
Meso-scale: evaluate the contact forces by a semi-analytical contact model



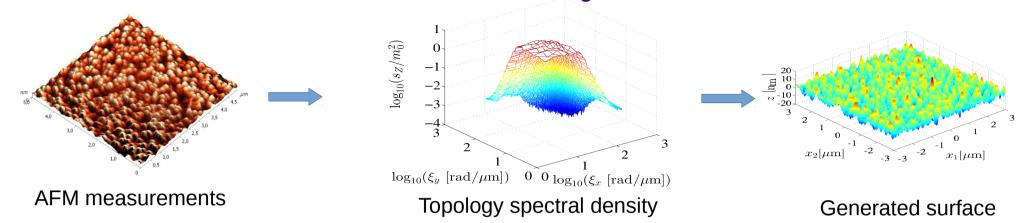
• Upper-sale: Integrate the contact forces as contact law into FE model

• Semi-analytical contact model for adhesive contact problem

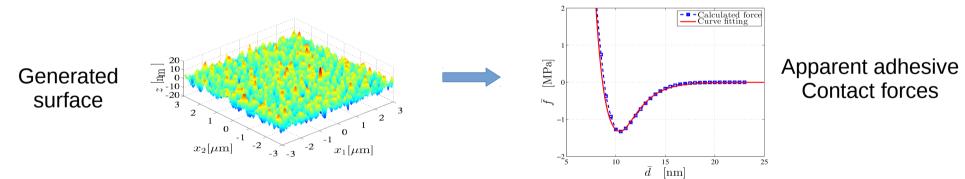
- Based on DMT spherical asperity contact model
- Rough surface contact model



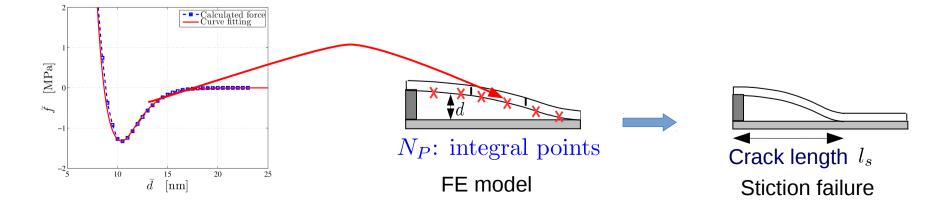
Lower scale: surface characterization and surface generator



Meso-scale: evaluate the contact forces by a semi-analytical contact model

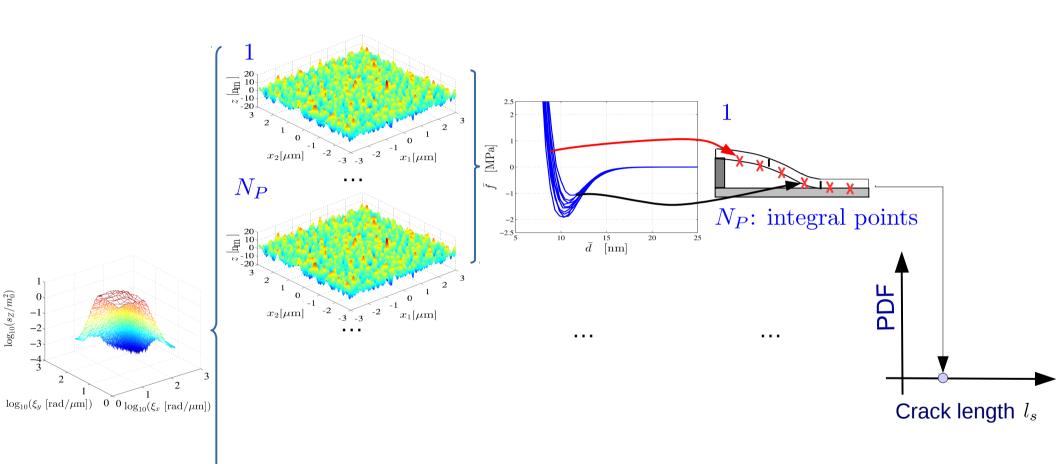


Upper-sale: Integrate the contact forces as contact laws into a FE model

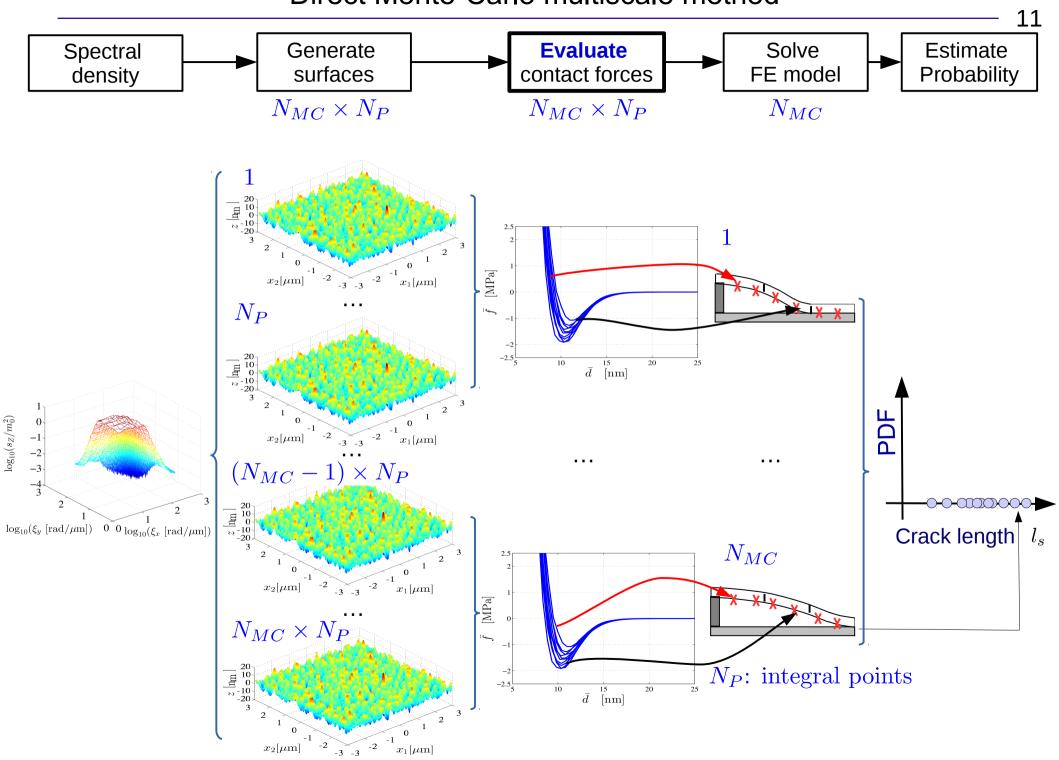


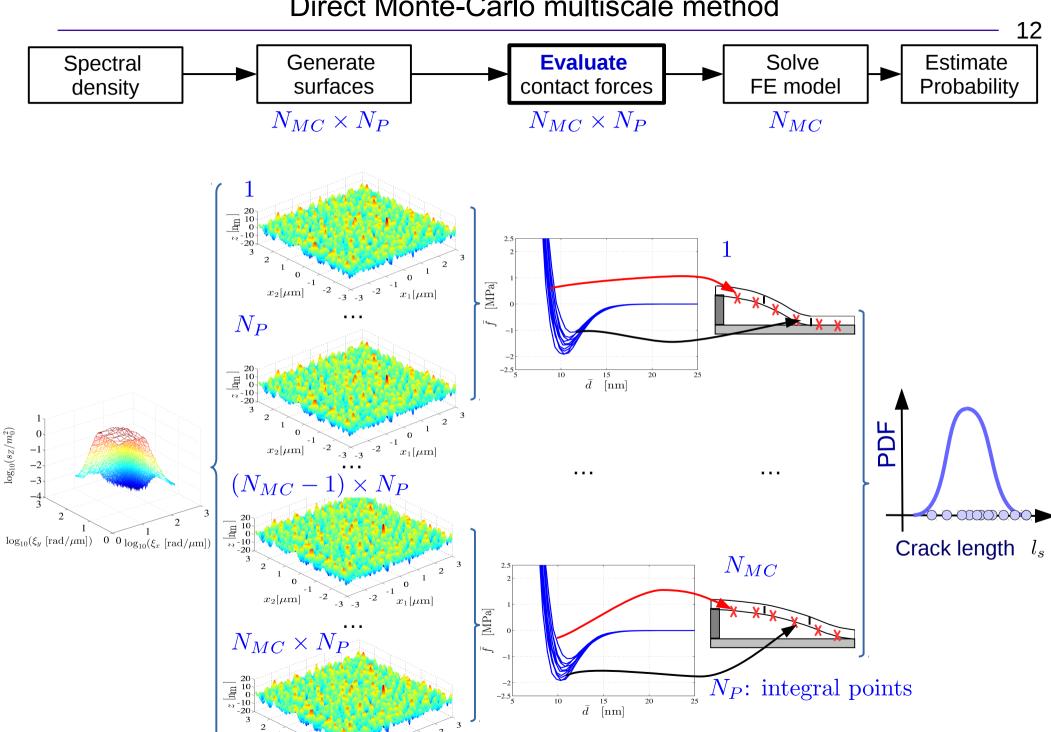
Direct Monte-Carlo multiscale method





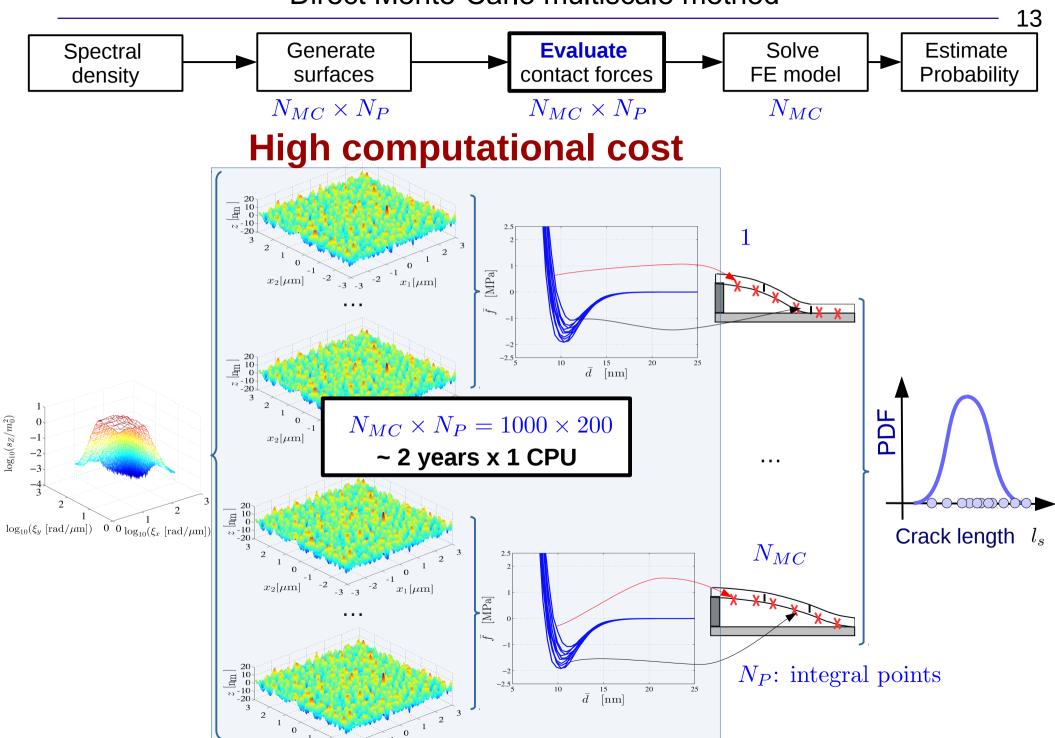
Direct Monte-Carlo multiscale method



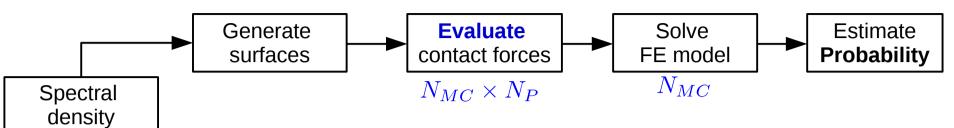


-1 -2 -3 -3 -2 -1

 $x_1[\mu\mathrm{m}]$

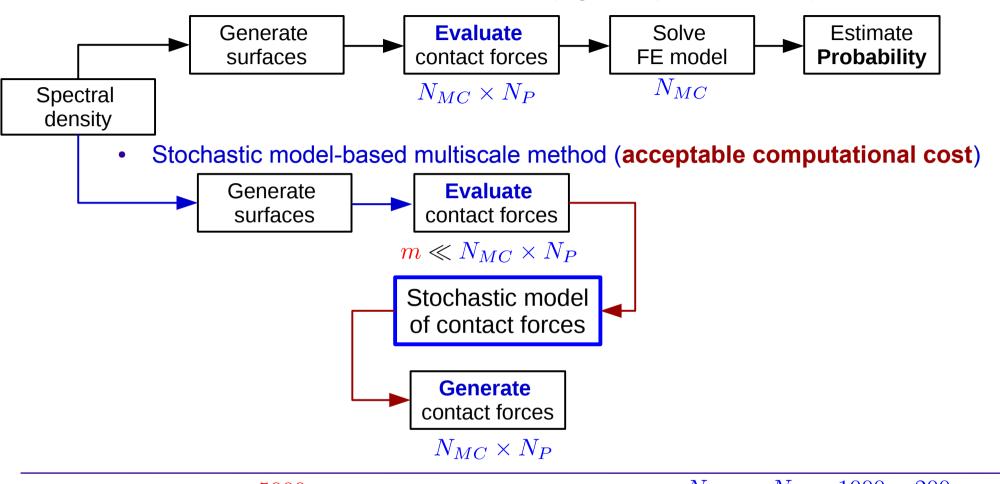


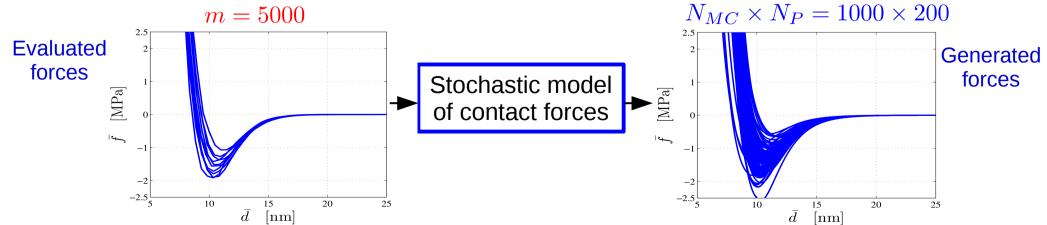
Direct Monte-Carlo multiscale method (high computational cost)



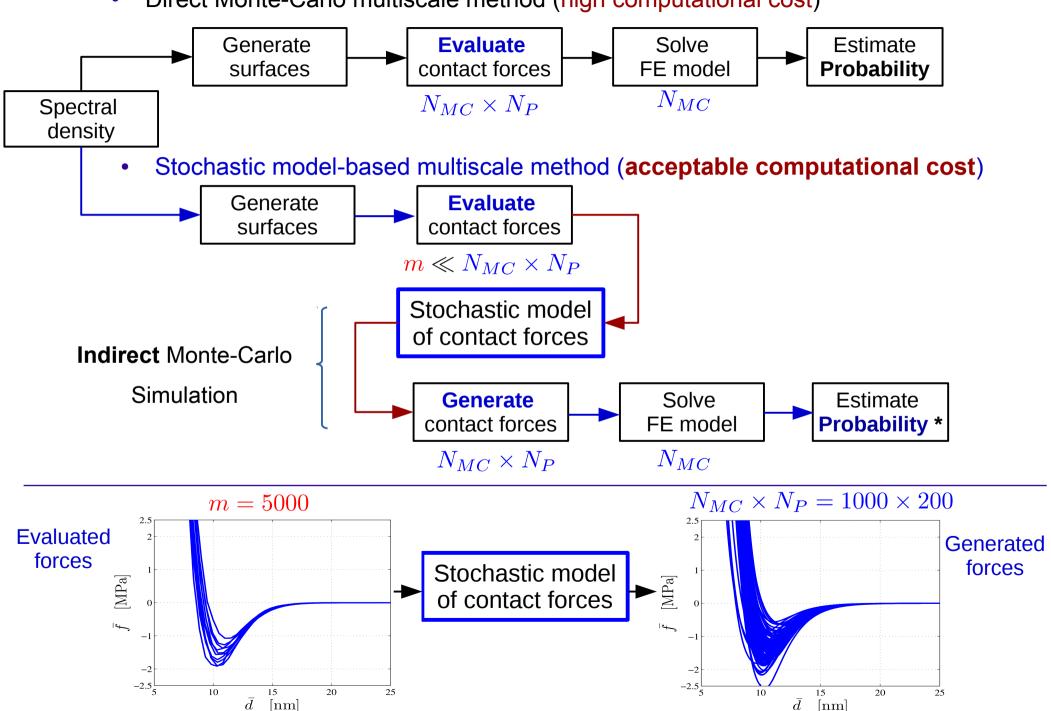
• Stochastic model-based multiscale method (acceptable computational cost)

Direct Monte-Carlo multiscale method (high computational cost)

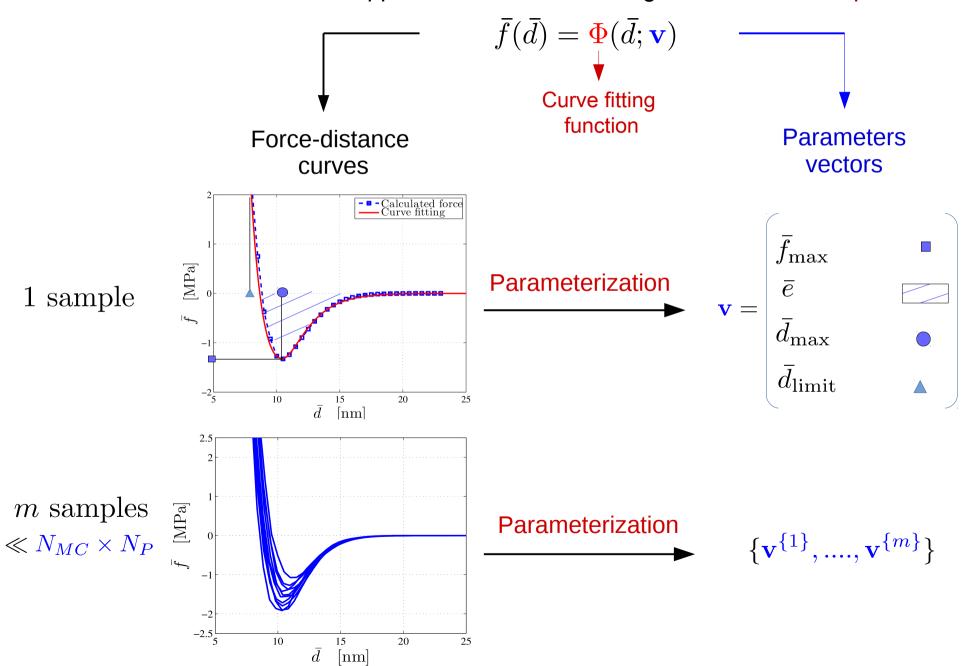




Direct Monte-Carlo multiscale method (high computational cost)



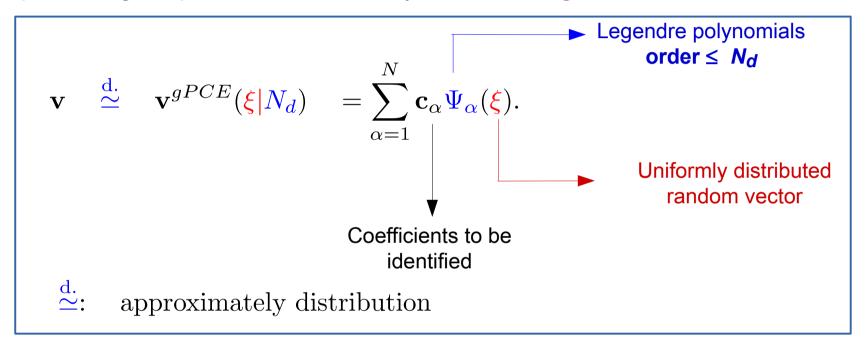
- Stochastic model for apparent contact forces: generalized polynomial chaos expansion (gPCE)
 - Parameterization of the apparent contact force using a modified Morse potential



- Stochastic model for apparent contact forces: generalized polynomial chaos expansion (gPCE)
 - Parameterization

$$\bar{f}(\bar{d}) = \Phi(\bar{d}; \mathbf{v})$$
 Input data: $\{\mathbf{v}^{\{1\}},, \mathbf{v}^{\{m\}}\}$

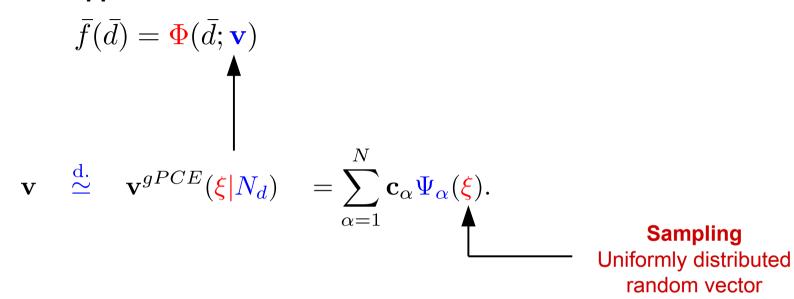
Representing the parameters vectors by a truncated gPCE model: order N_d

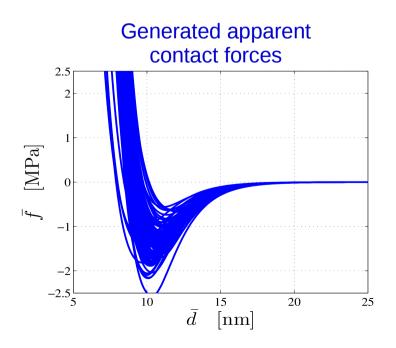


- The higher the order N_d , the better the approximation
- The coefficients are identified by the projection of the iso-probabilistic transformation on the polynomial chaos system

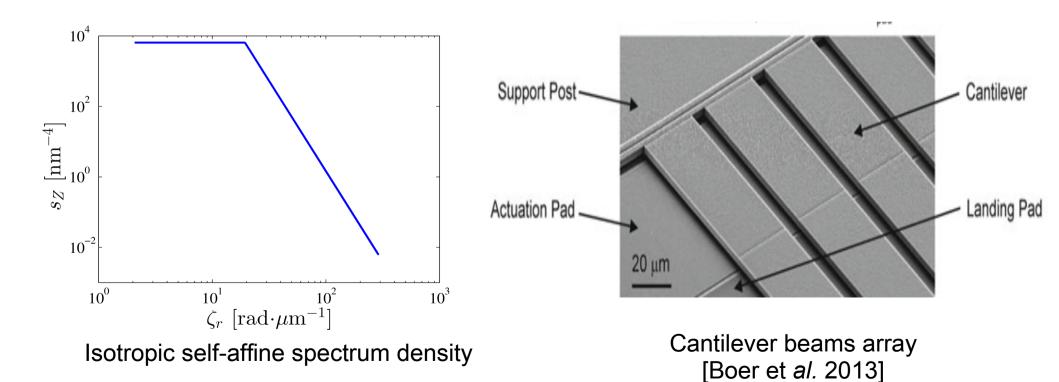
$$\mathbf{c}_{\alpha}: \quad \text{Distribution}(\{\mathbf{v}^{\{1\}},....,\mathbf{v}^{\{m\}}\}) \simeq \text{Distribution}(\mathbf{v}^{gPCE})$$

- Stochastic model for apparent contact forces: generalized polynomial chaos expansion (gPCE)
 - Generate apparent contact forces

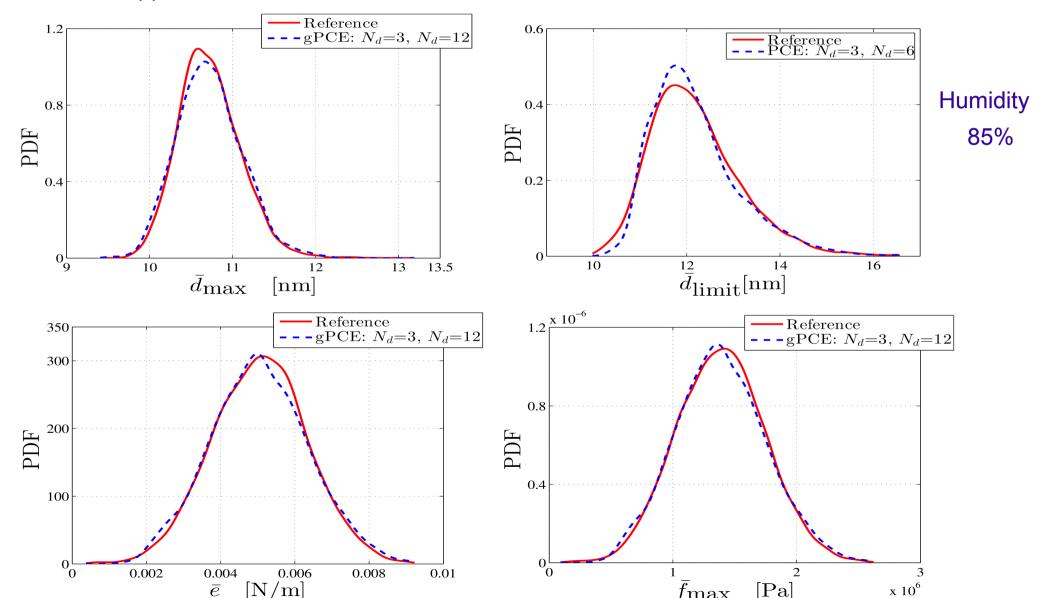




- Using the two methods: direct Monte-Carlo multiscale vs Stochastic-model based multiscale method to evaluate
 - Distribution of apparent contact forces parameters
 - Distribution of crack lengths
- Input: a constructed isotropic self-affine spectrum density function of the contacting surfaces based on experiments report in [Boer et al. 2013]



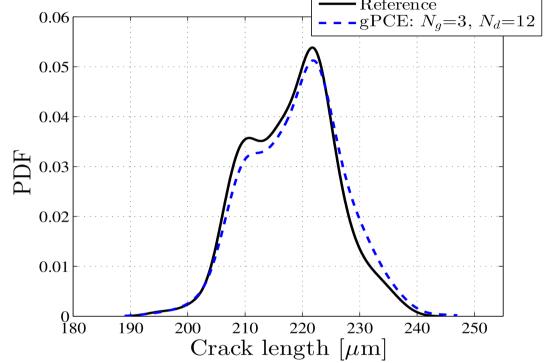
- Direct Monte-Carlo multiscale vs Stochastic-model based multiscale method
- Distribution of contact forces parameters
 - The stochastic model-based method **approximates well the probabilistic distribution** of the apparent contact forces



- Direct Monte-Carlo multiscale vs Stochastic-model based multiscale method
- Distribution of crack lengths

The stochastic model-based method approximates well the probabilistic distribution of

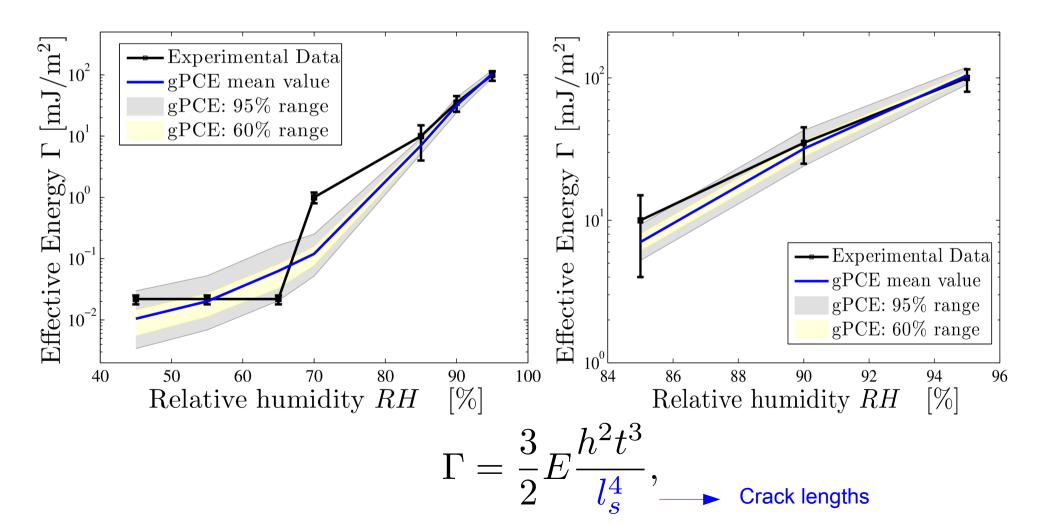
the crack lengths



Humidity 85%

- Computational efficiency: Time to evaluate 1 realization in one CPU:
 - Direct Monte-Carlo method: ~16 hours (due to the evaluation of the apparent contact forces)
 - Stochastic model-based method: ~ 5 minutes.
 - To obtain the distribution: 1000 realizations are required
- Good approximation and acceptable computation cost

- Validation: comparison with experimental results [Boer et al. 2013]
 - Accurate prediction of the experimental results at high humidity levels
 - At low humidity levels:
 - The van der Waals forces need to be accounted for
 - The rough surface needs to be directly measured



- A Stochastic model-based multiscale method for stiction problems taking the surface topology into account by
 - Using spectral density to characterize the AFM surface measurements and generate numerical surfaces
 - Using multi-scale approach, and a semi-analytical contact model
 - To define and evaluate the meso-scale apparent contact forces
 - To integrate them as contact laws to FE model
 - Applying gPCE to build a stochastic model of the random meso-scale contact forces
 - To reduce efficiently the computational cost
- The model is validated by a comparison with experimental results
- Applying the model to evaluate the failure percentage of MEMS designs such as
 - The stiction of MEMS accelerometers under shock
 - The stiction of MEMS gears system



Documents can be downloaded at: www.ltas-cm3.ulg.ac.be

http://www.sandia.gov/mstc/mems/

- T.-V. Hoang et al., A computational stochastic multi-scale methodology for mems structures involving adhesive contact. (submitted to Tribology International)
- T.-V. Hoang et al., A probabilistic model for predicting the uncertainties of the humid stiction phenomenon on hard materials
- L Wu et al., A micro-macro approach to predict stiction due to surface contact in microelectromechanical systems
- W. Merlijn van Spengen, MEMS reliability from a failure mechanisms perspective
- M. de Boer, Capillary adhesion between elastically hard rough surfaces
- · F. W. DelRio et al. Van der waals and capillary adhesion of polycrystalline silicon micromachined surface
- F. Poirion, C. Soize, Numerical Methods and Mathematical Aspects For Simulation of Homogeneous and Inhomogeneous Gaussian Vector Fields
- A. Clément, C. Soize, J. Yvonnet, *Uncertainty quantification in computational stochastic multiscale analysis of nonlinear elastic materials*
- C. Soize, R. Ghanem, Physical Systems with Random Uncertainties: Chaos Representations with Arbitrary Probability Measure
- · C. Desceliers, R. Ghanem, C. Soize, *Maximum likelihood estimation of stochastic chaos representations from experimental data*
- M. Arnst, R. Ghanem, C. Soize, *Identification of Bayesian posteriors for coefficients of chaos expansions*
- S. Das et al., Polynomial chaos representation of spatio-temporal random fields from experimental measurements
- D. W. Scott, Multivariate Density Estimation: Theory, Practice, and Visualization

Thank you for your attention