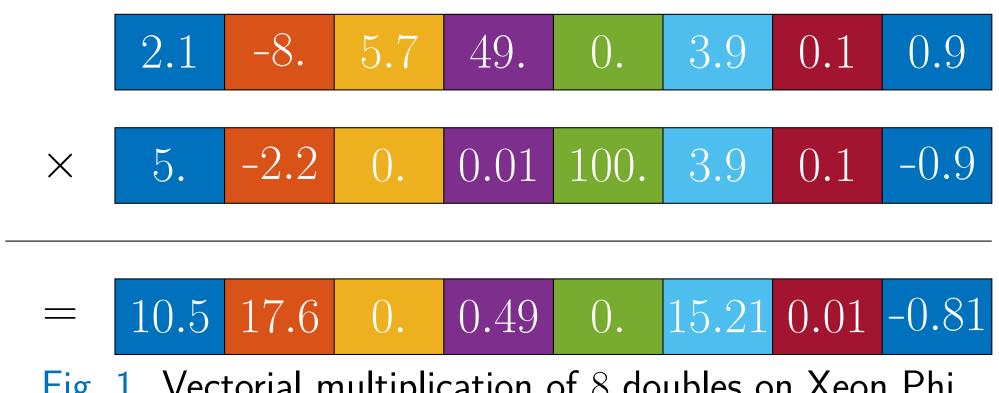
### Ensemble propagation for efficient uncertainty quantification on emerging architectures: Application to thermomechanical contact K. Liegeois<sup>1</sup>, R. Boman<sup>1</sup>, Ph. Mertens<sup>2</sup>, A. Panin<sup>2</sup>, E. T. Phipps<sup>3</sup>, M. Arnst<sup>1</sup> <sup>1</sup>Aerospace and Mechanical Engineering, Université de Liège, Liège, Belgium **THIS LIEGE** université <sup>2</sup>Institute of Energy and Climate Research - Plasma Physics, Forschungszentrum Jülich GmbH, Jülich, Germany <sup>3</sup>Center for Computing Research, Sandia National Laboratories, Albuquerque, NM **Ensemble propagation** Objective Efficient sampling-based uncertainty quantifica-Instead of individually evaluating each instance of tion with multiphysics models on emerging archithe model, ensemble propagation (EP) [1] consists in simultaneously evaluating a subset of samples of tectures. the model. **Emerging architectures** Model Model **Vectorization** (Single Instruction Multiple Data) Fig. 3. Sample and ensemble propagation with 8 samples. and **efficient memory usage** for optimal perfor-EP has the following advantages: mance Improved probability of auto-vectorization, 0.1 0.9 • Improved memory usage,





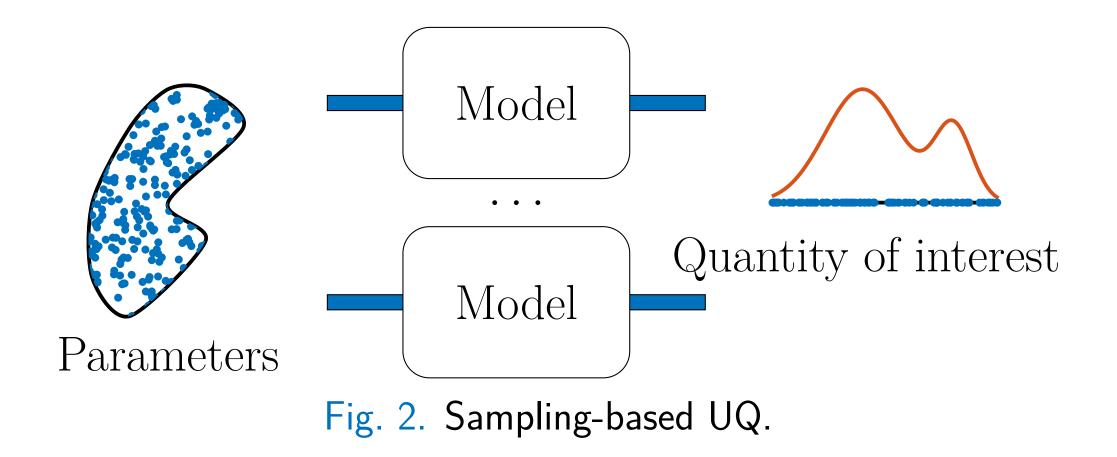
Vectorial multiplication of 8 doubles on Xeon Phi. Fig.

On Xeon Phi KNL, the maximum speed up is 244 and 1952 without and with double vectorization respectively.

# **Sampling-based** uncertainty quantification (UQ)

The Monte Carlo method is well-adapted to problems of uncertainty quantification:

- with nonsmooth responses,
- in very high dimension.

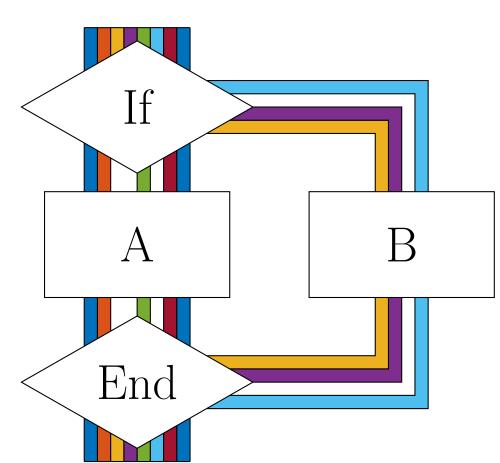


- Reuse of common variables,
- Reduction of Message Passing Interface (MPI) latency per sample.

Advantage: EP can improve the performance of sampling-based UQ on emerging architectures.

EP raises new challenges:

- Increased memory usage,
- Code divergence.



Code divergence of a sample dependent If. Fig. 4

Challenge: Code divergence inside an ensemble may reduce the computational performance.

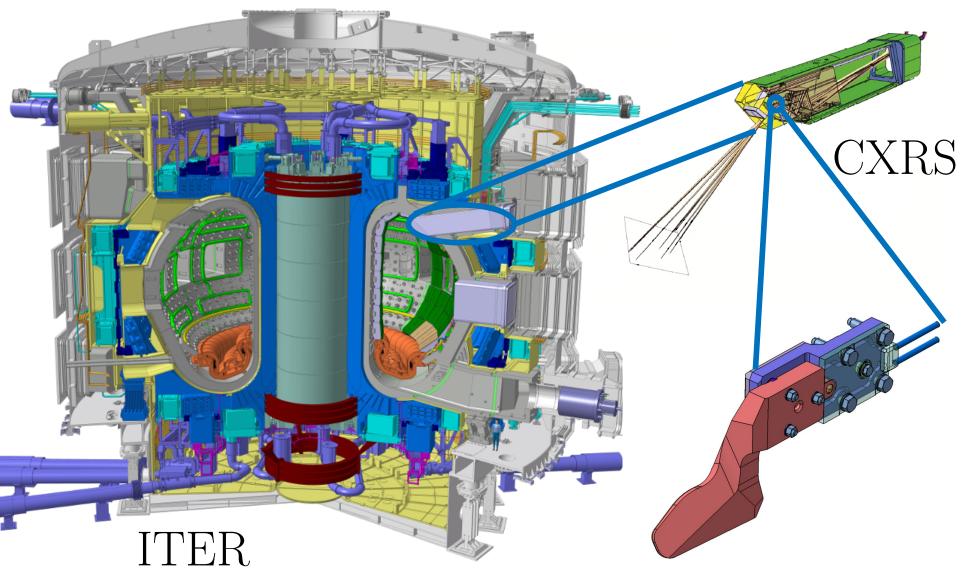
## Trilinos

Such an embedded EP is currently being implemented based on the use of C++ templating in Stokhos, a software component of Trilinos [2].



## ITER spectroscope

One of the possible options of the first mirror of the CXRS ITER spectroscope [3] will be bolted to a holder which will cool the mirror heated by the plasma. The deformation of the mirror surface depends on the temperature, which itself depends on the contact status of the bolts.





### Contact

Contact constraints can be either **active** or inactive.

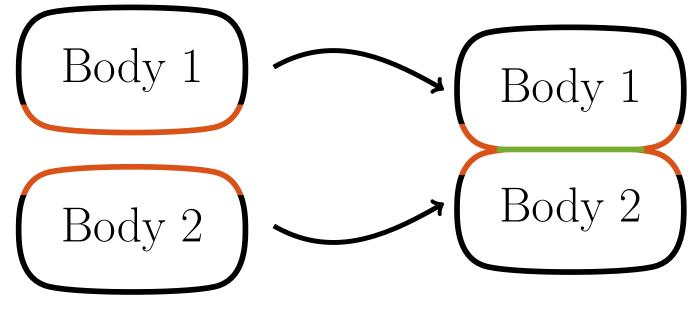
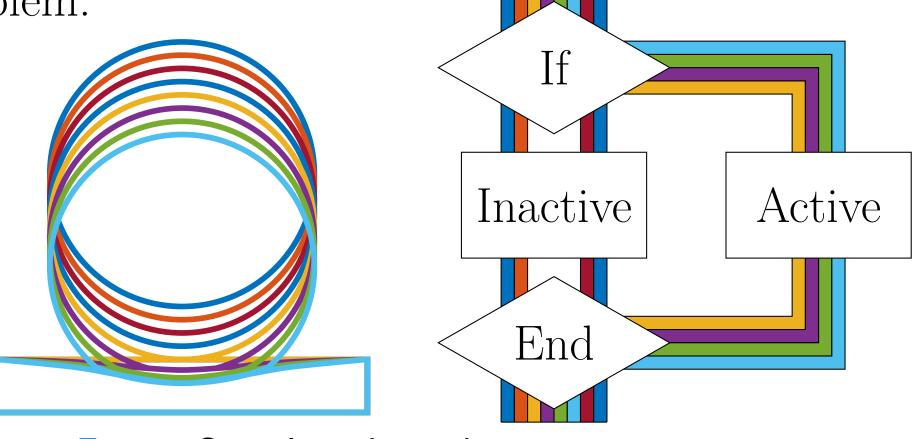


Fig. 6. Contact local status.

The active and inactive constraints are not known a priori and may modify the graph of the discretized problem.



Sampling dependent contact status. Fig.





same ensemble? • Grouping samples according to their predicted contact status,

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[3] Yury Krasikov, Anatoly Panin, Wolfgang Biel, Andreas Krimmer, Andrey Litnovsky, Philippe Mertens, Olaf Neubauer, and Michael Schrader. Major aspects of the design of a first mirror for the ITER core CXRS diagnostics. Fusion engineering and design, 96:812–816, 2015.

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# **Ongoing study**

• How to solve the model efficiently for any given ensemble of sufficiently similar samples?

• Select the most appropriate contact formulation for EP, • Select the most appropriate Krylov solver adequate for contact with EP,

• Suppress the code divergences using sample independent criteria and using masking.

Fig. 8. Masking of the If condition of Fig. 4 using mask m.

• How to group sufficiently similar samples in the

rouping samples according to their estimated number linear iterations of Krylov method.

### References

[1] Eric Phipps, Marta D'Elia, H Carter Edwards, Mark Hoemmen, Jonathan Hu, and Sivasankaran Rajamanickam.

Embedded ensemble propagation for improving performance,

portability, and scalability of uncertainty quantification on emerging computational architectures.

SIAM Journal on Scientific Computing, 39(2):C162–C193, 2017.

### Acknowledgements