# Uncertainty quantification and sensitivity analysis for a steam generator clogging simulation code



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- 1. Industrial context
  - The steam generator (SG) → heat exchanger between the primary and secondary circuits of a nuclear power plant (NPP).



- Corrosion in the secondary circuit produces iron oxide impurities  $\rightarrow$  clogging of the SG over time.
- Clogging is measured during NPP outages using image processing techniques.
- $\bullet$  Chemical cleanings can be performed to reduce it  $\to$  aging, environmental and economic impacts.
- Challenges for maintenance planning and optimization.
- 2. Clogging physical modelling



- 1. Thermohydraulic stationary two-state vector quantities are computed.
- 2. Solid and soluble iron-oxide particles are transported to the flow hole.
- 3. Clogging kinetics is characterised by the clogging rate  $\tau_c$ .

## 3. UQ methodology



## 4. DEPOTHYC module

- The numerical implementation of block 2. is the DEPOTHYC module [1].
- This module allows to predict clogging kinetics on **short periods of time**.
- $_{\bullet}$  Unitary call  $\sim 15$  min.

## 5. Input variables

## $\mathbf{X} = (\alpha, \beta, d_p, \epsilon_e, \epsilon_c, \Gamma_p(0), \Gamma_s^{\max}, a_v)$

- $\alpha, \beta$ : empirical correlation parameters.
- $d_p$ : diameter of particles.
- $\epsilon_e, \epsilon_c$ : clogging deposit porosities.
- $\Gamma_p(0), \Gamma_s^{\max}$ : initial solid particles fraction and solubility map.
- $a_v$ : calibration parameter.

### 6. Monte Carlo simulations



Cobweb graph -  $[\tau_c]$  vs  $[\alpha, \beta, d_p, \epsilon_e, \epsilon_c, \Gamma_p(0), \Gamma_s^{max}, a_v]$ 



- 7. Sensitivity analysis (SA)
- We studied Sobol' indices as in [3] working with PCE metamodels.
- HSIC indices [2] allow to uncover deep input-output dependency structures.
- We can perform **global** and **target** with HSIC indices [2].

## 8. Numerical results





## 9. Conclusion and perspectives

- We have recovered part of the SA results obtained in a previous study [3] with Sobol' indices.
- We refined the analysis by computing target HSIC indices and observe changes in the input sensitivity hierarchy for different thresholds.
- Perspectives: applying UQ methodology to a more advanced clogging simulation chain on long periods of time (to develop better physical understanding and a robust decision making tool).

#### References

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- [3] L. Lefebvre and al., Improving the predictivity of a steam generator clogging numerical model by GSA and Bayesian calibration techniques, Conference paper, Physics of Reactors, 2022