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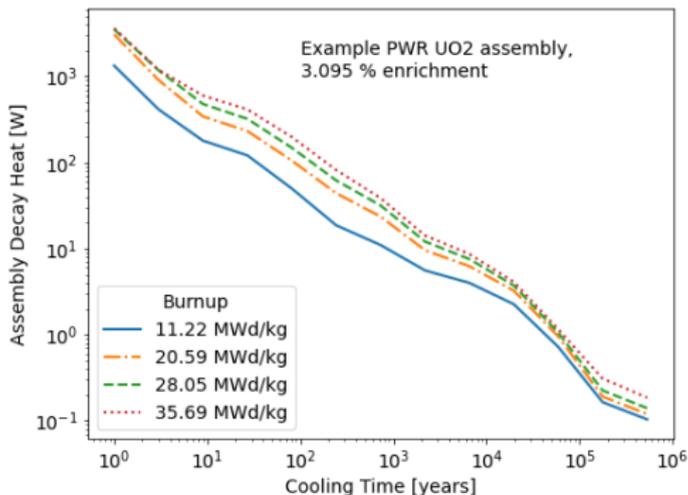
# Surrogate Models for Uncertainties on Decay Heat and Nuclide Inventory in Nuclear Depletion Calculations

CEIDEN Workshop, May 27th 2021

Spent Nuclear Fuel (SNF) remains a safety hazard up to a million years due to:

- Radiation
- Decay heat
- Criticality

It is thus critical to **quantify the risks** to ensure safe handling, storage, and long-term disposal.



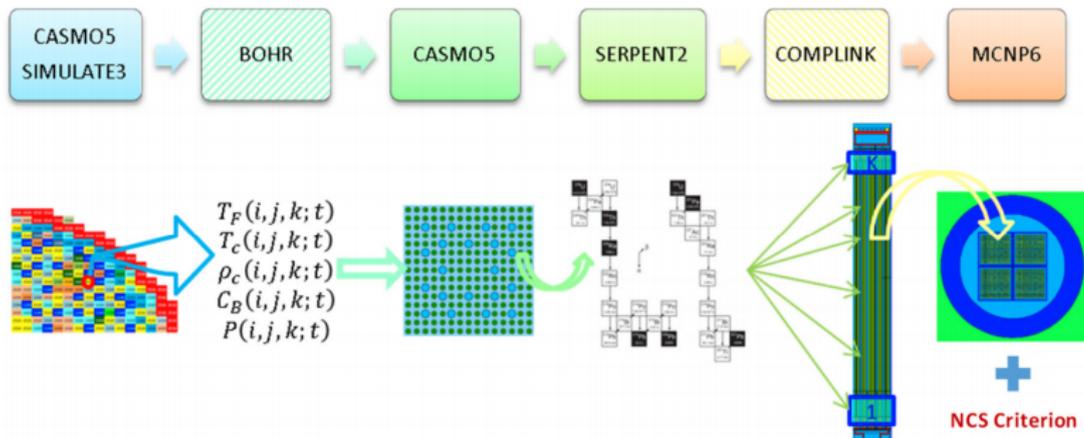
SNF is characterised by:

- Isotopic inventory [ $cm^{-3}$ ]
- Decay heat [ $W/t$ ]
- Criticality  $k_{eff}$

Risk assessment is computationally expensive, and is carried out for:

- Between 12 and 13 thousand FAs expected in Switzerland
- Many time steps into the future
- Quantities of interest need to include **uncertainties**

Procedure developed at PSI [1] to compute quantities of interest:



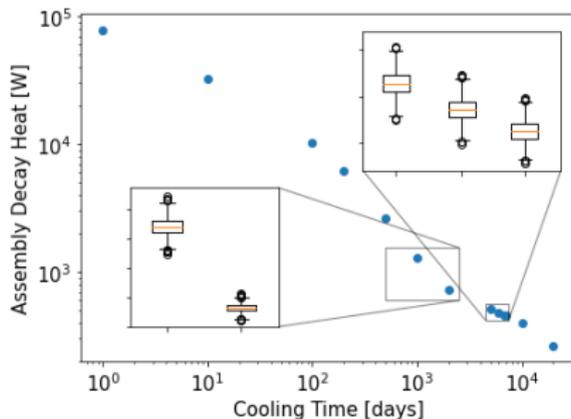
Nominal values of decay heat, isotopic content, and  $k_{eff}$  are not enough.

We require **Best Estimate Plus Uncertainty (BEPU)** to:

- better assess risks
- reduce exposure
- minimise costs

Uncertainty in deterministic codes? Yes!  
Through uncertainty propagation from input uncertainties in:

- FA characteristics (e.g. pin radius, enrichment)
- Operating history (e.g. burn-up, cooling cycles)
- Nuclear data

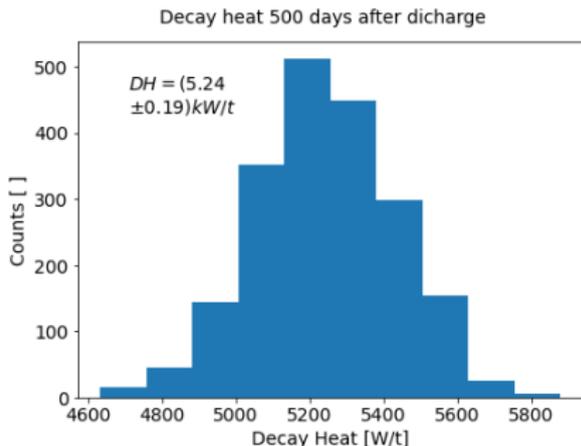


CASMO5 is a reactor lattice code, here used to simulate depletion in the reactor:

$$f : \underbrace{\begin{pmatrix} \text{Fresh fuel parameters} \\ \text{Irradiation history} \\ \text{Reactor parameters} \\ \text{Nuclear Data} \end{pmatrix}}_{\text{Uncertain input parameters}} \rightarrow \begin{pmatrix} \text{DH w.r.t. cooling time} \\ \text{Isotopic inventory at discharge time} \end{pmatrix}$$

Propagation of Nuclear Data uncertainty to CASMO5 output:

- (1) Sample Nuclear Data points from a covariance perturbation matrix [2]
- (2) Perform hundreds of CASMO5 simulations with random sampling of the Nuclear Data
- (3) Reconstruct histogram from output [3]



However,

- Stochastic sampling is computationally expensive
- Unfeasible for the over 12 thousand SNFs [4] expected in Switzerland ( $\sim 900$  years of computational time!)
- Impractical for optimisation tasks

Solution: train fast surrogate models that can replace CASMO5 calculations

# Surrogate Model for Nuclear Data Uncertainties

Training data:

- CASMO5 simulations of one FA with a specific irradiation history with **different nuclear data** sampled from a covariance matrix [2].
- Training set is **deterministic** and without uncertainty.

Output:

- Decay heat  $DH(t_j)$ , at  $t_j$  timesteps during cooling time
- Isotope densities  $\rho_i$

$$f' : \underbrace{\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_{15557} \end{pmatrix}}_{\text{nuclear data}} \rightarrow \begin{pmatrix} \{DH(t_j)\}_{j=1,2,\dots} \\ \{\rho_i\}_{i=1,2,\dots} \end{pmatrix} \quad (1)$$

Trained model is used it for stochastic sampling in a fraction of the time

Challenges:

- Large input might require dimensionality reduction e.g. PCA [5], encoder-decoder [6]
- Only applicable to a specific FA. Transfer learning as an option.



Thank you to PSI and Swiss-Nuclear for co-sponsoring this project.

Thank you for your attention.  
Questions?

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