

Deep Learning Surrogate Models for Neutronic Irradiation Optimisation in IFMIF-DONES

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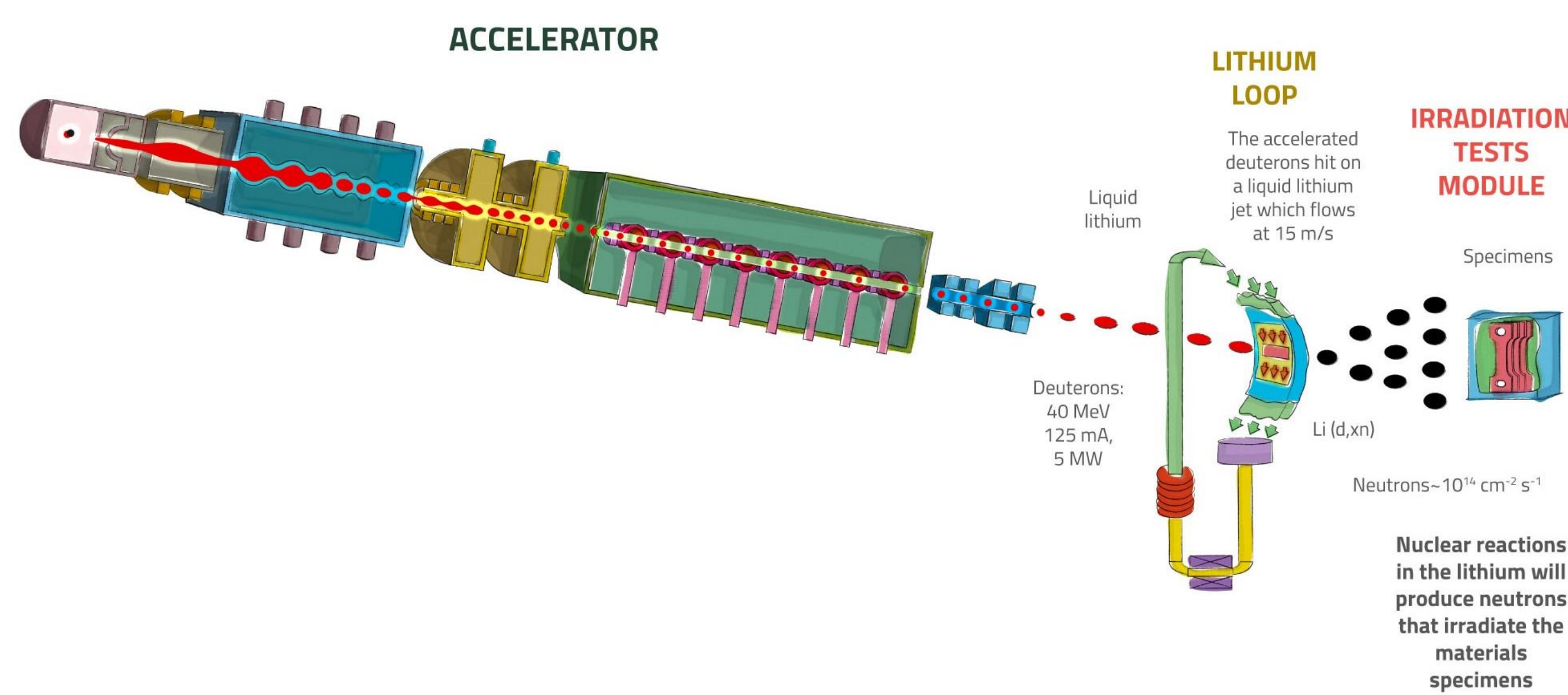
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INTRODUCTION

The **IFMIF-DONES** project, which aims to pave the way for understanding the effects of high-energy and high-flux neutronic irradiation on materials, is a crucial step for the future of fusion reactors.

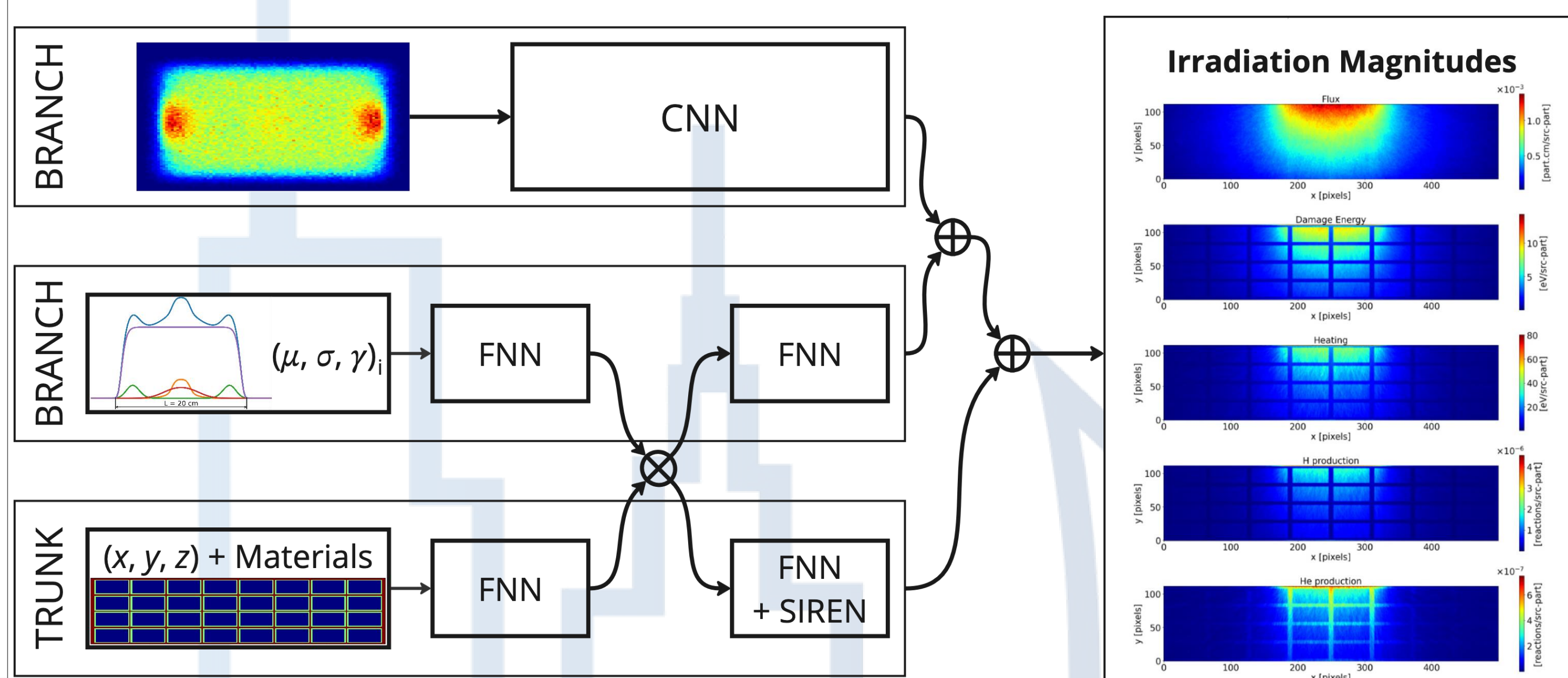


DESCRIPTION OF THE RESEARCH PROBLEM

Monte Carlo simulations are traditionally used to study neutronic transport studies; however, their computational cost limits the ability to optimise and control elements such as geometry or accelerator configuration, in the case of IFMIF-DONES. To alleviate this problem, the usage of differentiable **Deep Learning Surrogate Models** (DLSM) are employed.

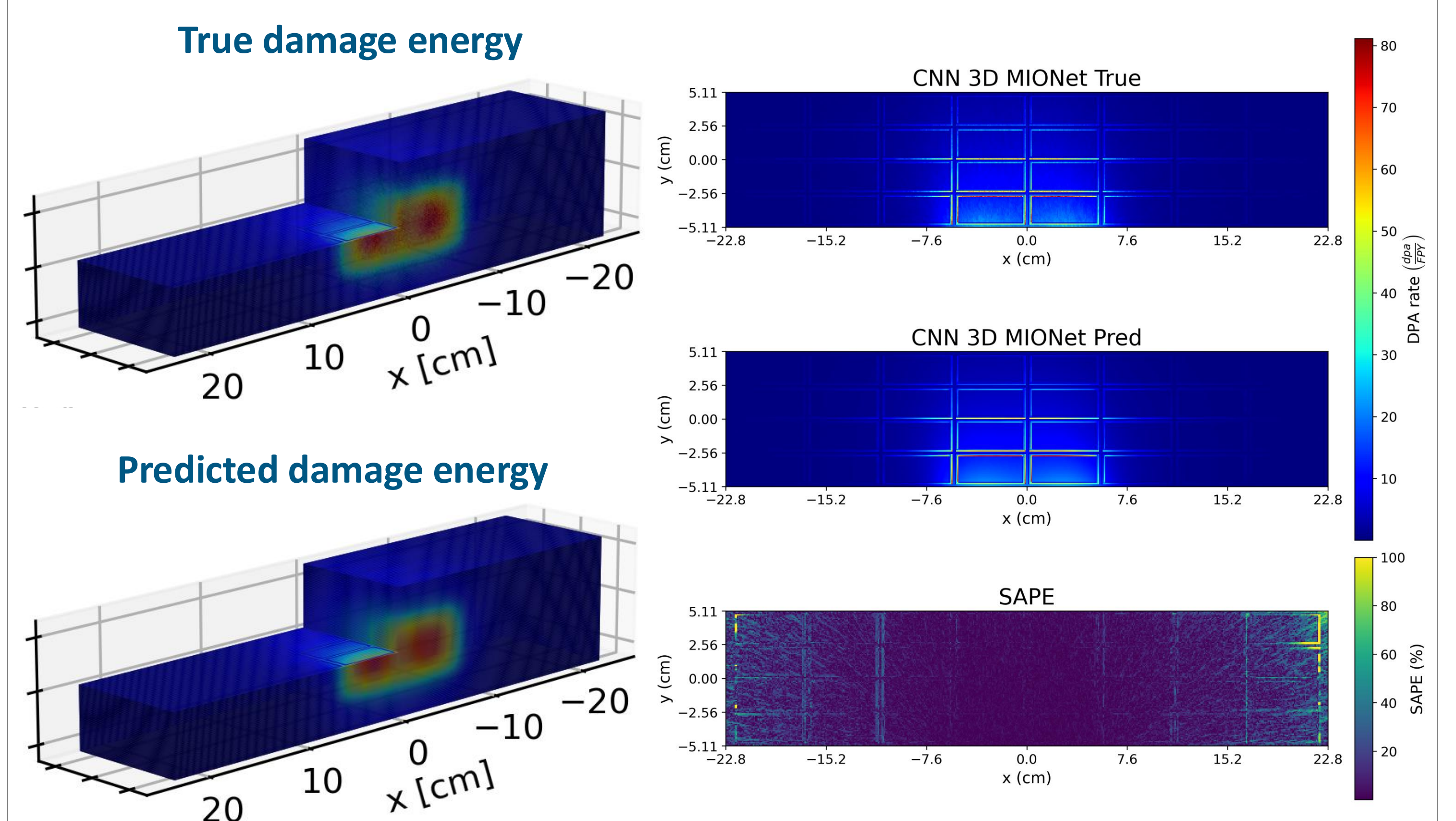
METHODOLOGY

1. Create a **soft-coupled Monte Carlo simulator** of the stripping reaction and neutron transport to predict neutron flux, damage energy, H and He production and heating.
2. Generate simulations for different inputs → deuteron beam lengths (L) and Gaussian parameters ($\sigma_i, \mu_i, \gamma_i$)
3. Train a **MIONet** (Trunk: geometry and materials, Branches: Gaussian parameters + footprint image)
4. Test predictions and their inference time
5. Use the nets differentiability to perform **Gradient Descent** (GD) and inverse design the Gaussian parameters

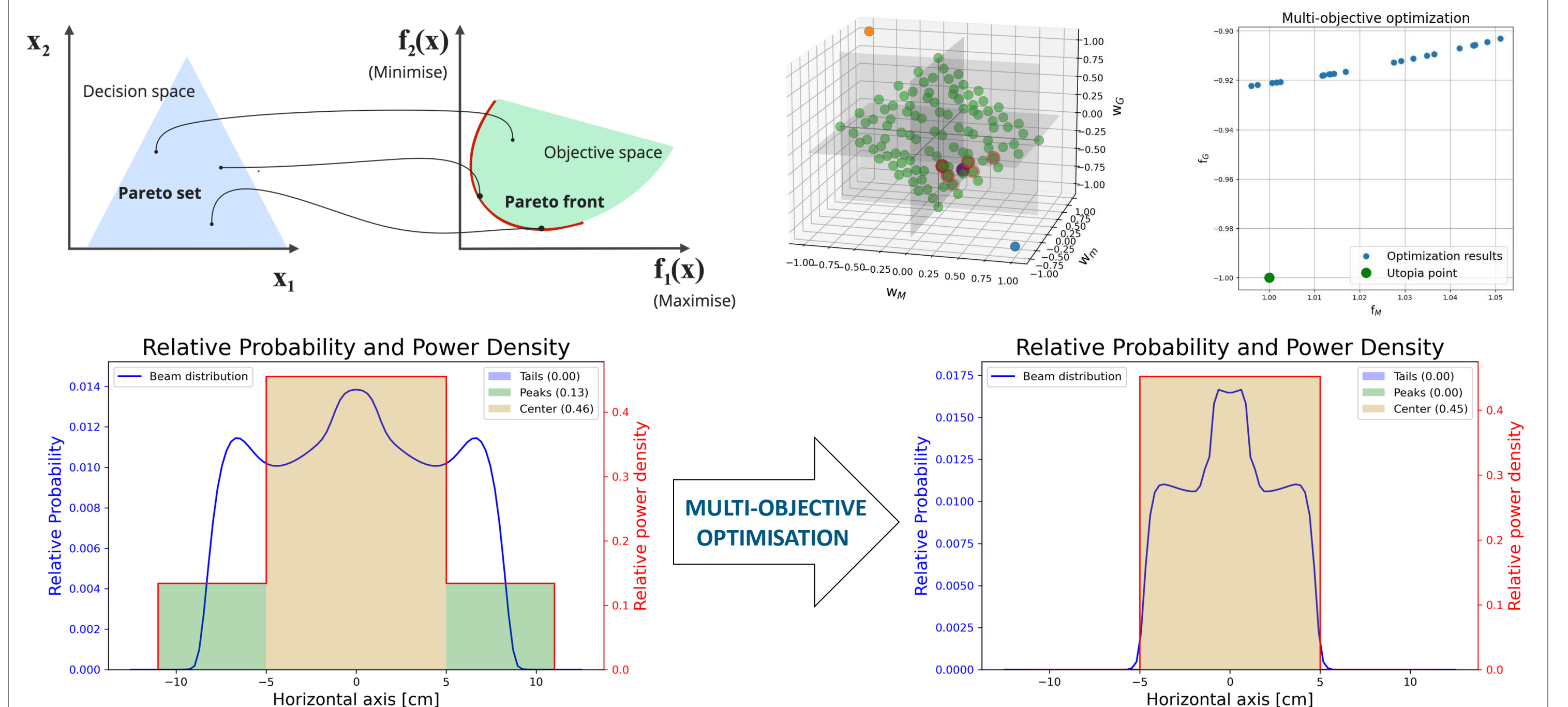


RESULTS

- **Speed:** speedup $\geq 10^6$, compared with the simulator
- **Accuracy:**
 - $R^2 = 0.994 \pm 0.00158$
 - sMAPE (%) = 8.74 ± 1.08



- **Optimisation:** GD, by leveraging differentiable model



CONCLUSIONS

- Computational efficiency of DLSM
- Potential for real-time control and design optimisation in fusion facilities
- **Future work:**
 - Improve training configuration with complete CLC geometric models
 - Couple with the accelerator's DRL control

ACKNOWLEDGEMENTS

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