

Introduction

- ITER requires accurate prediction of **burning plasma behavior**, where **heating, transport, radiation, and impurities** interact in complex nonlinear ways.
- Traditional **empirical models** lack flexibility; **NeuralPlasmaODE** provides a **physics-informed, data-driven** neural ODE framework for **core-edge plasma dynamics**.
- This study applies **sensitivity analysis** to identify which **parameters** (transport, ECR, impurities, ion orbit loss) most strongly affect **ITER's inductive scenario**.

Model Overview

- Plasma** is partitioned into **core and edge regions**, each governed by **particle and energy balance equations** solved with **neural ODEs**:

$$\frac{dn_{\sigma}^{\text{node}}}{dt} = s_{\sigma,\text{ext}}^{\text{node}} + s_{\sigma,\text{fus}}^{\text{node}} + s_{\sigma,\text{tran}}^{\text{node}} + s_{\sigma,\text{IOL}}^{\text{node}}$$

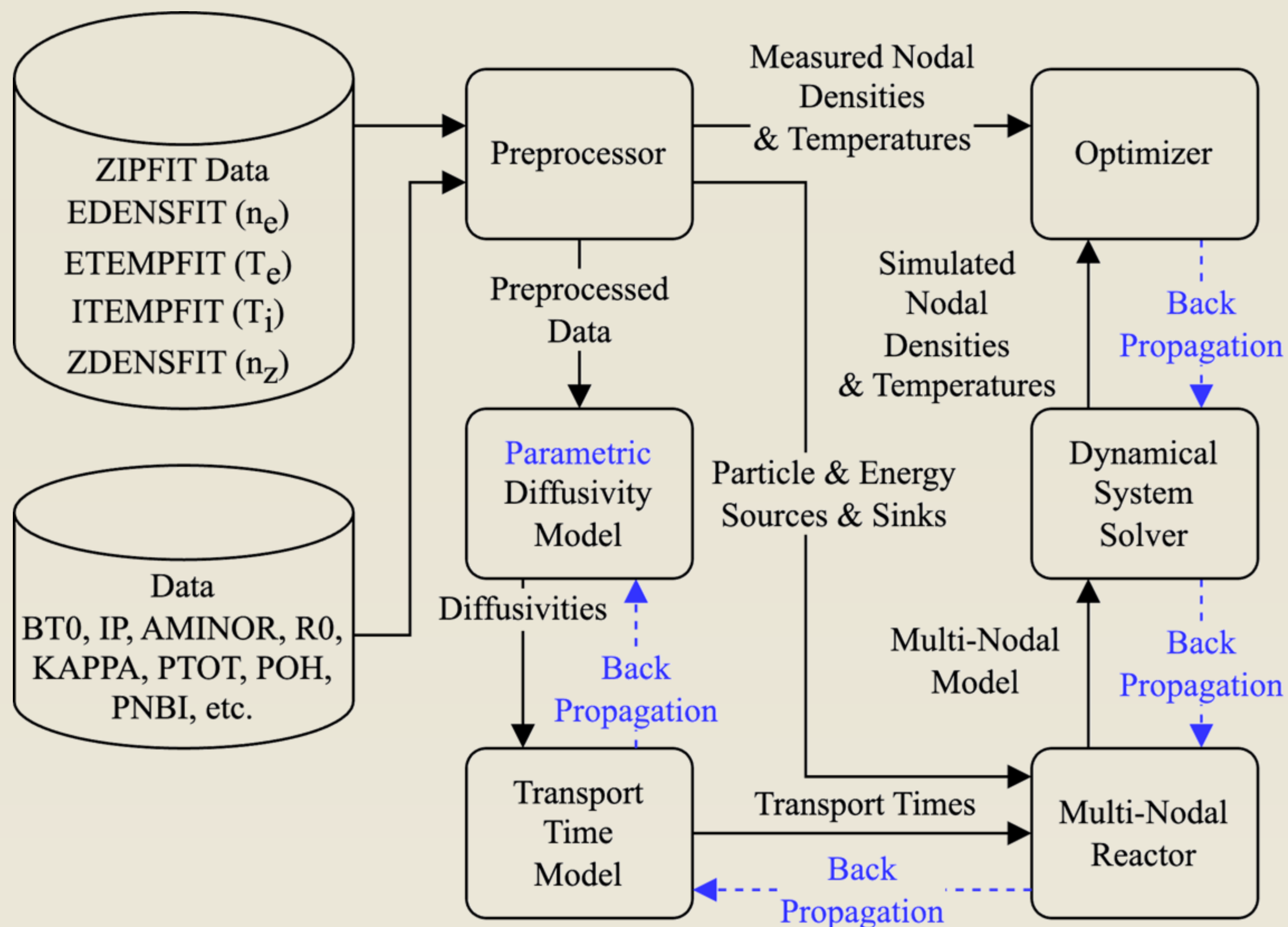
$$\frac{dU_{\sigma}^{\text{node}}}{dt} = p_{\sigma,\text{aux}}^{\text{node}} + p_{\sigma,\text{fus}}^{\text{node}} + Q_{\sigma}^{\text{node}} + p_{\sigma,\text{tran}}^{\text{node}} + p_{\sigma,\text{IOL}}^{\text{node}}$$

$$\frac{dU_e^{\text{node}}}{dt} = p_{\Omega}^{\text{node}} + p_{e,\text{aux}}^{\text{node}} + p_{e,\text{fus}}^{\text{node}} - p_{\text{rad}}^{\text{node}} + Q_e^{\text{node}} + p_{\sigma,\text{tran}}^{\text{node}}$$

- The model provides a **transparent yet data-informed** framework that balances **physics fidelity** with **flexibility**.

Computational Framework

- NeuralPlasmaODE** combines **Neural ODEs** with the **burning plasma dynamics model** to enhance predictive accuracy.



Sensitivity Analysis

- We apply **normalized sensitivity analysis** to measure the effect of small parameter changes on plasma responses:

$$S(y|p) = \frac{p}{y} \frac{\partial y}{\partial p}$$

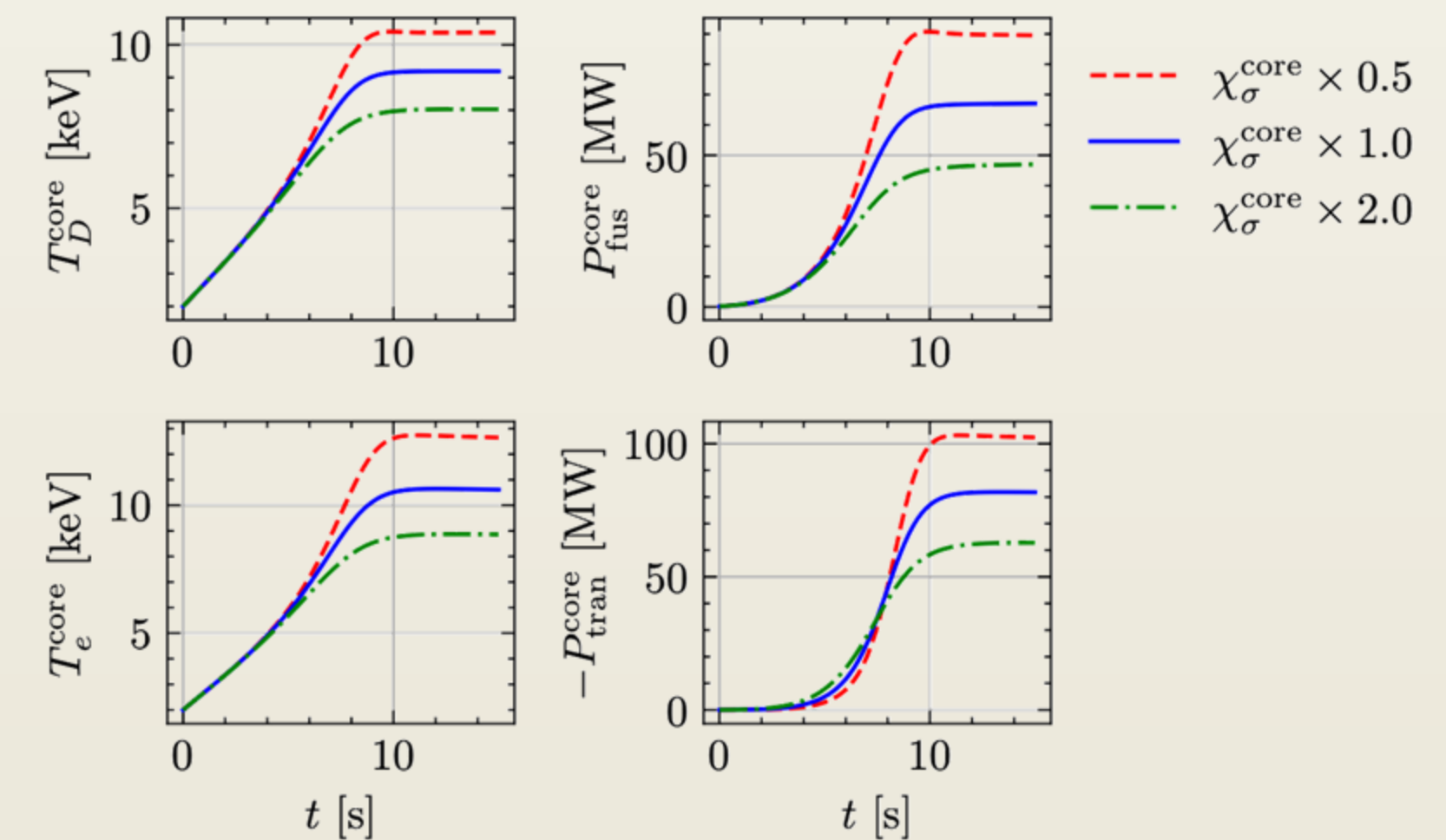
- Both **steady-state shifts** and **dynamic time evolution** were examined to capture **stability** and **transient effects**.

Conclusion

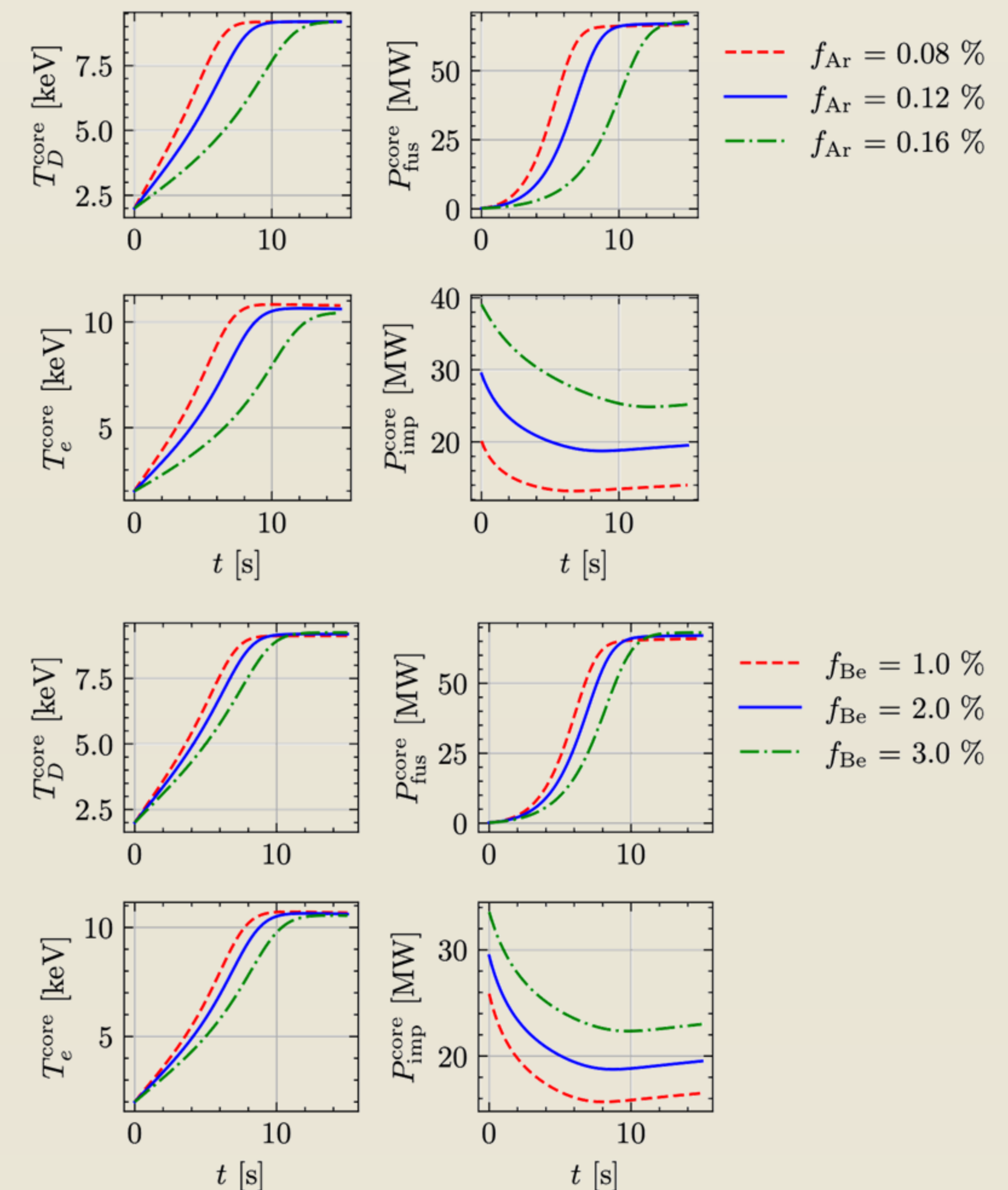
- Transport physics** (magnetic field, safety factor, temperature gradients) are the dominant control parameters for ITER performance.
- Impurity fractions**, especially argon, strongly impact confinement through enhanced radiative losses.
- NeuralPlasmaODE** with **sensitivity analysis** offers a predictive, interpretable tool for optimizing and controlling burning plasma scenarios.

Key Results

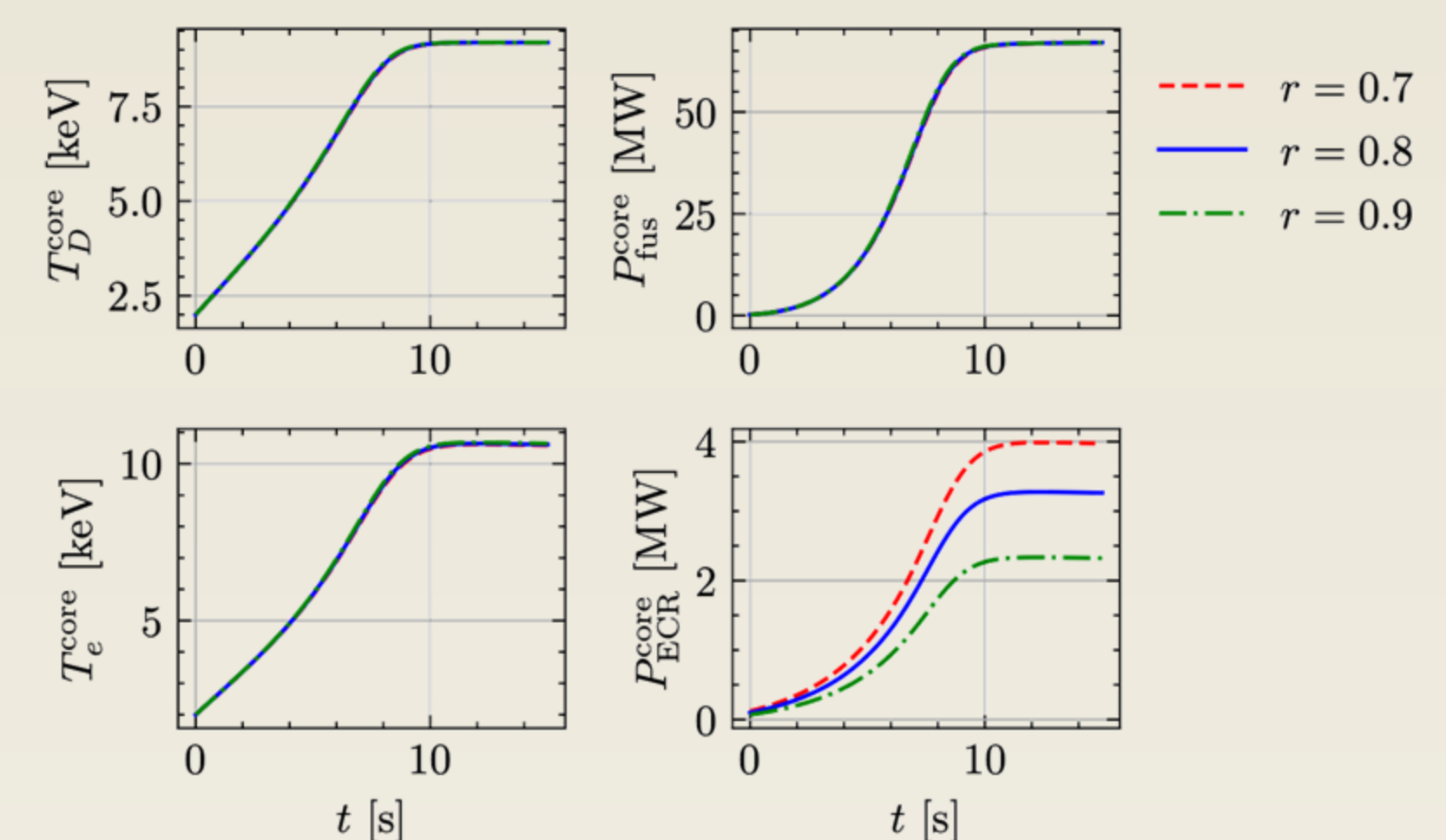
- Transport diffusivities** are the strongest drivers of plasma behavior. Magnetic field, safety factor, and temperature scaling exert dominant control, with temperature-dependent transport providing a stabilizing self-regulation against runaway heating.



- Impurity effects** are significant: argon impurities sharply reduce electron temperature via radiative losses, while beryllium has a weaker but still noticeable impact. Impurity management is thus critical for ITER performance.



- ECR parameters** (particularly wall reflectivity) influence electron energy balance but have modest overall effect compared to transport and impurities, though uncertainties in wall conditions remain important.



- Ion orbit loss** plays a negligible role in ITER's inductive scenario, confirming that transport and impurities are the key levers for confinement optimization.

References

- Stacey, Weston M. "A Nodal Model for Tokamak Burning Plasma Space-Time Dynamics." *Fusion Science and Technology* 77.2 (2021): 109-118.
- Liu, Zefang, and Stacey, Weston M. "Application of Neural Ordinary Differential Equations for Tokamak Plasma Dynamics Analysis." *ICLR 2024 Workshop on AI4DifferentialEquations In Science*.
- Liu, Zefang, and Weston M. Stacey. "Application of Neural Ordinary Differential Equations for ITER Burning Plasma Dynamics." *AAAI 2025 Workshop on AI to Accelerate Science and Engineering*.



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