Bayesian inference of Experimental Particle Transport in Tokamak Plasmas

F. Sciortino^{1,2}

T. Odstrčil³, N.T. Howard², E.S. Marmar², C. Chrystal³, A. Foster⁴, J.W. Hughes², S. Mordijck⁵, O. Meneghini³, S. P. Smith³, R. Reksoatmodjo⁵, J. E. Rice², K. E. Thome³

1. MPI-IPP

2. MIT

3. GA

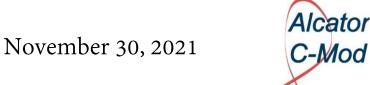
4. Harvard-Smithsonian

5. W&M

4th IAEA Technical Meeting on Fusion Data Processing, Validation and Analysis









Impurities and neutrals are key to core-edge integration

This talk is about inference of D's and v's and comparison to theory models

Parameter Estimation
Uncertainty Quantification
Model Selection

Question: are any discrepancies due to inadequate experimental analysis or wrong theory?

Approach: enable models to disagree with theory

advance forward models

integrated analysis of multiple measurements

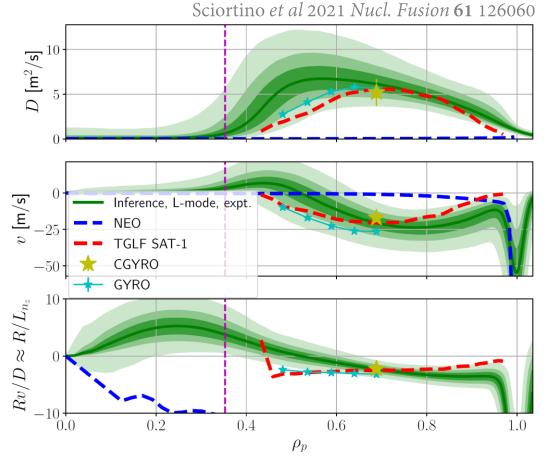
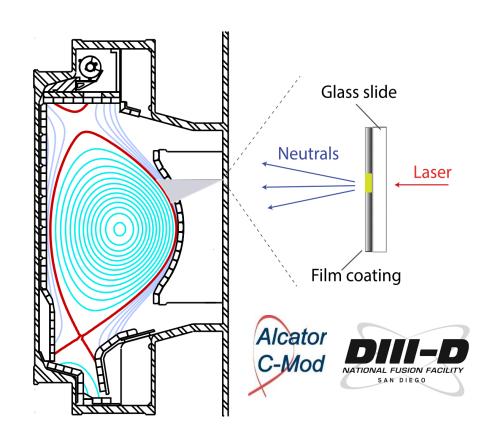


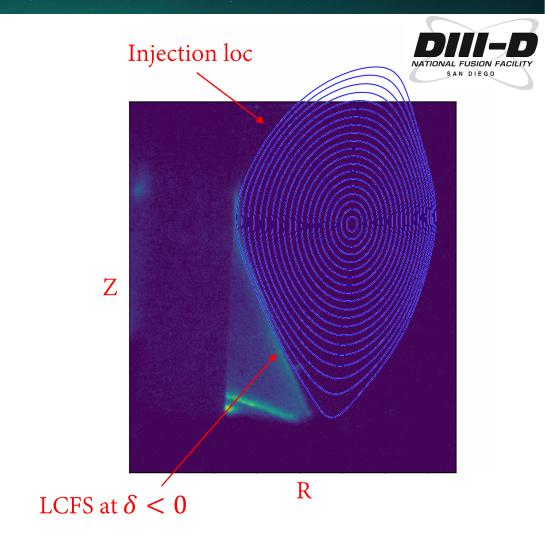
Fig: C-Mod example of D and v inference & comparison to theory models

Inferring particle transport from Laser Blow-Off (LBO) injections



Non-perturbative *trace* amounts of *non-recycling*, *non-intrinsic* ions

Alternative method for core transport: boron modulation via ICRH [Bruhn PPCF 2018, McDermott NF 2021]



unfiltered fast camera in DIII-D $\delta < 0$ experiment

I. Forward models for spectroscopy

II. Bayesian inference of particle transport coefficients

I. Forward models for spectroscopy

- <u>Simulation capabilities</u>: modeling with **Aurora**
- Physics fidelity: Charge Exchange (CX) with background neutrals
- Synthetic diagnostics: high-resolution spectral analysis

Aurora: a toolbox for particle transport and radiation modeling

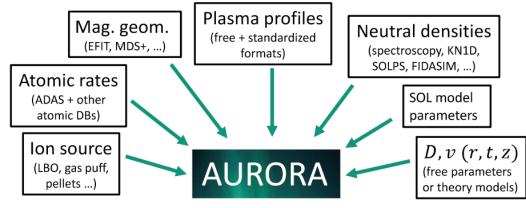
• Initially based on the STRAHL code, used for benchmarks

• Modern high-level interface between Python, Fortran and Julia Open-source (MIT license): https://aurora-fusion.readthedocs.io

• User-friendly and flexible radiation modeling using ADAS rates [Summers *et al* 2006 PPCF 48 263]

- 1.5D impurity transport simulations
 - Efficiently parallelized to infer transport coefficients
 - Options to model ion superstages, arbitrary Z dependences...





1.5D ion transport forward model

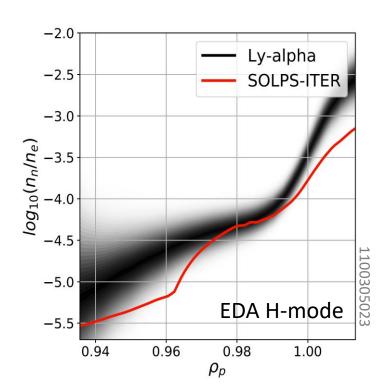
$$\downarrow \\ n_z, \epsilon_z, L_z(r, t, z)$$

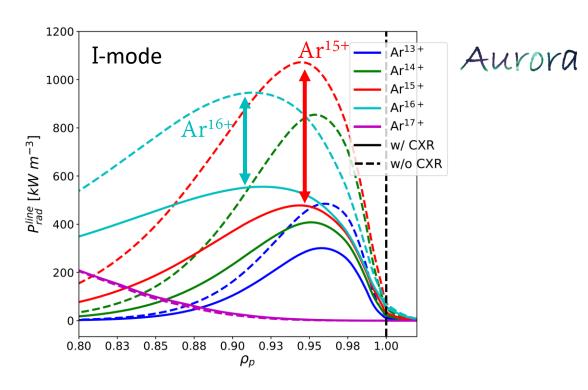
Physics fidelity: Charge Exchange (CX) between impurities and neutrals

CX can strongly affect the ionization balance of impurities – must be considered for particle transport studies [Dux, NF 2020]

• Direct neutrals measurements are rare, but available on C-Mod via Ly_{α} spectroscopy at the outer midplane

Favorable comparison to SOLPS-ITER [Wiesen JNM 2015] using the EIRENE [Reiter FST 2005] Monte Carlo neutral model

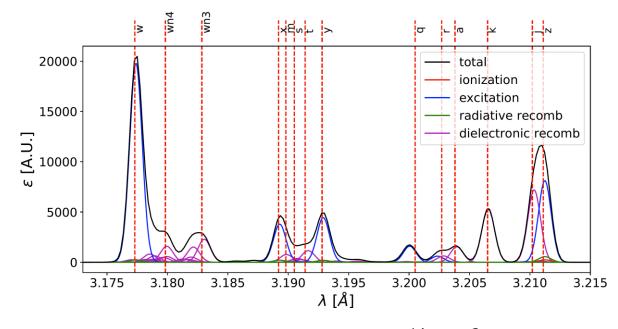






Advanced synthetic diagnostics: Modeling the full Ca K_{α} spectrum of XICS

- On C-Mod, the X-ray Imaging Crystal Spectroscopy (XICS) diagnostic measures the entire Ca K_{α} (n=2-1) spectrum
- New atomic data compilation from the atomDB database enables forward modeling of the entire spectral range





 $n_e = 10^{14} \ cm^{-3}; T_e = 1 \ keV$

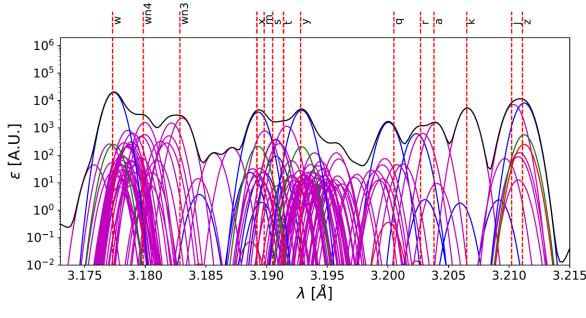
Advanced synthetic diagnostics: Modeling the full Ca K_{α} spectrum of XICS

- On C-Mod, the X-ray Imaging Crystal Spectroscopy (XICS) diagnostic measures the entire Ca K_{α} (n=2-1) spectrum
- New atomic data compilation from the atomDB database enables forward modeling of the **entire spectral range**

Includes hundreds of satellite lines from Li-like Ca

• Line-integrated spectra can now be used to infer D & v

Stronger constraints on particle transport



 $n_e = 10^{14} \ cm^{-3}; T_e = 1 \ keV$

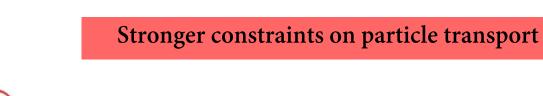


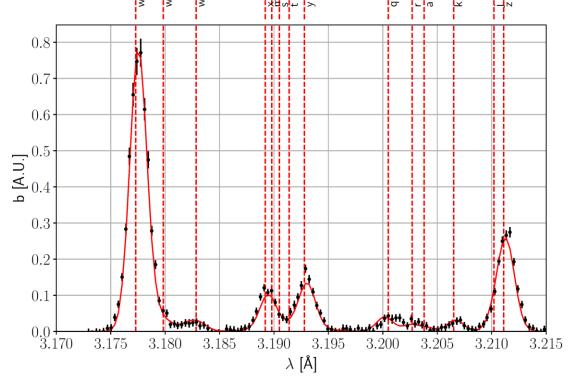
Advanced synthetic diagnostics: Modeling the full Ca K_{α} spectrum of XICS

- On C-Mod, the X-ray Imaging Crystal Spectroscopy (XICS) diagnostic measures the entire Ca K_{α} (n=2-1) spectrum
- New atomic data compilation from the atomDB database enables forward modeling of the **entire spectral range**

Includes hundreds of satellite lines from Li-like Ca

Line-integrated spectra can now be used to infer D & v





Experimental XICS Ca K_{α} spectrum matched by forward model

C-Mod

I. Forward models for spectroscopy

II. Bayesian inference of particle transport coefficients

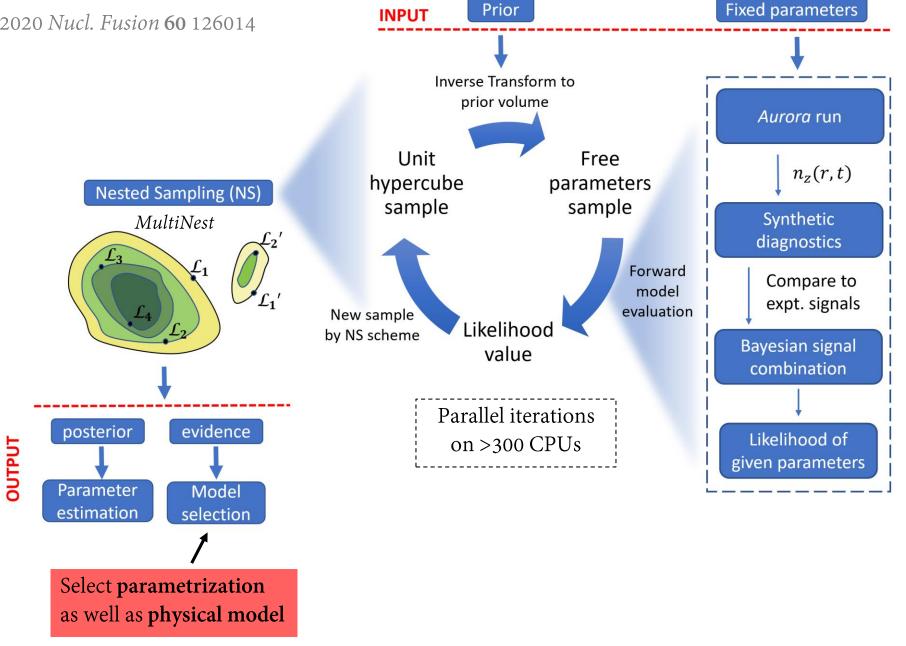
II. Bayesian inference of particle transport coefficients

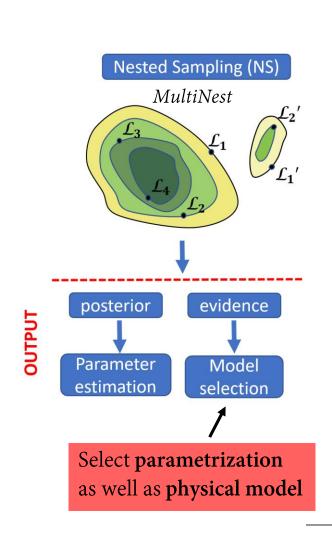
Methods:

- Model selection
- Physically-correlated priors
- Free spline knots
- Multi-species constraints

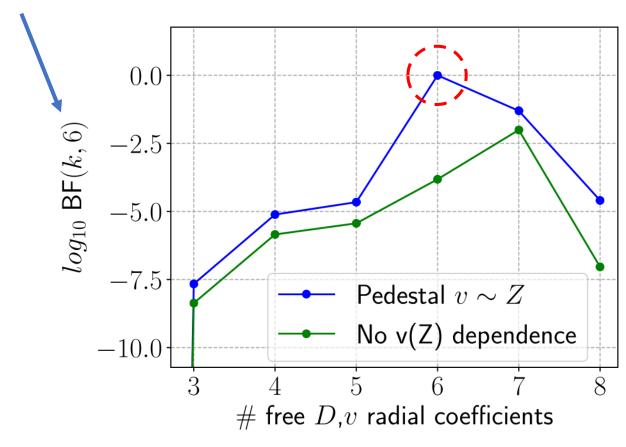
Applications:

- On C-Mod: L-, H-, and I-mode
- On DIII-D: RMP H-mode, $\delta < 0$





Bayesian evidence normalized to best case



Physical Bayesian Sampling of Transport Coefficients

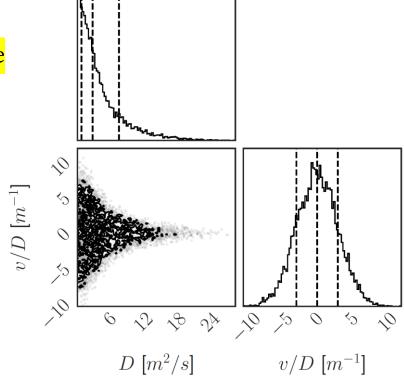
Problem:

How to ensure that sampling 2 parameters (e.g. D and v) within priors gives physical ratios/products (e.g. v/D)?

Approach/solution:

Constrain ratios/products of parameters by sampling in the complex plane

• Simple idea: sample $tan(\theta) = \chi v/D$ and $r = \sqrt{|\chi v|^2 + D^2}$ with χ a parameter that separates D and v scales



Free spline knots with forced identifiability

Problem:

Fixing spline knots can limit models & lead to under-/over-fitting if not combined with model selection

Approach #1:

Use non-parametric description of profiles (high-dim, hard time-dependent) – see *T. Nishizawa's earlier talk*

Approach #2:

Free spline knots sampled within a hyper-triangle, with prior $\pi(\theta) = \begin{cases} \frac{1}{n!(\theta_{\text{max}} - \theta_{\text{min}})^n} & \text{for } \theta_{\text{min}} < \theta_1 < \dots < \theta_n < \theta_{\text{max}} \\ 0 & \text{otherwise.} \end{cases}$

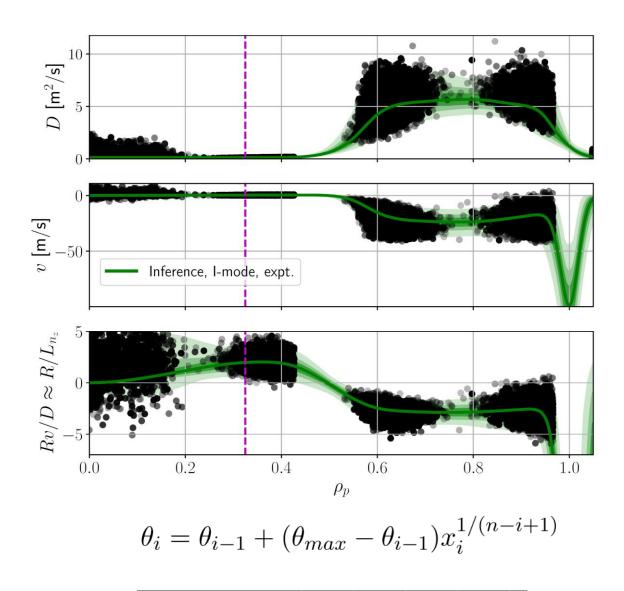
After some algebra, one finds that

$$\theta_i = \theta_{i-1} + (\theta_{max} - \theta_{i-1}) x_i^{1/(n-i+1)}$$

unit-hypercube sample

transforms unit-hypercube samples into ordered, identifiable, free spline knots

Free spline knots with forced identifiability

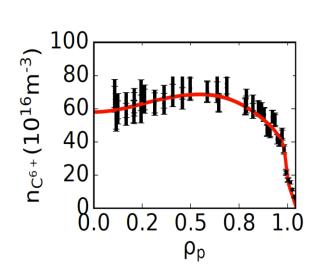


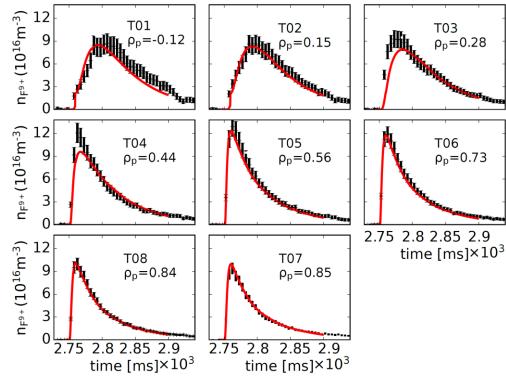
Constraints from multiple diagnostics + species: new important tools for validation



New OMFIT ImpRad module is now generalized to handle an arbitrary number of species

We constrain DIII-D transport using both intrinsic (quasi-steady) C and (rapidly-evolving) LBO-injected ions



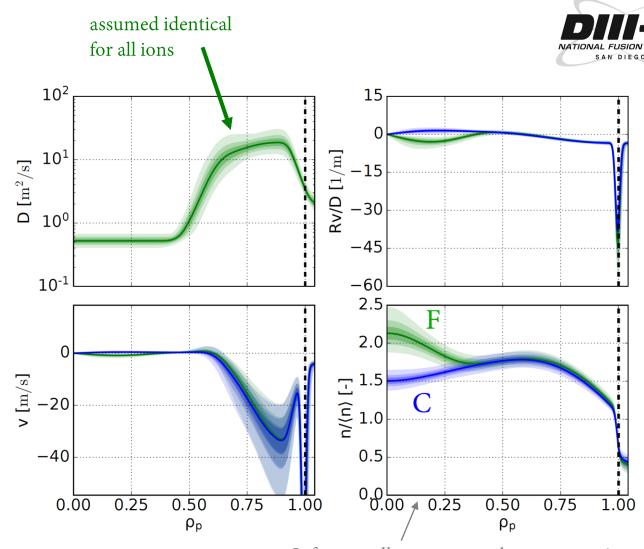


Further constraints on impurity transport from SXR and core-SPRED

Impurity transport inference in diverted δ < 0 low-power L-mode

Multi-species impurity transport inference

• F (LBO) and C (intrinsic)



Inference allows v to vary between species near axis and in pedestal

Impurity transport inference in diverted δ < 0 low-power L-mode

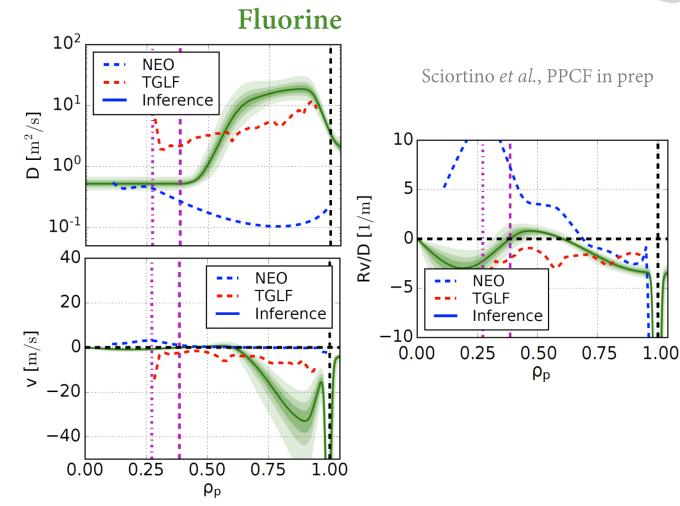


Multi-species impurity transport inference

• F (LBO) and C (intrinsic)

Comparison to theoretical models across multiple discharges suggests peaking mismatch

• Experimental flat/hollow profiles often not captured by theory



Summary

Bayesian inference of particle transport coefficients on C-Mod and DIII-D





- Parameter Estimation, Uncertainty Quantification, Model Selection
- Focus on high physics-fidelity, including CX, multiple diagnostics, multiple species, etc.
- Techniques to improve Bayesian inferences beyond particle/impurity transport

| | Alcator C-Mod | DIII-D |
|-----------------------|---|--|
| Regimes | L-, I-, and EDA H-mode | Diverted $\delta < 0$ RMP ELM-suppressed H-mode |
| Experimental analysis | Modeling of Ca \mathbf{K}_{α} spectrum + EUV line ratios | Multi-species inferences with quasi- steady C + LBO-injected ions |

• Aurora + ImpRad make similar analysis accessible and reproducible by multiple researchers

Sciortino *et al* 2021 *PPCF* **63** 112001

Sciortino et al 2021 Nucl. Fusion 61 126060