Applying self-consistent electron heat transport and ECH deposition profile estimation in DIII-D

27th Joint EU-US Transport Task Force Meeting

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Introduction

- Transport is often studied perturbatively^[1-5]
- Power deposition and transport estimates are traditionally separated
- However, edge turbulence can spoil deposition localization^[6-10]
- Need for simultaneous, self-consistent estimate of transport coefficients and power deposition profile^[6,10-13]



[1] J.D. Callen et al. (1977). Phys. Rev. Lett. 38 491-4 [2] H.J. Hatfuss et al. (1994). Plasma Phys. Control. Fusion. 36 B17–B37 [7] C. Tsironis et al. (2009). Phys. Plasmas. 16 112510 [3] N.J. Lopes Cardozo. (1995). Plasma Phys. Control. Fusion. 37 799 [4] F. Ryter et al. (2010). Plasma Phys. Control. Fusion. 52 124043 [5] C.C. Petty et al. (2015). Nucl. Fusion. 55 083011

[6] K.K. Kirov et al. (2002). Plasma Phys. Control. Fusion. 44 2583 [8] A. Snicker et al. (2018). Nucl. Fusion. 58 016002 [9] O. Chellaï et al. (2021). Nucl. Fusion. 61 066011 [10] M.W. Brookman et al. (2023). Nucl. Fusion. 63 044001

[11] E.A. Lerche et al. (2008). Plasma Phys. Control. Fusion. 50 035003 [12] A. Manini et al. (2003). Nucl. Fusion. 43 490-511 [13] F. Ryter et al. (2003) Nucl. Fusion. 43 1396

- Spatially varying, joint transport and power deposition estimates
- Assumptions and experimental checks
- Sensitivity analysis
- Experimental power deposition estimates in DIII-D
- Impact for ITER

Linearization through perturbative experiments – the model



Linearization through perturbative experiments – frequency domain



Estimation



Justification of model assumptions

• Density is not modulated ($\frac{\partial}{\partial t}n_e = 0$)



Justification of model assumptions

• Linearity of the plasma response



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Sensitivity of model coefficients

- Sensitivity of estimates of one parameter to errors in the estimates of others
- Result: P estimate is insensitive to variations in D and V



Sensitivity of model coefficients

- Small errors in P get compensated in other parameters
- Large deviations in D and V
- Due to different sensitivity of the temperature profile to different coefficients



Sensitivity of model coefficients

- Sensitivity of the temperature profile to different parameters under noisy conditions
- Depends on modulation frequency
- The effect of **P dominates**
- Future, large devices: more deposition broadening, so joint estimation required
- Depending on the coefficient of interest, magnitude or phase response in different frequency ranges



- Maximum Likelihood Estimator [1] (MLE)
 - Estimate D,V,K,P constant on subdomain
 - Taking noise covariances into account
 - Nonlinear optimization
- Frequency Domain Least Squares ^[2] (FDLS)
 - Parametrize spatial dependence of D,V,K,P in terms of basis functions
 - Derive closed-form linear least squares solution for hyperparameters
- Flux fit ^[3,4] (FF)
 - Parametrization of P as a modified Gaussian
 - Parametrize heat flux q in terms of D,V as basis functions
 - Fit the flux term to the deposition term using nonlinear optimization

• Break-in-slope^[5] (BIS)

- Trusted method, used as reference
- Time-domain, no transport
- Requires step in power; deposition proportional to break in temperature slope

 M. Van Berkel *et al.* (2014). *Automatica*. **50**. 2113–2119.
 R. J. R. van Kampen, *et al.* (2020) *IEEE Control Syst. Lett.*. **5**. 1681-1686.
 M. W. Brookman *et al.* (2021). *Phys. Plasmas.* **28**. 42507.
 Slief *et al.* (2022). *Phys. Plasmas.* **29**. 010703.
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$$P(\rho', \sigma, \mu, A, \alpha) \widetilde{U}(\omega) - \frac{3}{2}n_e(\rho')i\omega\Theta(\rho', \omega)$$

$$\tilde{q}_e(\rho,\omega) = -D(\rho)n_e(\rho)\frac{\partial\Theta}{\partial\rho}(\rho,\omega) + V(\rho)n_e(\rho)\Theta(\rho,\omega)$$

ρ

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Results – DIII-D

- Estimation of P in DIII-D: factor 1.5 3
 broadening with respect to ray-tracing^[1]
- Broad agreement between different experimental methods



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Results – Implications for ITER

- Estimation of P in DIII-D: factor 1.5 3
 broadening with respect to ray-tracing^[1]
- No broadening: business-as-usual



Results – Implications for ITER

- Estimation of P in DIII-D: factor 1.5 3
 broadening with respect to ray-tracing^[1]
- Edge case at factor 2.4 broadening: all $\frac{\sqrt{m}}{m}$ available power required for suppression $\frac{m}{m}$



Results – Implications for ITER

- Estimation of P in DIII-D: factor 1.5 3
 broadening with respect to ray-tracing^[1]
- At factor 3 broadening, more power required than available



Conclusions & Outlook

- Several methods can simultaneously estimate any combination of D,V,K,P
- Key assumptions hold in relevant experimental conditions
- Coefficients have different sensitivities to errors in the others
- Temperature profile sensitivity is **modulation frequency dependent**
- Significant broadening in DIII-D P estimates
- This could have **consequences for ITER NTM control**

Outlook:

- Joint transport coefficient estimation
- Time varying effects using Kalman filters