



NSTX-U modeling and control

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0. RT-MPTS Diagnostics (M. Kaur, F. Laggner, G. Tchilinguirian, R. Rozenblat)





Rt-MPTS for Control and Rt-Kinetic-EFIT

Project Objectives:

a) Compute n_e and T_e profiles in real-

time (2021)

b) Share this information with PCS (2021)

c) Enhance rt-EFIT with rt-MPTS data (2022)

d) Develop control algorithms to achieve and stabilize scenarios with prescribed edge and core structures (2022-2023)

e) Improved disruption avoidance (2023-2025)



Control Example: Pedestal Density (DIII-D)



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NSTX-U RT-MPTS Copy Running at LHD System



The digitizer cards and the real time server (right) as implemented at LHD. The rt-MPTS T_e result is compared with the offline TS system (bottom left), with the rt-MPTS system in red and reference in blue.

NSTX-U RT-Thomson: Help Advanced Scenarios





Comparison of pre-lithium ELMy discharge (black), and two postlithium discharges with different NBI power (blue, red)

Demonstration of pedestal density feedback for super H-mode. As the gas is increased from discharge to discharge (feed-forward), the MP coil current is feedback controlled such that a constant pedestal top electron density (ne,ped) is achieved.

Control/Physics Overview

- 1. Snowflake divertor (SFD) feedback control
- 2. Optimization of SFD power and particle exhaust
- 3. Improving SFD reconstruction via infrared thermography
- 4. Shape control model validation
- 5. Optimization of rampup feedforward trajectories
- 6. Neural networks for fast shape reconstruction and modeling



1. Snowflake divertor feedback control



Snowflake divertor control – (Vail)

- Snowflake divertor:
 - Second order null
 - High flux expansion, heat flux splitting phenomenon, detachment access
- Control algorithm:
 - Proportional control for isoflux shape targets
 - LQI for snowflake targets (dr, dθ)

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High-fidelity closed loop simulation indicates need for time-varying model in control

- Developed and tested with nonlinear closed loop simulation environment.
 - Linear time-invariant (LTI) system insufficient for moderate-large changes. Must use time-varying (LTV) model







2. Optimization of SFD power and particle exhaust (Vail, Izacard)



Snowflake divertor heat exhaust

- Study of power and particle exhaust capabilities using in NSTXU using cryopump + snowflake
- Develop simple, fast heat flux diffusion model [Vail, NME] and validated with UEDGE
 - Heat flux diffuses across ψ but in separate domains for SFD-minus





SFD power exhaust with cryopumping

- To pump 10MW NBI power, need P > 0.83 mTorr at pump inlet [Vail, NME]
 - Assume 24 kL/s volumetric pump rate for liquid helium cooled cryopump
- At pump optimal location, 83% of SFD equilibria in database meet this condition.
- UEDGE simulation: with pumping
 - Te rises at strike points due to reduced collisionality
 - SOL power is redistributed among strike points
 → changes the ideal 'power balanced' SFD configuration







3. Improving SFD reconstruction via infrared thermography (Wai, NME, 2020)



Infrared thermograpy (IRTV) for SFD reconstruction

- Snowflake plus: secondary x-point lies in the private flux region. Scrape-off layer (SOL) fieldlines directly intersect divertor in 2 locations ⇒ 2 heat flux peaks.
- Snowflake minus: secondary x-point lies in the SOL.
 Fieldlines directly intersect divertor in 3 locations ⇒ 3 heat flux peaks.
- Equilibrium vs. IRTV inconsistencies
 - Strike point location mismatch
 - Occasionally, incorrect # of heat flux peaks for the snowflake type
 - IRTV used to improve equilibrium
 - Useful for control (feedback on x-point locations [Kolemen, 2018])
 - Geometry sensitive to unmeasured divertor currents. Potential use as diagnostic for bootstrap current.







Heat flux power fraction in the SFD

- Power fraction f_{sp4} measured from the divertor heat flux profile. At each peak, $P_{pk} = \int 2\pi R(s)q_{\perp}(s)ds$
- Secondary separatrix position $r_{mid, xp2}$ measured from EFIT equilibrium.
- Data is selected from subset of shots that have wide range of x-point separation and fit to:

$$f_{sp4} := \frac{P_4}{P_2 + P_4}$$

$$=\frac{\int_{r_{mid,xp2}}^{\infty}q_{\perp}^{mid,peak}e^{-r/\lambda_{q}^{eff}}dr}{\int_{0}^{\infty}q_{\perp}^{mid,peak}e^{-r/\lambda_{q}^{eff}}dr}$$











IRTV used to constrain strike points, x-points

- Strike points and outboard power fraction are mapped to x-point locations [?].
- Final heat flux profiles match consistently.
- Modification to edge current observed symbolic application to edge current

100





 q_{\perp}^{div}/\sum

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0.2

Extension to NSTXU

- Technique was developed with DIII-D but principles can extend to NSTXU
 - Fewer constraints available to lack of visibility on inner wall, CHI gap





The 2 observable LFS heat flux peaks. Inner peak not observed, and outermost peak obscured by CHI gap.



4. Shape control model validation (Wai, Boyer)



Model validation for shaping feedforward control

• Shape control based on the toroidal circuit equation which can be transformed to time-varying state-space system.

$$\dot{\Psi}_{ss,plasma} = \frac{\partial \Psi_{s,plasma}}{\partial I_s} \dot{I}_s$$

$$\dot{I} = A(t)I + B(t)v$$

- Shape control algorithm relies entirely on PID feedback with no feedforward.
- Large shape errors at startup, and small errors during flattop lead to poor performance.
- A design tool that translates a target shape evolution into approximate feedforward current evolution is needed



Model validation

- First step is to validate the model versus experiment, so that current evolution can be simulated.
 - Vessel currents play a strong role on equilibrium, especially with NSTXU short pulse length
- Use a greybox fitting procedure to identify: coil power supply internal inductances, vacuum vessel resistances, plasma resistance Rp(t).



Vacuum vessel currents show inconsistency with measured.



Vessel fitting results modify resistance in bellows, passive plates

- Fitted model parameters give much better match to vessel currents, plasma current.
- Resistances that changed the most with fitting are consistent with expectations
 - Bellows
 - Passive plates, 'effective' resistances difficult to measure because of nontoroidal eddy currents





NSTY-II

F

1.2 1.4 1.6 1.8

r [m]

Identify plasma resistance Rp(t)

- Plasma resistance an important timevarying parameter to identify
- Sets the trajectory for OH coil
- Currently, using values fit from the dynamics model.
 - In future, could couple with evolution predictors (Nubeam net, current profile evolution, Te/ne modelling)
 - Fitted values not far from simple Te modeling with Spitzer resistivity

$$Res = \frac{2\pi R_0 \eta}{\pi \kappa a^2} \qquad \eta \approx \frac{\pi e^2 m^{1/2}}{(4\pi \epsilon_0)^2 (KT_e)^{3/2}} \ln \Lambda$$





5. Optimization of rampup feedforward trajectories (Wai, Boyer)



Optimization of feedforward trajectories -

Iterative time slice algorithm:





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Optimization to find feedforward trajectories

• Define a cost function of the form

$$J = \sum_{i=1}^{N} (I_{k+i} - r_{k+i})^T Q (I_{k+i} - r_{k+i}) + \Delta I_{k+i}^T Q_v \Delta I_{k+i}$$

Subject to: (dynamics constraint)

 $\dot{I} = A(t)I + B(t)v$

 The reference trajectory r depends on vessel currents, and A(t)/B(t) depend on the equilibrium, so this problem should be solved iteratively.





6. Neural networks for fast shape reconstruction and modeling (Boyer, Wai)



Eqnet finds equilibrium from coil currents

- Feedforward trajectory planner could be useful as an operator tool, especially if results can be obtained quickly! ~1 min
- Several steps currently in optimization take ~1hour
 - Free boundary GS solutions for all equilibria timeslices
 - Identify plasma flux response for all times
- **Eqnet:** finds approximate flux map based on coil currents, vessel currents, q and p profiles.
 - Use PCA reduction of inputs, n_components selected for 99% explained variance





Eqnet

- Uses separate PCA components for rampup and flattop
 - Allows accurate for accurate estimation during rampup (~ t < 300ms) since rampup samples underrepresented in database
- Eqnet has standard multi-layer perceptron (MLP) framework.



Generic MLP network







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Future extensions to include estimation of response

• Future extension

- Train NN to identify the plasma response
- In theory, identifying plasma response does not require much more representation capacity than estimating the equilibrium
- Targets could be identified from code (gspert) or from actual data (derivative of the equilibrium wrt time, minus the vacuum response)



Plasma response calculated from the gspert code



References

- P.J. Vail, et al. "Design and simulation of the snowflake divertor control for NSTX-U", PPCF, 2019.
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