Model Predictive Control with Integral Action for the Rotational Transform Profile Tracking in NSTX-U

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MPC of the rotational transform profile in NSTX-U



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Introduction to Model Predictive Control (MPC)

- A dynamic model of the system is used to predict the system output for a future time horizon.
- Ontrol sequence is calculated to optimize an objective function.
- Receding strategy: Only first element of the control sequence is applied at each step!



Hu, C. et al., Energies (2015)

Camacho and Bordons, Springer-Verlag (1999)

Magnetic Diffusion Equation

• The evolution of the **poloidal magnetic flux**, ψ is given by the **Magnetic Diffusion Equation**:

$$\frac{\partial \psi}{\partial t} = \frac{\eta(T_e)}{\mu_0 \rho_b^2 \hat{F}^2} \frac{1}{\hat{\rho}} \frac{\partial}{\partial \hat{\rho}} \left(\hat{\rho} D_{\psi} \frac{\partial \psi}{\partial \hat{\rho}} \right) + R_0 \hat{H} \eta(T_e) \frac{\langle \bar{j}_{NI} \cdot \bar{B} \rangle}{B_{\phi,0}},$$



Control – Oriented Current Profile Evolution Model

Reduction of Control-Oriented FPD Model

"FPD Model"

- $\frac{\partial \psi}{\partial t} = f\left(\psi, \frac{\partial \psi}{\partial \hat{\rho}}, \frac{\partial^2 \psi}{\partial \hat{\rho}^2}, \bar{u}, t\right)$ $\dot{\theta}(t) = g(\theta(t), \bar{u}(t))$ $\tilde{\theta}(t) = A(t)\tilde{\theta}(t) + B(t)\tilde{u}(t)$ $\tilde{\tilde{\theta}}(t) = A\tilde{\theta}(t) + B\tilde{u}(t)$ $\dot{\tilde{\iota}}(t) = \bar{A}\tilde{\iota}(t) + \bar{B}\tilde{u}(t)$ $\tilde{\iota}(k+1) = \bar{A}_d \tilde{\iota}(k) + \bar{B}_d \tilde{u}(k)$

- MDE combined with the simplified models of n_e , T_e , η , and \overline{j}_{ni} can be written as an infinite-dimensional PDE, where $\psi(\hat{\rho}, t)$ is the poloidal magnetic flux, and \overline{u} is the nonlinear iputs, i.e., $\overline{u} = p(u)$.
- FPD model is discretized in space to generate a set of nonlinear ODEs, where $\theta(t) = [\theta_1(t), \dots, \theta_n(t)]^T$, with $\theta(\hat{\rho}, t) = \partial \psi / \partial \hat{\rho}$ is the poloidal flux gradient.
- So The model is linearized around a set of reference physical inputs u_r , and states θ_r , yielding an LTV model, where $\tilde{\theta}(t) = \theta(t) \theta_r(t)$, and $\tilde{u}(t) = u(t) u_r(t)$.
- Further simplification leading to an LTI model is possible by setting $A = A(t_s)$ and $B = B(t_s)$, where t_s is some time during the flat-top phase of the discharge.
- Since $\iota(\hat{\rho}, t) = -\theta(\hat{\rho}, t)/B_{\phi,0}\rho_b^2\hat{\rho}$, the LTI model for $\tilde{\theta}$ can be converted into an LTI model for $\tilde{\iota}$, where $\bar{A} = T^{-1}AT$, $\bar{B} = T^{-1}B$, and $T = -\text{diag}(B_0\rho_b^2\hat{\rho}_i)$.
 - Finally, the *i* model is converted to discrete-time, and an output equation is added to select the reference-tracking states.

MPC Formulation with Integral Action

 $\tilde{i}(k+1) = \tilde{i}(k)$

• Rewrite the discrete, LTI model of the ι -profile in terms of the state increment, $\Delta \tilde{\iota}(k+1)$ and output increment, $\Delta y(k+1)$ so that input is the control increment, $\Delta \tilde{u}(k)$.

$$\Delta \tilde{\iota}(k+1) = \bar{A}_d \Delta \tilde{\iota}(k) + \bar{B}_d \Delta \tilde{\iota}(k)$$
(2)

$$\underbrace{\Delta y(k+1)}_{y(k+1)-y(k)} = \overline{C}_d \overline{A}_d \Delta \tilde{\iota}(k) + \overline{C}_d \overline{B}_d \underbrace{\Delta \tilde{\iota}(k)}_{\tilde{\iota}(k)-\tilde{\iota}(k-1)}$$
(3)

• Defining an enlarged state variable as $x(k) = [\Delta \tilde{\iota}(k) \ y(k)]^T$, equations (2) and (3) are combined together to form

$$\underbrace{\begin{bmatrix} \Delta \tilde{\iota}(k+1) \\ y(k+1) \end{bmatrix}}_{x(k+1)} = \underbrace{\begin{bmatrix} \bar{A}_d & 0_{n \times m} \\ \bar{C}_d \bar{A}_d & I_{m \times m} \end{bmatrix}}_{\tilde{A}} \underbrace{\begin{bmatrix} \Delta \tilde{\iota}(k) \\ y(k) \end{bmatrix}}_{x(k)} + \underbrace{\begin{bmatrix} \bar{B}_d \\ \bar{C}_d \bar{B}_d \end{bmatrix}}_{\tilde{B}} \Delta \tilde{\iota}(k)$$
(4)

The enlarged plant can then be written as

$$x(k+1) = \widetilde{A}x(k) + \widetilde{B}\Delta\widetilde{u}(k),$$
(5)

$$y(k) = \widetilde{C}x(k), \tag{6}$$

where,
$$\widetilde{C} = \begin{bmatrix} 0_{m \times n} & I_{m \times m} \end{bmatrix}$$

 Future feedback control increments (Δũ^{*}_{k|N}) are obtained by minimizing the quadratic performance index while satisfying the input constraints, i.e.,

$$\Delta \tilde{u}_{k|N}^* = \arg \min_{\Delta \tilde{u}_{k|N}} \left\{ \Delta \tilde{u}_{k|N}^T H \Delta \tilde{u}_{k|N} + 2x^T (k) f^T \Delta \tilde{u}_{k|N} \right\}$$
(7)
subject to $\mathcal{A} \Delta \tilde{u}_{k|N} \leq b_k$ (8)

- (7)-(8) define a standard Quadratic Programming (QP) problem.
- A *receding horizon strategy* is used and only the first control increment $\Delta \tilde{u}^*(k)$ in the calculated $\Delta \tilde{u}^*_{k|N}$ is used for control.
- Optimal feedback control action becomes

$$\tilde{u}(k) = \Delta \tilde{u}^*(k) + \tilde{u}(k-1).$$
(9)

Closed-Loop Integral MPC Simulation Study in MATLAB

• The target state trajectory $\iota_r(\rho, t)$ is generated through an open-loop TRANSP simulation with the following constant reference inputs.

$n_{e}(m^{-3})$	$5.0 imes 10^{19}$	
P ₁ (W)	0.2×10^{6}	
P ₂ (W)	0.4×10^{6}	
P ₃ (W)	$0.6 imes 10^6$	

P ₄ (W)	$0.8 imes 10^6$
P ₅ (W)	$1.0 imes 10^{6}$
P ₆ (W)	1.2×10^{6}
I _p (A)	$0.7 imes 10^{6}$

- The prediction horizon is set to N = 5 to guarantee closed-loop stability.
- The initial condition perturbation rejection capability is tested by setting

$$\iota(t_0) = \iota_r(t_0) + \delta\iota \tag{10}$$

 The controller is also tested against constant input disturbances starting from t = 2.5 s. i.e.,

$$\tilde{u}(k) = \begin{cases} \Delta \tilde{u}^*(k) + \tilde{u}(k-1), & t < 2.5 \,\text{s.} \\ \Delta \tilde{u}^*(k) + \tilde{u}(k-1) + u_d, & t \ge 2.5 \,\text{s.} \end{cases}$$
(11)

where $u_d = 0.15u_r$ stands for the constant disturbance inputs.

Results of the Closed-Loop Integral MPC Simulation Study



Upper Figures: (left)Time evolution of the optimal plasma current, (center) time evolution of the optimal n_e regulation, and (right) time evolution of the optimal neutral beam injection powers.

Lower Figures: Time evolution of the optimal outputs (solid) with their respective targets (dashed).

Conclusion and Future Work

- An NSTX-U-tailored plasma response model is obtained by combining the MDE with simplified models for various plasma variables.
- A constrained MPC algorithm is formulated based on the reduced-order, LTI model to regulate the rotational transform (*ι*-profile).
- An integrator is added to the MPC formulation to achieve offset-free tracking against modeling uncertainties and external disturbances.
- The proposed MPC control scheme is tested via closed-loop numerical simulations based on the control-oriented MDE solver.
- First MPC design for NSTX-U for current density profile control.
 - explicitly handles input and state constraints
 - predicts plasma future state in real time based on current plasma state
 - may be crutial in achieving current profile control + MHD instability avoidance

Future work includes:

- **Refinement** of the FPD control-oriented model using actual experimental data once NSTX-U achieves relevant plasma scenarios.
- Implementation of MPC algorithm in TRANSP's Expert routine and PCS.
- TRANSP closed-loop simulations \Rightarrow Experimental testing in NSTX-U.