MODEL-BASED OPTIMAL CURRENT PROFILE CONTROL DURING THE RAMP-UP PHASE

Y. Ou, C. Xu, E. Schuster

Laboratory for Control of Complex Physical Systems

Mechanical Engineering and Mechanics





Workshop on Improved MHD Control Configurations Columbia University, November 18-20, 2007 Collaborators involved in this project:

- John Ferron, Tim Luce, MikeWalker, DaveHumphreys General Atomics
- Tom Casper, Bill Mayer
 - Lawrence Livermore National Laboratory

Current Profile Control: Motivation

- A key goal in control of an AT discharge is to maintain safety factor (q) and pressure profiles that are compatible with both MHD stability at high toroidal beta and a high fraction of the self-generated bootstrap current in order to realize a stable, highly efficient, steady-state fusion reactor.
- In present experimental tokamaks, the development of current profile controllers is aimed at saving long trial-and-error periods of time currently spent by fusion experimentalists trying to manually adjust the time evolutions of the actuators to achieve a desired current profile.
- The high dimensionality of the problem and the strong coupling between the different variables describing the current profile evolution of the plasma call for a model-based, multivariable approach to obtain improved closed-loop performance.



Current Profile Control: Objective



During "Phase I" the control goal is to drive the current profile from any arbitrary initial condition to a prescribed target profile at some time $T \in (T_1, T_2)$ in the flat-top phase of the total current I(t) evolution. The prescribed target profile is not an equilibrium profile during "Phase I."



Outline

- Models for Control Design
- Open-loop Optimal Control
- Closed-loop Optimal Control



Current Profile Control: Model

V

Magnetic diffusion equation: $\frac{\partial \psi}{\partial t} = \frac{\eta(T_e)}{\mu_o \rho_b^2 \hat{F}^2} \frac{1}{\hat{\rho}} \frac{\partial}{\partial \hat{\rho}} \left(\hat{\rho} \hat{F} \hat{G} \hat{H} \frac{\partial \psi}{\partial \hat{\rho}} \right) - R_o \hat{H} \eta(T_e) \underbrace{\begin{pmatrix} j_{NI} \cdot B \\ B_{\phi,o} \end{pmatrix}}_{B_{\phi,o}}$

Boundary conditions:



4.5



Workshop on Improved MHD Control Configurations, Columbia Univ., Nov. 18-20, 2007

F

poloidal magnetic flux

Current Profile Control: Model

Model for Control Design

GOAL: Develop a model based controller to be used in achieving desirable current profiles during the plasma current ramp-up. A necessary prior task is the development of a dynamic model to use for controller design.

Modeling Alternatives:

HIGH

•The magnetic diffusion equation is accompanied by transport equations for the density and the temperature .

• The magnetic diffusion equation is accompanied by simplified, scenariooriented, models for the density and temperature.

• The magnetic diffusion equation is evaluated with real-time measurements of the density and the temperature.

Model reduction (PDE \rightarrow ODE) may be necessary. Particularly for closed-loop control.

Workshop on Improved MHD Control Configurations, Columbia Univ., Nov. 18-20, 2007

Current Profile Control: Model



Density:

$$n(\hat{\rho},t) = n^{profile}(\hat{\rho})u_{n}(t)$$

$$\bar{n}(t) = \int_{D}^{0} n(\hat{\rho},t)d\hat{\rho}$$
Temperature:

$$T_{e}(\hat{\rho},t) = k_{T_{e}}T_{e}^{profile}(\hat{\rho})\frac{I(t)\sqrt{P_{tot}(t)}}{\bar{n}(t)}$$

$$\eta(\hat{\rho},t) = \frac{k_{eff}Z_{eff}}{T_{e}^{3/2}(\hat{\rho},t)}$$
Parallel Current:

$$\frac{\langle \bar{j}_{NI} \quad \bar{B} \rangle}{B_{\varphi,o}} = k_{NI_{par}}j_{NI_{par}}^{profile}(\hat{\rho})\frac{I(t)^{1/2}P_{tot}(t)^{5/4}}{\bar{n}(t)^{3/2}}$$

We consider I(t), $\overline{n}(t)$ and $P_{tot}(t)$ the physical actuators of the system.



Workshop on Improved MHD Control Configurations, Columbia Univ., Nov. 18-20, 2007

-- Te

0.4

0.6

0.8

J_{NIpar}

Current Profile Control: Model Validation

The simplified model has been compared with experiment shot #129412.



Current Profile Control: Model Validation



The initial validating results show qualitative agreement between simplified model and experiment. More validation experiments will be carried out during the 2008 campaign.

Outline

- Models for Control Design
- Open-loop Optimal Control
- Closed-loop Optimal Control



Open-loop Optimal Control

GOAL: During "Phase I" an optimal control problem must be solved, where time evolution for three actuators $(I(t), \overline{n}(t) (u_n(t)), P_{tot}(t))$ are sought to minimize the functional.





The physical ranges for I(t), $\overline{n}(t)$ and $P_{tot}(t)$, are given by

$$\begin{cases} 0 \le I(t) \le I_{\max} & I(MA) \le \frac{\overline{n}(t)}{10^{19}} \le 5I(MA) \\ \left| \frac{dI(t)}{dt} \right| \le dI_{\max} & 0 \le P_{tot}(t) \le P_{\max} \end{cases}$$

To accurately reproduce experimental discharges, we must add constraints for I(t) and $\overline{n}(t)$, at the initial time of "Phase I", i.e., $\begin{cases} I(t=0s)=I_0\\ \overline{n}(t=0s)=\overline{n}_0 \end{cases}$

In addition, a value of the total current I(t) is prescribed for the flattop phase, i.e.,

$$I(t \ge T_1) = I_{t \arg et}$$





J = quadratic error between ι profile and the prescribed target profile ι_{des}

In each iteration of extremum seeking procedure, θ is used to construct the time evolution of the three physical actuators, I(t), $\overline{n}(t)$ and $P_{tot}(t)$.

• The vector parameter θ has 10 components

$$\theta = \begin{cases} I(0.4s), I(0.8s), \\ P_{tot}(0s), P_{tot}(0.4s), P_{tot}(0.8s), P_{tot}(1.2s), \\ \overline{n}(0.3s), \overline{n}(0.6s), \overline{n}(0.9s), \overline{n}(1.2s) \end{cases}$$

By taking into account that $I(0s) = I_0$ and $I(T_1) = I_{target}$, and using curve fitting for the points I(0s), I(0.4s), I(0.8s), I(1.2s) we can reconstruct the profile for I(t) for $t \in [0, T_1]$. In addition, we make $I(t) = I_{target}$ for $t \in [T_1, T_2]$.

• Following similar procedure, we can construct the law for $P_{tot}(t)$.

By considering that $\overline{n}(0s) = \overline{n}_0$, and using linear interpolation, we can define the law for $\overline{n}(t)$.

• The reconstructed control laws are in turn fed into the PDE model. Given initial ψ , the PDE system is integrated to obtain $\psi(\hat{\rho},t)$, and finally $\iota(\hat{\rho},t)$, which are necessary to evaluate the cost function



In this simulation, we consider the time interval [0, 2.4s]. The initial poloidal flux ψ is shown in Figure (a) and the target *t* profile is shown in Figure (b). The current I(t), average density n(t) and total power $P_{tot}(t)$ are reconstructed as previous description.



The besting matching for *i* profile

Normalized cost function evolution





Experiment July 6, 2007



Conclusions:

-Redefine constraintsfor actuators- Continue effort onmodel validation





What is the potential of ES?

Simple implementation for complex or unknown plants

Integration of CORSICA into MATLAB SIMULINK environment



Outline

- Models for Control Design
- Open-loop Optimal Control
- Closed-loop Optimal Control



Closed-loop Optimal Control

GOAL: Identical to open-loop control. During "Phase I" an optimal control problem must be solved, where time evolution for three actuators ($I(t), \overline{n}(t) (u_n(t)), P_{tot}(t)$) are sought to minimize the functional. Closed-loop control is expected to be more effective in dealing with model and IC uncertainties, and measurement noise.





Workshop on Improved MHD Control Configurations, Columbia Univ., Nov. 18-20, 2007



Closed-loop control scheme

This figure shows a closed-loop, receding-horizon, optimal controller based on an extremum-seeking optimization framework.



- 1. Select the tolerance $\varepsilon > 0$ and the maximum number of iterations for the extremum seeking control algorithm.
- 2. Define $t_i = t_0$. Provide the off-line actuator trajectories u(t), for $t > t_i = t_0$, and the actual initial poloidal flux profile $\psi(t_0)$ to the PDE model.
- 3. Compute the predicted $\iota(T)$ (control target) from the output sequence $\psi(t)_i$, for $t > t_i$, obtained from the PDE model.
- 4. Calculate the cost function. If it is less than ε_i , go to step 6.
- 5. Adjust the parameters θ (or u(t)) of the extremum seeking algorithm, until the cost function is less than ε or the maximum number of iteration is reached.
- 6. Implement the calculated actuator trajectories on the actual system for $[t_i + \Delta t, t_i + 2\Delta t]$.
- 7. Move the control horizon one sampling interval Δt ahead, measure the output of the actual system $\psi(t_i + \Delta t)$, make $t_i = t_i + \Delta t_i$, and go to step 3.





In this simulation, we consider the time interval [0, 2.4s]. The initial poloidal flux ψ is shown in Figure (a) and the target *i* profile is shown in Figure (b). The current I(t), average density n(t) and total power $P_{tot}(t)$ are reconstructed as previous description.



Computed best matching in open-loop control with nominal initial profile.







• This is some preliminary work that does not take into account the dynamics of the actuators.





Figure (a) shows the disturbed initial poloidal flux profile, and compares it with the nominal initial poloidal flux profile. Figure (b) shows the difference between the obtained ι profile and the desirable ι profile. As expected, the matching is worsen due to the disturbance in the initial poloidal flux profile.



Figure (a) shows the disturbed initial poloidal flux profile. Figure (b) shows the difference between the obtained t profile and the desirable t profile. The closed-loop approach provides a better matching than the open-loop control.





• This is some preliminary work that does not take into account the dynamics of the actuators.

Limitation: Computational demand



Workshop on Improved MHD Control Configurations, Columbia Univ., Nov. 18-20, 2007

Conclusions and Future Work

- We have demonstrated the existence of model-based solutions to the problem of defining actuator trajectories that achieve desired current profiles during the plasma current ramp.
- Simplified model validation
 - Absolutely necessary for closed-loop control
 - The simpler the model, the faster the convergence for open-loop control
- Incorporation of predictive codes (CORSICA, CRONOS) for model validation, controller validation, and controller design.
 - CORSICA in the loop for open-loop control design (ES, NLP)
- Reduced order modeling
 - Necessary for closed-loop control (Bilinear Opt. Cont., Receding Hor. Cont.)
 - Useful for open-loop control design (ES, NLP)
- Incorporation of transport equations for density and temperature
 - Exploit time-scale separation \rightarrow Algebraic-Differential Equations
- Magnetic diffusion equation with density and temperature measurements
 - This defines a different control problem (probably more difficult)
- Extension to Phase II of the discharge (flat-top)

Workshop on Improved MHD Control Configurations, Columbia Univ., Nov. 18-20, 2007