

# Integrated Robust Control of the Global Toroidal Rotation and Total Plasma Energy in Tokamaks

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**Abstract**—Integrated-control solutions will play a significant role in future tokamaks, in which a variety of coupled control problems will need to be solved simultaneously by means of a limited number of actuators. In this article, the problem of simultaneously regulating the global toroidal rotation,  $\Omega_\phi$ , and total plasma energy,  $W$ , is tackled. These two 0-D variables,  $\Omega_\phi$  and  $W$ , depend on the ion toroidal rotation and electron temperature profiles, respectively. Both  $\Omega_\phi$  and  $W$  also depend on the electron density and safety factor profiles. The actuation methods considered in this article are co-current and counter-current neutral beam injection. A nonlinear, robust controller that makes use of Lyapunov redesign techniques is synthesized based on 0-D, control-oriented models of the  $\Omega_\phi$  and  $W$  dynamics. In addition, an actuator management scheme is designed to handle variations in the control priorities and availability of the neutral beam injectors. The actuator manager solves an optimization problem in real time in order to find the most appropriate course of action when unexpected changes occur. The integrated control architecture is tested for a DIII-D scenario by means of the 1-D code Control-Oriented Transport Simulator (COTSIM), which predicts the time evolution of the electron temperature, electron density, ion toroidal rotation, and safety factor profiles.

**Index Terms**—Integrated control, nonlinear robust control, tokamaks.

## I. INTRODUCTION

IN TOKAMAK plasmas, the thermal energy per volume unit,  $E$ , and the ion toroidal angular velocity,  $\omega_\phi$ , are profiles of substantial interest [1]. The first one,  $E$ , is closely related to the ion and electron density and temperature profiles and, therefore, to the pressure, plasma  $\beta$ , and fusion triple product. All these variables characterize the plasma performance and are key for tokamaks to become economically competitive means of producing energy. The second one,  $\omega_\phi$ , is related to the ions toroidal momentum and, therefore, to their density and temperature as well. In addition, the triggering of numerous magnetohydrodynamic (MHD) instabilities is related to  $E$  and  $\omega_\phi$ , such as neoclassical

tearing modes (NTMs) or resistive wall modes (RWMs), which may substantially decrease the plasma performance and/or terminate the plasma discharge.

As a result, active control of  $E$  and  $\omega_\phi$  is a problem of relevance in current nuclear-fusion research. Although rotation control may be limited in large tokamaks, such as ITER, due to their higher volume and inertia, it may be a relevant problem for compact fusion reactors, which will play a significant role in the future nuclear-fusion research. In any case, this control problem offers significant challenges. First, modeling the  $E$  and  $\omega_\phi$  dynamics for control design is not an easy task. Although complex models providing a significant degree of accuracy might be available, using such models for control design may be impractical or just impossible. Instead, employing reduced, control-oriented models may make the controller synthesis possible. Moreover, the number of available actuators is limited, and the tokamak's actuation capability may not allow for controlling whole 1-D profiles (such as  $E$  or  $\omega_\phi$ ). Alternatively, it can be more realistic to control 0-D variables related to 1-D variables. For example, the stored thermal energy, denoted as  $W$ , and the bulk toroidal rotation, denoted as  $\Omega_\phi$ , are 0-D variables related to  $E$  and  $\omega_\phi$ , respectively.

Previous work on simultaneous control of  $W$  and  $\Omega_\phi$  can be found in [2]. In such work, a linear, data-driven model was employed, and a controller was synthesized using linear techniques. Previous work on controlling  $\omega_\phi$  (at least at one point) simultaneously with  $W$  can be found in [3] and [4] for DIII-D and in [5] for NSTX-U. Linear robust techniques are utilized in [3] and [4], whereas real-time optimization is used in [3], [4] after approximate linearization of the plasma dynamics.

In this article, a nonlinear, 0-D coupled model of the  $\Omega_\phi - W$  dynamics is employed to design a multi-input multi-output (MIMO) nonlinear controller. Model uncertainties are included to account for the unknown and/or unmodeled dynamics neglected in the 0-D modeling process. A nominal stabilizing controller is designed by using nonlinear design techniques and later robustified by using Lyapunov redesign techniques, ensuring stability and performance despite model uncertainties. Regulation of  $\Omega_\phi - W$  is carried out by means of neutral beam injection (NBI). The fact that this actuator affects both  $W$  and  $\Omega_\phi$  (by means of the heating power and torque injected), together with the coupled  $\Omega_\phi - W$  dynamics, suggest that MIMO controllers will have improved performance compared with other approaches. No other auxiliary sources are considered (e.g., radio frequency antennas,

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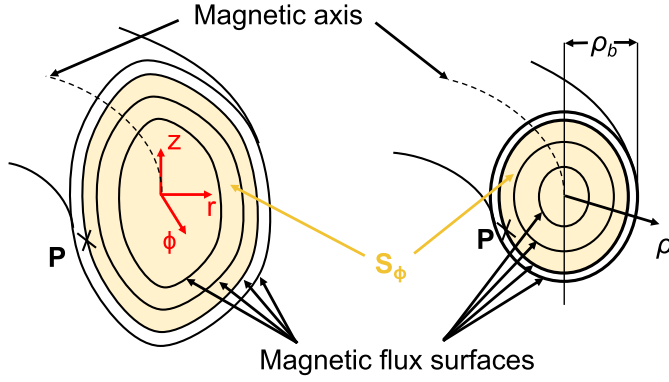


Fig. 1. Magnetic configuration in a tokamak under ideal MHD conditions.

such as electron-cyclotron (EC) heating) since assuming the availability of a larger number of actuators for this type of control problem, which would indeed make the control problem decoupled and rather trivial, is simply unrealistic in present and future tokamaks where these additional actuators need to be reserved for different control goals. In addition, future tokamaks will face changes in the plasma conditions, control objectives, and actuator availability that will require the implementation of supervisory systems with actuator management capabilities. For this reason, actuator management is added to the control scheme in order to handle possible variations in the control priorities and NBI availability (see our previous work [6]).

This article is organized as follows. The nonlinear model for the  $\Omega_\phi$  and  $W$  evolutions is described in Section II. The control design is presented in Section III. A simulation study to illustrate the performance of the 0-D controller in 1-D simulations for a DIII-D scenario is included in Section IV. Finally, a brief summary and possible future work are presented in Section V.

## II. GLOBAL TOROIDAL ROTATION AND TOTAL PLASMA ENERGY MODEL

Under ideal MHD conditions, the magnetic-flux surfaces in a tokamak form toroidally nested surfaces around the magnetic axis (see Fig. 1). A magnetic-flux surface is defined by points with the same value of the toroidal flux,  $\Phi$ , mean effective minor radius,  $\rho$ , poloidal magnetic flux,  $\Psi$ , pressure  $p$ , or others [1]. Only one of these functions (known as flux functions) is needed to index the magnetic-flux surfaces. This, together with the assumption of toroidal symmetry, reduces the 3-D problem in space ( $r$ - $z$ - $\phi$  coordinates) to a 1-D problem. In this model,  $\hat{\rho} = \rho/\rho_b$  is used as the flux function to label the magnetic-flux surfaces (where  $\rho \triangleq (\Phi/(\pi B_{\phi,0}))^{1/2}$ ,  $B_{\phi,0}$  is the toroidal magnetic field on axis, and  $\rho_b$  is the value of  $\rho$  at the last closed magnetic-flux surface, see Fig. 1).

The pressure field,  $p$ , is given by

$$p = n_e K T_e + n_i K T_i \quad (1)$$

where  $n_e$  and  $n_i$  are the electron and ion densities, respectively,  $T_e$  and  $T_i$  are the electron and ion temperatures, respectively, and  $K$  is Boltzmann's constant. It is assumed that the plasma

is purely hydrogenic; therefore, the quasi-neutrality condition can be written as  $n_e \approx n_i$ , and (1) becomes  $p = n_e K (T_e + T_i)$ . The thermal energy density,  $E$ , is given by

$$E = \frac{3}{2} n_e K (T_e + T_i) = \frac{3}{2} p \quad (2)$$

so  $E$  is a constant in a magnetic-flux surface as well.

Although it has been shown that  $E$  is a flux function, it cannot be shown in general that the ion toroidal velocity,  $v_{\phi,i}$ , or  $\omega_\phi$  are flux functions as well. The ion toroidal angular velocity,  $\omega_\phi$ , is taken as  $\omega_\phi = \langle v_{\phi,i} \rangle / R_0$ , where  $R_0$  is the tokamak's major radius (which is considered constant) and  $\langle \cdot \rangle$  denotes the flux-surface average.

### A. Thermal Stored Energy

The thermal stored energy,  $W$ , is the thermal energy contained within the last-closed magnetic-flux surface, and it is defined as

$$W(t) \triangleq \int_{V_p} E dV = \int_{\hat{\rho}=0}^{\hat{\rho}=1} E(\hat{\rho}, t) \frac{\partial V(\hat{\rho})}{\partial \hat{\rho}} d\hat{\rho} \quad (3)$$

where  $V_p$  is the plasma region enclosed within the last magnetic-flux surface and  $V(\hat{\rho})$  is the plasma volume enclosed by the magnetic-flux surface at  $\hat{\rho}$ .

The model proposed in this article for the dynamics of  $W$  is based on a 0-D power balance in the plasma

$$\frac{dW}{dt} = -\frac{W}{\tau_E} + P_{\text{tot}} \quad (4)$$

where  $P_{\text{tot}} = \sum_{i=1}^{N_{\text{NBI}}} P_{\text{NBI},i} + P_{\text{EC}}$  is the total injected power,  $P_{\text{NBI},i}$  is the power corresponding to the  $i$ -th NBI,  $N_{\text{NBI}}$  is the total number of NBIs,  $P_{\text{EC}}$  is the total EC power, and  $\tau_E$  is the energy-confinement time, which is modeled by the IPB98(y,2) scaling [7]

$$\tau_E = H_H k P_{\text{tot}}^{-0.69}, \quad (5)$$

$$k = 0.0562 I_p^{0.93} B_T^{0.15} R_0^{1.97} \kappa^{0.78} \epsilon^{0.58} M^{0.19} \bar{n}_{e,19}^{-0.41} \quad (6)$$

where  $H_H$  is the so-called confinement factor,  $I_p$  is the total plasma current which must be given in MA,  $P_{\text{tot}}$  must be given in MW,  $B_T$  is the toroidal magnetic field,  $\kappa$  is the plasma elongation,  $\epsilon = a/R_0$  is the inverse aspect ratio, where  $a$  is the tokamak minor radius,  $M$  is the effective mass of the plasma in a.m.u., and  $\bar{n}_{e,19}$  is the line-average electron density in  $10^{19} \text{ m}^{-3}$ . The confinement factor  $H_H$  is modeled as an uncertain parameter given by

$$H_H = H_H^{\text{nom}} + \delta_{H_H} \quad (7)$$

where  $H_H^{\text{nom}}$  is a constant, the nominal value of  $H_H$  which is known, and  $\delta_{H_H}$  is an uncertain term. In addition, an uncertain source/sink of power,  $\delta_P$ , is added to the right-hand side of (4) to account for neglected dynamics such as inefficient heat deposition, unexpected radiative losses, or others. Equation (4) can be rewritten as

$$\frac{dW}{dt} = -\frac{W}{H_H^{\text{nom}} k P_{\text{tot}}^{-0.69}} + P_{\text{tot}} + \delta_W \quad (8)$$

where  $\delta_W = W/(k P_{\text{tot}}^{-0.69}) [(1/H_H^{\text{nom}}) - (1/H_H^{\text{nom}} + \delta_{H_H})] + \delta_P$  is a term that bundles all the uncertain terms of the  $W$  subsystem.

### B. Global Toroidal Rotation

The bulk toroidal rotation,  $\Omega_\phi$ , is the figure of merit used in this model to characterize the toroidal plasma rotation. It is defined as the average toroidal rotation of the ions inside the plasma, and it is given by

$$\Omega_\phi(t) \triangleq \frac{1}{N_p(t)} \int_{V_p} \langle n_e(\hat{\rho}, t) \rangle \omega_\phi(\hat{\rho}, t) dV \quad (9)$$

where  $N_p$  is the total number of ions within the plasma, which is approximately calculated as

$$N_p = \bar{n}_e V(\rho_b) \quad (10)$$

where  $\bar{n}_e$  is the line-average electron density, and  $V(\rho_b)$  is the total plasma volume.

The model proposed in this article for the dynamics of  $\Omega_\phi$  is simplified but based on first principles. In the same way that a particle of mass  $m$  and position vector  $\vec{r}$  would have an angular momentum given by  $\vec{L} = \vec{r} \times m(d\vec{r}/dt)$ , the plasma is assumed to have an angular momentum  $\Gamma$  in the  $z$ -direction given by  $\Gamma = m_p R_0^2 \Omega_\phi$ , where  $m_p$  is the plasma mass contained within the last closed magnetic-flux surface, which neglecting the electrons contribution can be estimated as  $m_p = m_D N_p / N_{Av} = m_D \bar{n}_e V(\rho_b) / N_{Av}$ , where  $m_D$  is the molar mass of the plasma and  $N_{Av}$  is Avogadro's number. In this control-oriented model, the plasma is regarded as a particle with the same mass as the total plasma mass, which is rotating around the  $z$ -axis with angular velocity  $\vec{\Omega}_\phi = \Omega_\phi \vec{z}$ , and at a distance  $R_0$ . The angular momentum varies in time due to the external torque sources as

$$\frac{d\Gamma}{dt} = R_0^2 \frac{d(m_p \Omega_\phi)}{dt} = \sum_{i=1}^{N_{NBI}} T_{NBI,i} + T_{int} \quad (11)$$

where  $T_{NBI,i}$  is the torque injected by the  $i$ -th NBI, and  $T_{int}$  is a plasma intrinsic-torque source. The model could be easily extended to include other torque sources, such as nonresonant magnetic fields. Expanding the time derivative in (11) and rearranging terms, it is possible to write

$$\frac{d\Omega_\phi}{dt} = -\Omega_\phi \frac{1}{m_p} \frac{dm_p}{dt} + \sum_{i=1}^{N_{NBI}} \frac{T_{NBI,i}}{m_p R_0^2} + \frac{T_{int}}{m_p R_0^2}. \quad (12)$$

The first term,  $-\Omega_\phi(1/m_p)(dm_p/dt)$ , models the decrease of toroidal rotation by diffusion. By defining a global rotation confinement time,  $\tau_{\Omega_\phi}$ , as  $\tau_{\Omega_\phi} \triangleq m_p/(dm_p/dt)$ , the  $\Omega_\phi$  dynamics (12) can be rewritten as

$$\frac{d\Omega_\phi}{dt} = -\frac{\Omega_\phi}{\tau_{\Omega_\phi}} + \sum_{i=1}^{N_{NBI}} \frac{T_{NBI,i}}{m_p R_0^2} + \frac{T_{int}}{m_p R_0^2}. \quad (13)$$

Control-oriented models for  $\tau_{\Omega_\phi}$ ,  $T_{NBI,i}$ , and  $T_{int}$  are employed. First,  $\tau_{\Omega_\phi}$  is taken as  $\tau_{\Omega_\phi} = k_\Omega \tau_E$ , where  $k_\Omega > 0$  is a dimensionless parameter, which is modeled as a constant. Second,  $T_{NBI,i}$  is taken as

$$T_{NBI,i} = k_{NBI,i} P_{NBI,i} \quad (14)$$

where  $k_{NBI,i}$  are modeled as constant parameters. Finally, the intrinsic rotation is modeled as a torque source [8],

$T_{int} = k_{int} W / I_p$ , where  $k_{int}$  is taken as a constant parameter in this modeling work.

In addition to the uncertainty found in  $\tau_\Omega$  because of  $H_H$ , an unknown source/sink of torque  $\delta_T$  is added on the right-hand side of (13) to account for neglected dynamics, such as inefficient torque injection, uncertainty in the model parameters ( $k_{NBI,i}$ ,  $k_{int}$ , and so on), inaccurate measurements of the state/inputs, or others. Equation (13) can be rewritten as

$$\frac{d\Omega_\phi}{dt} = -\frac{\Omega_\phi}{k_\Omega H_H^{\text{nom}} k P_{\text{tot}}^{-0.69}} + \sum_{i=1}^{N_{NBI}} \frac{k_{NBI,i} P_{NBI,i}}{m_p R_0^2} + k_{int} \frac{W}{I_p m_p R_0^2} + \delta_{\Omega_\phi} \quad (15)$$

where  $\delta_{\Omega_\phi} = \Omega_\phi / (k_\Omega k P_{\text{tot}}^{-0.69}) [1/H_H^{\text{nom}} - (1/H_H^{\text{nom}}) + \delta_{H_H}] + \delta_T$  is a term that bundles all the uncertain terms of the  $\Omega_\phi$  subsystem.

### C. Summary: State-Space Model

The state-space model for the  $\Omega_\phi + W$  system is given by

$$\begin{bmatrix} \dot{\Omega}_\phi \\ \dot{W} \end{bmatrix} = \begin{bmatrix} -\frac{\Omega_\phi}{\tau_{\Omega_\phi}^{\text{nom}}} + \sum_i \frac{k_{NBI,i} P_{NBI,i}}{m_p R_0^2} + k_{int} \frac{W}{I_p m_p R_0^2} + \delta_{\Omega_\phi} \\ -\frac{W}{\tau_E^{\text{nom}}} + \sum_i P_{NBI,i} + P_{EC} + \delta_W \end{bmatrix}$$

or simply as

$$\dot{x} = \begin{bmatrix} f_{\Omega_\phi}(x, u, t, \delta) \\ f_W(x, u, t, \delta) \end{bmatrix} \quad (16)$$

where  $x \triangleq [\Omega_\phi, W]^T$  is the state vector,  $u \triangleq [P_{NBI,1}, \dots, P_{NBI,N_{NBI}}]^T$  is the input vector,  $\delta \triangleq [\delta_{H_H}, \delta_P, \delta_T]^T$  is the uncertainties vector, and  $\tau_E^{\text{nom}}$  and  $\tau_{\Omega_\phi}^{\text{nom}}$  are the values of  $\tau_E$  and  $\tau_{\Omega_\phi}$  with  $H_H = H_H^{\text{nom}}$ . The explicit dependence with  $t$  in the state equation (16) is due to the time-dependent inputs to the model that are assumed to be noncontrollable in this article, i.e.,  $P_{EC}$ ,  $I_p$ , and  $\bar{n}_e$ .

## III. CONTROL DESIGN

### A. Nominal Control Law via Lyapunov Theory

A nominal control law (i.e., with  $\delta = 0$ ) denoted by  $u^{\text{nom}}$  is designed with the purpose of simultaneously regulating  $\Omega_\phi$  and  $W$ . The control objective is to drive  $x$  to a target value,  $\bar{x} = [\bar{\Omega}_\phi, \bar{W}]^T$ , where  $\bar{\Omega}_\phi$  and  $\bar{W}$  are the rotation and energy targets, respectively. A control law  $u^{\text{nom}} = [P_{NBI,1}^{\text{nom}}, \dots, P_{NBI,N_{NBI}}^{\text{nom}}]^T$  that stabilizes the nominal system can be obtained by solving the following system of two nonlinear equations:

$$\begin{aligned} f_{\Omega_\phi}(x, u^{\text{nom}}, t, 0) &= -\frac{\bar{\Omega}_\phi + \tilde{\Omega}_\phi}{k_\Omega H_H^{\text{nom}} k (\sum_i P_{NBI,i}^{\text{nom}} + P_{EC})^{-0.69}} \\ &\quad + \sum_i \frac{k_{NBI,i} P_{NBI,i}^{\text{nom}}}{m_p R_0^2} + k_{int} \frac{\bar{W} + \tilde{W}}{I_p m_p R_0^2} \\ &= -k_{P,\Omega} \tilde{\Omega}_\phi + \frac{d\tilde{\Omega}_\phi}{dt} \end{aligned} \quad (17)$$

$$f_W(x, u^{\text{nom}}, t, 0) = -\frac{\bar{W} + \tilde{W}}{H_H^{\text{nom}} k (\sum_i P_{\text{NBI},i}^{\text{nom}} + P_{EC})^{-0.69}} + \sum_i P_{\text{NBI},i}^{\text{nom}} + P_{EC} = -k_{P,W} \tilde{W} + \frac{d\tilde{W}}{dt} \quad (18)$$

where  $k_{P,\Omega} > 0$  and  $k_{P,W} > 0$  are constants determined during the controller design,  $\tilde{\Omega}_\phi = \Omega_\phi - \bar{\Omega}_\phi$  is the error variable associated with  $\Omega_\phi$ , and  $\tilde{W} = W - \bar{W}$  is the error variable associated with  $W$ . The solution of (17) and (18) provides two constraints for  $N_{\text{NBI}}$  unknowns.

Under the control law defined by the solution of (17) and (18), the nominal dynamics is reduced to

$$\frac{d\tilde{x}}{dt} = -\begin{bmatrix} k_{P,\Omega} & 0 \\ 0 & k_{P,W} \end{bmatrix} \tilde{x} \quad (19)$$

where  $\tilde{x} = [\tilde{\Omega}_\phi, \tilde{W}]^T$ . Using the Lyapunov function  $V = 1/2 \tilde{x}^T \tilde{x} = 1/2 (\tilde{\Omega}_\phi^2 + \tilde{W}^2)$  for the reduced nominal system (19), it is found that

$$\dot{V} = -\tilde{x}^T \begin{bmatrix} k_{P,\Omega} & 0 \\ 0 & k_{P,W} \end{bmatrix} \tilde{x} = -k_{P,\Omega} \tilde{\Omega}_\phi^2 - k_{P,W} \tilde{W}^2. \quad (20)$$

Therefore, the evolution of the nominal system is exponentially stable [9].

### B. Robust Control Law via Lyapunov Redesign

In order to ensure robust stability for (16) under the presence of uncertainties ( $\delta \neq 0$ ), an extra term  $u^{\text{rob}}$  is added to  $u^{\text{nom}}$  so that  $u = u^{\text{nom}} + u^{\text{rob}}$  (i.e.,  $P_{\text{NBI},i} = P_{\text{NBI},i}^{\text{nom}} + P_{\text{NBI},i}^{\text{rob}}$ ), where  $u^{\text{rob}}$  is obtained by means of Lyapunov redesign techniques [9]. First, it is convenient to rewrite (16) as

$$\begin{bmatrix} \dot{\tilde{\Omega}}_\phi \\ \dot{\tilde{W}} \end{bmatrix} = \begin{bmatrix} -\frac{d\bar{\Omega}_\phi}{dt} + \frac{k_{\text{int}} W}{I_p m_p R_0^2} \\ -\frac{d\bar{W}}{dt} + P_{EC} \end{bmatrix} + \begin{bmatrix} -\frac{\Omega_\phi}{\tau_{\Omega_\phi}^{\text{nom}}} + \sum_i \frac{k_{\text{NBI},i} P_{\text{NBI},i}}{m_p R_0^2} \\ -\frac{W}{\tau_E^{\text{nom}}} + \sum_i P_{\text{NBI},i} \end{bmatrix} + \begin{bmatrix} \delta_{\Omega_\phi} \\ \delta_W \end{bmatrix} = F(\tilde{x}, t) + \hat{u} + \hat{\delta}(\tilde{x}, u, t) \quad (21)$$

where the matrix  $F \in \mathbb{R}^{2 \times 1}$  lumps the terms that do not depend on  $u$  or  $\delta$ ,  $\hat{u} \triangleq [-\Omega_\phi / \tau_{\Omega_\phi}^{\text{nom}} + \sum_i k_{\text{NBI},i} P_{\text{NBI},i} / (m_p R_0^2), -W / \tau_E^{\text{nom}} + \sum_i P_{\text{NBI},i}]^T$ , and  $\hat{\delta} \triangleq [\delta_{\Omega_\phi}, \delta_W]^T$  lumps the uncertain terms. Taking (21) and the same Lyapunov function  $V$  as before, it is found that

$$\begin{aligned} \dot{V} &= \tilde{x}^T F + \tilde{x}^T \hat{u}^{\text{nom}} + \tilde{x}^T \hat{u}^{\text{rob}} + \tilde{x}^T \hat{\delta} \\ &= -\tilde{x}^T \begin{bmatrix} k_{P,\Omega} & 0 \\ 0 & k_{P,W} \end{bmatrix} \tilde{x} + \tilde{x}^T \hat{u}^{\text{rob}} + \tilde{x}^T \hat{\delta} \end{aligned} \quad (22)$$

where  $\tilde{x}^T (F + \hat{u}^{\text{nom}}) = -\tilde{x}^T \begin{bmatrix} k_{P,\Omega} & 0 \\ 0 & k_{P,W} \end{bmatrix} \tilde{x}$  has been employed,  $\hat{u}^{\text{nom}}$  is the value of  $\hat{u}$  with  $P_{\text{NBI},i} = P_{\text{NBI},i}^{\text{nom}}$ , and  $\hat{u}^{\text{rob}}$  is the value of  $\hat{u}$  with  $P_{\text{NBI},i} = P_{\text{NBI},i}^{\text{rob}}$ . Taking

$\hat{u}^{\text{rob}} = -\eta(\tilde{x}) \tilde{x} / \|\tilde{x}\|_2$ , for some scalar function  $\eta(\tilde{x}) > 0$ , (22) becomes

$$\begin{aligned} \dot{V} &= -\tilde{x}^T \begin{bmatrix} k_{P,\Omega} & 0 \\ 0 & k_{P,W} \end{bmatrix} \tilde{x} - \eta(\tilde{x}) \|\tilde{x}\|_2 + \tilde{x}^T \hat{\delta}, \\ &\leq -\tilde{x}^T \begin{bmatrix} k_{P,\Omega} & 0 \\ 0 & k_{P,W} \end{bmatrix} \tilde{x} - \eta(\tilde{x}) \|\tilde{x}\|_2 + \|\tilde{x}\|_2 \|\hat{\delta}\|_2 \end{aligned} \quad (23)$$

and setting  $\eta(\tilde{x}) \geq \|\hat{\delta}\|_2 > 0$ , it is found that

$$\dot{V} \leq -\tilde{x}^T \begin{bmatrix} k_{P,\Omega} & 0 \\ 0 & k_{P,W} \end{bmatrix} \tilde{x} \quad (24)$$

which ensures that the uncertain system remains exponentially stable [9]. Although  $\hat{\delta}$  is not exactly known, it is assumed that an absolute bound to  $\|\hat{\delta}\|_2$ , denoted by  $\hat{\delta}_{\text{max}}$ , can be estimated. Therefore, taking  $\eta = \hat{\delta}_{\text{max}}$ , a robust control law given by  $u^{\text{rob}} = -\hat{\delta}_{\text{max}} \tilde{x} / \|\tilde{x}\|_2$  would ensure the exponential stability of the system for all values of  $\hat{\delta}$ . To make the control law continuous at  $\tilde{x} = 0$ , using a similar process as showed in [9], the control law is modified as

$$u^{\text{rob}} = -\hat{\delta}_{\text{max}} \frac{\tilde{x}}{\|\tilde{x}\|_2}, \quad \text{if } \hat{\delta}_{\text{max}} \|\tilde{x}\|_2 \geq \epsilon_R \quad (25)$$

$$u^{\text{rob}} = -\hat{\delta}_{\text{max}}^2 \frac{\tilde{x}}{\epsilon_R}, \quad \text{if } \hat{\delta}_{\text{max}} \|\tilde{x}\|_2 < \epsilon_R \quad (26)$$

where  $\epsilon_R > 0$  is a design parameter. The control law (25) and (26) does not ensure that the system remains exponentially stable, but it does ensure that  $\|\tilde{x}\|_2$  is bounded by a class  $\mathcal{K}$  function<sup>1</sup> of  $\epsilon_R$  as  $t \rightarrow \infty$ . Therefore, it is convenient that  $\epsilon_R$  is chosen small.

### C. Actuator Management via Optimization

It can be noted that (17) and (18) together with (25) and (26) is a nonlinear system of four equations with 2  $N_{\text{NBI}}$  unknowns,  $P_{\text{NBI},i}^{\text{nom}}$ , and  $P_{\text{NBI},i}^{\text{rob}}$ , for  $i = 1, \dots, N_{\text{NBI}}$ , which are the components of  $u^{\text{nom}}$  and  $u^{\text{rob}}$ . The solution of the system<sup>2</sup> depends on  $N_{\text{NBI}}$ . If  $N_{\text{NBI}} = 1$ , the system is overconstrained, indicating that, in general, it is not possible to control both  $\Omega_\phi$  and  $W$  simultaneously. If  $N_{\text{NBI}} = 2$ , there are the same number of unknowns as equations, and it is possible to regulate both  $\Omega_\phi$  and  $W$  at the same time as long as the NBI configuration allows to do so. Finally, if  $N_{\text{NBI}} > 2$ , the system is underconstrained, and extra conditions may need to be imposed in order to univocally determine  $P_{\text{NBI},i}$ .

A way of imposing extra conditions is to write the problem as an optimization problem. It is chosen that, at every time step, an objective function of the norm of  $\tilde{u} \triangleq u - u_{\text{FF}}$  is minimized ( $u_{\text{FF}}$  is a reference or feedforward input), i.e., the difference between  $u$  and  $u_{\text{FF}}$  is taken as small as possible. In addition, the constraints from the control laws (17)–(18), (25)–(26), and the physical saturation levels are imposed. The components of  $\tilde{u}$  (i.e.,  $\tilde{P}_{\text{NBI},i}$ ) are free variables of the optimization problem, which is given by

<sup>1</sup>A function  $f(x)$  belongs to class  $\mathcal{K}$  iff: 1)  $f(0) = 0$  and 2) it is a strictly increasing function of  $x$ .

<sup>2</sup>The discussion and/or formal demonstration of the existence and/or unicity of solutions for this nonlinear system is left aside in this work. It is assumed that the system has a unique solution if  $N_{\text{NBI}} > 1$ .



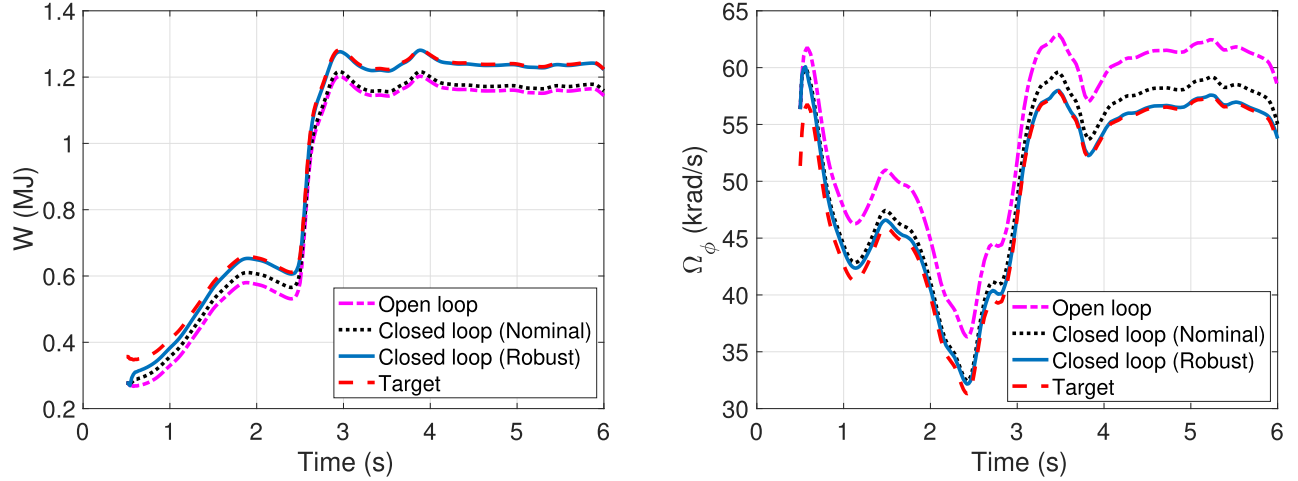


Fig. 2. State evolution in open loop (magenta dashed-dotted line) and closed loop under the nominal (black dotted line) and robust (blue solid line) control laws in 1-D simulations, together with the corresponding targets (red dashed line).

$$\min_{\tilde{u}} J(\tilde{u}) = \min_{\tilde{u}} \tilde{u}^T Q \tilde{u} \quad (27)$$

$$f_{\Omega_\phi}(x, u^{\text{nom}}, t, 0) = -k_{P,\Omega} \tilde{\Omega}_\phi + \frac{d\tilde{\Omega}_\phi}{dt} \quad (28)$$

$$f_W(x, u^{\text{nom}}, t, 0) = -k_{P,W} \tilde{W} + \frac{d\tilde{W}}{dt}, \quad (29)$$

$$-\frac{\tilde{\Omega}_\phi}{\tau_{\Omega_\phi}^{\text{nom}}(P_{\text{NBI},i}^{\text{rob}})} + \sum_i \frac{k_{\text{NBI},i} P_{\text{NBI},i}^{\text{rob}}}{m_p R_0^2} = -\hat{\delta}_{\text{max}} \frac{\tilde{\Omega}_\phi}{\epsilon} \quad (30)$$

$$-\frac{W}{\tau_E^{\text{nom}}(P_{\text{NBI},i}^{\text{rob}})} + \sum_i P_{\text{NBI},i}^{\text{rob}} = -\hat{\delta}_{\text{max}} \frac{\tilde{W}}{\epsilon} \quad (31)$$

$$\tilde{u} \in \tilde{U}, \quad (32)$$

where  $Q \in \mathbb{R}^{N_{\text{NBI}} \times N_{\text{NBI}}} \geq 0$  is a design matrix,  $\epsilon = \|\tilde{\Omega}_\phi, \tilde{W}\|_2$  if  $\hat{\delta}_{\text{max}} \|\tilde{\Omega}_\phi, \tilde{W}\|_2 \geq \epsilon_R$ ,  $\epsilon = \epsilon_R / \hat{\delta}_{\text{max}}$  if  $\hat{\delta}_{\text{max}} \|\tilde{\Omega}_\phi, \tilde{W}\|_2 < \epsilon_R$ , and  $\tilde{U}$  is the set of feasible NBI power deviations. The optimization problem (27)–(32) serves as an actuator manager that determines the inputs that fulfill the control law and saturation constraints and minimize  $J(\tilde{u})$ .

#### IV. SIMULATION STUDY

In this section, the control scheme previously introduced is tested in simulations for a DIII-D scenario. The scenario corresponds to shot 147634, in which eight NBIs are available. These are grouped in three sets of NBIs (i.e.,  $N_{\text{NBI}} = 3$ ) as follows.

- 1) *Group 1*: It consists of four co-current on-axis NBIs. This group's total power,  $P_{\text{NBI},1}$ , is denoted by  $P_{\text{CO-ON}}$ .
- 2) *Group 2*: It consists of two co-current off-axis NBIs. This group's total power,  $P_{\text{NBI},2}$ , is denoted by  $P_{\text{CO-OFF}}$ .
- 3) *Group 3*: It consists of two counter-current NBIs. This group's total power,  $P_{\text{NBI},3}$ , is denoted by  $P_{\text{COUNTER}}$ .

Grouping the NBIs is convenient to reduce the size of (27)–(32) and is necessary to not overestimate the actuation capability in DIII-D. Relevant machine parameters and model constants are  $R_0 = 1.80$  m,  $B_T = 1.65$  T,  $a = 0.60$  m,  $\kappa = 1.7$ ,  $M = 2$  a.m.u.,  $H_H^{\text{nom}} = 1.4$ ,  $k_\Omega = 1$ ,  $k_{\text{int}} = 3$  N·m MA/MJ,  $k_{\text{NBI},i} = 1$  N·m/MW for each cocurrent on-axis NBI,  $k_{\text{NBI},i} = 0.7$  N·m/MW for each cocurrent off-axis NBI, and

$k_{\text{NBI},i} = -1$  N·m/MW for each counter-current NBI. The saturation limits are  $P_{\text{NBI},i} \in [0, 2.5]$  MW.

For simulation testing, the Control-Oriented Transport Simulator (COTSIM) code is employed. COTSIM is a 1-D code for control testing and simulation that evolves  $T_e$  and  $\omega_\phi$  using the electron heat-transport equation (EHTE) and the toroidal rotation equation (TRE), which are given by

$$\frac{\partial (\frac{3}{2} n_e T_e)}{\partial t} = \frac{1}{\rho_b^2 \hat{\rho} \hat{H}} \frac{\partial}{\partial \hat{\rho}} \left( \hat{G} \hat{H}^2 \chi_e n_e \frac{\partial T_e}{\partial \hat{\rho}} \right) + Q_e \quad (33)$$

$$m_i \langle r^2 \rangle \frac{\partial (n_i \omega_\phi)}{\partial t} = \frac{1}{\hat{\rho} \hat{H}} \frac{\partial}{\partial \hat{\rho}} \left( f_\phi \chi_\phi n_i \frac{\partial \omega_\phi}{\partial \hat{\rho}} \right) + t_\omega \quad (34)$$

where  $\chi_e$  and  $\chi_\phi$  are the electron heat and toroidal momentum diffusivities, respectively,  $m_i$  is the ion mass,  $Q_e$  and  $t_\omega$  are the electron-heat and ion-torque deposition, respectively, and  $\hat{F}$ ,  $\hat{G}$ ,  $\hat{H}$ ,  $\langle r^2 \rangle$ , and  $f_\phi$  are fixed profiles corresponding to a particular magnetic configuration. Approximate models for  $n_e$  and  $n_i$  are employed,  $n_e = n_e^{\text{prof}} \bar{n}_e$  and  $n_i = n_i^{\text{prof}} \bar{n}_i$ , where  $n_e^{\text{prof}}$  and  $n_i^{\text{prof}}$  are fixed profiles that characterize the spatial distribution of the electron and ion densities, respectively, and  $\bar{n}_e$  and  $\bar{n}_i$  are the line-average electron and ion densities, respectively. For  $Q_e$  and  $t_\omega$ , COTSIM employs the following control-oriented models for this simulation study,  $Q_e = \sum_i Q_{\text{NBI},i}^{\text{prof}} P_{\text{NBI},i} + Q_{\text{EC}}^{\text{prof}} P_{\text{EC}}$ , and  $t_\omega = \sum_i t_{\text{NBI},i}^{\text{prof}} P_{\text{NBI},i} + t_{\text{int}}$ , where  $Q_{\text{NBI},i}^{\text{prof}}$  and  $Q_{\text{EC}}^{\text{prof}}$  are fixed profiles that characterize the electron-heating spatial deposition for the  $i$ -th NBI and EC, respectively,  $t_{\text{NBI},i}^{\text{prof}}$  is a profile that depends on both  $T_e$  and  $n_i$  and characterizes the torque spatial deposition, and  $t_{\text{int}}$  is the intrinsic torque, which is a function of  $E$  and, therefore, of  $T_e$  as well. Also, Bohm-like models [10] are utilized for  $\chi_e$  and  $\chi_\phi$ , as given by  $\chi_{(\cdot)} = T_e / B_T (a \nabla p_e / p_e)^{\lambda_1} q^{\lambda_2}$ , where  $p_e \triangleq n_e K T_e$  is the electron pressure,  $q$  is the safety factor, a measure of the helicity of the magnetic field lines in a tokamak [1] whose evolution is estimated in COTSIM by means of the magnetic diffusion equation [11], and  $\lambda_1$  and  $\lambda_2$  are constant parameters. Therefore, it can be appreciated that the EHTE and TRE models are coupled by means of  $\chi_e$ ,  $\chi_\phi$ ,  $Q_e$ , and  $t_\omega$ , representing a substantial increase in complexity with respect to the models employed for control synthesis.

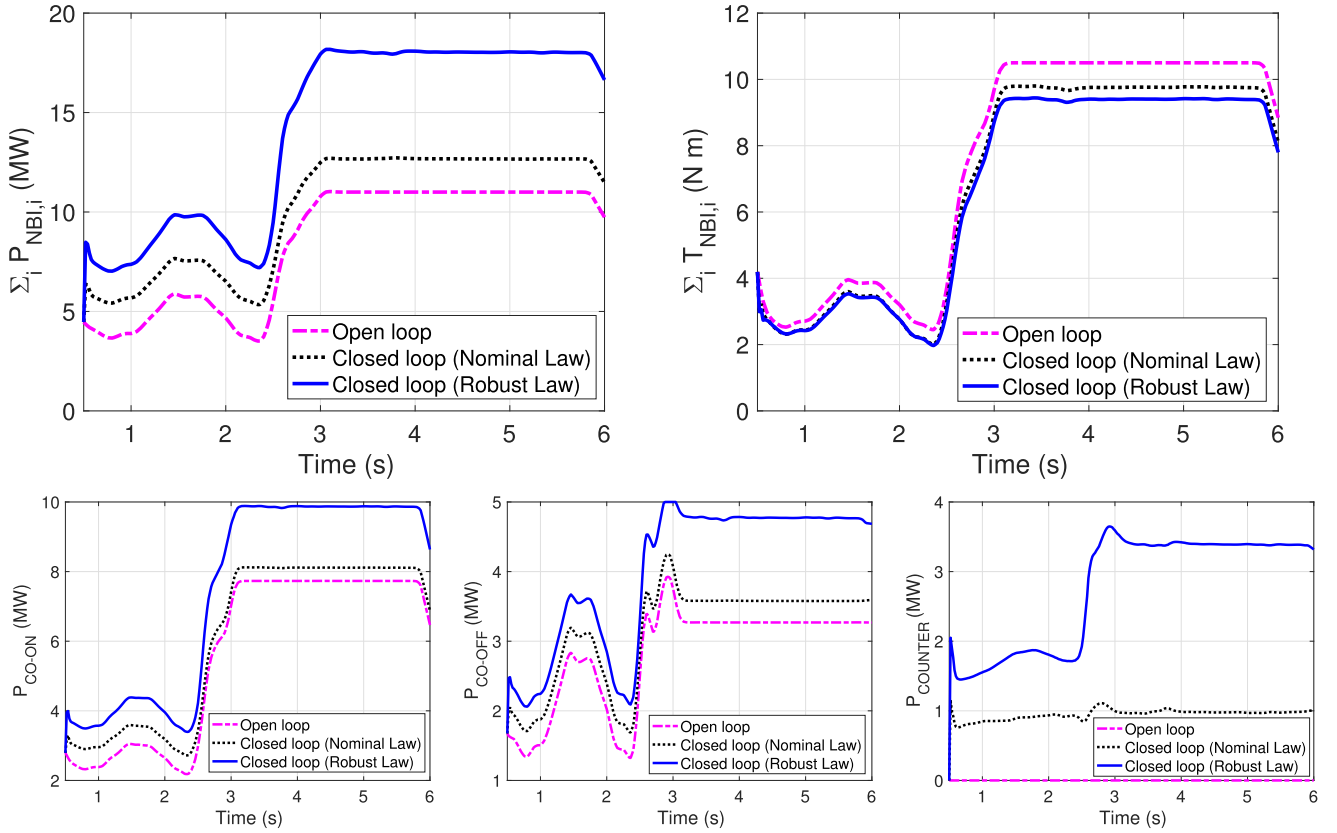


Fig. 3. NBI power ( $P_{\text{NBI}}$ ) and torque ( $T_{\text{NBI}}$ ) evolutions in open loop (magenta dashed-dotted line) and closed loop under the nominal (black dotted line) and robust (blue solid line) control laws, together with ON-axis cocurrent NBIs ( $P_{\text{CO-ON}}$ ), OFF-axis cocurrent NBIs ( $P_{\text{CO-OFF}}$ ), and counter-current NBI's ( $P_{\text{COUNTER}}$ ) in 1-D simulations.

Several simulations are run for this article. First, an open-loop simulation is carried out with the experimental inputs,  $u_{\text{exp}}$ , corresponding to shot 147634. The state evolution obtained in this open-loop simulation is denoted by  $x_{\text{exp}}$ . A target  $\bar{x} = [\bar{\Omega}_\phi, \bar{W}]^T$  is created based on the open-loop simulation. The  $\bar{\Omega}_\phi$  target is given by  $\bar{\Omega}_\phi = (\Omega_\phi)_{\text{exp}} - 5$  krad/s, whereas the  $\bar{W}$  target is given by  $\bar{W} = W_{\text{exp}} + 0.08$  MJ. Second, a closed-loop simulation is run under the nominal control law ( $u = u^{\text{nom}}$ ). Finally, a closed-loop simulation is run under the robust control law ( $u = u^{\text{nom}} + u^{\text{rob}}$ ). To solve (27)–(32),  $u_{\text{FF}} = u_{\text{exp}}$  is taken.

Fig. 2 shows the state evolution,  $x$ , in open loop (i.e.,  $x_{\text{exp}}$ ) and closed loop under the nominal and robust control laws, together with the target  $\bar{x}$ . Fig. 3 shows the time evolution of the total NBI power,  $\sum_i P_{\text{NBI},i}$ , and total NBI torque,  $\sum_i T_{\text{NBI},i}$  (as computed by the 0-D model (14)), in open-loop (i.e., calculated from  $u_{\text{exp}}$ ) and closed-loop simulations. Fig. 3 also shows the time evolution of  $P_{\text{NBI},i}$  grouped, as indicated earlier (co-current on-axis NBI,  $P_{\text{CO-ON}}$ , co-current off-axis NBI,  $P_{\text{CO-OFF}}$ , and counter-current NBI,  $P_{\text{COUNTER}}$ ). Fig. 2 shows that the nominal law drives  $x$  closer to  $\bar{x}$  compared with the open-loop evolution, but  $x$  does not converge to the target  $\bar{x}$  within the simulation time. It can be seen that the robust controller drives  $x$  much closer to the target  $\bar{x}$  (with a small error associated with  $\epsilon_R$ ) than the nominal controller. Fig. 3 shows how  $\sum_i P_{\text{NBI},i}$  is increased by both the nominal and robust controllers to achieve the target  $\bar{W} > W_{\text{exp}}$ . The

robust controller requests a higher  $\sum_i P_{\text{NBI},i}$  than the nominal controller (about 5 MW more), which seems to be the reason for its better performance. Also, as  $\bar{\Omega}_\phi < \Omega_{\text{exp}}$ , the NBI torque is decreased by both control laws. As before, the robust controller is more aggressive and requests a lower NBI torque than the nominal controller. Fig. 3 shows that, when using the robust law, all  $P_{\text{NBI},i}$  are very close to saturation, and  $P_{\text{CO-OFF}}$  even saturates for a small period of time. This happens as a result of a very demanding target  $\bar{x}$ .

## V. CONCLUSION

A robust controller for simultaneous energy and toroidal rotation control in tokamaks has been presented. It is a model-based, 0-D controller synthesized from nonlinear, control-oriented models. The controller is composed of a nominal control law and a robust control law, both designed using Lyapunov techniques. Both the nominal and the robust controllers have been tested in 1-D simulations using the COTSIM code for a particular DIII-D scenario. However, the controller design procedure is independent of the tokamak and/or scenario in question. If the robust control law is activated, the controller shows a better performance compared with the nominal control law working on its own. Future work may include the integration of new scalar variables in the control scheme and experimental testing in the DIII-D tokamak.

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