
ROBUST CONTROL OF RESISTIVE WALL MODES IN TOKAMAK PLASMAS

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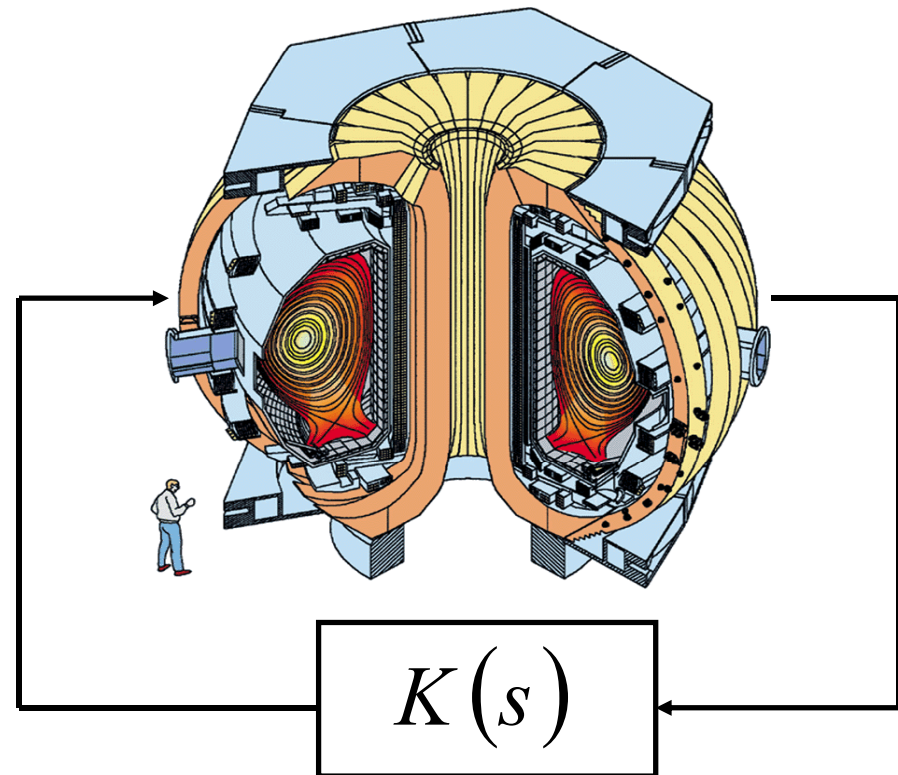
Laboratory for Control of
Complex Physical Systems

Mechanical Engineering
and Mechanics



LEHIGH
UNIVERSITY

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Collaborators involved in this project:

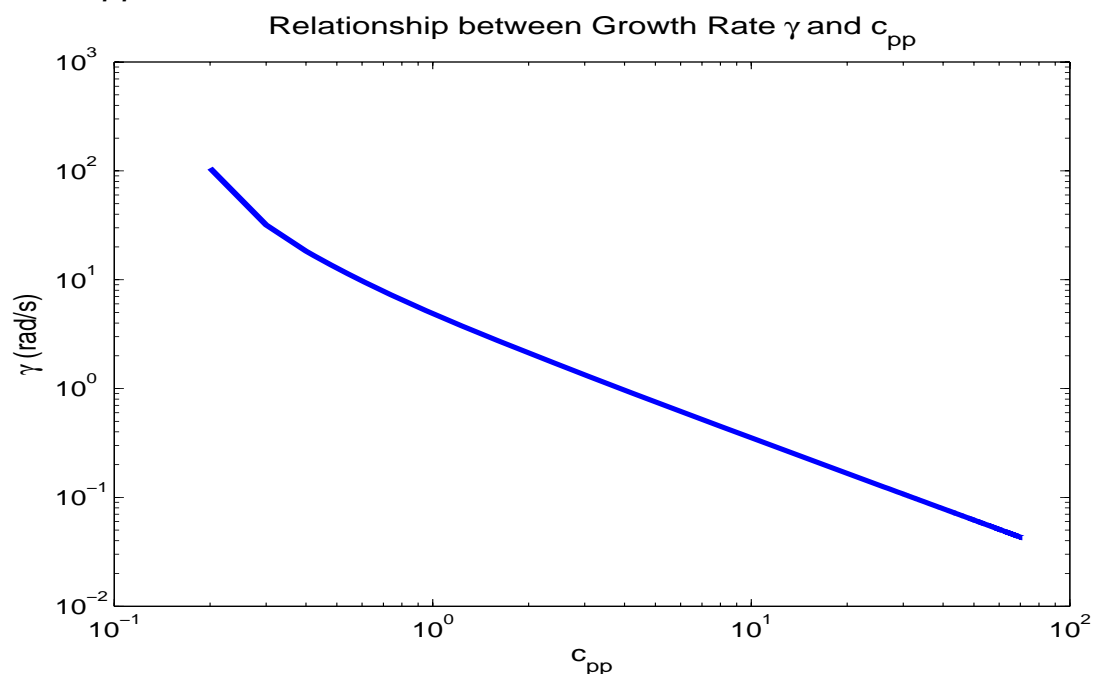
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Introduction

- Resistive Wall Mode (RWM) is a plasma kink instability whose growth rate is moderated by the influence of a resistive wall.
- FAR-TECH DIII-D/RWM model
- Plasma surface modeled as a toroidal current sheet and the wall modeled with an eigenmode approach.
- State-space model of the plant, whose states are the surrounding wall current and the external control coil currents.

Introduction

- 22 poloidal field probes and saddle loops.
- 12 in-vessel coils are used to oppose the deformation.
- State space model is parameterized with a scalar coupling coefficient c_{pp} , which is directly related to the growth rate γ



- 12 input, 22 output reduced to 3 input, 2 output using a typical quartet configuration for the I-coils and matched filter on the field probes.

System Model

The state matrices are given below, where each matrix has a physical correlation to a parameter in DIII-D which is well known, except for the uncertain parameter c_{pp} .

$$\dot{x} = Ax + Bu$$

$$y = Cx$$

$$A = \left(M_{ss} - M_{sp} c_{pp} M_{ps} \right)^{-1} R_{ss}$$

$$C = C_{ss} - C_{yp} c_{pp} M_{ps}$$

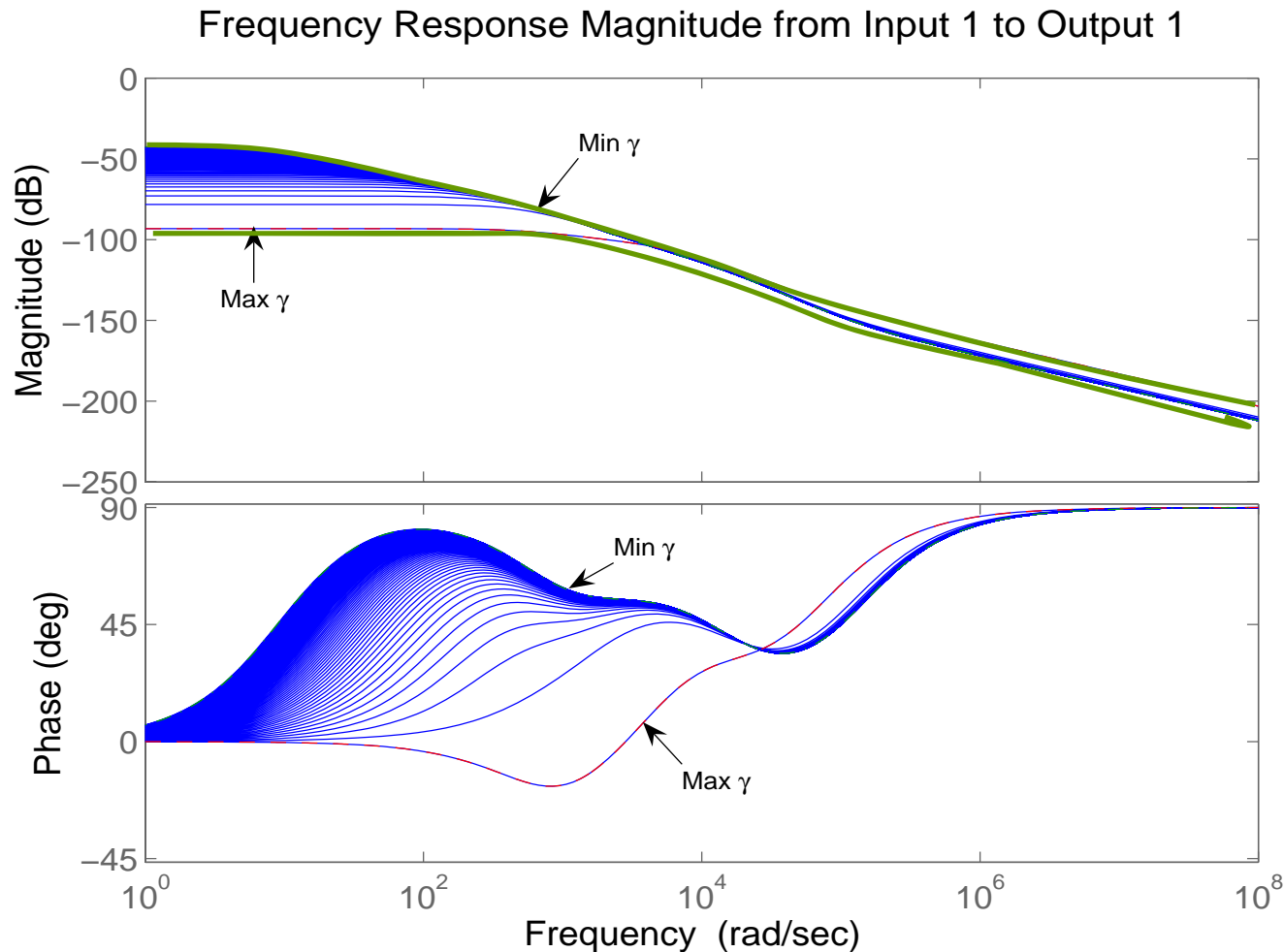
$$B = \left(M_{ss} - M_{sp} c_{pp} M_{ps} \right)^{-1}$$

The transfer function representation of the state matrices

$$\frac{Y(s)}{U(s)} = G(s) = D + C(sI - A)^{-1} B = G(s, c_{pp})$$

How does the dynamic response change as c_{pp} changes?

Frequency Response for Varying c_{pp}



We are interested in designing one controller that stabilizes all the systems → ROBUST CONTROL

Design Goal

- To design a model-based controller that stabilizes the system and meets performance criteria over a large range of growth rate values.

| Condition | Target Value | Maximum Constraint |
|---------------|--------------|--------------------|
| Rise Time | 1.0 ms | 5.0 ms |
| Settling Time | 5.0 ms | 10 ms |
| Overshoot | 15 % | 50 % |
| Input Voltage | N/A | ± 100 V |

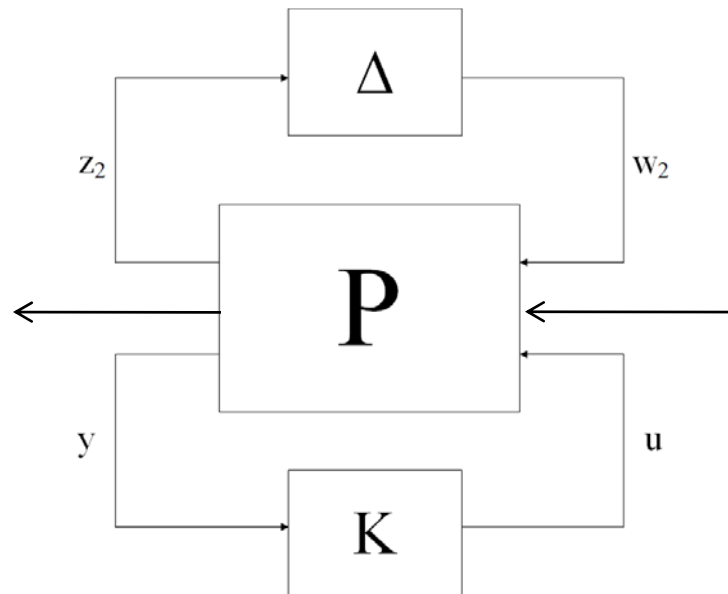
- Physical growth rate (γ) range: 10-5,000 rad/sec
- Corresponding c_{pp} range: 0.3325-71
- The growth rate relationship through c_{pp} is treated as an uncertain parameter that acts as a disturbance to a nominal system.
- Robust control tools are applied to the model to stabilize the system over the design range.
- Results in a single controller that ensures stability and performance within the desired growth rate range.

Controller Design

The goal is to design a feedback controller K that robustly stabilize the system for the applicable range of Δ . Defining

$$N = F_l(P, K) \quad \mu(N_{11}) = \frac{1}{\min\{k_m \mid \det(I - k_m N_{11} \Delta) = 0\}} \quad \bar{\sigma}(\Delta) \leq 1$$

where μ is the structured singular value and the term N_{11} isolates the uncertainty from the input and output of the system.



Assuming N and Δ are stable robust stability is given by

$$\mu(N_{11}(j\omega)) < 1, \forall \omega$$

and robust performance is given by

$$\mu(N(j\omega)) < 1, \forall \omega$$

System Space Model

The state matrices are given below, where each matrix has a physical correlation to a parameter in DIII-D which is well known, except for the uncertain parameter c_{pp} .

$$\dot{x} = Ax + Bu$$

$$y = Cx$$

$$A = \left(M_{ss} - M_{sp} c_{pp} M_{ps} \right)^{-1} R_{ss}$$

$$C = C_{ss} - C_{yp} c_{pp} M_{ps}$$

$$B = \left(M_{ss} - M_{sp} c_{pp} M_{ps} \right)^{-1}$$

Using the Sherman-Morrison formula

$$\left(A_T - b_T C_T D_T \right)^{-1} = A_T^{-1} + \frac{b_T \left(A_T^{-1} C_T \right) \left(D_T A_T^{-1} \right)}{1 - b_T D_T A_T^{-1} C_T}$$

The state matrices can be rewritten as

$$A = A_0 + \sum_{i=1}^4 \alpha_i A_i \quad B = B_0 + \sum_{i=1}^4 \alpha_i B_i \quad C = C_0 + \alpha_5 C_5$$

where each α term is a nonlinear function of the uncertain parameter c_{pp} and every other term is constant.

Linear Fractional Transformation

Using the linear fractional transformation, which is defined by the upper and lower transform for a matrix M as:

$$F_l(M, \Delta_l) = M_{11} + M_{12}\Delta_l(I - M_{22}\Delta_l)^{-1}M_{21} \quad F_u(M, \Delta_u) = M_{22} + M_{21}\Delta_u(I - M_{11}\Delta_u)^{-1}M_{12}$$

The transfer function representation of the state matrices

$$G(s) = D + C(sI - A)^{-1}B$$

$$M = \begin{bmatrix} M_{11} & M_{12} \\ M_{21} & M_{22} \end{bmatrix}$$

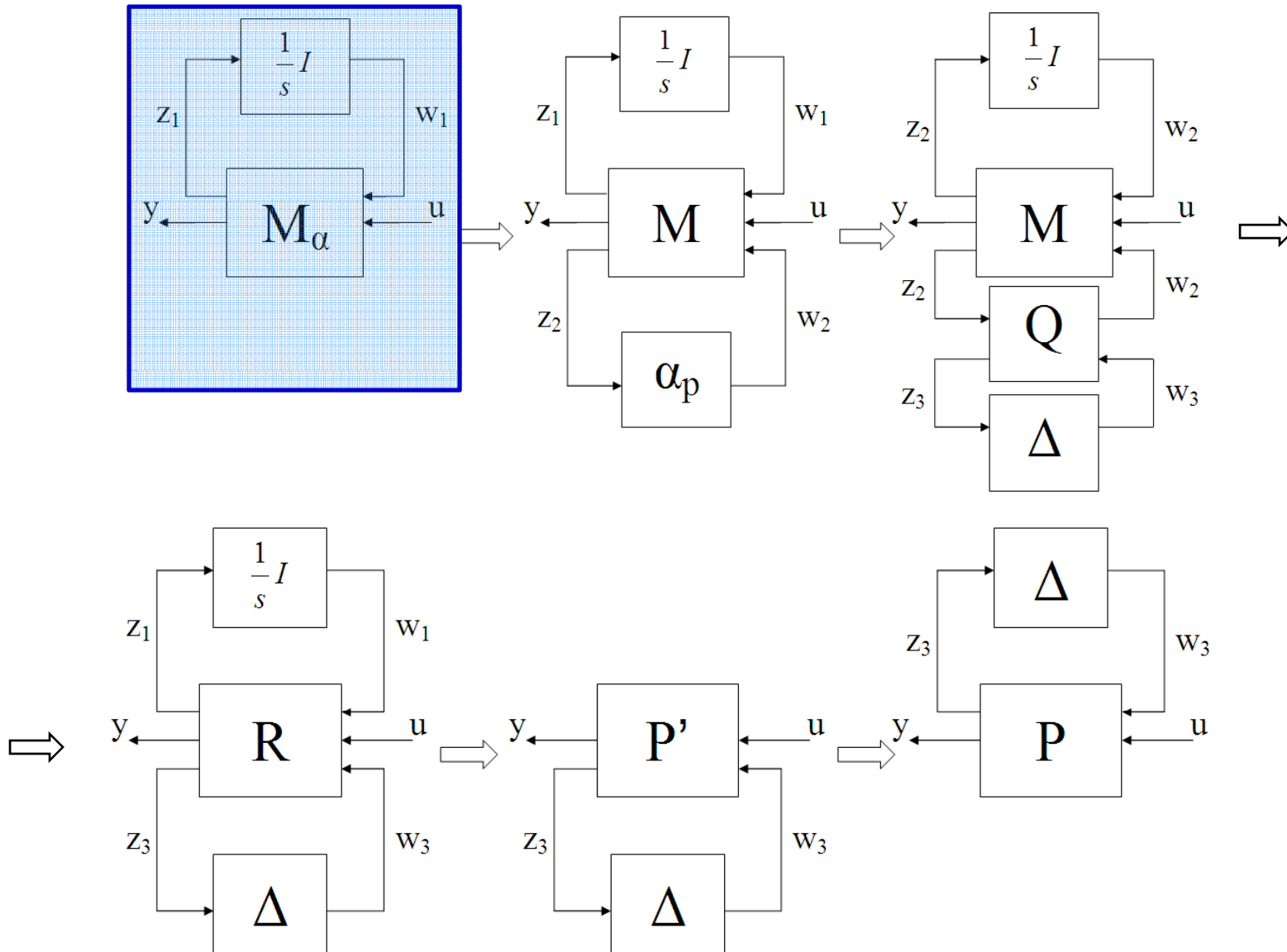
can be written as

$$G(s) = F_u\left(M_\alpha, \frac{1}{s}I\right) = M_{\alpha_{22}} + M_{\alpha_{21}} \frac{1}{s} \left(I - M_{\alpha_{11}} \frac{1}{s}\right)^{-1} M_{\alpha_{12}} = M_{\alpha_{22}} + M_{\alpha_{21}} (sI - M_{\alpha_{11}})^{-1} M_{\alpha_{12}}$$

where

$$M_\alpha = \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} A_0 + \sum_{i=1}^4 \alpha_i A_i & B_0 + \sum_{i=1}^4 \alpha_i B_i \\ C_0 + \alpha_5 C_5 & 0 \end{bmatrix}$$

Model Progression



Singular Value Decomposition

The uncertainty α can be formulated into a linear fractional transform by achieving the smallest possible repeated blocks.

$$J_i = \begin{bmatrix} A_i & B_i \\ C_i & D_i \end{bmatrix} = U_i \Sigma_i V_i^* = (U_i \sqrt{\Sigma_i}) (\sqrt{\Sigma_i} V_i^*) = \begin{bmatrix} L_i \\ W_i \end{bmatrix} \begin{bmatrix} R_i \\ Z_i \end{bmatrix}^*$$

Denoting q_i as the rank of each matrix J_i , introducing the uncertainty

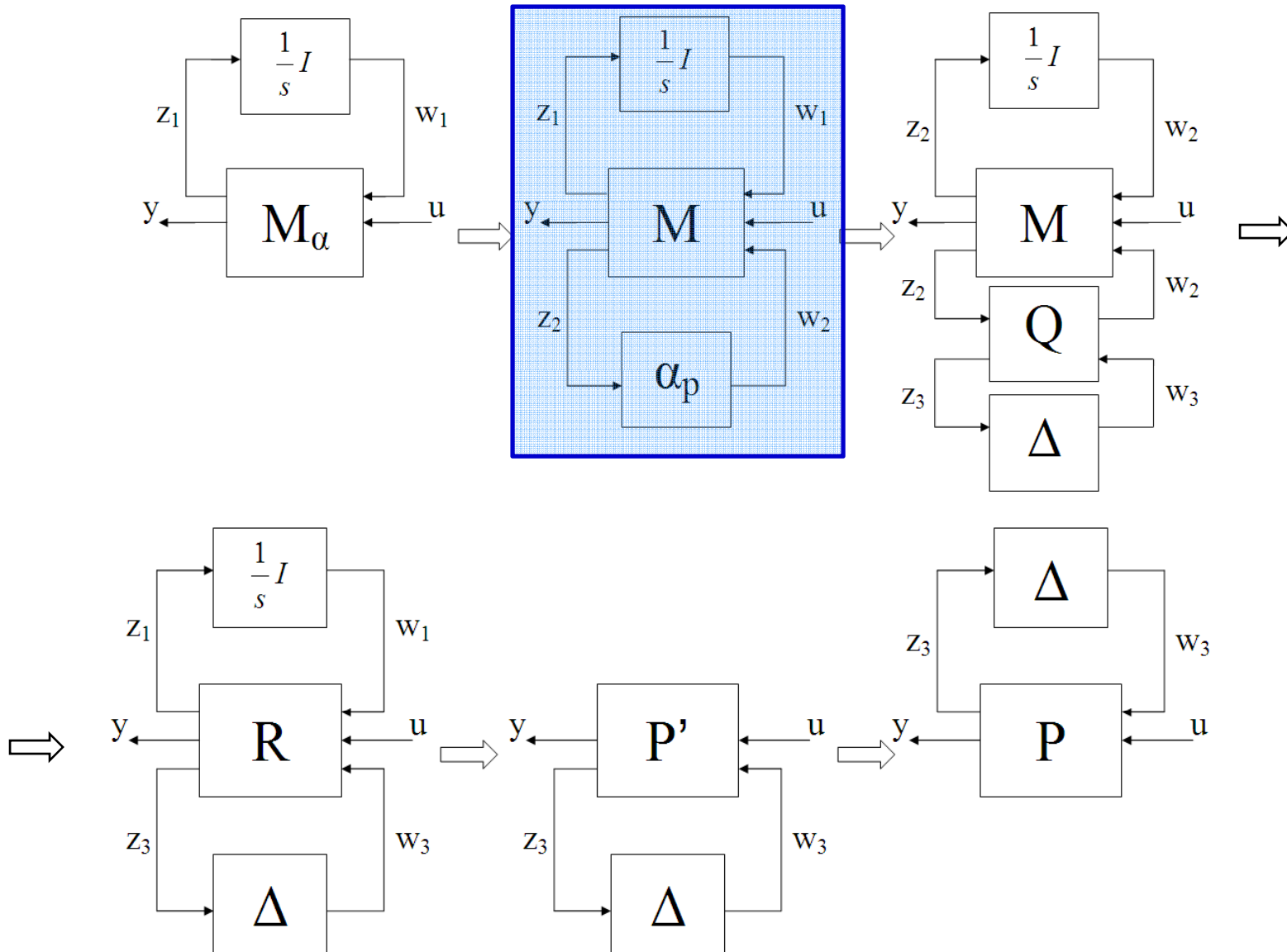
$$\alpha_i J_i = \begin{bmatrix} L_i \\ W_i \end{bmatrix} \left[\alpha_i I_{q_i} \begin{bmatrix} R_i \\ Z_i \end{bmatrix} \right]^*$$

$$M_\alpha = M_{11} + M_{12} \alpha_p M_{21} = F_l(M, \alpha_p)$$

$$M_{11} = \begin{bmatrix} A_0 & B_0 \\ C_0 & 0 \end{bmatrix} \quad M_{12} = \begin{bmatrix} L_1 & \cdots & L_5 \\ W_1 & \cdots & W_5 \end{bmatrix} \quad M_{21} = \begin{bmatrix} R_1^* & Z_1^* \\ \vdots & \vdots \\ R_5^* & Z_5^* \end{bmatrix} \quad \alpha_p = \begin{bmatrix} \alpha_1 I_{q_1} & & 0 \\ & \ddots & \\ 0 & & \alpha_5 I_{q_5} \end{bmatrix}$$

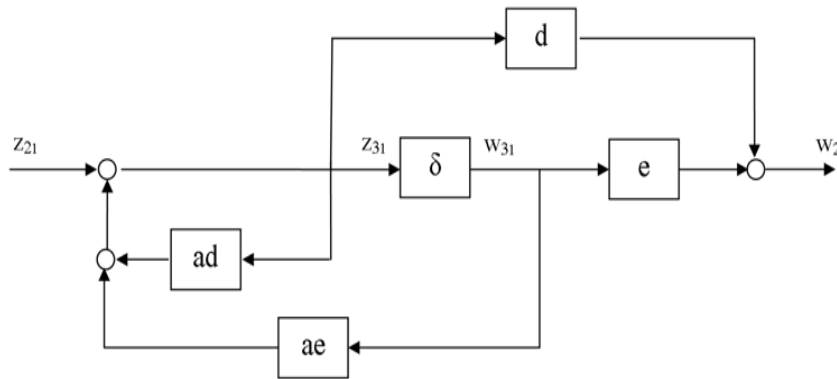
$$G(s) = F_u \left(M_\alpha, \frac{1}{s} I \right) = F_u \left(F_l(M_\alpha, \alpha_p), \frac{1}{s} I \right)$$

Model Progression



Pulling out the “ δ ”

The uncertainty α_p is still a nonlinear matrix based on the uncertain parameter c_{pp} . This parameter is extracted using by “pulling out the δ ”, where δ is the normalized c_{pp} .



$$\alpha_1 = \frac{d + \delta e}{1 - a(d + \delta e)}$$

$$\begin{bmatrix} w_{2_1} \\ z_{3_1} \end{bmatrix} = Q_1 \begin{bmatrix} z_{2_1} \\ w_{3_1} \end{bmatrix}$$

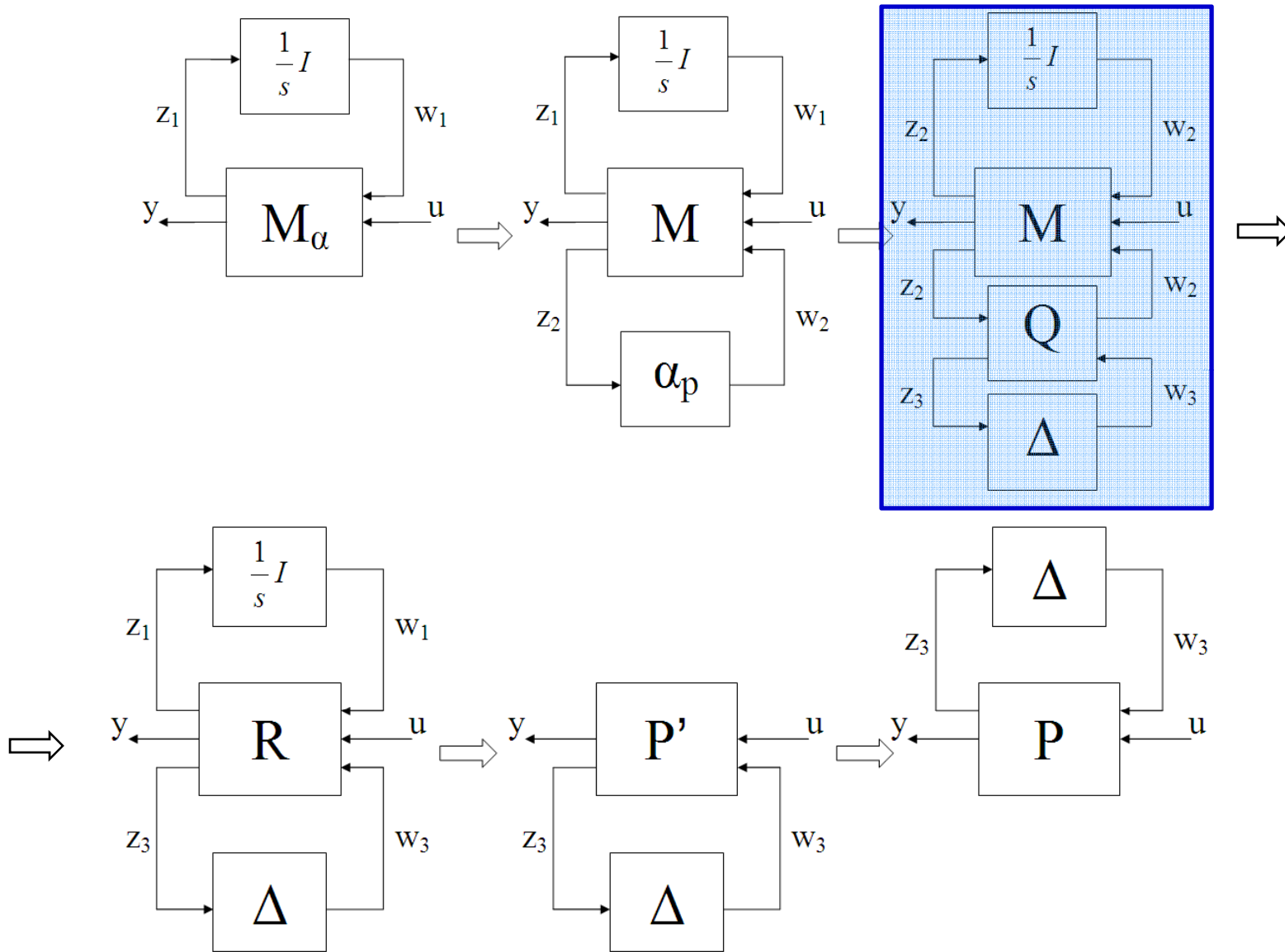
$$Q_1 = \begin{bmatrix} \frac{d}{1-ad} & e \left(1 + \frac{ad}{1-ad} \right) \\ \frac{1}{1-ad} & \frac{ae}{1-ad} \end{bmatrix}$$

$$\alpha_1 = F_l(Q_1, \delta)$$

The final result is a linear fractional transform

$$\alpha_p = F_l(Q, \Delta) \quad \Delta = \delta I \quad |\delta| \leq 1$$

Model Progression



Pulling out the “ δ ”

Finally this can be substituted into the plant and simplified using several properties of the linear fractional transform

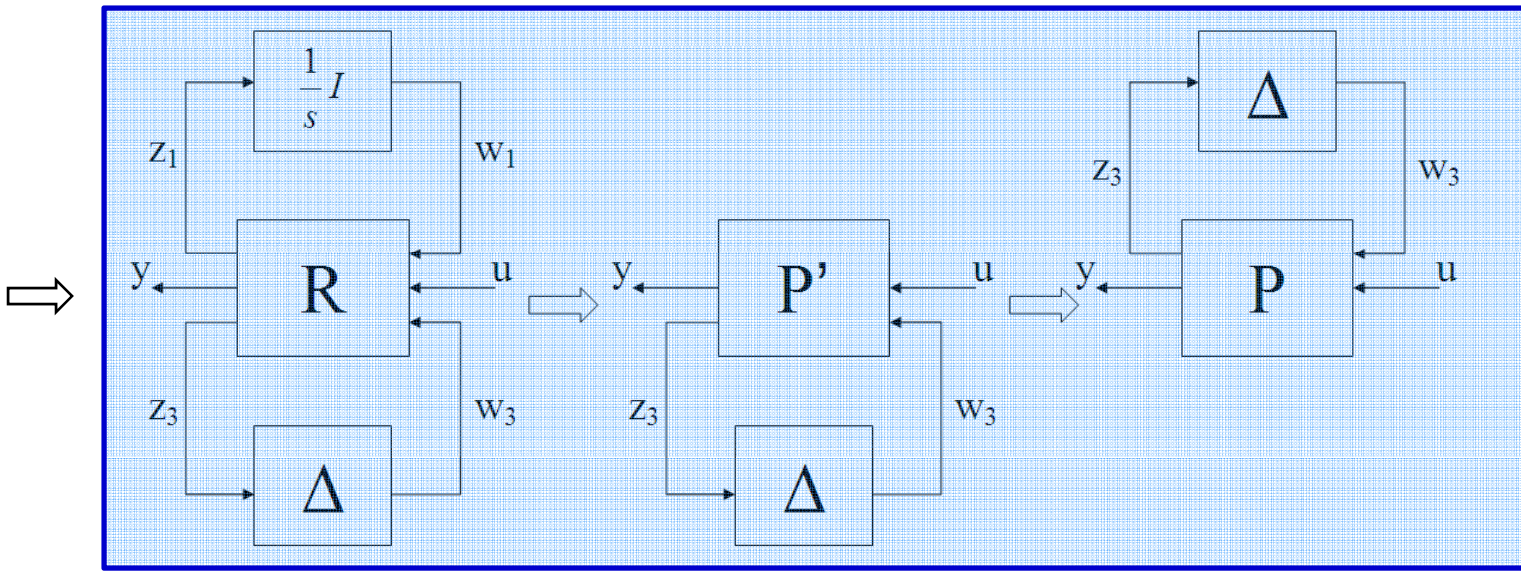
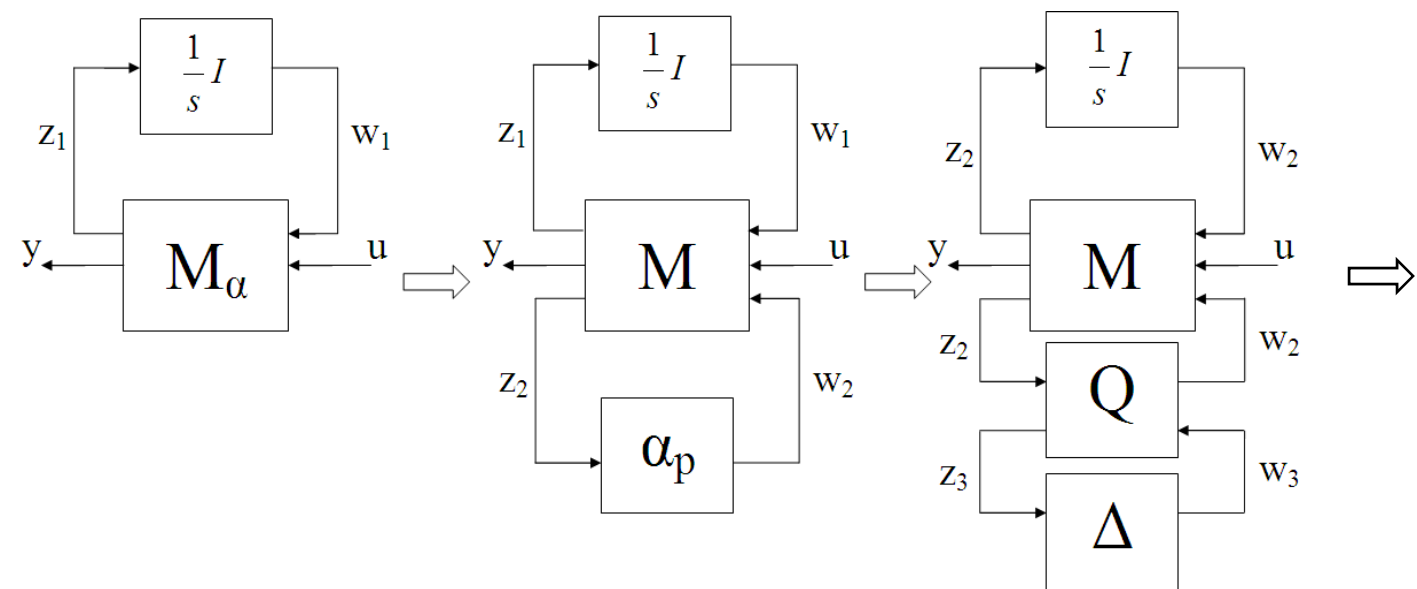
$$G(s) = F_u\left(M_\alpha, \frac{1}{s}I\right) = F_u\left(F_l(M_\alpha, \alpha_p), \frac{1}{s}I\right) = F_u\left(F_l(M_\alpha, F_l(Q, \Delta)), \frac{1}{s}I\right)$$

$$G(s) = F_u\left(F_l(R, \Delta), \frac{1}{s}I\right) = F_l\left(F_u\left(R, \frac{1}{s}I\right), \Delta\right) = F_l(P', \Delta) = F_u(P, \Delta)$$

where

$$R = \begin{bmatrix} M_{11} + M_{12}Q_{11}M_{21} & M_{12}Q_{12} \\ Q_{21}M_{21} & Q_{22} \end{bmatrix} \quad P' = F_u\left(R, \frac{1}{s}I\right) = \begin{bmatrix} P'_{11} & P'_{12} \\ P'_{21} & P'_{22} \end{bmatrix}$$

Model Progression



D-K Iteration

- No direct method to synthesize a μ -optimal controller.
- DK-iteration combines H_∞ synthesis and μ -analysis.
- This method starts with the upper bound on μ in terms of the scaled singular value

$$\mu(N) \leq \min_{D \in \mathcal{D}} \bar{\sigma}(DND^{-1})$$

where \mathcal{D} is the set of matrices D which commute with Δ ,
i.e., $D\Delta = \Delta D$.

- Then, the controller that minimizes the peak value over frequency of this upper bound is found, namely

$$\min_K \left(\min_{D \in \mathcal{D}} \|DN(K)D^{-1}\|_\infty \right)$$

D-K Iteration

- The controller is designed by alternating between the two minimization problems until reasonable performance is achieved.
- Follow the steps until $\|DN(K)D^{-1}\|_{\infty} < 1$, or H_{∞} norm doesn't decrease.
 1. **K-step.** Design an H_{∞} controller for the scaled problem with fixed $D(s)$.
 2. **D-step.** Find $D(j\omega)$ to minimize upper bound at each frequency with fixed N .
 3. Fit the magnitude of each element of $D(j\omega)$ to a stable and minimum-phase transfer function $D(s)$ and go to step 1.

Controller Simulation Results

- High rotating plasma, the growth rate γ ranges from 10 *rad/s* to 5,000 *rad/s*. This results in a range for the uncertain parameter c_{pp} that goes from 71 to 0.3325.
- Controllers designed with a smaller, more unstable nominal c_{pp} value produce the widest range of stability for c_{pp} .

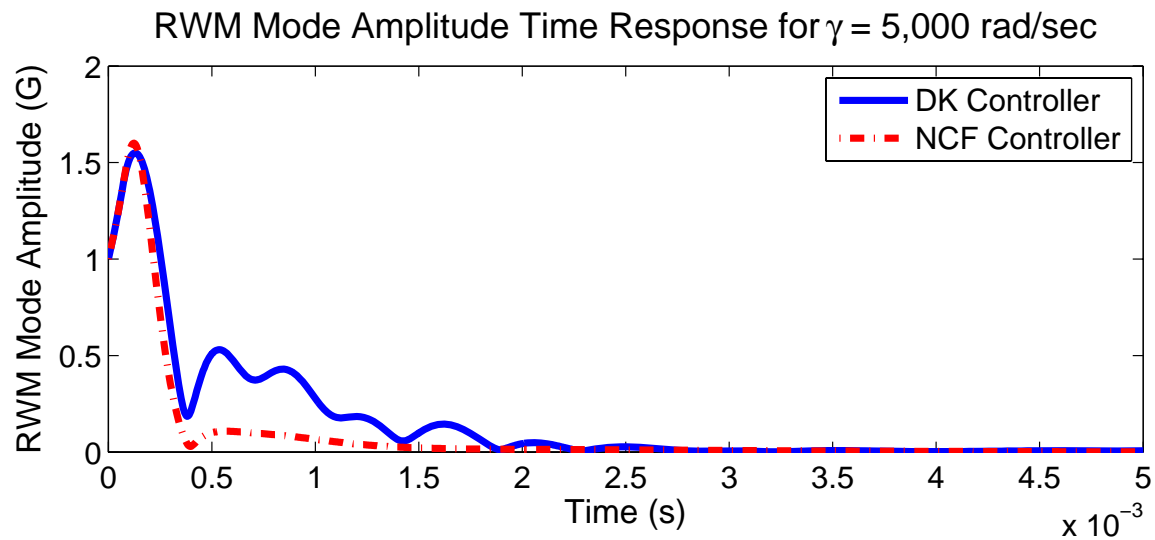
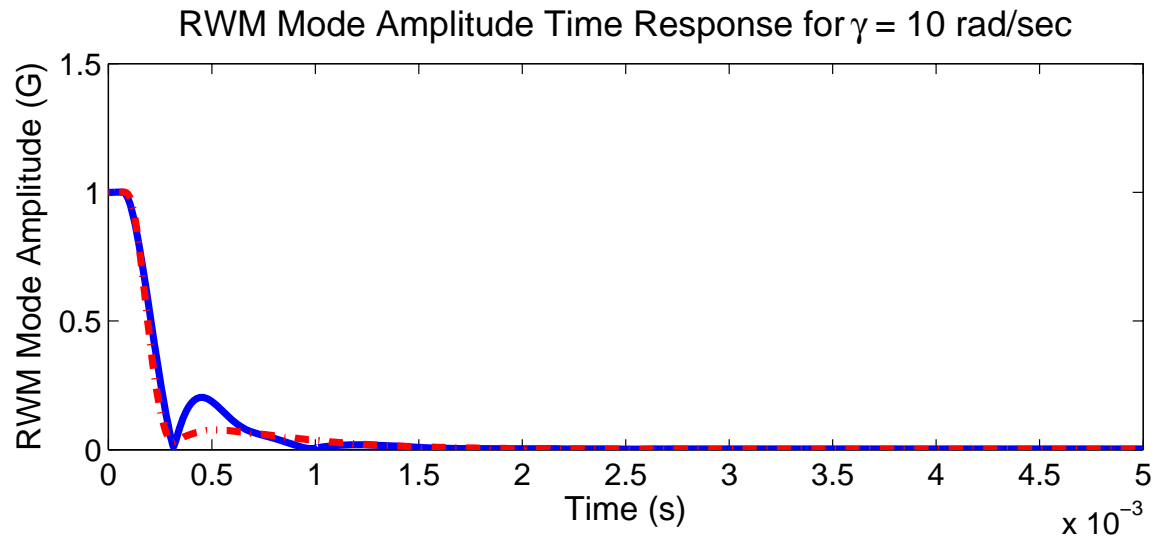
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Performance Constraints

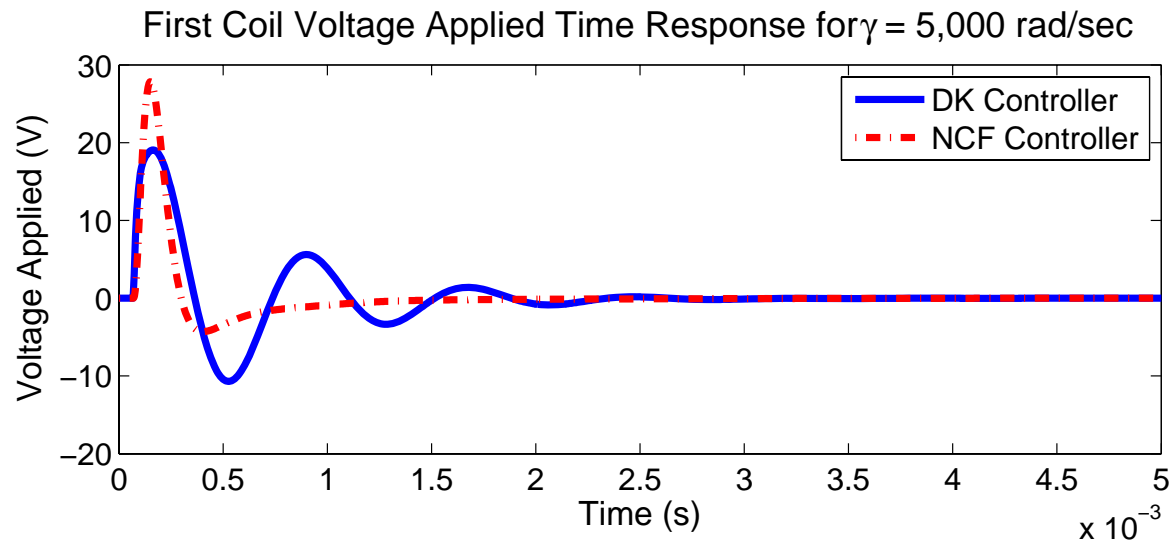
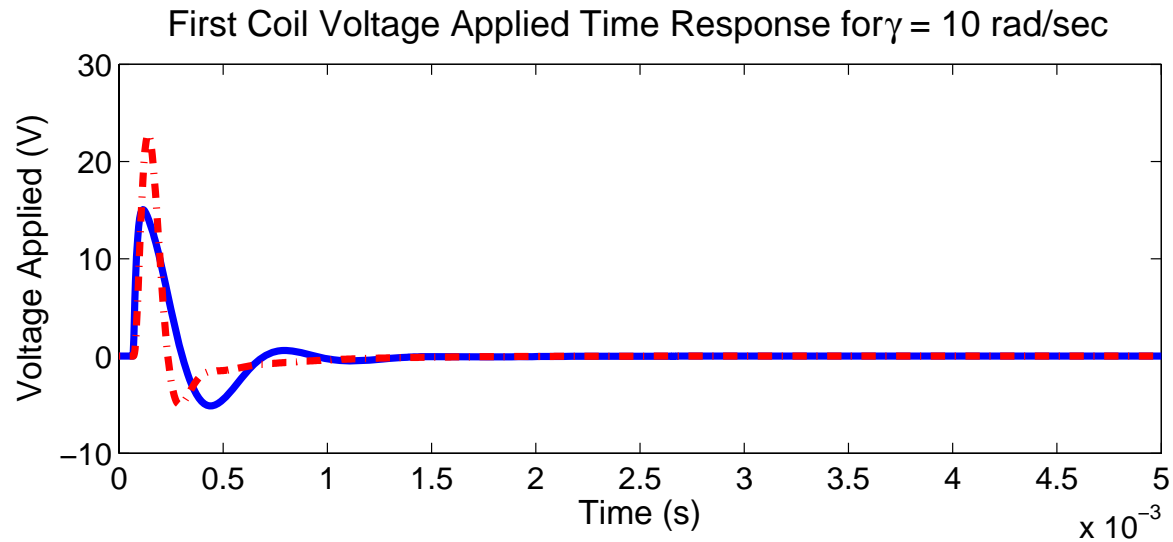
| Controller | DK |
|-----------------------|-----------------|
| Stability Range | 0 - 7,437 rad/s |
| Perf. Range (Step) | 0 - 7,254 rad/s |
| Perf. Range (Initial) | 0 - 6,459 rad/s |

Stability\Performance Ranges

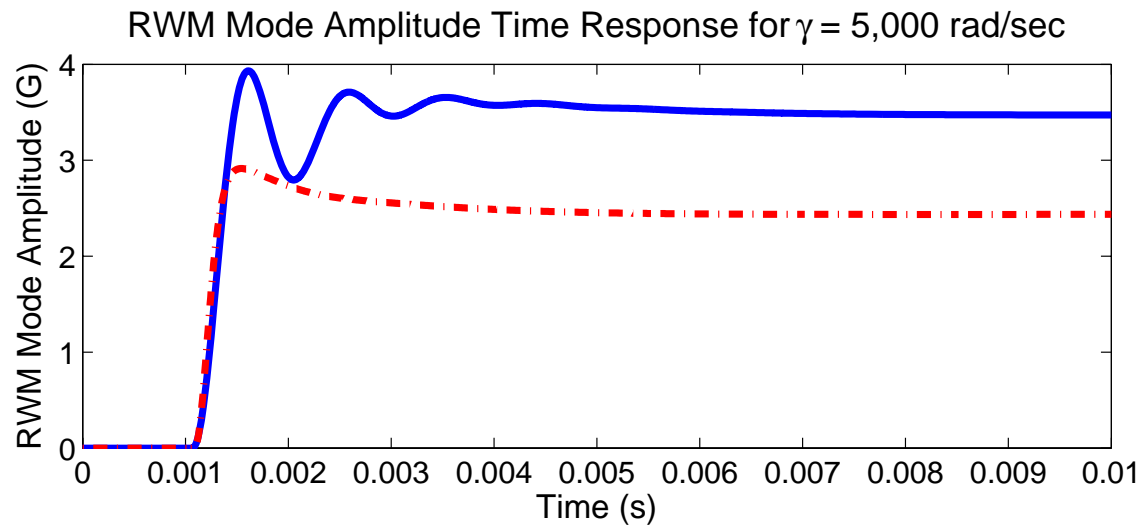
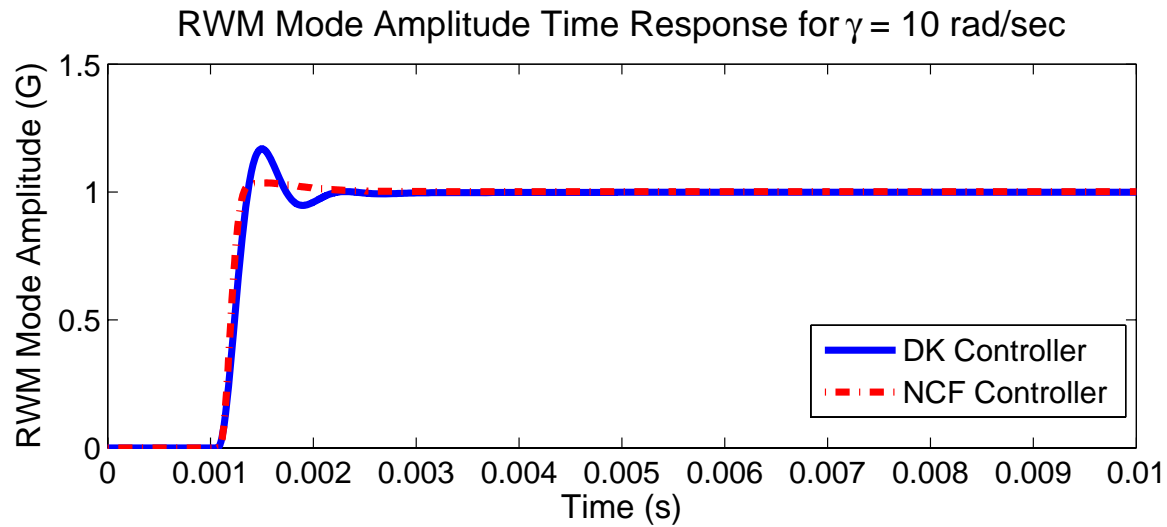
Initial Condition Response



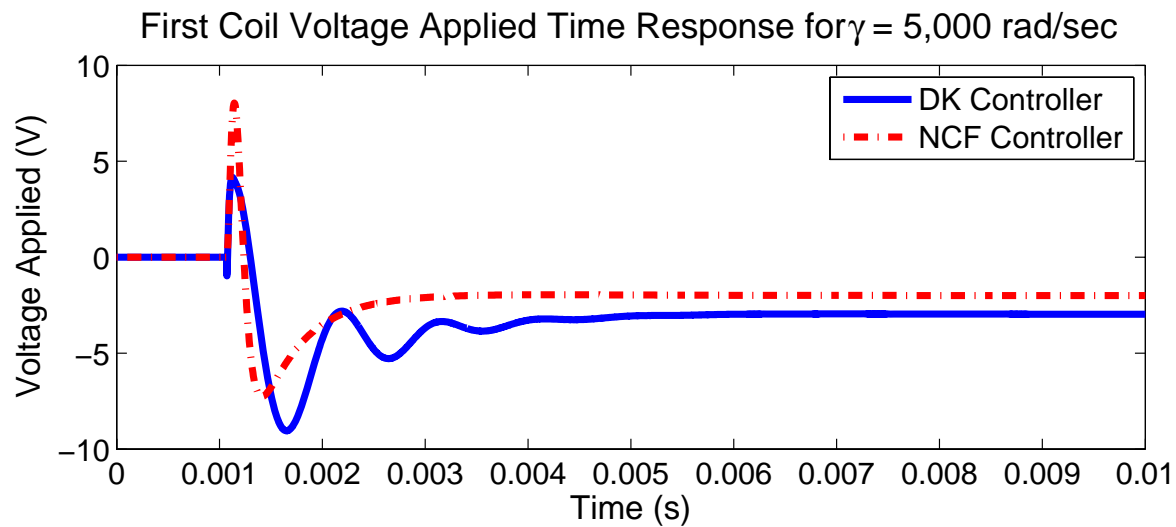
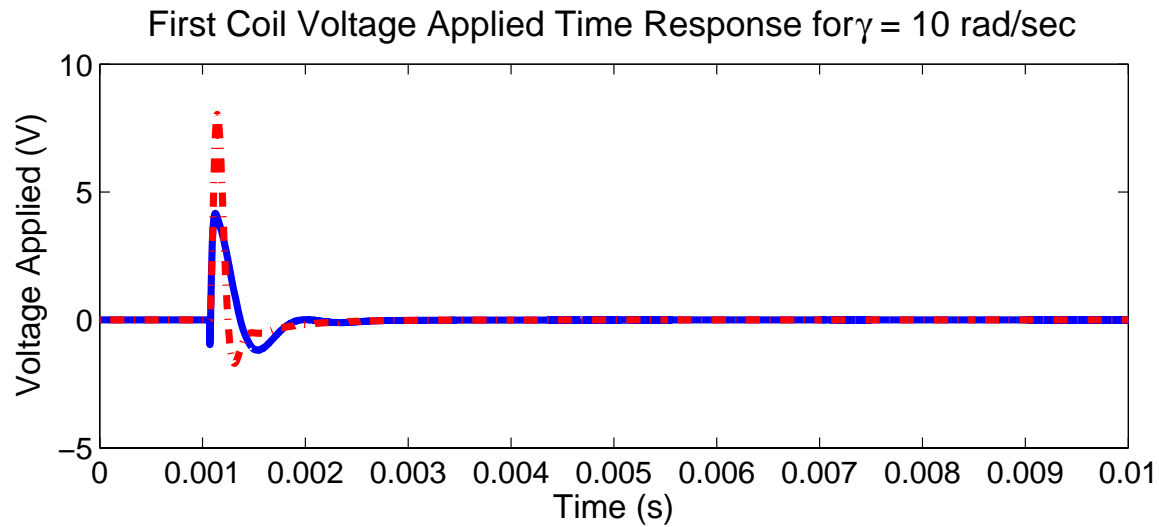
Initial Condition Response



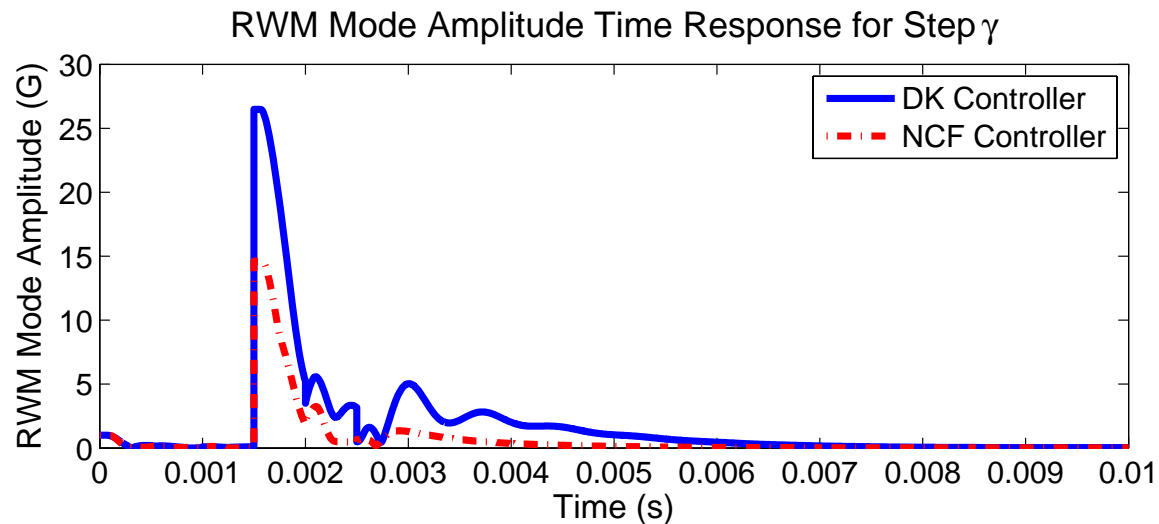
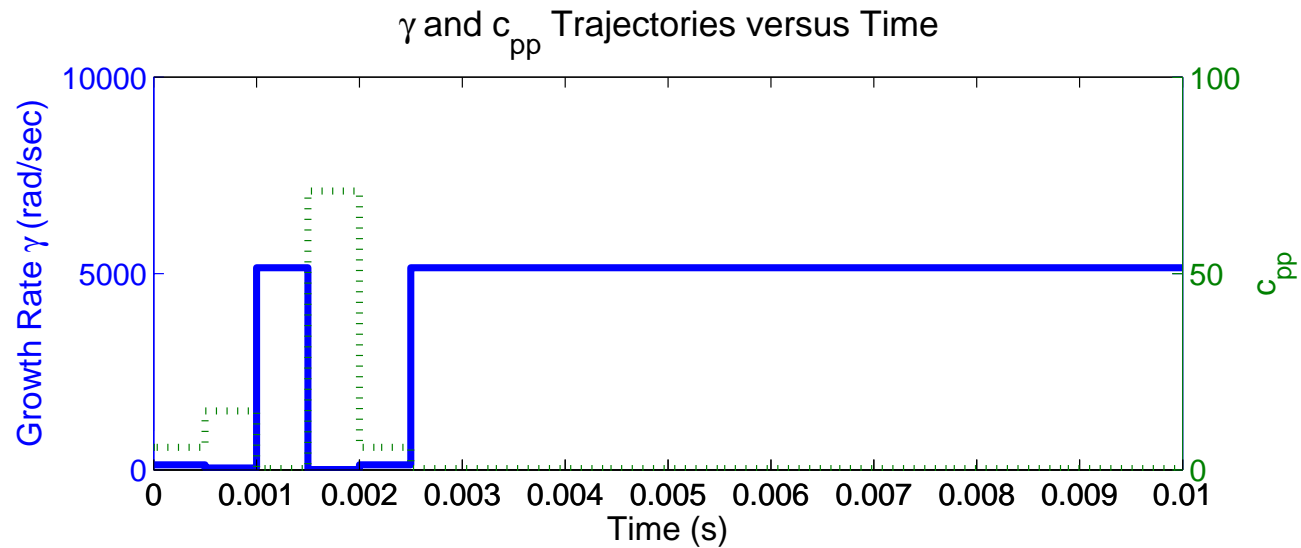
Step Response



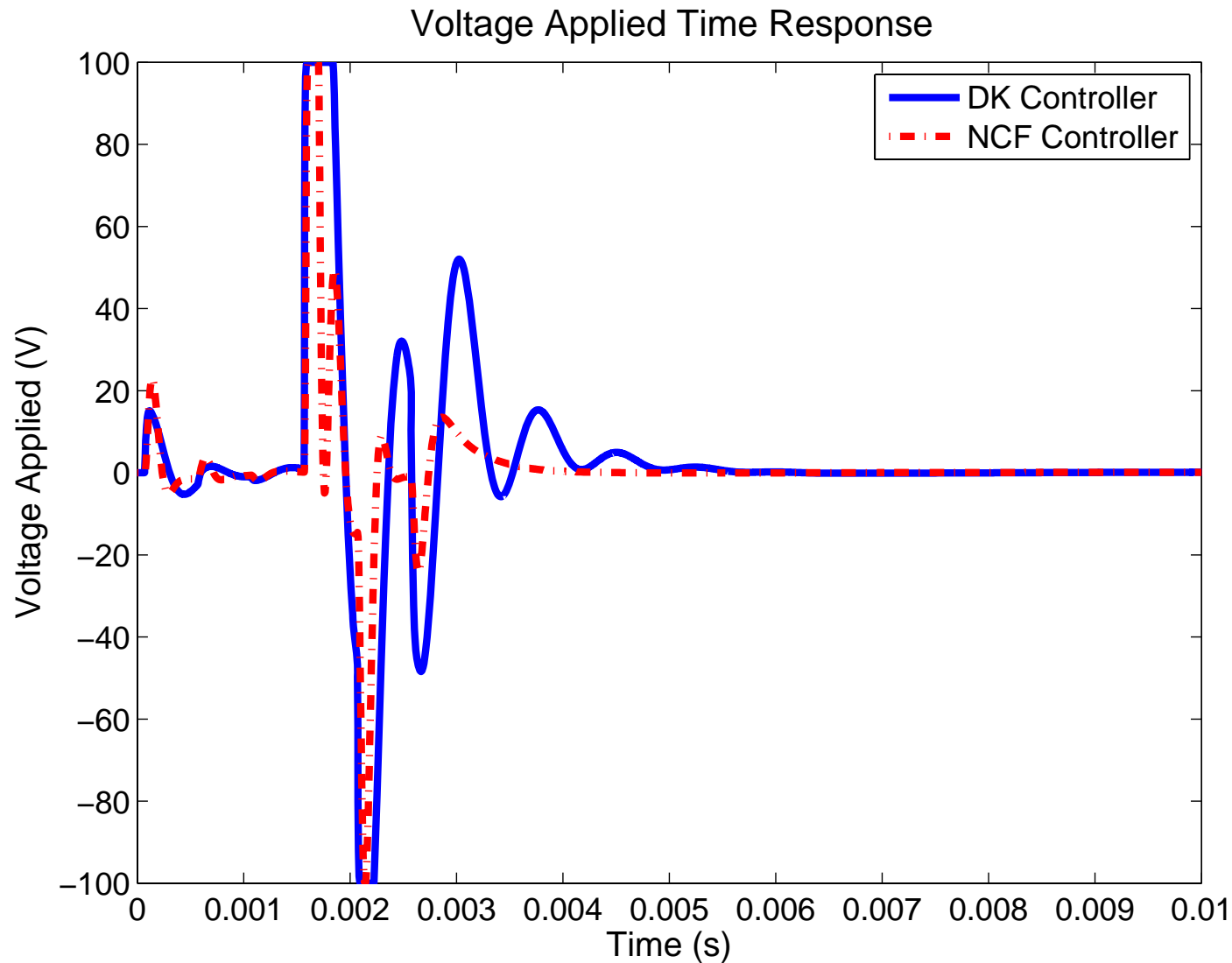
Step Response



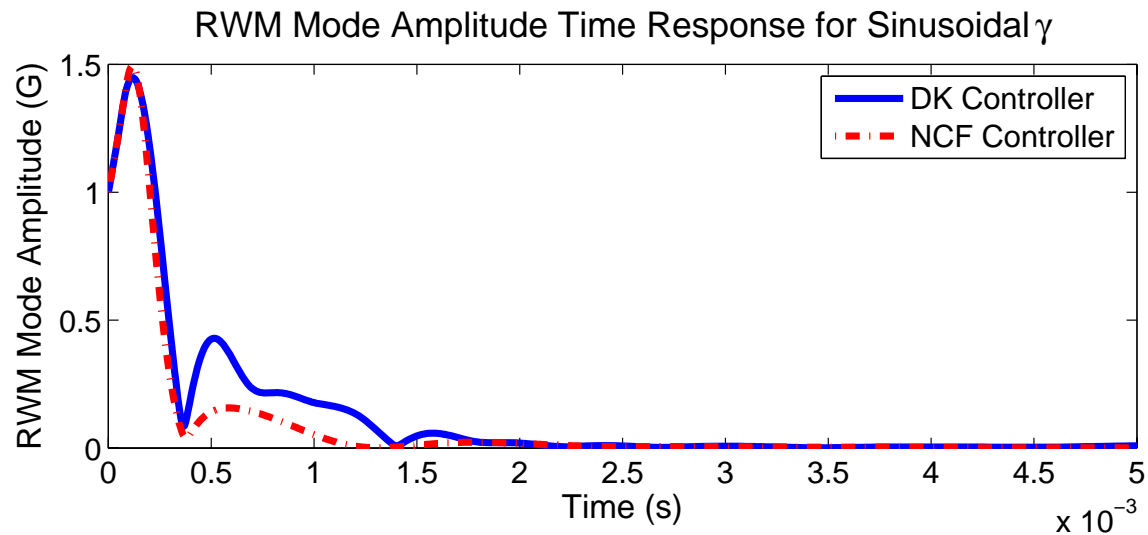
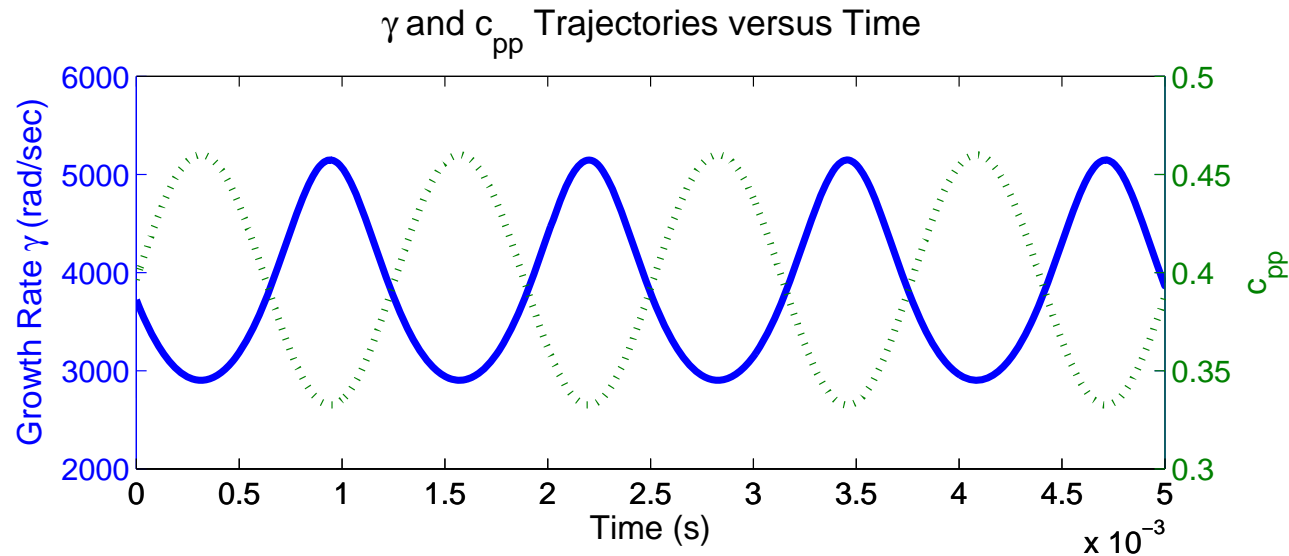
Time-varying Growth Rate Response



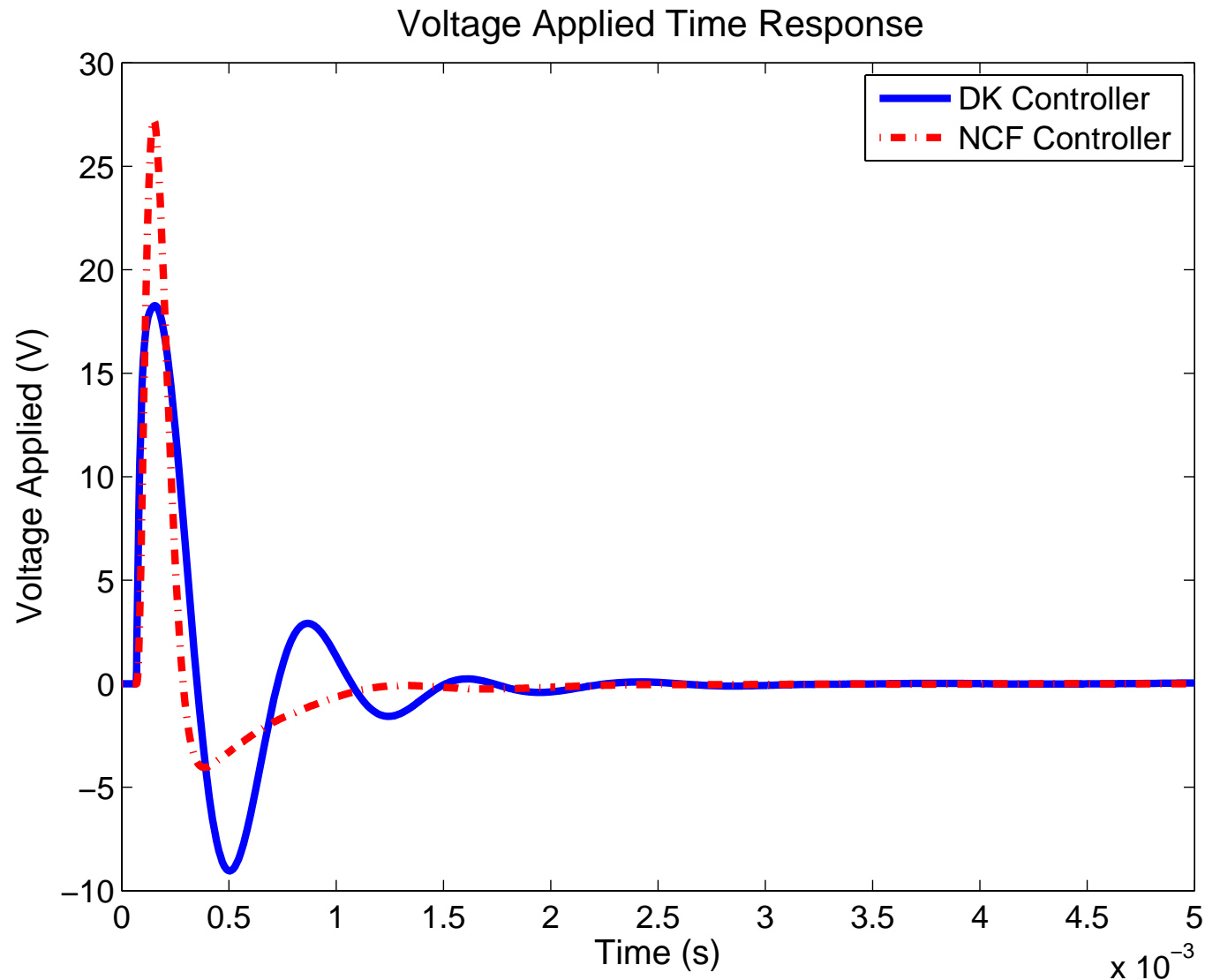
Time-varying Growth Rate Response



Sinusoidal Growth Rate Response



Sinusoidal Growth Rate Response



Conclusions

- Toroidal current sheet model for the DIII-D tokamak plasma was restructured into a robust control framework, isolating the RWM growth rate $\gamma (c_{pp})$, the key term of RWM instability.
- With the system model in this framework, the parametric *DK*-iteration method was applied to develop a structured-singular-value based robust controller for a pre-determined range of γ .
- Since the plasma RWM growth rate can vary throughout the operation of the DIII-D tokamak, the design of a controller that can stabilize the system over the entire physical range of γ is critical.
- In terms of robust stability, this method eliminates the need of online identification and controller scheduling.