

Robust Pole Placement for Plants with Semialgebraic Parametric Uncertainty

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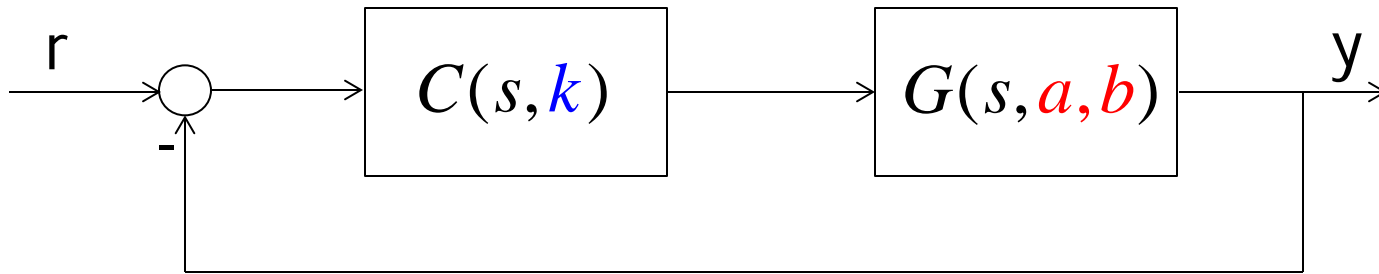


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Problem Formulation

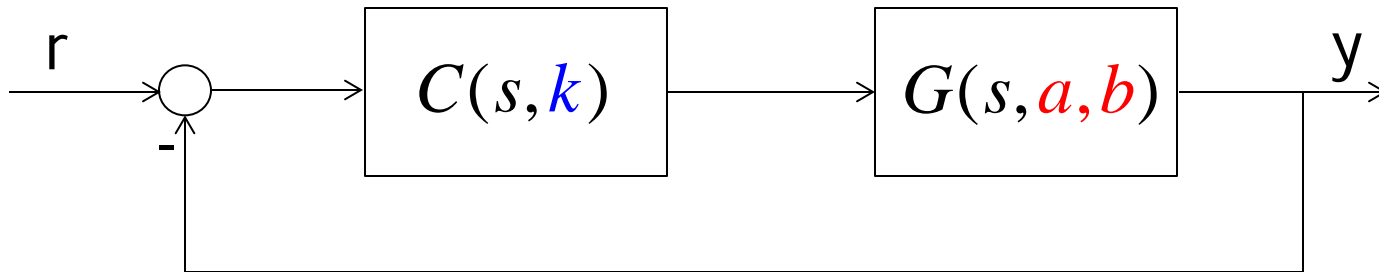


$$G(s, a, b) = \frac{N_G(s, b)}{D_G(s, a)} = \frac{b_{n-1}s^{n-1} + \dots + b_1s + b_0}{a_ns^n + \dots + a_1s + a_0}$$

(a, b) belong to an uncertainty set P

P : semialgebraic set

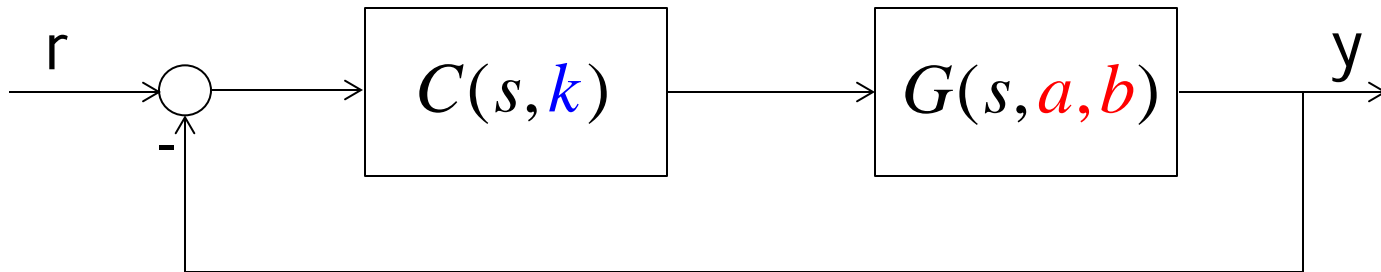
Problem Formulation



$$C(s, k) = \frac{N_C(s, k)}{D_C(s, k)} = \frac{k_r s^r + \cdots + k_1 s + k_0}{k_{2r+1} s^r + \cdots + k_{r+2} s + k_{r+1}} \quad r \geq n - 1$$

k : design parameters

Problem Formulation

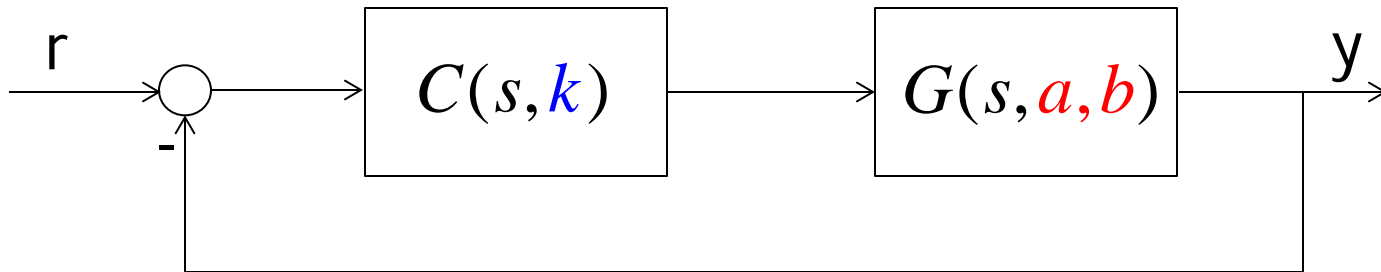


$$F(s, k, a, b) = D_G(s, a, b)D_C(s, k) + C_G(s, a, b)C_C(s, k) = f_{n+r}s^{n+r} + \cdots f_1s + f_0$$

$$\underline{f}_j \leq f_j \leq \overline{f}_j, \quad j = 0, \dots, n+r \quad \text{for all } (a, b) \in P$$

$F(s, k, a, b)$: (uncertain) closed-loop characteristic polynomial

Problem Formulation

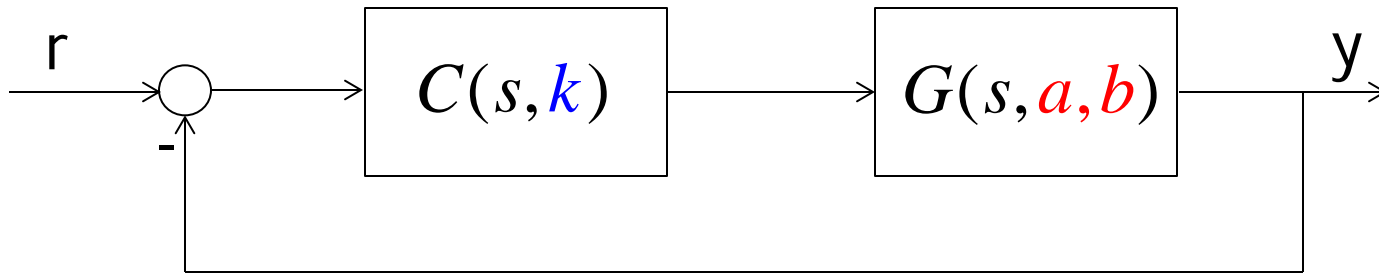


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$$\underline{f}_j \leq f_j \leq \overline{f}_j, \quad j = 0, \dots, n+r \quad \text{for all } (a, b) \in P$$

$$S(a, b)k = f, \quad \underline{f} \leq f \leq \overline{f} \quad \text{for all } (a, b) \in P$$

Problem Formulation



$$S(a, b)k = f, \quad \underline{f} \leq f \leq \overline{f} \quad \text{for all } (a, b) \in P$$

Robust Diophantine equation

Robust Pole Placement

Two problems have to be considered in robust pole placement:

- checking robust nonsingularity of the Sylvester matrix $S(a,b)$ over the uncertainty set P
- controller parameters computation

Checking robust nonsingularity of uncertain matrices

- ❑ Checking robust nonsingularity of uncertain matrices is an NP-hard problem
- ❑ Previously proposed approaches:
 - are only able to deal with the case that P is a box
 - provide conservative nonsingularity conditions
- ❑ New approach:
 - is able to deal with the case that P is a semialgebraic set
 - in the case P is a box, it provides less conservative nonsingularity conditions w.r.t. previously proposed approaches

Checking robust nonsingularity of uncertain matrices

Result 1

The uncertain matrix $S(a, b)$, with $(a, b) \in P$, is robustly nonsingular **if and only if** the following optimization problem is bounded:

$$h^* = \max_{\substack{x \in \mathbb{R}^{n+r} \\ a, b \in P}} \|x\|_2^2 \quad s.t. \quad S(a, b)x = 0$$

Checking robust nonsingularity of uncertain matrices

$$h^* = \max_{\substack{x \in \mathbb{R}^{n+r} \\ a, b \in P}} \|x\|_2^2 \quad s.t. \quad S(a, b)x = 0$$

An upper bound \bar{h} of h^* can be computed by exploiting LMI-relaxation techniques for polynomial optimization problems:

- Theory of moments (*Lasserre 2001*)
- SOS decomposition (*Chesi et al. 1999; Parrillo 2003*)

Result 2

The uncertain matrix $S(a, b)$, with $(a, b) \in P$, is robustly nonsingular if \bar{h} is finite

Robust nonsingularity of uncertain matrices: example

$$S(a,b) = \begin{bmatrix} 0 & b_2 & b_1 & b_0 & 0 & 0 \\ 0 & 0 & b_2 & b_1 & b_0 & 0 \\ 0 & 0 & 0 & b_2 & b_1 & b_0 \\ 1 & a_2 & a_1 & a_0 & 0 & 0 \\ 0 & 1 & a_2 & a_1 & a_0 & 0 \\ 0 & 0 & 1 & a_2 & a_1 & a_0 \end{bmatrix}$$
$$\begin{aligned} b_2 &\in [3 - \Delta, 3 + \Delta] & a_2 &\in [4 - \Delta, 4 + \Delta] \\ b_1 &\in [8 - \Delta, 8 + \Delta] & a_1 &\in [3 - \Delta, 3 + \Delta] \\ b_0 &\in [-5 - \Delta, -5 + \Delta] & a_0 &\in [-6 - \Delta, -6 + \Delta] \end{aligned}$$

Obtained results: matrix $S(a,b)$ is guaranteed to be robustly nonsingular for:

- $\Delta \leq 0.201$ (Beek, 1975)
- $\Delta \leq 0.208$ (Jansoon and Rohn, 1999)
- $\Delta \leq 0.420$ (LMI-relaxation approach)

For $\Delta = 0.427$ $S(a,b)$ is not robustly nonsingular

Design of robust controller

$$S(a, b)k = \underline{f}, \quad \underline{f} \leq \underline{f} \leq \overline{f} \quad \text{for all } (a, b) \in P$$

$$\underline{f} \leq S(a, b)k \leq \overline{f}, \quad \text{for all } (a, b) \in P$$

$$\underline{f}_i \leq S_i(a, b)k \leq \overline{f}_i, \quad i = 0, \dots, n + r, \quad \text{for all } (a, b) \in P$$

$$\begin{cases} S_i(a, b)k - \underline{f}_i \geq 0 \\ \overline{f}_i - S_i(a, b)k \geq 0 \end{cases} \quad i = 0, \dots, n + r, \quad \text{for all } (a, b) \in P$$

Design of robust controller

Problem 1

Find the controller parameters k such that

$$\begin{cases} S_i(a, b)k - \underline{f_i} \geq 0 \\ \overline{f_i} - S_i(a, b)k \geq 0 \end{cases} \quad i = 0, \dots, n + r, \quad \text{for all } (a, b) \in P$$

Robust Feasibility Problem

Design of robust controller

Problem 2

Find the controller parameters k and a set of SOS polynomials

$\sigma_j^2(a, b, \Theta_j)$ and $\chi_j^2(a, b, \Phi_j)$ such that

$$\begin{cases} S_i(a, b)k - \underline{f_i} = \sigma_0^2(a, b, \Theta_0) + \sum_{j=1}^{np} \sigma_j^2(a, b, \Theta_j) p_j(a, b) \\ \overline{f_i} - S_i(a, b)k = \chi_0^2(a, b, \Phi_0) + \sum_{j=1}^{np} \chi_j^2(a, b, \Phi_j) p_j(a, b) \end{cases} \quad i = 0, \dots, n + r$$

$$P = \{(a, b) : p_j(a, b) \geq 0, \quad j = 1, \dots, np\}$$

If k is feasible for Problem 2, then k is feasible for Problem 1

Design of robust controller

Find the controller parameters k and a set of SOS polynomials

$\sigma_j^2(a, b, \Theta_j)$ and $\chi_j^2(a, b, \Phi_j)$ such that

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- Enforcing $\sigma_j^2(a, b, \Theta_j)$ and $\chi_j^2(a, b, \Phi_j)$ to be SOS polynomials leads to LMI constraints in their coefficients
- Enforcing $S_i(a, b)k - \underline{f}_i = \sigma_0^2(a, b, \Theta_0) + \sum_{j=1}^{np} \sigma_j^2(a, b, \Theta_j) p_j(a, b)$ and $\overline{f}_i - S_i(a, b)k = \chi_0^2(a, b, \Phi_0) + \sum_{j=1}^{np} \chi_j^2(a, b, \Phi_j) p_j(a, b)$ leads to linear equality constraints in the design parameters k and in the coefficients of $\sigma_j^2(a, b, \Theta_j)$ and $\chi_j^2(a, b, \Phi_j)$

~~Robust~~ Feasibility Problem

Design of robust controller

Find the controller parameters k and a set of SOS polynomials

$\sigma_j^2(a, b, \Theta_j)$ and $\chi_j^2(a, b, \Phi_j)$ such that

$$\begin{cases} S_i(a, b)k - \underline{f}_i = \sigma_0^2(a, b, \Theta_0) + \sum_{j=1}^{np} \sigma_j^2(a, b, \Theta_j) p_j(a, b) \\ \overline{f}_i - S_i(a, b)k = \chi_0^2(a, b, \Phi_0) + \sum_{j=1}^{np} \chi_j^2(a, b, \Phi_j) p_j(a, b) \end{cases} \quad i = 0, \dots, n + r$$

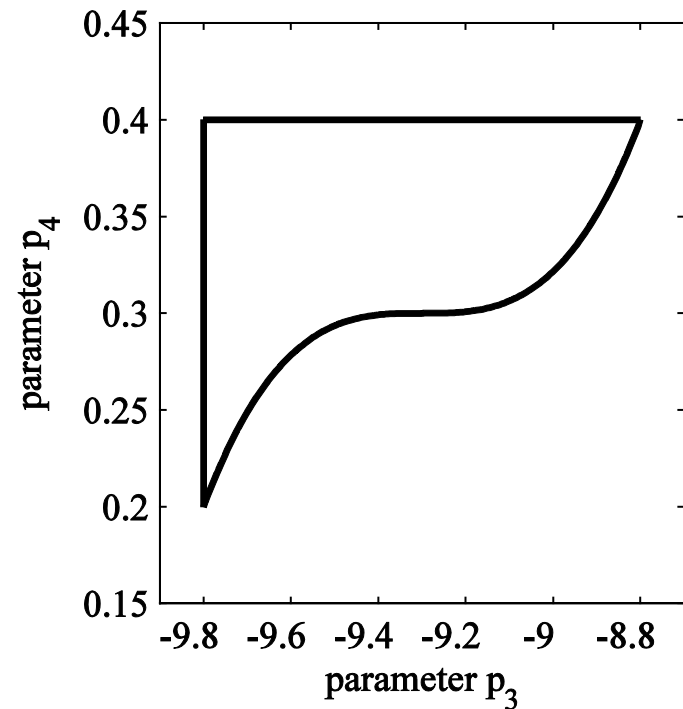
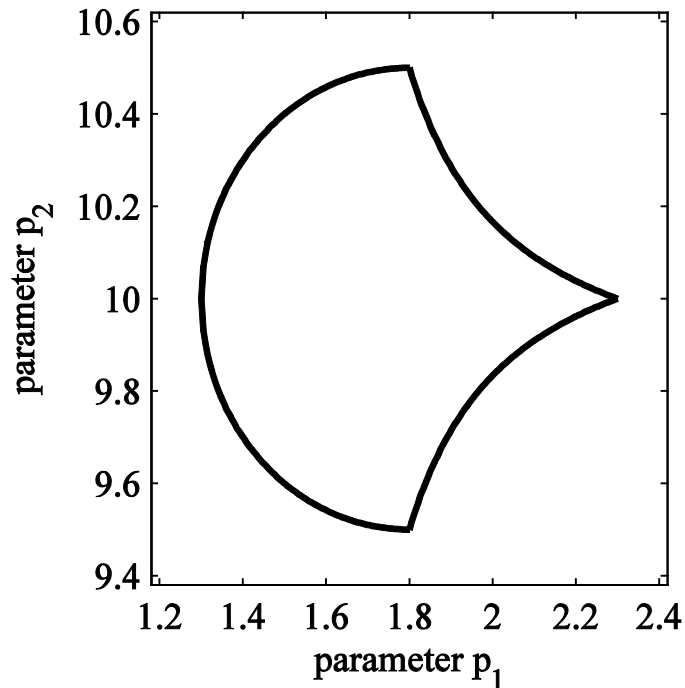
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Convex Feasibility Problem

Design of robust controller

Example

$$G(s) = \frac{p_1 s + p_2}{s^2 + p_3 p_4 s + p_2}$$



Design of robust controller

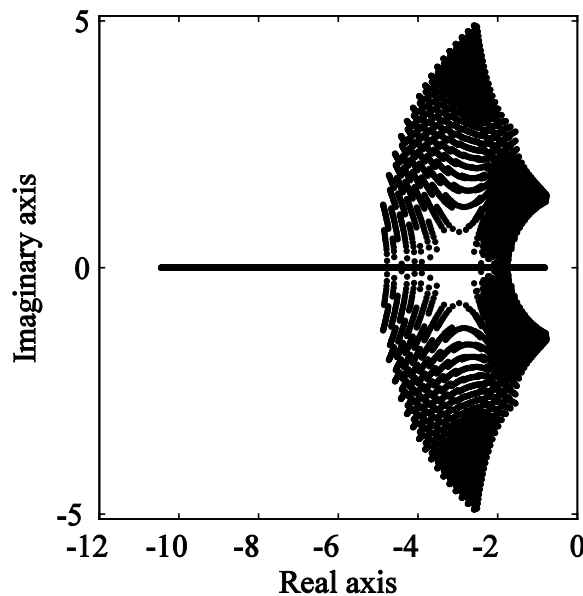
Example

$$F(s) = s^3 + f_2 s^2 + f_1 s + f_0$$

$$f_2 \in [6, 12]$$

$$f_1 \in [19, 35]$$

$$f_0 \in [25, 29]$$



$$C(s) = \frac{3.68s + 2.54}{s + 5.186}$$

Conclusions

- ❑ A novel approach to **check robust nonsingularity** of uncertain matrices is proposed
 - the considered problem can be formulated as a polynomial optimization problem
 - the case when the entries belong to an uncertain semialgebraic set can be treated
 - in the case of interval matrices, the presented method provides less conservative nonsingularity conditions w.r.t. previously published techniques
- ❑ The problem of **designing robust feedback controllers** is considered
 - the design problem is reduced to a robust feasibility problem over a nonconvex uncertainty set
 - SOS representation of positive polynomials is used to compute a robust controller through convex optimization