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## $H_2$ controller design using frequency-domain Data by convex optimization

Assumption:  $G(j\omega) = NM^{-1}$  is given for a SISO-LTI system  
when  $N(s)$  and  $M(s)$  are in  $R_\infty H_\infty$ .

Objective: compute a controller  $K(s) = X(s)Y(s)^{-1}$  that  
stabilizes the closed loop system and minimizes  $\|H\|_2^2$   
when  $H$  is a weighted closed-loop sensitivity function.

Idea: suppose that at <sup>every</sup>  $\omega$  we have  $H = W_1 S$  and

$$|H| = \frac{|W_1 M Y|}{|N X + M Y|} < \gamma(\omega)$$

Then we have the following optimization problem:

$$\min_{X, Y, \gamma} \int \gamma^2(\omega) d\omega$$

subject to:

$$|W_1 M Y| < \gamma(\omega) |N X + M Y| \quad \forall \omega$$

problem: the optimization is not convex!

For a finite set of frequencies we have:

$$\min_{X, Y, \gamma} \sum_{k=1}^N \gamma_k^2$$

$$\text{subject to } |W_1 M Y| < \gamma_k \operatorname{Re}\{N X + M Y\} \quad k=1, \dots, N$$

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the approximation of the constraint by replacing the absolute value of  $NX+MY$  with  $\text{Re}\{NX+MY\}$  has the interest that with increasing the order of  $X$  and  $Y$  we can converge to the real constraint. However, the new constraint is not convex because of multiplication with  $\delta_k$ , it is not quasi-convex because we have  $N$  variables ( $\delta_k \ k=1, \dots, N$ ).

Main research question: How we can convexify the constraint and keeping the property of convergence to the real constraint when order of  $X$  and  $Y$  increases

Note that if we have a stabilizing initial controller say  $K_0 = X_0 Y_0^{-1}$  we can convexify the constraints and solve the problem but we cannot show that the solution converges to the optimal one by increasing the order.

In the next page we show how we can convexify the problem by linearizing the nonconvex part around the initial controller.

③ let's define the following optimization problem

$$\min_{X, Y, \gamma} \sum_{k=1}^N \gamma_k$$

$$\text{subject to } |W_1 M Y|^2 < \gamma_k |N X + M Y|^2 \quad k=1, \dots, N$$

In order to convexify the constraints around  $K_0 = K_0 X_0^{-1}$ , take

$$Z = N X + M Y \Rightarrow |N X + M Y|^2 = Z^* Z$$

$$\text{we have } (Z - Z_0)^*(Z - Z_0) \geq 0 \quad \text{for any } Z_0 = N X_0 + M Y_0$$

$$\Rightarrow Z^* Z \geq Z^* Z_0 + Z_0^* Z - Z_0^* Z_0$$

$$\Rightarrow (W_1 M Y)^* \gamma_k^{-1} (W_1 M Y) < Z^* Z_0 + Z^* Z - Z_0^* Z_0$$

using the schur complement lemma the above constraints is equivalent to the following LMI,

$$\begin{bmatrix} Z^* Z_0 + Z^* Z - Z_0^* Z_0 & (W_1 M Y)^* \\ W_1 M Y & \gamma_k \end{bmatrix} > 0$$

so this LMI can be used as an inner convex approximation of the nonconvex constraint  $|W_1 M Y|^2 < \gamma_k |N X + M Y|^2$ .

The optimization problem can be improved iteratively by replacing  $K_0$  with the solution of the previous optimization. This algorithm will converge to a local optimal solution.