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# NUMERICAL METHODS FOR H<sub>2</sub> RELATED PROBLEMS

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#### Abstract

Recent results have shown that several  $H_2$  and  $H_2$ related problems can be formulated as convex programs with a *finite* number of variables. We present an interior point algorithm for the solution of these convex programs and illustrate its application with the standard LQR design.

#### 1. Introduction

It has been shown recently that a number of  $H_2$  and  $H_2$ -related problems can be formulated as convex programs with a finite number of variables — quadratic stabilization [1], mixed  $H_2/H_{\infty}$  and multicriterion LQG problems (see [2] and references therein). The common idea underlying these results is that though the original problem is not convex, a clever change of variables [3] makes it convex.

In this paper, we present a systematic procedure for transforming the convex programs resulting from  $H_2$ -related problems above into optimization over Affine Matrix Inequalities. We then present a simple interior point method for their solution. Though our presentation is through the simple LQR design example, the techniques readily extend to the more complicated problems cited above.

### 2. The LQR problem

Consider the linear time invariant system described by the state equations

$$\dot{x} = Ax + Bu + w z = \begin{bmatrix} R^{\frac{1}{2}} & 0 \\ 0 & Q^{\frac{1}{2}} \end{bmatrix} \begin{bmatrix} u \\ x \end{bmatrix}$$
(1)

where u is the control input, w is unit intensity white noise and z is the output signal of interest. The LQR problem is to design a feedback controller from the state x to the control input u which minimizes the L. El Ghaoui Ecole Nationale Supérieure de Techniques Avancées 32, Blvd. Victor, 75015 Paris, France.

 $H_2$  norm between w and z [4]. It is known that the optimal feedback law is a constant state feedback u = -Kx and optimizes the following program:

$$\min_{P,K} \mathbf{Tr}(QP) + \mathbf{Tr}(R^{1/2}KPK^TR^{1/2})$$

subject to

$$(A - BK)P + P(A - BK)^{T} + I < 0 \text{ and } P = P^{T} > 0.$$

By defining a new quantity Y = KP, the above problem can be written as

$$\min_{P,Y} \mathbf{Tr}(QP) + \mathbf{Tr}(R^{1/2}YP^{-1}Y^{T}R^{1/2})$$
(2)

subject to

$$AP + PA^{T} - BY - Y^{T}B^{T} + I < 0 \text{ and } P = P^{T} > 0,$$

which is a convex program (see, for example, [1], [2]).

# 3. Transformation to Optimization over AMI's

The general structure of the convex program (2) is

$$\min_{(P,Y)\in\mathcal{A}}J(P,Y)$$

where  $\mathcal{A}$  is some convex constraint set, and J a performance index. We show how this can be transformed into the problem

$$\min_{C(\gamma,Z)>0}\gamma\tag{3}$$

where  $(\gamma, Z)$  is a new set of variables and C is symmetric and an affine matrix function of  $(\gamma, Z)$ . The inequality C(z) > 0 is called an Affine Matrix Inequality (AMI).

#### Example: The LQR problem

The objective function of program (2) consists of the sum of two terms. It is easily shown that the second term

$$\Phi(P,Y) = \mathbf{Tr}(R^{1/2}YP^{-1}Y^TR^{1/2})$$

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can be expressed as

$$\Phi(P, Y) = \min(\mathbf{Tr}(X))$$
$$\begin{bmatrix} X & R^{1/2}Y \\ Y^T R^{1/2} & P \end{bmatrix} > 0$$

Then let

$$\begin{split} &C_1(\gamma, P, Y, X) := -\mathbf{Tr}(QP) - \mathbf{Tr}(X) + \gamma, \\ &C_2(\gamma, P, Y, X) := -AP - PA^T + BY + Y^TB^T - I, \\ &C_3(\gamma, P, Y, X) := \begin{bmatrix} X & R^{1/2}Y \\ Y^TR^{1/2} & P \end{bmatrix}, \\ &C(\gamma, P, Y, X) := \mathrm{diag}(C1, C2, C3). \end{split}$$

The optimization problem (2) can now be written

$$\begin{array}{l} \text{minimize } \gamma \\ C(\gamma, P, Y, X) > 0 \end{array} \tag{4}$$

which indeed is of the form (3).

# Well-Posedness

We say that the convex program (3) is well-posed if for every real  $\gamma$  the set  $\{Z \mid C(\gamma, Z) > 0\}$  is compact or empty. We observe the following without proof:

**Proposition 1:** The program (4) corresponding to the LQR problem is well-posed if R is positive definite, (A, B) is controllable and (Q, A) is observable.

Under similar assumptions, all the other problems cited in the introduction enjoy the same property.

# 4. Computational aspects

Problem (3) is a convex non-differentiable optimization program, and the ellipsoid algorithm or Kelley's cutting-plane algorithm [4] may be used to solve it. Recently, the work of Nemirovski *et al.* has led to the development of interior point algorithms based on the notion of the *analytic center* for a set of convex constraints [5]; these algorithms seem to hold great promise.

We will describe one such interior point algorithm, called the *method of centers*. Given an initial feasible point  $(\gamma_u, Z_u)$  for constraint C in program (3), and a desired absolute accuracy  $\epsilon$  on the optimum, the algorithm is as follows:

while 
$$\gamma_u - \gamma_l > \epsilon$$
,  
 $\gamma_0 := \gamma_u + \epsilon$ ,  
 $(\gamma^*, Z^*) := a\_center (diag(C, \gamma_0 - \gamma) > 0),$   
 $\gamma_u := \gamma^*$ ,

 $Z_u := Z^*,$ Compute a lower bound  $\gamma_l$ . end

**Remark 1:** An initial lower bound  $\gamma_l$  can be chosen to be 0.

**Remark 2:** Computing the analytic center of a convex bounded set (a\_center(diag( $C, \gamma_0 - \gamma) > 0$ )) needs an initial point interior to the constraint.  $(\gamma_u, Z_u)$  is such a point.

**Remark 3:** The lower bound  $\gamma_l$  is computed from  $\gamma^*$  and the Hessian of the barrier function of the constraint expressed at  $(\gamma^*, Z^*)$ . For more information, see [5], [6] and also [7] (these proceedings).

# 5. Conclusion

Through the LQR example, we have outlined a systematic procedure for transforming convex optimization programs arising from  $H_2$ -related problems into optimization over AMI's. We have also briefly described a simple interior point method for their solution. Our procedure easily applies to problems in [1] and [2], and more generally to many quadratic Lyapunov function shaping problems.

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