# MECH 6327 - Homework 4

Jonas Wagner

2021, April 12

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## **BV** Textobook Problems

### 0.1 Problem 5.43

The dual a SOCP defined as:

minimize 
$$f^T x$$
  
subject to  $\|A_i x + b_i\|_2 \le c_i^T x + d_i, \ i = 1, \dots, m$  (1)

with  $x \in \Re^n$  can be expressed as:

maximize 
$$\sum_{i=1}^{m} (b_i^T u_i - d_i v_i)$$
subject to 
$$\sum_{i=1}^{m} (A_i^T u_i - c_i v_i) + f = 0$$

$$\|u_i\|_2 \le v_i, \ i = 1, \dots, m$$

$$(2)$$

with variables  $u_i \in \Re_i^n$ ,  $v_i \in \Re$ , i = 1, ..., m and problem data  $f \in \Re^n$ ,  $A_i \in \Re^{n_i \times n}$ ,  $b_i \in \Re^{n_i}$ ,  $c_i \in \Re$ , i = 1, ..., m.

#### 0.1.1 Part a

**Problem:** Derive the dual by defining  $y_i \in \mathbb{R}^{n_i}$  and  $t_i \in \mathbb{R}$  and the equalities  $y_i = A_i x + b_i$ ,  $t_i = c_i^T x + d_i$  then deriving the Lagrange dual.

**Solution:** The problem can first be written in a standard form as:

minimize 
$$f^T$$
  
subject to  $y_i = A_i x + b_i$   
 $t_i = c_i^T x + d_i$   
 $\|y_i\|_2 \le t_i, \ \forall i = 1, \dots, m$  (3)

The lagrange can then be defined with

$$L(x, y, t, \lambda_{i}, \nu_{i}, \mu_{i}) = f^{T} + \sum_{i=1}^{m} \lambda_{i} (\|y_{i}\|_{2} - t_{i}) + \sum_{i=1}^{m} \mu_{i}^{T} (t_{i} - C_{i}x + d_{i})$$

$$= \left( f + \sum_{i=1}^{m} \left( A_{i}^{T} \mu_{i} - c_{i}\nu_{i} \right) \right)^{T} x + \sum_{i=1}^{m} \lambda_{i} \|y_{i}\|_{2} + \mu_{i}^{T} y_{i}$$

$$+ \sum_{i=1}^{m} \left( -\lambda_{i} + \nu_{i} \right) t_{i} - \left( \sum_{i=1}^{m} b_{i}^{T} \mu_{i} - d_{i}\nu_{i} \right)$$

$$(5)$$

Since the definition of the dual optimization problem is to maximize

$$g(\lambda_i, \nu_i, \mu_i) = \inf_{x, y_i, t_i} L(x, y, t, \lambda_i, \nu_i, \mu_i)$$

the inf can be found by determining when a min/max would occur for each of the variables.

For the critical point on x the direvative can be set to zero and thus the following equality must hold:

$$f + \sum_{i=1}^{m} \left( A_i^T \mu_i - c_i \nu_i \right) = 0 \tag{6}$$

For the  $y_i$  related term, it is known

$$\sum_{i=1}^{m} \lambda_i \|y_i\|_2 + \mu_i^T y_i$$

will be bounded below if it is within the cone defined by  $\lambda_i ||y_i||_2 \ge ||\mu_i||_2 y_i$  which can be rewritten as:

$$\|\mu_i\|_2 \leq \lambda_i$$

For the critical point over  $t_i$  the equality  $\nu_i = \lambda_i$ .

From this the dual problem can be obtained when the quantity  $\left(\sum_{i=1}^{m} b_i^T \mu_i - d_i \nu_i\right)$  is maximized. Thus the dual problem is defined as:

maximize 
$$\sum_{i=1}^{m} (b_i^T u_i - d_i v_i)$$
subject to 
$$\sum_{i=1}^{m} (A_i^T u_i - c_i v_i) + f = 0$$

$$\|u_i\|_2 \le v_i, \ i = 1, \dots, m$$

$$(7)$$

#### 0.1.2 Part b

**Problem:** Start with the conic formulation of the SOCP and use the conic dual to prove the equivalence. Use the fact that the second-order dual is self-dual.

Solution: Starting with the SOCP given as

minimize 
$$f^T x$$
  
subject to  $\|A_i x + b_i\|_2 \le c_i^T x + d_i, \ i = 1, \dots, m$  (8)

a standard form can be defined by

minimize 
$$f^T x$$
  
subject to  $(A_i x + b_i, c_i^T x + d_i) \leq_2 0$  (9)

Since the conic dual transformation is known to transform

minimize 
$$f^T x$$
  
subject to  $-(A^T x + b, c^T x + d) \leq_K 0$  (10)

into its dual according to its dual cone definition

maximize 
$$b^T u + d v$$
  
subject to  $A^T u + vc = f$  (11)  
 $(u, v) \succeq_{K^*} 0, i = 1, ..., m$ 

and from the fact that the 2-norm is self-dual, the dual program is given as

maximize 
$$\sum_{i=1}^{m} (b_i^T u_i - d_i v_i)$$
subject to 
$$-\sum_{i=1}^{m} (A_i^T u_i - c_i v_i) = f$$

$$(u_i, v_i) \succeq_2 0, \ i = 1, \dots, m$$

$$(12)$$

or equivalently

maximize 
$$\sum_{i=1}^{m} (b_i^T u_i - d_i v_i)$$
subject to 
$$\sum_{i=1}^{m} (A_i^T u_i - c_i v_i) + f = 0$$

$$\|u_i\|_2 \le v_i, \ i = 1, \dots, m$$

$$(13)$$

#### Problem 1: Robust control design 1

For the standard DT dynamical system defined as:

$$x_{t+1} = Ax_t + Bu_t \tag{14}$$

with dynamic matrix A unknown but assumed to belong to a set:

$$A \in \mathcal{A} = \operatorname{conv}(A_1, \dots, A_m) \tag{15}$$

with  $A_i$  and B known.

**Problem:** For a state-feedback controller  $u_t = Kx_t$  use Lyapunov techniques to design it so the system is globally asymptotically stable (GAS) by solving a semi-definite program (SDP).

Solution: The closed-loop system for the DT dynamical system can be defined by the dynamics

$$x_{t+1} = \hat{A}x = (A + BK)x \tag{16}$$

where  $\hat{A} = A + BK$ .

In order for the closed-loop system to be Globally Asymptotically Stable, a quadratic Lyapnov Function could be used to prove that if the following inequality is true then the system is GAS:

$$\hat{A}^T P \hat{A} - P \prec 0 \tag{17}$$

Since the system dynamics themselves are uncertain, this inequality will not be enough to prove GAS. This can be be address, however, by considering all  $A \in \mathcal{A}$  to be a linear combination of the individual corner matrices. Since this is a convex hull, it is known that following this to its conclusion, GAS can be guaranteed for all  $A \in \mathcal{A}$  if  $A_i$  is GAS  $\forall i = 1, ..., m$ .

Following this, a stabilizing gain can then be found as follows:

$$\hat{A}_i^T P \hat{A}_i - P \prec 0 \tag{18}$$

recognizing the Schur's compliment form, the following is true

$$\begin{bmatrix} P & \hat{A}_i^T \\ \hat{A}_i P^{-1} \end{bmatrix} \succ 0 \tag{19}$$

$$\begin{bmatrix}
P & \hat{A}_i^T \\
\hat{A}_i P^{-1}
\end{bmatrix} \succ 0$$

$$\begin{bmatrix}
P & (A_i + BK)^T \\
(A_i + BK) & P^{-1}
\end{bmatrix} \succ 0$$
(20)

A SDP feasibility problem can then be done to solve the problem such that

$$\begin{bmatrix}
P & \hat{A}_i^T \\
\hat{A}_i P^{-1}
\end{bmatrix} \succ 0$$

$$B^{-1} (\hat{A}_i - A_i) - K = 0, \ \forall i = 1, \dots, m$$
(21)

with variables P,  $\hat{A}_i$ , and K, along with problem data  $A_i$  and B. This is a problem that can now be easily implemented using CVX or YALMIP in MATLAB for given problem data.

## 2 Problem 2: Nonnegative and sum of squares polynomials

The Motzkin polynomial is defined as:

$$M(x,y) = x^2y^4 + x^4y^2 + 1 - 3x^2y^2$$
(22)

**Problem:** Show that the Motzkin polynomial is nonnegative but can be expressed as sum of squares. It is sufficient to show this using numerical and/or symbolic solvers.

**Solution:** Nonegativity of the Motzkin polynomial can be proven using the AM-GM inequality using n=3 with  $x^4y^2, x^2y^4, 1$ .

$$\sqrt[n]{\prod_{i=1}^{n} x_i} \le \frac{1}{n} \sum_{i=1}^{n} x_i \tag{23}$$

$$\sqrt[3]{(x^4y^2)(x^2y^4)(1)} \le \frac{1}{3}(x^4y^2 + x^2y^4 + 1) \tag{24}$$

$$x^{2}y^{2} \le \frac{1}{3}(x^{4}y^{2} + x^{2}y^{4} + 1) \tag{25}$$

$$3x^2y^2 \le x^4y^2 + x^2y^4 + 1 \tag{26}$$

$$x^4y^2 + x^2y^4 + 1 - 3x^2y^2 \ge 0 (27)$$

Therefore the Motzkin polynomial is non-negative.

However, it is not possible to put this into sum of square form using the standard solver as demonstrated by the infeasability result from the solvesos() command in yalmip (shown in AppendixA)

## A MATLAB Code:

All code I write in this course can be found on my GitHub repository: https://github.com/jonaswagner2826/MECH6327

Script 1: MECH6327\_HW4

```
% MECH 6327 - HW 4
3
   x = sdpvar(1,1);
   y = sdpvar(1,1);
   p = x^2 * y^4 + x^4 * y^2 + 1 - 3 * x^2 * y^2;
   F = sos(p);
   solvesos(F);
   sdisplay(sosd(F))
8
9
11
   % Results:
   % -----
12
   % -----
13
   % YALMIP SOS module started...
14
   % -----
15
  % Detected 0 parametric variables and 2 independent variables.
16
  % Detected O linear inequalities, O equality constraints and O LMIs.
17
   % Using kernel representation (options.sos.model=1).
18
   \% Initially 8 monomials in R^2
19
20
   % Newton polytope (2 LPs)......Keeping 4 monomials (0.20313sec)
   % Finding symmetries......Found 3 symmetries (Osec)
   % Partitioning using symmetry....1x1(4)
22
23
  %
24
25
  % Problem is unbounded.
26
27
  % -> Solver reported unboundness of the dual problem.
28
  % -> Your SOS problem is probably infeasible (SOS is dualized).
```

**Refrences:** \* not bibtex becouse of time...

 $https://people.eecs.berkeley.edu/\ elghaoui/Teaching/EE227A/lecture10.pdf$ 

https://people.orie.cornell.edu/miketodd/iccopt.pdf