

Using Linear Matrix Inequalities to Optimize Anti-windup Gain for Convex Systems

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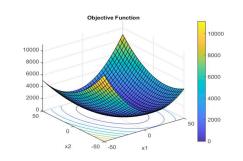
Introduction

This research sought out to determine a method for finding an optimal anti-windup gain for a convex system. The combination of two past works led to this method. The first publication, Static Anti-Windup Design, used a linear matrix inequality, or LMI, to determine an anti-windup gain for a nonlinear system. This LMI guaranteed regional stability paired with performance. The second paper, Saddle Point Convergence of Constrained Primal-Dual Dynamics, showed how a convex primal dual system can be implemented with control. This research bridges the gap between these works.

Convex Optimization Problem

We consider convex optimization problem of the form: $\min_{x} f(x)$ Subject to $g(x) = Cx - e \le 0$

- Convex objective function *f*(*x*)
- Linear inequality constraints $g(x) \le 0$



Classical Primal Dual Dynamics

Classical primal dual dynamic for solving convex optimization problems comprises both gradient descent flow in the primal variable x and gradient ascent flow in the Lagrangian (dual)variable λ towards the optimal solution.

$$\dot{x} = -\nabla f(x) + C^T \lambda$$
$$\dot{\lambda} = Cx - e$$

Methodology

The approach adopted here is to reformulate the primal-dual dynamics as a combination of three parts:

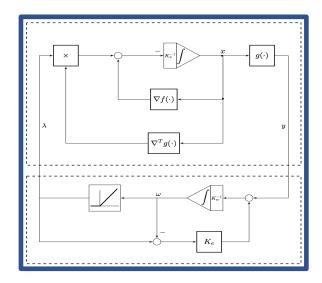
- An optimization plant corresponding to the gradient descent flow
- An optimization controller corresponding to the gradient ascent flow and
- An antiwindup augmentation that allows for affecting the behavior of the solver to a desired performance level.

The main task then is the determination of an appropriate antiwindup gain that encapsulates such performance specification.

The ensuing dynamical solver takes the form:

$$\begin{split} \dot{x} &= -\nabla f(x) + Bv; \ y = Cx\\ \dot{w} &= Aw + B(y-e) + \xi; \ \lambda = Cw + D(y-e)\\ u &= -\lambda; \ v &= \phi(u); \ \xi = f(x) = K_e \ (u-v) \end{split}$$

Where K_e is the antiwndup gain to be determined.



Primal-dual dyanmics with antiwindup augmentation

Anti-Windup Gain Computation

Below is the reduced LMI for convex systems. This LMI contains LMI variables, parameters derived from the system, and antiwindup gain. This LMI is solved using MATLAB, and the result is used in simulation.

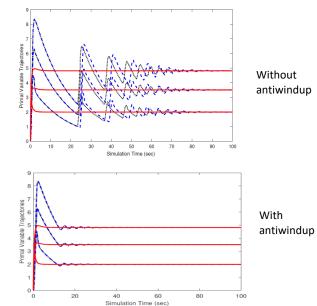
<pre>Q = sdpvar(n,n,'symmetric'); M = sdpvar(nu,nu,'diag'); X = sdpvar(size(Bsigma,2),nu); gamma = sdpvar(1,1);</pre>						
LMI = [[A*Q + Q*A'],	Bw,	Bq*M + Bsig	ma*X + Q*Cu', Q*Cz';			
Bw',	-gamma*eye(4),	Duw',	Dzw';			
M*Bq'+X'*Bsigma'+Cu*Q', Duw,		-2*M,	0;			
Cz*Q,	Dzw,	Ο,	-gamma]	<=0, Ç	⊋>=0, I	M>=0]

Computational Example

A quadratic programming problem:

minimize $\frac{1}{2}(4x_1^2 + 10x_2^2 + 3x_3^2 - 12x_1x_2 - 6x_2x_3 + 2x_1x_3)$ subject to $x_1 + 0.8x_2 + 1.2x_3 \ge 10$

Transforming this into a primal dual dynamic and simulating yields the following solution trajecories with and without antiwindup augmentation.



Conclusion

An application that showcases the design is solving a quadratic program. These problems have extensive applications in control and the speed at which the quadratic program is solved is a vital component to performance. These involve the minimization of a multivariable expression as well as constraints. A systems ability to rapidly solve a quadratic program is indicative of its ability to perform control, so if the proposed anti-windup gain solves quadratic programs, it will also have control applications.

Future Work

Further work with this model would include implementing the model on a physical device. One possible candidate for the device could be the Quasar 2 DOF Helicopter. This system models air or underwater vehicles and would provide a reliable model to test all three controllers.

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