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1 Introduction

The purpose of this paper is to present a general framework for the study of control problems using the behavioral formalism and to specialize such a setting to the \( \ell_1 \) optimal control problem. Especially in the control community, dynamical systems are dominantly viewed as operators acting on inputs and producing output signals. It has been argued in [9, 10] that for many applications in modeling, control and simulation, the traditional input-output framework may not be a natural starting point. Also, the causality structure which is often assumed in feedback configurations imposes constraints on the design of control systems which may not be necessary or which may not correspond to a physical structure. It is a distinguishing feature of the behavioral theory that systems are described in terms of equations rather than input-output operators. Behavioral equations define relationships among system variables in which input and output signals are not necessarily distinguished. System variables are therefore treated in a symmetric way which may have major conceptual advantages for theoretical and practical considerations in control.

Control of dynamical systems concerns the manipulation of a selected set of variables so as to achieve some kind of desirable behavior. Here, by 'selected variables' we will mean a distinguished set of system variables which can be interconnected with a control system. Desirable behavior will be expressed in terms of qualitative or quantitative properties of the system which is obtained by interconnecting plant and controller. For such an interconnection, we will make a crucial distinction between interconnection and external variables. Interconnection variables describe the interaction between plant and controller. These variables have been selected a priori and can be used for control purposes. External variables are the variables by means of which the controlled plant interacts with its environment. Typically, actuator inputs and measured process outputs are interconnection variables; external variables include reference signals, disturbances and to-be-controlled system variables. A plant is viewed as a dynamical system that imposes constraints on both external and interconnection variables. We will view a controller as a set of laws which imposes constraints on the interconnection variables only.

This paper is motivated by earlier work on control in a behavioral context [5, 6, 11, 14], and by papers on \( \mathcal{H}_2 \) and \( \mathcal{H}_\infty \) optimal control in this formalism [8, 12, 13]. The papers [8, 12, 13] concentrate on the full-information case, while in this paper we study the partial-information case since we explicitly specify the variables on which the controller can impose constraints. The \( \ell_1 \) optimal control problem has not been considered in the behavioral formalism before. For the main ideas of \( \ell_1 \) optimal control in the more common input-output framework we refer to [2].

2 System interconnections

The analysis of system interconnections is the core of many problems in modeling, simulation and control. In this section we concentrate on the interconnection of a plant with a controller. We will consider dynamical systems \( \Sigma = (T, W, B) \) with discrete time set \( T = \mathbb{Z}_+ \) or \( T = \mathbb{Z} \), finite dimensional real valued signal spaces \( W = \mathbb{R}^q \), \( q > 0 \) and behaviors \( B \) which are linear shift-invariant and complete subsets of \( W^T \). This class of linear systems will be denoted by \( \mathcal{L}^q \).

It has been shown in [9, 10] that \( \mathcal{L}^q \) admits a parametrization by means of real polynomial matrices with \( q \) columns. Precisely, for every system \( \Sigma \in \mathcal{L}^q \) there exists a polynomial \( R \in \mathbb{R}^{q \times q}[z] \), i.e. a polynomial of the form \( R(z) = \sum_{i=0}^{\ell} R_i z^i \) with \( R_i \) real matrices with \( q \) columns, such that the behavior \( B \) of \( \Sigma \) can...
be written as

\[ B = B_{ssr}(R) := \{ w : \mathbb{Z}^+ \to \mathbb{R}^q \mid R(\sigma)w = 0 \}. \]

Here, \( R(\sigma) \) is to be interpreted as a polynomial operator in the left-shift \( \sigma \) (i.e., \( \sigma w(t) = w(t + 1) \)) acting on the signal space \( (\mathbb{R}^q)^T \).

Let \( \Sigma_p = (T, W, B_p) \) be a dynamical system and suppose that its signal space \( W = \mathbb{R}^q \) is partitioned as

\[ W = W_\ast \times W_i \]

where \( W_\ast = \mathbb{R}^{q\ast} \) is the external signal space and \( W_i = \mathbb{R}^{q_i} \) is a non-empty set called the interconnection space. Here, \( q = q_\ast + q_i \) and \( q_i > 0 \). We refer to \( \Sigma_p \) as the plant. A controller for \( \Sigma_p \) is a dynamical system \( \Sigma_c = (T, W_c, B_c) \) which, when interconnected with \( \Sigma_p \) imposes constraints on the interconnection variables only. We formalize this as follows.

**Definition 2.1** The interconnection of the systems \( \Sigma_p = (T, W_\ast \times W_i, B_p) \) and \( \Sigma_c = (T, W_c, B_c) \) is the system

\[ \Sigma_p \cap \Sigma_c := (T, W_\ast \times W_i, B_p \cap B_c) \]

where

\[ B_p \cap B_c := \{(w_\ast, w_i) \mid (w_\ast, w_i) \in B_p \text{ and } w_i \in B_c \}. \tag{2.1} \]

If \( W_i \) is void then \( \Sigma_p \cap \Sigma_c \) is called a full interconnection.

Note that in a full interconnection \( B_p \cap B_c = B_p \cap B_c \). Further, it is easily seen that the interconnection \( \Sigma_p \cap \Sigma_c \in \mathcal{L}^q \) if \( \Sigma_p \in \mathcal{L}^q \) and \( \Sigma_c \in \mathcal{L}^q \). In what follows, we will view \( \Sigma_c \) as a dynamical system that is (or needs to be) designed to be interconnected with the plant \( \Sigma_p \). The interconnection \( \Sigma_{pc} := \Sigma_p \cap \Sigma_c \) is referred to as the controlled plant.

Not all system interconnections will qualify for the purpose of control. A well-posed interconnection is defined as follows.

**Definition 2.2** Let \( \Sigma_p = (T, W_\ast \times W_i, B_p) \) and \( \Sigma_c = (T, W_c, B_c) \) be two dynamical systems. Their interconnection \( \Sigma_p \cap \Sigma_c \) is said to be well-posed if there exists \( t_0 \in T \) such that

\[ \{(w_\ast, w_i'), (w_\ast, w_i'') \in B_p \cap B_c, w_i'(t) = w_i''(t) \text{ for } t \leq t_0 \} \]

\[ \Rightarrow \{w_i' = w_i'' \}. \]

This means that once the external trajectories \( w_\ast \) in an interconnected system are specified, the set of all interconnection variables \( w_i \) for which \( (w_\ast, w_i) \in B_p \cap B_c \) define an autonomous behavior. In other words, a well-posed interconnection does not allow inputs in the interconnection variables for the interconnected system.

Note that the interconnection variables \( w_i \) do not need to be partitioned in inputs (actuators) and outputs (measurements). This is one of the reasons that we avoid the usage of more classical terminology like ‘feedback’ and ‘closed-loop’ as the causality structure of the interconnection variables is irrelevant in this setting.

Let \( \Sigma_p \in \mathcal{L}^q \) be a dynamical system and suppose that its behavior is represented in polynomial form by

\[ B_p = B_{ssr}(\{(R_\ast, R_i)\}) \]

where \( R_\ast \in \mathbb{R}^{q\ast}[z] \) and \( R_i \in \mathbb{R}^{q_i}[z] \). If \( \Sigma_c \in \mathcal{L}^q \) is a controller, then its behavior can be represented as

\[ B_c = B_{ssr}(R_c) \]

where \( R_c \in \mathbb{R}^{q\ast}[z] \) and the resulting interconnected system admits an autoregressive representation of the form

\[ \begin{bmatrix} R_\ast(\sigma) & R_i(\sigma) \\ 0 & R_c(\sigma) \end{bmatrix} \begin{bmatrix} w_\ast \\ w_i \end{bmatrix} = 0. \tag{2.2} \]

Well-posedness of such an interconnection is easily checked.

**Proposition 2.3** The interconnection of \( \Sigma_p \) and \( \Sigma_c \) is well-posed if and only if the polynomial matrix

\[ \begin{bmatrix} R_\ast \\ R_i \end{bmatrix} \]

is injective as viewed as a matrix over the field of rational functions.

In a well-posed interconnection the controller has to add a minimal number of laws to guarantee that the interconnection variables are uniquely determined once the external variables are fixed. But we do not want to add too many control laws either. In a classical input-output framework a controller should guarantee that the interconnecting variables are uniquely determined once the external variables are fixed but it should not impose any constraints on the external inputs. To guarantee that also in the present setting we do not impose undue constraints on the external variables we introduce the concept of a minimal controller. For this, the input dimension of a dynamical system is the relevant integer invariant to consider.

**Definition 2.4** The complexity of a dynamical system \( \Sigma \in \mathcal{L}^q \) is the pair of integers \( (m(\Sigma), n(\Sigma)) := (m, n) \) which satisfy

\[ \dim(B|[0,t-1]) = mt + n \tag{2.3} \]

for all \( t \geq n \).
The numbers $m(\Sigma)$ and $n(\Sigma)$ are well defined this way and correspond to the number of inputs and the minimal number of states in an input-state-output representation of $\Sigma$.

**Definition 2.5** The system $\Sigma_\epsilon \in L^q$ is said to be a minimal controller for a plant $\Sigma_p \in L^q$ if the interconnection $\Sigma_p \cap \Sigma_\epsilon$ is well-posed and if for any other controller $\Sigma'_\epsilon$ which makes the interconnection well-posed we have

$$m(\Sigma_p \cap \Sigma_\epsilon) \geq m(\Sigma_p \cap \Sigma'_\epsilon).$$

Intuitively $m(\Sigma_p \cap \Sigma_\epsilon)$ is the number of free variables or inputs of the controlled systems and hence we require that we do not reduce the number of free variables any more than necessary to make the controlled system well-posed. A minimal controller therefore adds a minimal number of constraints to obtain a well-posed interconnection. Also minimality of an interconnection is easily checked.

**Proposition 2.6** Assume that the interconnection of $\Sigma_p \in L^q$ and $\Sigma_\epsilon$ in $L^q$ is well-posed. Then the interconnection is minimal if and only if

$$q_e = \text{normrank} R + \text{normrank} R_e.$$

### 3 Control objectives

Let $\Sigma_\epsilon \in L^{n+q}$, $\Sigma_p \in L^q$, and consider the controlled plant $\Sigma_{cp} := (T, \mathbb{R}^{n+q}, B_{cp}) = \Sigma_p \cap \Sigma_\epsilon$. Control objectives are usually specified as functionals defined on specific signals of the controlled plant behavior. Indeed, if initial conditions are known, if disturbances are known to be bounded in magnitude or if reference signals are specified, then only subsets of the controlled plant behavior are relevant for the specification of system performance. These restrictions will be formalized by considering trajectories $w \in B_{cp}$, which satisfy

$$w \in \mathcal{R} \quad (3.1)$$

where $\mathcal{R}$ is a subset of $(\mathbb{R}^q)^T$. Thus, the intersection $B_{cp} \cap \mathcal{R}$ is considered as the relevant set to verify control objectives. A **control objective** is another subset $\mathcal{S}$ of $(\mathbb{R}^q)^T$ which is assumed to be specified either in a qualitative or in a quantitative way. The controlled system $\Sigma_{cp}$ achieves the control objective if its behavior $B_{cp}$ satisfies the inclusion

$$B_{cp} \cap \mathcal{R} \subseteq \mathcal{S}. \quad (3.2)$$

In that case, the controller $\Sigma_\epsilon$ is said to achieve the control objective $\mathcal{S}$ for the plant $\Sigma_p$. We will outline how many different control objectives can be formulated in this framework.

#### 3.1 Stability

External stability can be formulated quite easily. In general external stability depends on an input output structure imposed on the external variables. Suppose therefore that $w_e$ is partitioned as

$$w_e = \begin{pmatrix} d \\ z \end{pmatrix} \quad (3.3)$$

where $d$ denotes an input and $z$ an output component of $w_e$. The classical definition of bounded-input, bounded-output stability then requires that $d \in \ell_{\infty}$ implies $z \in \ell_{\infty}$ in the controlled plant. This is imposed quite easily in our framework by choosing:

$$\mathcal{R} := \{ w = (d, z, w_i) \mid d \in \ell_{\infty} \}$$

$$\mathcal{S} := \{ w = (d, z, w_i) \mid z \in \ell_{\infty} \}$$

Similarly, **internal stability** can be imposed by considering well-posed interconnections and requiring (3.2) with

$$\mathcal{R} := \{ w = (d, z, w_i) \mid d = 0 \}$$

$$\mathcal{S} := \{ w = (d, z, w_i) \mid \lim_{t \to \infty} z(t) = 0, \text{ and } \lim_{t \to \infty} w_i(t) \}.$$
We define the $H_2$ control problem in our framework by considering the control objective (3.2) with

$$\mathcal{R} := \{ w \mid \sum_{t=-\infty}^{t=0} w'(t)Q_1w(t) \leq 1 \}$$

and

$$\mathcal{S} := \{ w \mid \sum_{t=0}^{t=\infty} w'(t)Q_2w(t) \leq 1 \}$$

for suitably chosen positive semi-definite matrices $Q_1$ and $Q_2$. The control objective therefore amounts to bounding a quadratic functional on those future trajectories which are compatible with pasts that satisfy a quadratic norm bound.

In a classical input-output setting this problem basically corresponds to a generalized $H_2$ control problem as studied in e.g. [7]. For non-fixed initial conditions, an $H_2$ problem can be formulated by introducing some measure on the size of the initial condition. With $u$ denoting a control input signal and $x$ the state of a linear, time-invariant system in input-state-output form, one can impose the following measure on initial states

$$m_1(x_0) = \min_u \{ \sum_{t=-\infty}^{-1} u'(t)u(t) \mid x(0) = x_0 \}.$$

For some given initial condition $x_0$ one could consider the following optimization criterion

$$m_2(x_0) = \min_u \{ \sum_{t=0}^{\infty} x'(t)Qx(t) \mid x(0) = x_0 \},$$

where $Q$ is a positive semi-definite matrix. The control objective amounts to finding a controller which achieves that for all $x_0$, $m_1(x_0) \leq 1$ implies $m_2(x_0) \leq 1$. For suitable choices of $Q_1$ and $Q_2$ this problem is equivalent to (3.2) with $\mathcal{R}$ and $\mathcal{S}$ as defined above.

### 3.4 H∞ optimal control

In $H_\infty$ control we impose conditions of the form

$$\|z\|_\infty \leq \|d\|_\infty$$

on the external variables $w$, which are assumed to be partitioned as in (3.3). This condition is obviously equivalent to

$$\sum_{t=0}^{\infty} \begin{pmatrix} (d(t))' & -I \\ x(t) & 0 \end{pmatrix} \begin{pmatrix} d(t) \\ x(t) \end{pmatrix} \leq 0.$$

In other words, we can formulate the $H_\infty$ control problem in our framework by choosing $\mathcal{R}$ and $\mathcal{S}$ as:

$$\mathcal{R} := \ell_2(\mathbb{Z}_+, \mathbb{R}^p)$$

and

$$\mathcal{S} := \{ w \mid \sum_{t=0}^{\infty} w'(t)Qw(t) \leq 0 \}$$

for some suitable chosen matrix $Q$ which will in general be indefinite.

### 4 The $\ell_1$ optimal control problem

$L_1$ optimal control is a more recent development (for extensive references see e.g. [2]). In its usual operator theoretic formulation, the $\ell_1$ optimal control problem amounts to minimizing the $\ell_\infty$ induced norm of a closed-loop operator mapping disturbances to a to-be-controlled output variable.

We will address this problem in a behavioral framework. Recall that the classical $\ell_1$ control problem amounts to finding a (stabilizing) controller for a plant so as to minimize

$$\sup_{d \in \ell_\infty} \|z\|_\infty$$

or so as to achieve that

$$\sup_{d \in \ell_\infty} \|z\|_\infty \leq 1.$$

Here, $d$ and $z$ denote a decomposition of the external variables in an input and an output component as in (3.3). The latter criterion will be formulated in our setting by imposing as an a priori condition that some components of the external variables, say $d$, have $\ell_\infty$ norm less than 1. This should imply that other components of the external variables, say $z$, have $\ell_\infty$ norm less than 1.

In addition, we would like to study the synthesis problem of constructing a controller which achieves this aim. For simplicity we consider in this paper a finite horizon version of this general problem.

Suppose that a polynomial $R \in \mathbb{R}^{p \times q}[z]$ of degree $L$ is given and partitioned as $R = [R_1 \ R_2]$. Let $N > L$ and let $T = [0, N]$ denote a finite time set. Then $R$ defines a finite-time dynamical system $\Sigma_R = (T, \mathbb{R}^q, B_R)$ with behavior

$$B_R := \{ w \mid [R(\sigma)w](t) = 0, t = 0, \ldots, N - L \}.$$

Then $B_R$ is a subspace of the linear space $\mathbb{R}^{q \times (N+1)}$. If we associate with $w : [0, N] \rightarrow \mathbb{R}^{n+6}$ the vector

$$(w_0'(0) \ldots w_N'(0))'$$

then the plant behavior can equivalently be described as

$$B_R = \{ w \mid Cw = 0 \} = \{ (w_s, w_t) \mid C_1 w_s + C_2 w_t = 0 \}$$

(4.1)

where $C$ is a real matrix acting on $\mathbb{R}^{n \times (N-L+1)}$ and is partitioned as $C = [C_1 \ C_2]$ conformally with the partitioning of $R$.

In such a setting, controllers are finite time dynamical systems $\Sigma_C = (T, \mathbb{R}^m, B_C)$ which can be represented by a real matrix $C_3$ acting on $\mathbb{R}^{m \times (N-L+1)}$, i.e.,

$$B_C = \{ w \mid C_3 w = 0 \}$$

(4.2)
The interconnected system $\Sigma_c \cap \Sigma_r$ is well defined in this way and its behavior $B_{\Sigma_c}$ is represented as
\[
\begin{bmatrix}
C_1 & C_2 \\
0 & C_3
\end{bmatrix}
\begin{bmatrix}
w_r \\
w_i
\end{bmatrix} = 0.
\]

The $t_1$ control objective is specified by taking both $S$ and $R$ polyhedral sets in the finite dimensional space $\mathbb{R}^{(N-L+1)}$ (which are not necessarily bounded). We will represent $S$ and $R$ by two real matrices $A_S$ and $A_R$, both having $q_r(N-L+1)$ columns, i.e.
\[
R = \{ w \ | \ A_R w_r \leq 1 \} \quad \text{and} \quad S = \{ w \ | \ A_S w_r \leq 1 \}
\]
where 1 denotes a vector of which each component is equal to 1 and of appropriate dimension. All inequalities should be interpreted element by element.

The $t_1$ control objective is now formulated as in (3.2). The condition (3.2) can be checked via a simple linear programming problem by using the following lemma (see [3, 15]):

**Lemma 4.1 (Farkas lemma)** Let a matrix $A$, column vectors $z$ and $a$, and some real number $c$ be given. Then $a'x < c$ for all $x$ such that $Ax = z$, if and only if

1. there exists a vector $t \geq 0$ such that $A't = a$ and $t'z < c$, or
2. there exists a vector $t \geq 0$ such that $A't = 0$ and $t'z < 0$.

Since our control objective translates into the implication
\[
\begin{bmatrix}
A_R & 0 \\
C_1 & C_2 \\
0 & C_3 \\
-C_1 & -C_2 \\
0 & -C_3
\end{bmatrix}
\begin{bmatrix}
w_r \\
w_i
\end{bmatrix} \leq 1,
\]
application of Farkas lemma yields the following result

**Corollary 4.2** Let $\Sigma_c$ and $\Sigma_r$ be finite time dynamical systems represented by (4.1) and (4.2), respectively. Let $B_{\Sigma_r}$ denote the behavior of the interconnection $\Sigma_c \cap \Sigma_r$. Then the control objective

$B_{\Sigma_c} \cap R \subseteq S$

is satisfied if and only if there exist matrices $T_1, T_2$ and $T_3$ such that:

\[
\begin{align*}
T_1 A_R + T_2 C_1 &= A_S \quad (4.3) \\
T_2 C_2 + T_3 C_3 &= 0 \quad (4.4) \\
T_1 1 &\leq 1 \quad (4.5) \\
T_1 &> 0 \quad (4.6)
\end{align*}
\]

To check for the existence of $T_1, T_2$ and $T_3$ is obviously a simple linear programming problem. This result therefore provides a linear programming type of test to check whether the interconnection of plant and control satisfy the control objective.

On the other hand, for the construction of a controller $\Sigma_r$, we have to make sure that there exists a suitable $C_3$ which guarantees the existence of appropriate $T_1, T_2$ and $T_3$ satisfying the conditions (4.3)-(4.6). We therefore need to search for $T_1$ and $T_2$ satisfying (4.3), (4.5) and (4.6) such that (4.4) is solvable for suitable $C_2$ and $C_3$. Noting that, except for dimensions, $C_3$ and $T_3$ are completely free, we obtain the following theorem:

**Theorem 4.3** There exists a controller $\Sigma_r$ which yields a minimal, well-posed interconnection such that (3.2) is satisfied for the interconnected system if and only if there exists matrices $T_1$ and $T_2$ satisfying (4.3), (4.5) and (4.6) such that

\[
\text{normrank}(T_2 C_2) \leq q_r - \text{normrank} C_2
\]

**Proof.** It is easy to check that to obtain a well-posed and minimal interconnection the number of rows of $C_3$ should be equal to $q_r - \text{normrank} C_2$. It is then easy to check that (4.4) is solvable for suitable $C_2$ and $C_3$ if and only if

\[
\text{normrank}(T_2 C_2) \leq q_r - \text{normrank} C_2.
\]

The condition in the above lemma is not easy to check because of the rank condition. However, it is interesting to note that in a special case we do not need this rank condition:

**Corollary 4.4** Assume that

\[
q_i \geq 2 \text{normrank} C_2.
\]

In that case, there exists a controller which yields a minimal, well-posed interconnection such that (3.2) is satisfied if and only if there exists matrices $T_1$ and $T_2$ satisfying (4.3), (4.5) and (4.6).

**Proof.** If (4.7) is satisfied then

\[
\text{normrank}(T_2 C_2) \leq \text{normrank} C_2 \leq q_r - \text{normrank} C_2.
\]

Hence the rank condition is always satisfied. The result then is an immediate consequence of theorem 4.3.
Remark 4.5 Throughout this paper, the interconnection variables $w_i$ have not been partitioned in input and output components. After a minimal controller has been designed, such a decomposition can be made *a posteriori*. This would yield the usual causality structure of closed-loop configurations in which certain interconnection variables can be identified as inputs (outputs) for the controller and outputs (inputs) for the plant.

Remark 4.6 The results of this section are derived by applying Farkas lemma for polyhedral sets in finite dimensional spaces. This yields a characterization for the $\ell_1$ control objective for finite time systems that has been translated in a controller synthesis procedure. We remark that a version of Farkas lemma for infinite dimensional vector spaces exists. See [1, 4]. A study of the infinite horizon $\ell_1$ control problem is a topic of current research.

5 Conclusion

In this paper we have formulated a structure for control system design in a behavioral setting. Currently we are extending this work for the $\ell_1$ optimal control problem. In particular, we are looking for techniques for the synthesis of minimal $\ell_1$-optimal controllers if the rank condition is needed. Also extensions to the infinite horizon case is a topic of future research.

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