A fractional representation approach to the robust regulation problem for multi-input multi-output systems

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Abstract
The aim of this article is in developing unifying frequency domain theory for robust regulation of multi-input multi-output systems. The main theoretical results achieved are a new formulation of the internal model principle, solvability conditions for the robust regulation problem, and a parameterization of all robustly regulating controllers. The main results are formulated with minimal assumptions and without using coprime factorizations, thus guaranteeing their applicability with a very general class of systems. In addition to theoretical results, the design of robust controllers is addressed. The results are illustrated by two examples involving a delay and a heat equation.

KEYWORDS
control design, factorization approach, feedback control, robust regulation

1 | INTRODUCTION

Controlling behavior of infinite-dimensional systems, for example, systems described by partial differential equations or time-delay systems, is of great interest in many applications. This article studies the frequency domain formulation of the control problem where a dynamic controller is to be found so that the output \( y(t) \) of the system asymptotically converges to the given reference signal \( y_r(t) \), that is, \( \| y(t) - y_r(t) \| \to 0 \) as \( t \to \infty \). The controllers achieving asymptotic convergence are said to be regulating. In addition, the controller is required to work despite small perturbations of the plant. This property is called robustness and it is important in real world control applications since the related system models, the controller design procedures, testing feasibility of the controller by simulations, and finally implementing the controllers in practice unavoidably involve some inaccuracies.\(^1\) The problem of finding a regulating controller that is robust to small perturbations is called the robust regulation problem. As explained by Paunonen and Pohjolainen,\(^2\) the robust regulation problem can be divided into two equally important parts: the robust stabilization and the robust regulation. In the stabilization part one is interested in finding a controller such that stability of the closed loop is maintained despite small perturbations of the plant. It involves the topological aspects of the problem, and it has been studied in the algebraic setting\(^3\)-\(^5\) as well as in many physically relevant algebraic structures\(^6\),\(^7\) This article focuses on the regulation part where one is interested in finding conditions under which stability implies regulation. This characterization of robust regulation has been used, for example, by Paunonen and Pohjolainen\(^2\) in the time domain and by Callier and Desoer\(^8\) in the frequency domain.

Regulation of systems modeled with ordinary differential equations achieved considerable attention in the 1970s.\(^9\)-\(^13\) The results have been generalized to infinite-dimensional systems since then.\(^14\)-\(^22\) A remarkably important result called the internal model principle was given by Francis and Wonham\(^13\) and Davison.\(^10\) It states that all robustly regulating...
controllers must contain an internal model, that is, a suitably reduplicated copy, of the unstable dynamics of the reference signals. The internal model principle has multiple different time domain characterizations. A frequency domain formulation of the internal model principle and solvability conditions for the robust regulation problem were given by Vidyasagar for rational transfer functions. These results were later generalized to specific classes of transfer functions suitable for infinite-dimensional systems and by using the fractional representation approach. Frequency domain methods for designing regulating controllers have also been considered by several authors.

In this article, the robust regulation problem is studied using the fractional representation approach. Fractional representations have two benefits. First, fractional representations allow considerations to be done only assuming that the original proof only stated that the internal model implies regulation. The preliminary results do not take causality into greater detail. In particular, the sufficiency part of the internal model principle now addresses robustness whereas introducing a new reformulation of the internal model principle and discussing the results and presenting their proofs resulted in the conference paper by the author. The solvability of the robust regulation problem, parameterization of robust controllers, or the controller design were not addressed. This article extends the preliminary results of Laakonen and Quadrat to multi-input multi-output (MIMO) systems. Unlike with the SISO systems, regulation does not imply robustness with the MIMO systems making the generalization of the results nontrivial. The results of this article show how the p-copy internal model for time domain systems can be understood within the general framework adopted in this article. MIMO formulation of the internal model principle in the general algebraic framework was first considered in the conference paper by the author. The solvability of the robust regulation problem, parameterization of robust controllers, or the controller design were not addressed. This article extends the preliminary results of Laakonen by introducing a new reformulation of the internal model principle and discussing the results and presenting their proofs in greater detail. In particular, the sufficiency part of the internal model principle now address also robustness whereas the original proof only stated that the internal model implies regulation. The preliminary results do not take causality into account, so the results addressing it are all new.

In addition to theoretical results, several controller design procedures related to the given existence results are proposed. They generalize the ideas for constructing robustly regulating controllers to the general algebraic framework. In addition, a new method of constructing the internal model one element at a time is proposed. It allows revising an already existing controller by including additional parts into its internal model thus extending the class of regulated signals.

Two examples are given to illustrate the proposed controller design procedures and theoretical results. The first example involves a reference signal with an infinite number of unstable poles making the design procedure complicated. This demonstrates how the choice of the ring depends on the problem at hand and underlines the importance of the general approach. In the second example with one dimensional heat equation, it is shown that one may be able to carry out the design procedure using approximations of the plant transfer matrix. This way one does not need to find the closed form of the plant transfer matrix which is a considerable benefit. In addition, the controller of this example is easily verified to be causal even though the controller design method does not directly imply causality. This shows that the results presented in this article that do not directly imply causality are also relevant and interesting.

The remaining part of this article is structured as follows. The preliminary definitions, notations, and stability results are introduced in Section 2. The problem formulation is given in Section 3. Section 4 is devoted to the internal model principle. In Section 5, simplification of the internal model is discussed in term of fractional ideals. Solvability of the robust regulation problem is studied in Section 6. In addition, controller design is addressed and a parameterization of all
robustly regulating controllers is proposed. The theoretical results and design procedures are illustrated by two examples in Section 7. Finally, the obtained results are summarized and discussed in Section 8.

2 NOTATIONS AND PRELIMINARY RESULTS

The set of stable causal SISO transfer functions is denoted by $\mathbb{R}$ and together with the summation $+$ and multiplication $\cdot$ operations it is assumed to be an integral domain, that is, a commutative ring with no zero divisors. The following integral domains appear in the examples. The Hardy space of bounded holomorphic functions in the right half plane $\mathbb{C}_+= \{ s \in \mathbb{C} \mid \text{Re}(s) > 0 \}$ is denoted by $H^\infty$. The set of all real rational functions and its subset of proper rational functions having no poles in $\mathbb{C}_+$ are denoted by $\mathbb{R}(s)$ and $\mathbb{R}H^\infty$, respectively. The integral domain $\mathbb{P}$ consists of functions $f(s)$ that are analytic and bounded in every right-half plane $\mathbb{C}_a = \{ s \in \mathbb{C} \mid \text{Re}(s) > a \}$ with $a > 0$ and polynomially bounded on the imaginary axis, that is, $|f(\text{i} \omega)| < M|\omega|^k$ for some $M, k > 0$. This integral domain corresponds to polynomial stability in the time domain.

The additive and multiplicative identities of $\mathbb{R}$ are denoted by 0 and 1, respectively. Invertible elements of $\mathbb{R}$ are called units. A set $A \subseteq \mathbb{R}$ is an ideal of $\mathbb{R}$, if it is an additive subgroup of $\mathbb{R}$ and $ab \in A$ whenever $a \in A$ and $b \in \mathbb{R}$. An ideal $A$ is prime if $A \neq \mathbb{R}$ and $a \in A$ or $b \in A$ if $ab \in A$. The field of fractions of $\mathbb{R}$ is denoted by $\mathbb{F}_\mathbb{R}$. The $\mathbb{R}$-module $f_1 \mathbb{R} + \cdots + f_n \mathbb{R}$, where $f_1, \ldots, f_k \in \mathbb{F}_\mathbb{R}$, is denoted by $\langle f_1, \ldots, f_k \rangle$ or $\langle f_i \mid i = 1, \ldots, k \rangle$.

Definition 1.

1. An $\mathbb{R}$-submodule $J$ of $\mathbb{F}_\mathbb{R}$ is called a fractional ideal if there exists $0 \neq a \in \mathbb{R}$ such that $aJ \subseteq \mathbb{F}_\mathbb{R}$.
2. A fractional ideal $J$ is finitely generated if $J = \langle f_1, \ldots, f_k \rangle$ for some elements $f_1, \ldots, f_k \in \mathbb{F}_\mathbb{R}$ and it is principal if it is generated by a single element, that is, $J = \langle f \rangle$ for some $f \in \mathbb{F}_\mathbb{R}$.

A matrix $H$ with elements $h_{ij}$ on the $i$th row and $j$th column is denoted by $H = (h_{ij})$ and its transpose is denoted by $H^T$. The set of all matrices with elements in a set $S$ is denoted by $\mathcal{M}(S)$ and the set of all $n \times m$ matrices by $S^{n \times m}$. The set of $n$-dimensional vectors with elements in $S$ is denoted by $S^n$.

The plant and the controller are assumed to be matrices over the field of fractions $\mathbb{F}_\mathbb{R}$. The control configuration considered in this article is depicted in Figure 1. The resulting closed loop transfer matrix from $(y_r, d)$ to $(e, u)$ is

$$H(P, C) := \begin{bmatrix} (I - PC)^{-1} & (I - PC)^{-1}P \\ C(I - PC)^{-1} & (I - CP)^{-1} \end{bmatrix} \in \mathcal{M}(\mathbb{F}_\mathbb{R}).$$

(1)

The transfer functions $(I - PC)^{-1}$ and $(I - PC)^{-1}P$ are called the sensitivity matrix and the load disturbance sensitivity matrix, respectively.

Remark 1. The choice $e = y + y_r$ is made in Figure 1 instead of the more intuitive $e = y - y_r$. It does not restrict the generality, and one can avoid some technical difficulties this way because the closed loop is then symmetric with respect to $P$ and $C$.

Definition 2.

1. A matrix or a vector $H \in \mathcal{M}(\mathbb{F}_\mathbb{R})$ is stable if $H \in \mathcal{M}(\mathbb{R})$, and otherwise it is unstable.
2. A controller $C \in \mathbb{F}_\mathbb{R}^{m \times n}$ stabilizes $P \in \mathbb{F}_\mathbb{R}^{n \times m}$ if the closed loop transfer matrix (1) is stable.

![Figure 1](image-url)
Remark 2. If $\mathbf{R} = H^\infty$, then Definition 2 of stability means that an input signal in the Lebesgue space $L^2(0, \infty, \mathbb{C}^m)$ produces $L^2(0, \infty, \mathbb{C}^n)$-output signal in the time-domain. Furthermore, under the standard assumptions that the system operator of the state-space representation generates a strongly stable semigroup, this means that the output converges to zero.\textsuperscript{35}

Definition 3.

1. The representation $H = ND^{-1}(H = D^{-1}N)$ is called a right (left) factorization of $H$ if $N, D \in \mathcal{M}(\mathbf{R})$ ($N, D \in \mathcal{M}(\mathbf{R})$) and $\det(D) \neq 0$ ($\det(D) \neq 0$).
2. The factorization $H = ND^{-1}(H = D^{-1}N)$ is right (left) weakly coprime if for any $X \in \mathcal{M}(\mathbf{F}_\mathbf{R})$ ($X \in \mathcal{M}(\mathbf{R})$) of suitable size one has

$$
\begin{bmatrix}
N \\
D
\end{bmatrix}
X \in \mathcal{M}(\mathbf{R}) \Rightarrow X \in \mathcal{M}(\mathbf{R}) \quad \left(X \begin{bmatrix}
\tilde{N} \\
\tilde{D}
\end{bmatrix} \in \mathcal{M}(\mathbf{R}) \Rightarrow \tilde{X} \in \mathcal{M}(\mathbf{R})\right).
$$

3. The factorization $H = ND^{-1}(H = D^{-1}N)$ is right (left) coprime if there exist $X, Y \in \mathcal{M}(\mathbf{R})$ ($X, Y \in \mathcal{M}(\mathbf{R})$) such that

$$
XD - YN = I (\tilde{D}\tilde{X} - \tilde{N}\tilde{Y} = I).
$$

Any right (left) coprime factorization is a weakly right (left) coprime factorization. It follows that the results assuming weakly coprime factorizations are valid if the factorization is coprime. In general, weakly coprime factorizations need not be coprime. However, a weakly right (left) coprime factorization of a stabilizing controller $C$ or a stabilizable plant $P$ is right (left) coprime.\textsuperscript{36}

In what follows the stability results given in the next theorem are used extensively. The first item is Theorem 3 of Quadrat\textsuperscript{39} reformulated using Proposition 4 of the same article. It gives a parameterization of all stabilizing controllers. The second part is obtained from the first one by changing the roles of $P$ and $C$. It holds by the symmetry of the closed loop control configuration of Figure 1.

Theorem 1. Let $C \in \mathbf{F}_\mathbf{R}^{m \times n}$ stabilize $P \in \mathbf{F}_\mathbf{R}^{m \times n}$. 

1. Denote

$$
\tilde{L} := \begin{bmatrix}
-(I - CP)^{-1}C & (I - CP)^{-1}
\end{bmatrix} \quad \text{and} \quad L := \begin{bmatrix}
(I - PC)^{-1}
\end{bmatrix}.
$$

All stabilizing controllers of $P$ are parameterized by

$$
C(W) = \left(C(I - PC)^{-1} + \tilde{L}WL\right) \left((I - PC)^{-1} + P\tilde{L}WL\right)^{-1} = \left((I - CP)^{-1} + \tilde{L}WLP\right)^{-1} \left((I - CP)^{-1}C + \tilde{L}WL\right),
$$

where the stable matrix $W \in \mathbf{R}^{(m+n) \times (m+n)}$ is such that $\det((I - PC)^{-1} + P\tilde{L}WL) \neq 0$ and $\det((I - CP)^{-1} + \tilde{L}WLP) \neq 0$.

2. Denote

$$
\tilde{M} := \begin{bmatrix}
-(I - PC)^{-1}P & (I - PC)^{-1}
\end{bmatrix} \quad \text{and} \quad M := \begin{bmatrix}
(I - CP)^{-1}
\end{bmatrix}.
$$

All plants that $C$ stabilizes are parameterized by

$$
P(X) = \left(P(I - CP)^{-1} + \tilde{M}XM\right) \left((I - CP)^{-1} + C\tilde{M}XM\right)^{-1} = \left((I - PC)^{-1} + \tilde{M}XMC\right)^{-1} \left((I - PC)^{-1}P + \tilde{M}XM\right)
$$

where the stable matrix $X \in \mathbf{R}^{(m+n) \times (m+n)}$ is such that $\det((I - CP)^{-1} + C\tilde{M}XM) \neq 0$ and $\det((I - PC)^{-1} + \tilde{M}XMC) \neq 0$. 

Next two lemmas are given for later use. The first one is Theorem 5.3.10 of Vidyasagar. The original proof of the theorem uses coprime factorizations, but it can easily be extended to the more general setting of this article.

**Lemma 1.** If \( C_s \) stabilizes \( P \) and \( C_0 \) stabilizes \( P_0 = P(I - C_s P)^{-1} \), then \( C_s + C_0 \) stabilizes \( P \).

**Proof.** The proof of the lemma is similar to the original proof by Vidyasagar. The only change required is to replace the arguments using coprime factorizations that show stability of \((I - (C_s + C_0)P)^{-1} C_s \). To this end, observe that

\[
(I - (C_s + C_0)P)^{-1} C_s = (I - C_s P)^{-1} (I - C_0 P)^{-1} C_s = (I - C_s P)^{-1} C_s + (I - C_s P)^{-1} C_0 (I - P_0 C_0)^{-1} P_0 C_s.
\]

This shows the claim since \((I - C_s P)^{-1} C_s, C_0 (I - P_0 C_0)^{-1} P_0 C_s \in \mathcal{M}(\mathbb{R})\) by the assumptions that \( C_s \) stabilizes \( P \) and \( C_0 \) stabilizes \( P_0 \).

**Lemma 2.** If \( X, Y, Z \in \mathcal{M}(\mathbb{R}) \) and \( \theta \in \mathbb{F}_R \) are such that \( I = \theta X - YZ \in \mathbb{F}_R^{\text{dixon}} \), then there exist \( \bar{X}, \bar{Z} \in \mathcal{M}(\mathbb{R}) \) such that \( \bar{X} \) is invertible over \( \mathbb{F}_R \) and \( I = \theta \bar{X} - YZ \).

**Proof.** The columns of \( X \) and \( YZ \) together span \( \mathbb{F}_R^n \). Thus, there exists a basis \( \{x_1, \ldots, x_k, h_{k+1}, \ldots, h_n\} \) of \( \mathbb{F}_R^n \) where \( x_1, \ldots, x_k \) are columns of \( X \) forming the basis of its column space and \( h_{k+1}, \ldots, h_n \) are columns of \( YZ \). For notational simplicity assume that \( x_1, \ldots, x_k \) are the first \( k \) columns of \( X \). Denote \( \theta = \frac{n}{d} \) where \( n, d \in \mathbb{R} \). Choose a matrix \( M \) that selects the columns \( h_{k+1}, \ldots, h_n \) of \( YZ \) so that

\[
X + dYZM = \left[ x_1, \ldots, x_k, x_{k+1} + dh_{k+1}, \ldots, x_n + dh_n \right] \in \mathcal{M}(\mathbb{R})
\]

The columns of \( X + dYZM \) are linearly independent and consequently it is invertible over \( \mathbb{F}_R^{\text{dixon}} \). Furthermore,

\[
I = \theta(X + dYZM) - Y(nZM + Z),
\]

which shows the claim.

Causality is the natural constraint that a system cannot depend on the future inputs. This can be considered in the chosen framework by following the approach by Vidyasagar et al. To this end, a prime ideal \( \mathbb{Z} \) of \( \mathbb{R} \) is fixed for the remaining part of this article, and it is assumed that any causal transfer function has a factorization whose denominator is not in \( \mathbb{Z} \). This leads to the following definition.

**Definition 4.**

1. A transfer function \( p \in \mathbb{F}_R \) is causal if it has a factorization \( p = \frac{n}{d} \) such that \( n \in \mathbb{R} \) and \( d \in \mathbb{R} \setminus \mathbb{Z} \).
2. A transfer function \( p \in \mathbb{F}_R \) is strictly causal if it has a factorization \( p = \frac{n}{d} \) such that \( n \in \mathbb{Z} \) and \( d \in \mathbb{R} \setminus \mathbb{Z} \).
3. A transfer matrix \( P \in \mathcal{M}(\mathbb{F}_R) \) is (strictly) causal if all of its elements are (strictly) causal.
4. A square transfer matrix \( P \in \mathcal{M}(\mathbb{F}_R) \) is \( \mathbb{Z} \)-nonsingular if its determinant is in \( \mathbb{R} \setminus \mathbb{Z} \).

It is natural to assume that the plants and controllers discussed are causal. However, the general framework chosen allows a causal plant to have noncausal stabilizing controllers, that is, the parameterized stabilizing controllers or plants stabilized by a controller in Theorem 1 may be noncausal. The approach adopted here is the same as by Mori and Abe, that is, noncausal plants and controllers are allowed in the general theoretical framework and causality issues are discussed separately. To the best of author’s knowledge, the most comprehensive results concerning the existence of causal controllers within the general framework are those of Propositions 6.1 and 6.2 in Mori and Abel. These results are recalled in the following theorem for later use.

**Theorem 2.**

1. If a causal plant is stabilizable, it has a causal stabilizing controller.
2. All stabilizing controllers of a strictly causal plant are causal.
Remark 3. Results similar to those stated in Theorem 2 are given by Vidyasagar et al.\textsuperscript{5} and Sule.\textsuperscript{37} In particular, Vidyasagar et al. give the second item of the theorem assuming that the plant has both left and right coprime factorizations, but dropping out the requirement that the ideal \( Z \) is prime.

Before proceeding to the problem formulation, a lemma concerning weakly coprime factorizations of causal transfer functions is given. It states that the denominator of any weakly coprime factorization of a causal transfer function in \( R \setminus Z \). An arbitrary factorization does not have this property. For example, given a factorization \( \frac{n}{d} \) and \( k \in Z \), then \( \frac{n}{d} k \) is a factorization such that \( dk \in Z \).

**Lemma 3.** If \( 0 \neq p \in F_R \) is a causal transfer function and \( \frac{n}{d} \) its weakly coprime factorization, then \( d \in R \setminus Z \).

**Proof.** Since \( p \) is causal, it has a factorization \( \frac{n_0}{d_0} \) where \( n_0 \in R \) and \( d_0 \in R \setminus Z \). The numerator \( n \neq 0 \) since \( p \neq 0 \), so \( \frac{n_0}{n} n = n_0 \in R \) and \( \frac{n}{d} d_0 = d_0 \in R \). This implies that \( \frac{n_0}{n} \in R \) since \( \frac{n}{d} \) was assumed to be a weakly coprime factorization. The claim follows since \( \frac{n}{d} d = d_0 \in R \setminus Z \) and all factors of an element in \( R \setminus Z \) belong to that set as well.\textsuperscript{32}

3 | PROBLEM FORMULATION

Consider the control configuration of Figure 1 where \( P \in F_{R}^{n \times m} \) and \( C \in F_{R}^{m \times n} \). The reference signals are assumed to be generated by a fixed signal generator \( \Theta_r \in F_{R}^{m \times q} \), that is, they are of the form \( y_r = \Theta_r y_0 \) where \( y_0 \in R^q \). This article is focused on regulation, so it is assumed that the disturbance signals contain only unstable dynamics that are already present in the signal generator. In other words, it is assumed that the disturbance signals are of the form \( d = \Theta_d d_0 \) where the vector \( d_0 \in R^q \) and \( \Theta_d = Q \Theta_r \in F_{R}^{m \times q} \) for some fixed matrix \( Q \in R^{m \times n} \). It is assumed throughout this article that the given nominal plant \( P \) and the signal generator \( \Theta_r \) are causal.

**Definition 5.**

1. The controller \( C \) regulates \( P \) if \( (I - PC)^{-1} \Theta_r y_0 \in R^n \) for all \( y_0 \in R^q \). Furthermore, a controller \( C \) robustly regulates \( P \) if
   (i) it stabilizes \( P \), and
   (ii) regulates every plant it stabilizes.
2. The controller \( C \) is causally robustly regulating for \( P \) if
   (i) it stabilizes \( P \), and
   (ii) regulates every causal plant it stabilizes.
3. A controller \( C \) is disturbance rejecting for \( P \) if \( (I - PC)^{-1} P \Theta_d d_0 \in R^n \) for all \( d_0 \in R^q \). Furthermore, a controller \( C \) is robustly disturbance rejecting for \( P \) if
   (i) it stabilizes \( P \), and
   (ii) is disturbance rejecting for every plant it stabilizes.

The problem of finding a controller \( C \) that (causally) robustly regulates a given nominal plant \( P \) is called the (causal) robust regulation problem.

**Remark 4.** By Remark 2, the time domain interpretation of regulation in Definition 5 is that the error between the output and the reference signal converges asymptotically to zero, that is, \( \| y(t) - y_r(t) \| \to 0 \) as \( t \to \infty \), for all reference signals.

**Remark 5.** Note that no topology is needed in the above definitions, and robustness of regulation simply means that stability of the closed loop implies regulation. This definition is justified since stability of the closed loop is a robust property under very general assumptions in the algebraic framework.\textsuperscript{3,5} This means that a controller solving the robust regulation problem regulates all plants sufficiently near the given plant \( P \).

**Remark 6.** It is not assumed here that the controller \( C \) is causal. This is because the parameterized controllers in Theorem 1 may be noncausal and some of the existence results are based on the theorem. However, for strictly causal plants such an assumption is unnecessary by Theorem 2. For causal plants the situation is more complicated and will be discussed later.

**Remark 7.** The plants appearing in applications can be assumed to be causal, so the causal robust regulation problem is the more natural one of the two problems posed here. It is obvious that all robustly regulating controllers are also
causally robustly regulating, so the more general theory involving also noncausal plants will produce controllers with the desired properties and is worth studying. On the other hand, a controller may stabilize also noncausal plants, so the robust regulation problem may lead to slightly too strong requirements for the controller. It will be shown in the next section that a strictly causal controller that is causally robustly regulating is also robustly regulating.

4 | THE INTERNAL MODEL PRINCIPLE

The aim of this section is to extend the classical frequency domain formulation given by Vidyasagar\(^7\) using fractional representations. The next theorem is a major step towards that goal. It shows that the unstable dynamics generated by any element of the signal generator \(\Theta_r\) must be blocked by every element of the sensitivity and load disturbance sensitivity matrices. Thus, it does not matter if some unstable dynamics appear, for example, only in the first element of any reference signal. For systems described by rational matrices this corresponds to the fact that the closed loop transfer matrix must vanish completely at each pole of the signal generator located in the closed right half plane \(\bar{C}_+\).

**Theorem 3.** A stabilizing controller \(C\) is robustly regulating for \(P\) with the signal generator \(\Theta_r = (\theta_{ij})\) if and only if for all \(i \in \{1, \ldots, n\}\) and \(j \in \{1, \ldots, q\}\)

\[
\theta_{ij} \left[ -(I - PC)^{-1} P (I - PC)^{-1} \right] \in \mathcal{M}(R). \tag{4}
\]

**Proof.** Denote \(M = \left( (I - CP)^{-1} (P(I - CP)^{-1} )^T \right)^T, U = (I - PC)^{-1}, \text{and} \ M = \left[ -UP \quad U \right]. \) These matrices are stable since \(C\) stabilizes \(P\). Since \(C\) stabilizes \(P\), the second part of Theorem 1 gives the parameterization of plants \(X\) the controller \(C\) stabilizes and

\[
(I - PC)^{-1} \Theta_r = (U + MXM - (UP + MXM) C)^{-1} (U + MXM) \Theta_r \tag{5a}
\]

\[
= (U - UPC)^{-1} (U + MXM) \Theta_r \tag{5b}
\]

\[
= U \Theta_r + MXM \Theta_r, \tag{5c}
\]

where \(\det (I - PC)^{-1} + MXM \neq 0\).

The sufficiency is shown first. Assume that (4) holds. Observe that the reference signal \(y_r = \Theta_r y_0\) with an arbitrary stable vector \(y_0\) of suitable size can be written in the form \(y_r = \sum_{ij} \theta_{ij} y_{ij}\) where \(y_{ij}\) are stable vectors. Substituting this into (5) yields

\[
(I - PC)^{-1} \Theta_r y_0 = \sum_{ij} \theta_{ij} U y_{ij} + \sum_{ij} \theta_{ij} MXM C y_{ij}. \tag{5d}
\]

This vector is stable since \(\theta_{ij} M, \theta_{ij} U \in \mathcal{M}(R)\) by (4) and \(MC \in \mathcal{M}(R)\) since \(C\) stabilizes \(P\). Thus, \(C\) is robustly regulating.

The necessity is shown next. Assume that \(C\) robustly regulates \(P\). Since \(C\) regulates all the plants it stabilizes, the matrix in (5) is stable for all matrices \(X\). Choosing \(X = 0\) yields \(U \Theta_r \in \mathcal{M}(R)\). This and (5) imply that \(MXM \Theta_r \in \mathcal{M}(R)\).

In particular,

\[
M \begin{bmatrix} 0 & 0 \\ 0 & X_0 \end{bmatrix} MC \Theta_r = \begin{bmatrix} -UP & U \\ 0 & X_0 \end{bmatrix} \begin{bmatrix} (I - CP)^{-1} \\ P(I - CP)^{-1} \end{bmatrix} C \Theta_r
\]

\[
= UX_0 P(I - CP)^{-1} C \Theta_r
\]

\[
= UX_0 (I - PC)^{-1} PC \Theta_r
\]

\[
= UX_0 (U - I) \Theta_r \in \mathcal{M}(R)
\]

if \(\det (U + UX_0 (U - I)) \neq 0\). Since \(U \Theta_r \in \mathcal{M}(R)\), it follows that \(UX_0 \Theta_r \in \mathcal{M}(R)\).

Choose \(X_0 = e_k e_i^T\) where \(e_i\) is the \(i\)th natural basis vector of \(F^*_R\). If \(\det (U + UX_0 (U - I)) \neq 0\) for some \(i, k \in \{1, \ldots, n\}\), then \(U e_k e_i \Theta_r \in \mathcal{M}(R)\). If for some \(i, k \in \{1, \ldots, n\}\)

\[
\det (U + U e_k e_i^T (U - I)) = \det (U) \det (I + e_k e_i^T (U - I)) = 0,
\]
then $\det(I + \mathbf{e}_k\mathbf{e}_k^T(U - I)) = 0$ since $\det(U) \neq 0$ by the stability of the closed loop system. Write $K := \mathbf{e}_k\mathbf{e}_k^T(U - I) = gh^T$ where $\mathbf{g}, \mathbf{h} \in \mathbb{F}_\mathbb{R}^n$. This is possible since $K$ is a rank one matrix. By the matrix determinant lemma, $\det(I + K) = 1 + h^Tg = 0$. This can happen only if $h^Tg = -1$, so choosing $X_0 = 2\mathbf{e}_k\mathbf{e}_k^T$ one has $\det(U + UX_0(U - I)) \neq 0$ and $U\mathbf{e}_k\mathbf{e}_k\Theta_r \in \mathcal{M} (\mathbf{R})$.

It has been shown that $U\mathbf{e}_k\mathbf{e}_k\Theta_r \in \mathcal{M} (\mathbf{R})$ for all $k, i \in \{1, \ldots, n\}$. This means that the $k$th column of $U$ multiplied by any element in the $i$th row of $\Theta_r$ must be stable. Thus, $\theta_y(I - PC)^{-1} \in \mathcal{M} (\mathbf{R})$ for all possible $i \in \{1, \ldots, n\}$ and $j \in \{1, \ldots, q\}$.

In order to complete the proof, choose $X = \begin{bmatrix} 0 & X_0 \\ 0 & 0 \end{bmatrix}$ in (5). Similar arguments as above show that $\theta_y(I - PC)^{-1}P \in \mathcal{M} (\mathbf{R})$, for all $i \in \{1, \ldots, n\}$ and $j \in \{1, \ldots, q\}$ so (4) follows.

Remark 8. Similar arguments as in the sufficiency part of the above proof show that (4) implies that $(I - P(X)C)^{-1}P(X)\Theta_d\theta_0 \in \mathcal{M} (\mathbf{R})$ for all suitable $X$. This means that (4) implies robust disturbance rejection as well. Thus, all robustly regulating controllers are also robustly disturbance rejecting. On the other hand, it was observed in Example 5.4 of 

Example 1. The condition (4) of Theorem 3 is equivalent of $C$ being regulating and disturbance rejecting for $P$ with all reference and disturbance signals of the form $y_r = \theta_y\mathbf{e}_k$ and $d = \theta_y\mathbf{e}_h^T$, where $\mathbf{e}_k$ and $\mathbf{e}_h^T$ are arbitrary natural basis vector of $\mathbb{F}_\mathbb{R}^n$ and $\mathbb{F}_\mathbb{R}^m$, respectively. This can be used to test if a controller is robustly regulating. For example, if one wants to find out if a controller $C$ achieves robust regulation for the single reference signal $y_r = (0, \theta, 0, 0, \ldots, 0) \in \mathbb{F}_\mathbb{R}^n$, then one needs to test if $C$ is regulating for all reference signals of the form $\theta_d\mathbf{e}_k$ with $k \in \{1, \ldots, n\}$ and disturbance rejecting for all disturbance signals $\theta_y\mathbf{e}_h^T$ with $h \in \{1, \ldots, m\}$.

The next theorem is the first main result of this article. It is a reformulation of the famous internal model principle using no coprime factorizations. It states that all the unstable dynamics produced by the signal generator must be built into the controller as an internal model in order to make it robustly regulating. It generalizes Theorem 3.2 of Laakkonen and Quadrat from SISO systems to MIMO systems. In Section 5, the new formulation is confirmed to be equivalent to the classical one of Vidyasagar when coprime factorizations exist.

**Theorem 4.** Denote $\Theta_r = (\theta_y)$. Controller $C$ solves the robust regulation problem for $P$ if and only if it stabilizes $P$ and for all $i \in \{1, \ldots, n\}$ and $j \in \{1, \ldots, q\}$ there exist $A_{ij}, B_{ij} \in \mathcal{M} (\mathbf{R})$ such that

$$\theta_yI = A_{ij} + B_{ij}C. \tag{6}$$

**Proof.** Denote $\tilde{M} = \begin{bmatrix} -(I - PC)^{-1}P & (I - PC)^{-1} \end{bmatrix}$. If $\theta_yI = A_{ij} + B_{ij}C$, it follows that $\theta_y\tilde{M} = A_{ij}\tilde{M} + B_{ij}C\tilde{M} \in \mathcal{M} (\mathbf{R})$. The matrices $\tilde{M}$ and $C\tilde{M}$ on the right-hand side of the equation are stable since $C$ stabilizes $P$. The controller $C$ is robustly regulating by Theorem 3. This shows sufficiency. The necessity follows since

$$\theta_yI = \theta_y(I - PC)^{-1}(I - PC) = \theta_y(I - PC)^{-1} - \theta_y(I - PC)^{-1}PC,$$

where $A_{ij} = \theta_y(I - PC)^{-1}$ and $B_{ij} = -\theta_y(I - PC)^{-1}P$ are stable matrices by Theorem 3 when $C$ is robustly regulating.

As mentioned in Remark 7, the definition requires a robustly regulating controller to regulate even the noncausal plants it stabilizes. By changing the roles of the plant and controller in Theorem 2, one has that a strictly causal controller stabilizes only causal plants. Combining it with the above theorem, the following corollary is obtained.

**Corollary 1.** A strictly causal stabilizing controller $C$ is causally robustly regulating for $P$ with the signal generator $\Theta_r = (\theta_y)$ if and only if it satisfies the two equivalent conditions (4) and (6) for all elements $\theta_y$.

**Remark 9.** Theorems 3 and 4 are necessary and sufficient conditions for a stabilizing controller to be robustly regulating. Thus, they can be thought as alternative characterizations of the internal model principle. However, only the latter one directly deals with the structure of the controller whereas the first one states a property of the closed loop system. Therefore the latter one is considered as the internal model principle in this article.

**Example 2.** For SISO plants, Theorem 4 takes the form $\langle \Theta_r \rangle \subseteq \langle 1, C \rangle$. The inclusion indicates that the signals generated by the generator can be divided into a stable part and an unstable part generated by the controller. This makes sense, since only the unstable dynamics need to be regulated by $C$. 

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Example 3. Consider linear systems described by ordinary differential equations with \( n \) inputs and outputs. The natural choice for the ring of stable transfer functions is \( \mathbb{R} = RH^\infty \) and then \( \mathbf{F}_{RH^\infty} = \mathbb{R}(s) \). The causal plants in \( \mathbb{R}(s) \) are the proper rational matrices. Such a matrix is stable if and only if it has no poles in the closed right half-plane \( \mathbb{C}_+ \), so the unstable dynamics of the reference signals are characterized by the poles in \( \mathbb{C}_+ \).

Consider a signal generator \( \Theta_r = (\theta_{ij}) \in \mathbf{F}_{RH^\infty}^{\infty x n} \). The condition (6) means that if \( \theta_{ij} \) has a pole of order \( k \) at \( z \in \mathbb{C}_+ \), then any robustly regulating controller \( C \) must have a pole of the same or higher order at \( z \). Furthermore, one can write \( C = UAV \) where \( \Lambda = \text{diag}(h_1, \ldots, h_n) \in \mathbf{F}_{RH^\infty}^{\infty x n} \) is the Smith–McMillan form of \( C \) and \( U \) and \( V \) are invertible over \( RH^\infty \).^7 Then (6) holds if and only if each of the \( n \) diagonal elements \( h_l \) in the Smith–McMillan form have a pole of order \( k \) or higher at \( z \). This corresponds to the well-known fact that a robustly regulating controller must contain a \( n \)-folded copy of any unstable dynamics of reference signals.^2,^13

5 | SIMPLIFICATION OF THE INTERNAL MODEL

The results of the previous section revealed that the unstable dynamics of each element in the signal generator must be included into a robustly regulating controller. The next theorem shows that the internal model is characterized by the fractional ideal generated by the elements of the signal generator. The theorem provides a way to characterize the internal model in a compact form as illustrated in Example 5.

Theorem 5. Let \( C \) stabilize \( P \). Consider \( \Theta_r = (\theta_{ij}) \) and the fractional ideal \( I = \langle \theta_{ij} \rangle 1 \leq i \leq n, 1 \leq j \leq q \).

1. If \( J \subset \langle f_1, \ldots, f_k \rangle \) and there exist \( A_l \) and \( B_l \) such that \( f_lA_l = A_l + B_lC \) for all \( l = 1, \ldots, k \), then \( C \) is robustly regulating.
2. If \( \langle f_1, \ldots, f_k \rangle \subset J \) and \( C \) is robustly regulating, then there exist \( A_l \) and \( B_l \) such that \( f_lA_l = A_l + B_lC \) for all \( l = 1, \ldots, k \).

Proof. Only the first part is proved. The second part can be proved using similar arguments. It is assumed that \( J \subset \langle f_1, \ldots, f_k \rangle \) and that there exist \( A_l \) and \( B_l \) such that \( f_lA_l = A_l + B_lC \) for all \( l = 1, \ldots, k \). Now \( \theta_{ij} \in \langle f_1, \ldots, f_k \rangle \) or equivalently \( \theta_{ij} = a_{ij}f_1 + \cdots + a_{ij}f_k \) for some \( a_{ij} \in \mathbb{R} \). Consequently,

\[
\theta_{ij}I = \sum_{l=1}^{k} a_{ij}f_l I = \sum_{l=1}^{k} a_{ij}(A_l + B_lC) = \left( \sum_{l=1}^{k} a_{ij}A_l \right) + \left( \sum_{l=1}^{k} a_{ij}B_l \right) C,
\]

where \( \sum_{l=1}^{k} a_{ij}A_l \) and \( \sum_{l=1}^{k} a_{ij}B_l \) are stable matrices. Since \( \theta_{ij} \) is an arbitrary element of \( \Theta_r \), the result follows by Theorem 4. \( \blacksquare \)

The above theorem shows that the instability generated by \( \Theta_r = (\theta_{ij}) \) is captured by the fractional ideal \( J \) generated by the elements \( \theta_{ij} \). In particular, if \( J \) is principal and \( I = \langle \theta \rangle \) where \( \theta \in \mathbf{F}_{R} \), then a stabilizing controller is robustly regulating if and only if there exist stable \( A \) and \( B \) such that \( \theta I = A + BC \). This in particular means that if \( \mathbb{R} \) is a Bezout domain, that is, a domain where all finitely generated ideals are principal, the required internal model is always characterized by a single element of \( \mathbf{F}_{R} \).

Example 4. The integral domain \( RH^\infty \) is a principal ideal domain.\(^7\) Consequently, it is a Bezout domain and the internal model is always captured by a single rational function in the field \( \mathbf{F}_{RH^\infty} = \mathbb{R}(s) \) if \( \mathbb{R} = RH^\infty \). Other common rings in system theory, for example, the Hardy space \( H^\infty \) or the convolution algebra \( A(\beta) \),\(^40\) are not typically Bezout. Consequently, there are signal generators generating instability that cannot be captured by any single fraction over the ring.

Example 5. Choose \( \mathbb{R} = H^\infty \) and consider the signal generator

\[
\Theta_r(s) = \begin{bmatrix}
\frac{1}{s^2 - 1} & 0 \\
0 & \frac{1}{s + \frac{1}{s^2 - 1}}
\end{bmatrix}.
\]

Then the reference signals are of the form

\[
y_r(s) = \begin{bmatrix}
\frac{1}{s^2 - 1} & 0 \\
0 & \frac{1}{s + \frac{1}{s^2 - 1}}
\end{bmatrix} \begin{bmatrix} a(s) \\ b(s) \end{bmatrix} = \begin{bmatrix}
\frac{1}{s^2 - 1}a(s) \\
\left( \frac{1}{s + \frac{1}{s^2 - 1}} \right) b(s)
\end{bmatrix}.
\]
where \(a(s), b(s) \in H^\infty\). The inclusion 
\[
\begin{pmatrix} (s+1)^2 \\
(e^{-s}-1)(s^2+\pi^2) \end{pmatrix} \subseteq \begin{pmatrix} 1 \\
\epsilon^{-s}-1 + \frac{1}{s^2+\pi^2} \end{pmatrix}
\]
holds since
\[
\begin{align*}
\frac{(s+1)^2}{(e^{-s}-1)(s^2+\pi^2)} &= \frac{1}{e^{-s}-1} a(s) + \left( \frac{1}{s} + \frac{1}{s^2+\pi^2} \right) b(s),
\end{align*}
\]
where
\[
\begin{align*}
\beta(s) &= \frac{3\pi^2 - 1 + s(\pi^2 - 3)}{2(s+1)} \in H^\infty,
\alpha(s) &= \frac{(s+1)^2}{s^2+\pi^2} - \beta(s) \left( \frac{e^{-s}-1}{s} + \frac{e^{-s}-1}{s^2+\pi^2} \right) = \frac{2(s+1)^3}{2(s+1)(s^2+\pi^2)} - \beta(s) \frac{e^{-s}-1}{s} \in H^\infty.
\end{align*}
\]

Note that \(a(s)\) is stable due to the pole-zero cancellations at \(\pm \pi i\) and 0. Similarly, 
\[
\begin{pmatrix} 1 \\
\epsilon^{-s}-1 + \frac{1}{s^2+\pi^2} \end{pmatrix} \subseteq \begin{pmatrix} (s+1)^2 \\
(e^{-s}-1)(s^2+\pi^2) \end{pmatrix}
\]

Thus,
\[
\begin{pmatrix} (s+1)^2 \\
(e^{-s}-1)(s^2+\pi^2) \end{pmatrix} \subseteq \begin{pmatrix} 1 \\
\epsilon^{-s}-1 + \frac{1}{s^2+\pi^2} \end{pmatrix}.
\]  

(7)

This means that the overall instability generated by the signal generator is captured by the single element \(\theta(s) = \frac{(s+1)^2}{(e^{-s}-1)(s^2+\pi^2)}\). This can be verified by observing that the pole locations and orders of \(\theta(s)\) are exactly the same that appear in the nonzero elements of the signal generator. Note that both diagonal elements of \(\Theta_i(s)\) have a first order pole at \(s = 0\), but this does not raise the order of the corresponding pole in \(\theta(s)\). In fact, if multiple elements of the signal generator have a pole at the same location, only the maximal order over them matters.

According to Theorems 4 and 5, a stabilizing controller is robustly regulating if and only if there exist stable matrices \(A, B \in \mathcal{M}(H^\infty)\) such that \(\theta I = A + BC\) where \(\theta = \frac{(s+1)^2}{(e^{-s}-1)(s^2+\pi^2)}\). Stable matrices \(A\) and \(B\) cannot have poles in \(\overline{C}_+\), so the unstable poles of the reference signals must be found in \(C\).

**Theorem 6.** Consider \(\Theta_r = (\theta_i)\), let \(C\) stabilize \(P\) and assume that the fractional ideal \(J = \langle \theta_i \rangle_{1 \leq i \leq n, 1 \leq j \leq q}\) is principal. Let \(\theta \in F_\mathbb{R}\) be such that \(J = \langle \theta \rangle\). If \(\theta = \frac{\alpha}{\delta}\) is a weakly coprime factorization, then \(C\) is robustly regulating if and only if there exist \(A_0, B_0 \in \mathcal{M}(\mathbb{R})\) such that
\[
d^{-1}I = A_0 + B_0 C.
\]  

(8)

If in addition to the above assumptions \(C\) has a right coprime factorization \(C = ND^{-1}\), then \(C\) is robustly regulating if and only if \(D = dD_0\) for some \(D_0 \in \mathcal{M}(\mathbb{R})\).

**Proof.** First, the controller \(C\) is shown to be robustly regulating if (8) holds. Multiplying both sides of (8) by \(n\) implies \(\theta I = A + BC\). This shows that \(C\) is robustly regulating by Theorem 5.

Next it is shown that robust regulation implies (8). To this end, set \(A_0 = d^{-1}(I - PC)^{-1}\) and \(B_0 = -d^{-1}(I - PC)^{-1}P\). The matrices \(nA_0\) and \(dA_0\) are stable since \(C\) is robustly regulating and stabilizing. Since \(\theta = \frac{\alpha}{\delta}\) is a weakly coprime factorization this implies that \(A_0\) is stable. Similar arguments show that \(B_0\) is stable. The claim follows since \(d^{-1}I = d^{-1}(I - PC)^{-1}(I - PC) = A_0 + B_0 C\).

The second statement of the theorem is shown by proving that (8) is equivalent to that \(D = dD_0\) for some \(D_0 \in \mathcal{M}(\mathbb{R})\). It is assumed that \(C = ND^{-1}\) is a right coprime factorization, so there exist \(X, Y \in \mathcal{M}(\mathbb{R})\) such that \(XD - YN = I\). If (8) holds, then \(d^{-1}D = (A_0 + B_0 C)D = A_0 D + B_0 N = D_0 \in \mathcal{M}(\mathbb{R})\) or equivalently \(D = dD_0\). The conclusion follows since, if \(D = dD_0\), then
\[
d^{-1}I = D_0 D^{-1} = D_0(XD - YN)D^{-1} = D_0 X - D_0 YC.
\]
Any weakly right coprime factorization of a stabilizing controller is also a right coprime factorization. Thus, it is sufficient to check that a given factorization of the controller is weakly coprime when applying the result of Theorem 6. Note that the assumption that $\theta$ of the above theorem has a coprime factorization does not follow by the existence of a weak coprime factorization and cannot be done without restricting generality.

**Remark 10.** If $J = \langle \theta_{ij} | 1 \leq i \leq n, 1 \leq j \leq q \rangle$ is principal and its generator has a weakly coprime factorization $\theta = d\ n$, then the internal model to be built into a robustly regulating controller is characterized by the stable element $d$. By Theorem 5, $d$ is unique up to multiplication by a unit of $R$. In this sense, one has a minimal internal model. For example, in Example 5, the stable element characterizing the minimal internal model is $d = \frac{(e^{-1}s^2 + \pi^1)}{(s+1)^2}$. The first item of Theorem 5, one may choose $d^{-1}$ to be the internal model even if $n$ and $d$ are not coprime. However, then the internal model is not minimal since $d^{-1}$ produces stronger instability than $\Theta_r$ is able to generate, or in other words $J \subsetneq \langle d^{-1} \rangle$.

Provided that the stable element $d$ characterizing the minimal internal model exists, it must divide all elements of the denominator $D$ in a coprime factorization of the controller. By Theorems 7.8 and 7.9 of Lang, if $R$ is a Bezout domain, $d$ is the largest invariant factor of the denominator of the coprime factorization of $\Theta_r$. This shows that Theorem 6 corresponds to Lemma 7.5.8 of Vidyasagar, that is, Theorem 4 is a generalization of the classical internal model principle for Bezout domains.

A consequence of Theorems 4 and 5 is that a robustly regulating controller maintains its functionality whenever the signal generator is perturbed so that no additional instabilities are generated. For example, one can multiply the signal generator by a stable matrix. This can be understood as a limited type of robustness to perturbations of the signal generator and is formulated in the next corollary. One should also note that having a nonminimal internal model in the sense of the above discussion allows some additional unstable dynamics in the perturbed signal generator.

**Corollary 2.** Let a controller $C$ robustly regulate $P$ with the signal generator $\Theta_r = (\theta_{ij})$. Then $C$ robustly regulates $P$ with the signal generator $\Phi_r = (\phi_{ij})$ if $\langle \phi_{ij} | 1 \leq i \leq n, 1 \leq j \leq q \rangle \subsetneq \langle \theta_{ij} | 1 \leq i \leq n, 1 \leq j \leq q \rangle$.

**Example 6.** Assume that $C$ solves the robust regulation problem with the signal generator $\Theta_r$ in Example 5 and let $\theta$ be the generating element found in the example. The controller is then robustly regulating with the signal generator $\theta I$ and Corollary 2 implies that the controller remains robustly regulating for a perturbed signal generator $\Phi_r$ whenever $\Phi_r = \theta \Phi_0$ for some stable $\Phi_0$. In such signal generators some of the unstable poles may cancel out, but no new unstable poles can appear.

### 6 SOLVABILITY CONDITIONS AND CONTROLLER DESIGN

Solvability of the robust regulation problem is studied in this section. The main theoretical results give solvability conditions with varying assumptions on the plant and the signal generator. The theoretical results lead to specific controller design procedures and a parameterization of robustly regulating controllers.

The solvability conditions of this section are given for the robust regulation problem and the design procedures can lead to noncausal controllers. Theorem 2 implies that the found controllers are causal whenever the given plant is strictly causal. Because of Remark 7, all controllers found by the design procedures solve the causal robust regulation problem, but the solvability conditions of Theorems 7, 8, 11, and 12 may not be necessary. These important observations are repeated in the following remark for future use. Additional observations specific to the different cases studied are presented in the subsections.

**Remark 11.** The following observations apply to the results of this section:

1. If the given plant $P$ is strictly causal, then the solvability conditions guarantee existence of a causal robustly regulating controller and the controllers constructed by the design procedures are causal.
2. A controller $C$ found by any of the proposed design procedures solves the causal robust regulation problem. However, the solvability conditions of Theorems 7, 8, 11, and 12 may not be necessary for the causal robust regulation problem and further study is needed.
6.1 | Case I: Stable plants

A solvability condition for stable plants is given first. It is inspired by the solvability condition suitable for rational transfer matrices given in Theorem 7.5.2 of Vidyasagar.\textsuperscript{7} Here the plant is assumed to be stable, but the signal generator does not need to possess a coprime factorization. The result shows that the robust regulation problem is solvable if and only if the plant does not block the unstable dynamics of the reference signals.

Theorem 7. Assume that $P \in \mathcal{M}(\mathbb{R})$. The robust regulation problem is solvable if and only if for all $i \in \{1, \ldots, n\}$ and $j \in \{1, \ldots, q\}$ the equation

$$I = \theta_{ij}^{-1}A_{ij} - PB_{ij}, \quad (9)$$

is solvable by some $A_{ij} \in \mathbb{R}^{m \times n}$ and $B_{ij} \in \mathbb{R}^{m \times q}$ whenever $\theta_{ij}$ is nonzero.

Proof. In order to show necessity, assume that $C$ is a robustly regulating controller. Then $A_{ij} := \theta_{ij}(I - PC)^{-1} \in \mathcal{M}(\mathbb{R})$ by Theorem 3 and $B_{ij} := C(I - PC)^{-1} \in \mathcal{M}(\mathbb{R})$ since $C$ is stabilizing. It follows that

$$I = \theta_{ij}^{-1}\theta_{ij}(I - PC)^{-1} - PC(I - PC)^{-1} = \theta_{ij}^{-1}A_{ij} - PB_{ij}. \quad \text{(10)}$$

By Lemma 2, without loss of generality, one may assume that $A_i$ is invertible over $\mathbb{F}_R$. Since $PB_i$ is stable as a product of two stable matrices, (10) reveals that $f_i^{-1}A_i \in \mathcal{M}(\mathbb{R})$ for every $l \in \{1, \ldots, k\}$. Thus, for $r \in \{1, \ldots, k\}$, $A_r := \prod_{l=0}^{r-1}f_l^{-1}A_{r-l}$ and $B_r := B_1 + \sum_{l=2}^{r}B_lA_{l-1}$ are stable matrices. In addition, one can show by induction that

$$I = A_r - PB_r \quad \text{(11)}$$

for all $1 \leq r \leq k$. Set $A = A_k$ and $B = B_k$. Since $A, AP, B, BP \in \mathcal{M}(\mathbb{R})$ and (11) holds for $r = k$, Proposition 6 of Quadrat\textsuperscript{36} implies that $C = BA^{-1}$ is stabilizing for $P$.

It remains to show that $C$ contains an internal model. Since $A = (I - PC)^{-1}$, Equation (11) is of the form $I = A - APC$. Multiplying it by $f_i$ and observing that $f_iA \in \mathcal{M}(\mathbb{R})$ shows that (6) holds and the claim follows by Theorem 4. \hfill\qed

The above theorem and its proof implies the following controller design procedure for a stable plant. The idea is to construct the unstable dynamics generated by the signal generator into the controller element by element.

Design procedure 1. Define the controller $C = BA^{-1}$ where the stable parameters $B$ and $A$ are chosen in the following way:

Step 1: Find a set of nonzero elements $f_1, \ldots, f_k \in \mathbb{F}_R$ such that $\langle f_1, \ldots, f_k \rangle = \langle \theta_{ij} \rangle_{1 \leq i \leq n, 1 \leq j \leq q}$.

Step 2: Set $A_0 = I$, $B_0 = 0$, and $l = 1$.

Step 3: If possible, find $A_l, B_l \in \mathcal{M}(\mathbb{R})$ such that $I = f_l^{-1}A_l - PB_l$ and $\det(A_l) \neq 0$. Define $A_l := f_l^{-1}A_lA_{l-1}$ and $B_l := B_{l-1} + B_lA_{l-1}$. If such matrices cannot be found, end the procedure since the robust regulation problem is not solvable.

Step 4: If $l = k$, set $A = A_l$ and $B = B_l$ and end the procedure. Otherwise, set $l = l + 1$ and return to Step 3.

Remark 12. Obviously one can always use the elements $\theta_{ij}$ of the signal generator in Step 1. However, the significance of Step 1 is that one can get rid of the unstable dynamics shared by one or more elements of the signal generator. If this is not done and the elements $\theta_{ij}$ share some unstable dynamics or if the generating elements $f_i$ are not chosen with care, the procedure would result into an oversized internal model, since the same unstable dynamics are constructed into the controller repeatedly. This increases the size of a state-space realization of the controller.
In order to illustrate this, consider Example 5. The unstable pole of order one at $s = 0$ appears in two elements of the signal generator. This pole would appear as a second order pole in the controller without the first step. The simplified internal model, that is, the element $\frac{(s+1)^2}{(s^2-1)s^2}$, only has a first order pole at $s = 0$. Consequently, only a first order pole at $s = 0$ is required in the controller.

It may not be easy to find the generating elements $f_i$ so that the same unstable dynamics are not repeated. At least they should form a minimal generating set and finding one is not a trivial task. Having a minimal generating set may not be enough as illustrated by Example 5 since the two nonzero elements of the signal generator form such a set for the corresponding fractional ideal. The situation is clear if one finds a single generating elements $f_i$ with a weakly coprime factorization as explained in Remark 10.

Remark 13. By Theorem 5, the above design procedure can be completed whenever the robust regulation problem is solvable since the fractional ideals in the first step are equal. Then the solvability conditions are checked in Step 3 and the failure of this step means that the robust regulation problem is not solvable. One can choose $\{f_1, \ldots, f_k\}$ having $\{\theta_{ij} | 1 \leq i \leq n, 1 \leq j \leq q\}$ as its proper subset in Step 1, but in that case the failure of Step 3 would not necessarily imply that the robust regulation problem is not solvable. Another downside of having inequality is that the resulting internal model is oversized if the procedure is successful.

Remark 14. One may add the internal model of new unstable dynamics into an existing controller using Design procedure 1. First observe that the existing controller gives the solution to (9) with the old dynamics by the proof of Theorem 7. The remaining task is to repeat Step 3 of the procedure with the unstable dynamics to be added into the controller.

Remark 15. A matrix is causal if it has a factorization whose denominator is $\mathbf{Z}$-nonsingular.32 The product of $\mathbf{Z}$-nonsingular matrices is $\mathbf{Z}$-nonsingular, so the controller found using Design procedure 1 is causal if the matrices $A_i$ can be chosen so that $f_i^{-1}A_i$ are $\mathbf{Z}$-nonsingular.

### 6.2 Case II: General plants

The following is the main result of this section, and it gives a necessary and sufficient condition for the solvability of the robust regulation problem using no coprime factorizations. It results to a straightforward controller design procedure, where the plant is first stabilized and then a stabilizing controller containing the internal model is determined by exploiting the parameterization of all stabilizing controllers.

**Theorem 8.** Let $C_s$ stabilize $P$ and write $\theta_{ij} = \frac{a_{ij}}{d_{ij}}$. Denote $U = (I - PC_s)^{-1}$, $V = C_s(I - PC_s)^{-1}$, $\tilde{L} = [-V \ (I - C_s)^{-1}]$, and $L = [U^T \ V^T]^T$. The robust regulation problem is solvable if and only if the system of equations

$$
\begin{pmatrix}
d_{ij}A_{ij} - \left[n_{ij}I \ 0\right] \\
\end{pmatrix}
\begin{pmatrix}
L \\
W
\end{pmatrix}
= 0,
$$

where $i \in \{1, \ldots, n\}$ and $j \in \{1, \ldots, q\}$, is solvable by $A_{ij} \in \mathbf{R}^{n \times (n+m)}$ and $W \in \mathbf{R}^{(n+m) \times (n+m)}$ such that $\det(U + P\tilde{L}WL) \neq 0$.

**Proof.** Writing $\tilde{A}_{ij} = \begin{bmatrix} A_{ij} & B_{ij} \end{bmatrix}$ where $A_{ij} \in \mathbf{R}^{n \times n}$ and $B_{ij} \in \mathbf{R}^{n \times m}$, Equation (12) can be written in the form

$$
\begin{bmatrix}
d_{ij}A_{ij} - n_{ij}I & d_{ij}B_{ij}
\end{bmatrix}
\begin{bmatrix}
U + P\tilde{L}WL \\
V + \tilde{L}WL
\end{bmatrix}
= 0,
$$

which is equivalent to

$$
\theta_{ij}I = A_{ij} + B_{ij}(V + \tilde{L}WL)(U + P\tilde{L}WL)^{-1}
$$

if $\det(U + P\tilde{L}WL) \neq 0$. The result follows by observing that any stabilizing controller $C$ can be expressed in the form $C = (V + \tilde{L}WL)(U + P\tilde{L}WL)^{-1}$ by Theorem 1 and that (13) is equivalent of it being robustly regulating by Theorem 4.
Design procedure 2. Define the controller

\[ C = (V + WL)(U + PLWL)^{-1}, \tag{14} \]

where the parameters \( U, L, \bar{L}, \) and \( W \) are chosen by the following procedure:

Step 1: Find a set of nonzero elements \( f_i, \ldots, f_k \in F \) such that \( \langle f_1, \ldots, f_k \rangle = \langle \theta_y | 1 \leq i \leq n, 1 \leq j \leq q \rangle \).

Step 2: Find a stabilizing controller \( C_s \) for \( P \) and let \( U, V, \bar{L}, \) and \( L \) be as in Theorem 8.

Step 3: Write \( f_i = n_i/d_i \) and find the parameter \( W \) by solving the matrix equation

\[
\begin{bmatrix}
  d_1I & 0 & \ldots & 0 \\
  0 & d_2I & \ldots & 0 \\
  \vdots & \vdots & \ddots & \vdots \\
  0 & 0 & \ldots & d_kI
\end{bmatrix}
\begin{bmatrix}
  A_1 & n_1I & 0 \\
  A_2 & n_2I & 0 \\
  \vdots & \vdots & \vdots \\
  A_k & n_kI & 0
\end{bmatrix}
\begin{bmatrix}
  P \bar{L} \\
  I
\end{bmatrix}
L = 0. \tag{15}
\]

Remark 13 concerning the first step applies to the above design procedure. In addition, the size of the matrix equation (15) is reduced if the number of elements \( f_i \) in the first step is small.

Remark 16. The point of using Equations (12) and (15) instead of (13) is that they involve only stable matrices. Then the proposed design procedure requires solving a matrix equation of the form \((M_1 + M_2 X)(M_3 + M_4 Y)M_5 = 0\) over the ring \( R \) where the matrices \( X, Y \) are to be solved.

The following necessary solvability condition shows that the robust regulation problem is solvable only if the plant does not block the unstable dynamics produced by the signal generator. The condition is not sufficient since it does not address the stabilizability of \( P \) in anyway.

Theorem 9. Assume that the robust regulation problem is solvable. Then for all \( i \in \{1, \ldots, n\} \) and \( j \in \{1, \ldots, q\} \) the equation \( \theta_y I = A_{ij} - \theta_y PB_{ij} \) is solvable by some \( A_{ij} \in R^{n \times n} \) and \( B_{ij} \in R^{m \times m} \).

Proof. If \( C \) is a robustly regulating controller, then \( \theta_y I = \theta_y (I - PC)^{-1} - \theta_y PC(I - PC)^{-1} \) where \( \theta_y (I - PC)^{-1} \in M (R) \) and \( C(I - PC)^{-1} \in M (R) \).

The next theorem generalizes the solvability condition of stable plants given in Theorem 7 to general plants. This condition is only sufficient. The restatement of this result for systems with a coprime factorization given later in Theorem 11 is both necessary and sufficient. Roughly speaking, the idea is that finding a numerator of the plant that does not block the unstable dynamics produced by the signal generator guarantees existence of a robustly regulating controller.

Theorem 10. The robust regulation problem is solvable if there exists a stabilizing controller \( C_s \) of \( P \) such that for all \( i \in \{1, \ldots, n\} \) and \( j \in \{1, \ldots, q\} \), the equation

\[ \theta_y I = A_{ij} - \theta_y P(I - C_s P)^{-1} B_{ij} \tag{16} \]

is solvable by some \( A_{ij} \in R^{n \times n} \) and \( B_{ij} \in R^{m \times m} \).

Proof. By Theorem 7, Equation (16) implies that there exists a controller \( C_r \) that robustly regulates \( P_0 = P(I - C_s P)^{-1} \). The claim follows if one can show that \( C = C_s + C_r \) is robustly regulating for \( P \). Lemma 1 implies that \( C \) is stabilizing. Now

\[
\theta_y (I - PC)^{-1} = \theta_y (I - (I - PC)^{-1} PC_r)^{-1} (I - PC_r)^{-1} = \theta_y (I - P_0 C_r)^{-1} (I - PC_r)^{-1} \in M (R)
\]
since \( C_s \) stabilizes \( P \) and \( C_r \) robustly regulates \( P_0 \), that is, \( \theta_{ij}(I - P_0C_r)^{-1} \in \mathcal{M}(\mathbb{R}) \). The matrix in the above equation remains stable if it is multiplied by \( P \) from the right since \((I - PC_s)^{-1}P \in \mathcal{M}(\mathbb{R})\). Theorem 3 implies that \( C \) solves the robust regulation problem.

The above theorem shows that one way of finding a robust controller is the two-stage controller design where one first stabilizes the plant and then constructs a robustly regulating controller for the stabilized plant. One way of completing the second part is given in Design procedure 1. Combining the two controllers leads to a robustly regulating controller for the given plant. This idea is summarized by the following controller design procedure.

**Design procedure 3.** Define the controller \( C = C_s + C_r \) where \( C_s \) and \( C_r \) are chosen in the following way:

1. Step 1: Find a stabilizing controller \( C_s \) for \( P \).
2. Step 2: Find a robustly regulating controller \( C_r \) for \( P_0 = P(I - C_sP)^{-1} \).

**Remark 17.** The above design procedure may fail because the stabilizing controller in the first step may already contain a partial internal model causing the stable plant \( P_0 \) to block some unstable dynamics of the reference signals and, consequently, Step 2 to fail. In such a case one may be able to complete the last step by ignoring the unstable dynamics already appearing in the stabilizing controller as is done in Example 7.

**Remark 18.** The controller found by using Design procedure 3 is causal if the stabilizing controller \( C_s \) and the robustly regulating controller \( C_r \) are both causal. The observations made in Remarks 11 and 15 apply to finding the causal robust controller.

### 6.3 Case III: Plants with right coprime factorizations

The following theorem generalizes the solvability condition of Theorem 7 to plants with a coprime factorization. The theorem generalizes the solvability condition of Laakonen by allowing a general signal generator.

**Theorem 11.** Provided that a stabilizable plant \( P \) has a right coprime factorization \( P = ND^{-1} \), the robust regulation problem is solvable if and only if for every nonzero \( \theta_{ij} \) where \( i \in \{1, \ldots, n\} \) and \( j \in \{1, \ldots, q\} \), the equation

\[
I = \theta_{ij}^{-1}A_{ij} - NB_{ij}
\]

is solvable by some \( A_{ij} \in \mathbb{R}^{m \times n} \) and \( B_{ij} \in \mathbb{R}^{m \times n} \).

**Proof.** Since \( P = ND^{-1} \) is a right coprime factorization, Lemma 8.3.2 of Vidyasagar and its proof imply that any stabilizing controller can be written in the form \( C_s = X^{-1}Y \) where \( X, Y \in \mathcal{M}(\mathbb{R}) \) are such that

\[
I = XD - YN
\]

and \( \det(X) \neq 0 \). If \( C_s = X^{-1}Y \) satisfying (18) solves the robust regulation problem, then a direct calculation shows that

\[
I = (I - PC_s)^{-1} - PC_s(I - PC_s)^{-1} = \theta_{ij}^{-1}\theta_{ij}(I - PC_s)^{-1} - NY,
\]

where \( \theta_{ij}(I - PC_s)^{-1} \in \mathcal{M}(\mathbb{R}) \). This implies necessity.

In order to show sufficiency, choose an arbitrary stabilizing controller \( C_s \) and let \( C_s = X^{-1}Y \) be its coprime factorization satisfying (18). By Theorem 7, Equation (17) implies that there exists a robustly regulating controller \( C_r \) for \( N \). Consider the controller \( C := C_s + X^{-1}C_r \). The chosen controller has the left coprime factorization \( C = (DRX)^{-1}(DRY + NR) \) where \( DR = (I - C_rN)^{-1} \) and \( NR = DRC_r \) are stable matrices since \( C_r \) stabilizes \( N \). To verify this, observe that

\[
DRXD - (DRY + NR)N = D_R(XD - YN) - N_RN = D_R - D_RC_rN = D_R(I - C_rN) = I.
\]
This implies that $C$ stabilizes $P$. In addition, $\theta_y(I - PC)^{-1} = \theta_y(I - NC_r)^{-1}(I - PC)^{-1} \in M(\mathbb{R})$ since $C_r$ robustly regulates $N$ and $C_s$ stabilizes $P$. The above matrix remains stable if it is multiplied by $P$, so $C$ is robustly regulating by Theorem 3.

The proof of the above theorem leads to the following design procedure. It is not specified how the robust controller for the stable transfer matrix in the last step is found since it can be done in various ways, for example, by using Design procedure 1 or the simple method applied in Example 8.

**Design procedure 4.** Define the controller $C = C_s + X^{-1}C_r$ where $C_s = X^{-1}Y$ and $C_r$ are chosen in the following way:

Step 1: Find a right coprime factorization $P = ND^{-1}$ of the plant and find a stabilizing controller $C_s = X^{-1}Y$ by solving the equation $XD - YN = I$.

Step 2: Find a robustly regulating controller $C_r$ for $N$.

Remark 19. In the above design procedure, the solvability of the robust regulation problem is verified in the last step when trying to construct a robustly regulating controller for the numerator matrix. If such a controller exists, then the problem is solvable and otherwise not.

### 6.4 Case IV: Simple signal generators

As the final result it is shown that the classical solvability condition of Theorem 7.5.2 in Vidyasagar\textsuperscript{7} and the related controller design method are applicable in the general framework provided that the internal model is captured by a single element with a weakly coprime factorization. In such a case, the parameterization of all stabilizing controllers can be applied to obtain a parameterization of all robustly regulating controllers.

**Theorem 12.** Assume that there exists $\theta \in \mathbb{F}_R$ with a weakly coprime factorization $\theta = \frac{n}{d}$ such that it satisfies the equality $\langle \theta \rangle = \langle \theta_y \mid 1 \leq i \leq n, 1 \leq j \leq q \rangle$. Then the robust regulation problem is solvable if and only if $P_0 = d^{-1}P$ is stabilizable and there exist $A, B \in M(\mathbb{R})$ such that

\[ I = dA + PB. \]  

(19)

Provided that the robust regulation problem is solvable, a controller $C$ solves it if and only if the controller is of the form $C = d^{-1}C_0$ where $C_0$ stabilizes $P_0 = d^{-1}P$.

**Proof.** The necessity parts of the claims are shown first. If a robustly regulating controller exists, Theorems 6 and 9 show that (19) holds. It remains to show that if $C$ solves the robust regulation problem then $C_0 = dC$ stabilizes $P_0$. Observe that

\[ (I - P_0C_0)^{-1} = (I - PC)^{-1} \in M(\mathbb{R}), \]  

(20a)

\[ C_0(I - P_0C_0)^{-1} = dC(I - PC)^{-1} \in M(\mathbb{R}), \]  

(20b)

\[ (I - C_0P_0)^{-1} = (I - CP)^{-1} \in M(\mathbb{R}), \]  

(20c)

\[ (I - P_0C_0)^{-1}P_0 = d^{-1}(I - PC)^{-1}P \in M(\mathbb{R}). \]  

(20d)

The stability in (20d) follows since $C$ is robustly disturbance rejecting for the signal generator $d^{-1}I$ by Theorems 3 and 6. The above equations imply that $C_0$ stabilizes $P_0$.

In order to show the sufficiency parts, it is shown that $C = d^{-1}C_0$ is stabilizing and robustly regulating for $P$ if $C_0$ stabilizes $P_0$ and (19) holds. In order to show stability, observe that (20a) and (20c) hold since $C_0$ stabilizes $P_0$. In addition,

\[ (I - PC)^{-1}P = d(I - P_0C_0)^{-1}P_0 \in M(\mathbb{R}). \]
and using (19) one observes that
\[
(C(I - PC)^{-1} = d^{-1} C_0(I - P_0 C_0)^{-1}(dA + PB)
= C_0 (I - P_0 C_0)^{-1} A + C_0 (I - P_0 C_0)^{-1} P_0 B \in \mathcal{M}(\mathbb{R}).
\]

This shows that C stabilizes P. By Equation (19), one has
\[
d^{-1} I = d^{-1} (I - PC)^{-1} (dA + PB) - d^{-1} (I - PC)^{-1} PC
= (I - PC)^{-1} A + (I - P_0 C_0)^{-1} P_0 B - (I - P_0 C_0)^{-1} P_0 C.
\]

Theorem 6 implies that C solves the robust regulation problem.

The signal generator is causal by assumption. It follows that the generating element \( \theta \) in the above theorem must be causal and Lemma 3 implies that the denominator \( d \) of its weakly coprime factorization is in \( \mathbb{R} \setminus \mathbb{Z} \). It follows that \( P_0 \) is causal as a product of two causal elements. Theorem 2 implies that \( P_0 \) has a causal stabilizing controller provided that it is stabilizable. This leads to the following corollary.

**Corollary 3.** The solvability condition presented in Theorem 12 is necessary and sufficient for the existence of a causal robustly regulating controller.

The above theorem implies the following straightforward design method, where one first includes the internal model and then stabilizes the resulting system. The order in which the stabilization and the construction of an internal model is done is reversed when compared to the design procedures proposed above.

**Design procedure 5.** Define the controller \( C = d^{-1} C_0 \) where \( C_0 \) and \( d \) are chosen in the following way:

Step 1: Find \( \theta \in \mathbb{F}_\mathbb{R} \) with a weakly coprime factorization \( \theta = n_d \) such that \( \langle \theta \rangle = \langle \theta_i | 1 \leq i \leq n, 1 \leq j \leq q \rangle \).

Step 2: Check the solvability by solving Equation (19). If this is not possible, end the procedure since the robust regulation problem is not solvable.

Step 3: Find a stabilizing controller \( C_0 \) for \( P_0 = d^{-1} P \).

**Remark 20.** It is important to notice that the procedure without Step 2 can produce a controller that is not a robustly regulating controller. For example, in the extreme case with \( \Theta_0 = d^{-1} I \) where \( d \) is a nonunit element of \( \mathbb{R} \) and \( P = 0 \), Equation (19) obviously has no solution, but \( P_0 = dP = 0 \) is stable already. Of course, solving the equation is not necessary if the solvability can be verified in some other way since the solution is not used in the design procedure. One possibility is to skip Step 2 in the first place and use Theorem 3 or 6 as a final step to verify that the resulting controller is indeed robustly regulating.

**Remark 21.** Finding \( \theta \) with a coprime factorization such that \( \langle \theta_i | 1 \leq i \leq n, 1 \leq j \leq q \rangle \subset \langle \theta \rangle \) is straightforward. For example, one can write \( \theta_i = n_d \) and set \( \theta = \frac{1}{d} \) where \( d = \prod_{1 \leq i \leq n, 1 \leq j \leq q} d_{ij} \). Then the controller constructed in the above design procedure is robustly regulating provided that (19) is satisfied. Again the internal model may be oversized, but this way one can apply the procedure even if \( \langle \theta_i | 1 \leq i \leq n, 1 \leq j \leq q \rangle \) is not originally principal or the generator of the principal fractional ideal does not possess a weakly coprime factorization.

Theorem 12 implies that all robust controllers are found by finding all the stabilizing controllers of \( P_0 \). Applying the parameterization for all stabilizing controllers given by Theorem 1 yields the following parameterization of robustly regulating controllers. For a strictly proper plant \( P \) all the controllers given by the parameterization are causal. If \( P \) is not strictly proper, then some of the controllers may be noncausal, but Corollary 3 guarantees that some of the controllers are causal.

**Corollary 4.** Assume that \( \langle \theta \rangle = \langle \theta_i | 1 \leq i \leq n, 1 \leq j \leq q \rangle \) where \( \theta \in \mathbb{F}_\mathbb{R} \) has a weakly coprime factorization \( \theta = n_d \) and that \( C \) solves the robust regulation problem. Denote \( L_0 := \left[ d^{-1} (I - PC)^{-1} \right]^T (C(I - PC)^{-1} )^T \) and \( \tilde{L}_0 := \left[ d^{-1} (I - PC)^{-1} \right]^T (C(I - PC)^{-1} )^T \)
\[-d(I - CP)^{-1}(I - CP)^{-1}\]. Then all controllers solving the robust regulation problem are given by the parameterization

\[
C(W) = (C(I - PC)^{-1} + \tilde{L}_0 WL_0) \left( (I - PC)^{-1} + P\tilde{L}_0 WL_0 \right)^{-1}
\]

(21a)

\[
= (I - CP)^{-1} + \tilde{L}_0 WL_0 P^{-1} \left( (I - CP)^{-1} + \tilde{L}_0 WL_0 \right),
\]

(21b)

where \( W \in \mathbb{R}^{(n+m)x(n+m)} \) is such an element that \( \det((I - PC)^{-1} + P\tilde{L}_0 WL_0) \neq 0 \) and \( \det(I - CP)^{-1} + \tilde{L}_0 WL_0 P) \neq 0 \).

7 EXAMPLES

In the first example, Design procedure 2 is applied to construct a robustly regulating controller for a delay system. The reference signals have complicated unstable dynamics which restricts the possible choices of the ring \( \mathbb{R} \) of stable transfer functions. This underlines the importance of the general approach.

Example 7. Let the given plant and the signal generator be

\[
P = \begin{bmatrix}
e^{-\frac{s}{2}} & 0 \\
\frac{e^{-s}}{1+2s} & 2e^{-s} \\
1+2s & 1+4s
\end{bmatrix}
\quad \text{and} \quad
\Theta_r = \begin{bmatrix}
\frac{1}{e^{-\frac{s}{2}}} & 0 \\
0 & \frac{1}{s} + \frac{1}{s^2 + \pi^2}
\end{bmatrix}.
\]

The aim is to find a robustly regulating controller by applying Design procedure 2.

Before proceeding a suitable ring of stable transfer functions and the ideal characterizing causality are fixed. The ring is chosen to be \( \mathbb{R} = \mathbb{P} \). The reason for this is that the signal generator has infinitely many poles on the imaginary axis, which poses some restriction on the stability type achievable. For example, it is not possible to solve the proposed robust regulation problem if \( \mathbb{R} \) is chosen to be \( H^\infty \). The ideal \( \mathbb{Z} \) is chosen to be

\[
\mathbb{Z} = \left\{ f \in \mathbb{P} \left| \lim_{\rho \to 0} \sup_{|s| > \rho} |f(s)| = 0 \right. \right\}.
\]

This mean that (strictly) causal stable elements are exactly the elements that are (strictly) proper in the sense of the definition given by Curtain and Morris. It is easy to verify that \( \mathbb{Z} \) is an ideal, but it is more complicated to show that it is prime. However, this is not needed since the plant possess right and left coprime factorizations. This follows since the stabilizing controller is rational and has coprime factorizations. Then strictly causal plants have only causal stabilizing controllers due to the results by Vidyasagar et al. mentioned in Remark 3. Finding the actual coprime factorizations is unnecessary.

The only unstable element of the plant can be written in the form \( \left( \frac{s}{s+1} \right)^{-1} e^{-\frac{s}{s+1}} \) where \( \frac{s}{s+1} \in \mathbb{R} \setminus \mathbb{Z} \) and \( e^{-\frac{s}{s+1}} \in \mathbb{Z} \), that is, this element is strictly causal. The other elements are clearly in \( \mathbb{Z} \), so the plant transfer matrix is strictly causal. This means that the controller found in the procedure will be causal. Similar arguments show that the signal generator is causal. It is not strictly causal since its first diagonal element does not vanish at infinity.

**Step 1:** The signal generator is simplified first. The observation that \( \frac{1}{e^{-\frac{s}{2}}} : \frac{e^{-\frac{s}{2}}}{s} = \frac{1}{s} \) and \( \frac{e^{-\frac{s}{2}}}{s} \in \mathbb{P} \) leads to the equality

\[
\left( \frac{1}{e^{-\frac{s}{2}}} - 1, 0, \frac{1}{s^2 + \pi^2} \right) = \left( \frac{1}{e^{-\frac{s}{2}}} - 1, \frac{1}{s^2 + \pi^2} \right).
\]

Denote \( f_1 = \frac{1}{e^{-\frac{s}{2}}} - 1 \) and \( f_2 = \frac{1}{s^2 + \pi^2} \). These elements have the fractional representations \( f_j = \frac{n_j}{d_j} \) where \( n_1 = 1, d_1 = e^{-\frac{s}{2}} - 1, n_2 = \frac{1}{(s+1)^{1}}, \) and \( d_2 = \frac{s^2 + \pi^2}{(s+1)^{1}} \) are elements of \( \mathbb{P} \).

The calculations of Example 5 show that the internal model is captured by the single element \( \frac{(s+1)^2}{(e^{-\frac{s}{2}} - 1)(s^2 + \pi^2)} \) also with \( \mathbb{R} = \mathbb{P} \). However, two elements are used in order to illustrate the matrix equation (15). This does not lead to an oversized internal model since the chosen elements \( f_1 \) and \( f_2 \) do not have common unstable dynamics, that is, unstable poles, unlike the original elements of the signal generator.
Step 2: The plant has a first order pole at \( s = 0 \). This means that the stabilized closed loop is likely to have a zero at this pole appearing also in the signal generator. This would lead to a situation where no robust controller in the next step is available since the stabilized plant would block some unstable dynamics and that is not allowed by Theorem 9. The part of the internal model containing this pole is therefore built into the controller already at this step. This can be done using the PI-controller

\[
C_s = -\frac{1}{16} \left( 4 + \frac{1}{s} \right) I
\]

having the desired pole at the origin. This leads to

\[
U = (I - PC_s)^{-1} = \begin{bmatrix}
\frac{16s^2 + (4s+1)e^{-2s}}{32s(16s^2 + (4s+1)e^{-2s})} & 0 \\
0 & \frac{8s}{8s + e^{-2s}}
\end{bmatrix}
\]

\[
V = UC_s = -\frac{4s+1}{16} U, \quad L = [-V \quad U], \quad \text{and} \quad L = \begin{bmatrix} U \\ V \end{bmatrix}. \]

The parameters of the controller are chosen so that the plant is stabilized and the resulting matrices are as simple as possible. Naturally, this step can be carried out by using other types of controllers as well.

Step 3: It remains to choose \( W \) and \( \tilde{A}_j, j \in \{1, 2\} \), that solve (15). The internal model should reproduce the poles of the signals generator. The pole at zero is already included in the stabilizing controller \( C_s \). Thus, \( W \) should be chosen so that it captures the remaining poles of the signal generator. After choosing \( W \) one should be able to choose \( \tilde{A}_j \) appropriately, which shows that the controller has an internal model of the signal generator.

To this end, a controller \( C_r \) containing all the unstable poles of the signal generator except the one at the origin is introduced. The controller is chosen to be of the form \( \sum \frac{\varepsilon}{s - \omega_k} M_k \) where \( \omega_k \) are the nonzero poles of the signal generator. It was shown by Laakkonen and Pohjalainen\(^{21} \) that aligning \( M_k \) appropriately with the stable plant to be regulated at each pole and choosing small enough gain \( \varepsilon \) results to a robustly regulating controller. This in mind the controller

\[
C_r = \sum_{n \in \{-1, 1\}} \frac{\varepsilon}{s + \pi n i} (P(\pi n i)U(\pi n i))^{-1} + \sum_{n \in \{-1\} \setminus \{0\}} \frac{\varepsilon}{n^2(s - 2\pi n i)} (P(2\pi n i)U(2\pi n i))^{-1}
\]

that stabilizes \( UP \) is chosen. The series converges outside the poles since the numerator is of second order with respect to \( n \) whereas the denominator is of fourth order. It follows that the matrix

\[
W = \begin{bmatrix}
0_{2 \times 2} & 0_{2 \times 2} \\
C_r(I - UPC_r)^{-1} & 0_{2 \times 2}
\end{bmatrix}
\]

is stable over \( P \). Choose

\[
\tilde{A}_j = f_j(I - UPC_r)^{-1} \begin{bmatrix} U & -UP \end{bmatrix},
\]

where \( j \in \{1, 2\} \). An analysis similar to that in Section 5.3 of Laakkonen and Pohjalainen\(^{21} \) shows that \( \tilde{A}_2 = f_2(I - UPC_r)^{-1} \in \mathcal{M}(P) \) and \( \frac{s}{(e^{-s} - 1)(s + 1)} (I - UPC_r)^{-1} \in \mathcal{M}(P) \). In addition, a direct calculation shows that \( \frac{s+1}{s} \begin{bmatrix} U & -UP \end{bmatrix} \in \mathcal{M}(P) \). Thus, \( \tilde{A}_1 \) is stable, that is,

\[
\tilde{A}_1 = \frac{s}{(e^{-s} - 1)(s + 1)} (I - UPC_r)^{-1} \cdot \frac{s + 1}{s} \begin{bmatrix} U & -UP \end{bmatrix} \in \mathcal{M}(P).
\]
It remains to show that the chosen matrices \( W \) and \( \tilde{A}_j \) satisfy the matrix equation (15) or equivalently the equations (12). To that end, calculate

\[
LWL = UC_r(1 - UPC_r)^{-1}U
\]

(23)

from which it follows that

\[
\left( I + \begin{bmatrix} pL \\ \bar{L} \end{bmatrix} W \right) L = \begin{bmatrix} U \\ C_s U \end{bmatrix} + \begin{bmatrix} PUC_r(1 - UPC_r)^{-1}U \\ UC_r(1 - UPC_r)^{-1}U \end{bmatrix}
\]

\[
= \begin{bmatrix} I - PUC_r + UPC_r \\ C_s(I - UPC_r) + UC_r \end{bmatrix} (1 - UPC_r)^{-1}U \
\]

\[
= \begin{bmatrix} I \\ C_s + C_r \end{bmatrix} (1 - UPC_r)^{-1}U,
\]

(24a)

(24b)

(24c)

where the property \( UP = PU \) was used. In addition, one can write

\[
\left( d_j \bar{A}_j - \begin{bmatrix} n_j I \\ 0 \end{bmatrix} \right) = n_j(I - UPC_r)^{-1}U \left[ I - U^{-1}(I - UPC_r) \right] - P
\]

\[
= n_j(I - UPC_r)^{-1}UP \begin{bmatrix} C_s + C_r \\ -I \end{bmatrix}.
\]

(25a)

(25b)

Showing that (12) holds can be done by substituting (24) and (25) into it.

Substitute \( U, V, \) and (23) into (14) to obtain

\[
C = (C_sU + UC_r(I - UPC_r)^{-1}U) \left( U + UPC_r(I - UPC_r)^{-1}U \right)^{-1}
\]

\[
= (C_s + UC_r(I - UPC_r)^{-1}) \left( I + UPC_r(I - UPC_r)^{-1} \right)^{-1}
\]

\[
= (C_s + UC_r(I - UPC_r)^{-1})(I - UPC_r)
\]

\[
= C_s + U(I - PC_r)C_r = C_s + C_r.
\]

Thus, the constructed robustly regulating controller is

\[
C = -\frac{1}{16} \left( 4 + \frac{1}{s} \right) I + \epsilon \begin{bmatrix} \frac{4x+1-16x^2}{8(x^2+\pi^2)} & 0 \\ \frac{4x^2(2x+1+8x^2)}{(1+4x^2)(x^2+\pi^2)} & \frac{12x+1-32x^2}{8(x^2+\pi^2)} \end{bmatrix} + \epsilon \sum_{n=1}^{\infty} \begin{bmatrix} \frac{4x+1-64x^2n^2}{8n^2(s^2+(2n)^2)} & 0 \\ \frac{16x^2(2x+1+32x^2n^2)}{(s^2+4n^2)(1+16x^2n^2)} & \frac{12x+1-128x^2n^2}{8n^2(s^2+(2n)^2)} \end{bmatrix}.
\]

Since Equation (7) holds with the choice \( R = P \), the fractional ideal (22) is principal with the generator \( \theta = \frac{(s+1)^2}{(e^{-s})^2(s^2+\pi^2)} \).

Furthermore, \( \theta \) has the coprime factorization \( 1/d \) where \( d = \frac{(e^{-s}-1)(s^2+\pi^2)}{(s+1)^2} \in P \), so all robustly regulating controllers are given by the parameterization (21). All of them are causal since the plant is strictly causal.

In the following example, Design procedure 4 is applied to calculate a finite-dimensional robust controller for a heat equation. The example is particularly important since it demonstrates that the design procedure may be carried out using standard techniques and without calculating a closed form expression of the plant transfer matrix.

**Example 8.** Consider the heat equation

\[
\frac{\partial z}{\partial t}(t,x) = \frac{\partial^2 z(t,x)}{\partial x^2}, \quad z(0,x) = z_0(x)
\]

\[
\frac{\partial z}{\partial x}(t,0) = -u_1(t), \quad \frac{\partial z}{\partial x}(t,1) = u_2(t)
\]

\[
y_1(t) = 4 \int_0^1 z(t,x) \, dx, \quad y_2(t) = \int_{\frac{1}{2}}^1 z(t,x) \, dx.
\]
on the unit interval. The measurements $y_1(t)$ and $y_2(t)$ should asymptotically track the reference signals $y_{ref}^1(t) = \sin(2t) + 1$ and $y_{ref}^2(t) = \cos(2t)$, respectively. A suitable choice for the ring of stable transfer functions is $R = H^\infty$. In what follows, the controller parameters of Design procedure 4 are chosen.

**Step 1.** The Laplace transforms of the reference signals are $\hat{y}_1(s) = \frac{1}{s^2+4} + \frac{1}{s}$ and $\hat{y}_2(s) = \frac{1}{s^2+4}$. The poles of the signals locate at $s = 0$, $s = 2i$, and $s = -2i$. This is the minimal set of poles the controller should have. Since the poles $s = \pm 2i$ appearing in both reference signals are simple, it is sufficient that the controller has only first order poles at these locations. This information is sufficient for constructing the internal model in Step 3, so the signal generator need not be given explicitly.

**Step 2.** A stabilizing controller and its left coprime factorization are found next. The eigenvalues of the plant are $\lambda_n = \pi^2 + 1 - n^2\pi^2$, $n = 0, 1, \ldots$ The system has no other spectrum points and there are only two unstable eigenvalues $\lambda_0 = \pi^2 + 1$ and $\lambda_1 = 1$. The finite-dimensional controller is defined as

$$\dot{v} = Mv + Hy, \quad u = Fv,$$

where

$$M = \begin{bmatrix} -200 & -150 \\ 0 & -50 \end{bmatrix}, \quad H = \begin{bmatrix} 120 & 0 \\ 1 & -1 \end{bmatrix}, \quad \text{and} \quad F = \begin{bmatrix} -50 & -75 \\ -50 & 75 \end{bmatrix}$$

are found by using the techniques presented in Curtain and Salamon. The approximated stability margin with this controller is $-1.70$. The transfer function of the stabilizing controller is

$$C_s(s) = F(sI - M)^{-1}H = \frac{1}{s^2 + 250s + 10,000} \begin{bmatrix} -6075s - 3.075 \cdot 10^5 & -5925s - 2.775 \cdot 10^5 \\ 75s + 7500 & -75s - 22,500 \end{bmatrix}.$$ 

Its left coprime factorization $C_s = X^{-1}Y$ can be found by using standard methods. Here it is found by using the Matlab function lncf. Only the denominator matrix

$$X(s) = \frac{1}{s^2 + 8605s + 9.911 \cdot 10^5} \begin{bmatrix} s^2 + 4324s + 2.094 \cdot 10^5 & -4108s - 1.838 \cdot 10^5 \\ -4108s - 2.063 \cdot 10^5 & s^2 + 4531s + 2.284 \cdot 10^5 \end{bmatrix}$$

is needed. Since the plant can be stabilized by a controller having a coprime factorization, it has a coprime factorization as well and the numerator of the right-coprime factorization needed in the next step is formally given by $N = P(I - C_sP)^{-1}X^{-1}$.

**Step 3.** Next a robustly regulating controller is constructed for the stable transfer matrix $N$. One such controller is given by

$$C_r(s) = -\epsilon \left( \frac{N^{-1}(0)}{s} + \frac{N^{-1}(2i)}{s - 2i} + \frac{N^{-1}(-2i)}{s + 2i} \right),$$

where $\epsilon > 0$ is to be chosen appropriately small. The idea behind this controller is exactly the same as that of the controller in Step 3 of Example 7. This verifies the solvability of the robust regulation problem. The designed controller is not only robust to the small perturbations in the plant, but also to small changes in the controller as long as the controller contains the internal model. Thus, it is possible to replace $P$ in $N = P(I - C_sP)^{-1}X^{-1}$ by an approximated system transfer function. This way one avoids calculation of the explicit plant transfer function. Here the plant is approximated using finite differences with ten points on $[0, 1]$. Finally, $N^{-1}(0)$, $N^{-1}(2i)$ and $N^{-1}(-2i)$ are calculated and their elements are rounded to one decimal. Choosing $\epsilon = 0.5$ yields the robust controller

$$C_r(s) = \frac{1}{s^3 + 4s} \begin{bmatrix} -0.9s^2 + 0.4s - 1.2 & -1.05s^2 - 0.4s - 1.4 \\ 0.9s^2 - 0.2s + 1.2 & -0.65s^2 + 0.2s - 1 \end{bmatrix}.$$
The closed loop eigenvalues with controller $C(s)$ in Example 8 of the numerator $N$. The robust controller

$$C(s) = C_r(s) + X^{-1}(s)C_r(s)$$

is obtained by substituting the above transfer matrices. This controller is causal since it is a proper rational transfer matrix. A minimal realization of $C$ is found using the Matlab function `minreal`. The closed-loop system has stability margin 0.4132. The eigenvalues closest to the imaginary axis are plotted in Figure 2. There are three pairs of eigenvalues in the figure two units or less away from the imaginary axis. These are the eigenvalues corresponding to the poles of $C_r$. They are moving to the left from the imaginary axis when increasing $\epsilon$ from zero to 0.5. The remaining eigenvalues shown in the figure are the rightmost eigenvalues of the stabilized closed-loop system of Step 2 that move to the right as $\epsilon$ is increased. Thus, the stability margin obtained with the proposed choice of $\epsilon$ is nearly optimal with the stabilizing controller constructed in Step 2. In comparison, the stability margin obtained by using the actual closed-loop transfer matrix $P$ instead of the approximated one when constructing $C_r$ is 0.4588.

Finally, the closed loop system is simulated. The approximation in the simulation is obtained by using finite differences with 150 points on $[0, 1]$. Figure 3 shows the behavior of the measured outputs. As expected, the outputs converge asymptotically to the reference signals. The oscillation is mainly due to the eigenvalues with the largest imaginary parts shown in Figure 2.

### 8 CONCLUDING REMARKS

This article introduced general frequency domain theory for robust regulation using fractional representations. The main theoretical contributions were the new formulation of the internal model principle, several conditions for solvability, and the parameterization of all robustly regulating controllers. Causality considerations were included. The usefulness of the results is due to the generality assured by the minimal set of standing assumptions and not requiring the existence of coprime factorizations. Unlike the results that are related to specific rings of stable transfer functions, the results presented in this article allow one to choose the stability type to work with. This is particularly important since the achievable stability type depends on the problem at hand as was demonstrated by Example 7 in which the choice of the ring of stable transfer functions was not trivial due to the challenging unstable dynamics of the reference signals.
The given conditions for solvability were accompanied by design procedures for robust controllers. Although it was not possible to give details on how to accomplish the steps of the design procedures due to the general approach, comparing the procedures reveals some main ideas. First, some of the procedures start with a step where the internal model is simplified. This step is compulsory in Design procedure 5 and is particularly important in the other procedures as well since without it the constructed internal model tends to be oversized, see Remark 12. Second, Design procedure 1 gives a recursive process to construct the internal model into the controller. This technique enables one to revise an existing robustly regulating controller by adding an internal model of new unstable dynamics so that it can handle a larger class of reference signals. Third, a two-step approach where one first stabilizes the plant and then constructs an internal model into the controller was used in Design procedures 2–4. This may be particularly handy since finding a robust controller for a stable plant can be straightforward as was seen in Example 8. In Design procedure 5, the order of stabilization and construction of internal model was reversed. Adding the internal model first and then stabilizing the resulting system is a straightforward method, but the downside is that one needs to stabilize the unstable dynamics of the plant and the signal generator at once.

The results of this article help in understanding some of the fundamental ideas in robust regulation such as the internal model principle. Whereas the results generalize the existing ones that are specific to some rings of stable transfer functions, they now provide a good starting point to go back from general to specific. In particular, several results require finding a generating set for a specific fractional ideal or solving matrix equations such as (6) or (12). These are not easy tasks in general and an interesting direction for future research would be to find out what one can say about their solvability in some of the most general rings of stable transfer functions such as $H^\infty$. On the other hand, one can try to find alternative formulations of the main results in order to obtain new insights. This has been done with SISO systems using fractional ideals.39 Two prominent frameworks for achieving further insights into robust regulation of MIMO systems are the lattice approach or the geometric systems theory.29,45 The results concerning causality were not complete for causal plants. Therefore, stronger results on causal stabilizing controllers such as parameterization of all causal controllers would be of great interest. Parameterizations for strictly causal controllers are already available.46

**CONFLICT OF INTEREST**
The author declares that there is no conflict of interest.

**DATA AVAILABILITY STATEMENT**
Data sharing is not applicable to this article as no datasets were generated or analyzed during the current study.
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