

# Zames-Falb multipliers for quadratic programming

W. P. Heath, *Member, IEEE* and A. G. Wills

## Abstract

In constrained linear model predictive control a quadratic program must be solved on-line at each control step, and this constitutes a nonlinearity. If zero is a feasible point for this quadratic program then the resultant nonlinearity is sector bounded. We show that if the nonlinearity is static then it is also monotone and slope restricted; hence we show the existence of Zames-Falb multipliers for such a nonlinearity. We express the results in terms of integral quadratic constraints. The multipliers may be used in a general and versatile analysis of the robust stability of input constrained model predictive control.

## I. INTRODUCTION

Recently we proposed a new stability test for constrained linear MPC (model predictive control) [8], [7]. If zero is feasible then the nonlinearity arising from the associated on-line quadratic program is sector bounded. Hence the multivariable circle criterion gives a sufficient condition for closed-loop stability. The results may be used to demonstrate stability against both structured and unstructured infinity-norm bounded model uncertainty [6].

In this paper we show the existence of Zames-Falb multipliers when such a nonlinearity is static—for example when the only constraints are fixed input constraints. In particular we show that the nonlinearity is bounded, monotone and slope restricted in the sense of [12] which summarizes the natural generalization of [17] to MIMO (multi-input multi-output) nonlinearities  $\phi : \mathbb{R}^N \rightarrow \mathbb{R}^N$  (in such analysis it is sufficient, without loss of generality, to consider only square systems). We express the results in terms of IQCs (integral quadratic constraints).

W. P. Heath is with Control Systems Centre, School of Electrical and Electronic Engineering, The University of Manchester, PO Box 88, Sackville Street, Manchester M60 1QD, UK [William.Heath@manchester.ac.uk](mailto:William.Heath@manchester.ac.uk)

A. G. Wills is with the School of Electrical Engineering and Computer Science, University of Newcastle, NSW 2308, Australia [Adrian.Wills@newcastle.edu.au](mailto:Adrian.Wills@newcastle.edu.au)

The multipliers, together with the results in [6], may be used for the analysis of the robust stability of input constrained MPC. The analysis is general and versatile, but we should note a caveat: the Zames-Falb stability criterion does not necessarily guarantee continuity of the input-output map [11].

**Notation:** Let  $L_2^N$  be the space of  $\mathbb{R}^N$ -valued functions  $f$  such that  $\int_{-\infty}^{\infty} |f(t)|^2 dt$  is finite; let  $l_2^N$  be the space of  $\mathbb{R}^N$ -valued sequences  $f_k$  such that  $\sum_{k=-\infty}^{\infty} |f_k|^2$  is finite. Write  $L_2$  for  $L_2^1$  and  $l_2$  for  $l_2^1$ . Let  $L_1$  be the space of real functions  $f$  such that  $\int_{-\infty}^{\infty} |f(t)| dt$  is finite; let  $l_1$  be the space of real sequences  $f_k$  such that  $\sum_{k=-\infty}^{\infty} |f_k|$  is finite. Let  $\langle x, y \rangle$  denote the inner product between  $x$  and  $y$ , and let  $x * y$  denote the convolution of  $x$  and  $y$ .

## II. ZAMES-FALB MULTIPLIERS FOR MIMO NONLINEARITIES

In [17] Zames and Falb derive multipliers in two parts:

ZF1: the first concerns bounded and monotone nondecreasing nonlinearities;

ZF2: the second concerns nonlinearities that are also slope-restricted.

Both results are derived for single-valued nonlinearities, where it is commented that they can be “easily” generalized to MIMO nonlinearities. It is observed in [15] that, in fact, the first result only generalizes to a MIMO (multi-input multi-output) nonlinearity  $\phi : \mathbb{R}^N \rightarrow \mathbb{R}^N$  under a further condition. Specifically it is shown that a necessary and sufficient further condition for Zames-Falb multipliers (ZF1) to exist is for the line integral  $\int_A^B \phi(x)^T dx$  to be independent of path, and this is equivalent to the condition that  $\phi$  is the gradient of some convex potential function. Hence in [12] the term “monotone” is reserved for this case. In this spirit, we say  $\phi$  is bounded (by  $c > 0$ ) if  $\|\phi(x)\| \leq c\|x\|$  for all  $x \in \mathbb{R}^N$ ;  $\phi$  is incrementally positive if  $[\phi(x) - \phi(y)]^T (x - y) \geq 0$  for all  $x, y \in \mathbb{R}^N$ ;  $\phi$  is monotone if it is incrementally positive and may be expressed as the gradient of a convex potential function;  $\phi$  is slope-restricted to the interval  $[a, b]$  if  $[\phi(x) - \phi(y) - a(x - y)]^T [\phi(x) - \phi(y) - b(x - y)] \leq 0$  for all  $x, y \in \mathbb{R}^N$ .

The Zames-Falb multipliers are then given as:

**Zames-Falb multipliers:** Let  $\phi : \mathbb{R}^N \rightarrow \mathbb{R}^N$  be bounded and monotone with  $\phi(0) = 0$ . Let  $h \in L_1$  satisfy  $\int_{-\infty}^{\infty} |h(t)| dt < 1$  (or let  $h \in l_1$  satisfy  $\sum_{t=-\infty}^{\infty} |h(t)| < 1$ ) and either let  $\phi$  be odd or let  $h(t) \geq 0$  for all  $t$ . Then

ZF1: For any  $x \in L_2^N$  (or  $x \in l_2^N$ ) we have

$$\langle \phi(x), x \rangle \geq \langle \phi(x), h * x \rangle \quad (1)$$

ZF2: If  $\phi$  is slope-restricted to the interval  $[\alpha, \beta - \varepsilon]$  for some real  $\alpha > 0$ ,  $\beta > 0$ ,  $\varepsilon > 0$  with  $\beta - \varepsilon > \alpha$  then for any  $x \in L_2^N$  (or  $x \in l_2^N$ ) we have

$$\langle \phi(x) - \alpha x, \beta x - \phi(x) \rangle \geq \langle \phi(x) - \alpha x, h * (\beta x - \phi(x)) \rangle \quad (2)$$

**Remark:** The multipliers can be constructed in the manner of [17], [3], [2]. Note that in [2] a restricted case is considered, and thus more general multipliers are obtained (see also [10]). When the condition ZF2 is derived in this manner, it is necessary to demonstrate that  $\tilde{\phi} = \hat{\phi}_1 \circ \hat{\phi}_2^{-1}$  is the gradient of a convex potential function with  $\hat{\phi}_1 = (\phi - \alpha I)$  and  $\hat{\phi}_2 = (\beta I - \phi)$ . Since  $\hat{\phi}_2$  is a one-to-one mapping and the gradient of a convex function, it follows from Theorem 26.6 of [14] that  $\hat{\phi}_2^{-1}$  is also the gradient of a convex function. We may express  $\tilde{\phi}$  as  $\tilde{\phi} = (\beta - \alpha)\hat{\phi}_2^{-1} - I$ . Hence  $\tilde{\phi}$  is the gradient of a potential function, and since  $\tilde{\phi}$  is non-decreasing it is the gradient of a convex potential: see [15].

### III. ZAMES-FALB MULTIPLIERS FOR QUADRATIC PROGRAMMING

Let the function  $f : \mathbb{R}^N \rightarrow \mathbb{R}^N$  be given by the quadratic program

$$\begin{aligned} f(x) &= \arg \min_n \frac{1}{2} n^T n - n^T x \\ &\text{subject to } Ln \preceq b \text{ with } b \succeq 0 \\ &\text{and } Mn = 0 \end{aligned} \quad (3)$$

with “ $\preceq$ ” and “ $\succeq$ ” denoting term-by-term inequality. We assume  $L$ ,  $M$  and  $b$  are fixed so that  $f$  is a static nonlinearity.

In [8] we showed that the nonlinearity that occurs in constrained linear MPC (model predictive control) can be expressed in this manner, provided 0 is feasible. Furthermore  $f$  lies in the sector  $[0, I]$  in the sense that

$$f(x)^T [f(x) - x] \leq 0 \text{ for all } x \in \mathbb{R}^N \quad (4)$$

Thus the multivariable circle criterion can be used to ascertain the closed-loop stability of MPC.

Here we show that Zames-Falb multipliers may be constructed for  $f$ . In particular, we show in the following two lemmas that  $f$  satisfies the conditions of [17] and [15] respectively.

**Lemma 1:** Let  $f$  be given by (3). Then  $f(0) = 0$ ,  $f$  is bounded by 1 and  $f$  is incrementally positive. Furthermore  $f$  is slope-restricted to the interval  $[0, 1]$ .

**Proof:** See Appendix.  $\square$

**Lemma 2:** Let  $f$  be given by (3). Then  $f$  is monotone.

**Proof:** See Appendix.  $\square$

Hence our main result:

**Theorem 1 (Zames-Falb multipliers for quadratic programming):** Let  $f$  be given by (3). Let  $h \in L_1$  satisfy  $\int_{-\infty}^{\infty} |h(t)| dt < 1$  (or let  $h \in l_1$  satisfy  $\sum_{t=-\infty}^{\infty} |h(t)| < 1$ ) and either let  $f$  be odd or let  $h(t) \geq 0$  for all  $t$ . Then

1) For any  $x \in L_2^N$  (or  $x \in l_2^N$ ) we have

$$\langle f(x), x \rangle \geq \langle f(x), h * x \rangle \quad (5)$$

2) For any  $x \in L_2^N$  (or  $x \in l_2^N$ ) and for any  $\varepsilon > 0$  we have

$$\langle f(x), (1 + \varepsilon)x - f(x) \rangle \geq \langle f(x), h * [(1 + \varepsilon)x - f(x)] \rangle \quad (6)$$

**Proof:**

1) This follows immediately from applying ZF1 to  $f$ .

2) This follows from applying the ZF2 to  $f(x) + \alpha x$  with  $\beta = 1 + \alpha + \varepsilon$ .

$\square$

The results can be extended to quadratic programs with more general positive definite Hessians via straightforward substitution. Let the function  $\tilde{f}: \mathbb{R}^N \rightarrow \mathbb{R}^N$  be given by the quadratic program

$$\begin{aligned} \tilde{f}(\tilde{x}) &= \arg \min_{\tilde{n}} \frac{1}{2} \tilde{n}^T G \tilde{n} - \tilde{n}^T \tilde{x} \\ &\text{subject to } \tilde{L} \tilde{n} \preceq \tilde{b} \text{ with } \tilde{b} \succeq 0 \\ &\text{and } \tilde{M} \tilde{n} = 0 \end{aligned} \quad (7)$$

with  $G = G^T > 0$ . This can be transformed to the form of (3) by the substitutions

$$\begin{aligned} f &= G_r \tilde{f} \\ x &= G_r^{-T} \tilde{x} \end{aligned} \quad (8)$$

where  $G_r^T G_r = G$ . Hence we may say:

**Corollary 1 (Zames-Falb multipliers for quadratic programming with generalized Hessian):** Let  $\tilde{f}$  be given by (7). Let  $h \in L_1$  satisfy  $\int_{-\infty}^{\infty} |h(t)| dt < 1$  (or let  $h \in l_1$  satisfy  $\sum_{t=-\infty}^{\infty} |h(t)| < 1$ ) and either let  $\tilde{f}$  be odd or let  $h(t) \geq 0$  for all  $t$ . Then

1) For any  $\tilde{x} \in L_2^N$  (or  $x \in l_2^N$ ) we have

$$\langle \tilde{f}(\tilde{x}), \tilde{x} \rangle \geq \langle \tilde{f}(\tilde{x}), h * \tilde{x} \rangle \quad (9)$$

2) For any  $\tilde{x} \in L_2^N$  (or  $x \in l_2^N$ ) and for any  $\varepsilon > 0$  we have

$$\langle \tilde{f}(\tilde{x}), (1 + \varepsilon)\tilde{x} - G\tilde{f}(\tilde{x}) \rangle \geq \langle \tilde{f}(\tilde{x}), h * [(1 + \varepsilon)\tilde{x} - G\tilde{f}(\tilde{x})] \rangle \quad (10)$$

□

#### IV. RESULTS EXPRESSED AS IQCS

In [13] a unified approach to robustness analysis via IQCs (integral quadratic constraints) was introduced. Such techniques may be used to analyze the robustness of MPC [6]. Here we express the various inequalities of the theorems in standard form. In particular, given  $\phi : L_2^N \rightarrow L_2^N$ , we adopt the notation of [9] and say

$$\phi \in \text{IQC}(\Pi) \quad (11)$$

when

$$\int_{-\infty}^{\infty} \begin{bmatrix} x(j\omega) \\ \phi(x)(j\omega) \end{bmatrix}^* \Pi(j\omega) \begin{bmatrix} x(j\omega) \\ \phi(x)(j\omega) \end{bmatrix} d\omega \geq 0 \quad (12)$$

for all  $x \in L_2^N$ . Similarly if  $\phi : l_2^N \rightarrow l_2^N$  we say

$$\phi \in \text{IQC}(\Pi) \quad (13)$$

when

$$\int_{-\pi}^{\pi} \begin{bmatrix} x(e^{j\omega}) \\ \phi(x)(e^{j\omega}) \end{bmatrix}^* \Pi(e^{j\omega}) \begin{bmatrix} x(e^{j\omega}) \\ \phi(x)(e^{j\omega}) \end{bmatrix} d\omega \geq 0 \quad (14)$$

for all  $x \in l_2^N$ .

Let  $H$  be the continuous (or discrete) Fourier transform of  $h$ . Then following [2] we may say:

**Corollary 2 (Results expressed as IQCs):**

1) Inequality (1) may be expressed as  $\phi \in \text{IQC}(\Pi)$  with

$$\Pi = \begin{bmatrix} 0 & (1-H^*)I \\ (1-H)I & 0 \end{bmatrix} \quad (15)$$

2) Inequality (2) may be expressed as  $\phi \in \text{IQC}(\Pi)$  with

$$\Pi = \begin{bmatrix} -\alpha\beta(2-H-H^*)I & \alpha(1-H)I + \beta(1-H^*)I \\ \alpha(1-H^*)I + \beta(1-H)I & (-2+H+H^*)I \end{bmatrix} \quad (16)$$

3) Inequality (6) may be expressed as  $f \in \text{IQC}(\Pi)$  with

$$\Pi = \begin{bmatrix} 0 & (1+\varepsilon)(1-H^*)I \\ (1+\varepsilon)(1-H)I & (-2+H+H^*)I \end{bmatrix} \quad (17)$$

4) Inequality (10) may be expressed as  $\tilde{f} \in \text{IQC}(\Pi)$  with

$$\Pi = \begin{bmatrix} 0 & (1+\varepsilon)(1-H^*)I \\ (1+\varepsilon)(1-H)I & (-2+H+H^*)G \end{bmatrix} \quad (18)$$

□

## V. CONCLUSION

We have shown that Zames-Falb multipliers may be applied to the nonlinearity that arises in constrained linear MPC, provided 0 is feasible and the nonlinearity is static. The result may be incorporated into a robustness analysis of input constrained MPC based on IQCs [6].

## APPENDIX: PROOF OF THE LEMMAS

### Proof of Lemma 1:

- 1) Since  $b \succeq 0$ , zero is feasible and it follows trivially that  $f(0) = 0$ .
- 2) The Karoush Kuhn Tucker (KKT) conditions [5] for  $f$  give

$$\begin{aligned} f(x) + M^T \zeta(x) + L^T \lambda(x) - x &= 0 \\ Mf(x) &= 0 \\ Lf(x) + s(x) &= b \\ s(x)^T \lambda(x) &= 0 \end{aligned} \quad (19)$$

with  $s(x) \succeq 0$  and  $\lambda(x) \succeq 0$ . Substituting for  $x$  (and  $y$ ) from the first condition gives:

$$\begin{aligned}
& [f(x) - f(y)]^T (x - y) \\
&= [f(x) - f(y)]^T [f(x) + M^T \zeta(x) + L^T \lambda(x) - f(y) - M^T \zeta(y) - L^T \lambda(y)] \\
&= \|f(x) - f(y)\|^2 + [f(x) - f(y)]^T M^T [\zeta(x) - \zeta(y)] + [(f(x) - f(y))]^T L^T [\lambda(x) - \lambda(y)] \\
&= \|f(x) - f(y)\|^2 + 0 + [s(y) - s(x)]^T [\lambda(x) - \lambda(y)] \\
&= \|f(x) - f(y)\|^2 + s(y)^T \lambda(x) + s(x)^T \lambda(y) \\
&\geq 0
\end{aligned} \tag{20}$$

- 3) Given  $x$ , suppose  $f_x$  is feasible with  $\|f_x\|^2 > \|x\|^2$  and hence  $\|f_x\|^2 > f_x^T x$ . Let  $\tilde{f}_x = (1 - \varepsilon)f_x$  with  $\varepsilon > 0$ . By convexity  $\tilde{f}_x$  is also feasible. Put  $J_x(f) = \frac{1}{2}f^T f - f^T x$ . Then

$$\begin{aligned}
J_x(\tilde{f}_x) &= \frac{1}{2}(1 - \varepsilon)^2 f_x^T f_x - (1 - \varepsilon)f_x^T x \\
&= J_x(f_x) - \varepsilon(f_x^T f_x - f_x^T x) + \frac{1}{2}\varepsilon^2 f_x^T f_x \\
&< J_x(f_x) \text{ for } \varepsilon \text{ sufficiently small}
\end{aligned} \tag{21}$$

Hence  $\tilde{f}_x$  cannot be equal to  $f(x)$  and so we must have  $\|f(x)\|^2 \leq \|x\|^2$ .

- 4) Given  $x, y$  suppose  $f_x$  and  $f_y$  are feasible with  $(f_x - f_y)^T (x - y) < \|f_x - f_y\|^2$ . Let  $\tilde{f}_x = f_x + \varepsilon(f_y - f_x)$  and  $\tilde{f}_y = f_y + \varepsilon(f_x - f_y)$  for some  $\varepsilon > 0$ . By convexity  $\tilde{f}_x$  and  $\tilde{f}_y$  are also feasible. Furthermore

$$\begin{aligned}
& J_x(\tilde{f}_x) + J_y(\tilde{f}_y) \\
&= J_x(f_x) + \varepsilon f_x^T (f_y - f_x) + \frac{1}{2}\varepsilon^2 \|f_x - f_y\|^2 - \varepsilon(f_y - f_x)^T x \\
&\quad + J_y(f_y) + \varepsilon f_y^T (f_x - f_y) + \frac{1}{2}\varepsilon^2 \|f_y - f_x\|^2 - \varepsilon(f_x - f_y)^T y \\
&= J_x(f_x) + J_y(f_y) - \varepsilon(1 - \varepsilon)\|f_x - f_y\|^2 + \varepsilon(f_x - f_y)^T (x - y) \\
&< J_x(f_x) + J_y(f_y) \text{ for } \varepsilon \text{ sufficiently small}
\end{aligned} \tag{22}$$

Hence  $\tilde{f}_x$  and  $\tilde{f}_y$  cannot be concurrently equal to  $f(x)$  and  $f(y)$ , so we must have  $[f(x) - f(y)]^T (x - y) \geq \|f(x) - f(y)\|^2$ , and hence  $[f(x) - f(y)]^T [f(x) - f(y) - (x - y)] \leq 0$ .

□

**Proof of Lemma 2:** It is well-known (e.g. [1]) that  $f$  is both piecewise affine [16] and continuous [4]. Within each affine region  $R_i$ , the KKT conditions give (c.f. [16], [5])

$$\begin{bmatrix} I & L_i^T & M^T \\ L_i & 0 & 0 \\ M & 0 & 0 \end{bmatrix} \begin{bmatrix} f_i(x) \\ \lambda_i(x) \\ \zeta(x) \end{bmatrix} = \begin{bmatrix} x \\ b_i \\ 0 \end{bmatrix} \quad (23)$$

where  $f = f_i$  within region  $R_i$ , where  $L_i$  and  $b_i$  correspond to the constraints which are active in the region (i.e.  $L_i$  is made up of rows of  $L$ , and  $b_i$  is made up of the corresponding entries in  $b$ , and we have the equality  $f_i(x) = b_i$  within  $R_i$ ), and where  $\lambda_i(x)$  are the corresponding Lagrange multipliers. Thus the piecewise affine control law is

$$f_i(x) = F_i x + a_i \quad (24)$$

with

$$\begin{aligned} F_i &= I - \begin{bmatrix} L_i \\ M \end{bmatrix}^T \left( \begin{bmatrix} L_i \\ M \end{bmatrix} \begin{bmatrix} L_i \\ M \end{bmatrix}^T \right)^{-1} \begin{bmatrix} L_i \\ M \end{bmatrix} \\ a_i &= \begin{bmatrix} L_i \\ M \end{bmatrix}^T \left( \begin{bmatrix} L_i \\ M \end{bmatrix} \begin{bmatrix} L_i \\ M \end{bmatrix}^T \right)^{-1} \begin{bmatrix} b_i \\ 0 \end{bmatrix} \end{aligned} \quad (25)$$

In particular  $F_i = F_i^T \geq 0$  and within each region  $R_i$  we can form the potential function

$$P_i(x) = \frac{1}{2} x^T F_i x + x^T a_i \quad (26)$$

such that

$$f_i(x) = \nabla P_i(x) \quad (27)$$

This is sufficient to show the integral is independent of path within each region [15].

So it suffices to show that if  $x_A$  and  $x_B$  lie on the border of adjoining regions  $R_i$  and  $R_j$ , then the integral from  $A$  to  $B$  is the same via region  $R_i$  or region  $R_j$ —see Fig 1.

Let  $I_i$  be the integral via region  $R_i$ , and  $I_j$  be the integral via region  $R_j$ . Thus

$$I_i = \int_A^B (F_i x + a_i)^T dx = \frac{1}{2} x_B^T F_i x_B + x_B^T a_i - \frac{1}{2} x_A^T F_i x_A - x_A^T a_i \quad (28)$$

and similarly

$$I_j = \int_A^B (F_j x + a_j)^T dx = \frac{1}{2} x_B^T F_j x_B + x_B^T a_j - \frac{1}{2} x_A^T F_j x_A - x_A^T a_j \quad (29)$$



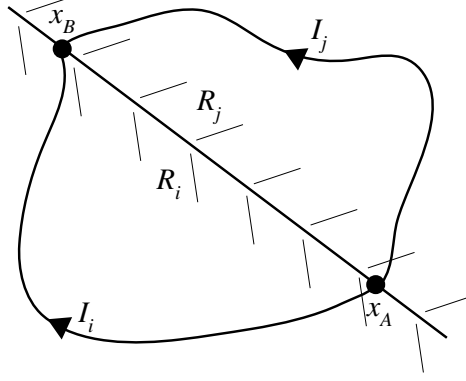


Fig. 1. This illustrates the final part of the proof of Lemma 4. Given  $x_A$  and  $x_B$  on the border of adjoining regions  $R_i$  and  $R_j$ , the integral from  $x_A$  to  $x_B$  is independent of choice of path.

Since  $x_A$  and  $x_B$  lie on the boundary of region  $R_i$  and  $R_2$  we have the relations

$$\begin{aligned} F_i x_A + a_i &= F_j x_A + a_j \\ F_i x_B + a_i &= F_j x_B + a_j \end{aligned} \quad (30)$$

In particular

$$a_i - a_j = (F_j - F_i)x_A = (F_j - F_i)x_B \quad (31)$$

Thus

$$\begin{aligned} I_j - I_i &= \frac{1}{2}x_B^T F_j x_B + x_B^T a_j - \frac{1}{2}x_A^T F_j x_A - x_A^T a_j - \frac{1}{2}x_B^T F_i x_B - x_B^T a_i + \frac{1}{2}x_A^T F_i x_A + x_A^T a_i \\ &= \frac{1}{2}x_B^T (F_j - F_i)x_B + \frac{1}{2}x_A^T (F_i - F_j)x_A + (x_A - x_B)^T (a_i - a_j) \\ &= \frac{1}{2}(x_A - x_B)^T (a_i - a_j) \\ &= \frac{1}{2}(x_A - x_B)^T (F_j - F_i)x_A \\ &= 0 \end{aligned} \quad (32)$$

as required.  $\square$

## REFERENCES

- [1] A. Bemporad, M. Morari, V. Dua, and E. N. Pistikopoulos. The explicit linear quadratic regulator for constrained systems. *Automatica*, 38:3–20, 2002.

- [2] F. J. D'Amato, M. A. Rotea, A. V. Megretski, and U. T. Jönsson. New results for analysis of systems with repeated nonlinearities. *Automatica*, 37(5):739–747, 2001.
- [3] C. A. Desoer and M. Vidyasagar. *Feedback systems: input-output properties*. Academic Press, New York, 1975.
- [4] A. V. Fiacco. *Introduction to sensitivity and stability analysis in nonlinear programming*. Academic Press, Orlando, Florida, 1983.
- [5] R. Fletcher. *Practical Methods of Optimization*. John Wiley & Sons Inc., New York, 1987.
- [6] W. P. Heath, G. Li, A. G. Wills, and B. Lennox. The robustness of input constrained model predictive control to infinity-norm bound model uncertainty. ROCOND06, 5th IFAC Symposium on Robust Control Design, Toulouse, France, July 5-7, 2006.
- [7] W. P. Heath and A. G. Wills. The inherent robustness of constrained linear model predictive control. 16th IFAC World Congress, Prague, July 3rd-8th, 2005.
- [8] W. P. Heath, A. G. Wills, and J. A. G. Akkermans. A sufficient stability condition for optimizing controllers with saturating actuators. *Int. J. Robust Nonlinear Control*, 15:515–529, 2005.
- [9] U. Jönsson. Lecture notes on integral quadratic constraints. Department of Mathematics, KTH, Stockholm. ISBN 1401-2294, 2000.
- [10] V. V. Kulkarni and M. G. Safonov. All multipliers for repeated monotone nonlinearities. *IEEE Trans-AC*, 47(7):1209–1212, 2002.
- [11] V. V. Kulkarni and M. G. Safonov. Incremental positivity nonpreservation by stability multipliers. *IEEE Trans-AC*, 47(1):173–177, 2002.
- [12] R. Mancera and M. G. Safonov. All stability multipliers for repeated MIMO nonlinearities. *Systems and Control Letters*, 54(4):389–397, 2005.
- [13] A. Megretski and A. Rantzer. System analysis via integral quadratic constraints. *IEEE Trans-AC*, 42(6):819–830, 1997.
- [14] R. T. Rockafellar. *Convex Analysis*. Princeton University Press, Princeton, 1970.
- [15] M. G. Safonov and V. V. Kulkarni. Zames-Falb multipliers for MIMO nonlinearities. *Int. J. Robust Nonlinear Control*, 10:1025–1038, 2000.
- [16] E. Zafiriou. Robust model predictive control of processes with hard constraints. *Computers Chem. Engng.*, 14(4–5):359–371, May 1990.
- [17] G. Zames and P.L. Falb. Stability conditions for systems with monotone and slope-restricted nonlinearities. *SIAM J. Control*, 6(1):89–108, 1968.