Robust Model Predictive Control via Scenario Optimization Technical Report # TR_CaFa_01122011, December 1st 2011 Proof of Theorem 4.1

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Preliminaries. Notice first that, for any $t \ge 0$, if $x_t \in \mathbb{X}_f$ then the optimal solution to problem $\mathcal{P}(x_t, \omega_t)$ is $(\mathcal{V}_t^*, z_t^*, q_t^*) = (0, 0, 0)$, since the terminal control law (i.e., with $\mathcal{V}_t = 0$) is able to keep the predicted state trajectory in the terminal set while satisfying all constraints. Also, if $z_t^* = 0$ is the optimal objective of problem $\mathcal{P}(x_t, \omega_t)$, then $x_t \in \mathbb{X}_f$, since z_t^* is an upper bound of $d(x_t, \mathbb{X}_f)$ (see also Remark 3.1), therefore $z_t^* = 0 \iff x_t \in \mathbb{X}_f$. Let then $x_t \notin \mathbb{X}_f$.

Proof of statement (a). At time t = 0, Proposition 3.1 guarantees with practical certainty that the first control correction satisfies the constraints on u_0 and x_1 with probability no less than p and constraint violation $q_t = q_t^*$. At any generic time step $t \ge 1$, the variables $(\tilde{\mathcal{V}}_t, \tilde{z}_t, \tilde{q}_t)$ are computed. Then, two cases may occur. If $z_t^* \le (z_{t-1} - \varepsilon d(x_{t-1}, \mathbb{X}_f))$, then case 3.c) is detected, and the first element $v_{0|t}^*$ of the optimal sequence \mathcal{V}_t^* is applied to the system. Being this sequence the solution of a scenario optimization problem, with practical certainty the probability of satisfying state and input constraints is no less than p, with constraint violation $q_t = q_t^*$. If, on the other hand, $z_t^* > (z_{t-1} - \varepsilon d(x_{t-1}, \mathbb{X}_f))$, then we are either in case 3.a) or 3.b), and in both cases the element $v_{k|t-k}^*$, for some $k \in [1, N - 1]$, is applied to the system. Being this value part of the solution sequence \mathcal{V}_{t-k}^* , with corresponding constraint violation $q_t^* = \tilde{q}_t = q_{t-k}^*$.

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Thus, in any case, with practical certainty, at each time step the MPCS algorithm guarantees satisfaction of state and input constraints with probability no less than p and constraint violation q_t .

Proof of statement (b). Each run of Algorithm 4.1 may have one of two possible behaviors, depending on whether or not there exists a finite time t > 0 such that $z_t^* > (z_{t-1} - \varepsilon d(x_{t-1}, \mathbb{X}_f))$ and $\tilde{z}_t < d(x_t, \mathbb{X}_f)$, that is, whether or not the situation in step 3.a is ever satisfied. We then name \mathcal{A} the situation when condition in step 3.a is met at some finite t > 0, and $\overline{\mathcal{A}}$ the complementary situation when this condition is not satisfied at any finite time, that is when $z_t^* \leq (z_{t-1} - \varepsilon d(\xi_{t-1}, \mathbb{X}_f))$ or $\tilde{z}_t \geq d(\xi_t, \mathbb{X}_f)$ holds for all t > 0.

I. Let us first consider the situation of case \overline{A} . Consider a generic time t. At step 3) of the MPCS algorithm, if $z_t^* > (z_{t-1} - \varepsilon d(x_{t-1}, \mathbb{X}_f))$, then, since it is assumed that we are in situation \overline{A} , it must hold that $\tilde{z}_t \ge d(x_t, \mathbb{X}_f)$, thus case 3.b) occurs, and the values $\mathcal{V}_t = \tilde{\mathcal{V}}_t$ and $z_t = \tilde{z}_t$ are set. Now, recalling that $\tilde{z}_t = \max(0, z_{t-1} - d(x_{t-1}, \mathbb{X}_f))$, two cases may occur: either $\tilde{z}_t = 0$ or $\tilde{z}_t = z_{t-1} - d(x_{t-1}, \mathbb{X}_f) > 0$. If $\tilde{z}_t = 0$, we have $0 = \tilde{z}_t \ge d(x_t, \mathbb{X}_f)$, i.e. $d(x_t, \mathbb{X}_f) = 0$, which would imply that the terminal set has been reached. Otherwise, if $\tilde{z}_t = z_{t-1} - d(x_{t-1}, \mathbb{X}_f) > 0$, then we have:

$$z_t = \tilde{z}_t \ge d(x_t, \mathbb{X}_f) \ge 0, \tag{.1}$$

and
$$z_t - z_{t-1} = \tilde{z}_t - z_{t-1} = z_{t-1} - d(x_{t-1}, \mathbb{X}_f) - z_{t-1} = -d(x_{t-1}, \mathbb{X}_f).$$

Thus, $z_t - z_{t-1} \le -\varepsilon d(x_{t-1}, \mathbb{X}_f), \ \forall x_{t-1} \notin \mathbb{X}_f,$ (.2)

and
$$z_t - z_{t-1} = 0 \iff x_{t-1} \in \mathbb{X}_f.$$
 (.3)

On the other hand, if at step 3) of the MPCS algorithm it happens that $z_t^* \leq (z_{t-1} - \varepsilon d(x_{t-1}, \mathbb{X}_f))$, then case 3.c) occurs, and the optimal values \mathcal{V}_t^* and z_t^* are retained, i.e. $z_t = z_t^*$, $\mathcal{V}_t = \mathcal{V}_t^*$. In this case, it is straightforward to note that equations (.1)-(.3) still hold true. The same reasoning can be repeated for any time step, as long as the case $z_t^* \leq (z_{t-1} - \varepsilon d(x_{t-1}, \mathbb{X}_f))$ or $\tilde{z}_t \geq d(x_t, \mathbb{X}_f)$ holds true as assumed, so that we can conclude that the variable z_t enjoys the following properties:

$$z_{t} \geq d(x_{t}, \mathbb{X}_{f}) \geq 0, \forall t \geq 0$$

$$z_{t} = 0 \iff x_{t} \in \mathbb{X}_{f}$$

$$z_{t+1} - z_{t} \leq -\varepsilon d(x_{t}, \mathbb{X}_{f}), \forall x_{t} \notin \mathbb{X}_{f}, \forall t \geq 0$$

$$z_{t+1} - z_{t} = 0 \iff x_{t} \in \mathbb{X}_{f}.$$
(.4)

Properties (.4) are sufficient to prove convergence of the state x_t to the set X_f :

$$0 \le \lim_{t \to \infty} d(x_t, \mathbb{X}_f) \le \lim_{t \to \infty} z_t = 0, \Rightarrow \lim_{t \to \infty} d(x_t, \mathbb{X}_f) = 0.$$

Therefore, we obtain that in case \overline{A} the MPCS algorithm guarantees that $\lim_{t\to\infty} d(x_t, \mathbb{X}_f) = 0$. II. Let us next analyze what happens in case A. Let $\overline{t} > 0$ be the time instant at which the case $z_t^* > (z_{t-1} - \varepsilon d(x_{t-1}, \mathbb{X}_f))$ and $\tilde{z}_t < d(x_t, \mathbb{X}_f)$ is met for the first time, and let $t^* < \overline{t}$ be the last time at which case $z_t^* \leq (z_{t-1} - \varepsilon d(x_{t-1}, \mathbb{X}_f))$ was satisfied, that is the last time previous to \overline{t} when an optimal command sequence was retained, together with its constraint violation q_t^* , according to case 3.c) of Algorithm 4.1; let $\ell = \overline{t} - t^* \geq 1$. According to step 3.a) of the MPCS algorithm, we set

$$\mathcal{V}_{\bar{t}} = \mathcal{V}_{\bar{t}}, \ z_{\bar{t}} = 0, \ q_{\bar{t}} = \tilde{q}_{\bar{t}}.$$
(.5)

Thus, at step 4) of the algorithm, the control move $u_{\bar{t}} = K_f x_{\bar{t}} + v_{0|\bar{t}}$ is applied to the system at time \bar{t} , where $v_{0|\bar{t}} = v^*_{\ell|t^*}$, i.e., $v_{0|\bar{t}}$ is the optimal correction predicted for time $t^* + \ell = \bar{t}$, computed at time t^* . At time step $t = \bar{t} + 1$, the state variable $x_{\bar{t}+1}$ is observed and $(\tilde{V}_{\bar{t}+1}, \tilde{z}_{\bar{t}+1}, \tilde{q}_{\bar{t}+1})$ are computed as $\tilde{z}_{\bar{t}+1} = \max(0, z_{\bar{t}} - d(x_{\bar{t}}, \mathbb{X}_f)), \tilde{q}_{\bar{t}+1} = q_{\bar{t}}, \tilde{\mathcal{V}}_{\bar{t}+1} = q_{\bar{t}}, \tilde{\mathcal{V}}_{\bar{t}+1}$ $\{v_{1:N-1|\bar{t}},0\} = \{v^*_{\ell+1|t^*}, v^*_{\ell+2|t^*}, \dots, v^*_{N-1|t^*}, 0, \dots, 0\}.$ Since (.5) holds, it must be $\tilde{z}_{\bar{t}+1} = 0.$ Then, z_{t+1}^* , q_{t+1}^* and \mathcal{V}_{t+1}^* are computed at step 2), and we notice that, by definition, $z_{t+1}^* \ge 0$. Therefore, at step 3) of the algorithm either (i) case 3.a) $\{z_{\bar{t}+1}^* > (z_{\bar{t}} - \varepsilon d(x_{\bar{t}}, \mathbb{X}_f)) \text{ and } \tilde{z}_{\bar{t}+1} < \varepsilon d(x_{\bar{t}}, \mathbb{X}_f)\}$ $d(x_{t+1}, \mathbb{X}_f)$ is detected again, or (*ii*) one of cases 3.b) or 3.c) are detected, which would imply, respectively, $0 = \tilde{z}_{\bar{t}+1} \ge d(x_{\bar{t}+1}, \mathbb{X}_f)$, or $0 \le d(x_{\bar{t}+1}, \mathbb{X}_f) \le z^*_{\bar{t}+1} = \tilde{z}_{\bar{t}+1} = 0$. Hence (in either case) $x_{t+1} \in \mathbb{X}_f$, so that convergence to the terminal set would be achieved. Consider then case (i): the values $\mathcal{V}_{\bar{t}+1} = \tilde{\mathcal{V}}_{\bar{t}+1}$, $z_{\bar{t}+1} = 0$ and $q_{\bar{t}+1} = q_{\bar{t}}$ are set in the algorithm, and the control move $u_{\bar{t}+1} = K_f x_{\bar{t}+1} + v_{\ell+1|t^*}^*$ is applied to the system. Now, the same circumstances actually reproduce for all time steps $t = \bar{t} + k$, $k \ge 0$, so the algorithm is such that the optimal input sequence $\mathcal{V}_{t^*}^*$, computed at time t^{*} by solving a scenario FHOCP, is the one actually next applied to the system, and the related constraint violation q_t^* is retained for all $t \ge t^*$. Thus, in case \mathcal{A} , there exists a finite time t^* such that the sequence $\mathcal{V}_{t^*}^*$ is applied to the system for all subsequent instants $t = t^* + k$, k = 0, ..., N - 1. Now, the sequence $\mathcal{V}_{t^*}^*$ is the result of the solution of the scenario-FHOCP $\mathcal{P}(x_{t^*}, \omega_{t^*})$, and Proposition 3.1 states that, with practical certainty, we have $R(\omega_{t^*}) \geq p$, where R is the reliability defined in Section III-A of the paper, which means that $\mathbb{P}\{\delta: h(s_{t^*}^*, x_{t^*}, \delta) \leq 0\} \geq p$. Therefore, in the situation \mathcal{A} , there exists a finite time t^* at which an optimal control sequence is computed by solving a scenario-FHOPC and next applied to the actual system for the subsequent N time instants: we can hence claim with practical certainty this sequence will satisfy the problem constraints and reach the terminal set within the time window from t^* to $t^* + N$, with probability at least p and constraint violation q_t^* .