CS159 Lecture 4 Supplementary Material: Synthesizing Terminal Components for MPC

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Spring 2021

Adapted from Berkeley ME231A Original slide set by F. Borrelli, M. Morari, C. Jones

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Polyhedra and polytopes

Polyhedra and polytopes

A polyhedron is the intersection of a *finite* number of closed halfspaces:

$$Z = \{ z \mid a_1^\top z \le b_1, \ a_2^\top z \le b_2, \dots, a_m^\top z \le b_m \}$$

= \{ z \left| Az \leftleq b \}

where $A := [a_1, a_2, \dots, a_m]^{\top}$ and $b := [b_1, b_2, \dots, b_m]^{\top}$.

A polytope is a bounded polyhedron.

Polyhedra and polytopes are always convex.



An (unbounded) polyhedron



A polytope

Polyhedra Representations

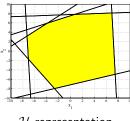
ightharpoonup An \mathcal{H} -polyhedron \mathcal{P} in \mathbb{R}^n denotes an intersection of a finite set of closed halfspaces in \mathbb{R}^n :

$$\mathcal{P} = \{ x \in \mathbb{R}^n : Ax \le b \}$$

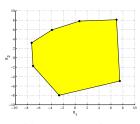
ightharpoonup A \mathcal{V} -polytope \mathcal{P} in \mathbb{R}^n is defined as

$$\mathcal{P} = \operatorname{conv}(V) = \{ v \in \mathbb{R}^n | \exists \lambda \in \mathbb{R}^k, v = V\lambda, 1_k^\top \lambda = 1, \lambda \ge 0 \}$$

for some $V = [V_1, \dots, V_k] \in \mathbb{R}^{n \times k}$ and the vector of ones $1_k \in \mathbb{R}^k$.



 \mathcal{H} -representation



 \mathcal{V} -representation

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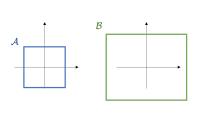
Basic Operations on Polytopes

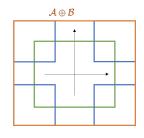
▶ Given two sets $\mathcal{A} \subset \mathbb{R}^n$ and $\mathcal{B} \subset \mathbb{R}^n$, the Minkowski sum of \mathcal{A} and \mathcal{B} is defined as

$$\mathcal{A} \oplus \mathcal{B} = \{ x + y \in \mathbb{R}^n \mid x \in \mathcal{A}, y \in \mathcal{B} \}$$

Furthermore, given the V-representations $\mathcal{A} = \operatorname{conv}([v_1^a, \dots, v_a^a])$ and $\mathcal{B} = \operatorname{conv}([v_1^b, \dots, v_b^b])$ the Minkowski sum

$$\mathcal{A} \oplus \mathcal{B} = \mathcal{A} = \mathsf{conv}([v_{1,1}^{ab}, \dots, v_{a,b}^{ab}]), \text{ where } v_{ij}^{ab} = v_i^a + v_j^b.$$





Basic Operations on Polytopes

▶ Projection Given a polytope $\mathcal{P} = \{[x'y']' \in \mathbb{R}^{n+m} : A^x x + A^y y \leq b\} \subset \mathbb{R}^{n+m} \text{ the projection onto the } x\text{-space } \mathbb{R}^n \text{ is defined as}$

$$\operatorname{proj}_x(\mathcal{P}) := \{ x \in \mathbb{R}^n \mid \exists y \in \mathbb{R}^m : A^x x + A^y y \le b \}.$$

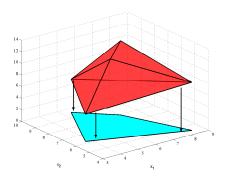


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Reachable Set

Reachable Set for a policy π

Consider the discrete-time system $x_{k+1} = f(x_k, u_k)$ and the state constraint set \mathcal{X} . The <u>Reachable Set for a policy π </u> from the set \mathcal{S} is defined as

$$\mathsf{Reach}_{\pi}(\mathcal{S}) \triangleq \{x \in \mathbb{R}^n \mid \exists \ x_0 \in \mathcal{S} \ \text{s.t.} \ x = f(x_0, \pi(x_0))\}$$

Consider the discrete-time system $x_{k+1} = f(x_k, u_k)$, the state constraint set \mathcal{X} and input constraint set \mathcal{U} . The <u>Reachable Set</u> from the set \mathcal{S} is defined as

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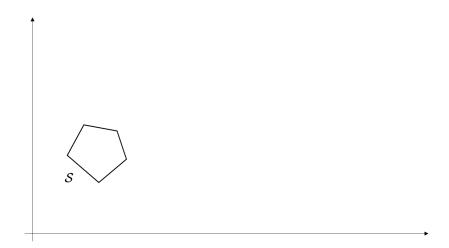
$$\operatorname{\mathsf{Reach}}_{\pi}(\mathcal{S}) \triangleq \{ x \in \mathbb{R}^n \mid \exists \ x_0 \in \mathcal{S} \text{ s.t. } x = f(x_0, \pi(x_0)) \}$$

Reachable Set

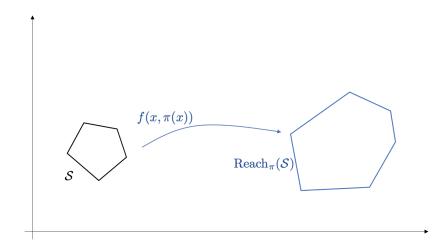
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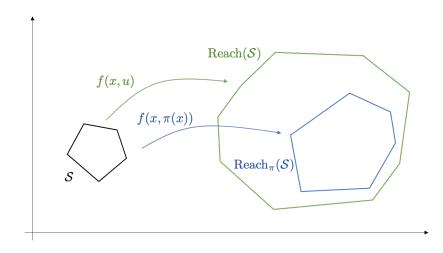
${\sf Reachable\ Set-Example}$



Reachable Set – Example



Reachable Set – Example



Reach Set Computation

▶ Consider the polyhedron $\mathcal{X} = \mathsf{conv}(V_{\mathsf{x}})$ and the linear discrete time system

$$x(t+1) = Ax(t) + Bu(t)$$

where the input $u \in \mathcal{U} = \operatorname{conv}(V_u)$ and define

$$A \circ \mathcal{X} = \operatorname{conv}(AV_x).$$

▶ Then for the policy $\pi(x) = Kx$

$$\mathsf{Reach}_\pi(\mathcal{X}) = (A - BK) \circ \mathcal{X}$$

and

$$\mathsf{Reach}(\mathcal{X}) = \{ \bar{x} + \bar{u} \mid \bar{x} \in A \circ V_x, \ \bar{u} \in B \circ \mathcal{U} \}$$
$$= (A \circ \mathcal{X}) \oplus (B \circ \mathcal{U}).$$

N-Step Reachable Sets

Definition (*N*-Step Reachable Set $\mathcal{R}_N(\mathcal{S})$)

Consider the discrete-time system $x_{k+1} = f(x_k, u_k)$, the state constraint set \mathcal{X} and input constraint set \mathcal{U} . For a given initial set $\mathcal{S} \subseteq \mathcal{X}$, the N-step reachable set $\mathcal{R}_N(\mathcal{S})$ is

$$\mathcal{R}_{i+1}(\mathcal{S}) \triangleq \mathsf{Reach}(\mathcal{R}_i(\mathcal{S})), \quad \mathcal{R}_0(\mathcal{S}) = \mathcal{S}, \quad i = 0, \dots, N-1$$

By definition all states $x_0 \in \mathcal{S}$ will evolve to the *N*-step reachable set $\mathcal{R}_N(\mathcal{S})$ in *N* time steps.

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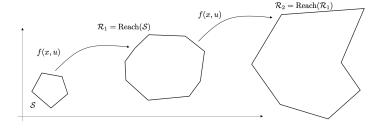


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Pre Set Definition

Pre Set for the policy π

Consider the discrete-time system $x_{k+1} = f(x_k, u_k)$ and the state constraint set \mathcal{X} . The <u>Pre Set for a policy π </u> from the set \mathcal{S} is defined as

$$\mathsf{Pre}_{\pi}(\mathcal{S}) \triangleq \{x \in \mathbb{R}^n \mid f(x, \pi(x)) \in \mathcal{S}\}$$

Pre Set

Consider the discrete-time system $x_{k+1} = f(x_k, u_k)$, the state constraint set \mathcal{X} and input constraint set \mathcal{U} . The <u>Pre Set</u> from the set \mathcal{S} is defined as

$$\mathsf{Pre}_{\pi}(\mathcal{S}) \triangleq \{x \in \mathbb{R}^n \mid \exists u \in \mathcal{U}, f(x, u) \in \mathcal{S}\}$$

Pre Set Definition

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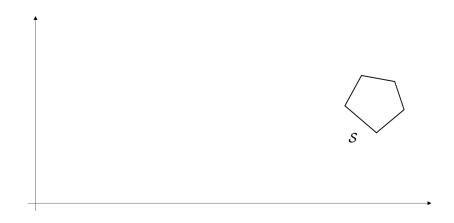
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Pre Set

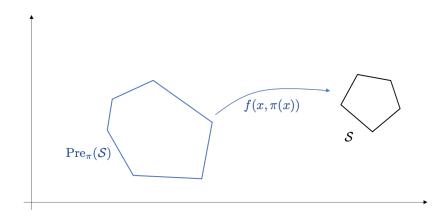
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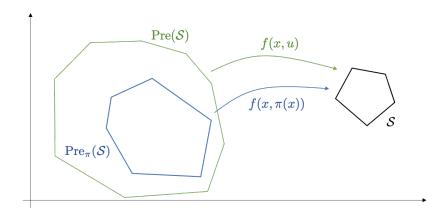
$\mathsf{Pre}\;\mathsf{Set}-\mathsf{Example}$



Pre Set – Example



Pre Set – Example



Pre Set Computation - Autonomous Systems

▶ Consider the polyhedron $\mathcal{X} = \{x \mid H_x x \leq h_x\}$ and the linear discrete time autonomous system

$$x(t+1) = Ax(t) + Bu(t)$$

► Then

$$\operatorname{Pre}(\mathcal{S}) = \{x \mid HAx \le h\}$$

Pre Set Computation - System with Inputs

▶ Consider the polyhedron $\mathcal{X} = \{x \mid H_x x \leq h_x\}$ and the linear discrete time system

$$x(t+1) = Ax(t) + Bu(t)$$

where the input $u \in \mathcal{U} = \{u \mid H_u u \leq h_u\}$ and define

$$A \circ \mathcal{X} = \operatorname{conv}(AV_x).$$

► Then

$$\operatorname{Pre}(\mathcal{S}) = \left\{ x \in \mathbb{R}^n \mid \exists u \in \mathbb{R} \mid \begin{bmatrix} H_x A & H_x B \\ 0 & H_u \end{bmatrix} \begin{pmatrix} x \\ u \end{pmatrix} \leq \begin{bmatrix} h_x \\ h_u \end{bmatrix} \right\}$$

which is the projection onto the *x*-space (with dimension \mathbb{R}^n) of the polyhedron

$$\mathcal{T} := \left\{ \begin{bmatrix} HA & HB \\ 0 & H_u \end{bmatrix} \begin{pmatrix} x \\ u \end{pmatrix} \le \begin{bmatrix} h \\ h_u \end{bmatrix} \right\}.$$

N-Step Controllable Sets

N-Step Controllable Set $\mathcal{K}_N(\mathcal{S})$

For a given target set $S \subseteq \mathcal{X}$, the N-step controllable set $\mathcal{K}_N(S)$ is defined as:

$$\mathcal{K}_N(\mathcal{S}) \triangleq \mathsf{Pre}(\mathcal{K}_{N-1}(\mathcal{S})) \cap \mathcal{X}, \ \mathcal{K}_0(\mathcal{S}) = \mathcal{S}, \ N \in \mathbb{N}^+.$$

By definition all states $x_0 \in \mathcal{K}_N(\mathcal{S})$ can be driven, through a time-varying control law, to the target set \mathcal{O} in N steps, while satisfying input and state constraints.

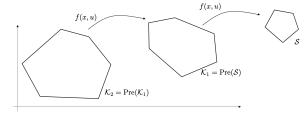
N-Step Controllable Sets

N-Step Controllable Set $\mathcal{K}_N(\mathcal{S})$

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By definition all states $x_0 \in \mathcal{K}_N(\mathcal{S})$ can be driven, through a time-varying control law, to the target set \mathcal{O} in N steps, while satisfying input and state constraints.



Maximal Controllable Set

Maximal Controllable Set $\mathcal{K}_{\infty}(\mathcal{S})$

For a given target set $\mathcal{O}\subseteq\mathcal{X}$, the maximal controllable set $\mathcal{K}_{\infty}(\mathcal{S})$ for the system x(t+1)=f(x(t),u(t)) subject to the constraints $x(t)\in\mathcal{X},\ u(t)\in\mathcal{U}$ is the union of all N-step controllable sets contained in \mathcal{X} ($N\in\mathbb{N}$).

As we will be discussing, Maximal Controllable Set characterize the MPC region of attraction. However, computing these set may be challenging as these sets are computed using projections.

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Invariant Sets

Invariant sets

- are computed for autonomous systems
- for a given feedback controller $u = \pi(x)$, will contain the evolution of the system for all times.

Positive Invariant Set

A set $\mathcal{O}\subseteq\mathcal{X}$ is said to be a positive invariant set for the autonomous system $x(t+1)=f(x(t),\pi(x(t)))$ subject to the constraints $x(t)\in\mathcal{X}$, if

$$x(0) \in \mathcal{O} \quad \Rightarrow \quad x(t) \in \mathcal{O}, \quad \forall t \in \mathbb{N}^+$$

Maximal Positive Invariant Set

The set \mathcal{O}_{∞} is the maximal invariant set if \mathcal{O}_{∞} is invariant and \mathcal{O}_{∞} contains all the invariant sets contained in \mathcal{X} .

Invariant Sets

Invariant sets

- are computed for autonomous systems
- for a given feedback controller $u = \pi(x)$, will contain the evolution of the system for all times.

Positive Invariant Set

A set $\mathcal{O} \subseteq \mathcal{X}$ is said to be a positive invariant set for the autonomous system $x(t+1) = f(x(t), \pi(x(t)))$ subject to the constraints $x(t) \in \mathcal{X}$, if

$$x(0) \in \mathcal{O} \quad \Rightarrow \quad x(t) \in \mathcal{O}, \quad \forall t \in \mathbb{N}^+$$

Maximal Positive Invariant Set \mathcal{O}_{∞}

The set \mathcal{O}_{∞} is the maximal invariant set if \mathcal{O}_{∞} is invariant and \mathcal{O}_{∞} contains all the invariant sets contained in \mathcal{X} .

Invariant Sets

Theorem (Geometric condition for invariance)

A set \mathcal{O} is a positive invariant set if and only if $\mathcal{O} \subseteq \mathsf{Pre}_{\pi}(\mathcal{O})$

$$\mathsf{NOTE} \colon \mathcal{O} \subseteq \mathsf{Pre}_{\pi}(\mathcal{O}) \Leftrightarrow \mathsf{Pre}_{\pi}(\mathcal{O}) \cap \mathcal{O} = \mathcal{O}$$

Algorithm

Input: System model f, control policy π , constraint set \mathcal{X}

Output: \mathcal{O}_{∞}

- 1. Let $\Omega_0 = \mathcal{X}$
- **2.** Let $\Omega_{k+1} = \operatorname{Pre}_{\pi}(\Omega_k) \cap \Omega_k$
- 3. If $\Omega_{k+1} = \Omega_k$ then $\Omega_{\infty} \leftarrow \Omega_{k+1}$
- 4. If else go to 2

The algorithm generates the set sequence $\{\Omega_k\}$ satisfying $\Omega_{k+1} \subseteq \Omega_k, \forall k \in \mathbb{N}$ and it terminates when $\Omega_{k+1} = \Omega_k$ so that Ω_k is the maximal positive invariant set \mathcal{O}_{∞} for $x(t+1) = f_a(x(t))$.

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Control Invariant Sets

Control invariant sets

- are computed for systems subject to external inputs
- provide the set of initial states for which there exists a controller such that the system constraints are never violated.

Control Invariant Set

A set $\mathcal{C} \subseteq \mathcal{X}$ is said to be a control invariant set if

$$x(t) \in \mathcal{C} \quad \Rightarrow \quad \exists u(t) \in \mathcal{U} \text{ such that } f(x(t), u(t)) \in \mathcal{C}, \quad \forall t \in \mathbb{N}^+$$

Maxımal Control İnvarian

The set \mathcal{C}_{∞} is said to be the maximal control invariant set for the system x(t+1)=f(x(t),u(t)) subject to the constraints in $x(t)\in\mathcal{X},\ u(t)\in\mathcal{U}$, if it is control invariant and contains all control invariant sets contained in \mathcal{X} .

Control Invariant Sets

Control invariant sets

- are computed for systems subject to external inputs
- provide the set of initial states for which there exists a controller such that the system constraints are never violated.

Control Invariant Set

A set $\mathcal{C} \subseteq \mathcal{X}$ is said to be a control invariant set if

$$x(t) \in \mathcal{C} \quad \Rightarrow \quad \exists u(t) \in \mathcal{U} \text{ such that } f(x(t), u(t)) \in \mathcal{C}, \quad \forall t \in \mathbb{N}^+$$

Maximal Control Invariant Set

The set \mathcal{C}_{∞} is said to be the maximal control invariant set for the system x(t+1)=f(x(t),u(t)) subject to the constraints in $x(t)\in\mathcal{X},\ u(t)\in\mathcal{U}$, if it is control invariant and contains all control invariant sets contained in \mathcal{X} .

Control Invariant Sets

Same geometric condition for control invariants holds: $\mathcal C$ is a control invariant set if and only if

$$\mathcal{C} \subseteq \mathsf{Pre}(\mathcal{C})$$

Algorithm

Input: System model f, constraint sets $\mathcal X$ and $\mathcal U$

Output: \mathcal{O}_{∞}

- 1. Let $\Omega_0 = \mathcal{X}$
- **2.** Let $\Omega_{k+1} = \text{Pre}(\Omega_k) \cap \Omega_k$
- 3. If $\Omega_{k+1} = \Omega_k$ then $\mathcal{C}_{\infty} \leftarrow \Omega_{k+1}$
- 4. If else go to 2

The algorithm generates the set sequence $\{\Omega_k\}$ satisfying $\Omega_{k+1} \subseteq \Omega_k, \forall k \in \mathbb{N}$ and it terminates if $\Omega_{k+1} = \Omega_k$ so that Ω_k is the maximal control invariant set \mathcal{C}_{∞} for the constrained system.

Invariant Sets and Control Invariant Sets

- ► The set \mathcal{O}_{∞} (\mathcal{C}_{∞}) is *finitely determined* if and only if $\exists i \in \mathbb{N}$ such that $\Omega_{i+1} = \Omega_i$.
- ► The smallest element $i \in \mathbb{N}$ such that $\Omega_{i+1} = \Omega_i$ is called the *determinedness index*.
- For all states contained in the maximal control invariant set \mathcal{C}_{∞} there exists a control law, such that the system constraints are never violated.

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Loss of Feasibility

MPC policies compute control actions by solving finite time optimal control problems over shifted time windows:

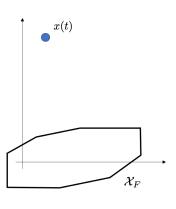
$$J_{t}^{*}(x(0)) = \min_{u_{t|t},...,u_{t+N-1|t}} \sum_{k=0}^{T-1} h(x_{k|t}, u_{k|t}) + V(x_{t+T|t})$$
such that
$$x_{k+1|t} = Ax_{k|t} + Bu_{k|t}, \forall k \in \{t, ..., t+N-1\}$$

$$x_{k|t} \in \mathcal{X}, u_{k|t} \in \mathcal{U}, \forall k \in \{t, ..., t+N-1\}$$

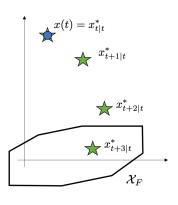
$$x_{t|t} = x(0), x_{N} \in \mathcal{X}_{F}$$

Solution: The terminal cost $V(x_{t+T|t})$ and terminal constraint \mathcal{X}_F , often referred to as <u>terminal components</u>, should approximate the tail of cost and constraints beyond the prediction horizon.

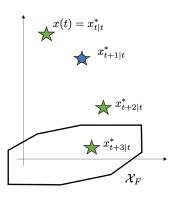
- At time step t assume that the MPC problem is feasible and let $\{u_{t|t}^*,\ldots,u_{t+N-1|t}^*\}$ and $\{x_{t|t}^*,\ldots,x_{t+N|t}^*\}$ be the optimal sequences of states and actions.
- At the next time step t+1, we have that $x(t+1) = x_{t+1|t}^*$.
- Therefore, at the next time step t+1 the sequences $\{u^*_{t+1|t}, \dots, u^*_{t+N-1|t}, 0\}$ and $\{x^*_{t+1|t}, \dots, x^*_{t+N|t}, 0\}$ are feasible, as $x^*_{t+N|t} = 0 \in \mathcal{X}_F$ and the origin is an unforced equilibrium point.



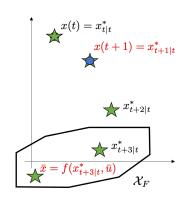
- At time step t assume that the MPC problem is feasible and let $\{u_{t|t}^*,\ldots,u_{t+N-1|t}^*\}$ and $\{x_{t|t}^*,\ldots,x_{t+N|t}^*\}$ be the optimal sequences of states and actions.
- At the next time step t+1, we have that $x(t+1) = x_{t+1|t}^*$.
- Therefore, at the next time step t+1 the sequences $\{u_{t+1|t}^*,\ldots,u_{t+N-1|t}^*,0\}$ and $\{x_{t+1|t}^*,\ldots,x_{t+N|t}^*,0\}$ are feasible, as $x_{t+N|t}^*=0\in\mathcal{X}_F$ and the origin is an unforced equilibrium point.



- At time step t assume that the MPC problem is feasible and let $\{u_{t|t}^*,\ldots,u_{t+N-1|t}^*\}$ and $\{x_{t|t}^*,\ldots,x_{t+N|t}^*\}$ be the optimal sequences of states and actions.
- At the next time step t+1, we have that $x(t+1) = x_{t+1|t}^*$.
- ▶ Therefore, at the next time step t+1 the sequences $\{u_{t+1|t}^*, \ldots, u_{t+N-1|t}^*, 0\}$ and $\{x_{t+1|t}^*, \ldots, x_{t+N|t}^*, 0\}$ are feasible, as $x_{t+N|t}^* = 0 \in \mathcal{X}_F$ and the origin is an unforced equilibrium point.



- At time step t assume that the MPC problem is feasible and let $\{u_{t|t}^*,\ldots,u_{t+N-1|t}^*\}$ and $\{x_{t|t}^*,\ldots,x_{t+N|t}^*\}$ be the optimal sequences of states and actions.
- At the next time step t+1, we have that $x(t+1) = x_{t+1|t}^*$.
- As $x_{t+N|t}^* \in \mathcal{X}_F$ there exists $\bar{u} \in \mathcal{U}$ such that $\bar{x} = f(x_{t+N|t}^*, \bar{u}) \in \mathcal{X}_F$. Thus, at the next time step t+1 the sequences $\{u_{t+1|t}^*, \dots, u_{t+N-1|t}^*, \bar{u}\}$ and $\{x_{t+1|t}^*, \dots, x_{t+N|t}^*, \bar{x}\}$ are feasible.



Stability - Assumptions

Let the following assumptions hold

▶ The stage cost satisfies

$$h(x, u) = 0 \forall x \in \mathcal{X} \setminus \{0\}, \forall u \in \mathcal{U} \setminus \{0\}$$

and h(0,0) = 0.

- ▶ The terminal set \mathcal{X}_F is a <u>control invariant set</u>
- ▶ The terminal cost function $V : \mathbb{R}^n \to \mathbb{R}$ is a control Lyapunov function for the set \mathcal{X}_F , i.e.,

$$\forall x \in \mathcal{X}_F, \exists u \in \mathcal{U} \text{ such that } V(f(x,u)) - V(x) \geq -h(x,u)$$
 and $f(x,u) \in \mathcal{X}_F$.

Next, we show by induction that the open-loop cost $J_t^*(x(t))$ is a Lyapunov function for the closed-loop system, i.e.,

$$J_{t+1}^*(x(t+1)) < J_t^*(x(t)), \forall x(t) \in \mathcal{X} \setminus \{0\}.$$

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Stability – Proof (1/2)

At time step t, assume that the MPC problem is feasible and let $\{u_{t|t}^*, \ldots, u_{t+N-1|t}^*\}$ and $\{x_{t|t}^*, \ldots, x_{t+N|t}^*\}$ be the optimal sequences of states and actions. Then the open-loop cost is

$$J_{t}^{*}(x(t)) = \sum_{k=t}^{N-1} h(x_{k|t}^{*}, u_{k|t}^{*}) + V(x_{t+N|t}^{*})$$

$$\geq \sum_{k=t}^{N-1} h(x_{k|t}^{*}, u_{k|t}^{*}) + h(x_{t+N|t}^{*}, \bar{u}) + V(f(x_{t+N|t}^{*}, \bar{u}))$$

for $\bar{u} \in \mathcal{U}$ such that $f(x_{t+N|t}^*, \bar{u})$.

Stability – Proof (2/2)

At the next time step t+1,

$$\bar{J} = \sum_{k=t+1}^{N-1} h(x_{k|t}^*, u_{k|t}^*) + h(x_{t+N|t}^*, \bar{u}) + V(f(x_{t+N|t}^*, \bar{u}))$$

is the cost associated with the feasible sequence of inputs $\{u^*_{t+1|t},\dots,u^*_{t+N-1|t},\bar{u}\}$, thus

$$J_t^*(x(t)) = h(x_{k|t}^*, u_{k|t}^*) + \bar{J} \ge h(x_{t|t}^*, u_{t|t}^*) + J_{t+1}^*(x(t+1)).$$

Concluding, the open-loop cost satisfies

$$J_{t+1}^*(x(t+1)) - J_t^*(x(t)) \le -h(x(t), u(t))$$

as $x_{t|t}^* = x(t)$ and $u_{t|t}^* = x(t)$, and it is a Lyapunov function for the closed-loop system.

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Constrained Linear Quadratic Regulator

Consider the following finite time optimal control problem:

$$J_{t}^{*}(x(0)) = \min_{u_{t|t}, \dots, u_{t+N-1|t}} \sum_{k=0}^{T-1} h(x_{k|t}, u_{k|t}) + x_{t+T|t}^{\top} P x_{t+T|t}$$
such that
$$x_{k+1|t} = A x_{k|t} + B u_{k|t}, \forall k \in \{t, \dots, t+N-1\}$$

$$x_{k|t} \in \mathcal{X}, u_{k|t} \in \mathcal{U}, \forall k \in \{t, \dots, t+N-1\}$$

$$x_{t|t} = x(0), x_{N} \in \mathcal{X}_{F}$$

where $h(x, u) = x^{\top}Qx + u^{\top}Ru$.

Next, we discuss how to construct the terminal cost $V(x) = x^{\top} P x$ and the terminal set \mathcal{X}_F to guarantee recursive feasibility and closed-loop stability.

Design Rules

1. Design unconstrained LQR control law

$$K_{\infty} = (B'P_{\infty}B + R)^{-1}B'P_{\infty}A$$

where P_{∞} is the solution to the discrete-time algebraic Riccati equation:

$$P_{\infty} = A'P_{\infty}A + Q - A'P_{\infty}B(B'P_{\infty}B + R)^{-1}B'P_{\infty}A$$

- 2. Choose the terminal weight $P = P_{\infty}$
- 3. Choose the terminal set \mathcal{X}_F to be the maximum invariant set for the closed-loop system $x_{k+1} = (A BK_{\infty})x_k$:

$$x_{k+1} = (A - BK_{\infty})x_k \in \mathcal{X}_F$$
, for all $x_k \in \mathcal{X}_F$

All state and input constraints are satisfied in \mathcal{X}_F :

$$\mathcal{X}_F \subseteq \mathcal{X}, F_{\infty} x_k \in \mathcal{U}, \text{ for all } x_k \in \mathcal{X}_F$$

Stability and Feasibility Proof

By construction all the Assumptions of the required to guarantee recursive feasibility and stability are verified:

- 1. The stage cost is a positive definite function
- 2. By construction the terminal set is **invariant** under the local control law $v=-K_{\infty}x$
- 3. Terminal cost is a continuous Lyapunov function in the terminal set \mathcal{X}_F and satisfies:

$$\begin{aligned} x_{k+1}^{\top} P x_{k+1} - x_k^{\top} P x_k \\ &= x_k' (-P_{\infty} + A' P_{\infty} A - A' P_{\infty} B (B' P_{\infty} B + R)^{-1} B' P_{\infty} A) x_k \\ &= -x_k' Q x_k \end{aligned}$$

Summary of Safety and Stability Properties

Key Message: When the MPC terminal components are not designed correctly, the closed-loop system may violate safety constraints and convergence to the goal state/set is not guaranteed

Solution: We have shown that given a terminal set set \mathcal{X}_F which is control invariant, and a terminal cost function V(x) which is a control Lyapunov function.

- The MPC problem is feasible at all times
- ▶ The closed-loop system is stable as for the positive definite open-loop cost we have $J_{t+1}^*(x(t+1)) < J_t^*(x(t)), \forall x(t) \notin \mathcal{X}_F$

Main drawback: These terminal components are hard to compute even for linear constrained deterministic systems.