

کنترل پیش بین

Model Predictive Control

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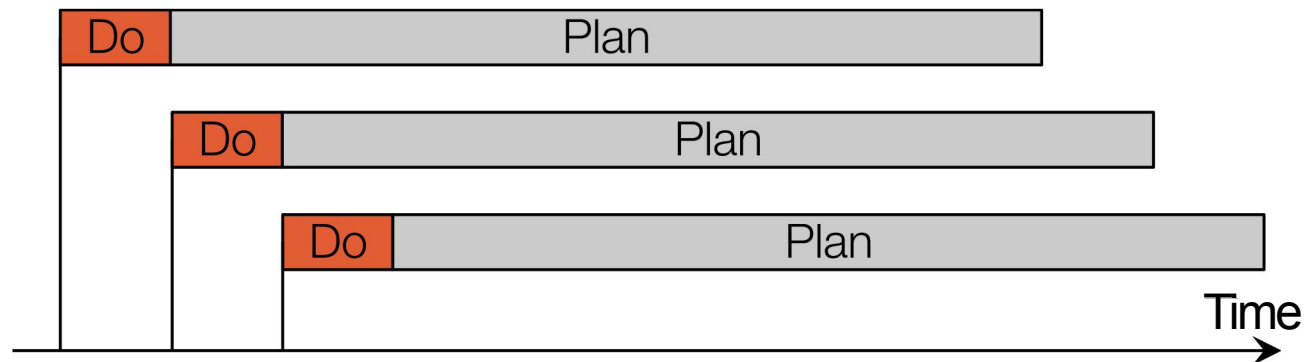
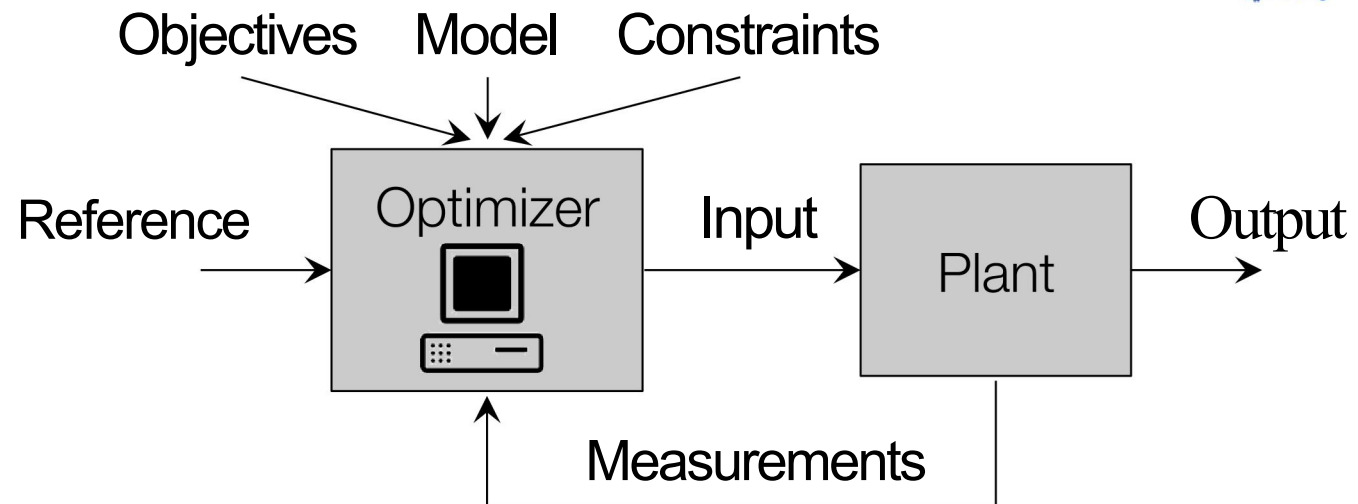
دانشگاه صنعتی خواجه نصیرالدین طوسی

Optimal Control



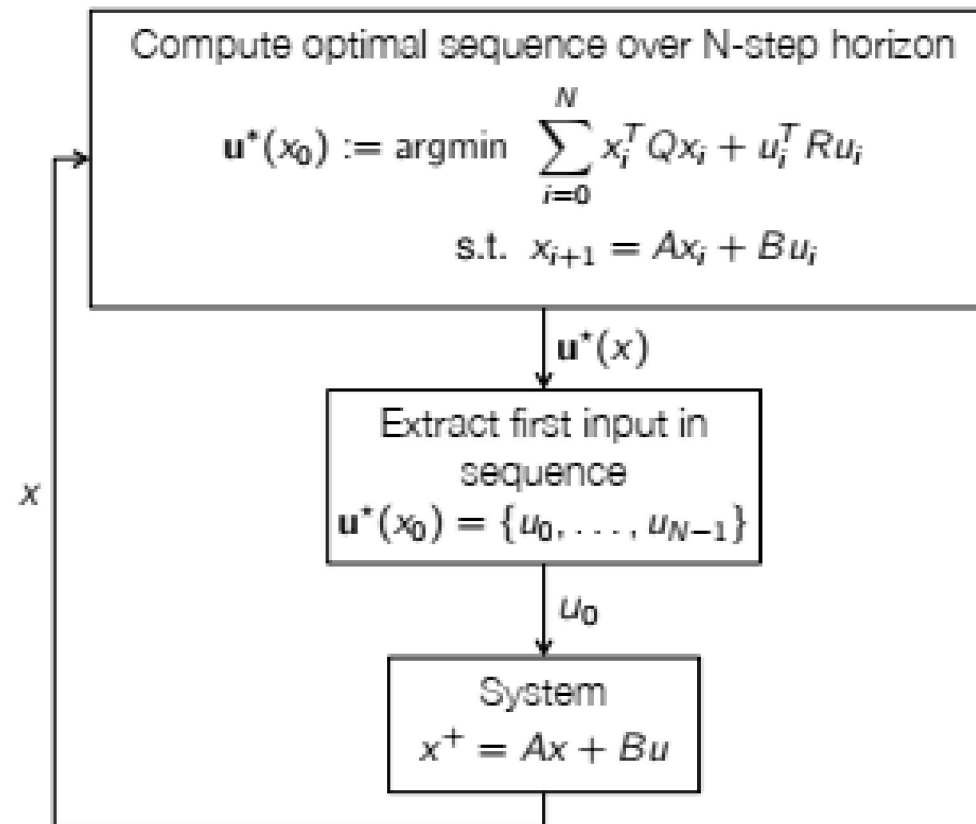
- ❑ Optimal Control
- ❑ Linear Quadratic Optimal Control
 - ❑ Batch Approach
 - ❑ Recursive Approach
 - ❑ Receding Horizon
 - ❑ Infinite Horizon Optimal Control
- ❑ Constrained Linear Optimal Control
- ❑ Constrained Optimal Control: 2-Norm
- ❑ Constrained Optimal Control: 1-Norm and ∞ -Norm

Receding horizon control



Receding horizon strategy introduces feedback.

Receding horizon control



For unconstrained systems, this is a **constant linear controller**
 However, can extend this concept to much more complex systems (MPC)

Example - Impact of Horizon Length

Consider the lightly damped, stable system

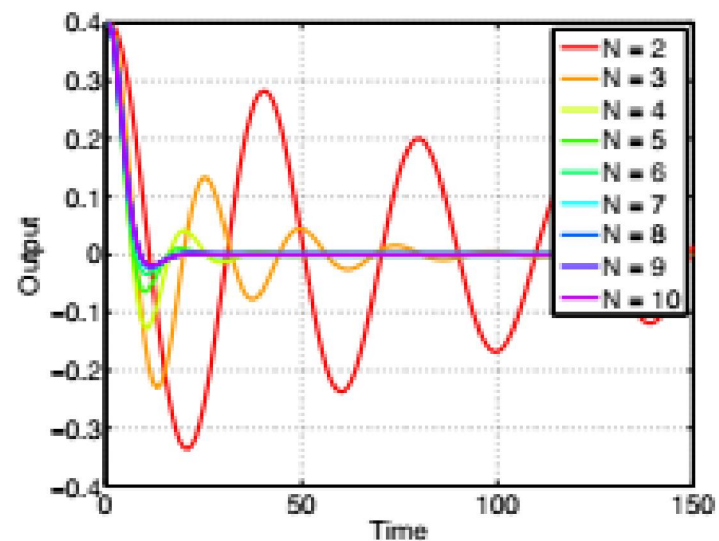
$$G(s) := \frac{\omega^2}{s^2 + 2\zeta\omega s + \omega^2}$$

where $\omega = 1$, $\zeta = 0.01$. We sample at 10Hz and set $P = Q = I$, $R = 1$.

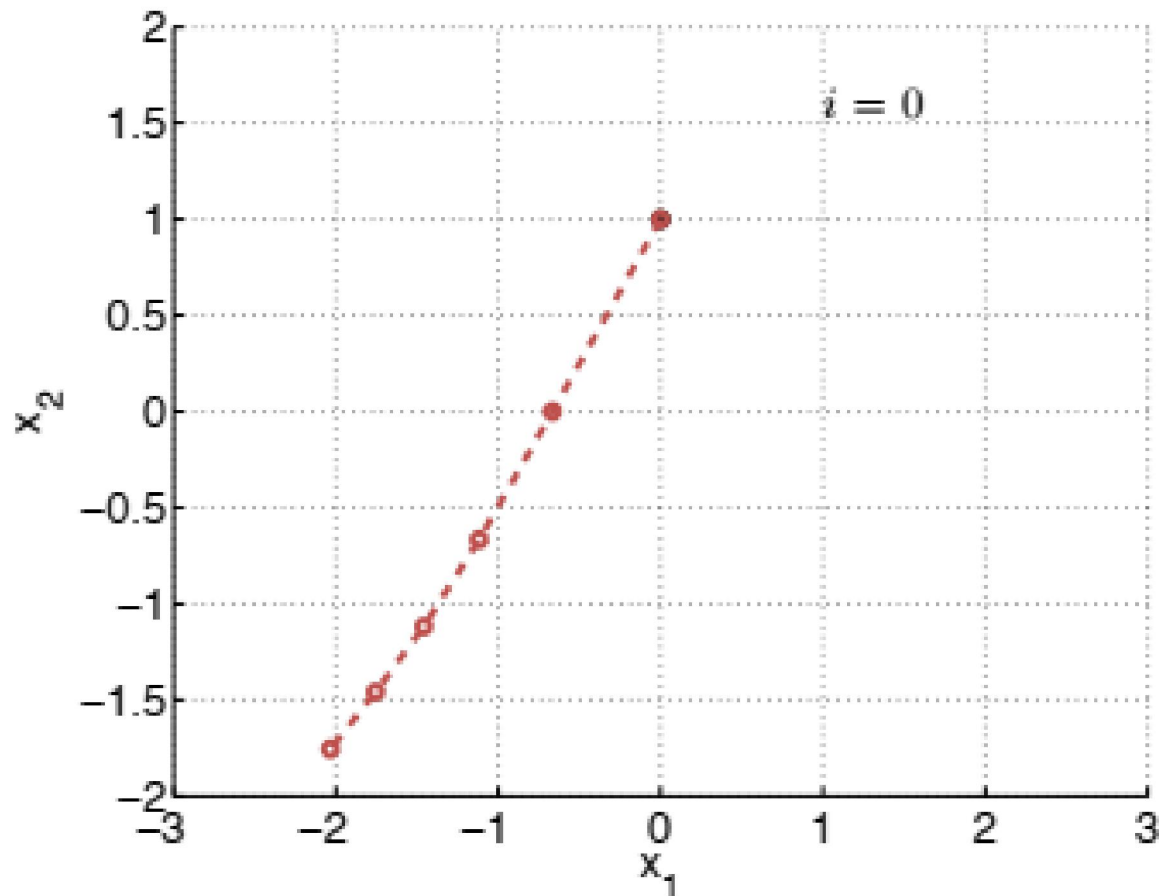
Discrete-time state-space model:

$$x^+ = \begin{bmatrix} 1.988 & -0.998 \\ 1 & 0 \end{bmatrix} x + \begin{bmatrix} 0.125 \\ 0 \end{bmatrix} u$$

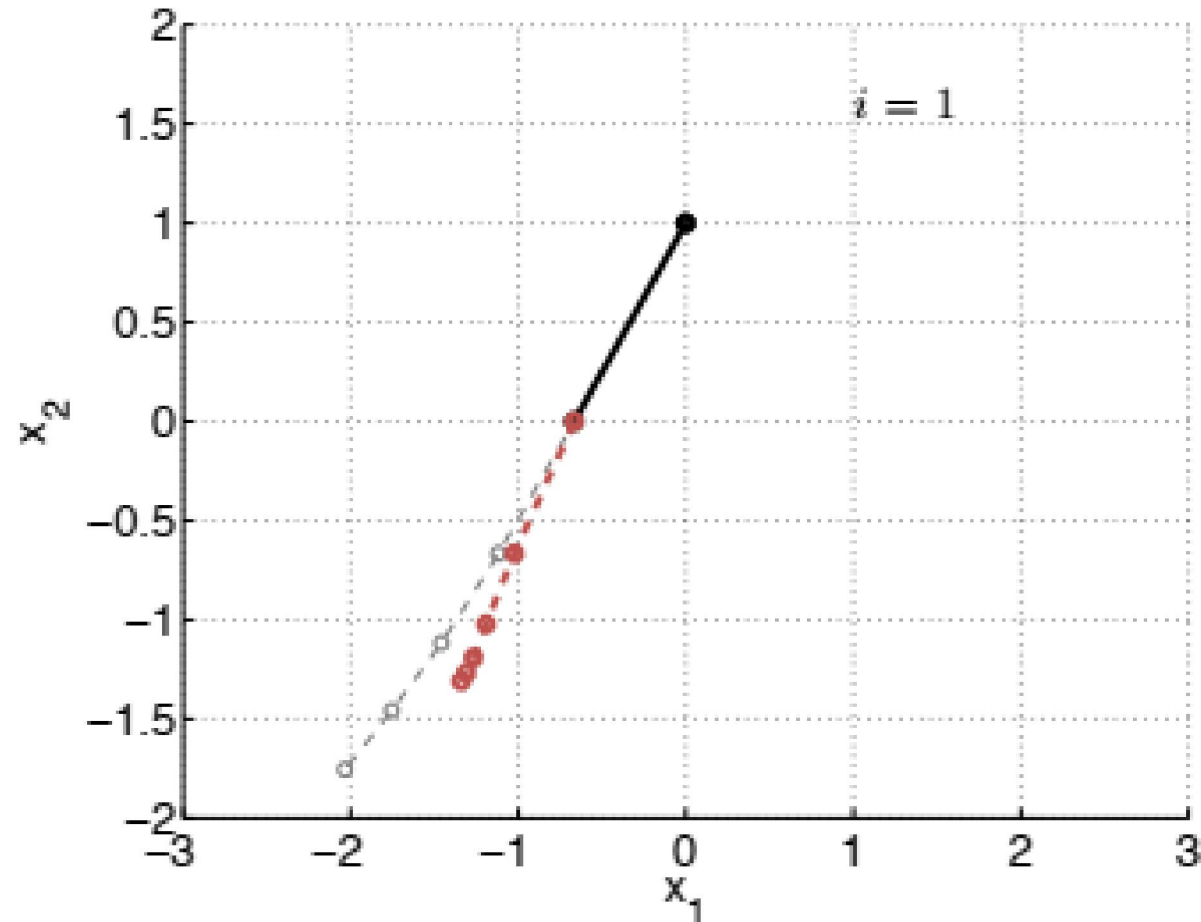
Closed-loop response



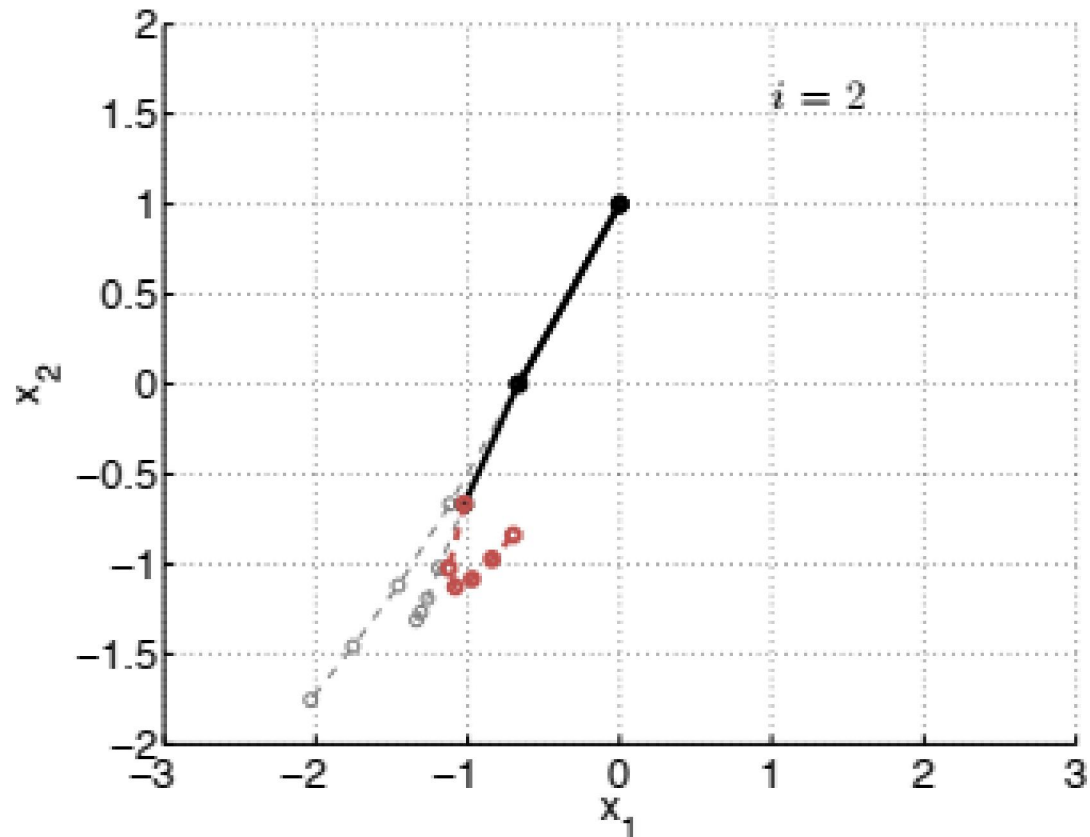
Example: Short horizon $N=5$



Example: Short horizon $N=5$

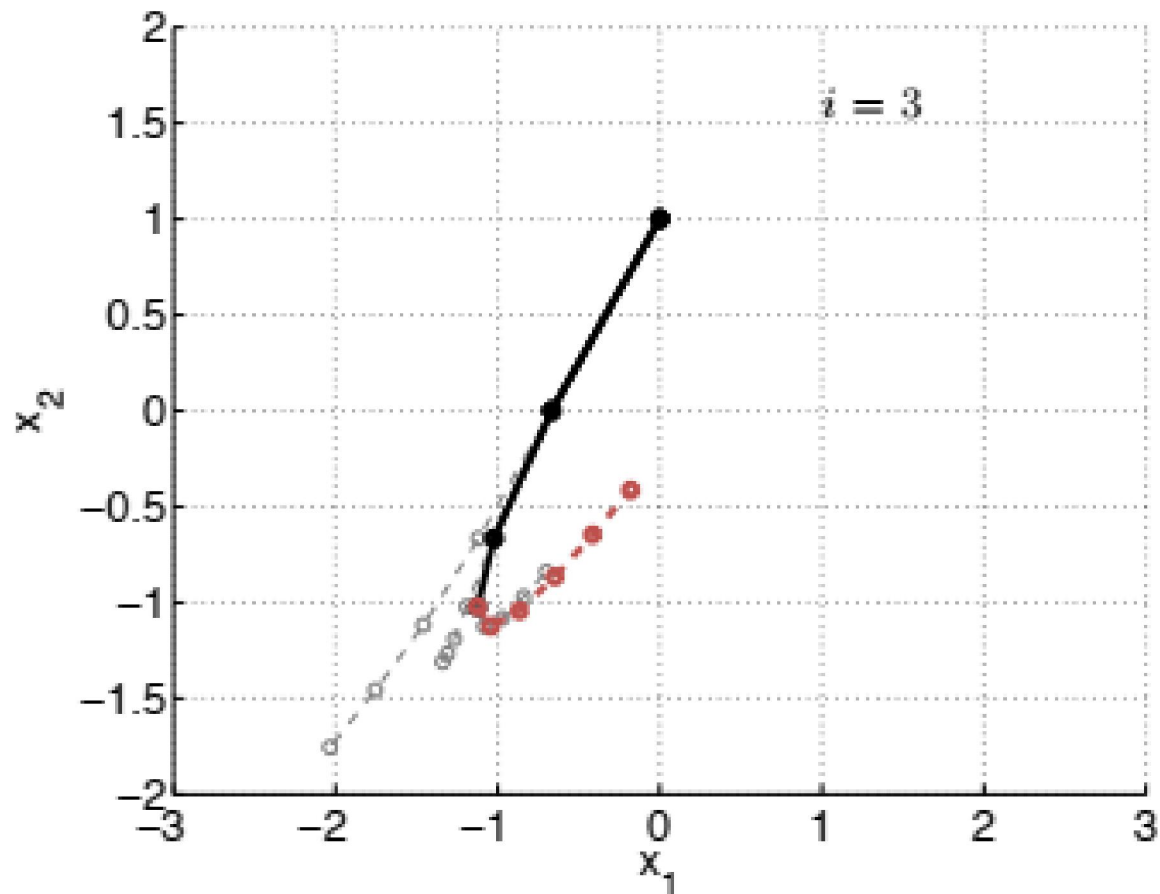


Example: Short horizon $N=5$

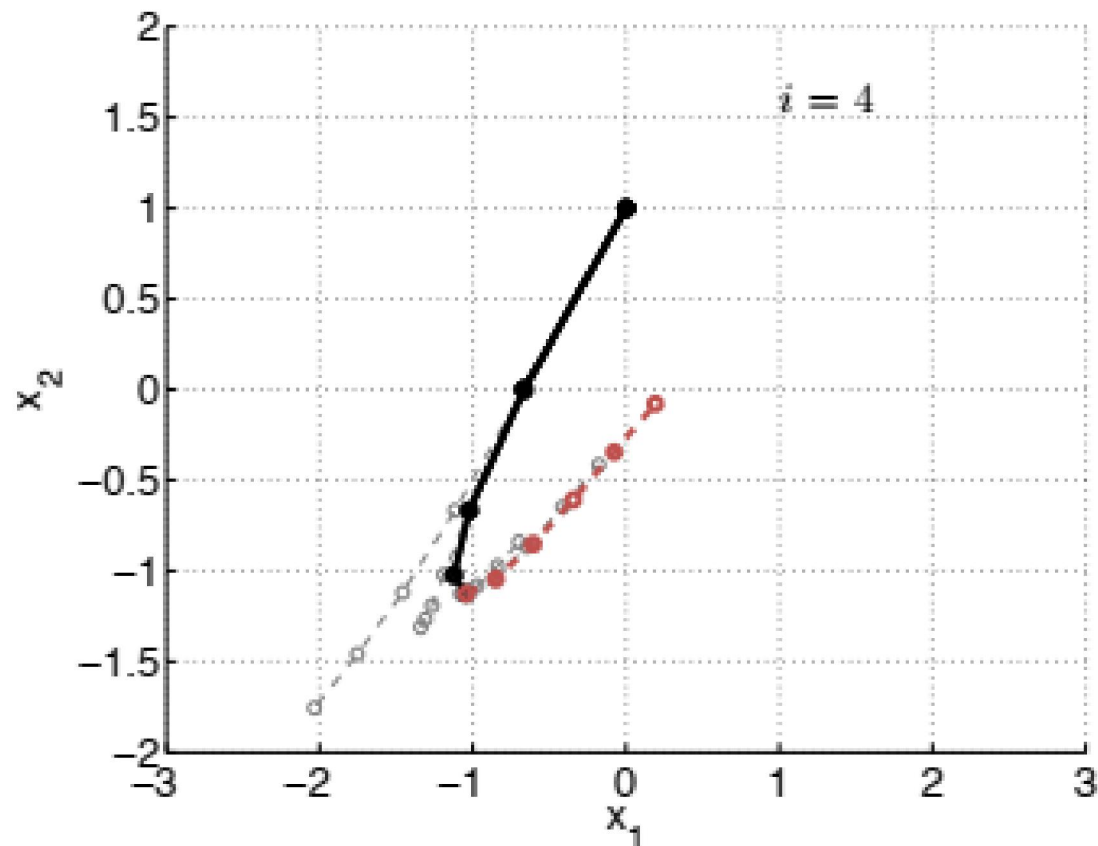


Short horizon: Prediction and closed-loop response differ.

Example: Short horizon $N=5$

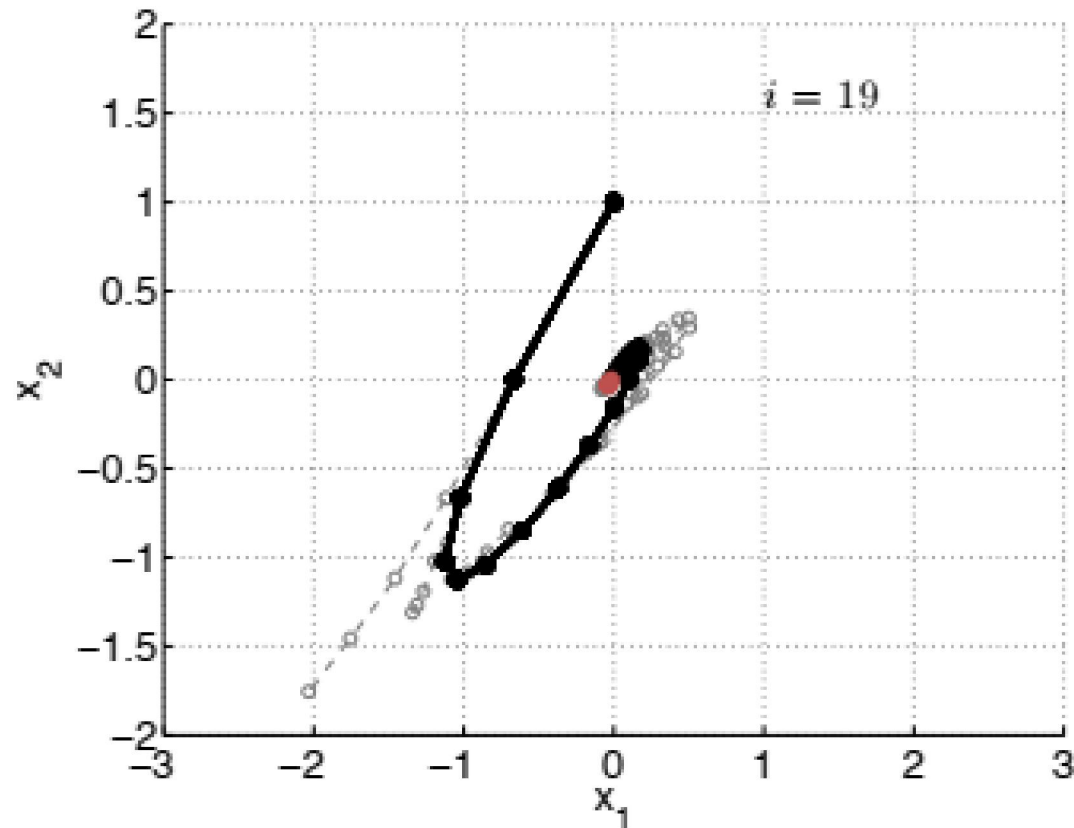


Example: Short horizon $N=5$



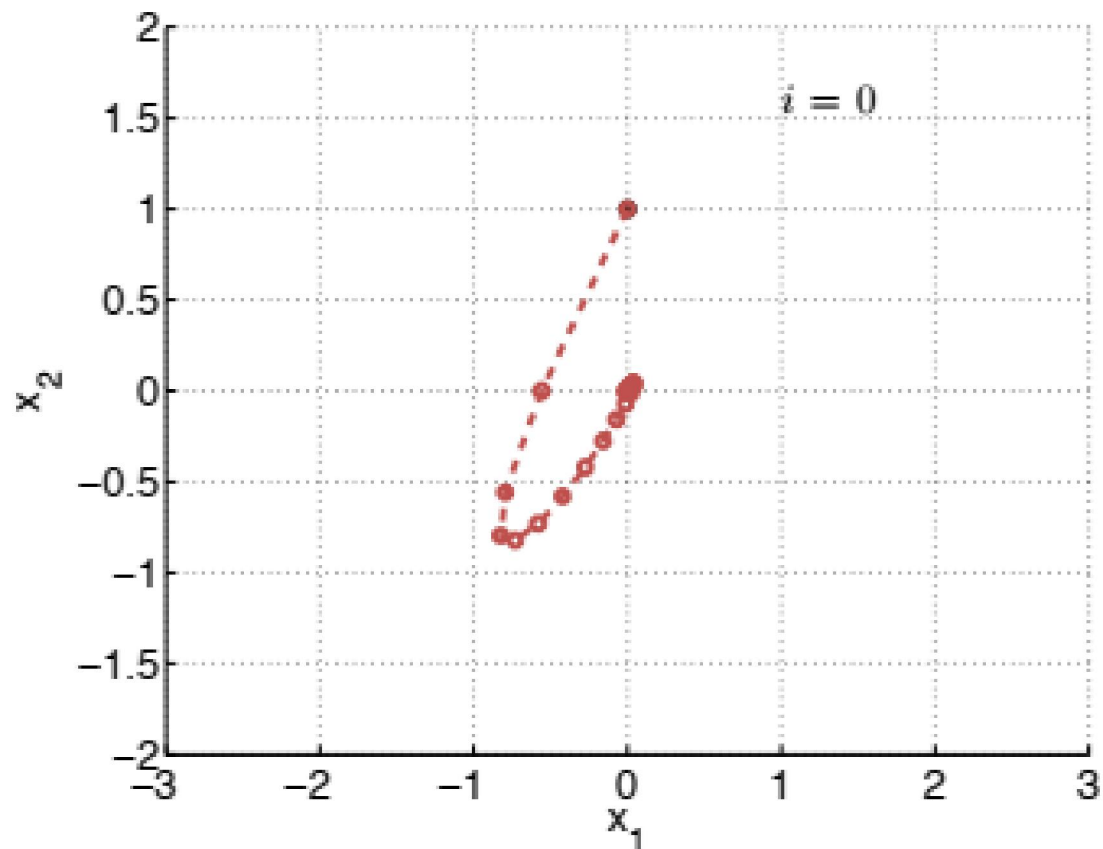
Short horizon: Prediction and closed-loop response differ.

Example: Short horizon $N=5$

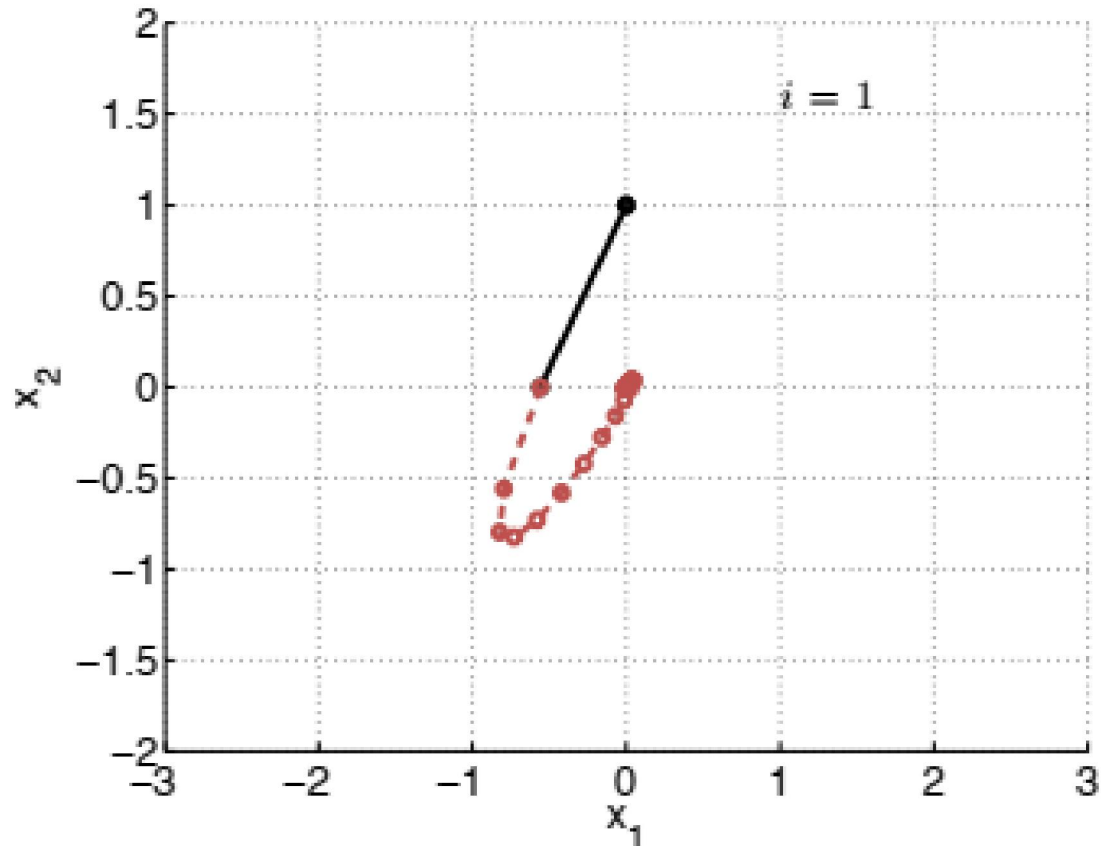


Short horizon: Prediction and closed-loop response differ.

Example: Long horizon $N=20$

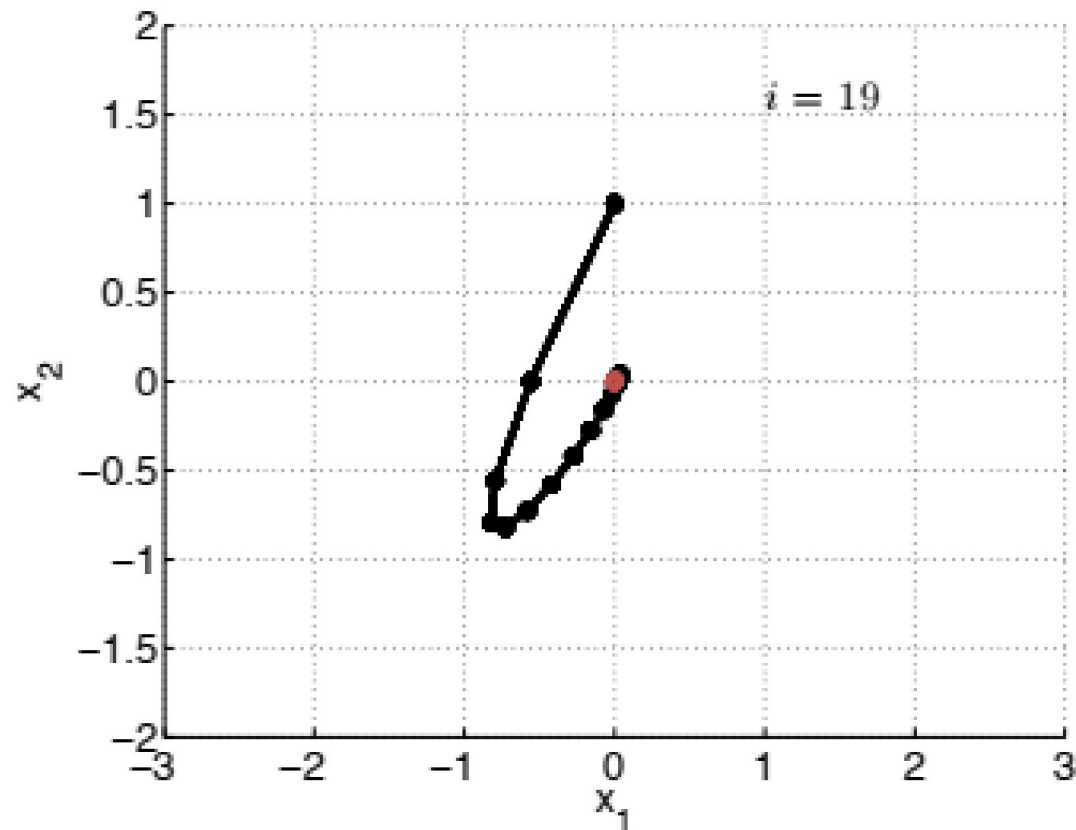


Example: Long horizon $N=20$



Long horizon: Prediction and closed-loop match.

Example: Long horizon $N=20$



Long horizon: Prediction and closed-loop match.

Linear Quadratic Optimal Control



Infinite Horizon Control Problem: Optimal Solution

- In some cases we may want to solve the same problem with an infinite horizon:

$$J_{\infty}(x(0)) = \min_{u(\cdot)} \left\{ \sum_{k=0}^{\infty} [x_k' Q x_k + u_k' R u_k] \right\}$$

$$\text{subject to } \begin{aligned} x_{k+1} &= A x_k + B u_k, \quad k = 0, 1, 2, \dots, \infty, \\ x_0 &= x(0) \end{aligned}$$

- As with the Dynamic Programming approach, the optimal input is of the form.

$$u^*(k) = -(B' P_{\infty} B + R)^{-1} B' P_{\infty} A x(k) \triangleq F_{\infty} x(k)$$

and the infinite-horizon cost-to-go is.

$$J_{\infty}(x(k)) = x(k)' P_{\infty} x(k).$$

Linear Quadratic Optimal Control



Infinite Horizon Control Problem: Optimal Solution

- The matrix P_∞ comes from an infinite recursion of the RDE, from a point infinitely far into the future.
- Assuming the RDE does converge to some constant matrix P_∞ , it must satisfy the following (from (7), with $P_k = P_{k+1} = P_\infty$)

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

which is called the Algebraic Riccati Equation (ARE).

- The constant feedback matrix F_∞ is referred to as the asymptotic form of the Linear Quadratic Regulator (LQR).
- In fact, if (A, B) is controllable and (Q, A) is observable, then the RDE (initialized with Q at $k = \infty$ and solved for $k \searrow 0$) converges to the unique positive definite solution P_∞ of the ARE



Linear Quadratic Optimal Control

Stability of Infinite-Horizon LQR

- In addition, the closed-loop system with $u(k) = F_{\infty}x(k)$ is guaranteed to be asymptotically stable, under the stabilizability and detectability assumptions of the previous slide.
- The latter statement can be proven by substituting the control law $u(k) = F_{\infty}x(k)$ into $x(k+1) = Ax(k) + Bu(k)$, and then examining the properties of the system

$$x(k+1) = (A + BF_{\infty})x(k). \quad (8)$$

The asymptotic stability of (8) can be proven by showing that the infinite horizon cost $J_{\infty}^*(x(k)) = x(k)'P_{\infty}x(k)$ is actually a Lyapunov function for the system, i.e. $J_{\infty}^*(x(k)) > 0, \forall k \neq 0, J_{\infty}^*(0) = 0$, and $J_{\infty}^*(x(k+1)) < J_{\infty}^*(x(k))$, for any $x(k)$. This implies that

Optimal Control



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Constrained Linear Optimal Control

Constrained Linear Optimal Control

Cost function

$$J_0(x(0), U_0) = p(x_N) + \sum_{k=0}^{N-1} q(x_k, u_k)$$

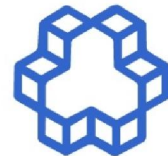
- $U_0 \triangleq [u'_0, \dots, u'_{N-1}]'$
- Squared Euclidian norm: $p(x_N) = x'_N P x_N$ and $q(x_k, u_k) = x'_k Q x_k + u'_k R u_k$.
- $p = 1$ or $p = \infty$: $p(x_N) = \|P x_N\|_p$ and $q(x_k, u_k) = \|Q x_k\|_p + \|R u_k\|_p$.

Constrained Finite Time Optimal Control problem (CFTOC)

$$\begin{aligned} J_0^*(x(0)) = & \min_{U_0} J_0(x(0), U_0) \\ \text{subj. to} & \quad x_{k+1} = A x_k + B u_k, \quad k = 0, \dots, N-1 \\ & \quad x_k \in \mathcal{X}, \quad u_k \in \mathcal{U}, \quad k = 0, \dots, N-1 \\ & \quad x_N \in \mathcal{X}_f \\ & \quad x_0 = x(0) \end{aligned} \quad (9)$$

N is the time horizon and \mathcal{X} , \mathcal{U} , \mathcal{X}_f are polyhedral regions.

Constrained Linear Optimal Control



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Feasible Sets

Set of initial states $x(0)$ for which the optimal control problem (9) is feasible:

$$\mathcal{X}_0 = \{x_0 \in \mathbb{R}^n \mid \exists(u_0, \dots, u_{N-1}) \text{ such that } x_k \in \mathcal{X}, u_k \in \mathcal{U}, \\ k = 0, \dots, N-1, x_N \in \mathcal{X}_f, \text{ where } x_{k+1} = Ax_k + Bu_k\}$$

In general \mathcal{X}_i is the set of states x_i at time i for which (9) is feasible:

$$\mathcal{X}_i = \{x_i \in \mathbb{R}^n \mid \exists(u_i, \dots, u_{N-1}) \text{ such that } x_k \in \mathcal{X}, u_k \in \mathcal{U}, \\ k = i, \dots, N-1, x_N \in \mathcal{X}_f, \text{ where } x_{k+1} = Ax_k + Bu_k\},$$

The sets \mathcal{X}_i for $i = 0, \dots, N$ play an important role in the the solution of the CFTOC problem. They are independent of the cost.

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Constrained Optimal Control: 2-Norm

Problem Formulation

Quadratic cost function

$$J_0(x(0), U_0) = x'_N P x_N + \sum_{k=0}^{N-1} x'_k Q x_k + u'_k R u_k \quad (10)$$

with $P \succeq 0$, $Q \succeq 0$, $R \succ 0$.

Constrained Finite Time Optimal Control problem (CFTOC).

$$\begin{aligned} J_0^*(x(0)) = \min_{U_0} \quad & J_0(x(0), U_0) \\ \text{subj. to} \quad & x_{k+1} = Ax_k + Bu_k, \quad k = 0, \dots, N-1 \\ & x_k \in \mathcal{X}, \quad u_k \in \mathcal{U}, \quad k = 0, \dots, N-1 \\ & x_N \in \mathcal{X}_f \\ & x_0 = x(0) \end{aligned} \quad (11)$$

N is the time horizon and \mathcal{X} , \mathcal{U} , \mathcal{X}_f are polyhedral regions.



Constrained Optimal Control: 2-Norm

Construction of the QP with substitution

- **Step 1:** Rewrite the cost as (see previous slides)

$$\begin{aligned} J_0(x(0), U_0) &= U_0' H U_0 + 2x(0)' F U_0 + x(0)' Y x(0) \\ &= [U_0' \ x(0)'] \begin{bmatrix} H & F' \\ F & Y \end{bmatrix} [U_0' \ x(0)']' \end{aligned}$$

- **Step 2:** Rewrite the constraints compactly as (details provided on the next slide)

$$G_0 U_0 \leq w_0 + E_0 x(0)$$

- **Step 3:** Rewrite the optimal control problem as

$$\begin{aligned} J_0^*(x(0)) &= \min_{U_0} [U_0' \ x(0)'] \begin{bmatrix} H & F' \\ F & Y \end{bmatrix} [U_0' \ x(0)']' \\ &\text{subj. to } G_0 U_0 \leq w_0 + E_0 x(0) \end{aligned}$$



Constrained Optimal Control: 2-Norm

Construction of the QP with substitution Solution

$$J_0^*(x(0)) = \min_{U_0} [U_0' x(0)'] \begin{bmatrix} H & F' \\ F & Y \end{bmatrix} [U_0' x(0)']'$$

subj. to $G_0 U_0 \leq w_0 + E_0 x(0)$

For a given $x(0)$ U_0^* can be found via a QP solver.



Constrained Optimal Control: 2-Norm

Construction of the QP with substitution

If \mathcal{X} , \mathcal{U} and \mathcal{X}_f are given by:

$$\mathcal{X} = \{x \mid A_x x \leq b_x\} \quad \mathcal{U} = \{u \mid A_u u \leq b_u\} \quad \mathcal{X}_f = \{x \mid A_f x \leq b_f\}$$

Then G_0 , E_0 and w_0 are defined as follows

$$G_0 = \begin{bmatrix} A_u & 0 & \dots & 0 \\ 0 & A_u & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & A_u \\ 0 & 0 & \dots & 0 \\ A_x B & 0 & \dots & 0 \\ A_x A B & A_x B & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ A_f A^{N-1} B & A_f A^{N-2} B & \dots & A_f B \end{bmatrix}, E_0 = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ -A_x \\ -A_x A \\ -A_x A^2 \\ \vdots \\ -A_f A^N \end{bmatrix}, w_0 = \begin{bmatrix} b_u \\ b_u \\ \vdots \\ b_u \\ b_x \\ b_x \\ b_x \\ \vdots \\ b_f \end{bmatrix}$$



Constrained Optimal Control: 2-Norm

Construction of the QP without substitution

To obtain the QP problem

$$J_0^*(x(0)) = \min_{U_0} [U_0' \ x(0)'] \begin{bmatrix} H & F' \\ F & Q \end{bmatrix} [U_0' \ x(0)']'$$

subj. to $G_0 U_0 \leq w_0 + E_0 x(0)$

we have substituted the state equations

$$x_{k+1} = Ax_k + Bu_k$$

into the state constraints $x_k \in \mathcal{X}$.

It is often more efficient to keep the explicit equality constraints.



Constrained Optimal Control: 2-Norm

Construction of the QP without substitution

We transform the CFTOC problem into the QP problem

$$\begin{aligned}
 J_0^*(x(0)) &= \min_z \quad [z' \ x(0)'] \begin{bmatrix} \bar{H} & 0 \\ 0 & Q \end{bmatrix} [z' \ x(0)']' \\
 &\text{subj. to } G_{0,\text{in}} z \leq w_{0,\text{in}} + E_{0,\text{in}} x(0) \\
 &G_{0,\text{eq}} z = E_{0,\text{eq}} x(0)
 \end{aligned}$$

- Define variable:

$$z = [x'_1 \ \dots \ x'_N \ u'_0 \ \dots \ u'_{N-1}]'$$

- Equalities from system dynamics $x_{k+1} = Ax_k + Bu_k$:

$$G_{0,\text{eq}} = \begin{bmatrix} I & & & & & & -B \\ -A & I & & & & & -B \\ & -A & I & & & & -B \\ & & \ddots & \ddots & & & \vdots \\ & & & -A & I & & -B \end{bmatrix}, E_{0,\text{eq}} = \begin{bmatrix} A \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$



Constrained Optimal Control: 2-Norm

2-Norm State Feedback Solution

Start from QP with substitution.

- **Step 1:** Define $z \triangleq U_0 + H^{-1}F'x(0)$ and transform the problem into

$$\hat{J}^*(x(0)) = \min_z z'Hz$$

subj. to $G_0z \leq w_0 + S_0x(0),$

where $S_0 \triangleq E_0 + G_0H^{-1}F'$, and

$$\hat{J}^*(x(0)) = J_0^*(x(0)) - x(0)'(Y - FH^{-1}F')x(0).$$

The CFTOC problem is now a **multiparametric quadratic program (mp-QP)**.

- **Step 2:** Solve the mp-QP to get explicit solution $z^*(x(0))$
- **Step 3:** Obtain $U_0^*(x(0))$ from $z^*(x(0))$

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Constrained Optimal Control: 2-Norm

Problem Formulation

Piece-wise linear cost function

$$J_0(x(0), U_0) := \|Px_N\|_p + \sum_{k=0}^{N-1} \|Qx_k\|_p + \|Ru_k\|_p \quad (12)$$

with $p = 1$ or $p = \infty$, P , Q , R full column rank matrices

Constrained Finite Time Optimal Control Problem (CFTOC)

$$\begin{aligned} J_0^*(x(0)) = & \min_{U_0} J_0(x(0), U_0) \\ \text{subj. to} & \quad x_{k+1} = Ax_k + Bu_k, \quad k = 0, \dots, N-1 \\ & \quad x_k \in \mathcal{X}, \quad u_k \in \mathcal{U}, \quad k = 0, \dots, N-1 \\ & \quad x_N \in \mathcal{X}_f \\ & \quad x_0 = x(0) \end{aligned} \quad (13)$$

N is the time horizon and \mathcal{X} , \mathcal{U} , \mathcal{X}_f are polyhedral regions.



Constrained Optimal Control: 2-Norm

Construction of the LP with substitution

The problem results in the following standard LP

$$\begin{aligned} \min_{z_0} \quad & c'_0 z_0 \\ \text{subj. to} \quad & \bar{G}_0 z_0 \leq \bar{w}_0 + \bar{S}_0 x(0) \end{aligned}$$

where $z_0 := \{\varepsilon_0^x, \dots, \varepsilon_N^x, \varepsilon_0^u, \dots, \varepsilon_{N-1}^u, u'_0, \dots, u'_{N-1}\} \in \mathbb{R}^s$,
 $s \triangleq (m+1)N + N + 1$ and

$$\bar{G}_0 = \begin{bmatrix} G_\varepsilon & 0 \\ 0 & G_0 \end{bmatrix}, \quad \bar{S}_0 = \begin{bmatrix} S_\varepsilon \\ S_0 \end{bmatrix}, \quad \bar{w}_0 = \begin{bmatrix} w_\varepsilon \\ w_0 \end{bmatrix}$$

For a given $x(0)$ U_0^* can be obtained via an LP solver (the 1-norm case is similar).

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