

# Performance bounds and robust control for constrained systems

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# Introduction

The optimal control problem:

$$V_\infty^*(x) := \min_{\pi} \mathbb{E} \left[ \sum_{i=0}^{\infty} \alpha^i (x_i^\top Q x_i + u_i^\top R u_i) \right]$$

subject to:

$$\left. \begin{aligned} x_{i+1} &= Ax_i + Bu_i + Gw_i \\ (x_i, u_i) &\in Z, \quad x_0 = x \end{aligned} \right\} \mathbb{P}\text{-a.s.}$$

Some assumptions:

- Disturbances:  $w_i \in W$ , i.i.d. and  $W$  compact.
- Constraints:  $Z$  polyhedral
- Control policy:  $\pi := \{\mu_0(\cdot), \mu_1(\cdot), \dots\}$ , arbitrary functions
- Measurements: states are directly measurable

# Introduction

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Some goals for this system:

- Find a computable suboptimal policy.
- Find a performance bound on the best possible controller.

Finite horizon problems:

- Upper/sub-optimal approximation with linear policies
- Lower/super-optimal approximation with linear multipliers

Infinite horizon problems:

- Control using upper bound terminal cost
- Performance bounds using lower bound terminal cost

Goal: a computable bound in the form

$$V^\ell(x) \leq V_\infty^*(x) \leq V^u(x)$$

# Finite-horizon problems

A finite horizon version:

$$V_N^*(x) := \min_{\pi} \mathbb{E} \left[ \sum_{i=0}^{N-1} \alpha^i (x_i^\top Q x_i + u_i^\top R u_i) + x_N^\top P x_N \right]$$

subject to:

$$\left. \begin{array}{l} x_{i+1} = Ax_i + Bu_i + Gw_i \\ (x_i, u_i) \in Z, \quad x_0 = x \\ x_N \in \mathcal{X}_f \end{array} \right\} \mathbb{P}\text{-a.s.}$$

Assumptions:

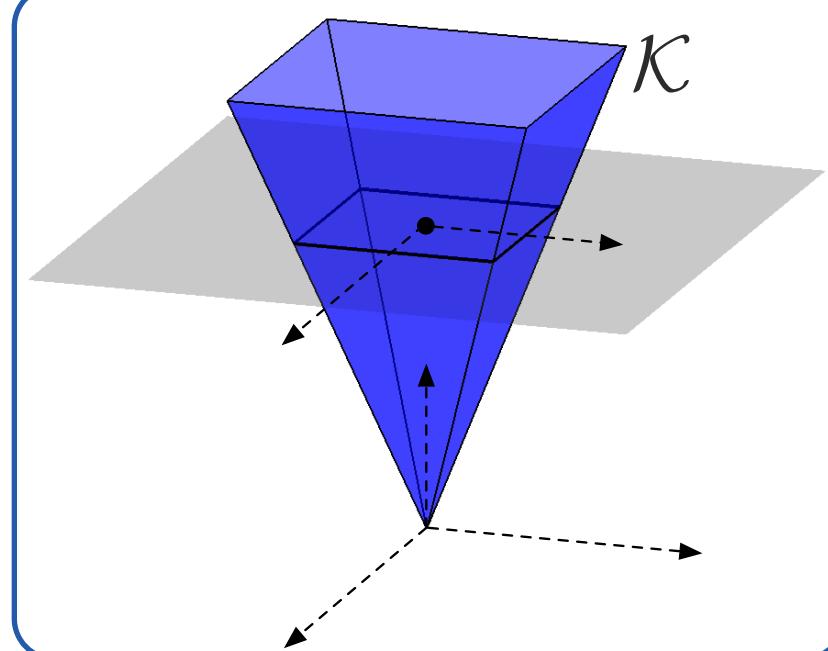
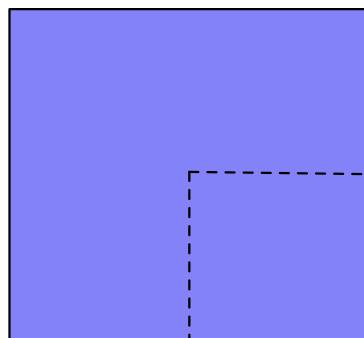
- All stage costs positive (semi)definite
- All constraints polyhedral

# Finite-horizon problems

Uncertainty set:

$$\mathbf{w} := (1; w_0; \dots; w_{N-1}), \quad \mathbb{E}[\mathbf{w}\mathbf{w}^\top] =: M$$

$$\mathbf{w} \in \mathcal{W} := \{\mathbf{w} \mid \mathbf{w} \succeq_{\mathcal{K}} 0, \ e_0^\top \mathbf{w} = 1\}$$



# Finite-horizon problems

A finite horizon version:

$$\begin{aligned} \inf \quad & \mathbb{E} [\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}] \\ \text{s.t.} \quad & \left. \begin{array}{l} \mathbf{u} \in \mathcal{N}, \mathbf{s} \in \mathcal{L}^2 \\ \mathbf{x} = \mathcal{B}\mathbf{u} + \mathcal{C}\mathbf{w} \\ F_u \mathbf{u} + F_x \mathbf{x} + \mathbf{s} = h \\ \mathbf{s} \geq 0 \end{array} \right\} \mathbb{P}\text{-a.s.} \end{aligned}$$

Decision variables:

- $\mathbf{u} \in \mathcal{N}$  : *causal* mapping from measurements to inputs
- $\mathbf{s} \in \mathcal{L}^2$  : Constraint slacks are a measurable random vector

Constraints must hold robustly for all  $\mathbf{w} \in \mathcal{W}$

# Finite-horizon problem (upper bound)

A suboptimal version:

$$\begin{aligned} \inf \quad & \mathbb{E} [\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}] \\ \text{s.t.} \quad & \left. \begin{array}{l} Q \in \mathcal{U}, \mathbf{s} \in \mathcal{L}^2 \\ \mathbf{x} = \mathcal{B}\mathbf{u} + \mathcal{C}\mathbf{w}, \mathbf{u} = Q\mathcal{G}\mathbf{w} \\ F_u \mathbf{u} + F_x \mathbf{x} + \mathbf{s} = h \\ \mathbf{s} \geq 0 \end{array} \right\} \mathbb{P}\text{-a.s.} \end{aligned}$$

Restrict the control policy to linear decision rules:

$$u = \underbrace{\begin{pmatrix} q_0 & | & Q_{0,0} & & & \\ \vdots & | & \vdots & \ddots & & \\ q_{N-1} & | & Q_{N-1,0} & \dots & Q_{N-1,N-1} & \end{pmatrix}}_{\in \mathcal{U}} \begin{pmatrix} 1 \\ \eta_0 \\ \vdots \\ \eta_{N-1} \end{pmatrix} = Q\mathcal{G}\mathbf{w}$$

# Finite-horizon problem (upper bound)

An equivalent form:

$$\begin{aligned} V_N^u(x) := \inf \mathbb{E} [\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}] \\ \text{s.t. } Q \in \mathcal{U}, \quad S \succeq_{\mathcal{K}^*} 0 \text{ (column-wise)} \\ (F_x \mathcal{B} + F_u) Q \mathcal{G} + F_x \mathcal{C} + S^\top = h e_0^\top \end{aligned}$$

Result is a conic program:

- QP if all sets polyhedral. Size scales with  $N^2$
- Provides upper bound on value and implementable policy.

$$V_N^*(x) \leq V_N^u(x)$$

**Upper bound makes primal variables linear in uncertainty**

# Finite-horizon problem again

The original finite horizon version:

$$V_N^*(x) := \inf \mathbb{E} [\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}]$$
$$\text{s.t. } \begin{aligned} & \mathbf{u} \in \mathcal{N}, \mathbf{s} \in \mathcal{L}^2 \\ & \mathbf{x} = \mathcal{B}\mathbf{u} + \mathcal{C}\mathbf{w} \\ & F_u \mathbf{u} + F_x \mathbf{x} + \mathbf{s} = h \\ & \mathbf{s} \geq 0 \end{aligned} \quad \left. \right\} \mathbb{P}\text{-a.s.}$$

Want to bound the degree of suboptimality caused by LDRs.

- Compute a *lower* bound on the value function:

$$V^\ell(x) \leq V_\infty^*(x) \leq V^u(x)$$

# Finite-horizon problem (partial Lagrangian)

A partially dualized version:

$$\inf_{\boldsymbol{\nu} \in \mathcal{L}^2} \sup \mathbb{E} [(\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}) + \boldsymbol{\nu}^\top (F_u \mathbf{u} + F_x \mathbf{x} + \mathbf{s} - h)]$$
$$\text{s.t. } \left. \begin{array}{l} \mathbf{u} \in \mathcal{N}, \mathbf{s} \in \mathcal{L}^2 \\ \mathbf{x} = \mathcal{B}\mathbf{u} + \mathcal{C}\mathbf{w} \\ \mathbf{s} \geq 0 \end{array} \right\} \quad \mathbb{P}\text{-a.s.}$$

A lower bound makes multipliers linear in uncertainty

# Finite-horizon problem (partial Lagrangian)

A partially dualized version:

$$\begin{aligned} \inf_{\boldsymbol{\nu} \in \mathcal{L}^2} \sup & \quad \mathbb{E} [(\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}) + \boldsymbol{\nu}^\top (F_u \mathbf{u} + F_x \mathbf{x} + \mathbf{s} - h)] \\ \text{s.t.} \quad & \left. \begin{array}{l} \mathbf{u} \in \mathcal{N}, \mathbf{s} \in \mathcal{L}^2 \\ \mathbf{x} = \mathcal{B}\mathbf{u} + \mathcal{C}\mathbf{w} \\ \mathbf{s} \geq 0 \end{array} \right\} \quad \mathbb{P}\text{-a.s.} \end{aligned}$$



$$\begin{aligned} V_N^\ell(x) := \inf_{\mathbf{Y}} \sup & \quad \mathbb{E} [(\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}) + \boldsymbol{\nu}^\top (F_u \mathbf{u} + F_x \mathbf{x} + \mathbf{s} - h)] \\ \text{s.t.} \quad & \left. \begin{array}{l} \mathbf{u} \in \mathcal{N}, \mathbf{s} \in \mathcal{L}^2 \\ \mathbf{x} = \mathcal{B}\mathbf{u} + \mathcal{C}\mathbf{w} \\ \mathbf{s} \geq 0, \quad \boldsymbol{\nu} = \mathbf{Y}\mathbf{w} \end{array} \right\} \quad \mathbb{P}\text{-a.s.} \end{aligned}$$

# Finite-horizon problem (lower bound)

First deal with the inner maximization:

$$\begin{aligned} & \sup_Y \mathbb{E} [\mathbf{w}^\top Y^\top [F_u \mathbf{u} + F_x \mathbf{x} + F_s \mathbf{s} - h]] \\ = & \sup_Y \mathbb{E} [\text{tr} \{ [F_u \mathbf{u} + F_x \mathbf{x} + F_s \mathbf{s} - h] \mathbf{w}^\top Y^\top \}] \\ = & \begin{cases} 0 & \text{if } \mathbb{E} [(F_u \mathbf{u} + F_x \mathbf{x} + F_s \mathbf{s} - h) \mathbf{w}^\top] = 0 \\ \infty & \text{otherwise.} \end{cases} \end{aligned}$$

Restricting multipliers to linear form results in an exp. constraint.

- Eliminate states  $\rightarrow$  equivalent to:

$$(F_u + F_x \mathcal{B}) \mathbb{E}[\mathbf{u} \mathbf{w}^\top] + \mathbb{E}[\mathbf{s} \mathbf{w}^\top] + (F_x \mathcal{C} - h e_0^\top) \mathbb{E}[\mathbf{w} \mathbf{w}^\top] = 0$$

# Finite-horizon problem (lower bound)

Then deal with expectations:

$$(F_u + F_x \mathcal{B}) \mathbb{E}[\mathbf{u}\mathbf{w}^\top] + \mathbb{E}[\mathbf{s}\mathbf{w}^\top] + (F_x \mathcal{C} - h e_0^\top) \mathbb{E}[\mathbf{w}\mathbf{w}^\top] = 0, \quad \mathbf{s} \geq 0$$

①                    ②                    ③                    ②

Some technical steps:

- ① For all  $\mathbf{u} \in \mathcal{N}$  there exists  $Q \in \mathcal{U}$  s.t.  $\mathbb{E}[\mathbf{u}\mathbf{w}^\top] = QGM$
- ② For all  $0 \leq \mathbf{s} \in \mathcal{L}^2$ , there exists  $S$  s.t.  $\mathbb{E}[\mathbf{s}\mathbf{w}^\top] = S^\top M$  and  $MS \succeq_{\mathcal{K}} 0$
- ③  $\mathbb{E}[\mathbf{w}\mathbf{w}^\top] = M$ , and right-multiply everything by  $M^{-1}$

# Finite-horizon problem (lower bound)

Then deal with expectations:

$$(F_u + F_x \mathcal{B}) Q G M + S^\top M + (F_x \mathcal{C} - h e_0^\top) M = 0$$

①                  ②                  ③

Some technical steps:

- ① For all  $\mathbf{u} \in \mathcal{N}$  there exists  $Q \in \mathcal{U}$  s.t.  $\mathbb{E}[\mathbf{u}\mathbf{w}^\top] = QGM$
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  - ③  $\mathbb{E}[\mathbf{w}\mathbf{w}^\top] = M$ , and right-multiply everything by  $M^{-1}$
- 4) Explicit minimization over  $\mathbf{u} \in \mathcal{N}$  results in  $\mathbf{u}^* = QG\mathbf{w}$ .

# Finite-horizon problem (lower bound)

An equivalent form:

$$\begin{aligned} V_N^\ell(x) := \inf \mathbb{E} [\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}] \\ \text{s.t. } Q \in \mathcal{U}, \quad M S \succeq_{\mathcal{K}} 0 \text{ (column-wise)} \\ (F_x \mathcal{B} + F_u) Q \mathcal{G} + F_x \mathcal{C} + S^\top = h e_0^\top \end{aligned}$$

Result is a conic program:

- QP if all sets polyhedral. Size scales with  $N^2$
- Provides lower bound on value:

$$V_N^\ell(x) \leq V_N^*(x)$$

# Finite-horizon bounds

Upper bounding problem:

$$V_N^u(x) := \inf \mathbb{E} [\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}]$$

$$\text{s.t. } Q \in \mathcal{U}, \quad S \succeq_{\mathcal{K}_*} 0 \text{ (column-wise)}$$

$$(F_x \mathcal{B} + F_u) Q \mathcal{G} + F_x \mathcal{C} + S^\top = h e_0^\top$$

Lower bounding problem:

$$V_N^\ell(x) := \inf \mathbb{E} [\mathbf{u}^\top \mathcal{R} \mathbf{u} + \mathbf{x}^\top \mathcal{Q} \mathbf{x}]$$

$$\text{s.t. } Q \in \mathcal{U}, \quad \textcolor{red}{M} S \succeq_{\mathcal{K}} 0 \text{ (column-wise)}$$

$$(F_x \mathcal{B} + F_u) Q \mathcal{G} + F_x \mathcal{C} + S^\top = h e_0^\top$$

# Infinite-horizon problems

An infinite horizon version:

$$V_{\infty}^*(x) := \min_{\pi} \mathbb{E} \left[ \sum_{i=0}^{\infty} \alpha^i (x_i^\top Q x_i + u_i^\top R u_i) \right]$$

subject to:

$$\left. \begin{aligned} x_{i+1} &= Ax_i + Bu_i + Gw_i \\ (x_i, u_i) &\in Z, \quad x_0 = x \end{aligned} \right\} \mathbb{P}\text{-a.s.}$$

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A **finite** horizon version:

$$V_\infty^*(x) := \min_{\pi} \mathbb{E} \left[ \sum_{i=0}^{N-1} \alpha^i (x_i^\top Q x_i + u_i^\top R u_i) + \alpha^N V_\infty^*(x_N) \right]$$

subject to:

$$\left. \begin{aligned} x_{i+1} &= Ax_i + Bu_i + Gw_i \\ (x_i, u_i) &\in Z, \quad x_0 = x \end{aligned} \right\} \mathbb{P}\text{-a.s.}$$

# Infinite-horizon problems

Upper bounds:  $V_0^u \geq V_\infty^* \implies V_N^u \geq V_\infty^*$

$$V_N^u(x) := \min_{\pi_N} \mathbb{E} \left[ \sum_{i=0}^{N-1} \alpha^i (x_i^T Q x_i + u_i^T R u_i) + \alpha^N V_0^u(x_N) \right]$$

subject to:

$$\left. \begin{aligned} x_{i+1} &= Ax_i + Bu_i + Gw_i \\ (x_i, u_i) &\in Z, \quad x_0 = x \end{aligned} \right\} \mathbb{P}\text{-a.s.}$$

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Lower bounds:  $V_0^l \leq V_\infty^* \implies V_N^l \leq V_\infty^*$

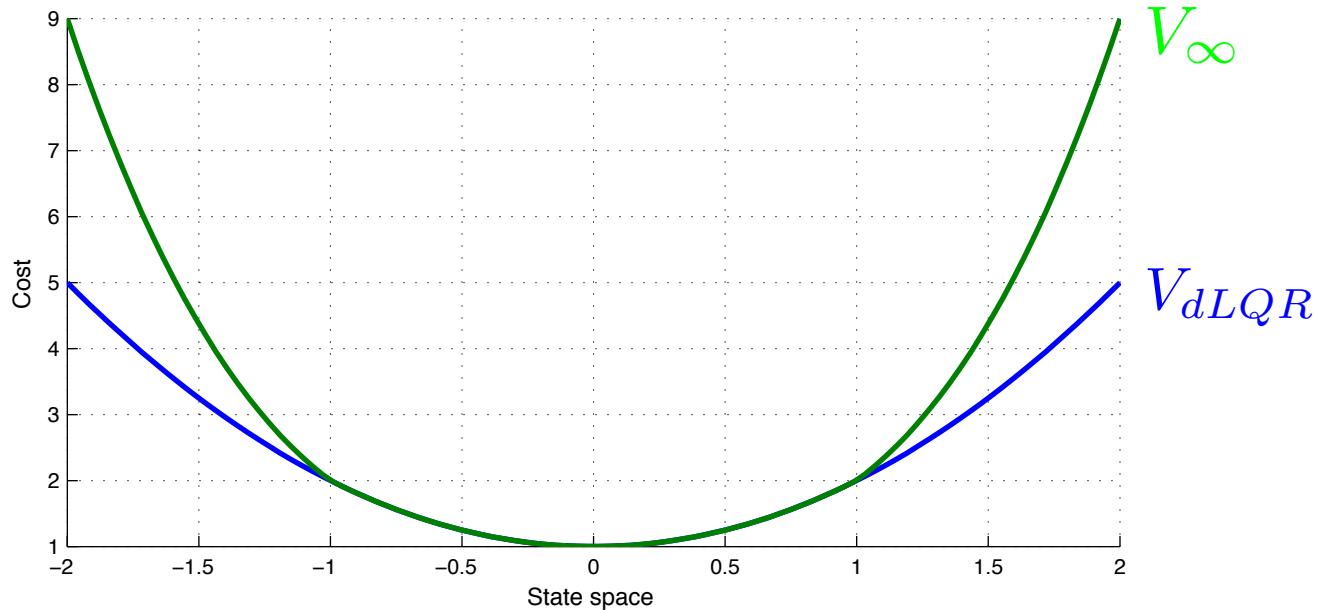
$$V_N^l(x) := \min_{\pi_N} \mathbb{E} \left[ \sum_{i=0}^{N-1} \alpha^i (x_i^T Q x_i + u_i^T R u_i) + \alpha^N V_0^l(x_N) \right]$$

subject to:

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# Infinite-horizon problems

An initial lower bound:



Unconstrained case (discounted LQR):  $V_{dLQR} = x^T P x + q$ , where

$$P = Q + \bar{A}^T P \bar{A} - \bar{A}^T P \bar{B} (R + \bar{B}^T P \bar{B})^{-1} \bar{B}^T P \bar{A}$$

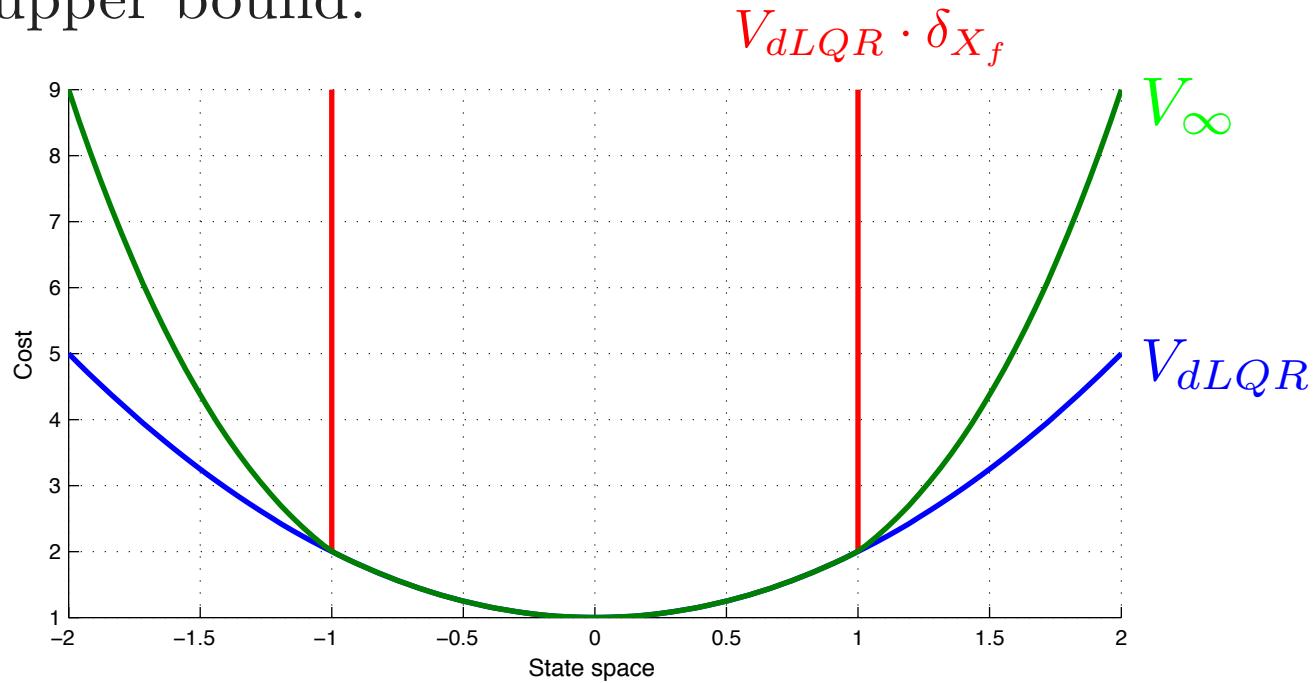
$$q = \frac{\text{tr}(\bar{C} P \bar{C} M)}{1 - \alpha}$$

$$K = -(R + \bar{B}^T P \bar{B})^{-1} \bar{B}^T P \bar{A}$$

and  $\bar{A} := \sqrt{\alpha} A$ ,  $\bar{B} := \sqrt{\alpha} B$  and  $C := \sqrt{\alpha} C$ .

# Infinite-horizon problems

An initial upper bound:



$X_f$  robust positive invariant (RPI) for  $K$

- $(A + BK)x + Cw \in X_f$  for all  $x \in X_f$  and  $w \in W$
- $(x, Kx) \in Z$  for all  $x \in X_f$

# Example (infinite-horizon)

- Two-state system:

$$x^+ = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} x + \begin{bmatrix} 0.2 \\ 1 \end{bmatrix} u + \begin{bmatrix} 0.1 \\ 0.1 \end{bmatrix} w$$

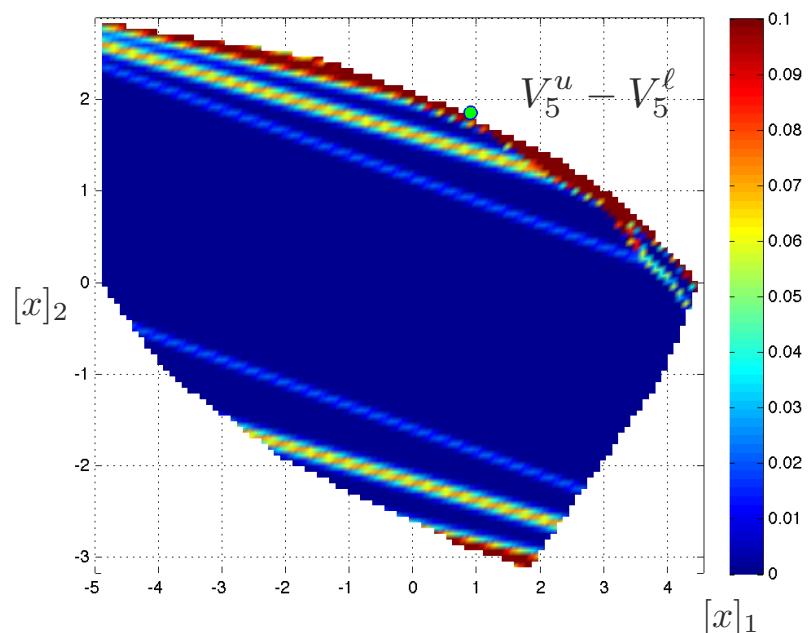
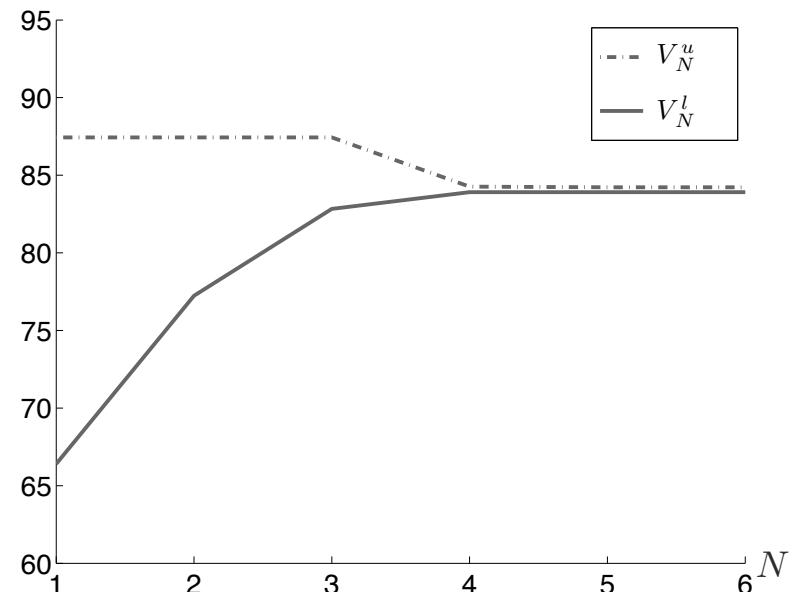
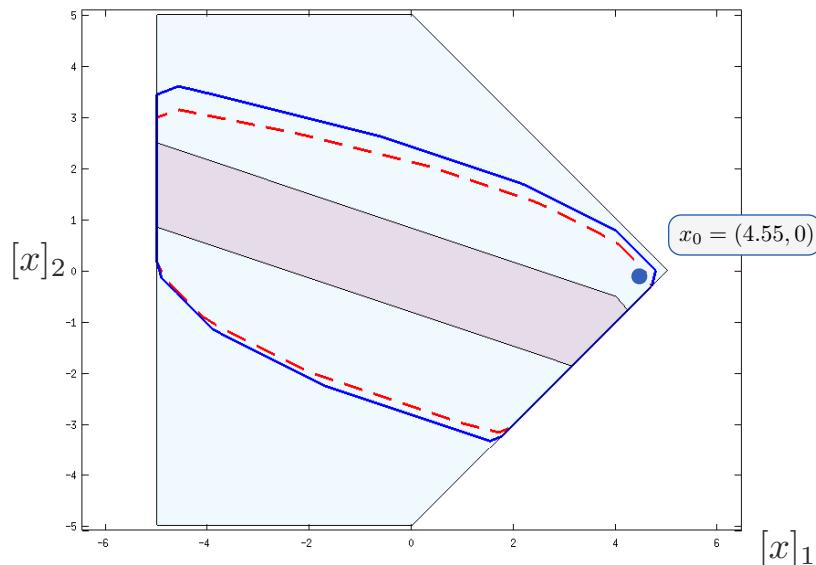
- Cost function:

$Q = I$ ,  $R = 1$ ,  $P = \text{dLQR solution}$ ,  $\alpha = 0.95$

- Disturbance/input constraints:

$|u| \leq 1$ ,  $w$  uniform on  $[0, 2]$

- State constraints:



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