

# The role of state constraints for turnpike behaviour and strict dissipativity of optimal control problems

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based on joint work with Roberto Guglielmi (GSSI, L'Aquila)

Control of state constrained dynamical systems  
Padova, 25–29 September 2017

# Outline

- The turnpike property
- Strict dissipativity
- Linear quadratic problems
- Main results

# System class

We consider **nonlinear discrete time** control systems

$$x_{\mathbf{u}}(n+1) = f(x_{\mathbf{u}}(n), \mathbf{u}(n)), \quad x_{\mathbf{u}}(0) = x$$

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$$\underset{\mathbf{u}}{\text{minimise}} \quad J_N(x, \mathbf{u}) = \sum_{n=0}^{N-1} \ell(x_{\mathbf{u}}(n), \mathbf{u}(n))$$

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We **illustrate** the property by two simple examples

# Example 1: minimum energy control

**Example:** Keep the state of the system inside a given interval  $X$  minimising the quadratic control effort

$$\ell(x, u) = u^2$$

with dynamics

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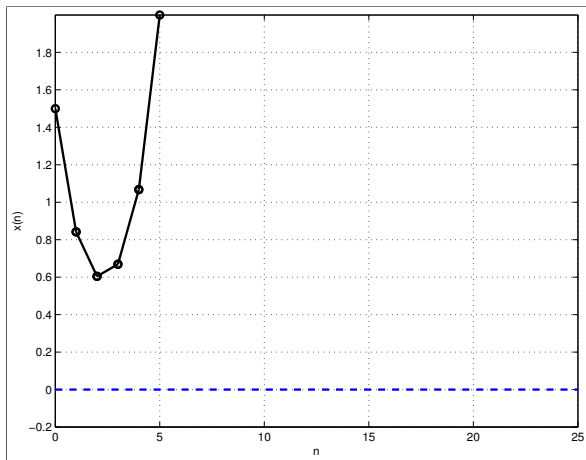
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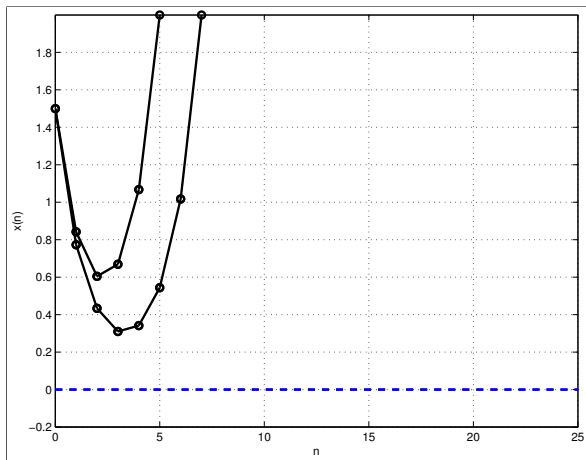
# Example 1: optimal trajectories



Optimal trajectory for  $N = 5$

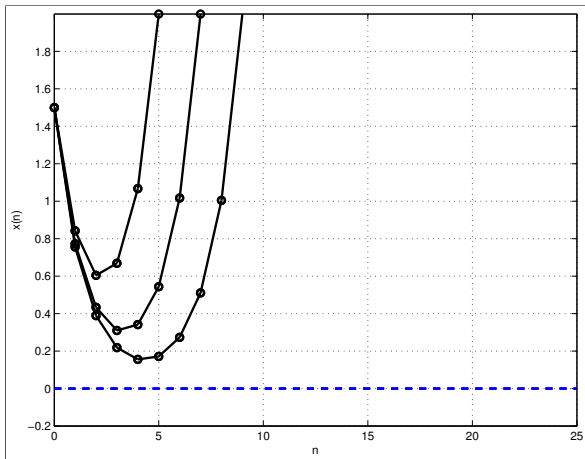


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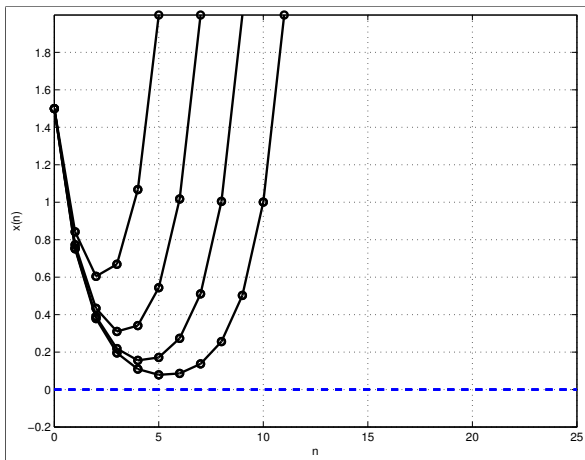
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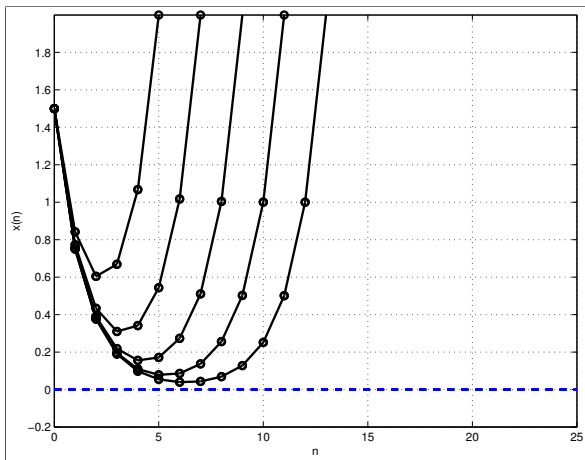
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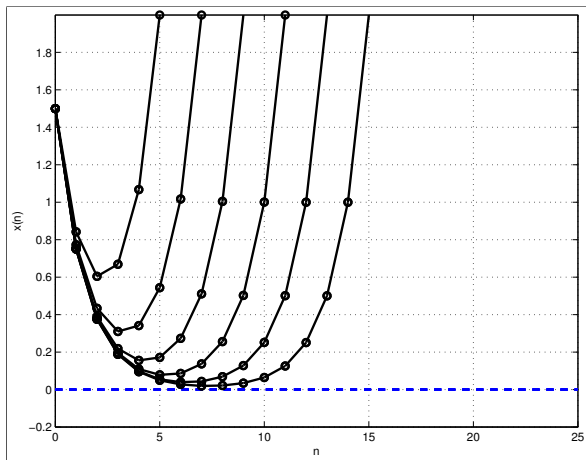
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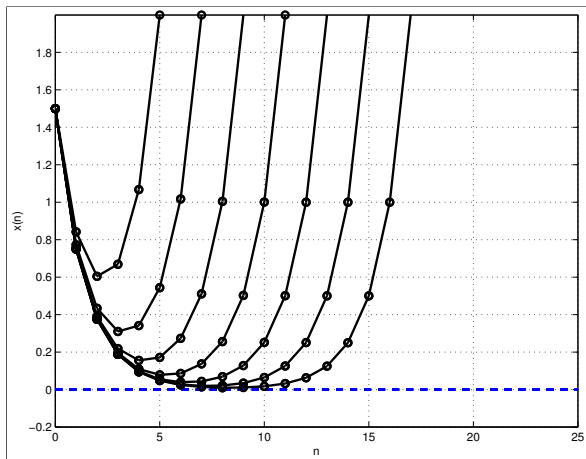
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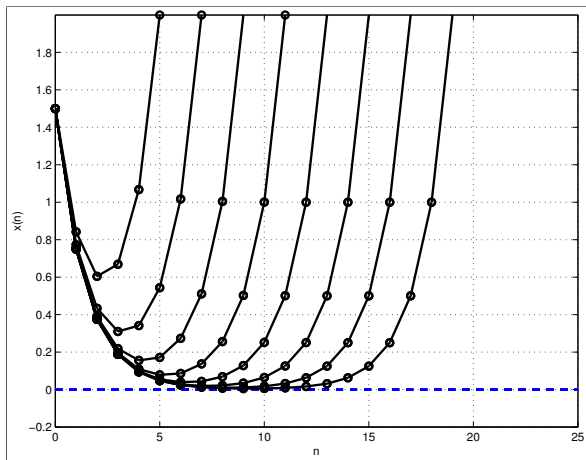
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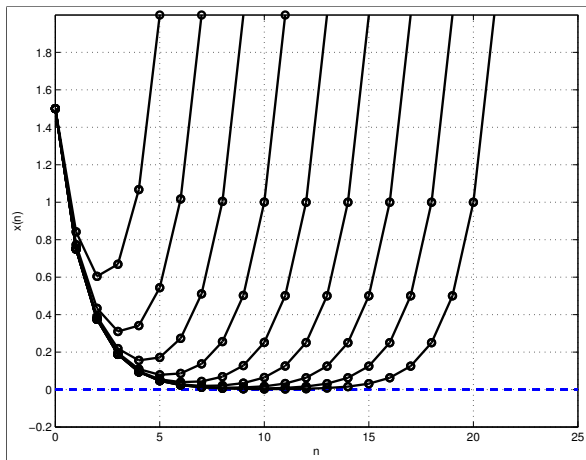
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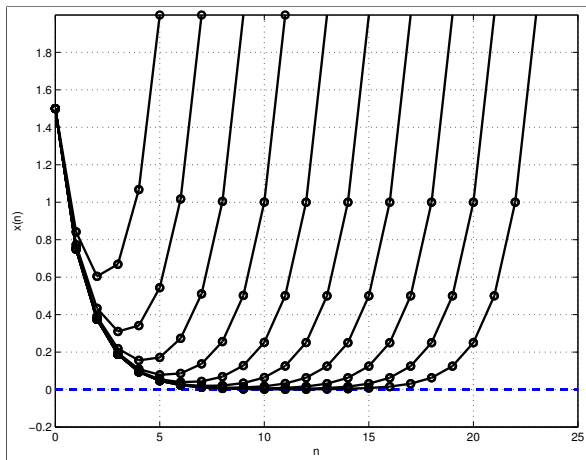
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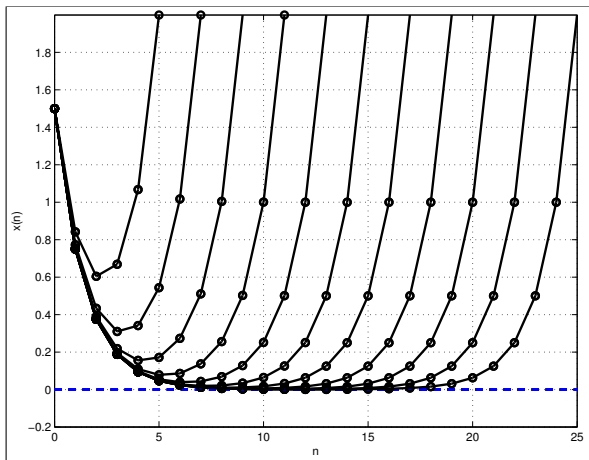


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## Example 2: a macroeconomic model

The second example is a 1d macroeconomic model

[Brock/Mirman '72]

Minimise the finite horizon objective with

$$\ell(x, u) = -\ln(Ax^\alpha - u), \quad A = 5, \alpha = 0.34$$

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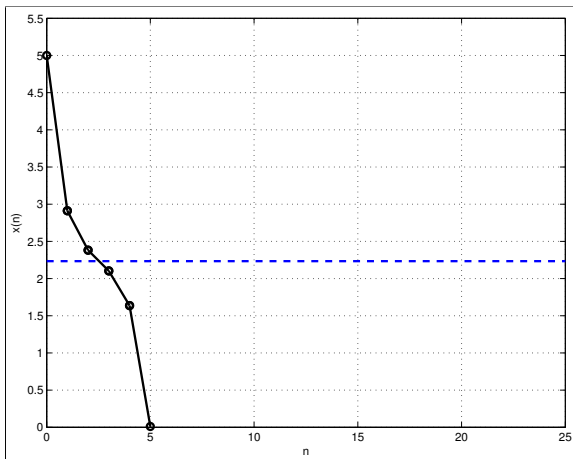
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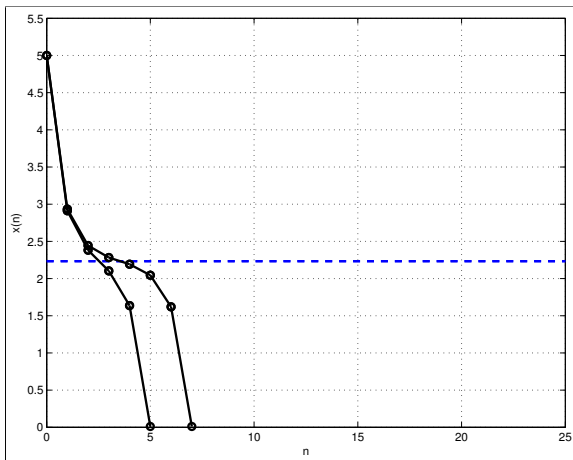
One may thus expect that finite horizon optimal trajectories also **stay for a long time** near that equilibrium

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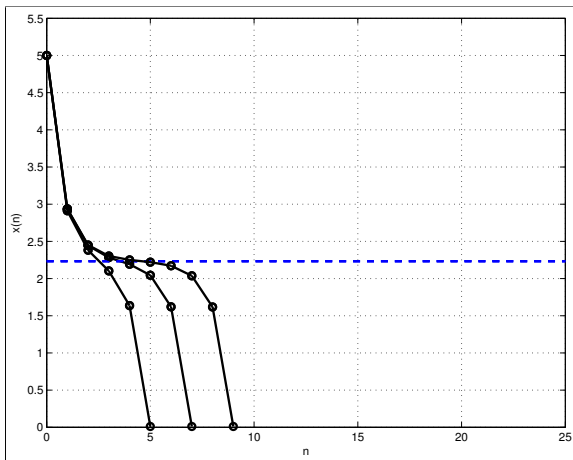
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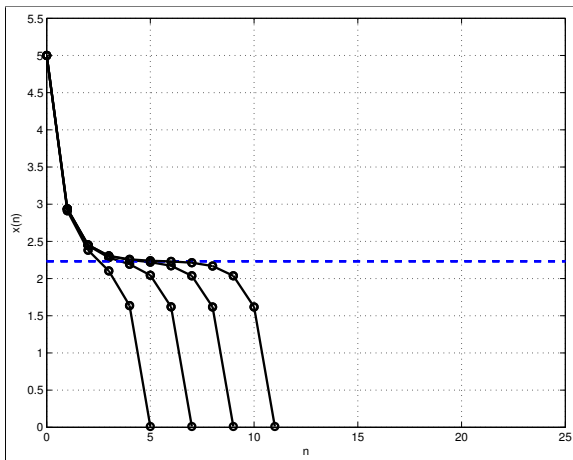


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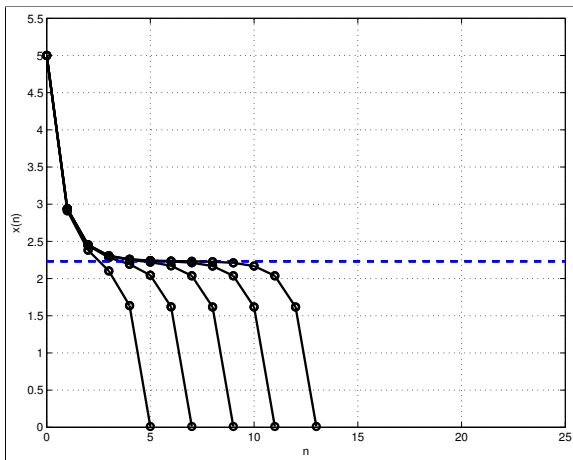
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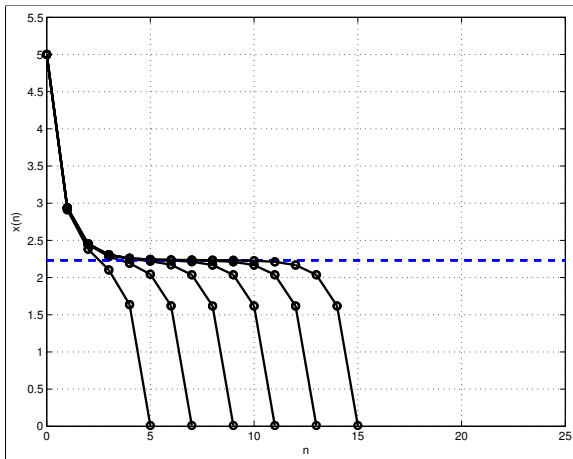
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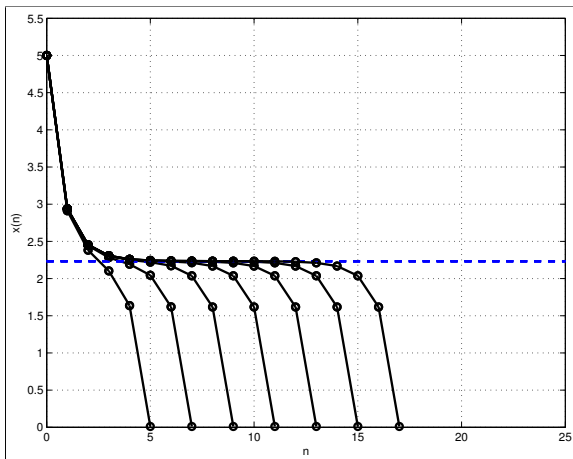
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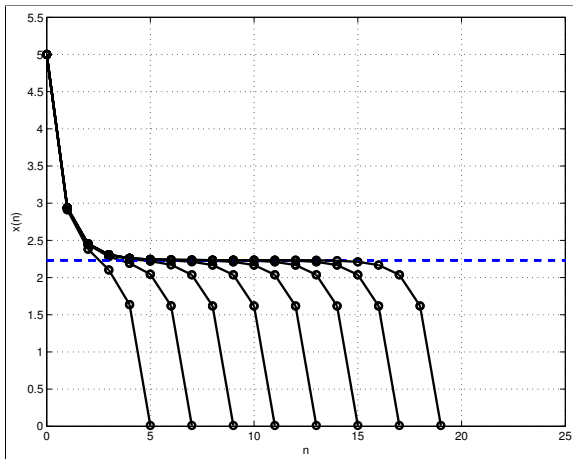
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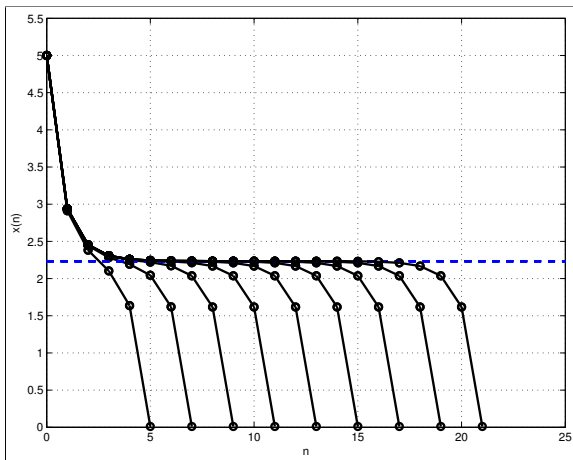
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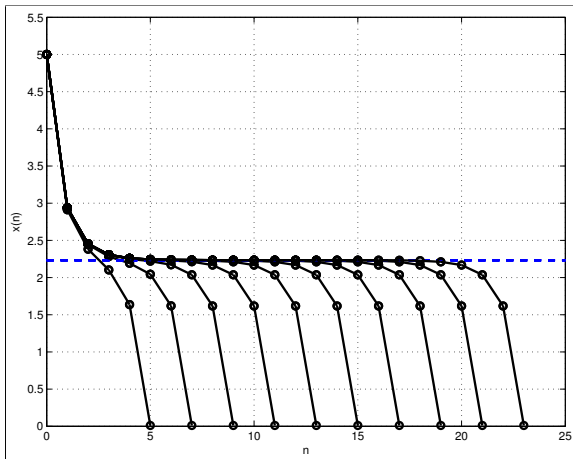
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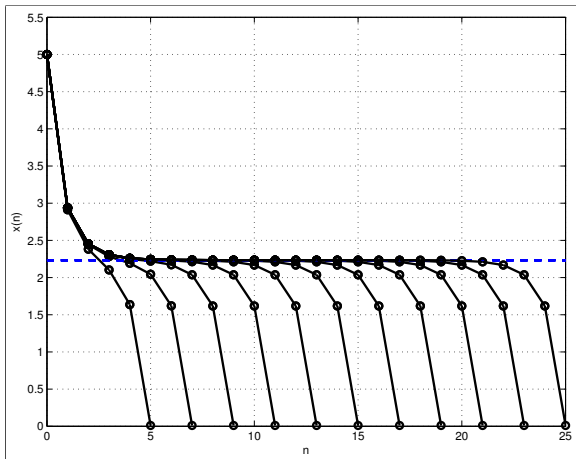
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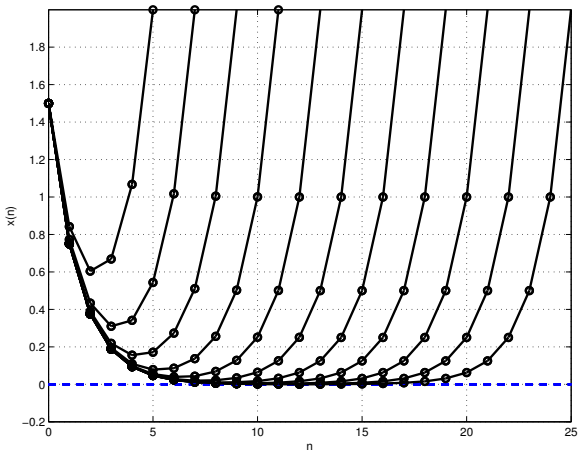


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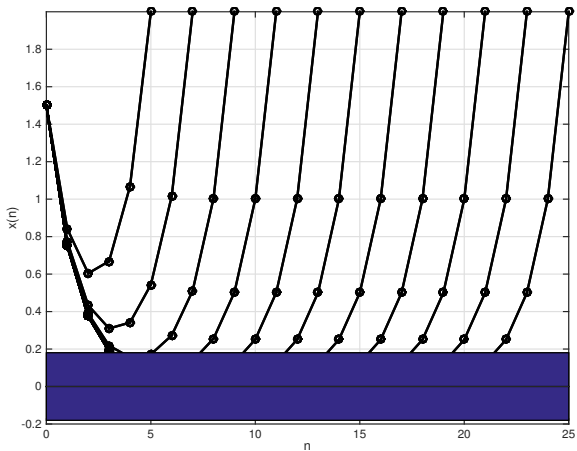


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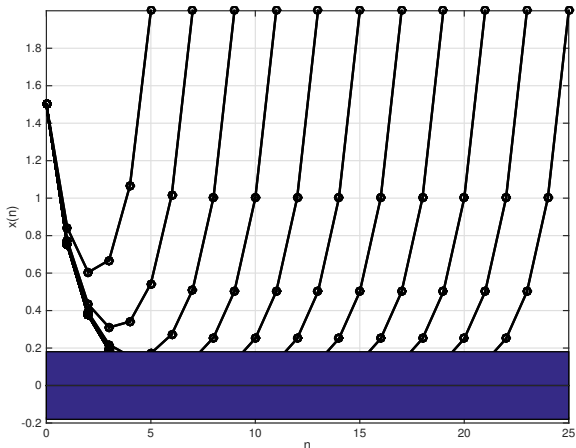
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Number of points outside the blue neighbourhood is **bounded**  
by a number independent of  $N$  (here: by 8)

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**Near equilibrium turnpike property:** For each  $\varepsilon > 0$ ,  $\delta > 0$  and  $\rho > 0$  there is  $C_{\rho, \varepsilon, \delta} > 0$  such that for all  $x \in \mathbb{X}$  and  $N \in \mathbb{N}$ , **all trajectories**  $x_{\mathbf{u}}$  with  $x_{\mathbf{u}}(0) = x \in B_\rho(x^e)$  and  $J_N(x, \mathbf{u}) \leq N\ell(x^e, u^e) + \delta$  satisfy the inequality

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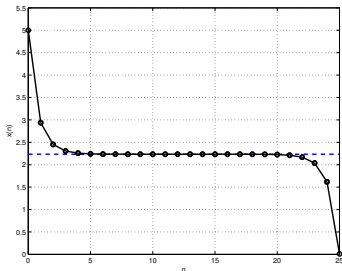


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- Many applications, e.g., **structural insight** in economic equilibria; **synthesis** of optimal trajectories [Anderson/Kokotovic '87]

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MPC is a method in which an **optimal control problem on an infinite horizon**

$$\underset{\mathbf{u}}{\text{minimise}} \quad J_{\infty}(x, \mathbf{u}) = \sum_{n=0}^{\infty} \ell(x_{\mathbf{u}}(n), \mathbf{u}(n))$$

is approximated by the **iterative** solution of **finite horizon problems**

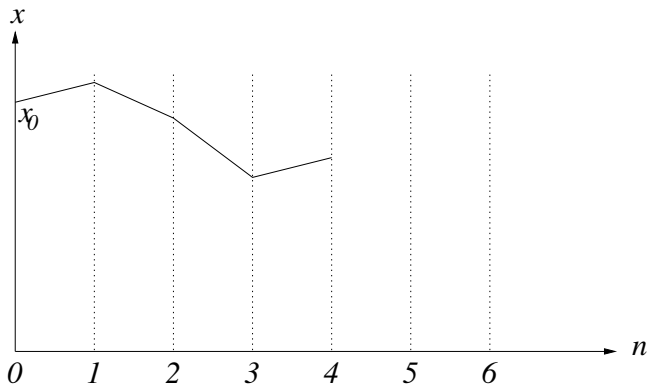
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with fixed  $N \in \mathbb{N}$

# MPC from the trajectory point of view

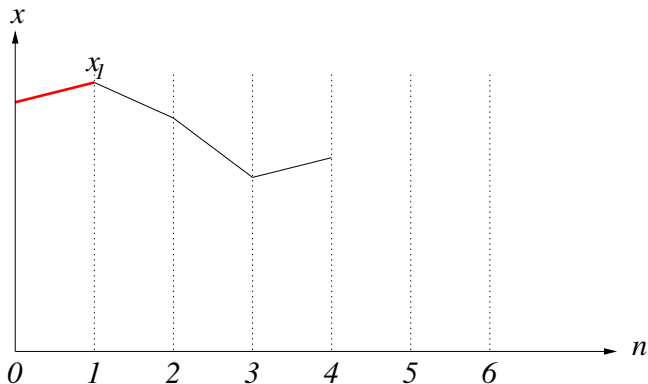


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black = predictions (open loop optimization)

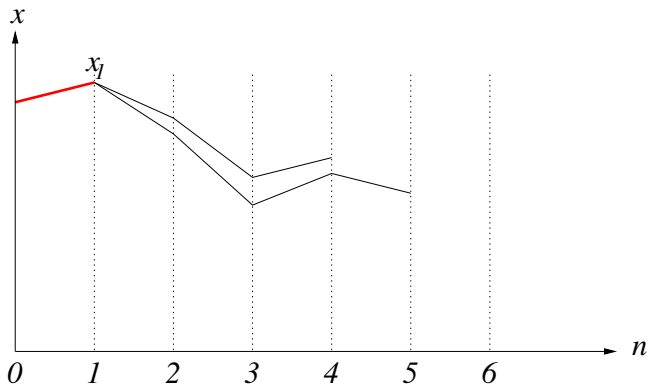
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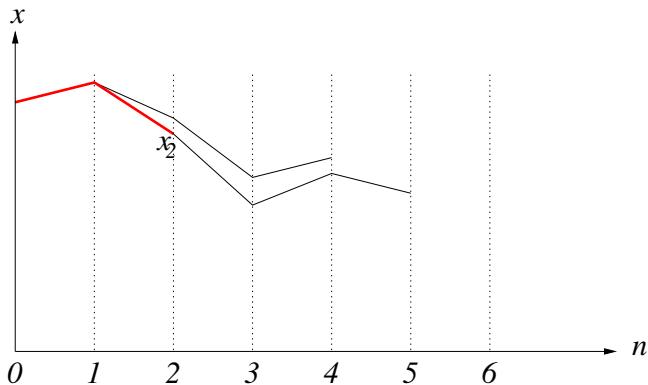
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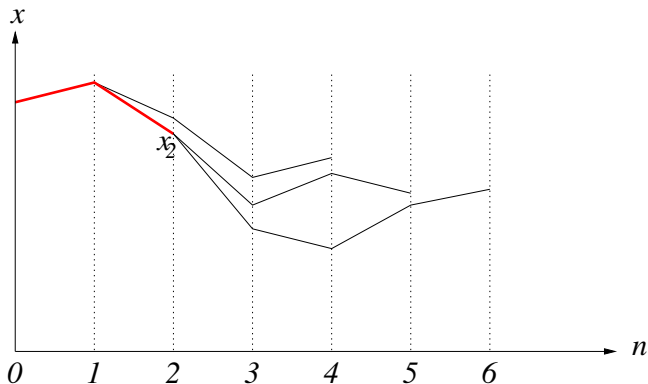
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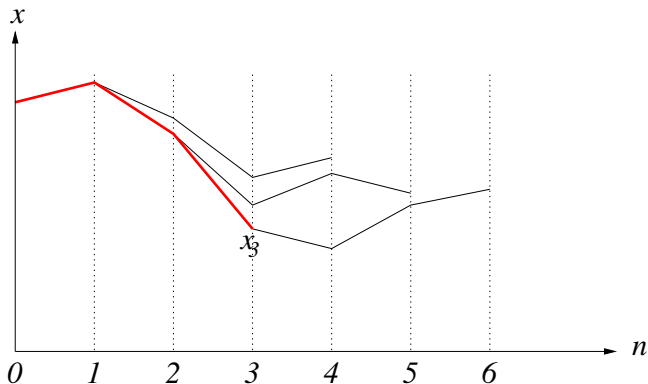
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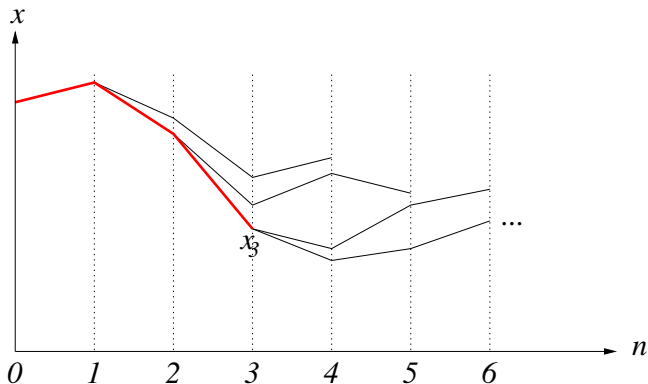
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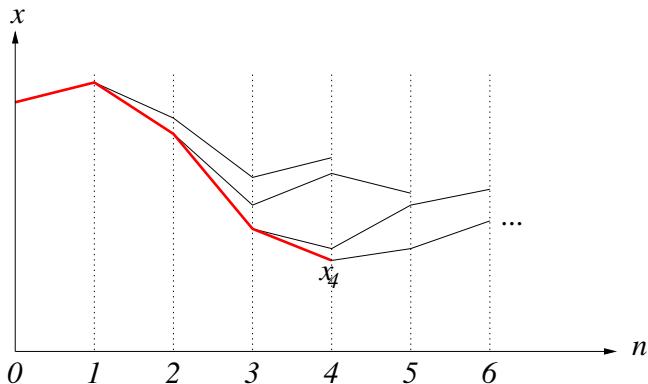
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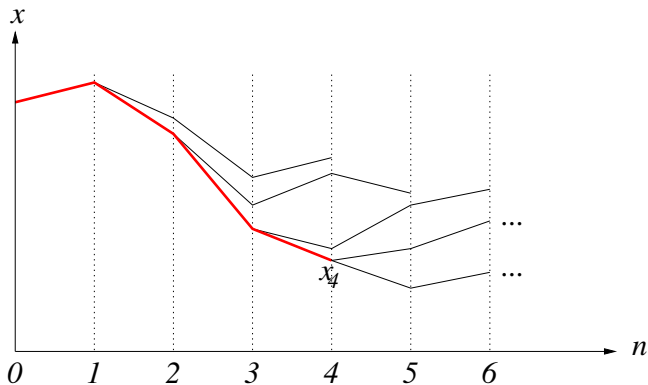


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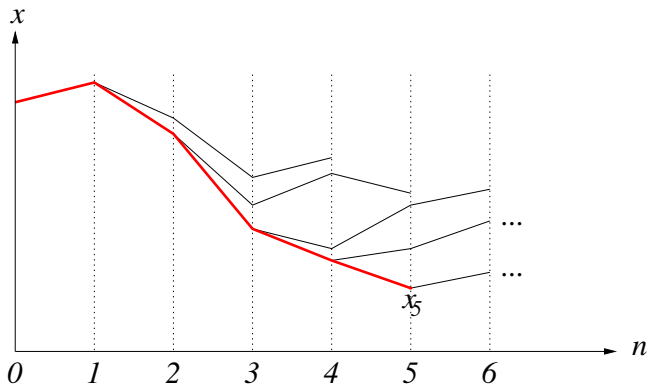
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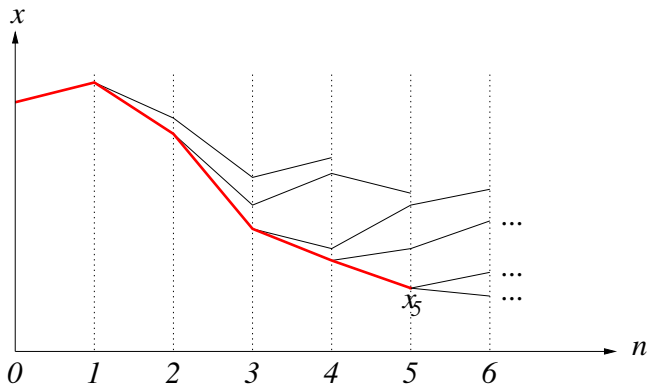
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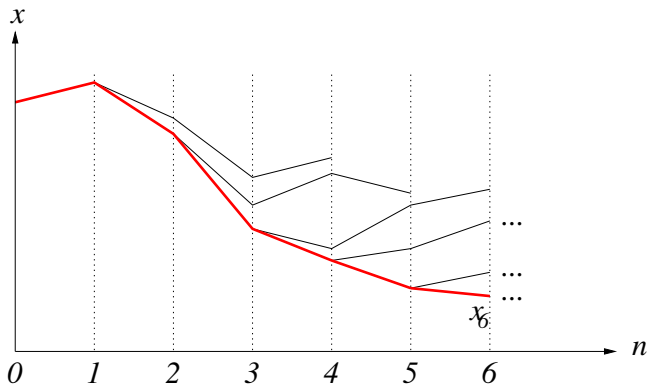
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The result exploits that the **red** closed loop trajectory approximately **follows the first part** of the **black** predictions up to the equilibrium

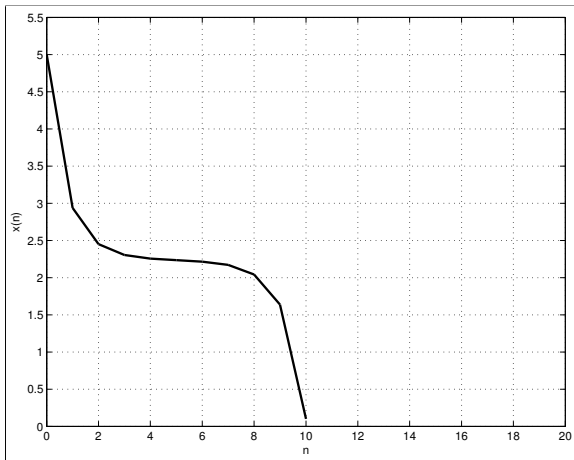
# Approximation result for MPC

If the finite horizon problems have the **turnpike property**, then a **rigorous approximation result** can be proved

The result exploits that the **red** closed loop trajectory approximately **follows the first part** of the **black** predictions up to the equilibrium

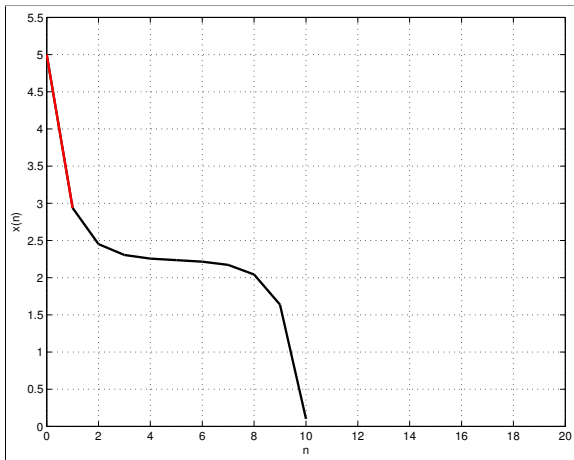
We **illustrate** this behaviour by our second example for  $N = 10$

## MPC for Example 2

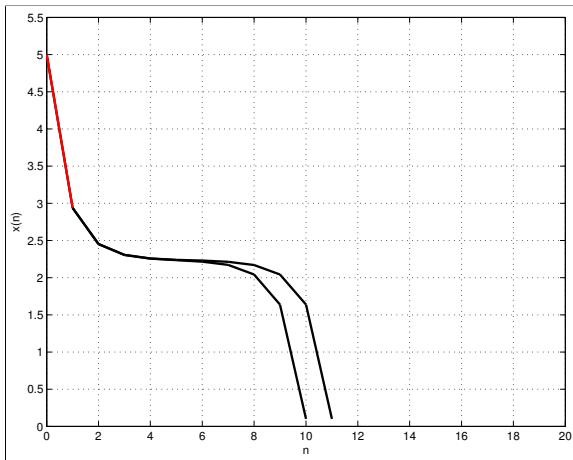




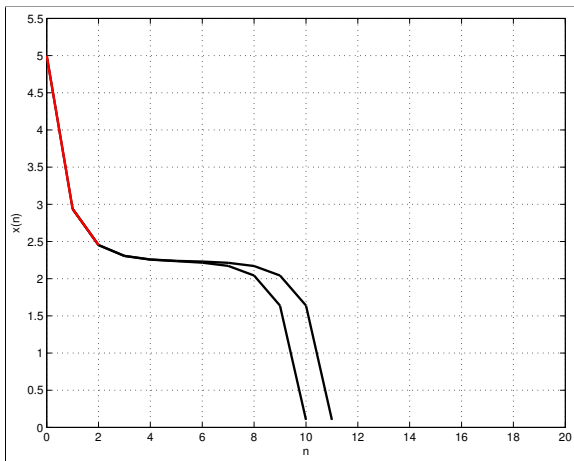
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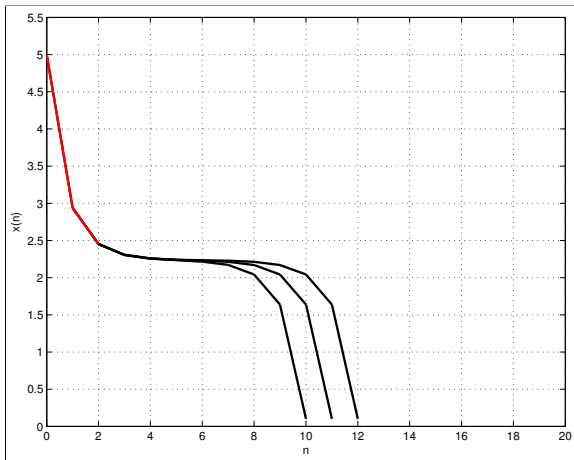
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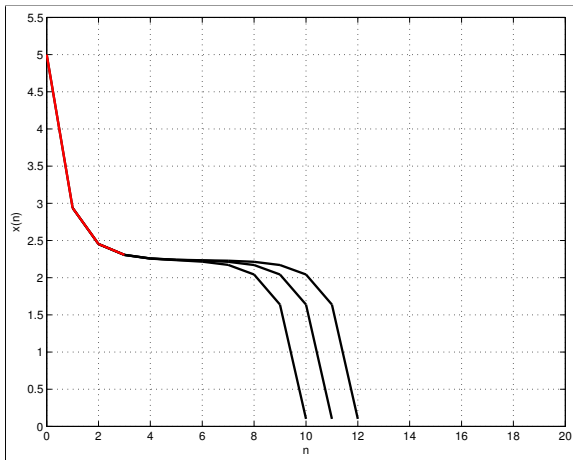
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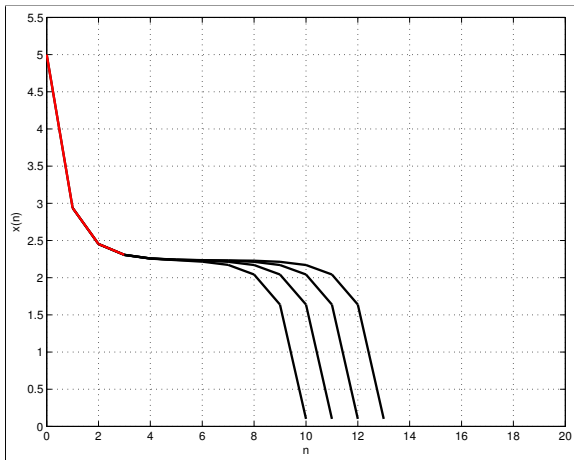
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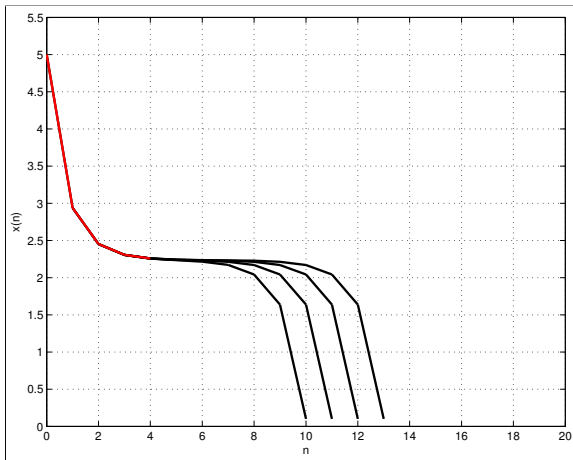
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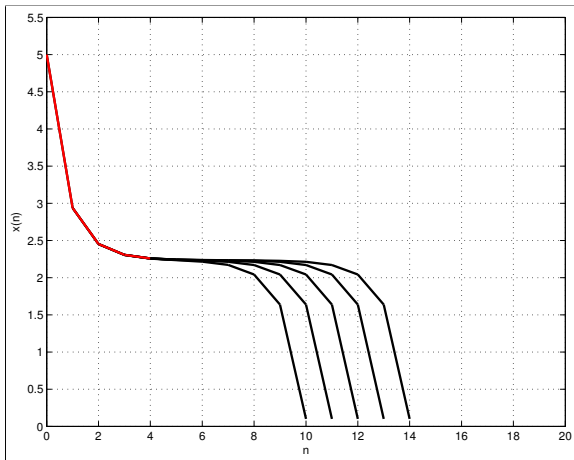
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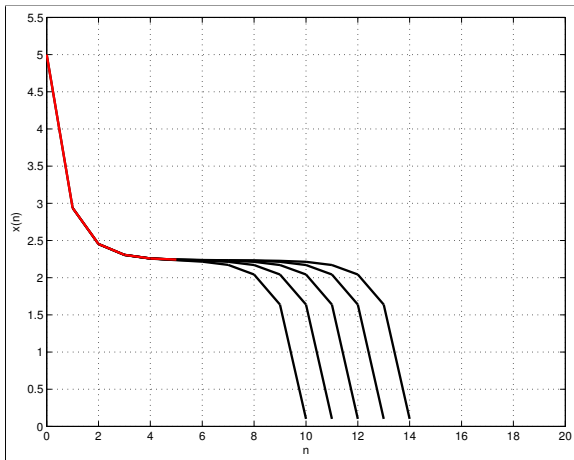


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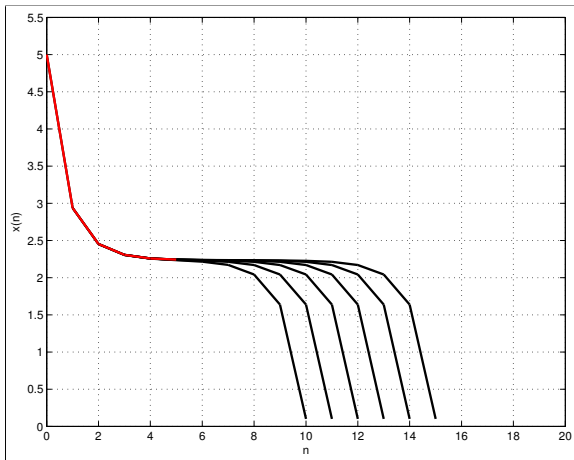




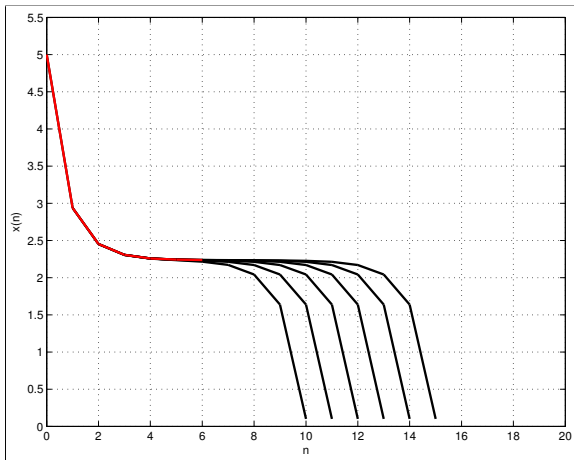
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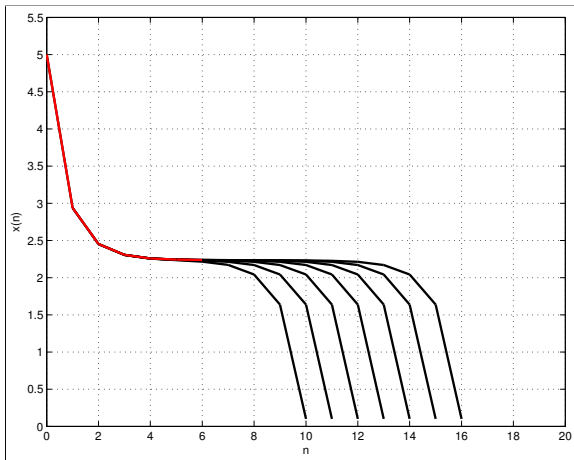
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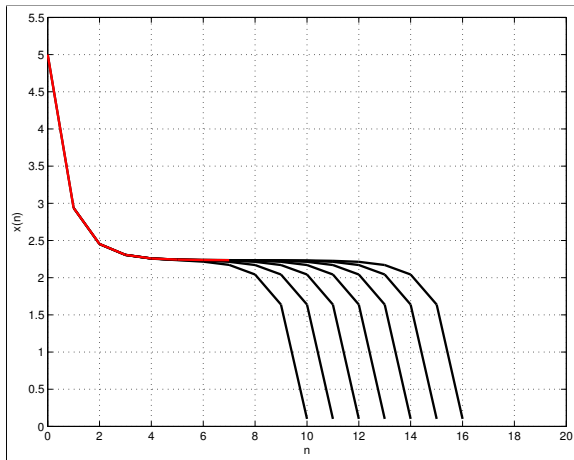
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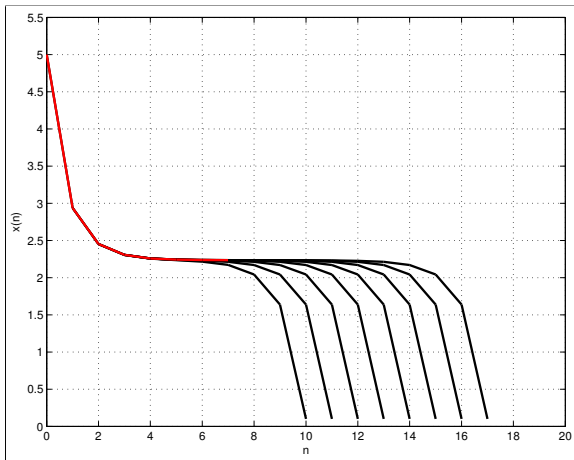
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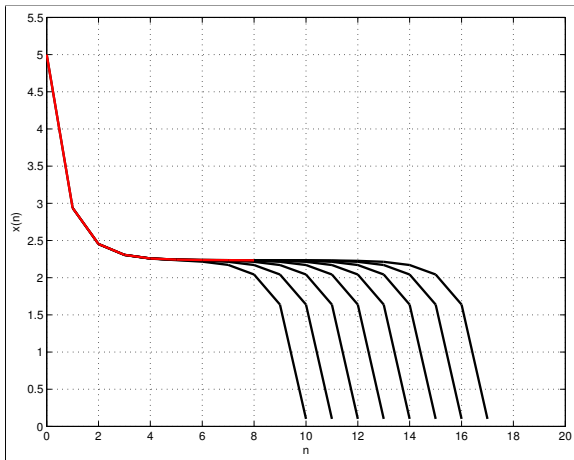
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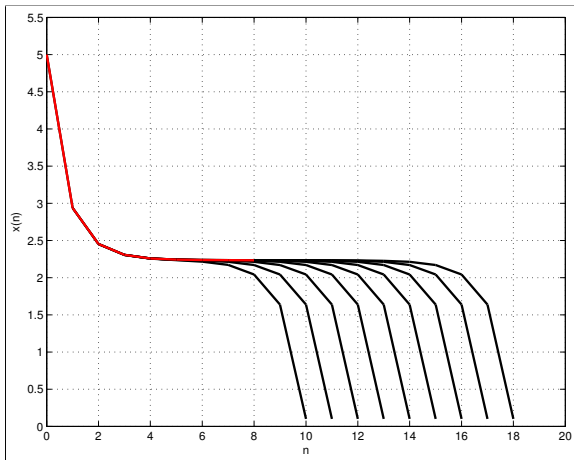
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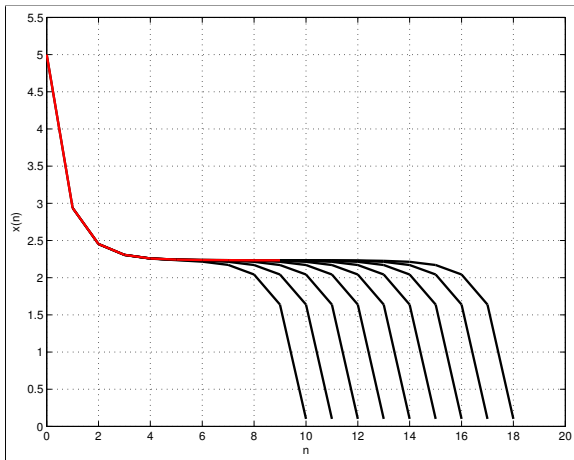


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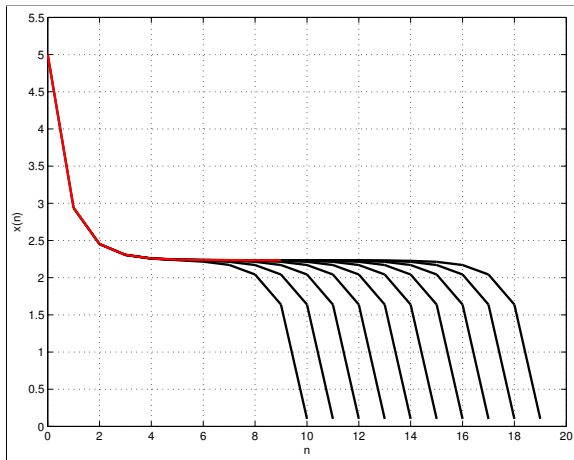




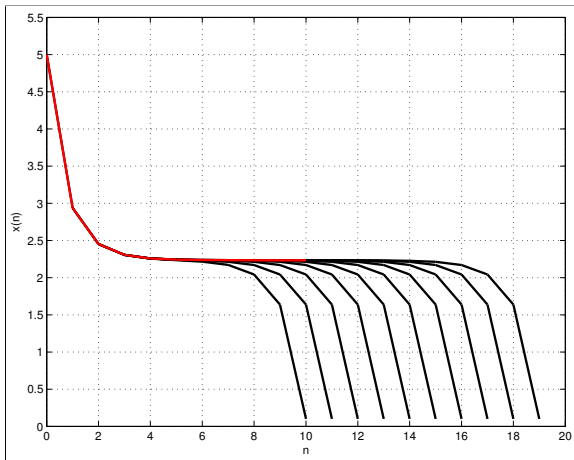
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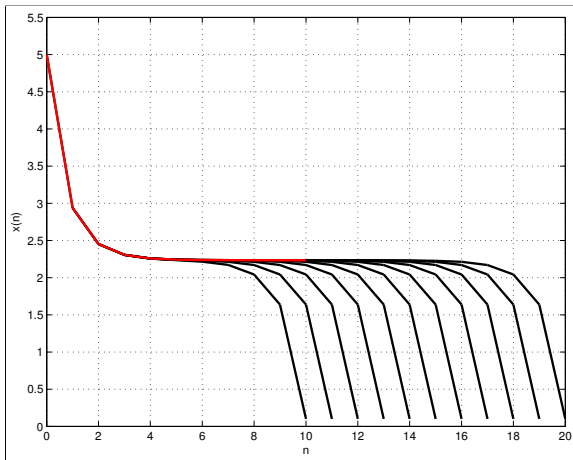
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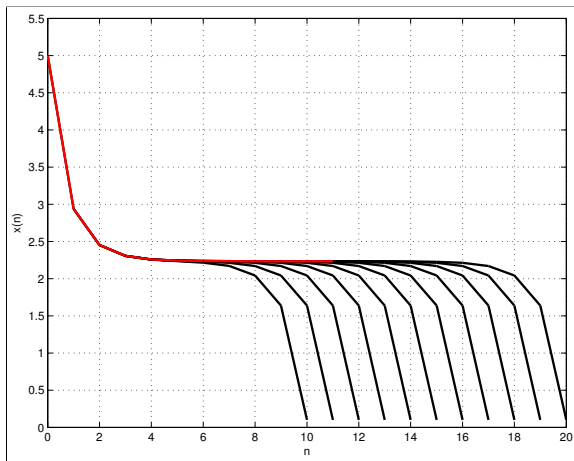
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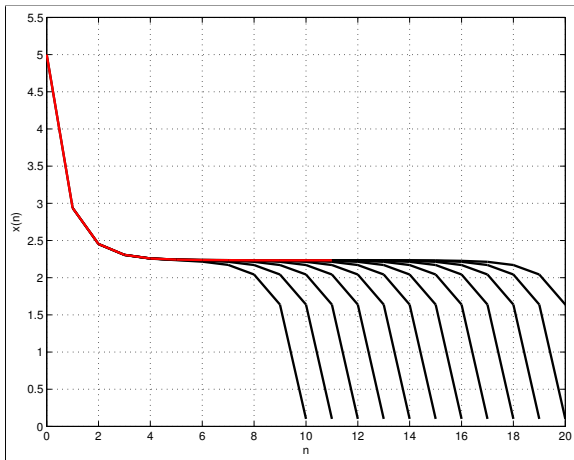
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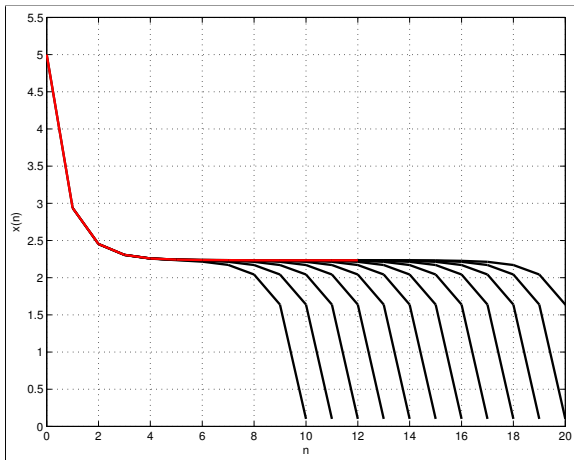
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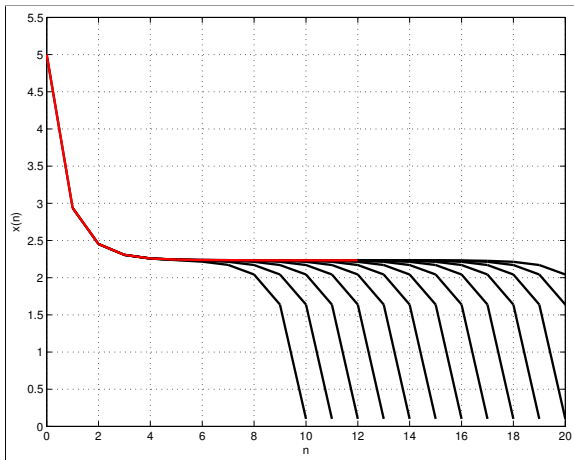
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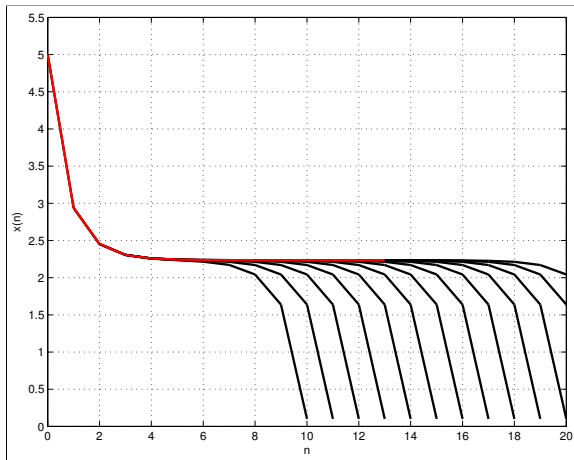


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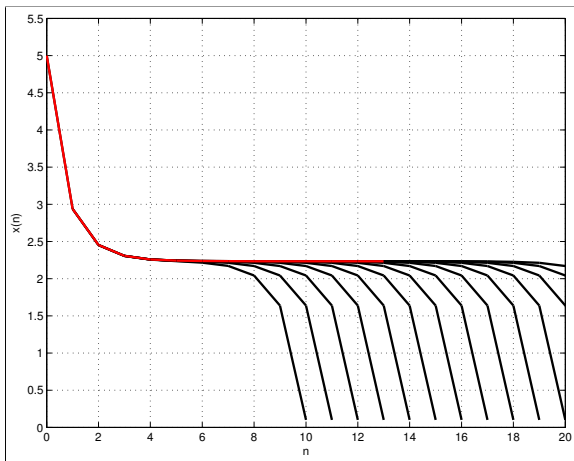




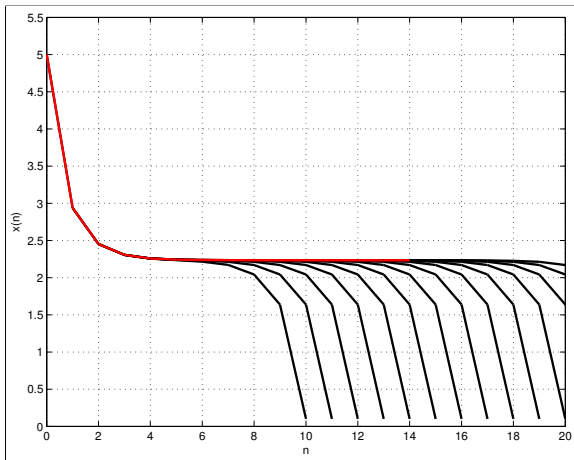
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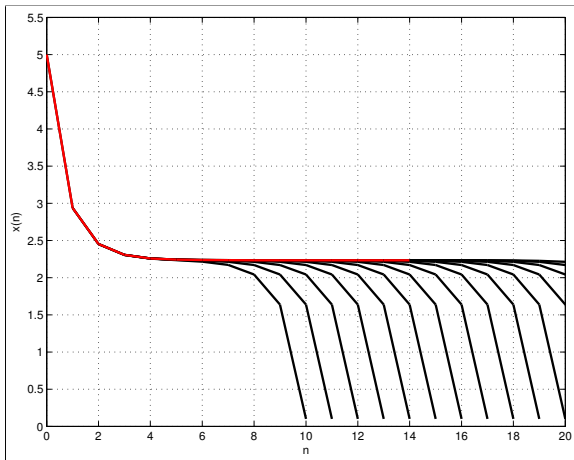
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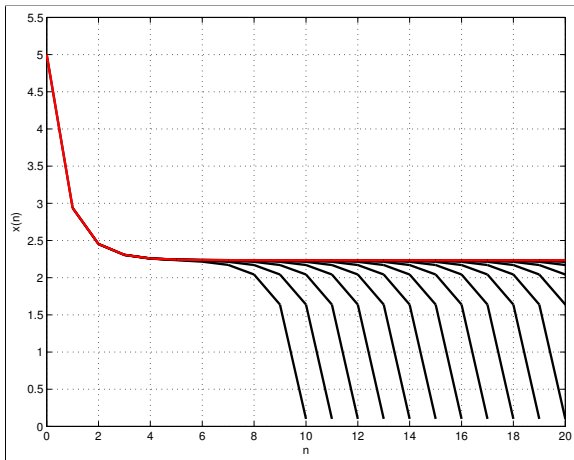
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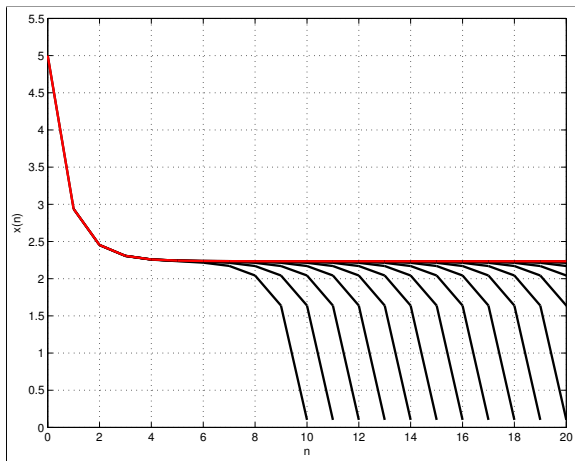
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## MPC for Example 2



Extension to **non equilibrium turnpikes** possible

[Zanon/Gr./Diehl '17, Gr./Müller '17, Gr./Pirkelmann '17]

Strict dissipativity

# Dissipativity

$$x^+ = f(x, u)$$

Introduce functions  $s : X \times U \rightarrow \mathbb{R}$  and  $\lambda : X \rightarrow \mathbb{R}$

$s(x, u)$  **supply rate**, measuring the (possibly negative) amount of energy supplied to the system via the input  $u$  in the next time step

$\lambda(x)$  **storage function**, measuring the amount of energy stored inside the system when the system is in state  $x$



# Dissipativity

**Definition** [cf. Willems '72] The system is called **strictly pre-dissipative** if there are  $x^e \in \mathbb{X}$ ,  $\alpha \in \mathcal{K}$  such that for all  $x \in \mathbb{X}$ ,  $u \in \mathbb{U}$  the inequality

$$\lambda(x^+) \leq \lambda(x) + s(x, u) - \alpha(\|x - x^e\|)$$

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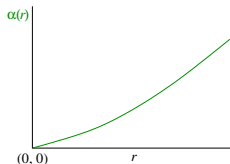
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$\alpha \in \mathcal{K}$ :  $\alpha : \mathbb{R}_0^+ \rightarrow \mathbb{R}_0^+$ , continuous,  
strictly increasing,  $\alpha(0) = 0$



The system is called **strictly dissipative** if it is strictly pre-dissipative with  $\lambda$  bounded from below

# Physical interpretation of dissipativity

$$\lambda(x^+) \leq \lambda(x) + s(x, u) - \alpha(\|x - x^e\|)$$

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strict dissipativity:

- energy can not be generated inside the system
- a certain amount of energy  $\alpha(\|x - x^e\|)$  **must be dissipated** (= given to the environment)

# History

Dissipativity was **defined** for continuous time systems in  
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Translation to **discrete time systems** is quite straightforward  
[Byrnes/Lin '94]

# Relation between strict dissipativity and turnpike

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Typical result:

**Theorem:** Assume  $\mathbb{X}$  is closed and bounded and  $\mathbb{U}$  is compact,  $\ell$  is continuous and bounded from below,  $x^e$  is an equilibrium around which the system is locally controllable and  $u^e \in \operatorname{argmin}\{\ell(x^e, u) \mid u \in \mathbb{U}, f(x^e, u) = x^e\}$

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Then the following statements are equivalent

- (a) The system is strictly dissipative with supply rate  $s(x, u) = \ell(x, u) - \ell(x^e, u^e)$  and a bounded storage function
- (b) The near equilibrium turnpike property holds



# Linear quadratic problems

## LQ problems

From now on we consider **linear quadratic** finite dimensional discrete time problems with  $X = \mathbb{R}^n$ ,  $U = \mathbb{R}^m$ ,

$$x^+ = Ax + Bu, \quad \ell(x, u) = x^T Qx + u^T Ru + b^T x + d^T u$$

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The same holds for **solutions**  $x(t)$  starting in  $v$  with  $u(t) \equiv 0$

# Storage functions for LQ problems

$$x^+ = Ax + Bu, \quad s(x, u) = x^T Qx + u^T Ru + b^T x + d^T u$$

**Lemma:** For LQ problems, a storage function  $\lambda$  can always be chosen of the form

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Moreover, the solution **satisfies**  $P > 0$  **if and only if** all unobservable eigenvalues satisfy  $|\mu| < 1$  (“ $(A, C)$  detectable”)



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“ $\Rightarrow$ ” needs some more work in order to show that  $P \geq 0, P \not> 0$  contradicts strict dissipativity for unbounded  $\mathbb{X}$

## Main results

# Main result without state constraints

Without state constraints:

**Theorem:** Consider the LQ problem with  $(A, B)$  stabilizable,  $Q = C^T C$  and state and control constraint sets  $\mathbb{X} = \mathbb{R}^n$  and  $\mathbb{U} \subseteq \mathbb{R}^m$ . Then the following properties are **equivalent**

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Moreover, if one of these properties holds, then the equilibria in (i) and (ii) **coincide**

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- (ii) The problem has the **near equilibrium turnpike property** at an equilibrium  $(x^e, u^e) \in \text{int}(\mathbb{X} \times \mathbb{U})$
- (iii) All unobservable eigenvalues  $\mu$  of  $A$  satisfy  $|\mu| \neq 1$

Moreover, if one of these properties holds, then the equilibria in (i) and (ii) **coincide**

# Main result with state constraints

With bounded state constraints:

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Moreover, if one of these properties holds, then the equilibria in (i) and (ii) **coincide**. If, in addition,  $(A, B)$  is stabilizable then the **turnpike property** holds.

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Is there an **intuitive explanation** for this fact?

## Example 1 reloaded

Cost function

$$\ell(x, u) = u^2$$

Dynamics

$$x^+ = 2x + u$$

Constraints

$$\mathbb{X} = [-2, 2], \mathbb{U} = [-3, 3]$$

## Example 1 reloaded

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Indeed, in this case **all optimal solutions grow exponentially**, because  $u \equiv 0$  is clearly the optimal control



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General principle for bounded constraints:

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## Outlook

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- What happens if  $x^e$  is at the **boundary of  $\mathbb{X}$** ?

# References

L. Grüne and R. Guglielmi, *Turnpike properties and strict dissipativity for discrete time linear quadratic optimal control problems*, submitted

L. Grüne, *Economic receding horizon control without terminal constraints*, *Automatica*, 49, 725–734, 2013

L. Grüne, M. Stieler, *Asymptotic stability and transient optimality of economic MPC without terminal conditions*, *Journal of Process Control*, 24 (Special Issue on Economic MPC), 1187–1196, 2014

L. Grüne, C.M. Kellett, S.R. Weller, *On a discounted notion of strict dissipativity*, *Proceedings of NOLCOS 2016*, *IFAC-PapersOnLine* 49, 247–252, 2016

L. Grüne, M.A. Müller, *On the relation between strict dissipativity and turnpike properties*, *Systems & Control Letters*, 90, 45–53, 2016