

# On economic model predictive control for time-varying systems

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algorithms, analysis and applications (Rome)

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# Outline

Introduction

Model predictive control

MPC convergence results

Application

Conclusion

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# Setting

- Consider a discrete-time, time-varying system

$$x(k+1) = f(k, x(k), u(k)), \quad x(0) = x_0$$

with  $x(k) \in X$ ,  $u(k) \in U$ .

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- Example: Simplified building model:

$$\textcolor{red}{x}(k+1) = \quad \textcolor{red}{x}(k) \quad + \quad \textcolor{blue}{u}(k) \quad + \quad \textcolor{green}{w}(k)$$

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- Example: Simplified building model:

$$x(k+1) = \underbrace{x(k)}_{\text{inside temperature}} + \underbrace{u(k)}_{\text{heating/cooling}} + \underbrace{w(k)}_{\text{outside temperature}}$$

- Goal: Keep temperature ( $x$ ) within a certain range  $\mathbb{X}(k)$ , using as little energy ( $u$ ) as possible.

# Setting

- Infinite horizon optimal control problem

$$\min_{u \in \mathbb{U}^\infty(k, x_0)} J_\infty(k, x_0, u) = \sum_{j=0}^{\infty} \ell(k+j, x_u(j, x_0), u(j)) \quad (1)$$

with stage cost  $\ell : \mathbb{N}_0 \times X \times U \rightarrow \mathbb{R}$ , and where  $\mathbb{U}^\infty(k, x_0)$  is the set of admissible control sequences.

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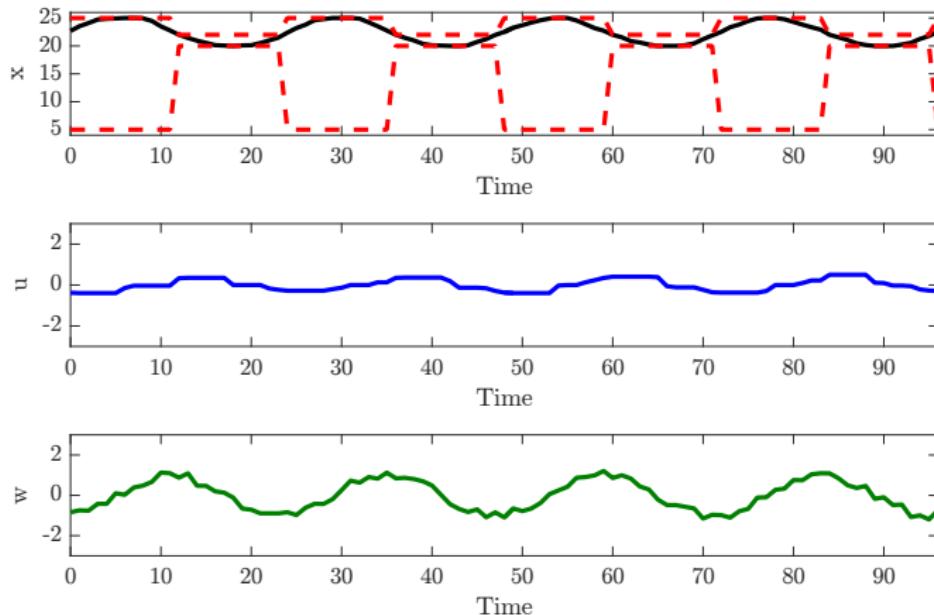
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- In the example:

Stage cost:

$$\ell(k, \textcolor{red}{x}, \textcolor{blue}{u}) = \textcolor{blue}{u}^2$$

# Example: optimal trajectory?



# What is optimal?

- $J_\infty(k, x_0, u)$  may not be finite for any  $u$ .  
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- Concept of overtaking optimality:  $(x_{u^*}, u^*)$  is called *overtaking optimal*<sup>a</sup> if

$$\liminf_{K \rightarrow \infty} \sum_{k=0}^{K-1} \ell(k, x_u(k, x_0), u(k)) - \ell(k, x_{u^*}(k, x_0), u^*(k)) \geq 0 \quad (2)$$

holds for all pairs  $(x_u, u)$ .

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<sup>a</sup> Joël Blot and Naïla Hayek. *Infinite-horizon optimal control in the discrete-time framework*. Springer, 2014.

## Optimal operation

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- Non-time-varying case:
  - Optimal equilibrium [Grüne '13]
  - Optimal periodic orbit [Grüne, Müller '16]
- In time-varying setting: more general optimal reference:  
system optimally operated at  $(x^*, u^*)$  if

$$\liminf_{K \rightarrow \infty} \sum_{k=0}^{K-1} \ell(k, x_u(k, x_0), u(k)) - \ell(k, x^*(k), u^*(k)) \geq 0 \quad (3)$$

(Initial value of  $x^* = x_{u^*}(\cdot; k, x)$  is free!)

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- $\rightsquigarrow$  Basic idea of MPC:  
Consider *finite horizon* optimization problems

$$\min_{u \in \mathbb{U}^N(k, x_0)} J_N(k, x_0, u) = \sum_{j=0}^{N-1} \ell(k + j, x_u(j, x_0), u(j)) \quad (4)$$

where  $\ell$  is the stage cost and  $N \in \mathbb{N}$  the *horizon length*.

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where  $\ell$  is the stage cost and  $N \in \mathbb{N}$  the *horizon length*.

- Solve (4) instead  $\rightsquigarrow u_N^*$ .
- Apply  $\mu_N(x_0) := u_N^*(0)$  as a feedback to the system.

## Why use *economic* MPC?

- Classical MPC: setpoint stabilization or tracking  
Stage cost penalizes distance to reference trajectory:

$$\ell(k, x, u) = \|x - x^*(k)\|^2 + \|u - u^*(k)\|^2$$

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  - Optimal reference is unknown
  - Tracking does not lead to optimal performance  $\rightsquigarrow$  [Grüne, Stieler '14]

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- Two problems:
  - Optimal reference is unknown
  - Tracking does not lead to optimal performance  $\rightsquigarrow$  [Grüne, Stieler '14]
- Solution: Use economic criterion in stage cost function of MPC.  
In the example  $\rightsquigarrow \ell(k, \textcolor{red}{x}, \textcolor{blue}{u}) = \textcolor{blue}{u}^2$

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- For now: optimality in a weak sense  
~~~ Look at relation between MPC closed loop cost

$$J_L^{\text{cl}}(k, x_0, \mu_N) = \sum_{j=0}^{L-1} \ell(k+j, x_{\mu_N}(j, x_0), \mu_N(x_{\mu_N}(j, x_0)))$$

and some optimal value “ $V_\infty(k, x_0) = \inf_{u \in \mathbb{U}^\infty(k, x_0)} J_\infty(k, x_0, u)$ ”.

## Some technicalities

- Modified cost function:

$$\hat{\ell}(k, x, u) = \ell(k, x, u) - \ell(k, x^*(k), u^*(k)) \quad (5)$$

$$\hat{J}_N(k, x_0, u) = \sum_{j=0}^{N-1} \hat{\ell}(k + j, x_u(j, x_0), u(j)) \quad (6)$$

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- Optimal value function

$$\hat{V}_N(k, x_0) := \inf_{u \in \mathbb{U}^N(k, x_0)} \hat{J}_N(k, x_0, u) \quad (7)$$

for  $N \in \mathbb{N}_0 \cup \{\infty\}$ .

# Main result

## Theorem 1

*Estimate for closed-loop cost:*

$$\hat{J}_L^{cl}(k, x, \mu_N) + \hat{V}_\infty(k+L, x_{\mu_N}(L, x)) \leq \hat{V}_\infty(k, x) + L\delta(N) \quad (8)$$

*for some function  $\delta \in \mathcal{L}$ .*

Requirements:

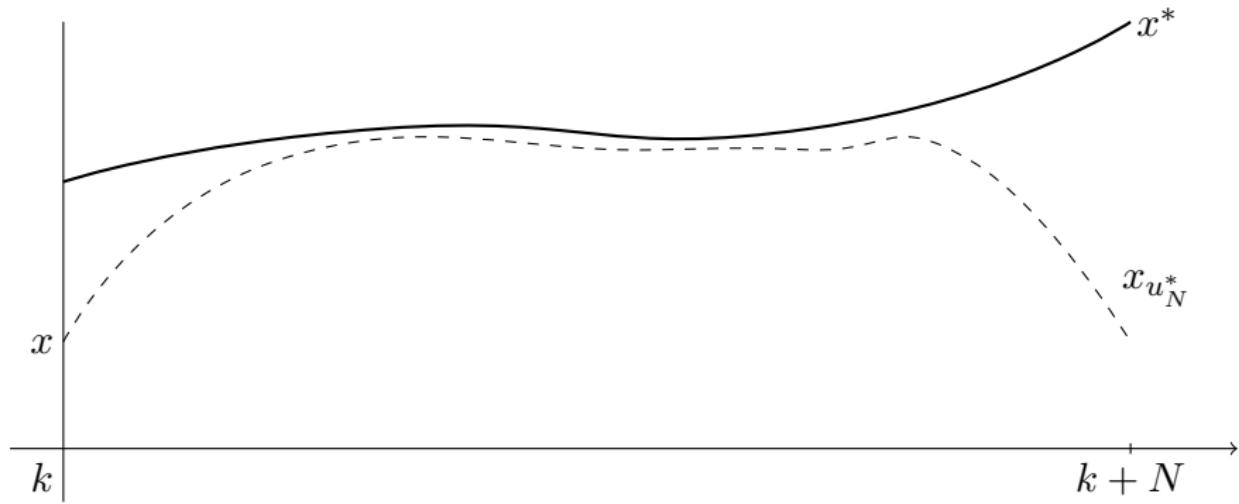
- Turnpike property
- Continuity of optimal value function

## Turnpike property (finite horizon)

The turnpike property holds if optimal trajectories satisfy

$$\begin{aligned} |(x_{u_N^*}(j, x), u_N^*(j))|_{(x^*(k+j), u^*(k+j))} &\leq \sigma(P) \\ \text{for all } j \in \{0, \dots, N\} \text{ with } j \notin \mathcal{Q}(k, x, P, N). \end{aligned} \tag{9}$$

for some  $\sigma \in \mathcal{L}$ .

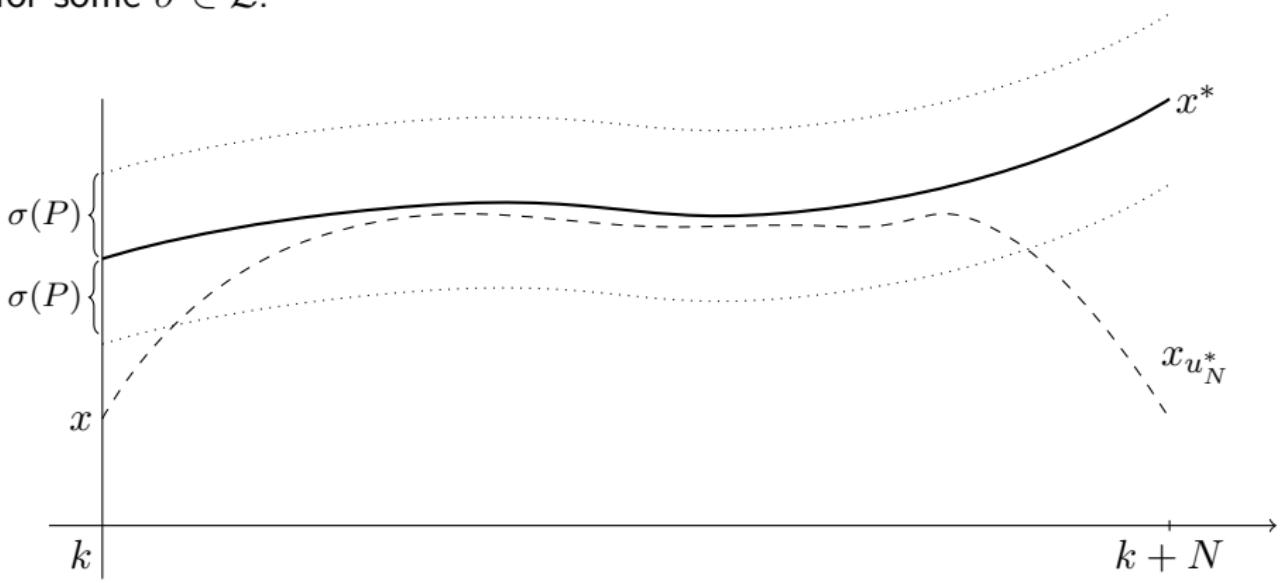


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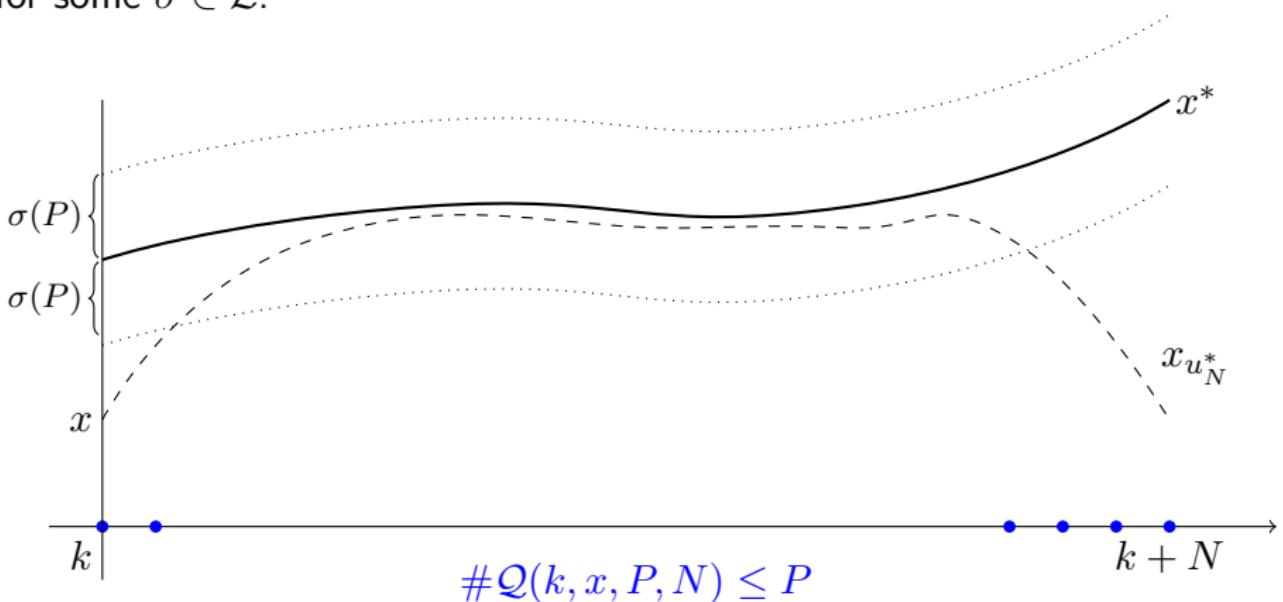


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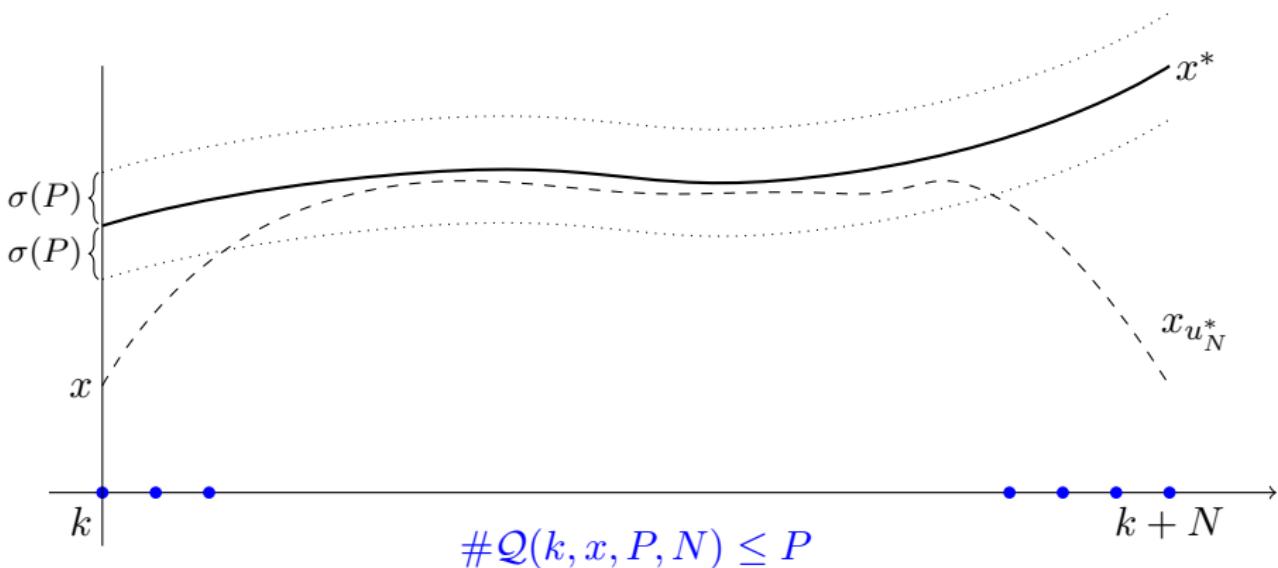


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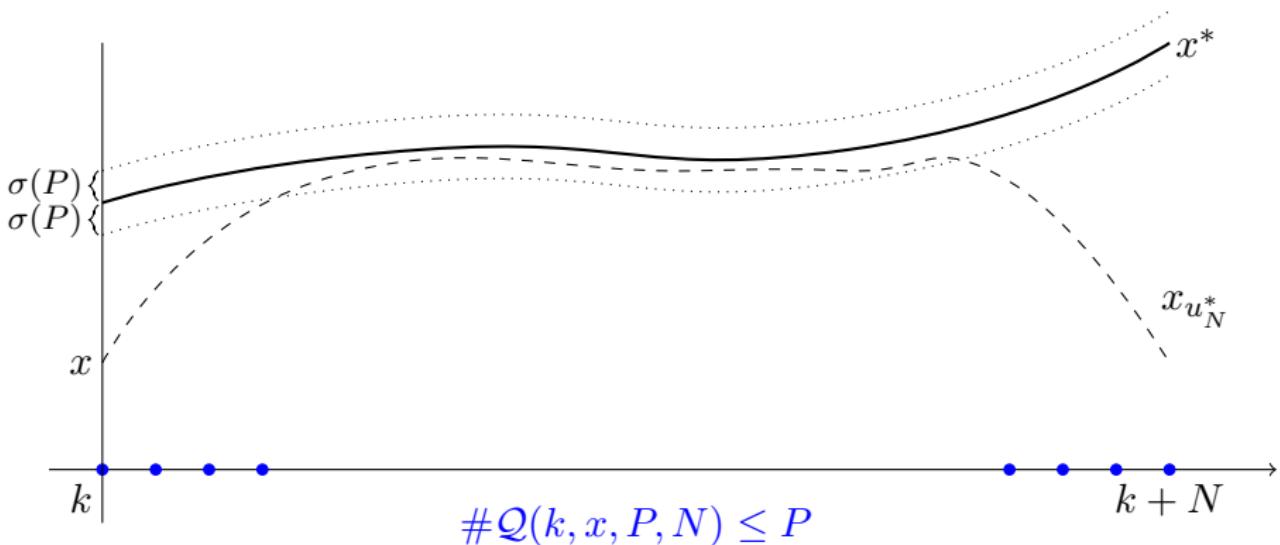


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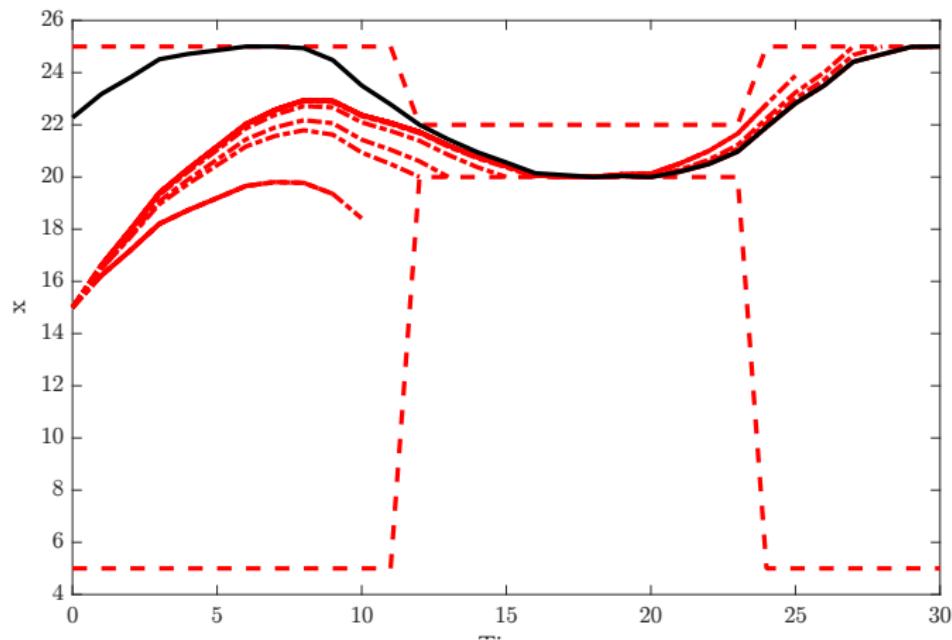
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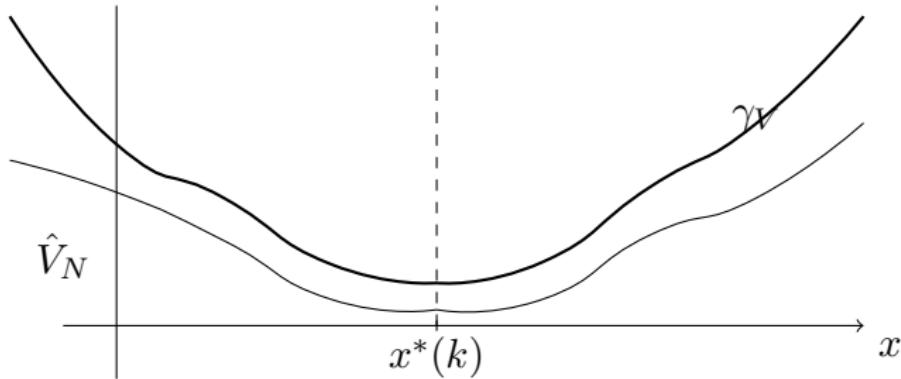
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# Example: turnpike property for different horizon length $N$



## Continuity of $\hat{V}_N$ and $\hat{V}_\infty$

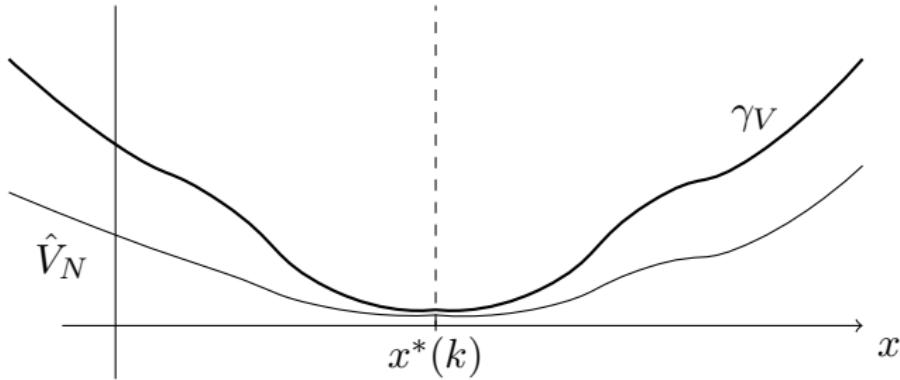


- We assume that the optimal value functions  $\hat{V}_N$  and  $\hat{V}_\infty$  are continuous in a neighbourhood of  $x^*$ :

$$|\hat{V}_N(k, x) - \hat{V}_N(k, x^*(k))| \leq \gamma_V(N, \|x - x^*(k)\|) \quad (10)$$

for all  $k \in \mathbb{N}$ ,  $N \in \mathbb{N} \cup \{\infty\}$ ,  $x \in \mathcal{B}_\varepsilon(x^*(k))$ .

## Continuity of $\hat{V}_N$ and $\hat{V}_\infty$

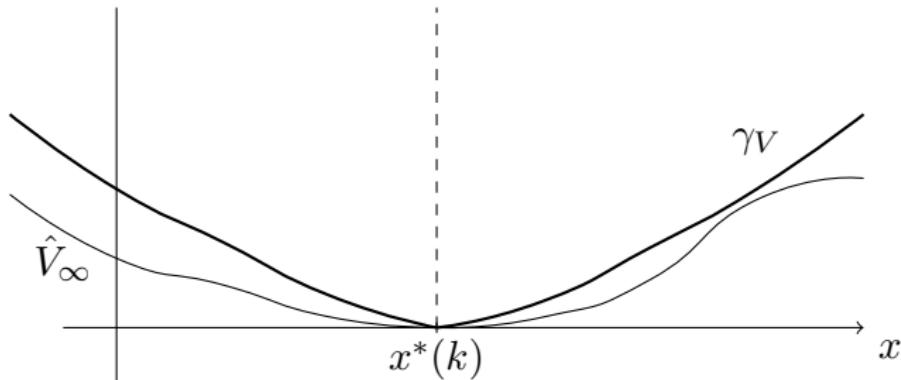


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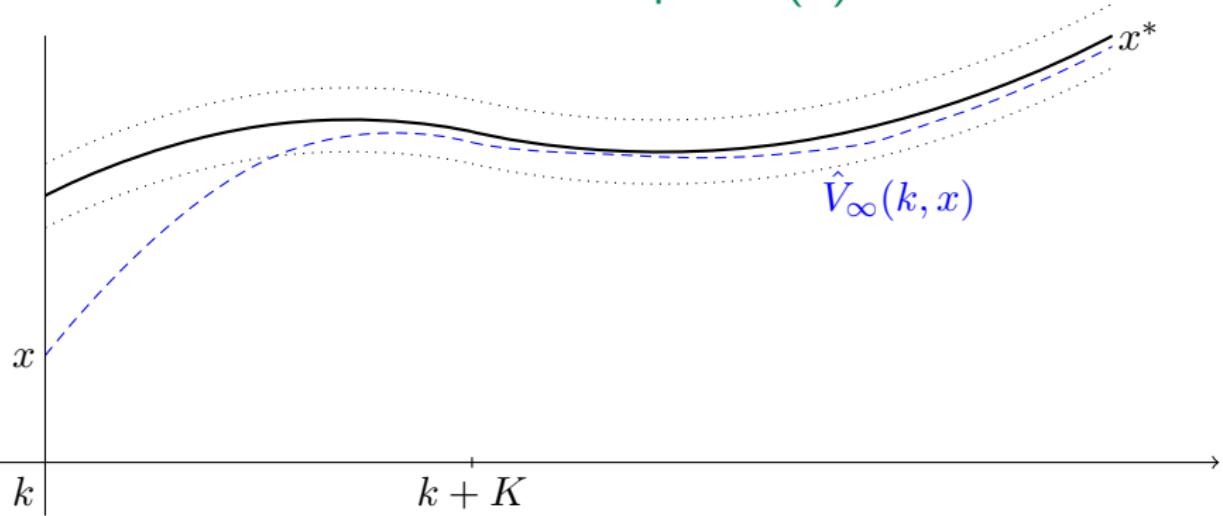
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for all  $k \in \mathbb{N}$ ,  $N \in \mathbb{N} \cup \{\infty\}$ ,  $x \in \mathcal{B}_\varepsilon(x^*(k))$ .

- Remark: for infinite horizon  $N = \infty$ :  $\hat{V}_\infty(k, x^*(k)) = 0$ .

# Outline of proof (1)

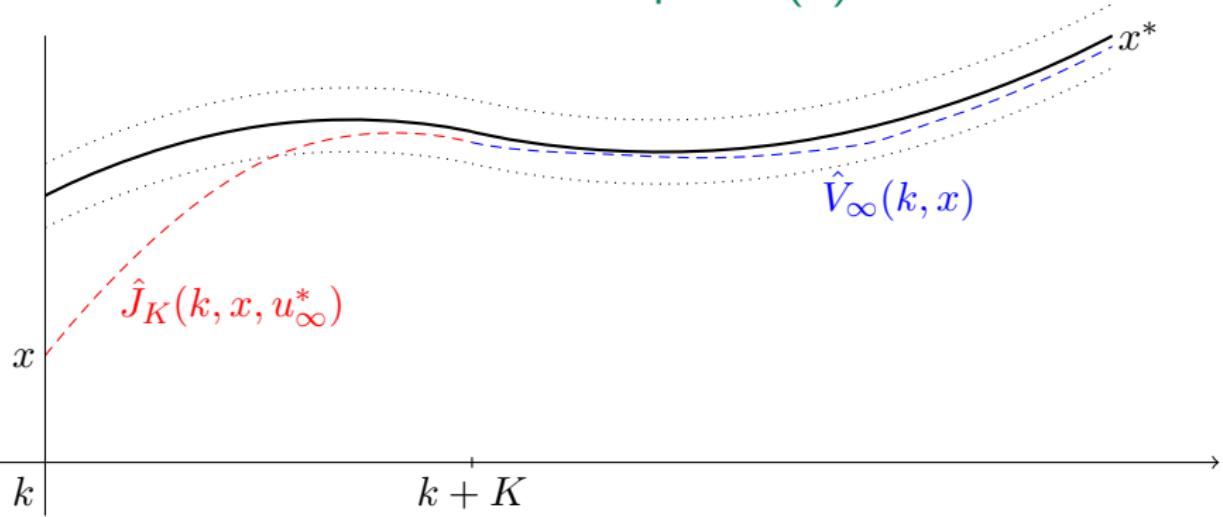


Show that

- $\hat{V}_\infty(k, x) = \hat{J}_K(k, x, u_\infty^*) + \text{"}\omega(N)\text{"}$
- $\hat{J}_K(k, x, u_\infty^*) = \hat{J}_K(k, x, u_N^*) + \text{"}\rho(N)\text{"}$

where  $\omega, \rho \in \mathcal{L}$ .

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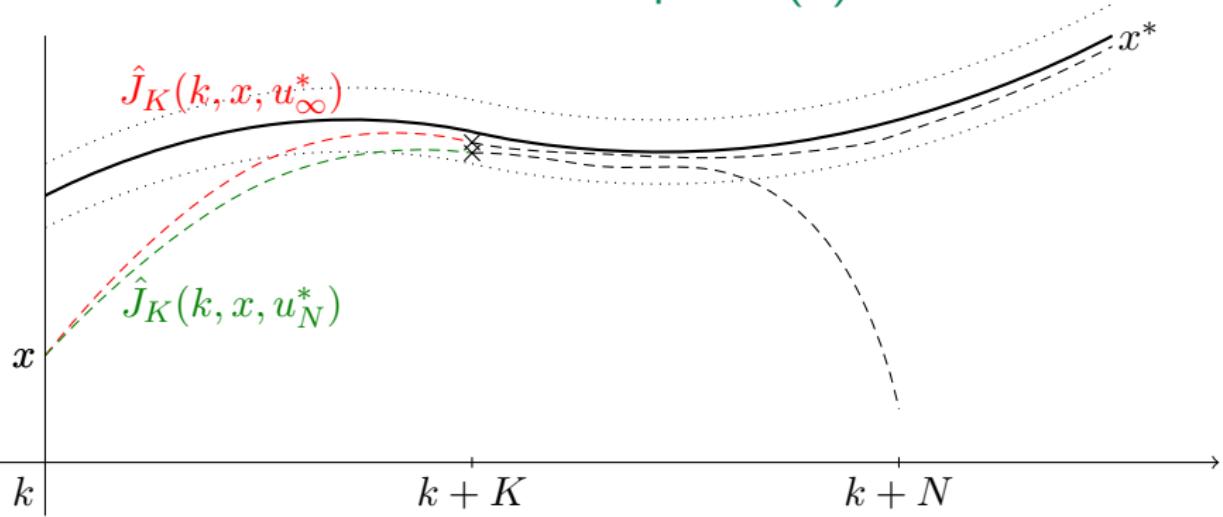


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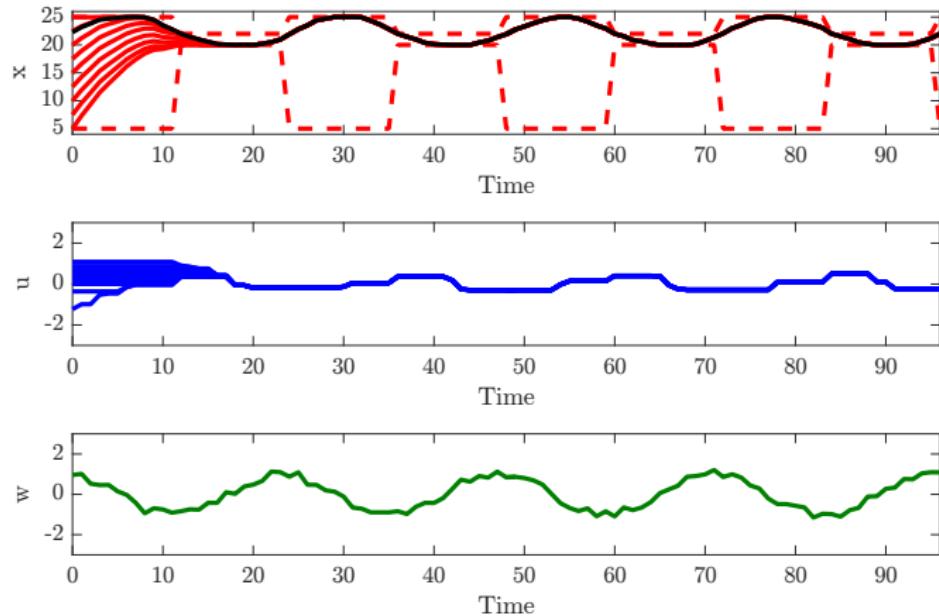
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- Summing up yields:

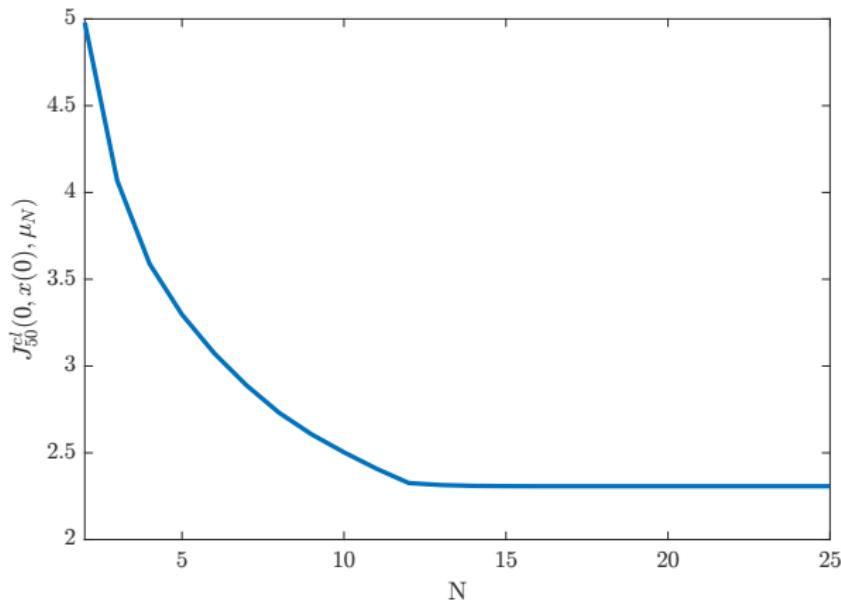
$$\hat{J}_L^{cl}(k, x, \mu_N) + \hat{V}_\infty(k+L, x_{\mu_N}(L, x)) \leq \hat{V}_\infty(k, x) + L\delta(N) \quad (13)$$

for  $\delta \in \mathcal{L}$ .

# Results: MPC closed loop for different initial values



## MPC closed loop cost for different horizon length



# How to verify requirements?

- Consider strictly dissipative systems:

There exists a storage function  $\lambda : \mathbb{R}_{\geq 0} \times \mathbb{R}^n \rightarrow \mathbb{R}_{\geq 0}$  such that

$$\lambda(k+1, f(k, x, u)) - \lambda(k, x) \leq \hat{\ell}(k, x, u) - \alpha(|(x, u)|_{(x^*(k), u^*(k))}) \quad (14)$$

for all  $k \in \mathbb{N}_0$ ,  $(x, u) \in \mathbb{X}(k) \times \mathbb{U}(k)$ .

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- Recent work (together with Marleen Stieler):

Verified dissipativity for example system:

- Stage cost needs to be strictly convex  $\rightsquigarrow$  modification necessary because  $\ell(k, x, u) = u^2$  is not convex w.r.t.  $x$
- Use  $\ell(k, x, u) = u^2 + \varepsilon(x - x_{ref})^2$ , for  $\varepsilon \ll 1$

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## Application: Energy efficient building operation

Boussinesq approximation<sup>1</sup> of airflow and temperatures inside a building:

$$\frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u} = -\frac{1}{\rho} \nabla p + \nu \Delta \mathbf{u} - \mathbf{g} \alpha (\mathcal{T} - \tilde{T}) \quad (15)$$

$$\nabla \cdot \mathbf{u} = 0 \quad (16)$$

$$\frac{\partial \mathcal{T}}{\partial t} + \mathbf{u} \cdot \nabla \mathcal{T} = \kappa \Delta \mathcal{T} \quad (17)$$

+ time-varying boundary conditions.

$\mathbf{u} : \Omega \times [0, \infty) \rightarrow \mathbb{R}^d$  is air velocity

where  $p : \Omega \times [0, \infty) \rightarrow \mathbb{R}$  is pressure

$\mathcal{T} : \Omega \times [0, \infty) \rightarrow \mathbb{R}$  is temperature

and parameters  $\rho, \nu, \mathbf{g}, \alpha, \tilde{T}, \kappa$ .

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<sup>1</sup>David J Tritton. *Physical fluid dynamics*. Springer Science & Business Media, 2012.

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# Summary, Remarks and open questions

Conclusion:

- Economic MPC solution approximates time-varying infinite horizon optimal solution
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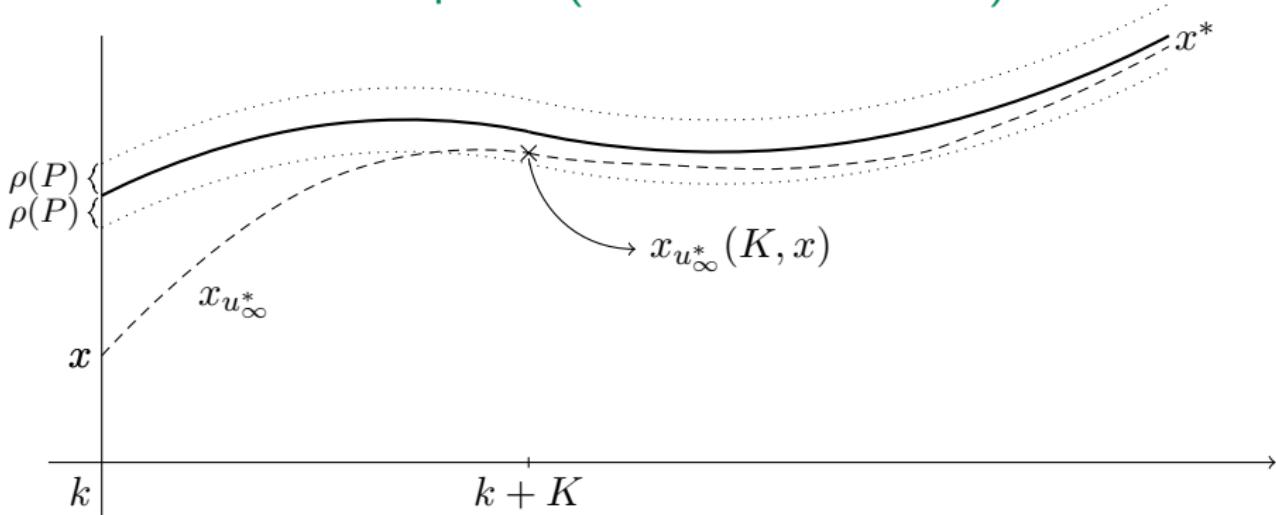
Outlook:

- More realistic PDE model of heat flow inside a building
- Can we still observe turnpike property?

## References

-  Joël Blot and Naïla Hayek. *Infinite-horizon optimal control in the discrete-time framework*. Springer, 2014.
-  David Gale. "On Optimal Development in a Multi-Sector Economy". In: *The Review of Economic Studies* 34.1 (1967), pp. 1–18. ISSN: 00346527, 1467937X. URL: <http://www.jstor.org/stable/2296567>.
-  Lars Grüne. "Approximation Properties of Receding Horizon Optimal Control". In: *Jahresbericht der Deutschen Mathematiker-Vereinigung* 118.1 (2016), pp. 3–37. ISSN: 1869-7135. DOI: 10.1365/s13291-016-0134-5. URL: <http://dx.doi.org/10.1365/s13291-016-0134-5>.
-  Lars Grüne and Marleen Stieler. "Asymptotic stability and transient optimality of economic MPC without terminal conditions". In: *Journal of Process Control* 24.8 (2014), pp. 1187–1196. URL: <http://www.sciencedirect.com/science/article/pii/S0959152414001401>.
-  Matthias A. Müller and Lars Grüne. "Economic model predictive control without terminal constraints for optimal periodic behavior". In: *Automatica* 70 (2016), pp. 128–139.
-  David J Tritton. *Physical fluid dynamics*. Springer Science & Business Media, 2012.

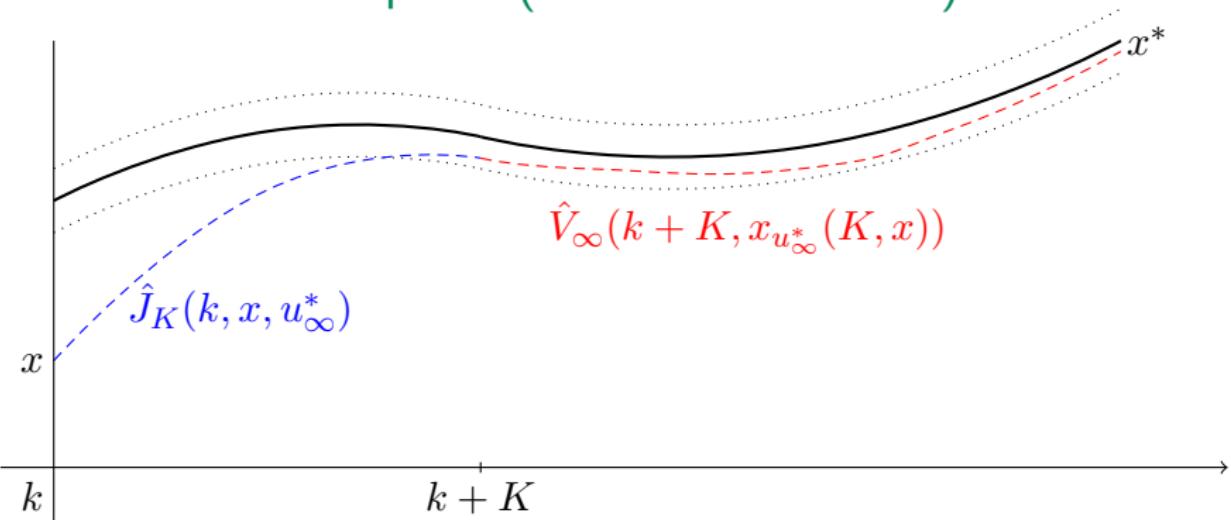
## Idea of proof (for infinite horizon)



Pick  $P \in \mathbb{N}$  such that  $\rho(P) \leq \varepsilon$  (possible because of turnpike property).  
 ↵ For  $K \notin \mathcal{Q}(k, x, P, \infty)$ :

$$|(x_{u_\infty^*}(K, x), u_\infty^*(K))|_{(x^*(k+K), u^*(k+K))} \leq \rho(P) \leq \varepsilon$$

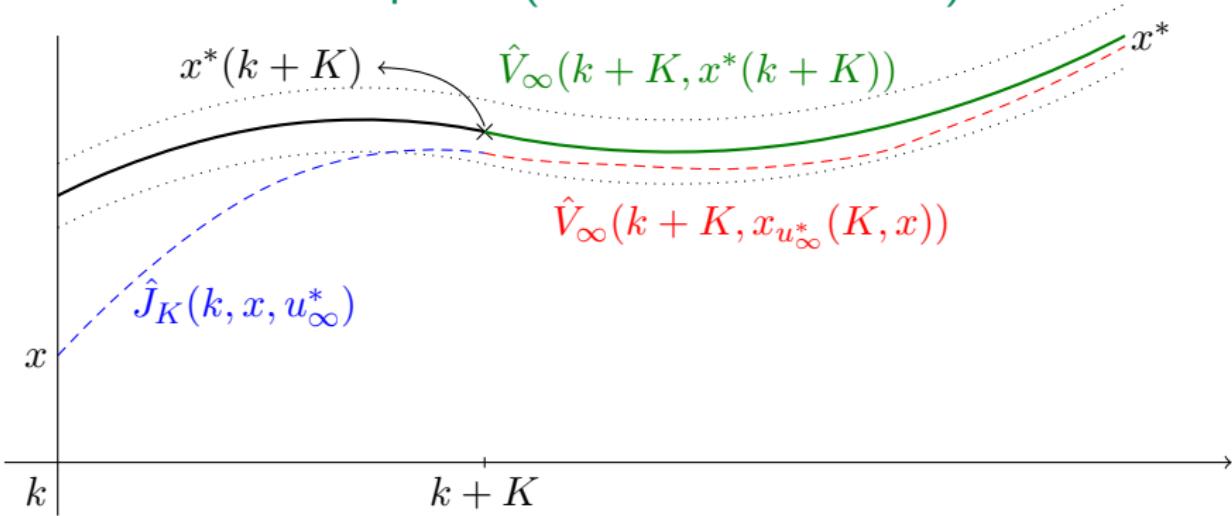
# Idea of proof (for infinite horizon)



Dynamic programming principle:

$$\hat{V}_\infty(k, x) = \hat{J}_K(k, x, u_\infty^*) + \hat{V}_\infty(k + K, x_{u_\infty^*}(K, x))$$

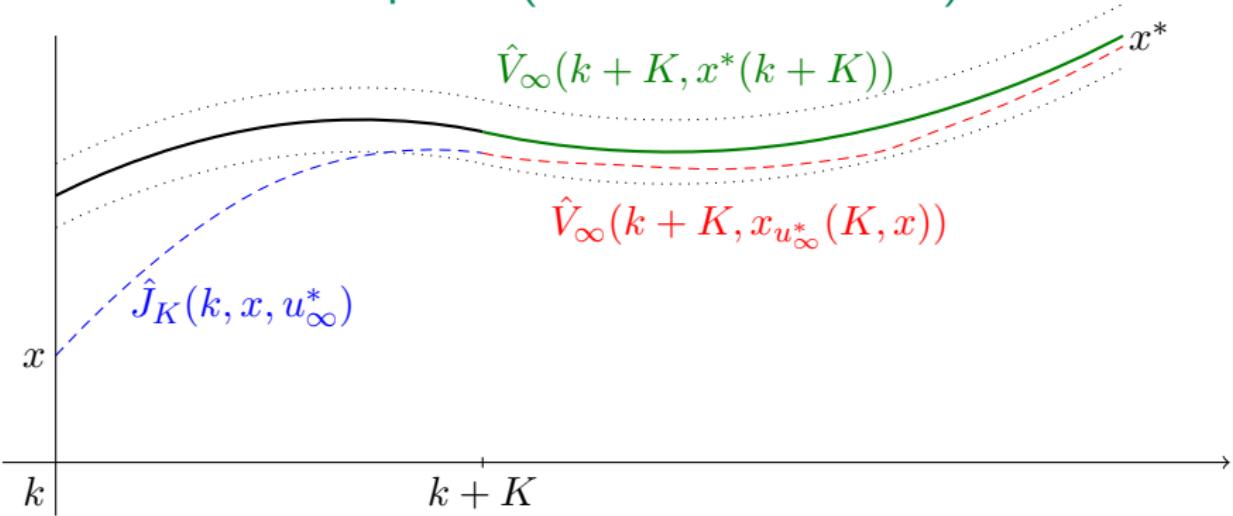
## Idea of proof (for infinite horizon)



$$\hat{V}_\infty(k + K, x^*(k + K)) = 0.$$

$$\begin{aligned} \Rightarrow \hat{V}_\infty(k, x) &= \hat{J}_K(k, x, u_\infty^*) \\ &\quad + \underbrace{\hat{V}_\infty(k + K, x_{u_\infty^*}(K, x)) - \hat{V}_\infty(k + K, x^*(k + K))}_{=: R_1(k, x, K)} \end{aligned}$$

## Idea of proof (for infinite horizon)



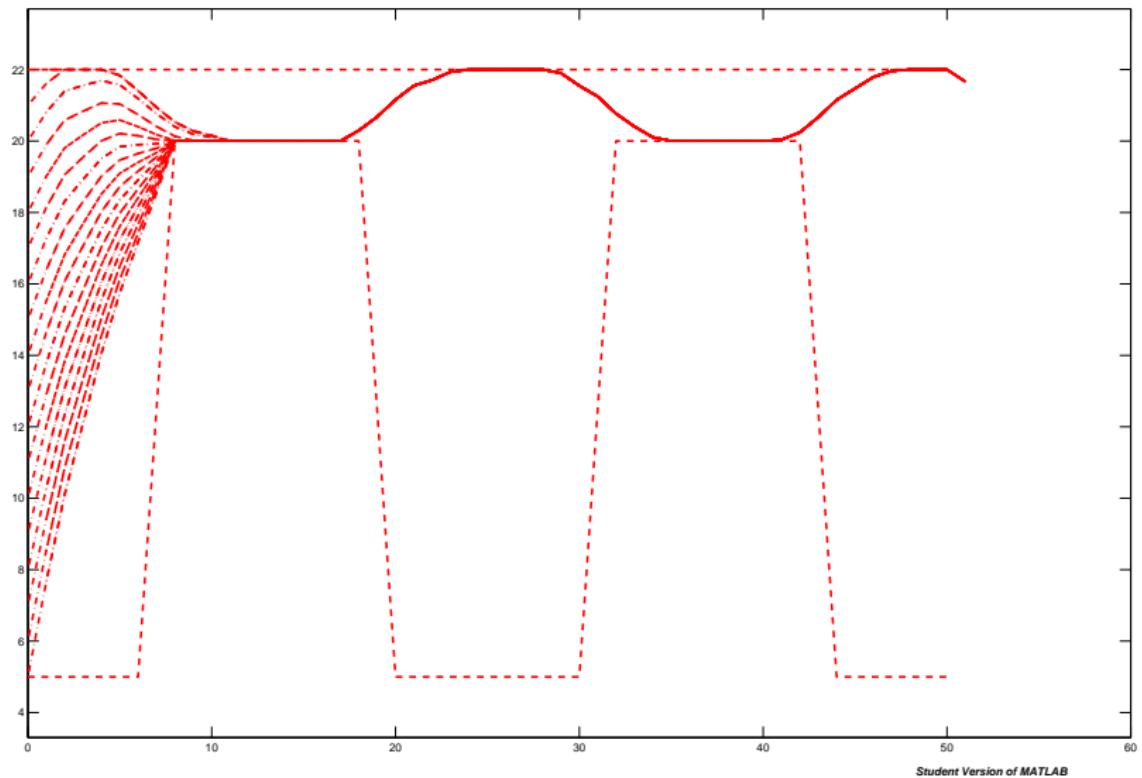
Continuity property:

$$|\hat{V}_\infty(k + K, x_{u_\infty^*}(K, x)) - \hat{V}_\infty(k + K, x^*(k + K))| \leq \omega_V(\rho(P))$$

$$\Rightarrow \hat{V}_\infty(k, x) = \hat{J}_K(k, x, u_\infty^*) + R_1(k, x, K)$$

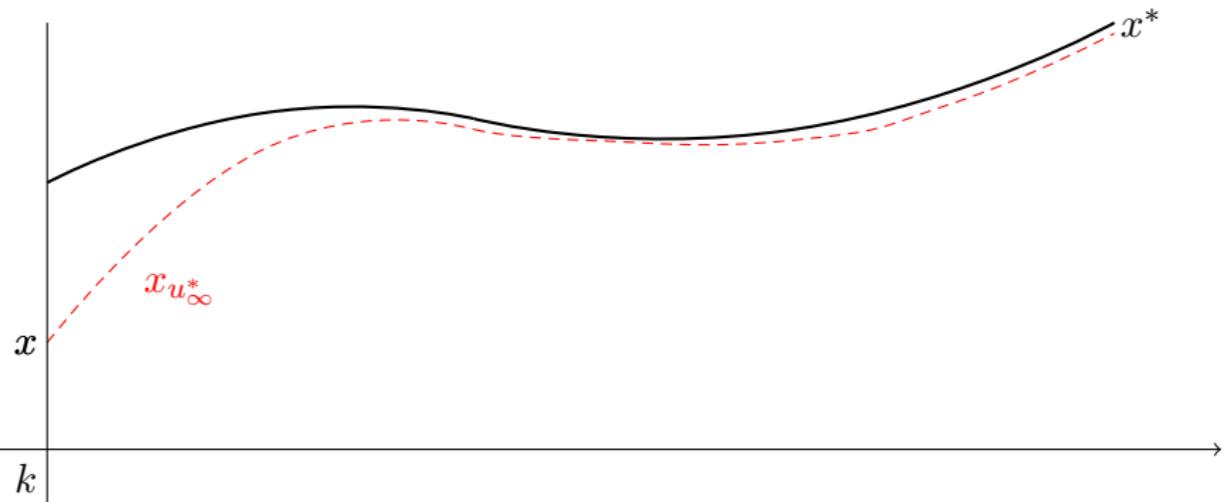
with  $|R_1(k, x, K)| \leq \omega_V(\sigma(P))$ .

# Example: turnpike property for different initial value $x_0$



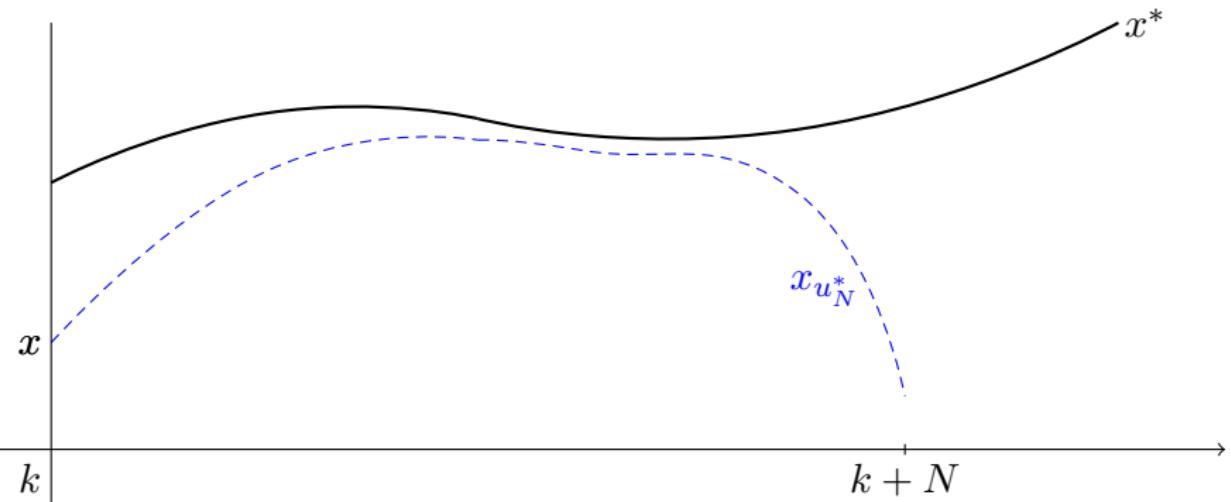
Student Version of MATLAB

## Idea of proof for Lemma 2



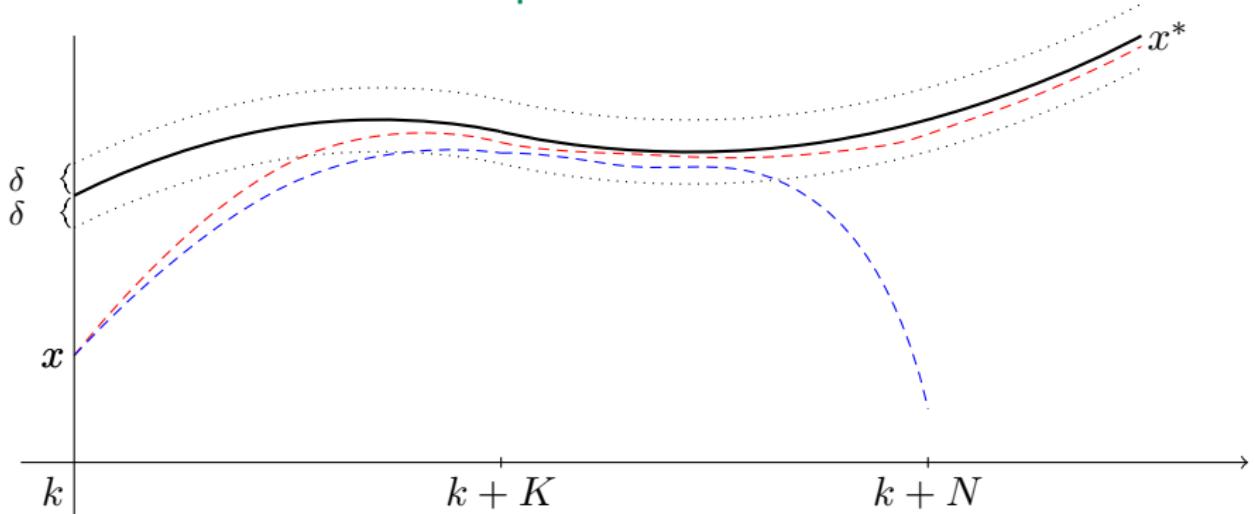
Open-loop on infinite horizon:  $x_{u_\infty^*}$

## Idea of proof for Lemma 2



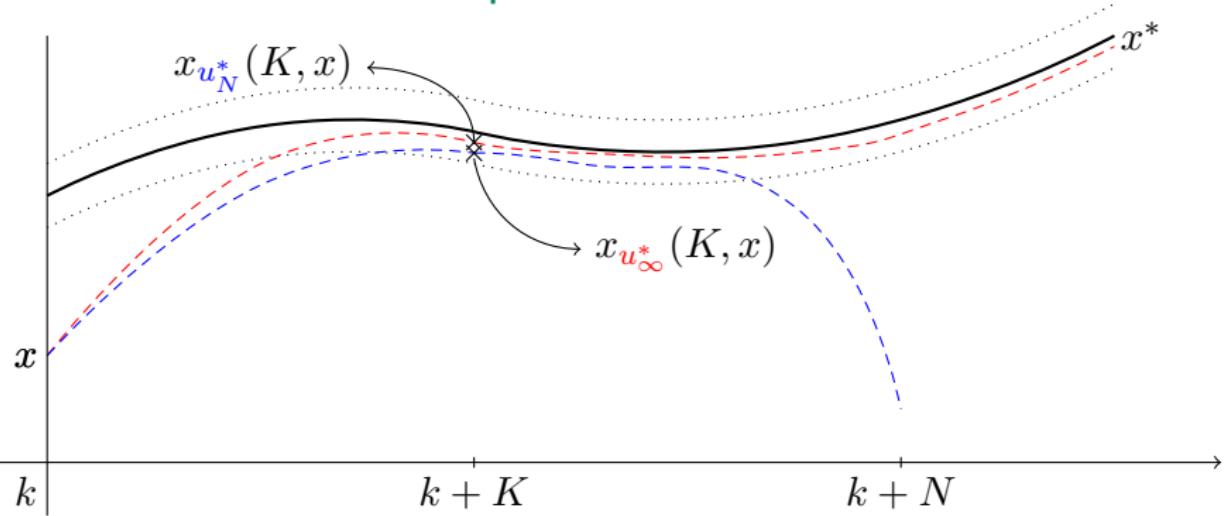
Open-loop on finite horizon:  $x_{u_N^*}$

## Idea of proof for Lemma 2



Choose  $P$  large enough, s.t.  $\delta := \max\{\sigma(P), \rho(P)\} < \varepsilon$ . Pick  $K \in \{0, \dots, N\} \setminus (\mathcal{Q}(k, x, P, N) \cup \mathcal{Q}(k, x, P, \infty))$ .

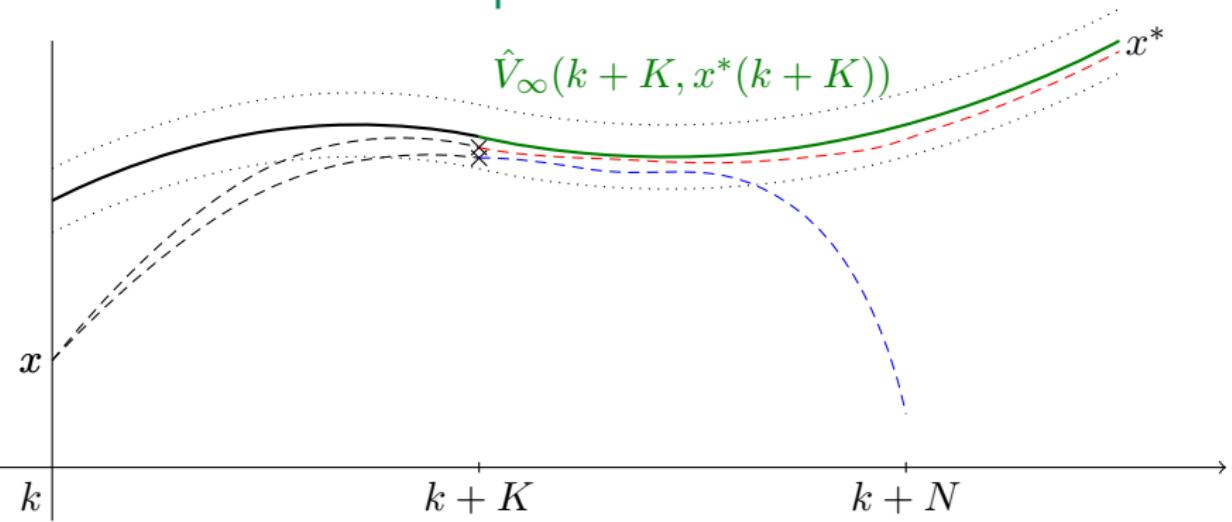
## Idea of proof for Lemma 2



Optimality of  $u_\infty^*$ :

$$\hat{J}_K(k, x, \mathbf{\hat{u}}_\infty^*) + \hat{V}_\infty(k+K, x_{u_\infty^*}(K, x)) \leq \hat{J}_K(k, x, \mathbf{\hat{u}}_N^*) + \hat{V}_\infty(k+K, x_{u_N^*}(K, x))$$

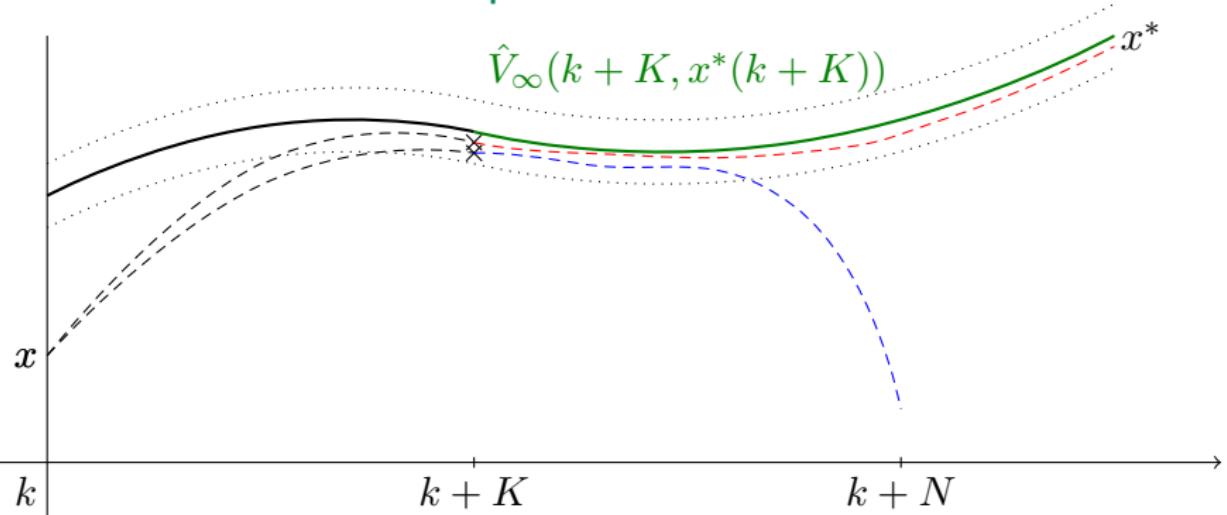
## Idea of proof for Lemma 2



Apply Lemma 1:

$$\hat{J}_K(k, x, u_\infty^*) + \underbrace{\hat{V}_\infty(k+K, x_{u_\infty^*}(K, x))}_{R_1(k, x, K) + \hat{V}_\infty(k+K, x^*(k+K))} \leq \hat{J}_K(k, x, u_N^*) + \underbrace{\hat{V}_\infty(k+K, x_{u_N^*}(K, x))}_{\tilde{R}_2(k, x, K, N) + \hat{V}_\infty(k+K, x^*(k+K))}$$

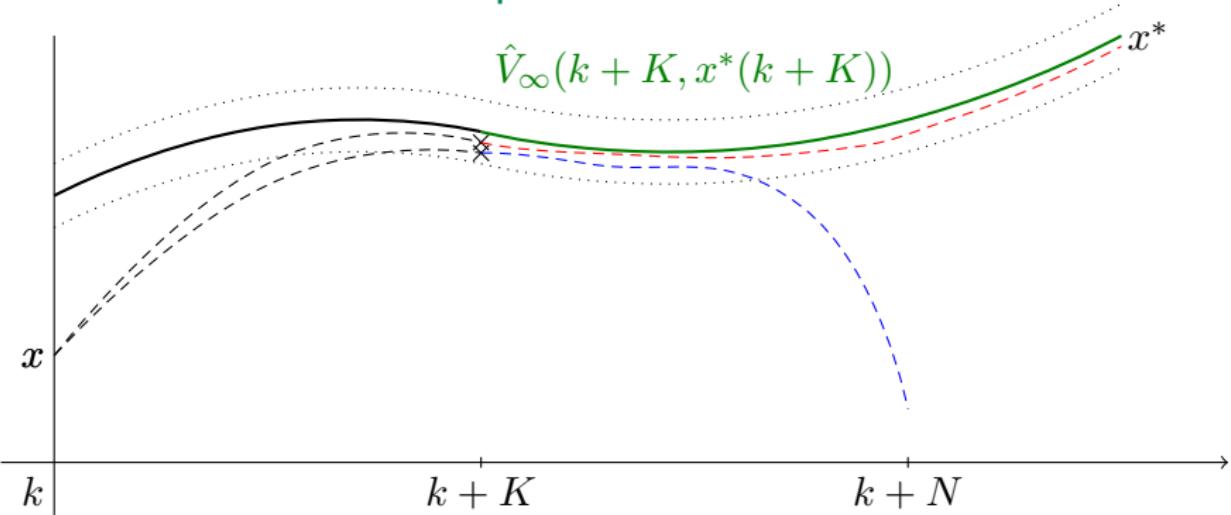
## Idea of proof for Lemma 2



Apply Lemma 1:

$$\begin{aligned}
 \hat{J}_K(k, x, u_\infty^*) + \underbrace{\hat{V}_\infty(k+K, x_{u_\infty^*}(K, x))}_{R_1(k, x, K) + \hat{V}_\infty(k+K, x^*(k+K))} &\leq \hat{J}_K(k, x, u_N^*) + \underbrace{\hat{V}_\infty(k+K, x_{u_N^*}(K, x))}_{\tilde{R}_2(k, x, K, N) + \hat{V}_\infty(k+K, x^*(k+K))} \\
 \Rightarrow \hat{J}_K(k, x, u_\infty^*) &\leq \hat{J}_K(k, x, u_N^*) - R_1(k, x, K) + \tilde{R}_2(k, x, K, N)
 \end{aligned}$$

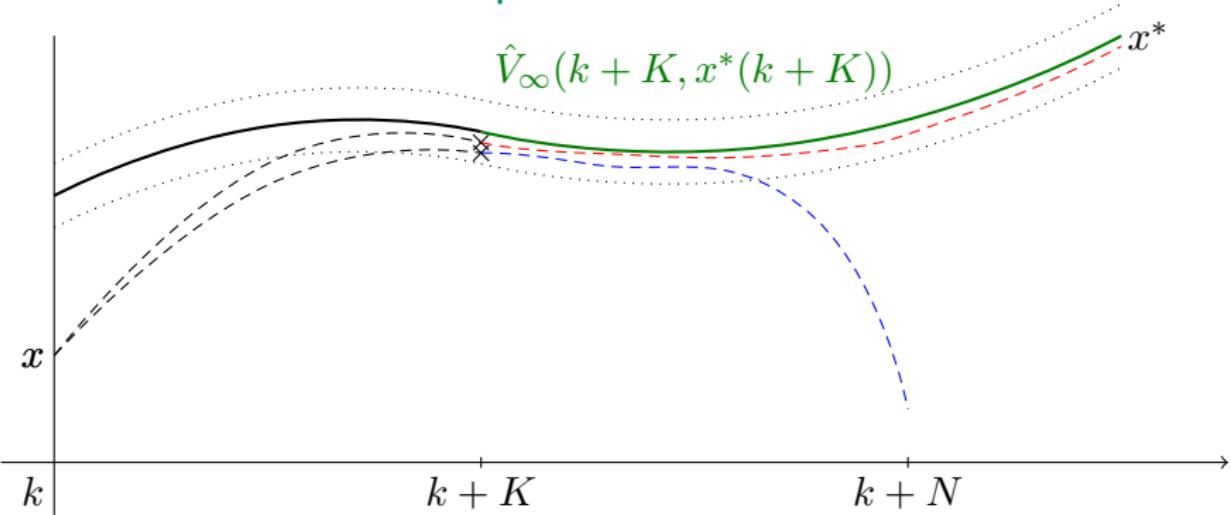
## Idea of proof for Lemma 2



Similarly: converse equality from Lemma 1 (2):

$$\hat{J}_K(k, x, u_N^*) \leq \hat{J}_K(k, x, u_\infty^*) - R_2(k, x, K, N) + \tilde{R}_1(k, x, K, N)$$

## Idea of proof for Lemma 2

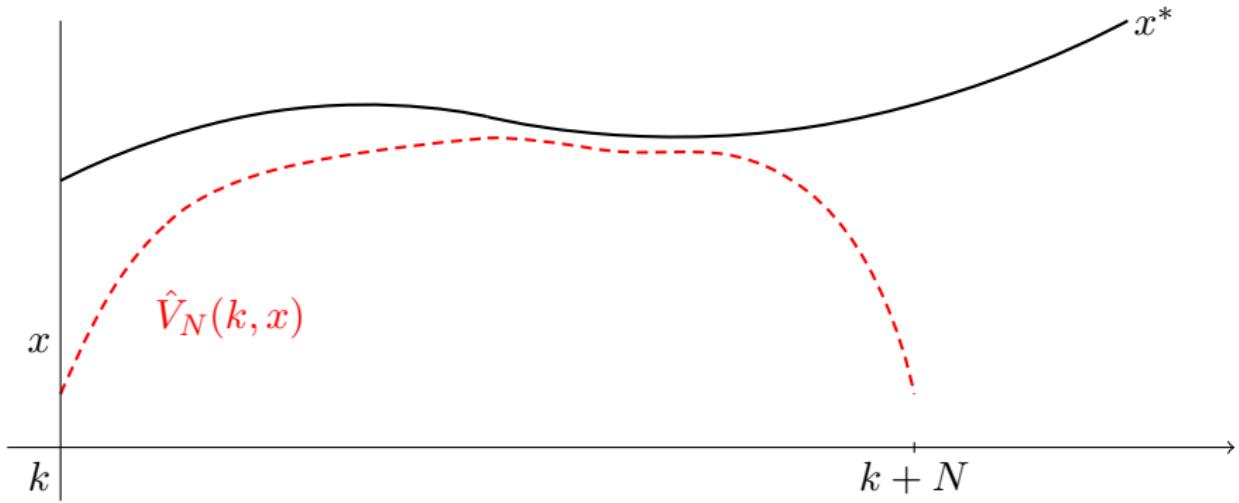


Similarly: converse equality from Lemma 1 (2):

$$\hat{J}_K(k, x, u_N^*) \leq \hat{J}_K(k, x, u_\infty^*) - R_2(k, x, K, N) + \tilde{R}_1(k, x, K, N)$$

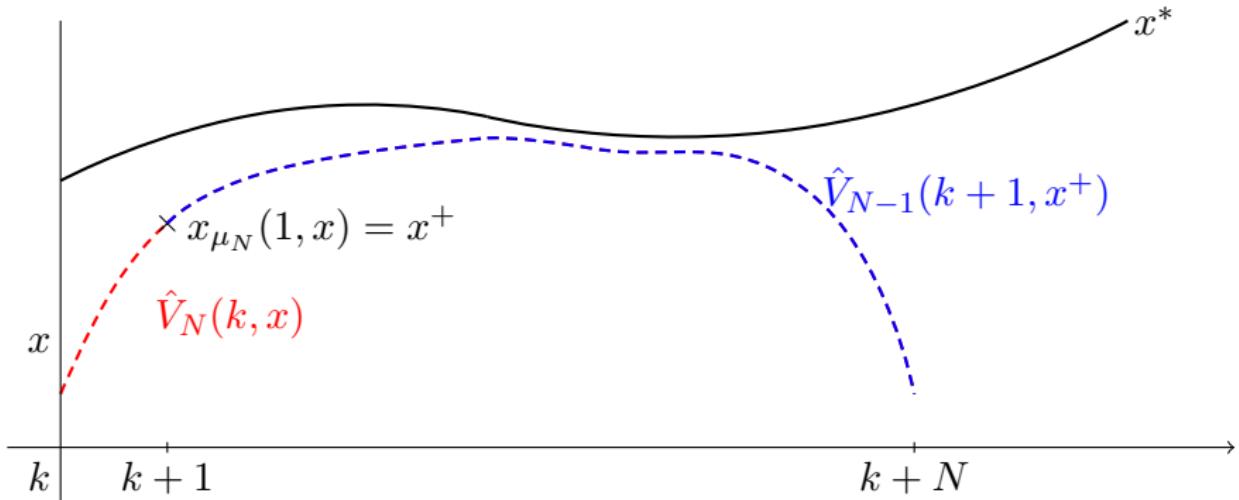
$$\Rightarrow \hat{J}_K(k, x, u_\infty^*) = \hat{J}_K(k, x, u_N^*) + R_3(k, x, K, N)$$

## Idea of proof for Theorem 1



- Optimal trajectory starting in  $x$  with horizon  $N$ .  $\rightsquigarrow \hat{V}_N(k, x)$

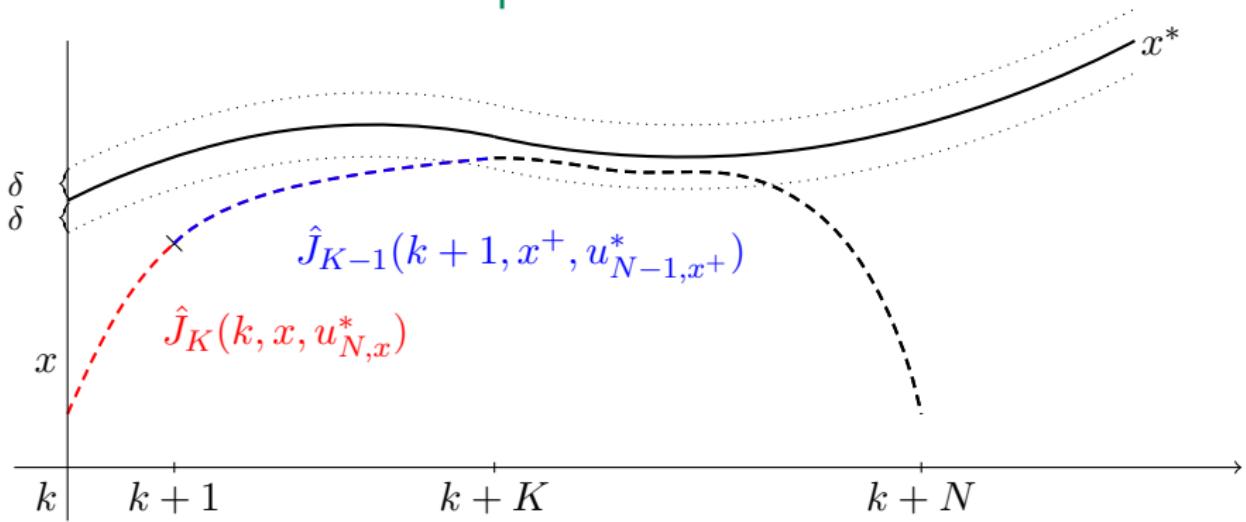
## Idea of proof for Theorem 1



- Optimal trajectory starting in  $x^+$  with horizon  $N - 1$ .  
 $\rightsquigarrow \hat{V}_{N-1}(k+1, x^+)$
- Stage cost along the MPC closed loop solution:

$$\hat{\ell}(k, x, \mu_N(x)) = \hat{V}_N(k, x) - \hat{V}_{N-1}(k+1, x^+)$$

## Idea of proof for Theorem 1



- Optimal trajectory starting in  $x^+$  with horizon  $N - 1$ .  
 $\rightsquigarrow \hat{V}_{N-1}(k+1, x^+)$
- Stage cost along the MPC closed loop solution:

$$\begin{aligned}\hat{\ell}(k, x, \mu_N(x)) &= \hat{V}_N(k, x) - \hat{V}_{N-1}(k+1, x^+) \\ &= \hat{J}_K(k, x, u_{N,x}^*) - \hat{J}_{K-1}(k+1, x^+, u_{N-1,x^+}^*)\end{aligned}$$

## Idea of proof for Theorem 1

- Stage cost along the MPC closed loop solution:

$$\begin{aligned}\hat{\ell}(k, x, \mu_N(x)) &= \hat{J}_K(k, x, \mathbf{u}_{N,x}^*) - \hat{J}_{K-1}(k+1, x^+, \mathbf{u}_{N-1,x^+}^*) \\ &\stackrel{\text{Lemma } 2}{=} \hat{J}_K(k, x, \mathbf{u}_{\infty,x}^*) - \hat{J}_{K-1}(k+1, x^+, \mathbf{u}_{\infty,x^+}^*) + R_3(\dots) - R_3(\dots)\end{aligned}$$

## Idea of proof for Theorem 1

- Stage cost along the MPC closed loop solution:

$$\begin{aligned}
 \hat{\ell}(k, x, \mu_N(x)) &= \hat{J}_K(k, x, u_{N,x}^*) - \hat{J}_{K-1}(k+1, x^+, u_{N-1,x^+}^*) \\
 &\stackrel{\text{Lemma } 2}{=} \hat{J}_K(k, x, u_{\infty,x}^*) - \hat{J}_{K-1}(k+1, x^+, u_{\infty,x^+}^*) + R_3(\dots) - R_3(\dots) \\
 &\stackrel{\text{Lemma } 1}{=} \hat{V}_{\infty}(k, x) - \hat{V}_{\infty}(k+1, x^+) \\
 &\quad + \underbrace{R_1(\dots) - R_1(\dots) + R_3(\dots) - R_3(\dots)}_{=: R_4(k, x, K, N)}
 \end{aligned}$$

## Idea of proof for Theorem 1

- Stage cost along the MPC closed loop solution:

$$\begin{aligned}
 \hat{\ell}(k, x, \mu_N(x)) &= \hat{J}_K(k, x, u_{N,x}^*) - \hat{J}_{K-1}(k+1, x^+, u_{N-1,x^+}^*) \\
 &\stackrel{\text{Lemma } 2}{=} \hat{J}_K(k, x, u_{\infty,x}^*) - \hat{J}_{K-1}(k+1, x^+, u_{\infty,x^+}^*) + R_3(\dots) - R_3(\dots) \\
 &\stackrel{\text{Lemma } 1}{=} \hat{V}_\infty(k, x) - \hat{V}_\infty(k+1, x^+) \\
 &\quad + \underbrace{R_1(\dots) - R_1(\dots) + R_3(\dots) - R_3(\dots)}_{=: R_4(k, x, K, N)}
 \end{aligned}$$

- One can show that  $|R_4(k, x, K, N)| \leq \delta(N)$  for some  $\delta \in \mathcal{L}$ .

## Idea of proof for Theorem 1

- Stage cost along the MPC closed loop solution:

$$\begin{aligned}
 \hat{\ell}(k, x, \mu_N(x)) &= \hat{J}_K(k, x, u_{N,x}^*) - \hat{J}_{K-1}(k+1, x^+, u_{N-1,x^+}^*) \\
 &\stackrel{\text{Lemma } 2}{=} \hat{J}_K(k, x, u_{\infty,x}^*) - \hat{J}_{K-1}(k+1, x^+, u_{\infty,x^+}^*) + R_3(\dots) - R_3(\dots) \\
 &\stackrel{\text{Lemma } 1}{=} \hat{V}_\infty(k, x) - \hat{V}_\infty(k+1, x^+) \\
 &\quad + \underbrace{R_1(\dots) - R_1(\dots) + R_3(\dots) - R_3(\dots)}_{=:R_4(k, x, K, N)}
 \end{aligned}$$

- One can show that  $|R_4(k, x, K, N)| \leq \delta(N)$  for some  $\delta \in \mathcal{L}$ .
- Sum  $\hat{\ell}(k+j, x, \mu_N(x))$  along the closed loop:

$$\begin{aligned}
 \hat{J}_L^{\text{cl}}(k, x, \mu_N) &= \sum_{j=0}^{L-1} \hat{\ell}(k+j, x_{\mu_N}(j, x), \mu_N(x_{\mu_N}(j, x))) \\
 &= \sum_{j=0}^{L-1} \hat{V}_\infty(j, x) - \hat{V}_\infty(j+1, x^+) + R_4(k, x, K, N) \\
 &\leq \hat{V}_\infty(k, x) - \hat{V}_\infty(k+L, x_{\mu_N}(L, x)) + L\delta(N)
 \end{aligned}$$