

# Economic model predictive control for time-varying systems

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# Outline

Introduction

Model predictive control

MPC convergence results

(Prospective) 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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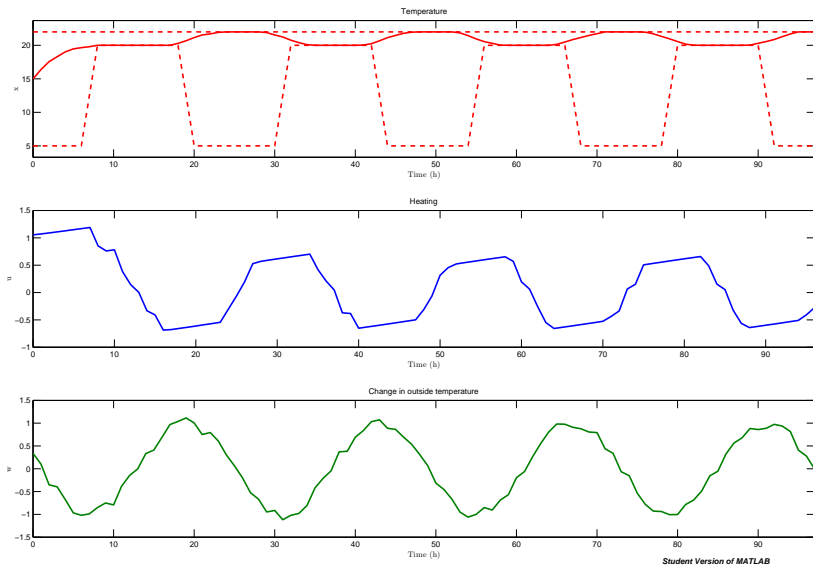
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- Example: Simplified building model:

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

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

## Example: optimal trajectory?



# 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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- In the example:

Stage cost:

$$\ell(k, \mathbf{x}, u) = u^2$$

## What is optimal?

- $J_\infty(k, x_0, u)$  may not be finite for any  $u$ .  
 $\rightsquigarrow J_\infty(k, x_0, u^*) \leq J_\infty(k, x_0, u)$  is not a meaningful definition of optimality.

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- $(x^*, u^*)$  is called *overtaking optimal*<sup>a</sup> if

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

holds for all pairs  $(x_u(\cdot, x_0), u(\cdot))$ .

Note: We do not demand that  $x^*(0) = x_0$ .

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

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# Model predictive control

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Consider *finite horizon* optimization problems

$$\min_{u \in \mathbb{U}^N(k, x_0)} J_N(k, x_0, u) = \sum_{j=0}^{N-1} \ell(k+j, x_u(j, x_0), u(j)) \quad (3)$$

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

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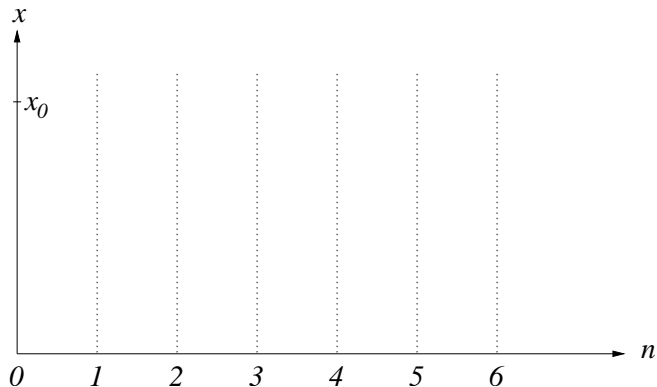
where  $\ell$  is the stage cost and  $N \in \mathbb{N}$  the *horizon length*.

- Solve (3) instead  $\rightsquigarrow u_N^*$ .
- Apply  $\mu_N(x_0) := u_N^*(0)$  as a feedback to the system.



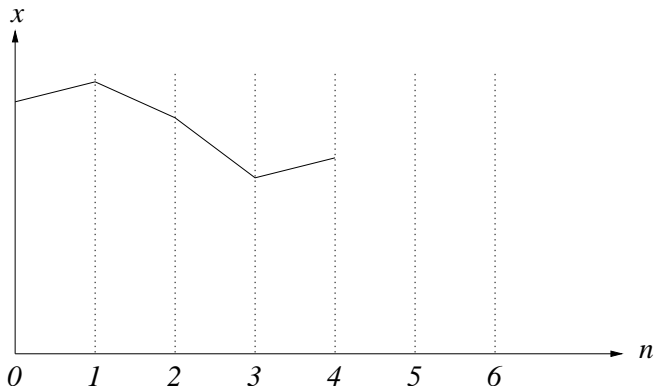
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The MPC principle for optimization horizon  $N = 4$ :



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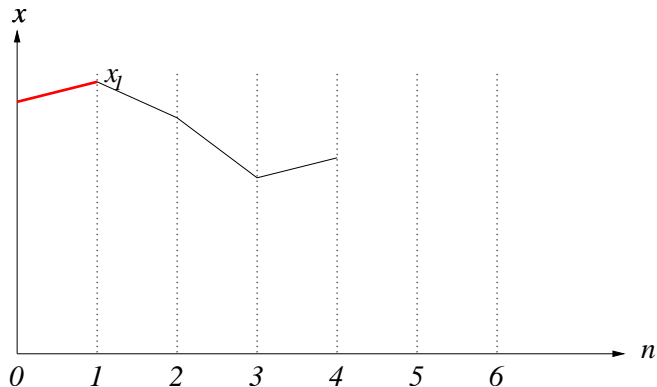
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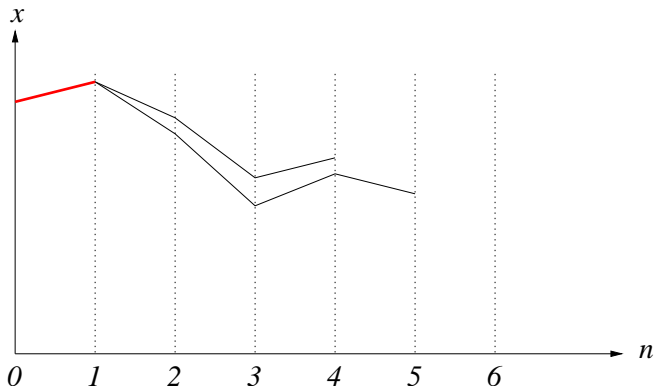


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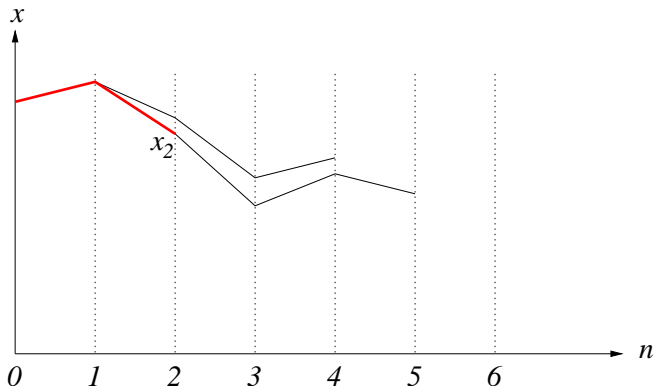


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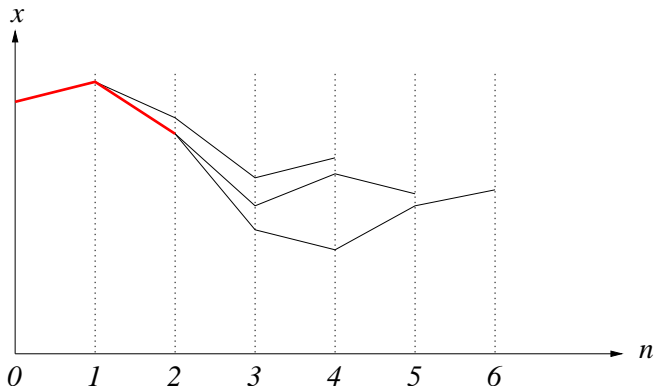


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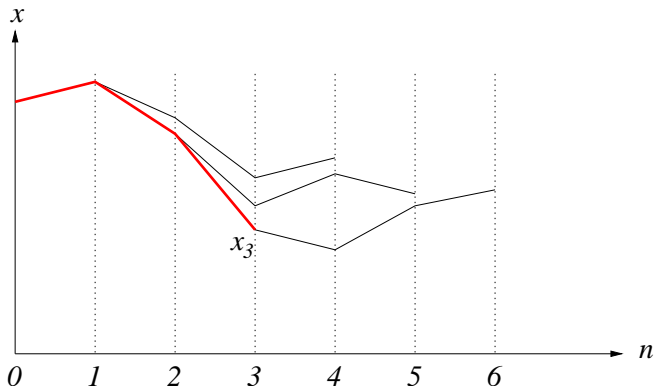


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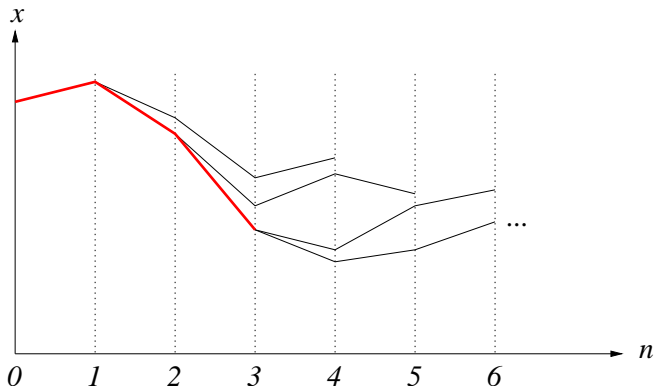


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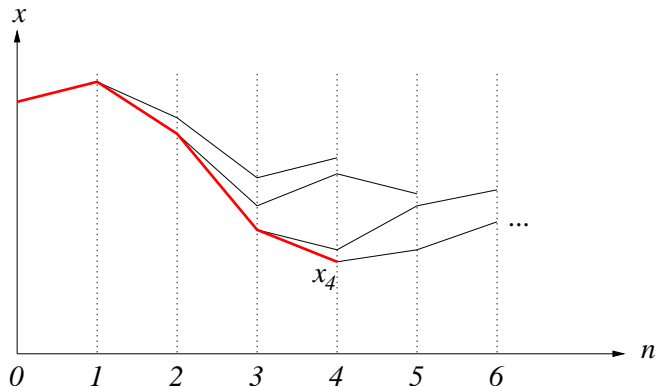
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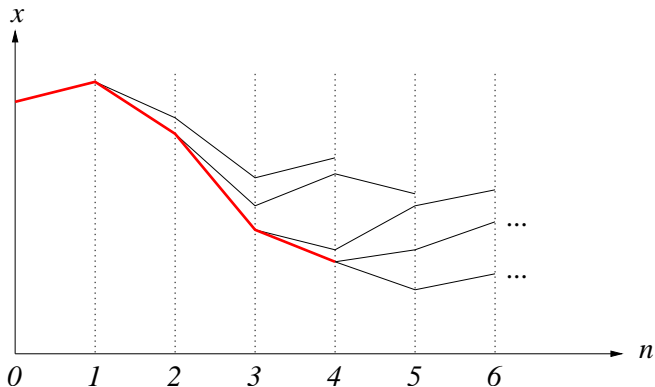


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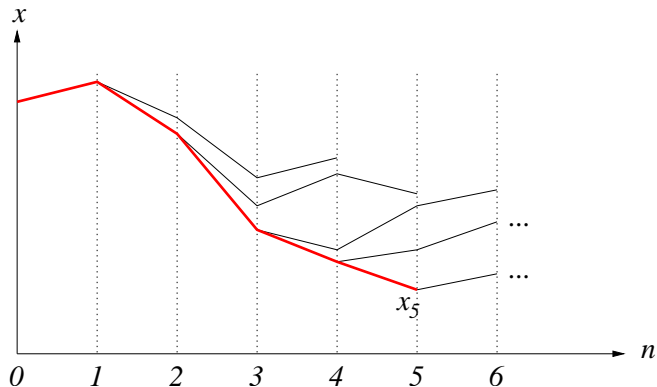


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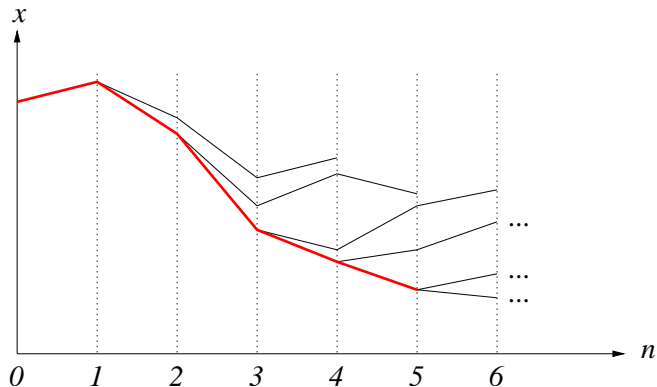


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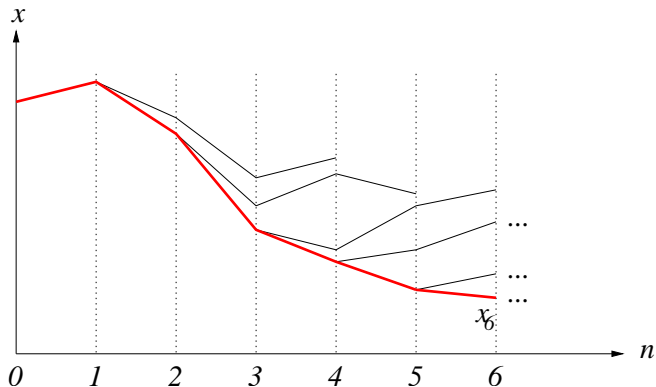


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## Why use economic MPC?

- Most commonly used type of MPC: setpoint stabilization or tracking  
Stage cost penalizes distance to (good) reference trajectory:

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

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Simplest case:  $(x^*(k), u^*(k)) = (x^e, u^e), k \in \mathbb{N}$  optimal equilibrium  
Also possible: periodic orbit as in Matthias' talk.

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- Two problems:
  - No way to compute optimal reference a priori.
  - Tracking does not lead to optimal performance  $\rightsquigarrow [3]^a$
- Solution: Use economic criterion in stage cost function of MPC.  
In the example  $\rightsquigarrow \ell(k, x, u) = u^2$

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- Prove, that the MPC closed loop  $x_{\mu_N}(\cdot, x_0)$  approximates the *overtaking optimal* trajectory  $x^*$
- 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 “ $V_\infty(k, x_0) = \inf_{u \in \mathbb{U}^\infty(k, x_0)} J_\infty(k, x_0, u)$ ”.

## Modified cost function

To avoid the possibly infinite cost we introduce

$$\hat{\ell}(k, x, u) = \ell(k, x, u) - \ell(k, x^*(k), u^*(k)) \quad (4)$$

and the corresponding cost

$$\hat{J}_N(k, x_0, u) = \sum_{j=0}^{N-1} \hat{\ell}(k+j, x_u(j, x_0), u(j)) \quad (5)$$

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

## Optimal value function

- Define the optimal value function

$$\hat{V}_N(k, x_0) := \inf_{u \in \mathbb{U}^N(k, x_0)} \hat{J}_N(k, x_0, u) \quad (6)$$

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- For  $N = \infty$ :  $\hat{V}_\infty(k, x^*(k)) = 0$ .
- Key assumptions:
  - Turnpike property
  - Continuity of optimal value function



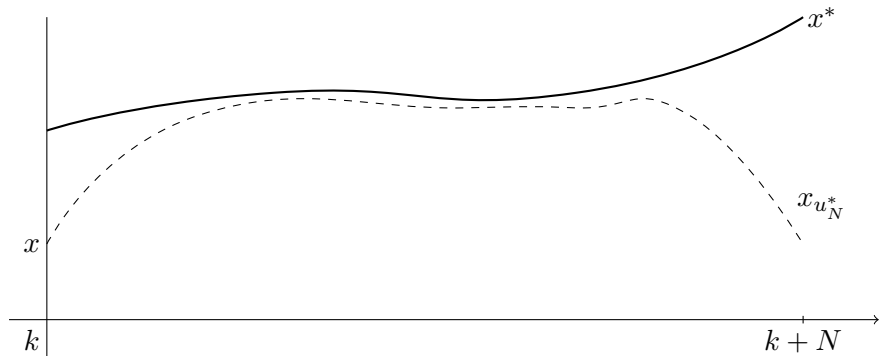
## Turnpike property (finite horizon)

The turnpike property holds if optimal trajectories satisfy

$$|(x_{u_N^*}(j, x), u_N^*(j))|_{(x^*(k+j), u^*(k+j))} \leq \sigma(P) \quad (7)$$

for all  $j \in \{0, \dots, N\}$  with  $j \notin \mathcal{Q}(k, x, P, N)$ .

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



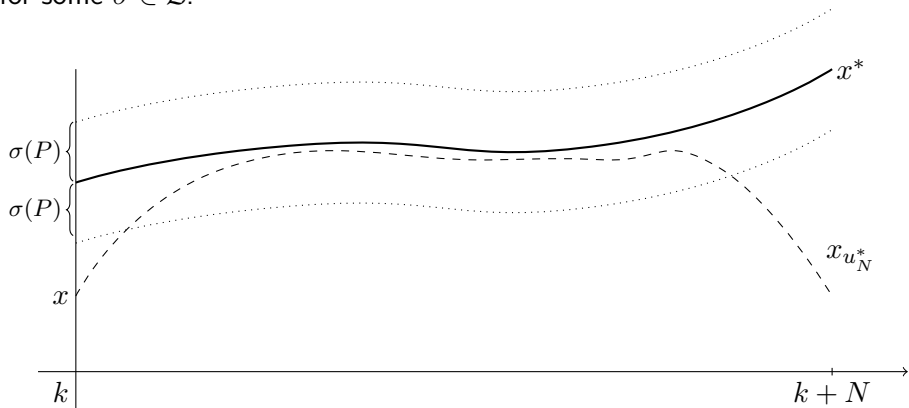
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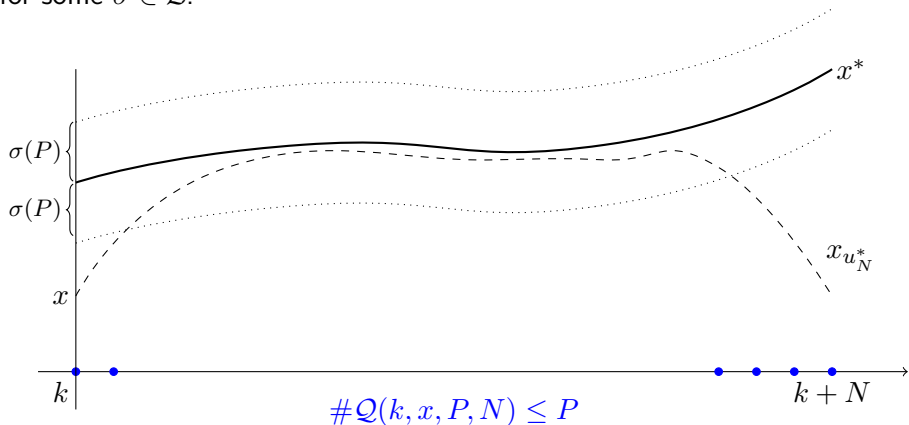
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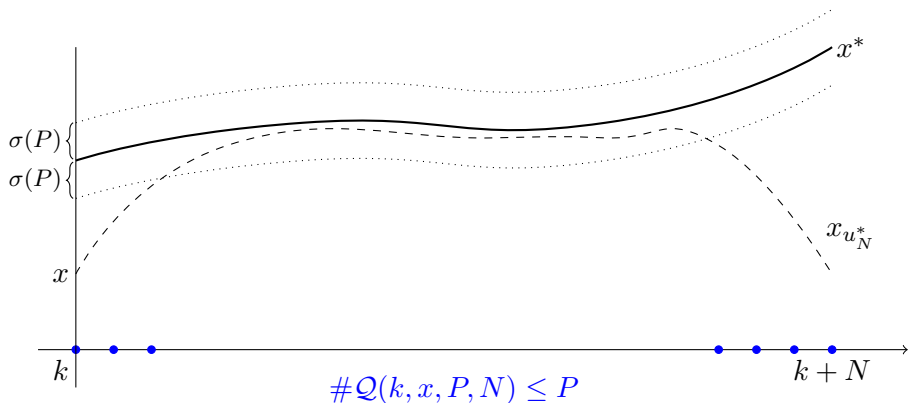
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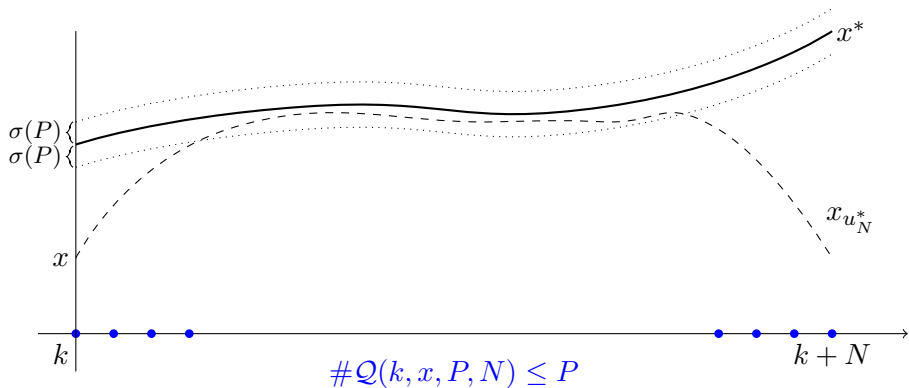
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## Continuity of $\hat{V}_N$ and $\hat{V}_\infty$

- We assume that the optimal value functions  $\hat{V}_N$  and  $\hat{V}_\infty$  are continuous in  $x^*$ , i.e. there is
  - $\varepsilon > 0$
  - $\gamma_V : \mathbb{N} \times \mathbb{R}_0^+ \rightarrow \mathbb{R}_0^+$  with  $\gamma_V(N, r) \rightarrow 0$  if both  $N \rightarrow \infty$  and  $r \rightarrow 0$ , and  $\gamma_V(\cdot, r)$ ,  $\gamma_V(N, \cdot)$  monotonous.

such that for all  $k \in \mathbb{N}$ ,  $N \in \mathbb{N} \cup \{\infty\}$

$$|\hat{V}_N(k, x) - \hat{V}_N(k, x^*(k))| \leq \gamma_V(N, \|x - x^*(k)\|) \quad (8)$$

for all  $x \in \mathcal{B}_\varepsilon(x^*(k))$ .

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- For  $N = \infty$ :  $\rightsquigarrow \hat{V}_\infty(k, x^*(k)) = 0$   
 $|\hat{V}_\infty(k, x)| \leq \omega_V(\|x - x^*(k)\|)$  with  $\omega_V \in \mathcal{K}_\infty$ , i.e.  $\hat{V}_\infty$  is bounded on  $\mathcal{B}_\varepsilon(x^*(k))$ .

## Lemma 1

*If the turnpike property holds and  $\hat{V}_N$  and  $\hat{V}_\infty$  are continuous in  $x^*$  then the following equations hold:*

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

*with  $|R_1(k, x, K)| \leq \omega_V(\sigma(P))$  for all  $k \in \mathbb{N}_0$ , for all  $x \in \mathbb{X}(k)$ , all  $N \in \mathbb{N}$ , all sufficiently large  $P \in \mathbb{N}$  and all  $K \notin Q(k, x, P, \infty)$ .*



## Lemma 2

If the turnpike property holds and  $\hat{V}_N$  and  $\hat{V}_\infty$  are continuous in  $x^*$  then the equation

$$\hat{J}_K(k, x, u_\infty^*) = \hat{J}_K(k, x, u_N^*) + R_3(k, x, K, N) \quad (9)$$

holds with  $|R_3(k, x, K, N)| \leq 2(\gamma_V(N - K, \sigma(P)) + \omega_V(\sigma(P)))$  for all  $k \in \mathbb{N}_0$ , all  $N \in \mathbb{N}$ , all sufficiently large  $P \in \mathbb{N}$ , all  $x \in \mathbb{X}(k)$  and all  $K \in \{0, \dots, N\} \setminus (\mathcal{Q}(k, x, P, N) \cup \mathcal{Q}(k, x, P, \infty))$ .

## Closed-loop cost estimate

### Theorem 1

*If the turnpike property holds and  $\hat{V}_N$  and  $\hat{V}_\infty$  are continuous in  $x^*$  then for each  $k \in \mathbb{N}_0$  and each sufficiently large  $N$ , the closed loop cost satisfies*

$$\hat{J}_L^{\text{cl}}(k, x, \mu_N) + \hat{V}_\infty(k + L, x_{\mu_N}(L, x)) \leq \hat{V}_\infty(k, x) + L\delta(N) \quad (10)$$

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

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## Application: Energy efficient building operation

- PDE model<sup>1</sup> of airflow and temperatures inside a building:

$$\frac{\partial \mathbf{y}_1}{\partial t} + \mathbf{y}_1 \cdot \nabla \mathbf{y}_1 = -\frac{1}{\rho} \nabla y_2 + \nu \Delta \mathbf{y}_1 - \mathbf{g} \alpha (y_3 - \tilde{y}_3) \quad (11)$$

$$\nabla \cdot \mathbf{y}_1 = 0 \quad (12)$$

$$\frac{\partial y_3}{\partial t} + \mathbf{y}_1 \cdot \nabla y_3 = \kappa \Delta y_3 \quad (13)$$

+ time-varying boundary conditions.

$\mathbf{y}_1 : \Omega \times [0, \infty) \rightarrow \mathbb{R}^d$  is **air velocity**

where  $y_2 : \Omega \times [0, \infty) \rightarrow \mathbb{R}$  is **pressure**

$y_3 : \Omega \times [0, \infty) \rightarrow \mathbb{R}$  is **temperature**

and parameters  $\rho, \nu, \mathbf{g}, \alpha, \tilde{y}_3, \kappa$ .

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$y_3 : \Omega \times [0, \infty) \rightarrow \mathbb{R}$  is **temperature**

and parameters  $\rho, \nu, \mathbf{g}, \alpha, \tilde{y}_3, \kappa$ .

- Joint work with Thomas Meurer, Julian Andrej (University of Kiel), and Stefan Volkwein, Luca Mechelli (University of Konstanz)

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

# Outline

Introduction

Model predictive control

MPC convergence results

(Prospective) Application

Conclusion

## Summary, Remarks and open questions

### Summary:

- Time-varying system dynamics leads to time-varying optimal operating behaviour.
- Economic MPC is natural approach.
- Under turnpike property and continuity assumptions we can show that economic MPC approximates infinite horizon optimal performance.

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- We can show that strict dissipativity (+ controllability assumptions) implies turnpike and continuity of the optimal value functions.



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- Time-varying system dynamics leads to time-varying optimal operating behaviour.
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- We can show that strict dissipativity (+ controllability assumptions) implies turnpike and continuity of the optimal value functions.

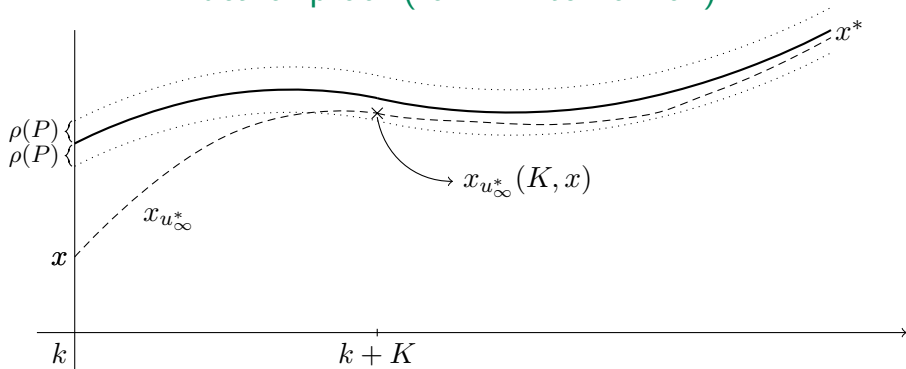
### Open questions:

- Convergence of the closed-loop to  $x^*$ ?
- How to verify dissipativity?

## References

-  Joël Blot and Naïla Hayek. *Infinite-horizon optimal control in the discrete-time framework*. Springer, 2014.
-  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: *SIAM Journal on Control and Optimization* 52.3 (2014), pp. 1935–1957.
-  Matthias Müller and Lars Grüne. “Economic model predictive control without terminal constraints for optimal periodic behavior”. In: *Automatica* (2016).
-  David J Tritton. *Physical fluid dynamics*. Springer Science & Business Media, 2012.

## Idea of proof (for infinite horizon)

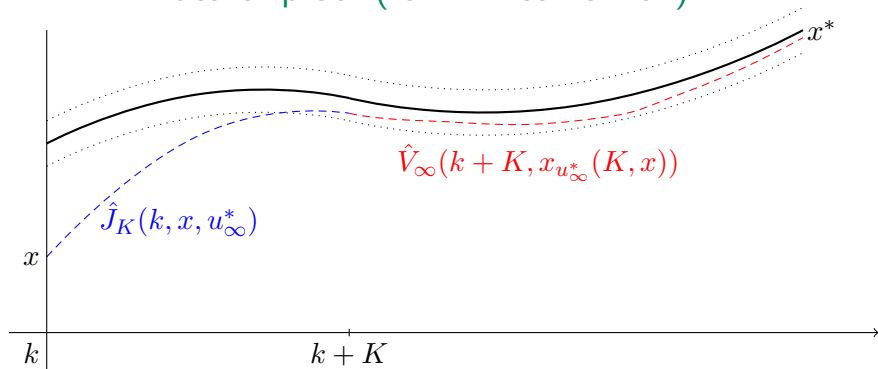


Pick  $P \in \mathbb{N}$  such that  $\rho(P) \leq \varepsilon$  (possible because of turnpike property).

$\rightsquigarrow$  For  $K \notin \mathcal{Q}(k, x, P, \infty)$ :

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

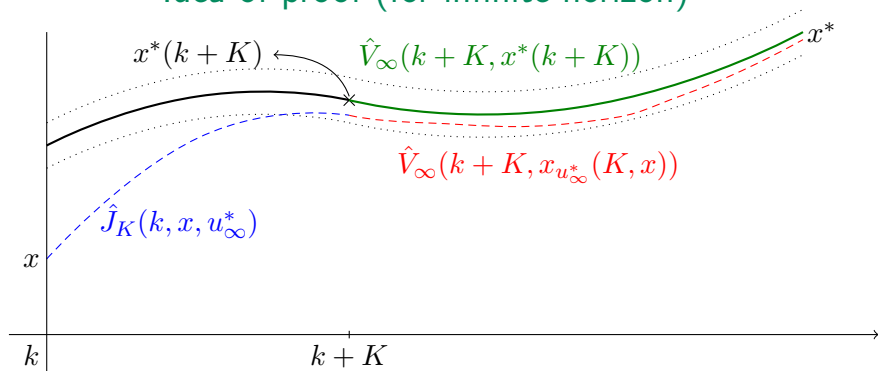
## Idea of proof (for infinite horizon)



Dynamic programming principle:

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

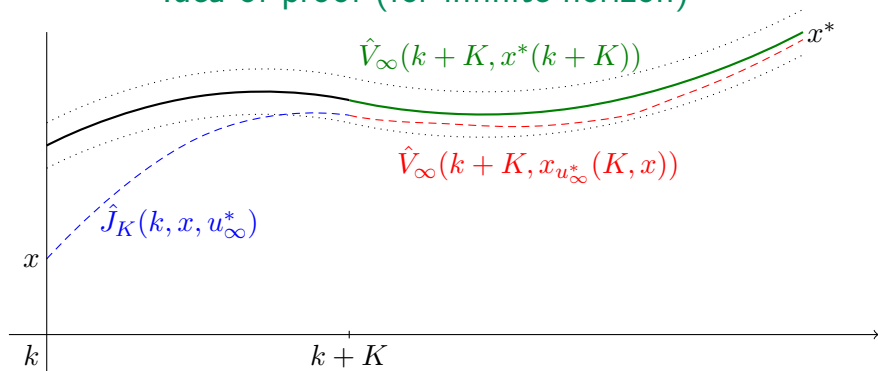
## Idea of proof (for infinite horizon)



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

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

## Idea of proof (for infinite horizon)

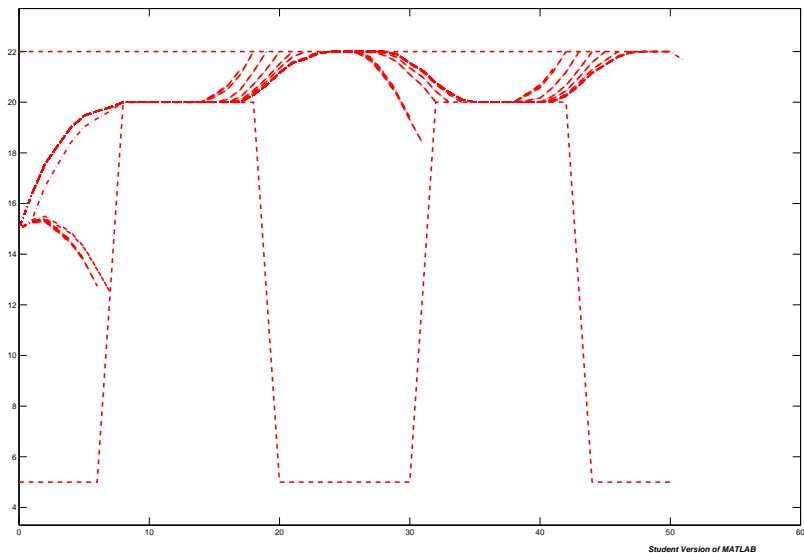


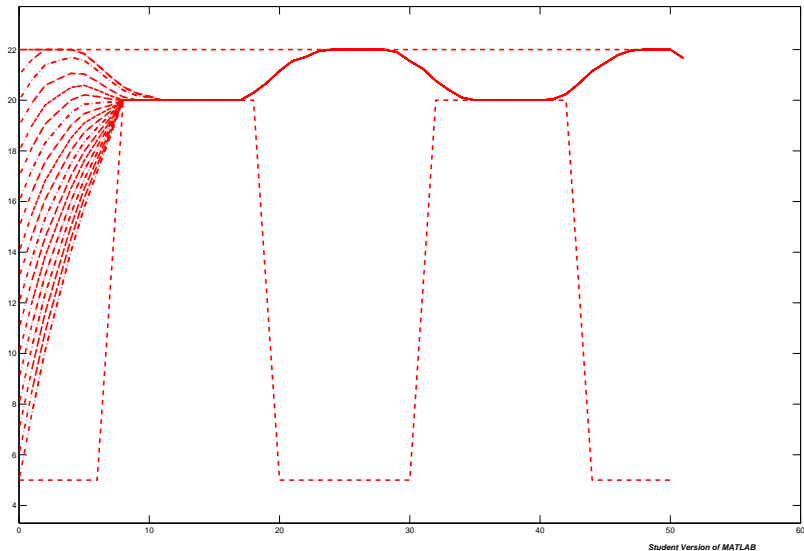
Continuity property:

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

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

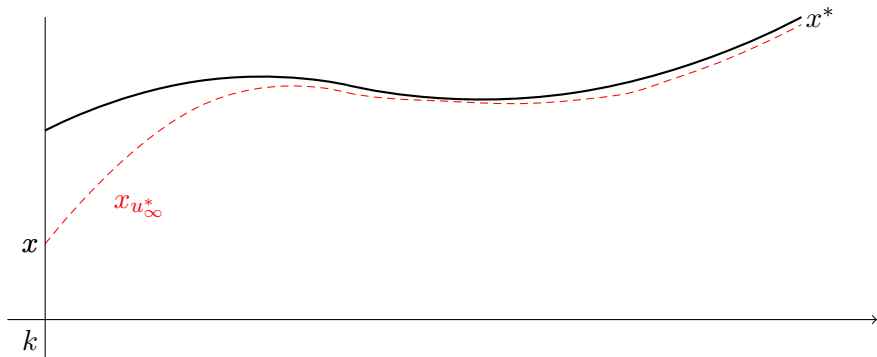
with  $|R_1(k, x, K)| \leq \omega_V(\sigma(P))$ .

Example: turnpike property for different horizon length  $N$ 

Example: turnpike property for different initial value  $x_0$ 

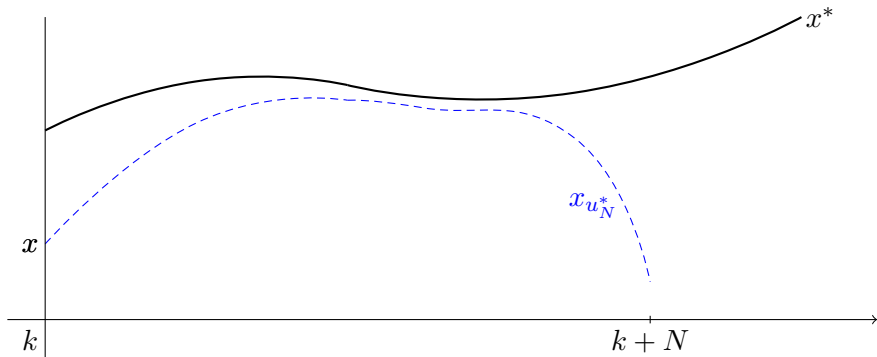


## Idea of proof for Lemma 2



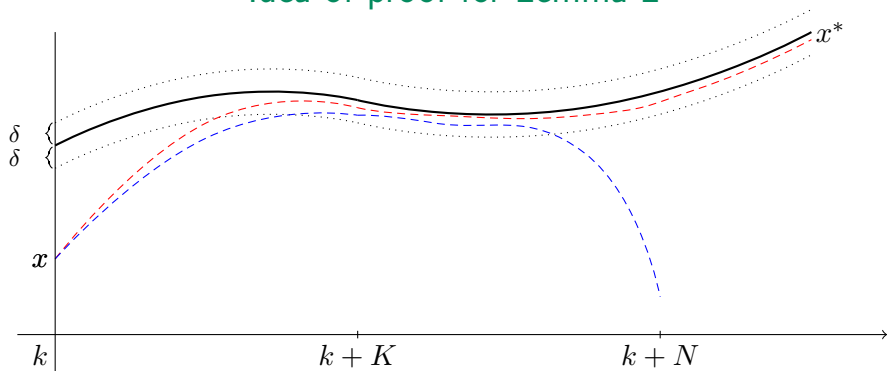
Open-loop on infinite horizon:  $x_{u_{\infty}^*}$

## Idea of proof for Lemma 2



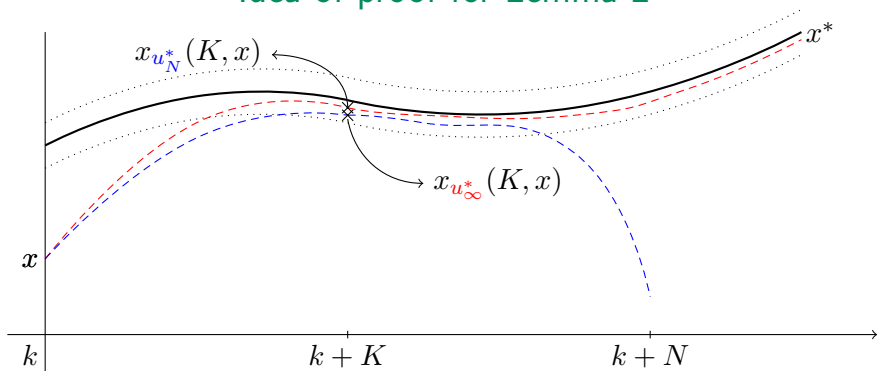
Open-loop on finite horizon:  $x_{u_N}^*$

## Idea of proof for Lemma 2



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

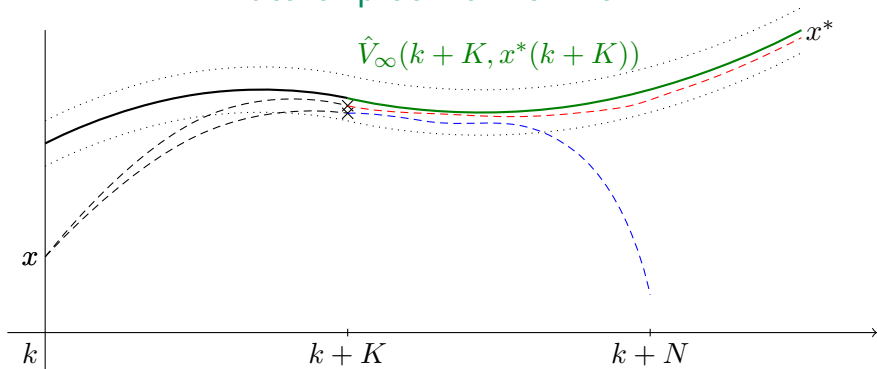
## Idea of proof for Lemma 2



Optimality of  $u_\infty^*$ :

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

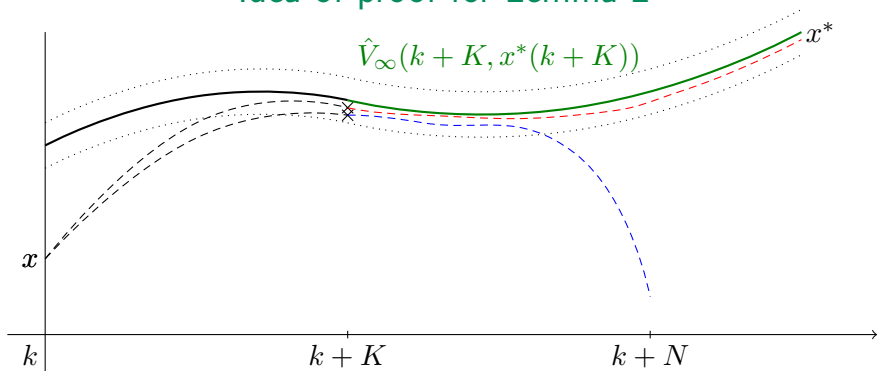
## Idea of proof for Lemma 2



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

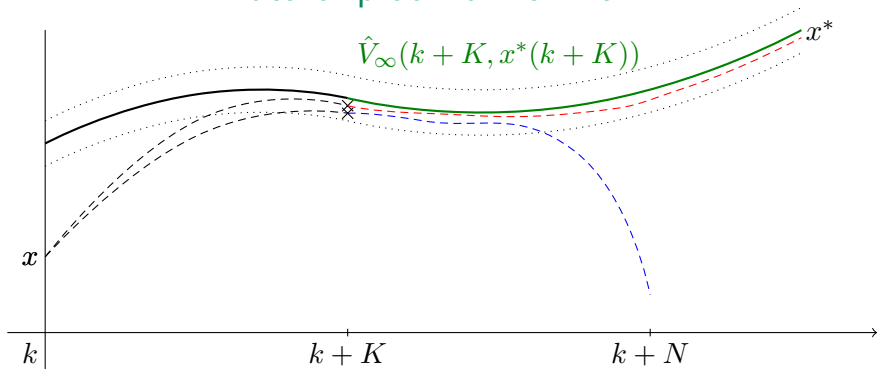


Apply Lemma 1:

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

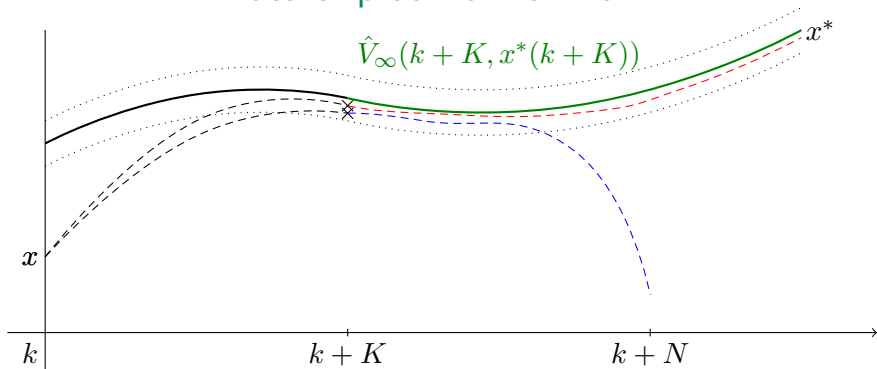
## Idea of proof for Lemma 2



Similarly: converse equality from Lemma 1 (2):

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

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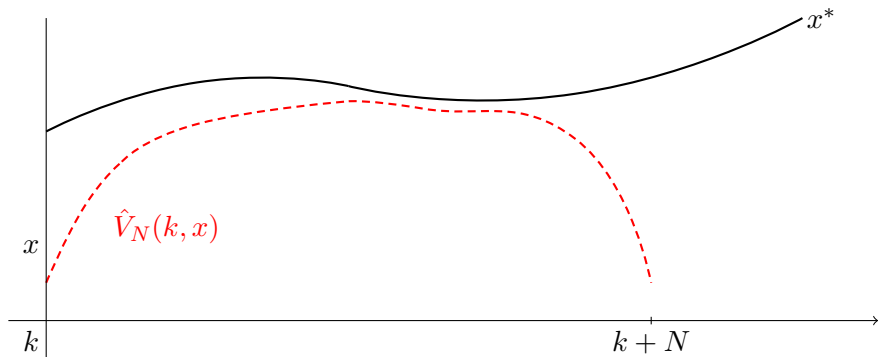
Similarly: converse equality from Lemma 1 (2):

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

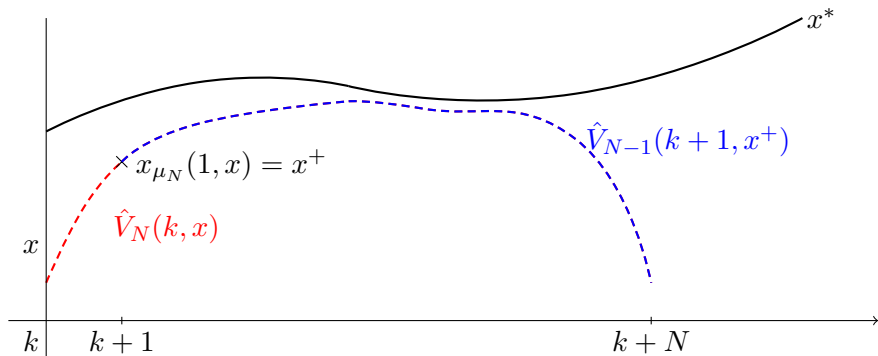


## Idea of proof for Theorem 1



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

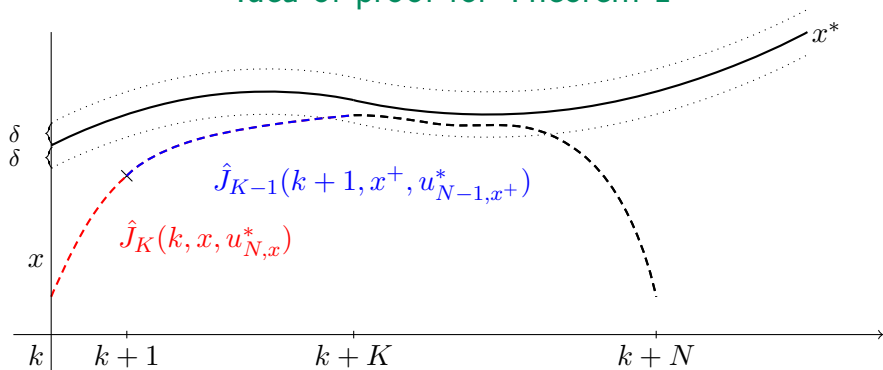
## Idea of proof for Theorem 1



- 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:

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

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

## Idea of proof for Theorem 1

- Stage cost along the MPC closed loop solution:

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

$$\stackrel{\text{Lemma 2}}{=} \hat{J}_K(k, x, u_{\infty,x}^*) - \hat{J}_{K-1}(k+1, x^+, u_{\infty,x^+}^*) + R_3(\dots) - R_3(\dots)$$

# Idea of proof for Theorem 1

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 &\quad + \underbrace{R_1(\dots) - R_1(\dots) + R_3(\dots) - R_3(\dots)}_{=: R_4(k, x, K, N)}
 \end{aligned}$$

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- Stage cost along the MPC closed loop solution:

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 \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:

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

## Boundary conditions

Temperature:

$$y_3 = u \text{ on } \Gamma_1$$

$$-\partial_n y_3 = \gamma(y_3 - y_{\text{out}}) \text{ on } \Gamma \setminus \Gamma_1$$

Air velocity:

$$\mathbf{y}_1 = 0 \text{ on } \Gamma \quad (\text{No-slip BC})$$



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 $-\partial_n y_3 = \gamma(y_3 - y_{\text{out}})$  on  $\Gamma \setminus \Gamma_1$

Air velocity:  $\mathbf{y}_1 = 0$  on  $\Gamma$  (No-slip BC)

Optimal control problem:

$$\begin{aligned}
 & \min_u \frac{1}{2} \int_0^\infty \int_\Gamma u^2 \, dx \, dt \\
 & \text{s.t. } (11), (12), (13), \\
 & \quad \underline{y}_3(t) \leq y_3(t) \leq \overline{y}_3(t) \quad \forall t \in [0, \infty), \\
 & \quad \underline{u}(t) \leq u(t) \leq \overline{u}(t) \quad \forall t \in [0, \infty).
 \end{aligned} \tag{14}$$