

Ökonomische prädiktive Regelung

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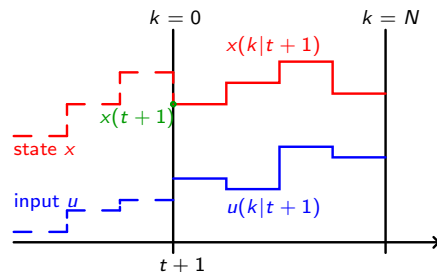
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- Model predictive control (MPC) and economic MPC
- The role of dissipativity
 - Classifying optimal operating behaviors
 - Closed-loop convergence
- Average constraints
- Applications
 - Economic dispatch in power systems
 - Cooperative control of self-interested agents
- Conclusions

Introduction - Model predictive control

Model predictive control (MPC)

- Modern, optimization-based control technique
- Successful applications in many industrial fields



Basic MPC scheme

At each time t ,

- solve finite horizon optimal control problem
- apply first part of optimal solution

Main advantages of MPC

- Can handle hard **constraints** on states and inputs
- Optimization of some **performance criterion**
- Applicable to **nonlinear, MIMO** systems

Model predictive control

- Nonlinear discrete time system $x(t+1) = f(x(t), u(t))$
- State and input constraints $x(t) \in \mathbb{X}, u(t) \in \mathbb{U}$

Standard MPC problem formulation

$$J_N^*(x(t)) = \min_{u(\cdot|t)} \sum_{k=0}^{N-1} \ell(x(k|t), u(k|t)) + V^f(x(N|t))$$

$$\text{s.t. } x(0|t) = x(t), \quad x(k+1|t) = f(x(k|t), u(k|t)), \quad k = 0, \dots, N-1$$

$$x(k|t) \in \mathbb{X}, \quad u(k|t) \in \mathbb{U}, \quad k = 0, \dots, N-1$$

$$x(N|t) \in \mathbb{X}^f$$

- Most results in MPC literature: classical control objective of **setpoint stabilization** is considered
- MPC controller design: determine V^f and \mathbb{X}^f s.t. closed loop is stable [Chen & Allgöwer '98, Mayne et al. '00, Grüne '09, ...]
- Basic assumption: stage cost ℓ is **positive definite** w.r.t. setpoint to be stabilized
- However: **different** control objective is of interest in many applications

- Maximization of product in process industry
- Minimization of energy consumption in building climate control
- Manufacturing industry: cost efficient scheduling of production process

⇒ Setpoint stabilization is **not** primary control objective

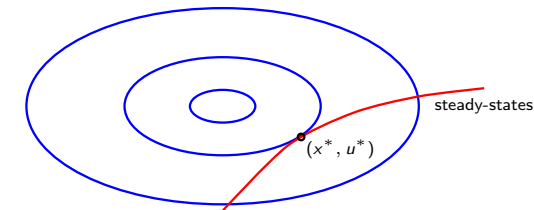
Economic MPC

- Stage cost ℓ can be **general cost function**, need not be positive definite
⇒ Closed-loop system does not necessarily converge to steady-state
- Resulting questions:
 - What is the **optimal operating regime** (steady-state, periodic, ...)?
 - Does the **closed-loop system** "find" optimal operating behavior?

Definition - optimal operation at steady-state

- Optimal steady-state: $(x^*, u^*) = \arg \min_{x \in \mathbb{X}, u \in \mathbb{U}, x=f(x,u)} \ell(x, u)$
- A system is **optimally operated at steady-state** if for each feasible state and input sequences $x(\cdot)$ and $u(\cdot)$ the following holds:

$$\liminf_{T \rightarrow \infty} \sum_{t=0}^{T-1} \frac{\ell(x(t), u(t))}{T} \geq \ell(x^*, u^*).$$



Definition - Dissipativity [Willems '72, Byrnes & Lin '94]

A system is **strictly dissipative** with respect to the **supply rate** s if there exists a **storage function** λ such that for all $x \in \mathbb{X}$ and $u \in \mathbb{U}$ it holds that

$$\lambda(f(x, u)) - \lambda(x) \leq s(x, u) - \alpha(\|x - x^*\|), \quad \alpha \in \mathcal{K}_\infty.$$

Dissipativity and optimal steady-state operation

additional controllability condition
[Müller et al. '13,15]



Optimal operation
at steady-state

Dissipativity w.r.t. supply rate
 $s(x, u) = \ell(x, u) - \ell(x^*, u^*)$



[Angeli et al. '12]

If steady-state operation is optimal, does closed-loop system converge to x^* ?

- Main idea for stability proof in stabilizing MPC: use optimal value function as Lyapunov function

$$J_N^*(x(t+1)) - J^*(x(t)) \leq -\ell(x(t), u(t)) - \ell(x^*, u^*) \leq -\alpha(\|x(t)\|)$$

- In economic MPC: second inequality **not** satisfied!
- Define rotated cost function

$$L(x, u) = \ell(x, u) - \ell(x^*, u^*) + \lambda(x) - \lambda(f(x, u))$$

- If system is strictly dissipative: $L(x, u) \geq \alpha(\|x - x^*\|)$

Modified optimization problem

$$J_N^*(x(t)) = \min_{u(\cdot|t)} \sum_{k=0}^{N-1} \ell(x(k|t), u(k|t)) + V^f(x(N|t))$$

- s.t. $x(0|t) = x(t), \quad x(k+1|t) = f(x(k|t), u(k|t)), \quad k = 0, \dots, N-1$
 $x(k|t) \in \mathbb{X}, \quad u(k|t) \in \mathbb{U}, \quad k = 0, \dots, N-1$
 $x(N|t) \in \mathbb{X}^f$

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Modified optimization problem

$$\tilde{J}_N^*(x(t)) = \min_{u(\cdot|t)} \sum_{k=0}^{N-1} L(x(k|t), u(k|t)) + \tilde{V}^f(x(N|t))$$

s.t. $x(0|t) = x(t), \quad x(k+1|t) = f(x(k|t), u(k|t)), \quad k = 0, \dots, N-1$
 $x(k|t) \in \mathbb{X}, \quad u(k|t) \in \mathbb{U}, \quad k = 0, \dots, N-1$
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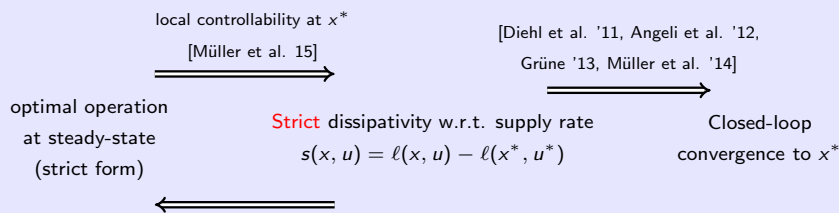
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$$L(x, u) = \ell(x, u) - \ell(x^*, u^*) + \lambda(x) - \lambda(f(x, u))$$

- If system is strictly dissipative: $L(x, u) \geq \alpha(\|x - x^*\|)$
- Key step: original and modified optimization problem have **same** solution
- Can use \tilde{J}_N^* as Lyapunov function:

$$\tilde{J}_N^*(x(t+1)) - \tilde{J}_N^*(x(t)) \leq -L(x(t), u(t)) \leq -\alpha(\|x(t)\|)$$

Strict dissipativity and optimal steady-state operation

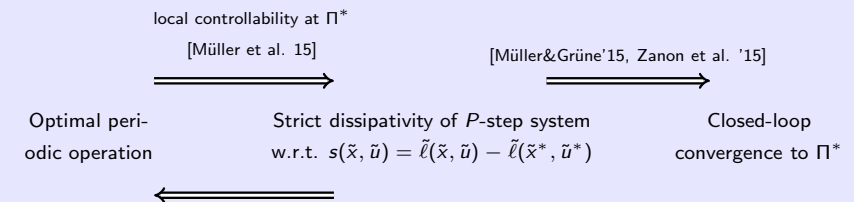


Discussion

- Closed-loop system “does the right thing”, i.e., “finds” optimal operating behavior
- Can be concluded **without** having to compute storage function λ

Results can be extended to optimal periodic behavior:

Dissipativity and optimal periodic operation

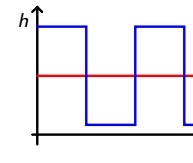


Discussion

- Dissipativity plays central role in economic MPC
- Closed-loop system “does the right thing”, i.e., “finds” optimal operating behavior
- Can be concluded **without** having to compute storage function
- Results hold for both optimal steady-state and periodic behavior

- Model predictive control (MPC) and economic MPC
- The role of dissipativity
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 - Economic dispatch in power systems
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- In economic MPC: closed-loop system does not necessarily converge to some steady-state
- → **Constraints on averages** of state/input variables are important and have to be considered **online**
- Examples:
 - Process industry: average amount of inflow or heat flux in chemical reactor
 - Building climate control: small average temperature deviations
 - Manufacturing industry: average amount of parts which can be processed



- **Transient** average constraints: $\frac{\sum_{k=t}^{t+\tau-1} h(x(k), u(k))}{\tau} \in \mathbb{Y}$ for some $\tau \geq 1$ and all $t \geq 0$
- **Asymptotic** average constraints: $Av[h(x, u)] \in \mathbb{Y}$, with $Av[h(x, u)] := \{\xi : \exists \{t_n\} \rightarrow +\infty : \lim_{n \rightarrow \infty} \frac{\sum_{k=0}^{t_n} h(x(k), u(k))}{t_n+1} = \xi\}$

$$\min_{u(t)} \sum_{k=0}^{N-1} \ell(x(k|t), u(k|t)) + V^f(x(N|t))$$

s.t. $x(0|t) = x(t), \quad x(k+1|t) = f(x(k|t), u(k|t)), \quad k = 0, \dots, N-1$
 $x(k|t) \in \mathbb{X}, \quad u(k|t) \in \mathbb{U}, \quad k = 0, \dots, N-1$
 $x(N|t) \in \mathbb{X}^f(t)$

$$\sum_{k=0}^{N-1} h(x(k|t), u(k|t)) \in \mathbb{Y}_t \quad \mathbb{Y}_{t+1} := \mathbb{Y}_t \oplus \mathbb{Y} \oplus \bar{\mathbb{Y}}(t) \oplus \{-h(x(t), u(t))\}$$

- Additional (time-varying) constraint is needed to ensure satisfaction of average constraints
- Systematic procedure available how $\mathbb{X}^f(t)$ and $\bar{\mathbb{Y}}(t)$ can be computed

Theorem [Müller et al. '14]

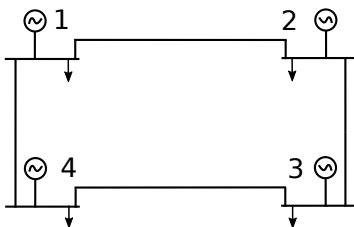
Suppose that $\mathbb{X}^f(t)$ and $\bar{\mathbb{Y}}(t)$ are appropriately defined. Then

- Initial feasibility implies recursive feasibility.
- Closed-loop system satisfies asymptotic average constraints $Av[h(x, u)] \subseteq \mathbb{Y}$.

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Application 1: Real Time Economic Dispatch for Power Systems



Power Dynamics:

$$x_i = [P_i^M, \omega_i, \delta_i], u_i = P_i^C, d_i = P_i^L$$

$$x_i^+ = A_{N_i} x_{N_i} + B_i u_i + E_i d_i$$

Economic Dispatch: Power load P_i^L changes
 \Rightarrow drive system to (new) economic steady state

$$\min \sum_{i=1}^M (P_i^M)^2 \quad \text{st.} \quad \sum_{i=1}^M P_i^M = \sum_{i=1}^M P_i^L$$



State of the art solution:

Economic Automatic Generation Control (EAGC) [Li et al. '14]

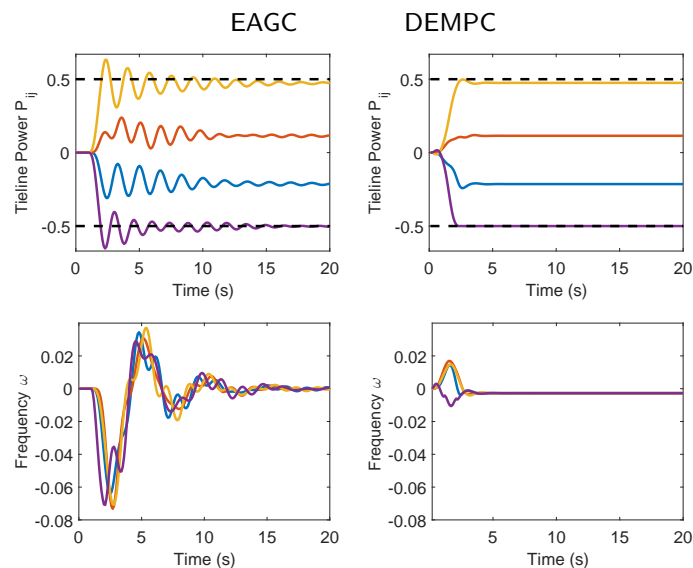
\Rightarrow converges to (new) economic optimal steady state

Distributed Economic MPC solution:

- Economic stage cost: $l_i = (P_i^M)^2 + c\omega_i^2$
- Constraints: $P_i^M \in [P_M^{\min}, P_M^{\max}]$, $P_{ij} \in [P_{ij}^{\min}, P_{ij}^{\max}]$
- Average constraints: $Av[\omega_i] = 0$
- Strict dissipativity: system is optimally operated at steady-state



Simulation Result



Application 2: Cooperative control of self-interested agents

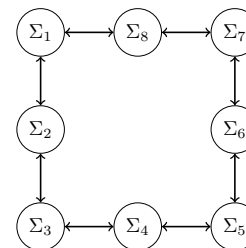
- Several independent systems pursuing a cooperative task; e.g. manufacturing robots, vehicle platoons, wind farms
- Each system has its own individual objective (self-interest), possibly conflicting with cooperative task
- Availability of information and communication between the systems is limited and requires time
- Centralized control not possible / not scalable / not desirable

\Rightarrow **Distributed economic MPC**

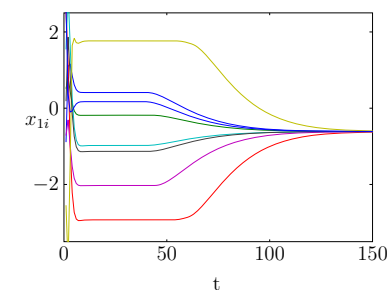


Example: Agreement of self-interested agents

- Double integrator systems: $A_i = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$, $B_i = [0 \ 1]^T$
- **Cooperative goal:** systems have to reach consensus asymptotically ($x_1 = x_2 = \dots = x_n$)
- **Individual goal:** each system has preferred state
→ Cost functions $\ell_i(x_i, u_i) = (x_{1i} - a_i)^2 + (x_{2i} - b_i)^2 + (u_i - c_i)^2$
- Individual objectives **conflicting** with cooperative task of agreement
- Cooperative goal achieved via **average constraints**



Communication topology: cycle



Interpretation of simulation results

- During transient phase, each system converges to its preferred state $x_{1i} = a_i$
 - Systems reach consensus asymptotically → ensured via average constraint
 - Speed of convergence depends on different parameters
- ⇒ Cooperative requirement is satisfied while systems can act according to individual objective during transient phase



Conclusions

- **Model predictive control:** optimization-based control technique with many successful applications
- **Economic MPC:** new approaches to consider general control objectives
- **Dissipativity** plays a crucial role in economic MPC:
 - Classify optimal operating behavior
 - Convergence analysis of closed-loop system
- **Average constraints** are of importance in economic MPC and have to be dealt with online
- **Applications** in distributed control

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