

Coordination of large collections of “uncertain” switched systems

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Johanna Mathieu

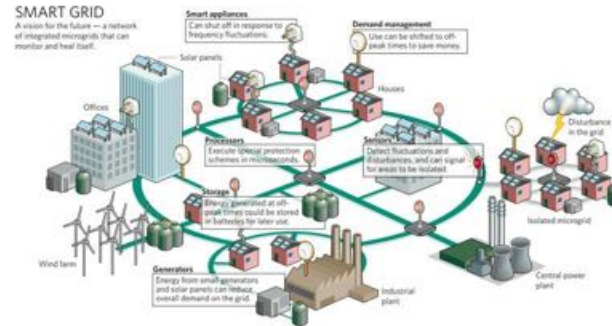


Contents

- What is in this talk?
 - Multi-agent systems
 - Positive systems
 - Switched systems
 - Optimization (relaxations)
 - “Uncertainty”
 - Graph theory
- What is not in this talk?
 - Stability
 - Frequency domain

Motivation and applications

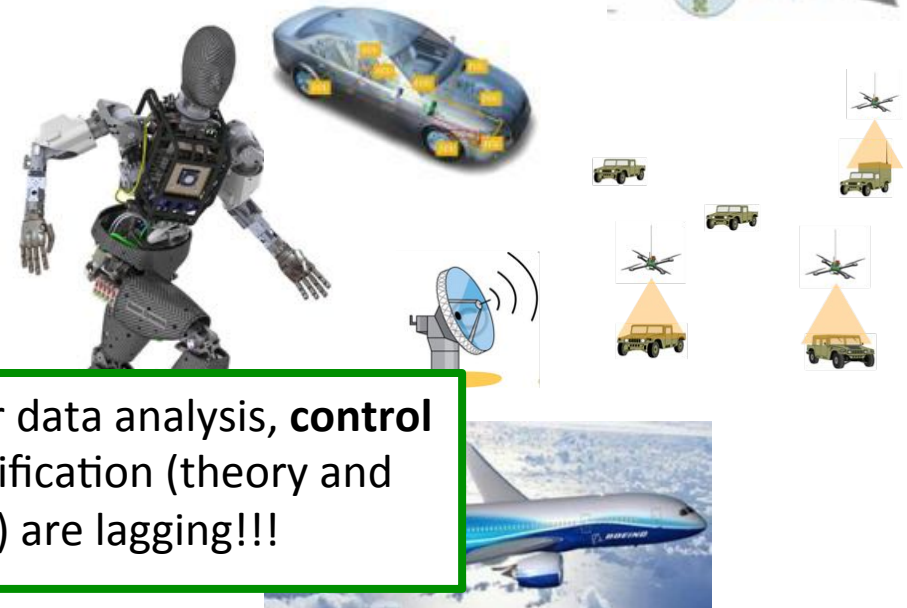
- Large-scale, complex, distributed sensing, actuation and control systems:
 - **Smart grid, Smart buildings, Aircraft systems, Automotive, Robotics, Manufacturing & Automation, Security & Surveillance**



Observations:

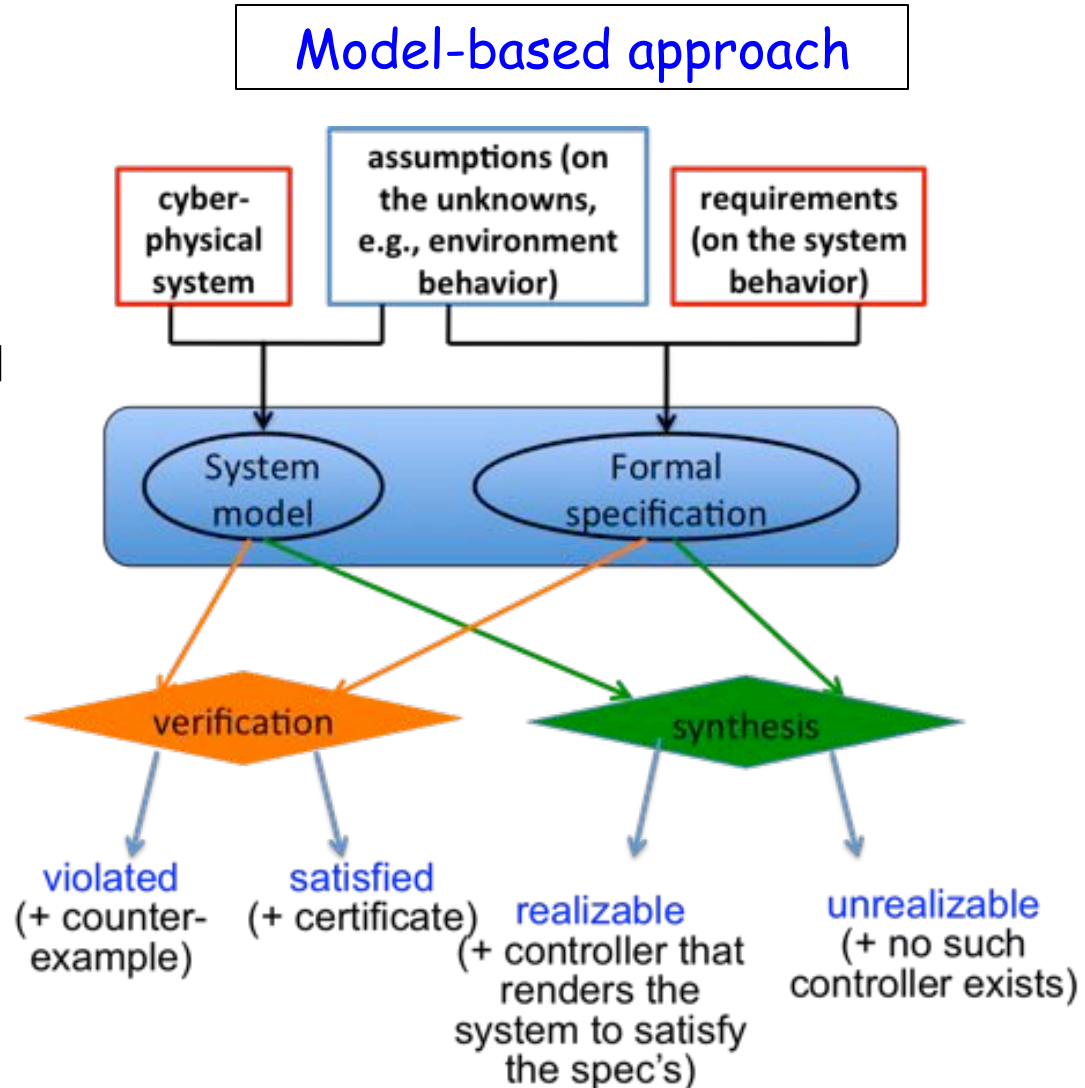
- A very large number of (discrete & continuous) states and decision variables
- Complex requirements → need controllers too complex to be designed/analyzed by

Scalable tools for data analysis, **control design** and verification (theory and software) are lagging!!!

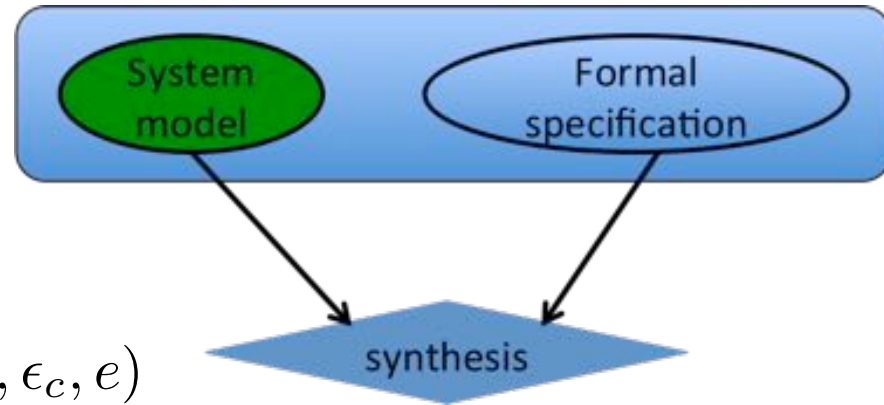


Formal methods in control

- Models for:
 - the system (usually hybrid/switched ODEs, with continuous/discrete inputs, disturbances and parametric uncertainty)
 - the environment (faults, external events)
- Formalized assumptions and requirements
 - linear temporal logic and its extensions
- Methods for verification and synthesis
 - algorithms that can process formal models and requirements to do analysis and control synthesis



System models



Differential equations (continuous-time):

$$\dot{x} = f(x, u_c, u_d, \epsilon_c, e)$$

Or, difference equations (discrete-time):

$$x(k+1) = f(x(k), u_c(k), u_d(k), \epsilon_c(k), e(k))$$

$x \in \mathcal{X}$: state

$u_c \in \mathcal{U}_c$: continuous control input

$u_d \in \mathcal{U}_d$: discrete control input

$\epsilon_c \in \mathcal{D}_c$: disturbance input

$e \in \mathcal{D}_d$: discrete uncontrollable input

$$\mathcal{X} \subset \mathbb{R}^N$$

Some characteristics:

- Hard constraints (on input and states)
- Infinite horizon specifications
- Hybrid (either the system or the controller or both)
- Robust/reactive

State-of-the-art in formal methods in control (incomplete list!)

- Hard state/input constraints, hybrid dynamics, complex specifications (e.g., temporal logics)
 - Belta, Fainekos, Girard, Murray, Pappas, Tabuada, Tomlin ...
- Applications (with “small” state-space dim.)
 - Robotics, building thermal management, adaptive cruise control, aircraft subsystems, traffic control
- “Medium”-scale systems
 - Monotonicity (Hafner & Del Vecchio 11, Coogan & Arcak 15)
 - Multi-scale abstractions for safety (Girard et al. 13)
 - Compositional synthesis (Nilsson & Ozay, Chen et al., Kim et al.), incremental abstractions (Nilsson & Ozay)

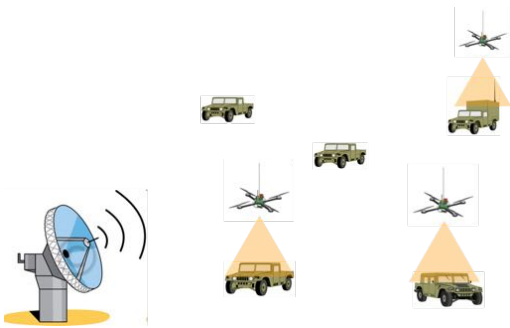
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- “Large”-scale (but not synthesis)
 - Parametric verification of rectangular hybrid automata (Johnson & Mitra 12)
 - Abstractions of large collections of stochastic systems (Soudjani & Abate 15)

Recurring theme:
structural properties

Large collections of systems

Example 1: Emergency response with a robotic swarm

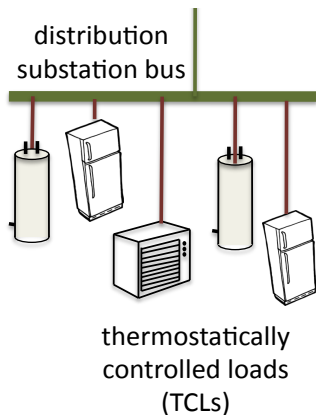
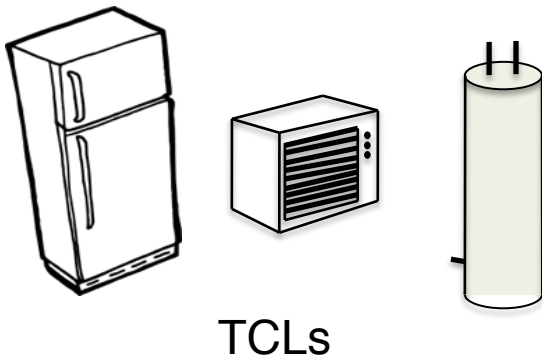


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- Deploy a large collection of robots (e.g., quadrotors, ground vehicles) for search and rescue mission
- Plan trajectories by taking dynamic constraints into account
- Requirements:
 - Sufficiently many robots in certain areas at any given time
 - Not too many robots in certain regions (danger zones)
 - Collision avoidance
 - Charging/reporting constraints

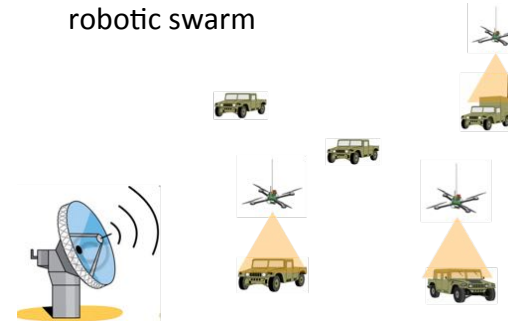
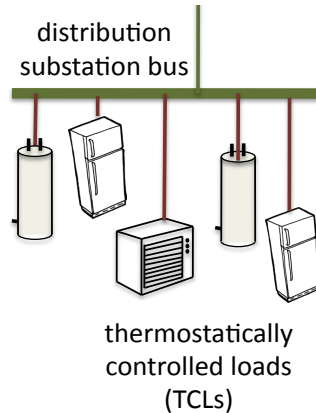
Large collections of systems

Example 2: Coordination of thermostatically controlled loads (TCLs)



- Thermostatically controlled loads (e.g., refrigerators, air conditioners, water heaters) for demand response
- Thermal dynamics can be controlled via ON/OFF switches
- Requirements:
 - Not too many TCLs ON at the same time (to avoid line overload)
 - Sufficiently many ON all the time (to utilize renewable energy)
 - Local temperature constraints (never out of desired temperature range)

Common structural properties



- Large number of systems, small number of classes
- Counting constraints: “how many in each mode?”, “how many in what region?”
- Identity of individual systems is not important

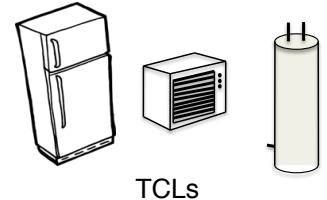
For simplicity, assume:

- dynamics are identical within each class
- (wlog) there is only one class

Mathematical formulation: TCLs

The temperature θ of a TCL has dynamics

$$\dot{\theta}_i = \begin{cases} f_{on}(\theta_i), & \text{if TCL is on} \\ f_{off}(\theta_i), & \text{if TCL is off} \end{cases}$$



Suppose we have a collection of TCL's $\{\theta_i\}_{i \in [N]}$.

- Customers: Want TCL temperature to be close to a desired temperature θ_i^{des} , but small deviations are allowed.

$$\|\theta_i - \theta_i^{des}\| \leq \Delta \quad (1)$$

- Utility company: Wants to control aggregate demand, i.e. the number of TCLs that are on

$$\sum_{i=1}^N \mathbb{1}_{\{\text{TCL } i \text{ is on}\}} \quad (2)$$

Goal: Find a switching (i.e., on/off) strategy that exploits the flexibility in (1) so that (2) can be controlled.

Mathematical formulation: General

- N identical switched system with M modes:

$$\dot{x}_i(t) = f_{\sigma_i(t)}(x_i(t)), \quad \sigma_i : \mathbb{R} \mapsto [M],$$

- Mode-specific unsafe sets: $\mathcal{U}_m, m \in [M]$
 - Equivalent to forced mode switches.
- Mode-counting bounds:

$$\underline{K}_m \leq \sum_{i=1}^N \mathbb{1}_m(\sigma_i(t)) \leq \overline{K}_m \quad (3)$$

Want to synthesize a switching strategy σ_i such that (3) satisfied over time.

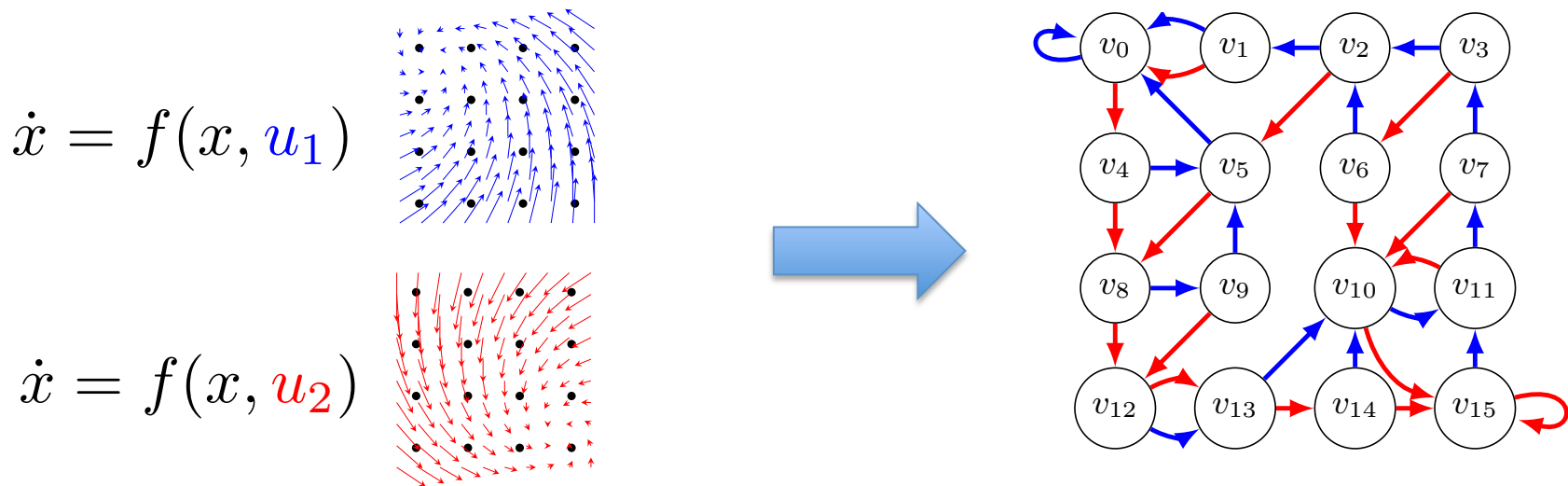
Structural property: both the dynamics and the specification (counting constraints) are **permutation invariant!**

Solution overview

- Construct symbolic abstractions (i.e., a finite transition system) and aggregate dynamics and define “equivalent” problems on these structures
- (Analyze abstractions to understand fundamental limitations if any)
- An optimization-based solution approach
- Analysis of the solution approach

Solution overview

- Construct symbolic abstractions (i.e., a finite transition system)
 - ϵ -approximate bisimilar abstraction



- for each path on the finite transition system, there is a piecewise constant input that generates a trajectory such that time-sampled trajectory remains ϵ -close to the discrete states

Abstraction of individual dynamics

- Assume dynamics are δ -GAS with \mathcal{KL} functions β_m

$$\|\phi_t^m(x) - \phi_t^m(y)\|_\infty \leq \beta_m(\|x - y\|_\infty, t). \quad (4)$$

Abstraction of individual dynamics

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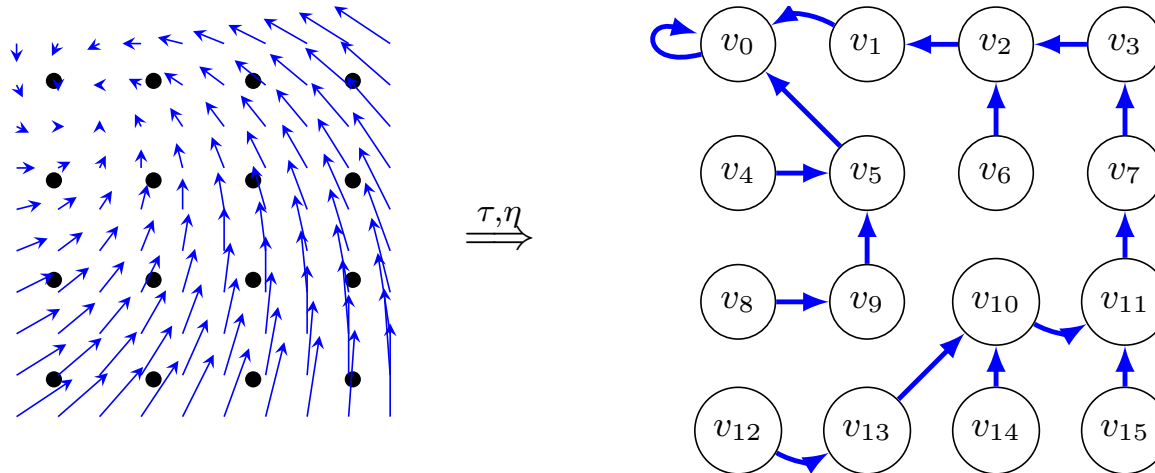
- With discretization in time (τ) and space (η), an ϵ -approximate bisimilar model is obtained if $\beta_m(\epsilon, \tau) + \frac{\eta}{2} \leq \epsilon$.

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 - Mode 1 abstraction

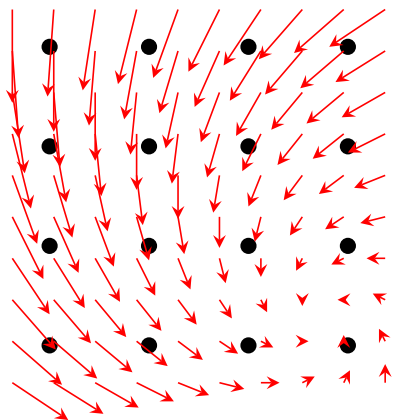


Abstraction of individual dynamics

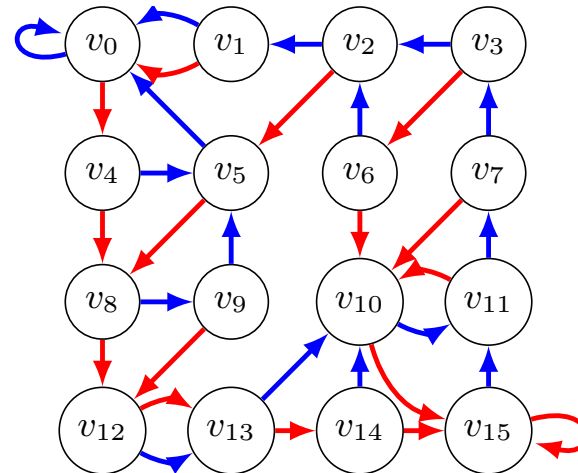
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 - Mode 2 abstraction



$\xrightarrow{\tau, \eta}$

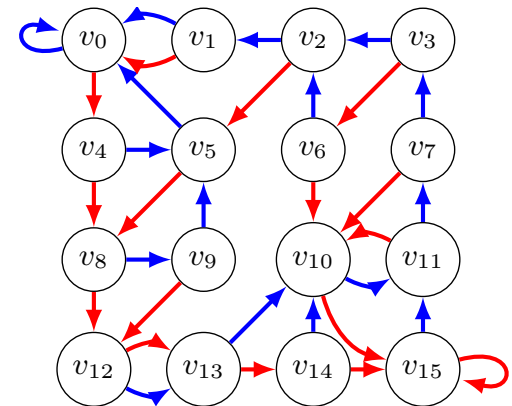


mode-transition graph $G = (V, E)$

Some observations

- For a homogeneous collection, each system will have an identical mode-transition graph
- Transition graphs are deterministic

mode-transition graph $G = (V, E)$



Some observations

- For a homogeneous collection, each system will have an identical mode-transition graph
- Transition graphs are deterministic
- Consider mild heterogeneity

mode-transition graph $G = (V, E)$

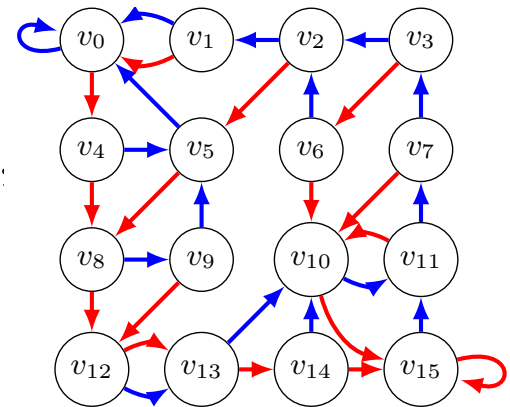
$$\dot{x}_i(t) = f_{\sigma_i(t)}(x_i(t), d_i(t)), \quad \sigma_i : \mathbb{R} \mapsto [M],$$

where $d_i \in \mathcal{D}$ (bounded parametric uncertainty or disturbance). If $f_m(x, d)$ is L_m -Lipschitz in x , and

$$\|f_m(x, d) - f_m(x, 0)\| \leq \delta_m \quad \text{for all } d_i \in \mathcal{D},$$

then, with discretization in time (τ) and space (η), an ϵ -approximate bisimilar model is obtained if

$$\beta_m(\epsilon, \tau) + \frac{\delta_m}{L_m} (e^{L_m \tau} - 1) + \frac{\eta}{2} \leq \epsilon.$$



Aggregate dynamics on graph

Let $V = \{v_1, \dots, v_K\}$ denote the nodes of mode-transition graph $G = (V, E)$. Introduce the states $w_k^{m_1}$ and $r_k^{m_1, m_2}$.

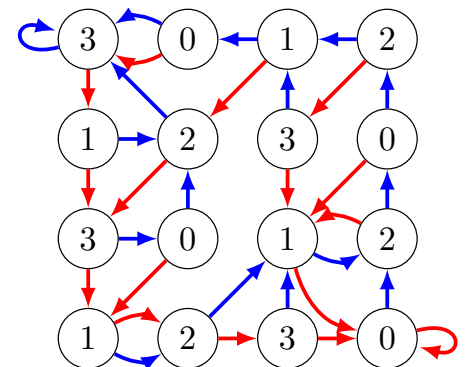
- w_i^m represents **number of systems in mode m at v_k** .
- $r_k^{m_1, m_2}$ represents **number of systems at v_k that switch from m_1 to m_2** .
- The dynamics become

$$(w_k^{m_1})^+ = \sum_{j \in \mathcal{N}_k^{m_1}} \left(w_j^{m_1} + \sum_{m_2} r_j^{m_2, m_1} - r_j^{m_1, m_2} \right),$$

- Constrained control actions:

$$0 \leq \sum_{m_2} r_k^{m_1, m_2} \leq w_k^{m_1},$$

- Compact description: $\mathbf{w}^+ = A\mathbf{w} + B\mathbf{r}$



Equivalent problem on aggregate dynamics

Theorem 1:

Consider aggregate dynamics $\Sigma_G : \mathbf{w}^+ = A\mathbf{w} + B\mathbf{r}$ with safety and mode-counting constraints:

$$w_k^m(t) = 0 \quad \forall k \in U_m, \quad (5)$$

$$\underline{K}_m \leq \sum_{i \in [N]} w_i^m(t) \leq \overline{K}_m. \quad (6)$$

Then,

- if \exists sequence of control inputs \mathbf{r}^ω for Σ_G that enforce (5) and (6) with $U_m + B_\epsilon$, then \exists a solution to the original problem.
- if \nexists a sequence of control input \mathbf{r}^ω for Σ_G that enforces (5) and (6) with $U_m - B_\epsilon$, then no solution to the original problem.

We will focus on aggregate dynamics. We need infinite horizon strategies!

Solution strategy: from a given initial state, steer the system, while respecting the constraints, to a **nice state** from which a periodic input suffices.

Controllability-like conditions

Solution strategy: from a given **initial state**, **steer the system**, while respecting the constraints, **to a nice state** from which a periodic input suffices.

- Let's put the mode-counting constraints aside.
- Are there any fundamental limitations on what states can be reached from an initial condition?

$$\Sigma_G : \mathbf{w}^+ = A\mathbf{w} + B\mathbf{r}$$

with local safety and
input constraints

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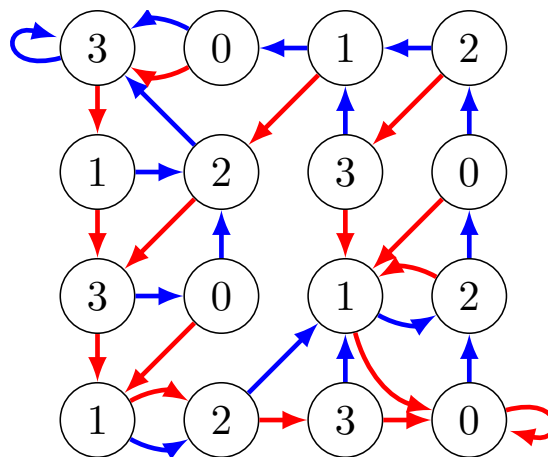
Definition: The period n of a strongly connected graph is the greatest common divisor of the lengths of its cycles.

Theorem 2: If the connected components of mode-transition graph has period $n=1$, any state is reachable from any other state (within the connected component). If $n>1$, then the reachable states live on a hyperplane arrangement with n hyperplanes.

Solution strategy

Solution strategy: from a given initial state, steer the system, while respecting the constraints, to a **nice state** from which a periodic input suffices.

- **Prefix:** for a fixed horizon T , given initial state, we will steer the state at time T to “**nice**” cycles
- **Suffix:** let individual systems circulate in the cycles



Cycle terminology

- Cycle $C = \{v_{c_1}, \dots, v_{c_{|C|}}\}$ in G
- A *cycle assignment* for C is a function $\alpha : C \mapsto \mathbb{R}^+$.

Mode-counts on for a cycle assignment:

- Max-count $\overline{\Psi}^m(C, \alpha)$: maximal number of individual systems simultaneously in mode m when circulating α in C :
- Min-count $\underline{\Psi}^m(C, \alpha)$: minimal number of individual systems simultaneously in mode m when circulating α in C :

Illustration: cycles

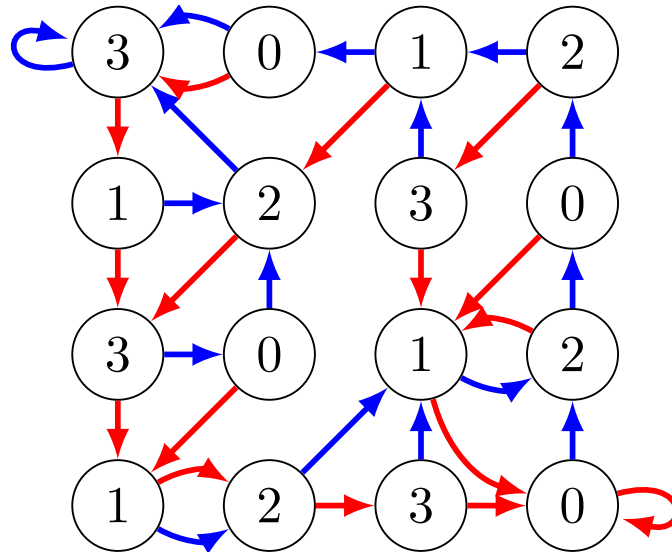


Illustration: cycles

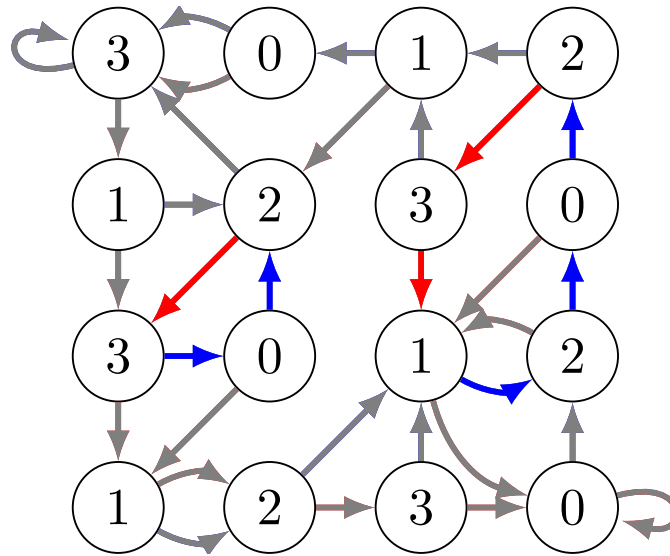
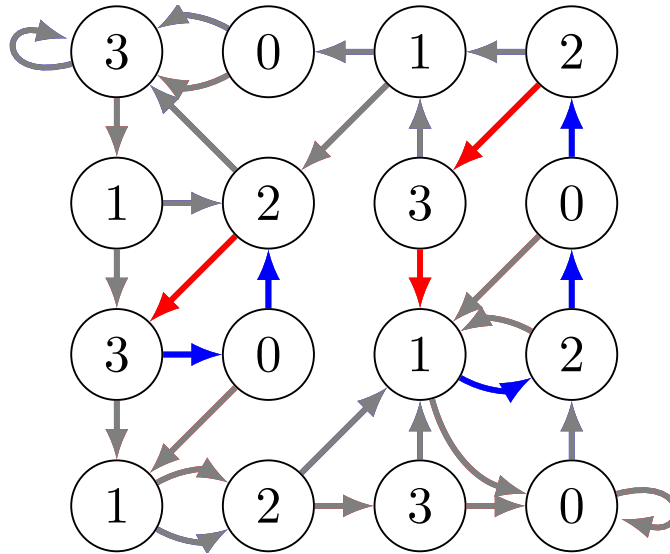


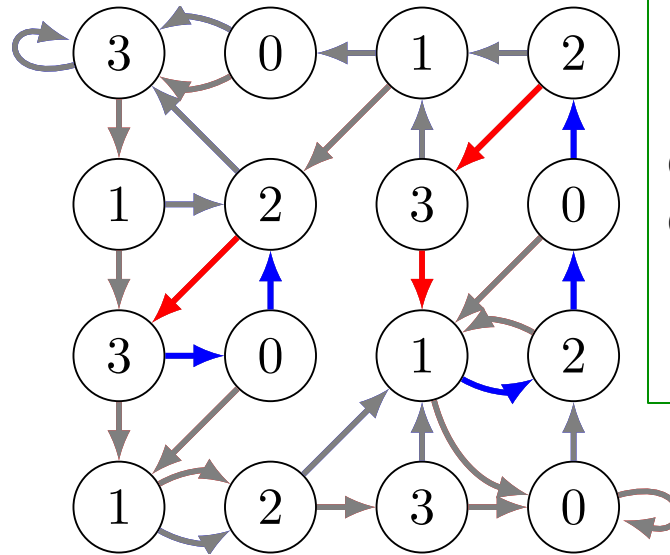
Illustration: cycles



- Big cycle C_1 , assignment $\alpha_1 = [1, 2, 0, 2, 3]$, gives red counts

$$\underline{\Psi}(C_1, \alpha_1) = 2, \quad \overline{\Psi}(C_1, \alpha_1) = 5$$

Illustration: cycles



Mode-counting constraints

$$\underline{\Psi}^m(C, \alpha) \geq \underline{K}_m, \quad \overline{\Psi}^m(C, \alpha) \leq \overline{K}_m,$$

can be represented as linear constraints

$$\underline{K}_m \mathbf{1} \leq Y_C^m \alpha \leq \overline{K}_m \mathbf{1}$$

Y_c^m is a circulant matrix.

- Big cycle C_1 , assignment $\alpha_1 = [1, 2, 0, 2, 3]$, gives red counts

$$\underline{\Psi}(C_1, \alpha_1) = 2, \quad \overline{\Psi}(C_1, \alpha_1) = 5$$

- Small cycle C_2 , assignment $\alpha_2 = [3, 0, 2]$, gives red counts

$$\underline{\Psi}(C_2, \alpha_2) = 0, \quad \overline{\Psi}(C_2, \alpha_2) = 3$$

Solution via linear programming

For cycles C_1, \dots, C_J , required mode-counts K_m , horizon T

find $\alpha_1, \dots, \alpha_J$ cycle assignments,

$\mathbf{r}(0), \dots, \mathbf{r}(T-1)$,

$\mathbf{w}(0), \dots, \mathbf{w}(T)$,

s.t. $\underline{K}_m < \sum w_j^{m_1}(t) < \overline{K}_m \quad 0 \leq t \leq T-1$ mode-counting during prefix

Feasibility problem with linear constraints:

- integrality constraints on the inputs (ILP)
- relaxing integrality (LP)

\underline{K}_m

\sum_j

$\Lambda(\mathbf{w}(0))$

Number of constraints and variables are independent of the number of systems N!

mode-counting during suffix

boundary conditions between prefix and suffix

$\mathbf{w}(t+1) = A\mathbf{w}(t) + B\mathbf{r}(t), \quad t = 0, \dots, T-1,$

system dynamics

$\Lambda(\mathbf{w}(0)) = \lambda_0,$

$\sum_{m_2} r_j^{m_1, m_2} = w_j^{m_1}$ for all $j \in \bigcup_{i \in U_{m_1}} \mathcal{N}_i^{m_1},$

local safety constraints

$r_j^{m_2, m_1} = 0$ for all $m_2 \in [M], j \in U_{m_1},$

control constraints.

Analysis

- **Integer solutions (ILP)**
 - **Completeness of prefix-suffix solutions:** There exists a finite T and some maximal cycle length L such that ILP with all cycles with length less than L provides a complete solution to the original problem
 - From any feasible ILP solution, we can extract a solution to the original problem
- **Non-integer solutions (LP):**
 - Enough to consider simple cycles
 - Gives certificates for non-existence of solutions
- **Rounding a non-integer solution:**
 - A non-integer solution over the cycles can be rounded to an integer feasible solution with mode counting loss at most

$$\underline{\Psi}^m(C, \alpha_{int}) \leq \underline{\Psi}^m(C, \alpha_{avg}) + \frac{|C|}{4}$$

Intuition behind cycles: TCLs

$$\dot{\theta}_i = -a(\theta_i - \theta_a) - bP_m$$

θ :room temperature

θ_a :ambient temperature

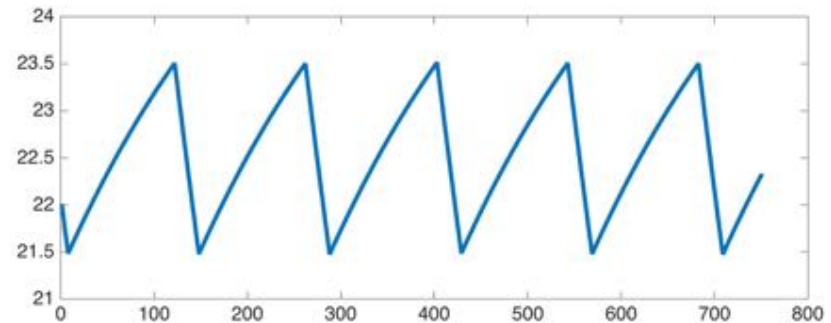
$P_m = 0$ when OFF

$P_m = 5.6$ when ON

local safety

$$\theta_i \in [21.5, 23.5]$$

For an individual system if only local ON/OFF control is used (no demand response for extra switching), the temperature evolves as follows:



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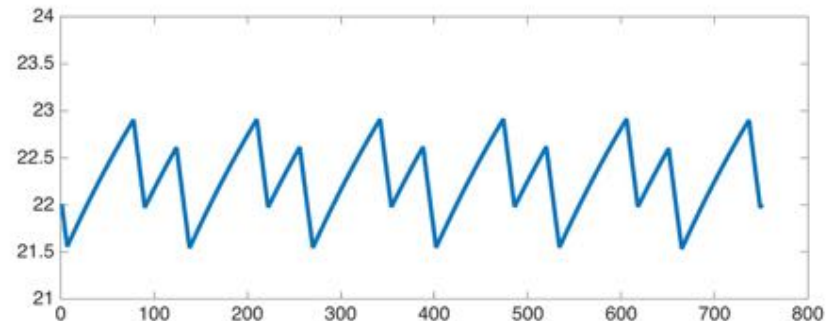
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Roughly, cycles are defining new “bands” within the dead-band allowed by the local safety constraints. That is, we are changing the duty cycle.

Results on TCLs

N = 10000 units

10000-D state-space with 2^{10000} modes!

$$\dot{\theta}_i = -a(\theta_i - \theta_a) - bP_m$$

θ :room temperature

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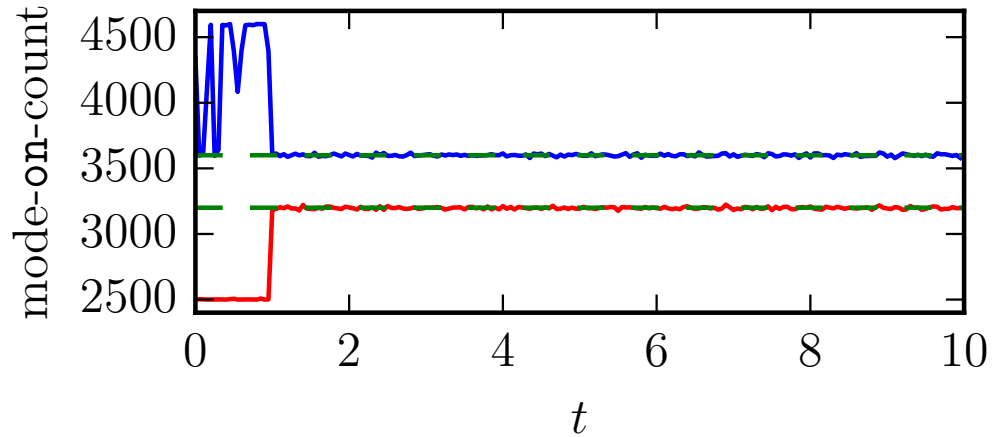
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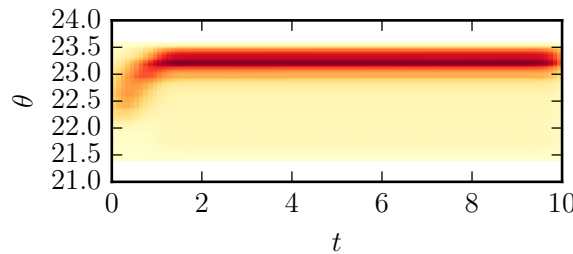
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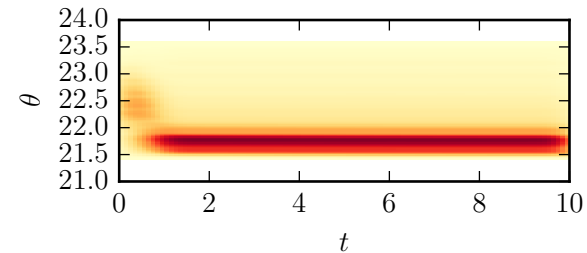
Two different runs with different mode-counting constraints (also stricter constraints at the suffix)



Lower mode-count:



Higher mode-count:

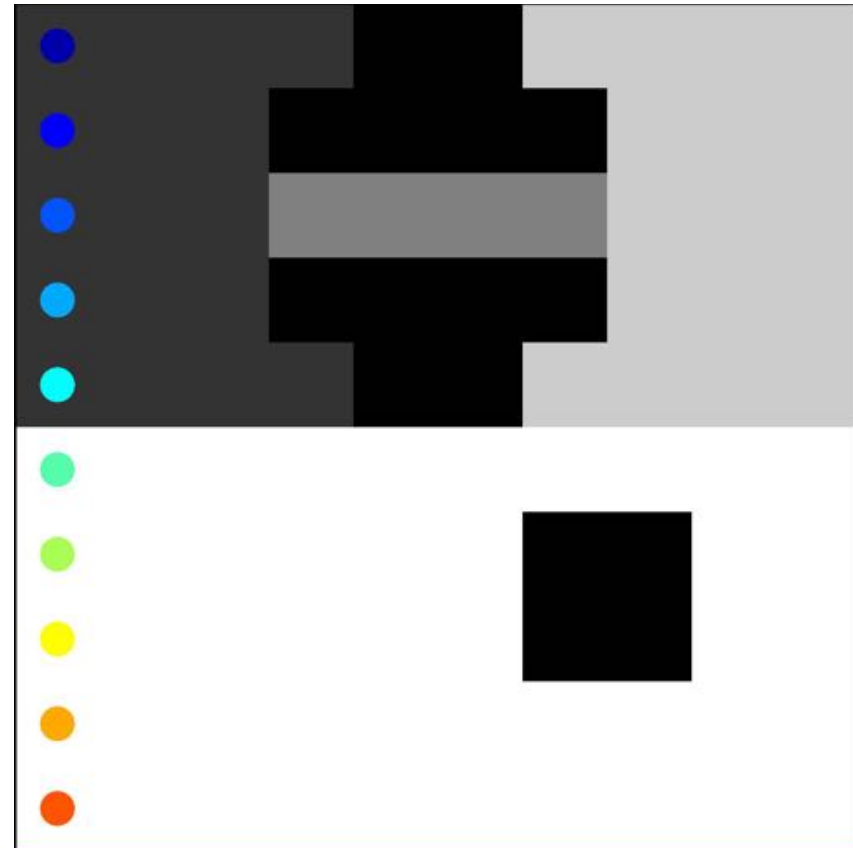


Beyond mode counting

- Counting the agents in a region of state-space
- Time-evolution of counting constraints (counting LTL)

$\varphi ::= \text{True} \mid cp \mid \varphi_1 \wedge \varphi_2 \mid \neg\varphi \mid \bigcirc\varphi \mid \varphi_1 \mathcal{U} \varphi_2,$

- $cp = [\text{atom prop.}, \text{count}]$
- Possible to encode asynchrony as well



With Yunus Emre Sahin & Petter Nilsson
ICCP17

Summary: structure for scalability

- A control synthesis method for large collections of systems with **counting constraints**
 - exploits the symmetry (**permutation invariance**) in the dynamics and in specifications
 - works across scales (10 to 10K or more systems)
 - with potential applications in different domains
 - extensions to asynchrony, counting temporal logic

Summary: structure for scalability

- A control synthesis method for large collections of systems with **counting constraints**
 - exploits the symmetry (**permutation invariance**) in the dynamics and in specifications
 - works across scales (10 to 10K or more systems)
 - with potential applications in different domains
 - extensions to asynchrony, counting temporal logic
- Current work
 - partial information
 - non-deterministic abstractions (for not incrementally stable systems), asynchronous switching
 - tighter rounding bounds between LP and ILP
 - other types of symmetries that can be exploited