

Model Predictive Control from Signal Temporal Logic Specifications: A Case Study

Vasu Raman¹

Alexandre Donzé² and Mehdi Maasoumy²

¹California Institute of Technology

²University of California at Berkeley



CyPhy
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Modern Cyber-Physical Systems



Caltech DUC vehicle



NASA/JPL-Caltech Rover

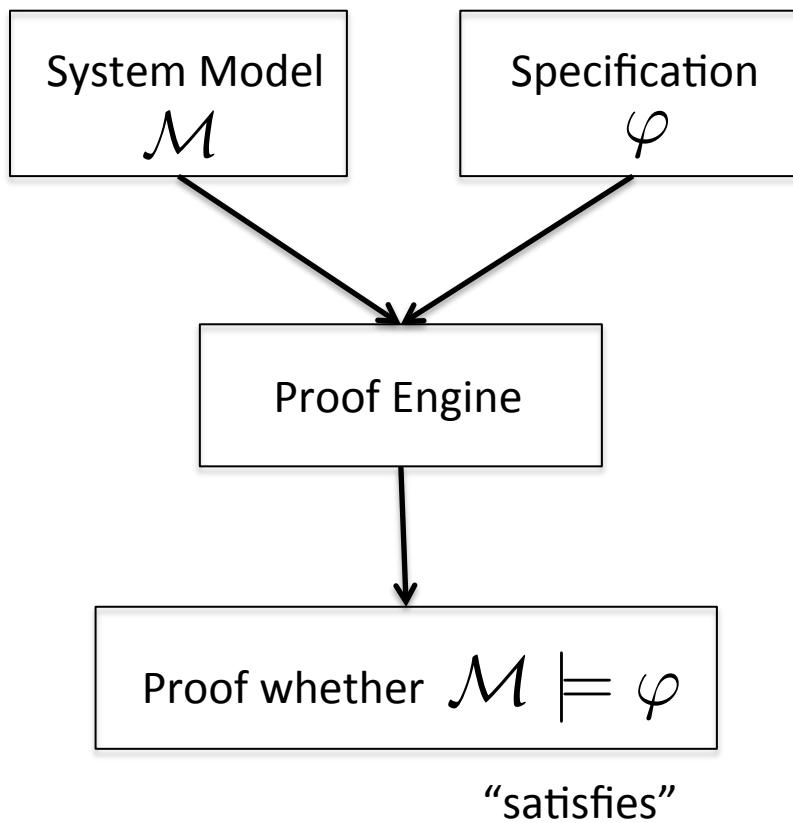


Smart Grid
(automationfederation.org)

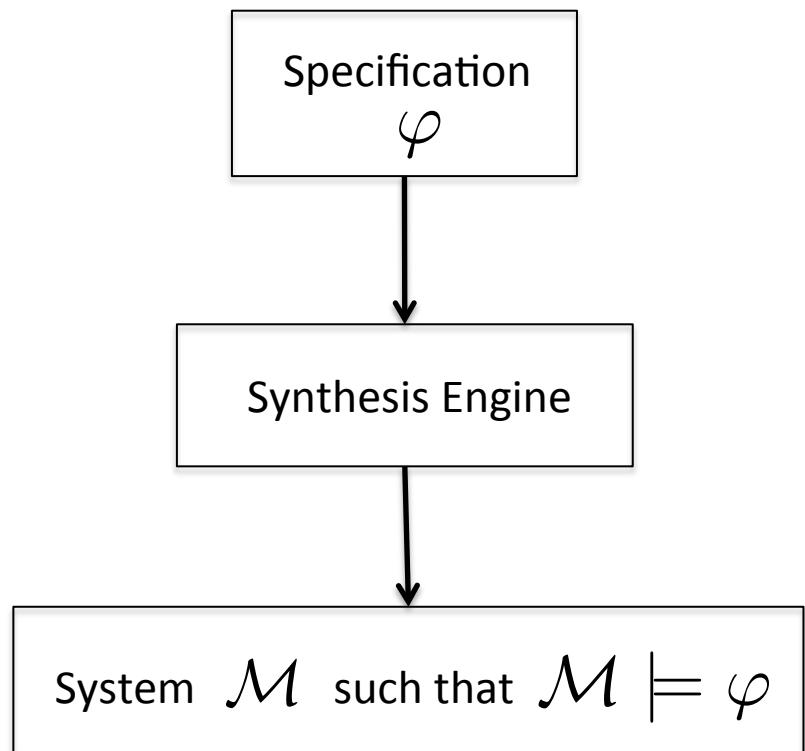
- Operate **autonomously**
- Fulfill **complex** requirements
- Easy to **specify** and **enforce** guarantees

Formal Methods: Two Perspectives

Verification

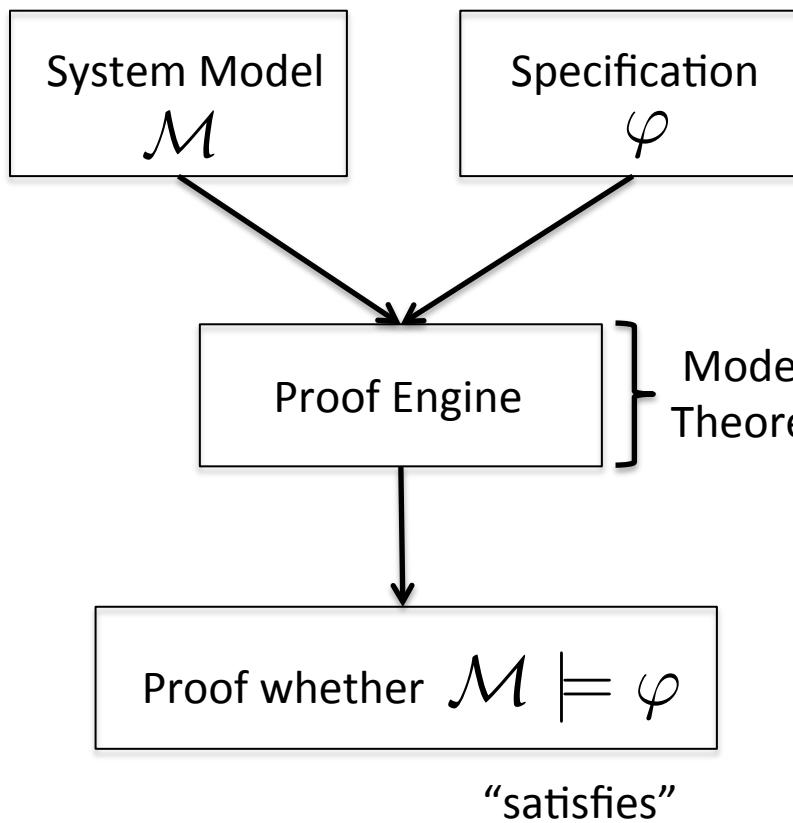


Synthesis

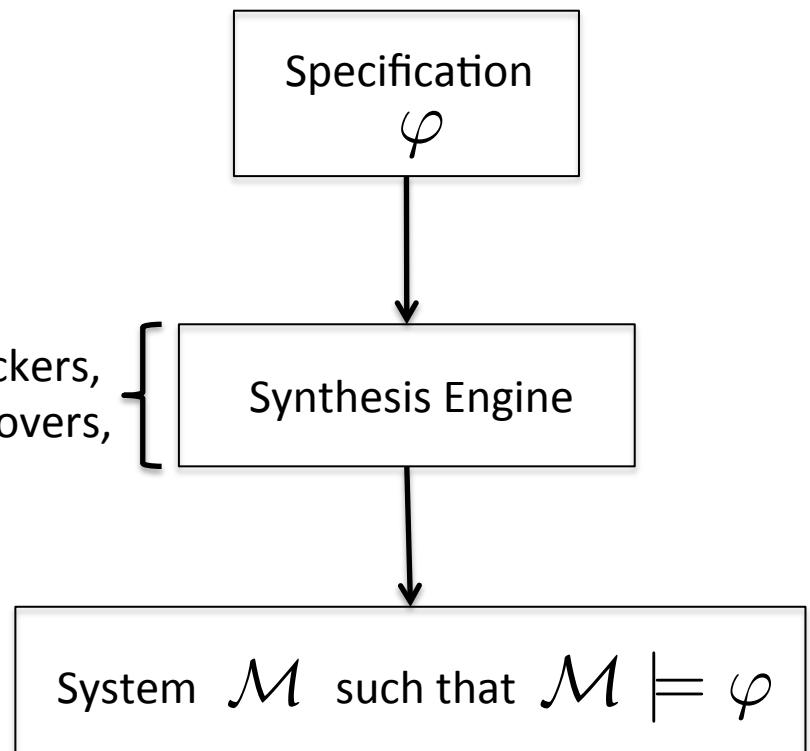


Formal Methods: Two Perspectives

Verification



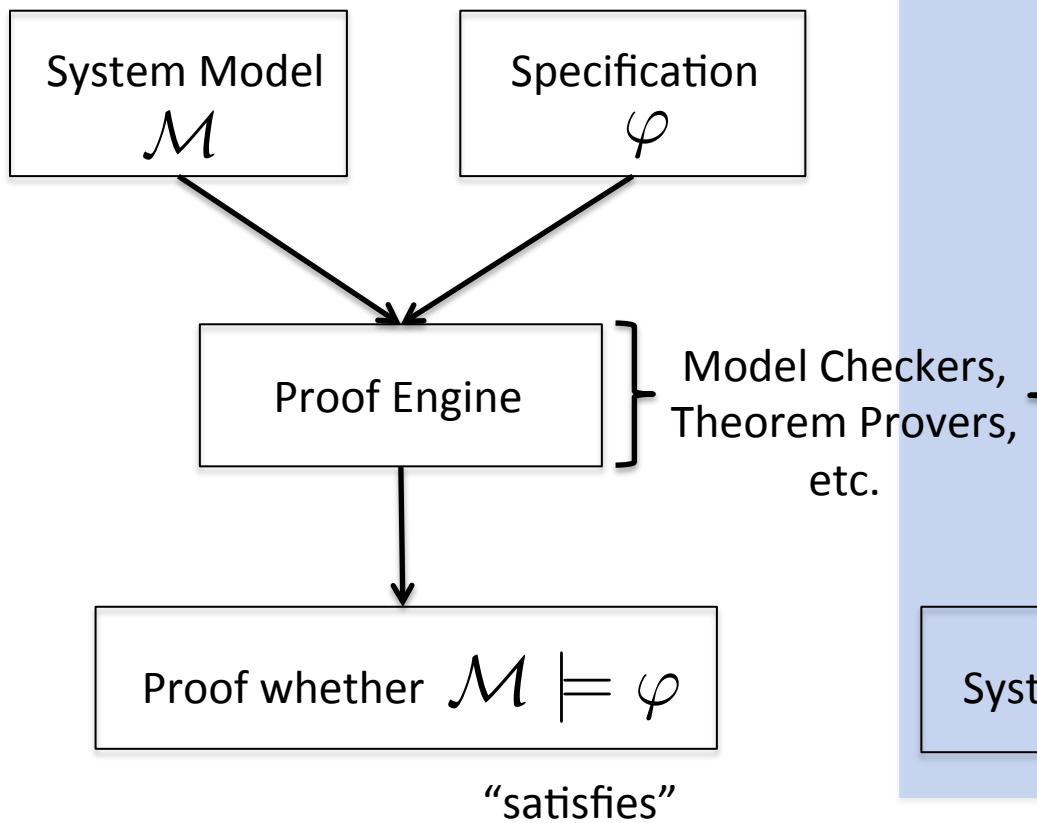
Synthesis



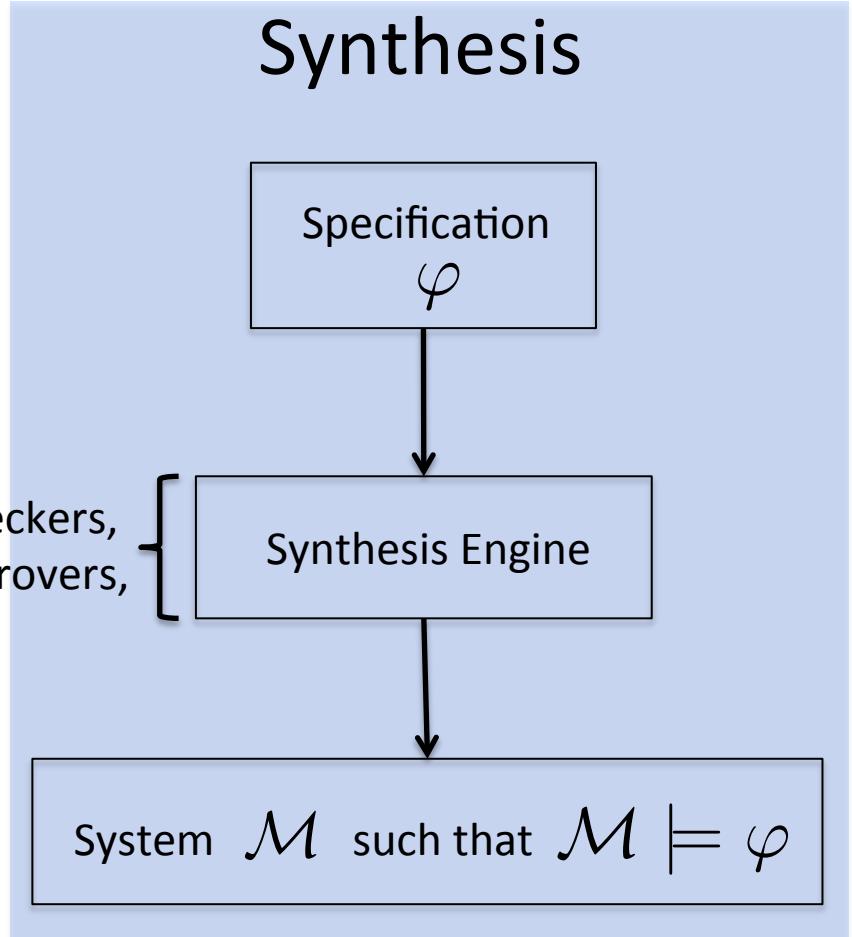
“satisfies”

Formal Methods: Two Perspectives

Verification

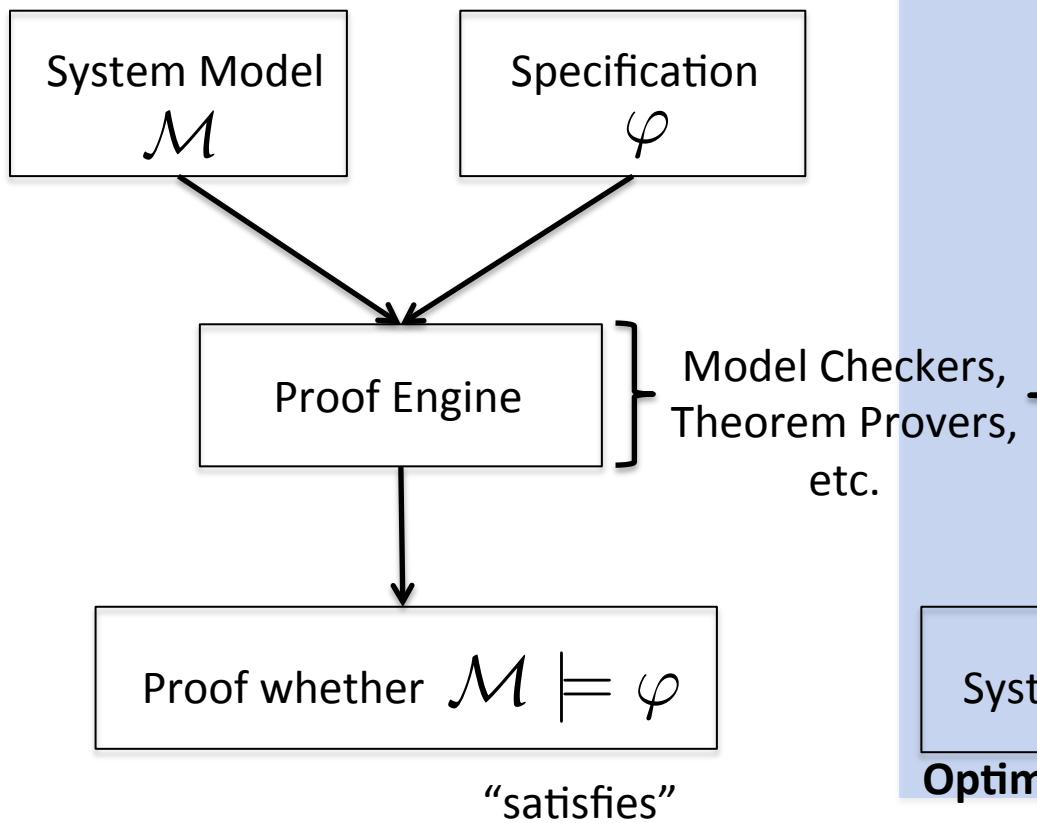


Synthesis

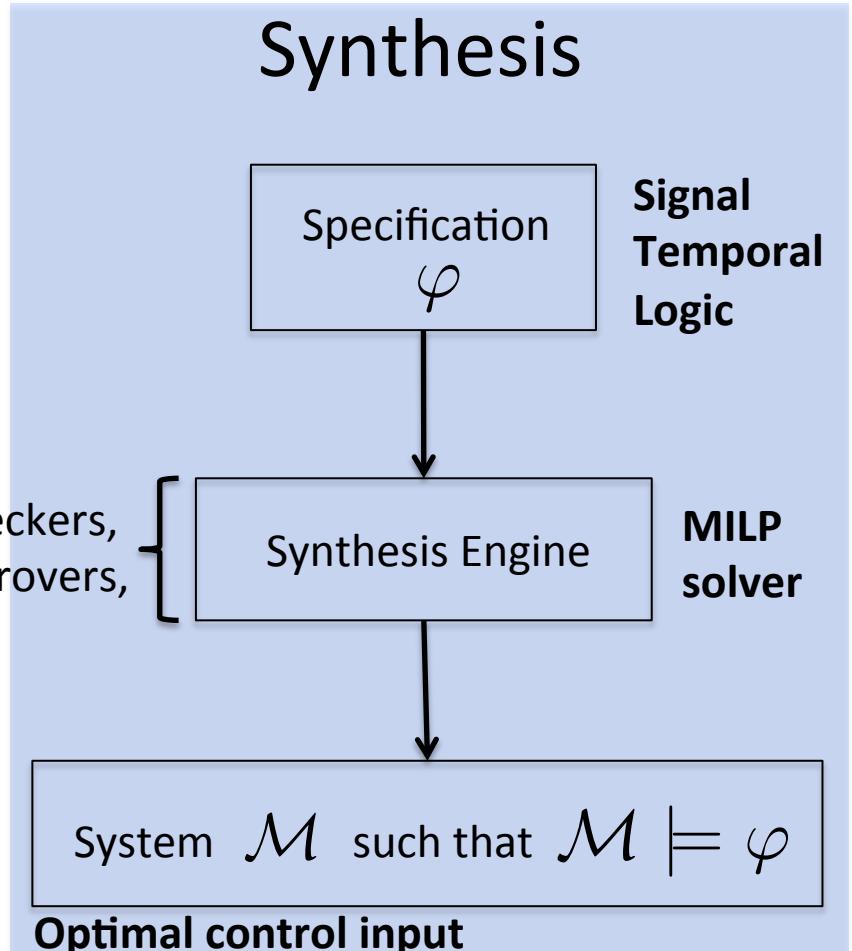


Formal Methods: Two Perspectives

Verification



Synthesis



Temporal Logic Synthesis for CPS (Related Work)

- Robotics
 - Kress-Gazit, Fainekos and Pappas, ICRA 2007
 - Kloetzer and Belta, TAC 2008
 - Karaman and Frazzoli, CDC 2009
 - Bhatia, Kavraki and Vardi, ICRA 2010
- Autonomous Cars
 - Wongpiromsarn, Topcu and Murray, HSCC 2010
- Aircraft Electric Power Systems
 - Nuzzo et al, IEEE Access 2013

Temporal Logic Synthesis for CPS (what is lacking?)

- Usually requires **discrete abstraction**
 - “If temperature falls below 20°C, get it back above 20°C in the next time step”

$$\square(T_{\text{less_than_20}} \implies \bigcirc(\neg T_{\text{less_than_20}}))$$

Temporal Logic Synthesis for CPS (what is lacking?)

- Temporal duration is often **cumbersome**
 - “Infinitely often visit A and no more than 5 time steps later visit B”

$$\square \diamond (A \wedge \bigcirc B \vee \bigcirc \bigcirc B \vee \bigcirc \bigcirc \bigcirc B \vee \bigcirc \bigcirc \bigcirc \bigcirc B \vee \bigcirc \bigcirc \bigcirc \bigcirc \bigcirc B)$$

- “All visits to A and B should be no more than 5.1s apart”

$$\square (A \implies \diamond (\text{clock_less_than_}5.1 \wedge B))$$

Signal Temporal Logic (STL)

- Continuous predicates: $\mu(\mathbf{x}) > 0$
- Boolean Operators: $\wedge, \vee, \implies, \neg$
- Bounded Temporal Operators:

$$\Box_{[a,b]} \varphi$$
$$\Diamond_{[a,b]} \varphi$$
$$\varphi_1 \mathcal{U}_{[a,b]} \varphi_2$$

φ holds at all $t \in [a, b]$ φ holds at some $t \in [a, b]$

- Synthesis undecidable for dense time
 - We'll restrict to discrete time (but continuous systems)

Signal Temporal Logic (STL)

Syntax

$$\varphi ::= \mu \mid \neg\mu \mid \varphi \wedge \psi \mid \varphi \vee \psi \mid \Box_{[a,b]} \psi \mid \varphi \mathcal{U}_{[a,b]} \psi$$
$$\mu \equiv \mu(\mathbf{x}) > 0$$

Semantics

$$(\mathbf{x}, t) \models \mu \iff \mu(\mathbf{x}(t)) > 0$$
$$(\mathbf{x}, t) \models \neg\mu \iff \neg((\mathbf{x}, t) \models \mu)$$
$$(\mathbf{x}, t) \models \varphi \wedge \psi \iff (\mathbf{x}, t) \models \varphi \wedge (\mathbf{x}, t) \models \psi$$
$$(\mathbf{x}, t) \models \varphi \vee \psi \iff (\mathbf{x}, t) \models \varphi \vee (\mathbf{x}, t) \models \psi$$
$$(\mathbf{x}, t) \models \Box_{[a,b]} \varphi \iff \forall t' \in [t + a, t + b], (\mathbf{x}, t') \models \varphi$$
$$(\mathbf{x}, t) \models \varphi \mathcal{U}_{[a,b]} \psi \iff \exists t' \in [t + a, t + b] \text{ s.t. } (\mathbf{x}, t') \models \psi \wedge \forall t'' \in [t, t'], (\mathbf{x}, t'') \models \varphi.$$

Examples

- If temperature falls below 20°C, get it back above 20°C within 5 time steps
 - ($T_{\text{less_than_20}} \implies \bigcirc(\neg T_{\text{less_than_20}})$)
- Infinitely often visit A and no more than five time steps later visit B
 - $\diamond(A \wedge \bigcirc B \vee \bigcirc \bigcirc B \vee \bigcirc \bigcirc \bigcirc B \vee \bigcirc \bigcirc \bigcirc \bigcirc B \vee \bigcirc \bigcirc \bigcirc \bigcirc \bigcirc B)$
- All visits to A and B should be no more than 5.1 seconds steps apart
 - ($A \implies \diamond(\text{clock_less_than_5.1} \wedge B)$)

Examples

- If temperature falls below 20°C, get it back above 20°C within 5 time steps
$$\Box(T < 20 \implies \Diamond_{[0,5]}(T > 20))$$
- Infinitely often visit A and no more than five time steps later visit B
$$\Box \Diamond(A \wedge \Diamond_{[0,5]} B)$$
- All visits to A and B should be no more than 5.1 seconds steps apart
$$\Box(A \implies \Diamond_{[0,5.1]} B)$$

Optimal Control Synthesis from STL

Given:

Discrete time continuous system $x_{t+1} = f(x_t, u_t)$

STL specification φ

Initial state x_0

Cost function J on runs of the system

Compute:

$$\begin{aligned} \arg \min_{\mathbf{u}} \quad & J(\mathbf{x}(x_0, \mathbf{u}), \mathbf{u}) \\ \text{s.t. } & \mathbf{x}(x_0, \mathbf{u}) \models \varphi \end{aligned}$$

Model Predictive Control from STL

Given:

Discrete time continuous system $x_{t+1} = f(x_t, u_t)$

STL specification φ

Initial state x_0

Cost function J on runs of the system

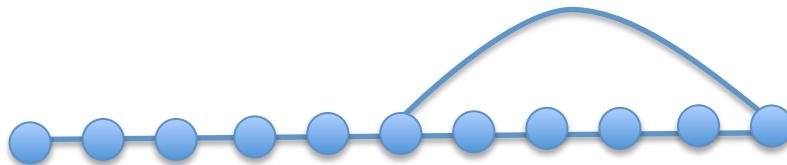
Horizon H

Compute:

$$\begin{aligned} \arg \min_{\mathbf{u}_t^H} \quad & J(\mathbf{x}^H(x_t, \mathbf{u}_t^H), \mathbf{u}_t^H) \\ \text{s.t. } & \mathbf{x}(x_0, \mathbf{u}) \models \varphi, \end{aligned}$$

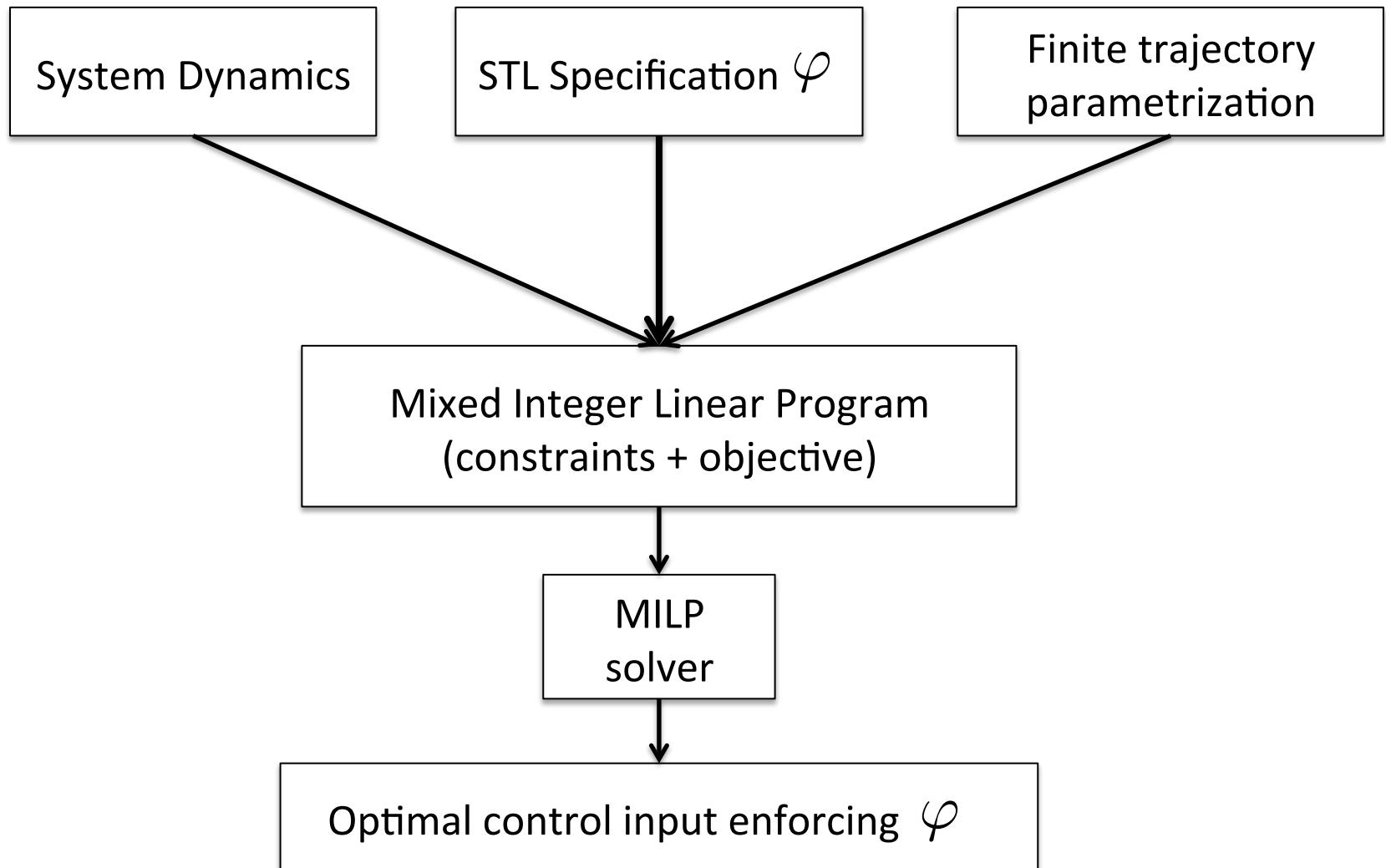
Finite Trajectory Parametrization

- Lasso-shaped parametrization for infinite executions

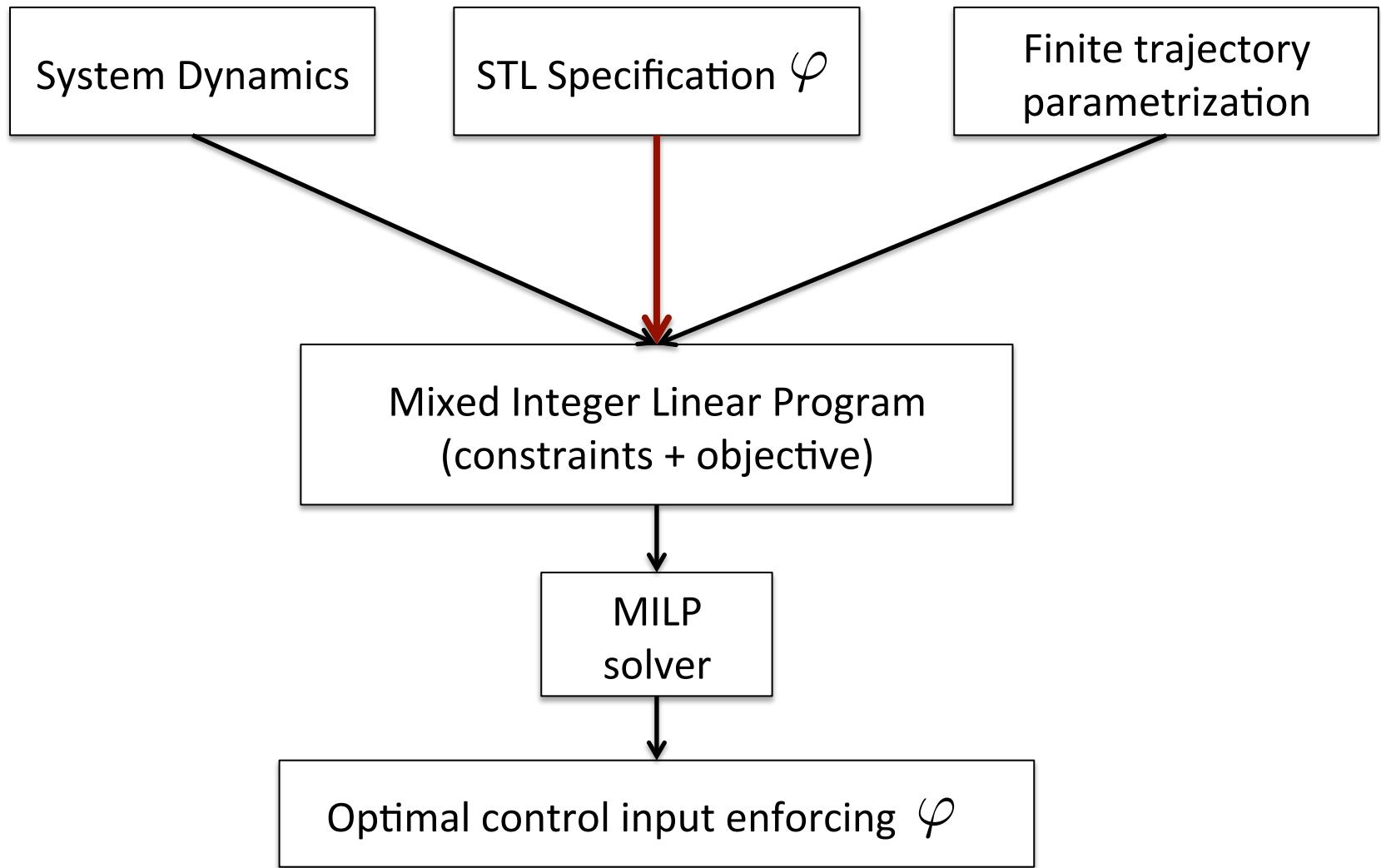


- Common approach in Bounded Model Checking

STL Synthesis for Control (Overview)



STL Synthesis for Control (Overview)



STL to MILP constraints

Given a formula ψ with subformulas denoted by φ

Introduce

$$z_t^\varphi$$

Constrained
such that

$$z_t^\varphi = 1 \Leftrightarrow (\mathbf{x}, t) \models \varphi$$

Enforce

$$z_0^\psi = 1$$

Recursively generate the MILP constraints corresponding to z_0^ψ .

STL to MILP constraints

Given a formula ψ with subformulas denoted by φ

Predicates

$$\begin{aligned}\mu(x_t) &\leq M_t z_t^\mu + \epsilon_t \\ -\mu(x_t) &\leq M_t(1 - z_t^\mu) - \epsilon_t\end{aligned}$$

Conjunction

$$\begin{aligned}z_t^\psi &\leq z_t^{\varphi_i}, i = 1, \dots, m, \\ z_t^\psi &\geq 1 - m + \sum_{i=1}^m z_t^{\varphi_i}\end{aligned}$$

Disjunction

$$\begin{aligned}z_t^\psi &\geq z_t^{\varphi_i}, i = 1, \dots, m, \\ z_t^\psi &\leq \sum_{i=1}^m z_t^{\varphi_i}\end{aligned}$$

STL to MILP constraints

Given a formula ψ with subformulas denoted by φ

Always

$$\psi = \square_{[a,b]} \varphi$$

$$a_t^N = \min(t + a, N), \quad b_t^N = \min(t + b, N)$$
$$z_t^\psi = \vee_{i=a_t^N}^{b_t^N} z_i^\varphi \wedge (\vee_{j=1}^N l_j \wedge \wedge_{i=j+\hat{a}_t^N}^{j+\hat{b}_t^N} z_i^\varphi)$$

Eventually

$$\psi = \diamondsuit_{[a,b]} \varphi$$

$$z_t^\psi = \wedge_{i=a_t^N}^{b_t^N} z_i^\varphi \wedge (\vee_{j=1}^N l_j \wedge \wedge_{i=j+\hat{a}_t^N}^{j+\hat{b}_t^N} z_i^\varphi)$$

Until

$$\psi = \varphi_1 \mathcal{U}_{[a,b]} \varphi_2$$

$$\varphi_1 \mathcal{U}_{[a,b]} \varphi_2 = \square_{[0,a]} \varphi_1 \wedge \diamondsuit_{[a,b]} \varphi_2$$
$$\wedge \diamondsuit_{[a,a]}(\varphi_1 \mathcal{U} \varphi_2)$$

Quantitative Semantics for STL

- How much can we **vary the signal** and still satisfy φ
- Robustness function $\rho^\varphi : \mathcal{X} \times \mathbb{N} \rightarrow \mathbb{R}$
 $(\mathbf{x}, t) \models \varphi \equiv \rho^\varphi(\mathbf{x}, t) > 0$

$$\begin{aligned}\rho^\mu(\mathbf{x}, t) &= \mu(\mathbf{x}(t)) \\ \rho^{\neg\mu}(\mathbf{x}, t) &= -\mu(\mathbf{x}(t)) \\ \rho^{\varphi \wedge \psi}(\mathbf{x}, t) &= \min(\rho^\varphi(\mathbf{x}, t), \rho^\psi(\mathbf{x}, t)) \\ \rho^{\varphi \vee \psi}(\mathbf{x}, t) &= \max(\rho^\varphi(\mathbf{x}, t), \rho^\psi(\mathbf{x}, t)) \\ \rho^{\square_{[a,b]} \varphi}(\mathbf{x}, t) &= \min_{t' \in [t+a, t+b]} \rho^\varphi(\mathbf{x}, t') \\ \rho^{\varphi \text{ } \mathcal{U}_{[a,b]} \text{ } \psi}(\mathbf{x}, t) &= \max_{t' \in [t+a, t+b]} (\min(\rho^\psi(\mathbf{x}, t'), \\ &\quad \min_{t'' \in [t, t']} \rho^\varphi(\mathbf{x}, t'')))\end{aligned}$$

Quantitative Semantics for STL

- How much can we **vary the signal** and still satisfy φ
- Robustness function $\rho^\varphi : \mathcal{X} \times \mathbb{N} \rightarrow \mathbb{R}$
 $(\mathbf{x}, t) \models \varphi \equiv \rho^\varphi(\mathbf{x}, t) > 0$
- Examples: $\mu_1 \equiv x - 3 > 0 \quad \varphi = \square_{[0,2]} \mu_1$

$$\rho^{\mu_1}(x, 0) = x(0) - 3$$

$$\rho^{\mu_1 \wedge \mu_2}(x, t) = \min(\rho^{\mu_1}, \rho^{\mu_2})$$

$$\rho^\varphi(x, t) = \min_{t \in [0,2]} \rho^{\mu_1}(x, t) = \min_{t \in [0,2]} x(t) - 3$$

Maximally Robust Synthesis from STL

Given:

Discrete time continuous system $x_{t+1} = f(x_t, u_t)$

STL specification φ

Initial state x_0

Robustness function $\rho^\varphi : \mathcal{X} \times \mathbb{N} \rightarrow \mathbb{R}$

Compute:

$$\begin{aligned} & \arg \max_{\mathbf{u}} \quad \rho^\varphi(x_0, 0) \\ & \text{s.t. } \mathbf{x}(x_0, \mathbf{u}) \models \varphi \end{aligned}$$

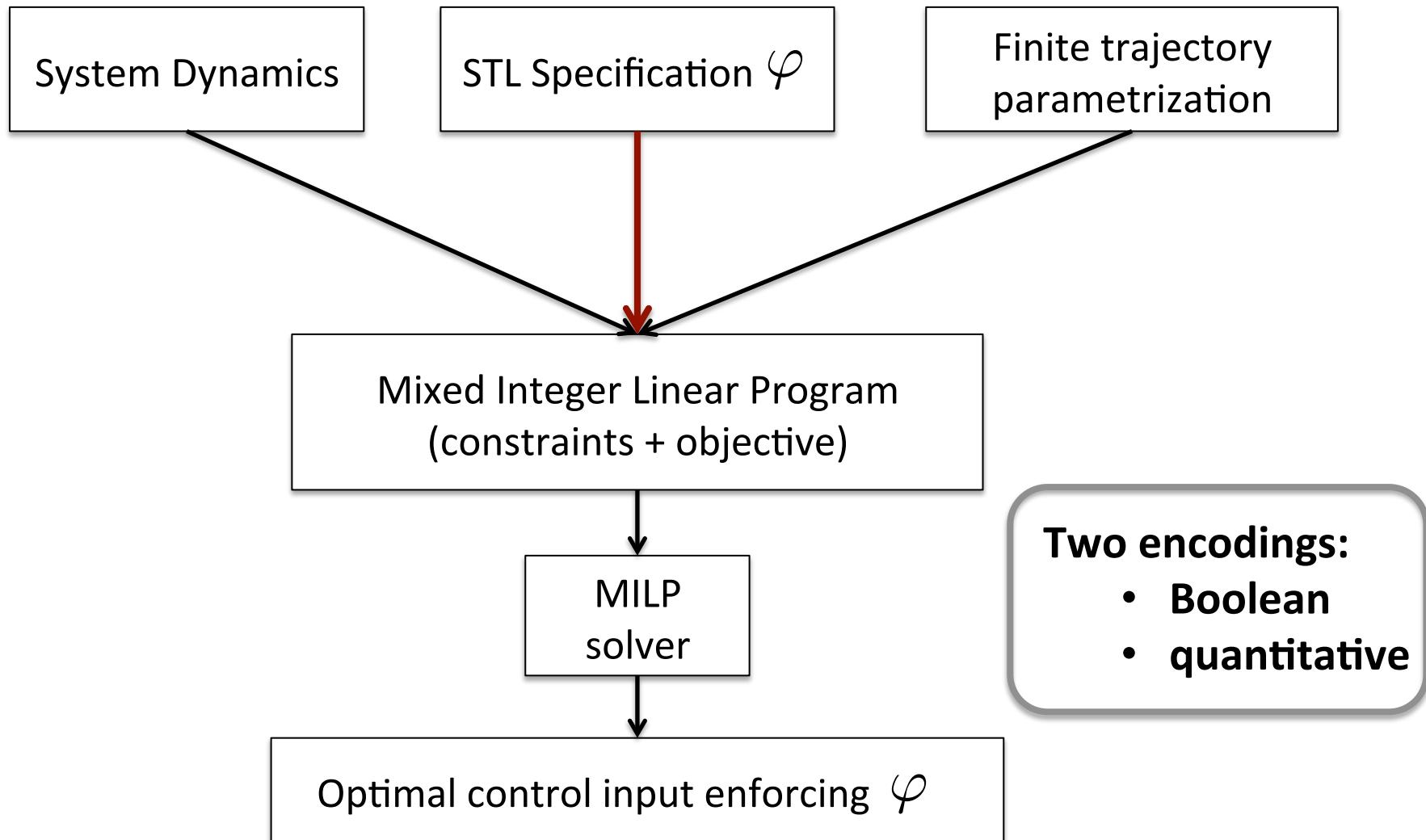
STL to MILP constraints

Given a formula ψ with subformulas denoted by φ

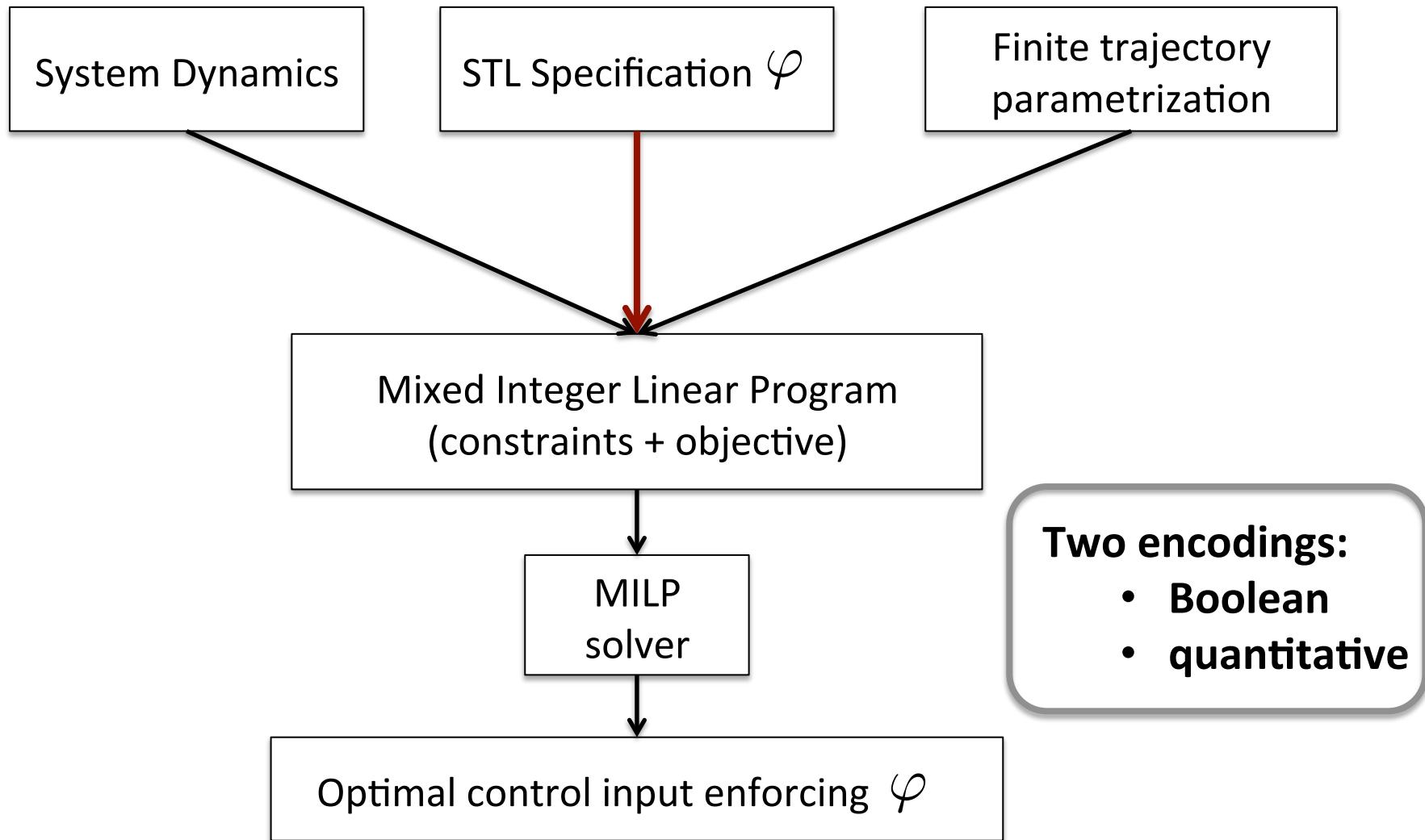
	Boolean encoding	Robustness encoding
Introduce	z_t^φ	r_t^φ
Constrained such that	$z_t^\varphi = 1 \Leftrightarrow (\mathbf{x}, t) \models \varphi$	$r_t^\varphi > 0 \Leftrightarrow (\mathbf{x}, t) \models \varphi$ In fact, $r_t^\varphi = \rho^\varphi(\mathbf{x}, t)$
Enforce	$z_0^\psi = 1$	$r_0^\psi > 0$

Recursively generate the MILP constraints corresponding to z_0^ψ or r_0^ψ

STL Synthesis for Control (Overview)



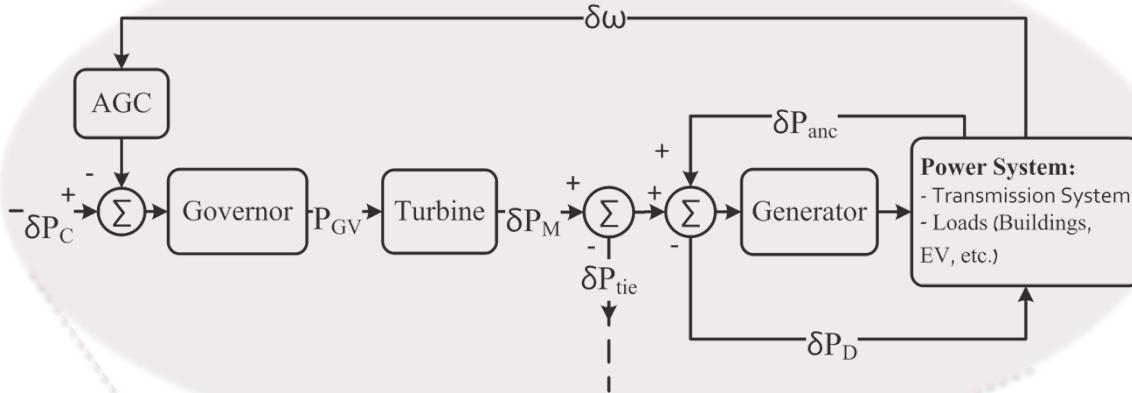
STL Synthesis for Control (Overview)



MPC/Receding Horizon Control (for bounded formulas)

- Pick H based on φ
 - conservative bound on trajectory length to decide satisfiability
 - e.g. for $\square_{[0,10]} \diamondsuit_{[1,6]} \varphi$ use $H \geq 10 + 6 = 16$
- Open-loop synthesis at each time step
 - STL constraints apply on the length- H prefix
- Store history of states and inputs
 - ensures φ is satisfied over the length- H prefix
- Extends to certain unbounded formulas
 - e.g. $\varphi = \square(\varphi_{MPC})$ for bounded φ_{MPC} .

Example: Grid regulation

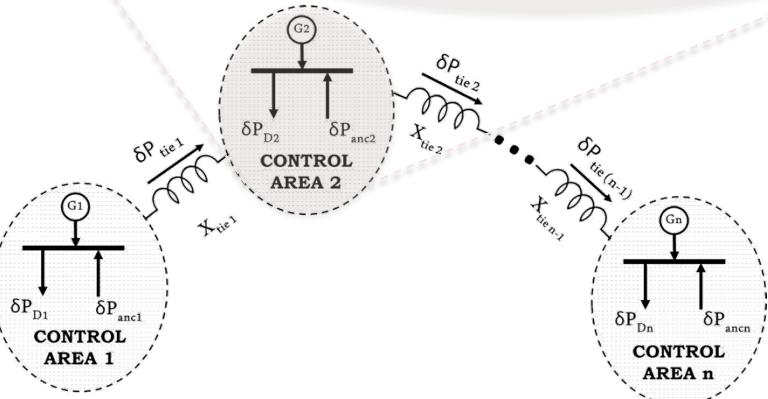


Controlling ancillary service power flow for grid frequency regulation

Minimize control input
subject to

"If the Area Control Error (ACE) increases above 0.01, it will decrease below 0.01 within τ time steps"

$$\varphi_t = \neg(|\text{ACE}^1| < .01) \Rightarrow (\Diamond_{[0,\tau]}(|\text{ACE}^1| < .01) \\ \wedge \neg(|\text{ACE}^2| < .01)) \Rightarrow (\Diamond_{[0,\tau]}(|\text{ACE}^2| < .01))$$



Example: Grid regulation

$$\min_{U_{\text{anc}}[k]}$$

$$J(\text{ACE}, U_{\text{anc}}) + \|x[k+H] - x_{\text{ref}}\|_Q$$

s.t.

$$x[k+j+1] =$$

$$Ax[k+j] + B_2 u_{\text{anc}}[k+j] + Ed[k+j]$$

Dynamics

$$\underline{u}_{\text{anc}} \leq u_{\text{anc}}[k+j] \leq \bar{u}_{\text{anc}}$$

$$|u_{\text{anc}}[k+j+1] - u_{\text{anc}}[k+j]| \leq \lambda$$

$$x[k+H] \in \mathcal{X}[H]$$

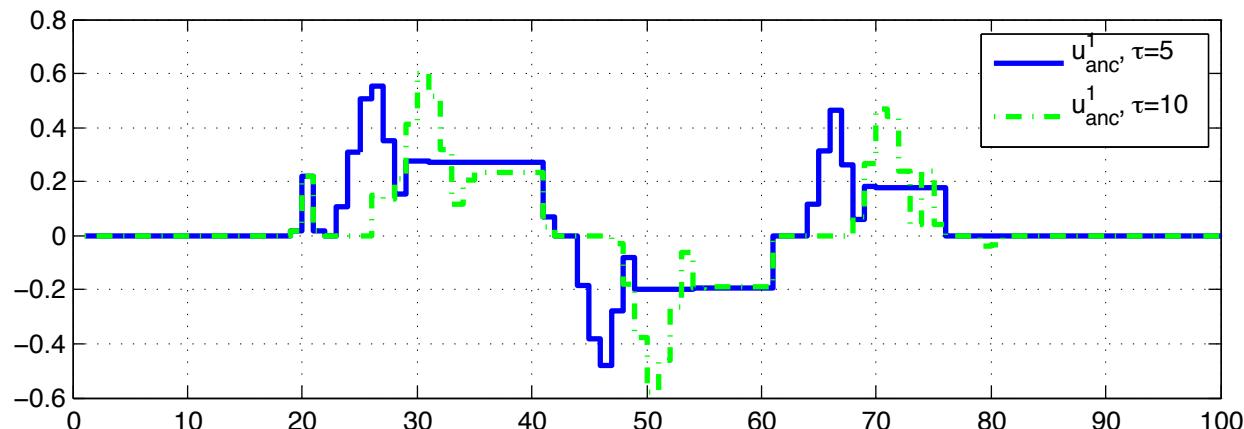
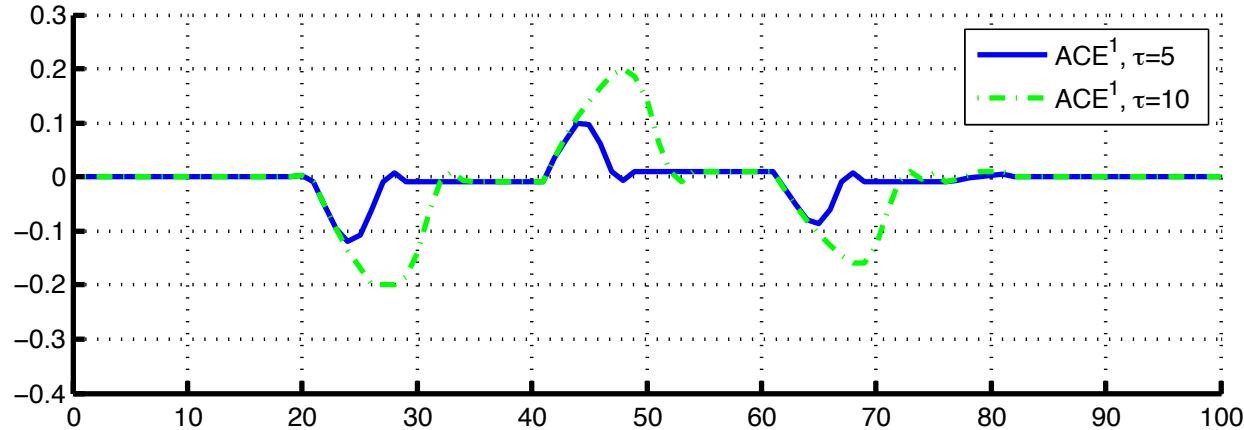
$$x[k] \models \varphi \quad \text{Specification}$$

$$J(\text{ACE}, U_{\text{anc}}) = \|U_{\text{anc}}\|_{\ell_2} = \sum_{i=1}^2 \sum_{j=0}^{H-1} (U_{\text{anc}}^i[k+j])^2$$

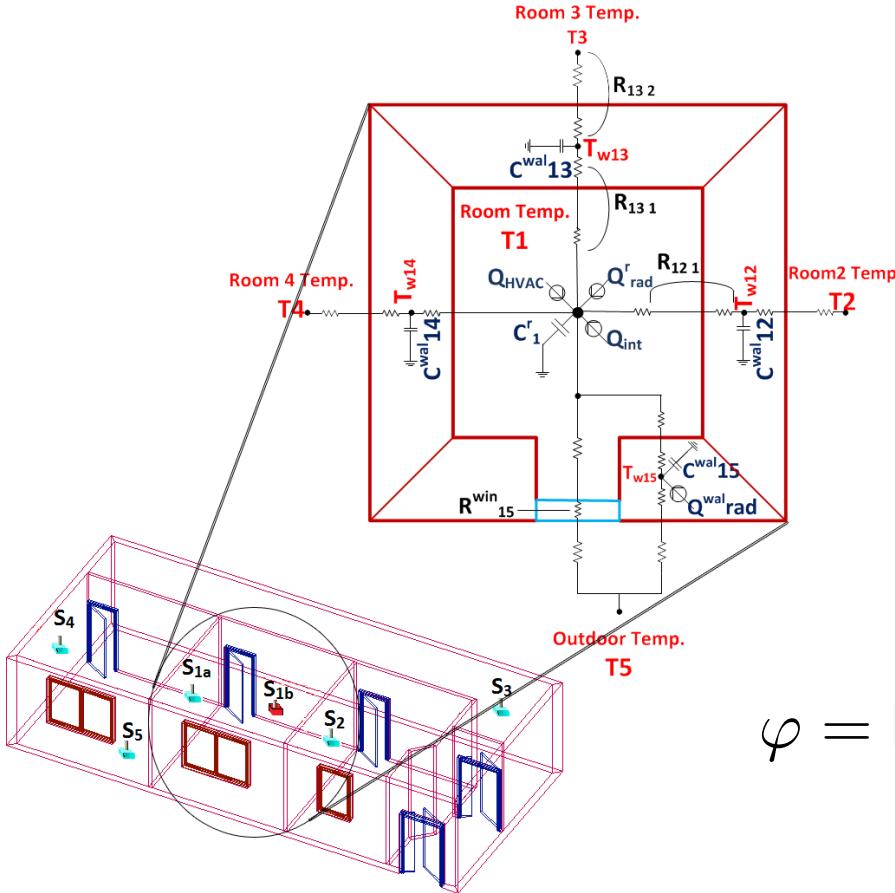
$$\varphi = \square(\varphi_t)$$

$$\begin{aligned} \varphi_t = & \neg(|\text{ACE}^1| < .01)) \Rightarrow (\diamondsuit_{[0,\tau]}(|\text{ACE}^1| < .01) \\ & \wedge (\neg(|\text{ACE}^2| < .01)) \Rightarrow (\diamondsuit_{[0,\tau]}(|\text{ACE}^2| < .01) \end{aligned}$$

Example: Grid regulation



Example: HVAC system



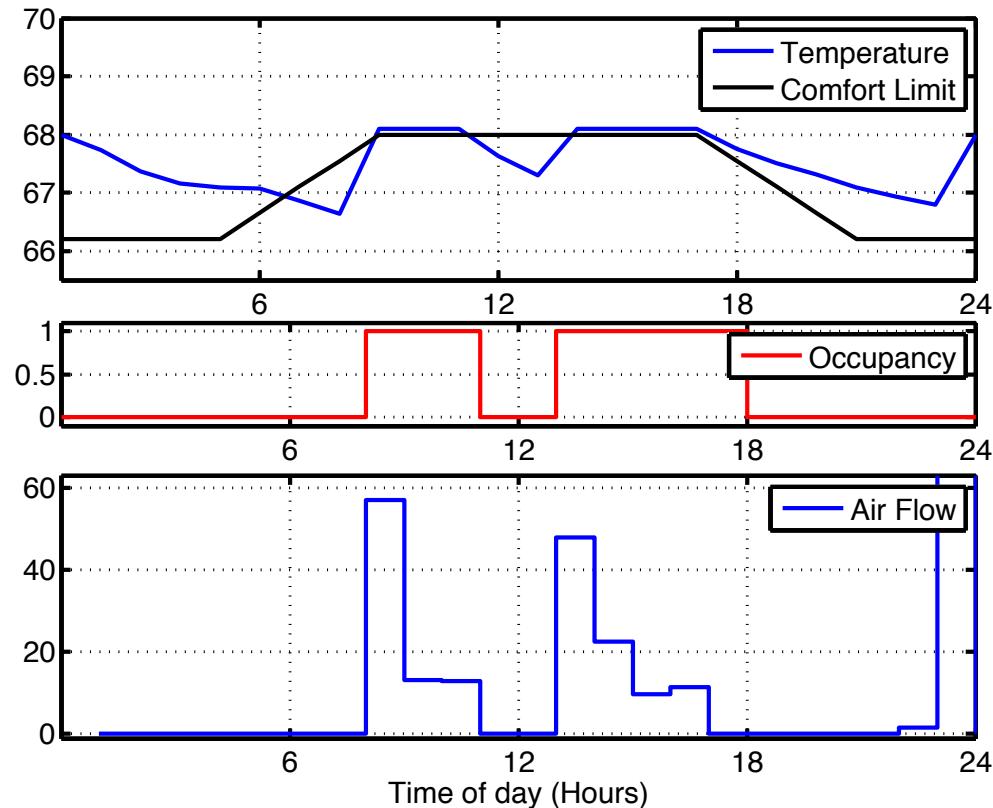
Minimize the input (total air flow)
subject to

*"If the occupancy of a room is > 0 ,
the temperature should be above
the comfort level"*

$$\varphi = \square_{[0,H]}((\text{occ}_t > 0) \Rightarrow (T_t > T_t^{\text{conf}}))$$

Example: HVAC system

$$\varphi = \square_{[0,H]} ((\text{occ}_t > 0) \Rightarrow (T_t > T_t^{\text{conf}}))$$



$$\min_{\vec{u}_t} \sum_{k=0}^{H-1} \|u_{t+k}\| \quad \text{s.t.}$$

$$x_{t+k+1} = f(x_{t+k}, u_{t+k}, d_{t+k}),$$

$$x_t \models \varphi$$

$$u_{t+k} \in \mathcal{U}_{t+k}, \quad k = 0, \dots, H - 1$$

Future Work



- Receding Horizon framework for unbounded STL properties
 - ties to online monitoring of STL properties
 - formalize connection with reactive synthesis
- Contract-based framework for specifying and designing components (e.g. of the smart-grid) and their interactions

Thank You!

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