

Parametric Identification of Temporal Properties

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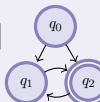
Context

- ▶ Properties monitoring of hybrid systems (embedded systems, mixed-signal circuits, etc)
- ▶ Behaviours have dense-time and real-valued components
- ▶ We specify and monitor properties using Signal Temporal Logic

In this work, we consider *Parametric STL* formulas, i.e., formulas with parameters

Hybrid System

$$\dot{x} = f_q(x, t) \parallel$$



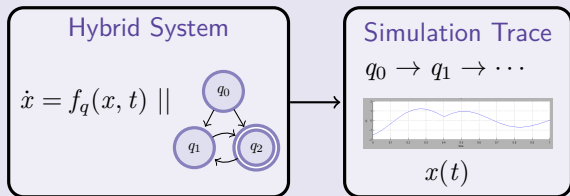
Goal: solving an inverse problem

For which values of parameters (e.g., p, s_1, s_2) does a given signal x satisfies φ ?

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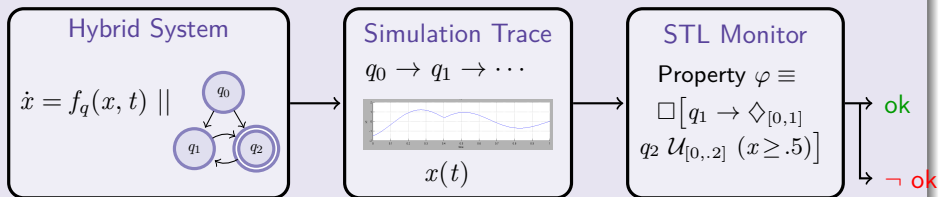
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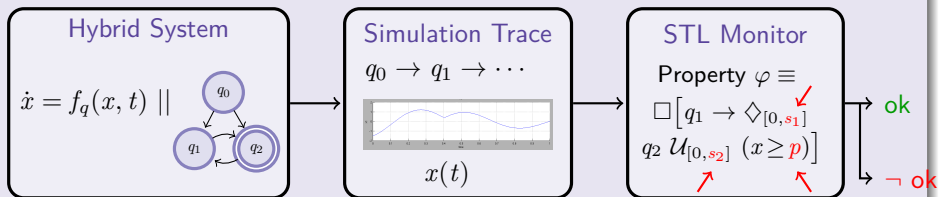
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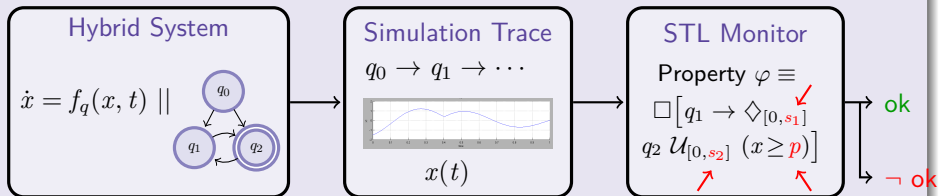
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Outline

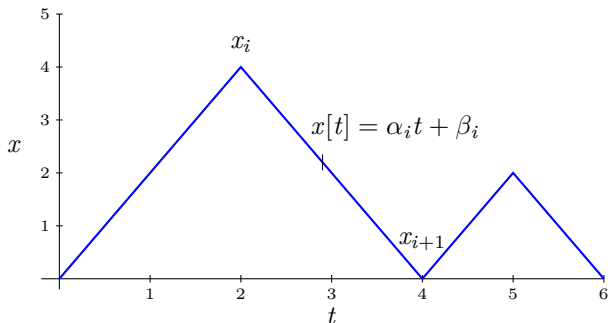
- 1 Parametric Signal Temporal Logic
- 2 Exact Computation of Validity Domains
- 3 Numerical Approximation of Validity Domains

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Signals

We consider piecewise affine signals given by sets of points ($x_i = x[t_i]$)



In this talk, we consider 1-dimensional signals for simplicity.

Signal Temporal Logic

Goal specifying temporal properties of signals

Definition (STL Syntax)

An STL formula is given by:

$$\varphi := \mu \mid \neg\varphi \mid \varphi_1 \wedge \varphi_2 \mid \varphi_1 \mathcal{U}_{[a,b]} \varphi_2$$

where μ is a predicate of the form $f(x) > p$.

Definition (STL Semantics)

The validity of a formula φ with respect to a signal x at time t is

$$\begin{aligned} (x, t) \models f(x) > p &\Leftrightarrow f(x[t]) > p \\ (x, t) \models \varphi \wedge \psi &\Leftrightarrow (x, t) \models \varphi \wedge (x, t) \models \psi \\ (x, t) \models \neg\varphi &\Leftrightarrow \neg((x, t) \models \varphi) \\ (x, t) \models \varphi \mathcal{U}_{[a,b]} \psi &\Leftrightarrow \exists t' \in [t+a, t+b] \text{ s.t. } (x, t') \models \psi \wedge \\ &\quad \forall t'' \in [t, t'], (x, t'') \models \varphi \end{aligned}$$

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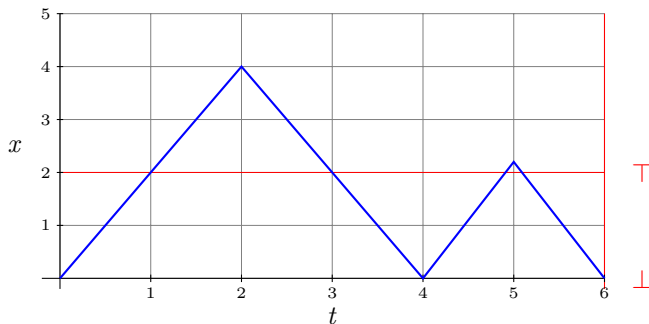
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Additionally: $\diamond_{[a,b]}\varphi = \top \mathcal{U}_{[a,b]}\varphi$ and $\square_{[a,b]}\varphi = \varphi \mathcal{U}_{[a,b]}\perp$.

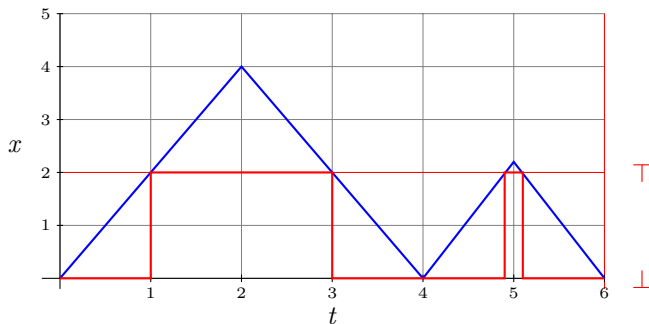
Examples



Truth value of :

▶ $\varphi = x > 2$

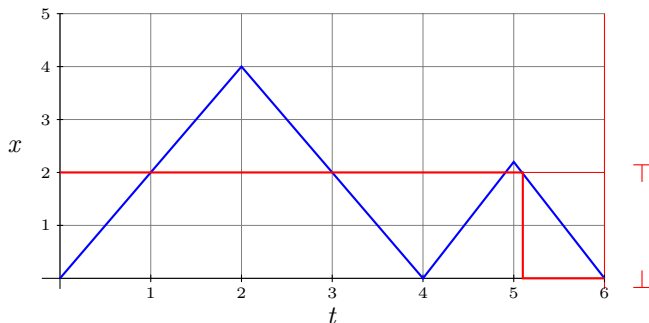
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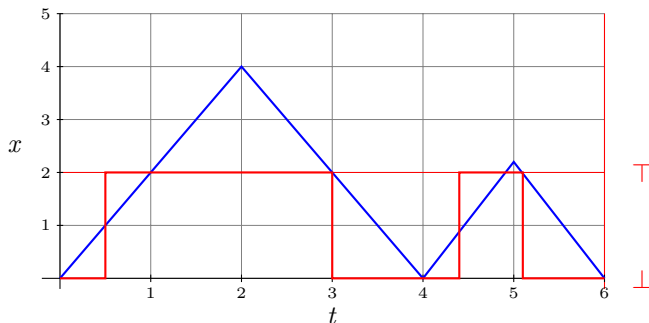


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▶ $\varphi = \diamond_{[0,\infty]}(x > 2)$

Examples

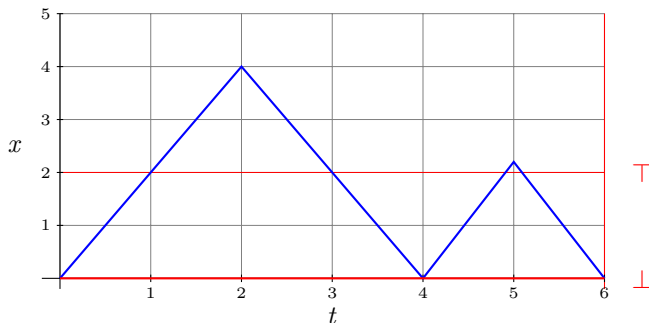


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Examples

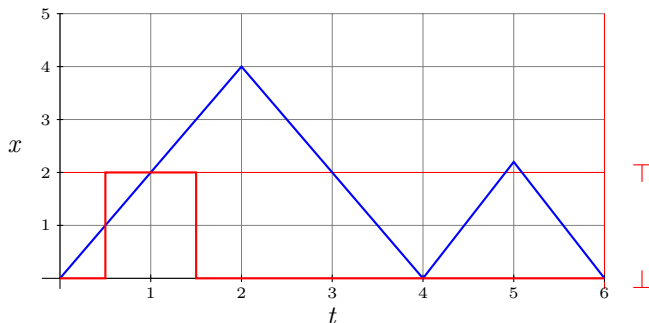


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▶ $\varphi = \square_{[0.5, 1.5]}(x > 2)$

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Validity Domains

Definition

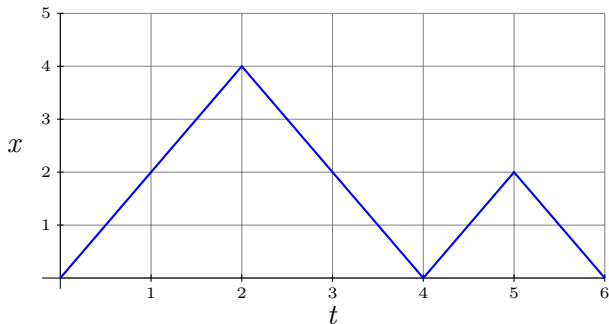
Let φ be a PSTL formula with amplitude parameters $p \in \mathcal{P}$ and timing parameters $s \in \mathcal{S}$.

The validity domain of a signal x for φ is the set of times and parameters values for which φ is satisfied by x

$$d(x, \varphi) = \{(p, s, t) \in \mathcal{P} \times \mathcal{S} \times \mathbb{R}^+ \text{ s.t. } (x, t) \models \varphi_{p,s}\}$$

Validity Domains of Predicates

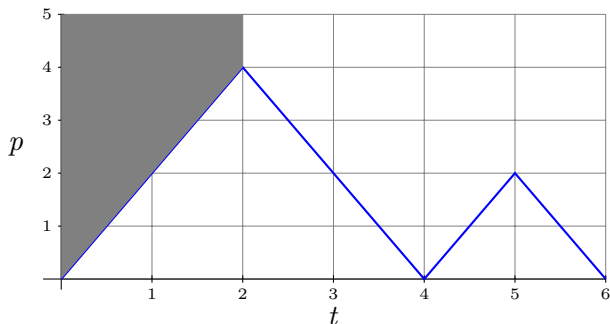
Consider again signal x and the predicate $x > p$.



$$\begin{aligned}d(x, x > p) = & (t \geq 0 \wedge t < 2 \wedge 2p > 4t) && \vee \\ & (t \geq 2 \wedge t < 4 \wedge 2p + 4t > 16) && \vee \\ & (t \geq 4 \wedge t < 5 \wedge p > 2t - 8) && \vee \\ & (t \geq 5 \wedge t < 6 \wedge p + 2t > 12)\end{aligned}$$

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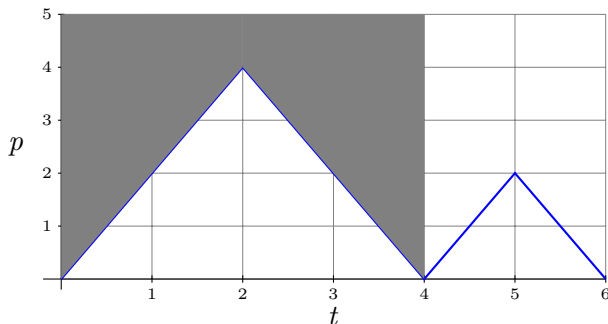
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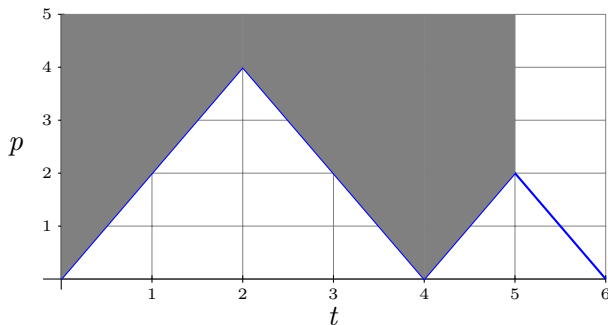
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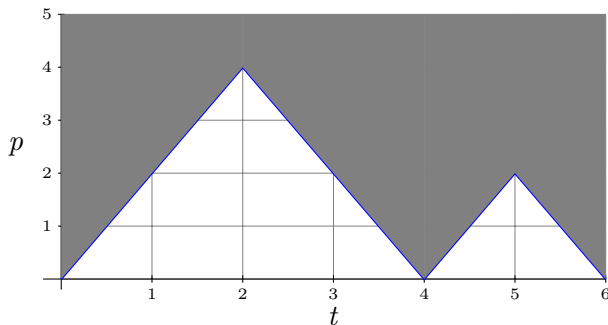
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Inductive Computation of Validity Domains

Validity domains can be computed inductively as follows:

$$\begin{aligned}d(x, f(x) > p) &= \{(t, p) : f(x[t]) > p\} \\d(x, \varphi \wedge \psi) &= d(x, \varphi) \cap d(x, \psi) \\d(x, \neg\varphi) &= \overline{d(x, \varphi)} \\d(x, \varphi \mathcal{U}_{[a,b]}\psi) &= \{(t, p, s) : \exists t' \in [t + a, t + b] \text{ s.t. } (t', p, s) \in d(x, \psi) \wedge \\ &\quad \forall t'' \in [t, t'](t'', p, s) \in d(x, \varphi)\}\end{aligned}$$

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Example

Validity domains of $\varphi = \Box_{[0,s_1]}(x < p)$?

$$\begin{aligned}d(x, \Box_{[0,s_1]}(x < p)) &= \{(t, s_1, p) \text{ s.t. } (x, t) \models \Box_{[0,s_1]}(x < p)\} \\ &= \{(t, s_1, p) \text{ s.t. } \forall \tau, t \leq \tau \leq t + s_1 \implies (\tau, p) \in d(x, x < p)\}\end{aligned}$$

Quantifier-free (QF) formula, knowing a QF formula for $d(x, x < p)$:

$$\begin{aligned}d(x, \Box_{[0,s_1]}(x < p)) &= QE(\forall \tau, t \leq \tau \leq t + s_1 \implies (\tau \geq 0 \wedge \tau < 2 \wedge 2p > 4\tau) \vee \dots) \\ &= p + 2s_1 + 2t < 12 \vee p + 2t > 12 \vee p > 0 \vee p \leq 0) \wedge \\ &\quad (p + 2s_1 + 2t < 8 \vee p + 2t > 8 \vee p + 4 \leq 0 \vee p > 4) \wedge \\ &\quad (s_1 + t \geq 6 \vee (p - 2s_1 - 2t > 0 \wedge s_1 + t < 2) \vee \\ &\quad \dots)\end{aligned}$$

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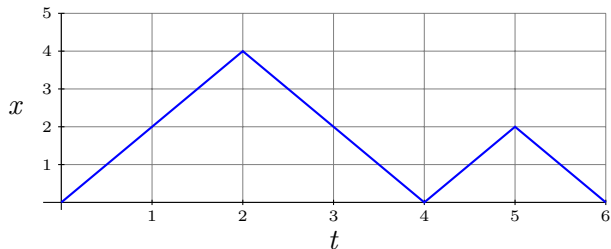
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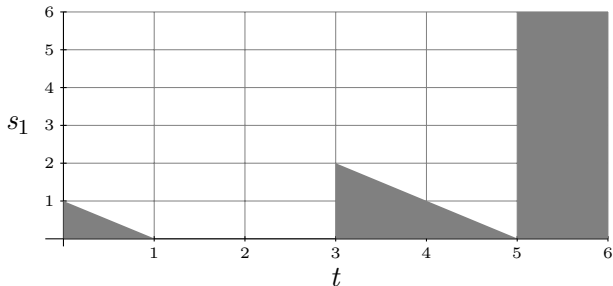
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Example



Validity domain $d(x, \square_{[0,s_1]}(x < 2))$



Experimental Results

Typical stabilization property

$$\varphi_{st} : \Box((x > p) \rightarrow \Diamond_{[0,s_2]}\Box_{[0,s_1]}(x < p)).$$

which we decompose into subformulas

$$\varphi_1 : \Box_{[0,s_1]}(x < p)$$

$$\varphi_2 : \Diamond_{[0,s_2]}\Box_{[0,s_1]}(x < p)$$

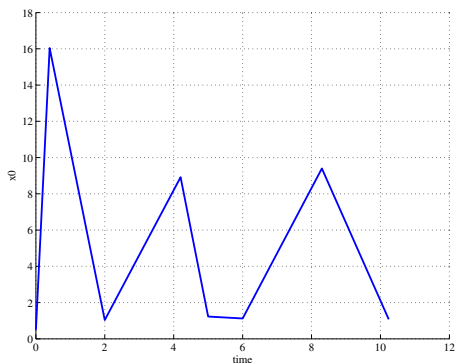
$$\varphi_3 : (x \geq p) \rightarrow \Diamond_{[0,s_2]}\Box_{[0,s_1]}(x < p)$$

We tested the computation of validity domains using QE for a signal with different number of samples.

Experimental Results

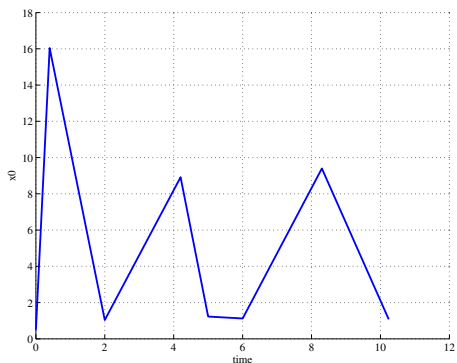
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8	0.02	38	0.11	197	0.17	207	3	4219
16	0.10	66	0.81	855	0.74	375	83.79	37709
32	0.26	86	19.07	6553	18.27	2885	*	*
64	4.16	144	341.95	23103	308.93	10258	*	*
128	68.29	895	*	*	*	*	*	*
256	386.72	3098	*	*	*	*	*	*

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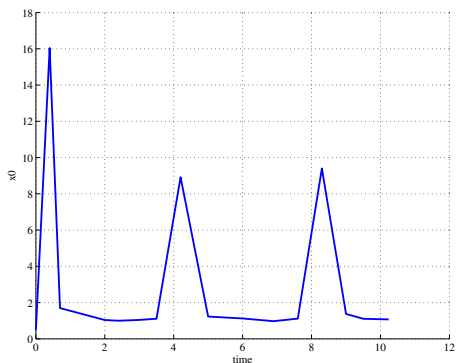
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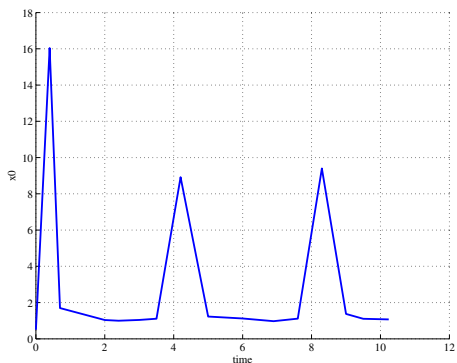
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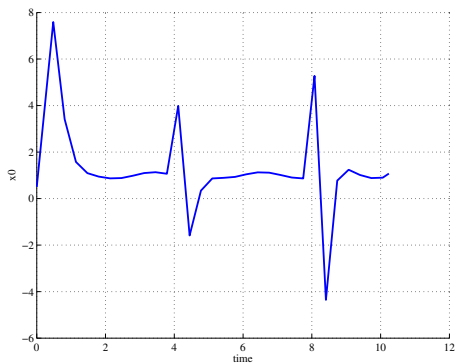
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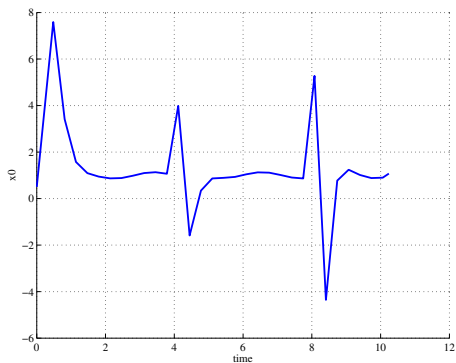
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256	386.72	3098	*	*	*	*	*	*

Experimental Results



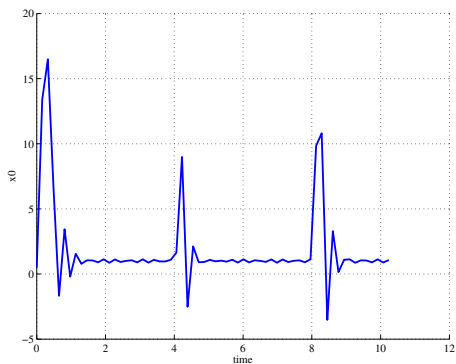
k	φ_1		φ_2		φ_3		φ_{st}	
	time(s)	size	time(s)	size	time(s)	size	time(s)	size
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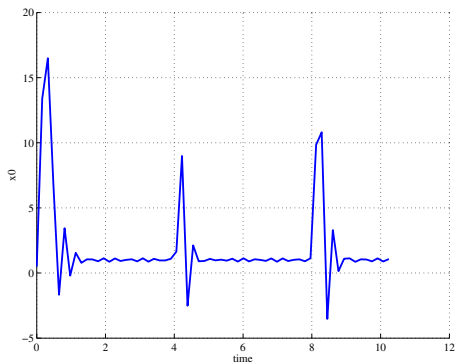
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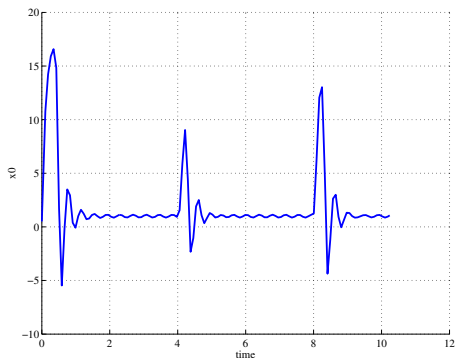
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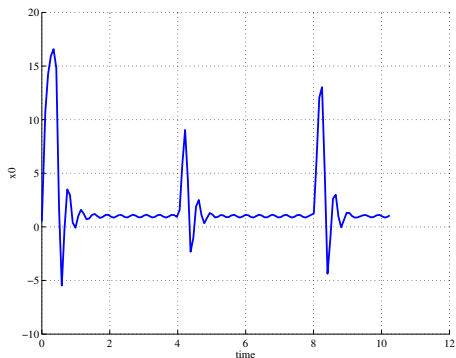
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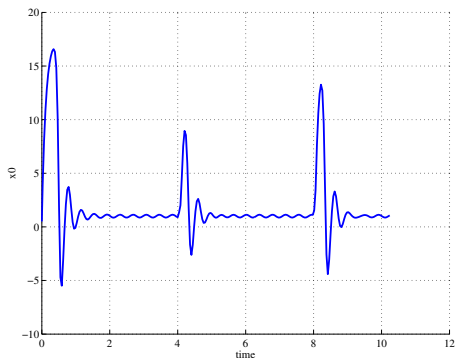
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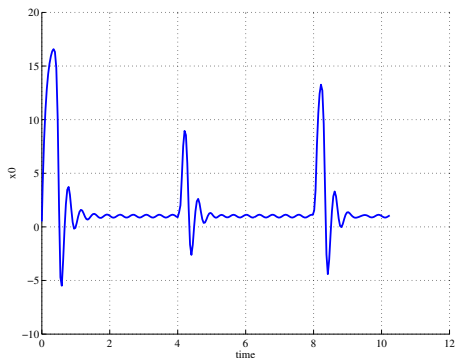
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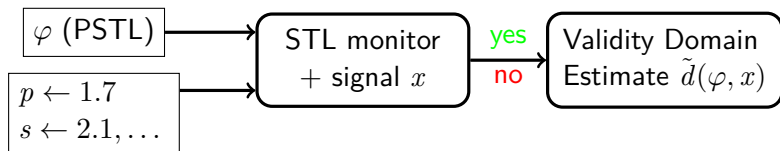
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Outline

- 1 Parametric Signal Temporal Logic
- 2 Exact Computation of Validity Domains
- 3 Numerical Approximation of Validity Domains

General Approach

Estimate the set $d(\varphi, x)$ using a finite number of queries to an STL monitor with different values for the parameters

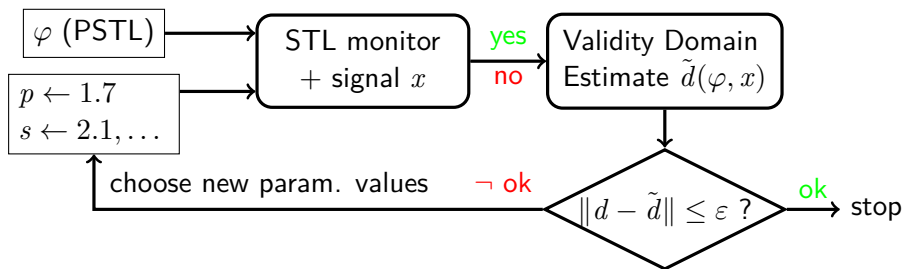


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- ▶ Scalability issues with the QE based approach
- ▶ STL (i.e., instantiated PSTL), formulas can be monitored efficiently

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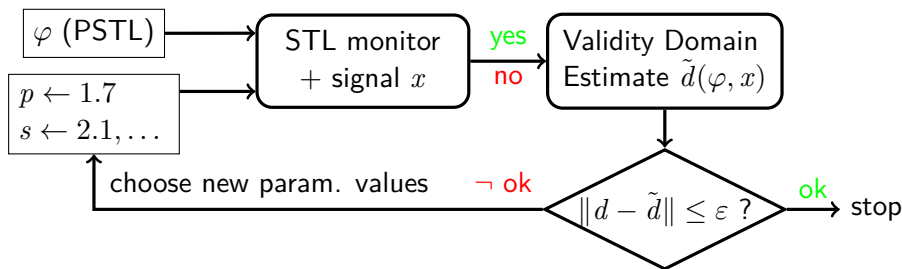


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Monitoring Computational Cost¹

(a) Signal x , formula $\varphi = (x > 0) \underbrace{\mathcal{U}_{[0,1]} (x > 0) \mathcal{U}_{[0,1]} (x > 0) \dots}_{i \text{ times}}$

(b) Formula φ_{st} with different number of samples

(a)

i	time(s)
1	0.34747
2	0.46335
3	0.60599
4	0.76067
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Nb of samples	time(s)
31416	0.18402
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- ▶ However, the approximation scales exponentially with the number of queries.
- ▶ In certain cases, efficient heuristics are possible

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Polarity

Definition (informal)

The *polarity* $\pi(p, \varphi)$ of a parameter p with respect to a formula φ is positive if it is easier to satisfy φ as we increase the value of p and negative otherwise

Examples and remarks

Example: standard parameters: simple

Example: $\forall p \in \mathbb{R}^+, \forall x \in \mathbb{R}, x > 0 \wedge x < p \rightarrow x < p$

Example: temporal parameters: subtle

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Monotonic Validity Domains

Definition

A subset $V \subseteq \mathcal{P} \times \mathcal{S}$ is monotonic if for every i , whenever a parameter valuation $(v_1, \dots, v_i, \dots, v_n)$ is in V so is any $(v_1, \dots, v'_i, \dots, v_n) \in \mathcal{P} \times \mathcal{S}$ satisfying $v'_i > v_i$ (when $\pi(p_i, \varphi) = +$) or $v'_i < v_i$ (when $\pi(p_i, \varphi) = -$).

- ▶ If φ has only one monotonic param. p then $d(x, \varphi) \subset [p^*, \infty)$ for some p^*
- ▶ More generally the *boundary* of $d(x, \varphi)$ has the same properties as *Pareto fronts*

Monotonic validity domains can be estimated using heuristics generalizing dichotomic search in $\dim \geq 1$

In the following, we sketch a methodology adapted from

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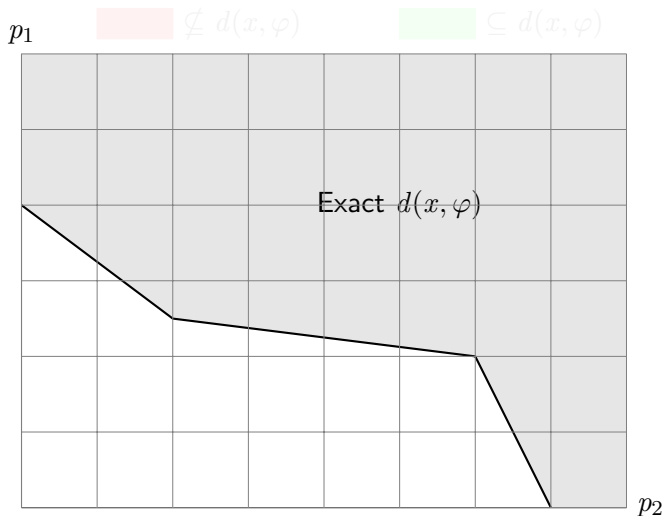
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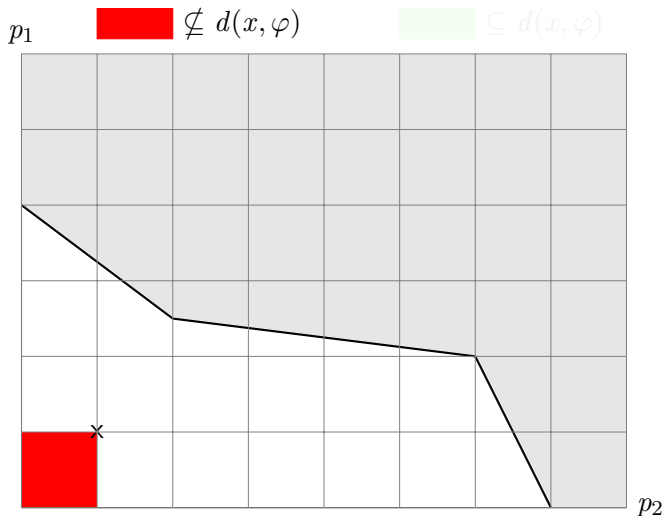
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Approximating Monotonic Validity Domains, Illustration



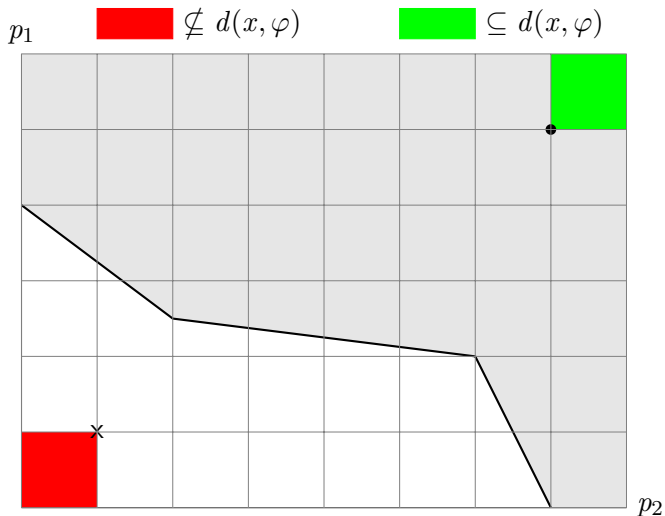
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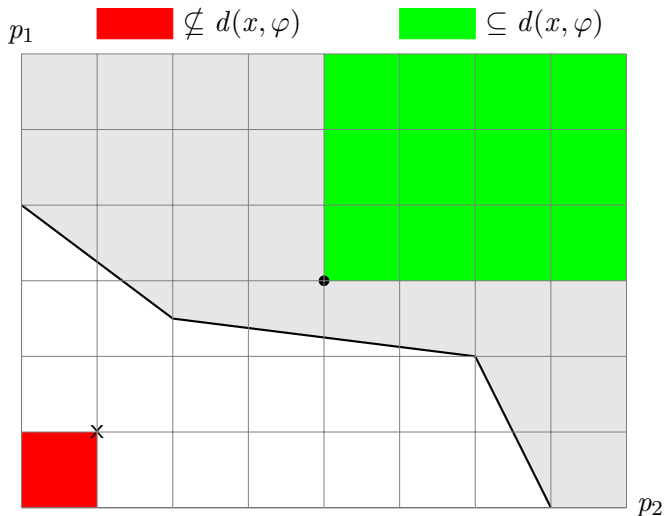
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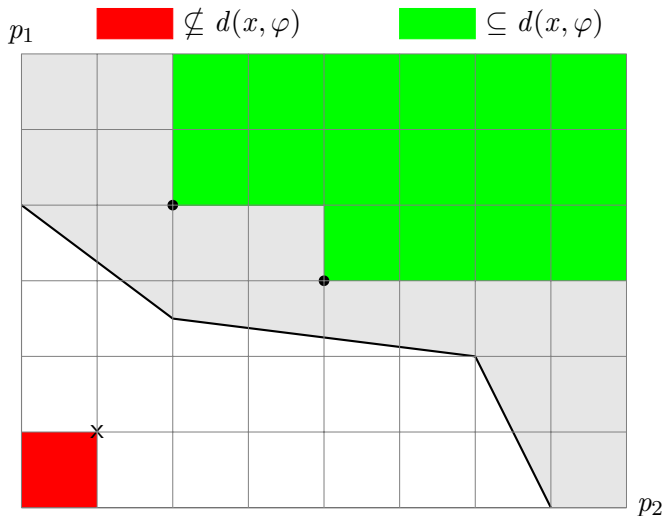
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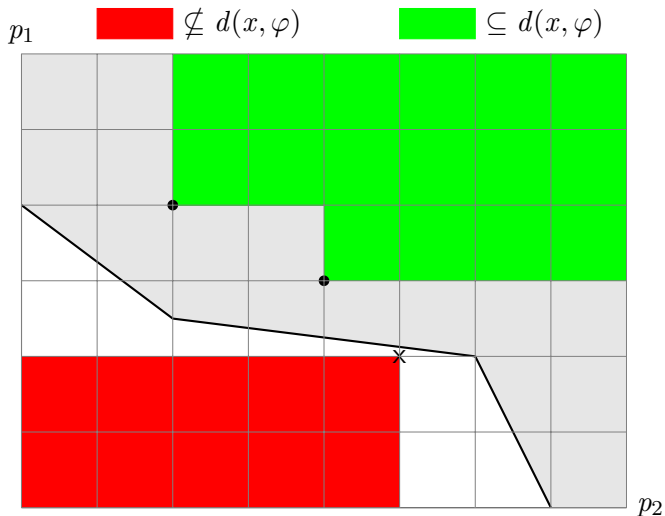
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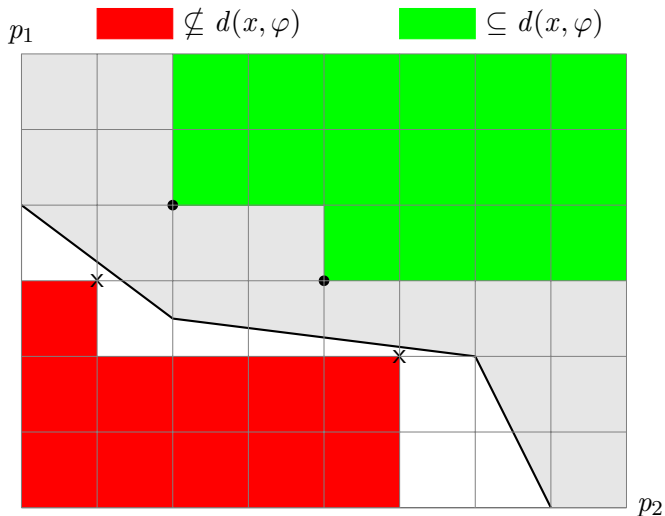
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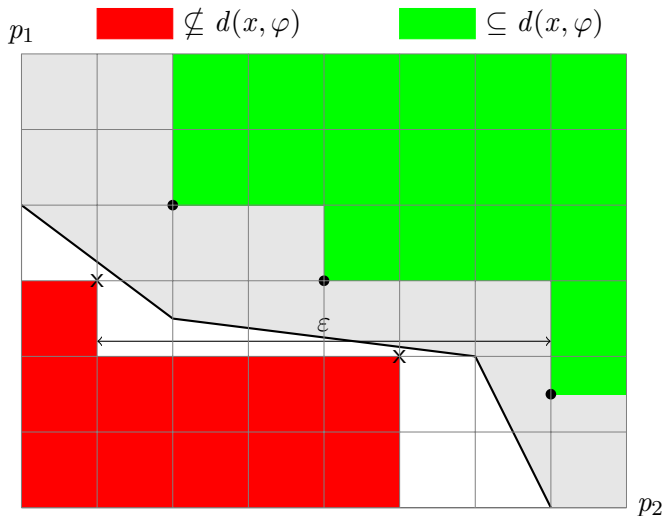
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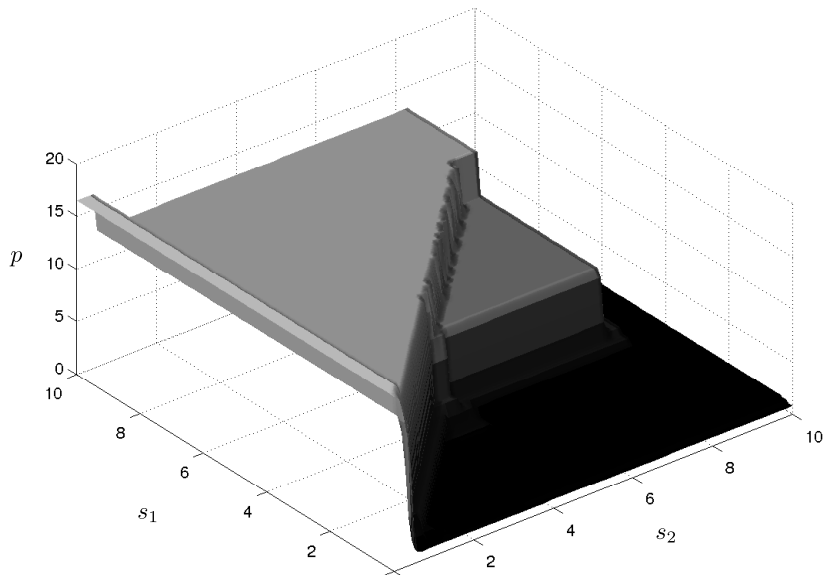
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Example

Stabilization property, $d(x, \varphi_{st})$ for x with 1024 samples



Conclusion

A problem potentially useful in many situations : given simulation traces, what properties do the system satisfy ?

Exact computation of validity domains using quantifier elimination

- ▶ Elegant but scalability problems with the size of x and φ
- ▶ What can be done beyond our “naive” approach ?

Numerical approximation using efficient monitoring

- ▶ Scales in $|x|$ and $|\varphi|$ but exponential in number of parameters
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