

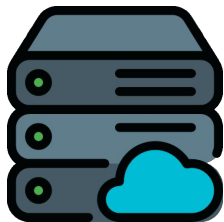
Time Robustness for Point-Based Semantics of Metric Interval Temporal Logic

Simone Silveti¹, Ivan Compagnucci², Francesca Cairoli¹
Catia Trubiani² and Laura Nenzi¹

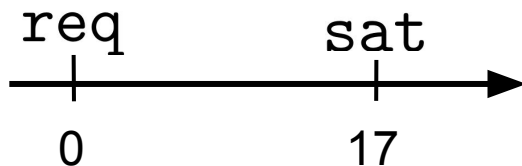
¹ University of Trieste – ² Gran Sasso Science Institute



System



Traces



Requirements

Every request must be acknowledged between 10 and 20 time units

MITL

$$\sigma \quad \varphi = \mathbf{G}(\text{req} \rightarrow \mathbf{F}_{[10,20]} \text{sat})$$

$\sigma \models \varphi$ Boolean Semantics

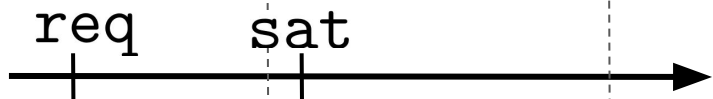
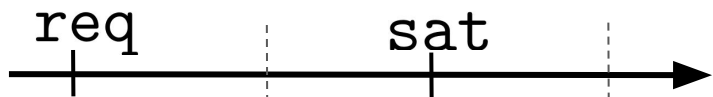
Beyond Boolean Semantics

$$\varphi = \mathbf{G}(\text{req} \rightarrow \mathbf{F}_{[10,20]}\text{sat})$$

Reliably under
perturbation

The earlier the
better

?



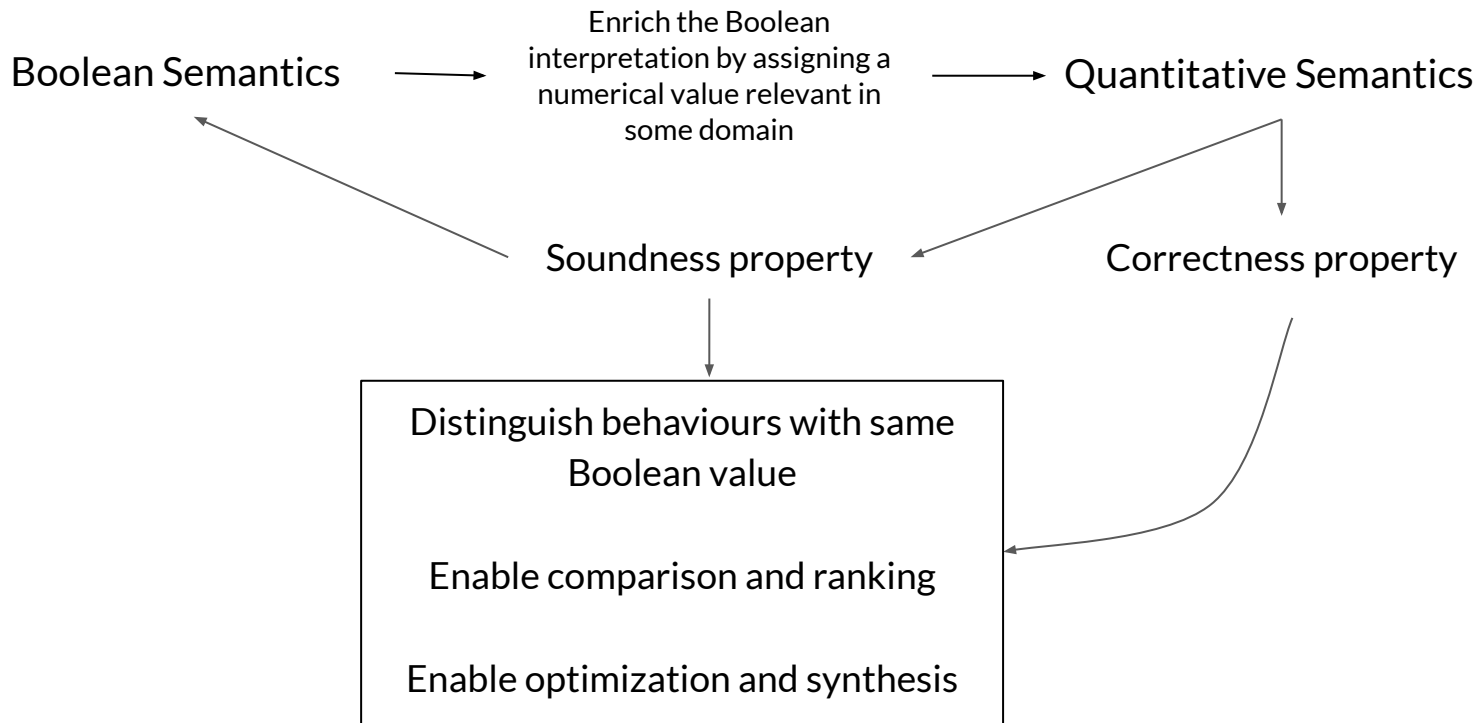
0

10

20

Beyond Boolean Semantics

$$\varphi = \mathbf{G}(\text{req} \rightarrow \mathbf{F}_{[10,20]}\text{sat})$$



Contribution

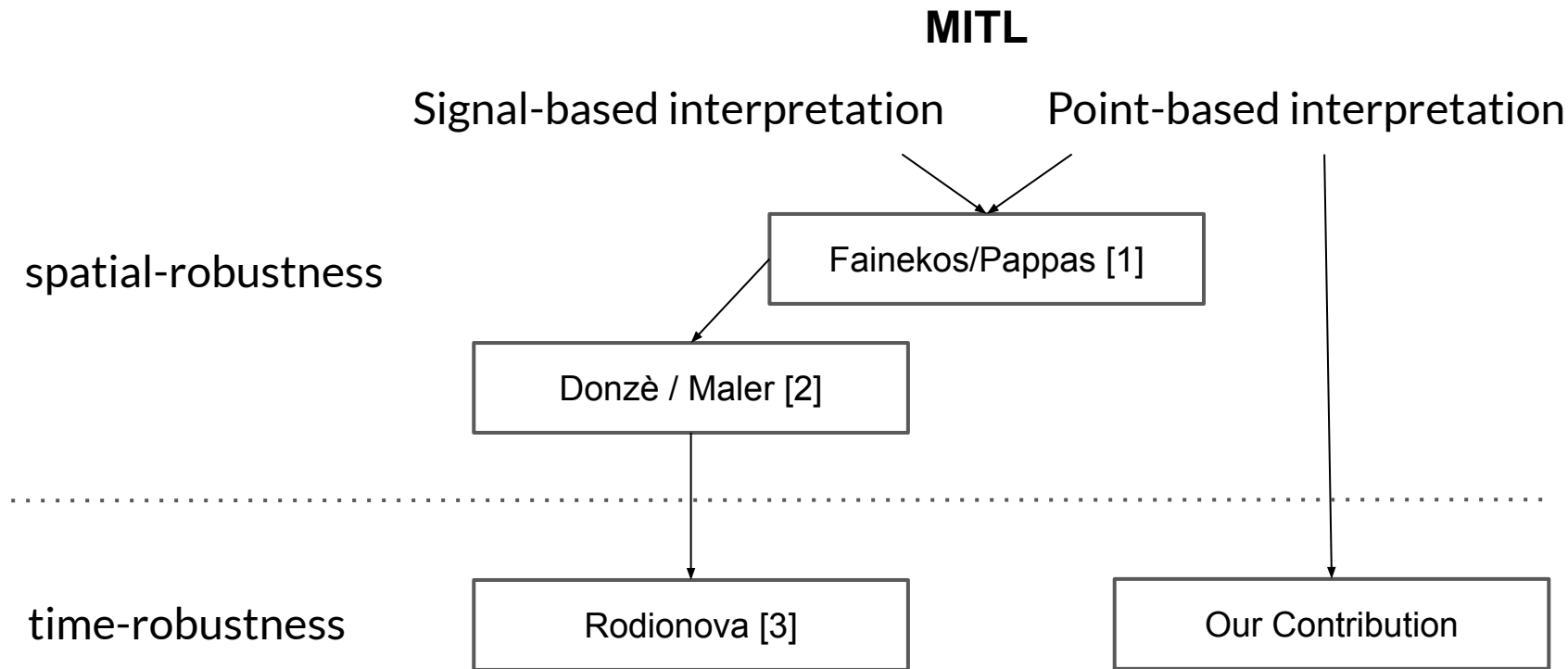
Definition of a new time-robustness semantics for the point-based interpretation of Metric Interval Temporal Logic (MITL)

Proof of soundness and stability guarantees wrt the Boolean Semantics of MITL

Evidence of relation between time robustness and tolerance under temporal uncertainty

2 use cases

A bit of history

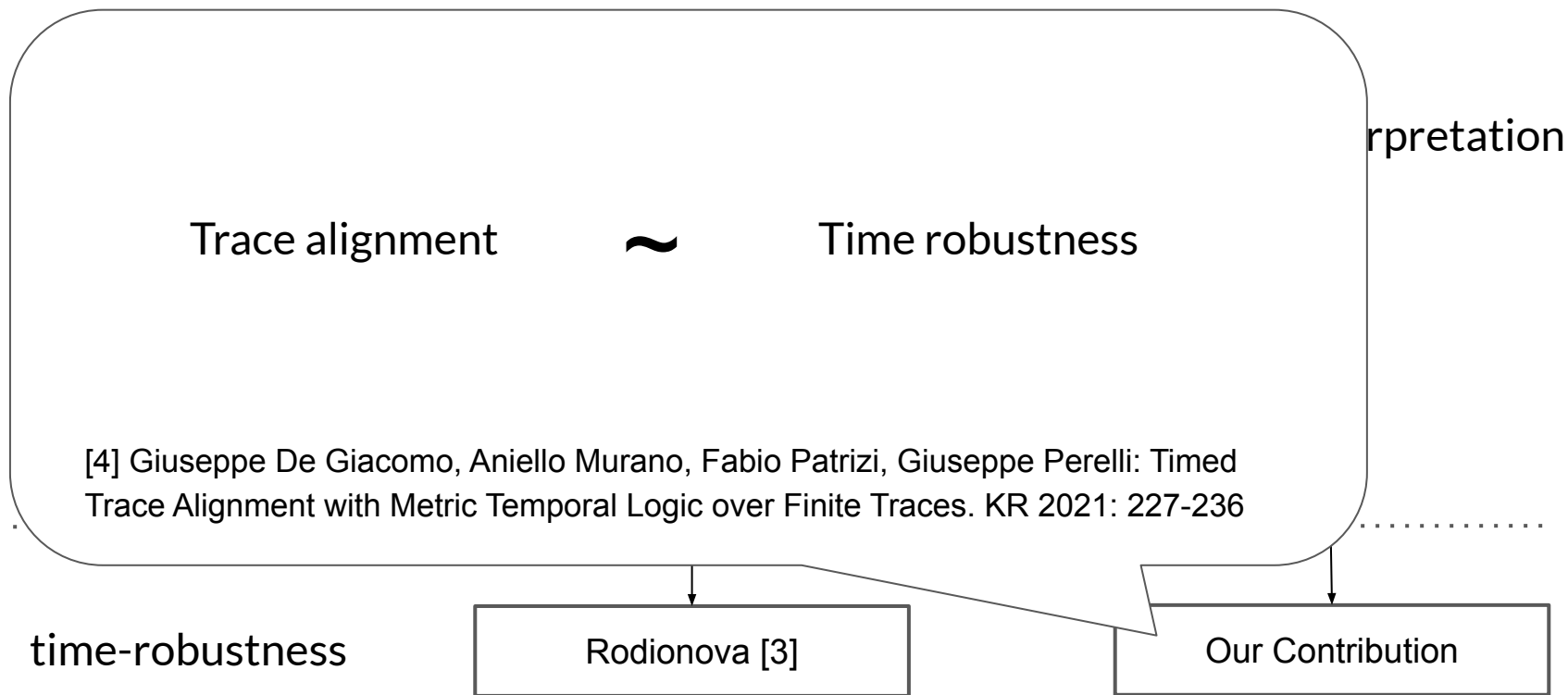


[1] Fainekos, Pappas (2009). Robustness of temporal logic specifications for continuous-time signals. Theor. Comput. Sci. 410(42))

[2] Donzé, Maler (2010). Robust Satisfaction of Temporal Logic over Real-Valued Signals. FORMATS 2010

[3] Rodionova et al. (2022). Temporal robustness of temporal logic specifications: Analysis and control design. ACM Transactions on Embedded Computing Systems 22(1):1–44

A bit of history



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Definition

Finite event trace $\sigma = (t_0, s_0)(t_1, s_1) \dots (t_{n-1}, s_{n-1})$

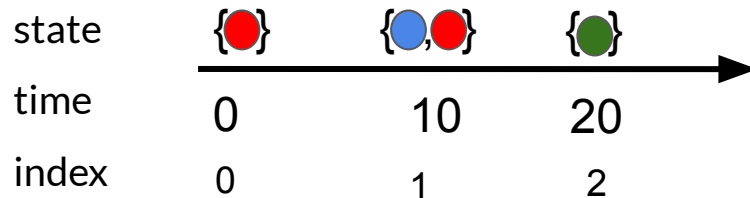
|

Event $\in \mathbb{R}_{\geq 0} \times 2^{\text{AP}}$

|

$\{p_0, p_1, \dots, p_{m-1}\}$

$$\sigma = (0, \{\text{red}\})(10, \{\text{blue}, \text{red}\})(20, \{\text{green}\})$$



MITL Syntax

$$\varphi := \top \mid \mathbf{p} \mid \neg\varphi \mid \varphi \wedge \varphi \mid \varphi \mathbf{U}_I \varphi \mid \mathbf{X}_I \varphi$$

$\downarrow$
Atomic predicate

$\downarrow$
Until

$\downarrow$
Next

MITL Syntax

$$\varphi := \top \mid \mathbf{p} \mid \neg\varphi \mid \varphi \wedge \varphi \mid \varphi \mathbf{U}_I \varphi \mid \mathbf{X}_I \varphi$$

|
Atomic predicate

| |
Until Next

|

$$\mathbf{F}_I \varphi := \top \mathbf{U}_I \varphi$$

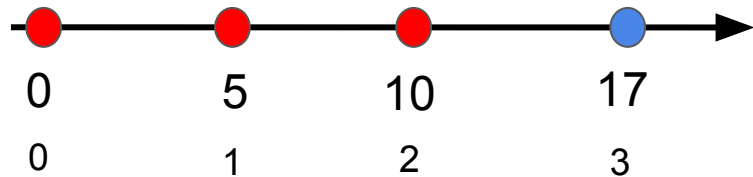
|

$$\mathbf{G}_I \varphi := \neg \mathbf{F}_I \neg \varphi$$

MITL Semantics

$$\varphi := \top \mid \mathbf{p} \mid \neg\varphi \mid \varphi \wedge \varphi \mid \varphi \mathbf{U}_I \varphi \mid \mathbf{X}_I \varphi$$

$$(\sigma, i) \models \varphi_1 \mathbf{U}_I \varphi_2 \text{ if } \exists j \geq i, (\sigma, j) \models \varphi_2, t_j - t_i \in I, \\ \forall k \in [i, j), (\sigma, k) \models \varphi_1$$

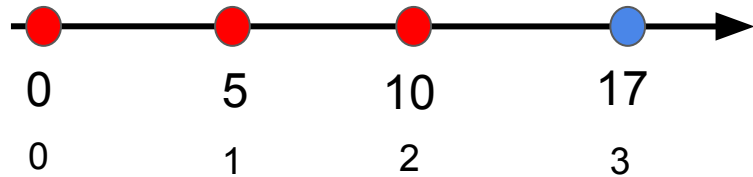


$$(\sigma, 0) \models \text{red circle } \mathbf{U}_{[15,20]} \text{ blue circle}$$

MITL Semantics

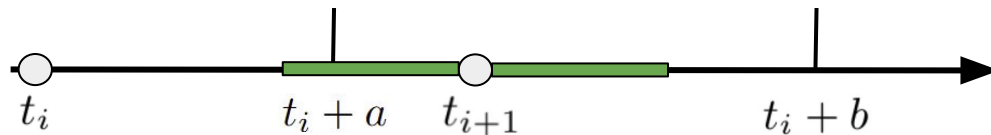
$$\varphi := \top \mid \mathbf{p} \mid \neg\varphi \mid \varphi \wedge \varphi \mid \varphi \mathbf{U}_I \varphi \mid \mathbf{X}_I \varphi$$

$$(\sigma, i) \models \mathbf{X}_I \varphi \text{ if } i < n - 1 \text{ and } (\sigma, i + 1) \models \varphi, \text{ and} \\ t_{i+1} - t_i \in I$$



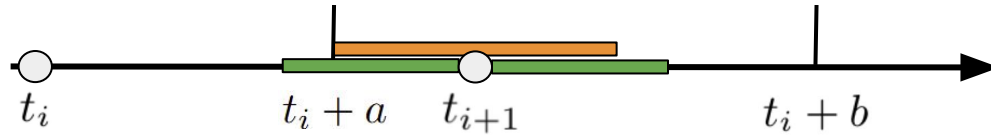
$$(\sigma, 2) \models \mathbf{X}_{[5,10]} \text{ (blue circle)}$$

From Boolean to Time Robustness



$(\sigma, i) \models \mathbf{X}_I \varphi$ if $i < n - 1$ and $(\sigma, i + 1) \models \varphi$, and
 $\underline{t_{i+1} - t_i \in I}$

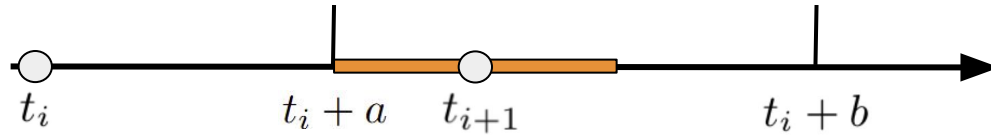
From Boolean to Time Robustness



$$(\sigma, i) \models \mathbf{X}_I \varphi \text{ if } i < n - 1 \text{ and } (\sigma, i + 1) \models \varphi, \text{ and}$$

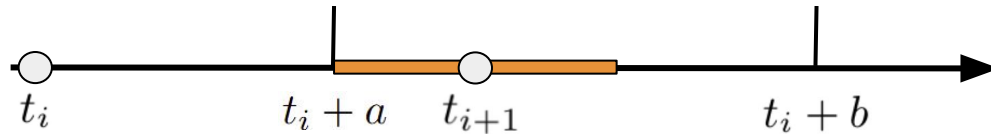
$$\underline{t_{i+1} - t_i \in I}$$

From Boolean to Time Robustness



$(\sigma, i) \models \mathbf{X}_I \varphi$ if $i < n - 1$ and $(\sigma, i + 1) \models \varphi$, and
 $t_{i+1} - t_i \in I$

From Boolean to Time Robustness

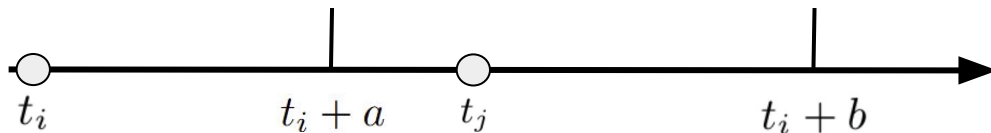


$$(\sigma, i) \models \mathbf{X}_I \varphi \text{ if } i < n - 1 \text{ and } (\sigma, i + 1) \models \varphi, \text{ and}$$

$$\underline{t_{i+1} - t_i \in I}$$

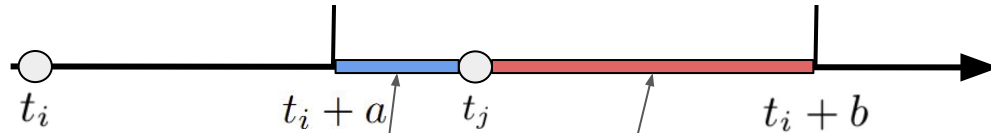
$$\rho(\mathbf{X}_I \varphi, \sigma, i) = \begin{cases} \min(\rho(\varphi, \sigma, i + 1), d(t_i, t_{i+1}, I)) & \text{if } i < n - 1, \\ -\infty & \text{otherwise.} \end{cases}$$

From Boolean to Time Robustness



$$d(t_i, t_j, [a, b]) = \begin{cases} t_i + b - t_j & \text{if } a = 0, \\ \min(t_j - t_i - a, t_i + b - t_j) & \text{otherwise.} \end{cases}$$

From Boolean to Time Robustness



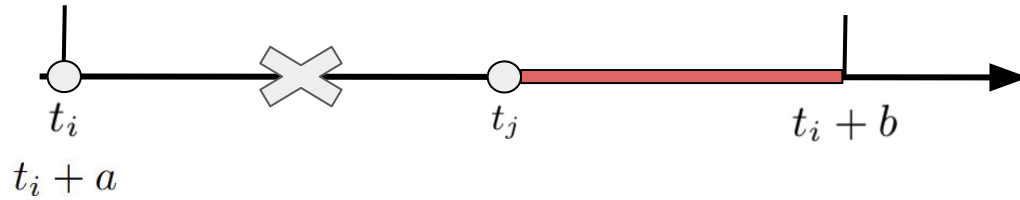
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From Boolean to Time Robustness



$$d(t_i, t_j, [a, b]) = \begin{cases} t_i + b - t_j & \text{if } a = 0, \\ \min(t_j - t_i - a, t_i + b - t_j) & \text{otherwise.} \end{cases}$$



We do not allow changing the order of events

Time Robustness

$$\rho(\top, \sigma, i) = +\infty$$

$$\rho(p, \sigma, i) = \begin{cases} +\infty & \text{if } (\sigma, i) \models p, \\ -\infty & \text{otherwise.} \end{cases}$$

$$\rho(\neg\varphi, \sigma, i) = -\rho(\varphi, \sigma, i)$$

$$\rho(\varphi_1 \wedge \varphi_2, \sigma, i) = \min(\rho(\varphi_1, \sigma, i), \rho(\varphi_2, \sigma, i))$$

$$\rho(\varphi_1 \mathbf{U}_I \varphi_2, \sigma, i) = \max_{j \geq i} \left(\min \left(\rho(\varphi_2, \sigma, j), d(t_i, t_j, I), \right. \right. \\ \left. \left. \min_{i \leq k < j} \rho(\varphi_1, \sigma, k) \right) \right)$$

$$\rho(\mathbf{X}_I \varphi, \sigma, i) = \begin{cases} \min(\rho(\varphi, \sigma, i+1), d(t_i, t_{i+1}, I)) & \text{if } i < n-1, \\ -\infty & \text{otherwise.} \end{cases}$$

Theorem 1 (Soundness of time robustness). *The time robustness is sound with respect to Boolean semantics of MITL, meaning that for each formula φ , event trace σ and $i < |\sigma|$ we have:*

$$\rho(\varphi, \sigma, i) > 0 \Rightarrow (\sigma, i) \models \varphi \quad (1)$$

$$\rho(\varphi, \sigma, i) < 0 \Rightarrow (\sigma, i) \not\models \varphi \quad (2)$$

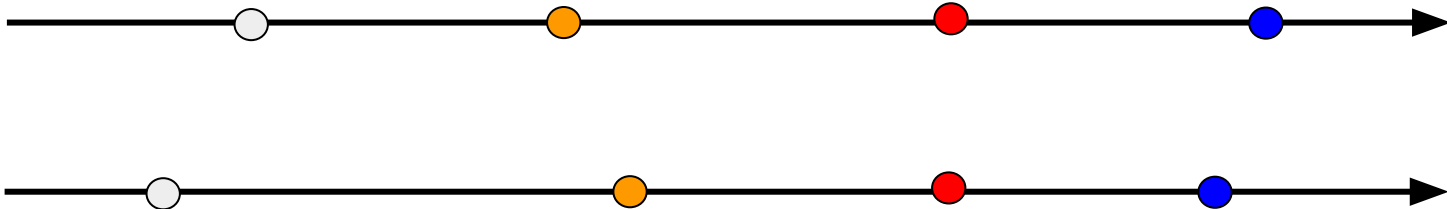
$$(\sigma, i) \models \varphi \Rightarrow \rho(\varphi, \sigma, i) \geq 0 \quad (3)$$

$$(\sigma, i) \not\models \varphi \Rightarrow \rho(\varphi, \sigma, i) \leq 0. \quad (4)$$

Stability

Theorem 2 (Stability property of time robustness). *Let us consider $\varphi \in \mathcal{L}_{MITL}$ and two trajectories σ and σ' such that $\vec{s}(\sigma) = \vec{s}(\sigma')$, then*

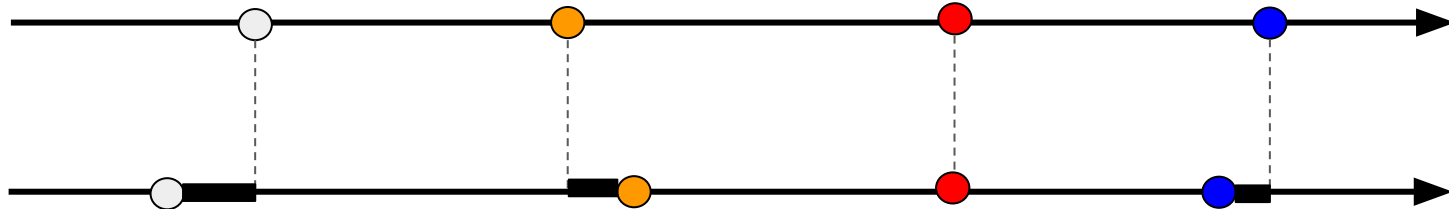
$$\forall i \leq |\sigma|, \quad |\rho(\varphi, \sigma', i) - \rho(\varphi, \sigma, i)| \leq \|\vec{t}(\sigma') - \vec{t}(\sigma)\|_1. \quad (5)$$



Stability

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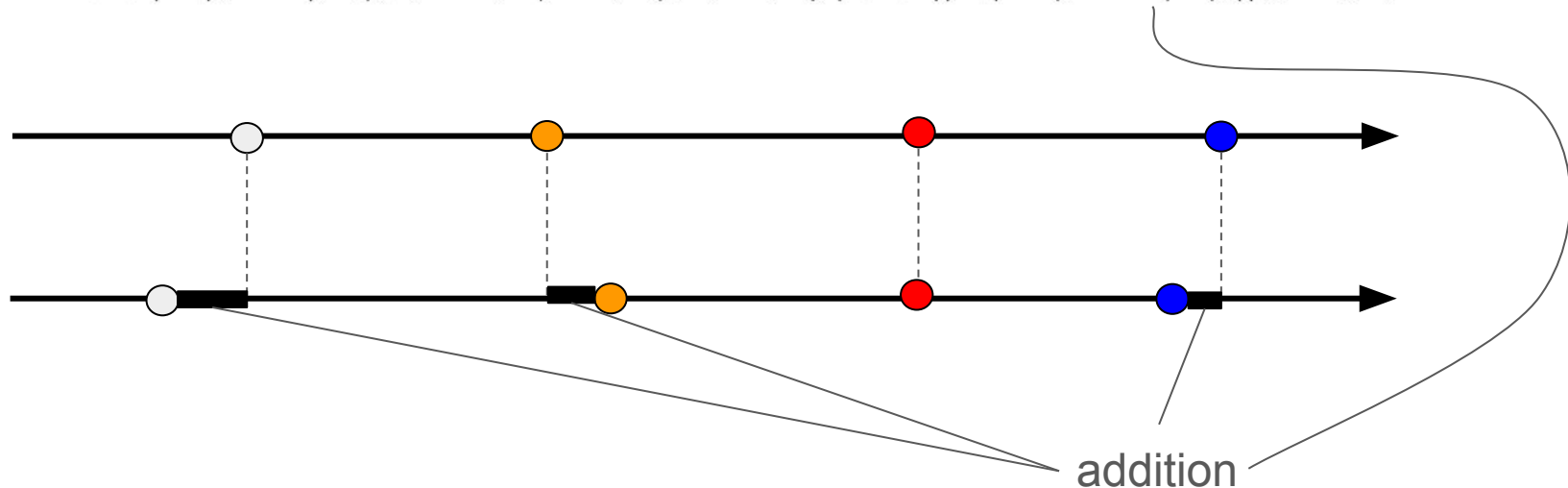
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Stability

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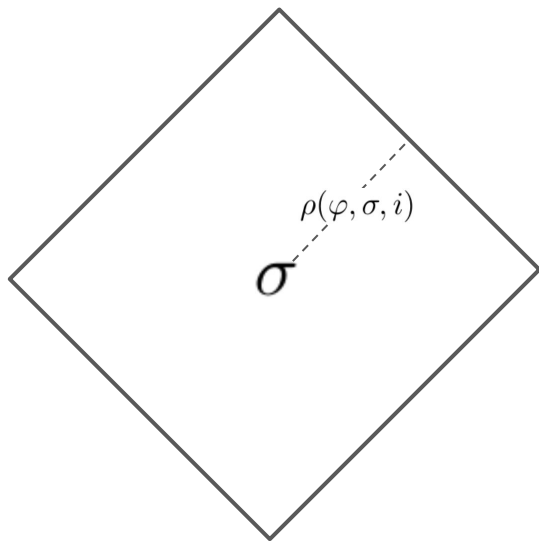


Stability

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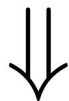
$< \rho(\varphi, \sigma, i)$



Stability

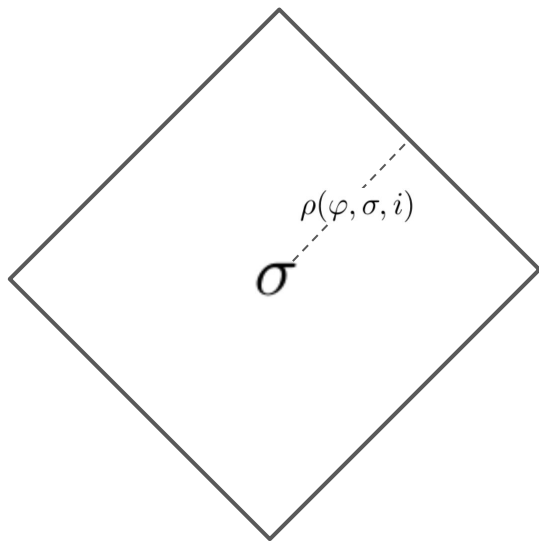
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$$\forall i \leq |\sigma|, \quad |\rho(\varphi, \sigma', i) - \rho(\varphi, \sigma, i)| \leq \|\vec{t}(\sigma') - \vec{t}(\sigma)\|_1 \quad (5)$$



$$\rho(\varphi, \sigma', i) > 0$$

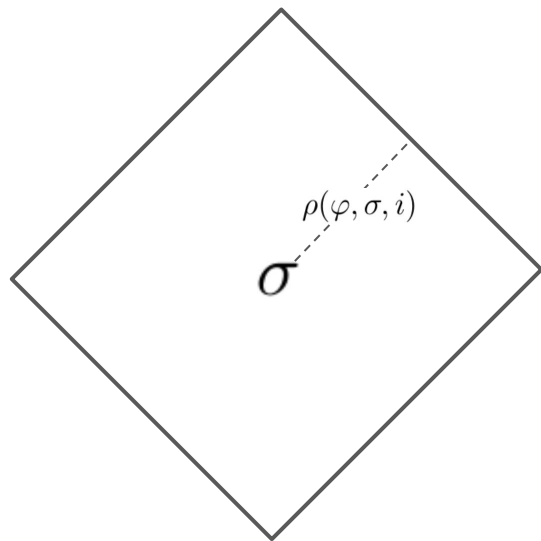
$$< \rho(\varphi, \sigma, i)$$



Stability

Theorem 2 (Stability property of time robustness). *Let us consider $\varphi \in \mathcal{L}_{MITL}$ and two trajectories σ and σ' such that $\vec{s}(\sigma) = \vec{s}(\sigma')$, then*

$$\forall i \leq |\sigma|, \quad |\rho(\varphi, \sigma', i) - \rho(\varphi, \sigma, i)| \leq \boxed{\|\vec{t}(\sigma') - \vec{t}(\sigma)\|_1} \quad (5)$$



$$\begin{array}{c} \Downarrow \\ \rho(\varphi, \sigma', i) > 0 \\ \Downarrow \text{ soundness} \\ (\sigma', i) \models \varphi \end{array}$$

$< \rho(\varphi, \sigma, i)$

Monitoring Algorithm

$$\rho(\varphi_1 \mathbf{U}_I \varphi_2, \sigma, i) = \max \left(\min_{j \geq i} \left(\rho(\varphi_2, \sigma, j), d(t_i, t_j, I), \right. \right. \\ \left. \left. \min_{i \leq k < j} \rho(\varphi_1, \sigma, k) \right) \right)$$

+

Tree structure of the formula

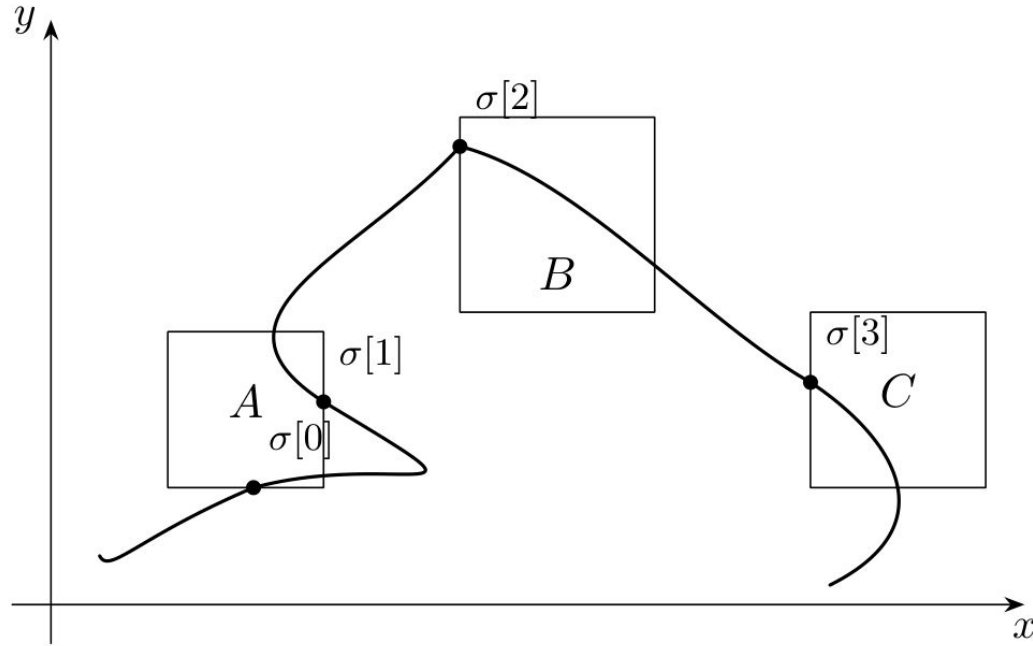


Complexity $\mathcal{O}(n^2 |\varphi|)$



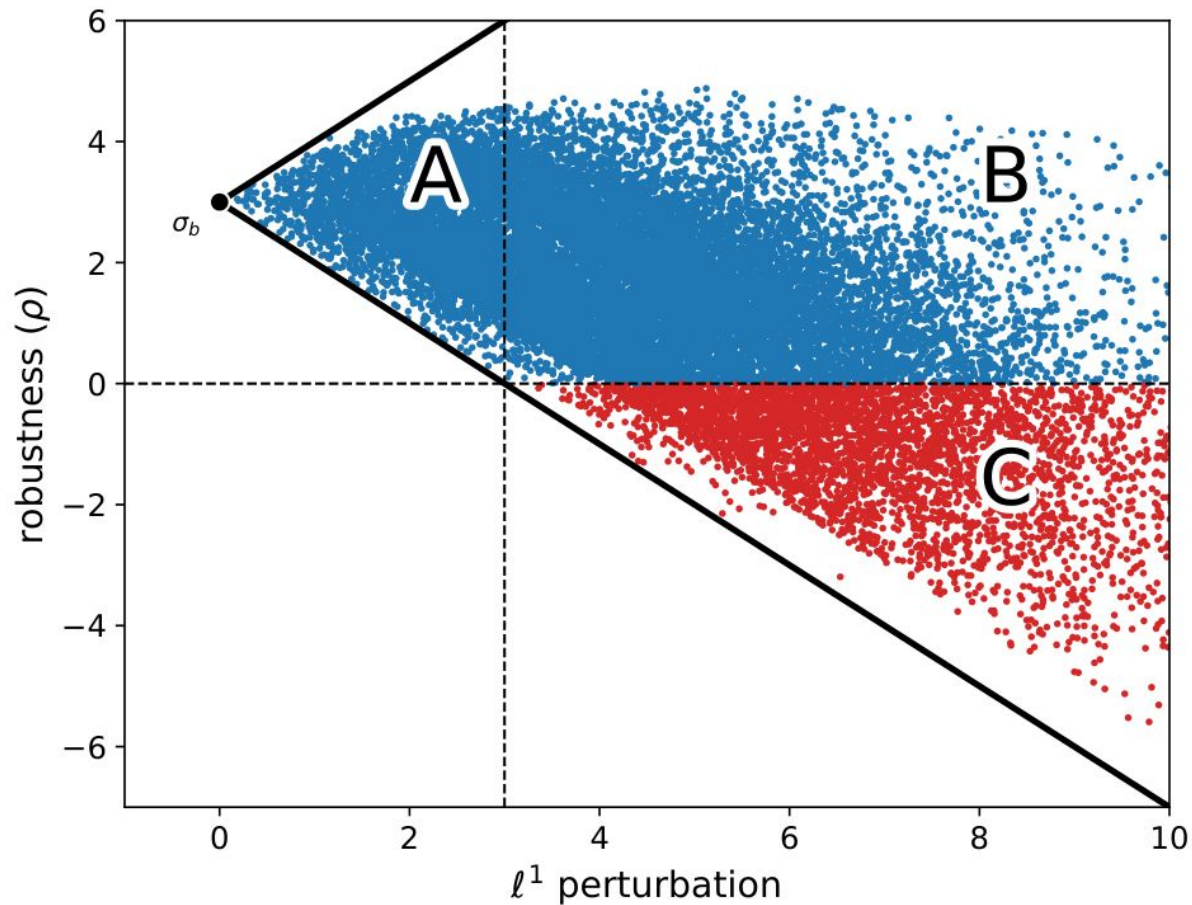
TIRO tool

Drone surveillance system

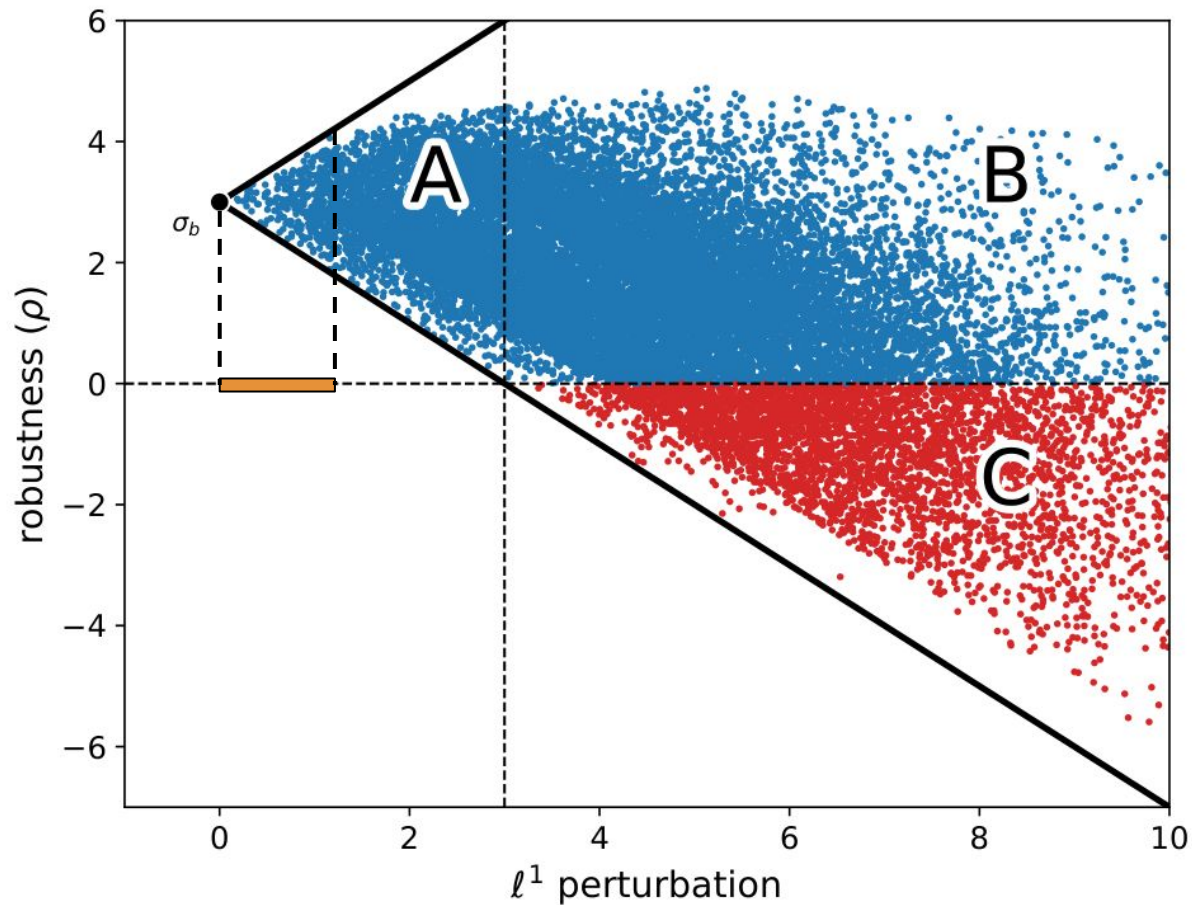


$$\psi_d = \mathbf{F}_{[10,20]}(\text{enter}A \wedge \mathbf{F}_{[40,50]}(\text{enter}B \wedge \mathbf{F}_{[20,30]} \text{enter}C))$$

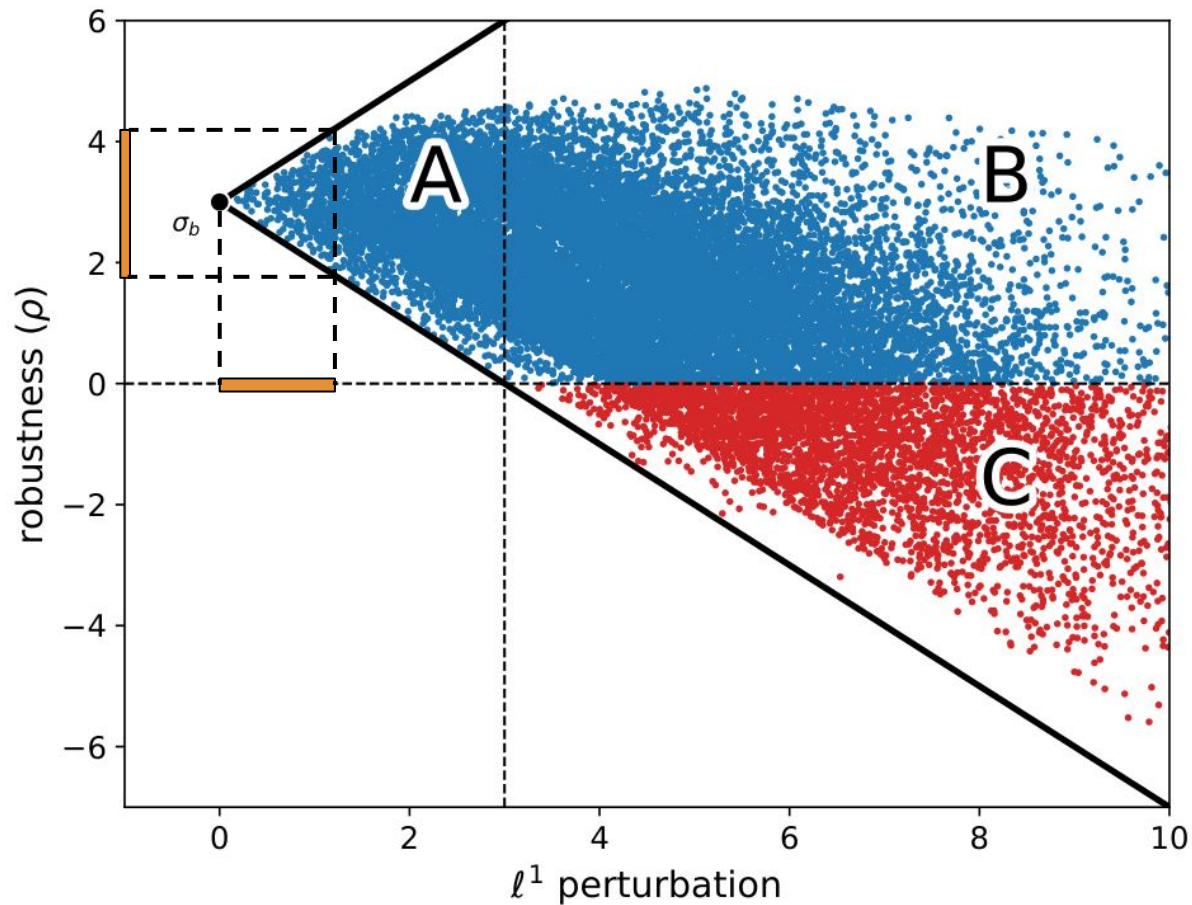
Drone surveillance system



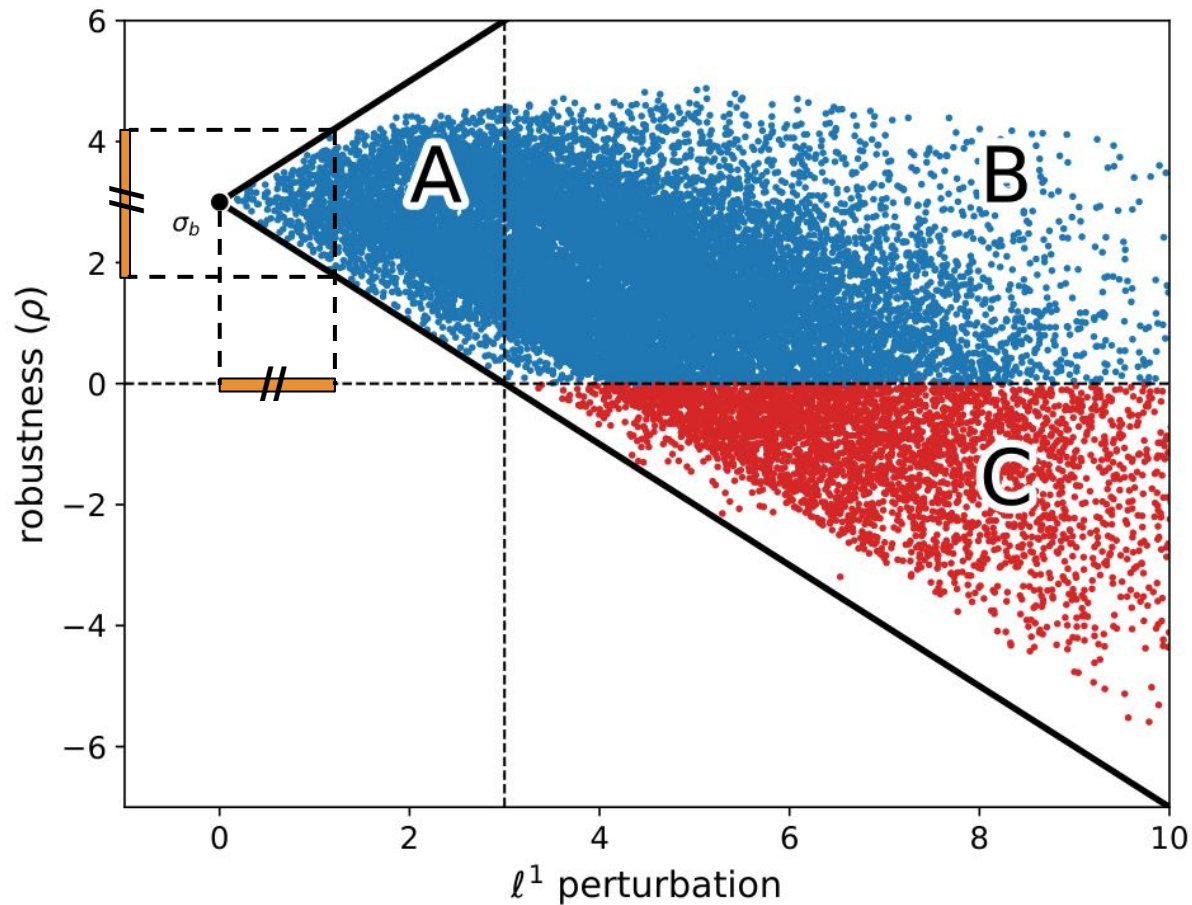
Drone surveillance system



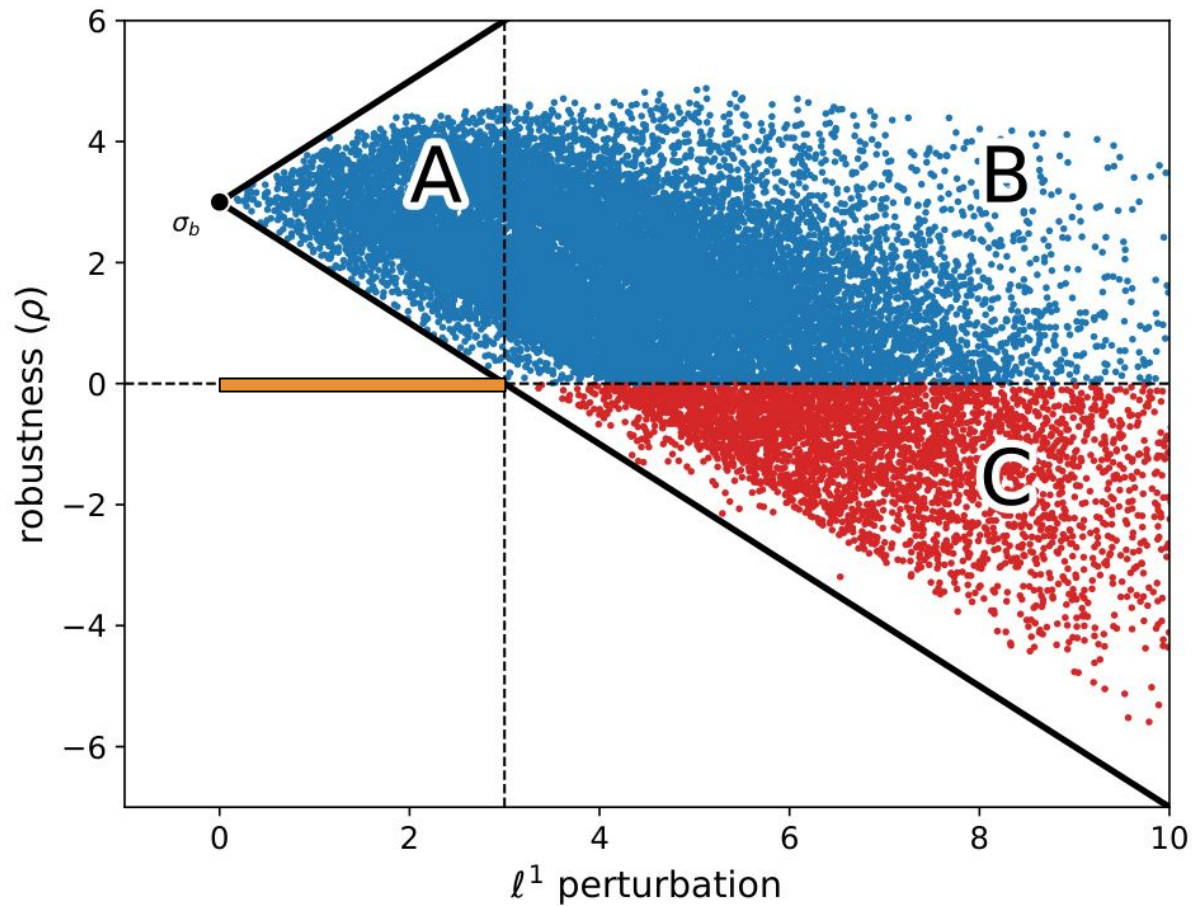
Drone surveillance system



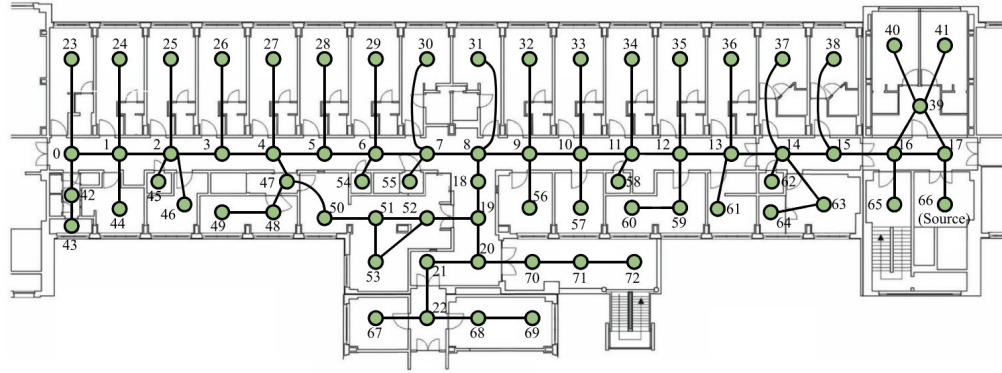
Drone surveillance system



Drone surveillance system



Smart Hospital



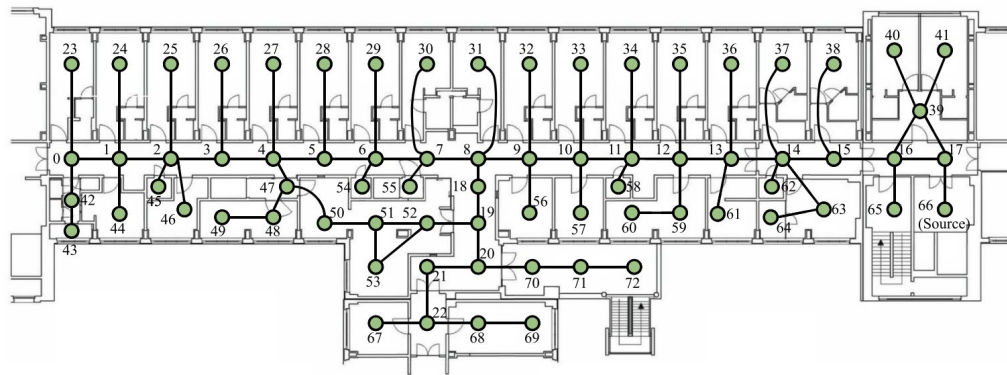
Robots delivering medical items from pharmacy to target rooms



might fail and ask for another robot to recover them

malicious: request fake recovering

Smart Hospital



$$\varphi_a = \mathbf{G}(\text{create} \rightarrow \mathbf{F}_{[0,20]} \text{load})$$

$$\varphi_b = \mathbf{G}(\text{load} \rightarrow \mathbf{F}_{[L,U]} \text{deliver})$$

$$\varphi_c = \mathbf{G}(\text{fail} \rightarrow X_{[0,2500]} \text{recover})$$

Robots delivering medical items from pharmacy to target rooms

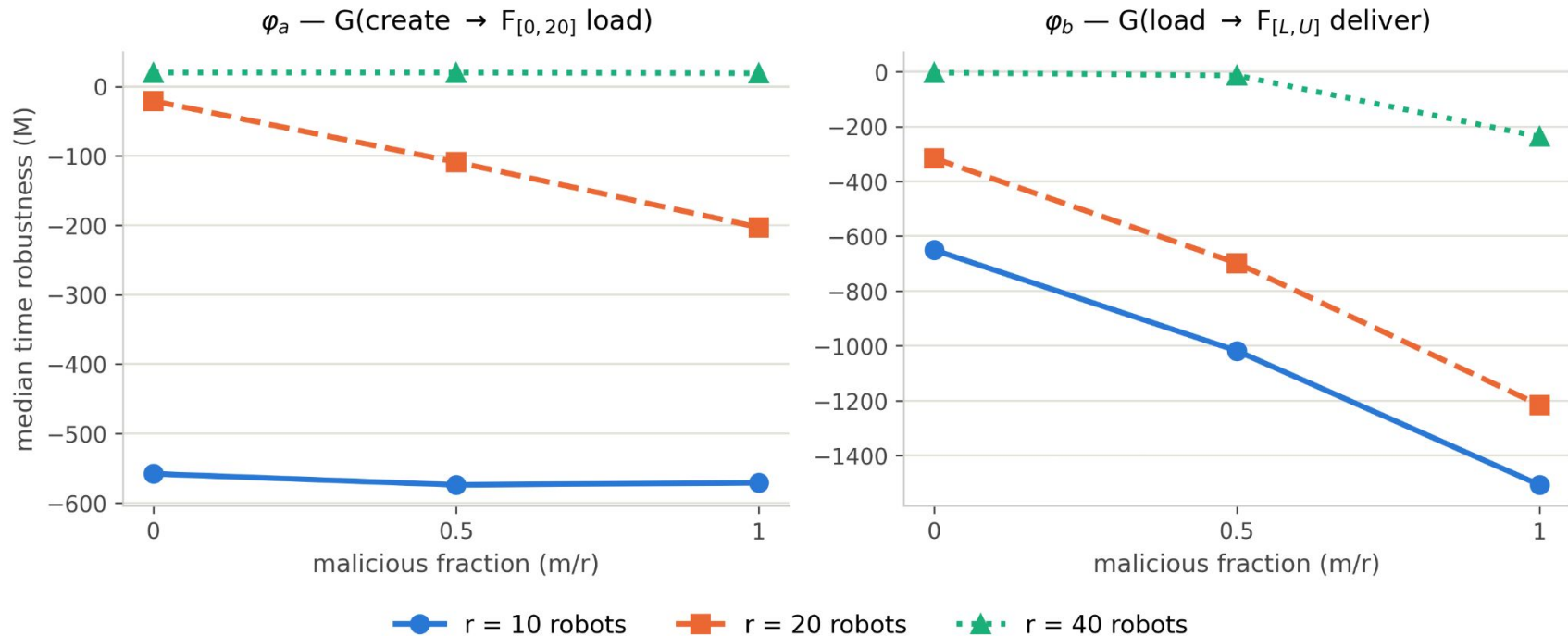


might fail and ask for another robot to recover them

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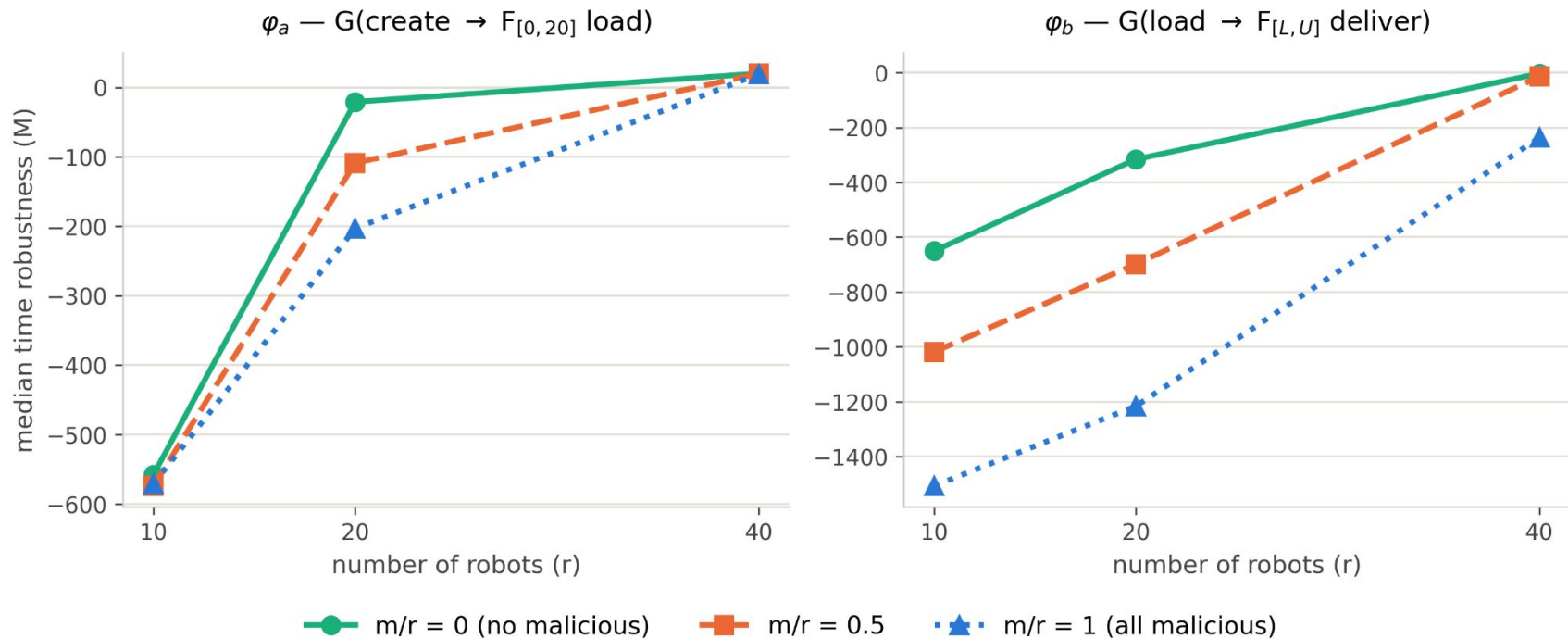
Smart Hospital

Median time robustness vs. fraction of malicious robots



Smart Hospital

Median time robustness vs. number of robots



Future work

Use time robustness for synthesis problem

Define an optimized algorithm for time robustness

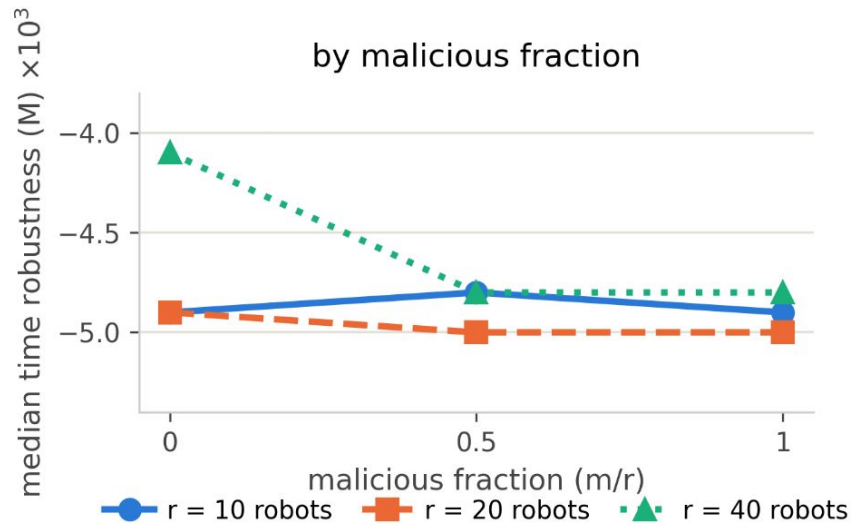
Remove the constraint of fixed ordering

Thanks

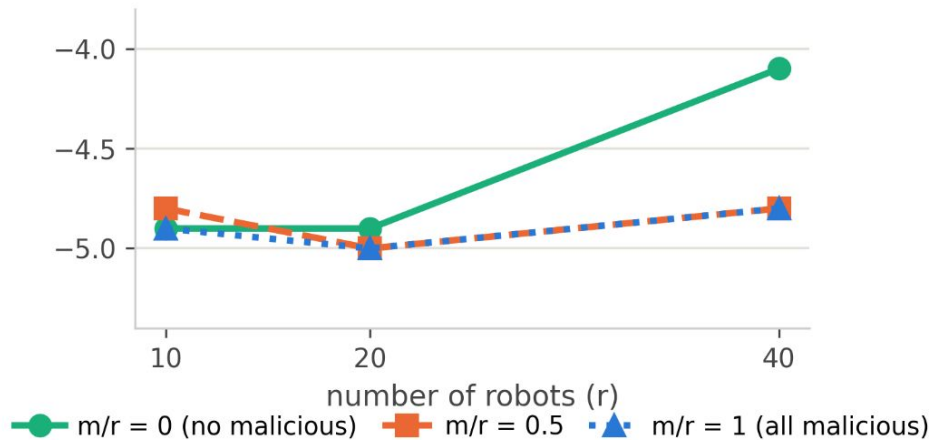
Smart Hospital

φ_c — $G(\text{fail} \rightarrow X_{[0, 2500]} \text{ recover})$: robustness is stable

by malicious fraction



by number of robots



Conclusions

Introduced a quantitative semantics for MITL under point-based interpretation extending the concept of time robustness to event-based systems

Proved soundness and stability property

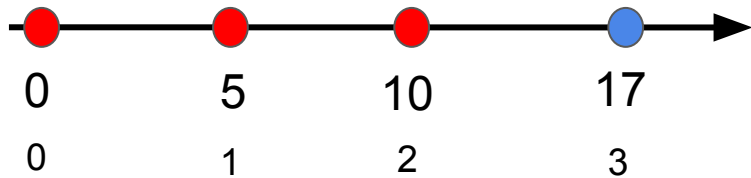
Defined a simple monitoring algorithm

Link time robustness to the resilience of a system to timing deviation with two experiments

MITL Semantics

$$\varphi := \top \mid \mathbf{p} \mid \neg\varphi \mid \varphi \wedge \varphi \mid \varphi \mathbf{U}_I \varphi \mid \mathbf{X}_I \varphi$$

$$(\sigma, i) \models \top$$

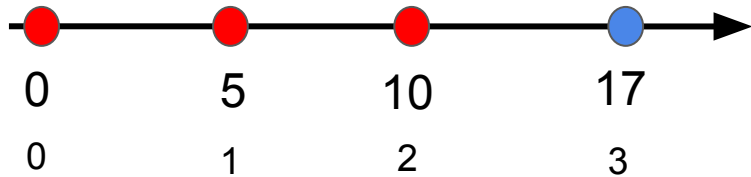


$$(\sigma, 0) \models \top \quad (\sigma, 1) \models \top \quad \dots$$

MITL Semantics

$$\varphi := \top \mid \mathbf{p} \mid \neg\varphi \mid \varphi \wedge \varphi \mid \varphi \mathbf{U}_I \varphi \mid \mathbf{X}_I \varphi$$

$$(\sigma, i) \models \mathbf{p} \text{ if } \mathbf{p} \in s_i$$



$$(\sigma, 0) \models \text{red circle}$$

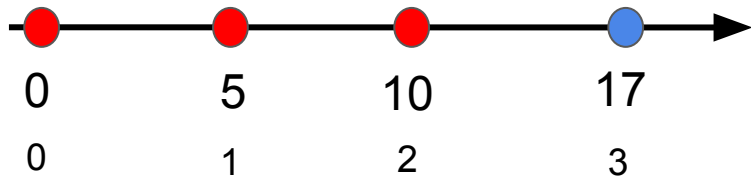
$$(\sigma, 3) \models \text{blue circle}$$

MITL Semantics

$$\varphi := \top \mid \mathbf{p} \mid \neg\varphi \mid \varphi \wedge \varphi \mid \varphi \mathbf{U}_I \varphi \mid \mathbf{X}_I \varphi$$

$$(\sigma, i) \models \neg\varphi \text{ if } (\sigma, i) \not\models \varphi$$

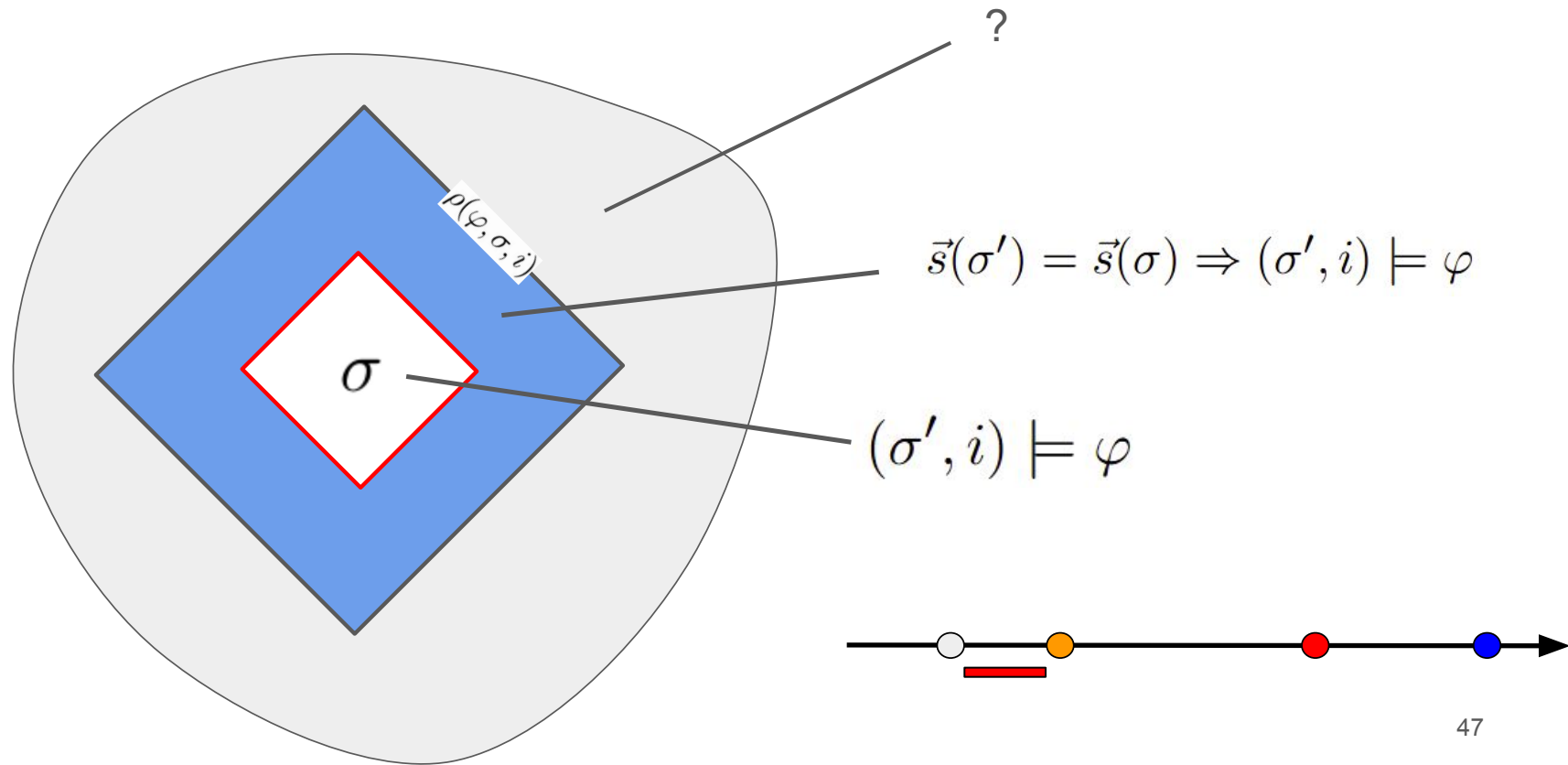
$$(\sigma, i) \models \varphi_1 \wedge \varphi_2 \text{ if } (\sigma, i) \models \varphi_1 \text{ and } (\sigma, i) \models \varphi_2$$



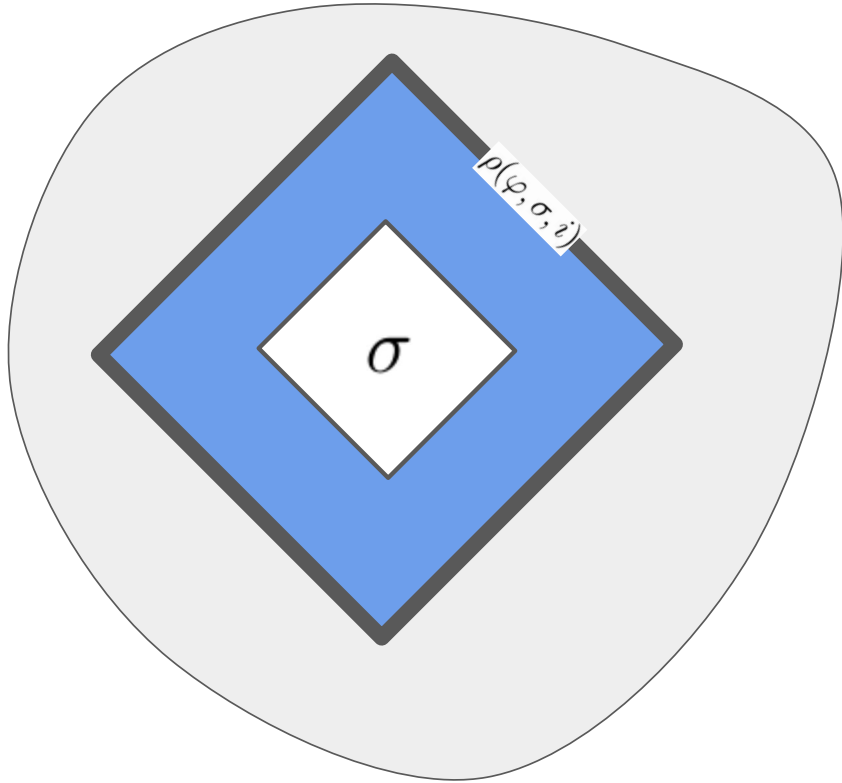
$$(\sigma, 0) \models \neg \text{blue circle}$$

$$(\sigma, 0) \models \neg (\text{blue circle} \wedge \text{red circle})$$

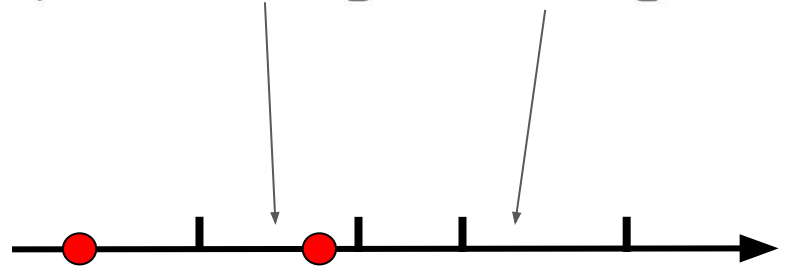
Topology of satisfaction



Under Approximation



$$\varphi = \mathbf{F}_I \neg p \vee \mathbf{F}_J \neg p$$



Smart Hospital

#robots	#malicious		median		φ_a			φ_b			$\varphi_c (\times 10^3)$		
	r	m	m/r	M	φ_a		M	φ_b		M	$\varphi_c (\times 10^3)$		
					Q_1	Q_3		Q_1	Q_3		Q_1	Q_3	
	10	0	0	-558	-55929	-178	-650	-5661	9	-4.9	-6.5	-3.8	
	10	5	0.5	-574	-3533	-189	-1018	-7247	8	-4.8	-6.1	-3.8	
	10	10	1	-571	-2033	-201	-1507	-9772	6	-4.9	-6.4	-3.9	
	20	0	0	-21	-204	20	-316	-4059	10	-4.9	-6.6	-3.8	
	20	10	0.5	-109	-424	-2	-698	-8275	5	-5.0	-6.4	-3.8	
	20	20	1	-203	-1093	-42	-1216	-18478	-12	-5.0	-6.6	-3.9	
	40	0	0	20	20	20	-3	-512	15	-4.1	-5.5	-2.9	
	40	20	0.5	20	20	20	-14	-606	13	-4.8	-6.2	-3.7	
	40	40	1	19	-23	20	-237	-1758	4	-4.8	-6.4	-3.8	

Smart Hospital

#robots	#malicious		median		φ_a			φ_b			$\varphi_c (\times 10^3)$		
	r	m	m/r	M	φ_a		M	φ_b		M	$\varphi_c (\times 10^3)$		
					Q_1	Q_3		Q_1	Q_3		Q_1	Q_3	
10	0	0	0	-558	-55929	-178	-650	-5661	9	-4.9	-6.5	-3.8	
10	5	0.5	0.5	-574	-3533	-189	-1018	-7247	8	-4.8	-6.1	-3.8	
10	10	1	1	-571	-2033	-201	-1507	-9772	6	-4.9	-6.4	-3.9	
20	0	0	0	-21	-204	20	-316	-4059	10	-4.9	-6.6	-3.8	
20	10	0.5	0.5	-109	-424	-2	-698	-8275	5	-5.0	-6.4	-3.8	
20	20	1	1	-203	-1093	-42	-1216	-18478	-12	-5.0	-6.6	-3.9	
40	0	0	0	20	20	20	-3	-512	15	-4.1	-5.5	-2.9	
40	20	0.5	0.5	20	20	20	-14	-606	13	-4.8	-6.2	-3.7	
40	40	1	1	19	-23	20	-237	-1758	4	-4.8	-6.4	-3.8	

Smart Hospital

#robots	#malicious		median		φ_a			φ_b			$\varphi_c (\times 10^3)$		
	r	m	m/r	M	φ_a		M	φ_b		M	$\varphi_c (\times 10^3)$		
					Q_1	Q_3		Q_1	Q_3		Q_1	Q_3	
10	0	0	0	-558	-55929	-178	-650	-5661	9	-4.9	-6.5	-3.8	
	10	5	0.5	-574	-3533	-189	-1018	-7247	8	-4.8	-6.1	-3.8	
	10	10	1	-571	-2033	-201	-1507	-9772	6	-4.9	-6.4	-3.9	
20	0	0	0	-21	-204	20	-316	-4059	10	-4.9	-6.6	-3.8	
	20	10	0.5	-109	-424	-2	-698	-8275	5	-5.0	-6.4	-3.8	
	20	20	1	-203	-1093	-42	-1216	-18478	-12	-5.0	-6.6	-3.9	
40	0	0	0	20	20	20	-3	-512	15	-4.1	-5.5	-2.9	
	40	20	0.5	20	20	20	-14	-606	13	-4.8	-6.2	-3.7	
	40	40	1	19	-23	20	-237	-1758	4	-4.8	-6.4	-3.8	

Smart Hospital

#robots	#malicious		median		φ_a			φ_b			$\varphi_c (\times 10^3)$		
	r	m	m/r	M	φ_a		M	φ_b		M	$\varphi_c (\times 10^3)$		
					Q_1	Q_3		Q_1	Q_3		Q_1	Q_3	
10	0	0	0	-558	-55929	-178	-650	-5661	9	-4.9	-6.5	-3.8	
	10	5	0.5	-574	-3533	-189	-1018	-7247	8	-4.8	-6.1	-3.8	
	10	10	1	-571	-2033	-201	-1507	-9772	6	-4.9	-6.4	-3.9	
20	0	0	0	-21	-204	20	-316	-4059	10	-4.9	-6.6	-3.8	
	20	10	0.5	-109	-424	-2	-698	-8275	5	-5.0	-6.4	-3.8	
	20	20	1	-203	-1093	-42	-1216	-18478	-12	-5.0	-6.6	-3.9	
40	0	0	0	20	20	20	-3	-512	15	-4.1	-5.5	-2.9	
	40	20	0.5	20	20	20	-14	-606	13	-4.8	-6.2	-3.7	
	40	40	1	19	-23	20	-237	-1758	4	-4.8	-6.4	-3.8	

Time robustness captures the resilience of a system to timing deviations