

# Generalized Barrier Functions: Integral Conditions & Recurrent Relaxations

Yue Shen, Hussein Sibai, [Enrique Mallada](#)

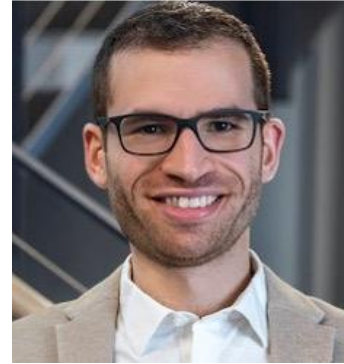


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September 25th, 2024

# Acknowledgements



Yue Shen

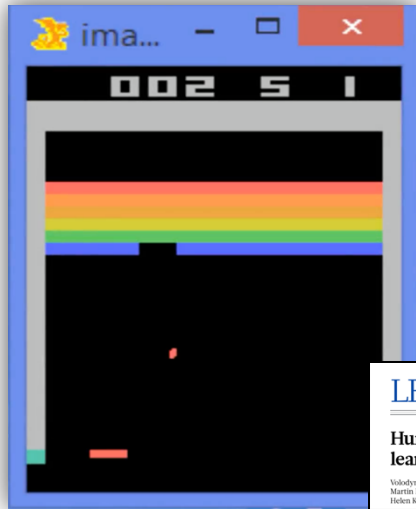


Hussein Sibai



# A World of Success Stories

2017 Google DeepMind's DQN



**LETTER**

doi:10.1038/nature14236

**Human-level control through deep reinforcement learning**

Vladimir Mnih<sup>1</sup>\*, Koray Kavukcuoglu<sup>2</sup>, David Silver<sup>1</sup>\*, Andrei A. Rusu<sup>1</sup>, Joel Veness<sup>1</sup>, Marc G. Bellemare<sup>1</sup>, Alex Graves<sup>1</sup>, Martin Riedmiller<sup>1</sup>, Andreas K. Faisal<sup>1</sup>, Georg Ostrovski<sup>1</sup>, Stig Petersen<sup>1</sup>, Charles Beattie<sup>1</sup>, Amir Sadik<sup>1</sup>, Ioannis Antonoglou<sup>1</sup>, Helen King<sup>1</sup>, Dhruv Kumaram<sup>1</sup>, Duan Wu<sup>1</sup>, Shuan Liang<sup>1</sup> & Demis Hassabis<sup>1</sup>

2017 AlphaZero – Chess, Shogi, Go



Boston Dynamics



2019 AlphaStar – Starcraft II



**Article**

**Grandmaster level in StarCraft II using multi-agent reinforcement learning**

<https://doi.org/10.1038/41586-019-1724-z>

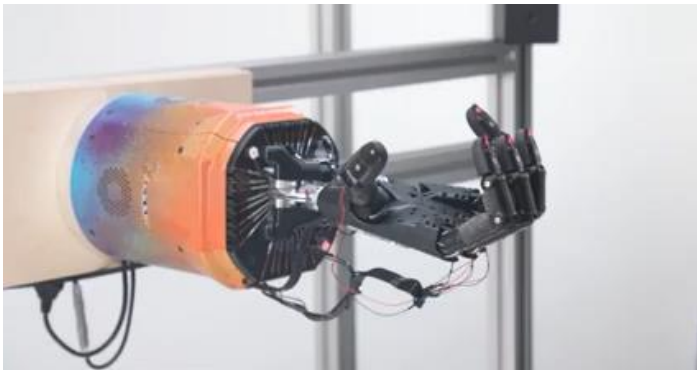
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Oleks Vinyals<sup>1</sup>\*, Igor Belobachkin<sup>1</sup>, Wojciech M. Czarnecki<sup>1</sup>, Michael Mattheu<sup>1</sup>, Andrew Dudzik<sup>1</sup>, Junyoung Chung<sup>1</sup>, David H. Choi<sup>1</sup>, Richard Powell<sup>1</sup>, Timo Schaul<sup>1</sup>, Perio Georgiev<sup>1</sup>, Junhyuk Oh<sup>1</sup>, Dan Horgan<sup>1</sup>, Manuel Krotke<sup>1</sup>, Ivo Danihelba<sup>1</sup>, Aki Huang<sup>1</sup>, Laurent Sifre<sup>1</sup>, Trevor Cai<sup>1</sup>, John F. Agapiou<sup>1</sup>, Alex Jankelberg, Alexander S. Wehner<sup>1</sup>, Alexey Levins<sup>1</sup>, Tobias Pohlen<sup>1</sup>, Valentin Dalibard<sup>1</sup>, David Budden<sup>1</sup>, Yury Sulsky<sup>1</sup>, James Molloy<sup>1</sup>, Tom L. Paine<sup>1</sup>, Caglar Gulcehre<sup>1</sup>, Ziyu Wang<sup>1</sup>, Tobias Pfaff<sup>1</sup>, Yakov Vevy<sup>1</sup>, Roman Ring<sup>1</sup>, Dani Yogatama<sup>1</sup>, Dario Wascel<sup>1</sup>, Karina Madhavi<sup>1</sup>, Oliver Smith<sup>1</sup>, Tom Schaul<sup>1</sup>, Timothy Lillicrap<sup>1</sup>, Koray Kavukcuoglu<sup>1</sup>, Demis Hassabis<sup>1</sup>, Chris Apps<sup>1</sup> & David Silver<sup>1</sup>\*

OpenAI – Rubik's Cube



Waymo



# The Need for Safety Guarantees

## Angry Residents, Abrupt Stops: Waymo Vehicles Are Still Causing Problems in Arizona

RAY STERN | MARCH 31, 2021 | 8:26AM

GARY MARCUS | BUSINESS | 08.14.2019 09:00 AM

## DeepMind's Losses and the Future of Artificial Intelligence

Alphabet's DeepMind unit, conqueror of Go and other games, is losing lots of money. Continued deficits could imperil investments in AI.

AARIAN MARSHALL | BUSINESS | 12.07.2020 04:06 PM

## Uber Gives Up on the Self-Driving Dream

The ride-hail giant invested more than \$1 billion in autonomous vehicles. Now it's selling the unit to Aurora, which makes self-driving tech.

## Tesla Recalls Nearly All Vehicles Due to Autopilot Failures

Tesla disagrees with feds' analysis of glitches

BY LINA FISHER, 2:54PM, WED. DEC. 13, 2023

## CRUISE KNEW ITS SELF-DRIVING CARS HAD PROBLEMS RECOGNIZING CHILDREN — AND KEPT THEM ON THE STREETS

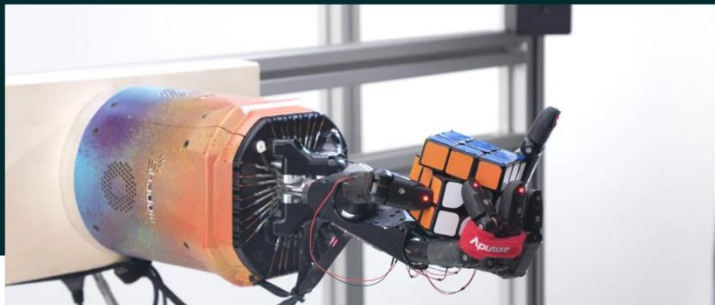
According to internal materials reviewed by The Intercept, Cruise cars were also in danger of driving into holes in the road.



## OpenAI disbands its robotics research team

Kyle Wiggers | @Kyle\_L\_Wiggers | July 16, 2021 11:24 AM

f t in



## Self-driving Uber car that hit and killed woman did not recognize that pedestrians jaywalk

The automated car lacked "the capability to classify an object as a pedestrian unless that object was near a crosswalk," an NTSB report said.



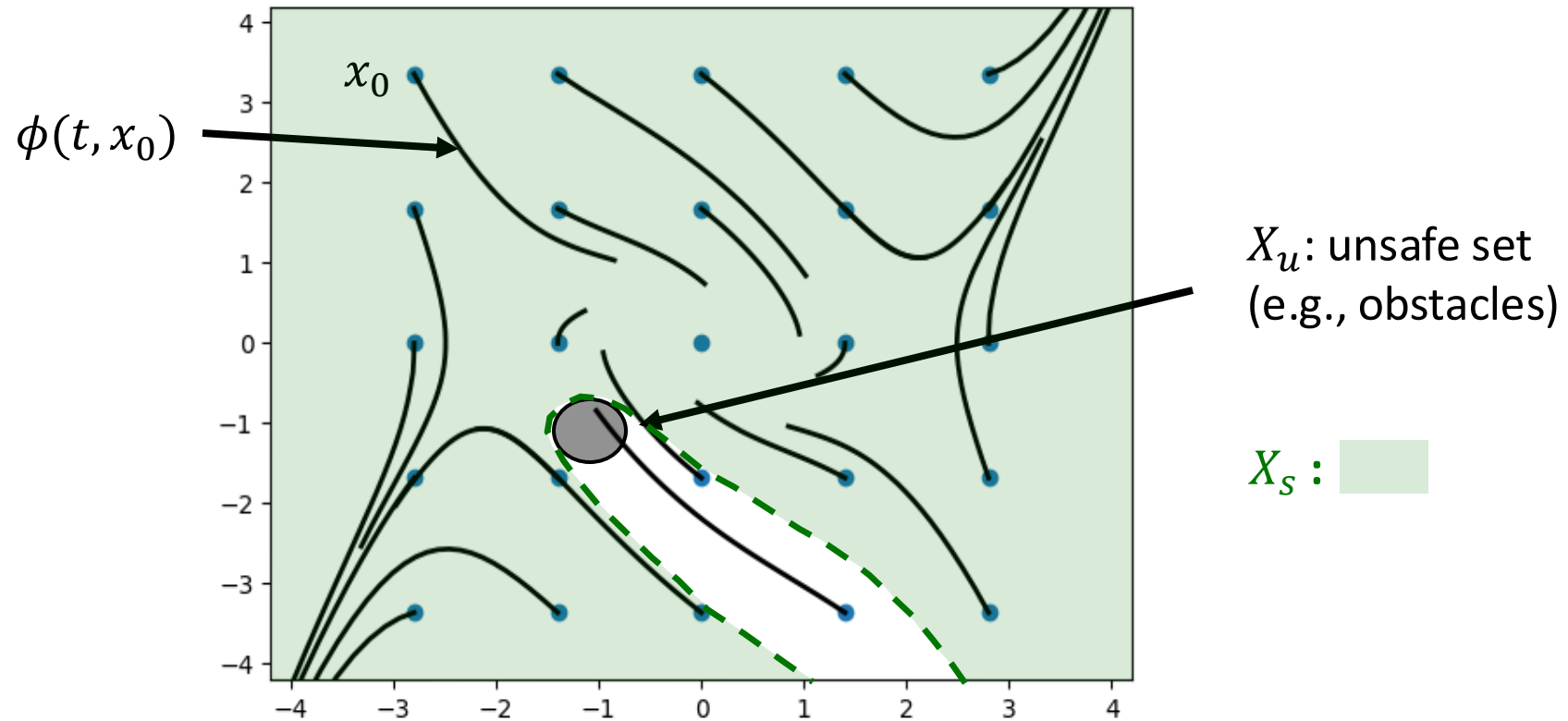


# Safety in Dynamical Systems

Consider the continuous-time dynamical system:  $\dot{x} = f(x)$

- $\phi(t, x_0)$ : solution at time  $t$  starting from  $x_0$
- $X_u$ : set of unsafe states

**Goal:** Find the safe set  $X_s := \{x_0 \in \mathbb{R}^d \mid \phi(t, x_0) \notin X_u, \forall t \geq 0\}$



# Safety in Dynamical Systems **via Invariant Sets**

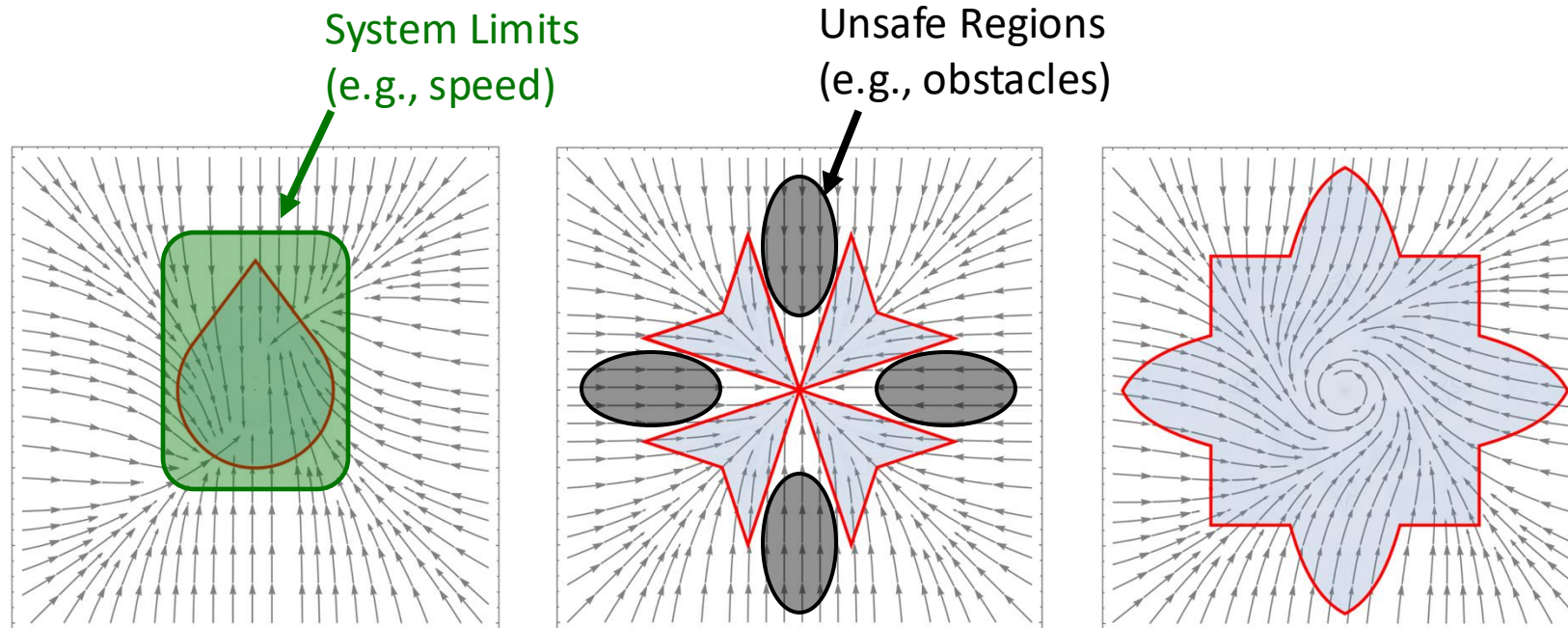
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**Goal:** Find the safe set  $X_s := \{x_0 \in \mathbb{R}^d \mid \phi(t, x_0) \notin X_u, \forall t \geq 0\}$

**General Approach:** Use invariant sets!

A set  $S \subseteq \mathbb{R}^d$  is **invariant** if and only if:  $x_0 \in S \rightarrow \phi(t, x_0) \in S, \forall t \geq 0$



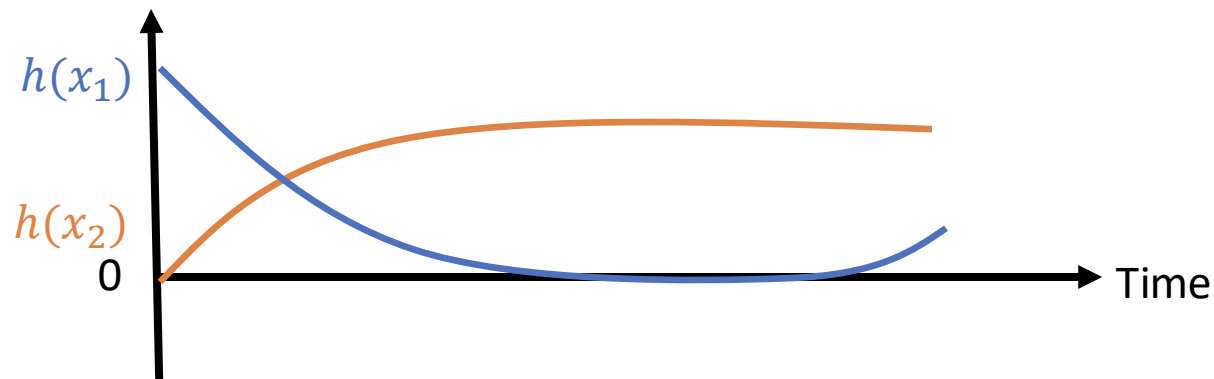
# Certifying Safety using Barrier Functions

## Theorem - Nagumo's Barrier Functions [Nagumo '42] :

Let  $h: \mathbb{R}^d \rightarrow \mathbb{R}$  be differentiable, with 0 being a *regular value*.  
Then  $h$  is a Nagumo's Barrier Function (NBF) satisfying:

$$L_f h(x) := \lim_{t \rightarrow 0} \frac{h(\phi(t, x)) - h(x)}{t} \geq 0, \quad \forall x \in h_{=0},$$

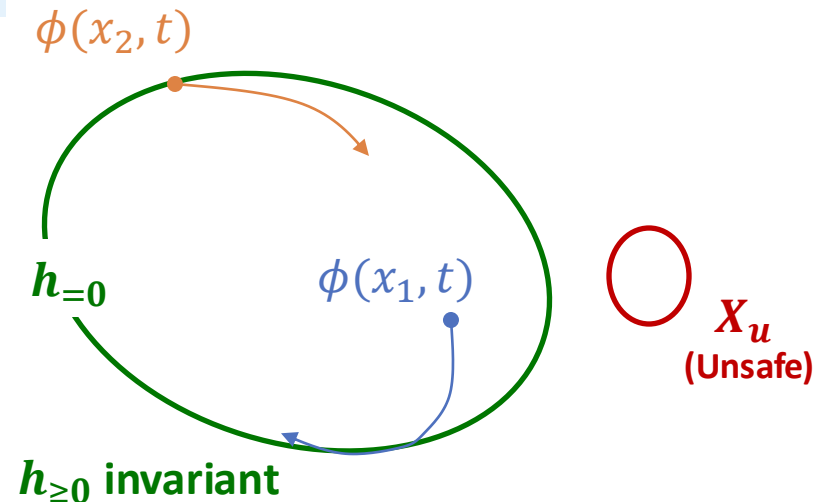
if and only if  $h_{\geq 0} := \{x \in \mathbb{R}^d \mid h(x) \geq 0\}$  is invariant.



Then  $h_{\geq 0}$  is a safe set whenever  $h_{\geq 0} \cap X_u = \emptyset$



Mitio Nagumo



# Shaping Safe Behavior using Barrier Functions (BFs)

Barrier functions provide a flexible framework to shape the behavior of trajectories

**Nagumo's (NBF)**

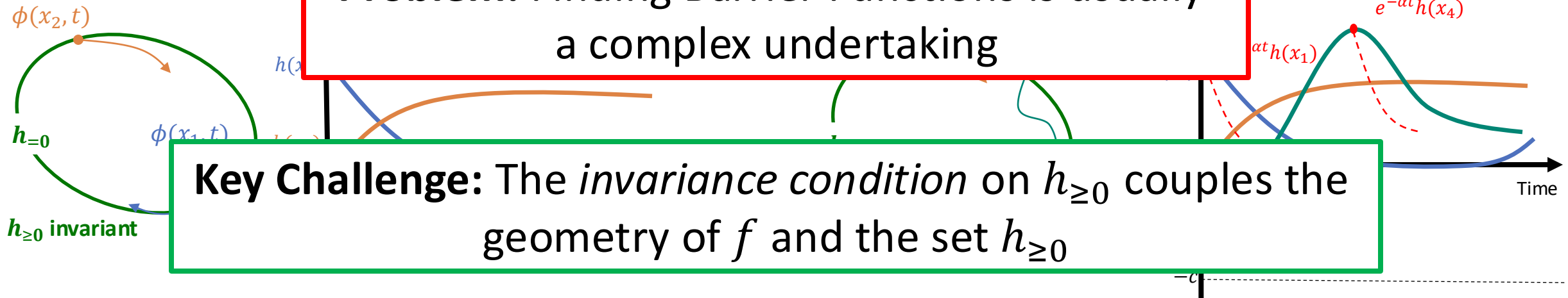
$$L_f h(x) \geq 0, \quad \forall x \in h=0$$

**Exponential Barrier Functions (EBF)**

$$L_f h(x) \geq -\alpha h(x), \quad \forall x \in h \geq -c$$

**Problem:** Finding Barrier Functions is usually a complex undertaking

**Key Challenge:** The *invariance condition* on  $h_{\geq 0}$  couples the geometry of  $f$  and the set  $h_{\geq 0}$



**Other:** Zeroing BFs (ZBFs), Minimal BFs (MBFs), Control BFs (CBFs), High Order CBFs (HOCBFs), ...

S. Prajna, A. Jadbabaie. *Safety Verification of Hybrid Systems Using Barrier Certificates*. HSCC 2004

P. Wieland, F. Allgöwer. *Constructive safety using control barrier functions*. IFAC Proceedings Volumes 2007

A. Ames, S. Coogan, M. Egerstedt, G. Notomista, K. Sreenath, P. Tabuada. *Control barrier functions: Theory and applications*. IEEE ECC 2019

R. Konda, A. Ames, S. Coogan. *Characterizing safety: Minimal control barrier functions from scalar comparison systems*. IEEE L-CSS 2020

W. Xiao, C. Belta. *High-order control barrier functions*. IEEE TAC 2021



# Outline

- Letting things go and come back: *Recurrent sets*
- Generalized barriers: Integral forms and recurrent relaxations
- Safety verification via Recurrent Barrier Functions

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- **Letting things go and come back: *Recurrent sets***
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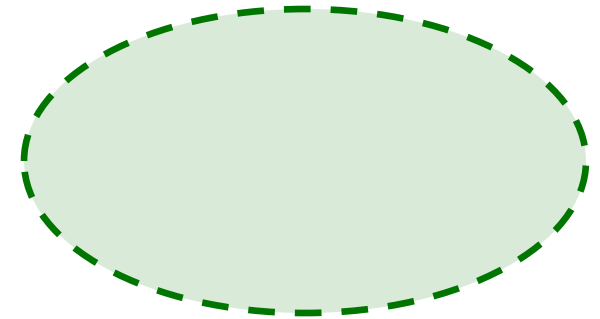
# Core challenge: Invariant safe sets are hard to find

A set  $\mathcal{S} \subseteq \mathbb{R}^d$  is **positively invariant** if and only if:  $x_0 \in \mathcal{S} \rightarrow \phi(t, x_0) \in \mathcal{S}, \forall t \geq 0$

Any trajectory starting in the set remains in inside it for all times

**Goal:** Given an unknown safe set  $X_S$ , find an invariant set  $\mathcal{S}$  such that  $\mathcal{S} \subset X_S$ .

Example:



$X_S : \text{[ ]}$

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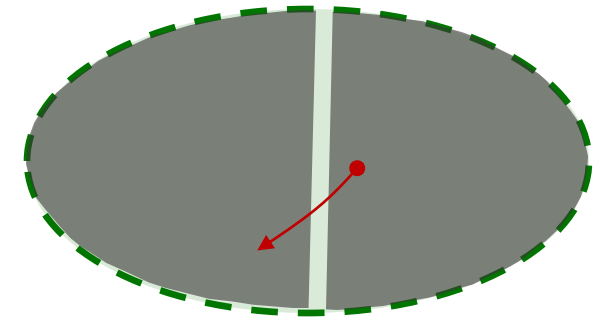
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## Challenges:

- $\mathcal{S}$  is topologically constrained
  - Trajectory cannot cross disconnected parts of  $\mathcal{S}$

**Example:**

$\mathcal{S} \subseteq X_S$  is not invariant!



$X_S$  : [ ]     $\mathcal{S}$  : [ ]

**A not invariant trajectory:** [ ]

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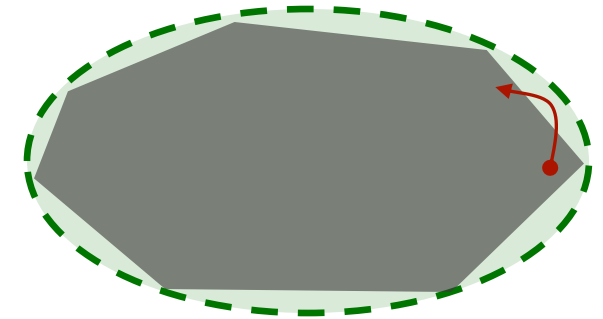
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  - $f$  should point inwards for  $x \in \partial\mathcal{S}$

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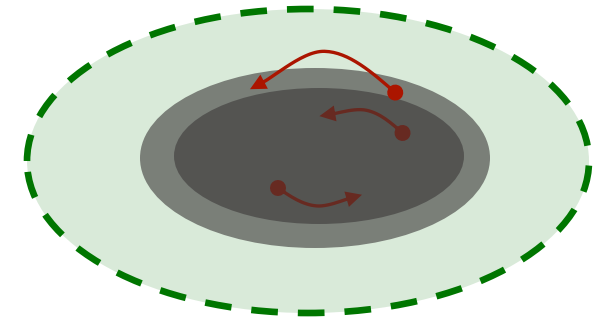
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- $\mathcal{S}$  is geometrically constrained
  - $f$  should point inwards for  $x \in \partial\mathcal{S}$
- $\mathcal{S}$  is hard to grow
  - $\mathcal{S}$  should adapt to new trajectories

Invariance introduces strict constraints on the shape, topology, and the future extension of the set  $\mathcal{S}$  !

**Example:**

$\mathcal{S} \subseteq X_S$  is not invariant!



$X_S$  : [ ]     $\mathcal{S}$  : [ ]

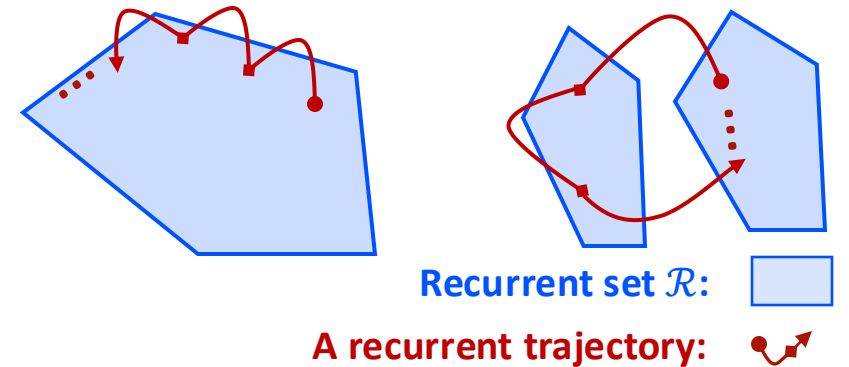
A not invariant trajectory: [ ]

# Recurrent sets: Letting things go, and come back

A set  $\mathcal{R} \subseteq \mathbb{R}^d$  is **recurrent** if for any  $x_0 \in \mathcal{R}$  and  $t \geq 0$ ,  $\exists t' \geq t$  s.t.  $\phi(t', x_0) \in \mathcal{R}$ .

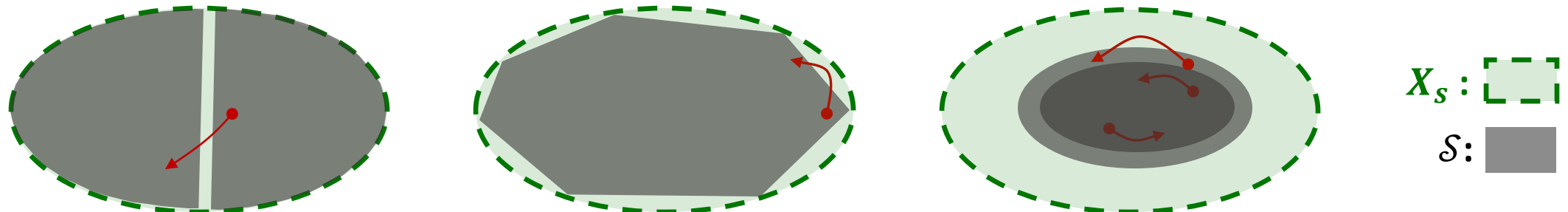
## Property of Recurrent Sets

- $\mathcal{R}$  need **not** be **connected**
- $\mathcal{R}$  does **not** require  $f$  to **point inwards** on all  $\partial\mathcal{R}$



Recurrent sets, while not invariant, guarantee that solutions that start in this set, will come back **infinitely often, forever!**

Previous good inner approximations of  $X_S$  are recurrent sets



# Outline

- Letting things go and come back: *Recurrent sets*
- **Generalized barriers: Integral forms and recurrent relaxations**
- Safety verification via Recurrent Barrier Functions

# Integral Nagumo's Barrier Function (INBF)

**Nagumo's Barrier Functions** [Nagumo '42]:  
Let  $h$  be differentiable, regular at zero, and

$$L_f h(x) \geq 0, \quad \forall x \in h_{=0}$$

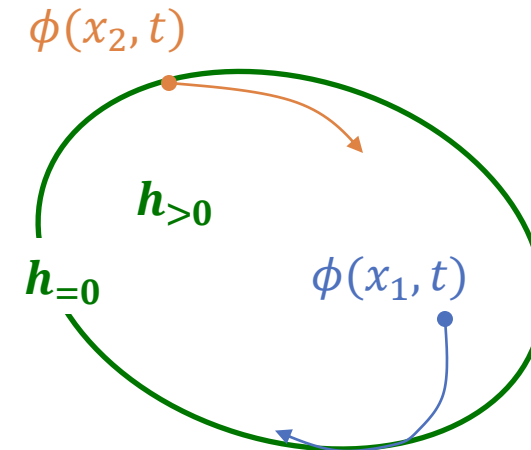
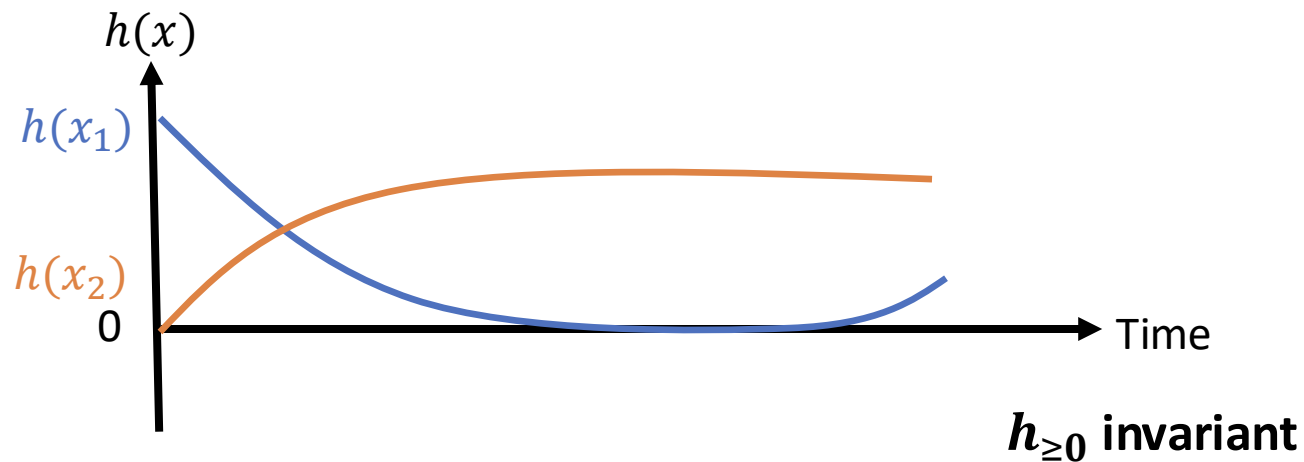
**Integral Nagumo's Barrier Function:**  
Let  $h$  be continuous, and

$$h(\phi(t, x)) \geq 0, \quad \forall x \in h_{=0}, t \geq 0$$



The super-level set  $h_{\geq 0}$  is invariant.

\* ← - - -  
requires more conditions on  $h$



# Recurrent Nagumo's Barrier Function (RNBF)

**Thm: Integral Nagumo's Barrier Function:**

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if and only if  $h_{\geq 0}$  is **invariant**

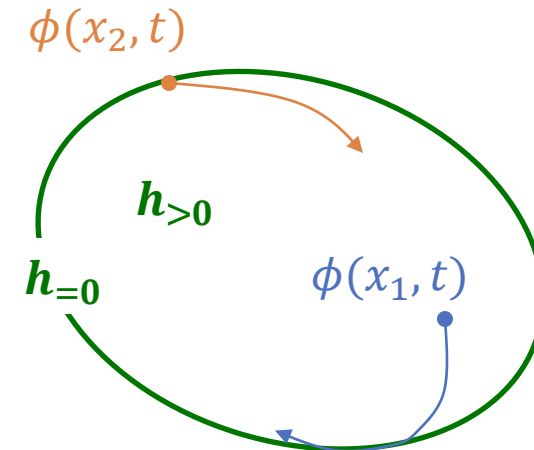
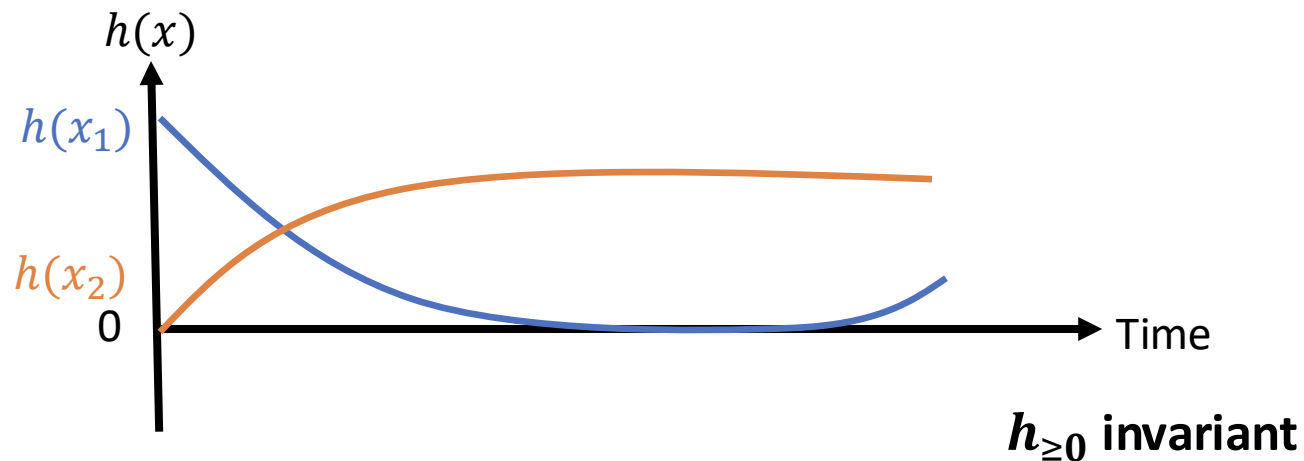
**Thm: Recurrent Nagumo's Barrier Function:**

Let  $h$  be continuous. Then:

$$\max_{t \in (0, \tau]} h(\phi(t, x)) \geq 0, \forall x \in h_{=0}$$

if and only if  $h_{\geq 0}$  is  **$\tau$ -recurrent**

**Definition:** A set  $\mathcal{R} \subseteq \mathbb{R}^d$  is  **$\tau$ -recurrent** if for any  $x_0 \in \mathcal{R}$  and  $t \geq 0$ ,  $\exists t' \in (t, t + \tau]$  s.t.  $\phi(t', x_0) \in \mathcal{R}$ .





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As  $\tau \rightarrow 0$   
 $\longleftrightarrow$   
 By definition

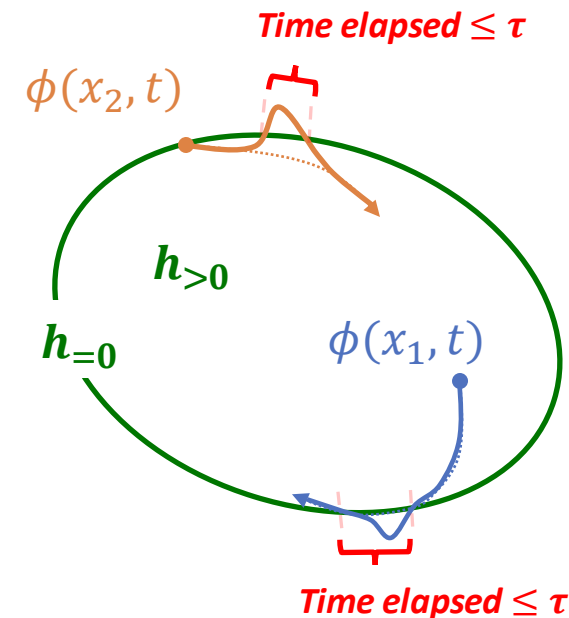
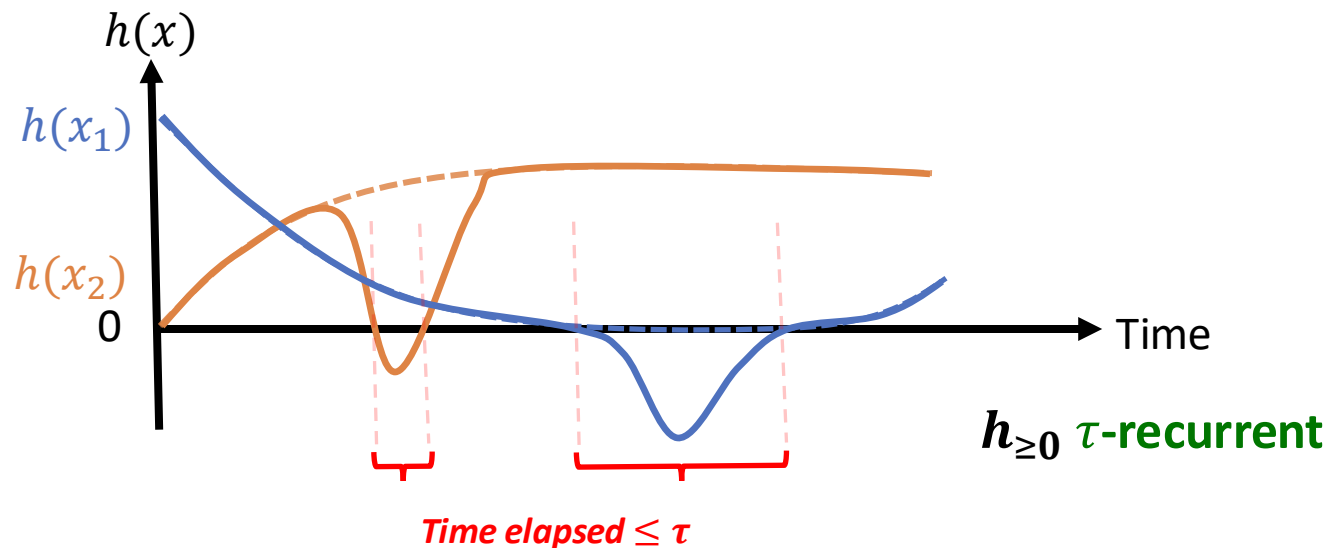
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# Integral Exponential Barrier Function (IEBF)

## Exponential Barrier Functions:

Let  $h$  be differentiable, and

$$L_f h(x) \geq -\alpha h(x), \quad \forall x \in h_{\geq -c}$$

## Integral Exponential Barrier Function:

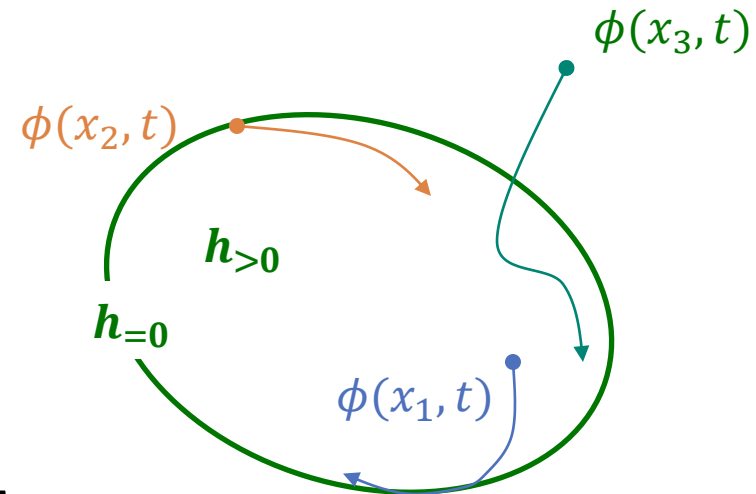
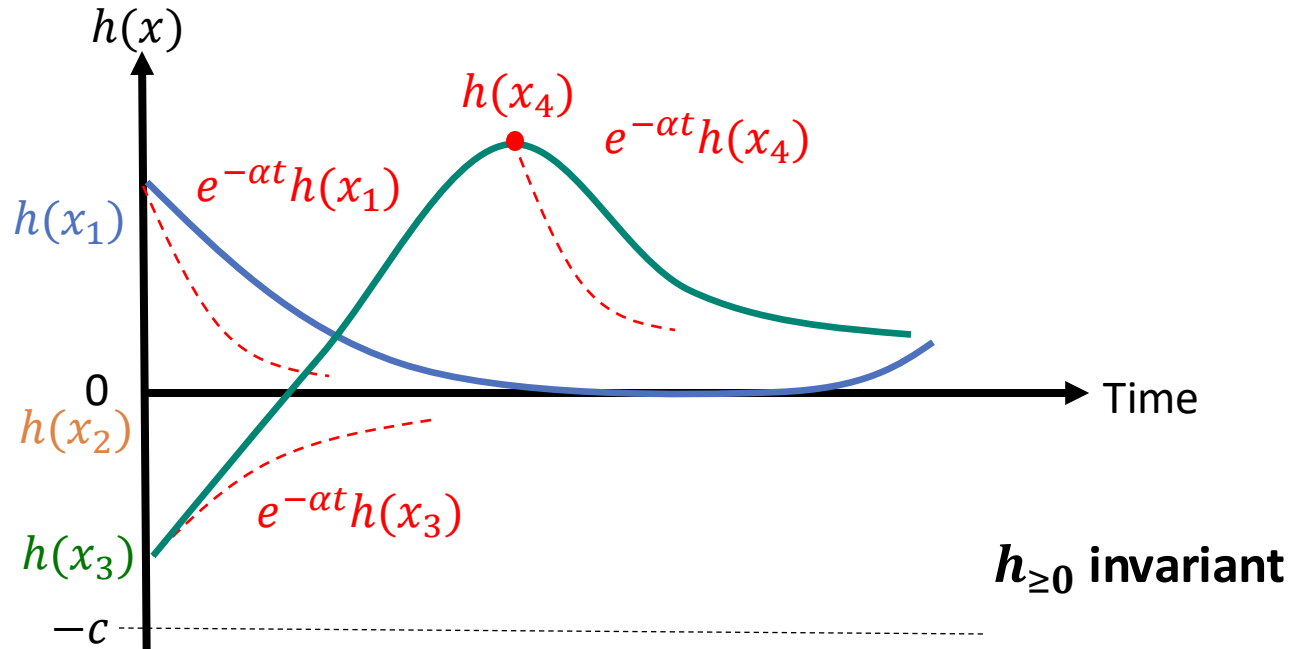
Let  $h$  be continuous, and

$$h(\phi(t, x)) \geq e^{-\alpha t} h(x), \quad \forall x \in h_{\geq -c}$$

for all  $t \geq 0$



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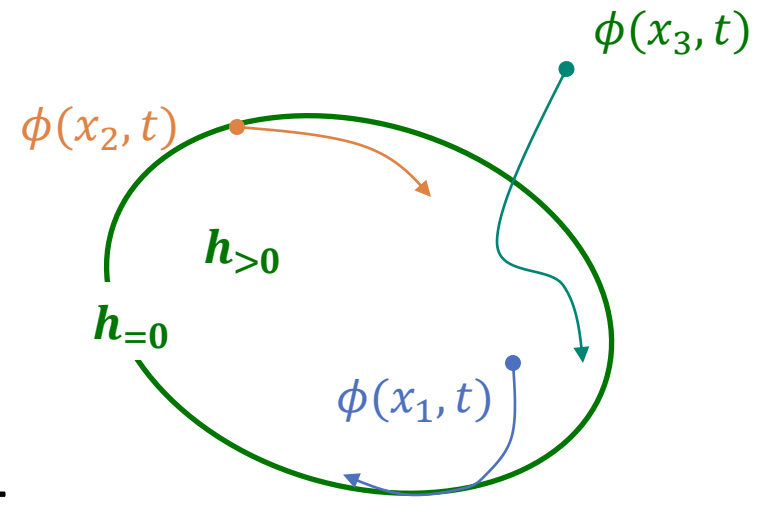
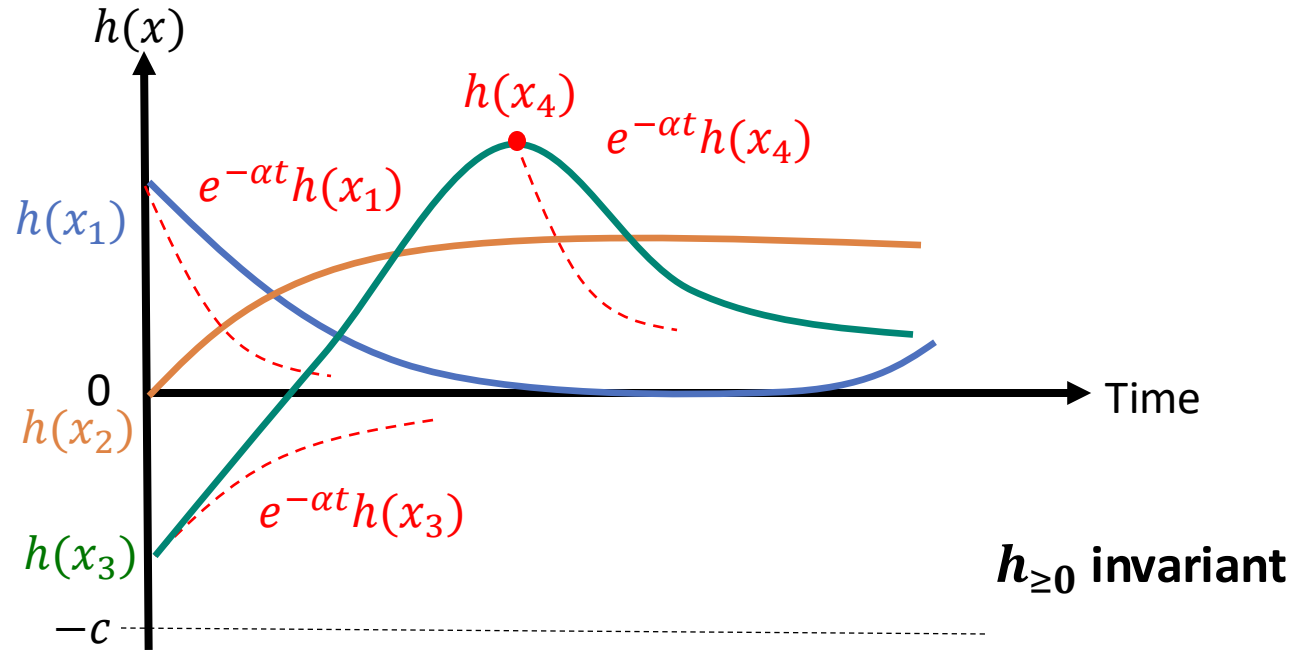
for all  $t \geq 0$ , then,  $h_{\geq 0}$  is **invariant**

**Thm: Recurrent Exponential Barrier Function:**

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$$\max_{t \in (0, \tau]} e^{\alpha t} h(\phi(t, x)) \geq h(x), \quad \forall x \in h_{\geq -c}$$

then,  $h_{\geq 0}$  is  **$\tau$ -recurrent**



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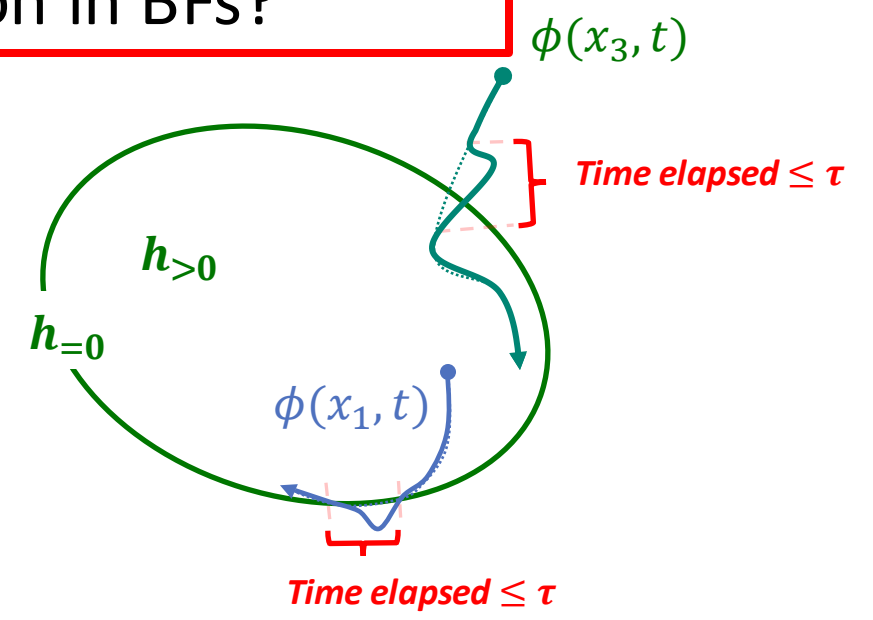
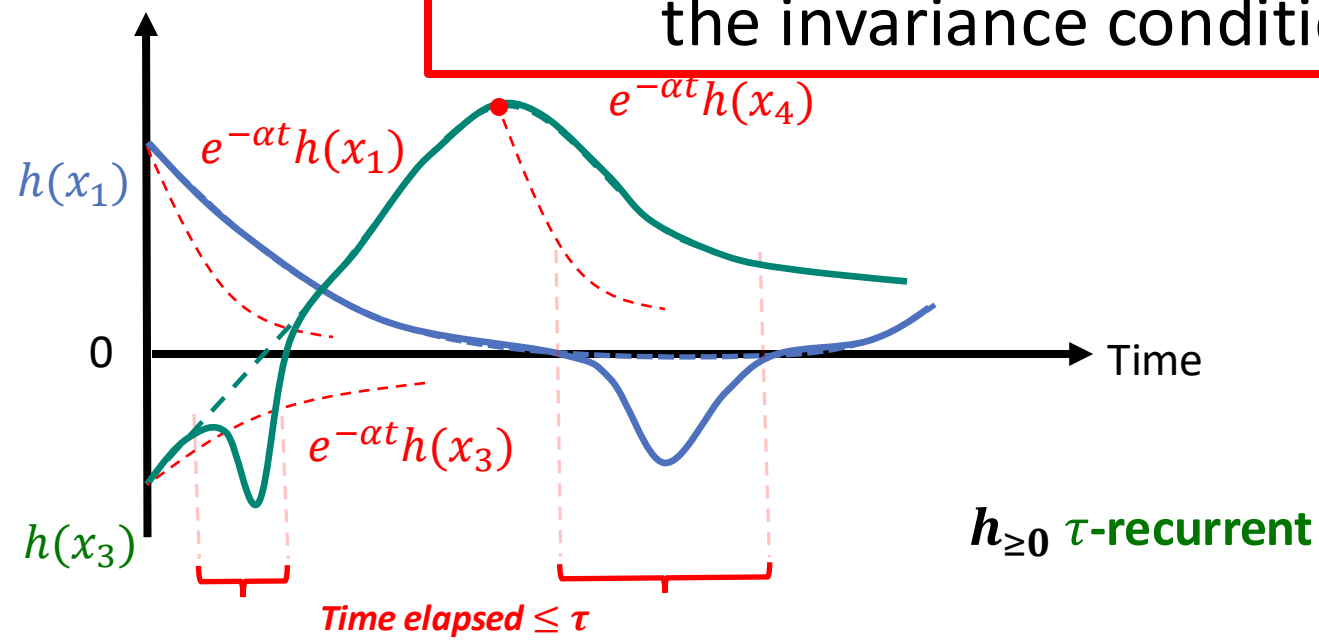
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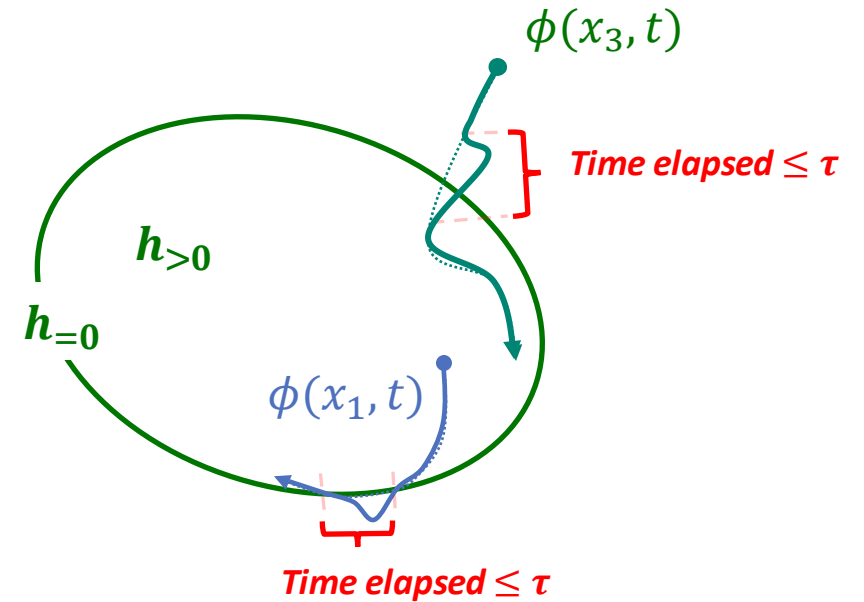
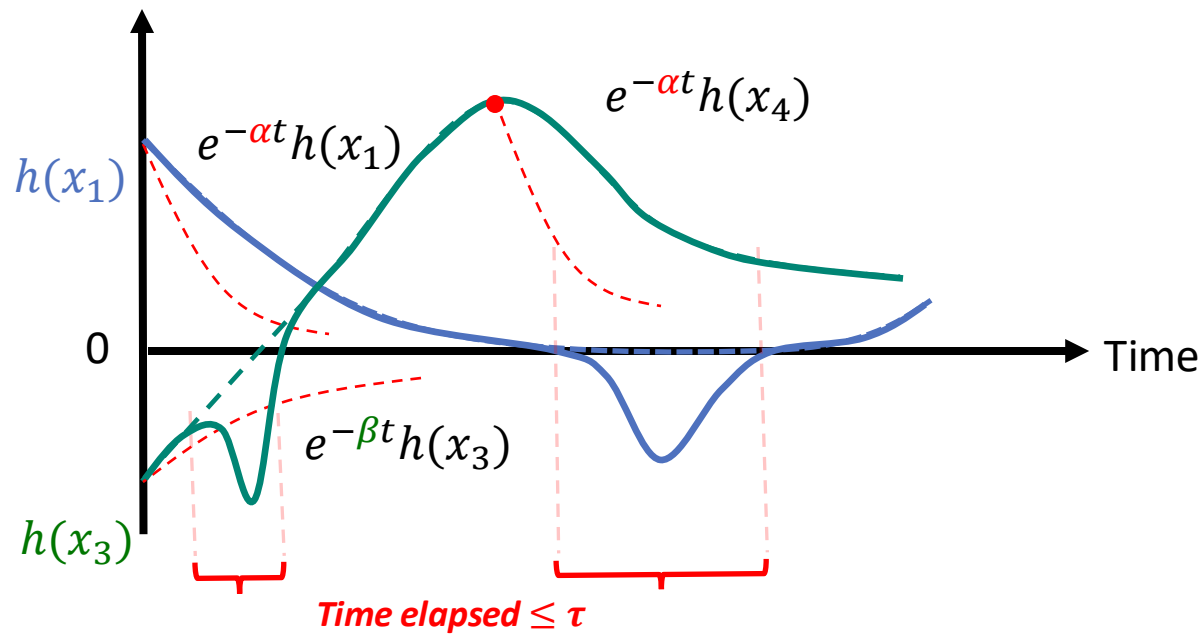
**Question:** Do we gain anything from relaxing the invariance condition in BFs?



# Signed Norms are Recurrent Barrier Functions!

We first generalize REBF using different exponential rates  $\alpha, \beta > 0$ :

$$\max_{t \in (0, \tau]} e^{\alpha t} [h(\phi(t, x))]_+ + e^{\beta t} [h(\phi(t, x))]_- \geq h(x), \quad \forall x \in h_{\geq -c}$$





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**Theorem:** Assume there exists an **Integral Exponential BF (IEBF)**,  $h$ , defined over  $D_0 := h_{\geq -c}$  for some  $c > 0$ . Then  $\exists \alpha > 0$  such that:

$$e^{\alpha t} h(\phi(t, x)) \geq h(x), \quad \forall x \in h_{\geq -c}$$

for all  $t \geq 0$ .

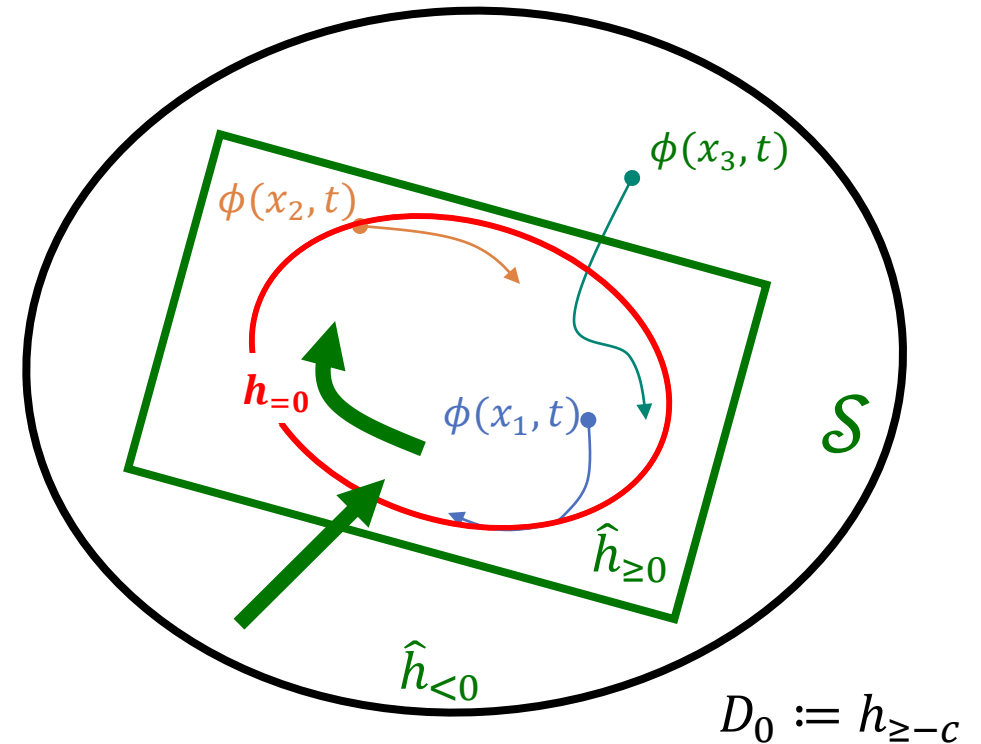
Then for any set  $\mathcal{S}$  with  $h_{\geq 0} \subseteq \mathcal{S} \subseteq h_{\leq -c}$ , the function

$$\hat{h}(x) := -\text{sd}(x, \mathcal{S})$$

is a **Recurrent Exponential Barrier Function (REBF)**:

$$\max_{t \in (0, \tau]} e^{\hat{\alpha} t} [h(\phi(t, x))]_+ + e^{\hat{\beta} t} [h(\phi(t, x))]_- \geq h(x), \quad \forall x \in h_{\geq -c}$$

with any parameters  $\hat{\beta} < \alpha < \hat{\alpha}$  whenever  $\tau \geq \bar{\tau}(\hat{\alpha} - \alpha, \hat{\beta} - \alpha)$



# Outline

- Letting things go and come back: *Recurrent sets*
- Generalized barriers: Integral forms and recurrent relaxations
- **Safety verification via Recurrent Barrier Functions**

# Safety Verification via Recurrent Sets

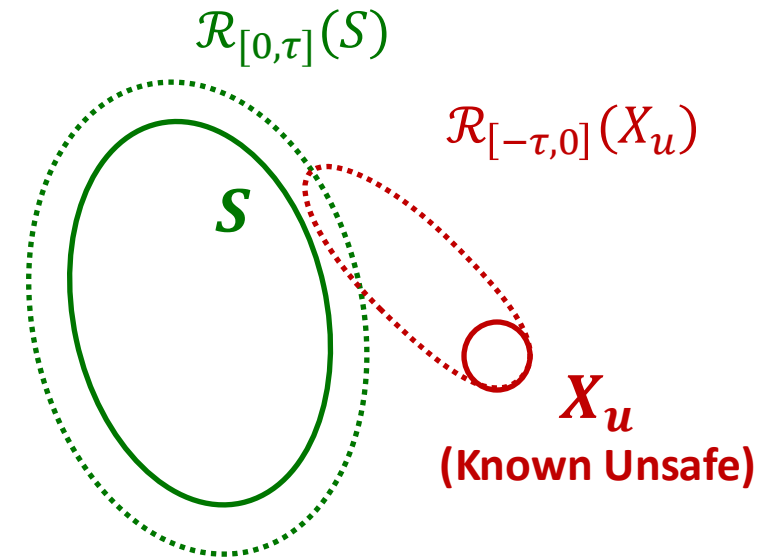
**Theorem** - Consider a closed set  $S$  that is  $\tau$ -recurrent.  
Then its  $\tau$ -reachable set:



is **invariant**.

Moreover,  $S$  is **safe** whenever:

1.  $\mathcal{R}_{[0,\tau]}(S) \cap X_u = \emptyset$ ,
2.  $S \cap \mathcal{R}_{[-\tau,0]}(X_u) = \emptyset$



# A GPU based algorithm

A set  $S$  is safe whenever:

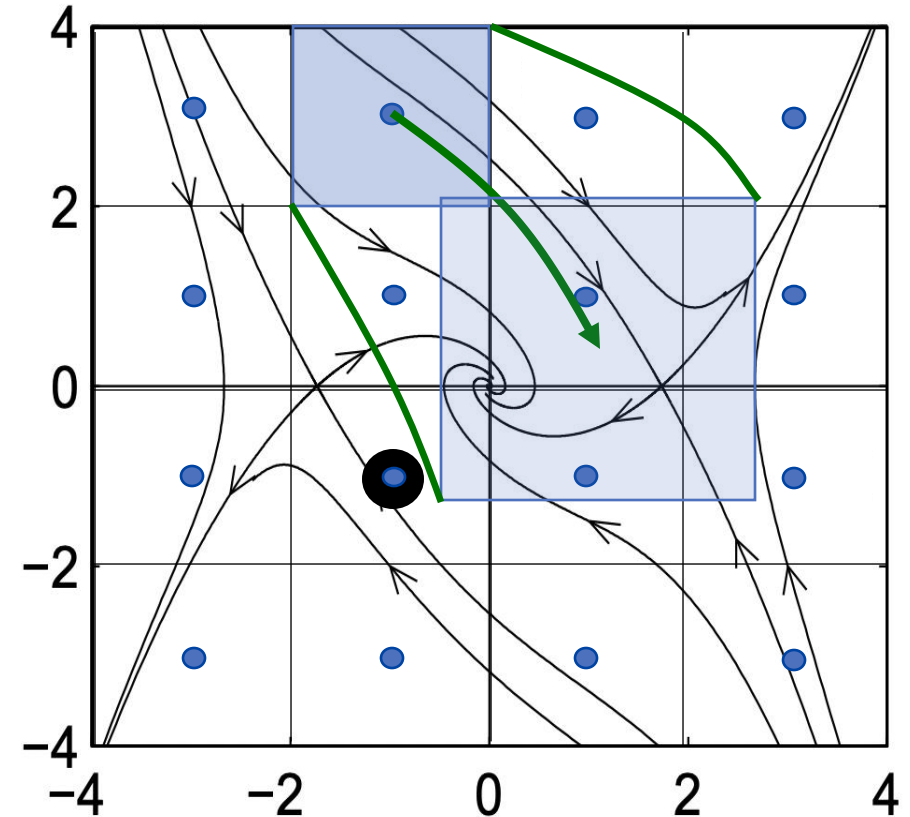
## Reachability Condition

- $\mathcal{R}_{[0,\tau]}(S) \cap X_u = \emptyset$

## Recurrent Condition

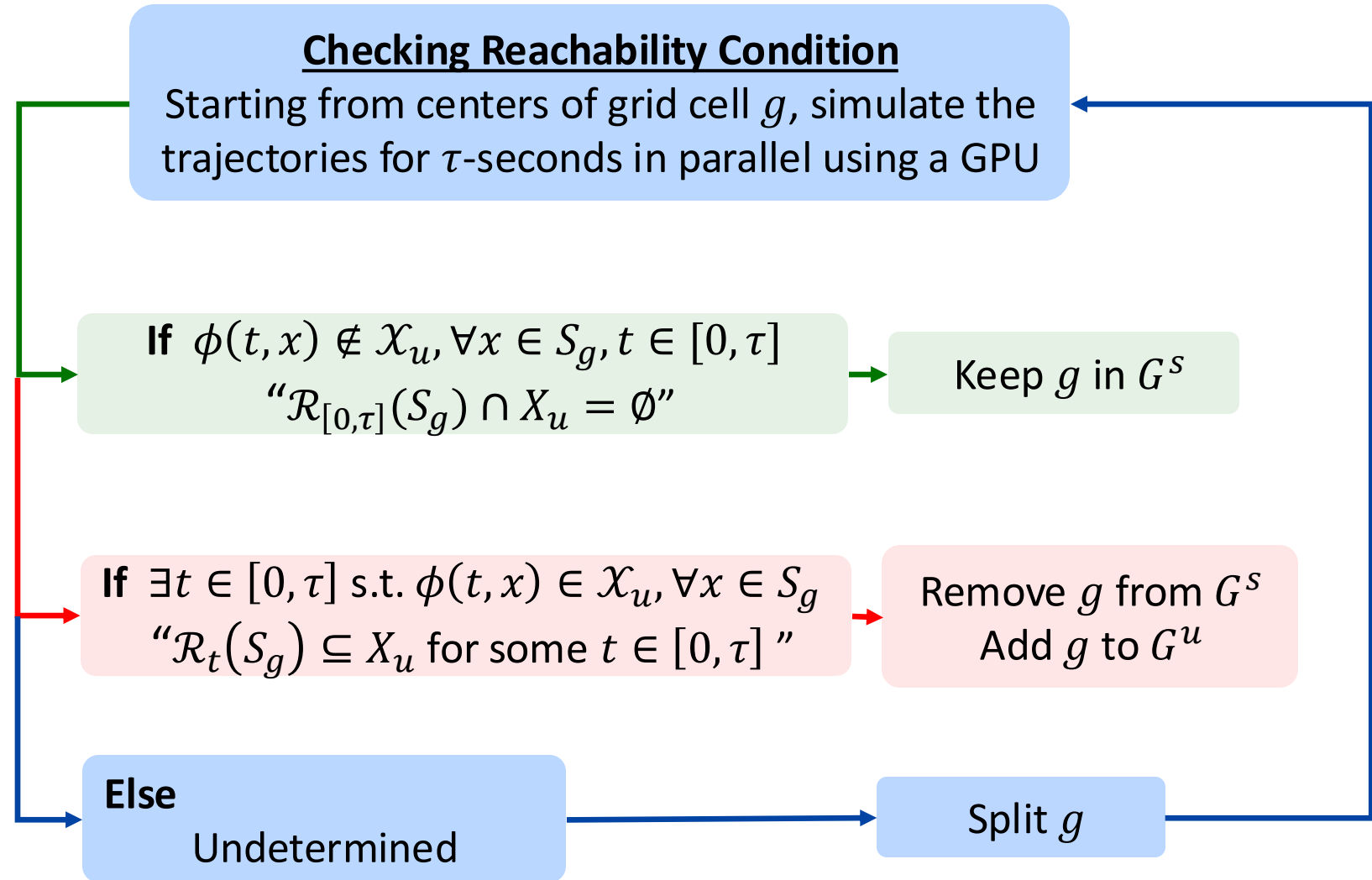
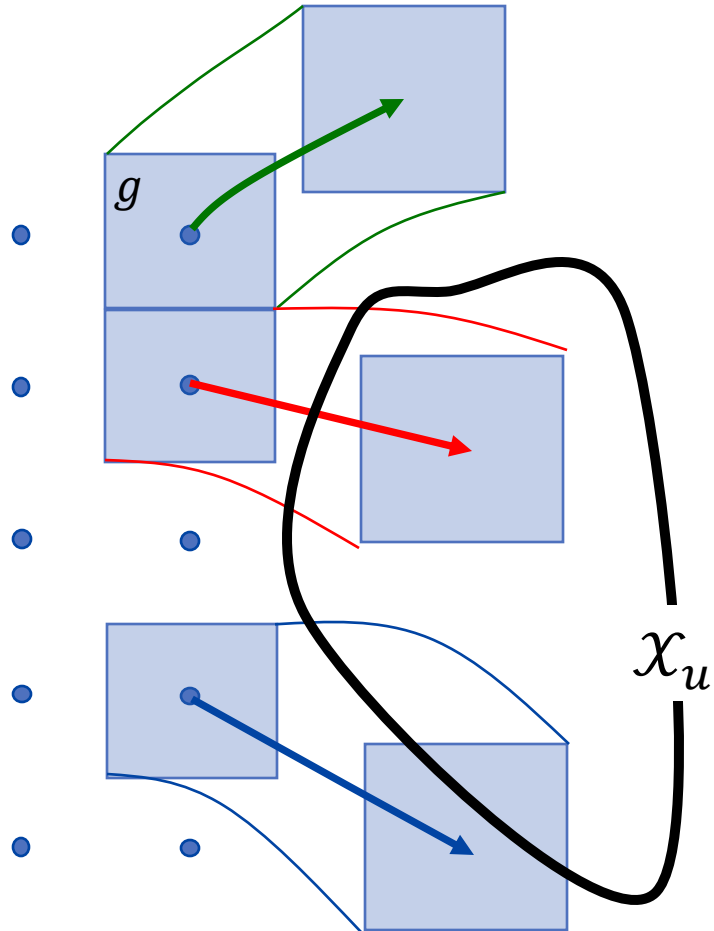
- $\hat{h}(x) := -sd(x, S)$  is a Recurrent Exponential Barrier Function

- Cover the region with a grid  $G$ 
  - For each point  $g \in G$ ,  $S_g$  represents its cell
- We build  $S = \bigcup_{g \in G^S} S_g$ , with  $G^S$  representing safe points
  - Initialize  $G^S \leftarrow G$
- Check both conditions using only one trajectory for each cell!

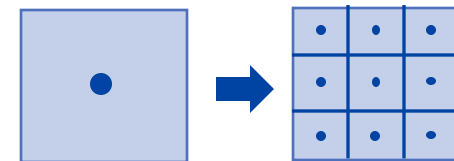


$X_u$  :

# A GPU based algorithm



\*Stop splitting  $g$  and mark it as unsafe whenever  $g$  is too small



# A GPU based algorithm

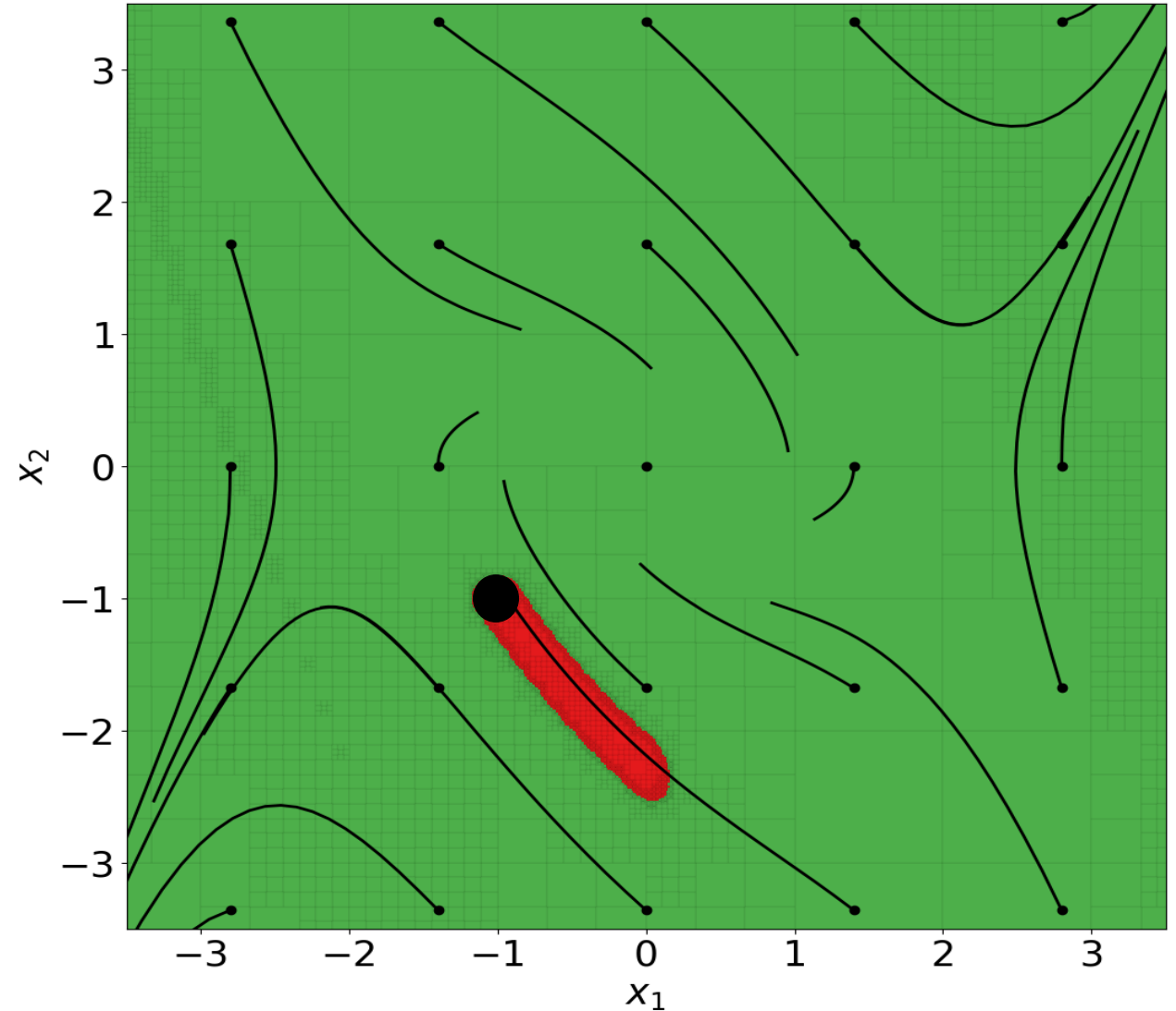
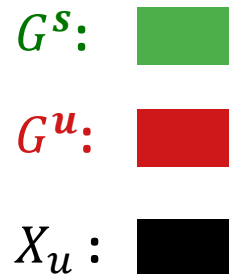
A set  $S = \bigcup_{S \in G^s} S_g$  is safe whenever:

## Reachability Condition

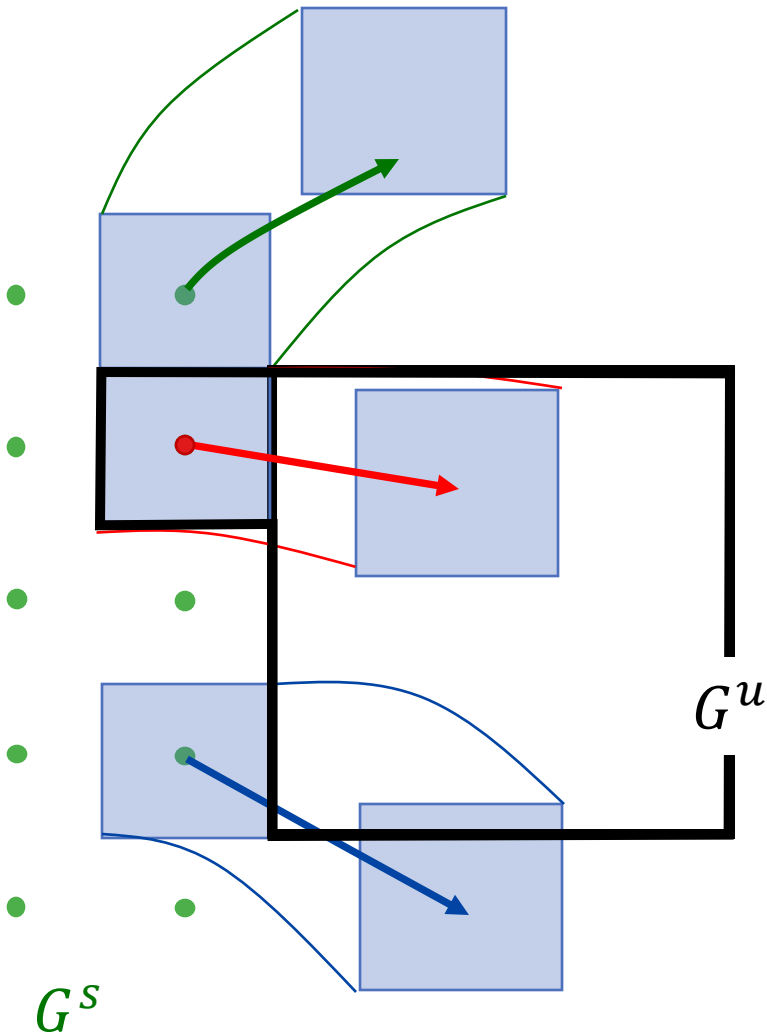
- $\mathcal{R}_{[0,\tau]}(S) \cap X_u = \emptyset$

## Recurrent Condition

- $\hat{h}(x) := -\text{sd}(x, S)$  is a Recurrent exponential barrier function



# A GPU based algorithm



## Check the Recurrent condition

- Let  $S = \cup_{g \in G^s} S_g$ , ██████████
- Starting from centers of grid cell  $g \in G^s$ , simulate the trajectories for  $\tau$ -seconds in parallel using a GPU

If  $\max_{t \in (0, \tau]} e^{\hat{\alpha} t} \hat{h}(\phi(t, x)) \geq \hat{h}(x), \forall x \in S_g$   
 "REBF condition is satisfied within  $g$ "

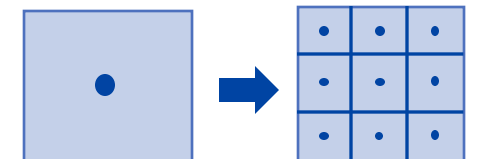
Keep  $g$  in  $G^s$

If  $\max_{t \in (0, \tau]} e^{\hat{\alpha} t} \hat{h}(\phi(t, x)) < \hat{h}(x), \forall x \in S_g$   
 "REBF condition is NOT satisfied within  $S_g$ "

Remove  $g$  from  $G^s$   
 Add  $g$  to  $G^u$

Else  
 Undetermined

Split  $g$



\*Stop splitting  $g$  and mark it as unsafe whenever  $g$  is too small

# A GPU based algorithm

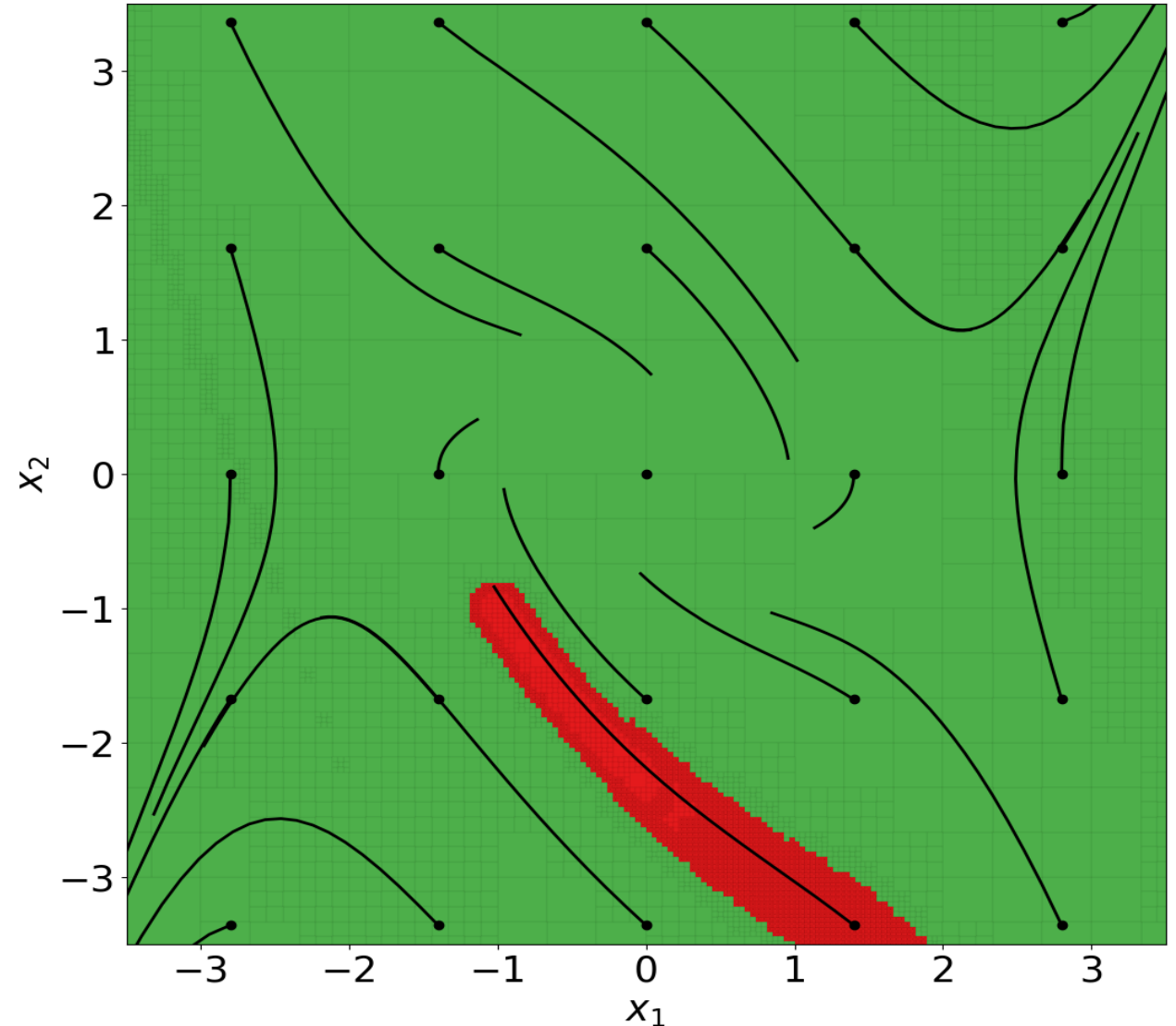
A set  $S = \bigcup_{S \in G^s} S_g$  is safe whenever:

## Reachability Condition

- $\mathcal{R}_{[0,\tau]}(S) \cap X_u = \emptyset$

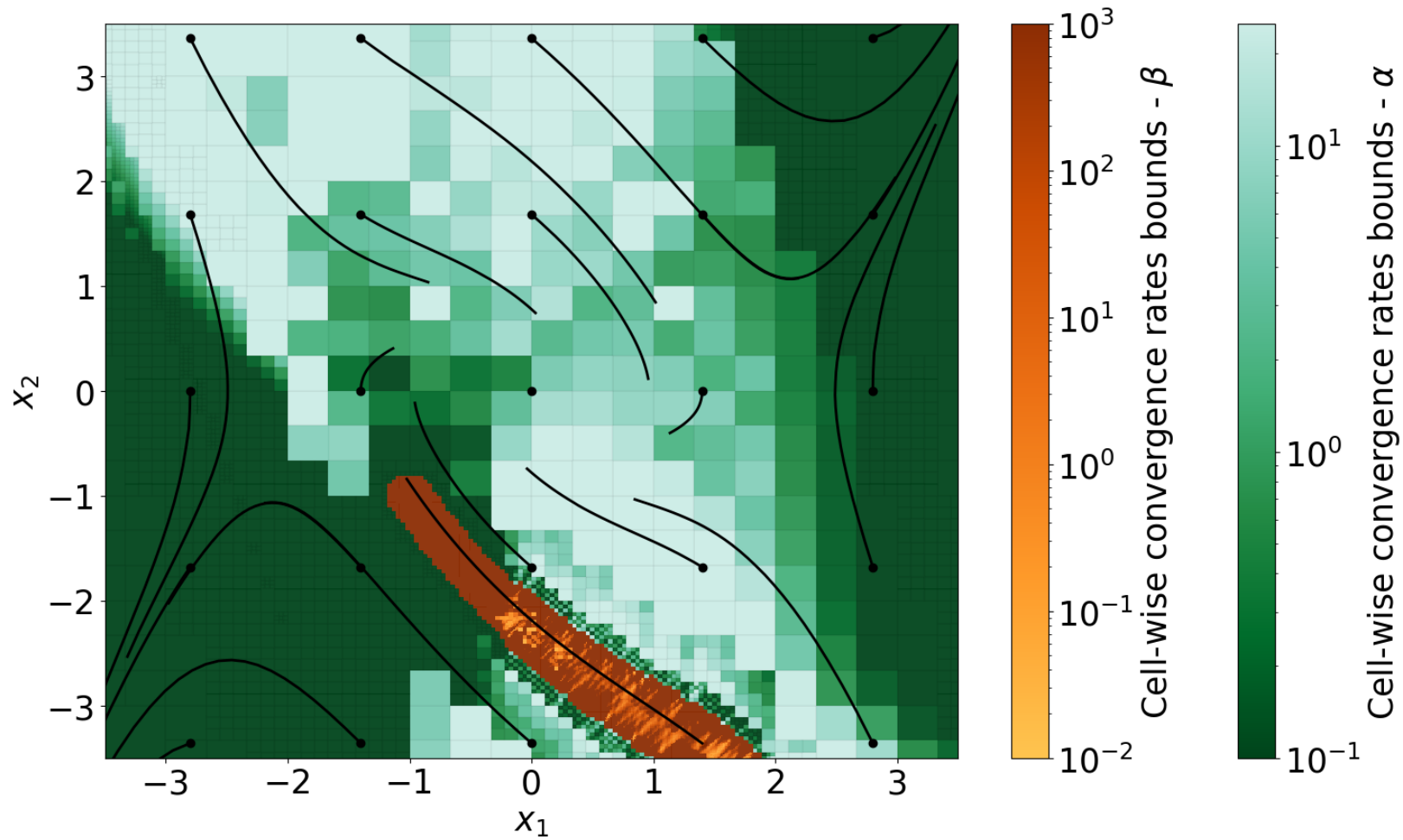
## Recurrent Condition

- $\hat{h}(x) := -sd(x, S)$  is a Recurrent exponential barrier function





# A GPU based algorithm



# Conclusions and Future work

- **Takeaways**

- Proposed a **relaxed notion of invariance** known as **recurrence**
- Introduced **Recurrent Barrier Functions** using recurrence ideas
- **Signed norms** on many sets are RBFs!
- Develop **parallelizable algorithms using GPUs**

- **Ongoing work**

- **Recurrent Sets:** Smart choice of multi-points, control recurrent sets, GPU implementation
- **Function Certificates:** Generalize other Lyapunov notions, Control Lyapunov Functions, Control Barrier Functions, Contraction, etc.
- **Recurrence Entropy:** Understanding the complexity of making a set recurrent when compared with invariance

# Thanks!

## Related Publications:

[arXiv 22] Shen, Bichuch, M, *Model-free Learning of Regions of Attraction via Recurrent Sets*, **CDC 2022**, journal preprint **arXiv:2204.10372**.

[CDC 23] Siegelmann, Shen, Paganini, M, *A recurrence-based direct method for stability analysis and GPU-based verification of non-monotonic Lyapunov functions*, **CDC 2023**

[HSCC 24] Sibai, M, *Recurrence of nonlinear control systems: Entropy and bit rates*, **HSCC, 2024**

[Allerton 24] Shen, Sibai, M, *Generalized Barrier Functions: Integral conditions and recurrent relaxations*, **Allerton 2024**

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