

# Guaranteed viability kernel enclosure

## SWIM 2015

**Dominique Monnet, Luc Jaulin, Jordan Ninin**  
*LAB-STICC / ENSTA-Bretagne*

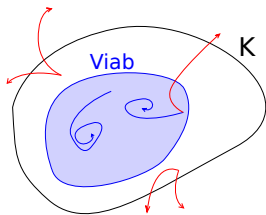
## What is viability?

System  $\mathcal{S}$  defined by:

$$\begin{aligned}\dot{\mathbf{x}} &= f(\mathbf{x}, \mathbf{u}), \\ f &: \mathbb{R}^n \times \mathbb{U} \rightarrow \mathbb{R}^n\end{aligned}$$

A state  $\mathbf{x}$  is viable if at least one evolution of  $\mathcal{S}$  from  $\mathbf{x}$  can stay indefinitely in a set of constraint  $\mathbb{K}$ .

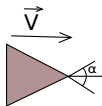
The viability kernel of  $\mathbb{K}$  under  $\mathcal{S}$  noted  $Viab_{\mathcal{S}}(\mathbb{K})$  is the set that contains every viable state.



## Why viability?

Example: management of renewable resources, economics, robotics,...

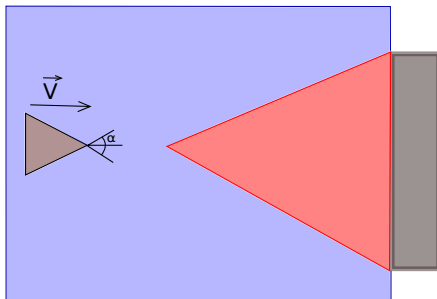
Is it possible to avoid the wall?



## Why viability?

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Is it possible to avoid the wall?



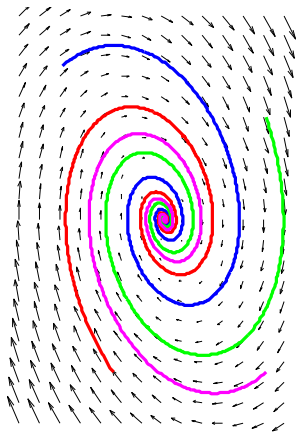
- 1 Attraction domains
- 2 Capture basin
- 3 Example
- 4 Conclusion

# Plan

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## Attraction domain of a system

Attraction domains of  $S$  are interesting for viability, if they are located in  $\mathbb{K}$ .



## Theorem on viability

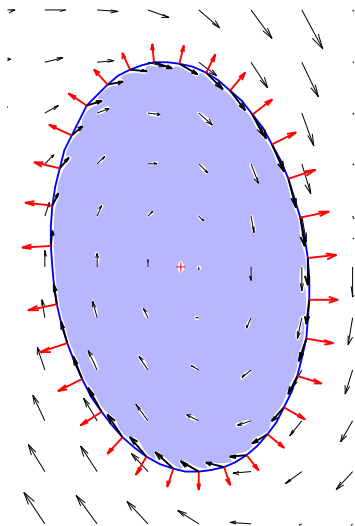
## Theorem

Let a dynamical system  $\dot{\mathbf{x}} = f(\mathbf{x}, \mathbf{u})$ ,  $\mathbb{U}$  the set of possible control and  $\mathbb{K}$  a closed subset of  $\mathbb{R}^n$ .

Let  $L \in \mathcal{C}^1(\mathbb{K}, \mathbb{R})$ , and  $\mathbb{B}_L(r) = \{\mathbf{x} \in \mathbb{R}^n \mid L(\mathbf{x}) \leq r\}$ , with  $r \in \mathbb{R}^+$ .  
If  $\mathbb{B}_L(r) \subseteq \mathbb{K}$  and  $\forall \mathbf{x} \in \mathbb{B}_L(r), \exists \mathbf{u} \in \mathbb{U}$  such as  $\langle f(\mathbf{x}, \mathbf{u}), \nabla L(\mathbf{x}) \rangle \leq 0$ ,  
then  $\mathbb{B}_L(r)$  is viable in  $\mathbb{K}$ .



## Illustration of the theorem



## Lyapunov function

## Definition

A function  $L : \mathbb{R}^n \rightarrow \mathbb{R}$  is said to be of Lyapunov for the dynamical system  $\dot{\mathbf{x}} = f(\mathbf{x})$  if:

- 1  $L(\mathbf{0}) = 0$ .
- 2  $\forall \mathbf{x} \in \mathbb{N}, L(\mathbf{x}) \geq 0$ .
- 3  $\forall \mathbf{x} \in \mathbb{N}, \langle f(\mathbf{x}), \nabla L(\mathbf{x}) \rangle \leq 0$ .

Where  $\mathbb{N}$  is a subset of  $\mathbb{R}^n$  and  $\mathbf{0} \in \mathbb{N}$ .

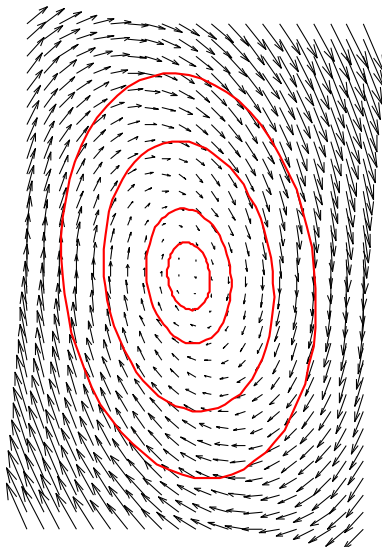
## How to find a Lyapunov function

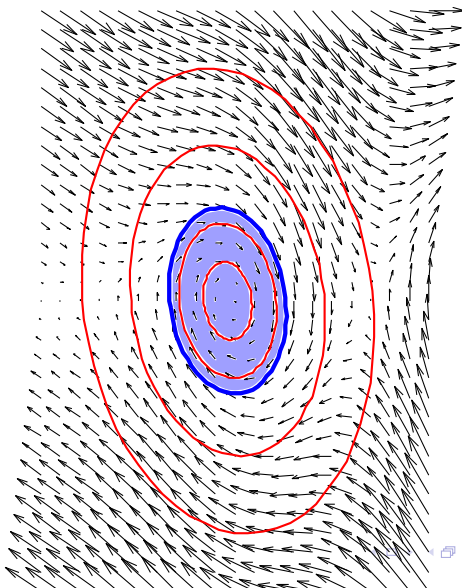
We choose a particular control  $\mathbf{u} \in \mathbb{U}$ .  $\mathcal{S}_{\mathbf{u}}$ :  $\dot{\mathbf{x}} = f(\mathbf{x}, \mathbf{u})$  is an autonomous system.

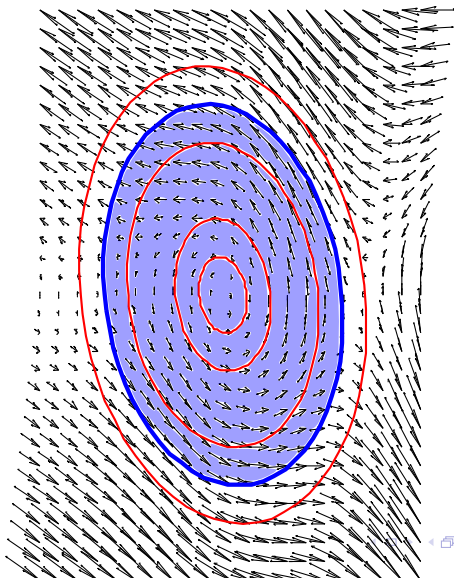
$\mathbf{x}^*$  an equilibrium point of  $\mathcal{S}_{\mathbf{u}} \iff f(\mathbf{x}^*, \mathbf{u}) = \mathbf{0}$ .

- Linearize  $\mathcal{S}_{\mathbf{u}}$  around  $\mathbf{x}^*$ , we get  $\mathcal{S}_{\mathbf{u}}^{\mathbf{x}^*}$  defined by
$$\dot{\tilde{\mathbf{x}}} = A\tilde{\mathbf{x}}, \tilde{\mathbf{x}} = \mathbf{x} - \mathbf{x}^*$$
- solve  $A^T P + PA = -I$ , where  $P$  is the unknown amount
- check whether  $P$  is positive definite
- If  $P$  is positive definite, then  $\frac{1}{2}\tilde{\mathbf{x}}^T P \tilde{\mathbf{x}}$  is a Lyapunov function for the linear system, and  $\mathbf{x}^*$  is stable.

If we do not find a Lyapunov function for  $\mathcal{S}_{\mathbf{u}}^{\mathbf{x}^*}$ , we compute the linear system  $\mathcal{S}_{ctrl}$  for which  $\mathbf{x}^*$  is a stable equilibrium point.

Lyapunov function and linearized system  $\mathcal{S}_u^x$ 

Lyapunov function and autonomous system  $\mathcal{S}_u$ 

Lyapunov function and system  $\mathcal{S}$ 

## Viable set characterization algorithm

- Choose a control  $\mathbf{u} \in \mathbb{U}$
- Find  $\mathbf{x}^* \in \mathbb{K}$
- Linearize  $\mathcal{S}_{\mathbf{u}}$  around  $\mathbf{x}^*$ .
- Try to find a Lyapunov function of  $\mathcal{S}_{\mathbf{u}}^{\mathbf{x}^*}$
- If no function found, compute  $\mathcal{S}_{ctrl}$
- Try to find a Lyapunov function for  $\mathcal{S}_{ctrl}$
- Find  $r \in \mathbb{R}^+$  such as conditions of the theorem are met

$\mathbb{E}$  is the union of viable sets found around equilibrium points of  $\mathcal{S}$  in  $\mathbb{K}$ .

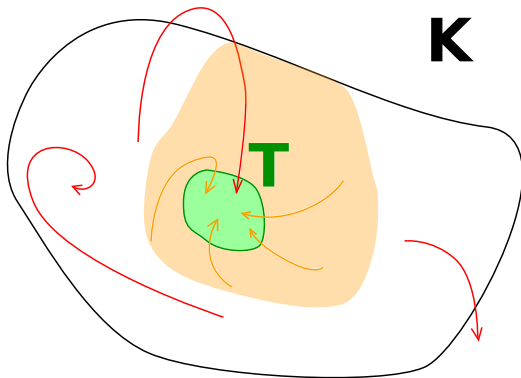
# Plan

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## Capture basin problem

The capture basin of a set  $\mathbb{T} \subset \mathbb{K}$  viable in  $\mathbb{K}$  noted  $Capt_{\mathcal{S}}(\mathbb{K}, \mathbb{T})$  is composed of every states  $\mathbf{x}$  such as  $\mathcal{S}$  can reach  $\mathbb{T}$  from  $\mathbf{x}$  in a finite time without leaving  $\mathbb{K}$ .



## Theorem on the viability of a capture basin

## Theorem

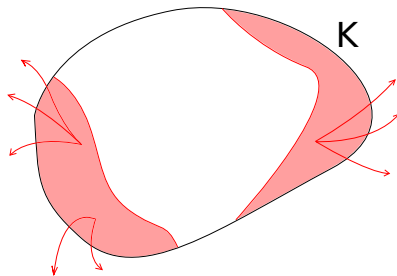
*Let  $S$  a dynamical system,  $\mathbb{K}$  a closed subset of the state space of  $S$  and  $\mathbb{T} \subset \mathbb{K}$ .*

*If  $\mathbb{T}$  is viable in  $\mathbb{K}$ ,  
then  $\text{Capt}_S(\mathbb{K}, \mathbb{T})$  is viable in  $\mathbb{K}$ .*

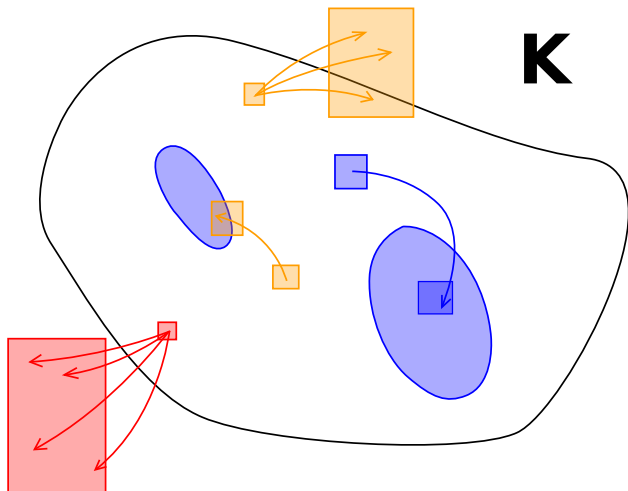
The set  $\mathbb{V}_{in} = \mathbb{T} \cup \text{Capt}_S(\mathbb{K}, \mathbb{T})$  is an under approximation of  $\text{Viab}_S(\mathbb{K})$ .

## Over approximation of the viability kernel

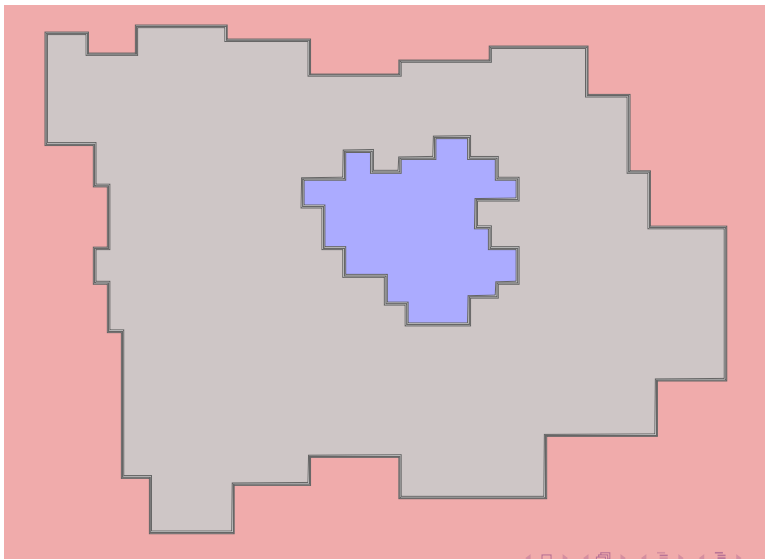
- We try to find an over approximation of  $Viab_S(\mathbb{K})$  to get an enclosure of  $Viab_S(\mathbb{K})$ .
- If  $\forall \mathbf{u} \in \mathbb{U}$ ,  $\mathcal{S}$  cannot stay in  $\mathbb{K}$  from a state  $\mathbf{x} \in \mathbb{K}$ , then  $\mathbf{x} \notin Viab_S(\mathbb{K})$ .



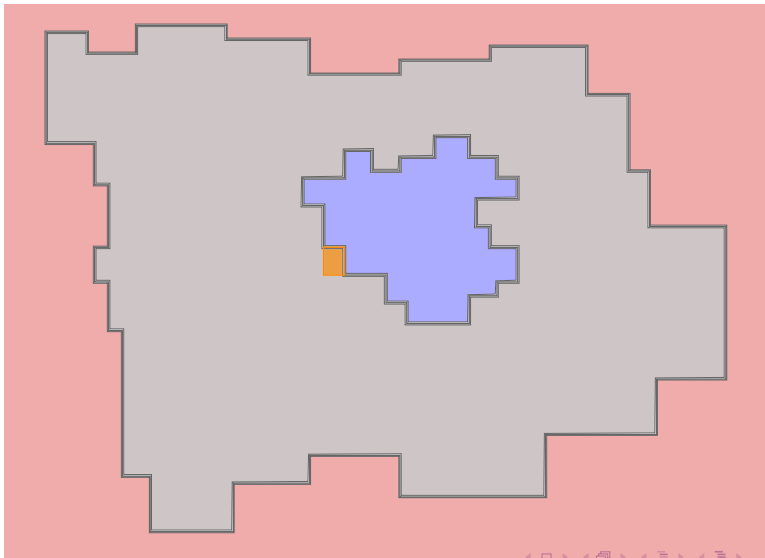
## Guaranteed integration of a box [Chaputot, 2015]



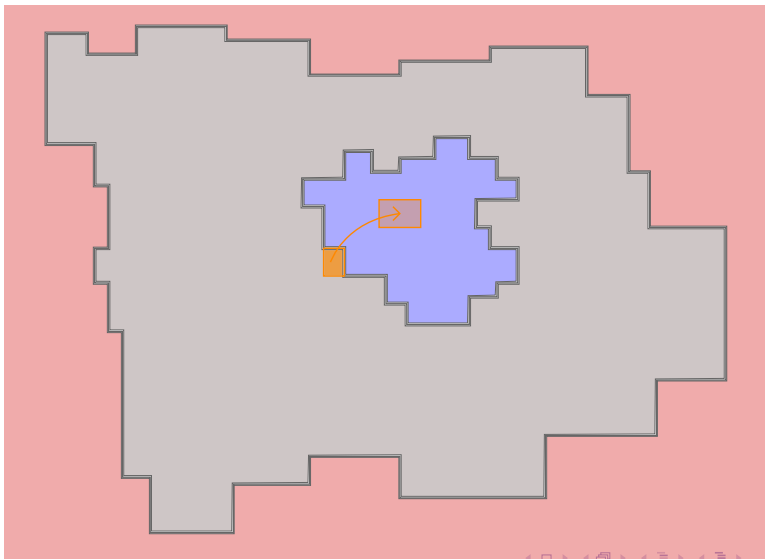
## Under approximation algorithm



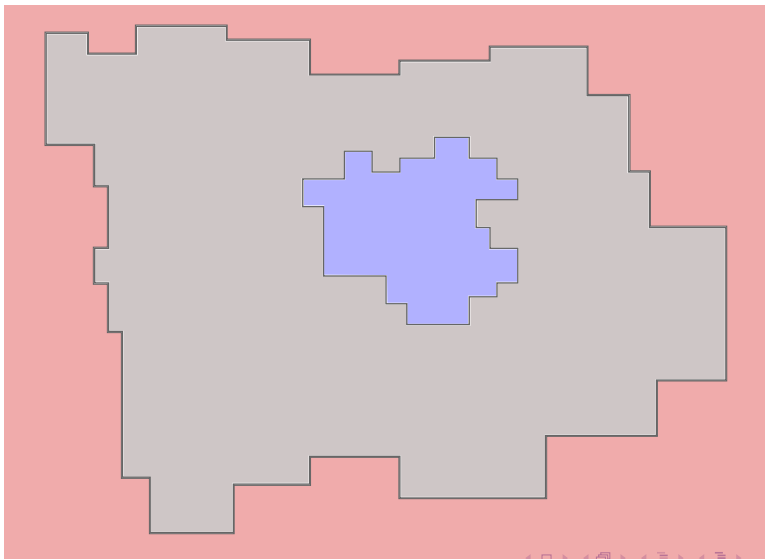
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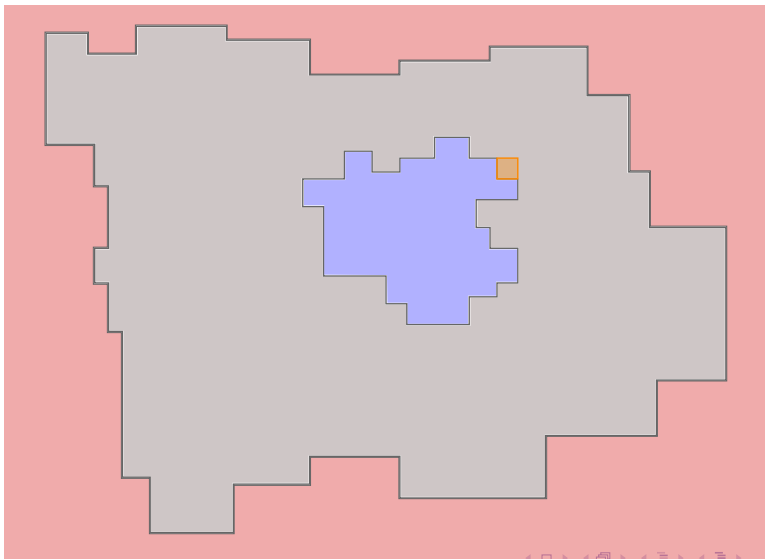


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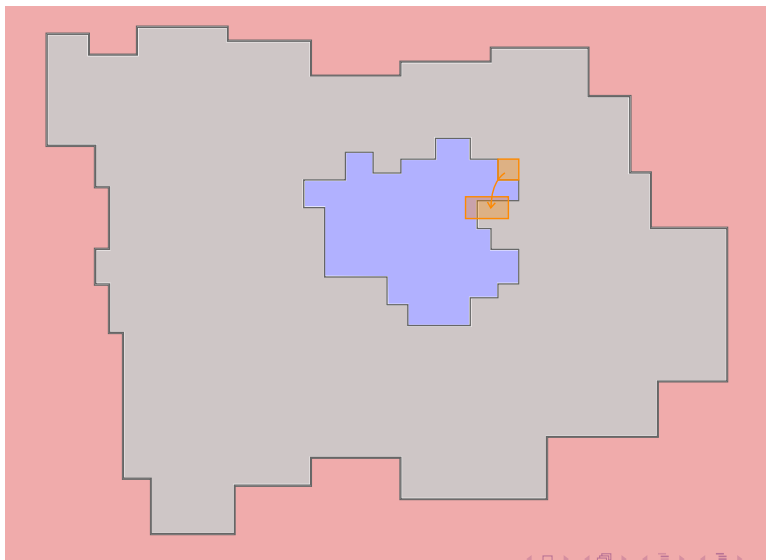




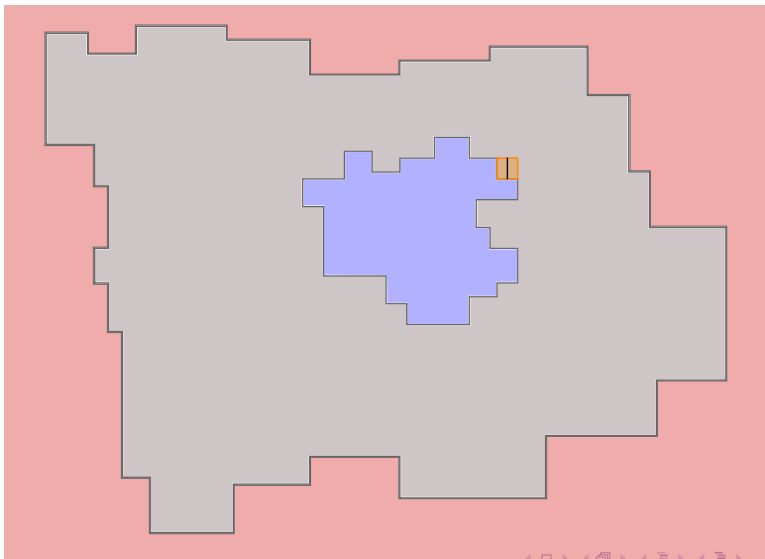
## Under approximation algorithm



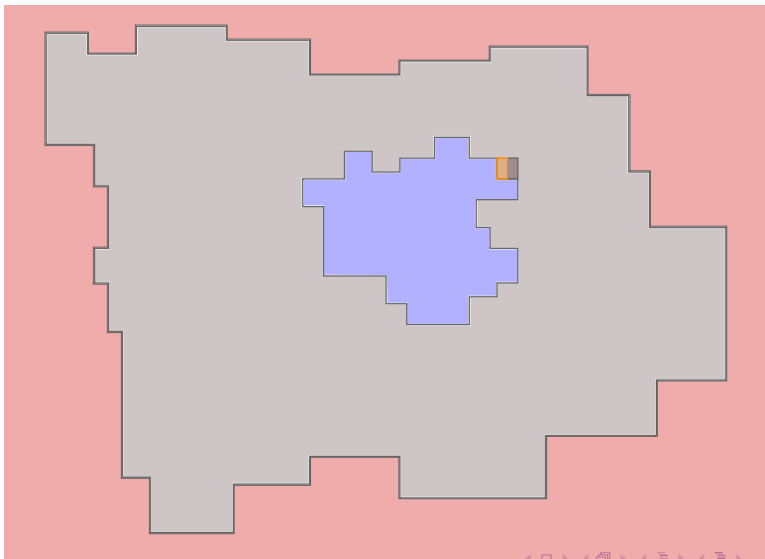
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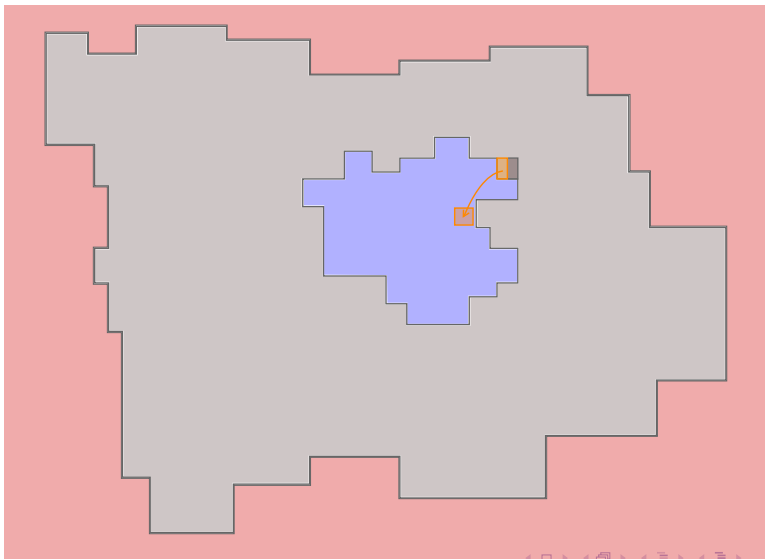
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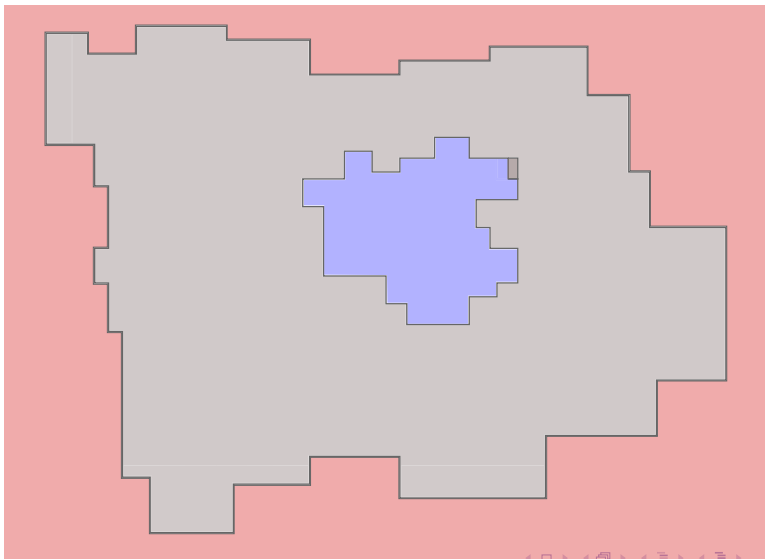
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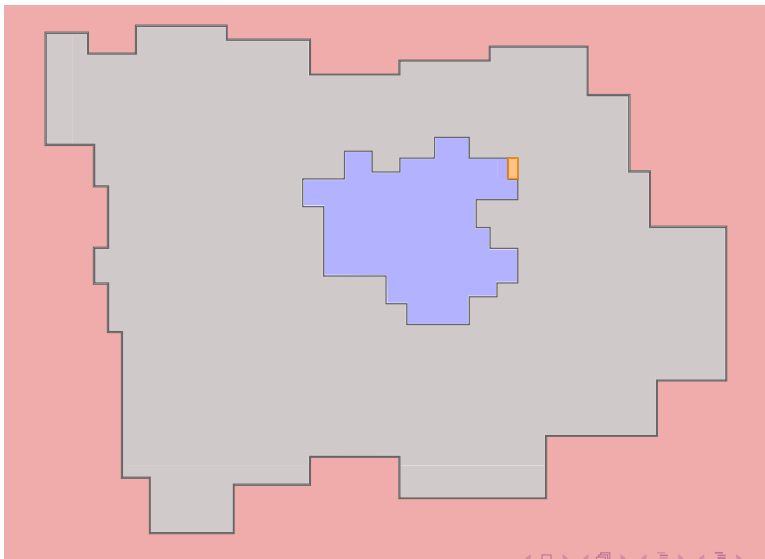
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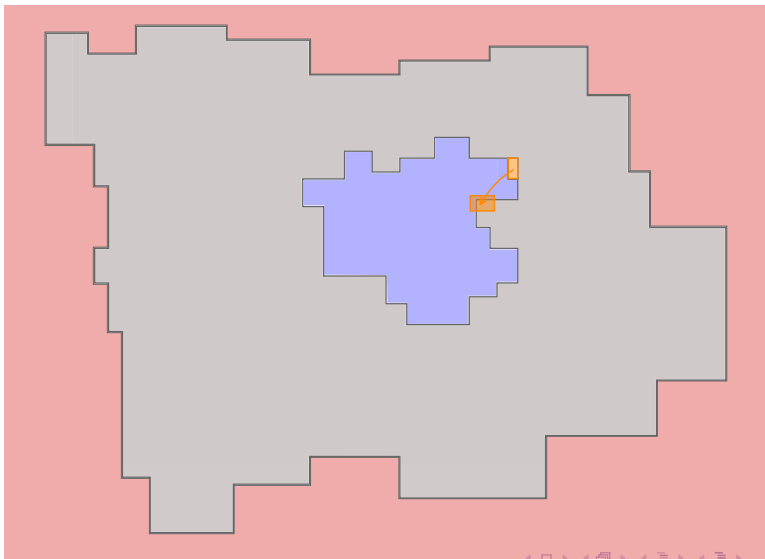
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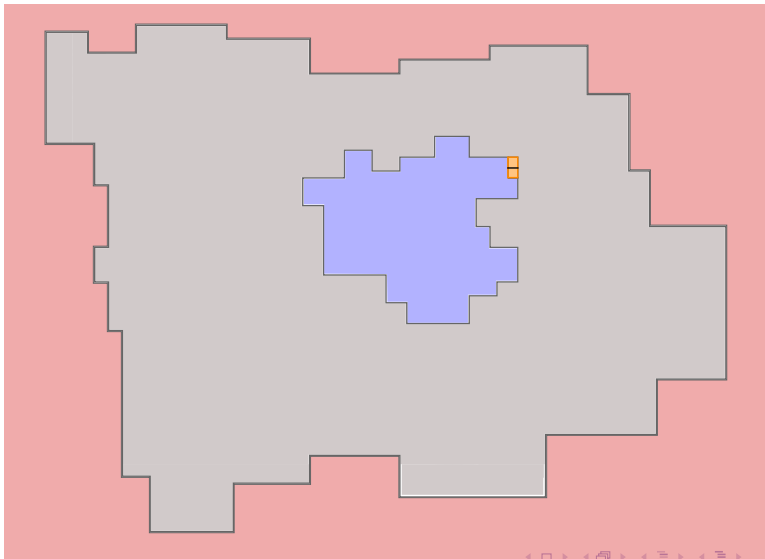


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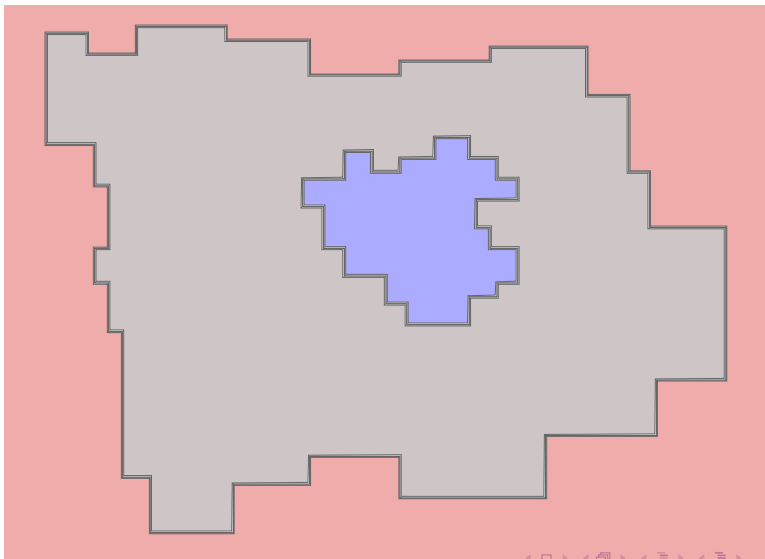




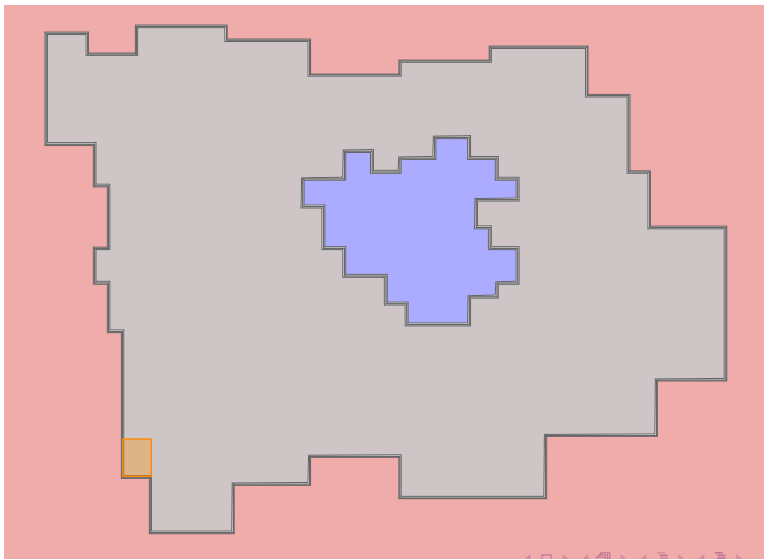
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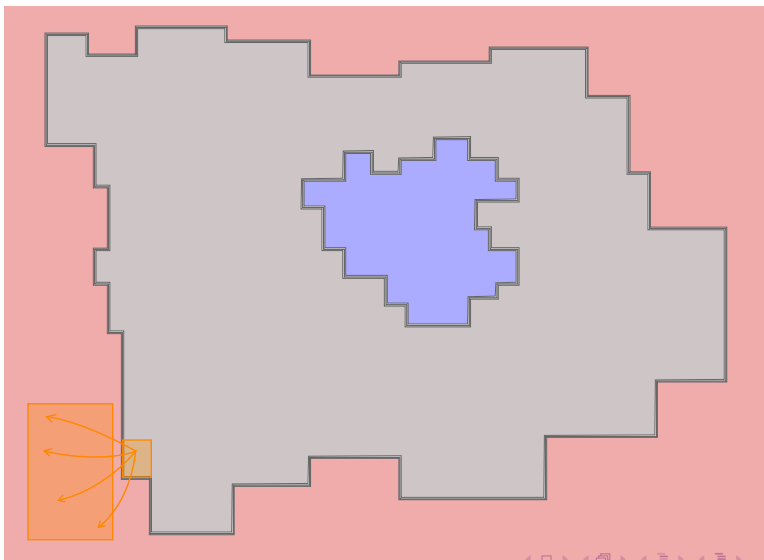
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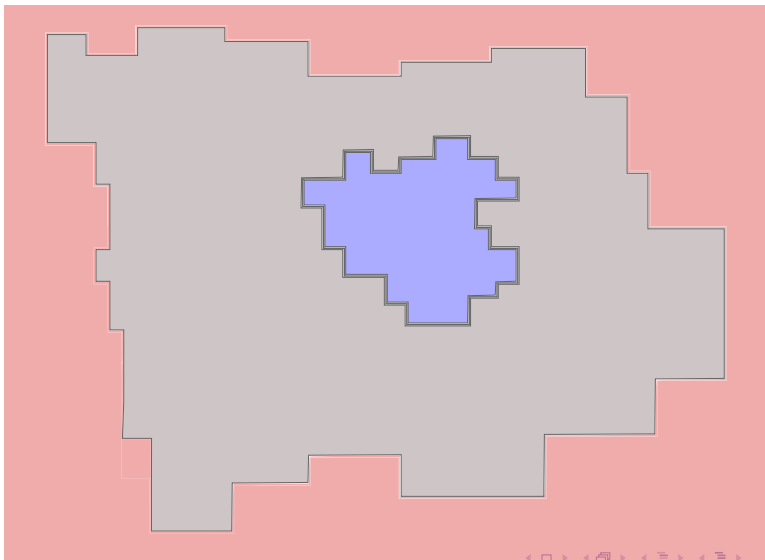
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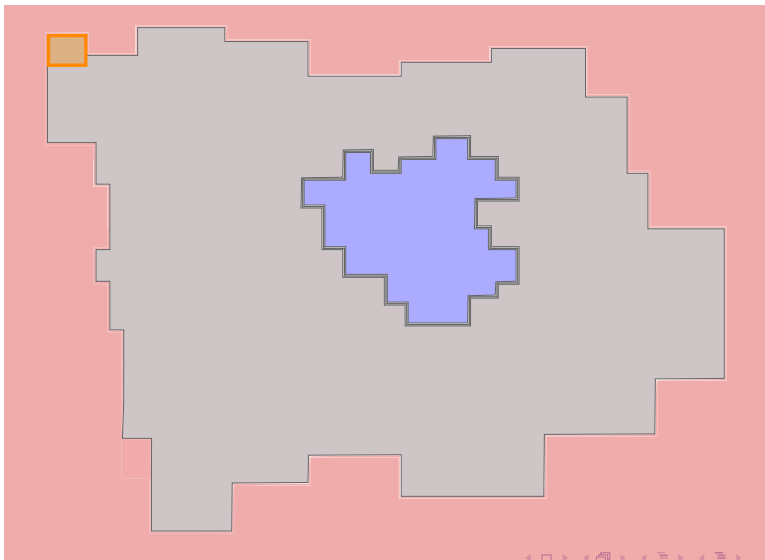
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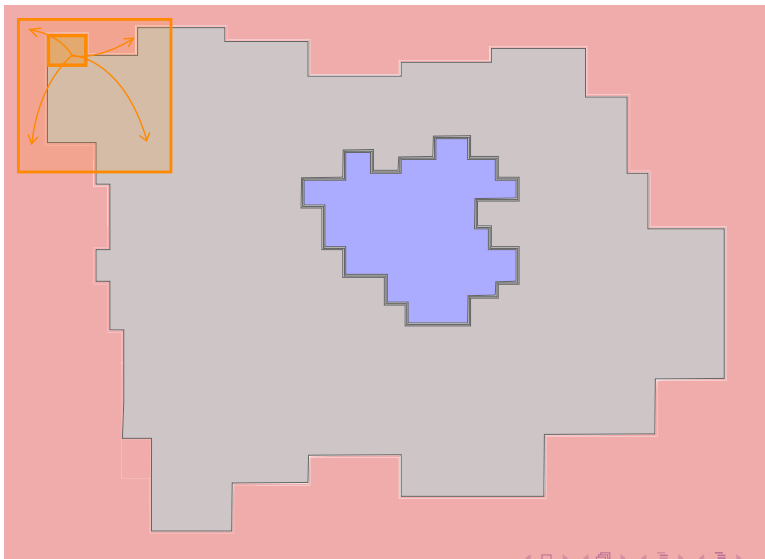
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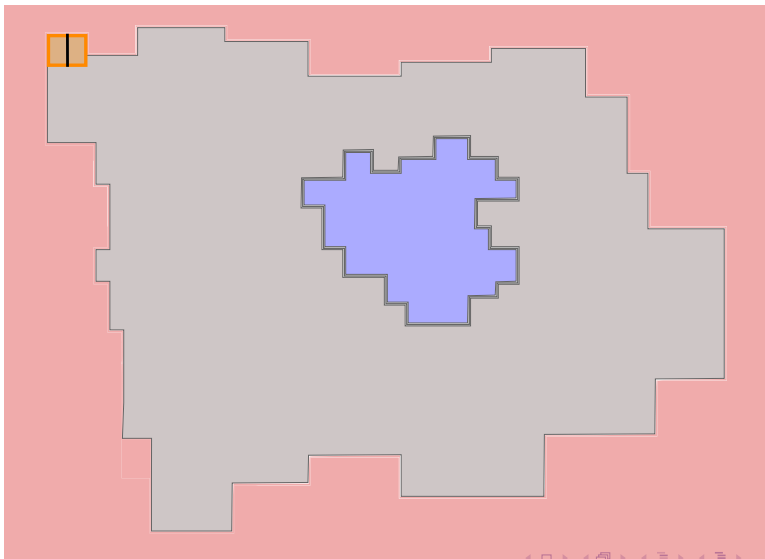
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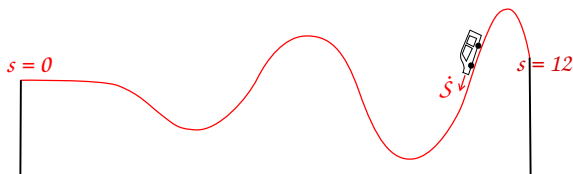
## Car on the hill problem

- The landscape is represented by a parametric function  $g(s)$
- State vector:  $\mathbf{x} = \begin{pmatrix} s \\ \dot{s} \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$
- Evolution function:

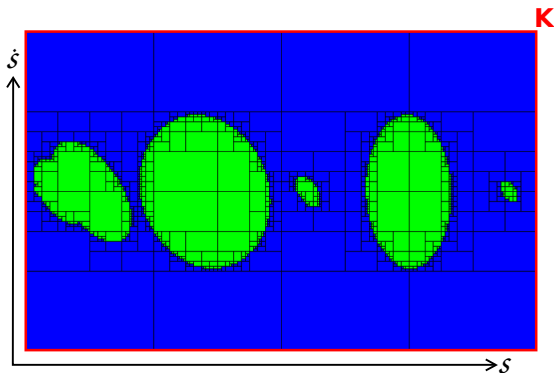
$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -9.81g'(x_1) - \alpha x_2 + u \end{cases}$$

$$u \in [-2, 2]$$

- The car must stay on the landscape, i.e.  $s \in [0, 12]$

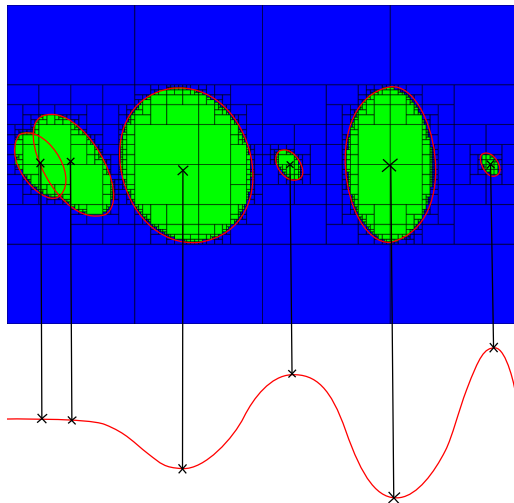


## Results of viable set characterization algorithm

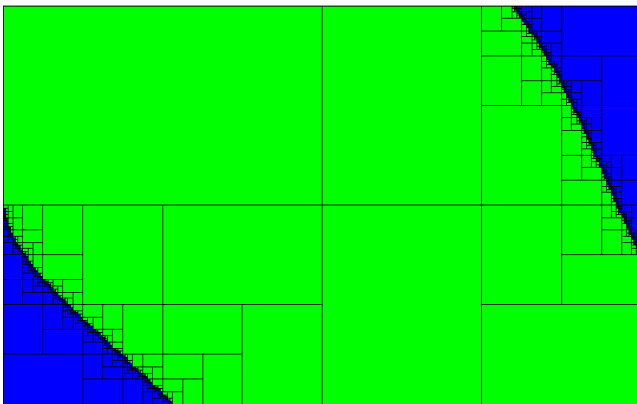


Computation time  $\approx$  1 minute.

## Results explained

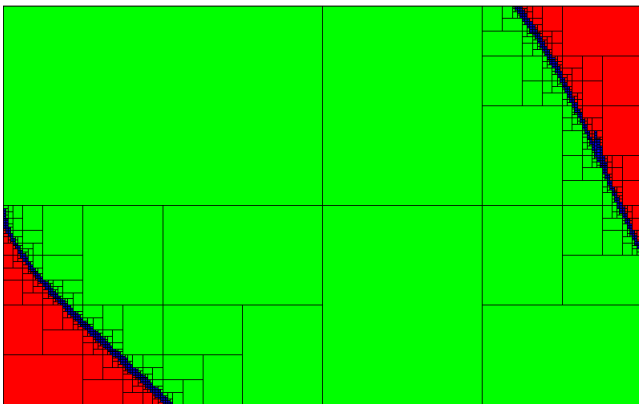


## Results of under approximation algorithm



Computation time  $\approx$  20 minutes

## Results of over approximation algorithm



Computation time  $\approx$  10 minutes

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## Conclusion

- We are able to deal with many viability problems in a guaranteed way.
- The system must have at least one equilibrium point
- We can deal with 2D problems, but under and over approximation algorithms are not efficient for higher dimensional problems