

# Lie Groups Applied to mobile robotics

Application to Reachability analysis

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16 May 2022



# Outline

- 1 Context
- 2 Guaranteed integration method using Lie symmetries
  - Why
  - Main Idea
  - Basic example
  - Example of the tank-like robot
- 3 Application to reachability
- 4 Conclusion

## Section 1

# Context

# Context of this research

- ▶ development of offshore wind farms

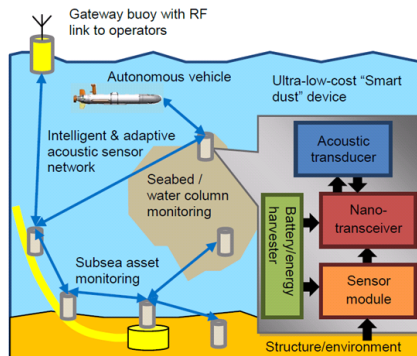


Figure: AUV used as an autonomous data mule

Source: USMART project

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- ▶ development of offshore wind farms
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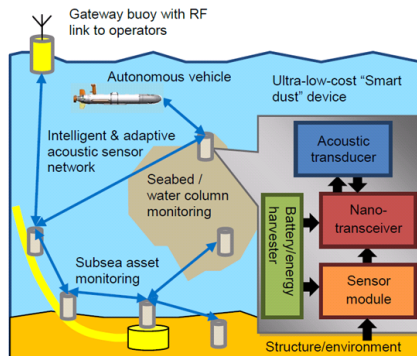


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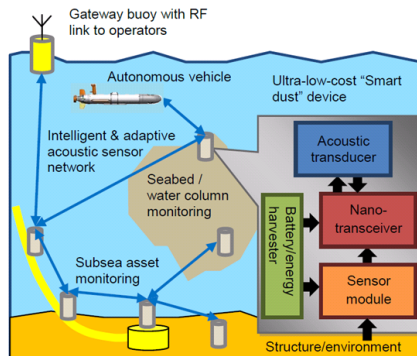


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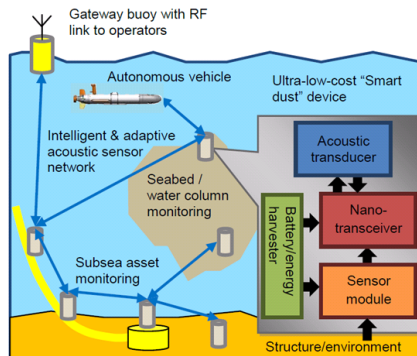


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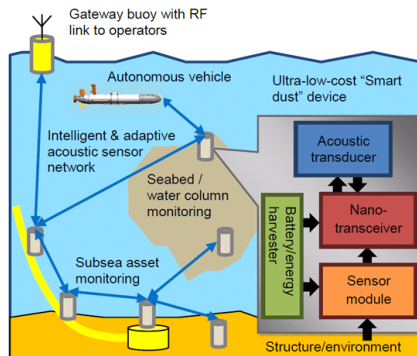


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Development of swarms of AUVs



Need for cheaper units with no expensive sensors



Improve algorithms of state estimation

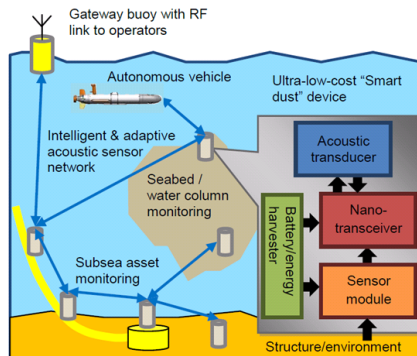


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## Section 2

# Guaranteed integration method using Lie symmetries

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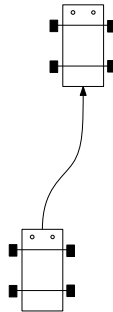
# Why a new guaranteed integration method ?

- ▶ Need for guarantee as we are working with complex systems
- ▶ Conventional tools can be quite slow when performing numerous integrations
- ▶ Conventional tools cannot deal with large initial conditions

# Principle

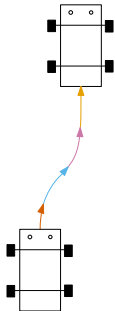


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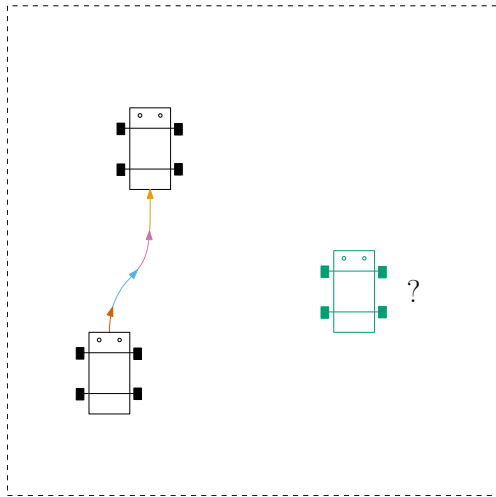
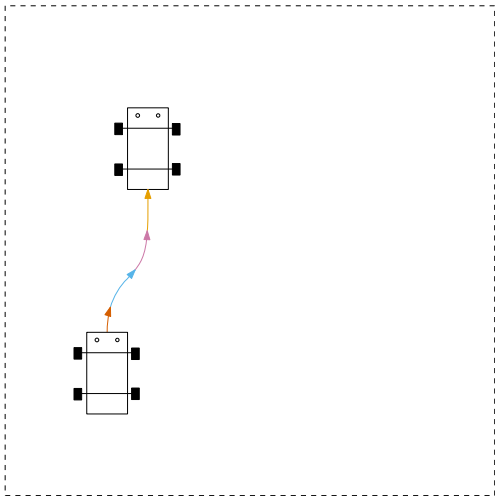




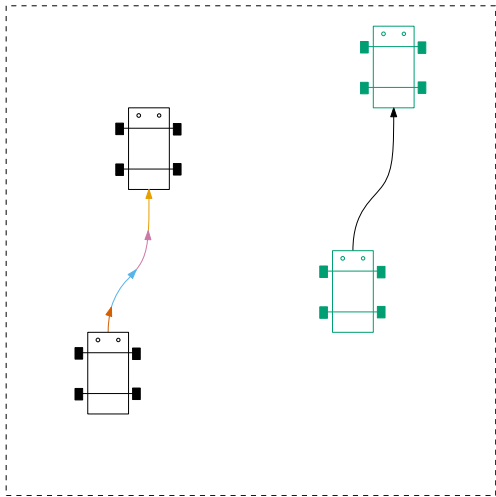
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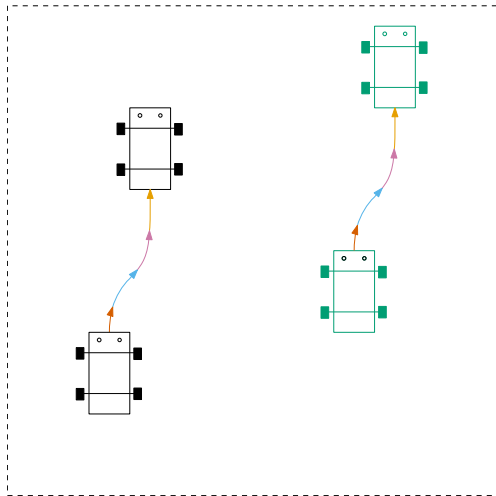
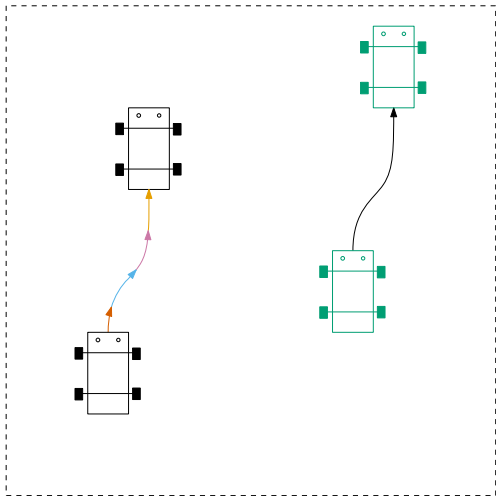
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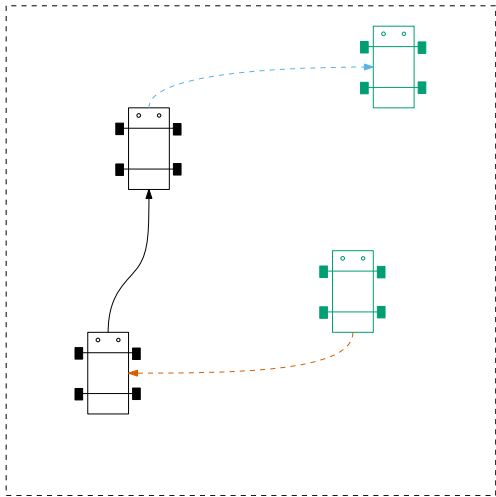
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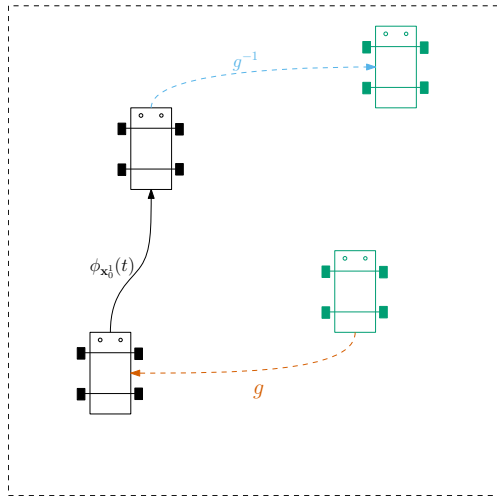
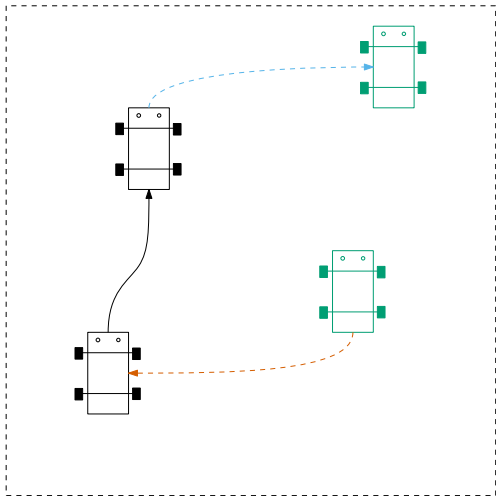
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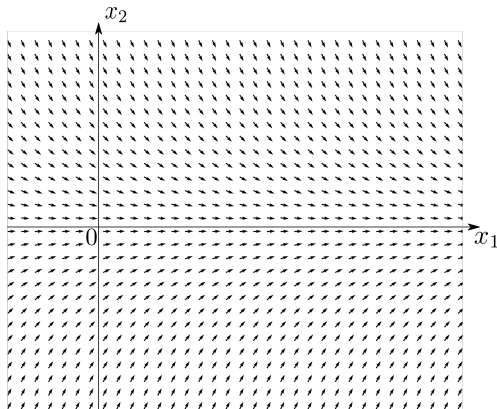


## Hints from the vector field

Let us consider the system defined by:

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}) = \begin{pmatrix} 1 \\ -x_2 \end{pmatrix}$$

From the associated vector field we can deduce a translation symmetry along  $Ox_1$  and mirror symmetry over  $Ox_1$  that does not affect the vector field. These symmetries are called **Lie** symmetries.



# Determine the flow function

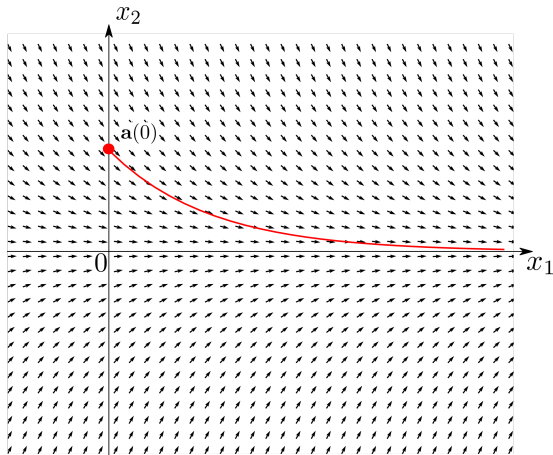
**Objective:** Determine the flow function  $\Phi_t(\mathbf{x})$   
i.e. the function that describes the evolution of  
the vector  $\mathbf{x}$  at time  $t$ .



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We already have a **reference** trajectory denoted  $\mathbf{a}(\cdot)$  (painted red) and a transformation function  $\mathbf{g}_p$  which can send any point of the state space to any other following the symmetries of the system.



# Determine the flow function

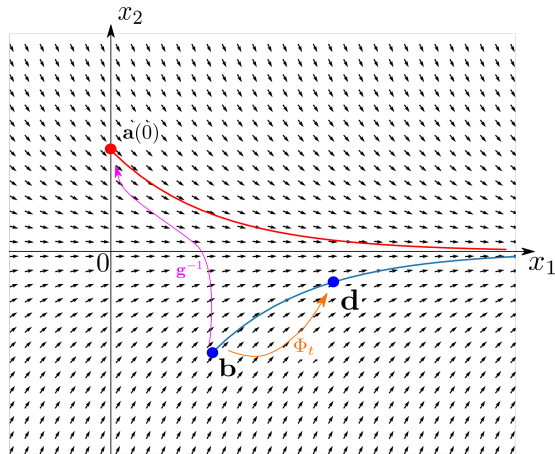
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Therefore

$$\Phi_t(\mathbf{x}) = \mathbf{g}_p \circ \mathbf{a}(t)$$

where  $\mathbf{p} = \mathbf{h}(\mathbf{x}, \mathbf{a}(0))$



$$\mathbf{d} = \Phi_t(\mathbf{b})$$

# Determine the flow function

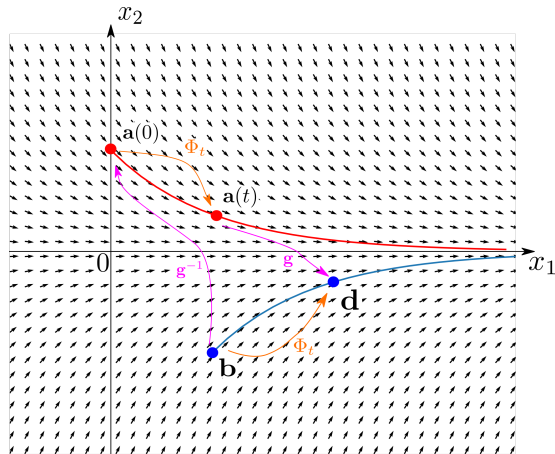
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# Guaranteed Integration, a set inversion problem

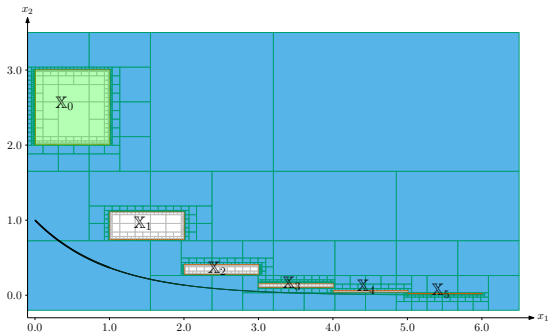
With the flow function  $\Phi_t$ , performing a guaranteed integration for an uncertain initial condition is equivalent to solving a set inversion problem.

Consider a uncertain initial box  $[\mathbf{x}_0]$  for which we want to find the image set by  $\Phi_t$ . We want to find the set  $\mathbb{X}_t$  such that

$$\mathbb{X}_t = \Phi_{-t}^{-1}([\mathbf{x}_0]).$$

# Applying the SIVIA

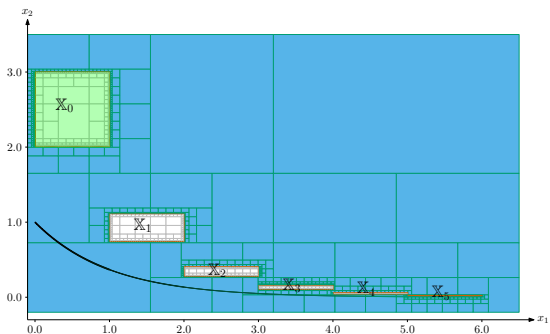
Initial condition:  $[\mathbf{x}_0] = [0, 1] \times [2, 3]$



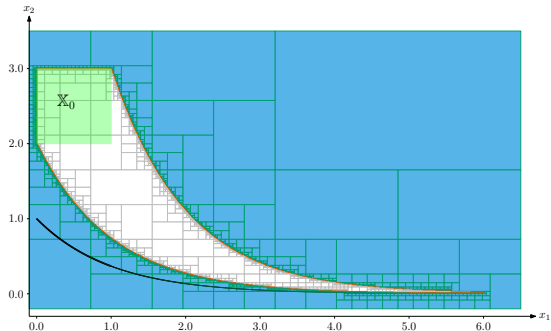
Discrete sets computation

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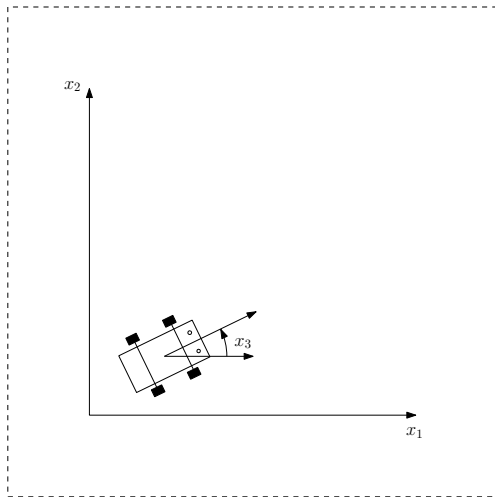


Continuous set computation

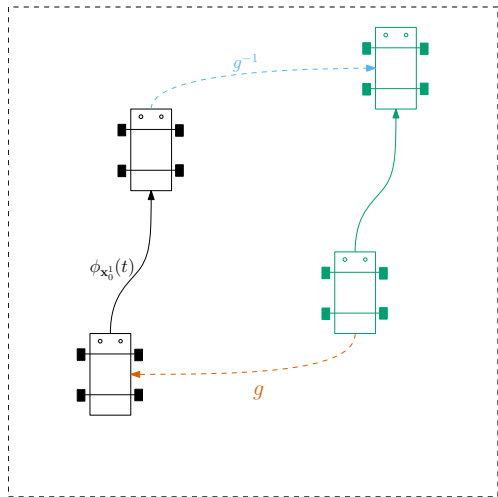
# Example of the tank-like robot: Introduction

Let us consider the system defined by:

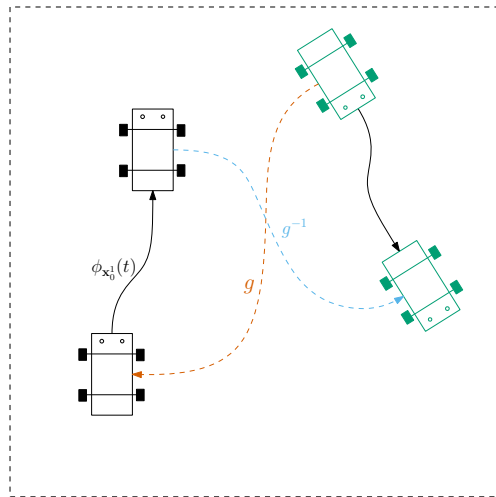
$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u}(t)) = \begin{pmatrix} u_1(t) \cos(x_3) \\ u_1(t) \sin(x_3) \\ u_2(t) \end{pmatrix}$$



# Example of the tank-like robot: Symmetries



Translation symmetries

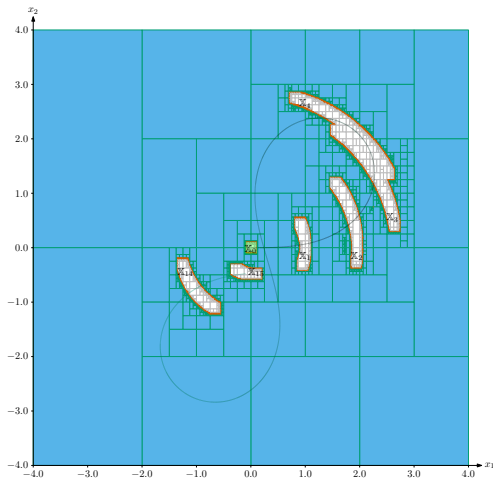


Rotation symmetry



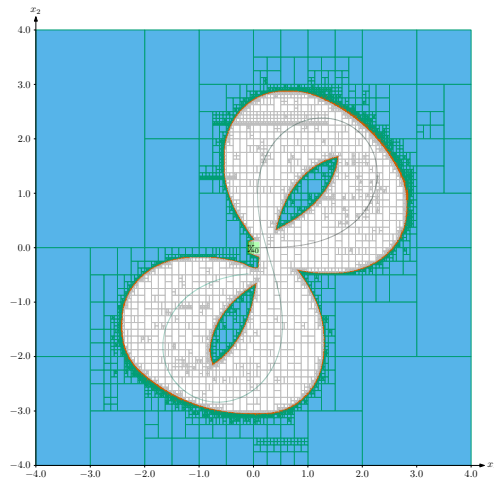
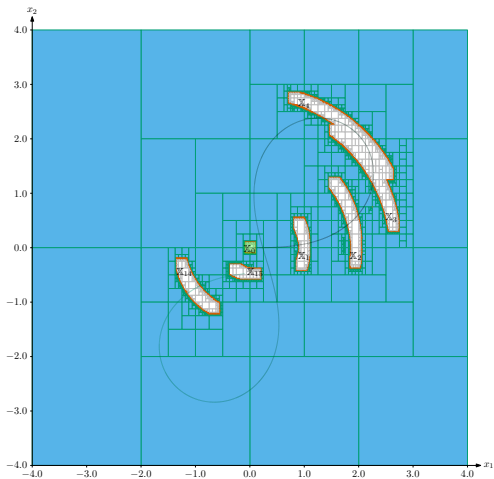
# Applying the SIVIA

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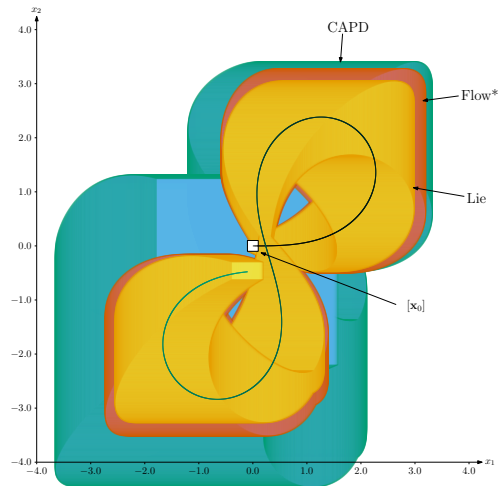
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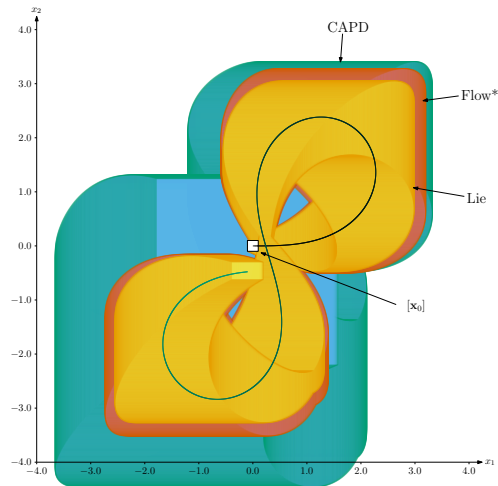
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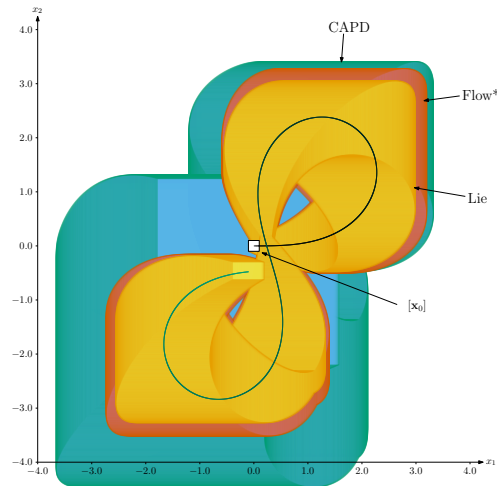
- ▶ able to deal with large initial condition (no bloating effect)
- ▶ less computation time (less steps of operations)



# Pros and limits of the method

## Pros:

- ▶ able to deal with large initial condition (no bloating effect)
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- ▶ Easily get an outer **and** inner approximation

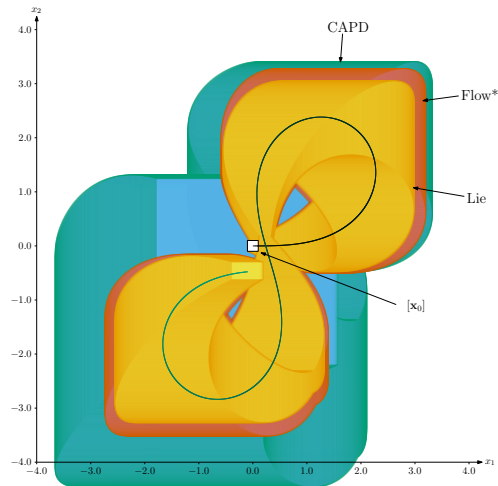


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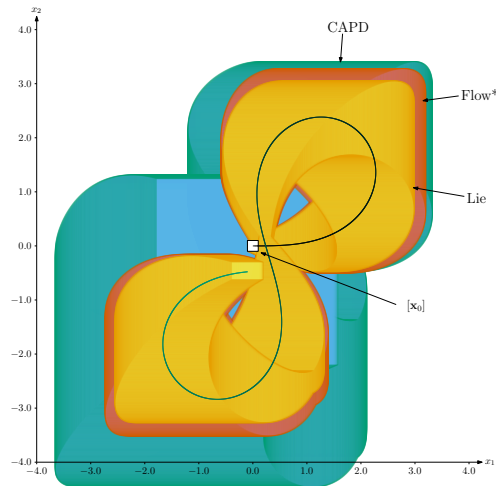
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- ▶ Need for enough symmetries, method will not work for every systems





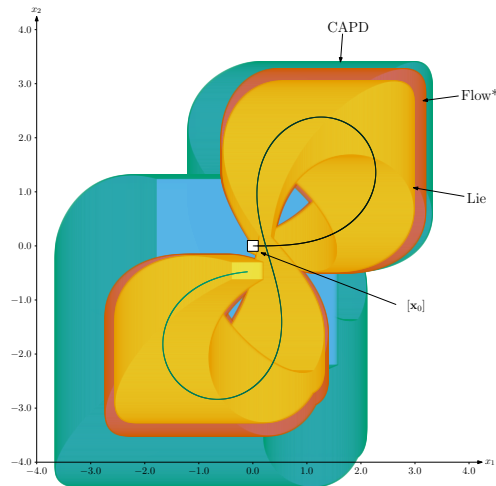
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- ▶ Need for a reference



## Section 3

## Application to reachability

# Reachability analysis

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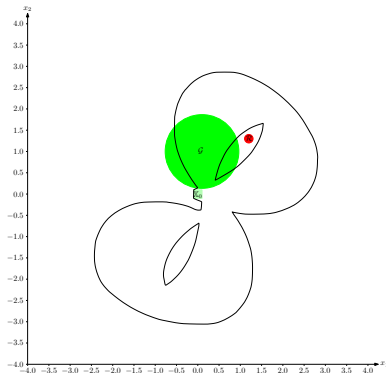
- ▶ Performance assessment
- ▶ Scheduling
- ▶ Controller design
- ▶ Deadlock

The following slides will focus on performance assessment in a mobile robotic context



# Reachable area problem

Common robotic problem: Ensure that a vehicle will either certainly reach a mandatory area or avoid (a dangerous) one.



# Reachable area problem

Checking that we reach a mandatory area:

$$\forall t \in [t], \begin{cases} 1. \Phi_{-t}(\mathbf{x}(t)) \in [\mathbf{x}_0] & \text{(evolution)} \\ 2. \mathbf{x}(t) \notin \mathcal{G} & \text{(area to reach)} \end{cases}$$

These two constraints cannot be both satisfied  $\Rightarrow$  The system will go through the mandatory area

# Reachable area problem

Checking that we avoid a forbidden area:

$$\begin{cases} 1. \Phi_{-t}(\mathbf{x}(t)) \in [\mathbf{x}_0] & \text{(evolution function)} \\ 2. \exists t \in [t], \mathbf{x}(t) \in \mathcal{R} & \text{(area to reach)} \end{cases}$$

These two constraints cannot be both satisfied  $\Rightarrow$  The system will avoid the forbidden area

# Reachable area problem

## Section 4

## Conclusion

# Conclusion

- ▶ A new guaranteed integration method based on Lie symmetries
- ▶ adapted to large initial conditions
- ▶ Not suitable for every kind of dynamical systems but suitable for robotics

Thank you for your attention

Proof  $\mathbb{X}_t = \Phi_{-t}^{-1}([\mathbf{x}_0])$  :

$$\begin{aligned}
 \mathbf{x} \in \mathbb{X}_t &\iff \exists \mathbf{x}_0 \in [\mathbf{x}_0], \mathbf{x} = \phi_t(\mathbf{x}_0) \\
 &\iff \exists \mathbf{x}_0 \in [\mathbf{x}_0], \mathbf{x}_0 = \phi_{-t}(\mathbf{x}) \\
 &\iff \phi_{-t}(\mathbf{x}) \in [\mathbf{x}_0] \\
 &\iff \mathbf{x} \in \phi_{-t}^{-1}([\mathbf{x}_0])
 \end{aligned}$$



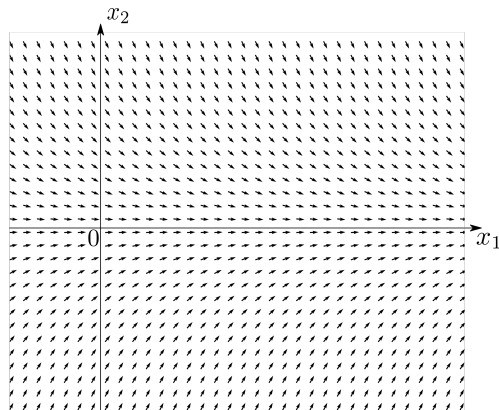
Definition (Lie symmetry) : A function  $\mathbf{g}_p$  is a symmetry of  $\mathbf{f}$  if the action  $\bullet$  of  $\mathbf{g}_p$  on  $\mathbf{f}$  leaves  $\mathbf{f}$  unchanged i.e

$$\mathbf{g}_p \bullet \mathbf{f} = \mathbf{f}.$$

It is also called a **stabiliser**.

Definition (Lie group of symmetry): Consider a state equation  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x})$  and an manifold  $\mathbb{P}$ . A Lie group  $G_p$  of symmetries is a family of diffeomorphisms  $\mathbf{g}_p \in \text{diff}(\mathbb{R}^n)$  parameterised by  $\mathbf{p} \in \mathbb{P}$  such that:

- ▶  $G_p$  is a Lie group with respect to the composition  $\circ$ ,
- ▶  $\forall \mathbf{p} \in \mathbb{P}, \mathbf{g}_p \bullet \mathbf{f} = \mathbf{f}$ .

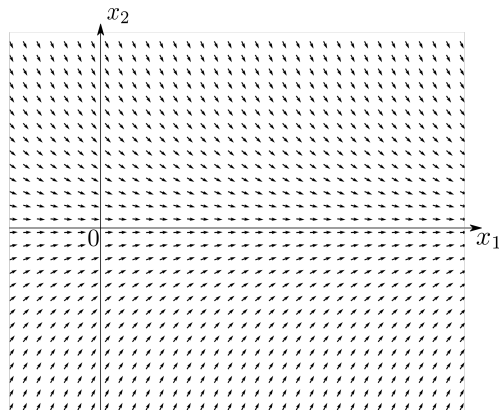


Proposition (Action of a diffeomorphism):

$$\mathbf{g} \bullet \mathbf{f} = \left( \frac{d\mathbf{g}}{d\mathbf{x}} \circ \mathbf{g}^{-1} \right) \cdot (\mathbf{f} \circ \mathbf{g}^{-1}).$$

Objective: Finding  $\mathbf{g}$  such that

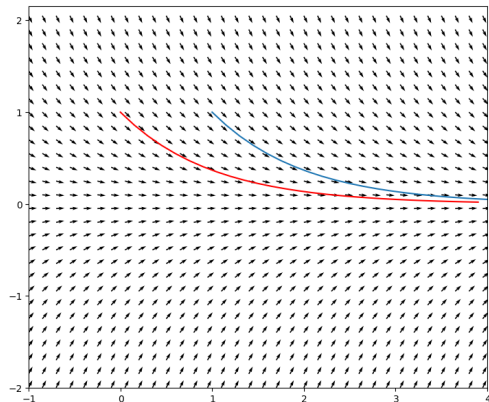
$$\mathbf{g} \bullet \mathbf{f} = \left( \frac{d\mathbf{g}}{d\mathbf{x}} \circ \mathbf{g}^{-1} \right) \cdot (\mathbf{f} \circ \mathbf{g}^{-1}) = \mathbf{f}$$



Translation symmetry:

$$\mathbf{g}_\alpha : \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \mapsto \begin{pmatrix} x_1 + \alpha \\ x_2 \end{pmatrix}$$

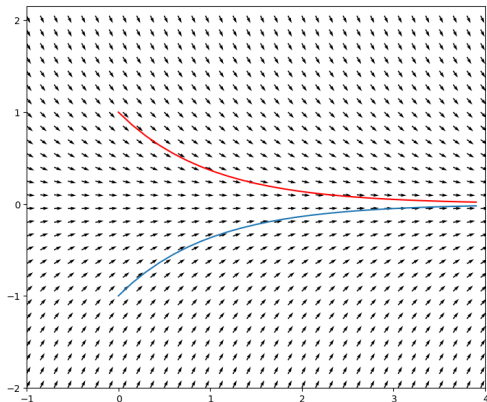
$$\begin{aligned} \mathbf{g}_\alpha \bullet \mathbf{f}(\mathbf{x}) &= \left( \frac{d\mathbf{g}_\alpha}{d\mathbf{x}} \circ \mathbf{g}_\alpha^{-1} \right) \cdot (\mathbf{f} \circ \mathbf{g}_\alpha^{-1})(\mathbf{x}) \\ &= \left( \frac{d\mathbf{g}_\alpha}{d\mathbf{x}} \cdot \mathbf{f} \right) \circ \mathbf{g}_\alpha^{-1}(\mathbf{x}) \\ &= \left( \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ -x_2 \end{pmatrix} \right) \circ \begin{pmatrix} x_1 - \alpha \\ x_2 \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ -x_2 \end{pmatrix} \\ &= \mathbf{f}(\mathbf{x}) \end{aligned}$$



Mirror symmetry:

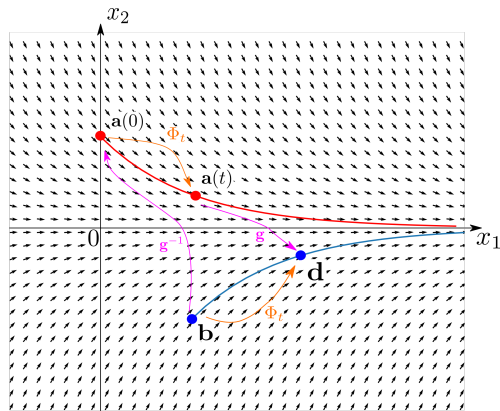
$$\mathbf{g}_\beta : \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \mapsto \begin{pmatrix} x_1 \\ \beta x_2 \end{pmatrix}$$

$$\begin{aligned} \mathbf{g}_\beta \bullet \mathbf{f}(\mathbf{x}) &= \left( \frac{d\mathbf{g}_\beta}{d\mathbf{x}} \circ \mathbf{g}_\beta^{-1} \right) \cdot (\mathbf{f} \circ \mathbf{g}_\beta^{-1})(\mathbf{x}) \\ &= \left( \frac{d\mathbf{g}_\beta}{d\mathbf{x}} \cdot \mathbf{f} \right) \circ \mathbf{g}_\beta^{-1}(\mathbf{x}) \\ &= \left( \begin{pmatrix} 1 & 0 \\ 0 & \beta \end{pmatrix} \cdot \begin{pmatrix} 1 \\ -x_2 \end{pmatrix} \right) \circ \begin{pmatrix} x_1 \\ \frac{x_2}{\beta} \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ -x_2 \end{pmatrix} \\ &= \mathbf{f}(\mathbf{x}) \end{aligned}$$



Complete symmetry:

$$\mathbf{g}_p : \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \mapsto \begin{pmatrix} x_1 + p_1 \\ p_2 x_2 \end{pmatrix}$$



To find the transport function, we must solve

$$\mathbf{g}_p(\mathbf{a}) = \mathbf{x},$$

in order to express  $\mathbf{p}$  using only  $\mathbf{a}$  and  $\mathbf{x}$ .

Using the previous example :

$$\begin{aligned} \mathbf{g}_p(\mathbf{a}) = \mathbf{x} &\iff \begin{pmatrix} a_1 + p_1 \\ p_2 a_2 \end{pmatrix} = \mathbf{x} \\ &\iff \mathbf{p} = \begin{pmatrix} x_1 - a_1 \\ \frac{x_2}{a_2} \end{pmatrix} \end{aligned}$$

Therefore,

$$\mathbf{h}(\mathbf{x}, \mathbf{a}) = \begin{pmatrix} x_1 - a_1 \\ \frac{x_2}{a_2} \end{pmatrix}.$$

Flow function of the first example:

$$\mathbf{a}(t) = \begin{pmatrix} t \\ e^{-t} \end{pmatrix} \text{ and } \mathbf{a}(0) = (0, 1)$$

$$\begin{aligned} \Phi_t(\mathbf{x}) &= \mathbf{g}_{\mathbf{h}(\mathbf{x}, \mathbf{a}_0)} \circ \mathbf{a}(t) \\ &= \mathbf{g}_{x_1, x_2} \circ \begin{pmatrix} t \\ e^{-t} \end{pmatrix} \\ &= \begin{pmatrix} t + x_1 \\ x_2 \cdot e^{-t} \end{pmatrix} \end{aligned}$$

Flow function for the tank-like robot:

$\mathbf{a}(0) = (0, 0, 0)$  and  $\mathbf{a}(t)$  obtained with CAPD

$$\Phi_{-t}(\mathbf{x}) = \begin{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \mathbf{R}_{x_3 - a_3(t)} \cdot \begin{pmatrix} -a_1(t) \\ -a_2(t) \end{pmatrix} \\ x_3 + a_3(t) \end{pmatrix}$$