

# Vector Field Guided Path Following Control: Singularity Elimination and Global Convergence

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# Introduction

The **path following** problem deals with finding a control law for a mobile vehicle to converge to and move along a desired geometric path.

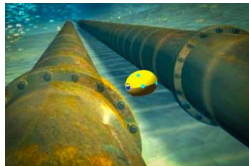
Path following is a basic function for many mobile robots, and related new applications have emerged, e.g., to probe atmospheric phenomena by drones<sup>1</sup>.



(a) wheeled robots



(b) aerial robots



(c) underwater robots

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<sup>1</sup>Lacroix, et al., 2016, "Fleets of enduring drones to probe atmospheric phenomena with clouds"

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Among many path following algorithms, we focus on those using a **guiding vector field**.

## Vector Field Guided Path Following

A **vector field** is designed such that the corresponding integral curves converge to the desired path. It acts as **guidance**.

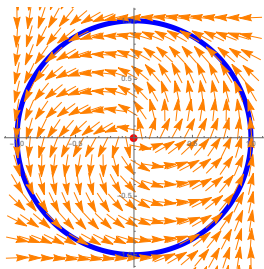
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Take a 2D *normalized* vector field as an example (Kapitanyuk et al., 2017).



$$\phi(x, y) = x^2 + y^2 - R^2 = 0$$

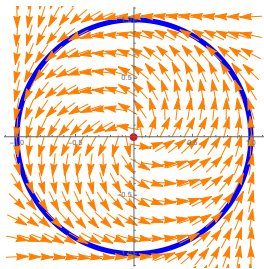
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### Advantages:

1. Mild restrictions on initial conditions;
2. Intuitive, practical;
3. Most accurate, least control effort among several algorithms (Sujit et al., 2014).

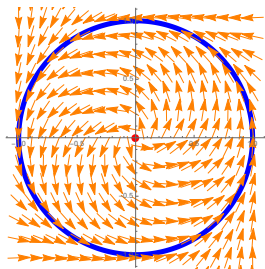
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### Challenges:

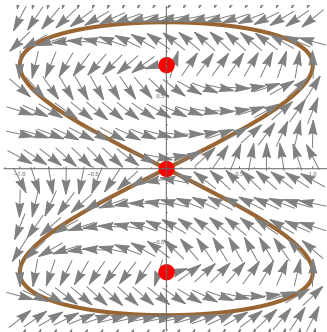
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### Challenges:

1. Singular points;
2. Self-intersected desired paths;

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How to deal with **self-intersected** desired paths?

How to guarantee **global convergence** to the **desired path**?

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# Preliminaries

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## High-dimensional Guiding Vector Field

Desired path  $\mathcal{P} \subseteq \mathbb{R}^n$

$$\mathcal{P} = \{\xi \in \mathbb{R}^n : \phi_i(\xi) = 0, i = 1, \dots, n - 1\},$$

where  $\phi_i \in C^2 : \mathbb{R}^n \rightarrow \mathbb{R}$ .

The value  $\phi_i(\xi)$  is called the *path-following error*. The desired path is a one-dimensional connected  $C^2$  manifold.

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Vector field  $\chi : \mathbb{R}^n \rightarrow \mathbb{R}^n$

$$\chi = \underbrace{\times(\nabla\phi_1, \dots, \nabla\phi_{n-1})}_{\text{propagation term}} + \underbrace{\sum_{i=1}^{n-1} -k_i\phi_i\nabla\phi_i}_{\text{converging term}},$$

# High-dimensional Guiding Vector Field

## Remark 1 (Categories of Desired Paths)

*Desired paths can generally be classified into those homeomorphic to the unit circle  $\mathbb{S}^1$  if they are compact, and those homeomorphic to the real line  $\mathbb{R}$  otherwise.*

*Notions interchangeably used throughout the talk:*

*simple closed  $\iff$  homeomorphic to  $\mathbb{S}^1$*

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*Note that self-intersected desired paths do not fall into either of the category, but can be transformed into unbounded and non-self-intersected desired paths introduced later.*



# High-dimensional Guiding Vector Field

## Remark 2 (Dimensions of Desired Paths)

*Topologically, the desired path  $\mathcal{P}$  itself is **one-dimensional**, independent of the dimensions of the Euclidean space where it lives.*

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*Topologically, the desired path  $\mathcal{P}$  itself is **one-dimensional**, independent of the dimensions of the Euclidean space where it lives.*

*However, we call the desired path  $\mathcal{P}$   $n$ -D (or  $nD$ ) if it lives in the  $n$ -dimensional Euclidean space  $\mathbb{R}^n$  and **not** in any lower-dimensional subspace  $\mathcal{W} \subseteq \mathbb{R}^n$  (i.e., the smallest subspace the desired path lives in).*

# Issues on the Global Convergence to Desired Paths

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## General Convergence Results

We investigate the convergence property of the integral curves of the vector field  $\chi$ ; that is, the solutions to the following **autonomous ODE**:

$$\dot{\xi}(t) = \chi(\xi(t)).$$

### Lemma 1

*Under some assumptions<sup>2</sup>, the integral curves of the guiding vector field  $\chi : \mathbb{R}^n \rightarrow \mathbb{R}^n$  converge either to the desired path  $\mathcal{P} \subseteq \mathbb{R}^n$ , or to the singular set  $\mathcal{C} \subseteq \mathbb{R}^n$ .*

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<sup>2</sup>Yao, Kapitanjuk, Cao, 2018, CDC

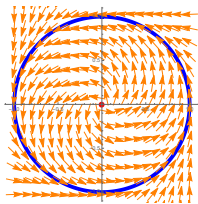
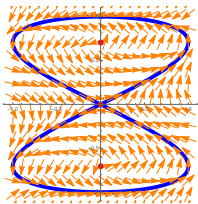
# Topological Obstruction

## Theorem 1 (Crossing Points are Singular Points)

*A crossing point of a self-intersected desired path is also a singular point of the guiding vector field.*

## Theorem 2 (Impossibility of global convergence)

*If a desired path is **simple closed**, then it is not possible to guarantee the global convergence to the desired path.*



# Topological Obstruction

## Remark 3

*Root causes:*

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- Based on the new desired paths, we can derive a **higher-dimensional** vector field of which the singular set is **empty**.

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- We construct **unbounded** and **non-self-intersected** desired paths from the originally simple closed or self-intersected desired paths by “**cutting**” and “**stretching**”.
- Based on the new desired paths, we can derive a **higher-dimensional** vector field of which the singular set is **empty**.
- To go back to the original desired path, we “**project**” it onto the original lower-dimensional space.

# Higher-dimensional Singularity-free Vector Field Construction

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## Singularity-free vector field

Suppose the 2D desired path  $\mathcal{P}^{\text{phy}}$  is parameterized by

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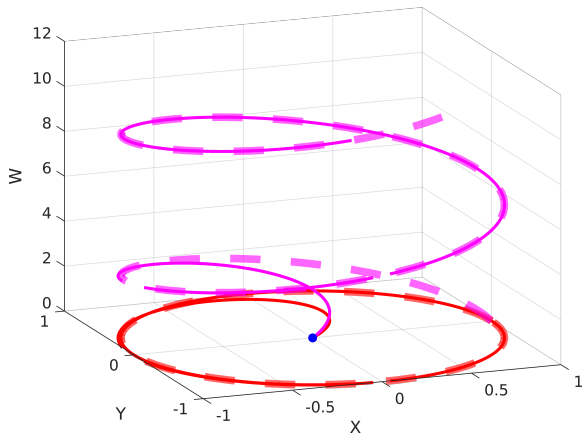
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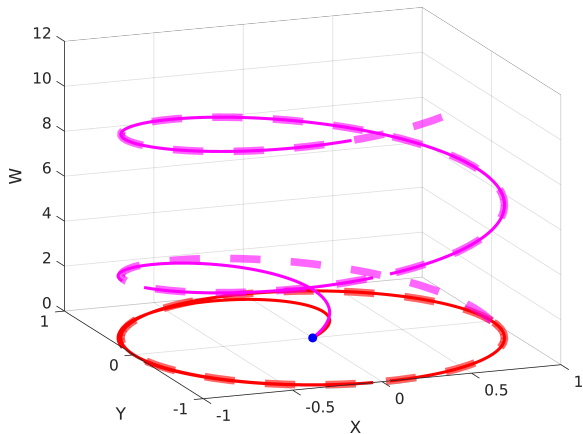
Intuitively, the higher-dimensional desired path  $\mathcal{P}$  is obtained by “cutting” and “stretching” the 2D desired path  $\mathcal{P}^{\text{phy}}$  along the virtual  $w$ -axis.



# Singularity-free vector field



## Singularity-free vector field



Are there singular points in the new **higher-dimensional** vector field  $\chi : \mathbb{R}^{2+1} \rightarrow \mathbb{R}^{2+1}$ ?

## Singularity-free vector field

The propagation term:

$$\nabla\phi_1 \times \nabla\phi_2 = \begin{bmatrix} f'_1(w) \\ f'_2(w) \\ 1 \end{bmatrix}.$$

No singular points; that is,  $\mathcal{C}^{\text{hgh}} = \emptyset$ .

## Singularity-free vector field

We use the linear transformation operator  $P_a = I - aa^\top = \begin{bmatrix} I_2 & 0 \\ 0 & 0 \end{bmatrix}$ , where  $a = (0, 0, 1)^\top$ . The effect is to “zero” the last entry of a vector.

Then we have:

$$\pi_{(1,\dots,n)}(\mathcal{P}^{\text{trs}}) = \mathcal{P}^{\text{phy}}, \text{ where } \mathcal{P}^{\text{trs}} = P_a(\mathcal{P}^{\text{hgh}}).$$

## Singularity-free vector field

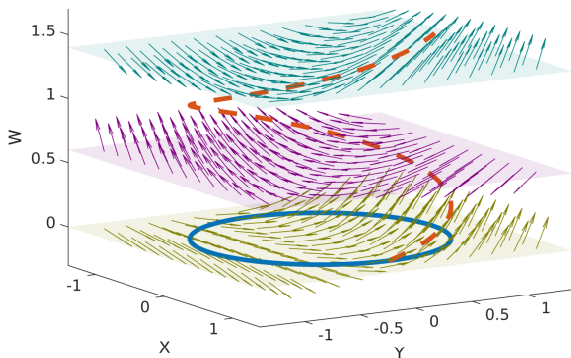
### Theorem 3

Given a 2D parameterized physical desired path  $\mathcal{P}^{\text{phy}} \subseteq \mathbb{R}^2$ . Let  $\phi_1, \phi_2$  be chosen as before. Then there are **no singular points** in the corresponding **(2 + 1)**-dimensional vector field  $\chi^{\text{high}} : \mathbb{R}^{2+1} \rightarrow \mathbb{R}^{2+1}$ . Let  $a = (0, 0, 1)^\top$  for the linear transformation operator  $P_a$ . Then the **projected transformed trajectory**

$$\xi^{\text{prj}}(t) = (x_1(t), x_2(t))^\top$$

globally asymptotically converges to the physical desired path  $\mathcal{P}^{\text{phy}}$  as  $t \rightarrow \infty$ .

## Singularity-free vector field



**Figure 1:** Three layers of the 3D vector field corresponding to a circle. The solid line is the 2D desired path while the dashed line is the corresponding 3D (unbounded) desired path. Three layers of the 3D vector field evaluated at  $w = 0, 0.6, 1.4$  respectively are illustrated.

# Simulations

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## Simulations

Red point:  $(f_1(w), f_2(w))$

Actual aircraft movement

Virtual aircraft movement (due to the  
virtual  $w$ -axis)



## Conclusions

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## Conclusions

- Transform simple closed or self-intersected to **unbounded and non-self-intersected** virtual desired paths in a higher-dimensional space;
- Then create a **high-dimensional** singularity-free vector field;
- **Global convergence** to the desired paths, which can be even **self-intersected**, is guaranteed.
- Extensions to  $n$ -dimensional, some other appealing features of the vector field and implementation details on a fixed-wing UAV can be found in<sup>3</sup>

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<sup>3</sup>W. Yao, H. G. de Marina, B. Lin, and M. Cao, "Singularity-free guiding vector field for robot navigation," IEEE Transactions on Robotics, 2020, accepted for publication.

# Thank you!

Feel free to contact me (Weijia Yao) via  
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