

**Stabilisierung kontroll-affiner Systeme
durch Lie-Klammer Approximationen mit
Anwendungen zur abstands-basierten Formationssteuerung**

Raik Suttner

Institute of Mathematics, University of Wuerzburg, Germany

Zhiyong Sun

Research School of Engineering, Australian National University, Australia

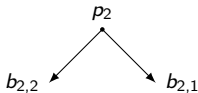
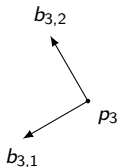
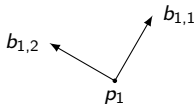
Motivation: distanced-based formation control

Model: N kinematic points in \mathbb{R}^n :

$$\dot{p}_i = \sum_{k=1}^n u_{i,k} b_{i,k}, \quad i = 1, \dots, N,$$

where

- $p_i = (p_{i,1}, \dots, p_{i,n}) \in \mathbb{R}^n$ position of agent i ,
- $b_{i,1}, \dots, b_{i,n} \in \mathbb{R}^n$ orthonormal basis of velocity direction,
- $u_{i,1}, \dots, u_{i,n}$ input channels for velocities.

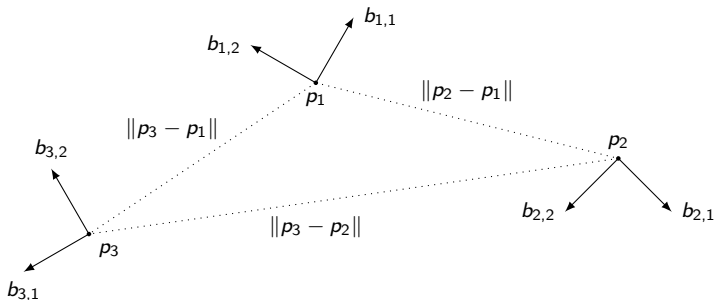


Motivation: distanced-based formation control

Measurements: described by an undirected graph $G = (V, E)$, where

- $V = \{1, \dots, N\}$ set of nodes,
- $E = \{\{i, j\} \subseteq V \mid |\{i, j\}| = 2\}$ set of edges, we write $ij := \{i, j\}$.

If $ij \in E$, then agent i can measure *distance* (\neq relative position) to agent j and vice versa.



Motivation: distanced-based formation control

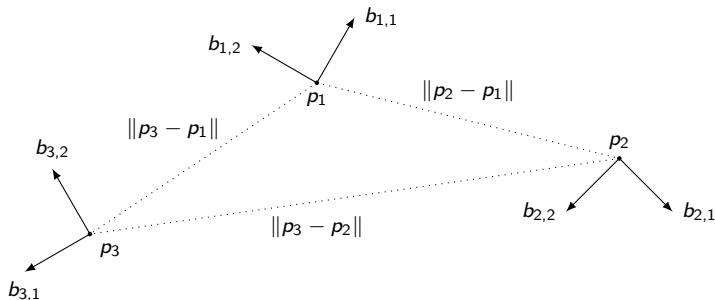
Objective: steer agents to state $p = (p_1, \dots, p_N) \in \mathbb{R}^{nN}$ such that

$$\forall ij \in E: \quad \|p_j - p_i\| = d_{ij},$$

where $d_{ij} > 0$ *desired distance* between agents i, j .

Constraints:

- autonomous agents \Rightarrow decentralized, distributed control law,
- each agent can only use its own distance measurements,
- no shared clock, no storage of data.



Motivation: distanced-based formation control

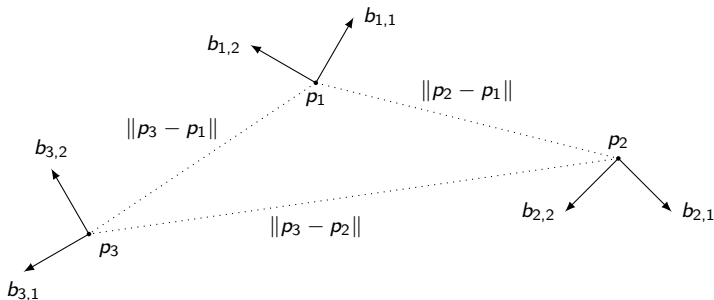
Standard approach: gradient-based control law.

Local potential function for agent i :

$$\psi_i(p) := \frac{1}{4} \sum_{j \in V: ij \in E} (\|p_j - p_i\|^2 - d_{ij}^2)^2$$

Control law for agent i :

$$u_i = (u_{i,1}, \dots, u_{i,n}) = -\nabla_{p_i} \psi_i(p) = \frac{1}{2} \sum_{j \in V: ij \in E} (p_j - p_i) (\|p_j - p_i\|^2 - d_{ij}^2)$$



Motivation: distanced-based formation control

Standard approach: gradient-based control law.

Local potential function for agent i :

$$\psi_i(p) := \frac{1}{4} \sum_{j \in V: ij \in E} (\|p_j - p_i\|^2 - d_{ij}^2)^2$$

Control law for agent i :

$$u_i = (u_{i,1}, \dots, u_{i,n}) = -\nabla_{p_i} \psi_i(p) = \frac{1}{2} \sum_{j \in V: ij \in E} (p_j - p_i)(\|p_j - p_i\|^2 - d_{ij}^2)$$

Then, set of *target formations*

$$F := \{p = (p_1, \dots, p_n) \in \mathbb{R}^{nN} \mid \forall ij \in E: \|p_j - p_i\| = d_{ij}\}$$

- is locally asymptotically stable (Krick et al., 2009; Oh, Ahn, 2014)
- is locally exponentially stable if frameworks (G, p) with $p \in F$ are *infinitesimally rigid* (Mou et al, 2016, Sun et al., 2016)

Motivation: distanced-based formation control

Standard approach: gradient-based control law.

Local potential function for agent i :

$$\psi_i(p) := \frac{1}{4} \sum_{j \in V: ij \in E} (\|p_j - p_i\|^2 - d_{ij}^2)^2$$

Control law for agent i :

$$u_i = (u_{i,1}, \dots, u_{i,n}) = -\nabla_{p_i} \psi_i(p) = \frac{1}{2} \sum_{j \in V: ij \in E} (p_j - p_i) (\|p_j - p_i\|^2 - d_{ij}^2)$$

Then, set of *target formations*

$$F := \{p = (p_1, \dots, p_n) \in \mathbb{R}^{nN} \mid \forall ij \in E: \|p_j - p_i\| = d_{ij}\}$$

- is locally asymptotically stable (Krick et al., 2009; Oh, Ahn, 2014)
- is locally exponentially stable if frameworks (G, p) in F are *infinitesimally rigid* (Mou et al, 2016, Sun et al., 2016)

Obstacle: implementation requires measurements of relative positions
actively controlled variable \neq sensed variable

Motivation: distanced-based formation control

Approach: approximate the gradient by Lie brackets.

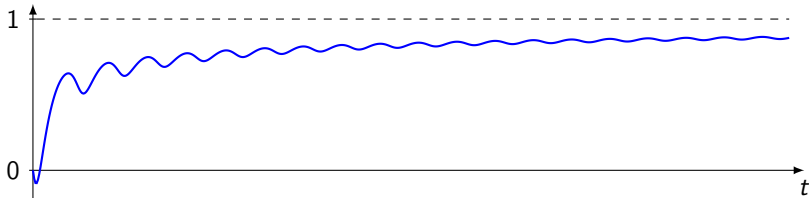
Toy model: $N = 1$ agent in $n = 1$ dimension, i.e., $\dot{p} = u \in \mathbb{R}$,
with potential $\psi(p) := \frac{1}{2}|p - 1|^2$.

Define $u_1, u_2, h_1, h_2: \mathbb{R} \rightarrow \mathbb{R}$ by

$$u_1(t) := \cos t, \quad u_2(t) := \sin t,$$

$$h_1(y) := \begin{cases} y \sin(\log y) & y > 0 \\ 0 & y \leq 0 \end{cases}, \quad h_2(y) := \begin{cases} y \cos(\log y) & y > 0 \\ 0 & y \leq 0 \end{cases}.$$

Control law: $u := u_1(t) h_1(\psi(p)) + u_2(t) h_2(\psi(p))$



Motivation: distanced-based formation control

Approach: approximate the gradient by Lie brackets.

Toy model: $N = 1$ agent in $n = 1$ dimension, i.e., $\dot{p} = u \in \mathbb{R}$, with potential $\psi(p) := \frac{1}{2}|p - 1|^2$.

Define $u_1, u_2, h_1, h_2: \mathbb{R} \rightarrow \mathbb{R}$ by

$$u_1(t) := \cos t, \quad u_2(t) := \sin t,$$

$$h_1(y) := \begin{cases} y \sin(\log y) & y > 0 \\ 0 & y \leq 0 \end{cases}, \quad h_2(y) := \begin{cases} y \cos(\log y) & y > 0 \\ 0 & y \leq 0 \end{cases}.$$

Control law: $u := u_1(t) h_1(\psi(p)) + u_2(t) h_2(\psi(p))$

“**Lie bracket**” of $f_1 := (h_1 \circ \psi)$ and $f_2 := (h_2 \circ \psi)$:

$$[f_1, f_2](p) := f_2'(p) f_1(p) - f_1'(p) f_2(p) = - \underbrace{\psi(p)}_{> 0 \text{ for } p \neq 1} \psi'(p)$$

Approximation of Lie Brackets

Closed-loop system: $\dot{p} = u_1(t) f_1(p) + u_2(t) f_2(p) = \sum_i u_i(t) f_i(p)$.

Integral equation, averaging procedure:

$$p(t) = p(t_0) + \sum_i \int_{t_0}^t \underbrace{u_i(s)}_{\uparrow} \underbrace{f_i(p(s))}_{\downarrow} ds$$

Approximation of Lie Brackets

Closed-loop system: $\dot{p} = u_1(t) f_1(p) + u_2(t) f_2(p) = \sum_i u_i(t) f_i(p)$.

Integral equation, averaging procedure:

$$\begin{aligned}
 p(t) &= p(t_0) + \sum_i \int_{t_0}^t \underbrace{u_i(s)}_{\uparrow} \underbrace{f_i(p(s))}_{\downarrow} ds \\
 &= p(t_0) + \sum_i [U_i(s) f_i(p(s))]_{t_0}^t - \sum_{i,j} \int_{t_0}^t U_i(s) u_j(s) f_i'(p(s)) f_j(p(s)) ds \\
 &= p(t_0) + \sum_i [U_i(s) f_i(p(s))]_{t_0}^t - \int_{t_0}^t \frac{1}{2} [f_1, f_2](p(s)) ds \\
 &\quad - \sum_{i,j} \int_{t_0}^t \underbrace{\widetilde{UV}_{j,i}(s)}_{\uparrow} \underbrace{f_i'(p(s)) f_j(p(s))}_{\downarrow} ds
 \end{aligned}$$

Approximation of Lie Brackets

Closed-loop system: $\dot{p} = u_1(t) f_1(p) + u_2(t) f_2(p) = \sum_i u_i(t) f_i(p)$.

Integral equation, averaging procedure:

$$\begin{aligned}
 p(t) &= p(t_0) + \sum_i \int_{t_0}^t \underbrace{u_i(s)}_{\uparrow} \underbrace{f_i(p(s))}_{\downarrow} ds \\
 &= p(t_0) + \sum_i [U_i(s) f_i(p(s))]_{t_0}^t - \sum_{i,j} \int_{t_0}^t U_i(s) u_j(s) f_i'(p(s)) f_j(p(s)) ds \\
 &= p(t_0) + \sum_i [U_i(s) f_i(p(s))]_{t_0}^t - \int_{t_0}^t \frac{1}{2} [f_1, f_2](p(s)) ds \\
 &\quad - \sum_{i,j} \int_{t_0}^t \underbrace{\widetilde{UV}_{j,i}(s)}_{\uparrow} \underbrace{f_i'(p(s)) f_j(p(s))}_{\downarrow} ds \\
 &= p(t_0) + [D_1(s, p(s))]_{t_0}^t + \int_{t_0}^t \left(\frac{1}{2} [f_1, f_2](p(s)) + D_2(s, p(s)) \right) ds
 \end{aligned}$$

with certain time-varying remainder vector fields D_1, D_2 .

Approximation of Lie Brackets

Closed-loop system: $\dot{p} = u_1(t) f_1(p) + u_2(t) f_2(p)$

Integral equation, averaging procedure:

$$p(t) = p(t_0) + [D_1(s, p(s))]_{t_0}^t + \int_{t_0}^t \left(\frac{1}{2} [f_1, f_2](p(s)) + D_2(s, p(s)) \right) ds$$

Approximation of Lie Brackets

Closed-loop system: $\dot{p} = u_1(t) f_1(p) + u_2(t) f_2(p)$

Integral equation, averaging procedure:

$$p(t) = p(t_0) + [D_1(s, p(s))]_{t_0}^t + \int_{t_0}^t \left(\frac{1}{2} [f_1, f_2](p(s)) + D_2(s, p(s)) \right) ds$$

A detailed analysis shows: D_1, D_2 vanish sufficiently fast compared to $[f_1, f_2]$ when $\psi(p) \rightarrow 0$. Thus, for $\psi(p)$ close to 0:

$$p(t) \approx p(t_0) + \int_{t_0}^t \frac{1}{2} [f_1, f_2](p(s)) ds = p(t_0) - \frac{1}{2} \int_{t_0}^t \psi(p(s)) \psi'(p(s)) ds$$

\Rightarrow trajectories follow approximately $[f_1, f_2]$.

Approximation of Lie Brackets

Toy model: system $\dot{p} = u_1(t) f_1(p) + u_2(t) f_2(p)$ on \mathbb{R} .

More general: averaging analysis by Kurzweil, Jarník (1988); Sussmann, Liu (1991). Control-affine system

$$\dot{p} = f_0(p) + \sum_{i=1}^m u_i^\omega(t) f_i(p),$$

with

- a drift vector field $f_0: \mathbb{R}^n \rightarrow \mathbb{R}^n$,
- suff. smooth control vector fields $f_1, \dots, f_m: \mathbb{R}^n \rightarrow \mathbb{R}^n$,
- time-varying functions $u_1^\omega, \dots, u_m^\omega: \mathbb{R} \rightarrow \mathbb{R}$ of the form

$$u_i^\omega(t) = \Lambda_i(t) \omega^{e_i} \sin(\omega \Omega_i t + \Phi_i),$$

where $\omega > 0$ is a frequency parameter

Approximation of Lie Brackets

Control-affine system

$$\dot{p} = f_0(p) + \sum_{i=1}^m u_i^\omega(t) f_i(p),$$

Integral equation, averaging procedure:

$$p(t) = p(t_0) + [D_1^\omega(s, p(s))]_{t_0}^t + \int_{t_0}^t \left(f_0(p(s)) + \sum_{0 < |l| \leq r} \frac{v_l(s)}{|l|} [f_l](p(s)) + D_2^\omega(s, p(s)) \right) ds,$$

where

- D_1^ω, D_2^ω certain time-varying remainder vector fields,
- $l = (i_1, \dots, i_k)$ multi-index of length $|l| = k$,
- $v_l: \mathbb{R} \rightarrow \mathbb{R}$ time-varying function (depending on u_i^ω),
- $[f_l] := [f_{i_1}, [f_{i_2}, \dots, [f_{i_{k-1}}, f_{i_k}] \dots]]$ iterated Lie bracket.

One can show that $D_1^\omega, D_2^\omega \rightarrow 0$ as $\omega \rightarrow \infty$ on compact sets.

Approximation of Lie Brackets

$$\omega\text{-dependent system} \quad \Sigma^\omega: \dot{p} = f_0(p) + \sum_{i=1}^m u_i^\omega(t) f_i(p)$$

$$\text{Limit system} \quad \Sigma^\infty: \dot{p} = f_0(p) + \sum_{0 < |I| \leq r} \frac{v_I(t)}{|I|} [f_I](p)$$

Theorem (Kurzweil, Jarník (1988); Sussmann, Liu (1991))

For every initial condition (t_0, p_0) , every $T > 0$, and every $\varepsilon > 0$, there exists $\omega_0 > 0$ such that for every $\omega \geq \omega_0$, and every $t \in [t_0, t_0 + T]$:

$$\|x^\infty(t) - x^\omega(t)\| \leq \varepsilon.$$

Application: motion planning for nonholonomic systems.

Approximation of Lie Brackets

$$\omega\text{-dependent system} \quad \Sigma^\omega : \dot{p} = f_0(p) + \sum_{i=1}^m u_i^\omega(t) f_i(p)$$

$$\text{Limit system} \quad \Sigma^\infty : \dot{p} = f_0(p) + \sum_{0 < |I| \leq r} \frac{v_I(t)}{|I|} [f_I](p)$$

Theorem (Dürr et al. (2013))

Case $r = 2$ and $u_i^\omega(t) = \Lambda_i(t) \sqrt{\omega} \sin(\omega \Omega_i t + \Phi_i)$, Λ_i bounded.

For every compact set $K \subseteq \mathbb{R}^n$, every $T > 0$, and every $\varepsilon > 0$, there exists $\omega_0 > 0$ such that for every $\omega \geq \omega_0$, every $(t_0, p_0) \in \mathbb{R} \times K$, and every $t \in [t_0, t_0 + T]$:

$$\|x^\infty(t) - x^\omega(t)\| \leq \varepsilon$$

as long as x^∞ stays in K .

Application: stabilization of control-affine systems.

Stabilization of Control-Affine Systems

$$\omega\text{-dependent system} \quad \Sigma^\omega: \dot{p} = f_0(p) + \sum_{i=1}^m u_i^\omega(t) f_i(p)$$

$$\text{Limit system} \quad \Sigma^\infty: \dot{p} = f_0(p) + \sum_{0 < |l| \leq r} \frac{v_l(t)}{|l|} [f_l](p)$$

Theorem (Dürr et al. (2013))

Case $r = 2$ and $u_i^\omega(t) = \Lambda_i(t) \sqrt{\omega} \sin(\omega \Omega_i t + \Phi_i)$, Λ_i bounded.

For a compact $K \subseteq \mathbb{R}^n$, the following implication holds:

$$K \text{ is LUAS for } \Sigma^\infty \quad \implies \quad K \text{ is } \omega\text{-PLUAS for } \Sigma^\omega$$

LUAS = **l**ocally **u**niformly **a**symptotically **s**table.

ω -PLUAS = **p**ractically LUAS if ω sufficiently large.

Stabilization of Control-Affine Systems

$$\omega\text{-dependent system} \quad \Sigma^\omega: \dot{p} = f_0(p) + \sum_{i=1}^m u_i^\omega(t) f_i(p)$$

$$\text{Limit system} \quad \Sigma^\infty: \dot{p} = f_0(p) + \sum_{0 < |l| \leq r} \frac{v_l(t)}{|l|} [f_l](p)$$

Theorem (S. (2017))

General $r \in \mathbb{N}$, and $K \subseteq \mathbb{R}^n$ compact.

Suppose that u_i^ω satisfy certain averaging conditions as $\omega \rightarrow 0$.

(a) K is LUAS for $\Sigma^\infty \implies K$ is ω -PLUAS for Σ^ω

(b) If vector fields satisfy certain boundedness conditions, then:

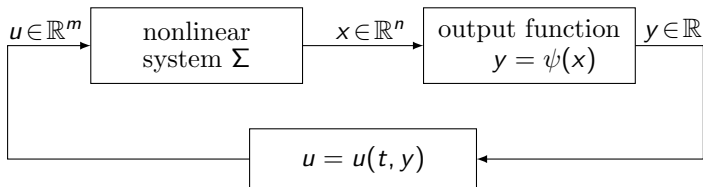
$$K \text{ is LUES for } \Sigma^\infty \implies K \text{ is } \omega\text{-LUES for } \Sigma^\omega$$

LUES=locally **u**niformly **e**xponentially **s**table.

ω -LUES=LUES if ω sufficiently large.

Stabilization of Control-Affine Systems

Application to extremum seeking control.

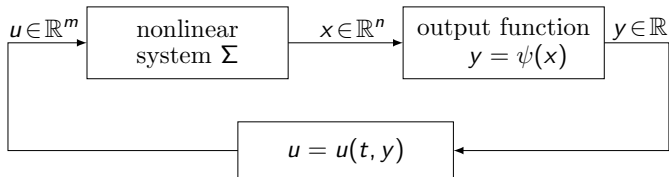


Objective: find a control law u that steers the system to an extremum of ψ (w.l.o.g. minimum of ψ)

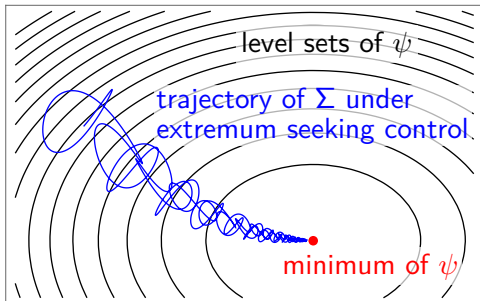
Restrictions:

- no information about the current system state x ,
- no explicit knowledge about the output function $x \mapsto \psi(x)$,
- only real-time measurements of $y = \psi(x)$ are available

Application to extremum seeking control.



Idea by Dürr et al.(2013): approximate descent directions of ψ by Lie brackets (feed in sinusoidal perturbations).



Application to extremum seeking control

Applicable to control-affine systems

$$\dot{p} = g_0(p) + \sum_{k=1}^m u_k g_k(p), \quad y = \psi(p).$$

Dürr et al. (2013):

- Define $u_{2k-1}^\omega(t) := \sqrt{\omega \Omega_k} \cos(\omega \Omega_k t)$, $u_{2k}^\omega(t) := \sqrt{\omega \Omega_k} \sin(\omega \Omega_k t)$.
- Define $h_1(y) := y$ and $h_2(y) := 1$ for every $y \in \mathbb{R}$.

Control law: $u_k = u_{2k-1}^\omega(t) h_1(\psi(p)) + u_{2k}^\omega(t) h_2(\psi(p))$.

Application to extremum seeking control

Applicable to control-affine systems

$$\dot{p} = g_0(p) + \sum_{k=1}^m u_k g_k(p), \quad y = \psi(p).$$

Dürr et al. (2013):

- Define $u_{2k-1}^\omega(t) := \sqrt{\omega \Omega_k} \cos(\omega \Omega_k t)$, $u_{2k}^\omega(t) := \sqrt{\omega \Omega_k} \sin(\omega \Omega_k t)$.
- Define $h_1(y) := y$ and $h_2(y) := 1$ for every $y \in \mathbb{R}$.

Control law: $u_k = u_{2k-1}^\omega(t) h_1(\psi(p)) + u_{2k}^\omega(t) h_2(\psi(p))$.

New vector fields $f_{2k-1} := (h_1 \circ \psi)g_k$ and $f_{2k} := (h_2 \circ \psi)g_k$.

$$\omega\text{-dependent system} \quad \Sigma^\omega : \dot{p} = g_0(p) + \sum_{i=1}^{2m} u_i^\omega(t) f_i(p),$$

$$\text{Limit system} \quad \Sigma^\infty : \dot{p} = g_0(p) + \frac{1}{2} \sum_{k=1}^m [f_{2k-1}, f_{2k}](p),$$

with descent directions $[f_{2k-1}, f_{2k}](p) = -(\mathcal{L}_{g_k} \psi)(p)g_k(p)$.

Application to extremum seeking control

$$h_1(y) := y, \quad h_2(y) := 1.$$

$$\omega\text{-dependent system} \quad \Sigma^\omega: \dot{p} = g_0(p) + \sum_{k=1}^m u_k g_k(p)$$

$$\text{under control law} \quad u_k = u_{2k-1}^\omega(t) h_1(\psi(p)) + u_{2k}^\omega(t) h_2(\psi(p)),$$

$$\text{Limit system} \quad \Sigma^\infty: \dot{p} = g_0(p) - \frac{1}{2} \sum_{k=1}^m (\mathcal{L}_{g_k} \psi)(p) g_k(p),$$

Theorem (Dürr et al. (2013))

Suppose that $K \subseteq \mathbb{R}^n$ is a compact set of minima of ψ . Then, the following implication holds:

$$K \text{ is LUAS for } \Sigma^\infty \quad \implies \quad K \text{ is } \omega\text{-PLUAS for } \Sigma^\omega.$$

Application to extremum seeking control

$$h_1(y) := \sqrt{y} \sin(\log y), \quad h_2(y) := \sqrt{y} \cos(\log y).$$

$$\omega\text{-dependent system} \quad \Sigma^\omega: \dot{p} = g_0(p) + \sum_{k=1}^m u_k g_k(p)$$

$$\text{under control law} \quad u_k = u_{2k-1}^\omega(t) h_1(\psi(p)) + u_{2k}^\omega(t) h_2(\psi(p)),$$

$$\text{Limit system} \quad \Sigma^\infty: \dot{p} = g_0(p) - \frac{1}{2} \sum_{k=1}^m (\mathcal{L}_{g_k} \psi)(p) g_k(p),$$

Theorem (S. (2017))

Suppose that $\psi \geq 0$ and that $K := \psi^{-1}(\{0\})$ is compact.

- (a) K is LUAS for $\Sigma^\infty \implies K$ is ω -PLUAS for Σ^ω
 (b) If the drift g_0 satisfies certain boundedness conditions, then:

$$K \text{ is LUES for } \Sigma^\infty \implies K \text{ is } \omega\text{-LUES for } \Sigma^\omega$$

Stabilization of control-affine systems

$$h_1(y) := \sqrt{y} \sin(\log y), \quad h_2(y) := \sqrt{y} \cos(\log y) \text{ for } y > 0.$$

Example

Let $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{m \times n}$. Consider the system

$$\Sigma^\omega: \quad \dot{x} = Ax + Bu, \quad y = \psi(x) := \|x\|^2$$

with $u = (u_1, \dots, u_m)^\top$ under the control law

$$u_k = \lambda u_{2k-1}^\omega(t) h_1(\psi(x)) + \lambda u_{2k}^\omega(t) h_2(\psi(x)).$$

with $\lambda, \omega > 0$. Assume that $\text{rank } B = n$. Limit system

$$\Sigma^\infty: \quad \dot{x} = (A - \lambda^2 B^\top B)x$$

If $\lambda, \omega > 0$ are sufficiently large, the $\{0\}$ is LUES for Σ^ω .

Stabilization of control-affine systems

Remark: choice of functions h_1, h_2 in

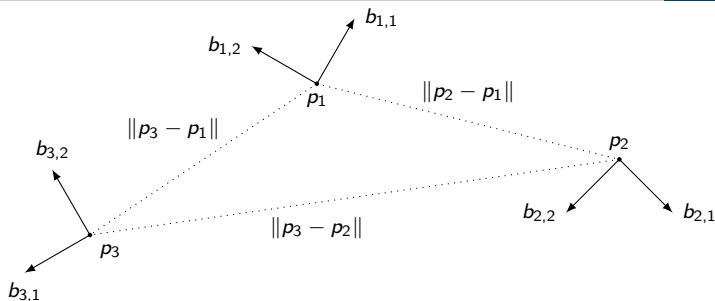
$$u_k = u_{2k-1}^\omega(t) h_1(\psi(p)) + u_{2k}^\omega(t) h_2(\psi(p))$$

- Dürr et al. (2013): $h_1(y) = y$ and $h_2(y) = 1$
→ ω -PLUAS
- Scheinker et al. (2014): $h_1(y) = \sin y$ and $h_2(y) = \cos y$
→ ω -PLUAS
- S. (2017): $h_1(y) = \sqrt{y} \sin(\log y)$ and $h_2(y) = \sqrt{y} \cos(\log y)$
→ ω -PLUAS and also ω -LUES
- S., Sun (2018): $h_1(y) = y \sin(\log y)$ and $h_2(y) = y \cos(\log y)$
→ LUAS (for every $\omega > 0$), but not LUES

Integral equation, averaging procedure:

$$p(t) = p(t_0) + [D_1^\omega(s, p(s))]_{t_0}^t + \int_{t_0}^t \left(f^\infty(p(s)) + D_2^\omega(s, p(s)) \right) ds,$$

Application to distanced-based formation control



“Team” of control-affine systems

$$\dot{p}_i = \sum_{k=1}^n u_{i,k} b_{i,k}, \quad y_i = \psi_i(p), \quad i = 1, \dots, N,$$

output functions = local potential functions

$$\psi_i(p) := \frac{1}{4} \sum_{j \in V: ij \in E} (\|p_j - p_i\|^2 - d_{ij}^2)^2$$

can be computed by individual distance measurements.

Application to distanced-based formation control

$h_1(y) := y \sin(\log y)$, $h_2(y) := y \cos(\log y)$ for $y > 0$.

ω -dependent system $\Sigma^\omega: \dot{p}_i = \sum_{k=1}^n u_{i,k} b_{i,k}$, $i = 1, \dots, N$

under control law $u_{i,k} = u_{i,k,1}^\omega(t) h_1(\psi_i(p)) + u_{i,k,2}^\omega(t) h_2(\psi_i(p))$,

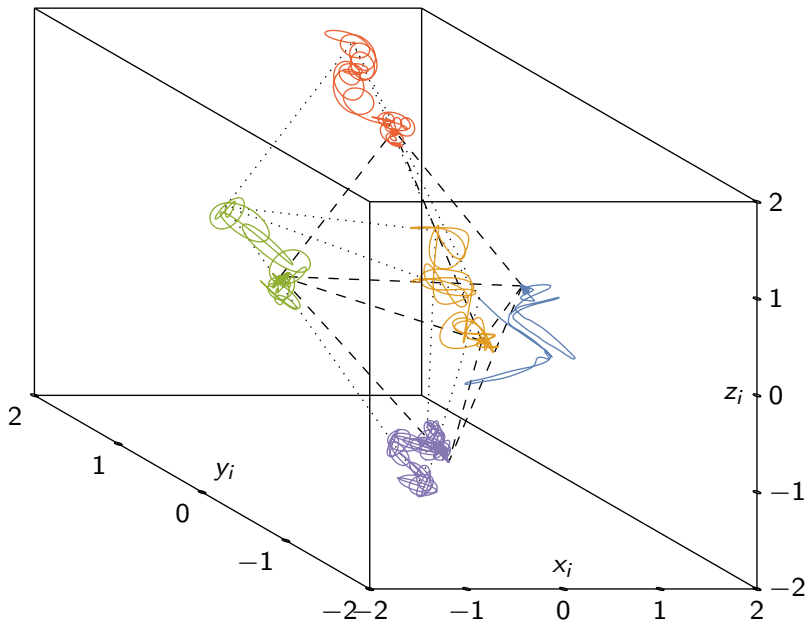
Limit system $\Sigma^\infty: \dot{p}_i = -\frac{1}{2} \psi_i(p) \nabla_{p_i} \psi_i(p)$, $i = 1, \dots, N$.

Theorem (S., Sun(2018))

*Suppose that the frameworks (G, p) with $p \in F$ are infinitesimally rigid. Then, **for every** $\omega > 0$, the set F is LUAS for Σ^ω .*

Remark: the domain of attraction increases when ω increases.

Application to distanced-based formation control



Thank you for your attention!