# Path Planning in Environments with Intermittent State Feedback

S. C. Edwards, D. M. Le, D. P. Guralnik and W. E. Dixon, "A Topologically Inspired Path-Following Method With Intermittent State Feedback," in IEEE Robotics and Automation Letters, vol. 6, no. 3, pp. 4449-4456, July 2021, doi: 10.1109/LRA.2021.3067295.

















# Motivation



- Systems often operate in environments that include:
  - A2AD (Anti Access/Area Denial) Environments
  - Unknown Terrain
  - Complex Environments (Mountains, Extreme Weather, Marine, etc.)
- Limitations:
  - Sensing (GPS, Cameras, Lidar, etc.)
  - Global Communication















### **Problem Statement**



- Task: Agent is to follow path  $(X_d)$  where state feedback is denied
- Agent must switch between two tasks:
  - Following  $X_d$
  - Obtaining state feedback to regulate tracking errors



H-Y. Chen, et. al.,"A Switched Systems Approach to Path Following with Intermittent State Feedback", IEEE TRO, 2019.















- 1. Depart from the feedback region to  $X_d$ .
- 2. Follow  $X_d$  until it is time to return to the feedback region.
- 3. Take a recommended return trajectory such to guarantee reentry into the feedback region.



















- Previous methods only considered circular feedback regions and circular paths to follow.
- This has issues...















## Jordan Curves



- Jordan Curve (*C*) separates plane into two regions
- A curve originating in the exterior with an odd number of intersections with *C* terminates in the interior
- An even number of intersections terminates in the exterior.



















- Bounding regions are used to upper bound potential trajectories
- Required for separation of the initial position state and terminal state















## Target Regions



- Using Jordan Curves, "Target Regions"  $(T_{C,R})$  are developed to guarantee an agent's re-entry.
- The return target is the center of the ball inscribed in  $T_{C,R}$ .
- This generalizes to higher dimensions.















### Maximum Allowed Uncertainty Regions (MAURs)









uke











- With a total of 46,855 data points, the meanaverage increase in the MAUR was found to be 233%.
- The largest increase was found to be 969%. This was observed in the "horseshoe" geometry (C).















### MAUR Computation Methods



- Current methods use a brute force approach
- New methods aim to reduce computation time from hours to fractions of a second
- Add other animation of old method





















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### **Topological Methods for Guaranteeing Transitions in Switched Systems**

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April 30, 2021

















### **Informal Challenge:** Given a controlled hybrid system with uncertain state, how to guarantee transitions into a desired operational mode?

 $\rightsquigarrow$  crucial for ensuring high-level plan execution

 $\rightsquigarrow$  complicated geometry of transition boundaries

 $\rightsquigarrow$  must be addressed with minimal specific geometric insight















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Guaranteed Transitions: Lack of symmetry makes the problem less intuitive...

→ It's not about thickness of the target domain

 $\rightsquigarrow$  It's about the transition boundary separating the error cone















A setup, successfully generalizing our work in [1]:

Incomplete information on state-environment interaction: agent dynamics over x ∈ F<sup>0</sup>, F ⊂ ℝ<sup>n</sup> closed, given by

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• A collection  $\mathscr{A}$  of curves  $\hat{x}: [0, \infty) \to \mathbb{R}^n$ , called *admissible plans*,

$$\mathscr{A}_p \triangleq \left\{ \hat{x} \in \mathscr{A} \colon \hat{x}(0) = p \right\},\tag{2}$$

 $\rightsquigarrow$  Usually, expect to have  $\mathscr{A}_p = p + \mathscr{A}_0 = \{p + \hat{x} : \hat{x} \in \mathscr{A}_0\}$  $\rightsquigarrow$  In Sage's problem  $\hat{x}(t) \triangleq p + t\mathbf{v}$ , with  $\|\mathbf{v}\| = v_0$  a known constant





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#### • Each plan is executed by applying a control u(t), $t \ge 0$ , guaranteeing

$$|x(t) - \hat{x}(t)|| \le \varrho(t), \ t \ge 0$$
 (3)

where  $\rho: [0,\infty) \to [0,\infty)$  is a known *error-bounding function* (EBF).

 $\rightsquigarrow$  There can be many EBFs, e.g., depending on  $p = \hat{x}(0)$ 





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- 𝒢<sub>p</sub>(ε) ⊆ 𝒢<sub>p</sub> is the collection of all plans satisfying (4).
- $p \in \mathbb{R}^n$  is  $\varepsilon$ -feasible, if  $\mathscr{G}_p(\varepsilon) \neq \emptyset$ .
- p ∈ ℝ<sup>n</sup> is *feasible* if it is ε-feasible for some ε > 0.
- ▶ What **actually** guarantees a plan?















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- ▶ What **actually** guarantees a plan?
- ▶ **Recall:** Let  $U \subset \mathbb{R}^n$  be an open domain and let  $p, q \in U$ . A closed set  $K \subset \mathbb{R}^n$ *separates* U *between* p *and* q, if p, q lie in distinct components of  $U \smallsetminus K$ .
- ▶  $\partial \mathcal{F}$  separates  $\mathbb{R}^n$  between  $p \in \mathcal{F}^{\complement}$  and any  $q \in int(\mathcal{F})$ .











Definition. The maximum allowed uncertainty radius (MAUR) at p:

$$\mathbf{M}_{\mathcal{F}}^{*}(p) \triangleq \begin{cases} \sup\left(\varepsilon > 0 \colon \mathscr{G}_{p}(\varepsilon) \neq \varnothing\right), & \text{if } p \text{ is feasible,} \\ 0 & \text{otherwise,} \end{cases}$$
(5)

and denote  $\mathcal{M}(\mathcal{F}) \triangleq \{f : \mathbb{R}^n \to \mathbb{R}_{\geq 0} \colon f \leq \mathbf{M}_{\mathcal{F}}^*\}$ , the *MAUR bounds*.

- **Problem 1.** Find explicit constructions of  $f \in \mathcal{M}(\mathcal{F})$ .  $\rightsquigarrow$  The inscribed ball criterion (IBC) leads to one such construction
- **Problem 2.** Given  $f \in \mathcal{M}(\mathcal{F})$  and  $p \in \mathbb{R}^n$ , compute f(p) and  $\hat{x} \in \mathscr{G}_p(f(p))$ .  $\rightsquigarrow$  Expect: the better f approximates  $\mathbb{M}^*_{\mathcal{F}}$ , the more complex this problem will be





• An *IBC certificate at* p is an EBF  $\varrho$  with

$$\blacktriangleright B_{\hat{x}(s)}(\varrho(s)) \subseteq \mathcal{F},$$

for some  $\hat{x} \in \mathscr{A}_p$ ,  $s \ge 0$ .









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► Then one defines the *IBC-MAUR*:

and  $\mathbf{M}_{\mathcal{F}}^{^{IBC}}(p) \triangleq 0$  otherwise.







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 $\blacktriangleright \text{ Clearly, } \mathbf{M}_{\mathcal{F}}^{^{IBC}} \leq \mathbf{M}_{\mathcal{F}}^{*}.$ 

The argument takes no account of the error cone!

 $\rightsquigarrow$  Information about  $\boldsymbol{x}(t)$  is thrown away













- Let U be a collection of open domains such that
  - $\blacktriangleright \ \mathbb{R}^n \in \mathcal{U},$
  - ▶  $0 \in R$  for all  $R \in \mathcal{U}$ .
  - $\label{eq:linear} \begin{array}{l} \rightsquigarrow \mbox{ In [1], $\mathcal{U}$ consists of all strips centered at $0 \in \mathbb{R}^2$} \\ \rightsquigarrow \mbox{ $\mathcal{U}$ expresses a bound on the growth of error cones} \end{array}$

















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Duke

▶ Given  $R \in \mathcal{U}$ ,  $p \in \mathbb{R}^n$  the *target at p*,  $\mathbf{T}_R(p)$ , is defined as the set of  $q \in \mathbb{R}^n$ separated from p by  $\partial \mathcal{F}$  in p + R.















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- A *U*-certificate at p is an EBF  $\rho$ , with:
  - ▶  $B \triangleq B_{\hat{x}(s)}(\varrho(s)) \subseteq \mathbf{T}_R(p)$ , and ▶  $\hat{x}(t) \in p + R$  for all  $0 \le t \le s$ .

for some  $\hat{x} \in \mathscr{A}_p$ ,  $s \ge 0$  and  $R \in \mathcal{U}$ .









### Topolo

### **Topological Transition Guarantee (TTG)**

► We can then define the *U*-MAUR:

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$$\mathbf{M}_{\mathcal{F}}^{\mathcal{U}}(p) \triangleq \sup \left( \varrho(0) : \frac{\varrho \text{ is a } \mathcal{U}}{\operatorname{certificate at } p} \right)$$

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▶ Since  $\mathbf{T}_{\mathbb{R}^n}(p) = \mathcal{F}$ , we have

$$\blacktriangleright \mathbf{M}_{\mathcal{F}}^{^{IBC}} \leq \mathbf{M}_{\mathcal{F}}^{\mathcal{U}} \leq \mathbf{M}_{\mathcal{F}}^{*}$$

















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 $\blacktriangleright \mathbf{M}_{\mathcal{F}}^{^{IBC}} \leq \mathbf{M}_{\mathcal{F}}^{\mathcal{U}} \leq \mathbf{M}_{\mathcal{F}}^{*}.$ 

- ▶ Hence a focus on computability of lower bounds  $f \leq \mathbf{M}_{\mathcal{F}}^{\mathcal{U}}$ , obtained by
  - restircting  $\mathcal{U}$ : strips, cylinders, cones
  - restricting *Q*: specific EBFs
  - restricting \(\partial \mathcal{F}\): collared spheres















▶ Jordan Curves [2] provide a "separation standard" in the plane.

- $\partial \mathcal{F}$  can be any continuous simple closed curve (SCC)
- $\blacktriangleright~\partial \mathcal{F}$  can be lower-approximated by polygonal SCCs













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- ▶  $\partial \mathcal{F}$  can be lower-approximated by polygonal SCCs
- ▶ For simple parametric  $\mathcal{U}$  (strips, cones), lower bounds on  $\mathbf{M}_{\mathcal{F}}^{\mathcal{U}}(p)$  are...
  - Piecewise regular for regular increasing  $\varrho$
  - Solutions to a constrained optimization problem

 $\rightsquigarrow$  Combinatorial/Topological aspects to be resolved

Computations are parallelizable.

 $\rightsquigarrow$  Target balls picked via edge-by-edge optimization















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- Higher-dimensional analog via collared spheres [3]:
  - An embedding  $\gamma : \mathbb{S}^{n-1} \times \{0\} \hookrightarrow \mathbb{R}^n$  is *collared*, if it extends to an embedding of  $\mathbb{S}^{n-1} \times [-1, 1] \hookrightarrow \mathbb{R}^n$ .
  - Generalized Schoenflies [4] implies: If γ : S<sup>n-1</sup> → S<sup>n</sup> is collared, then S<sup>n</sup> \ γ(S<sup>n-1</sup>) is the disjoint union of two open balls.













#### Future Directions: Reactive TTGs

Returning to Single-Agent Relay Tracking,

- Available plans: For all p,  $\mathscr{A}_p \equiv \mathbb{S}^{n-1}$ ;
- Desired TTG:  $f \in \mathcal{M}(\mathcal{F})$  with  $f \leq \mathbf{M}_{\mathcal{F}}^{\mathcal{U}}$ ;
- Task:  $X_d$ , parametric or subdivided.















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- ► Observe: TTG direction x̂<sub>p</sub> is well-defined for all p ∈ 𝓕<sup>C</sup> except for a null set.
  - f increases in the direction  $\hat{x}_p$ ;
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- ▶ Problem 3: Compute approximations  $f \leq \mathbf{M}_{\mathcal{F}}^{\mathcal{U}}$  in closed form.
- ► Problem 4: Determine the relationship between ∇f and TTG plans.
  - Characterizing TTG plans *locally*?
  - Value tradeoffs à-la [5, 6] between tracking X<sub>d</sub> and detours into F?

















#### Future directions: Learning TTGs

In the absence of complete knowledge of  $\mathcal{F}.\,.\,$ 

▶ Instead of learning  $\mathcal{F}$ , learn  $\mathbf{M}_{\mathcal{F}}^{\mathcal{U}}$  (or possibly  $\nabla \mathbf{M}_{\mathcal{F}}^{\mathcal{U}}$ ), as a model.









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• With  $f_t$  approximating  $\mathbf{M}_{\mathcal{F}}^{\mathcal{U}}$  at time t,

- $f_t$  corresponding to a polygonal approximation  $\mathcal{F}_t$  of  $\mathcal{F}$ ;
- Monotonicity,  $t \leq s \implies f_t \leq f_s$ , ensure a valid TTG for all time.
- ▶ Challenge: Maintain a connected model of *F*, or—
- ▶ Challenge: Extend methods over disconnected *F*, and—
- **Challenge:** Directed *exploration* of  $\partial \mathcal{F}$  may be required.











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• Closed parametric form  $\mathbf{M}_{\mathcal{F}}^{\mathcal{U}} = H(c^*)$ ,  $c^* \in \mathbb{R}^N$  means...

- $f_t$  takes the form of  $H(c_t)$ ,  $c_t \in \mathbb{R}^N$ ;
- Think of N as a bound on the complexity of  $\mathcal{F}_t$ ;
- We could attempt learning  $c^*$ ;
- A natural loss function is, e.g.,  $\mathcal{L}(c) \triangleq \left\| \mathbf{M}_{\mathcal{F}}^{\mathcal{U}} H(c) \right\|_{2}^{2}$ ;
- ▶ Question: Does GD over *L* respond well to the above challenges?
- Question: Could exploration be guided so that GD responds adequately to these challenges?













#### THANK YOU FOR YOUR ATTENTION!













#### References



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