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# Convex Optimal Control Synthesis Under Safety Constraints

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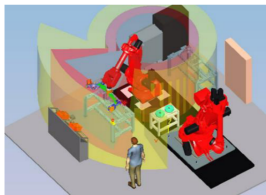
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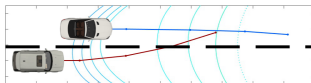
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# Motivation

- Safety in human-robot interaction<sup>1</sup>:



- Safety in autonomous car planning<sup>2</sup>:



- Unsafe set: the collection of states that we do not want our system to be at.

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1: G. Michalos, S. Makris, P. Tsarouchi, T. Guasch, D. Kontovrakis, G. Chryssolouris "Design Considerations for Safe Human-robot Collaborative Workplaces", Procedia CIRP 2015

2: B. Landry, M. Chen, S. Hemley and M. Pavone "Reach-Avoid Problems via Sum-of-Squares Optimization and Dynamic Programming", ArXiv 2018

# Task

Reach-avoid problem for dynamical system

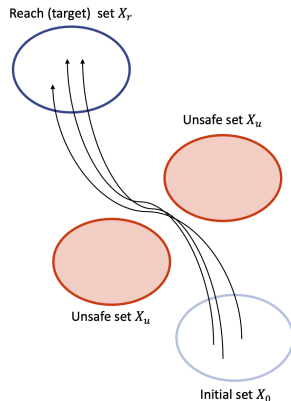
$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}) + \mathbf{g}(\mathbf{x})\mathbf{u}$$

Initial sets, unsafe sets, and reach sets

$$\mathbf{X}_0, \mathbf{X}_u, \mathbf{X}_r$$

Optimal control problem, cost function

$$\mathbb{E}[\int_0^\infty (q(\mathbf{x}(t)) + \mathbf{u}^\top(t)\mathbf{R}\mathbf{u}(t)) dt]$$



# Koopman and Perron-Frobenius Operator

- Koopman (composition) operator

$$[\mathbb{K}_t \varphi](\mathbf{x}) = \varphi(\mathbf{s}_t(\mathbf{x})), \quad \forall \varphi$$

- Perron-Frobenius (transfer) operator

$$\int_{\mathbf{s}_{-t}(A)} \psi(\mathbf{x}) d\mathbf{x} = \int_A \mathbb{P}_t[\psi](\mathbf{x}) d\mathbf{x}, \quad \forall A \subset \mathbf{X}, \quad \forall \psi$$

Duality:  $\forall \varphi, \psi :$

$$\langle [\mathbb{K}_t \varphi](\mathbf{x}), \psi(\mathbf{x}) \rangle = \langle \mathbb{P}_t \psi(\mathbf{x}), \varphi(\mathbf{x}) \rangle$$

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$$\mathbb{K}_t : \mathcal{L}_\infty(\mathbf{X}) \rightarrow \mathcal{L}_\infty(\mathbf{X}), \mathbb{P}_t : \mathcal{L}_1(\mathbf{X}) \rightarrow \mathcal{L}_1(\mathbf{X}).$$

$\mathbf{s}_t(\mathbf{x})$  denotes the solution of  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x})$  starting from  $\mathbf{x}$

# Intuition and occupation measure

- Evolution of distribution  $h_0(\mathbf{x})$  under dynamics

$$\int_{\phi_t(A)} h_t(\mathbf{x}) d\mathbf{x} = \int_A h_0(\mathbf{x}) d\mathbf{x} \triangleq \int_{\phi_t(A)} \mathbb{P}_t[\psi](\mathbf{x}) d\mathbf{x}, \forall A$$

- Occupation measure  $\rho(\mathbf{x})$  - accumulated 'mass'

$$\rho(\mathbf{x}) := \int_0^\infty h_t(\mathbf{x}) dt$$

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$\phi_t(A)$  represents the evolution of set  $A$  under dynamics  $\dot{\mathbf{x}} = \mathbf{f} + \mathbf{g}\mathbf{u}$

# Problem reformulation

- Cost function reformulated:

$$J(\mu_0) = \int_{\mathbf{x}} (q + \mathbf{k}^\top \mathbf{R} \mathbf{k})(\mathbf{x}) \rho(\mathbf{x}) d\mathbf{x}$$

- Initial distribution:

$$\nabla \cdot (\rho(\mathbf{f} + \mathbf{g}\mathbf{u})) = h_0$$

- Safety constraint: zero occupation in  $\mathbf{X}_u$

$$\int_{\mathbf{X}_u} \rho(\mathbf{x}) d\mathbf{x} = \int_{\mathbf{x}} \mathbb{1}_{\mathbf{X}_u}(\mathbf{x}) \int_0^\infty h_t(\mathbf{x}) dt d\mathbf{x} = 0$$

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Assume feedback control  $\mathbf{u} = \mathbf{k}(\mathbf{x})$ .

$q$  is the state cost, and  $\mathbf{R}$  is the penalty control magnitude,  $h_0$  is the initial distribution

# Convex formulation

- Is this a convex optimization problem?

$$\nabla \cdot (\rho(\mathbf{f} + \mathbf{g}\mathbf{u})) = h_0$$

is bi-linearity in  $(\rho, \mathbf{u})$ .

- Define  $\bar{\rho} = \rho\mathbf{u}$ , then convex in  $(\rho, \bar{\rho})$ :

$$\begin{aligned} \inf_{\rho, \bar{\rho}} \quad & \int_{\mathbf{x}} q(\mathbf{x})\rho(\mathbf{x}) + \frac{\bar{\rho}(\mathbf{x})^\top \mathbf{R}\bar{\rho}(\mathbf{x})}{\rho(\mathbf{x})} d\mathbf{x} \\ \text{s.t.} \quad & \nabla \cdot (\mathbf{f}\rho + \mathbf{g}\bar{\rho}) = h_0 \geq 0 \\ & \int_{X_u} \rho(\mathbf{x}) d\mathbf{x} = 0 \end{aligned}$$

## Solving: fixed Lagrangian multiplier

- Choose and fix a large Lagrangian multiplier  $\bar{\lambda}$  and solve the problem

$$\inf_{\rho, \bar{\rho}} \int_{\mathbf{x}} \left( q + \bar{\lambda} \mathbf{1}_{x_u} \right) \rho + \frac{\bar{\rho}^\top \mathbf{R} \bar{\rho}}{\rho} d\mathbf{x}$$
$$\text{s.t. } \nabla \cdot (\mathbf{f} \rho + \mathbf{g} \bar{\rho}) \geq 0$$

- How to solve?

Using polynomials and SOS.

- How to deal with the term  $\frac{\bar{\rho}^\top \mathbf{R} \bar{\rho}}{\rho}$ ?

Construct an upper bound  $w(\mathbf{x})$  and minimize  $w$ .



# Polynomial parameterization

- Parameterize

$$\rho = \frac{a}{b^\alpha}, \bar{\rho} = \frac{c}{b^\alpha}$$

$a, c$  are unknown polynomial variables,  $b(\mathbf{x}) > 0$

- Construct  $w(\mathbf{x})$ . Define

$$\mathbf{M} \triangleq \begin{bmatrix} w & \mathbf{c}^T \\ \mathbf{c} & a\mathbf{R}^{-1} \end{bmatrix}$$

$$\mathbf{M} \succeq 0 \Leftrightarrow w \geq 0, w \geq \frac{\mathbf{c}^T \mathbf{R} \mathbf{c}}{a} = \mathbf{u}^T \mathbf{R} \mathbf{u}$$

# Sum of squares (SOS)

- Polynomial  $p(\mathbf{x})$  is SOS:

$$p(\mathbf{x}) \in \Sigma$$

- SDP formulation

$$\begin{aligned} \min_{C_a, C_c, C_w} \quad & C_a^T (\mathbf{d}_1 + \bar{\lambda} \mathbf{d}_3) + C_w^T \mathbf{d}_2 \\ \text{s.t.} \quad & a \in \Sigma[\mathbf{x}], b^{\alpha+1} h_0 \in \Sigma[\mathbf{x}] \\ & \begin{bmatrix} w & \mathbf{c}^T \\ \mathbf{c} & a\mathbf{R}^{-1} \end{bmatrix} \succeq 0, \end{aligned}$$

which is a standard SDP.

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$C_a, C_c, C_w$  are corresponding polynomial coefficients in a common monomial basis.

# Safety verification

- Safety: For almost all  $\mathbf{x}_0 \in X_0$ , the solution  $\mathbf{s}_t(\mathbf{x})$  starting at  $\mathbf{x}_0$  satisfies  $\mathbf{s}_T(\mathbf{x}) \in X_r$ , for some  $T > 0$ ,  $\mathbf{s}_t(\mathbf{x}) \notin X_u$ , and  $\mathbf{s}_t(\mathbf{x}) \in \mathbf{X}$  for all  $t \in [0, T]$
- Criteria<sup>1</sup>: System safe if  $\exists$  polynomial  $\rho_s$ :

$$\begin{aligned}\rho_s(\mathbf{x}) &\geq 0, \forall \mathbf{x} \in \mathbf{X}_0 \\ \nabla \cdot (\rho_s(\mathbf{f} + \mathbf{g}\mathbf{u}))(\mathbf{x}) &> 0, \forall \mathbf{x} \in cl(\mathbf{X} \setminus \mathbf{X}_r) \\ \rho_s &< 0, \forall \mathbf{x} \in cl(\partial \mathbf{X} \setminus \partial \mathbf{X}_r) \cup \mathbf{X}_u\end{aligned}$$

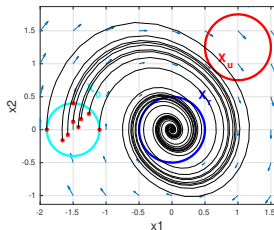
- Find positive polynomial  $\rho_s$  inside compact sets defined by polynomials (semi-algebraic sets): an SDP feasibility problem.

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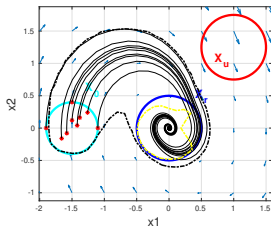
1. S. Prajna, and A. Rantzer "Convex programs for temporal verification of nonlinear dynamical systems." SIAM Journal on Control and Optimization, vol.46, no.3, pp.999–1021, 2007

# Simulation results

Dynamical system: Van Der Pol

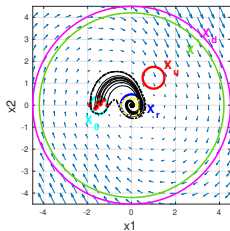


(a) Zoomed view,  $\bar{\lambda} = 0$

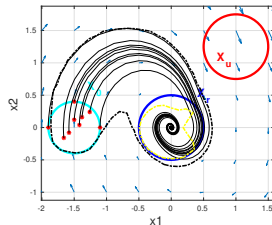


(b) Zoomed view,  $\bar{\lambda} = 1e^5$ .

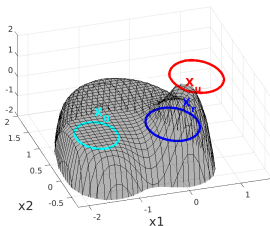
# Simulation results



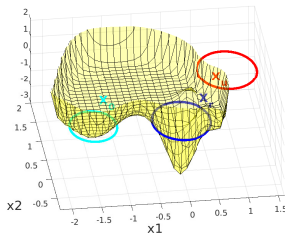
(a) Phase portrait,  $\bar{\lambda} = 1e^5$ .



(b) Zoomed view.



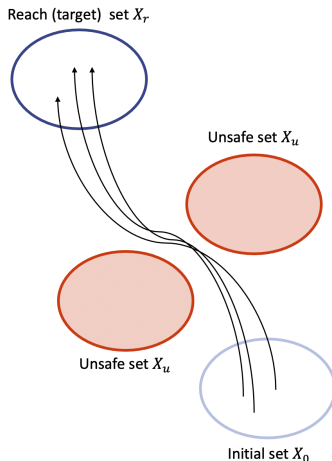
(c) Surface of  $\rho_s(\mathbf{x})$



(d) Surface of  $\nabla \cdot (\rho_s(\mathbf{f} + \mathbf{g}\mathbf{u}))$

# Takeaway

- Introduced the notion of occupation measure
- Reformulated the optimal control problem under safety constraints
- Verified the solution in terms of safety constraints





# Thank you!

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