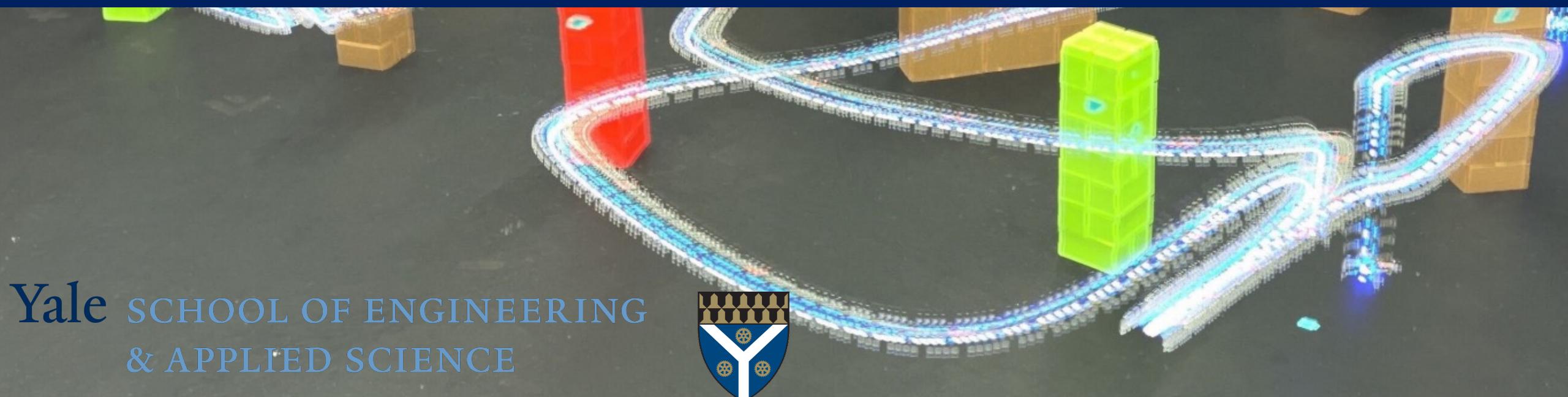


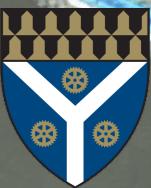


Robustness and Optimality in Ergodic Control

Ian Abraham



YALE SCHOOL OF ENGINEERING
& APPLIED SCIENCE



Acknowledgements



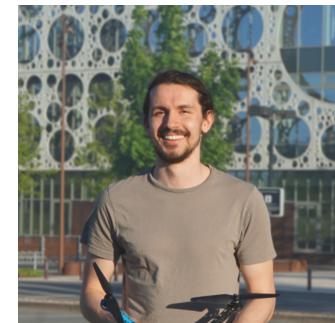
Dayi Dong



Henry Berger



Elena Wittemyer



Adam Seewald



Cameron Lerch

[Berger In Prep] Reachability Analysis for Ergodic Coverage Problems

[Seewald ICRA 24'] Energy-Aware Ergodic Search: Continuous Exploration for Multi-Agent Systems with Battery Constraints

[Wittemyer IROS 23'] Bi-Level Image-Guided Ergodic Exploration with Application to Planetary Rovers

[Lerch ICRA 23'] Safety-critical ergodic exploration in cluttered environments via control barrier functions

[Dong RSS 23'] Time Optimal Ergodic Search

[Abraham TASE 21'] An Ergodic Measure for Active Learning from Equilibrium.

[Abraham WAFR 20'] Active Area Coverage from Equilibrium

[Abraham RSS 18'] Data-Driven Measurement Models for Active Localization in Sparse Environments

[Abraham RAL 17'] Ergodic Exploration using Binary Sensing for Nonparametric Shape Estimation



Motivation: Optimal and Robust Exploration

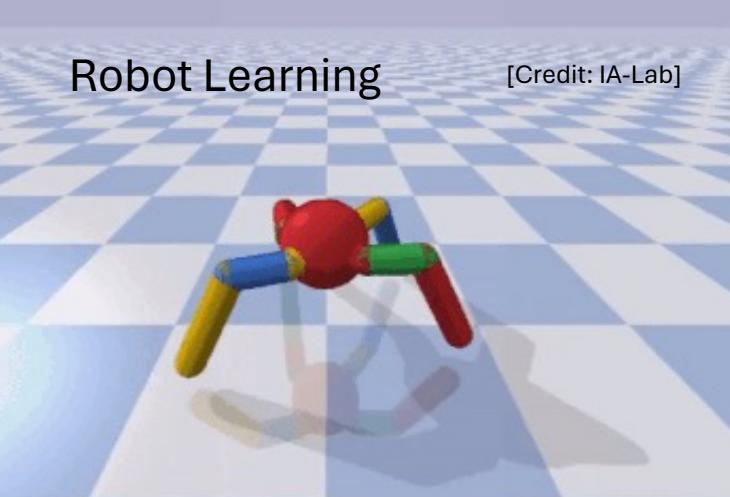
Mapping



Search and rescue

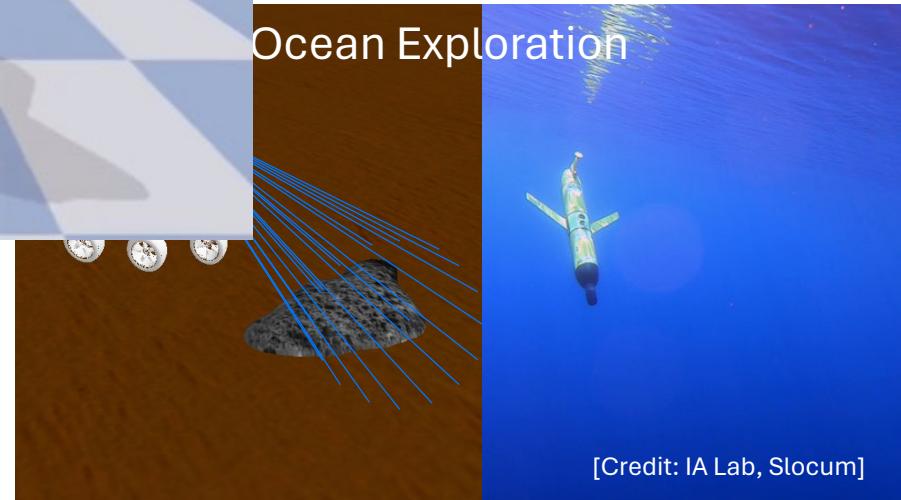


Robot Learning



Environmental monitoring

Ocean Exploration



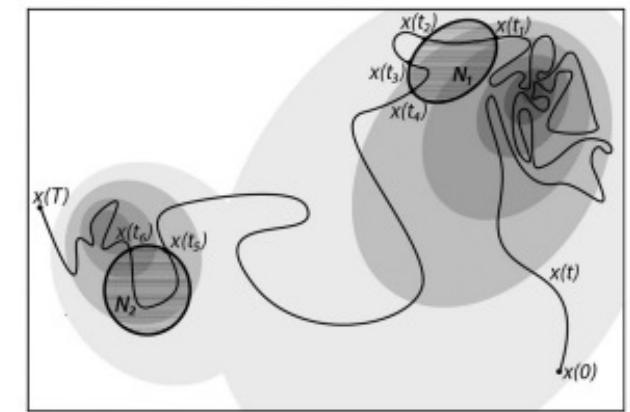
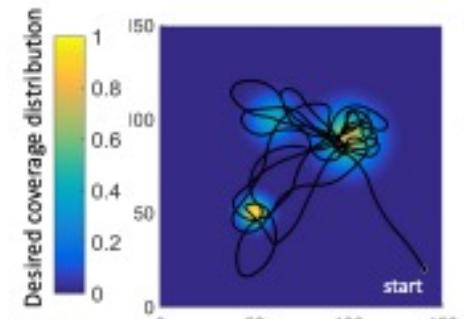
- Safety Guarantees in Ergodic Control
 - Set-Invariance with Control Barrier Functions
 - Hamilton Jacobi Isaacs Reachability
- Time-Optimal Ergodic Control
- Energy-Optimal Ergodic Control

Introduction to Ergodicity

Ergodicity

- A trajectory is said to be ergodic if, **on average**, it spends time in regions **proportional to the utility of exploring** said area
- For $t_f \rightarrow \infty$, ergodic trajectories guarantee complete coverage

$$\lim_{t_f \rightarrow \infty} \frac{1}{t_f} \int_0^{t_f} \phi(g \circ x(t)) dt = \underbrace{\int_{\mathcal{W}} \phi(w) \mu(w) dw}_{\text{---}}$$



[1] Ayvali, Elif, Hadi Salman, and Howie Choset. "Ergodic coverage in constrained environments using stochastic trajectory optimization." 2017 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS). IEEE, 2017.

[2] Miller, Lauren M., et al. "Ergodic exploration of distributed information." IEEE Transactions on Robotics 32.1 (2015): 36-52.

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Safety Guarantees in Ergodic Control (CBF)

Problem Formulation

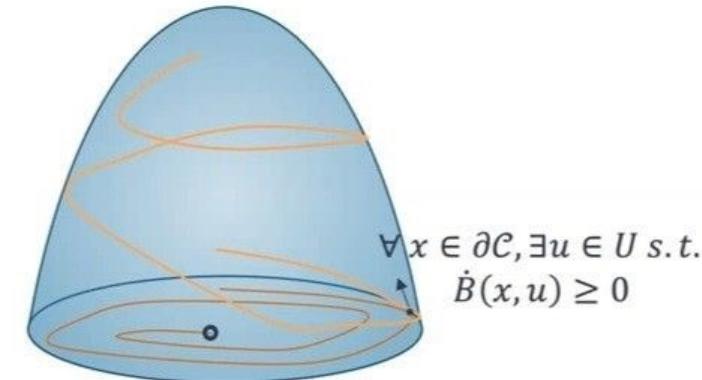
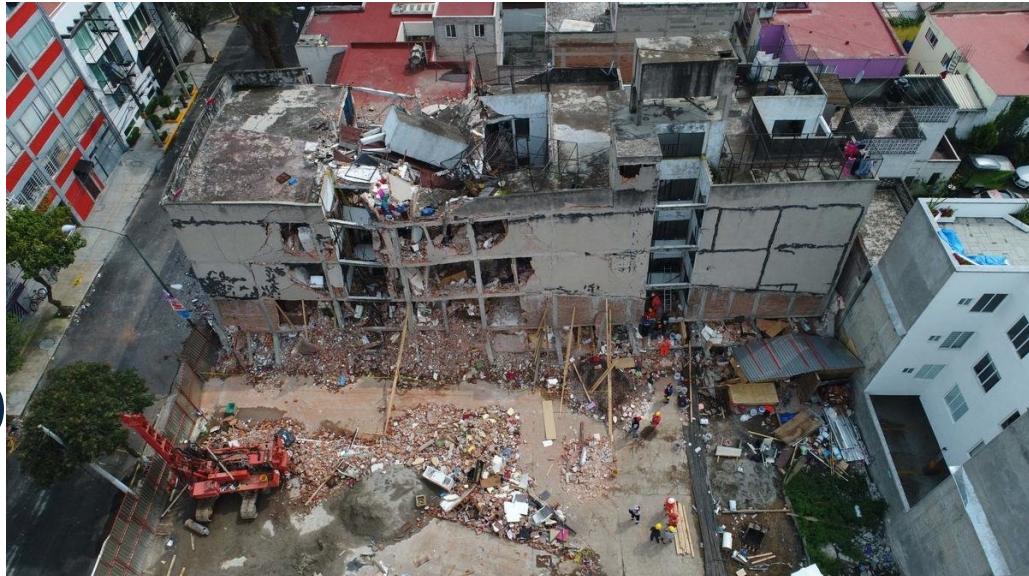
- Given an area with many obstacles, guarantee safe navigation without loss of coverage quality
- Safety defined as guaranteeing *set-invariance* over safe-set (using a Control Barrier Function¹)

$$\mathcal{C} = \{x \in \mathcal{X} \mid B(x) \geq 0\}$$

$$\min_u \|u - u_{\text{nom}}\|$$

$$\text{s.t. } \dot{B}(x, u) = \nabla B(x) \cdot f(x, u) \geq -\gamma(B(x)) \quad \forall u \in \mathcal{U}$$

$$\text{and } \dot{x} = f(x, u)$$



Safety Guarantees in Ergodic Control (CBF)

Problem Formulation

Issue:

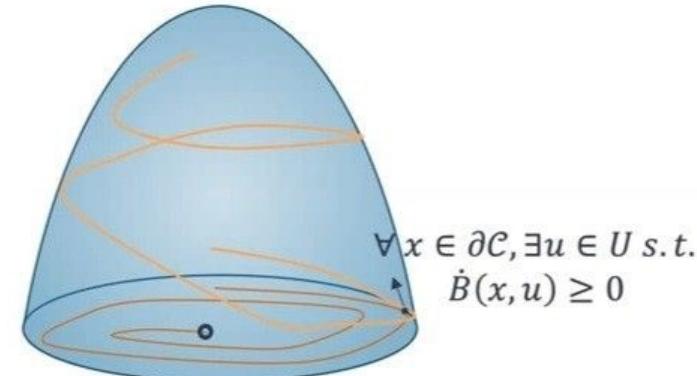
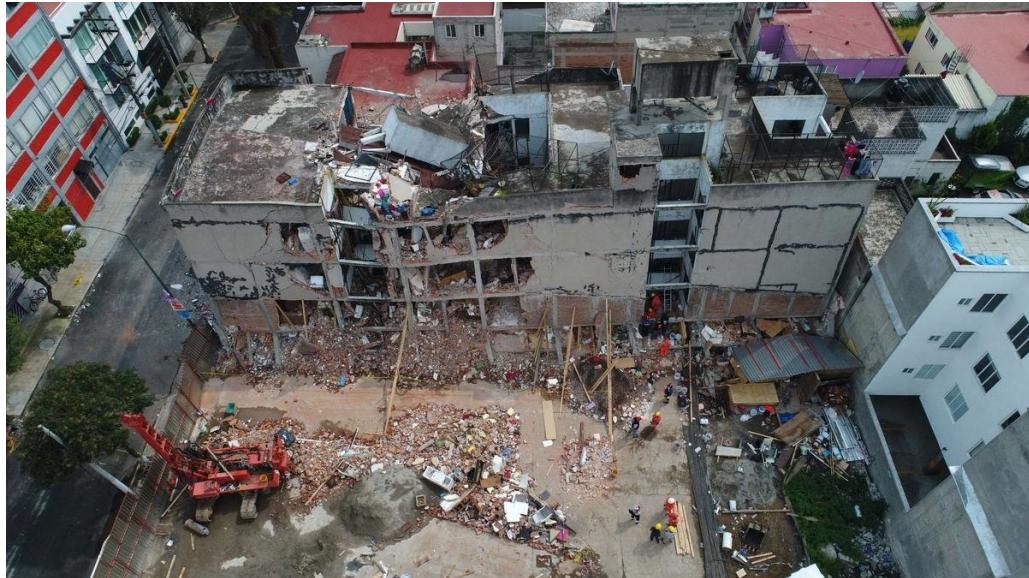
- Safety “filter” overrides future actions that inhibit effective exploration

$$\mathcal{C} = \{x \in \mathcal{X} \mid B(x) \geq 0\}$$

$$\min_u \|u - u_{\text{nom}}\|$$

$$\text{s.t. } \dot{B}(x, u) = \nabla B(x) \cdot f(x, u) \geq -\gamma(B(x)) \quad \forall u \in \mathcal{U}$$

$$\text{and } \dot{x} = f(x, u)$$



Safety Guarantees in Ergodic Control (CBF)

Solution

- Jointly solve for coverage + safety
 - We solve problem over **discrete** trajectories

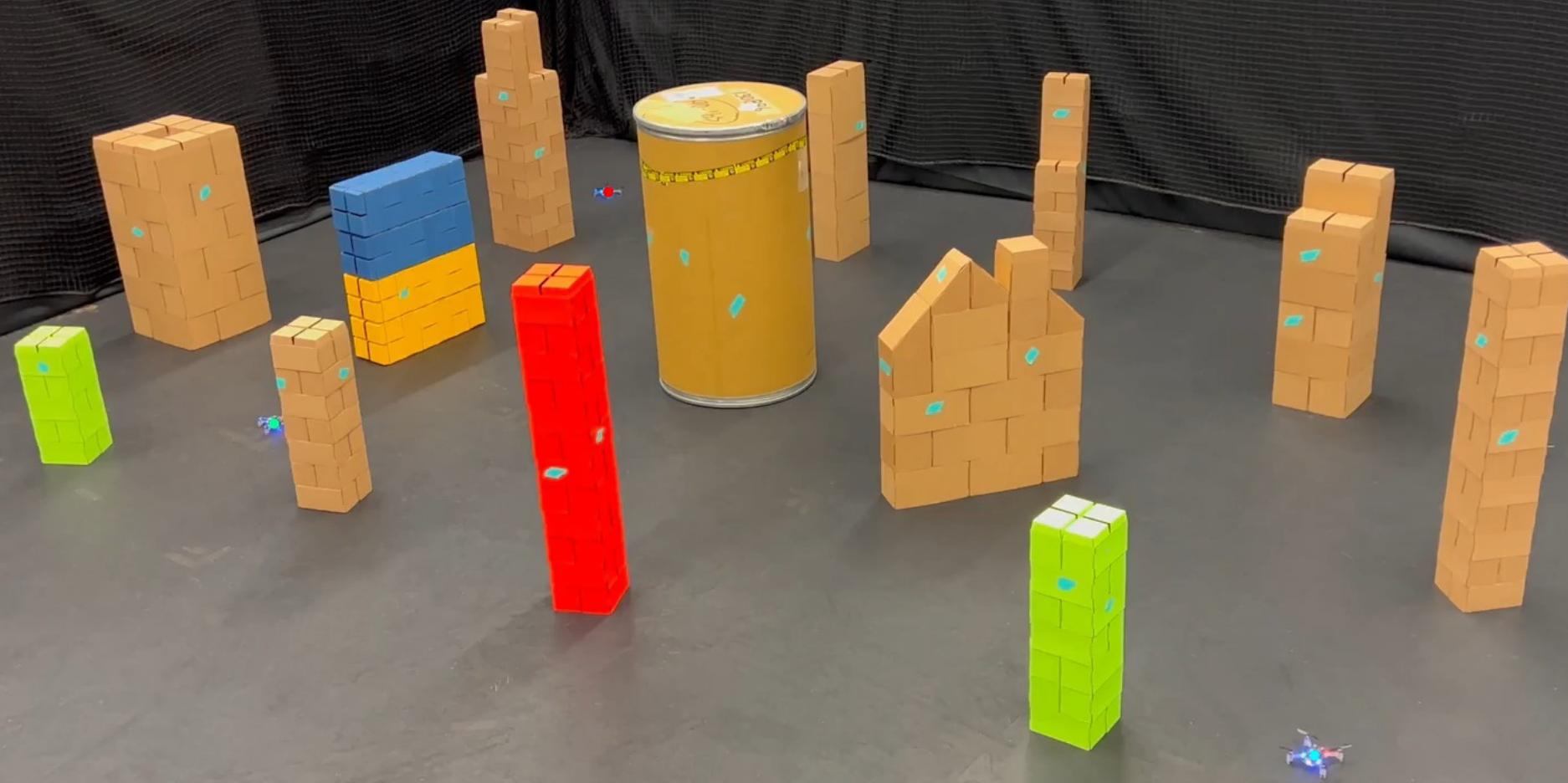
$$\min_{\mathbf{x}, \mathbf{u}} \mathcal{E}(\mathbf{x}, \phi) + \sum_0^{T-1} \mathbf{u}_t^\top R \mathbf{u}_t$$

s.t.
$$\begin{cases} x_{t+1} = f(x_t, u_t), x_t \in \mathcal{X}, u_t \in \mathcal{U} \\ x_0 = \bar{x}_0, x_{T-1} = \bar{x}_f, g(x) \in \mathcal{W} \\ \underline{\Delta B(x_t, u_t) \geq -\gamma B(x_t)} \end{cases}$$

$$\begin{aligned} \mathcal{E}(\mathbf{x}, \phi) &= \sum_{k \in \mathbb{N}^v} \Lambda_k (c_k(\mathbf{x}) - \phi_k)^2 \\ &= \sum_{k \in \mathbb{N}^v} \Lambda_k \left(\frac{1}{T} \sum_{t=0}^{T-1} F_k(g(x_t)) - \underline{\int_{\mathcal{W}} \phi(w) F_k(w) dw} \right)^2 \end{aligned}$$

Safety Guarantees

- Satisfied through barrier constraint
- Ergodic metric local minima provide additional flexibility
- Generic to various notions of safety



Multi-drone ergodic search with drone-drone and drone-obstacle avoidance using¹⁰
QRE

- Safety Guarantees in Ergodic Control
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Reachability in Ergodic Control

Problem Formulation

- Can we guarantee coverage fidelity under dynamic disturbances?
- Problem can be thought of as a reachability problem on achieving levels of ergodicity

Hamilton-Jacobi-Isaacs Reachability Problem

Find $V : \mathcal{X} \times [t_0, t_f] \rightarrow \mathbb{R}$ that satisfies

$$\frac{\partial V}{\partial t} = - \min_{u(t)} \max_{d(t)} \left\{ H(x, u, d, t) \right\}$$

$$\text{and } V(x, t_f) = m(x),$$

where the Hamiltonian H is defined as

$$H(x, u, d, t) \triangleq \ell(x, u, d, t) + \left(\frac{\partial V}{\partial x} \right)^\top f(x, u, d, t).$$

$$\begin{aligned} & \min_{x(t), u(t)} \left\{ \mathcal{E}(x(t), \phi) + \int_{t_0}^{t_f} u(t)^\top \mathbf{R} u(t) dt \right\} \\ & \text{s.t.} \quad \begin{cases} x \in \mathcal{X}, u \in \mathcal{U}, g(x) \in \mathcal{W} \\ x(t_0) = \bar{x}_0, x(t_f) = \bar{x}_f \\ \dot{x} = f(x, u) \\ h_1(x, u) \leq 0, h_2(x, u) = 0 \end{cases} \end{aligned}$$

Reachability in Ergodic Control

Extended Ergodic State

- We can define an extended state that converts problem into Bolza-form for HJI

Definition 1 *Extended ergodic state.* The ergodic metric can be equivalently expressed as

$$\begin{aligned}\mathcal{E}(x(t), \phi, t_f) &= \sum_{k \in \mathcal{K}^v} \Lambda_k (c_k(x(t), t_f) - \phi_k)^2 \\ &= \frac{1}{t_f^2} \|z(t_f)\|_{\Lambda}^2\end{aligned}$$

where $z(t_f) = [z_0, z_1, \dots, z_{|\mathcal{K}^v|}]^\top$ is the solution to

$$\dot{z}_k = F_k(g(x(t))) - \phi_k \tag{1}$$

with initial condition $z(t_0) = \mathbf{0}$, and $\Lambda = \text{diag}(\Lambda)$ is a diagonal matrix consisting of the weights $\Lambda = [\Lambda_0, \dots, \Lambda_{|\mathcal{K}^v|}]$.

Extended Ergodic Control Problem w/ Disturbance

Find $u : [t_0, t_f] \rightarrow \mathcal{U}$ that minimizes

$$\mathcal{J}_{\text{worst case}} = \max_d \left\{ \frac{1}{t_f^2} \|z(t_f)\|_{\Lambda}^2 + \int_{t_0}^{t_f} \ell(x(\tau), u(\tau)) d\tau \right\},$$

subject to

$$\begin{aligned}\dot{x}(t) &= f(x(t), u(t), d(t)) \\ \dot{z}_k(t) &= F_k(g(x(t))) - \phi_k, \forall k \in \mathcal{K} \\ \text{and } z_k(t_0) &= 0\end{aligned}$$

- [1] De La Torre, Gerardo, et al. "Ergodic exploration with stochastic sensor dynamics." *2016 American Control Conference (ACC)*. IEEE, 2016.
- [2] Mathew, George, and Igor Mezić. "Metrics for ergodicity and design of ergodic dynamics for multi-agent systems." *Physica D: Nonlinear Phenomena* 240.4-5 (2011): 432-442.

Reachability in Ergodic Control

Extended Ergodic State

- We can define an extended state that converts problem into Bolza-form for HJI

Extended Ergodic Control Problem w/ Disturbance

Find $u : [t_0, t_f] \rightarrow \mathcal{U}$ that minimizes

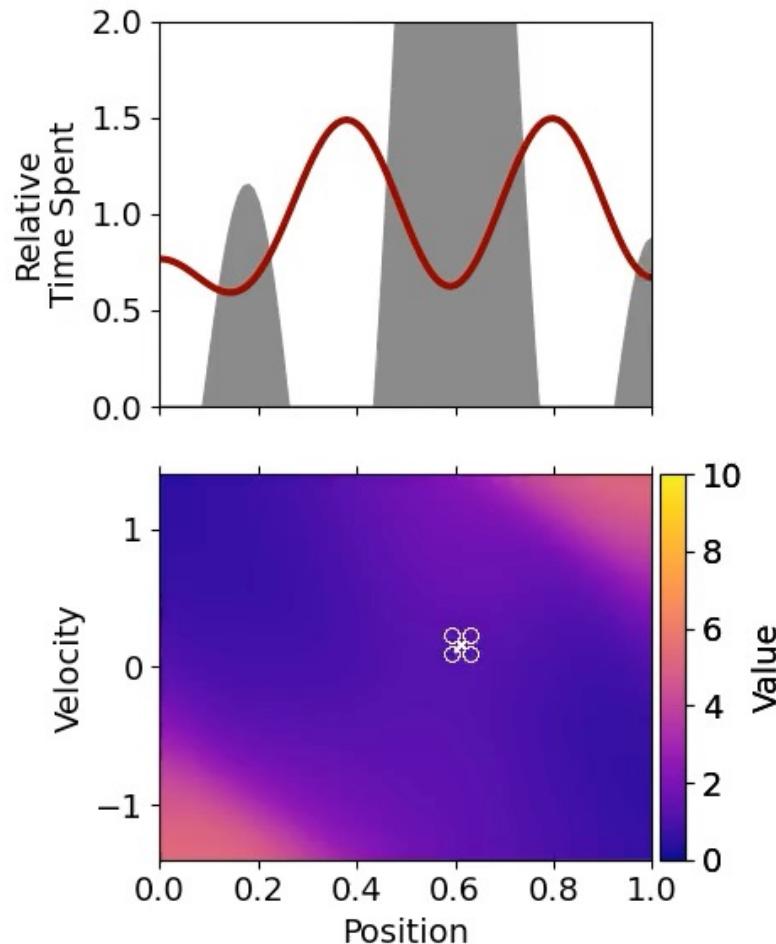
$$\mathcal{J}_{\text{worst case}} = \max_d \left\{ \frac{1}{t_f^2} \|z(t_f)\|_{\Lambda}^2 + \int_{t_0}^{t_f} \ell(x(\tau), u(\tau)) d\tau \right\}, \quad (1a)$$

subject to

$$\dot{x}(t) = f(x(t), u(t), d(t)) \quad (1b)$$

$$\dot{z}_k(t) = F_k(g(x(t))) - \phi_k, \forall k \in \mathcal{K} \quad (1c)$$

$$\text{and } z_k(t_0) = 0 \quad (1d)$$



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Time-Optimality in Ergodic Control

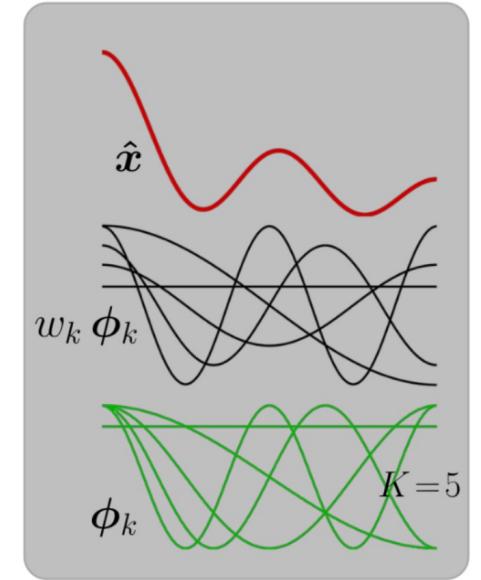
Problem Formulation

- What is the minimum time a robot needs to explore an area?
 - Ergodic metric based on average time trajectory visits area
 - Metric calculated using Fourier spectral decomposition
 - promotes uniform coverage based on utility measure

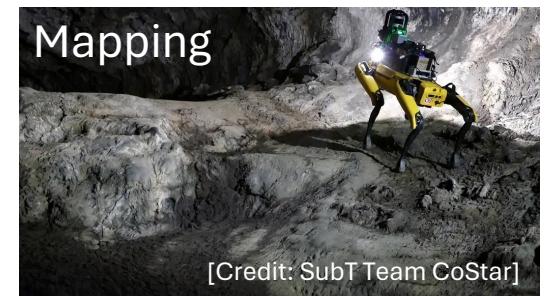
$$\begin{aligned}\mathcal{E}(x(t), \phi) &= \sum_{k \in \mathcal{K}^v} \Lambda_k (c_k - \phi_k)^2 \\ &= \sum_{k \in \mathcal{K}^v} \Lambda_k \underbrace{\left(\frac{1}{t_f} \int_{t_0}^{t_f} F_k(g(x(t))) dt - \int_{\mathcal{W}} \phi(w) F_k(w) dw \right)^2}_{\text{Mapping}}\end{aligned}$$

- Ergodic metric is inversely proportional to time and can act as constraint to bound time

[1] Calinon, Sylvain. "Mixture models for the analysis, edition, and synthesis of continuous time series." *Mixture Models and Applications* (2020): 39-57.
[2] Dong, Dayi, Henry Berger, and Ian Abraham. "Time Optimal Ergodic Search." *Robotics: Science and Systems* (2023).



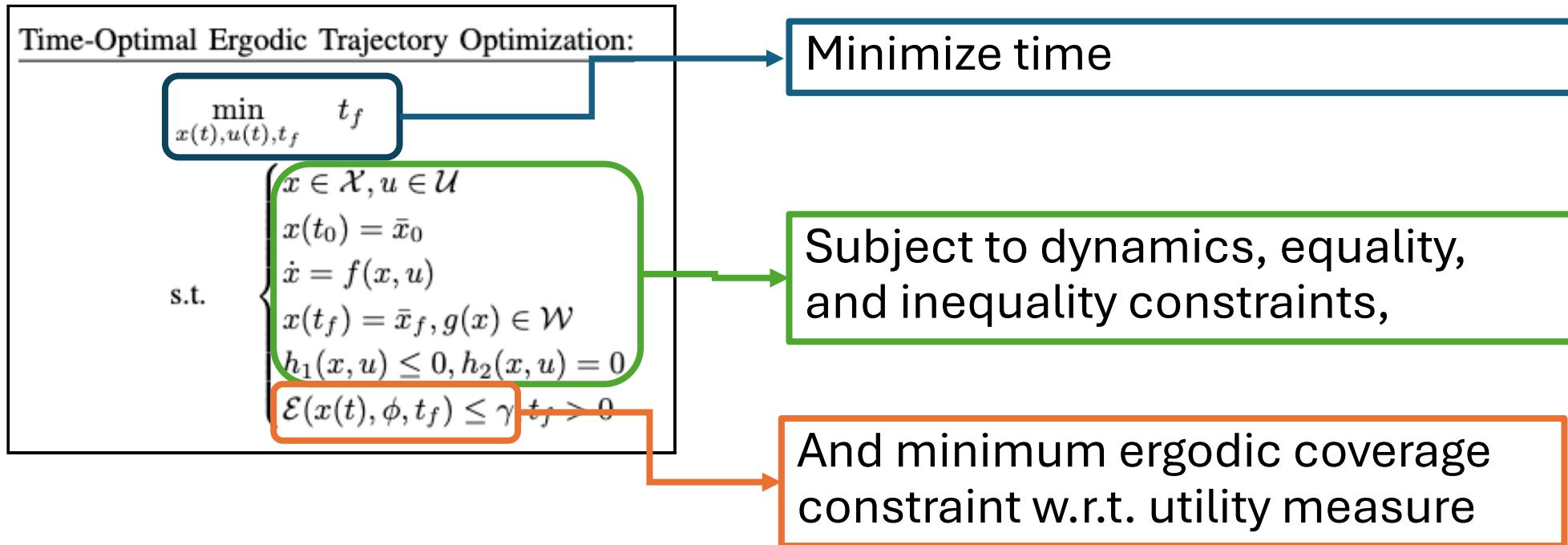
[1] Calinon, 2020



Time-Optimality in Ergodic Control

Problem Formulation

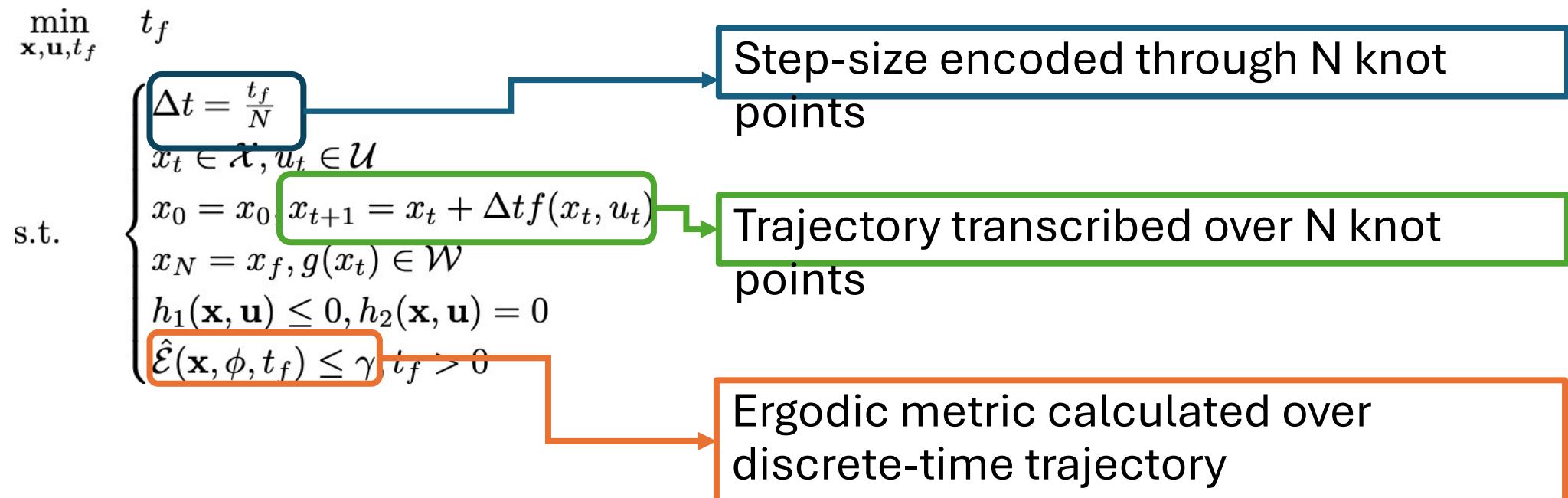
- We can rewrite ergodic trajectory optimization as a minimum-time problem with the ergodic metric as a constraint!



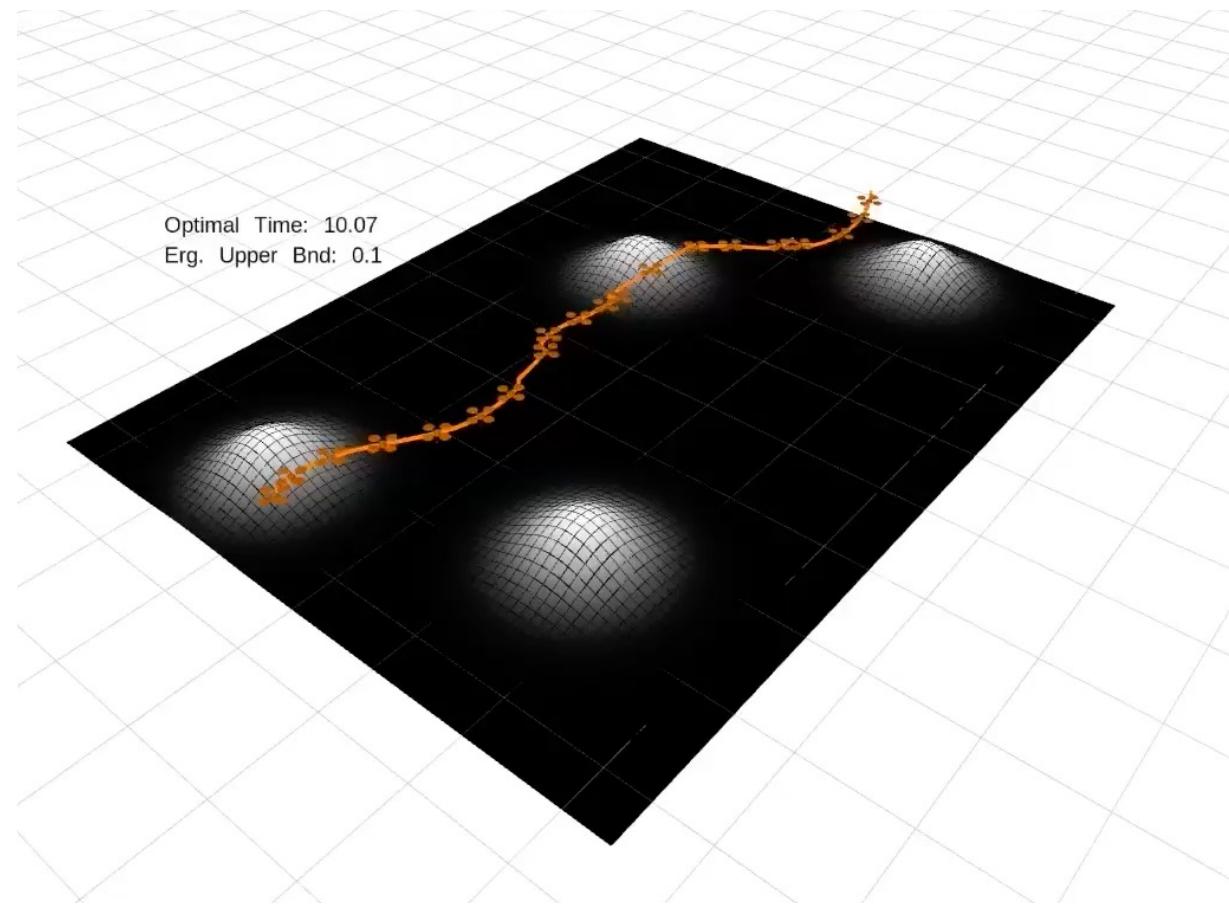
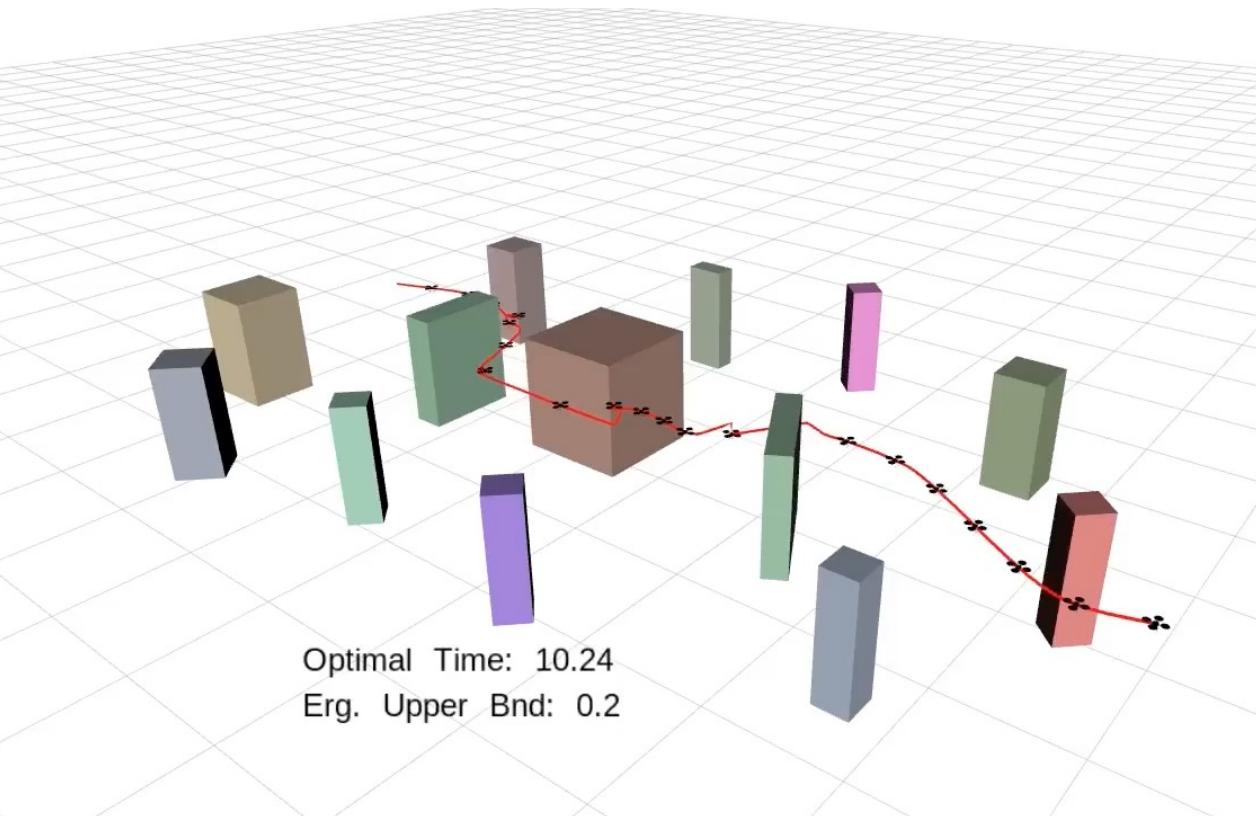
Time-Optimality in Ergodic Control

Numerical Solutions

- Locally optimal conditions proven to exist (based on Maximum Principle)
 - Computationally hard to compute
- Direct transcription can be used to compute solutions (based on KKT conditions)



Example Time-Optimal Ergodic Trajectories



Experimental Drone Results: Time-Optimal Uniform Coverage in Cluttered Environment

Coarse Minimum-Time Search

Erg. Upper Bnd: 0.1
Elapsed Time: 14.08 s
Optimized Time: 13.56 s

