

Multiple Pursuer-Based Intercept via Forward Stochastic Reachability

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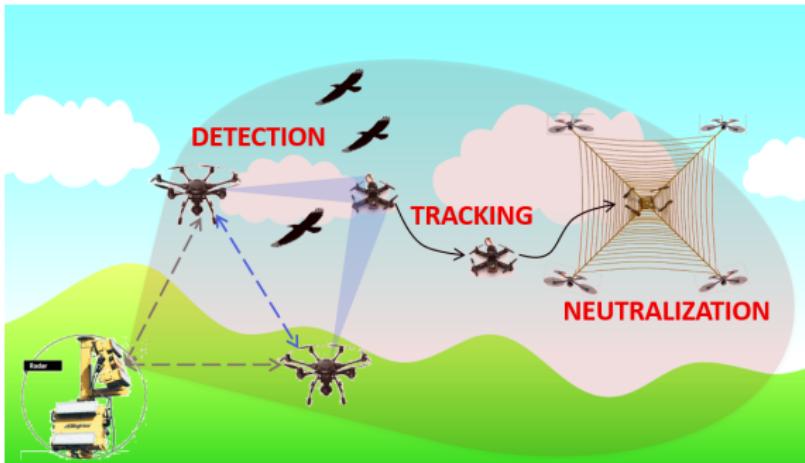
*Sandia National Laboratories



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Motivation: Aerial Suppression of Airborne Platform



- ▶ Protect assets from an intruder UAV
 - ▶ Modeled as a (non-reactive) stochastically moving goal/threat
- ▶ Pursue the intruder with UAV teams
 - ▶ Stochastic reachability-based pursuit

Maximize capture probability via a team of pursuer UAVs

Related work

- ▶ **Multi-agent differential game-based pursuit**
 - ▶ Vidal, Shakernia, Kim, Shim, & Sastry (2002); Tomlin, Mitchell, Bayen, & Oishi(2003); Mitchell, Bayen, & Tomlin (2005); C. F. Chung, T. Furukawa, & A. H. Goktogan (2006); C. F. Chung, & A. H. Goktogan (2008); Huang, Zhang, Ding, Stipanović, & Tomlin (2011);
- ▶ **Stochastic reachability**
 - ▶ Abate, Prandini, Lygeros, & Sastry (2008); G. Hollinger, S. Singh, J. Djugash, & A. Kehagias (2009); Summers, & Lygeros (2011); Kariotoglou, Raimondo, Summers, & Lygeros (2011); Lesser, Oishi, & Erwin (2013); Vinod, HomChaudhuri, & Oishi (2017); N. Malone, K. Lesser, M. Oishi, & L. Tapia (2017);

System description

- ▶ Linearized UAV dynamics ($n = 12$)

- ▶ Based on Asctec Hummingbird

- ▶ Goal UAV G

$$\dot{\mathbf{x}}_G[t+1] = A\mathbf{x}_G[t] + BK(\mathbf{x}_G[t] - \bar{\mathbf{x}}_a) + B\mathbf{w}[t]$$



- ▶ LQR to an asset location $\bar{\mathbf{x}}_a$

- ▶ Model uncertainty via $\mathbf{w} \sim \mathcal{N}(\bar{\mu}_{\mathbf{w}}, \Sigma_{\mathbf{w}})$

- ▶ Uncontrolled stochastic dynamics

- ▶ N pursuers UAV P_i

$$\dot{\mathbf{x}}_{P_i}[t+1] = A\bar{\mathbf{x}}_{P_i}[t] + B\bar{\mathbf{u}}_{P_i}[t]$$

- ▶ Bounded control authority $\bar{\mathbf{u}}_{P_i} \in \mathcal{U}_P$

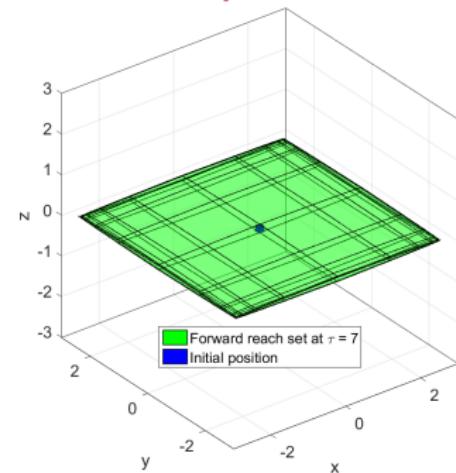
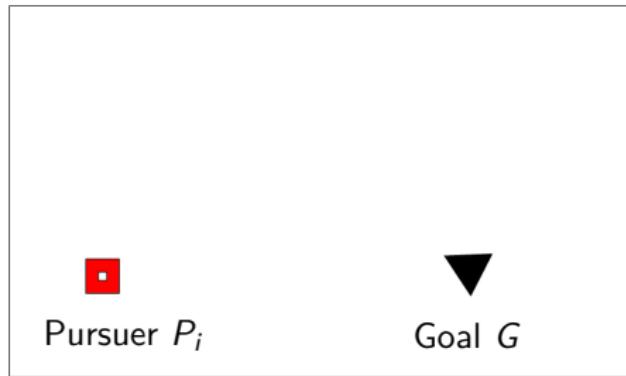
- ▶ Controlled deterministic dynamics

- ▶ Capture set for each P_i

- ▶ Known initial states for goal and pursuer UAVs

Mahony, Kumar, & Corke, IEEE Rob. Auto. Mag., 2012

Capture of stochastic target via convex optimization

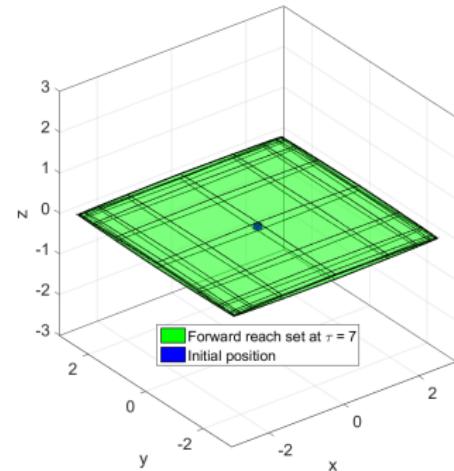
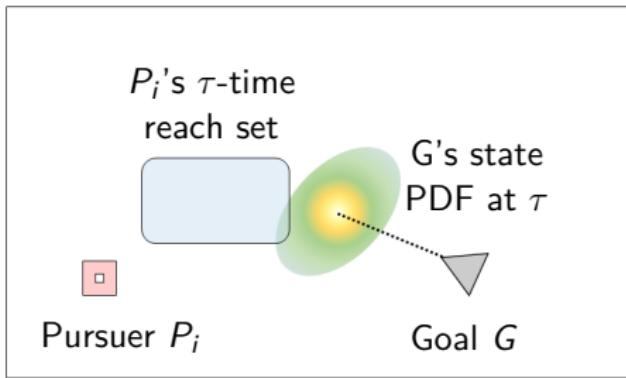


Single Pursuer Problem :

$$\left\{ \begin{array}{l} \text{maximize}_{\tau, \bar{x}_{P_i}[\tau]} \\ \text{subject to} \end{array} \right. \begin{array}{l} \text{CatchPr}_{\bar{x}_{P_i}}(\tau, \bar{x}_{P_i}[\tau]; \bar{x}_0) \\ \tau \in \mathbb{Z}_{[1, N]} \\ \bar{x}_{P_i}[\tau] \in \text{Reach}_{P_i}(\tau; \bar{x}_{P_i}[0]) \end{array}$$

Vinod, HomChaudhuri, and Oishi, HSCC 2017

Capture of stochastic target via convex optimization

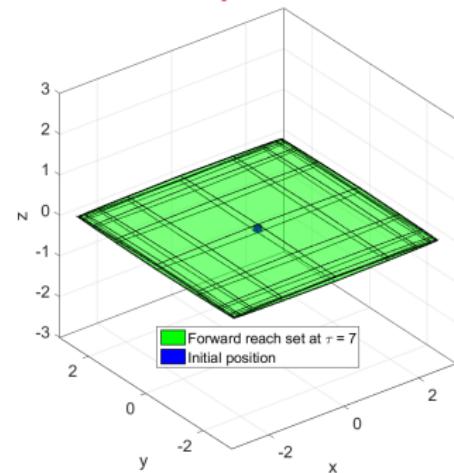
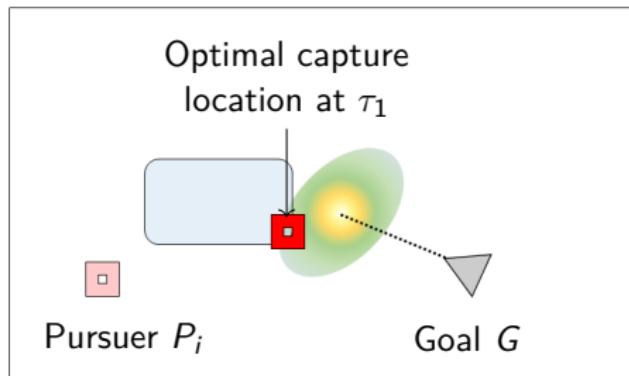


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Vinod, HomChaudhuri, and Oishi, HSCC 2017

Capture of stochastic target via convex optimization



Single Pursuer Problem :

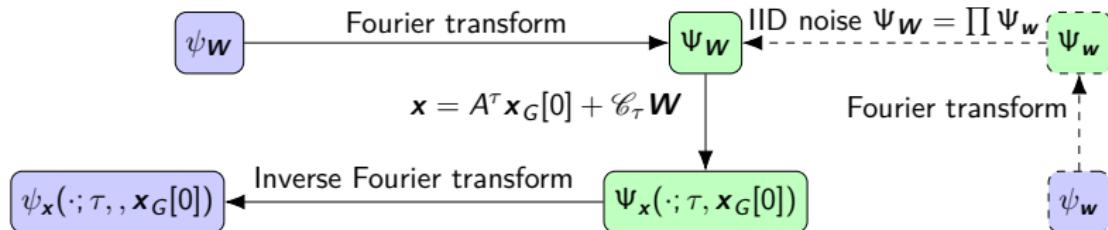
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Vinod, HomChaudhuri, and Oishi, HSCC 2017

Forward reachability for goal and pursuer UAVs

► Catch probability

$$\text{CatchPr}_{\bar{x}_{P_i}}(\tau, \bar{x}_{P_i}[\tau]; \mathbf{x}_G[0]) = \int_{\text{CatchSet}(\bar{x}_{P_i}[\tau])} \psi_{\mathbf{x}_G}(\bar{y}; \tau, \mathbf{x}_G[0]) d\bar{y}$$



$$\Psi_w(\bar{\alpha}) = \mathbb{E}_w \left[\exp \left(j \bar{\alpha}^\top w \right) \right] = \int_{\mathbb{R}^p} e^{j \bar{\alpha}^\top \bar{z}} \psi_w(\bar{z}) d\bar{z}$$

$$\psi_w(\bar{z}) = \left(\frac{1}{2\pi} \right)^p \int_{\mathbb{R}^p} e^{-j \bar{\alpha}^\top \bar{z}} \Psi_w(\bar{\alpha}) d\bar{\alpha}$$

Forward reachability for goal and pursuer UAVs

- Log-concave (over $\bar{x}_{P_i}[\tau]$) τ -time catch probability

$$\text{CatchPr}_{\bar{x}_{P_i}}(\tau, \bar{x}_{P_i}[\tau]; \mathbf{x}_G[0]) = \int_{\text{CatchSet}(\bar{x}_{P_i}[\tau])} \psi_{\mathbf{x}_G}(\bar{y}; \tau, \mathbf{x}_G[0]) d\bar{y}$$

- Goal UAV G's state is Gaussian ($\bar{x}_0 = \mathbf{x}_G[0]$)

$$\mathbf{x}_G[\tau; \bar{x}_0] \sim \mathcal{N}(\bar{\mu}[\tau], \Sigma[\tau])$$

- Deterministic forward reach set for pursuer P_i

$$\begin{aligned}\text{Reach}_{P_i}(\tau; \bar{x}_{P_i}[0]) &= \{\bar{y} \in \mathcal{X} | \exists \bar{\pi}_\tau \in \mathcal{U}_P^\tau \text{ s.t. } \bar{x}_{P_i}[\tau] = \bar{y}\} \\ &= \{A^\tau \bar{x}_{P_i}[0]\} \oplus \mathcal{C}_\tau \mathcal{U}_P^\tau\end{aligned}$$

Single Pursuer Problem- τ :

$$\begin{cases} \underset{\bar{x}_{P_i}[\tau]}{\text{minimize}} & -\log \left(\text{CatchPr}_{\bar{x}_{P_i}}(\tau, \bar{x}_{P_i}[\tau]; \bar{x}_0) \right) \\ \text{subject to} & \bar{x}_{P_i}[\tau] \in \text{Reach}_{P_i}(\tau; \bar{x}_{P_i}[0]) \end{cases}$$

Single Pursuer Problem- τ is convex!

Vinod, HomChaudhuri, and Oishi, HSCC 2017

Problem statements

- Q1** Maximize the team capture probability over a finite time horizon
- Q2** Perform experimental validation of forward stochastic reachability-based interception

Extension to multiple pursuers w/ convexity preserved

- For $\bar{X}_P[\tau] = [(\bar{x}_{P_1}[\tau])^\top, (\bar{x}_{P_2}[\tau])^\top, \dots, (\bar{x}_{P_N}[\tau])^\top]^\top$,

$$\text{TeamCatchPr}(\tau, \bar{X}_P; \mathbf{x}_G[0]) = \max_{P_i} [\text{CatchPr}_{\bar{x}_{P_i}}(\tau, \bar{x}_{P_i}[\tau]; \mathbf{x}_G[0])]$$

Multiple Pursuer Problem :

$$\left\{ \begin{array}{ll} \underset{\tau, \bar{X}_P[\tau]}{\text{maximize}} & \text{TeamCatchPr}(\tau, \bar{X}_P[\tau]; \bar{x}_0) \\ \text{subject to} & \left\{ \begin{array}{l} \tau \in \mathbb{Z}_{[1, N]} \\ \bar{X}_P[\tau] \text{ from dynamics} \end{array} \right. \end{array} \right.$$

Single Pursuer Problem- i :

$$\left\{ \begin{array}{ll} \underset{\tau, \bar{x}_{P_i}[\tau]}{\text{maximize}} & \text{CatchPr}_{\bar{x}_{P_i}}(\tau, \bar{x}_{P_i}[\tau]; \bar{x}_0) \\ \text{subject to} & \left\{ \begin{array}{l} \tau \in \mathbb{Z}_{[1, N]} \\ \bar{x}_{P_i}[\tau] \in \text{Reach}_{P_i}(\tau, \bar{x}_{P_i}[0]) \end{array} \right. \end{array} \right.$$

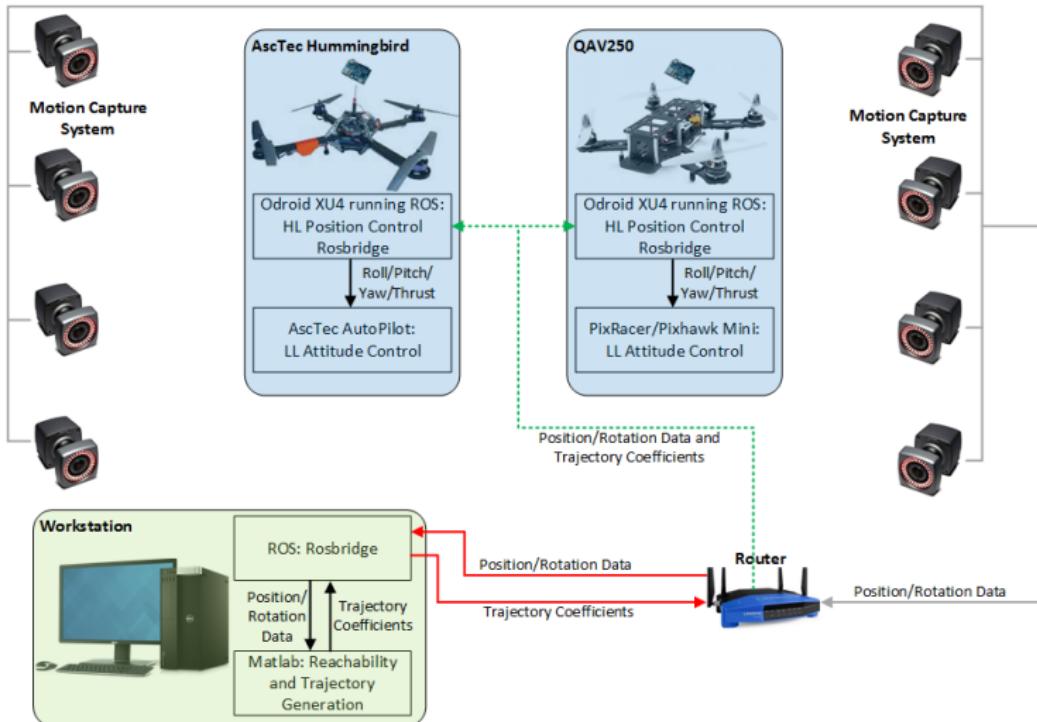
Single Pursuer Problem-(τ, i) :

$$\left\{ \begin{array}{ll} \underset{\bar{x}_{P_i}[\tau]}{\text{minimize}} & -\log \left(\text{CatchPr}_{\bar{x}_{P_i}}(\tau, \bar{x}_{P_i}[\tau]; \bar{x}_0) \right) \\ \text{subject to} & \bar{x}_{P_i}[\tau] \in \text{Reach}_{P_i}(\tau; \bar{x}_{P_i}[0]) \end{array} \right.$$

Single Pursuer Problem-(τ, i) is convex!

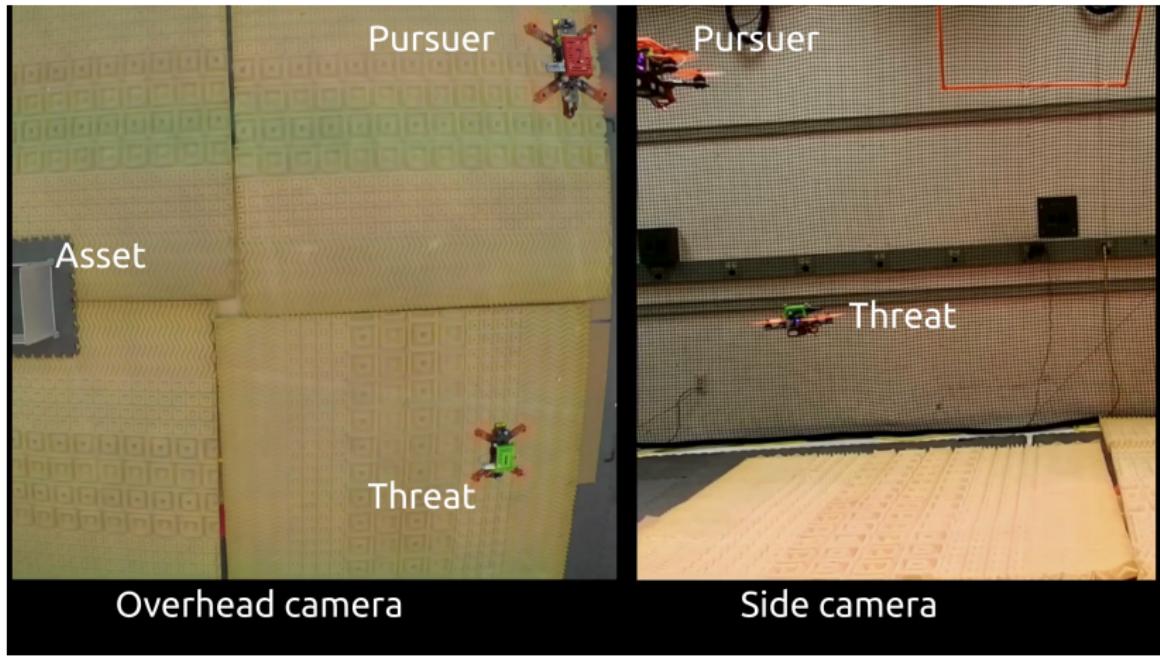
- Enforce coordination via other definitions for TeamCatchPr

Experimental Setup



Multi-Agent, Robotics, Heterogeneous Systems (MARHES) Lab
(<http://marhes.unm.edu/>)

Robustness to model uncertainty



- ▶ Mismatch in dynamics (human-controlled)
- ▶ Robustness due to receding horizon control

Robustness to model uncertainty

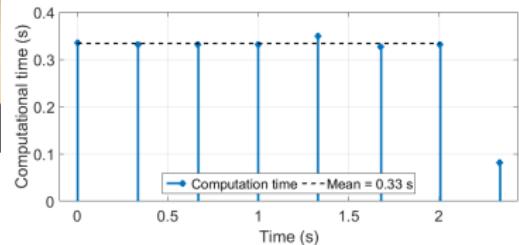
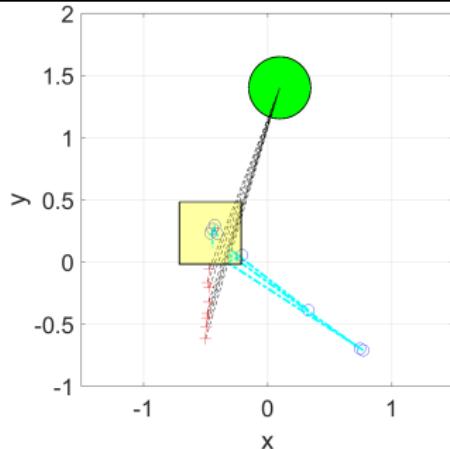
<https://www.youtube.com/watch?v=eFGg7U7gEQw>

Robustness to model uncertainty

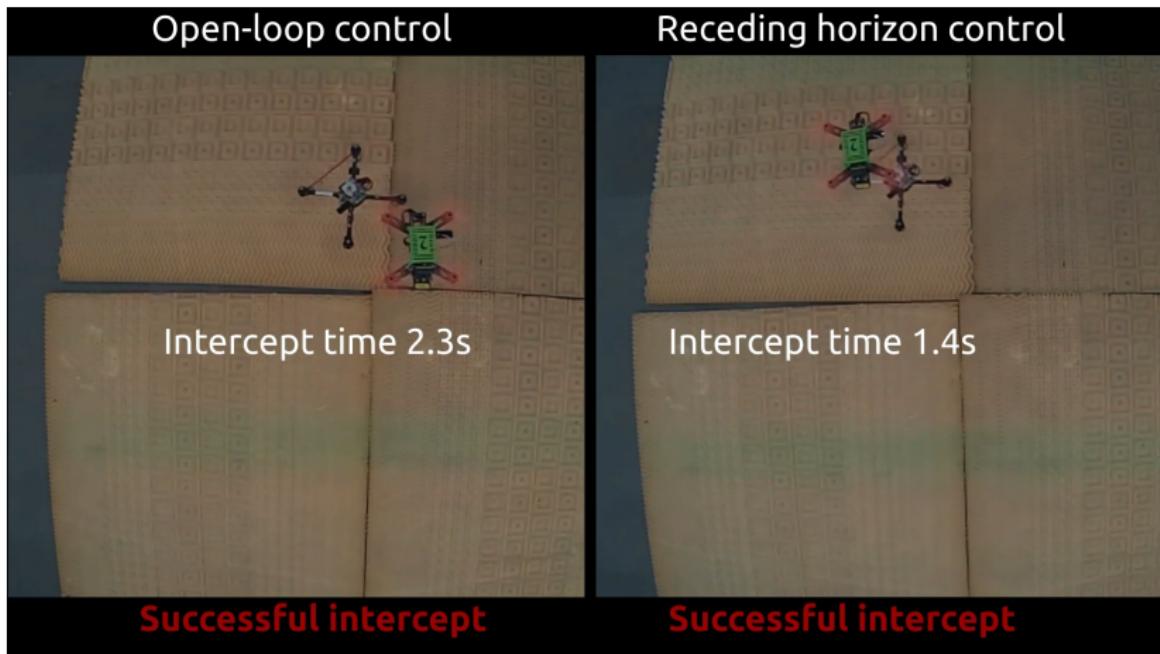


Successful intercept --- Threat within 0.25m of the pursuer in x-y

- + Threat
- o Pursuer (RHC)
- Threat predicted $\bar{\mu}_{x_G}[k; \cdot]$
- Pursuer desired $p_i(t)$
- [green] Asset
- [yellow] Capture Set



Open-loop vs receding horizon control

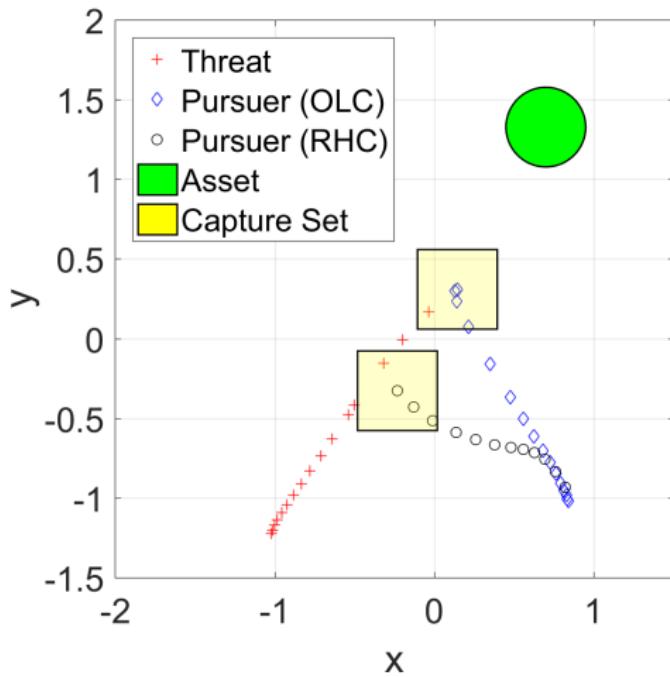


- Threat UAV follows the LQR-based stochastic dynamics
- Receding horizon control > open-loop control

Open-loop vs receding horizon control

<https://www.youtube.com/watch?v=H0BZrk9Goxg>

Open-loop vs receding horizon control



Summary, future work, and acknowledgments

Summary

- ▶ Stochastic reachability-based pursuit w/ multiple pursuers
- ▶ Experimental validation of forward stochastic reachability

Future work

- ▶ Experiments with multiple pursuers
- ▶ Enforcing more coordination
- ▶ Enforcing keep-out regions

Acknowledgments

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- ▶ NSF CNS-1329878

