

Set theoretic approach for target localization and tracking using a fleet of drones

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Outline

1. Introduction
2. Accounting for decoys
3. Detailed Framework
4. Conclusion

Context

Considered problem:

- Searching, detecting, and tracking mobile targets
- Potentially large region of interest
- Cooperative agents (*e.g.* UAVs)

Search process usually based on probabilistic information

[Bertuccelli and How, 2005, Khodayi-mehr et al., 2019, Li and Duan, 2017, Moon, 2008, Sun et al., 2014, Yao et al., 2016, Yang et al., 2007, Zhang et al., 2017]

- Efficiency depends on
 - availability,
 - quality,
 - and reliability of the information.

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Problem statement

Performance sensitive to *a priori* assumptions on Probability Density Function of process [Gu et al., 2015].

Alternative assumption:

- Noises and uncertainties bounded

Approach:

- Set-membership state estimation [Reynaud et al., 2018, Reboul et al., 2019]
- Set-membership compliant Model Predictive Control (MPC)

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Objectives

Targets search considering the following issues:

- Limited field of view and communication range
- Unknown displacement of targets
- Presence of decoys/false targets
- Confusion between target tracks
- Non-detection occurrences
- Uncertainties of the position of each drone

Main objective:

- Developing set-membership estimator
- Designing cooperative guidance scheme

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Limited FoV and presence of decoys

Considering localization within search area:

- of *a priori* unknown number of targets
- in presence of static decoys
- search area uncluttered
- non-detection not considered

Robust state estimator processes sets guaranteed to contain state of true targets already detected.¹

¹J. Ibenthal, L. Meyer, M. Kieffer, H. Piet-Lahanier, Bounded-error target localization and tracking in presence of decoys using a fleet of UAVs, presented at 21st IFAC World Congress in Berlin, Germany, 2020

Limited communication, obstacles and moving decoys

Considering localization within search area:

- of *a priori* unknown number of **indistinguishable** targets
- in presence of **moving** decoys
- search area with **known obstacles**
- **limited communication**
- non-detection not considered herein²

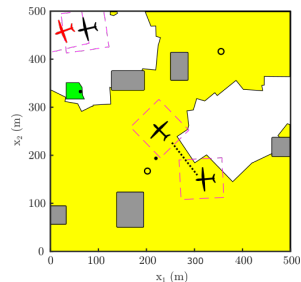


Figure: Simulations

²J. Ibenthal, L. Meyer, H. Piet-Lahanier, M. Kieffer, Target search and tracking using a fleet of UAVs in presence of decoys and obstacles, presented at IEEE CDC, Jeju Island, 2020.

Generalization of the prior approaches

Considering localization within search area:

- of *a priori* unknown number of targets
 - in presence of moving decoys
 - with limited and **delayed** communication
 - non-detection still not considered³
- **Identification condition modeled by unknown geometric conditions**
 - **Extensive study of the parameters in simulations**

³J. Ibenthal, M. Kieffer, L. Meyer, H. Piet-Lahanier, Bounded-error target localization and tracking using a fleet of UAVs, Automatica, accepted 2021.

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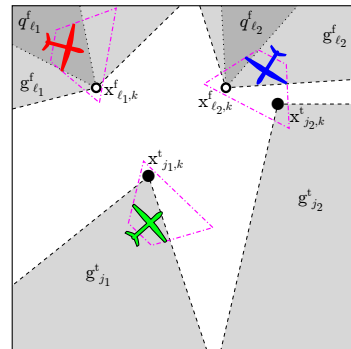
Outline

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2. Accounting for decoys
3. Detailed Framework
 - Problem formulation
 - Evolution of the set estimates
 - Cooperative control design
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Problem formulation

Consider

- fleet of UAVs equipped with optical seekers
- moving true targets
- moving false targets
- detection and identification under specific conditions



UAVs and targets

Region of interest and possible target locations $\mathbb{X}_0$

Time is discretized: at time instant $t = kT$

- state vector $\mathbf{x}_{i,k}^u$ of UAV $i = 1, \dots, N_u$ evolves as

$$\mathbf{x}_{i,k+1}^u = \mathbf{f}_k^u \left(\mathbf{x}_{i,k}^u, \mathbf{u}_{i,k} \right),$$

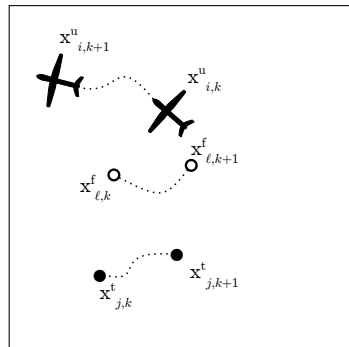
with control input $\mathbf{u}_{i,k} \in \mathbb{U}$

- state vector $\mathbf{x}_{j,k}^t$ of target j evolves as

$$\mathbf{x}_{j,k+1}^t = \mathbf{f}_k^t \left(\mathbf{x}_{j,k}^t, \mathbf{v}_{j,k} \right),$$

with state perturbation $\mathbf{v}_{j,k} \in [\mathbf{v}_k]$

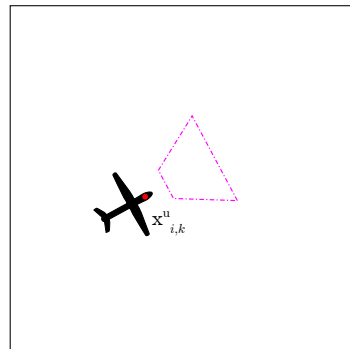
- false targets may be static or moving



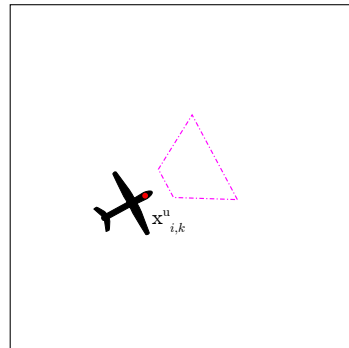
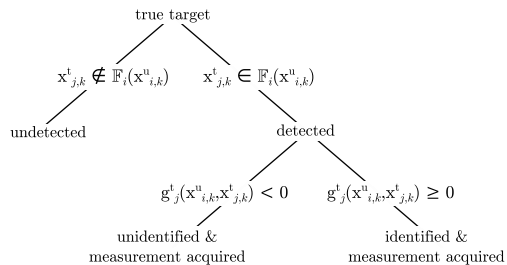
Measurements

UAVs equipped with optical seekers:

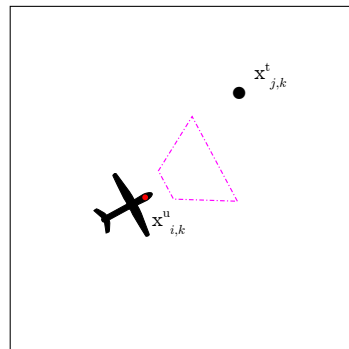
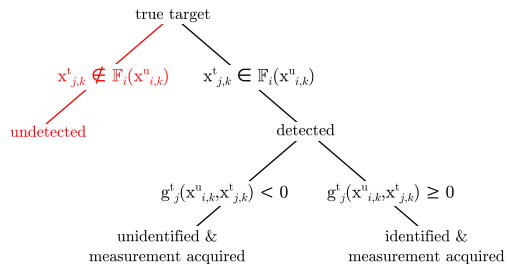
- Field of view $\mathbb{F}_i(\mathbf{x}_i^u)$



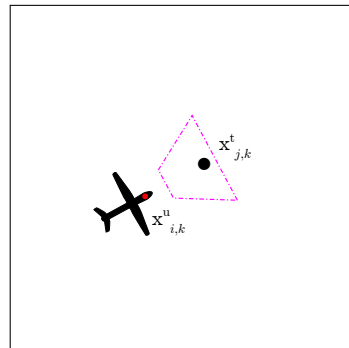
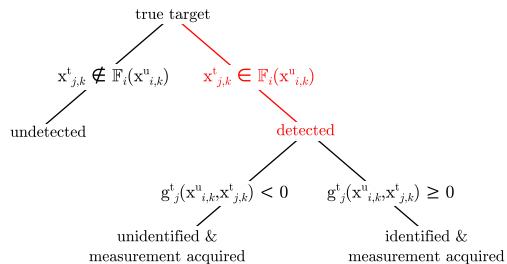
Measurements for true targets



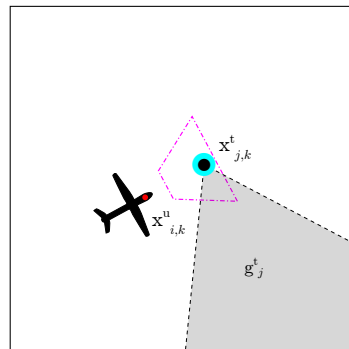
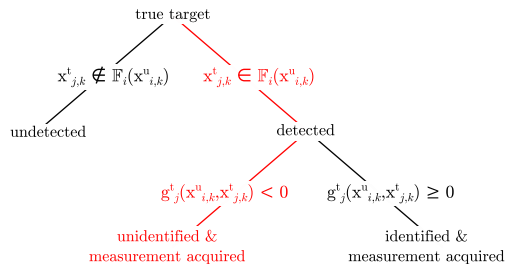
Measurements for true targets



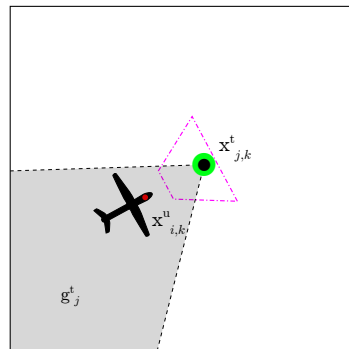
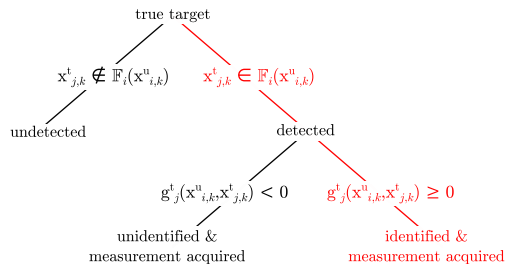
Measurements for true targets



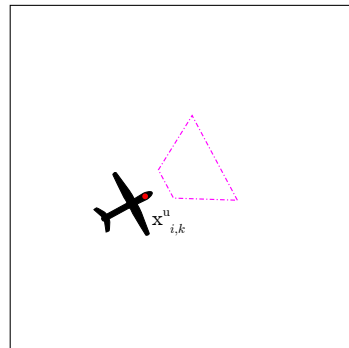
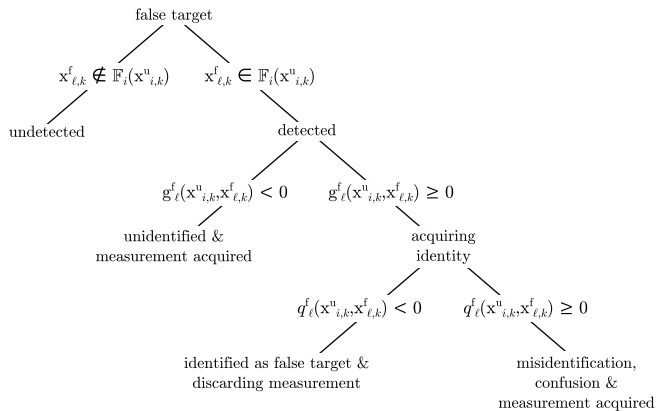
Measurements for true targets



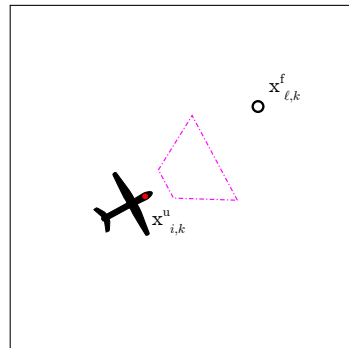
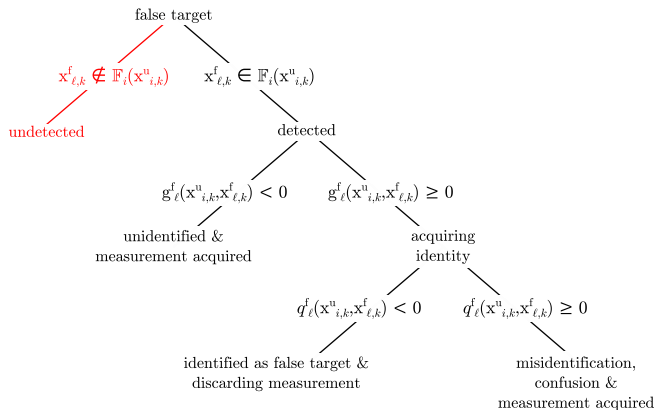
Measurements for true targets



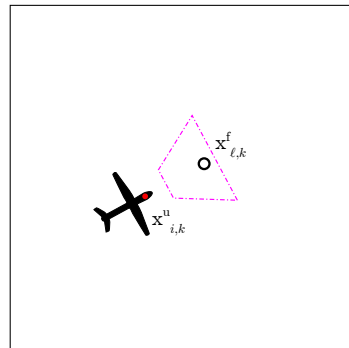
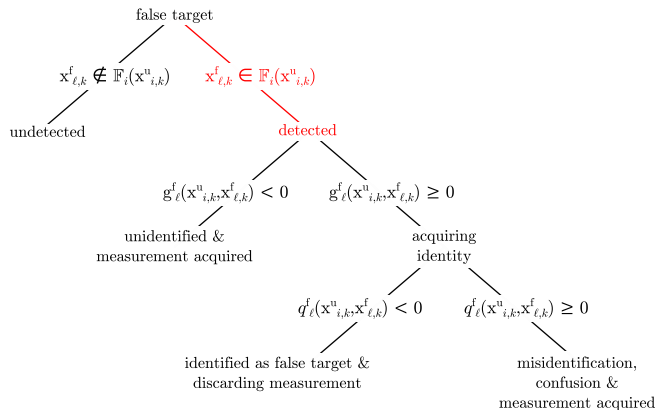
Measurements for false targets



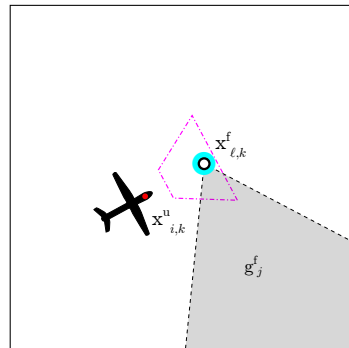
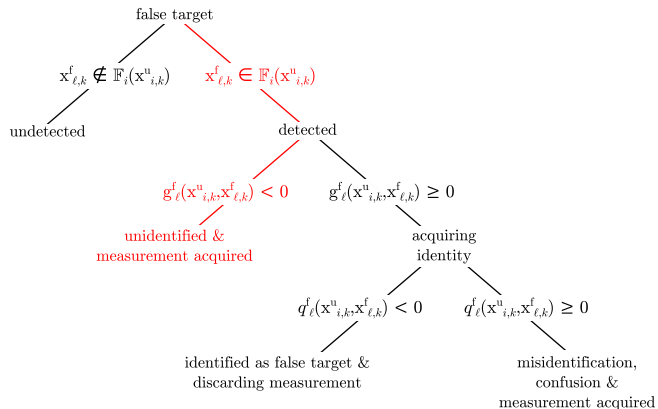
Measurements for false targets



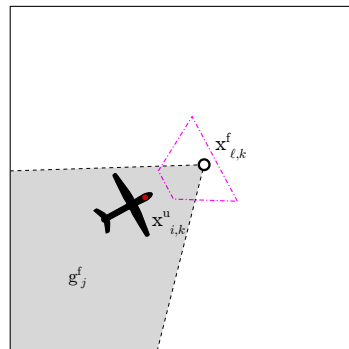
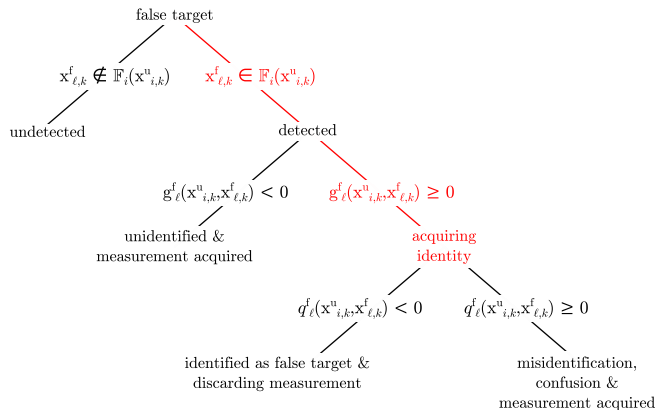
Measurements for false targets



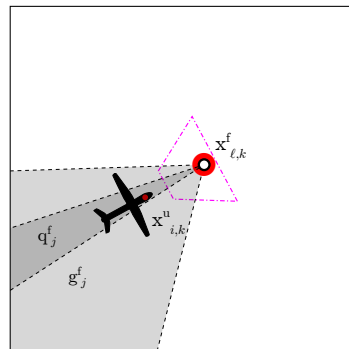
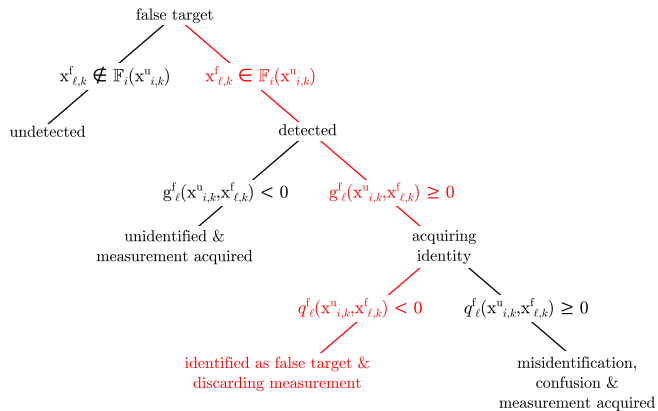
Measurements for false targets



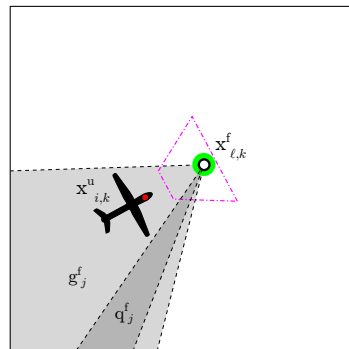
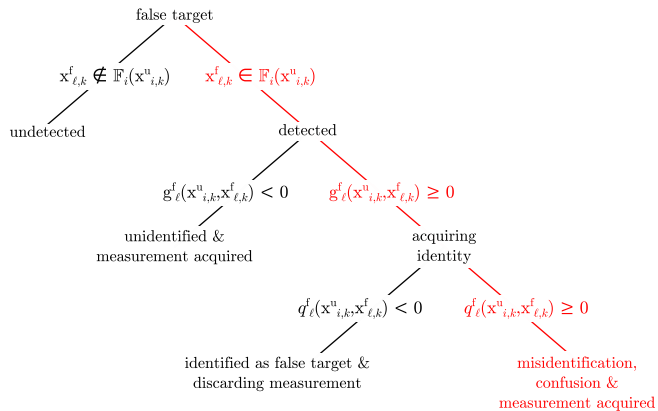
Measurements for false targets



Measurements for false targets



Measurements for false targets



Observation

Noisy observation of target state

$$\mathbf{y}_{i,j,k} = \mathbf{h}_i(\mathbf{x}_{i,k}^u, \mathbf{x}_{j,k}^t) + \mathbf{w}_{i,j,k}$$

$\mathbf{w}_{i,j,k}$ represents the measurement noise, bounded in some known box $[\mathbf{w}]$

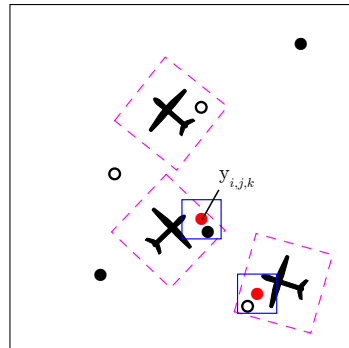


Figure: Collecting observations

Communication

Two UAVs in the vicinity are able to communicate.

Exchanged Information:

- UAV state value
- Control inputs
- Set estimates

Assumption:

- communication without error

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Estimates

Unexplored set $\bar{\mathbb{X}}_{i,k}$

- Possible states of not yet detected targets

Target state estimate $\mathcal{X}_{i,k} = \{\mathbb{X}_{i,j,k}\}_{j \in \mathcal{L}_{i,k}}$ for **identified** targets

- may not contain a target due to presence of decoys
- $\mathcal{L}_{i,k}$: list of indices of targets already detected

Target state estimate $\mathbb{X}_k^U$ for **unidentified** targets

- may contain targets and decoys

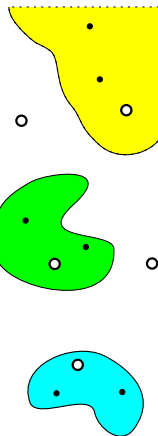


Figure: Estimates

Nonlinear recursive set-membership state estimator

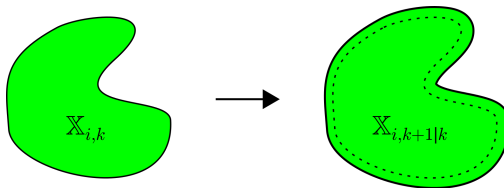
Initialization at $k = 0$:

- $\mathcal{L}_{i,0} = \emptyset$,
- $\mathcal{X}_{i,0} = \emptyset$,
- $\mathbb{X}_{i,0}^U = \emptyset$
- and $\bar{\mathbb{X}}_{i,0} = \mathbb{X}_0$ for $i = 1, \dots, N_u$

Prediction

$$k \rightarrow k+1 | k \rightarrow k+1 | k+1 \rightarrow k+1$$

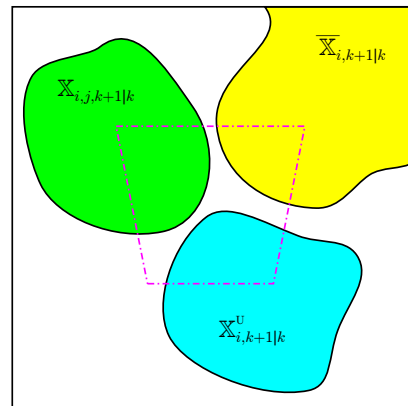
$$\mathbb{X}_{i,j,k+1|k} = \mathbf{f}_k^t(\mathbb{X}_{i,j,k}, [\mathbf{v}_k]) \cap \mathbb{X}_0$$



Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Set estimate update from
measurements

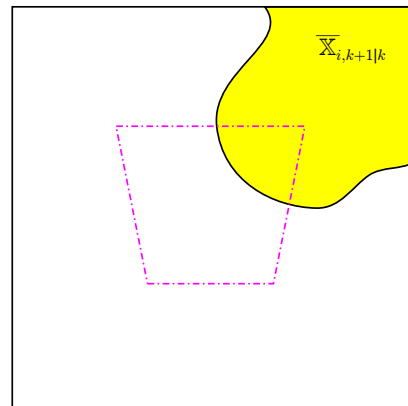


Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating unexplored set $\bar{\mathbb{X}}_{i,k+1|k}$

When no target is detected



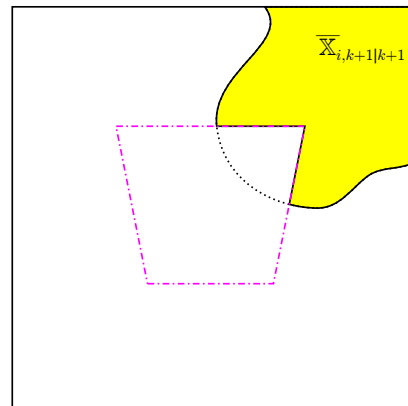
Correction from measurements

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Updating unexplored set $\bar{\mathbb{X}}_{i,k+1|k}$

When no target is detected

$$\bar{\mathbb{X}}_{i,k+1|k+1} = \bar{\mathbb{X}}_{i,k+1|k} \setminus \mathbb{F}_i(\mathbf{x}_{i,k+1}^u)$$



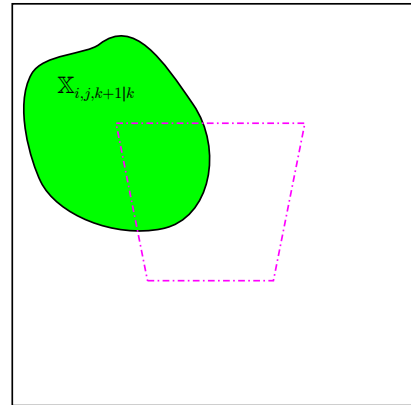
Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating target set estimate

$$\mathbb{X}_{i,j,k+1|k}$$

When no target is detected



Correction from measurements

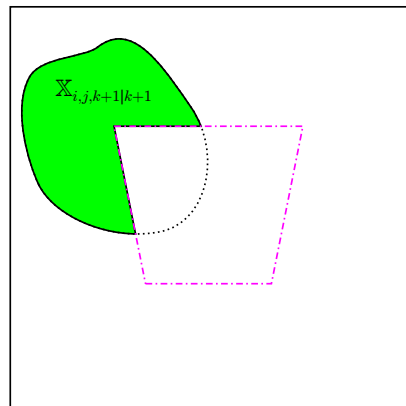
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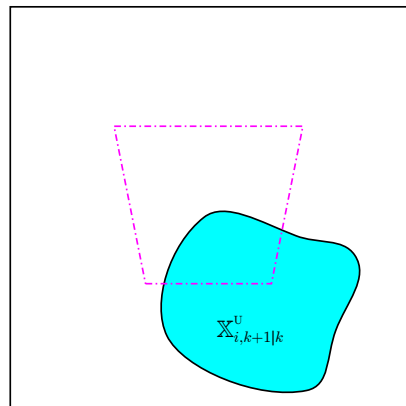


Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating unidentified target set
estimate $\mathbb{X}_{i,k+1|k}^U$

When no target is detected



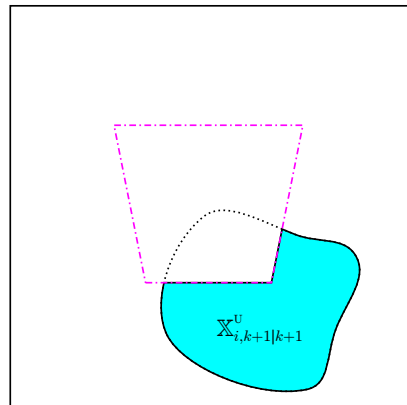
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Updating unidentified target set
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When no target is detected

$$\mathbb{X}_{i,k+1|k+1}^U = \mathbb{X}_{i,k+1|k}^U \setminus \mathbb{F}_i(\mathbf{x}_{i,k+1}^u)$$



Correction from measurements

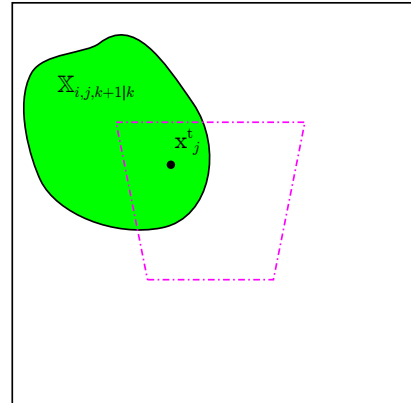
$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating target set estimate

$$\mathbb{X}_{i,j,k+1|k}$$

When a target is

- detected again
- and identified



Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating target set estimate

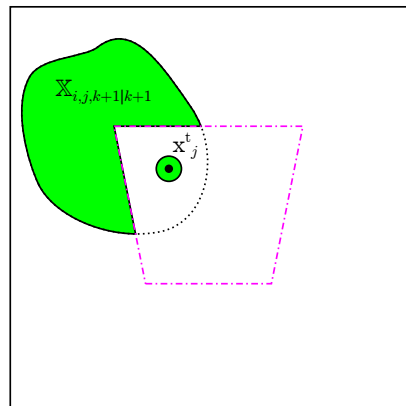
$$\mathbb{X}_{i,j,k+1|k}$$

When a target is

- detected again
- and identified

$$\mathbb{S}_1 = \left\{ \mathbf{x} \in \mathbb{X}_{i,j,k+1|k} \mid \mathbf{h}_i \left(\mathbf{x}_{i,k+1}^u, \mathbf{x} \right) \in \mathbf{y}_{i,j,k+1}^l - [\mathbf{w}_{i,k}] \right\}$$

$$\mathbb{X}_{i,j,k+1|k+1} = \left(\mathbb{X}_{i,j,k+1|k} \setminus \mathbb{F}_i \left(\mathbf{x}_{i,k+1}^u \right) \right) \cup \mathbb{S}_1$$



Correction from measurements

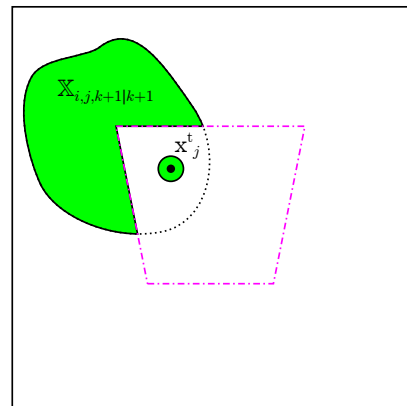
$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating target set estimate

$$\mathbb{X}_{i,j,k+1|k}$$

When a target is

- detected again
- and **not** identified



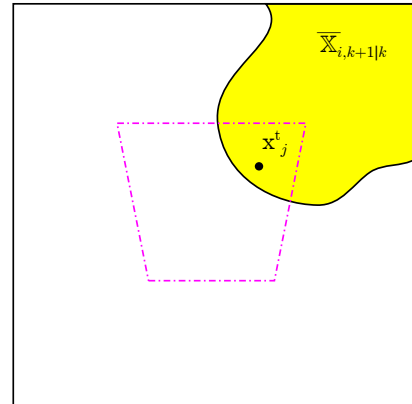
Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating target set estimate

$$\bar{\mathbb{X}}_{i,j,k+1|k}$$

When a new target is identified



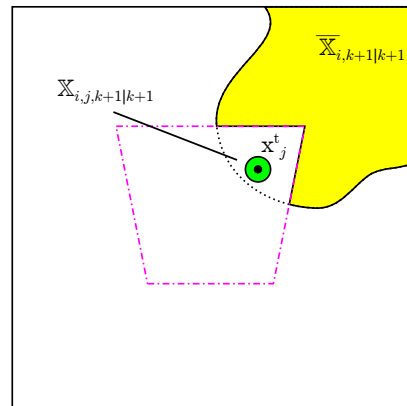
Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating target set estimate

$$\mathbb{X}_{i,j,k+1|k}$$

When a new target is identified



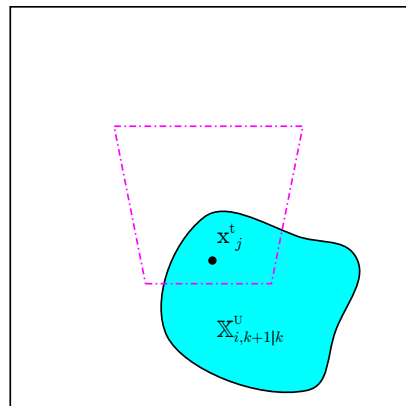
Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating target set estimate

$$\mathbb{X}_{i,j,k+1|k}$$

When a unidentified target is identified



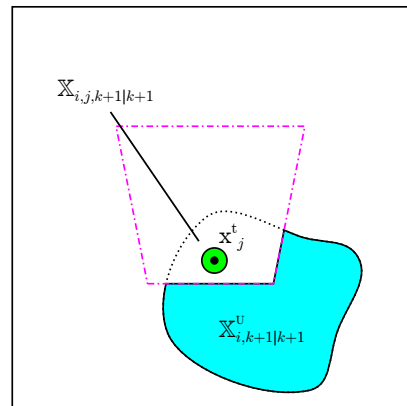
Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating target set estimate

$$\mathbb{X}_{i,j,k+1|k}$$

When a unidentified target is identified



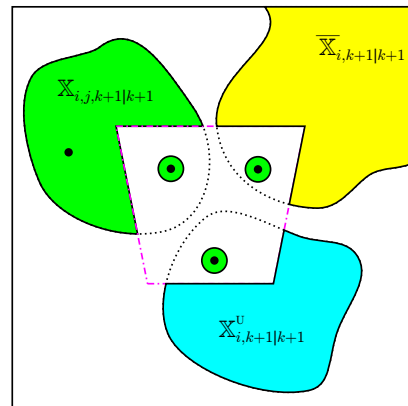
Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating target set estimate

$$\mathbb{X}_{i,j,k+1|k}$$

Accounting for all cases



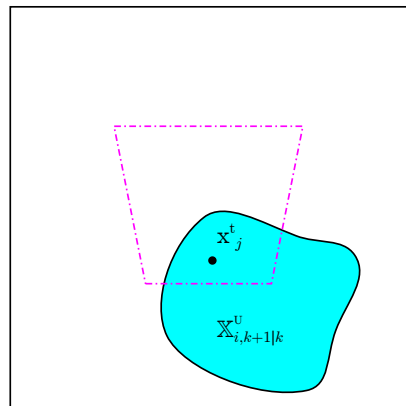
Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating unidentified target set
estimate $\mathbb{X}_{i,k+1|k}^U$

When a target is

- detected again
- and not identified



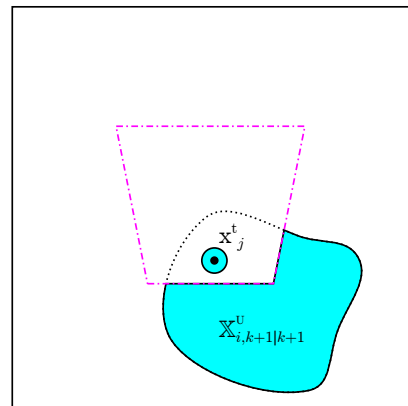
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$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

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When a target is

- detected again
- and not identified

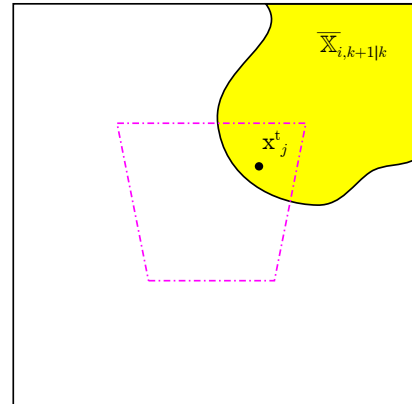


Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating unidentified target set
estimate $\mathbb{X}_{i,k+1|k}^U$

When new target is not identified

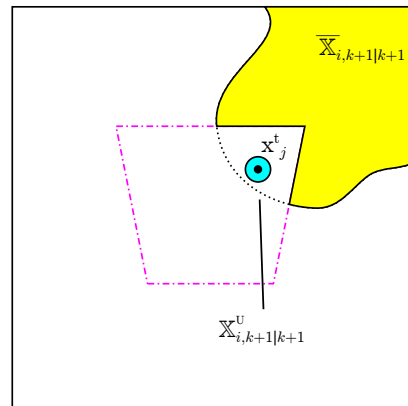


Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating unidentified target set
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When new target is not identified

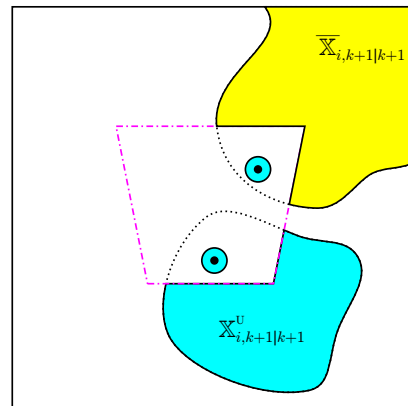


Correction from measurements

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$

Updating unidentified target set
estimate $\mathbb{X}_{i,k+1|k}^U$

Accounting for all cases



Correction from communication

$$k \rightarrow k+1 | k \rightarrow k+1 | k+1 \rightarrow k+1$$

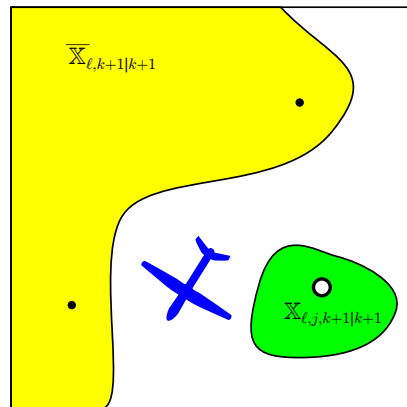
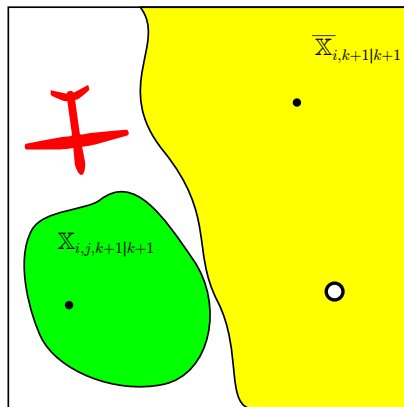
At the end of each time step k UAV i communicates with its neighbors

Exchanged information:

- Target set estimate
- Unexplored set
- Receiving the corresponding sets from its neighbors.

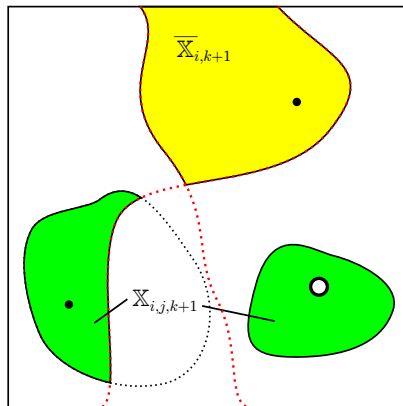
Correction from communication

$$k \rightarrow k+1|k \rightarrow k+1|k+1 \rightarrow k+1$$



Correction from communication

$$k \rightarrow k+1 | k \rightarrow k+1 | k+1 \rightarrow k+1$$



Outline

1. Introduction
2. Accounting for decoys
3. Detailed Framework
 - Problem formulation
 - Evolution of the set estimates
 - Cooperative control design
 - Simulation
4. Conclusion

Cooperative distributed control design

Target state estimation uncertainty at time k for UAV i

$$\Phi \left(\mathcal{X}_{i,k}, \mathbb{X}_{i,k}^U, \bar{\mathbb{X}}_{i,k} \right) = \phi \left(\left(\bigcup_{\mathbb{X}_{i,j,k} \in \mathcal{X}_{i,k}} \mathbb{X}_{i,j,k} \right) \cup \mathbb{X}_{i,k}^U \cup \bar{\mathbb{X}}_{i,k} \right)$$

Each UAV

- has access to $\mathbb{X}_{i,j,k}$, $\mathbb{X}_{i,k}^U$, and $\bar{\mathbb{X}}_{i,k}$
- determines sequence of control inputs minimizing estimation uncertainty

We rely on a model predictive control approach

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Simulation conditions

The performance of the proposed approach is evaluated via simulations:

- Targets move on the ground with a constant speed module V^t
- UAVs fly at constant altitude and with constant speed
- UAVs are equipped with identical optical sensors
- The control input applied to yaw rate derivative
- Area of interest $\mathbb{X}_0$: cube of $[0, 400] \times [0, 400] \times [0, 100] \text{ m}^3$

Results - Impact of the fleet size

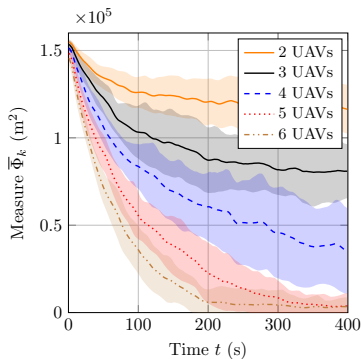


Figure: Mean values (line) and standard deviation (shaded area) of $\bar{\Phi}_k$ evaluated for 30 simulations with 3 true targets, 3 false targets, and 2 to 6 UAVs.

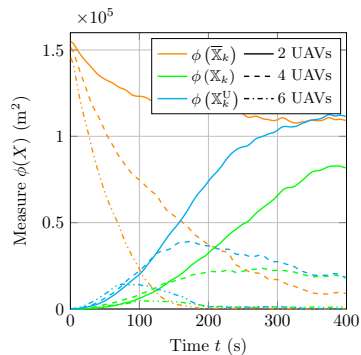


Figure: Mean values of $\phi(\bar{\mathbb{X}}_k)$, $\phi(\mathbb{X}_k^U)$, and $\phi(\mathbb{X}_k)$ evaluated with 3 true and 3 false targets, considering 2, 4 and 6 UAVs.

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Conclusion

Development of several approaches solving the following issues:

- Cooperative target localization and tracking
- Static and dynamic targets
- Identification or detection of a target depends deterministically on the point of view
- Presence of static and dynamic decoys
- Presence of obstacles
- Distributed UAV control scheme

Conclusion

Other developments include

- Study of non-detection with moving targets
- Improve displacement strategies of UAVs accounting for their points of view
- Compare proposed approaches with probabilistic target search
- Implement on test platform (quadrotors)

Perspectives:

- Investigate game theory for cooperative guidance

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