



# Confidence-Based Buffer for Strategic Deconfliction with Probabilistic Operational Intent

José Ignacio de Alvear Cárdenas

Priyank Pradeep

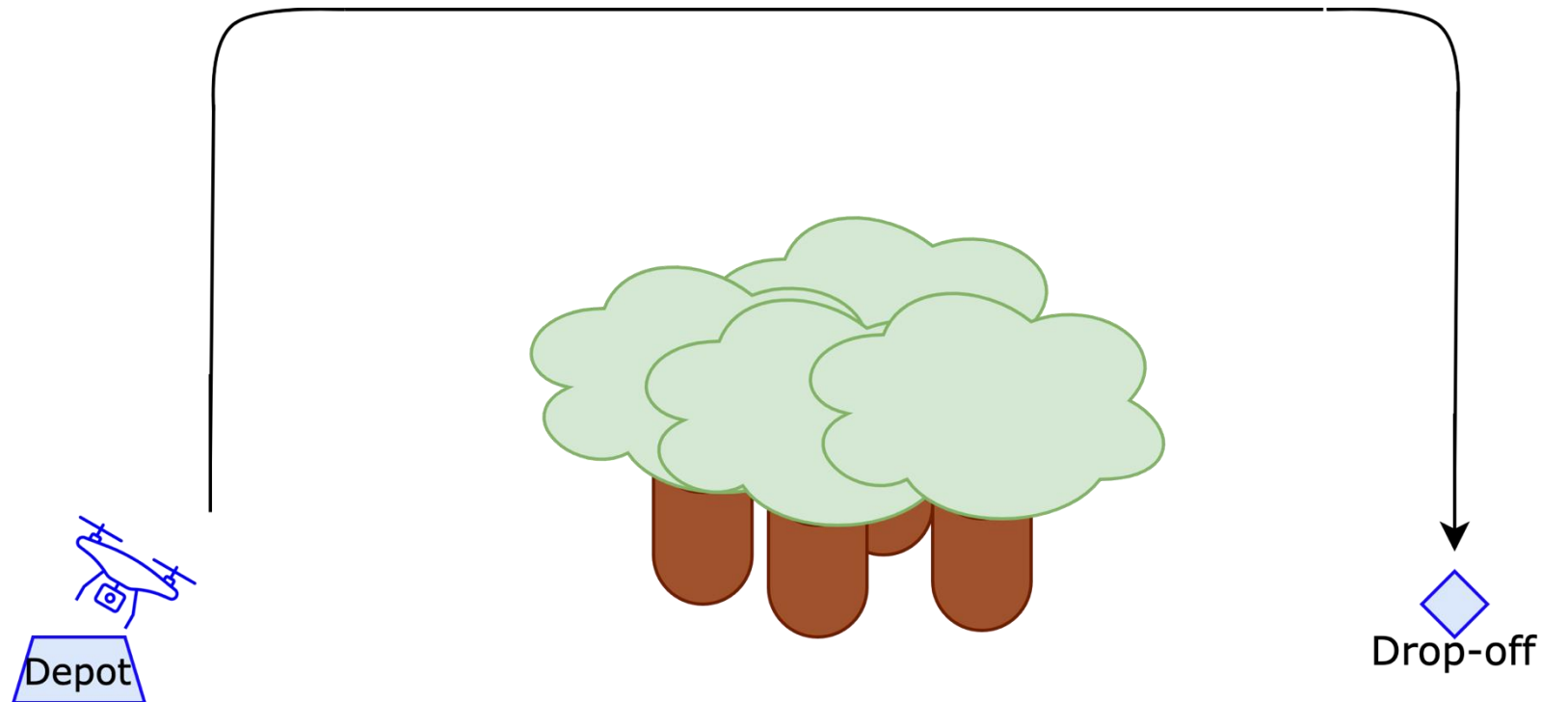
Min Xue

Vincent H. Kuo

Seungman Lee

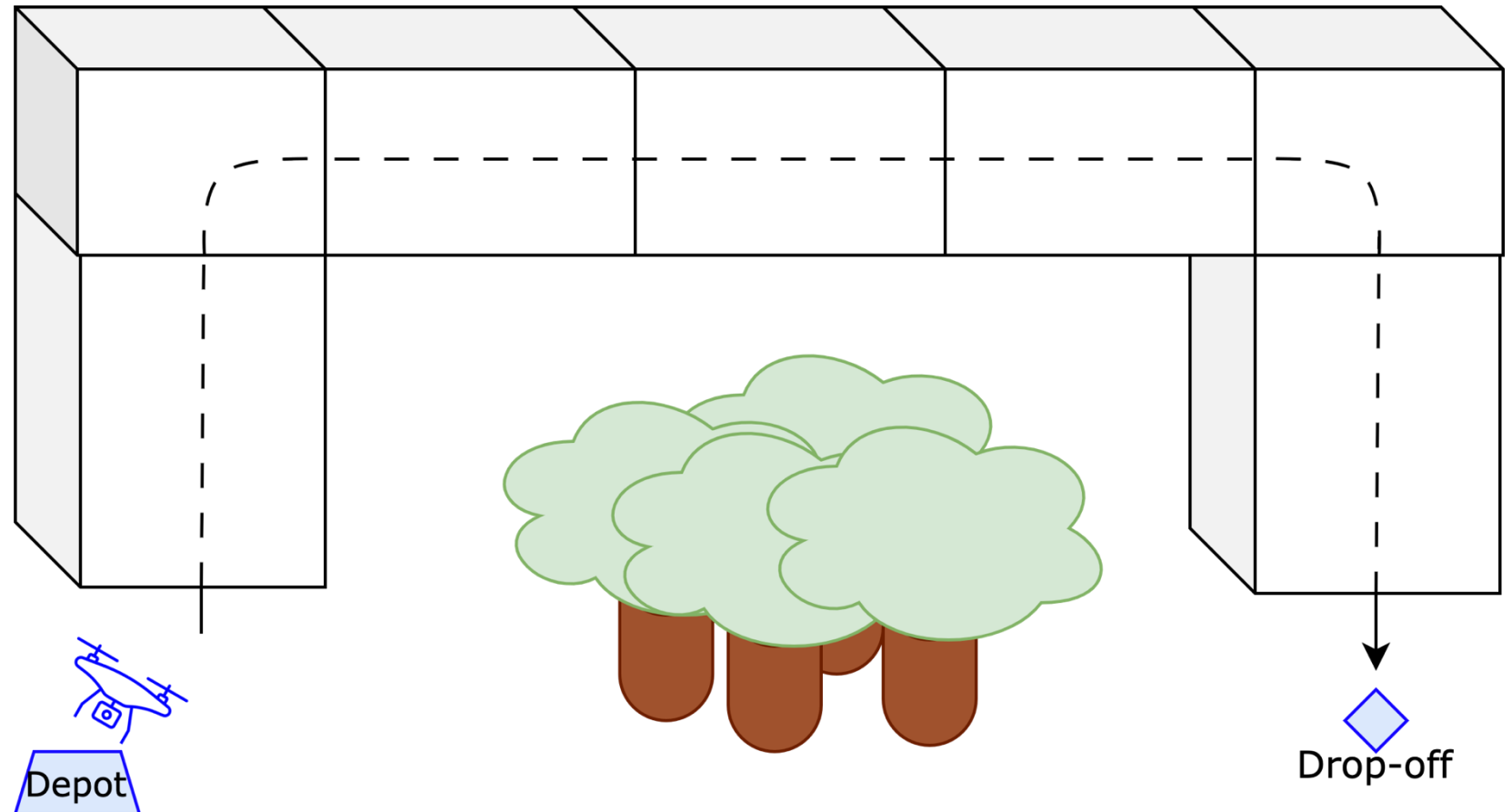
## UTM BVLOS

- Trajectory-based operation



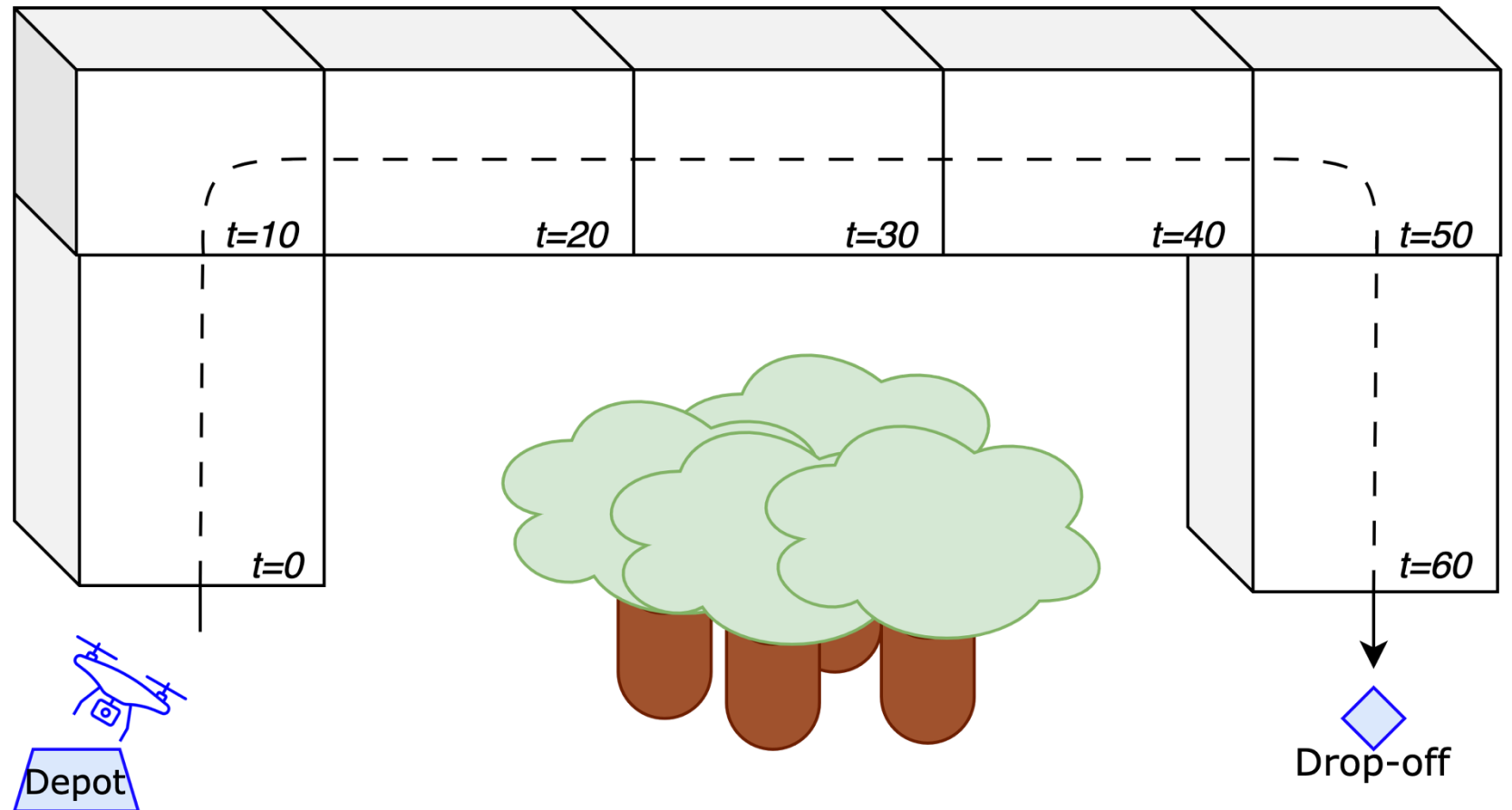
## UTM BVLOS

- Trajectory-based operation
- 4D Operational Volume Blocks (OVBs)



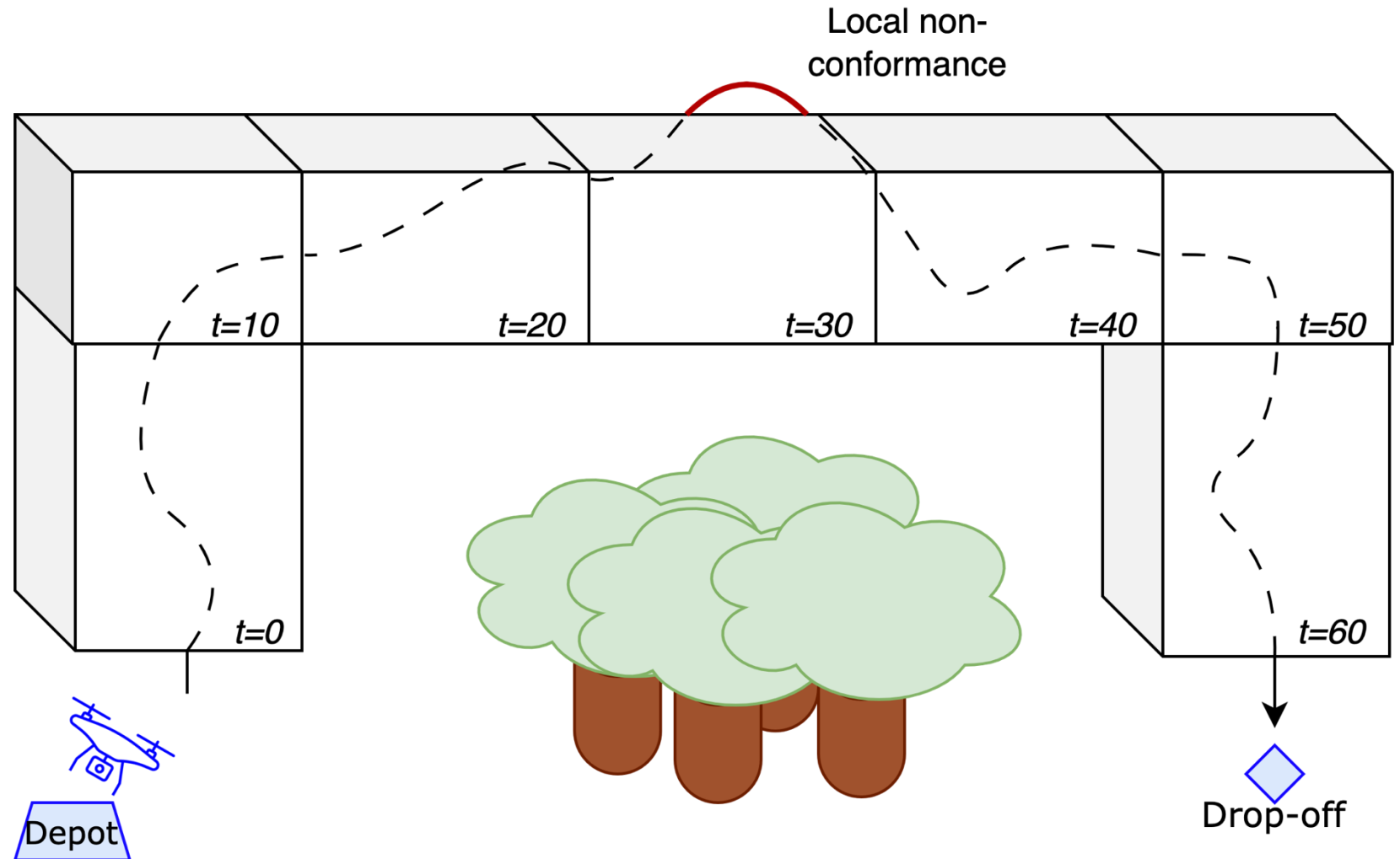
## UTM BVLOS

- Trajectory-based operation
- 4D Operational Volume Blocks (OVBs)
- Fixed in space and specified entry and exit times



## UTM BVLOS

- Trajectory-based operation
- 4D Operational Volume Blocks (OVBs)
- Fixed in space and specified entry and exit times
- ASTM International: 95% conformance

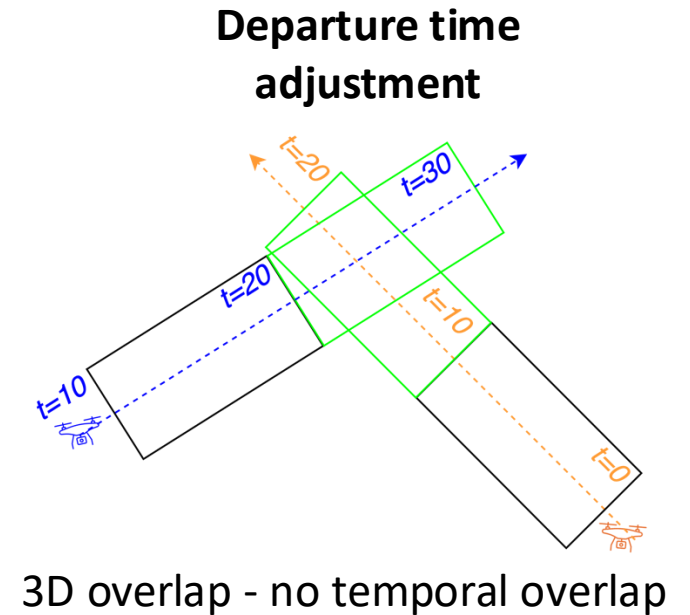
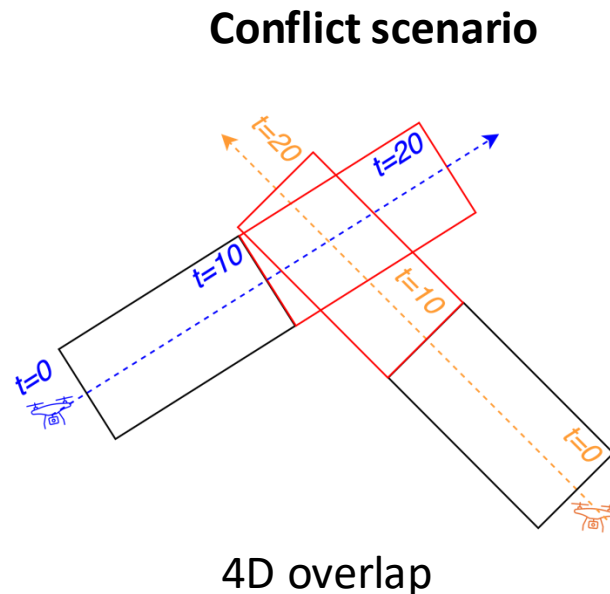
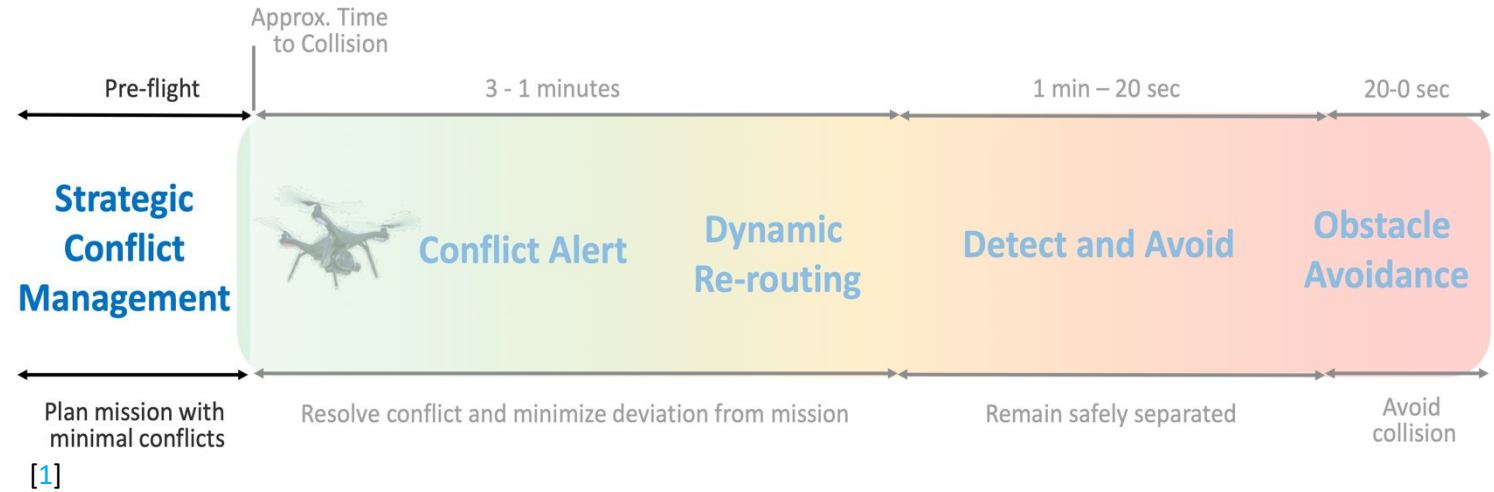


## UTM BVLOS

- Multi-layer conflict management model

## Strategic deconfliction

- Time, Velocity, Altitude, Heading



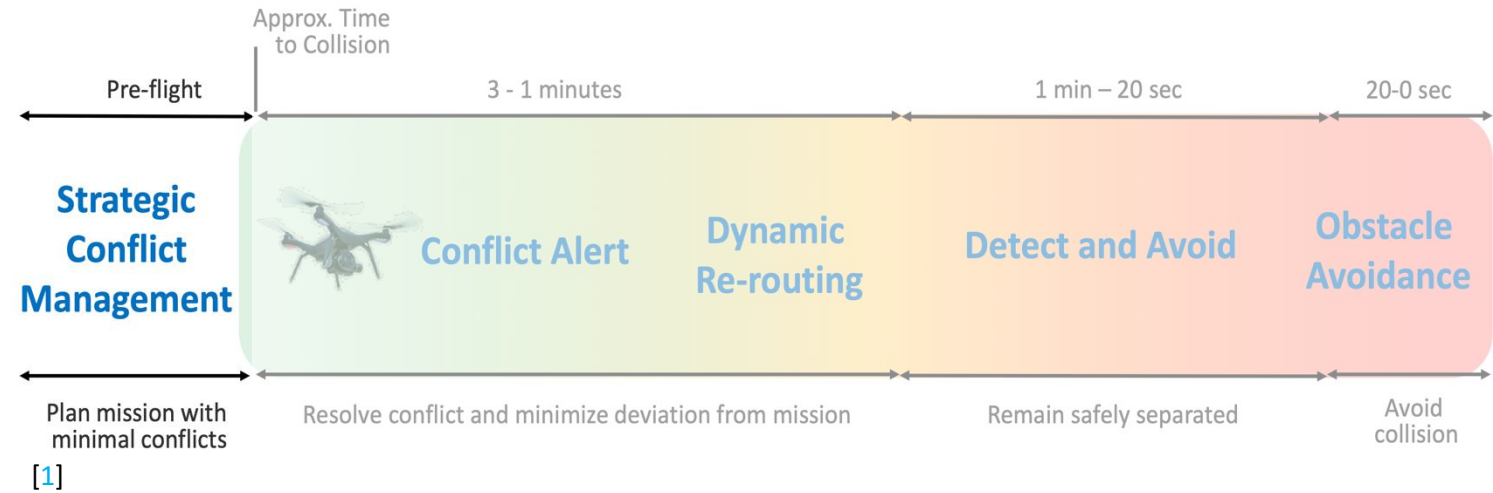
\* Top-down view of cruise flight phase

## UTM BVLOS

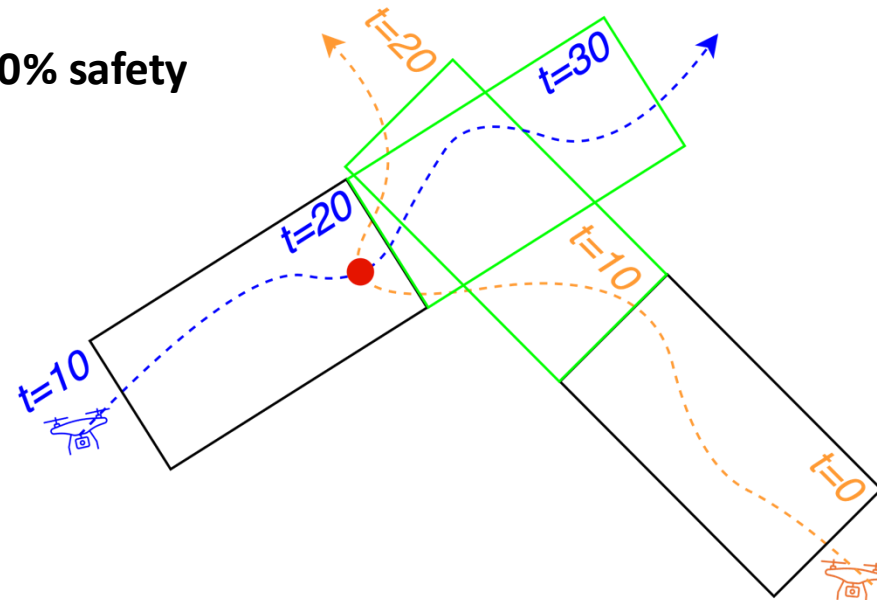
- Multi-layer conflict management model

## Strategic deconfliction

- Time, Velocity, Altitude, Heading



No overlap ≠ 100% safety



\* Top-down view of cruise flight phase

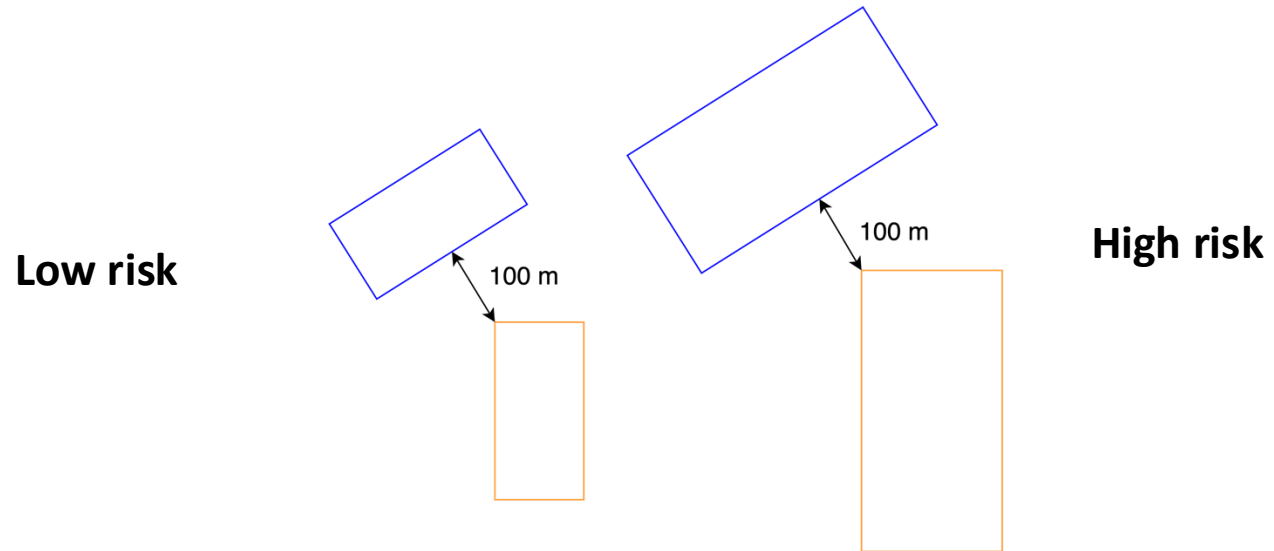
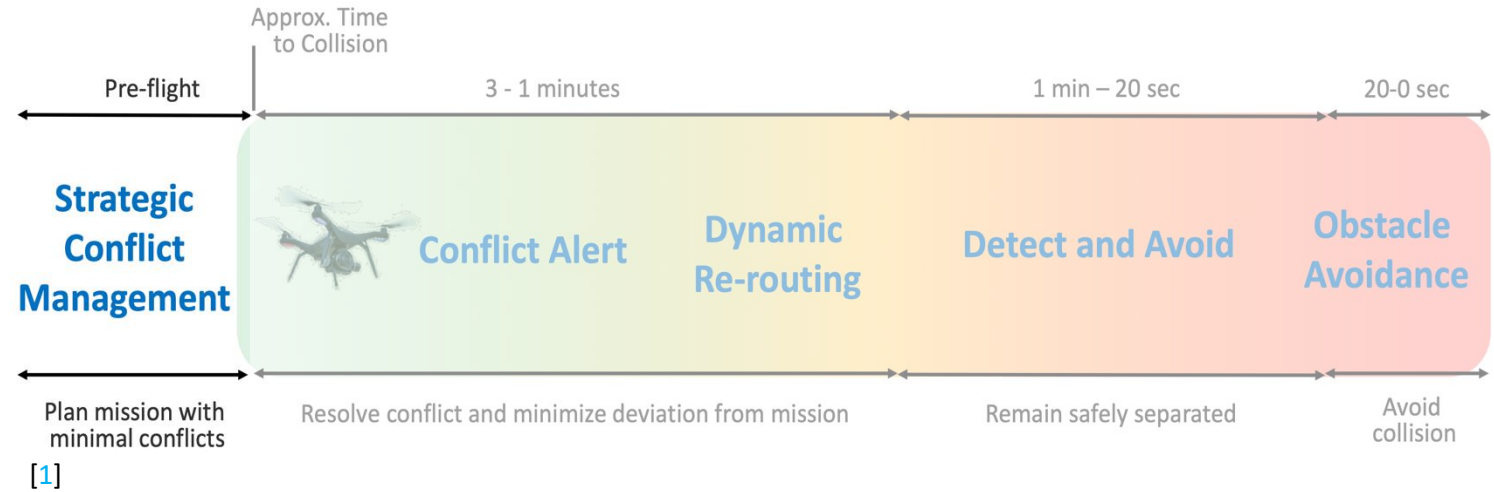
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## Fix size buffer



\* Top-down view of cruise flight phase

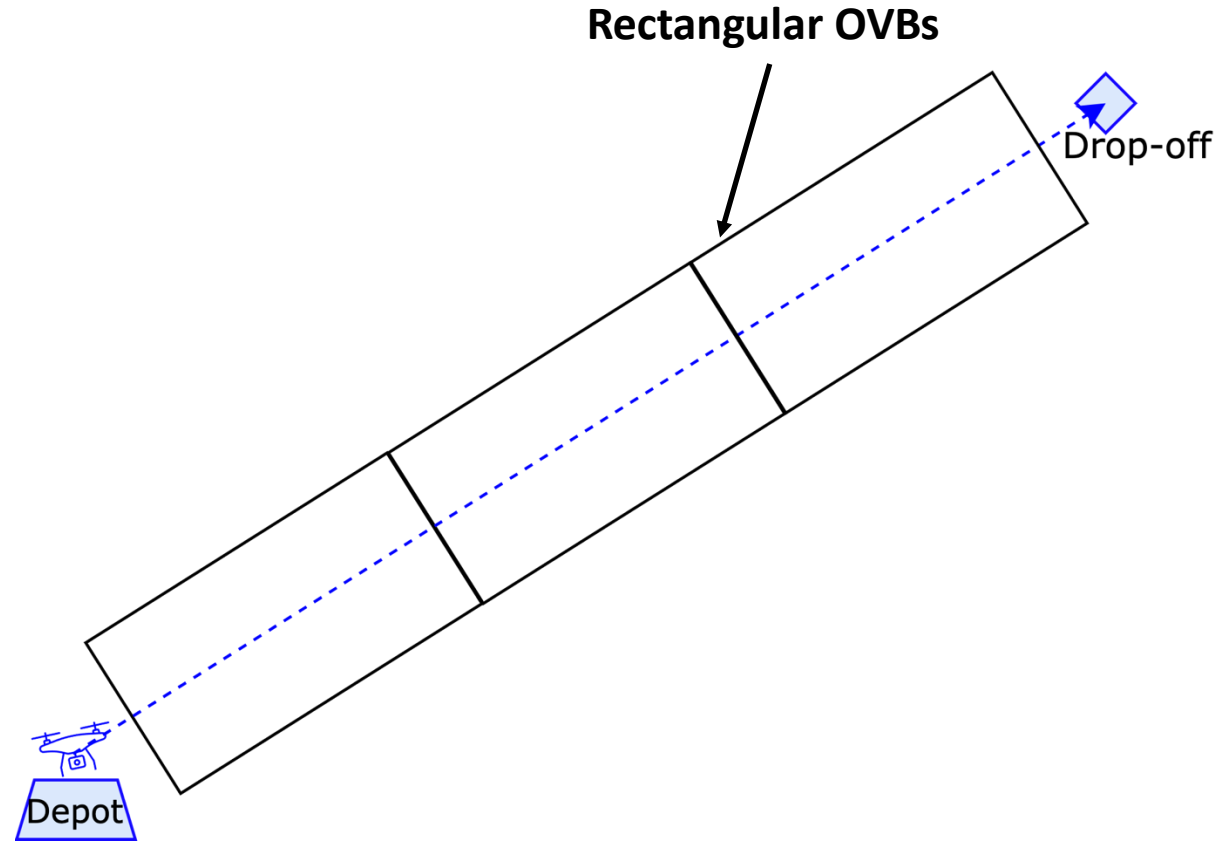
## 1. Paper research contributions

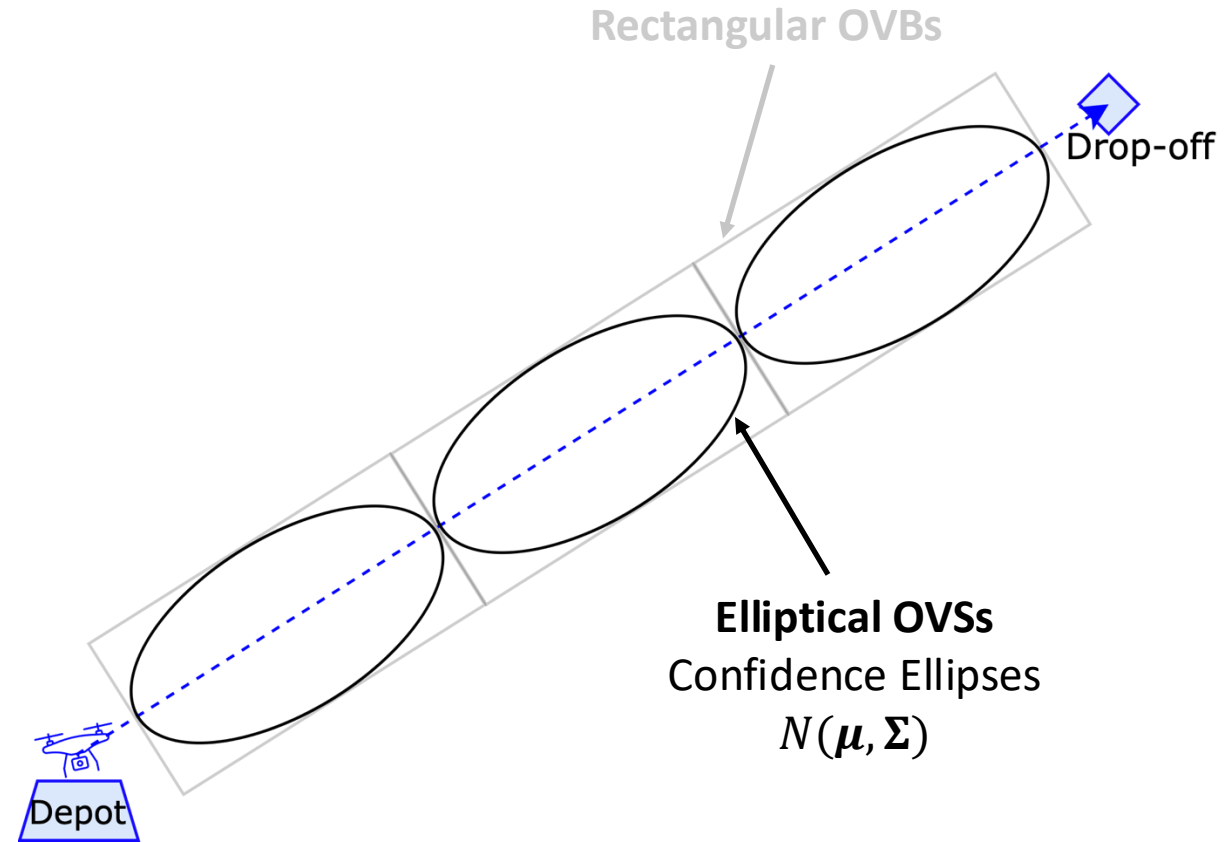
- Adaptive buffer 
- Strategic deconfliction approach 
- Safety metric 

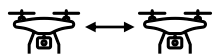
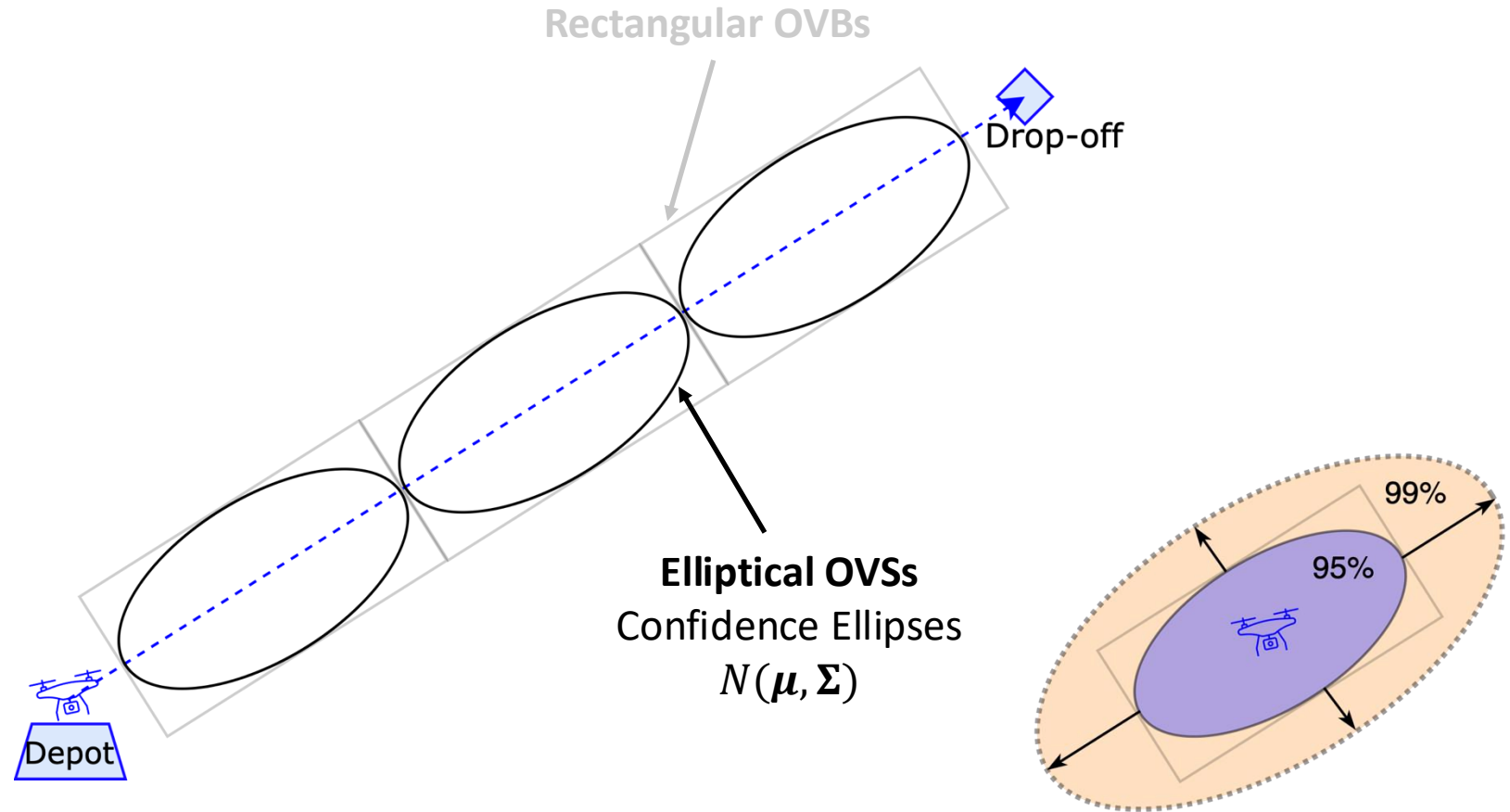
## 2. Simulation scenarios

## 3. Results

## 4. Conclusions



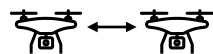
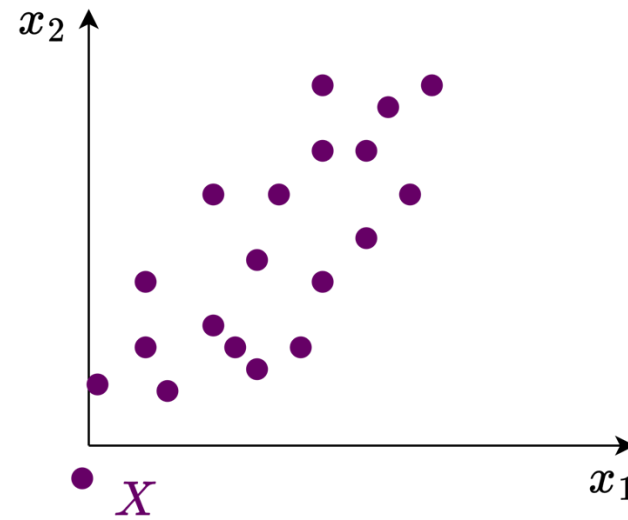




## Mahalanobis distance

Vehicle position samples

$$X \sim N(\mu, \Sigma)$$



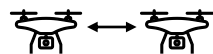
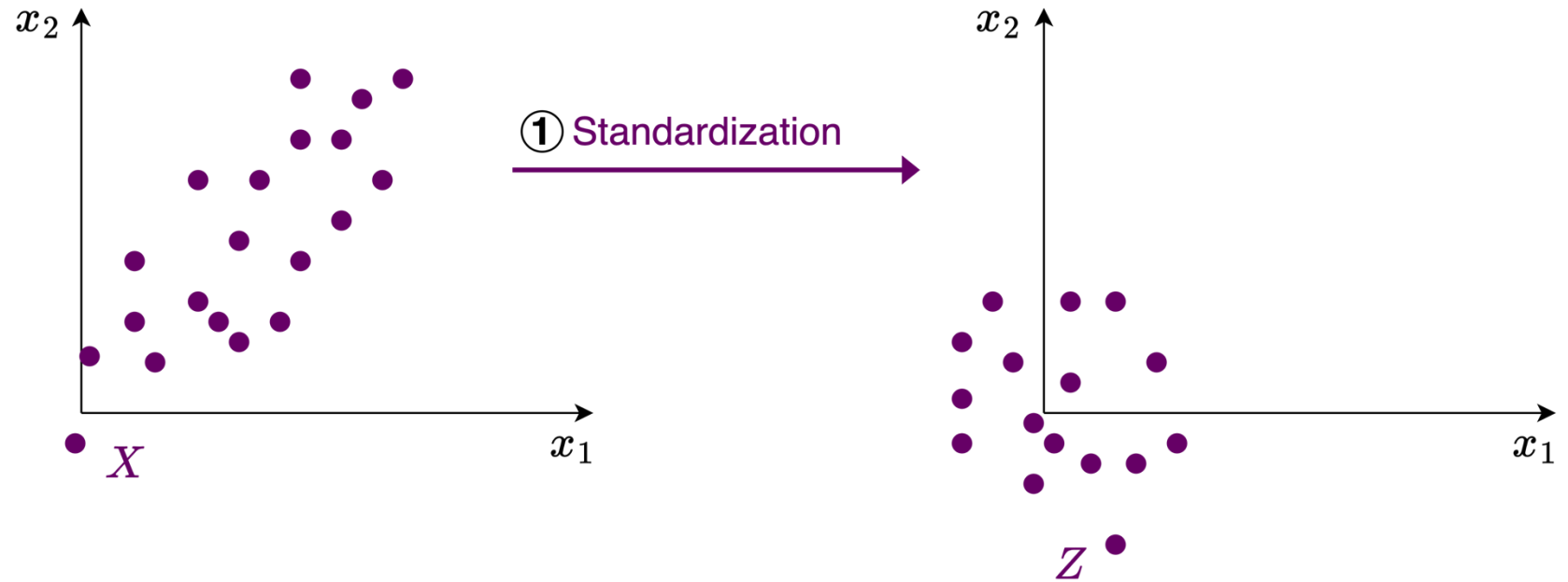
## Mahalanobis distance

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Standardized samples

$$Z = \Sigma^{-0.5} (X - \mu)$$



## Mahalanobis distance

Vehicle position samples

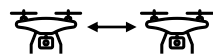
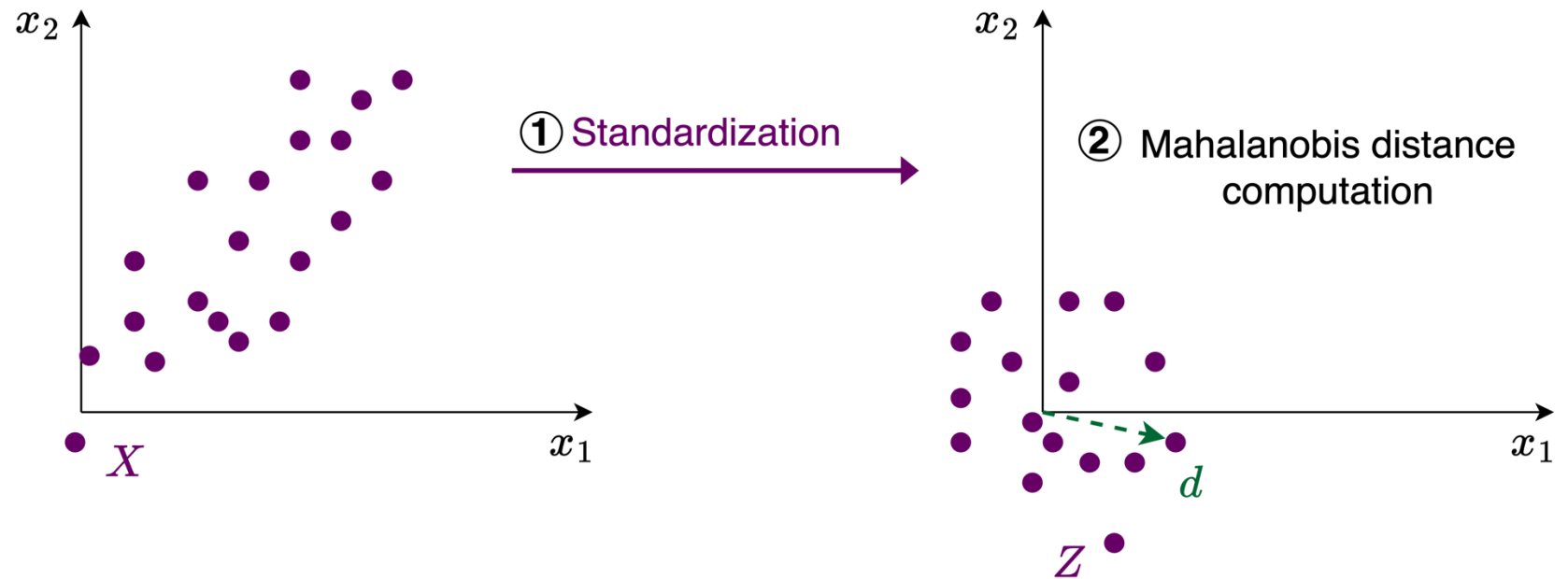
$$\mathbf{X} \sim N(\boldsymbol{\mu}, \boldsymbol{\Sigma})$$

Standardized samples

$$\mathbf{Z} = \boldsymbol{\Sigma}^{-0.5} (\mathbf{X} - \boldsymbol{\mu})$$

Chi-square distribution

$$\mathbf{z}^T \mathbf{z} = \sum_i^2 z_i^2 = d^2 \sim \chi_2^2$$



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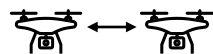
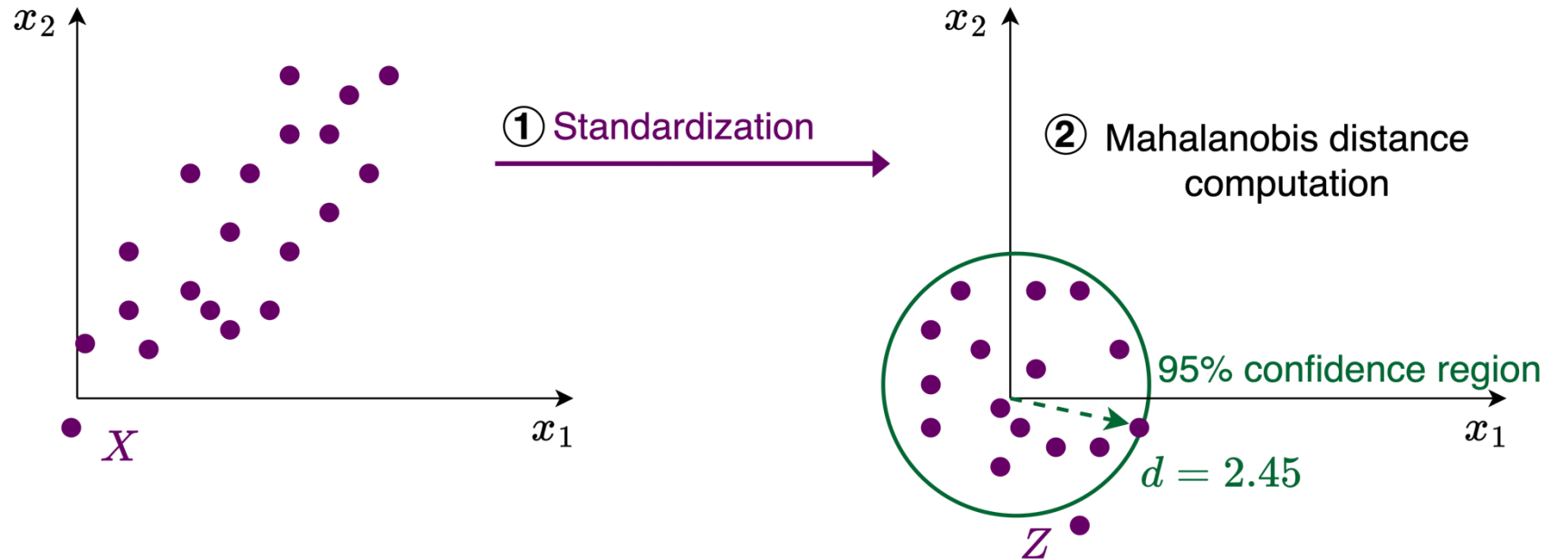
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Mahalanobis distance

$$d = \sqrt{(\chi_2^2)^{-1}(0.95)}$$



## Mahalanobis distance

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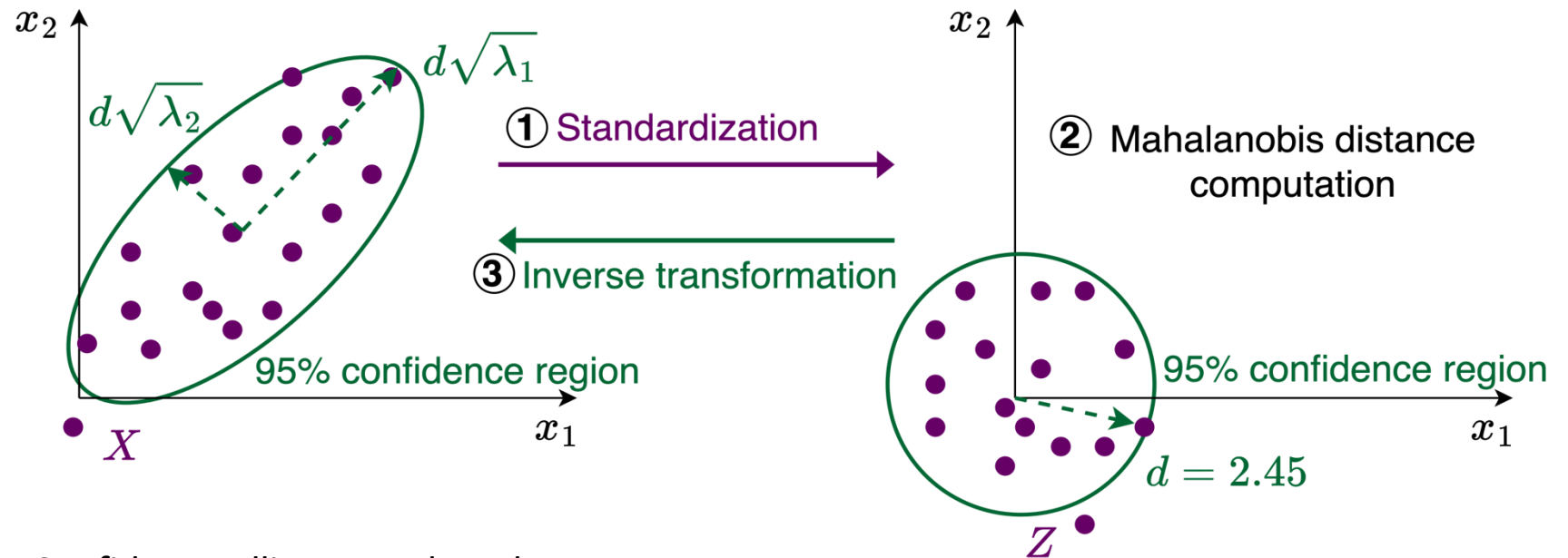
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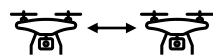
Mahalanobis distance

$$d = \sqrt{(\chi_2^2)^{-1}(0.95)}$$



Confidence ellipse axes lengths

$$\begin{bmatrix} a \\ b \end{bmatrix} = d\sqrt{\lambda}$$

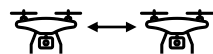
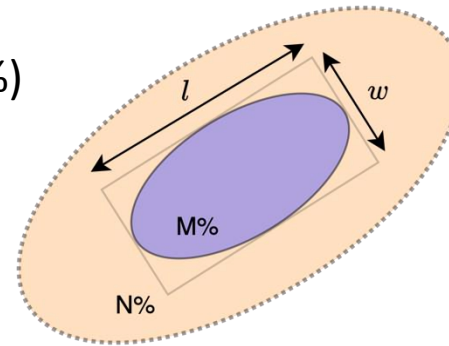


## Recipe for adaptive buffer (confidence ellipse expansion)

$M\%$ : initial confidence level from operator (e.g. 95%)

$N\%$ : target confidence level (e.g. 99%)

$l, w$ : rectangular OVB dimensions

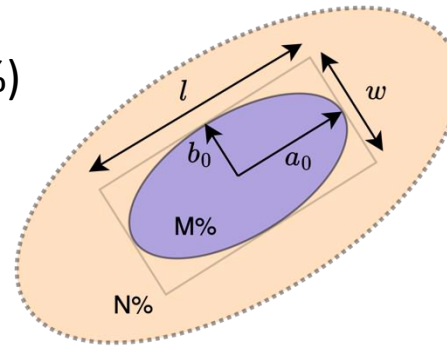


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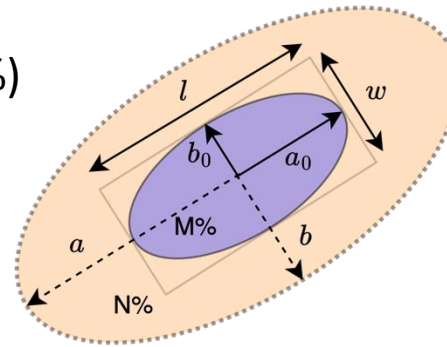
1. Compute axes lengths of OVB inscribed ellipse  $1. \begin{bmatrix} a_0 \\ b_0 \end{bmatrix} = \begin{bmatrix} l \\ w \end{bmatrix}$

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1. Compute axes lengths of OVB inscribed ellipse

$$1. \begin{bmatrix} a_0 \\ b_0 \end{bmatrix} = \begin{bmatrix} l \\ w \end{bmatrix}$$

2. Compute Mahalanobis distance of the ellipse

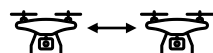
$$2. d = \sqrt{(\chi_2^2)^{-1} \left( \frac{N}{100} \right)}$$

3. Compute the Mahalanobis distance of the desired ellipse

$$3. d_0 = \sqrt{(\chi_2^2)^{-1} \left( \frac{M}{100} \right)}$$

4. Compute the axes lengths of expanded ellipse

$$4. \begin{bmatrix} a \\ b \end{bmatrix} = \frac{d}{d_0} \begin{bmatrix} a_0 \\ b_0 \end{bmatrix}$$



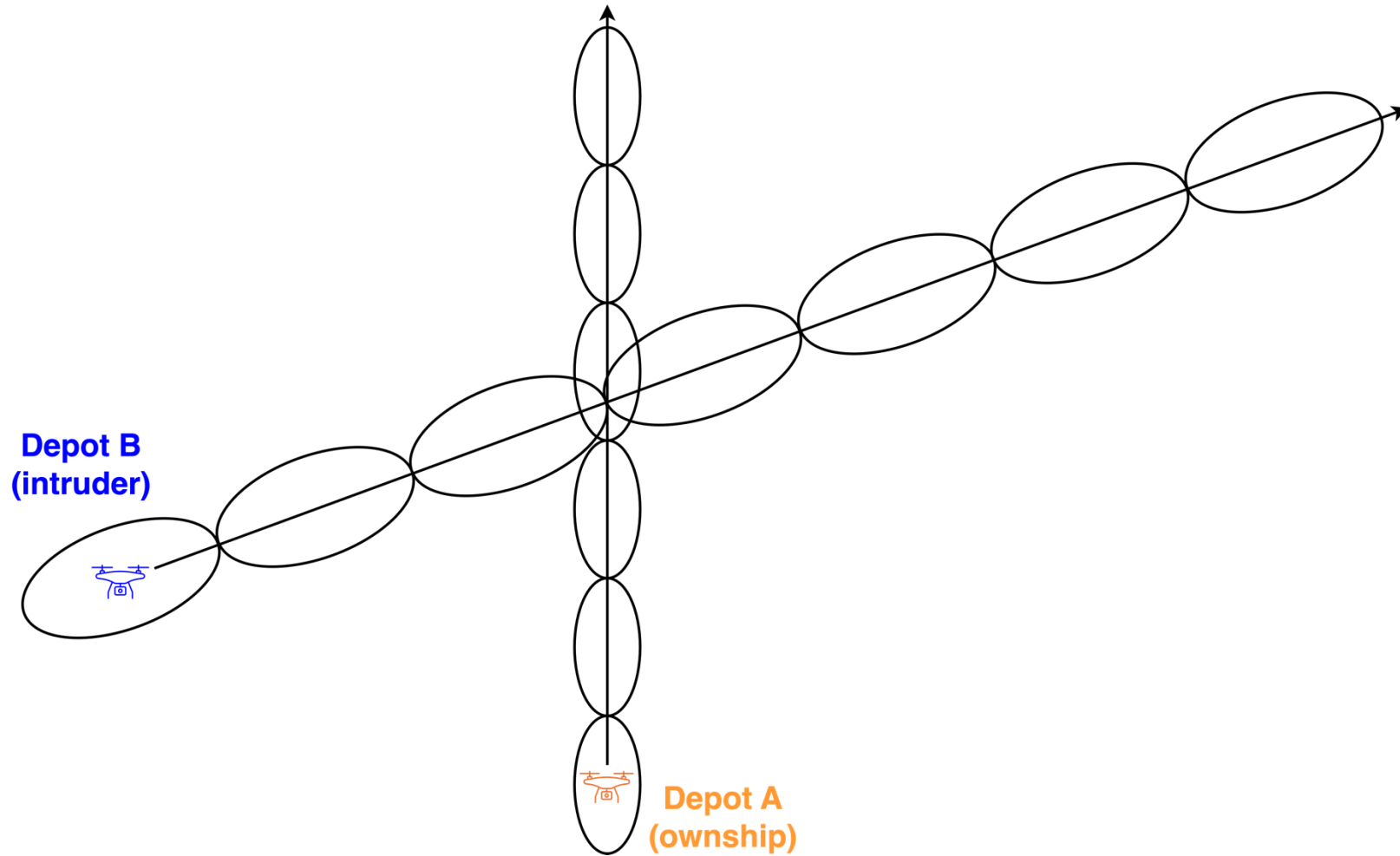
## Rolling Horizon with K-Position Search

- Mixed Integer Linear Programming formulation
- Objective function: minimize total ground delay
- Constraints:
  - Departure temporal separation
  - Crossing waypoints temporal separation

Modification: deconfliction of ellipses instead of rectangles

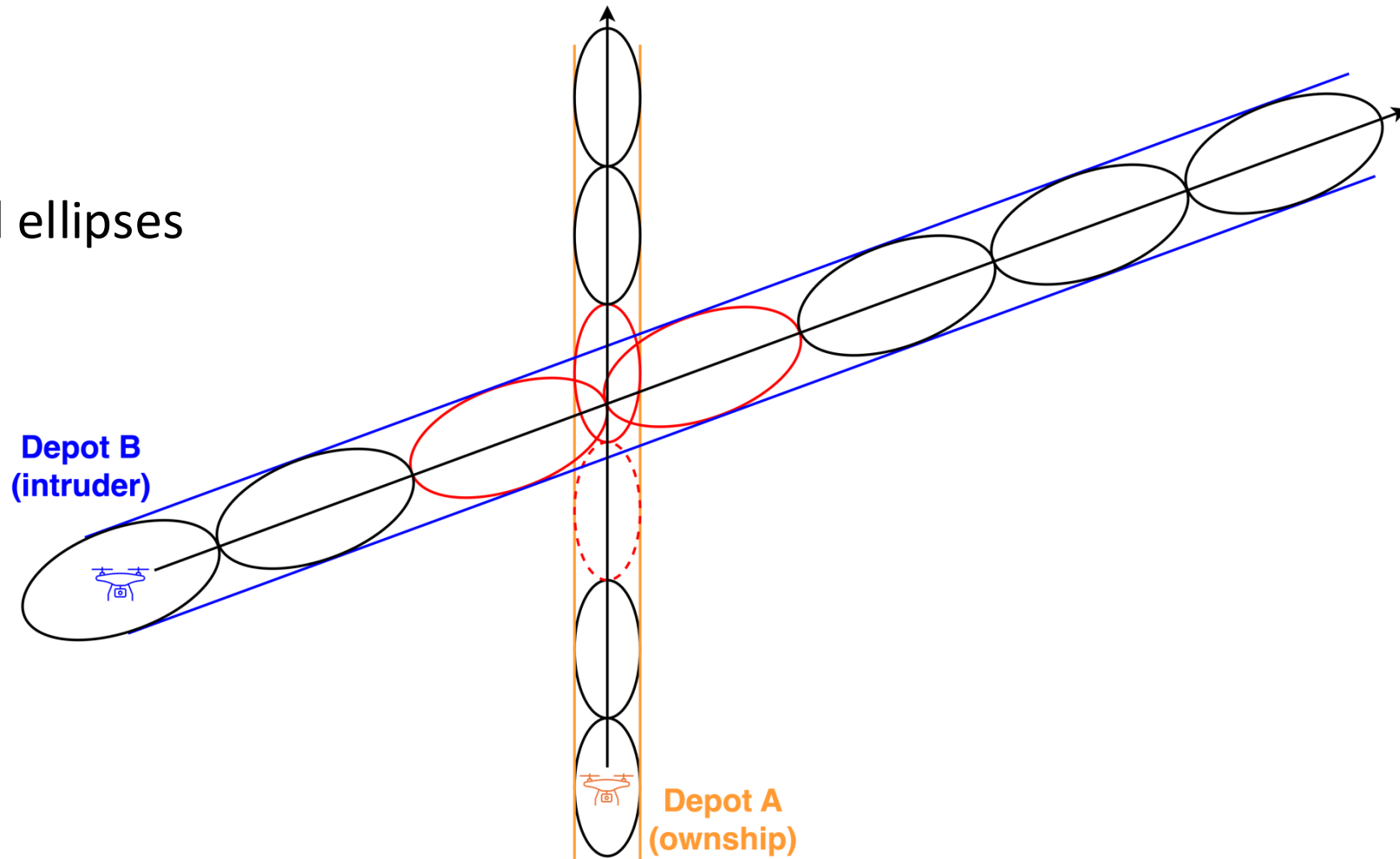


## SD modification



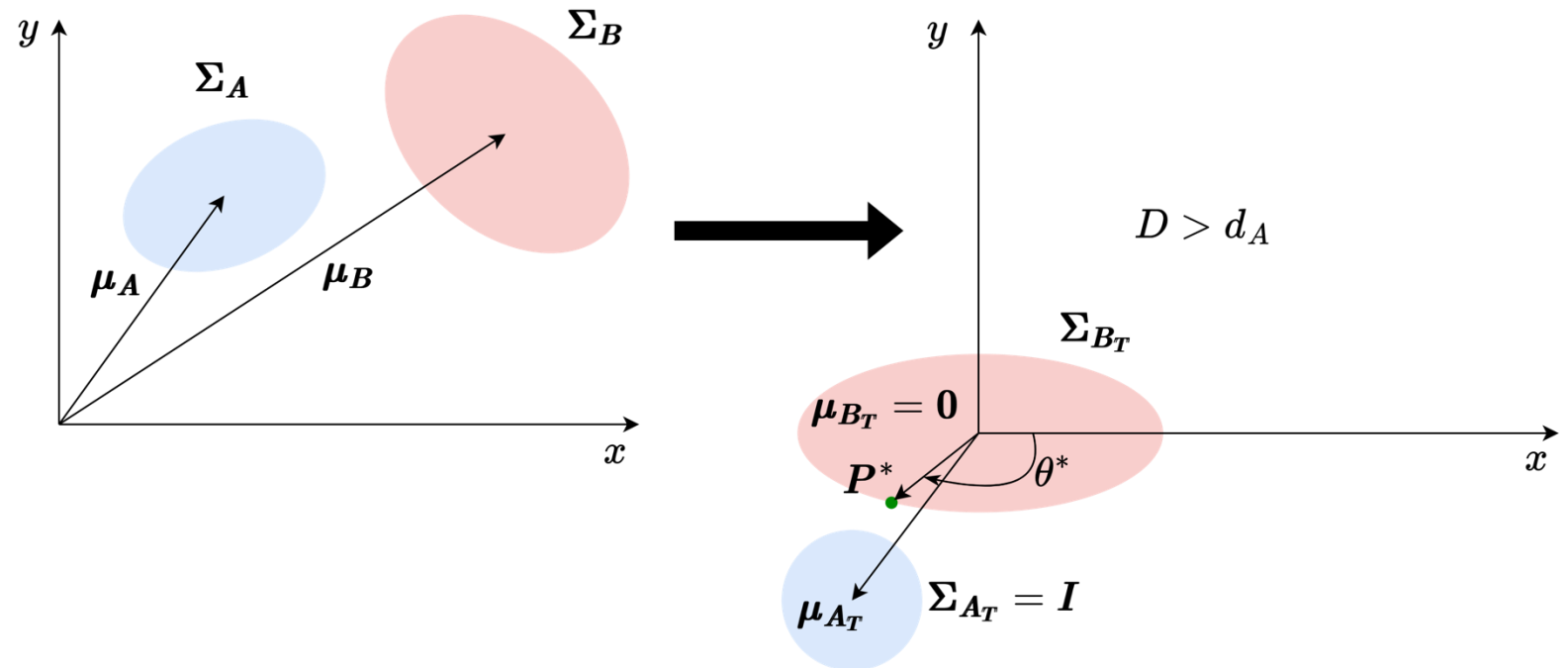
## SD modification

- Heuristic to discard ellipses



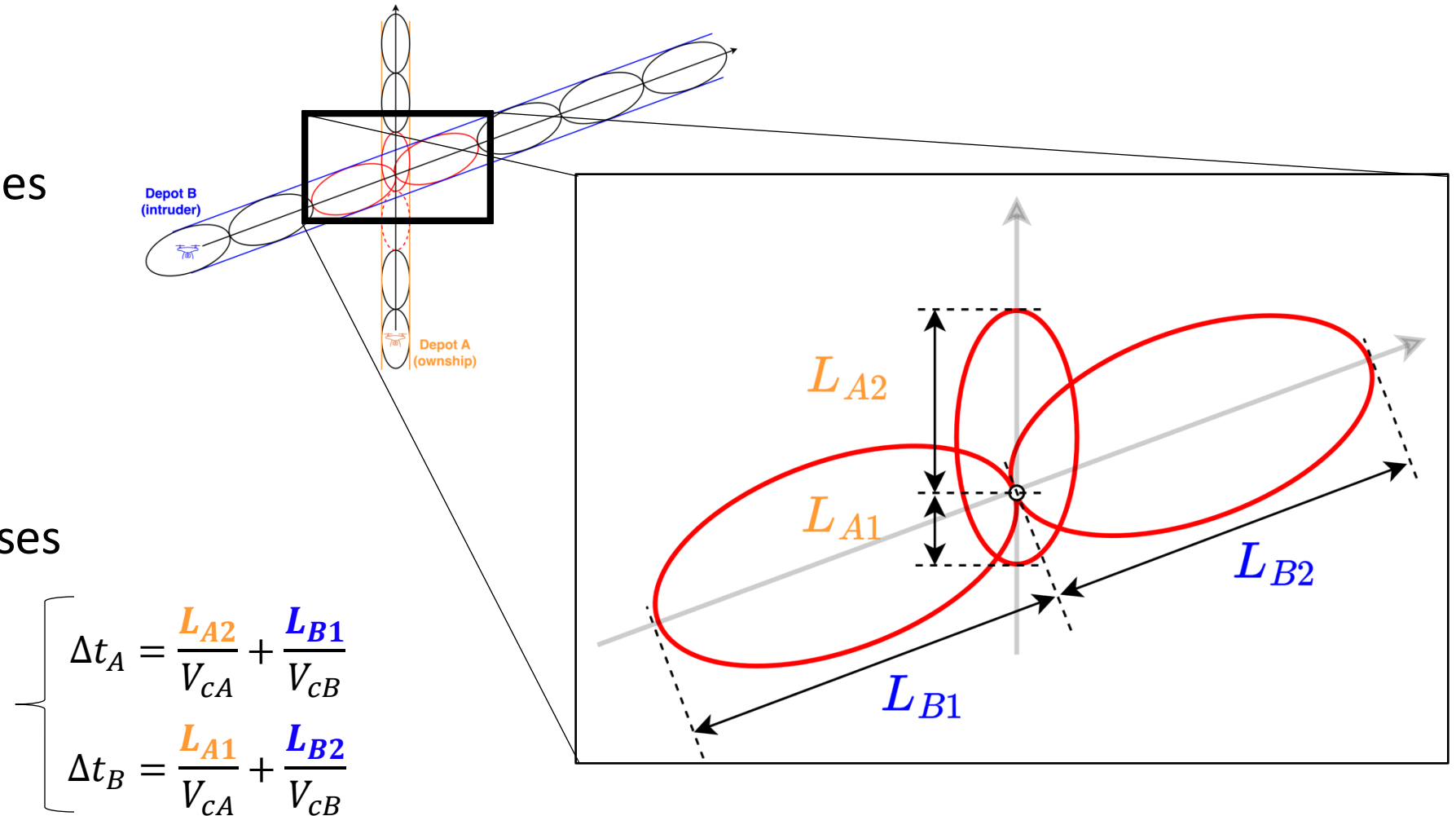
## SD modification

- Heuristic to discard ellipses
- Check intersection for candidate ellipses



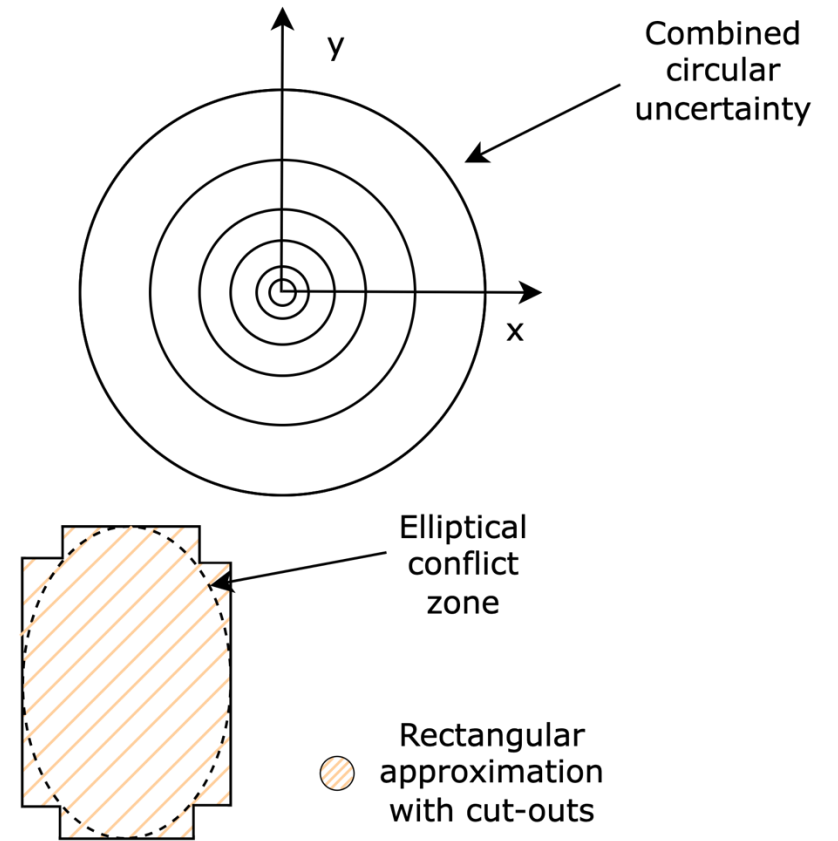
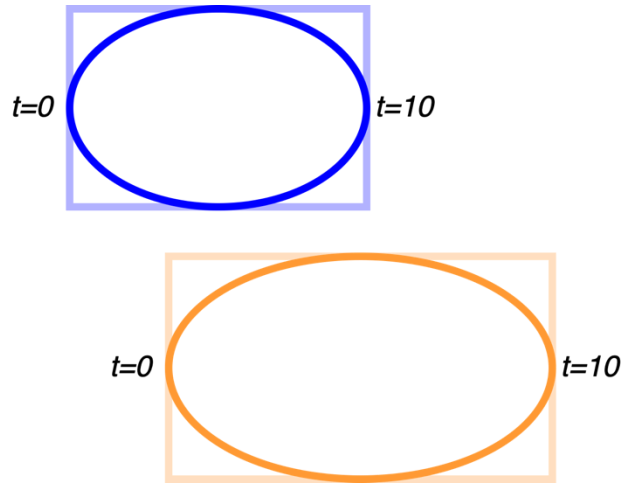
## SD modification

- Heuristic to discard ellipses
- Check intersection for candidate ellipses
- Compute temporal separation between ellipses at intersections



➤ **Hwang's method: 1-to-1 ellipse perfect temporal overlap**

Probability of NMAC between ellipses  
 $P_{\alpha i, \beta j}$

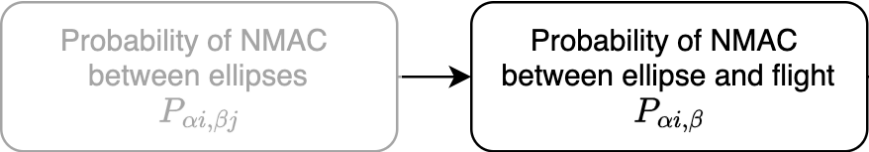


[2]

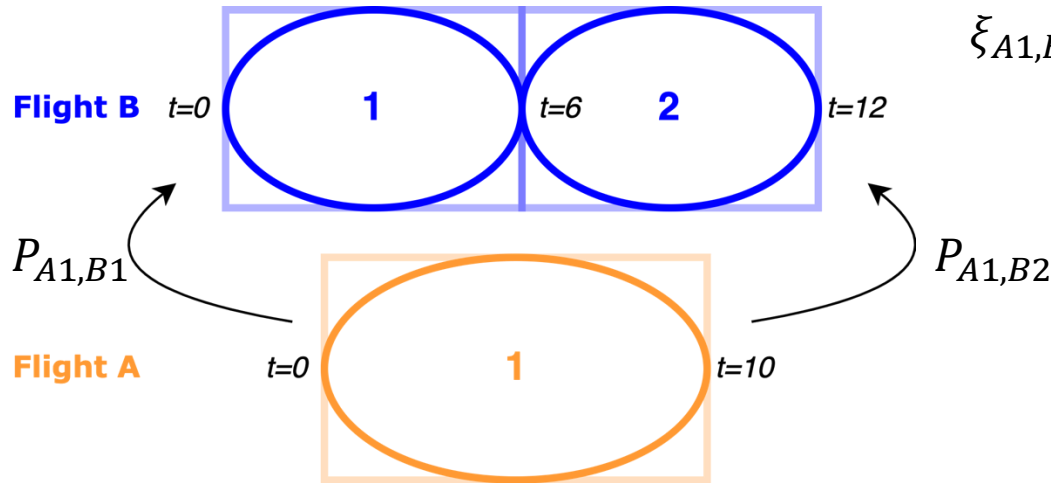


➤ **Hwang’s method:** 1-to-1 ellipse perfect temporal overlap

➤ **Time weighted average:** 1-to-many overlap



$$P_{A1,B} = \xi_{A1,B1} \cdot P_{A1,B1} + \xi_{A1,B2} \cdot P_{A1,B2} = \sum_{j=1}^{n_B} \xi_{A1,Bj} \cdot P_{A1,Bj}$$

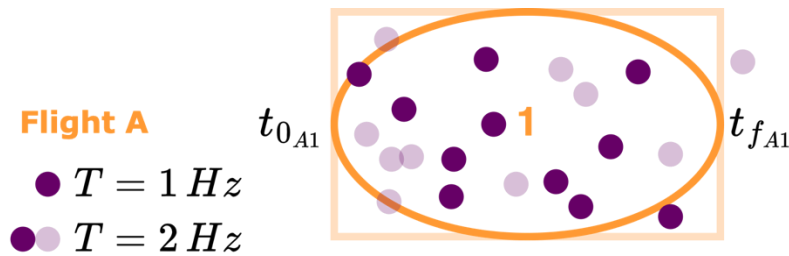
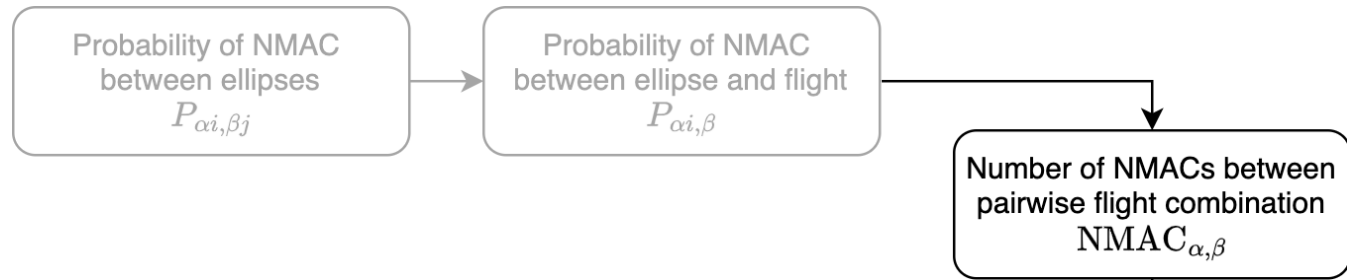


$$\xi_{A1,B1} = \frac{6}{10} \quad \xi_{A1,B2} = \frac{4}{10}$$

➤ **Hwang’s method:** 1-to-1 ellipse perfect temporal overlap

➤ **Time weighted average:** 1-to-many overlap

➤ **Sampling frequency:** conversion from probability to number of NMAC



$$NMAC_{AB} = \sum_{i=1}^{n_A} P_{A1,B} \cdot N_{Ai}$$

$$N_{Ai} = \frac{t_{f,Ai} - t_{0,Ai}}{T}$$

$$N_{A1} = \frac{10 - 0}{2} = 5$$

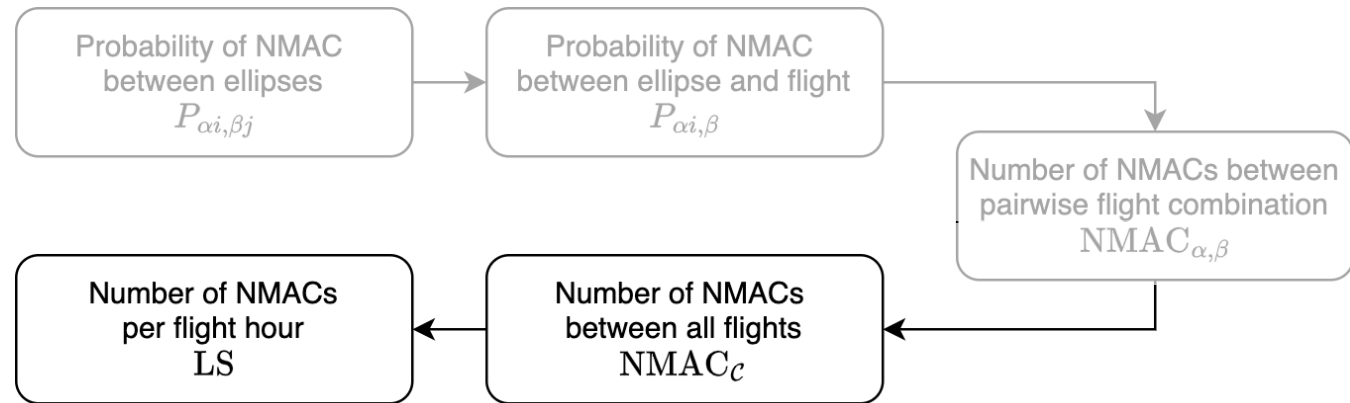
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➤ **# NMAC per flight hour:** Level of Safety (LS)

➤ **Average ground delay**



$$NMAC_{AB} = \sum_{i=1}^{n_A} P_{A1,B} \cdot N_{Ai}$$

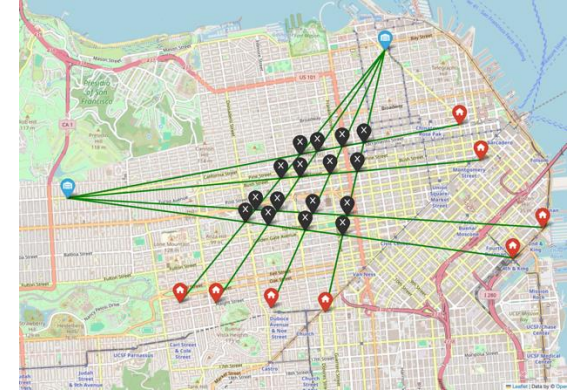
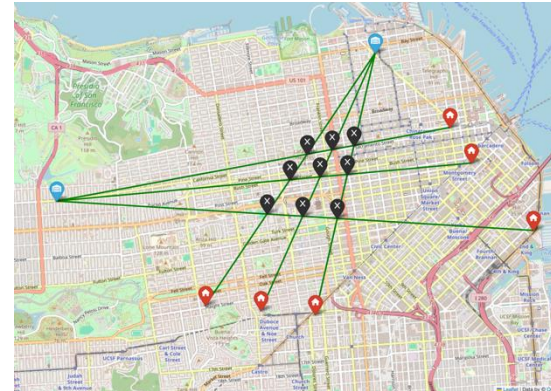
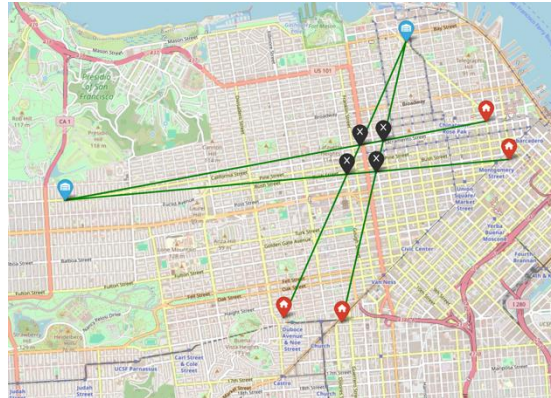
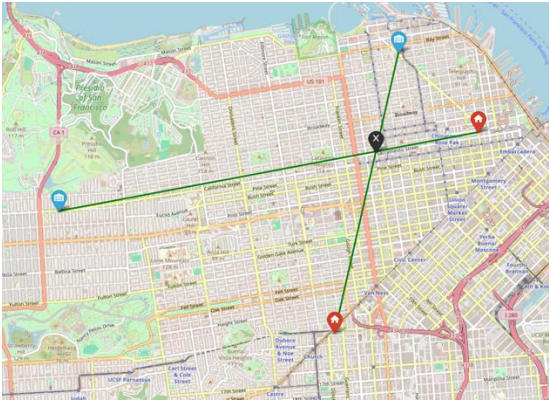
$$NMAC_C = \sum_{(f_A, f_B) \in C} NMAC_{A,B}$$

$$GD = \frac{1}{n_f} \sum_{f \in \mathcal{F}} Dep(f) - Req(f)$$

$$LS = \frac{NMAC_C}{\sum_{f \in \mathcal{F}} Arr(f) - Dep(f)}$$



## Increasing Complexity

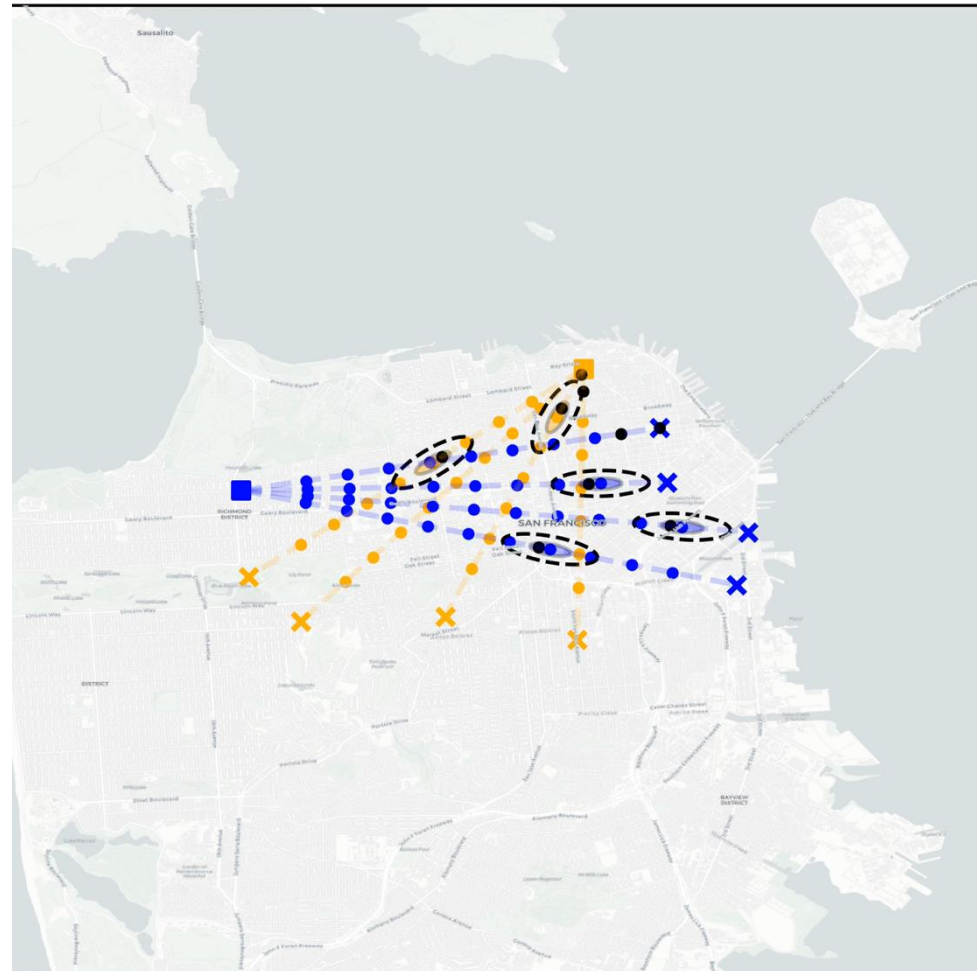


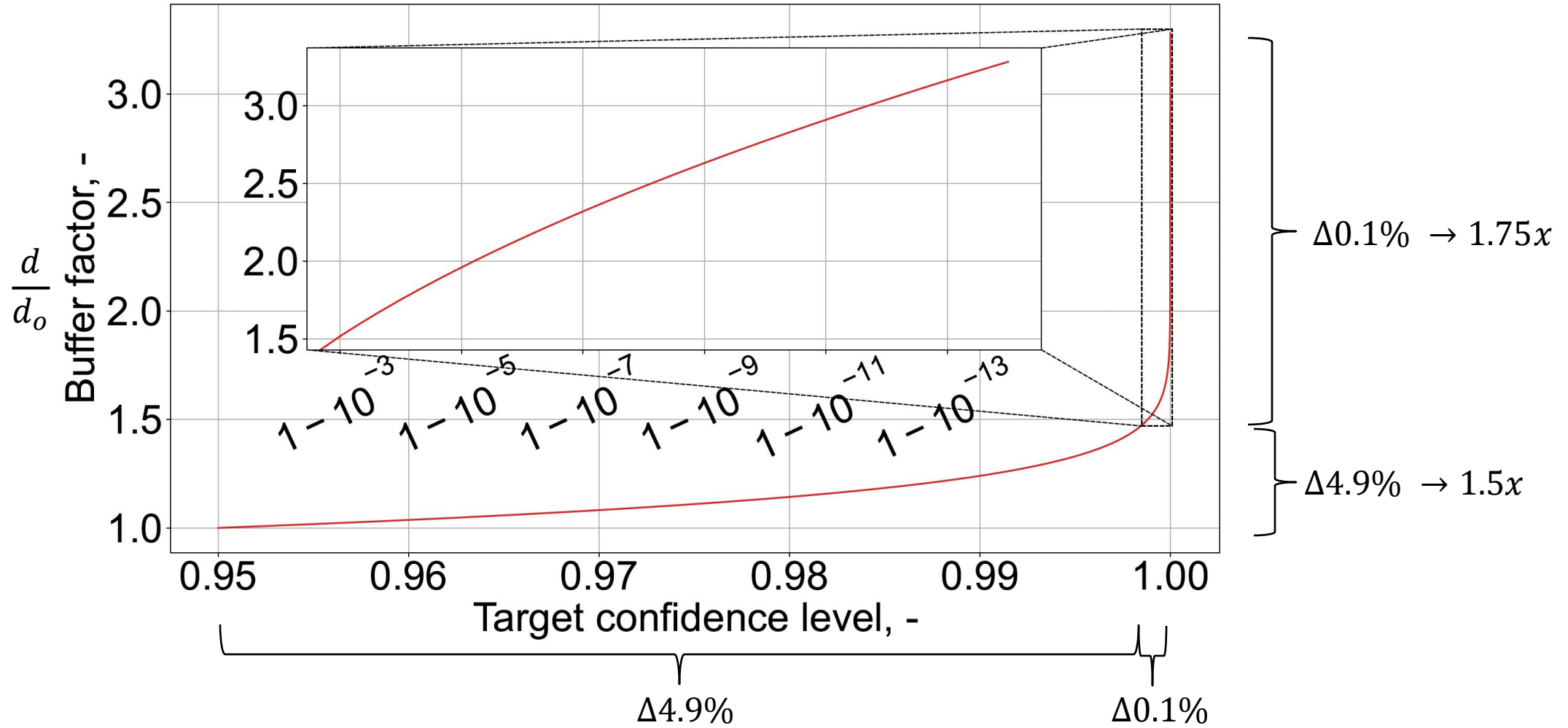
[3]

Parameter	Values	Parameter	Values
Number of flights per route	10 [-]	Loss of Separation radius (NMAC)	42.5 [m]
Poisson demand rate	$60^{-1}$ [s <sup>-1</sup> ]	OVB length and width	600 [m] and 200 [m]

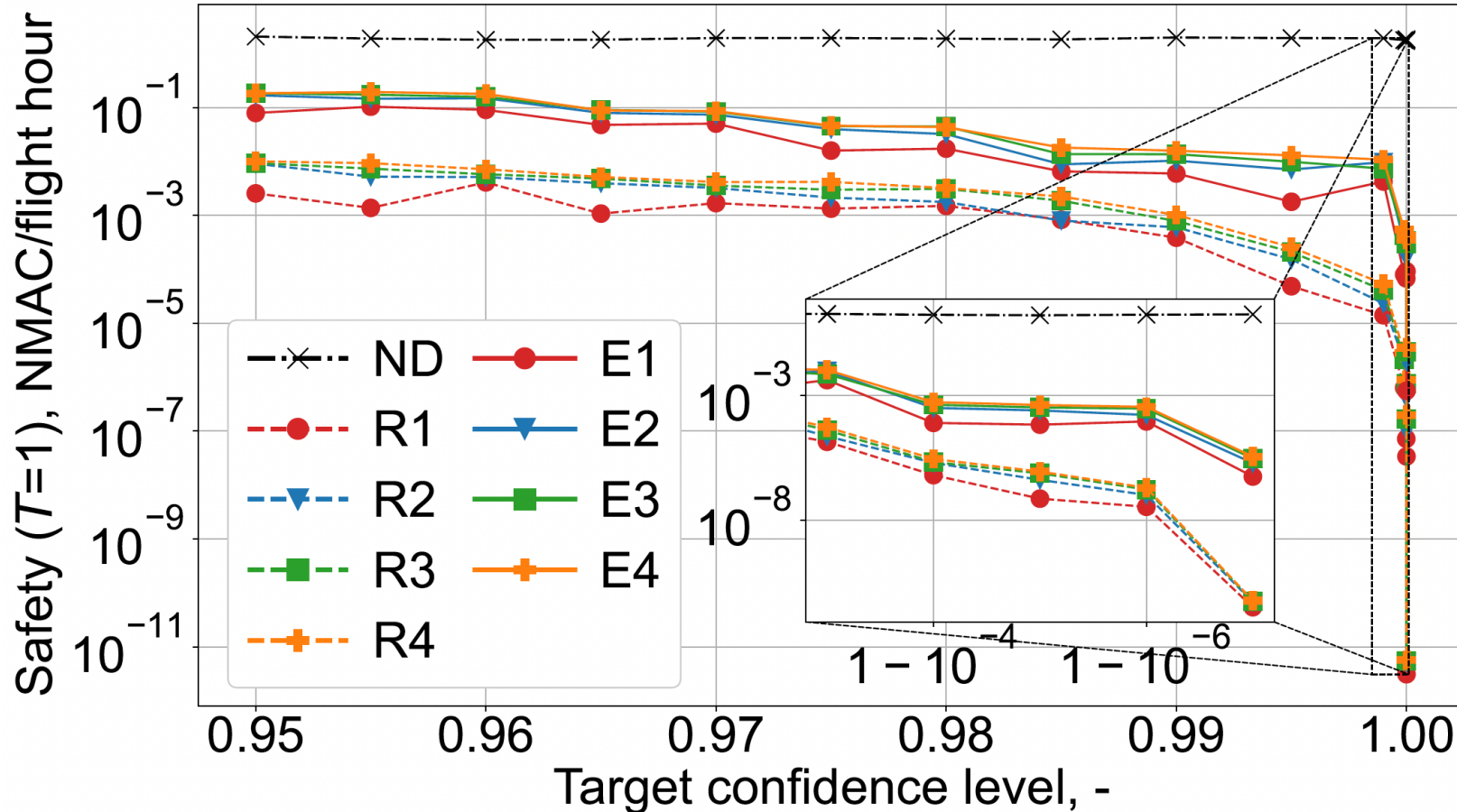


Time: 4625



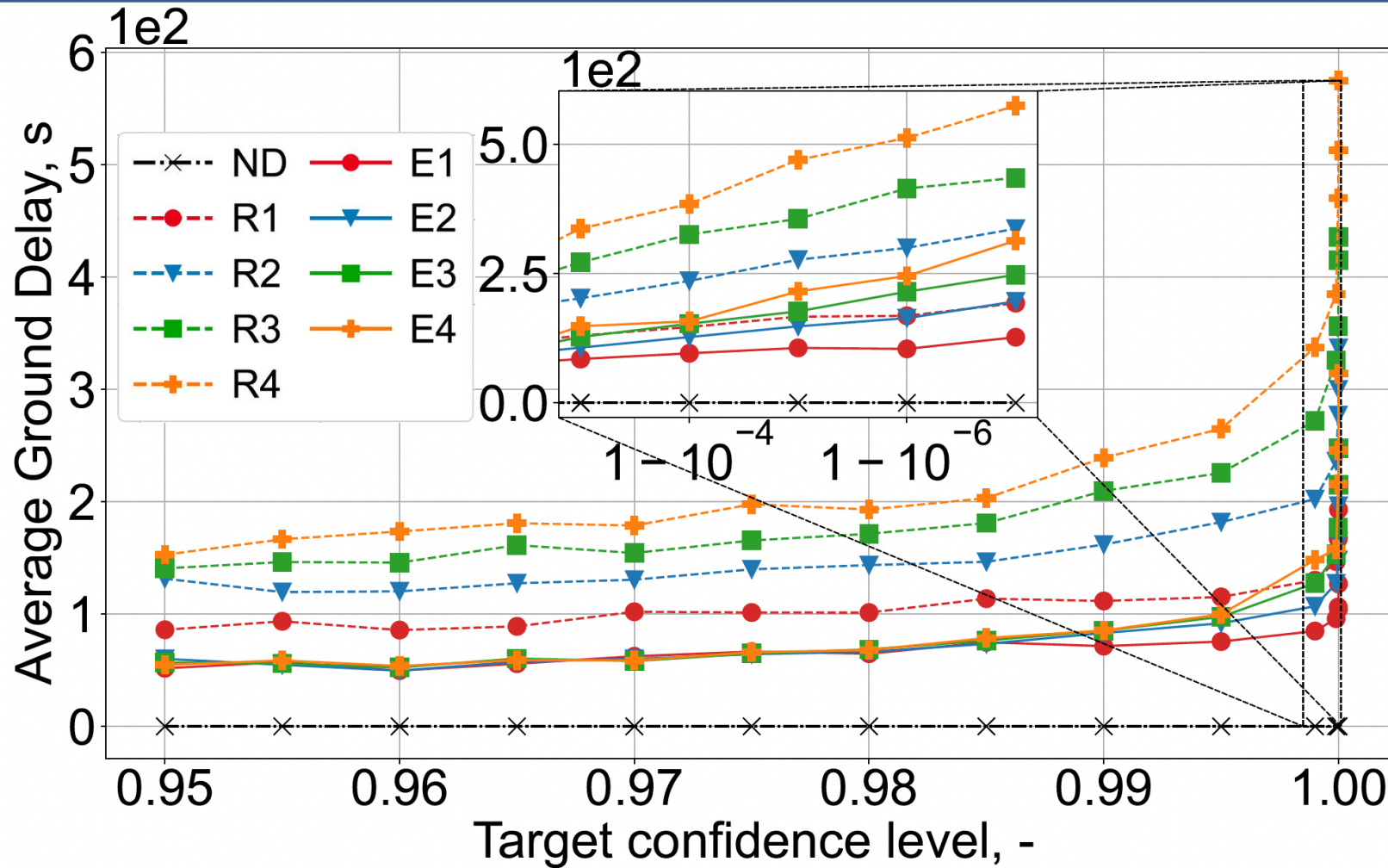


- 1. No deconfliction: highest risk
- 2. Complexity
- 3. Buffer
- 4. Geometry



Increasing Safety

1. No deconfliction: no delay
2. Complexity
3. Buffer
4. Geometry

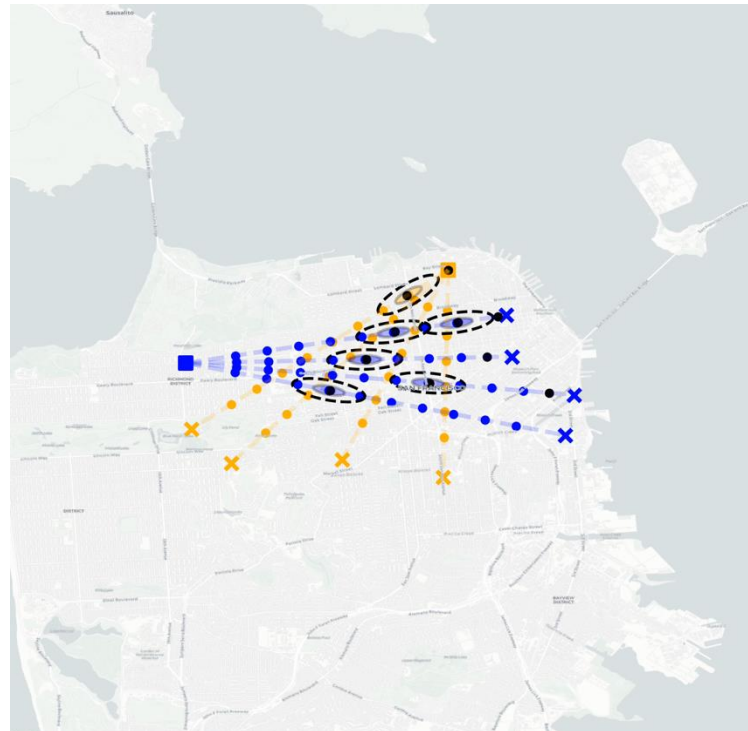


Increasing efficiency

- **Adaptive buffer** with **confidence level** as input
- New flavor of Rolling Horizon K-Position search strategic deconfliction approach with **accurate temporal separation** for **elliptical OVBs**
- **Safety metric** for risk estimation even without OVB overlap
- Outlook: safety metric + TLS → confidence level requirement definition



# Thanks for your attention!



**José Ignacio de Alvear Cárdenas**  
[joseignacio.dealvearcardenas@nasa.gov](mailto:joseignacio.dealvearcardenas@nasa.gov)



- [1] NASA UTM technical interchange meeting 2021
- [2] Hwang, I., and Seah, C. E., “Intent-Based Probabilistic Conflict Detection for the Next Generation Air Transportation System,” Proceedings of the IEEE, Vol. 96, No. 12, 2008, pp. 2040–2059. doi:10.1109/JPROC.2008.2006138.
- [3] Pradeep, P., Yarramreddy, G. S., Amirsoleimani, N., Munishkin, A. A., Morris, R. A., Xue, M., Kalyanam, K. M., and Chour, K., “Rolling Horizon with K-Position Search Method for Strategic Deconfliction of Package Delivery UAS,” AIAA AVIATION Forum and ASCEND 2024, 2024. <https://doi.org/10.2514/6.2024-4456>.