

DEPARTMENT OF INDUSTRIAL SYSTEMS ENGINEERING AND PRODUCT DESIGN

Geozone-Aware Unmanned Aerial Vehicles (UAV) Path Planning Using RRT* and Jellyfish-Inspired Optimization for Urban Air Mobility (UAM)

Authors: Judit Salvans Baucells, Elham Fakhraian, Ivana Semanjski

Presenter: Judit Salvans Baucells

Master's Student in Industrial Engineering and Operations Research, Ghent University

Judit Salvans Baucells



Judit Salvans Baucells is a master's student in Industrial Engineering at Universitat Politècnica de Catalunya (UPC), currently completing this paper at Ghent University through the Erasmus+ program. Her academic interests include operations research, autonomous systems, and sustainable urban logistics.



Her thesis focuses on geozone-aware UAV path planning for Urban Air Mobility (UAM), integrating RRT* algorithms with bio-inspired optimization. She also has hands-on experience in process optimization, mechanical system design, and regulatory-compliant engineering.

OVERVIEW



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APPLICATIONS

Motivation



- Agriculture
- Search and Rescue

- Delivery Services
- Weather



- Logistics
- Emergency Management

UAV OPERATIONS

Autonomus Navigation

Sensors
Cameras
Radars
...



Real-time Data

Path Planning

Lenght
Security
Viability
...



Optimal Path

Motivation



REGULATIONS

Motivation



Technologies for drones and Vertical Take-Off and Landing (VTOL) aircraft are evolving all the time. As drones and VTOLs are becoming a fixture in all our lives, we at **EASA set up rules for flying drones and VTOLs in Europe**, to define how they are designed, manufactured and operated. We will **also set up rules and standards for a specific type of airspace called U-Space, which will enable drones & VTOLs to fly safely across European skies**. In short, we are enabling innovative air mobility with drones and VTOLs.

OBJECTIVES

First Goal

Integrate geozone constraints directly into the path planning process.

Second Goal

Optimization process to improve paths quality.

Third Goal

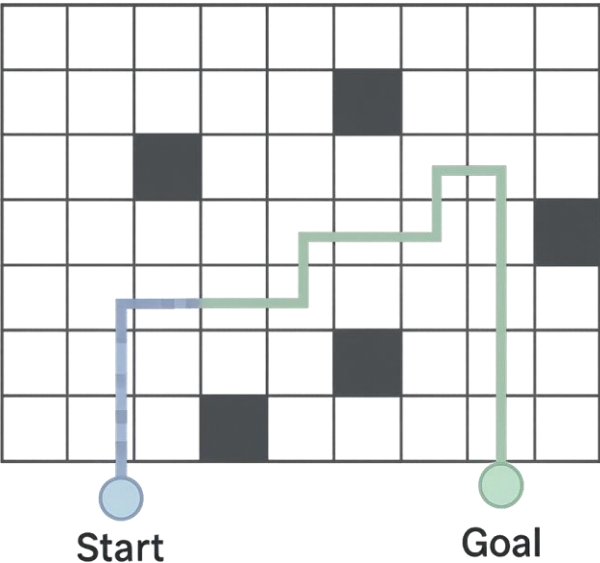
Test the method in real urban environment.

Fourth Goal

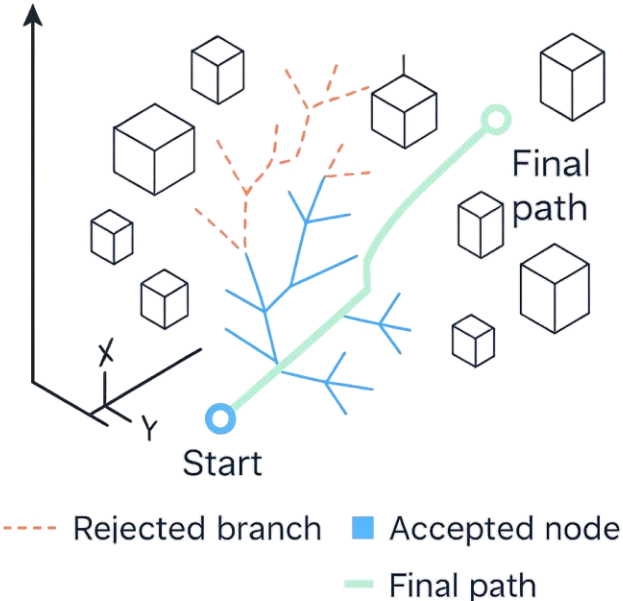
Support future U-space integrations.

LITERARY REVIEW

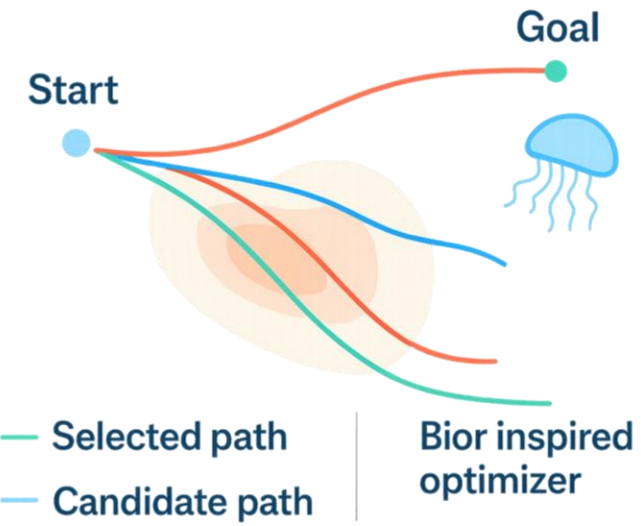
Graph-Based Algorithms: Dijkstra and A*

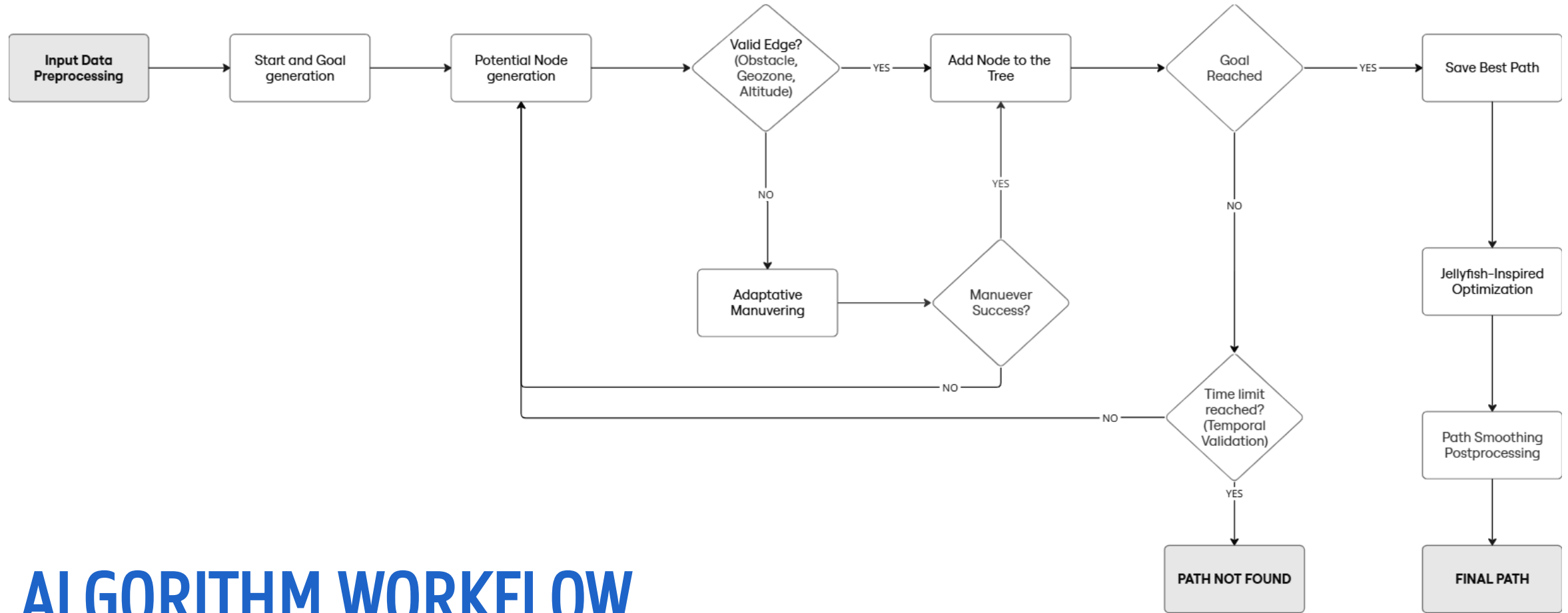


Sampling-Based Algorithms: RRT and RRT*



Metaheuristic and Bio-Inspired Algorithms





ALGORITHM WORKFLOW

IMPLEMENTATION



Start & Goal

- <80m height
- Ground Level
- Rooftop Point



Node Candidate

- 30% Goal Point
- 60% Existing Nodes Surroundings
- 10% Random



RRT* Tree

- Collision Free
- Between height limits
- Outside Geozones



Adaptive Maneuver

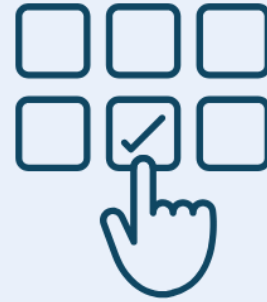
- Take off maneuver
- Landing maneuver
- Fly over an obstacle

IMPLEMENTATION



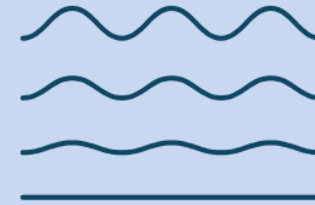
Path Found

- Goal Reached
- Time limit NOT reached



Jellyfish-Inspired Optimization

- Path Length
- Threat Cost
- Smoothness



Smoothing Process

- Sharp Turns
- Abrupt Direction Changes
- Respect Viability

RRT* Tree Constraints

$$isValid(e) = \begin{cases} 1, & \text{if } e \cap O = \emptyset, e \cap G = \emptyset, \quad h(e) \in [h, H] \\ 0, & \text{otherwise} \end{cases}$$

$e \rightarrow$ candidate edge,

$O \rightarrow$ set of obstacle volumes

$G \rightarrow$ restricted geozones

$[h, H] \rightarrow$ edge's altitude range.

Jellyfish-Inspired Optimization

$$Score_i = \sum_{k=1}^3 w_k \cdot normalized_{k,i}$$

$w = [w_1, w_2, w_3] \rightarrow$ randomly sampled weight vector

$$\sum_{k=1}^3 w_k = 1$$

CASE STUDY

Piombino, Italy



DJI Matrice 300 RTK

WHAT TYPE OF DRONE CAN I FLY?							
Operation			Drone Operator / pilot				
C-Class	Max Take off mass	Subcategory	Operational restrictions	Drone Operator registration?	Remote pilot qualifications	Remote pilot minimum age	
Privately build	<250g	A1 Not over assemblies of people (can also fly in subcategory A3)	Operational restrictions on the drone's use apply (follow the QR code below)	Yes No if toy or not fitted with camera/sensor	Read user's manual	No minimum age (certain conditions apply)	
legacy < 250g							
C0							
C1	<900g						
C2	<4kg	A2 Fly close to people (can also fly in subcategory A3)			Yes	Check out the QR code below for the necessary qualifications to fly these drones	16
C3	<25kg	A3 Fly far from people					
C4							
Privately build							
Legacy drones (art 20)							



For more details go to <https://www.easa.europa.eu/domains>





TERRAIN & OBSTACLES DATA

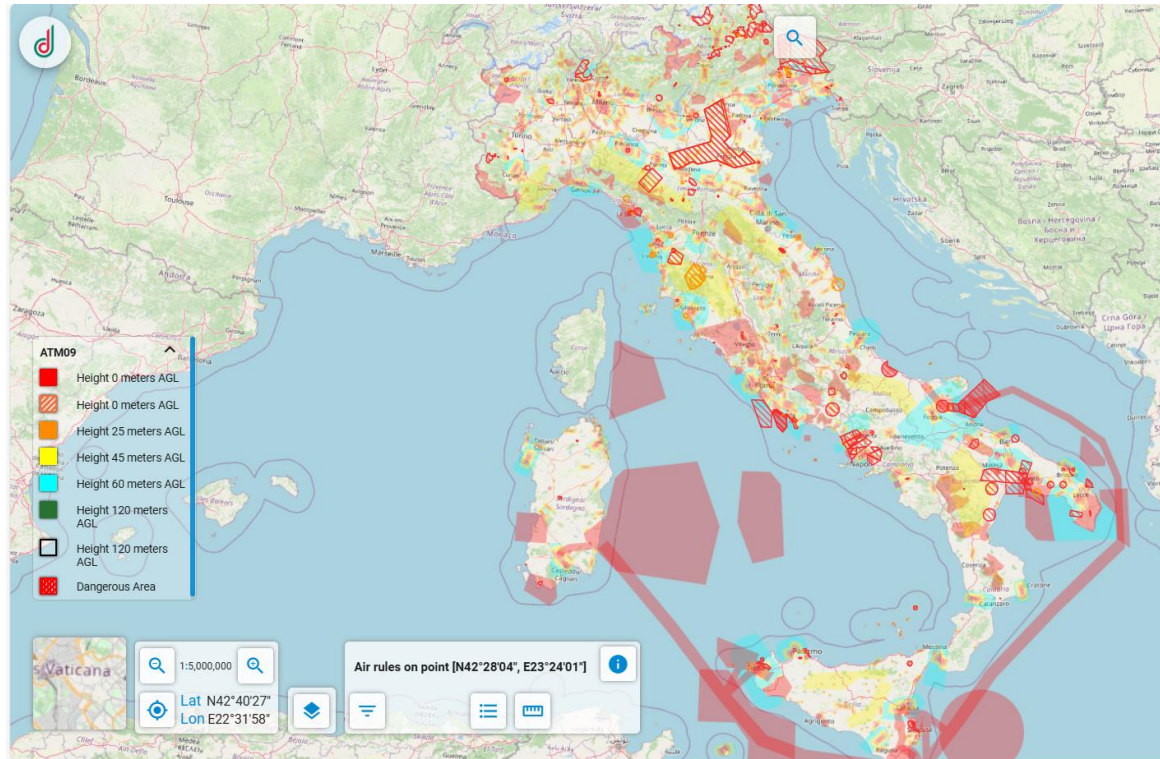
```
{'NAME': 'sisw-test-area-piombino',  
'CLOSED': 'YES',  
'PERIMETER': '4.401 km',  
'ENCLOSED_AREA': '0.94 sq km',  
'coordinates': [(10.536867583333333,  
42.939028638888885), (10.5340785,  
42.93482538888889), (10.528657500000001,  
42.93246844444444),  
...,  
(10.536867583333333,  
42.939028638888885)]}
```

Input Data

```
{{ 1: {'CLOSED': 'YES', 'suolo_id':  
'RT0201010300340938',  
..., 'coordinates':  
↪ [(10.53534725, 42.93442638888889,  
40.19), (10.535718833333334,  
42.93507280555555, 40.19)  
...]}},  
13: {'CLOSED': 'YES', 'suolo_id':  
'RT0201010300341633',  
..., 'coordinates':  
↪ [(10.533123777777778,  
42.93253788888889,  
37.74), (10.533304222222222,  
42.93134927777778, 37.74)  
... ]}}
```

GEOZONES DATA

Input Data



{'title': 'ITA ZoneVersion 31/03/2025 07:25:02', 'description':

↪ 'ITA ZoneVersion - GeoZones[4317] - ATM09[4013]/NFZ[304]/NOTAM[0]',

↪ 'features': {'A557281': { 'identifier': 'A557281',

.....

'geometry': [{

'lowerLimit': 45,

'upperLimit': 120,

'horizontalProjection': {'type': 'Polygon',

'coordinates': [[[10.5552691, 42.942228],

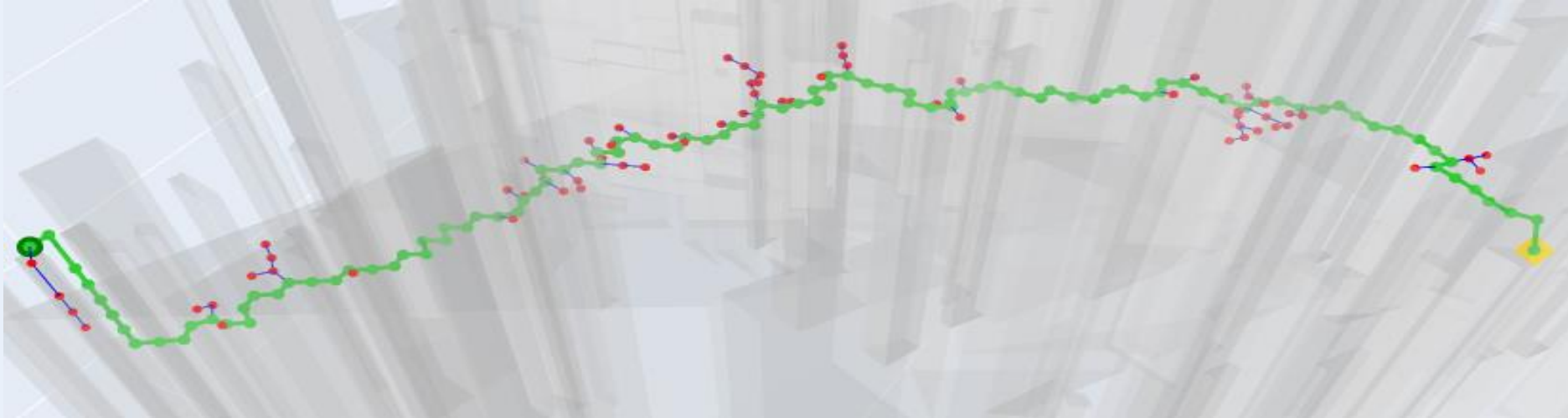
...,

↪ [10.5430489, 42.941751]]]

}}]}}

EXPERIMENTS

A	(10.5375 , 42.9361, 30.9) (10.5313, 42.931, 22.7)	G	(10.5332, 42.9303, 32.1) (10.5384, 42.9339, 0.5)	L	(10.5335, 42.9289, 40.9) (10.5296, 42.9288, 26.1)
B	(10.535, 42.9307, 25.2) (10.5335, 42.9295, 0.5)	H	(10.5398, 42.9298, 0.5) (10.5314, 42.9289, 34.4)	M	(10.5339, 42.9343, 0.5) (10.5399, 42.9261, 0.5)
C	(10.5326, 42.933, 27.7) (10.5301, 42.932, 22.9)	I	(10.5363, 42.9293, 21.5) (10.5336, 42.9349, 28.4)	N	(10.5320, 42.9301, 0.5) (10.5334, 42.9303, 32.1)
D	(10.536, 42.9308, 29.5) (10.5377, 42.9376, 20.5)	J	(10.5381, 42.9345, 30.6) (10.5327, 42.9329, 0.5)	O	(10.5375, 42.9361, 30.9) (10.5377, 42.9296, 28.1)
E	(10.536, 42.9308, 29.5) (10.5377, 42.9376, 20.5)	K	(10.5353, 42.9292, 0.5) (10.5333, 42.9320, 0.5)	P	(10.5358, 42.9369, 23.3) (10.5374, 42.9308, 25.7)
F	(10.536, 42.9308, 29.5) (10.5377, 42.9376, 20.5)				

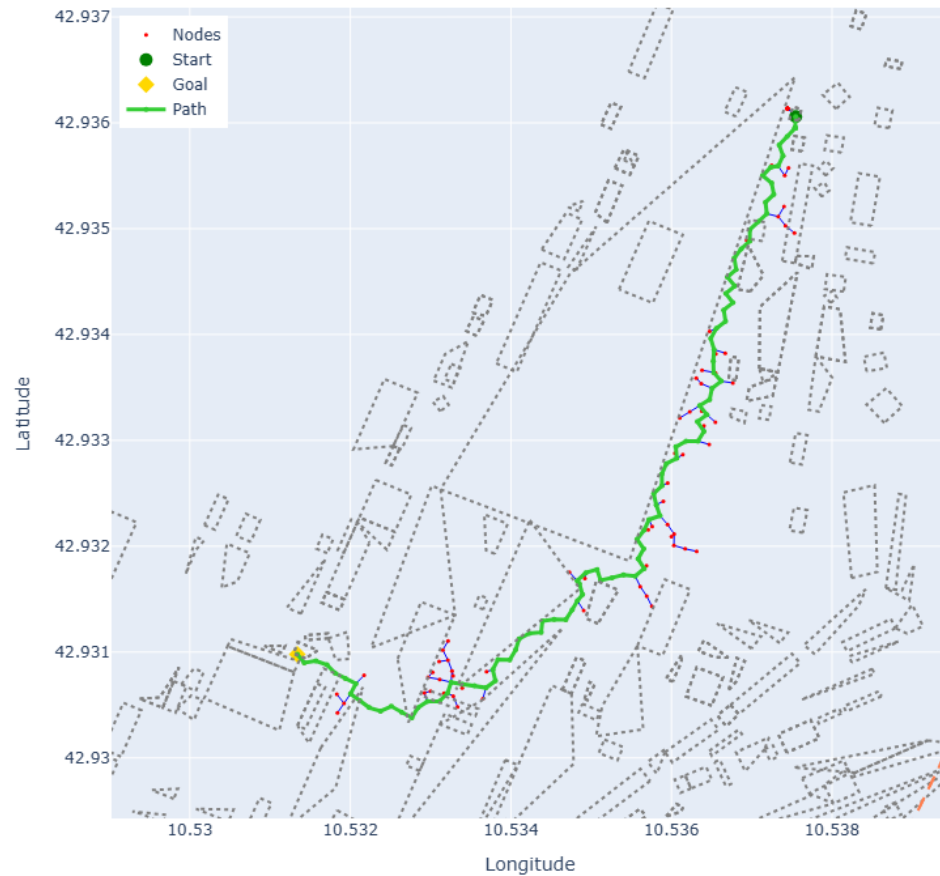


STEP - SIZE ANALYSIS

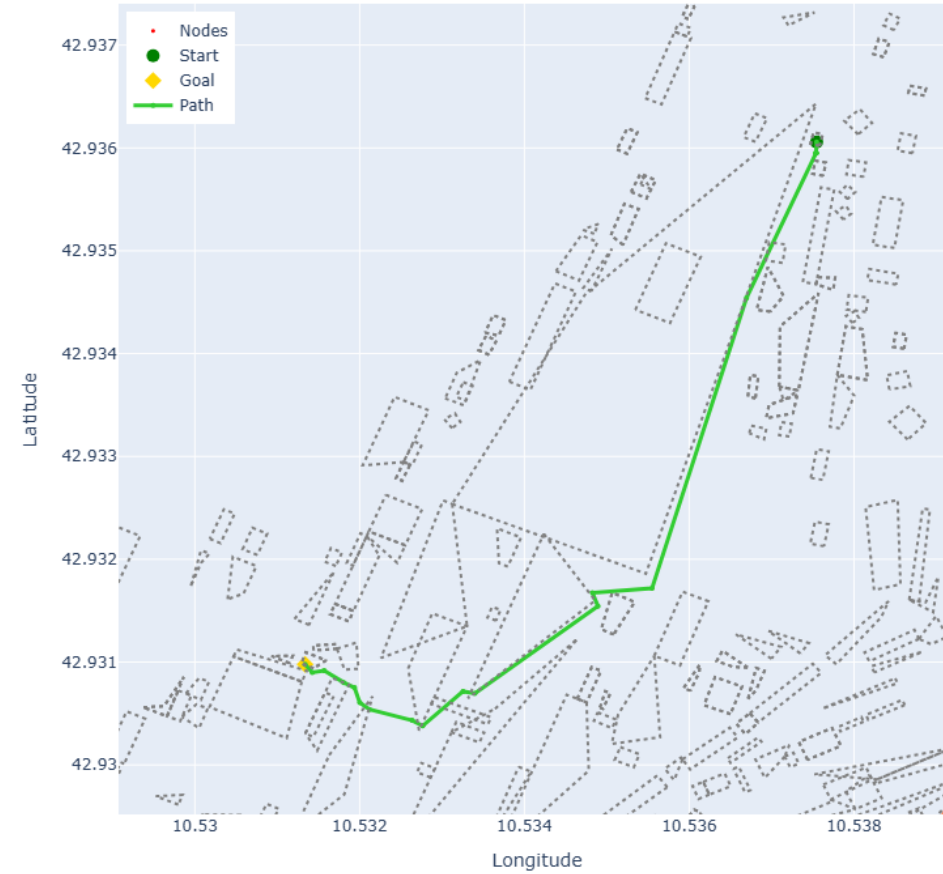
Metric	4	8	10	12	14
Success Rate (%)	40	70.62	71.88	79.38	77.5
Execution Time (s)	68.06	45.22	43.55	24.81	25.26
Path Length	4.34*	4.02*	4.0*	4.22*	4.07*
Threat Cost	2.67*	1.32*	0.96*	0.69*	0.6*
Node Count	2.01*	2.01*	1.62*	1.71*	1.39*
Smoothness	6.15*	3.16*	2.64*	2.42*	2.03*

PATH SMOOTHING

RRT Tree 2D - Version vA_best_raw

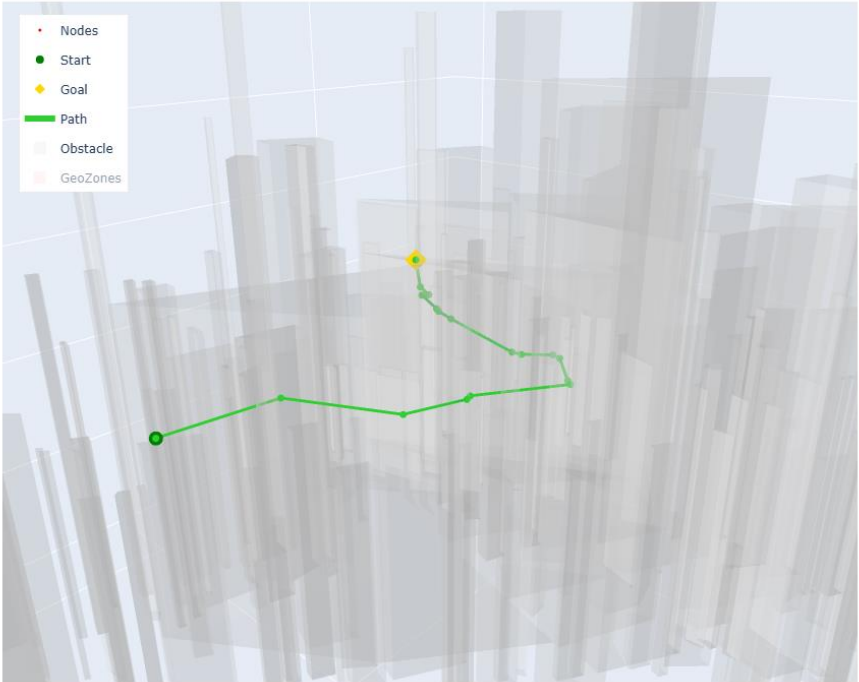


RRT Tree 2D - Version vA_best_smooth



PATH SMOOTHING

RRT Tree Version vP_best_smooth



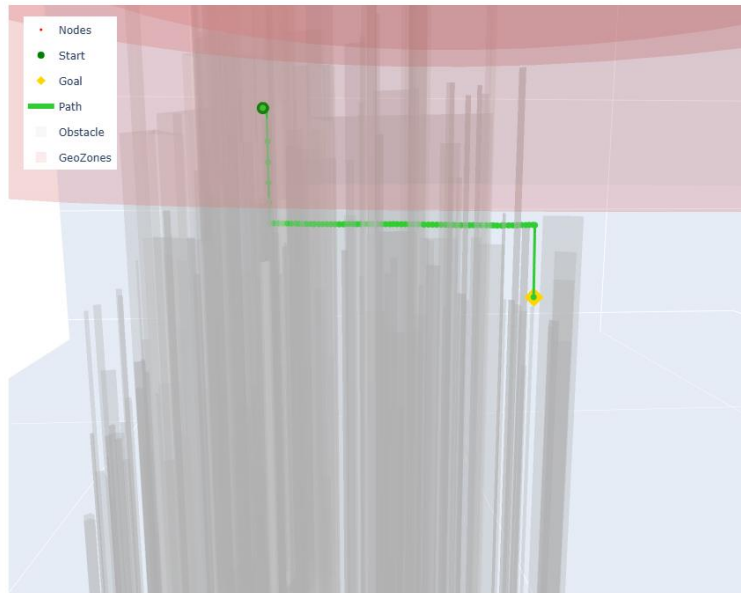
Results

Trajectory	Version	Threat*	Nodes*	Smoothness*
A	raw	4.286	2.355	8.167
A	smooth	2.857	0.242	1.085
B	raw	7.143	10	2.484
B	smooth	5.714	0.081	0.487
C	raw	4.286	0.431	0.66
C	smooth	1.429	0.027	0.121
...	...	...	...	...
K	raw	2.857	7.51	3.279
K	smooth	2.857	0.188	0.896
L	raw	5.714	1.373	1.8
L	smooth	2.857	0.296	0.283
...	...	...	...	...
O	raw	2.857	2.032	4.081
O	smooth	1.429	0.121	0.396
P	raw	0	4.361	6.722
P	smooth	0	0.175	0.92

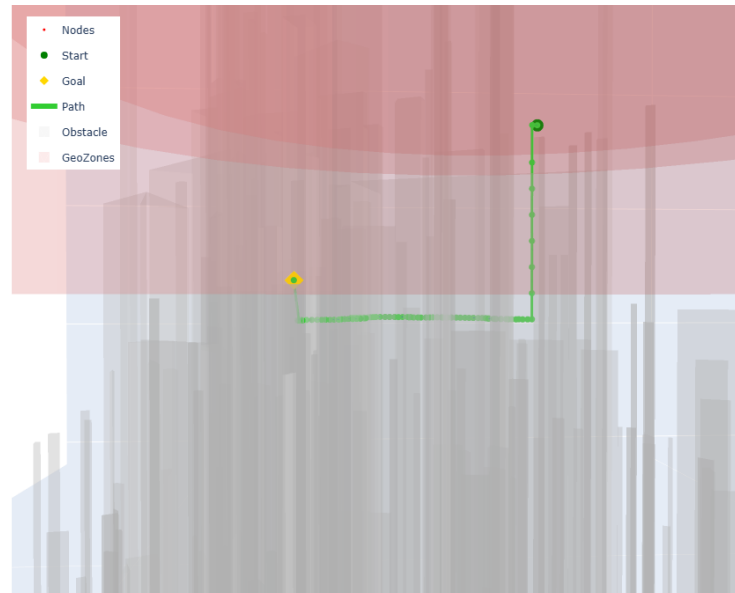
GEOZONE-AWARE PATH PLANNING

Results

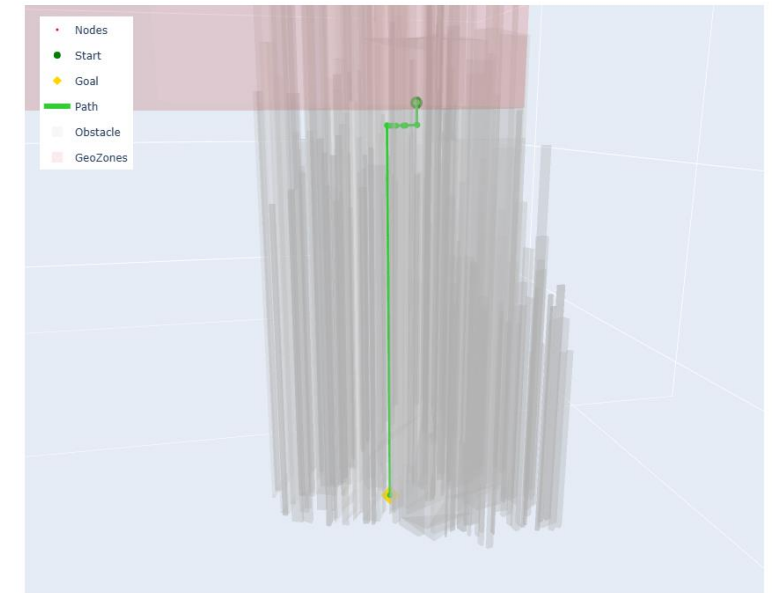
RRT Tree Version vD_best_raw



RRT Tree Version vE_best_raw



RRT Tree Version vB_best_smooth



CONCLUSIONS

1 We developed a geozone-aware RRT* algorithm for safe UAV navigation.

2 Jellyfish-inspired multi-objective decision finds the most optimal path.

3 Tests over Piombino showed strong performance in dense environments.

4 The method supports scalable, regulation-compliant urban UAV flights.

FUTURE WORK

1 Define complex zones where the algorithm shows low success rates.

2 Use zone analysis to plan UAV infrastructures like take-off platforms.

3 Integrate real-time weather data to adapt paths dynamically.

4 Extend the algorithm to multi-drone coordination scenarios.

5 Test in other urban environments to validate scalability.

THANK YOU

ACKNOWLEDGMENT

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JUDIT SALVANS BAUCCELLS

judit.salvansbaucells@ugent.be

ELHAM FAKHRAIAN

elham.fakhraian@ugent.be



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