

ICMASS 2024

Lidar-Based Obstacle Detection and Path Prediction for Unmanned Surface Vehicles

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Lidar-Based Obstacle Detection for Unmanned Surface Vehicles

1. Introduction
2. Why USVs?
3. Motivation
4. Developed Pipeline
 - Overview
 - Detail each Task
5. Results
6. Discussion
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Introduction

Digital test field Elbe



1

Integration of inland vessels in digital logistics chains
Digitization of ship and land systems for the transmission of logistics-relevant data.

2

Improving the river Elbe's navigability
Recording and provision of fairway and navigation channel information based on highly precise digital data.

3

Reduction of downtimes and energy requirements
AI-based analysis and assistance systems for inland vessels.

This project was funded by the Federal Ministry of Digital and Transport under the funding code 45DTW2V09A.

Maritime Infrastructure

—
Backbone of Global Trade economy

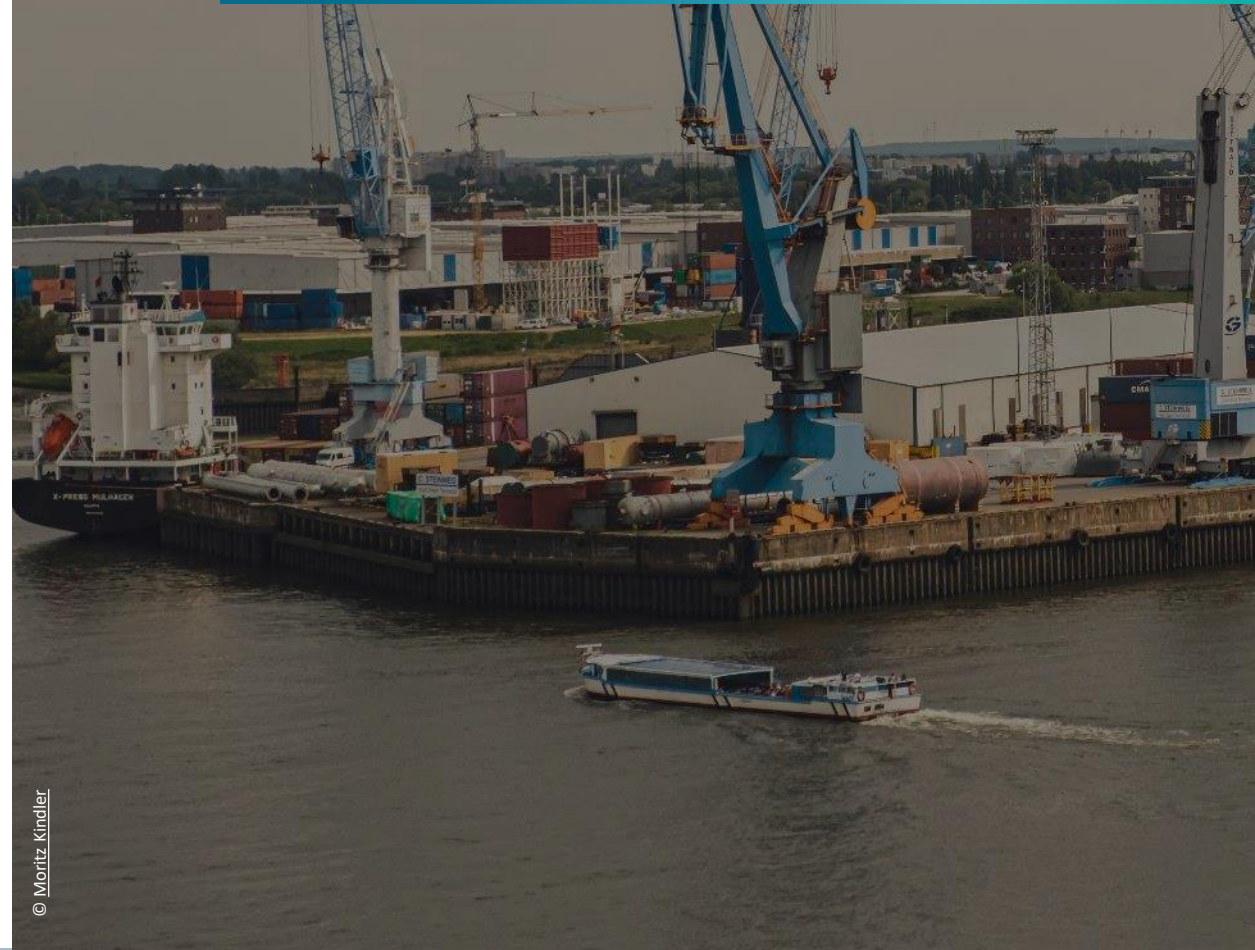


Image: <https://unsplash.com/de/fotos/schiff-in-der-nahe-von-seehafen-shdC9zPtPKU>

Why USVs?

Benefits

Efficiency

- Operate 24/7 without fatigue.
- Cover vast areas swiftly.

Safety

- Reduce human exposure to maritime hazards.
- Reliable in unpredictable sea conditions.

Cost-Effective

- Minimize long-term human operational costs.
- Efficient maintenance and operation.

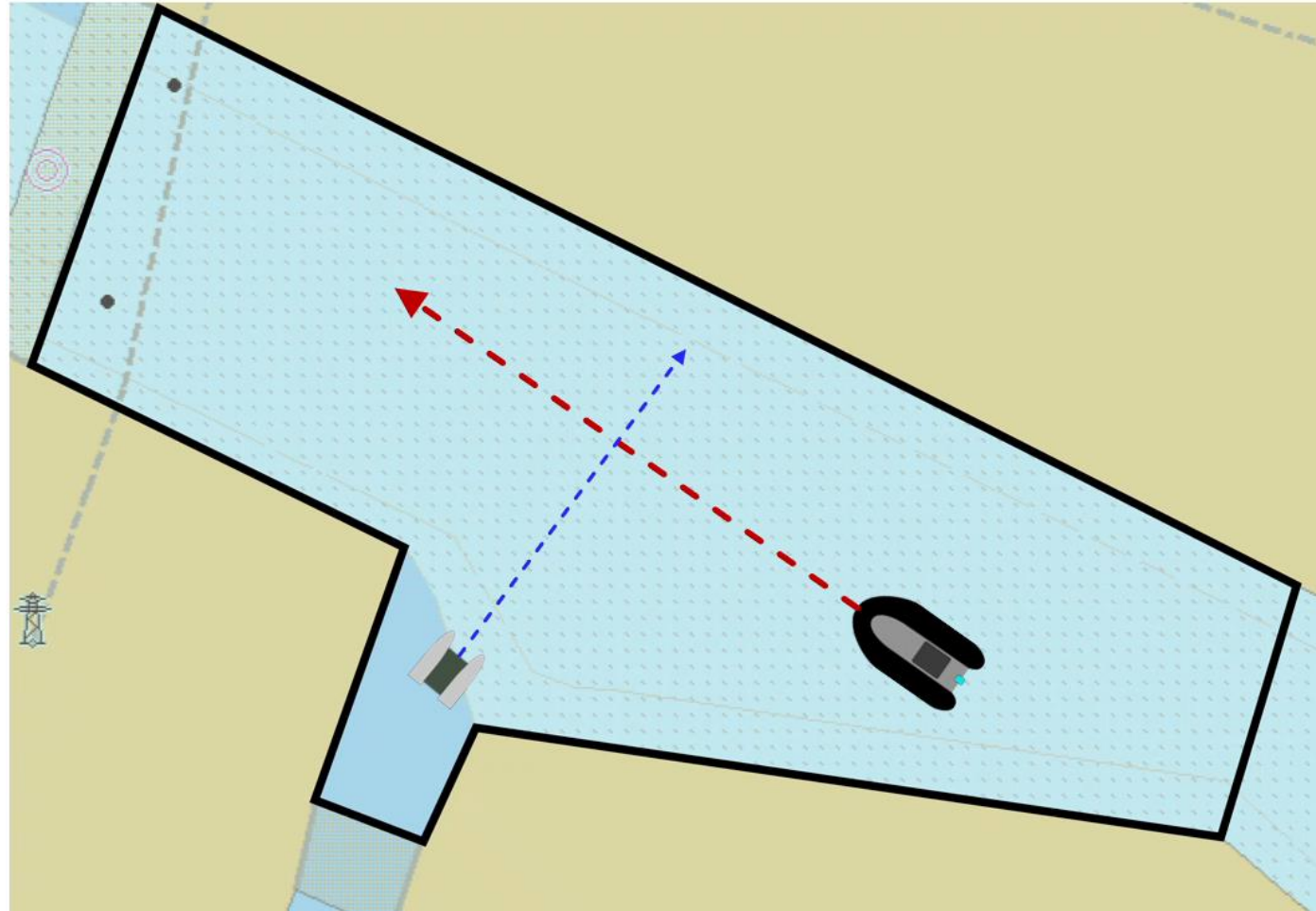
Advanced Technology

- AI-powered surveillance and monitoring.
- Real-time data collection and analysis.



Motivation

Why do we need tracking of dynamic objects?



With Tracking

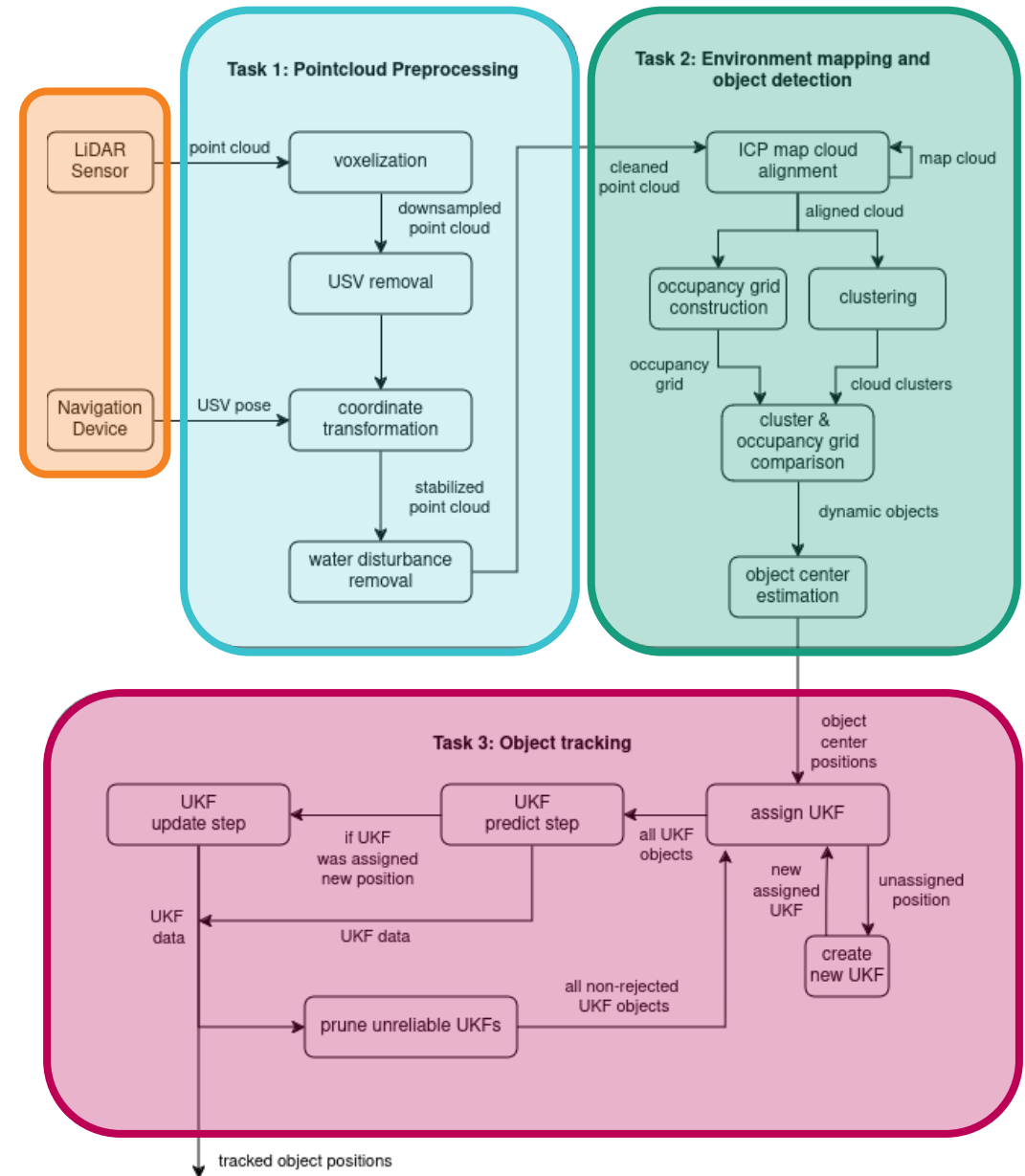
Developed Pipeline Overview

Data Sources

Task 1: Pointcloud Preprocessing

Task 2: Environment Mapping and Object Detection

Task 3: Object Tracking

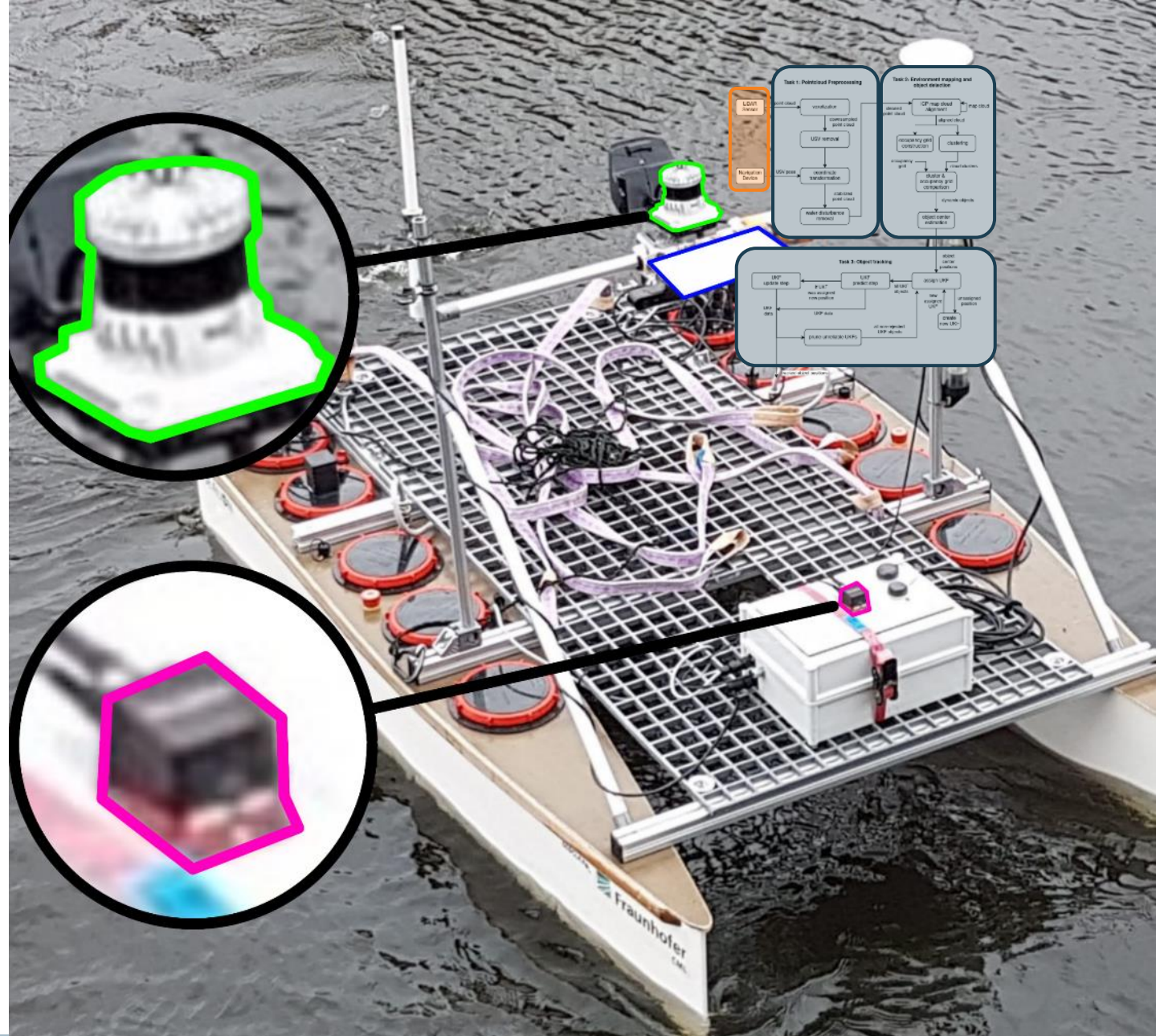


SeaML:SeaLion

Data sources

Vehicle Facts

- Catamaran Design
- Construction dimensions - 2.2m x 1.5m
- Interface for different sensors and payloads
 - Payload up to 100kg
- 2 x 3HP electric outboard motors
- 2 x 48V batteries (50 Ah 2.3 kWh each)
- Modular power grid (5/12/24/48V Output power)
- Sensors



Developed Pipeline

Task 1

Pointcloud Preprocessing

Down-sampling for Computational Efficiency:

- Reduces point-cloud size for lower computational load.

Pass-through Filtering:

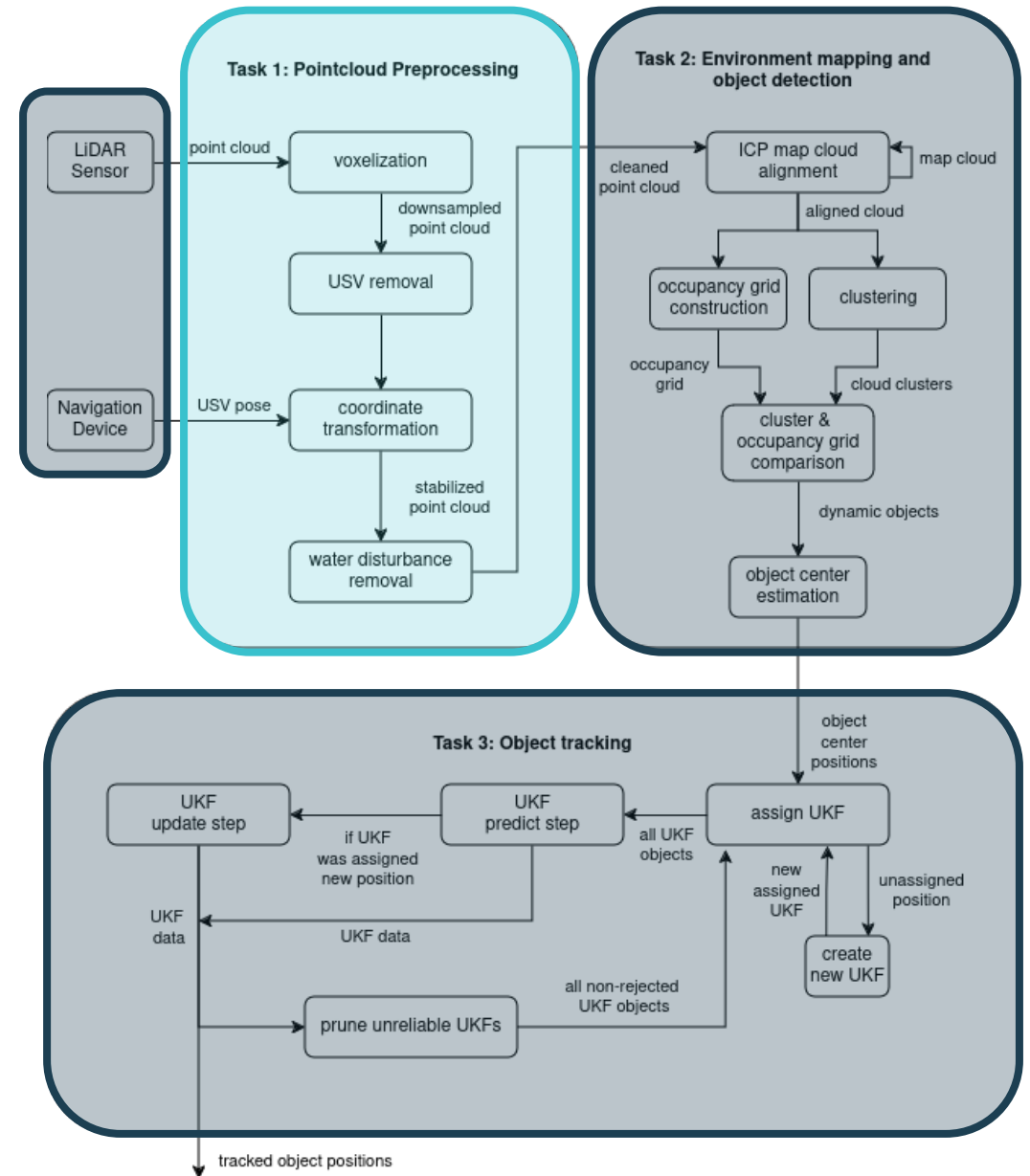
- Removes points from the observing USV (e.g., masts in LiDAR view).

Global Coordinate Alignment:

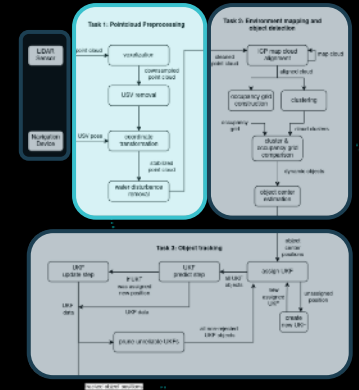
- Aligns point cloud using USV pose data (roll, pitch, yaw stabilization).
- Translates to global coordinates with GNSS data for fixed positioning.

Surface Noise Removal:

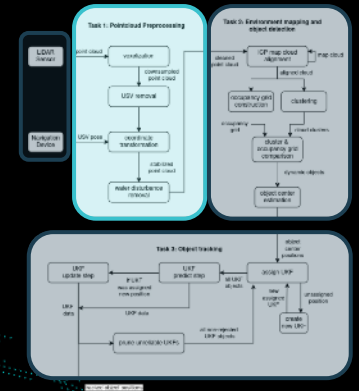
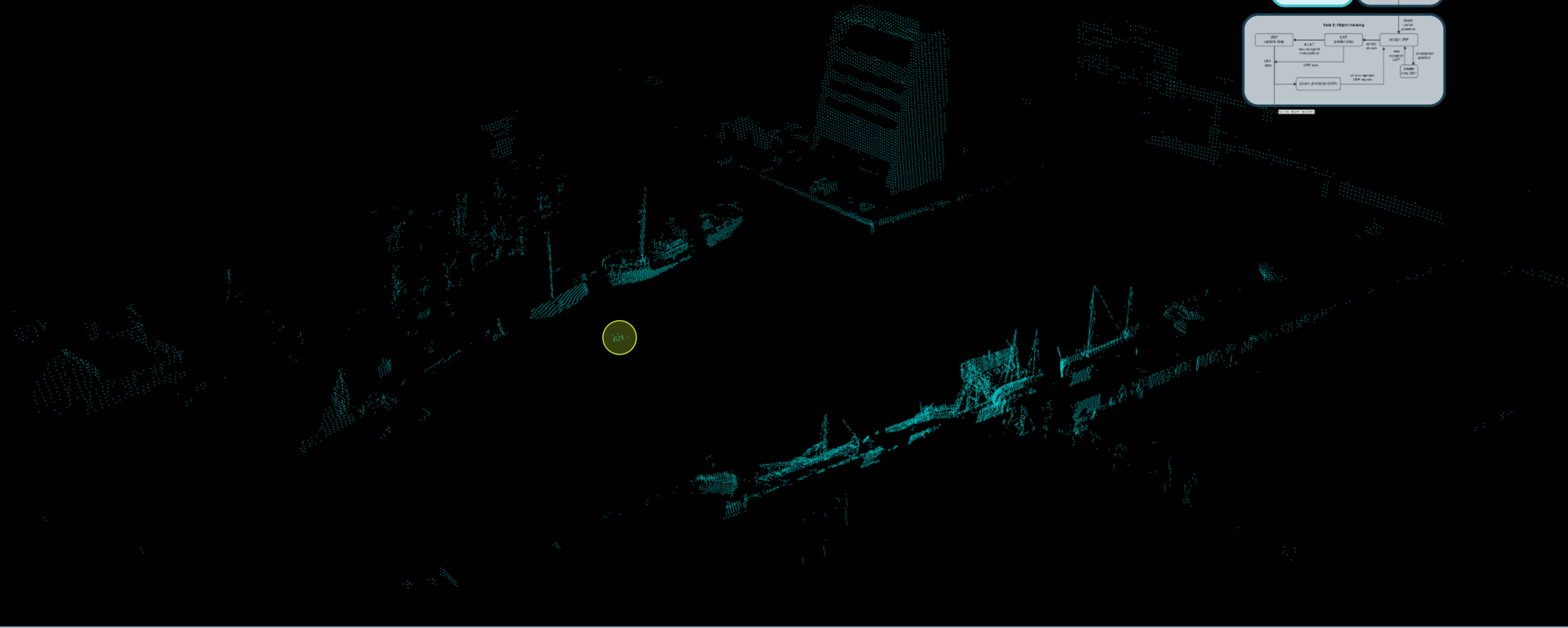
- Filters out water splashes and wakes, aligned after stabilization.



Pipeline Pre-processing



Pipeline Pre-processing



Developed Pipeline

Task 2

Environment Mapping and Object Detection

Initial Point Cloud Alignment:

- ICP Algorithm: Iteratively aligns incoming point clouds to a base map.
- Initial Cloud: Sets the reference point for subsequent data alignment.

Dynamic Occupancy Grid Construction:

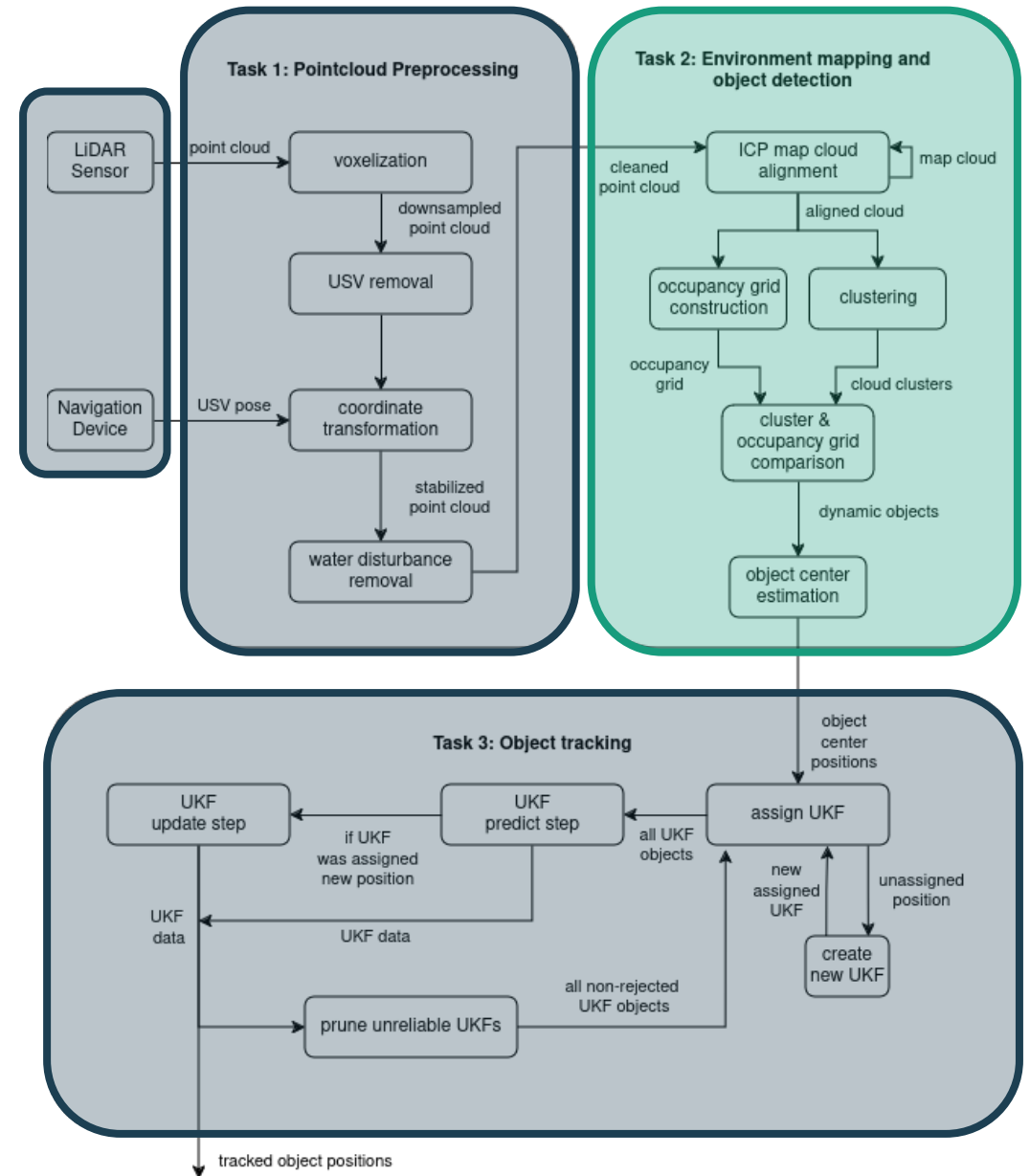
- Grid Cell Values: Range from -1 (unknown) to 100 (static obstacle).
- Dynamic Updates: Grid cells updated with each new aligned cloud.

Static vs. Dynamic Object Identification:

- Static Objects: Cells with high values (≥ 80) mark.
- Dynamic Objects: Moving clusters identified and tracked if not covering marked land cells.

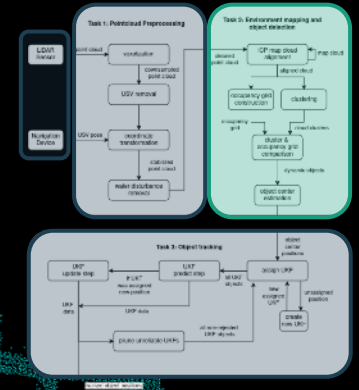
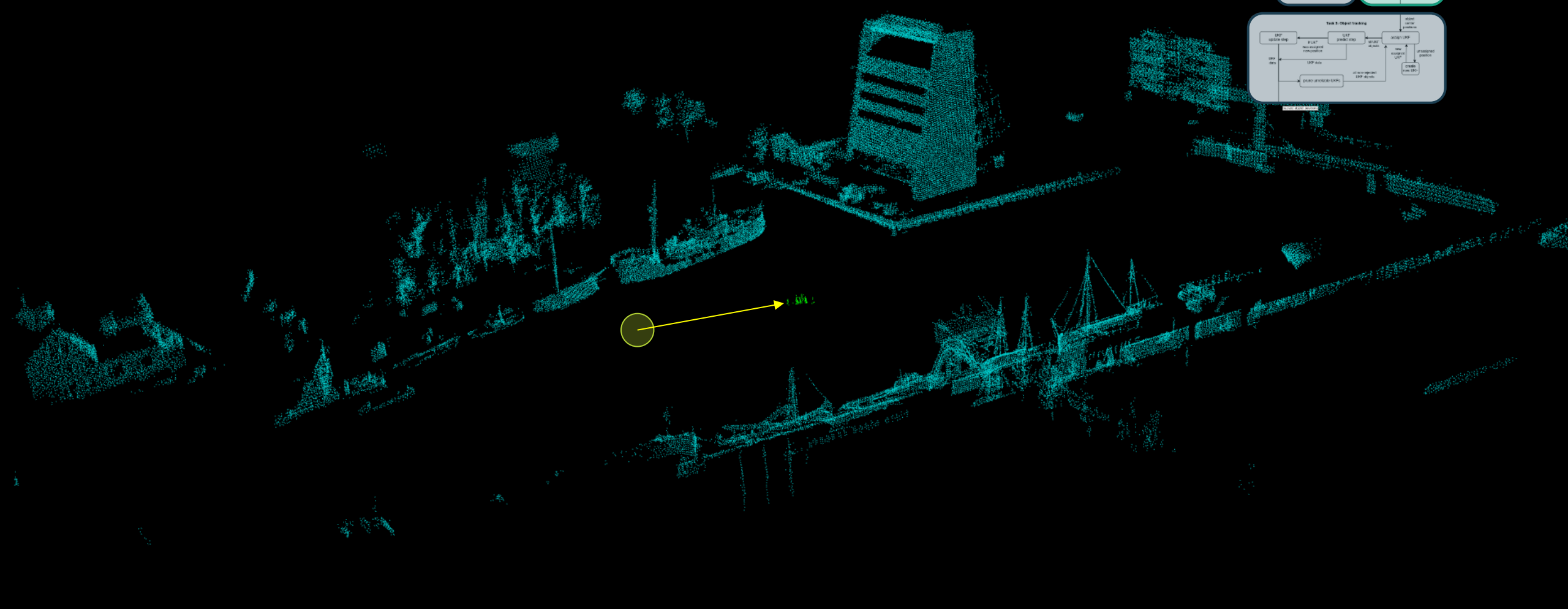
Dynamic Obstacle Detection:

- Cluster centroids calculated and passed to object tracker for dynamic tracking.



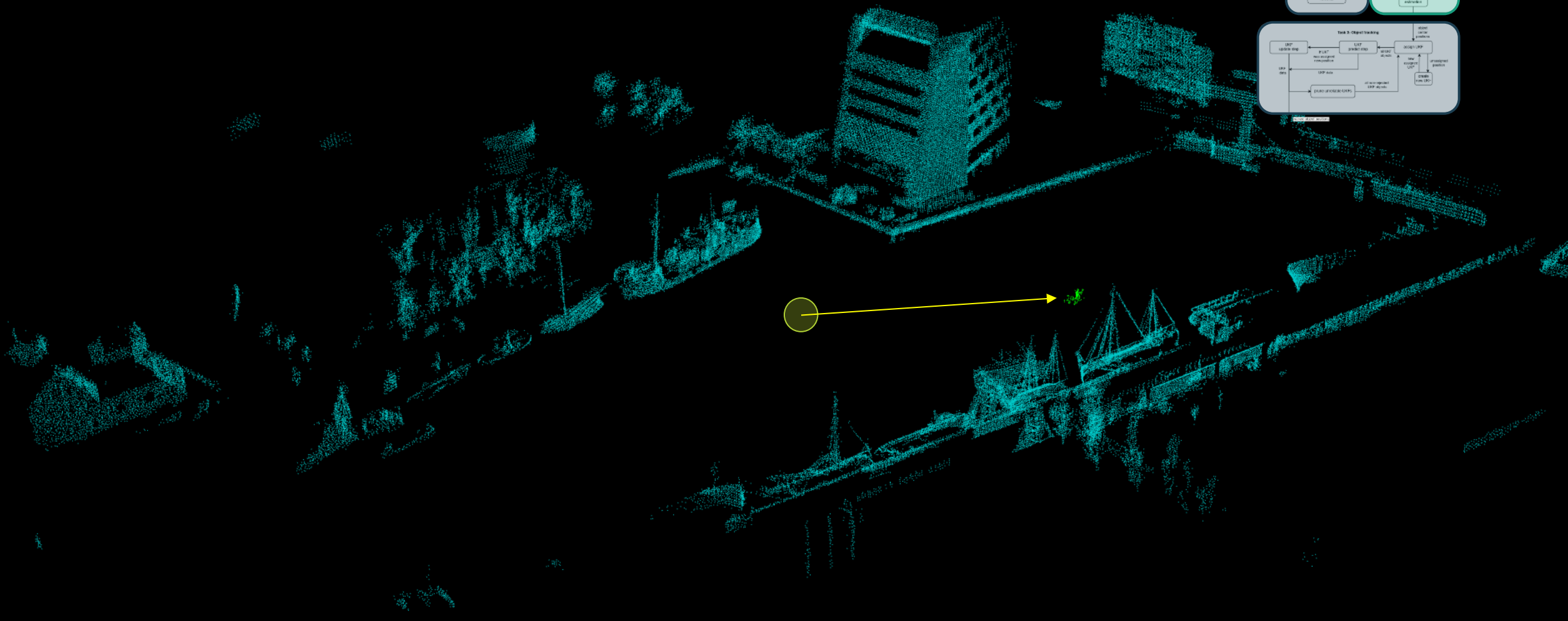
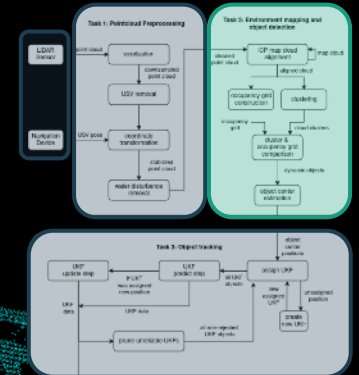
Pipeline

Mapping – 15 seconds



Pipeline

Mapping – 30 seconds



Pipeline

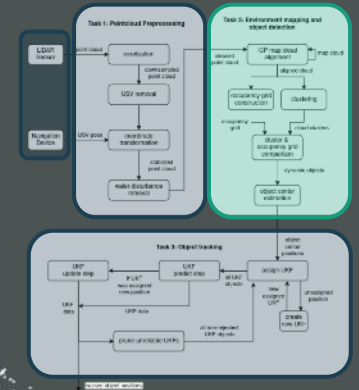
WATER

Confidence-Rating



LAND

USV
Position



Developed Pipeline

Task 3

Object Tracking

Vessel Movement Modeling:

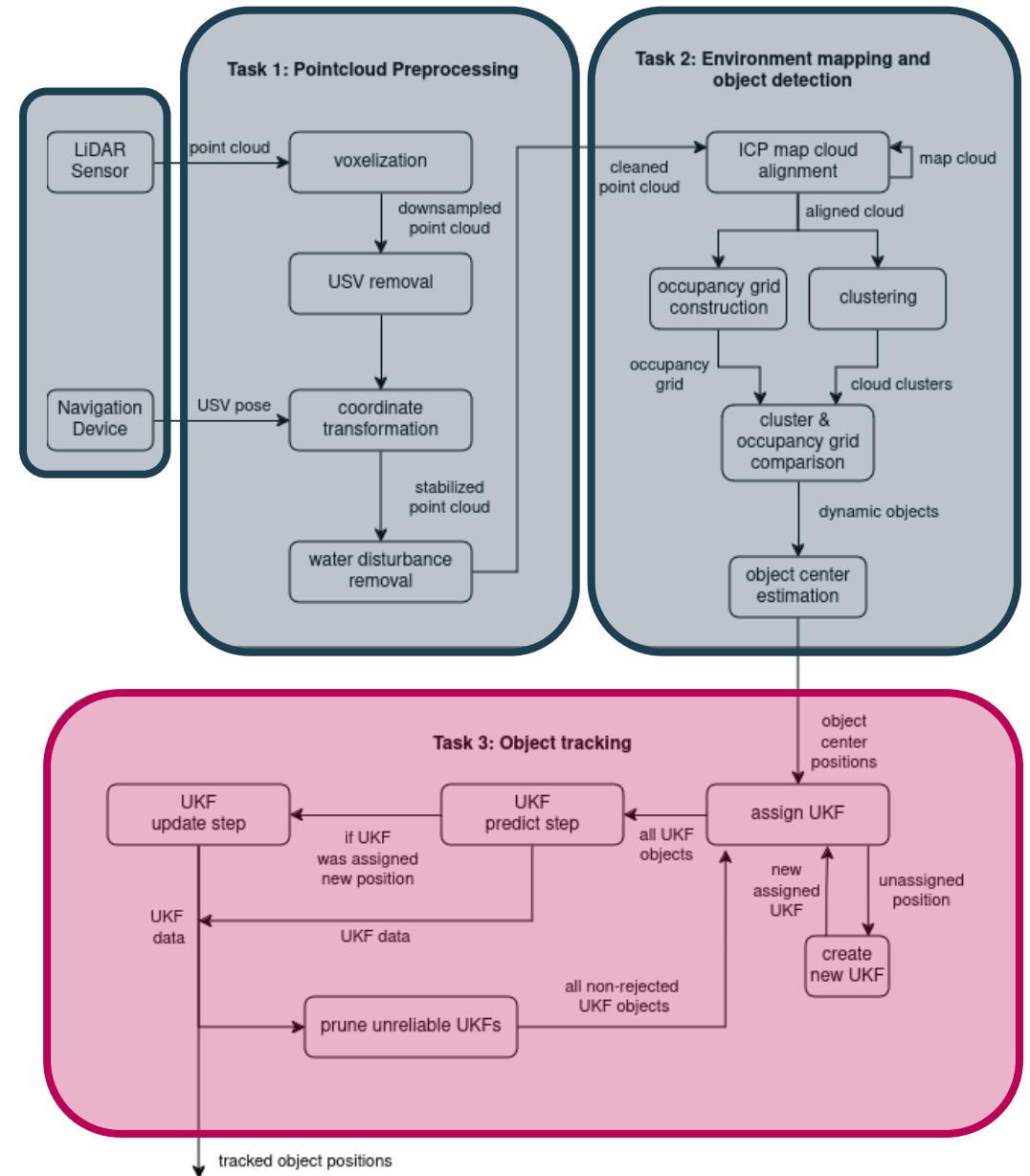
- Constant Turn Rate and Velocity (CTRV) Model
- Key tracked states: Position, Velocity, Heading.

Dynamic UKF Tracking Process:

- Centroid Matching: Checks new centroids against existing UKF instances.
- Measurement Update: Updates tracker if an existing UKF is within range.
- New Object Detection: Creates new UKF tracker if no match is found.

Tracker Management:

- Position Estimation: Continues estimation if no new data is available.
- Tracker Pruning: Removes UKF instances when objects go out of view.
- Allows for independent and concurrent tracking of multiple targets.



Experimental Setup

Test Location

Lotsekanal (Pilot Channel), Hamburg, Germany

- Enclosed, controlled channel in the southern port.
- Protected by a lock—minimal current and tidal effects.
- Trial conducted in February 2024.

Test Area Dimensions:

- **200 m long and 65 m wide** (restricted by docked vessels).

Environmental Conditions:

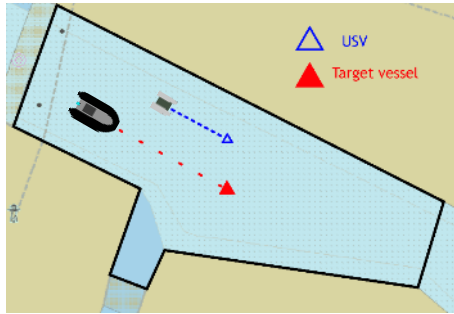
- Wind Speeds: 6 m/s to 8 m/s (11 to 16 knots).
- Wind Directions: N, NE, E.
- Main disturbance: Wind, with minimal current influence.

Target Boat:

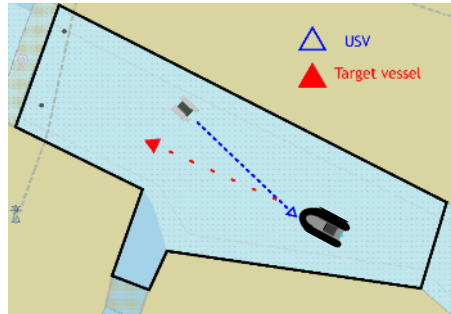


Experimental Setup

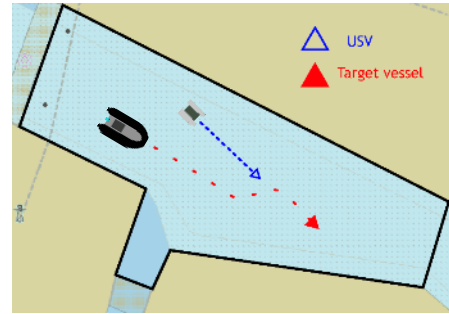
Test Scenarios



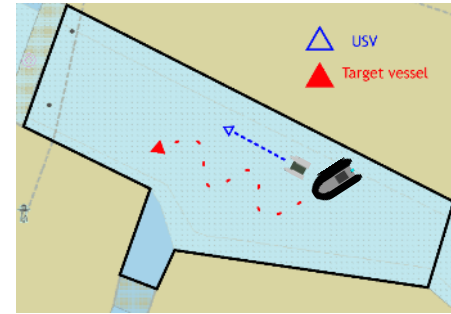
Scenario 1:
Target vessel moving in
the Same Direction



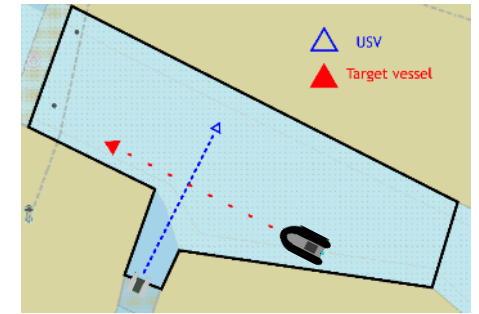
Scenario 2:
Target Vessel Moving in
the Opposite Direction



Scenario 3:
Target Vessel Overtaking
the USV



Scenario 4:
Target Vessel Moving in a
Sinusoidal Curve



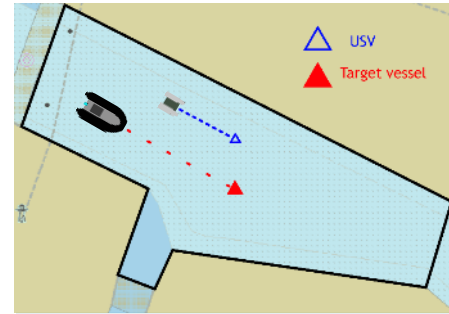
Scenario 5:
Target Vessel Moving
Perpendicularly

Results

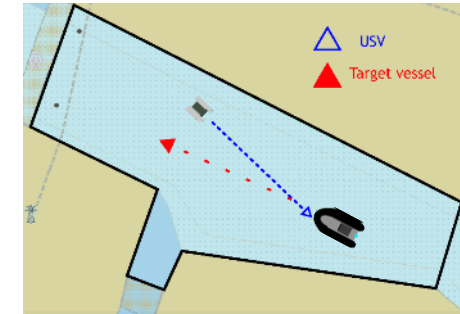
Tracking

Overview of the tracking errors across the five different scenarios.

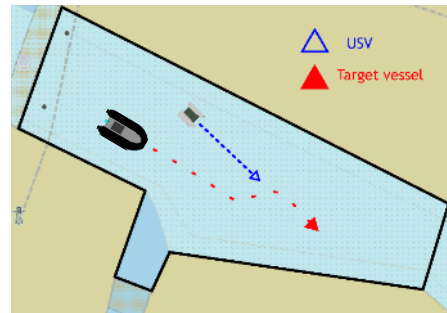
Scenario	RMSE Tracking Error [m]
1	3.3
2	5.4
3	2.9
4	4.1
5	3.5
Average	3.8



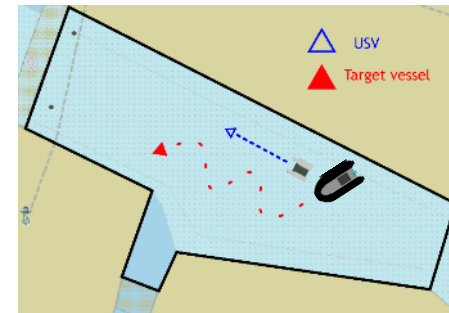
Scenario 1



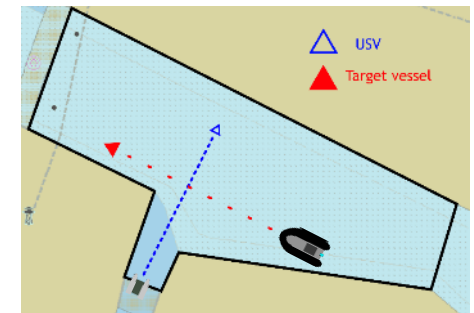
Scenario 2



Scenario 3:



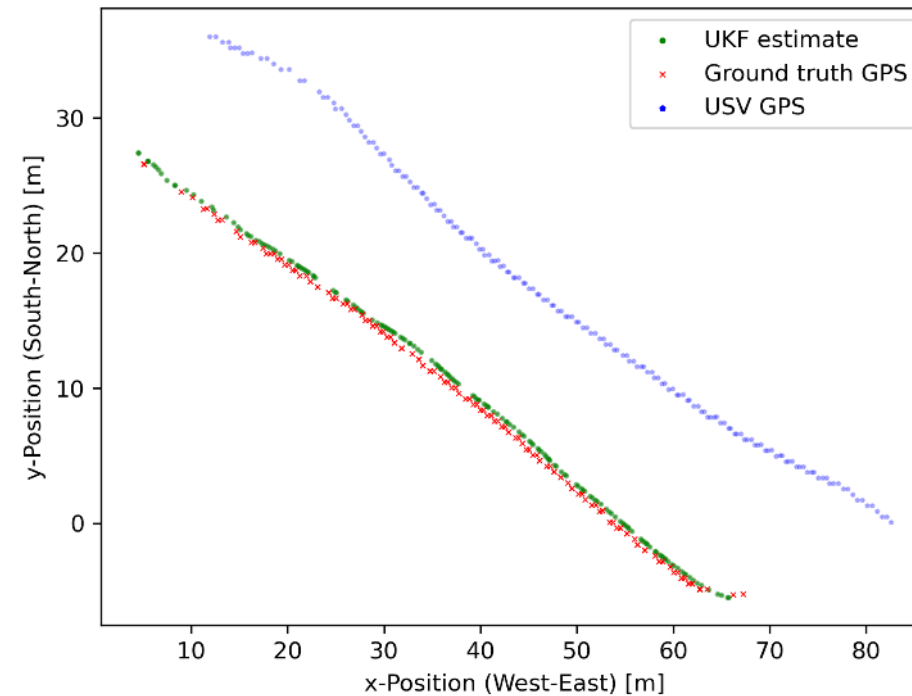
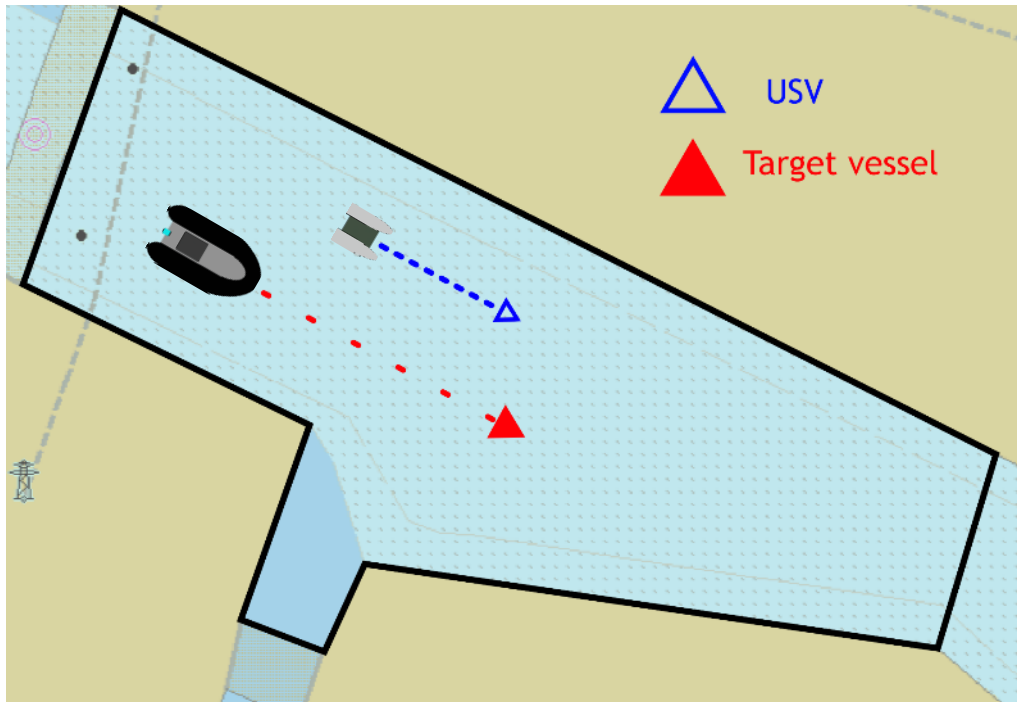
Scenario 4



Scenario 5

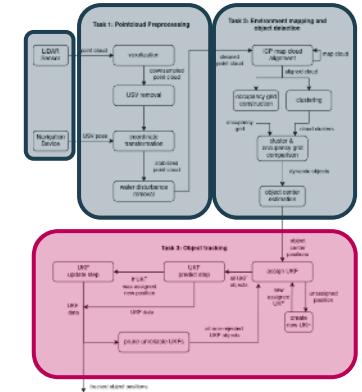
Results

Tracking Scenario 1



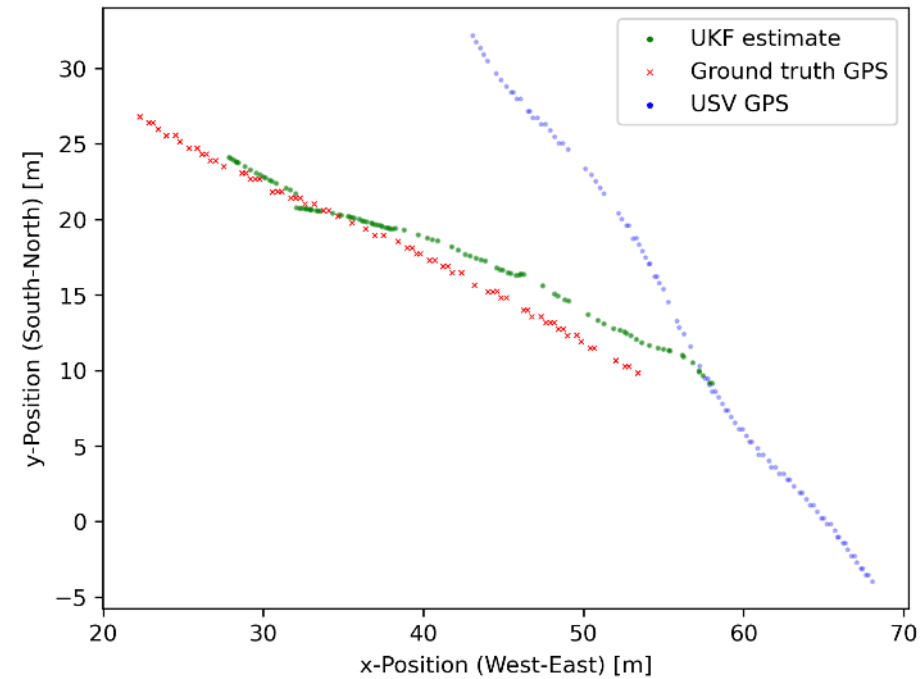
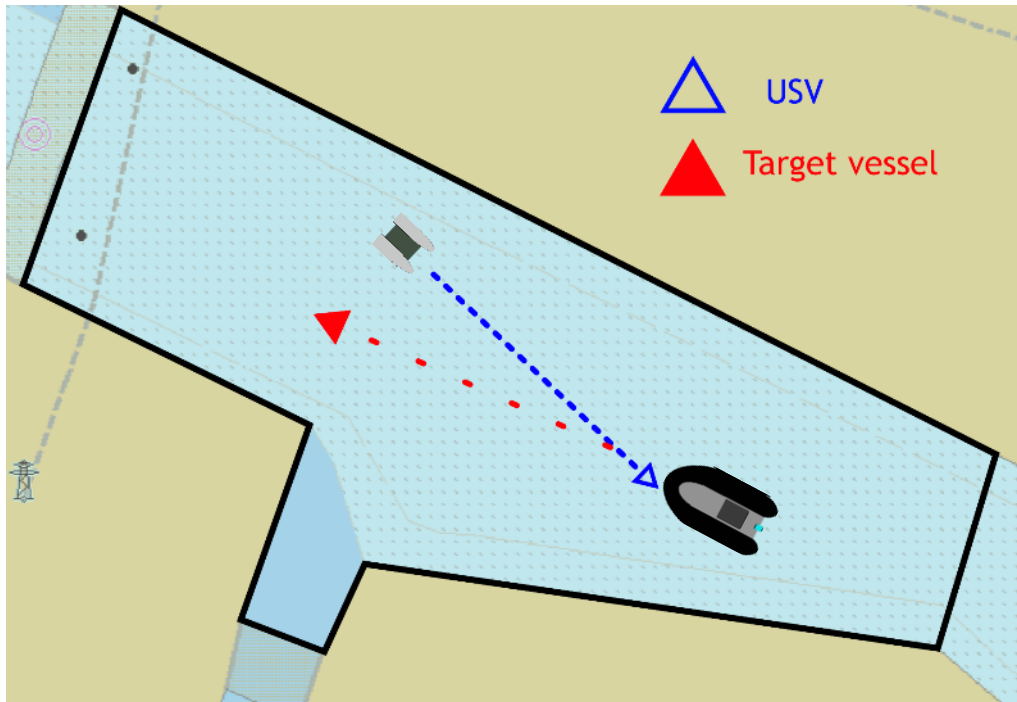
Tracking error
green position vs
red position

3.3 m



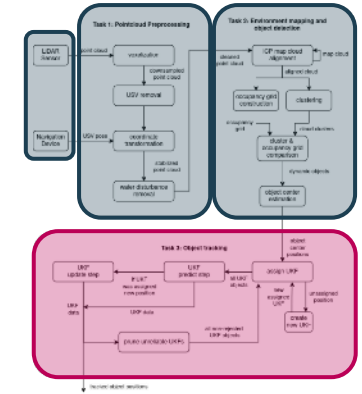
Results

Tracking Scenario 2



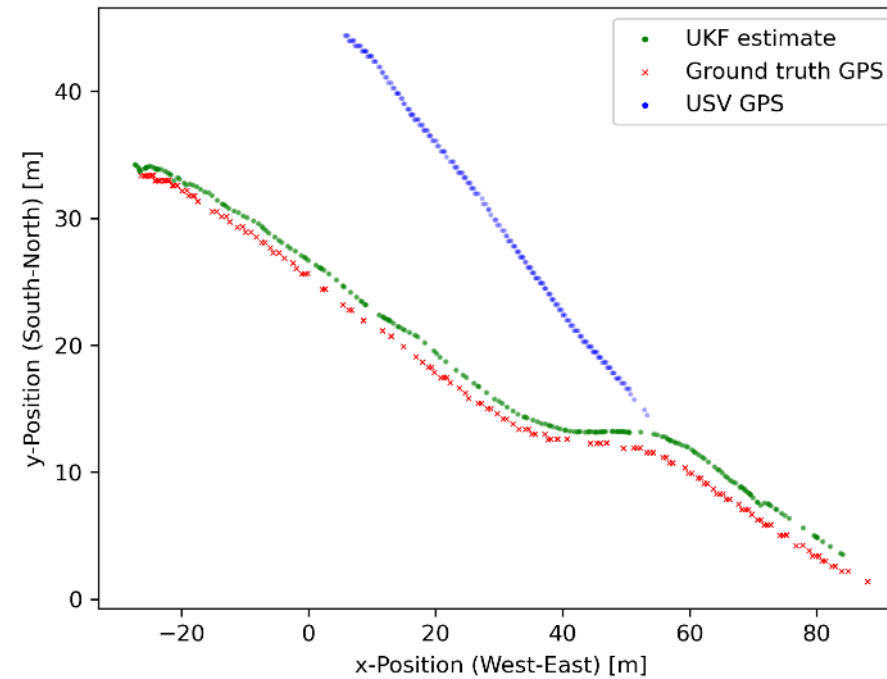
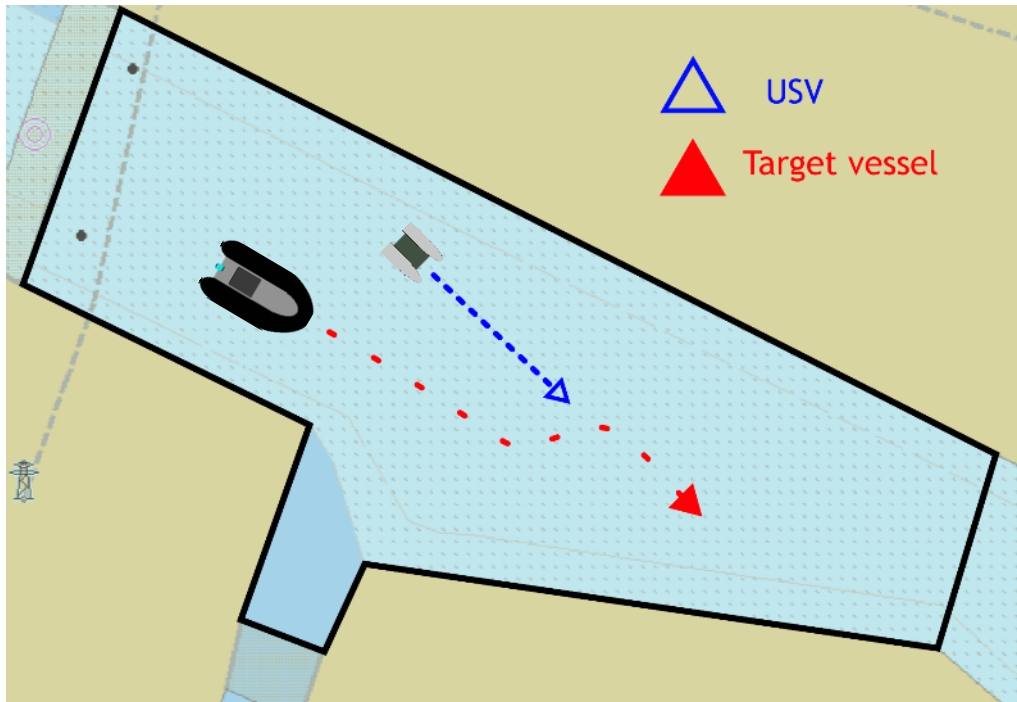
Tracking error
green position vs
red position

5.4 m



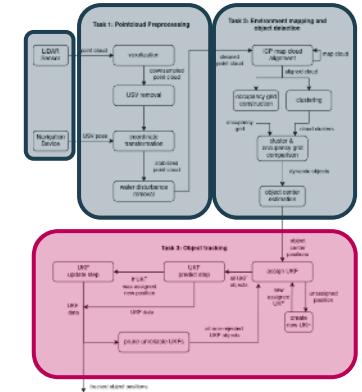
Results

Tracking Scenario 3

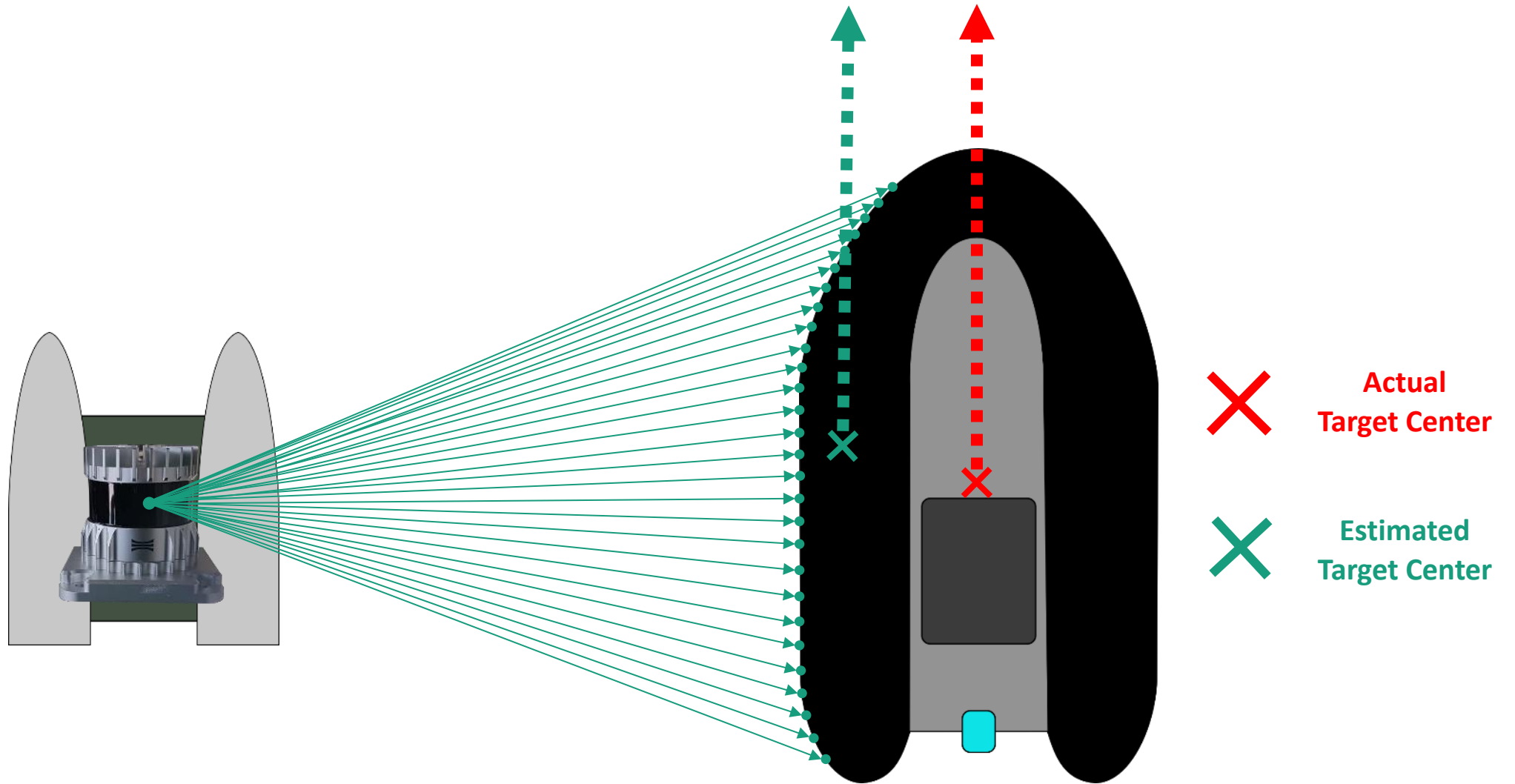


Tracking error
green position vs
red position

2.9 m



Results Discussion



Conclusion

Lidar-Based Obstacle Detection and Path Prediction

Key Achievements:

- Developed a novel perception pipeline for obstacle detection and path prediction.
- Overcame limitations of traditional methods in narrow, crowded environments like the Lotsekanal.

Innovative Approach:

- Used occupancy grid data for static land removal.
- Improved obstacle avoidance by integrating dynamic obstacle trajectories.

Future Enhancements:

- Advanced path planning with extrapolated dynamic trajectories.
- Explore AI-based object detection with 2D CNN PointPillars for greater accuracy and speed.



Contact

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