# Wireless Localization of Laboratory Devices



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## **PROJECT GOAL**

- Evaluate wireless localization methods (e.g., RSSI, ToF, UWB) without external receivers.
- Assess feasibility for < **50 cm accuracy** under real lab conditions.
- Develop and test a software prototype to demonstrate the selected approach.

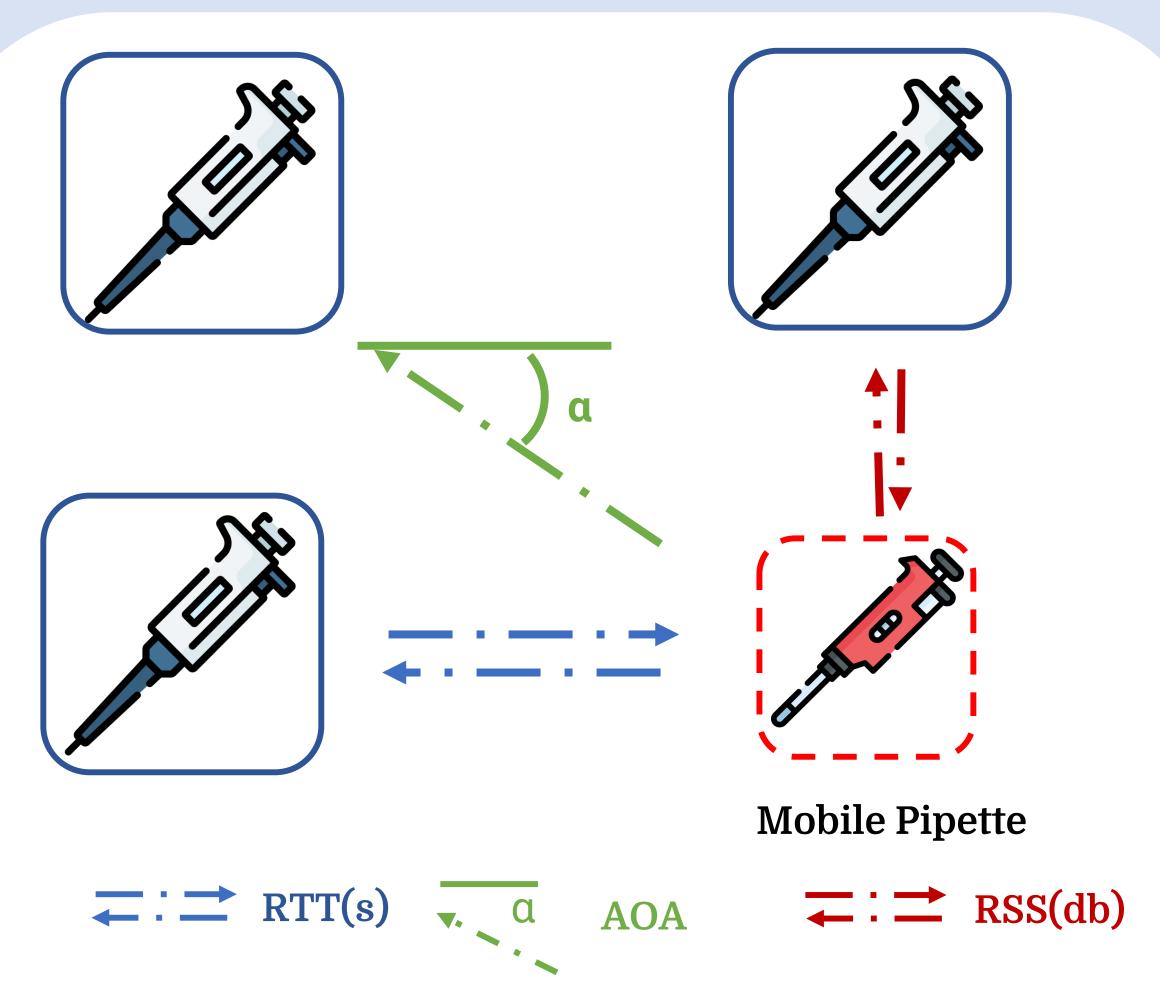
## **METHODOLOGY**

- A literature review on Ultra-Wideband (UWB), Received Signal Strength (RSS), Angle of Arrival (AoA), Bluetooth Low Energy (BLE), Wi-Fi, and Round-trip time (RTT) technologies.
- Select a use-case scenario and define the final setup to meet  $\leq 50$  cm accuracy.
- Implement a simulation environment to test the selected localization approach.
- Evaluate performance, analyze results

#### **APPROACH**

- Setup: Use three fixed pipettes (anchors) and one mobile pipette (target).
- Data Collection:
  - *Phase 1:* Gather RTT and AoA measurements from known positions to create a reference database and train the positioning model.
  - Phase 2: Use real-time RTT and AoA in simulation workspace to estimate the mobile pipette's position via the trained model.
- Validation: Compare estimated and actual positions to quantify accuracy.

#### LOCALIZATION TOPOLOGY



\*RTT: Round Trip Time RSS: Received Signal Strength AOA: Angle of Arrival

# **SIMULATION RESULTS**

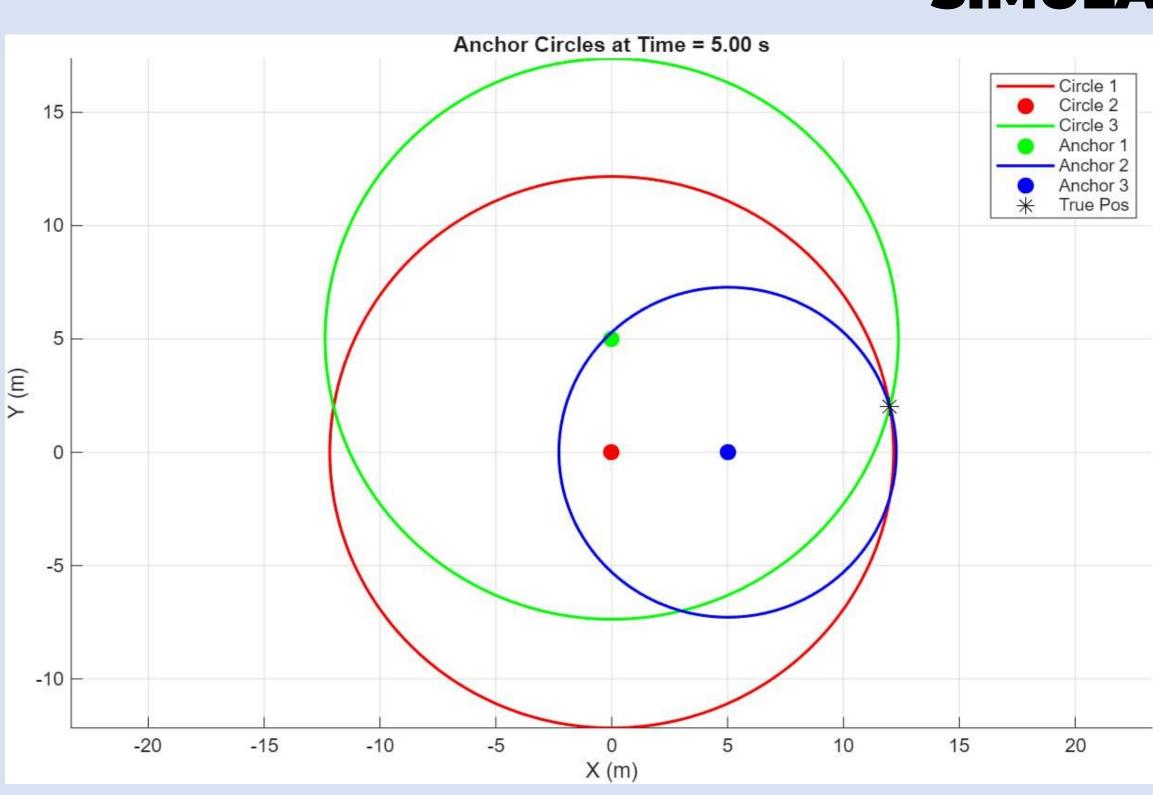


Fig 1: Trilateration of True Position from Anchor Circles at Time t = 5s

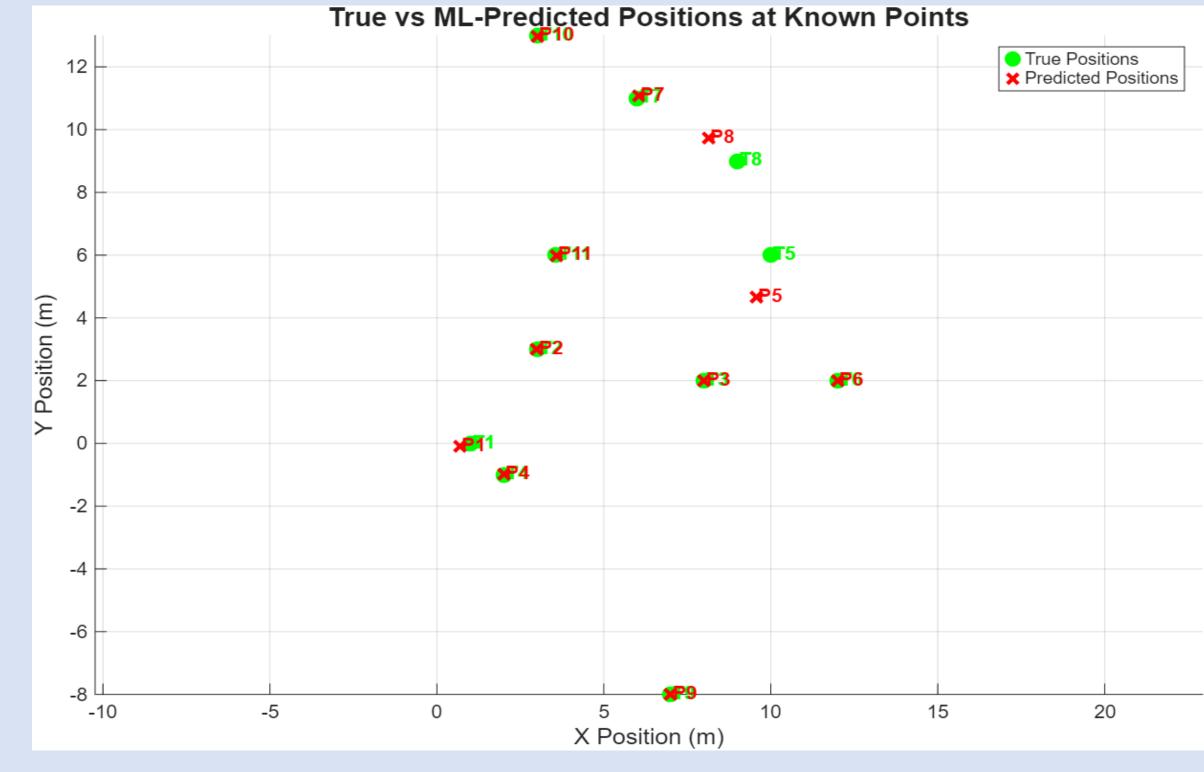


Fig 2: Comparison of true positions with ML-predicted positions at known points, showing localization accuracy.

# **ACCURACY RESULTS**

Min Error: ± 3.30 cm Max Error: ± 113.88 cm Mean Error: ± 38.27 cm Std: ± 30.75 cm

## CONCLUSION

- UWB was excluded due to hardware limitations.
- RSS showed poor accuracy and was not suitable for precise localization.
- RTT and AoA performed best based on simulation results and accuracy comparison.
- Machine learning improves accuracy to 88.12% and enables scalable localization in larger lab environments

# **FUTURE WORK**

- Build a real-world dataset using actual lab devices in live environments.
- Compute localization parameters under real lab conditions, including noise and signal collisions.
- Develop a user-friendly app for labs and customers to simplify setup and usage.
- Refine accuracy metrics by testing in dynamic, realworld scenarios.