

MATLAB EXPO 2019

Sensor Fusion and Tracking for Autonomous Systems

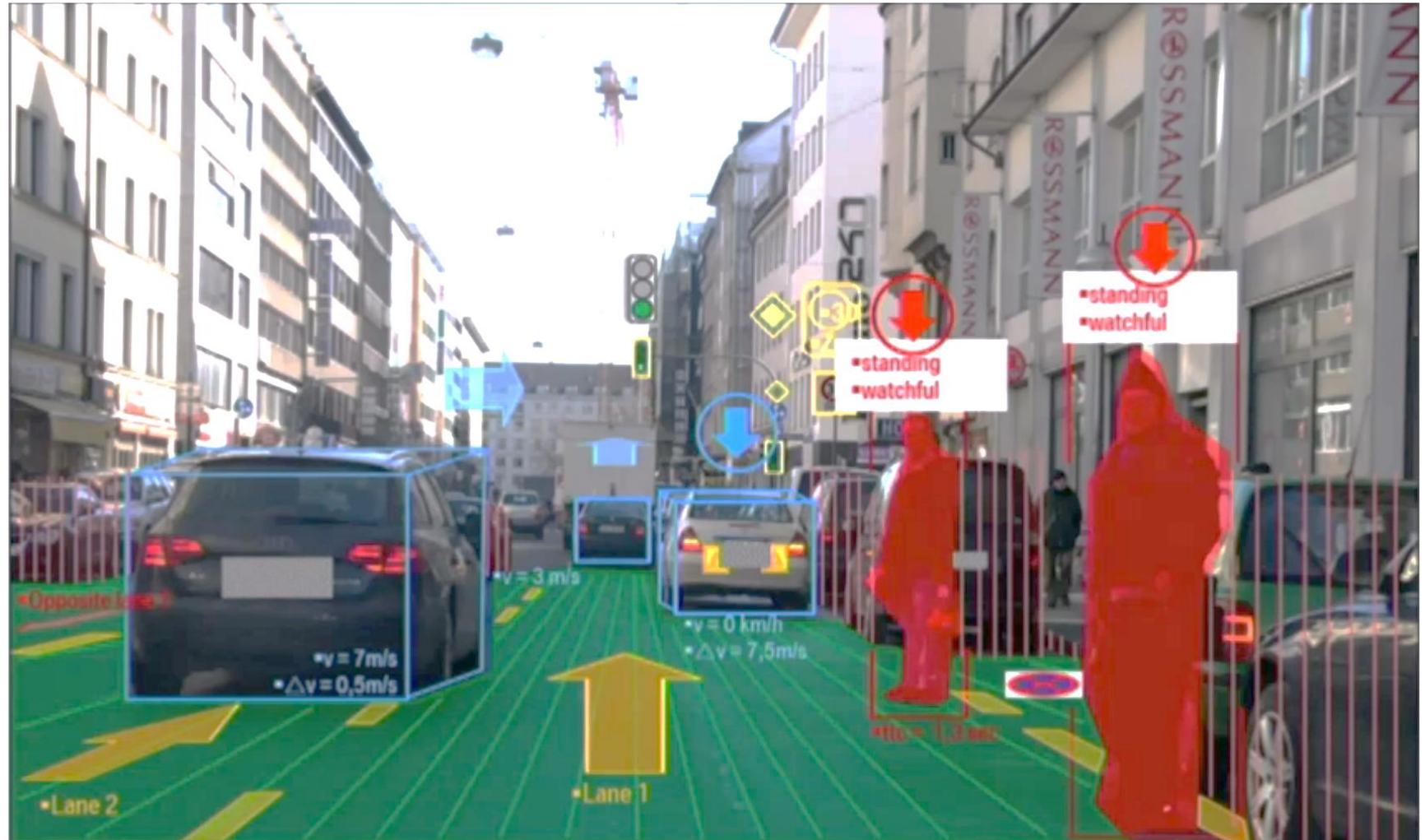
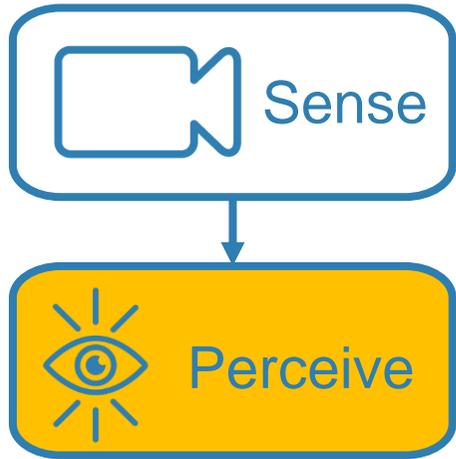
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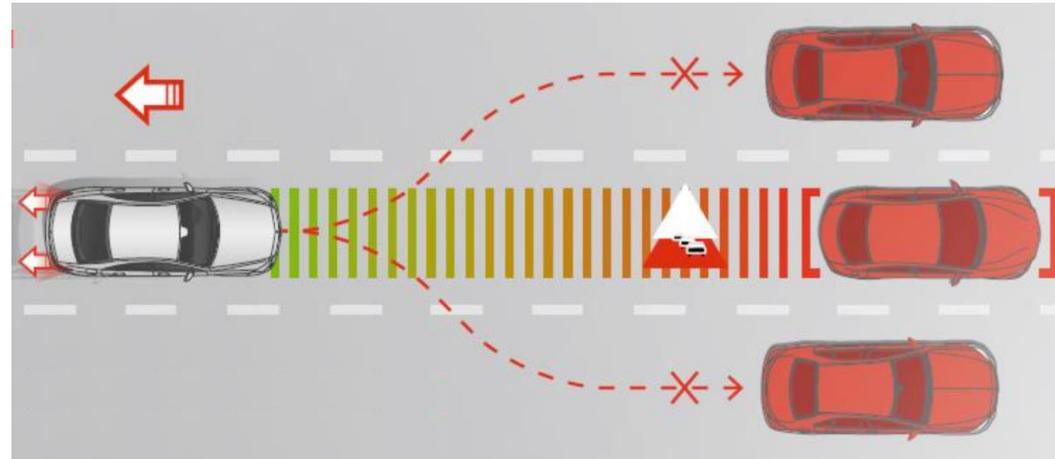
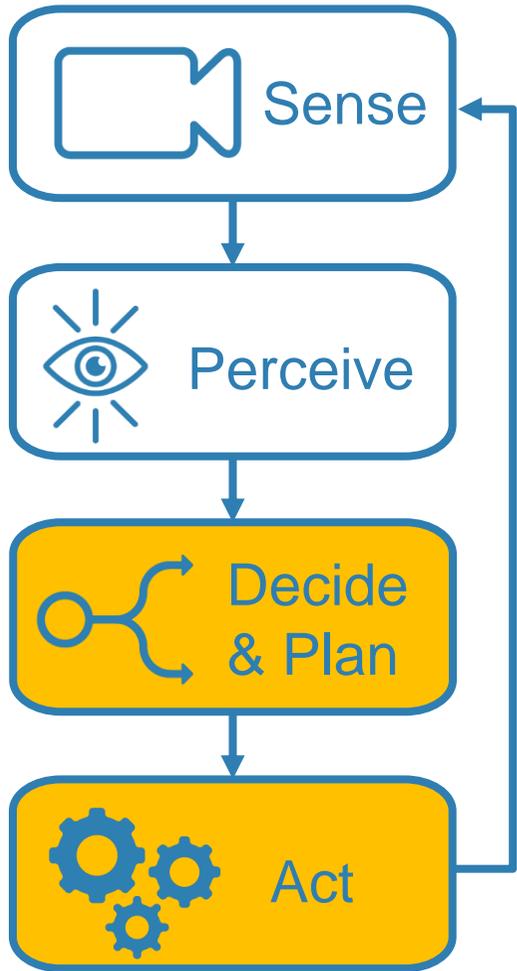
Capabilities of an Autonomous System



Capabilities of an Autonomous System



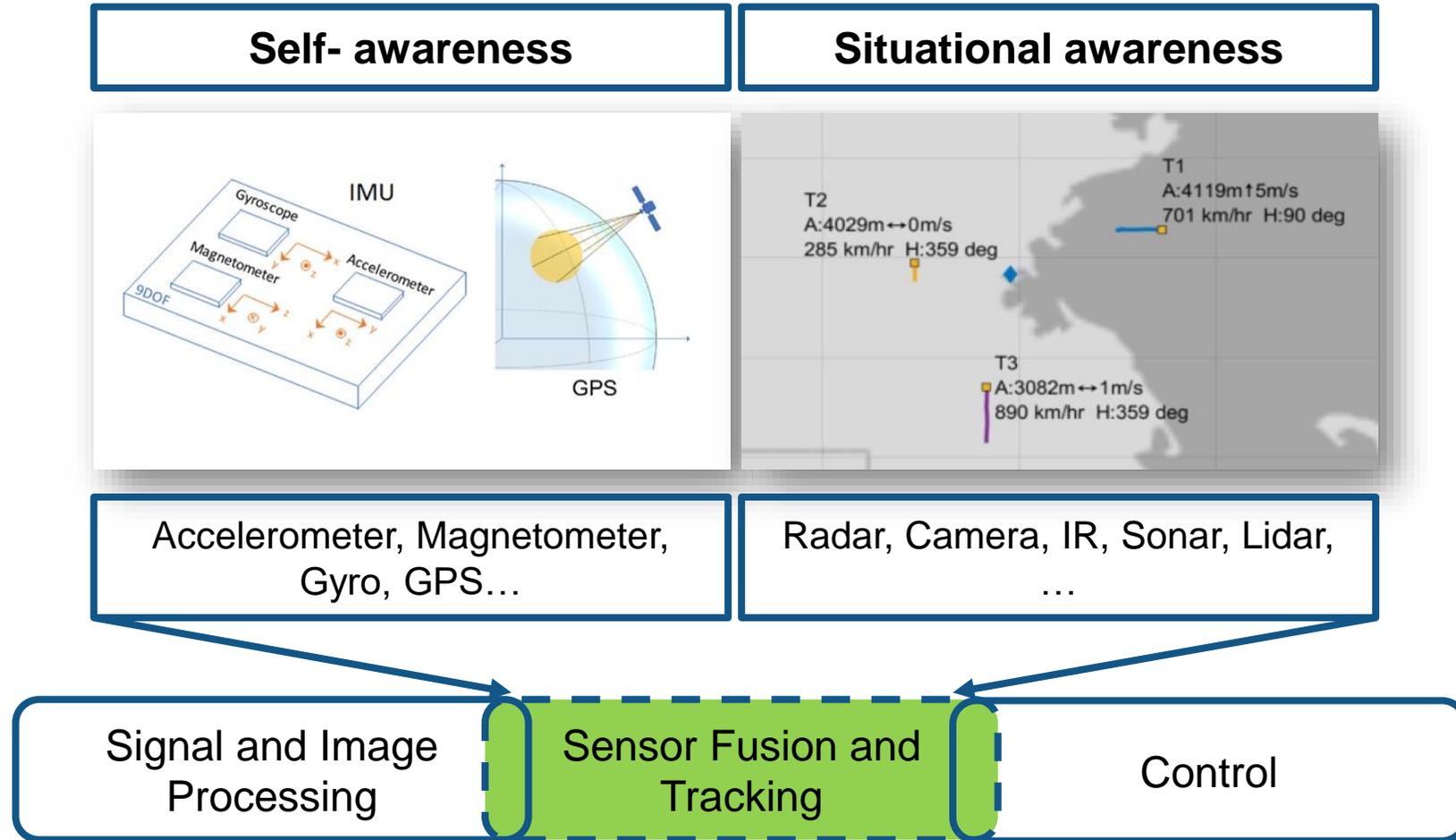
Capabilities of an Autonomous System



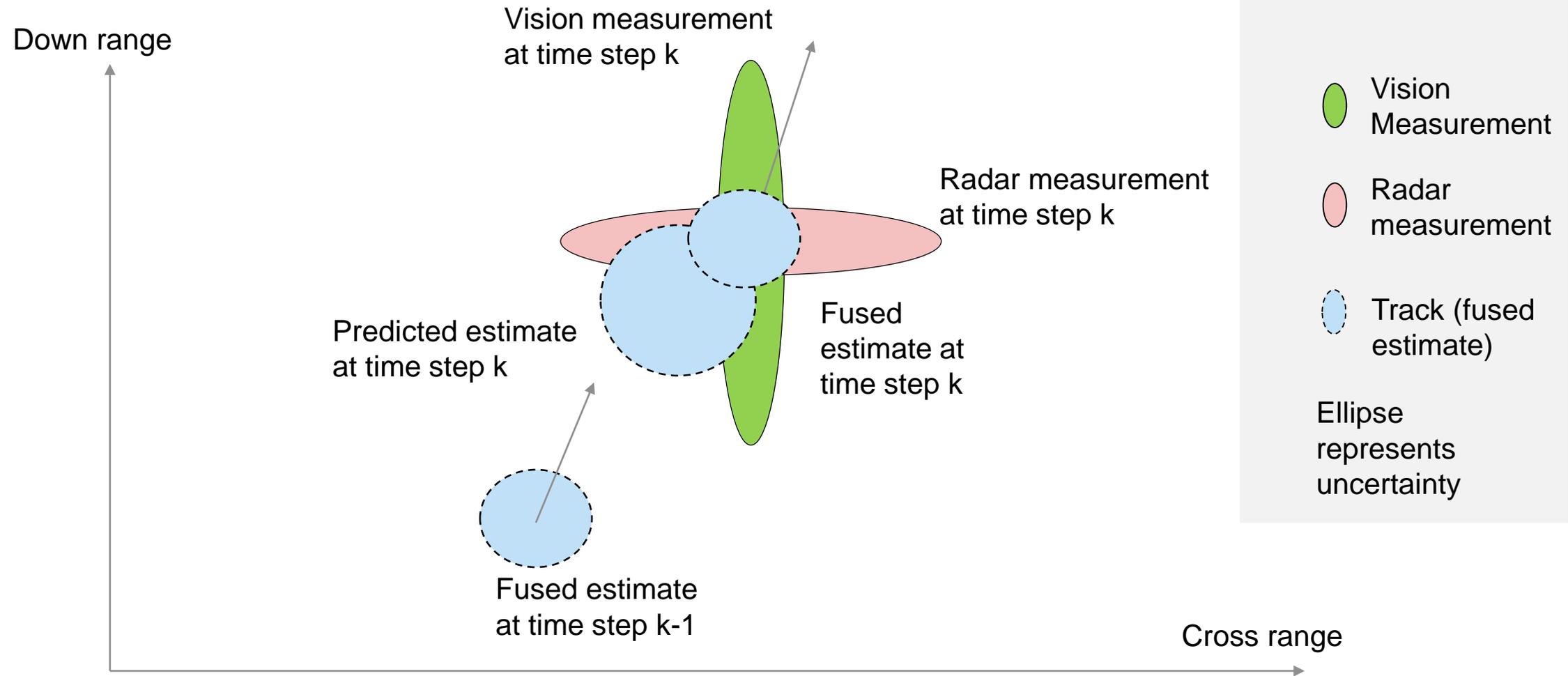
Agenda

- Introduction
- Technology overview of perception
- Algorithm development for sensor fusion and tracking
- Q&A
- Resources for further exploration

Sensor fusion and tracking is...

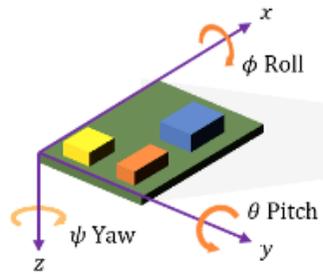


Fusion combines the strengths of each sensor

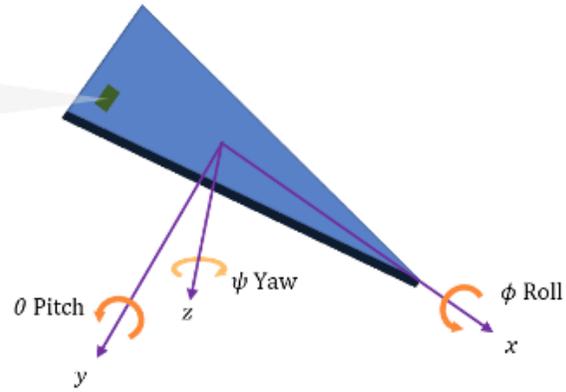


What is localization?

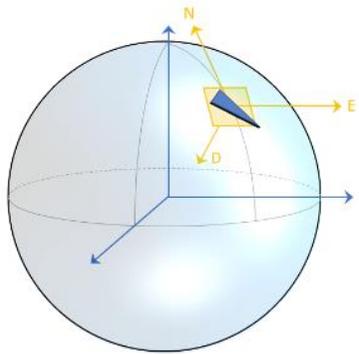
Sensor Frame



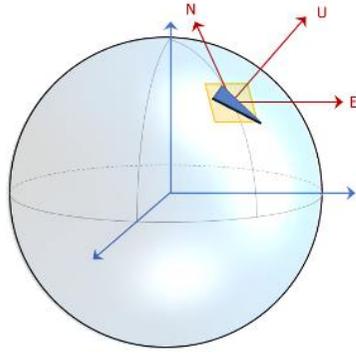
Body Frame



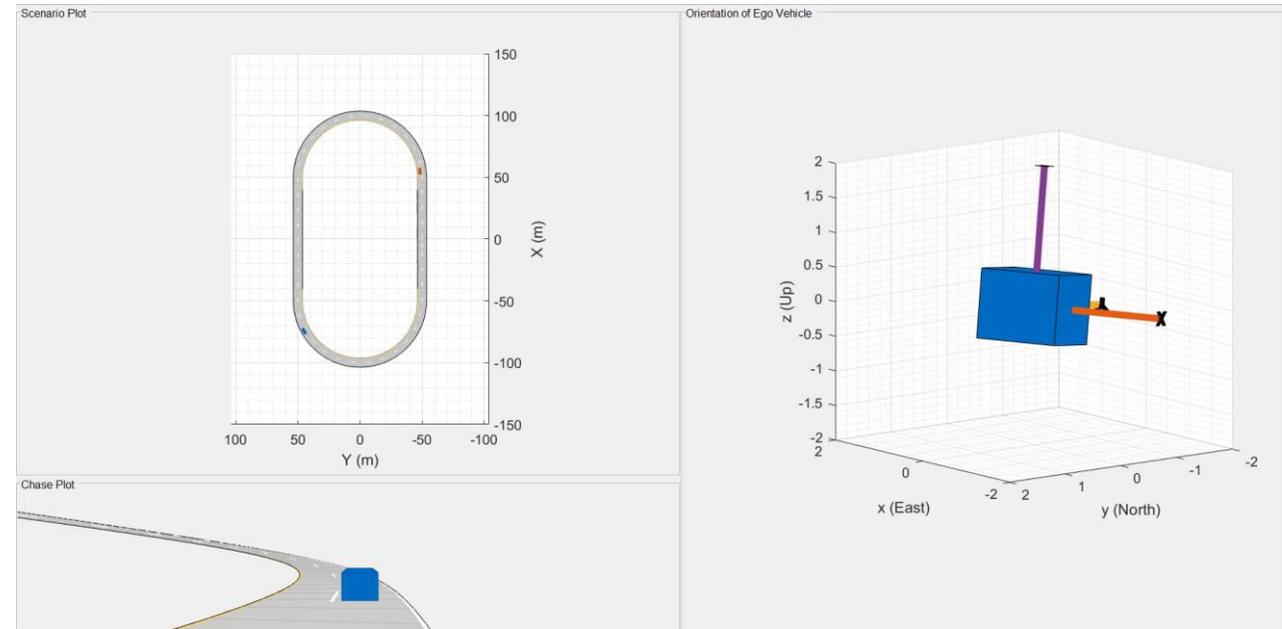
Ground Reference Frame



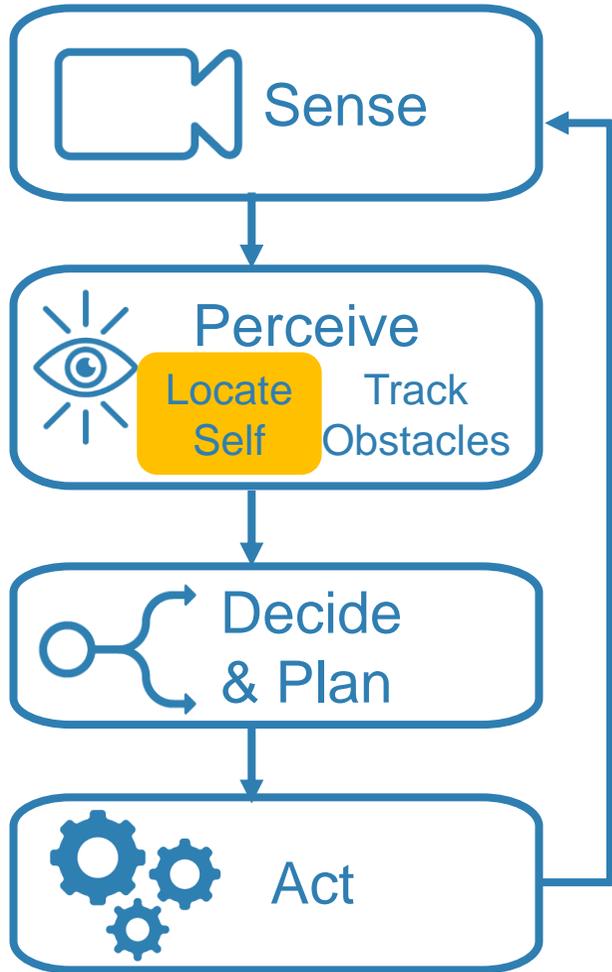
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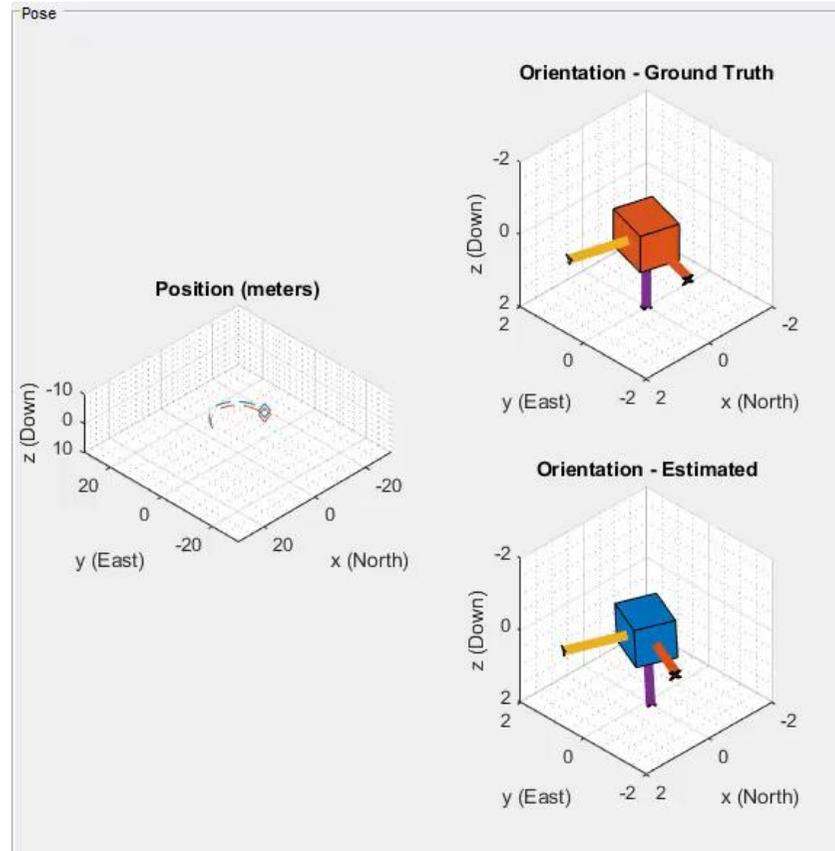
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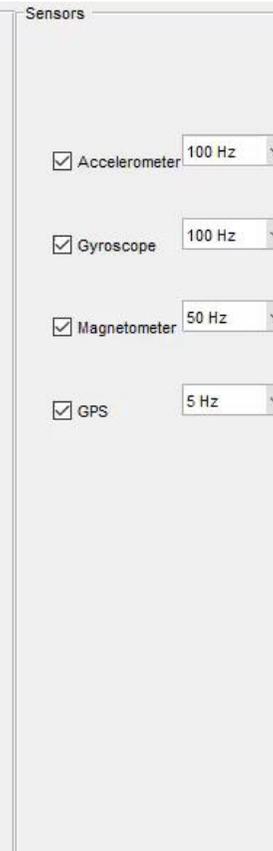
Fusing sensor data improves localization



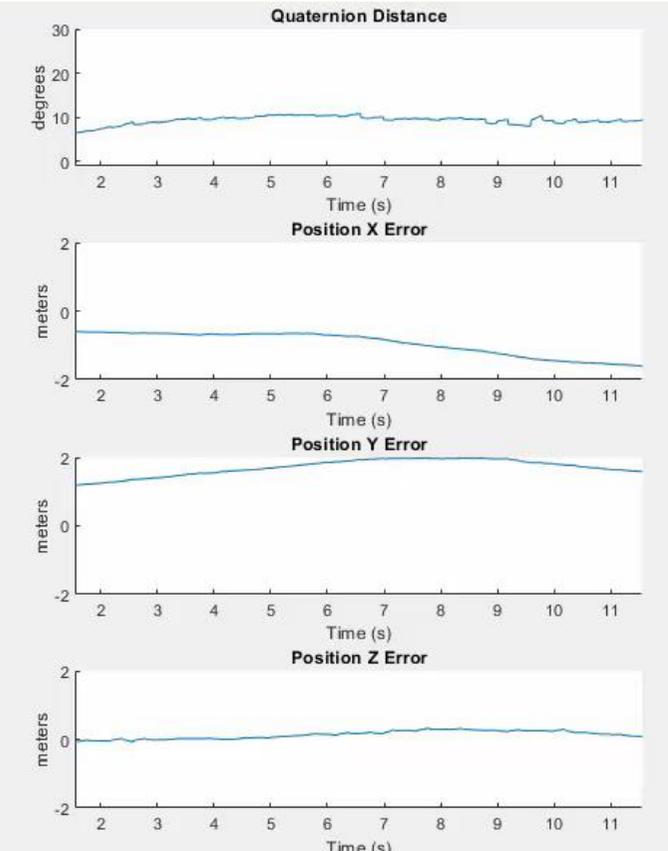
Ground truth vs. Estimate



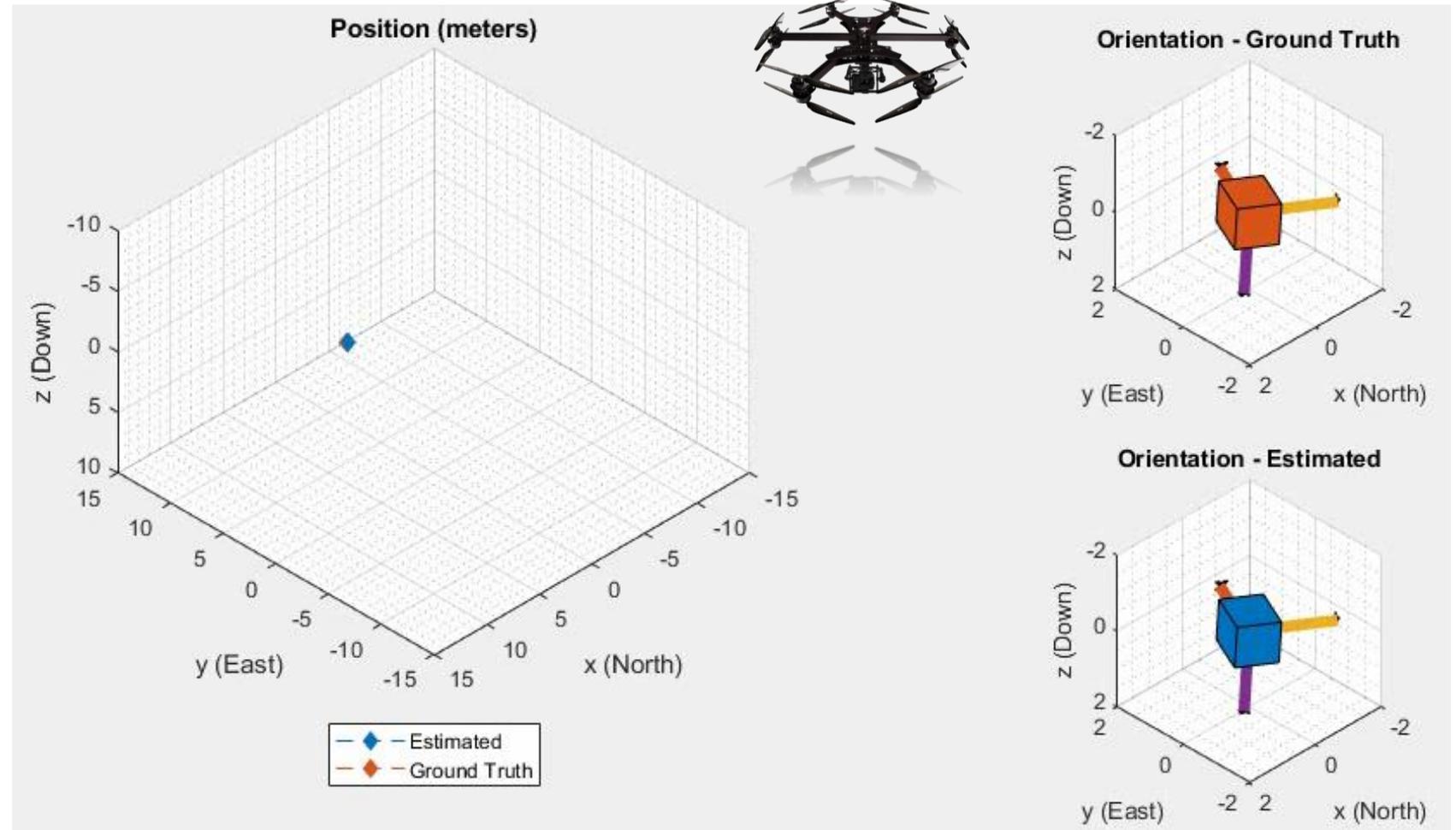
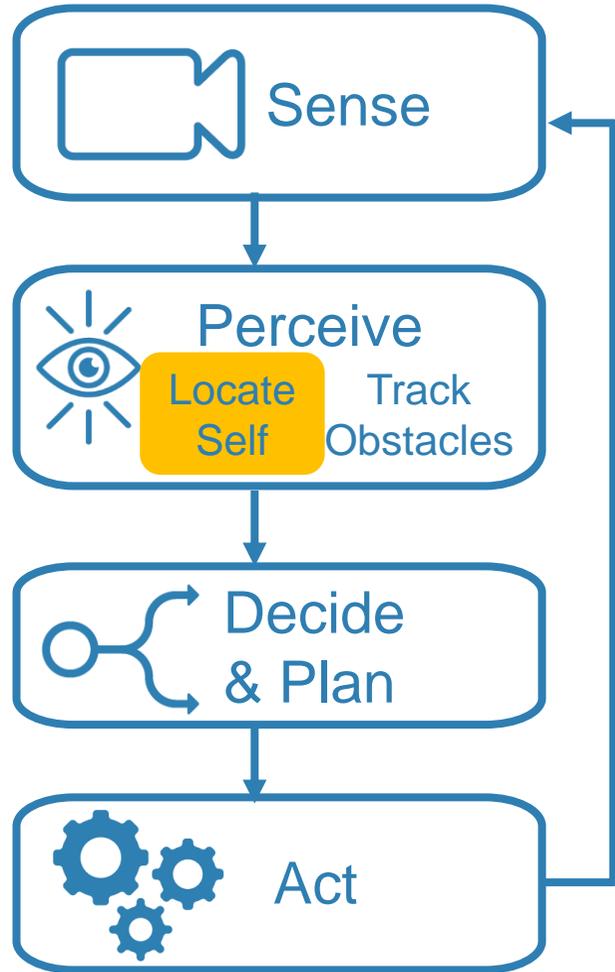
Sensors



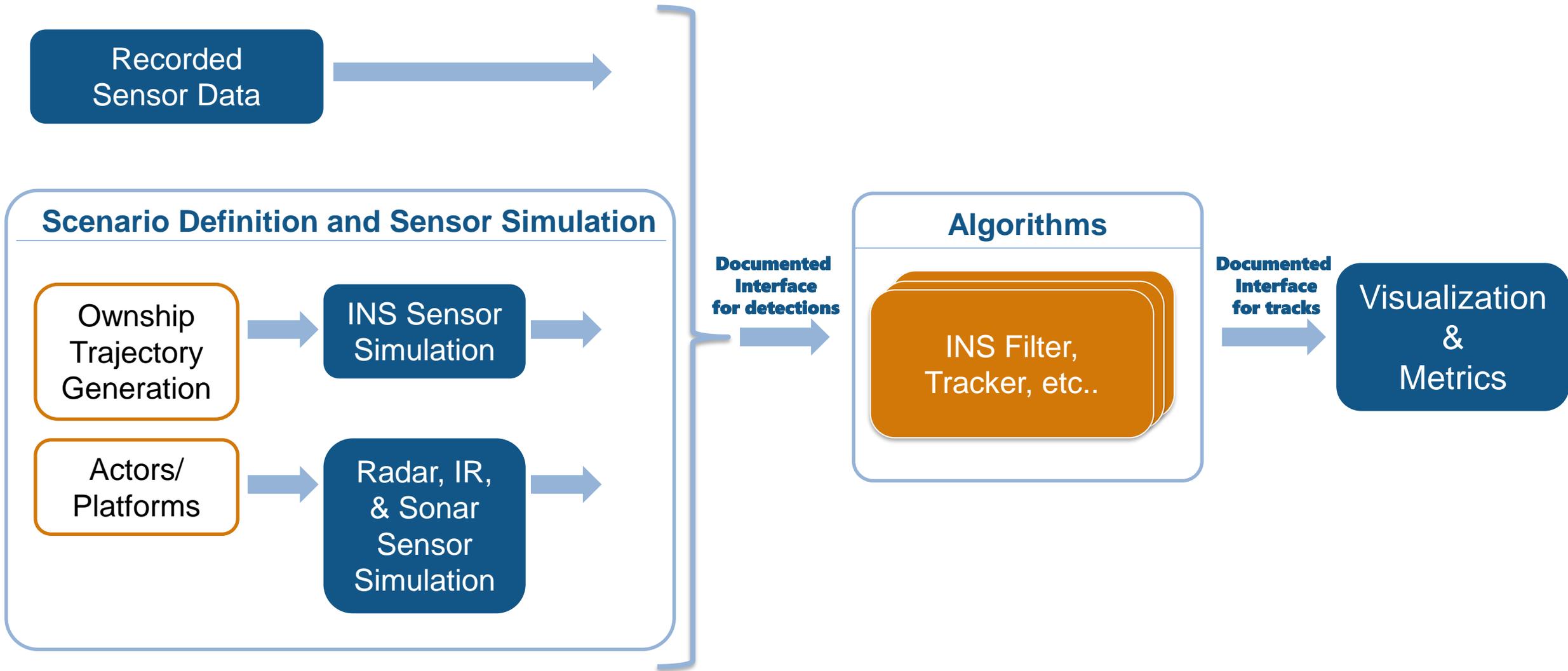
Error Measurements



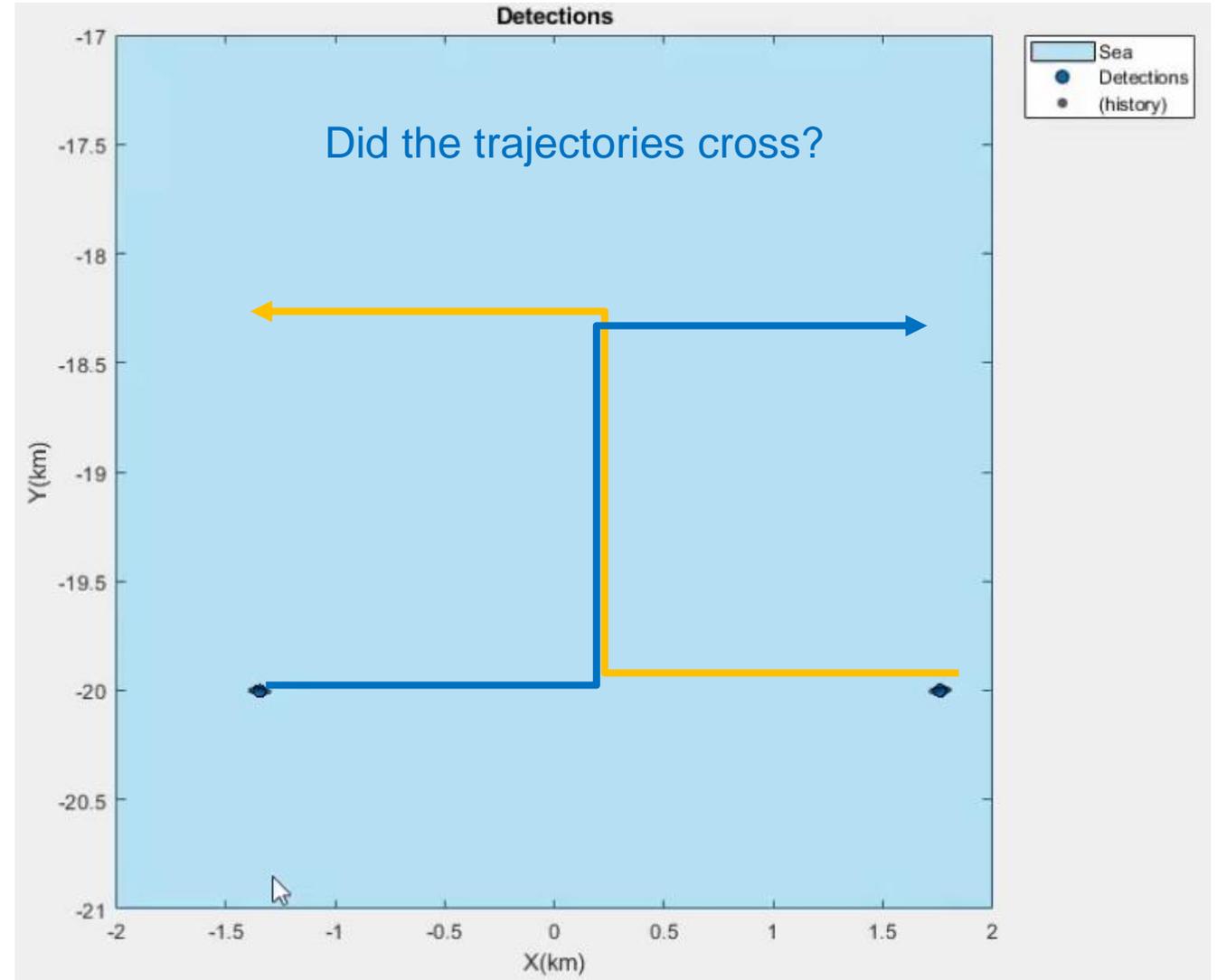
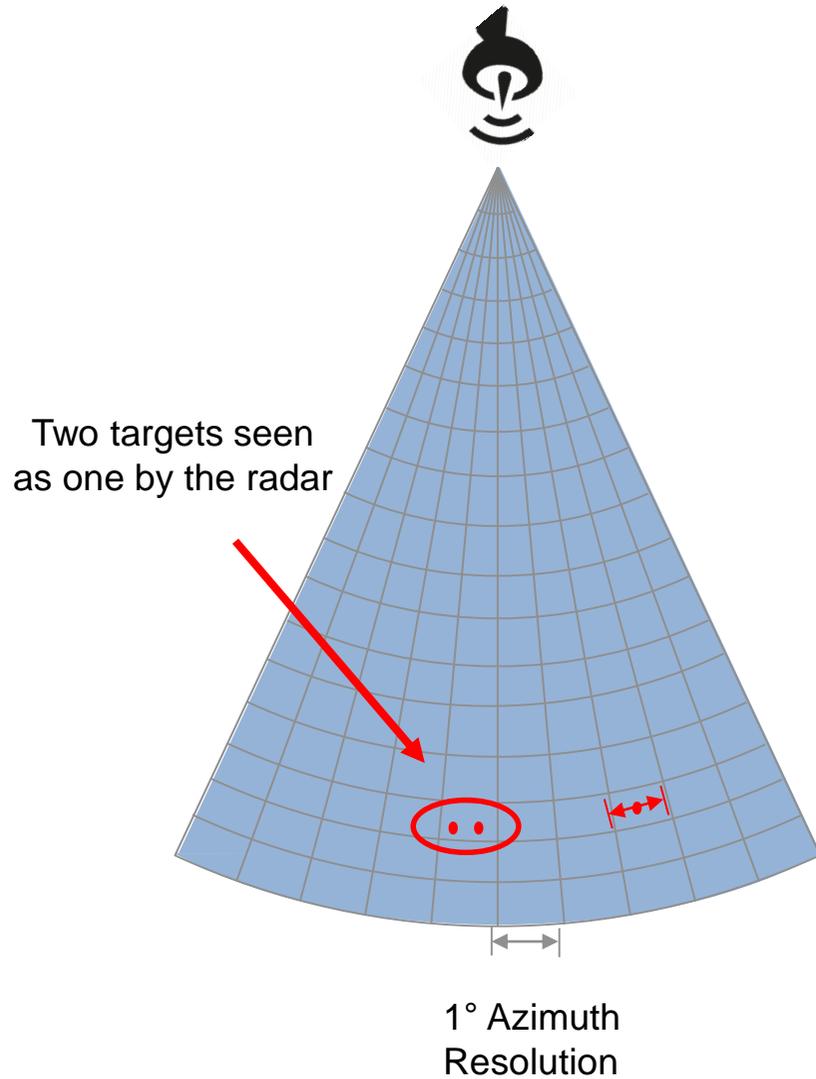
Self-awareness is needed to create situational awareness



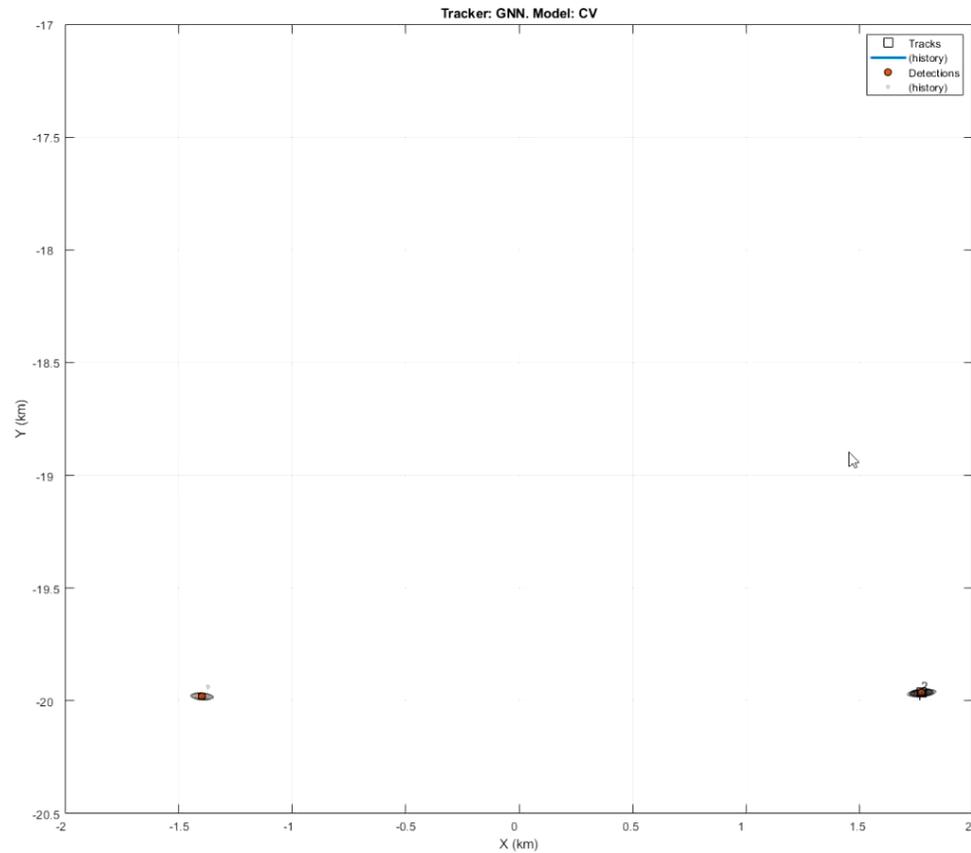
Flexible Workflows Ease Adoption: Wholesale or Piecemeal



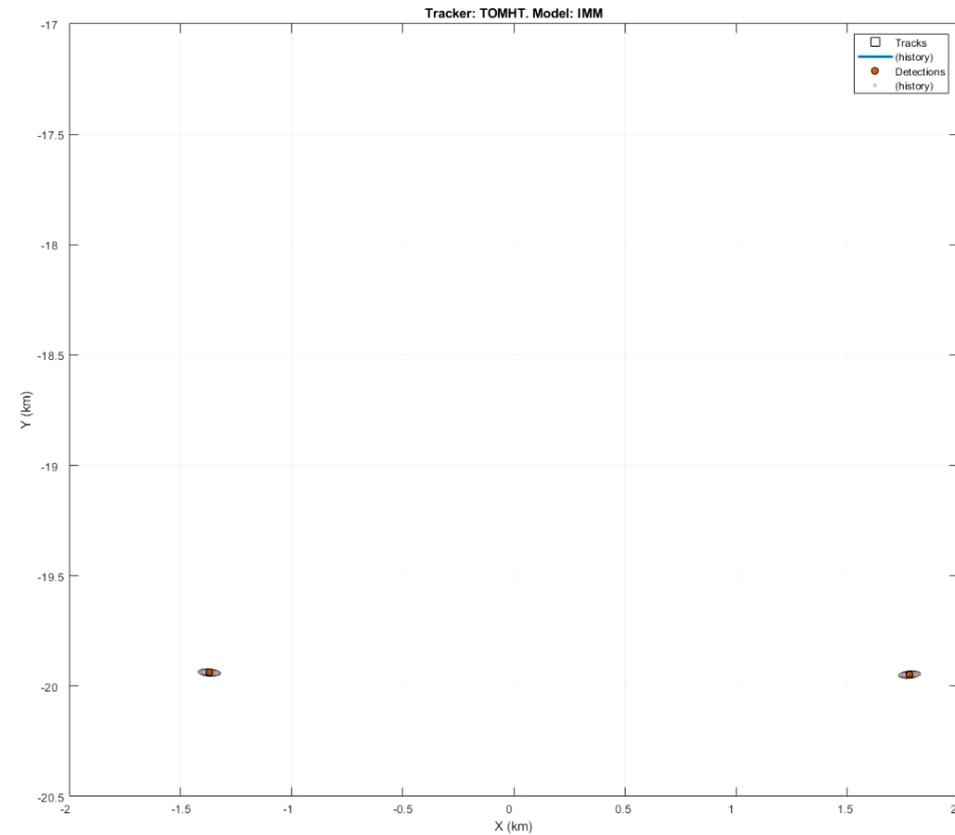
You can synthesize test cases that may be difficult to re-create



Perform quick what-if analyses between different trackers

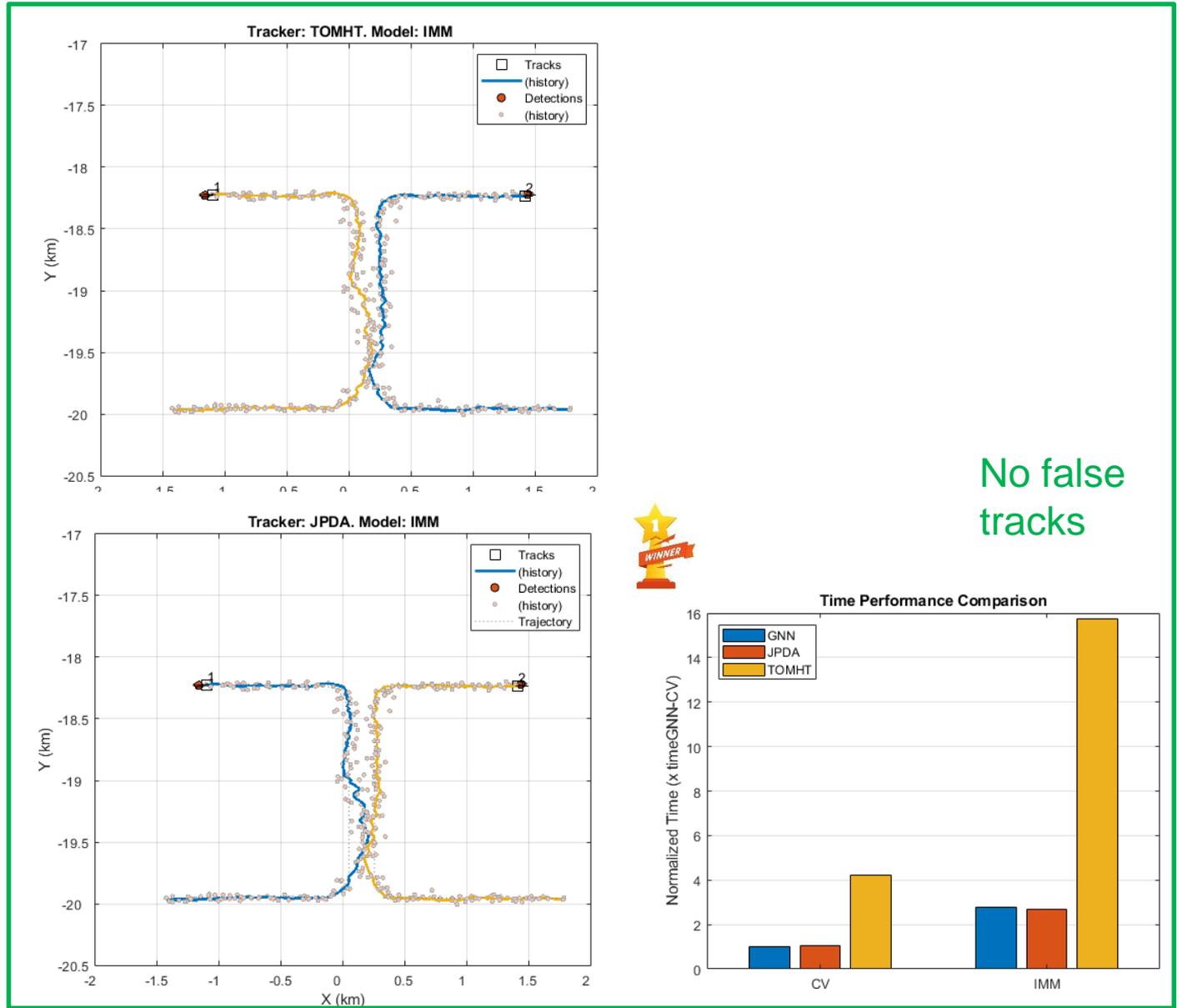
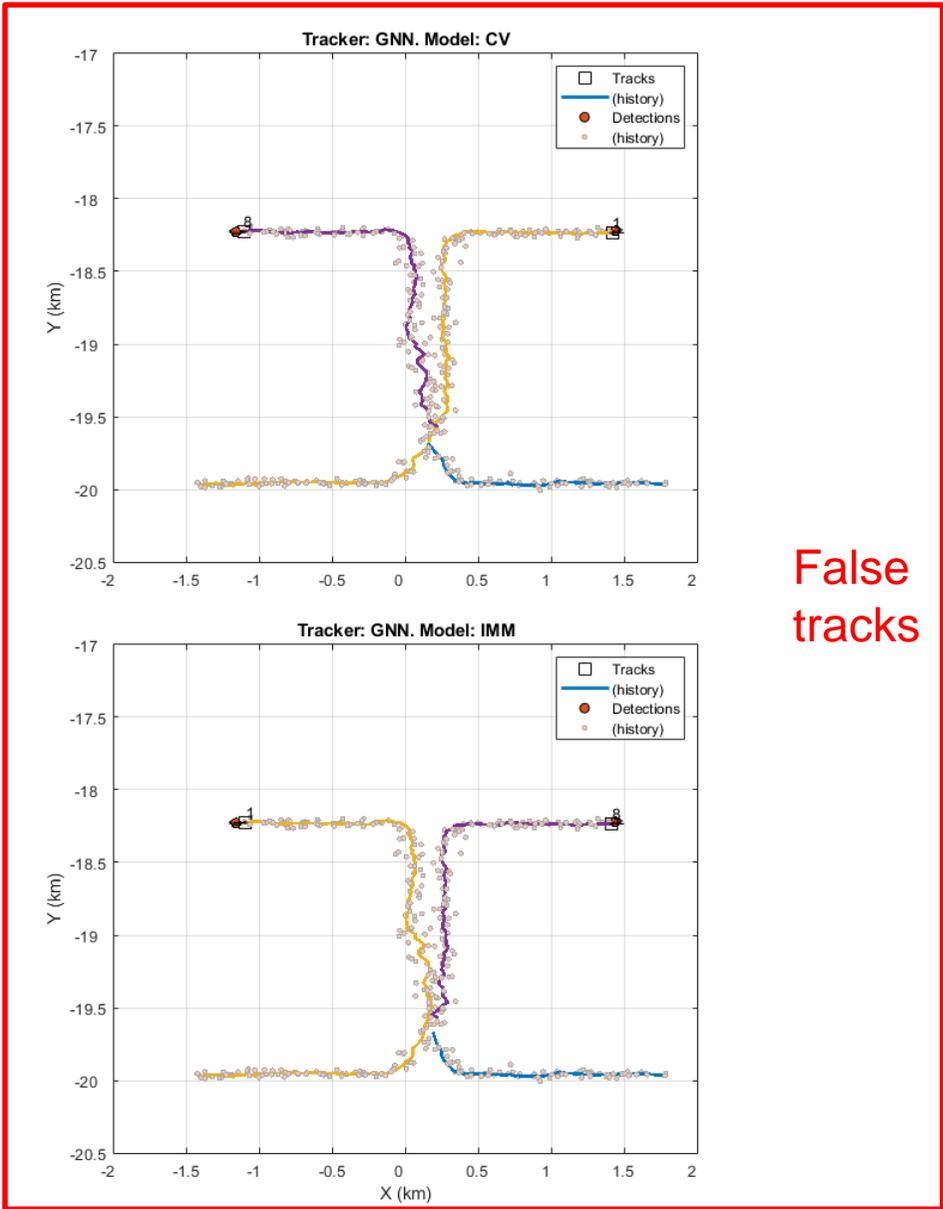


>> trackerGNN, CV



>> trackerTOMHT, IMM

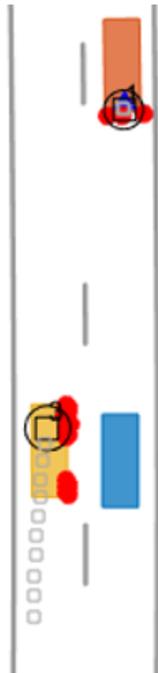
Evaluate results based on performance metrics



Point object vs. Extended object

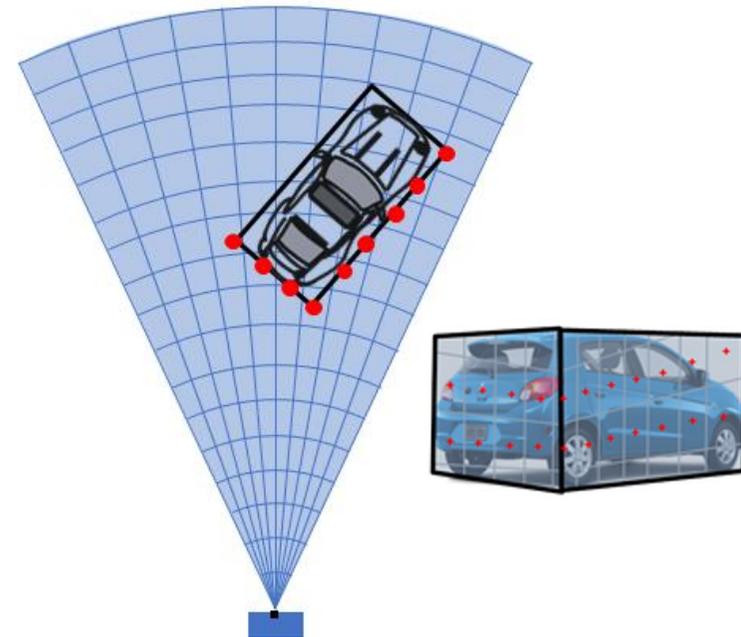
- **Point object**

- Distant object represented as a single point
- One detection per object per scan

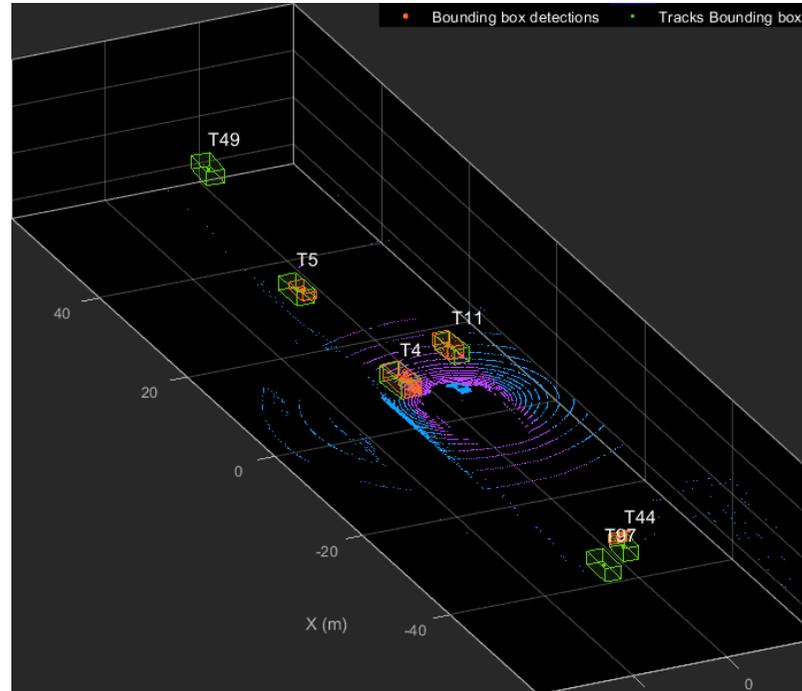
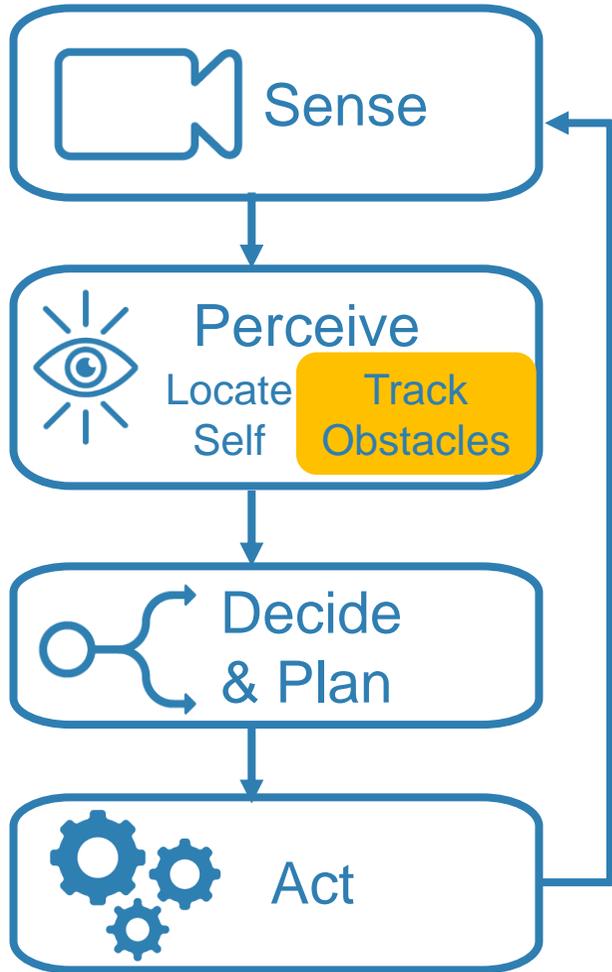


- **Extended object**

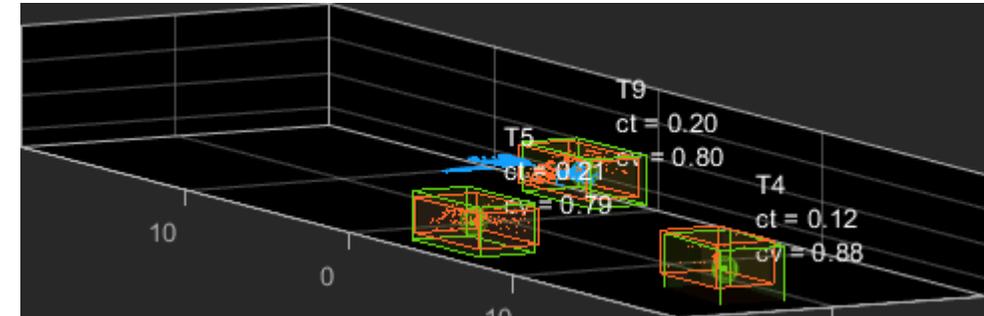
- High resolution sensors generate multiple detections per object per scan



Tracking with Lidar



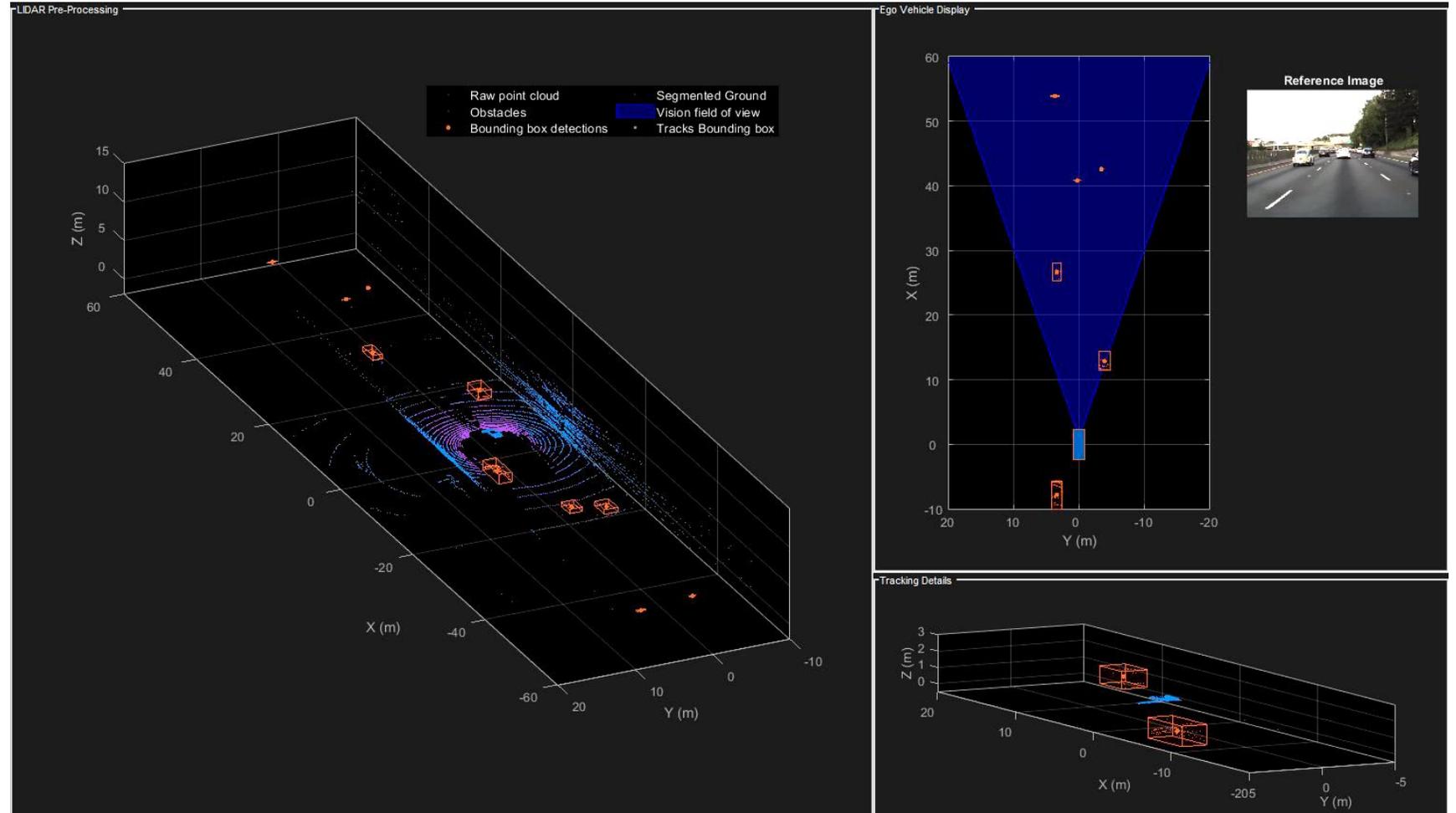
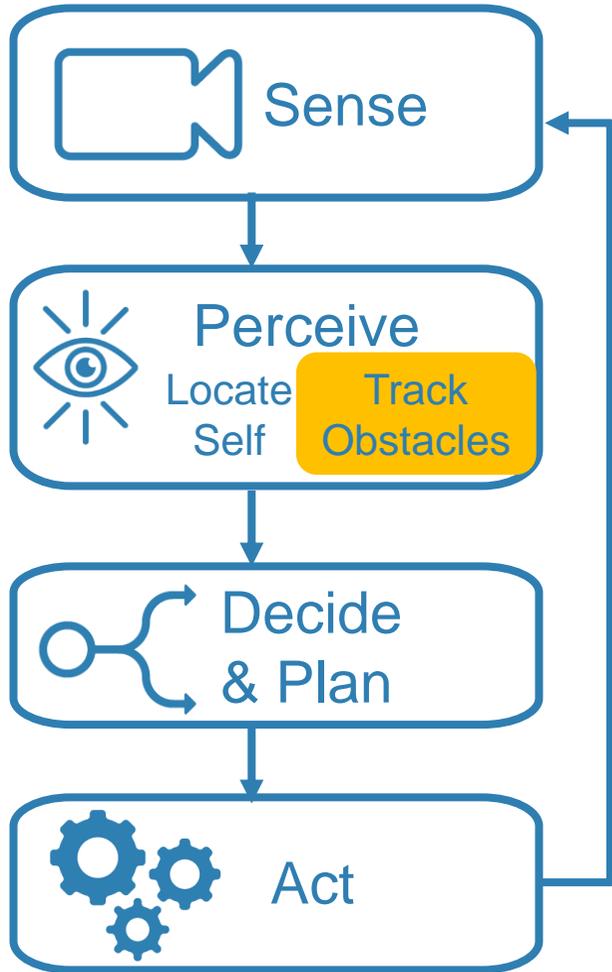
Maintain tracks at edge of coverage



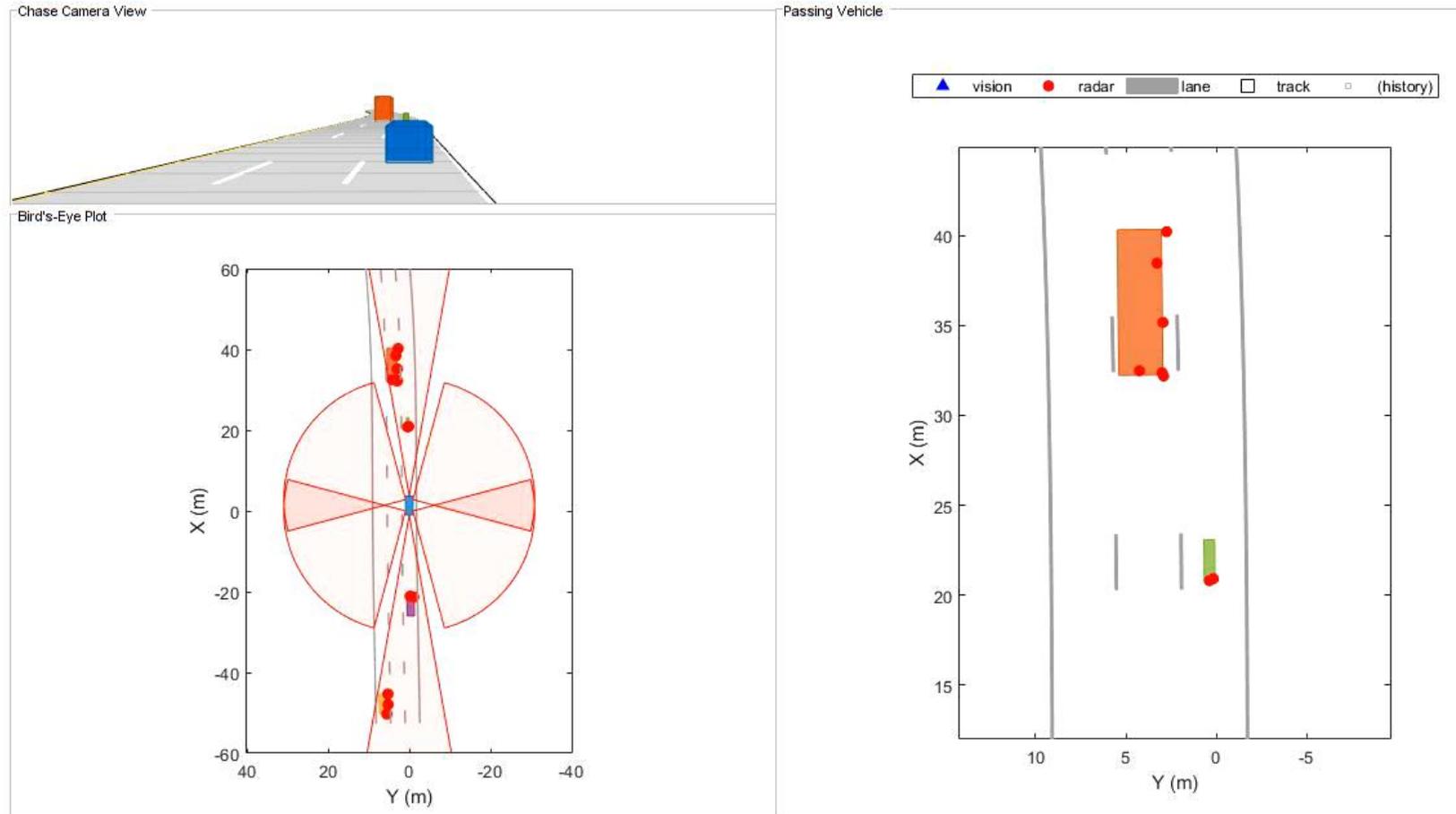
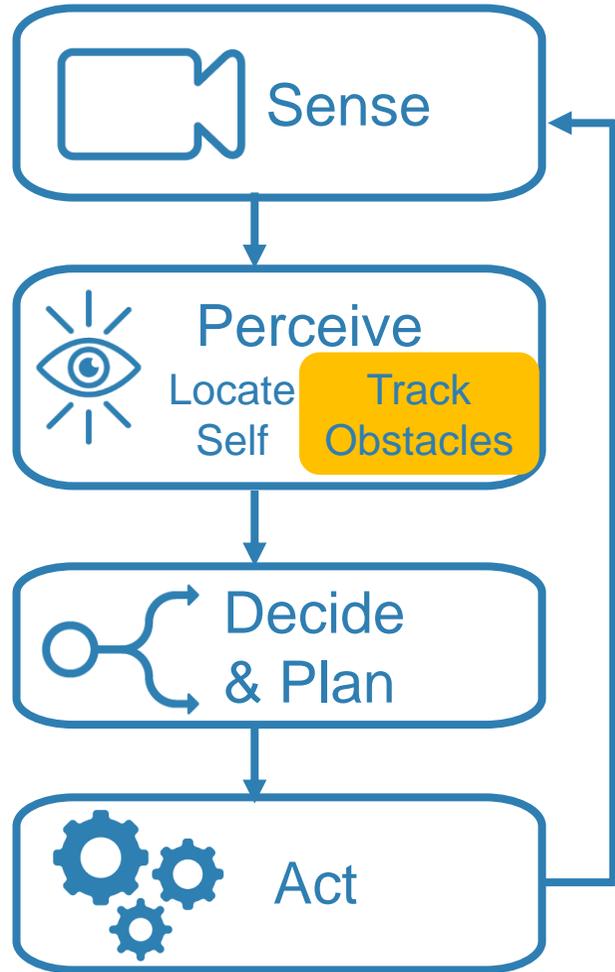
Track maneuvering vehicles (during lane changes)

- Design 3-D bounding box detector
- Design IMM-JPDA tracker
- Generate C/C++ code for tracker

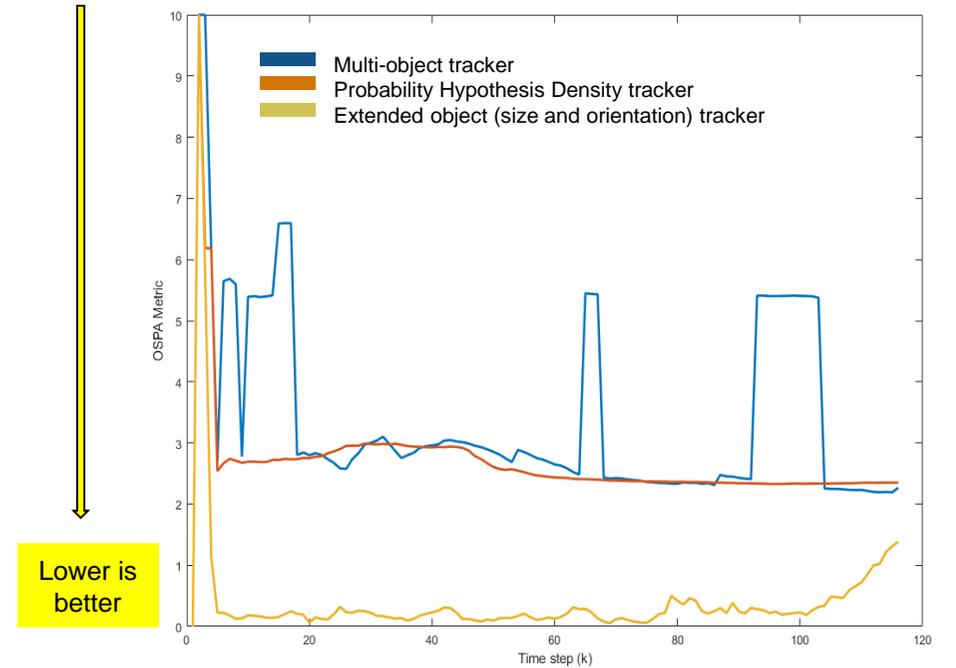
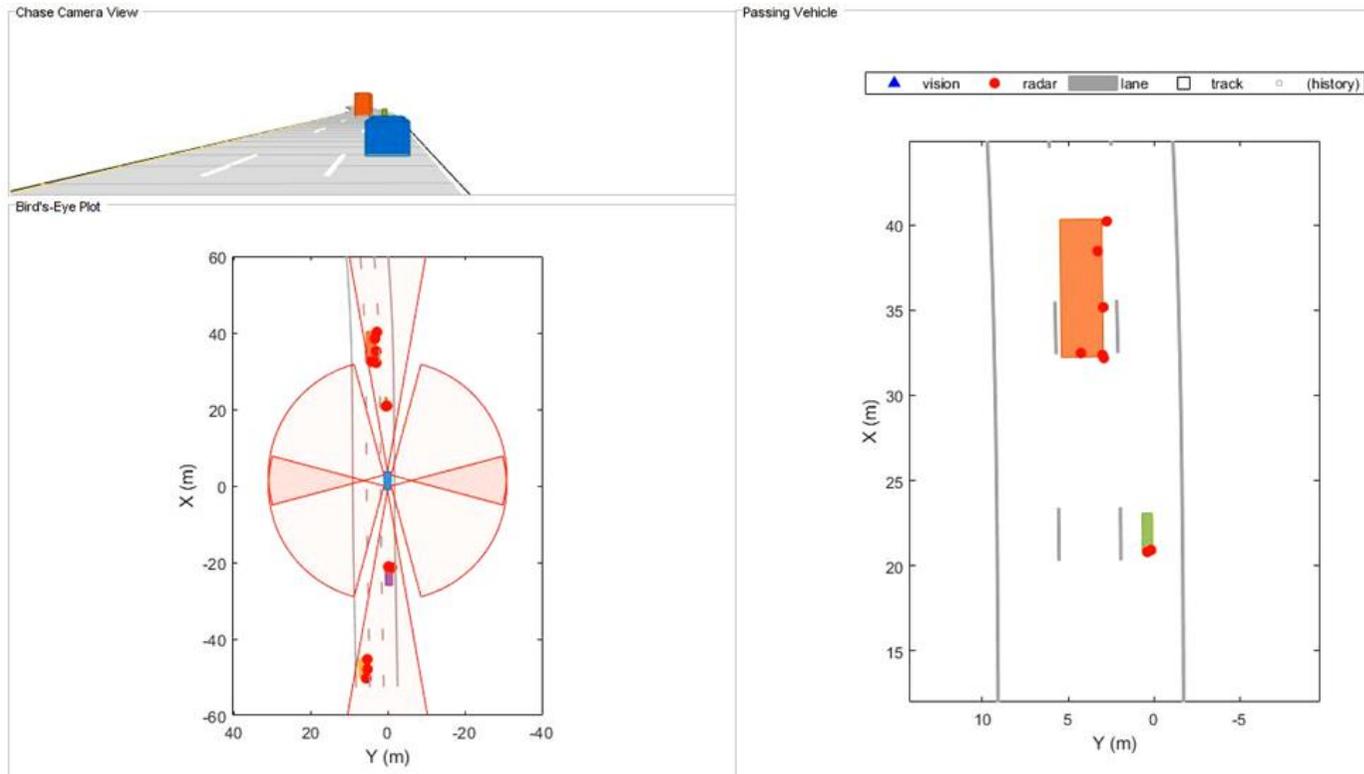
Tracking with Lidar (with ground truth shown)



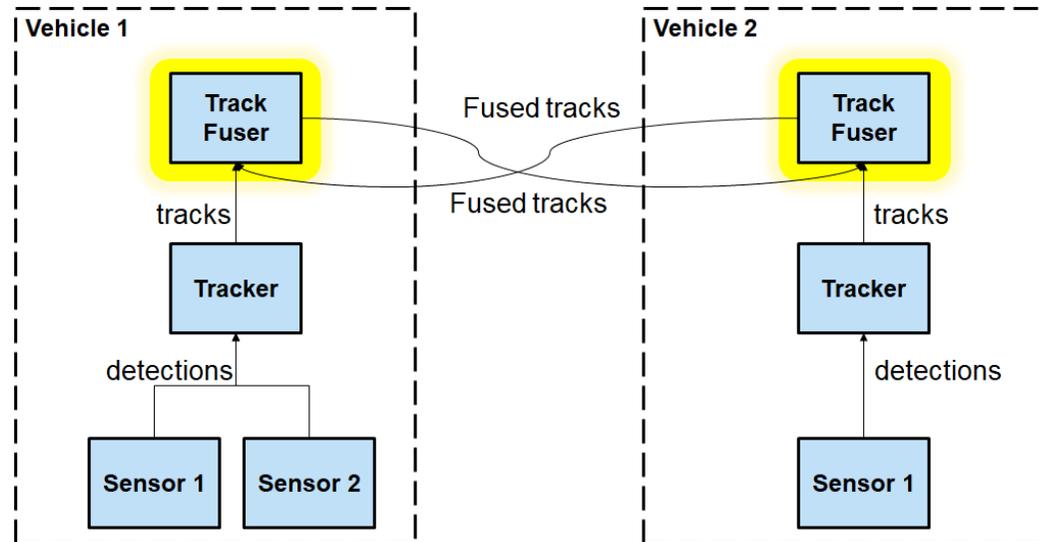
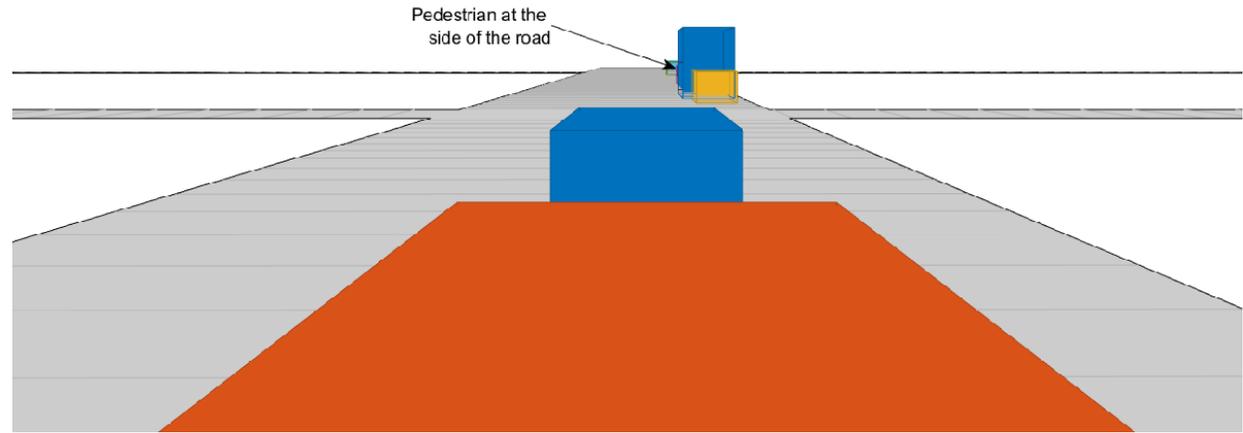
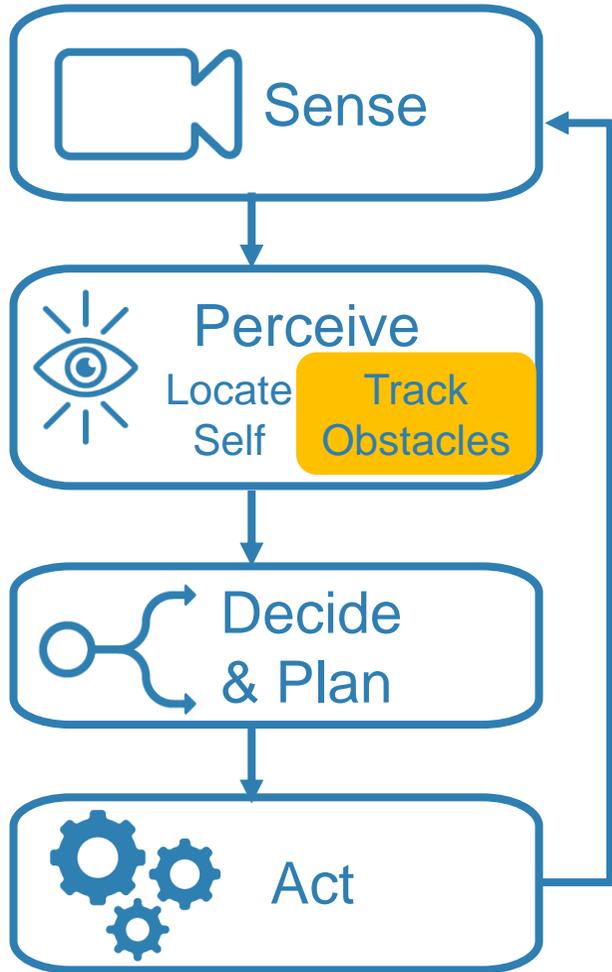
Extended Object Tracking: Estimate position, velocity, and size



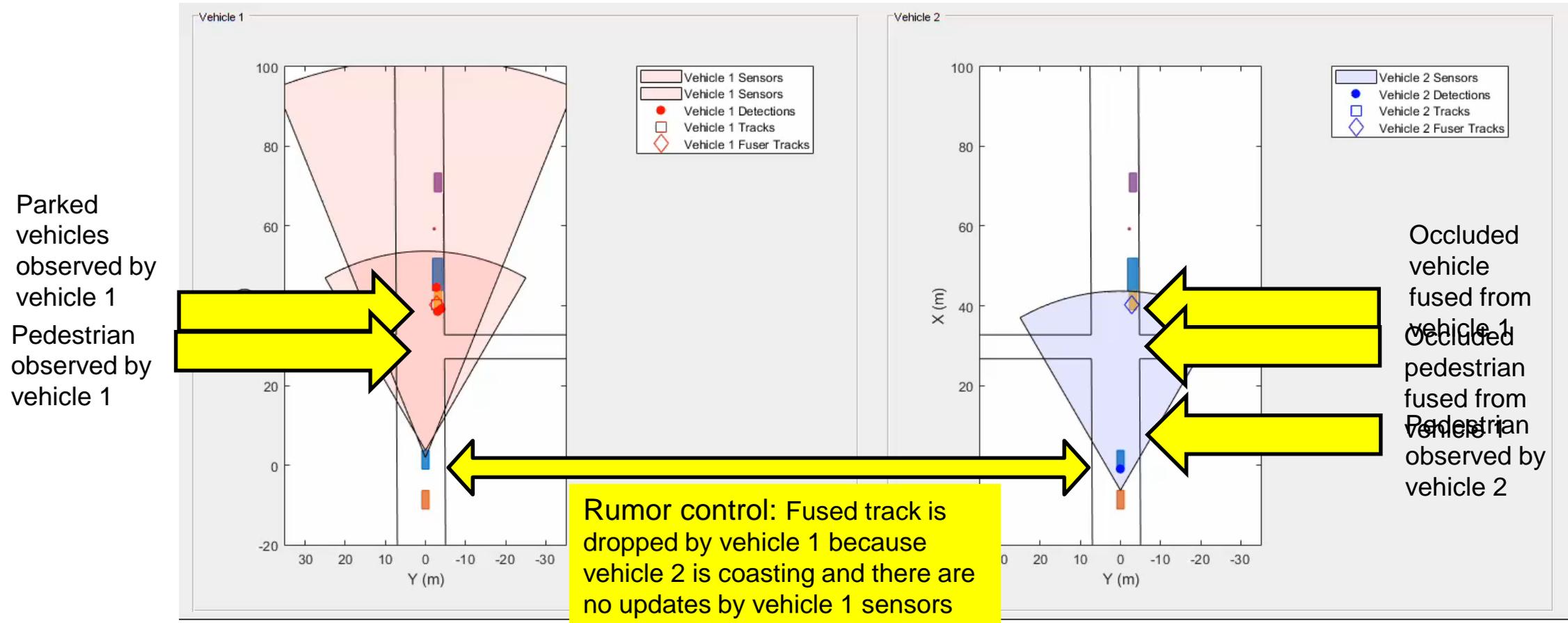
Evaluate tracker performance of Extended Objects



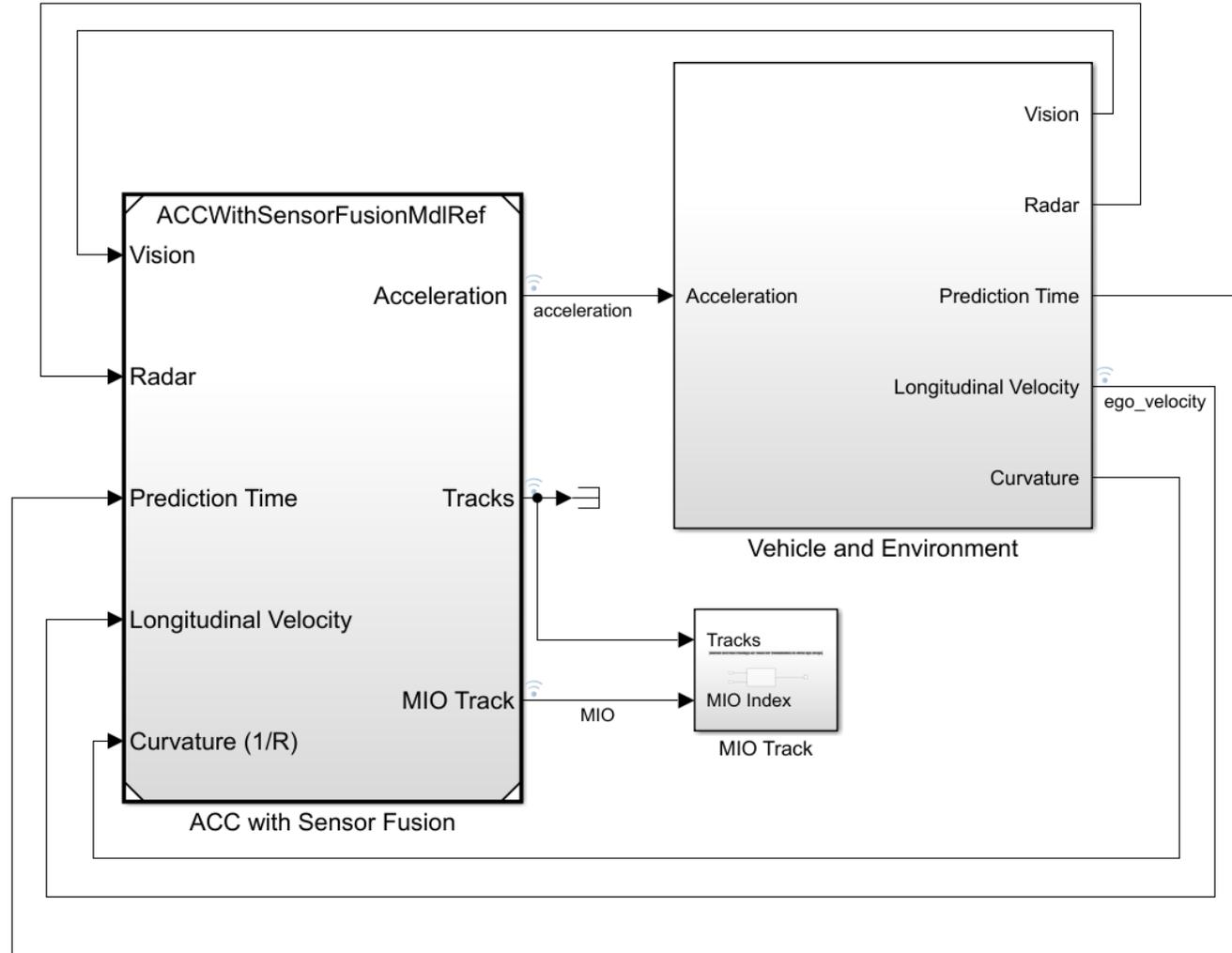
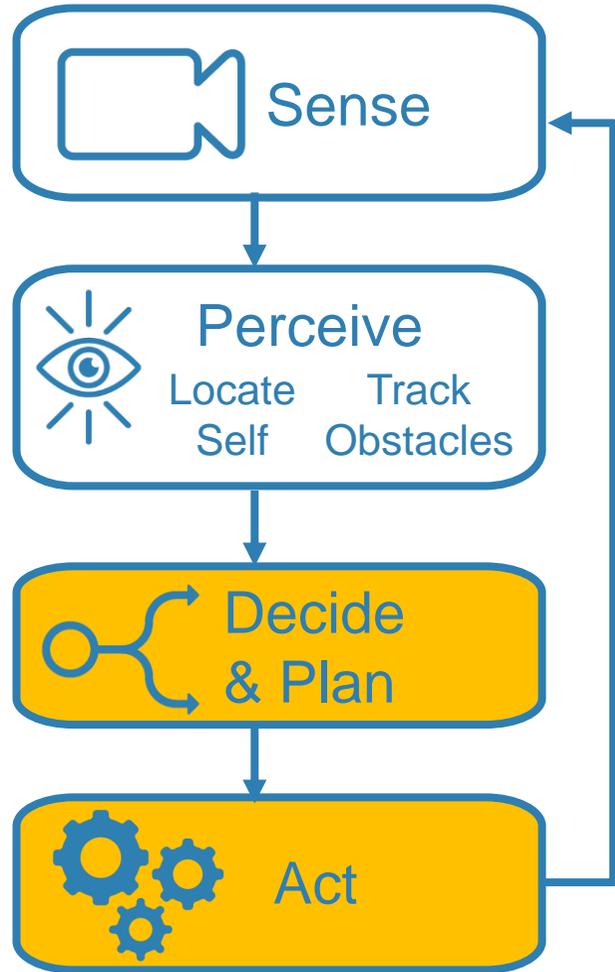
Fusing tracks is also easy



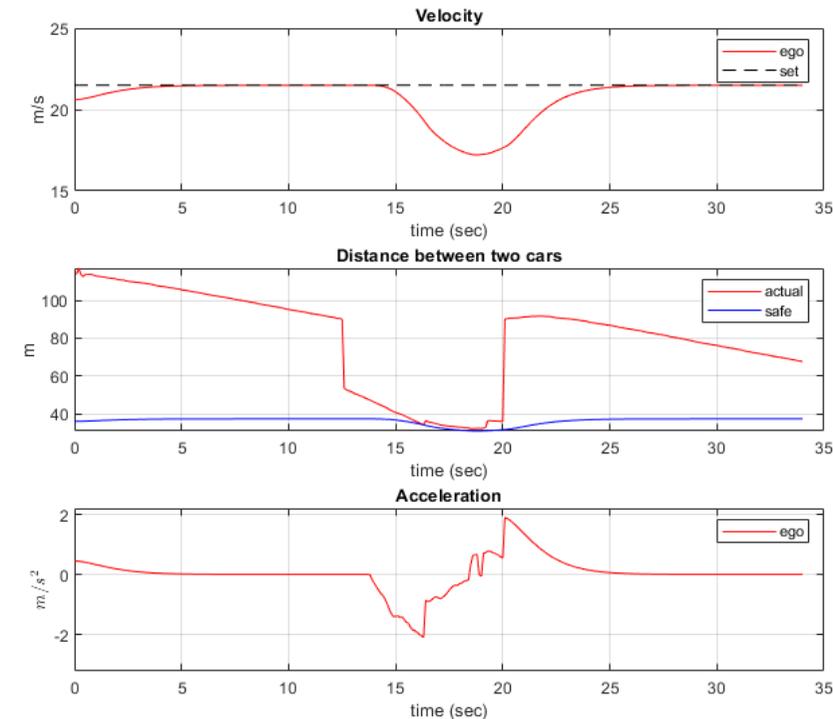
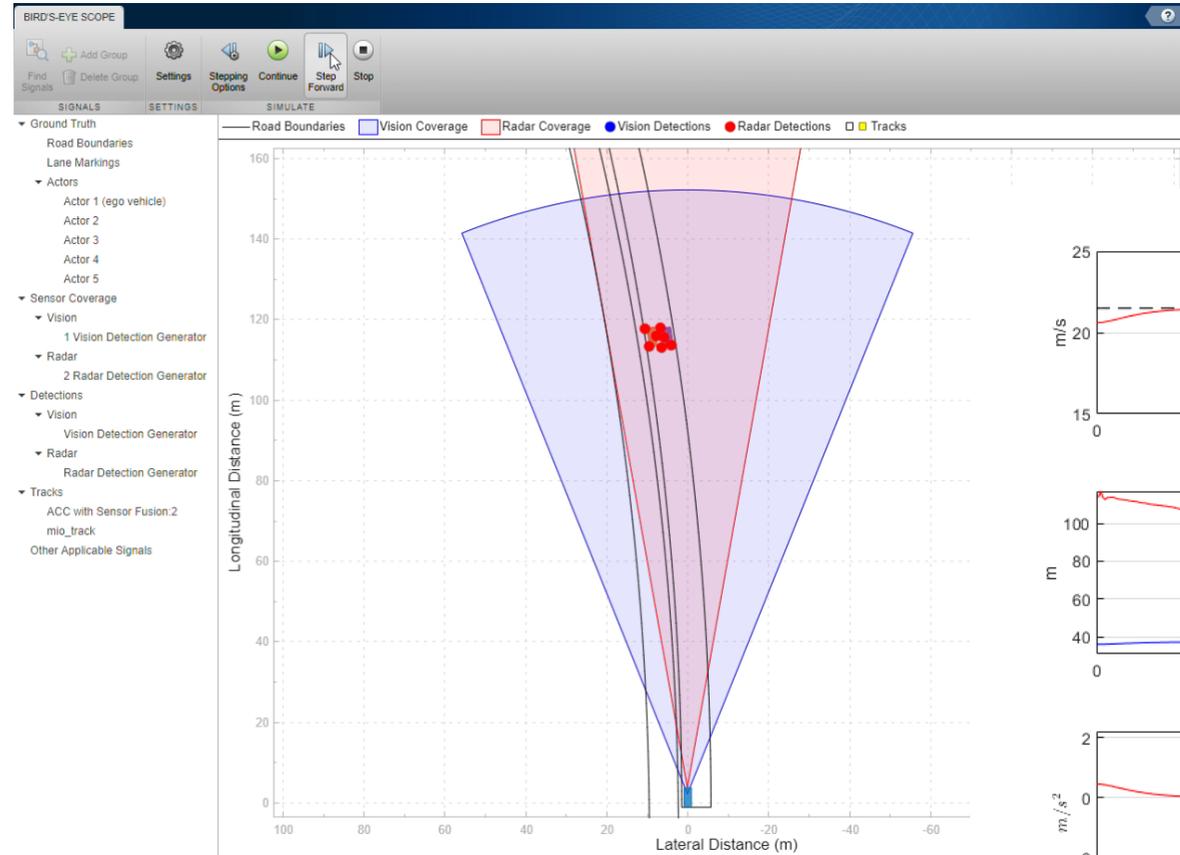
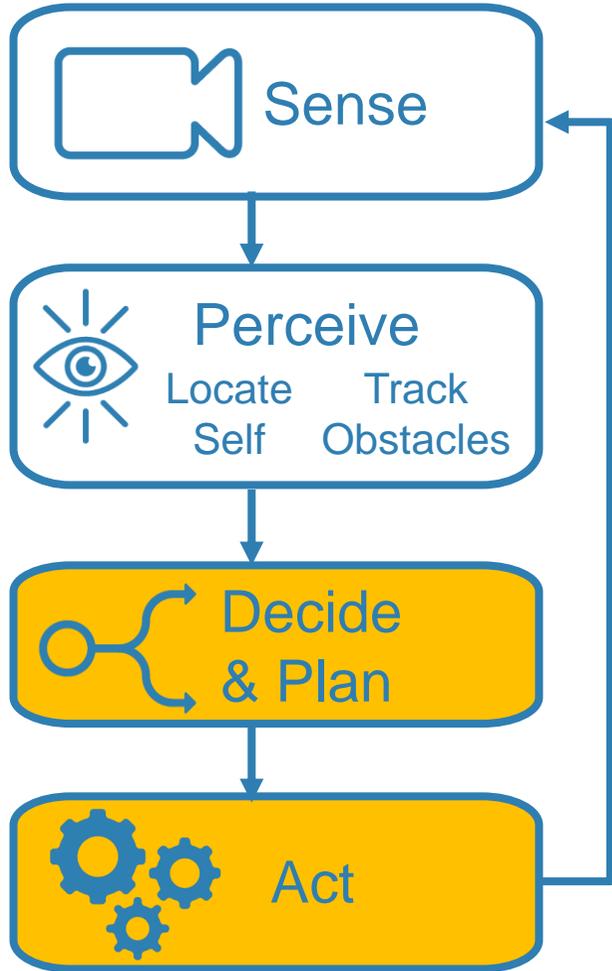
Track-to-Track Fusion



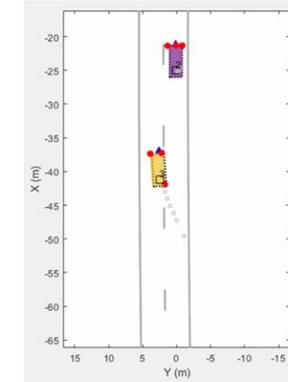
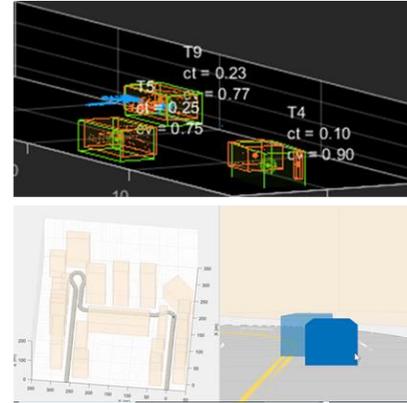
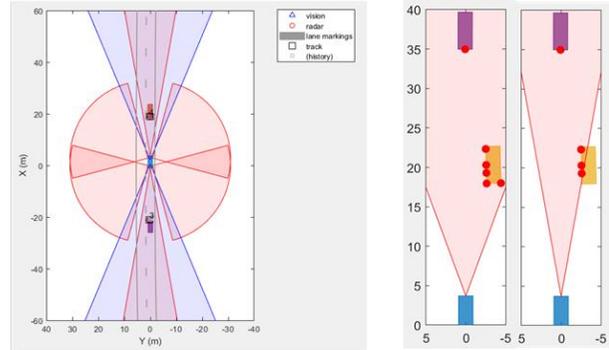
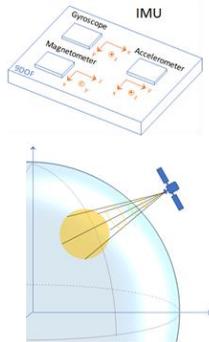
Connect to Decision Making / Control: Adaptive Cruise Control



Connect to Decision Making / Control: Adaptive Cruise Control

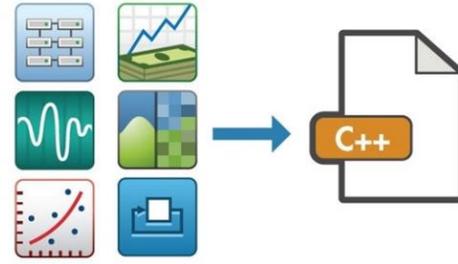
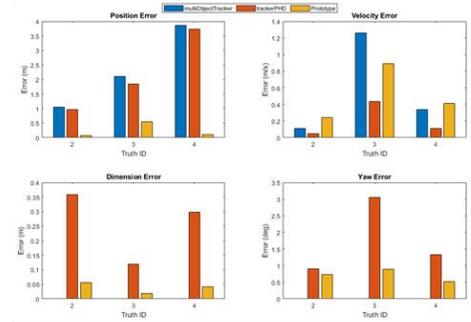
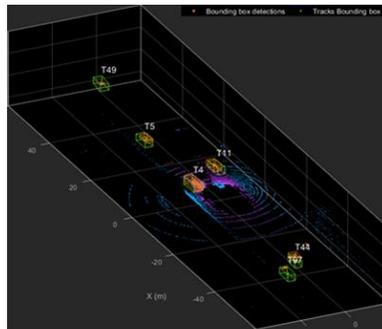


Sensor Fusion and Tracking Toolbox Summary



Scenarios and Sensors Simulation

Tracking and Localization Algorithms



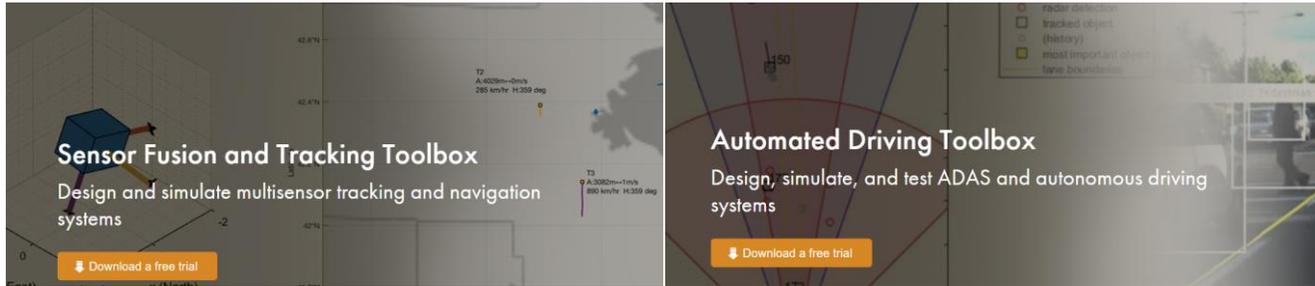
Visualization and Metrics

Code Generation



There are many resources to get started with

Tech Talks



MathWorks

QUICK START GUIDE

Getting Started with Sensor Fusion and Tracking Toolbox™

Definitions of Localization-Related Terms

Accelerometer: a sensor that measures the object acceleration.
 Gyroscope: a sensor that measures the object angular velocity.
 Magnetometer: a sensor that measures the magnetic field around the object.
 IMU: inertial measurement unit. A device that consists of accelerometers and gyroscopes.
 MARG: magnetic, angular rate, and gravity (aka magnetometer, gyroscope, and accelerometer).
 ARB: attitude and heading reference system. A system that fuses accelerometers, gyroscopes, and magnetometers and provides object attitude information. (AKA AHRS) Real-time algorithms.
 GPS: global positioning system. A satellite-based system that provides accurate positioning.
 INS: inertial navigation system. A system that fuses data from accelerometers, gyroscopes, magnetometers, and sometimes ultrameters to continuously calculate the position, orientation, and velocity of moving objects without an external source.
 GPS/INS: A system that fuses GPS information with INS information.

Types of Tracking Filters and How to Choose the Right One

Filter Name	Supports Non-Linear Models	Gaussian Noise	Computational Complexity	Comments
Alpha-Beta				Sub-optimal.
Kalman		✓		Optimal for linear systems.
Extended Kalman	✓	✓		Use linearized models to propagate uncertainty covariance.
Uncentred Kalman	✓	✓		Sample the uncertainty covariance to propagate it. May become numerically unstable in single-precision.
Cubature Kalman	✓	✓		Sample the uncertainty covariance to propagate it. Numerically stable.
Gaussian-Sum	✓	Assume a weighted sum		Good for partially observable case (e.g., single-object tracking).
Interacting Multiple Models (IMM)	Multiple Models	Assume a weighted sum of distributions		Manoeuvring objects (e.g., accelerate, turn).
Particle	✓	Can be any distribution		Sample the uncertainty distribution using weighted particles.

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Quick Start Guide

Series: Understanding Sensor Fusion and Tracking



Part 1: What is Sensor Fusion?

This video provides an overview of what sensor fusion is and how it helps in the design of autonomous systems. It also covers a few scenarios that illustrate the various ways in which sensor fusion can be implemented.



Part 2: Fusing a Mag, Accel, and Gyro to Estimate Orientation

This video describes how we can use a magnetometer, accelerometer, and a gyro to estimate an object's orientation. The goal is to show how these sensors contribute to the solution, and to explain a few things to watch out for along the way.



Part 3: Fusing a GPS and IMU to Estimate Pose

This video describes how we can use a GPS and an IMU to estimate an object's orientation and position. We'll go over the structure of the algorithm and show you how the GPS and IMU both contribute to the final solution.



Part 4: Tracking a Single Object With an IMM Filter

This video describes how we can track a single object by estimating state with an interacting multiple model filter. We build up some intuition about the IMM filter and show how it is a better tracking algorithm than a single model Kalman filter.



Part 5: How to Track Multiple Objects at Once

This video describes two common problems that arise when tracking multiple objects: data association and track maintenance. We cover a few ways to solve these issues and provide a general way to approach all multi-object tracking problems.