

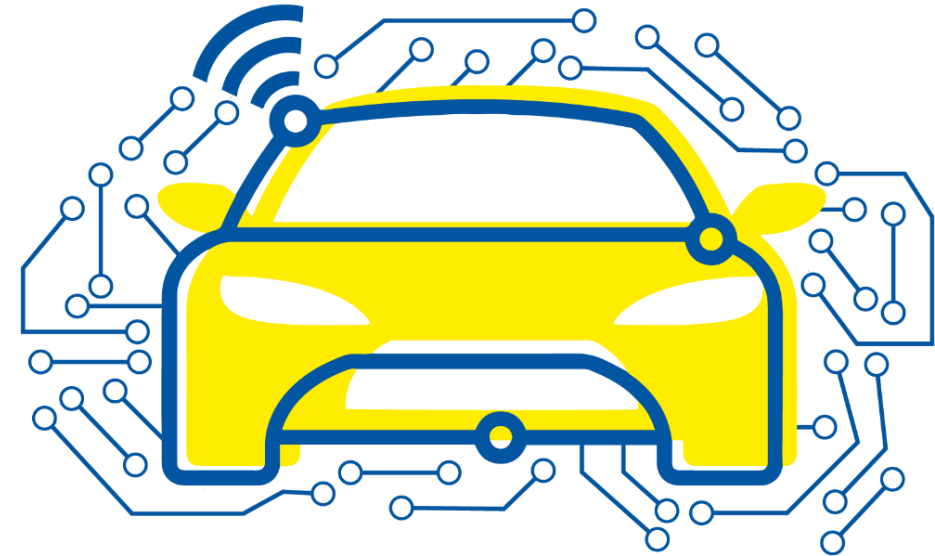
Automated and Connected Driving Challenges

Section 3 – Object Fusion and Tracking

Introduction Challenges

Prof. Dr.-Ing. Lutz Eckstein

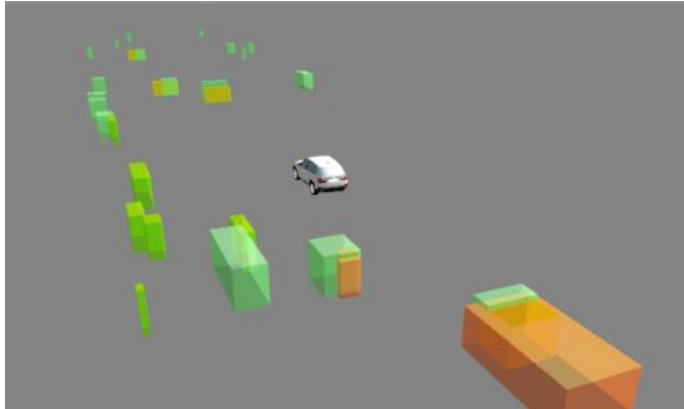
Institute for Automotive Engineering





Object Fusion and Tracking

Multi-Instance Kalman Filter: Input / Output and Challenges

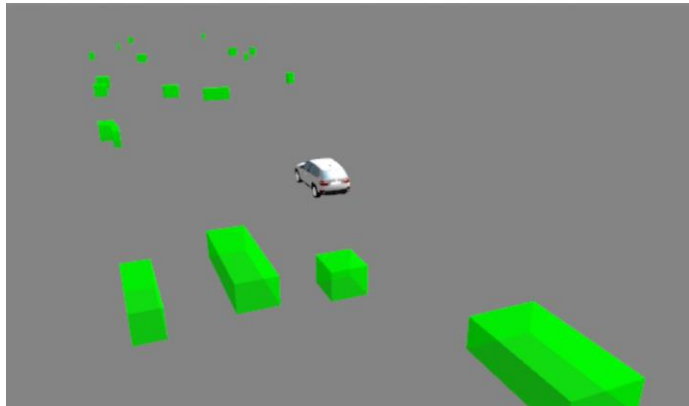


Input

- Detected objects of multiple sensors

Challenges

- Different sensor types with different measurement characteristics
- False detections (false positives)
- Missing detections (false negatives)
- Partially correct objects (incorrect orientation / classification / ...)



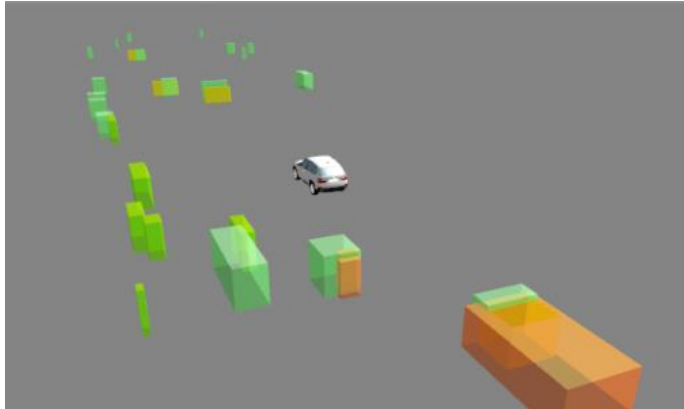
Output

- Single list of tracked objects



Object Fusion and Tracking

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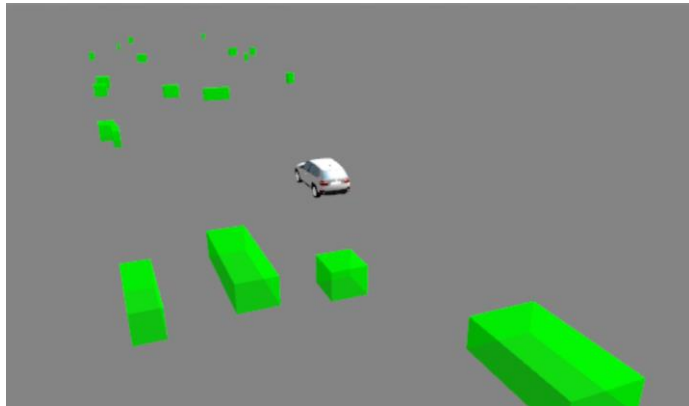


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Object Fusion and Tracking

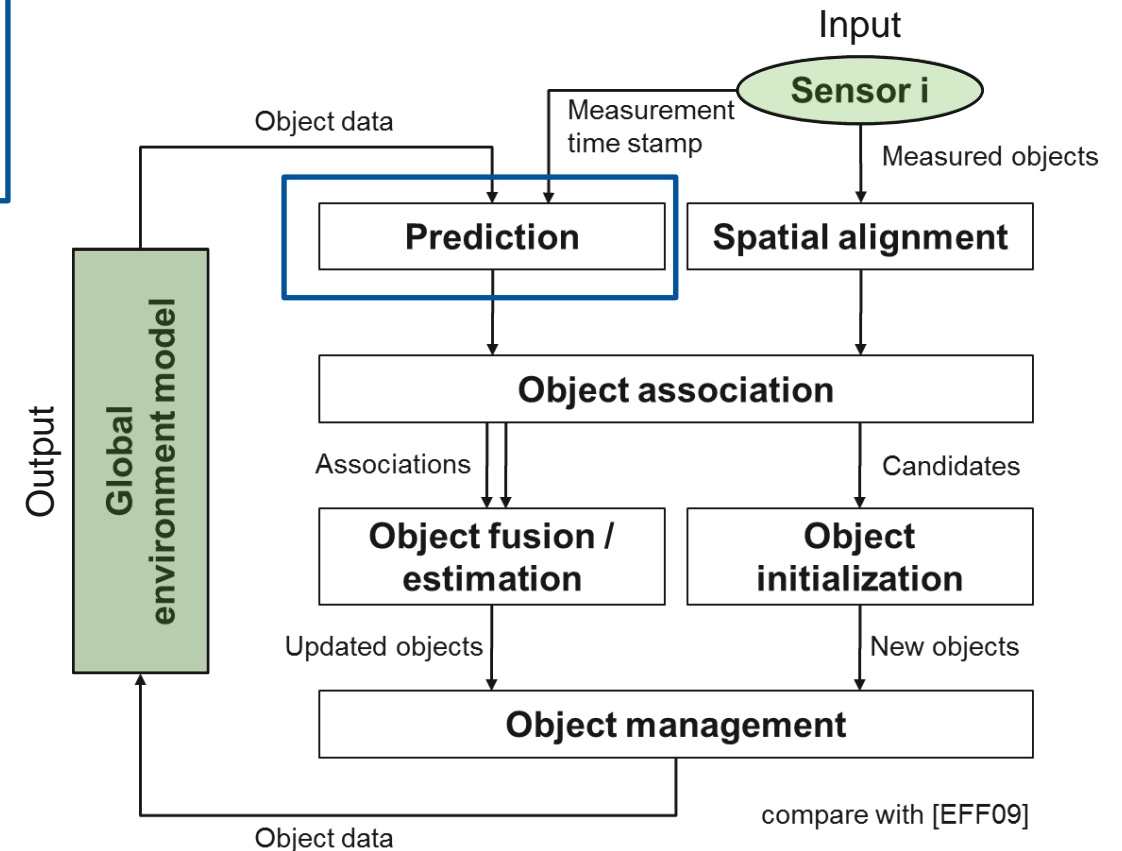
Challenges

→ Three subsections

Object Prediction

- Temporally align sensor objects with global objects
- Use a motion model to predict state of the vehicle

Multi-Instance Kalman Filter





Object Fusion and Tracking

Challenges

→ Three subsections

Object Prediction

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Object association

- Efficiently associate sensor objects to global objects
- Find candidates for new global objects
- Mark undetected global objects

Object fusion

- Fuse all object states
- Minimize uncertainties

Multi-Instance Kalman Filter

