



## T5: On-board autonomous navigation

# Sensor fusion for on-board autonomous navigation of AAM aircraft

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#### Motivation

The integration of highly automated Advanced Air Mobility (AAM) aircraft in urban areas places high demands on positioning and navigation in terms of accuracy and robustness:

- The problem of navigation using global navigation satellite systems (GNSS) in an urban environment; purely GNSS-based positioning is not accurate [1]
- The obvious solution is to use additional sensors (e.g. camera, light detection and ranging, LiDAR) [2]
- Camera and LiDAR sensors as well as semantic mapping [3] of the environment (Simultaneous Localization and Mapping (SLAM)) are intended to ensure positioning for safe and reliable flight operations

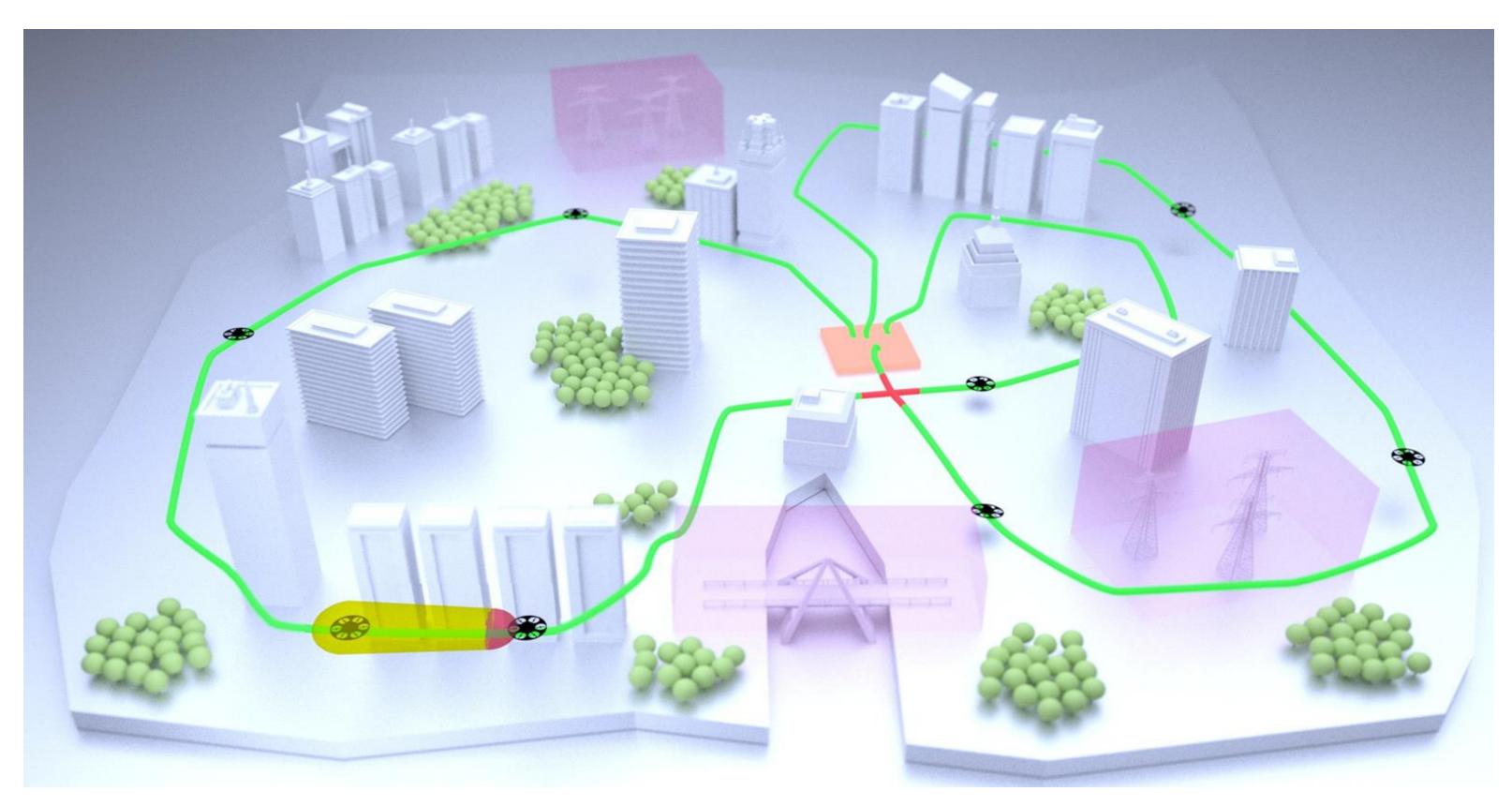


Figure 1: Scenario with geozones, buildings and potential conflicts between participating AAM aircraft

#### Methods

#### Evaluation of optical SLAM algorithms

- Integration of semantic image features relevant for AAM aircraft by integrating neural networks
- Integration of additional sensors such as Inertial Measurement Unit (IMU) and LiDAR to increase reliability
- Integration and testing of the algorithms in the closed-loop control system of an AAM aircraft

#### Results

- Determination of own position for AAM aircraft using semantic SLAM
- Fusion with other sensor data such as IMU, LiDAR
- Testing of the method in the context of AAM with a closed control loop; comparison with alternative methods
- Goal: accurate and GNSS-interferenceindependent self-positioning for the navigation of AAM aircraft in an urban environment

### Networking in the RTG

- T1 to T3 and T7 (boundary conditions)
- Reliable, robust self-position for T6 and postdoc project
- Description of landmarks (semantics)
  and initial maps for T8
- DLR: Regular scientific exchange with the department of Unmanned Aircraft of the Institute of Flight Systems, use of test equipment/facilities and joint flight tests, use of VAST



Figure 2: Virtual representation of a village in the mountains with road, people, cars, buildings and vegetation

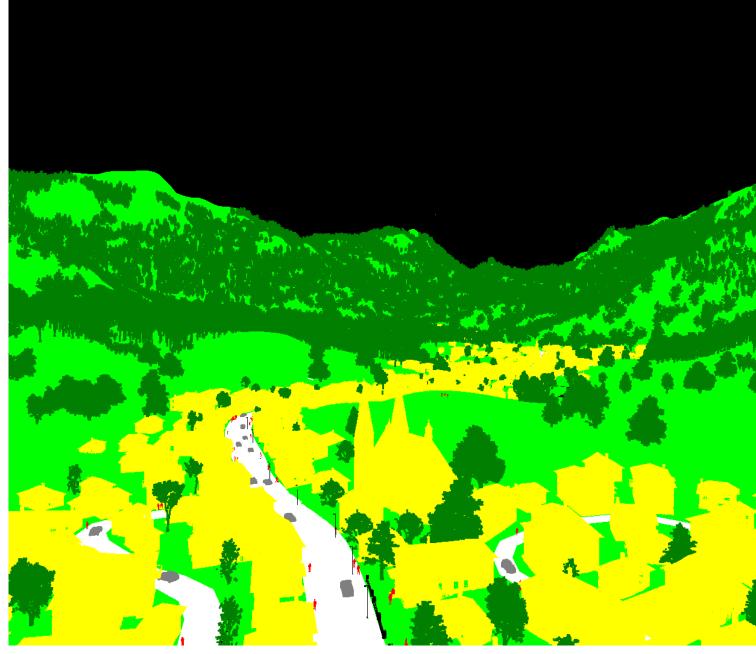


Figure 3: Semantic representation of the scene from Figure 2, different object classes are shown in different colors



Figure 4: DLR test aircraft CDO DJI M600 (max. take-off weight: approx. 15 kg, diagonal propeller axes distance: 1,133 mm, propeller diameter 21")

#### Literatur:

- [1] F. Andert, N. Ammann, J. Puschel und J. Dittrich, "On the safe navigation problem for unmanned aircraft: Visual odometry and alignment optimizations for UAV positioning", in 2014 International Conference on Unmanned Aircraft Systems (ICUAS), 2014, doi: 10.1109/icuas.2014.6842318.
- [2] F. Andert und S. Krause, "Optical aircraft navigation with multi-sensor SLAM and infinite depth features", in 2017 International Conference on Unmanned Aircraft Systems (ICUAS), 2017, doi: 10.1109/icuas.2017.7991319.
- [3] C. Hinniger, J. Rüter: "Synthetic Training Data for Semantic Segmentation of the Environment from UAV Perspective", Aerospace. 2023; 10 (7): 604. https://doi.org/10.3390/aerospace10070604





