



## **T6: Robust Motion Trajectories** Development of Robust Motion trajectories under Consideration of Model and Data Uncertainty

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### Motivation

The optimization of movement trajectories, taking into account polymorphic uncertainty, is the basis for:

- the optimization of energy consumption and time requirements,
- maintaining the high safety level of aviation.

The focus on robustness is essential due to the large variety of sources of uncertainty:

- natural variability of environmental conditions,
- uncertainty, incompleteness and inaccuracy of
  - sensor and geodata,
  - data from mobility surveys,
- available expert knowledge and
- microclimatic models.



Figure 1: Optimization of trajectories by considering polymorphic uncertainty

### Methods

The quantification of the uncertainty and the update of the movement trajectory in real time is carried out by:

- Application of Bayesian filters [1] to take into account time-varying aleatoric uncertain measurement data and simulation parameters (inherent, objective uncertainty),
- extension by additionally including epistemic fuzziness (e.g. measurement errors) and
- recursive quantification of polymorphically uncertain model parameters.

### Results

Enabling the consideration of polymorphic fuzziness in combination with the use of extended Bayesian filters:

- Consideration of all available information,
- realistic uncertainty quantification of the simulation parameters [3], [4], [5],
- robustness evaluation [2] of the calculated trajectories and
- Adaptative adjustment of the trajectory in real time based on new data/measurements.

The following collaborations are planned for the determination and application of robust movement trajectories:

Uncertainty quantification in the context of trajectory simulation (T10).

Required input variables are:

- Pre-processed position data from sensor and geodata (T5, T8),
- data from mobility surveys (T11) and
- predictions from microclimatic models (T7).

# previous model measurement model

### Literature:

- [1] Banerjee, P.; Corbetta, M.: In-Time UAV Flight-Trajectory Estimation and Tracking Using Bayesian Filters. IEEE Aerospace Conference (2020), 1-9
- [2] Schietzold, F. N.; Leichsenring, F.; Götz, M.; Graf, W.; Kaliske, M.: Robustness versus Performance-Nested Inherence of Objectives in Optimization with Polymorphic Uncertain Parameters. Advances in Engineering Software 156

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Figure 2: Scheme Bayesian filters



Figure 3: Time-dependent polymorphic fuzzy quantity, modelled by p-boxes [5]

(2021)

0.8

0.6

0.2

 $F(\sigma)$ 

[3] Götz, M.; Graf, W.; Kaliske, M.: Enhanced uncertain structural analysis with time- and spatial-dependent (functional) fuzzy results. Mechanical Systems and Signal Processing 119 (2019), 23-38

[4] Zschocke, S.; Leichsenring, F.; Graf, W.; Kaliske, M.: A concept for data-driven computational mechanics in the presence of polymorphic uncertain properties. Engineering Structures 267 (2022), 114672

[5] Zschocke, S.; Graf, W.; Kaliske, M.: Polymorphic Uncertain Structural Analysis: Challenges in Data-Driven Inelasticity. Proceedings in Applied Mathematics & Mechanics 22 (2022), e202200023

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