

ME 780
Nonlinear State Estimation for Robotics and Computer Vision
Applications
SPRING 2017

Instructor: Professor Steven Waslander

Overview and Objectives

The course will be addressing two concepts:

1. **Nonlinear State Estimation:** Present a mathematical foundation which is required to solve state estimation problems using nonlinear models.
2. **State-of-the-art Applications:** Review the latest work within the areas of Simultaneous Localization and Mapping (SLAM), sensor calibration, visual odometry, and other robotics and computer vision applications.

Schedule

Two 1.5 hour long weekly meeting will be held in room TBD. Students in the class will alternate in presenting formal talks on the mathematical foundation for nonlinear state estimation and applications of said estimation to robotics, well as their own findings in simulation using real data sets. Topics for each week of lecture will be agreed upon in the first week of class, and the class is expected to take the standard 12 weeks.

Prerequisites

Students are expected to have good knowledge of linear algebra, probability theory, calculus, numerical computation, and basic knowledge of computer vision techniques.

Course Outline

Foundation

1. Lie Group theory for $\text{SO}(3)$ and $\text{SE}(3)$, parameterizations for rotations and transformations.
2. Review of Probability theory, maximum likelihood estimation, and maximum a posteriori estimation.
3. Kinematic modeling of different vehicles [1, 2].

4. Review of computer vision techniques for landmark tracking and projection [3].
5. Differential calculus of 3D orientations, representing uncertainty on $\mathbb{SO}(3)$ and $\mathbb{SE}(3)$ [4, 5].
6. Factor graph representation, solving estimation problems on manifolds [6, 7].
7. Batch estimation, solving sparse linear systems in information and square root form [8, 9, 10].
8. Marginalization and sliding window estimation [11, 12, 13, 14].

Applications for Robotics and Computer Vision

1. Modeling camera residual terms using landmark re-projection and photometric error [3, 15].
2. Modeling IMU residual terms and noise characterization [16, 17, 18].
3. Modeling Absolute and Relative pose residual terms, point-cloud scan registration [19, 20].
4. Application: Bundle Adjustment [21, 22, 23].
5. Application: Sensor Calibration [24, 25, 26, 27].
6. Application: Landmark Based Visual-Inertial Odometry and SLAM [28, 29, 30, 31].
7. Application: Direct Visual-Inertial Odometry and SLAM [32, 15, 33, 34].
8. Application: Lidar Localization and SLAM [35, 36].

Grade Distribution

Presentation 1:	20%
Presentation 2:	20%
Project Report:	60%
Total:	100%

References

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