



Improved Methods for Tracking and Characterizing Inactive Space Objects

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Moriba Jah⁽¹⁾, Kyle DeMars⁽²⁾, and Robert Bishop⁽³⁾

(1) Air Force Research Laboratory, Space Vehicles Directorate

(2) NRC Research Fellow (AFRL)

(3) Marquette University



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Advanced Sciences and Technology Research Institute for Astrodynamics*



Charter: Research entity combining government in-house and contractor expertise in astrodynamics with a consortium of universities performing research in relevant areas of interest to the U.S. Air Force Research Laboratory and Department of Defense.

Research Topics:

- Data Association and Tracking
- Initial Orbit Determination
- Maneuver Detection and Reconstruction
- Satellite Characterization
- Data/Sensor Fusion for SSA
- Orbital Debris Tracking and Characterization
- Astrodynamics Applications to High Performance Computing

*Dr. Moriba Jah, Director



Astrodynamics Tools:

- General Mission Analysis Tool (GMAT)
- Satellite Tool Kit (STK)
- Orbit Determination Tool Kit (ODTK)
- Goddard Trajectory Determination System (GTDS)
- MATLAB (various analysis models and algorithms)

Current Academic Members:

- Purdue University
- University of Colorado, Boulder
- Texas A&M
- University of Texas, Austin
- State University of New York, Buffalo
- Utah State University
- Embry-Riddle Aeronautical University
- Missouri University of Science & Technology
- Penn State University
- Georgia Tech
- US Air Force Academy
- Naval Postgraduate School



Background/Motivation



- Every Space Object (SO) is not exhaustively surveyed and the number of objects is increasing
 - Data Exploitation: We must maximize the information we can extract from all sources of collected data (i.e. minimize ambiguity)
 - Data Collection: We must gather data in a way that maximizes its information content (i.e. maximize efficiency)
- Unrealistic and/or too large of an ambiguity leads to erroneous decisions, inability to discern, and overall lack of knowledge and confidence. All decisions are based upon knowledge.
 - The knowledge that we have need not be perfect, but must be realistic and predictable



Knowledge Realism: Why?



- Decisions are based upon what we believe to know
 - Inferring space environment effects from object behavior
 - Gravity fields, Solar Activity, Atmospheric Density, etc.
 - Identifying Conjunctions
 - Computing Collision Probabilities
 - Correlating tracked objects to a catalog
 - Correctly Identifying and Discriminating between one and other Space Objects
- Flawed knowledge leads to flawed or no decisions



Knowledge Realism: How?



- **Developed method based upon probabilistic approach**
- **Developed a realistic and quantifiable measure of association based upon SO ambiguity**
- **Employed a method that can readily process data of various types and disparate sensors**
- **Developed and implemented a method that can adaptively approximate the actual SO state error distribution**
- **Developed the Adaptive Entropy Gaussian Information Synthesis (AEGIS) method as an enabling technology**
 - Exploits Gaussian Sum approximations for a given probability density function (pdf)
 - Adaptive coarsening and refining of components
 - Exploits Information-Theoretic measures of ambiguity (i.e. entropy)
 - Leverages contemporary advanced estimation strategies (e.g. Sigma-Point Filters)



Propagation Test



Dynamical models taken to be

- ▶ 8×8 spherical harmonics gravity
- ▶ 3rd body perturbations from Sun and Moon
- ▶ SRP acceleration and moment on 8-plate macro model

Nominal orbit described by

$$a = 42165.91 \text{ km}, \quad e = 0.0002429, \quad i = 0.83^\circ, \quad \Omega = 0^\circ, \quad \omega = 0^\circ, \quad M = 0^\circ$$

Apply the EKF, UKF, and SGMUKF and evaluate performance of each method

- ▶ note that $\text{trace}\{\mathbf{F}(\hat{\mathbf{x}}(t), t)\} = 0$
- ▶ the differential entropy of the linearized system is constant
- ▶ implement 3-component and 5-component splitting libraries

Initial uncertainty is taken to be

- ▶ 1 km for position
- ▶ 1 m/s for velocity
- ▶ 1° for attitude
- ▶ $0.1^\circ/\text{hr}$ for angular velocity



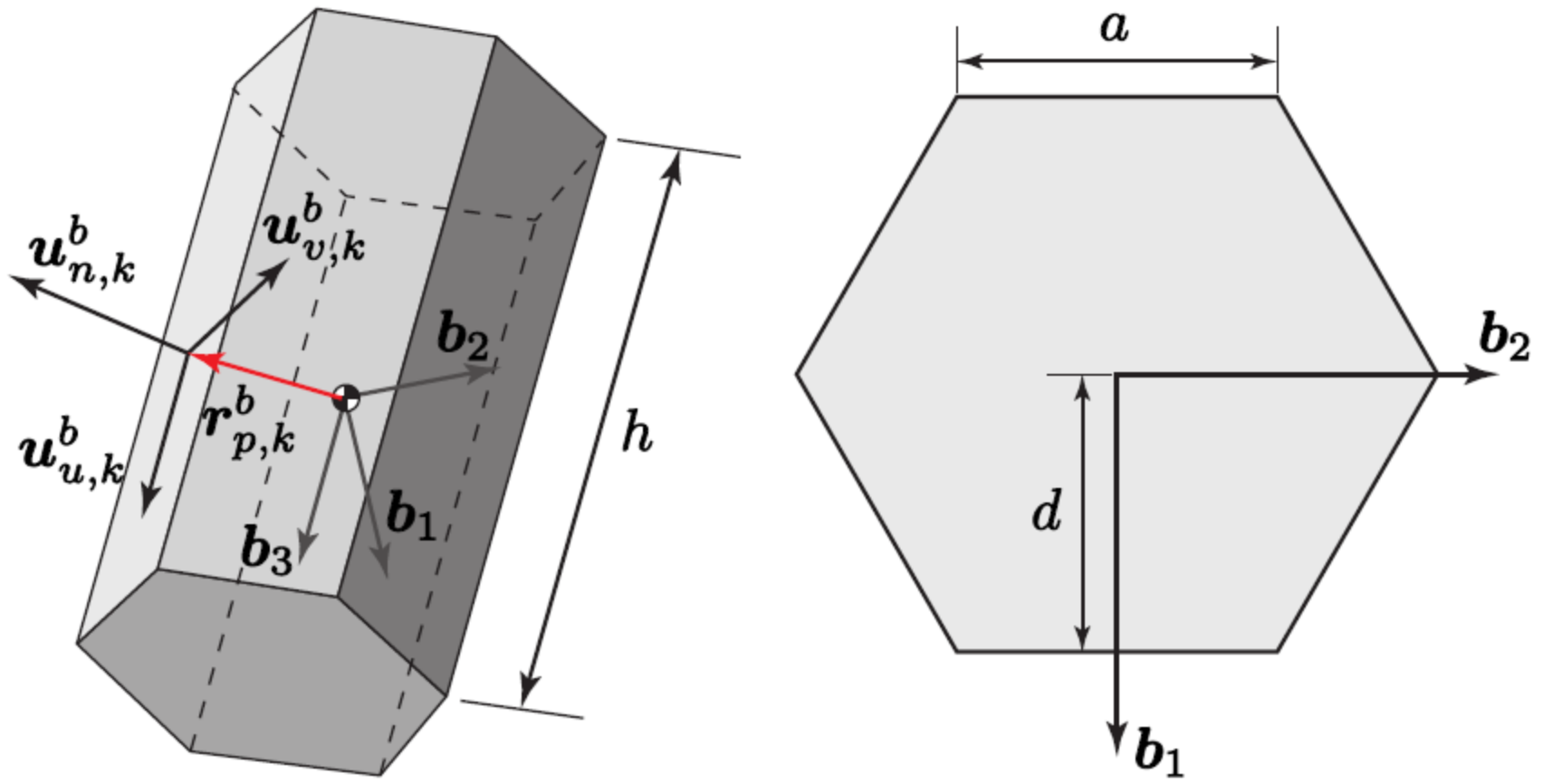
Strategy Definitions



- **Extended Kalman Filter (EKF)**
 - Based upon linearized dynamics
 - Linear covariance (assumes Gaussian error distribution)
 - Observations update propagated mean and covariance
- **Unscented Kalman Filter (UKF)**
 - Non-linear dynamics
 - Non-linear covariance (computes error distribution second moment)
 - Observations update computed mean and covariance (second order effects or higher captured)
- **Adaptive Entropy Gaussian Information Synthesis (AEGIS)**
 - Non-linear dynamics
 - Does not assume a Gaussian error distribution, but error distribution is approximated by sum of individually weighted Gaussian components
 - Each Gaussian component is a UKF
 - Observations update each Gaussian component based upon Bayesian likelihoods



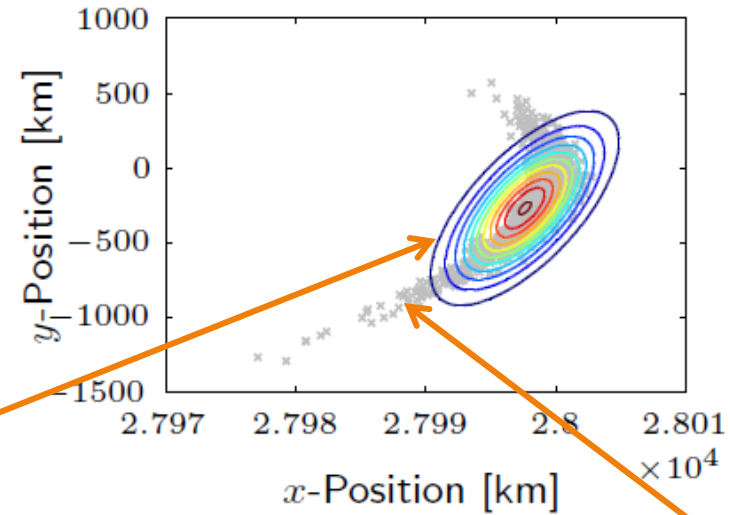
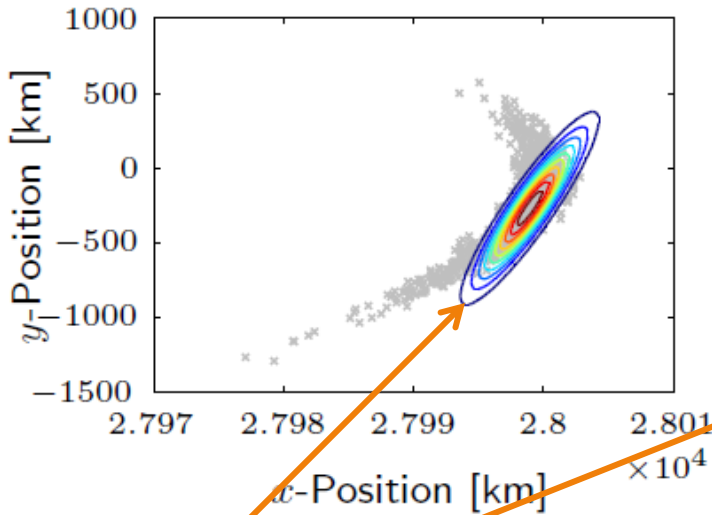
Spacecraft Model





True vs Predicted SO Ambiguity for Various Strategies: 1 Rev Propagation

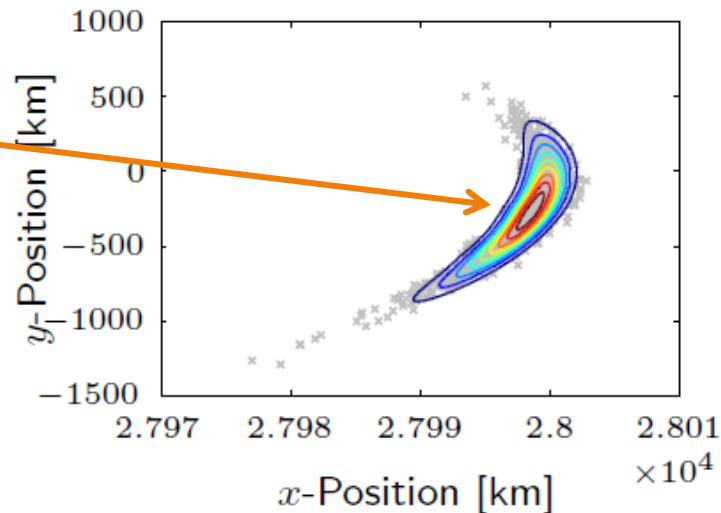
EKF



UKF

Estimated Knowledge Contours

True Error Gray Dots



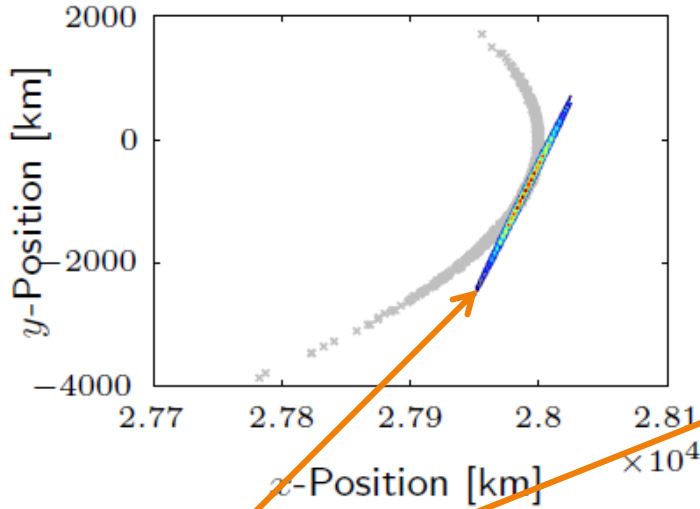
AEGIS



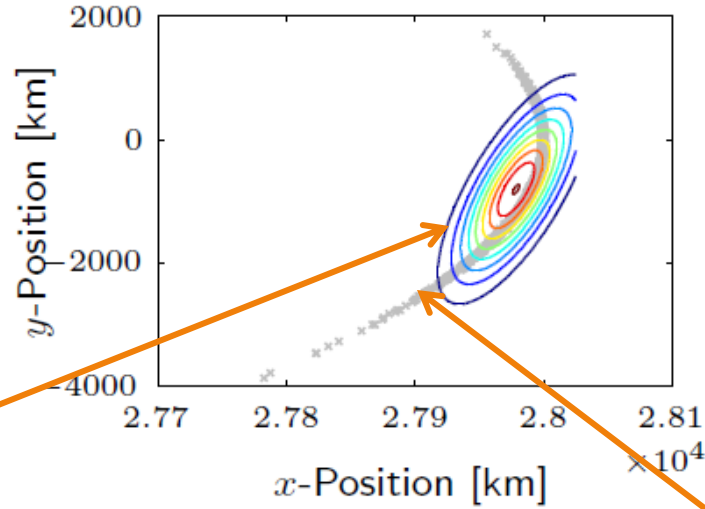
True vs Predicted SO Ambiguity for Various Strategies: 3 Rev Propagation



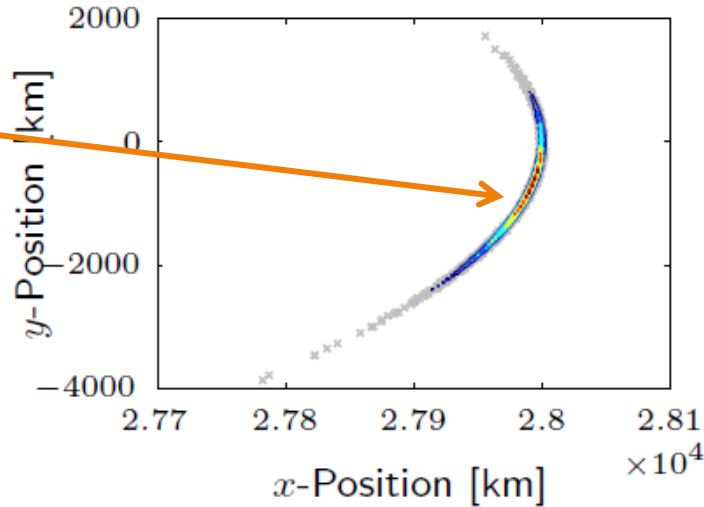
EKF



UKF



AEGIS

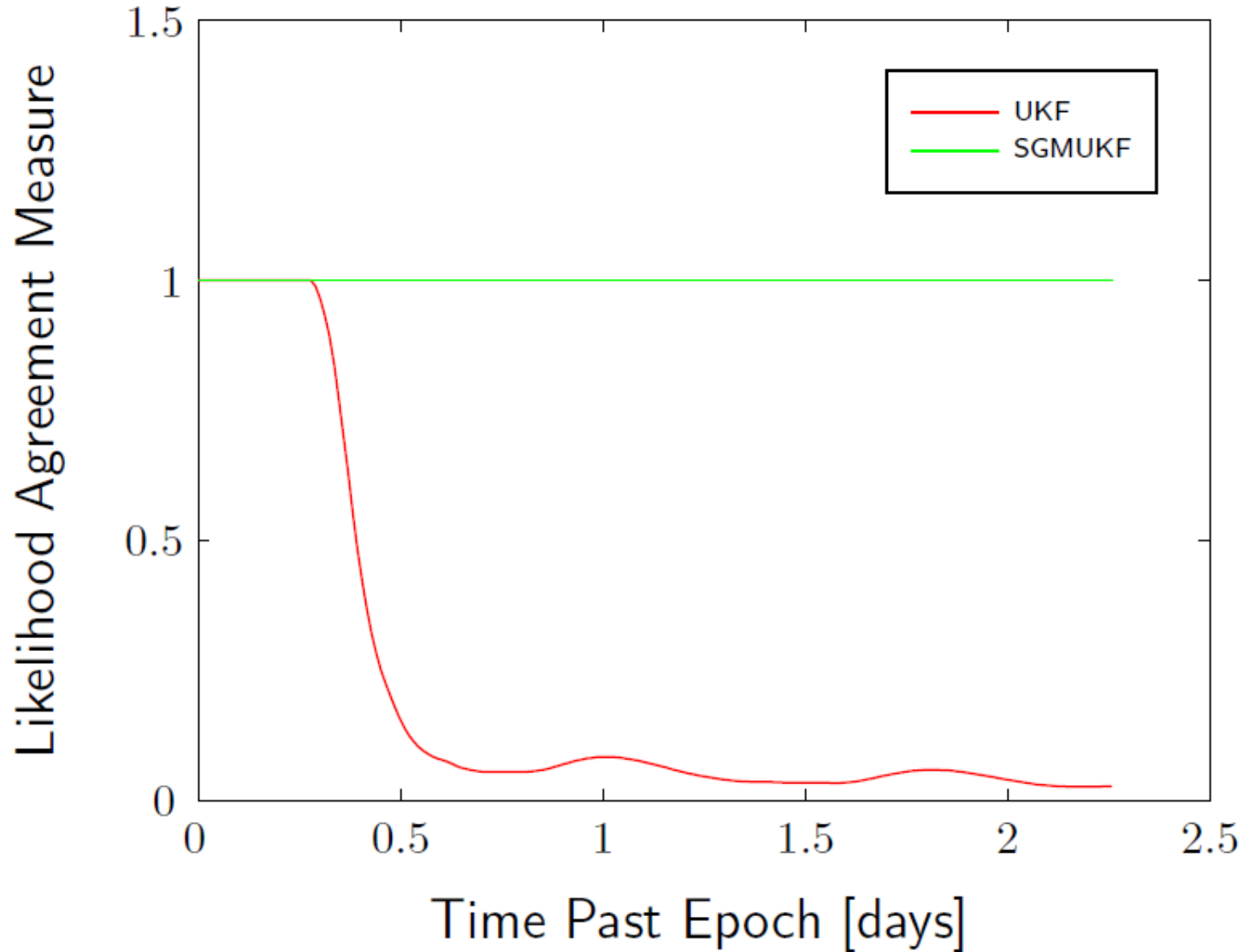


Estimated Knowledge

True Error Gray Dots



Likelihood Agreement Measure: UKF and AEGIS PDFs vs Monte Carlo





Propagation and Data Reduction Test



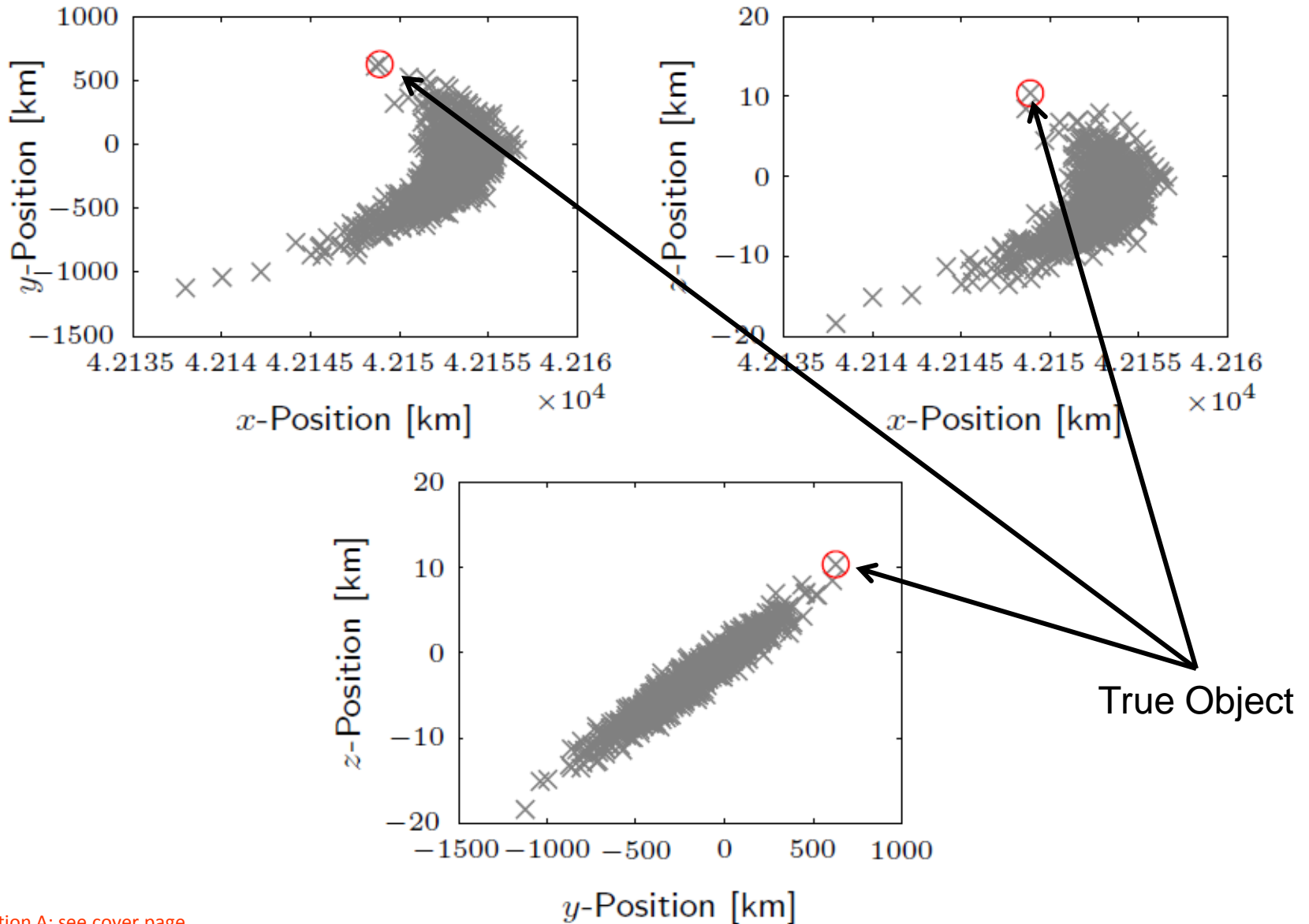
- ▶ Generate measurement data of topocentric right ascension and declination
- ▶ Observer located at the Maui GEODSS complex

$$\phi = 20.708074^\circ, \quad \lambda = -156.257486^\circ, \quad \text{and} \quad h = 3060.74 \text{ m}$$

- ▶ Measurement arc is taken to be
 - ▶ 20 minutes in duration
 - ▶ 20 seconds between measurements
 - ▶ total of 61 measurements
- ▶ Each of the measurements is subjected to a Gaussian, white-noise sequence with a standard deviation of 1 arc-second on both the right ascension and declination angles.
- ▶ Test the update capabilities of the UKF, and 5-component SGMUKF on a single stress-case.



Selected Truth Point for Analysis





Estimated SO Ambiguity vs True SO: 1 Rev Propagation

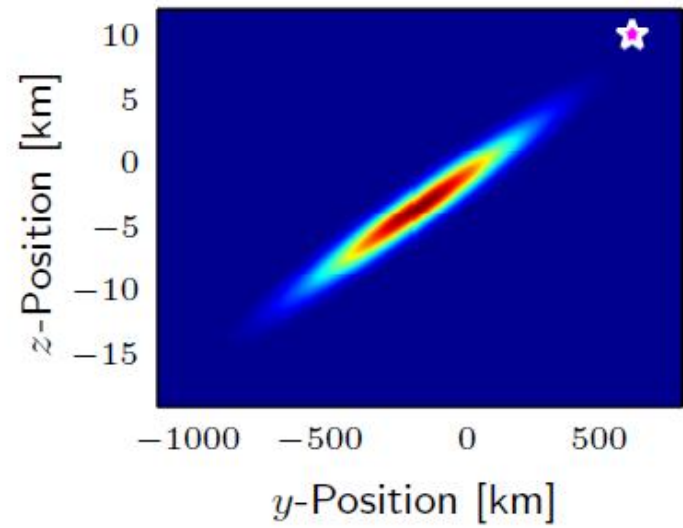
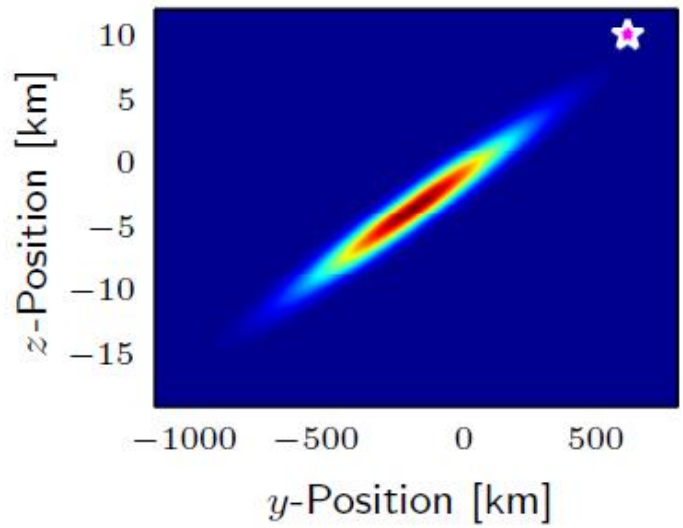
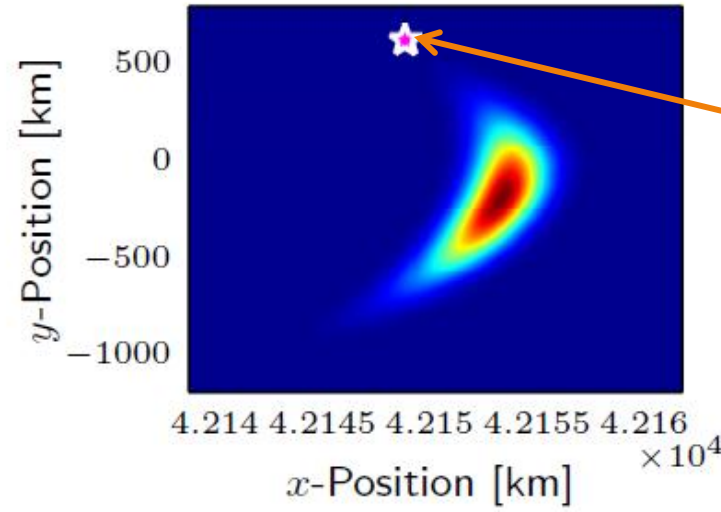
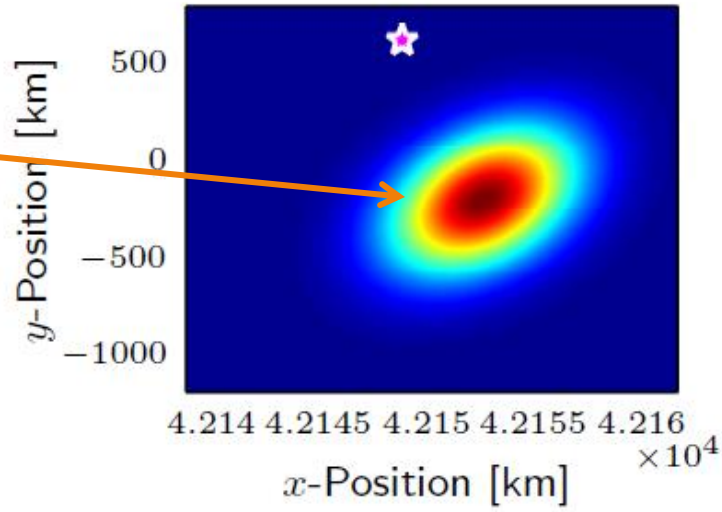


Estimated Knowledge

True Object

UKF

AEGIS



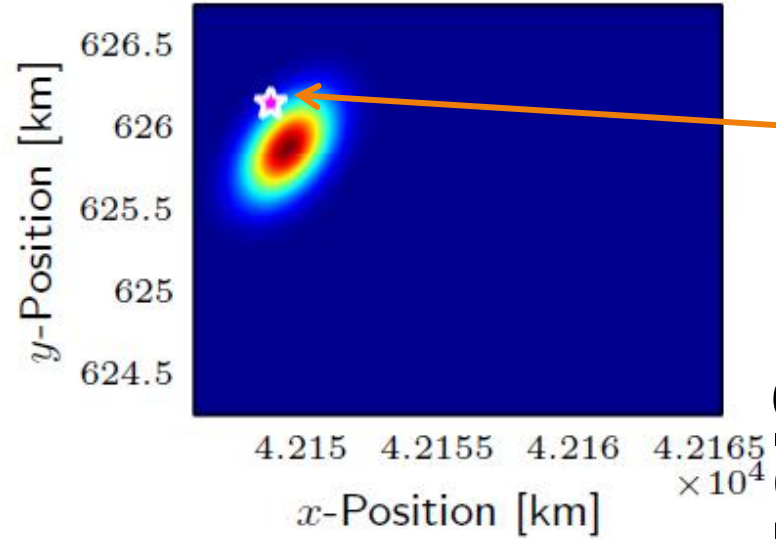
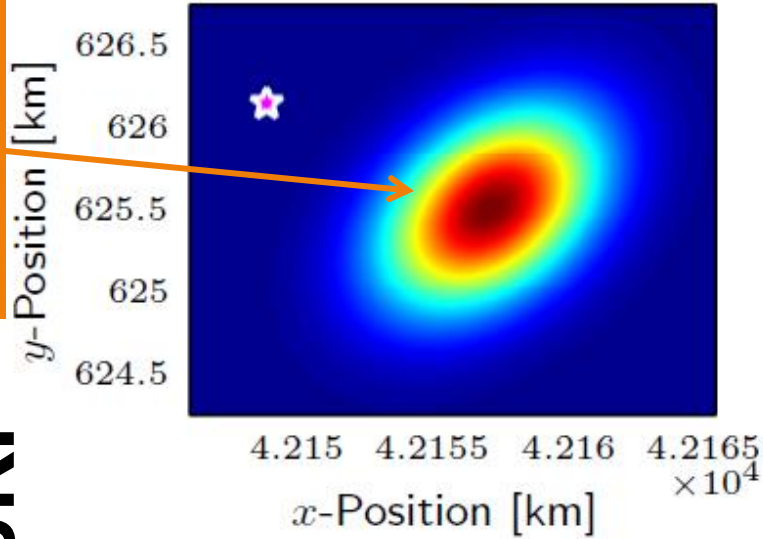


Estimated SO Ambiguity vs True SO After Measurement Update: 1 Measurement



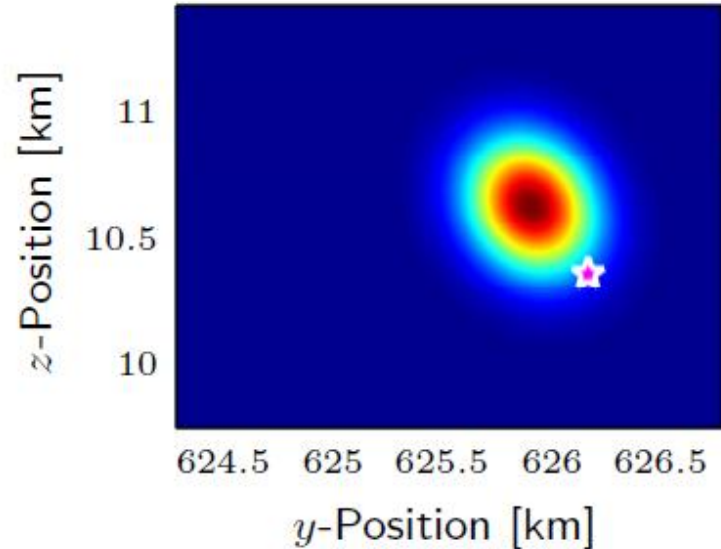
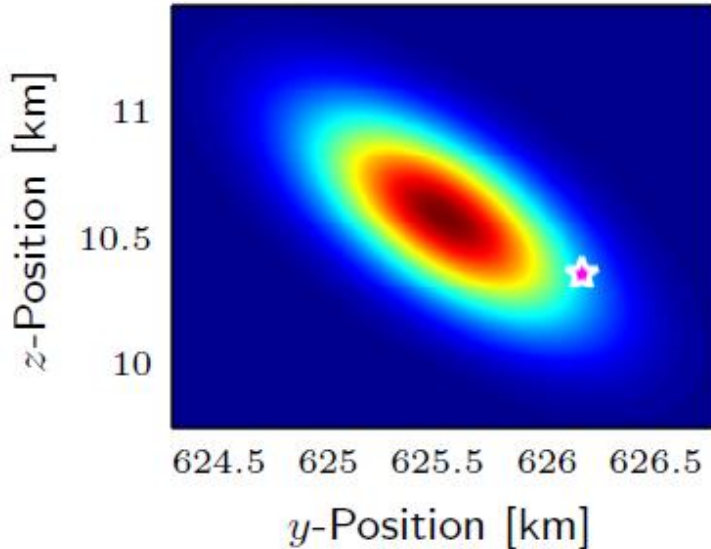
Estimated Knowledge

UKF



True Object

AEGIS



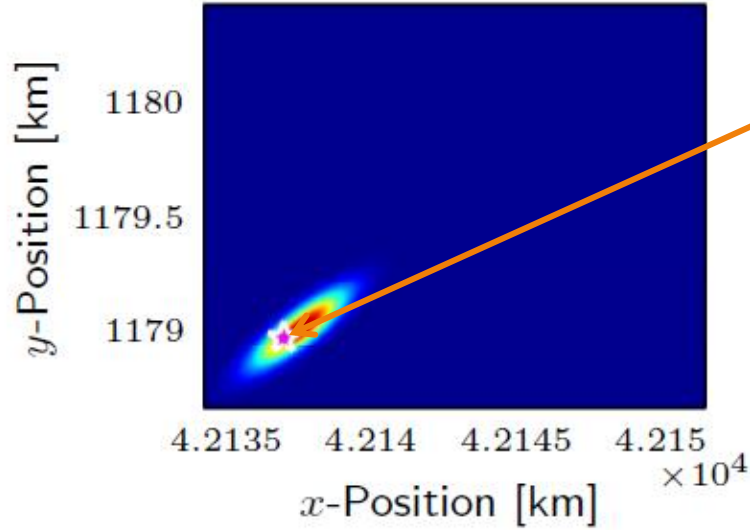
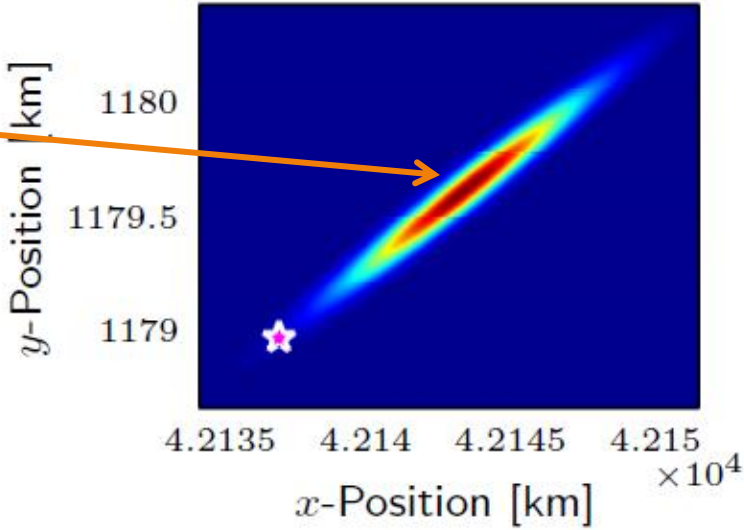


Estimated SO Ambiguity vs True SO After Measurement Update: 10 Measurements



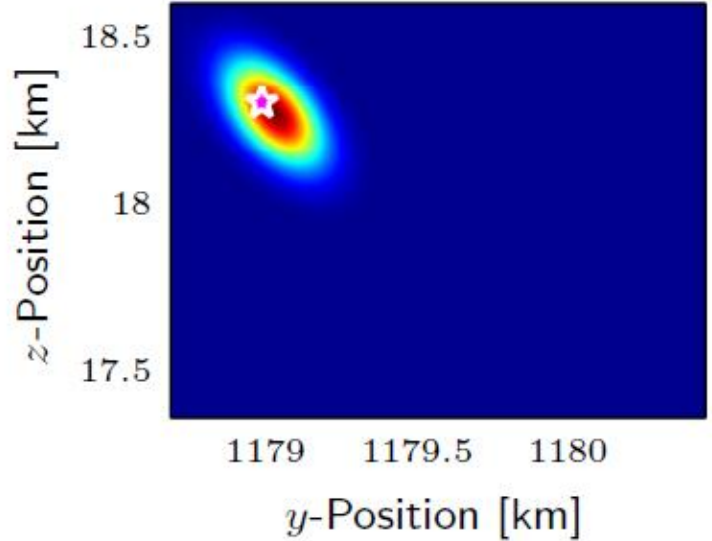
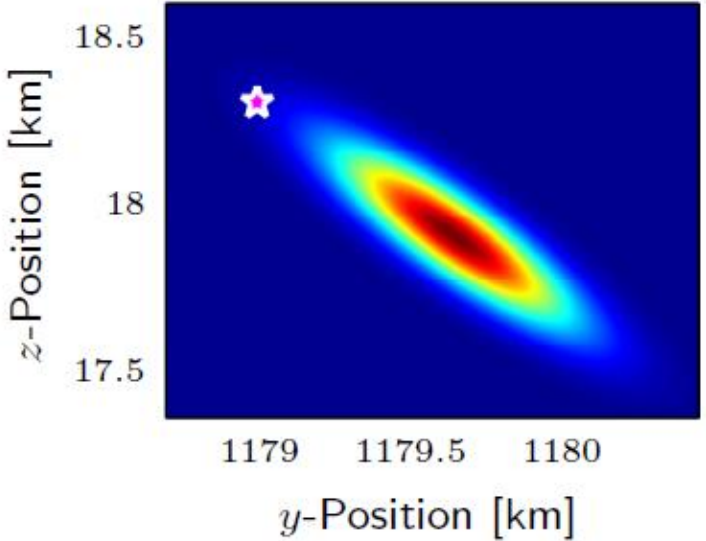
Estimated Knowledge

UKF



True Object

AEGIS



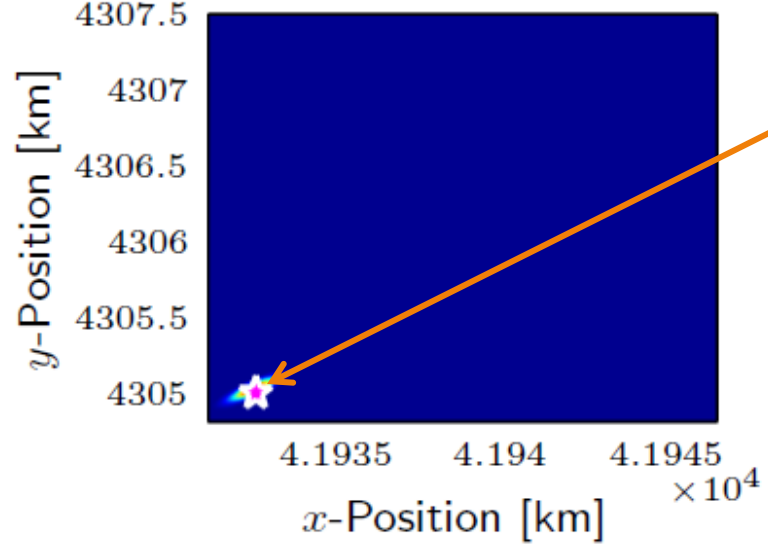
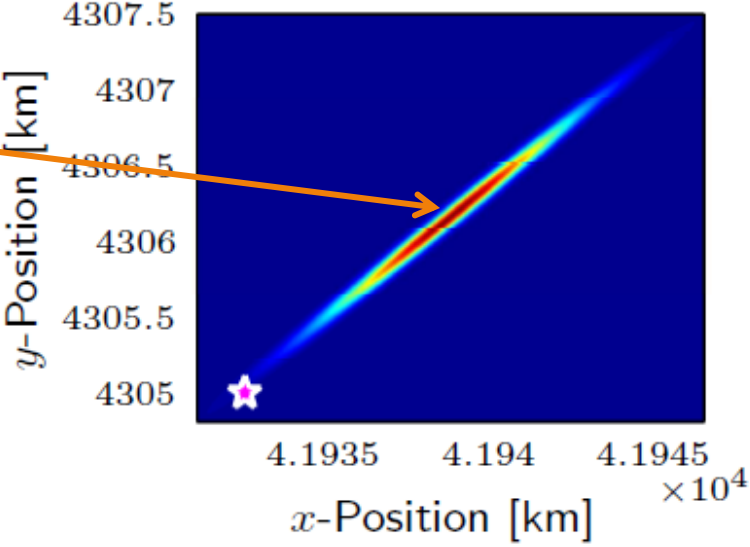


Estimated SO Ambiguity vs True SO After Measurement Update: 61 Measurements



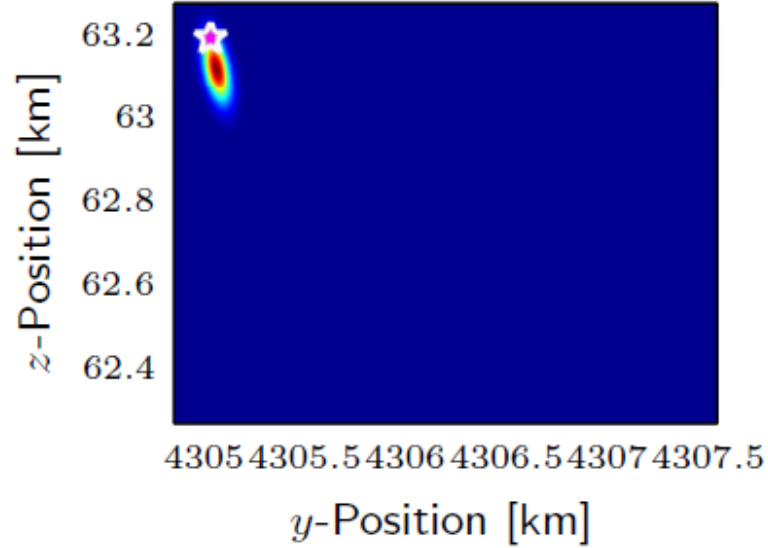
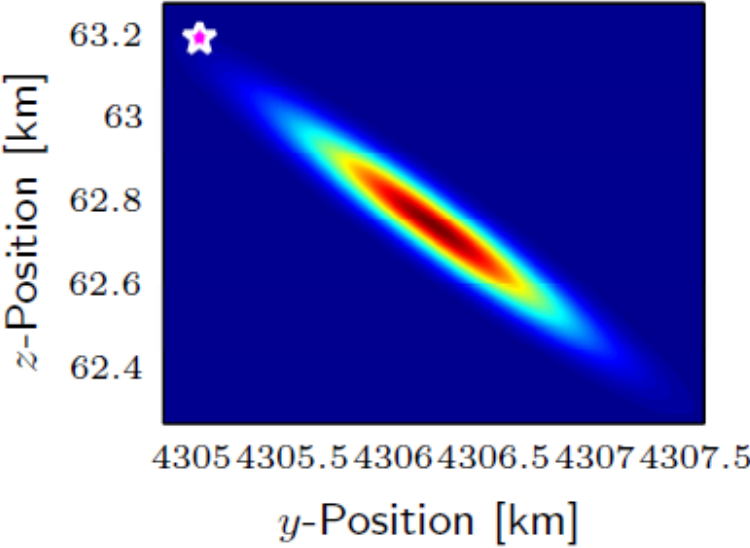
Estimated Knowledge

UKF



True Object

AEGIS





Summarizing Statements



- **Action:** Developed the Adaptive Entropy Gaussian Information Synthesis (AEGIS) method as an enabling technology
 - Based upon probabilistic approach
 - Yields realistic and quantifiable measure of SO knowledge (ambiguity)
 - Adaptively approximates the actual SO state error distribution
- **Impact:** Realistic measure of ambiguity improves SO data/track correlation and allows for more accurate trajectory estimation and prediction