

# **Techniques for assessing space object cataloguing performance during design of surveillance systems**

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- Europe builds up surveillance capabilities → requires analysis of sensor design
- Typically detection performance is assessed during design but ***not*** the cataloguing performance

**Cataloguing tasks:** combine observations to improve the orbital states (e.g. for collision avoidance)

- Determine initial orbits from observations
- Associate observations
- Improve states and determine accuracy

*Develop a simulation framework for all individual steps and assess sensor cataloguing performance*

# SIMULATION

# Simulation (1)



1. Propagate subset of MASTER reference population with SGP4
2. Subset selected with object size / perigee height envelope
3. Radar setup: south pointing and 60° elevation, 100° × 30° field-of-regard in deflection angles
4. Create track whenever object passes field-of-regard and is detectable

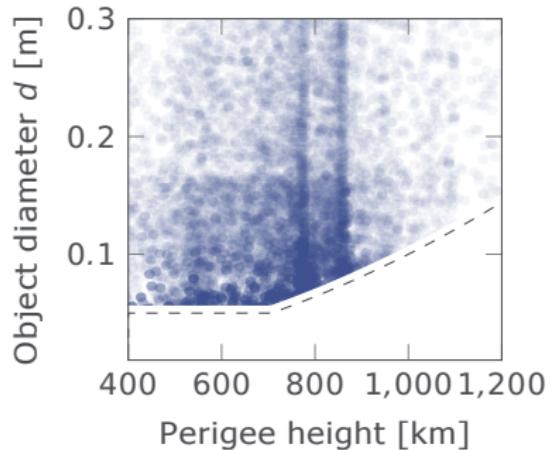
## Simulation (2)

Assure minimum  
signal-to-noise ratio

$$\frac{d^2}{\rho^4} \geq \frac{d_{\min}^2}{\rho_{\text{ref}}^4} \propto \text{SNR}_{\min}$$

Generate random errors

$$\sigma_\rho^2 \propto \frac{1}{\text{SNR}} \quad \text{and} \quad \sigma_{\alpha,\beta}^2 \propto \frac{1}{\text{SNR}}.$$



Parameter	$\rho_{\text{ref}}$	$d_{\min}$	$\sigma_{\rho,\max}$	$\sigma_{\alpha,\beta,\max}$
Value	1000 km	10 cm	5 m	0.2°

# INITIAL ORBIT DETERMINATION

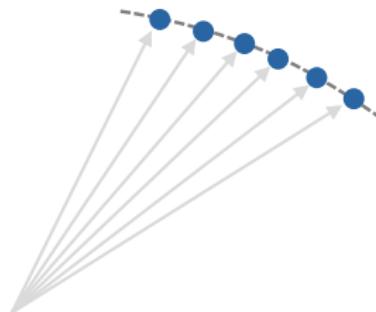
# Initial orbit determination



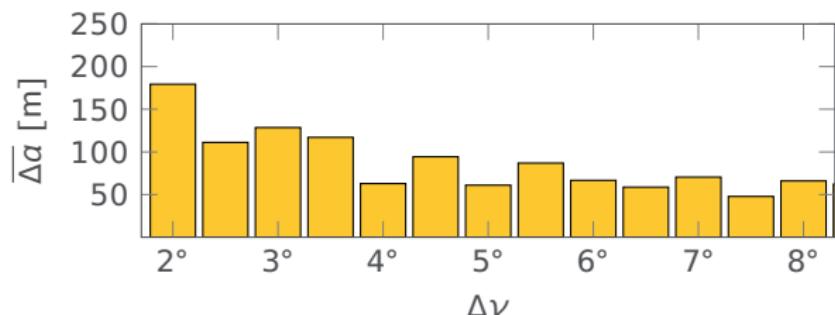
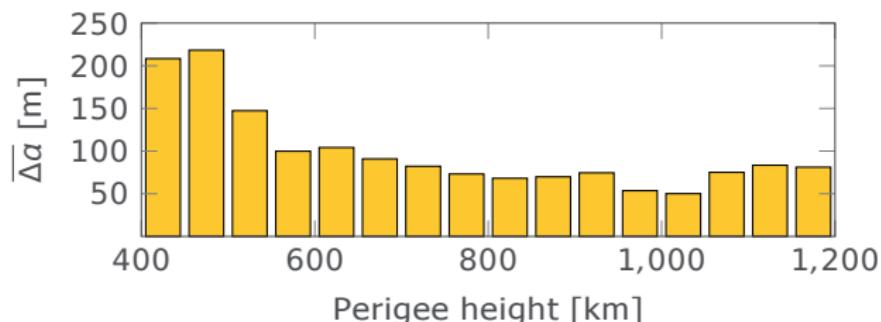
Stable fixed-point iterations to get a state  $\mathbf{y}$

$$\mathbf{y}^{(i+1)} = \mathbf{H}(\mathbf{y}^{(i)})\mathbf{z}$$

with observations  $\mathbf{z}$  and update-matrix  $\mathbf{H}$  from series expansion around  $\mathbf{y}^{(i)}$   
(adapted from the GTDS ranges and angles method)



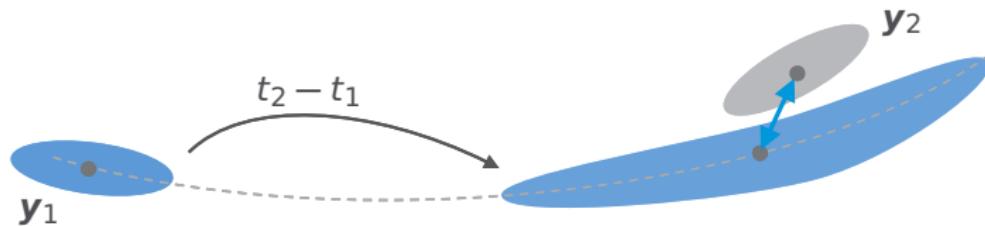
# Initial orbit accuracy



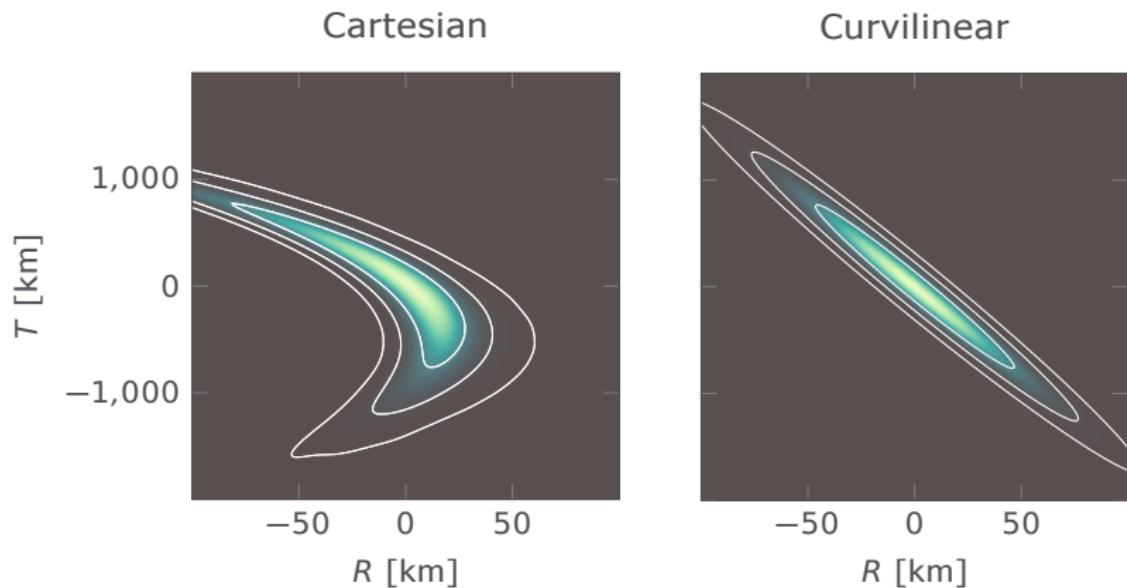
# ASSOCIATION

# Association test

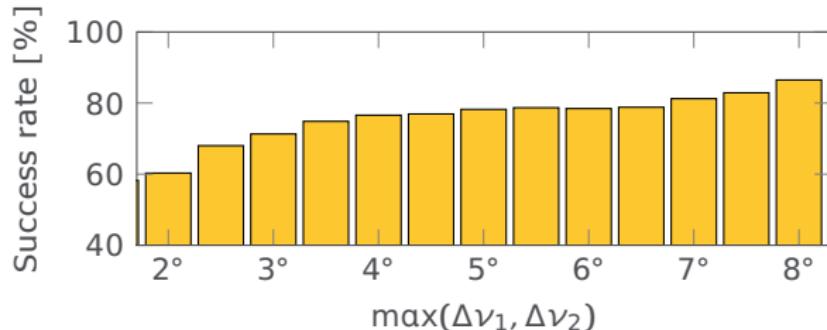
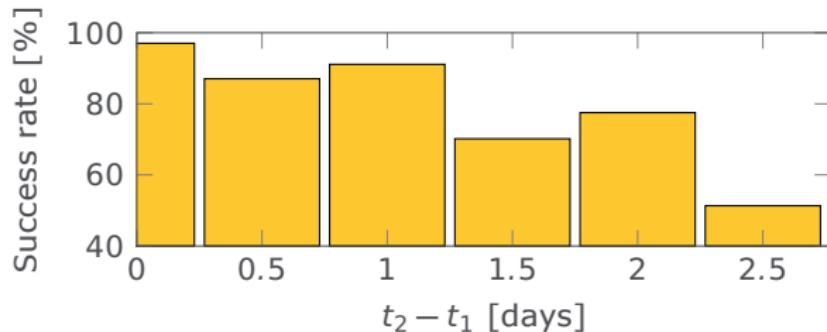
- Test two consecutive tracks → estimate if they belong to the same object
- Propagate to a common epoch
- Covariance-based association: accept if distance measure below threshold



# Uncertainty transformation



# Association performance



# ACHIEVABLE ACCURACY

# Consider covariance analysis



Model insufficiencies cause uncertainty: e.g. density variations in atmosphere, bias terms and other error sources

→ account erroneous model in covariance estimation

Most dominant error source is the unknown atmospheric variation

# Consider parameter

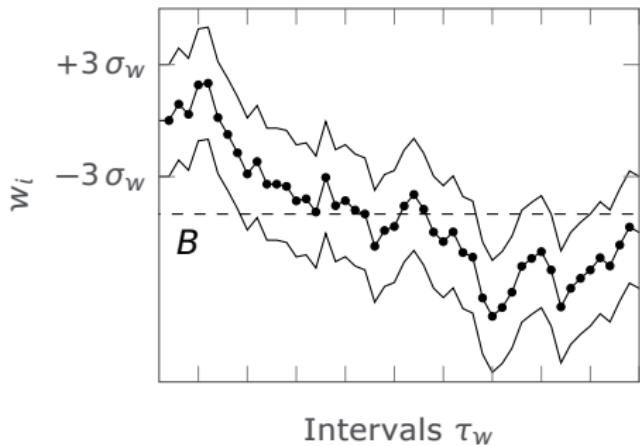


Augment drag acceleration  
with correction  $w$

$$\mathbf{a}_D(\mathbf{y}, t, B)(1 + w(t))$$

Model  $w$  as a random walk  
using discrete steps  $w_i$   
with  $E(w_i) = 0$  and the  
covariance

$$E(w_i w_j) = \sigma_w^2 \exp\{-|i-j|\tau_w\}$$



# Fit span selection

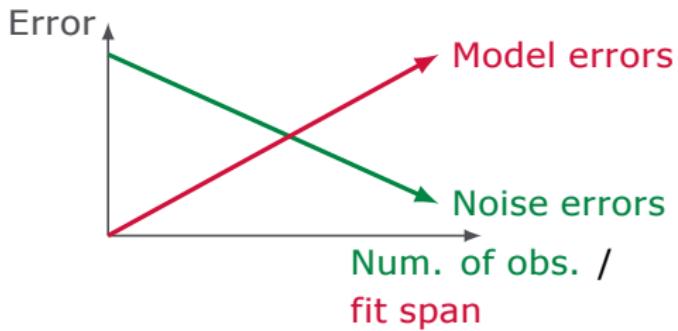


Trade-off between information gain and model capabilities

- More measurements → better accuracy
- Larger time span → model cannot be adjusted

*Difficult to simulate*

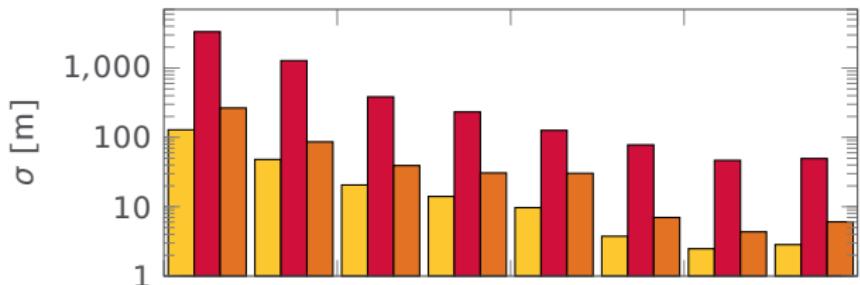
Use a fit span depending  
on energy-dissipation-rate  
(ranging from few days to  
two weeks)



# Achievable accuracy (*preliminary*)

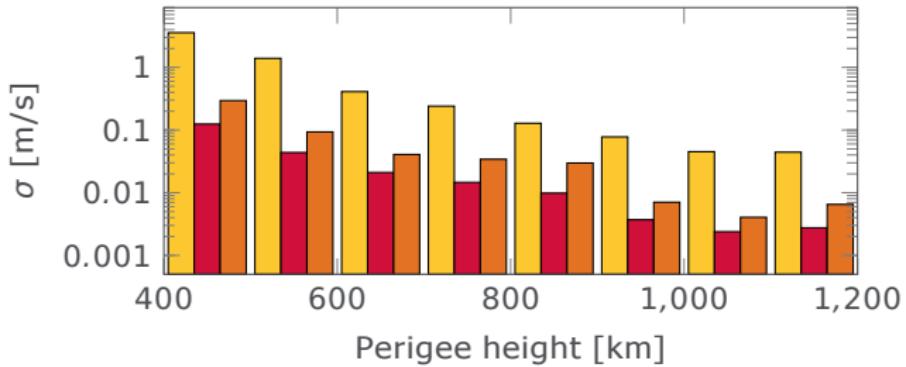


Position



Velocity

- Radial
- Along-track
- Cross-track



- Simulation of a surveillance radar
- Initial orbit determination using fixed-point iterations
- Covariance-based measurement association
- Preliminary study on achievable accuracies

## Open tasks

- Incorporate more elaborate detection model
- Special treatment of short radar tracks
- Use framework for the design: analyse influence of location, field-of-regard, pointing, and other design parameters

ANY  
QUESTIONS ?

