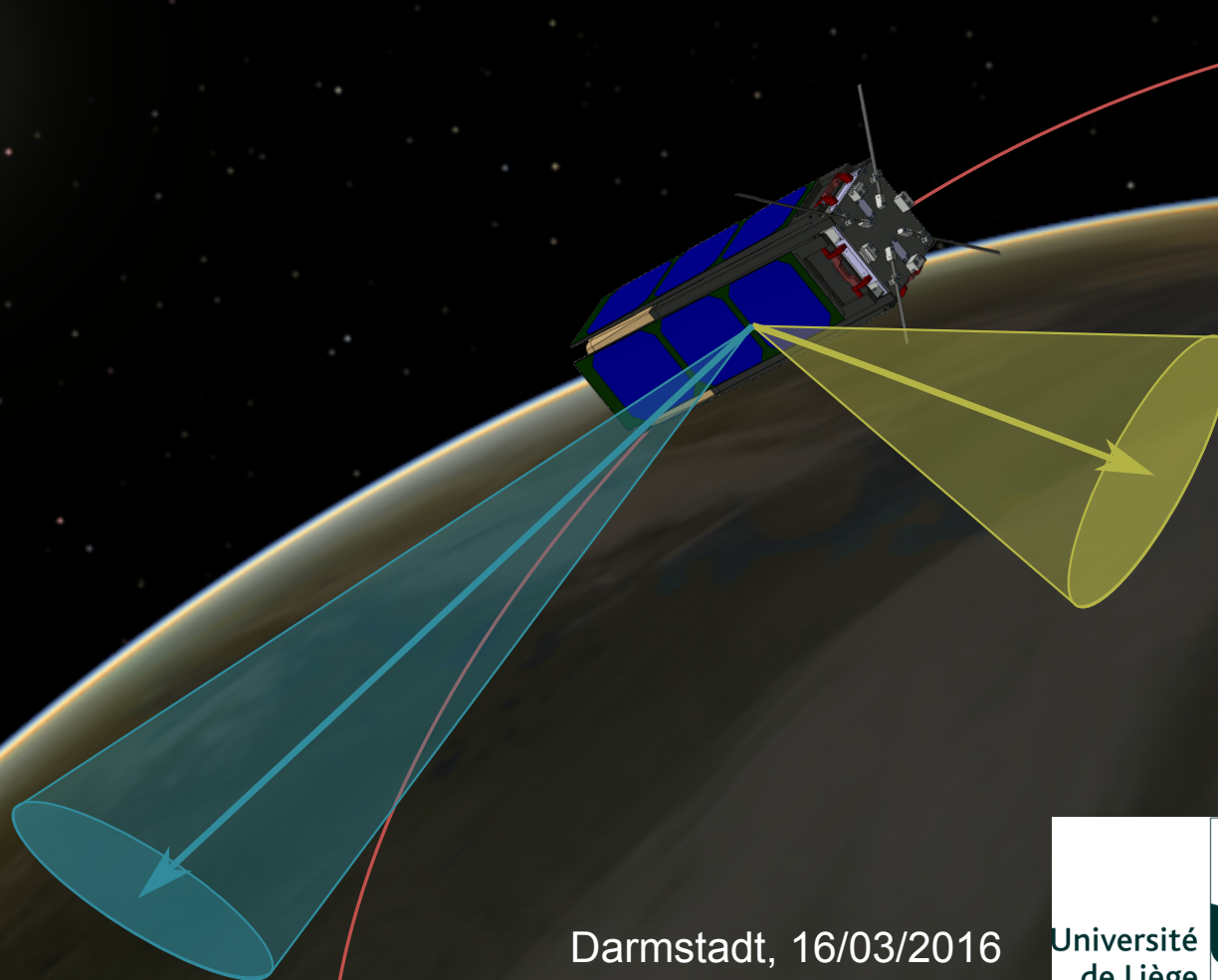


Recursive Estimation of Non-gravitational Perturbations from Satellite Observations

L. Dell'Elce¹, O. Ben-Yaacov², P. Gurfil²

¹Université de Liège

²Technion - Israel Institute of Technology



Darmstadt, 16/03/2016



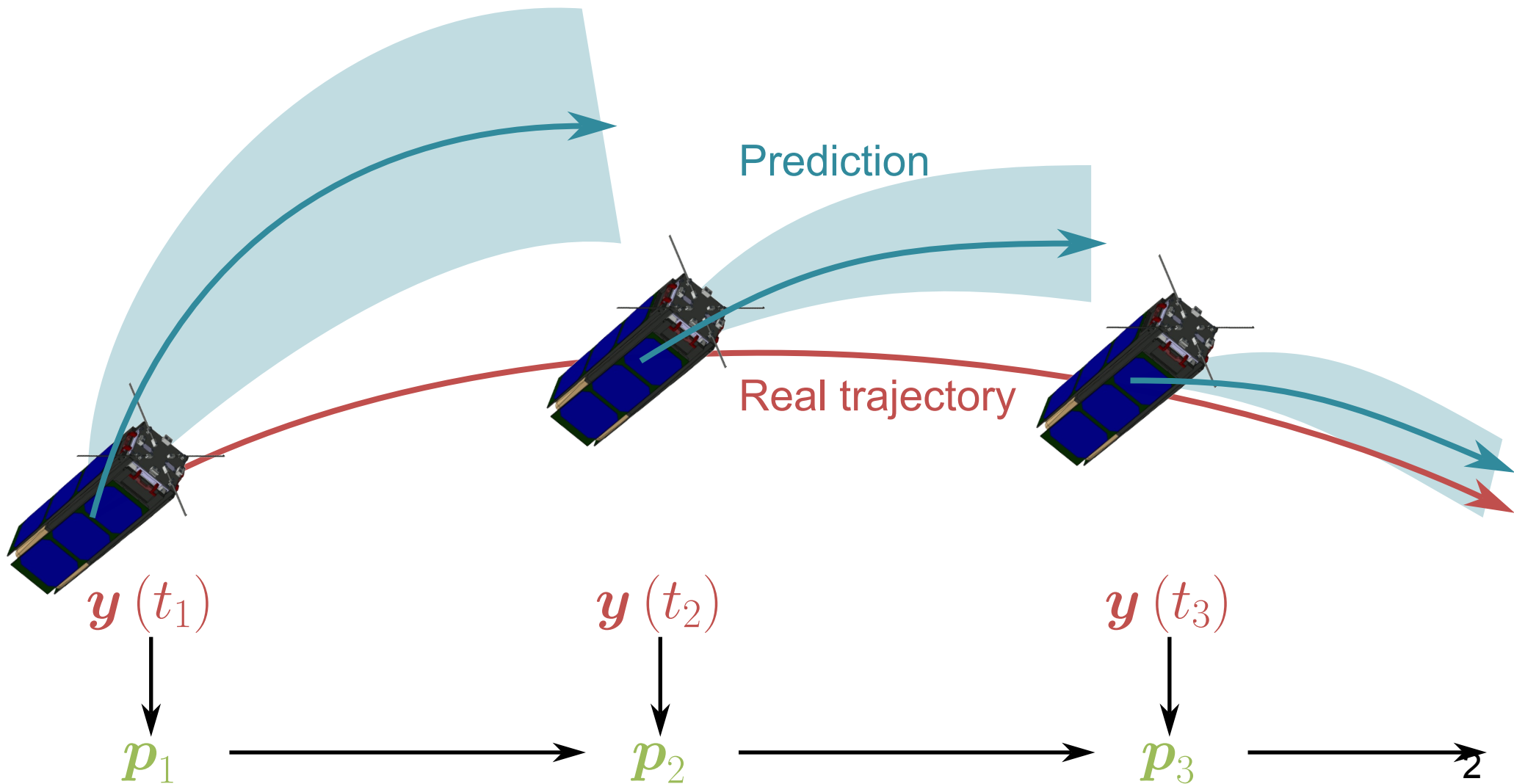
Technion
Israel Institute of Technology

Inferring non-gravitational forces from observations

\boldsymbol{x} \longrightarrow Orbital elements

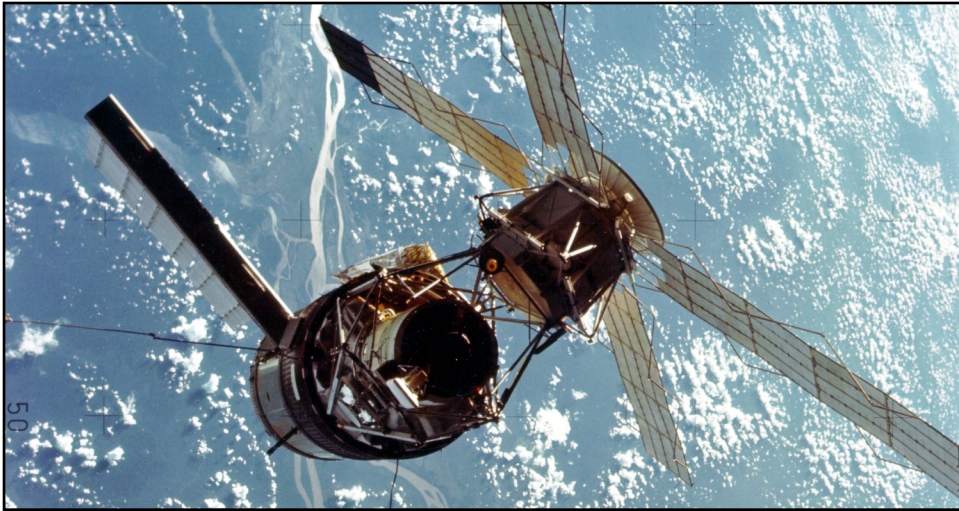
\boldsymbol{y} \longrightarrow Noisy observations

$$\boldsymbol{p} \longrightarrow \begin{cases} \boldsymbol{f}_{drag}(\boldsymbol{x}|\boldsymbol{p}) \\ \boldsymbol{f}_{srp}(\boldsymbol{x}|\boldsymbol{p}) \end{cases}$$

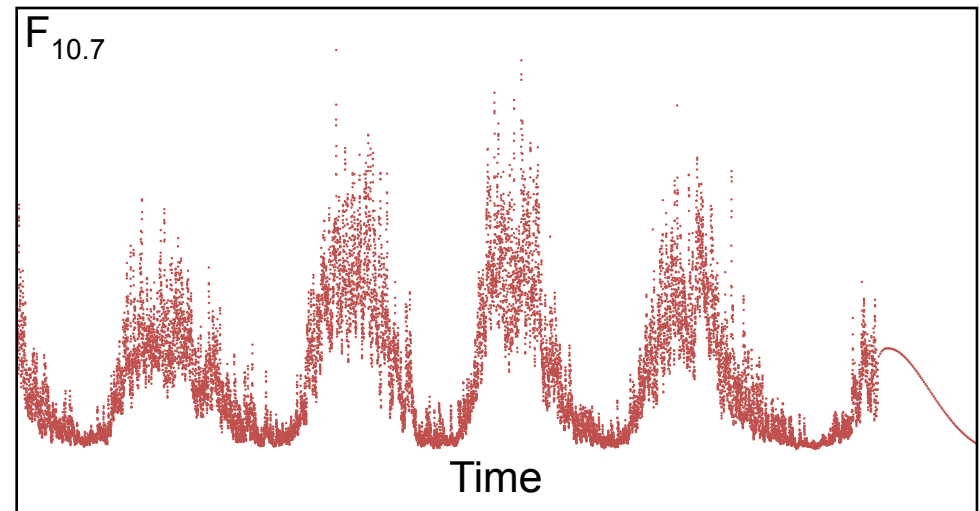


Challenging modeling of non-gravitational forces

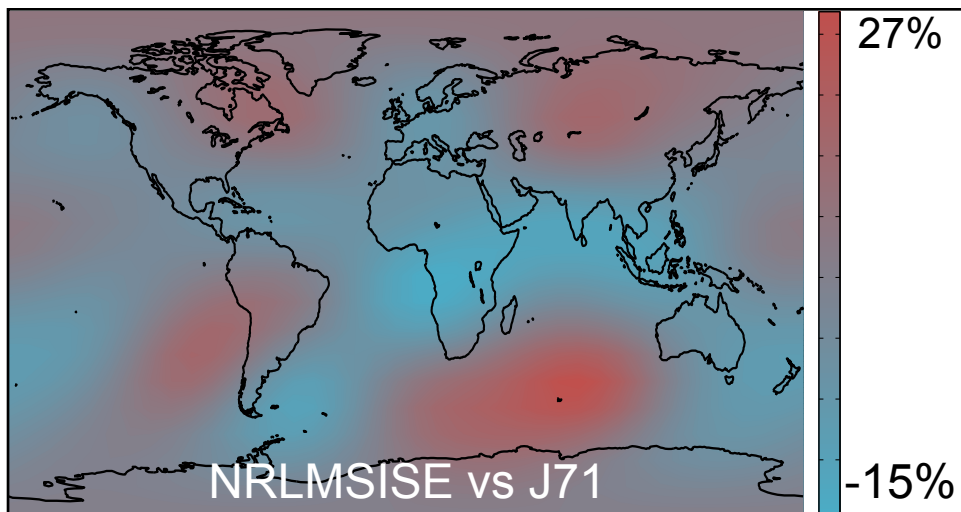
Spacecraft dependent



Stochastic space weather

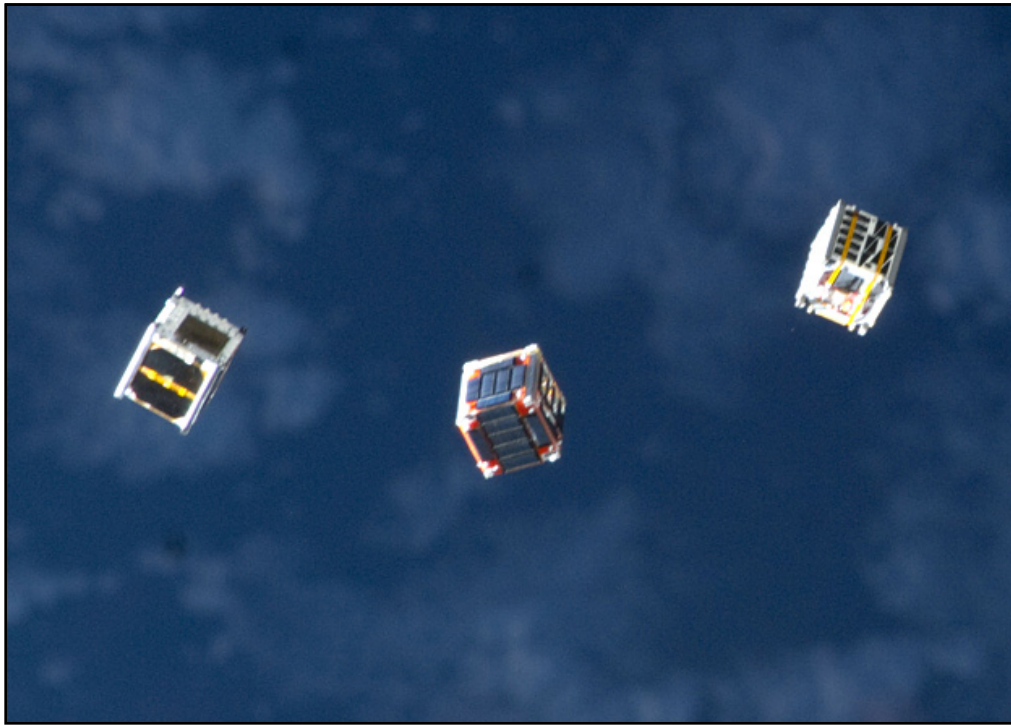


Biased atmospheric models



Why not use high-sensitivity accelerometers?

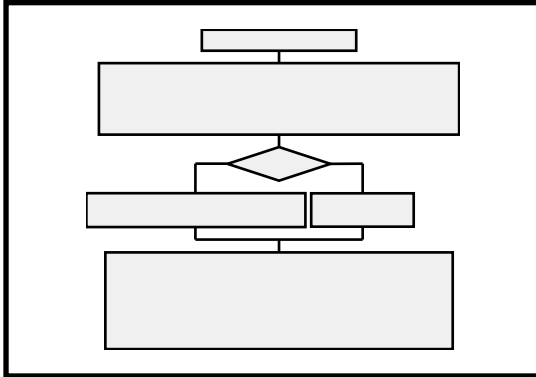
Small satellites



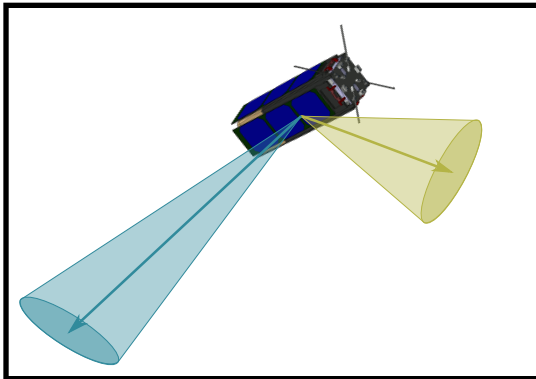
Ground-based estimation



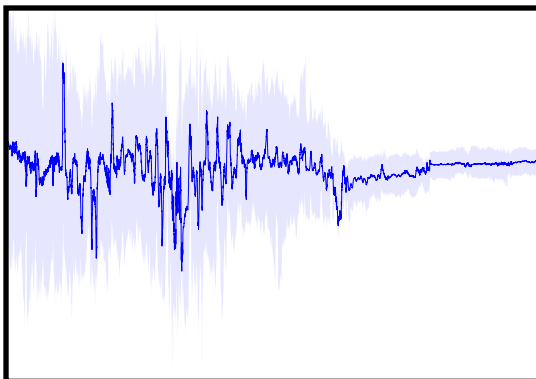
Outline



1. Sequential Monte Carlo for parameter estimation



2. Estimation of non-gravitational perturbations



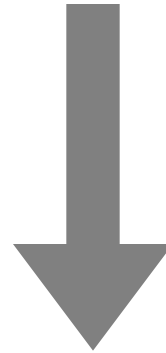
3. Numerical simulations

1. Recursive estimation of hidden Markov models

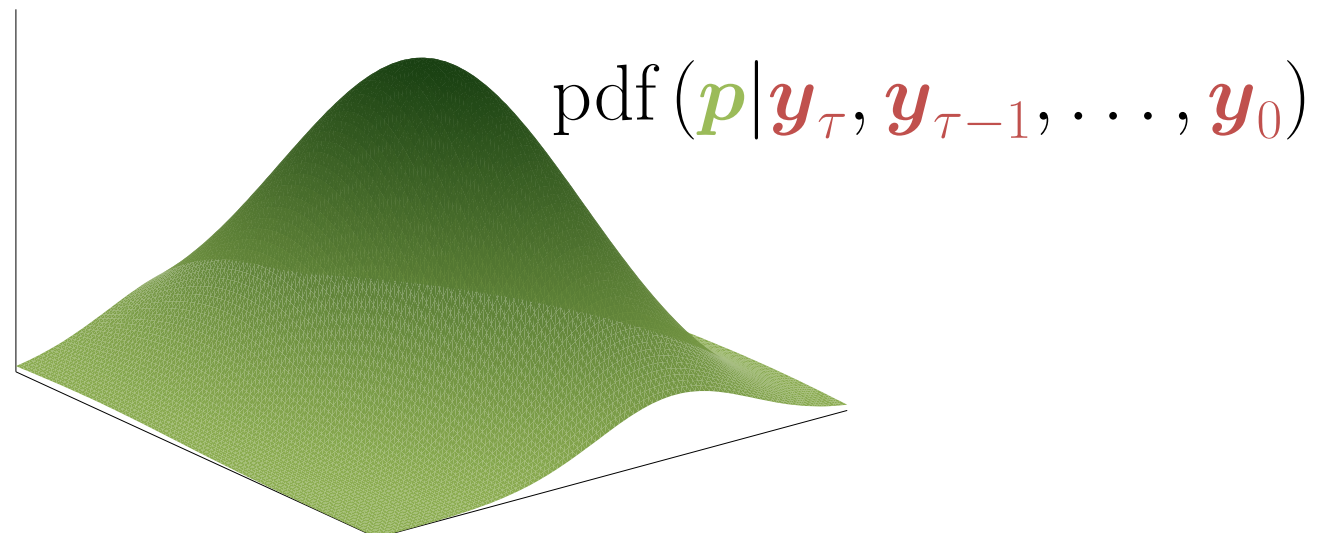
Hidden Markov model

$$\mathbf{X}_{\tau+1} \sim f(\mathbf{x}_{\tau+1} \mid \mathbf{x}_{\tau}, \dots, \mathbf{x}_{\tau-m}, \mathbf{p})$$

$$\mathbf{Y}_{\tau} \sim g(\mathbf{y}_{\tau} \mid \mathbf{x}_{\tau}, \mathbf{p})$$



Posterior distribution



1. Sequential Monte Carlo to estimate the posterior

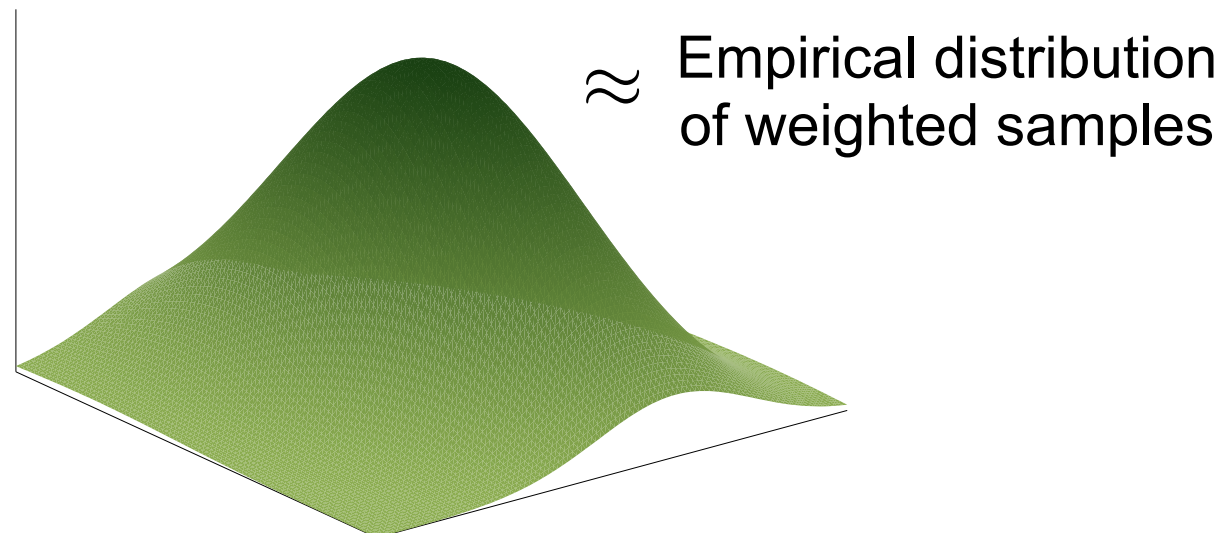
Hidden Markov model

$$\mathbf{X}_{\tau+1} \sim f(\mathbf{x}_{\tau+1} \mid \mathbf{x}_{\tau}, \dots, \mathbf{x}_{\tau-m}, \mathbf{p})$$

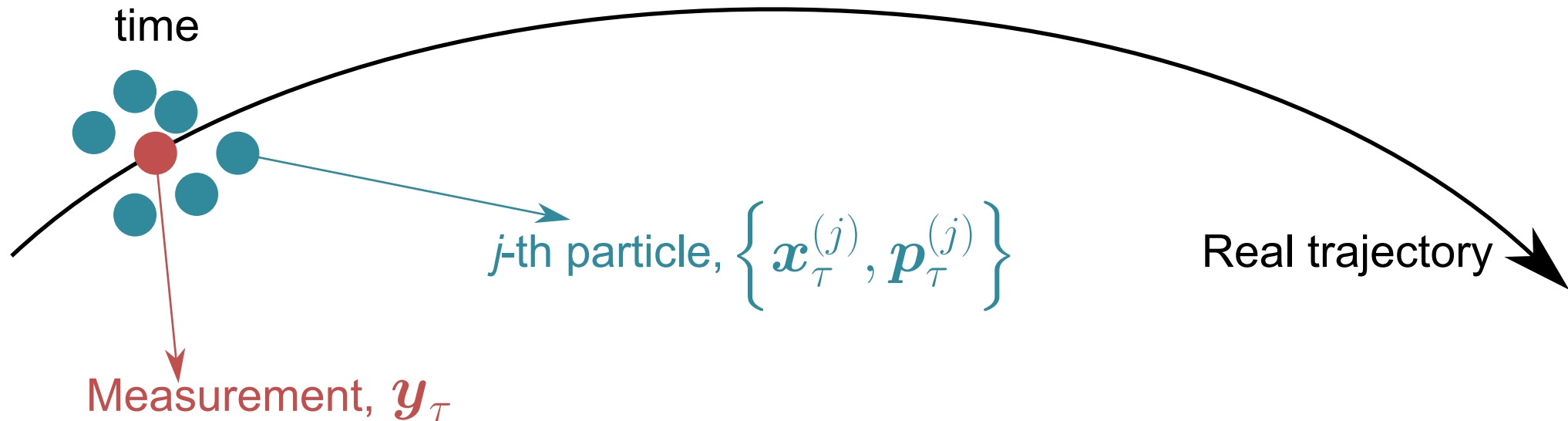
$$\mathbf{Y}_{\tau} \sim g(\mathbf{y}_{\tau} \mid \mathbf{x}_{\tau}, \mathbf{p})$$

Particle filter

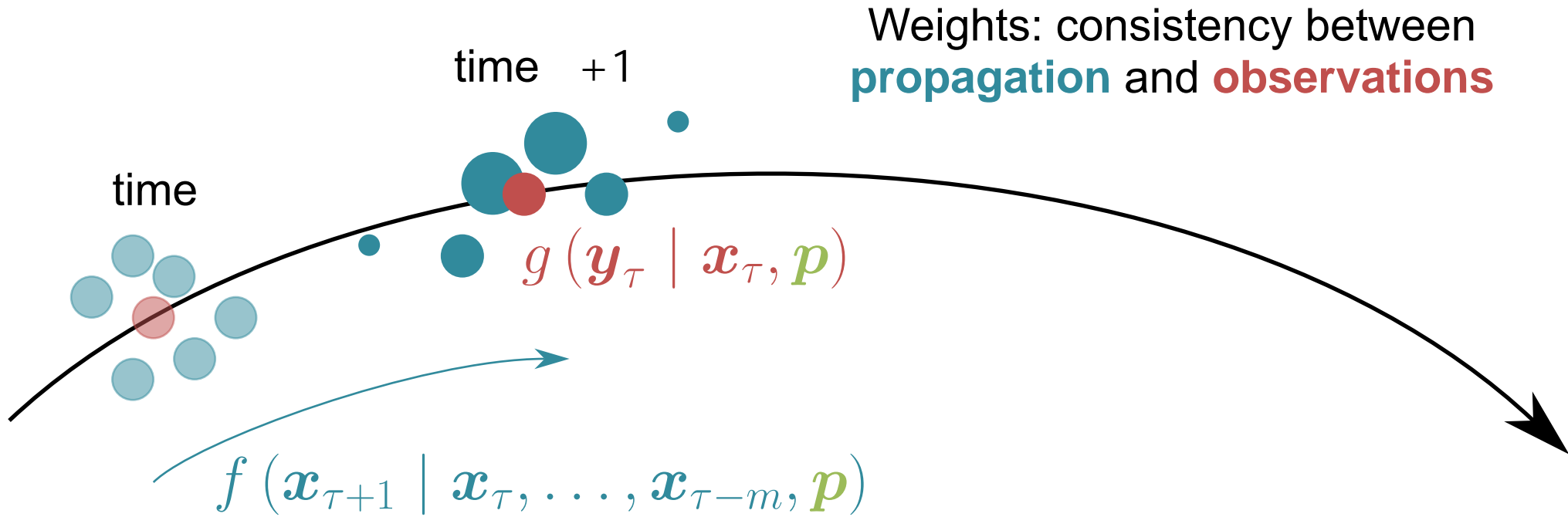
Posterior distribution



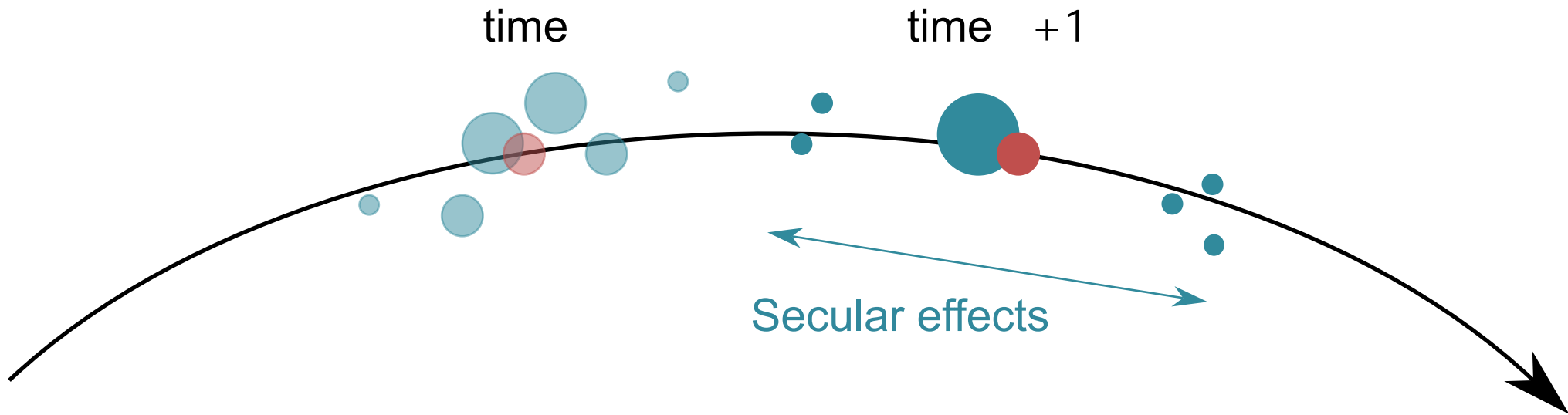
1. Each particle has a set of parameters and states





1. Integrate, don't differentiate



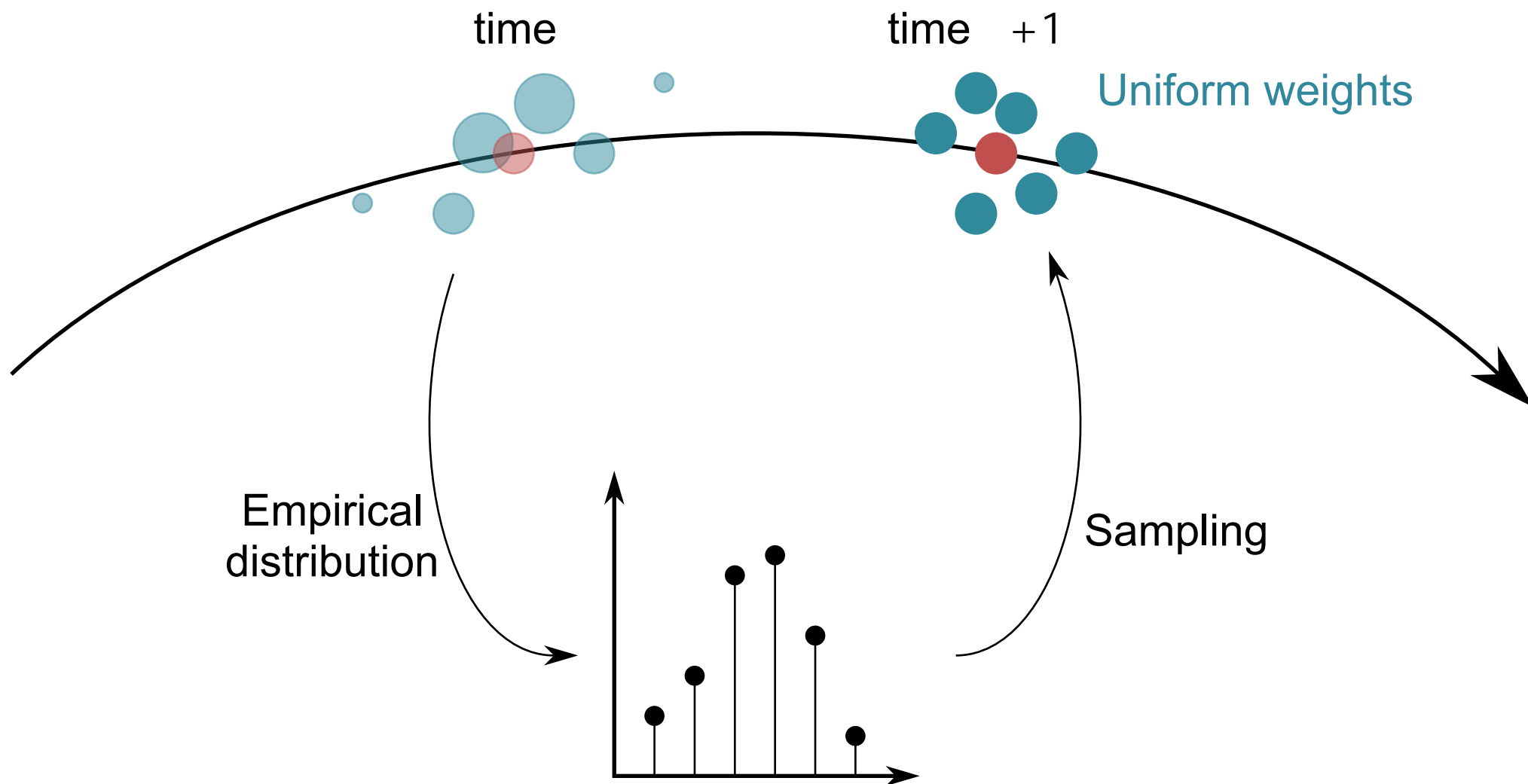
1. Good particles are identified via recursive updates



Multiple updates:

-  Identify good particles
-  Degeneracy

1. Resampling prevents degeneracy of the weights



2. How does it apply to satellite force estimation?

\mathbf{x} \longrightarrow Averaged Orbital elements

\mathbf{y} \longrightarrow Noisy observations

\mathbf{p} \longrightarrow Drag/reflectivity coefficient, density, ...

Measurement noise
(GPS, contact transformation)

$$g(\mathbf{y}_\tau \mid \mathbf{x}_\tau, \mathbf{p})$$

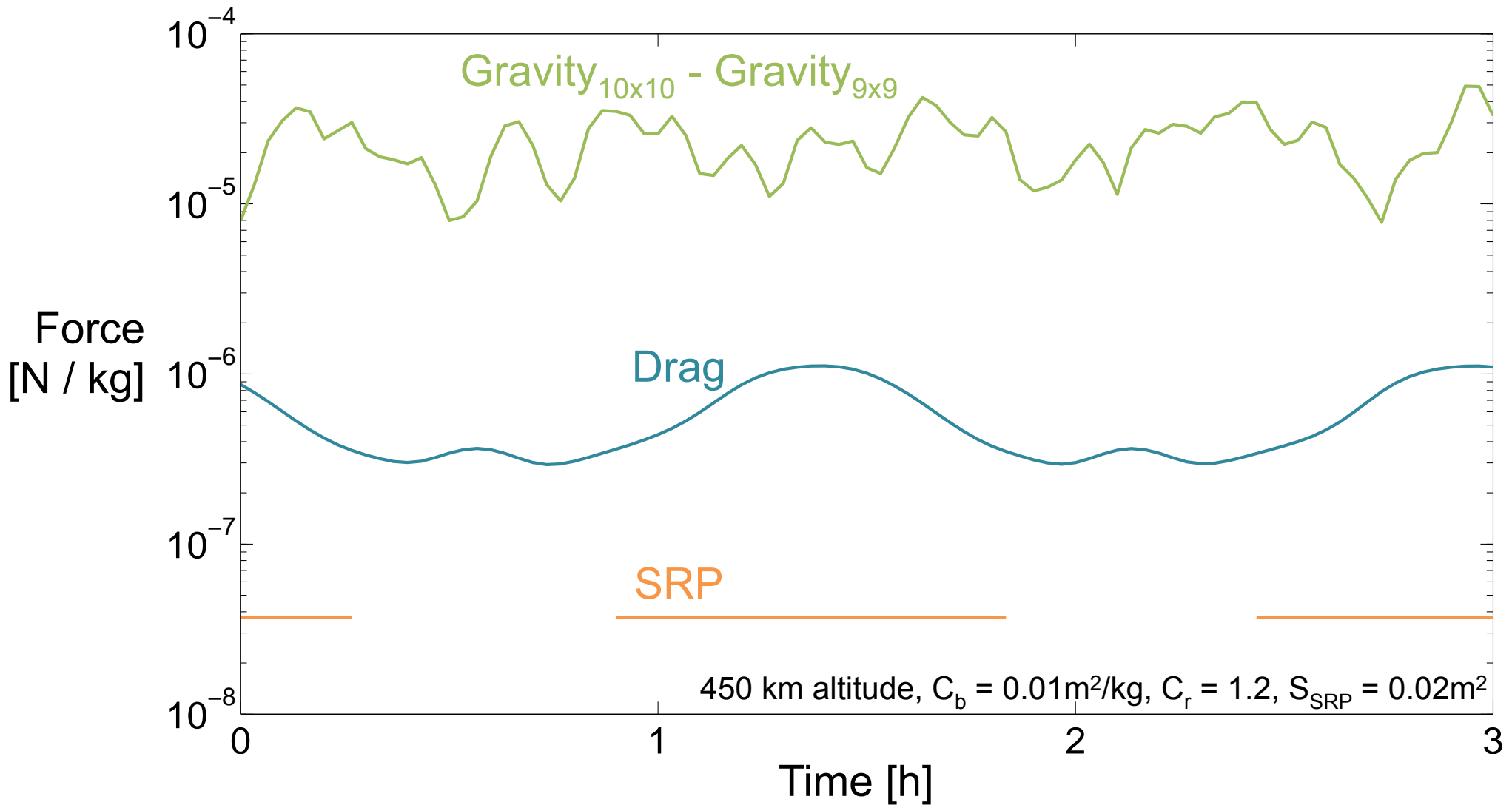
Parametric force model

$$\begin{cases} \mathbf{f}_{drag}(\mathbf{x} \mid \mathbf{p}) \\ \mathbf{f}_{srp}(\mathbf{x} \mid \mathbf{p}) \end{cases}$$

Orbital propagator & process noise

$$f(\mathbf{x}_{\tau+1} \mid \mathbf{x}_\tau, \dots, \mathbf{x}_{\tau-m}, \mathbf{p})$$

2. Why averaged elements?



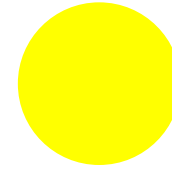
Robustness to unmodeled dynamics
Analytical & semi-analytical techniques



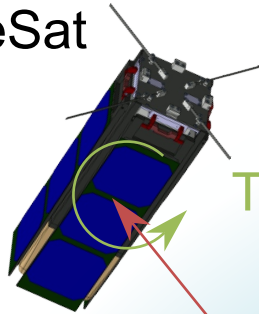
Measurement noise

3. Simulations in a high-fidelity environment

Solar radiation pressure



3U CubeSat



Time-varying pitch angle

450 km



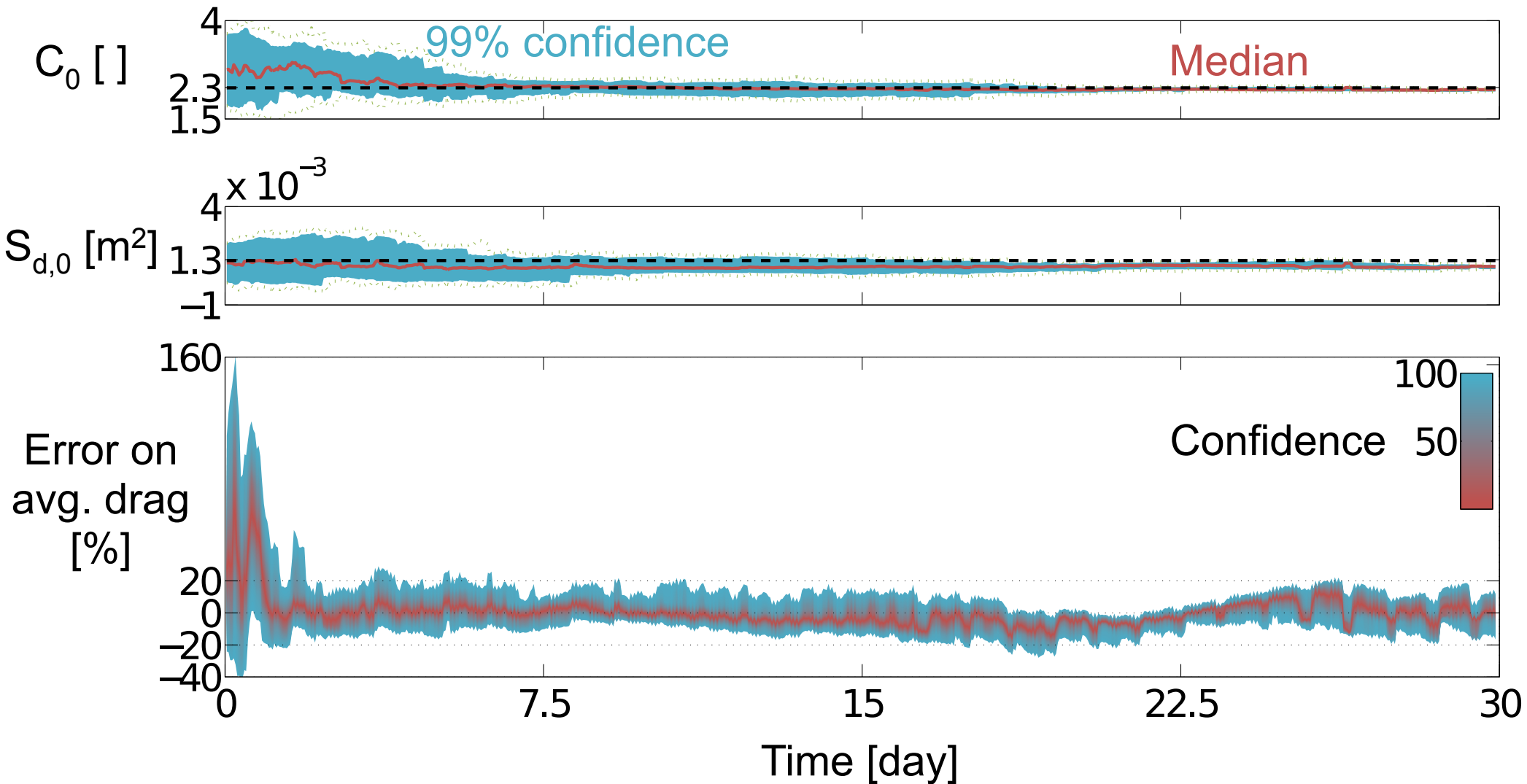
NRLMSISE,
variable accommodation,
thermal flow,
time-varying solar activity

10x10 gravitational harmonics
Third-body (Sun and Moon)

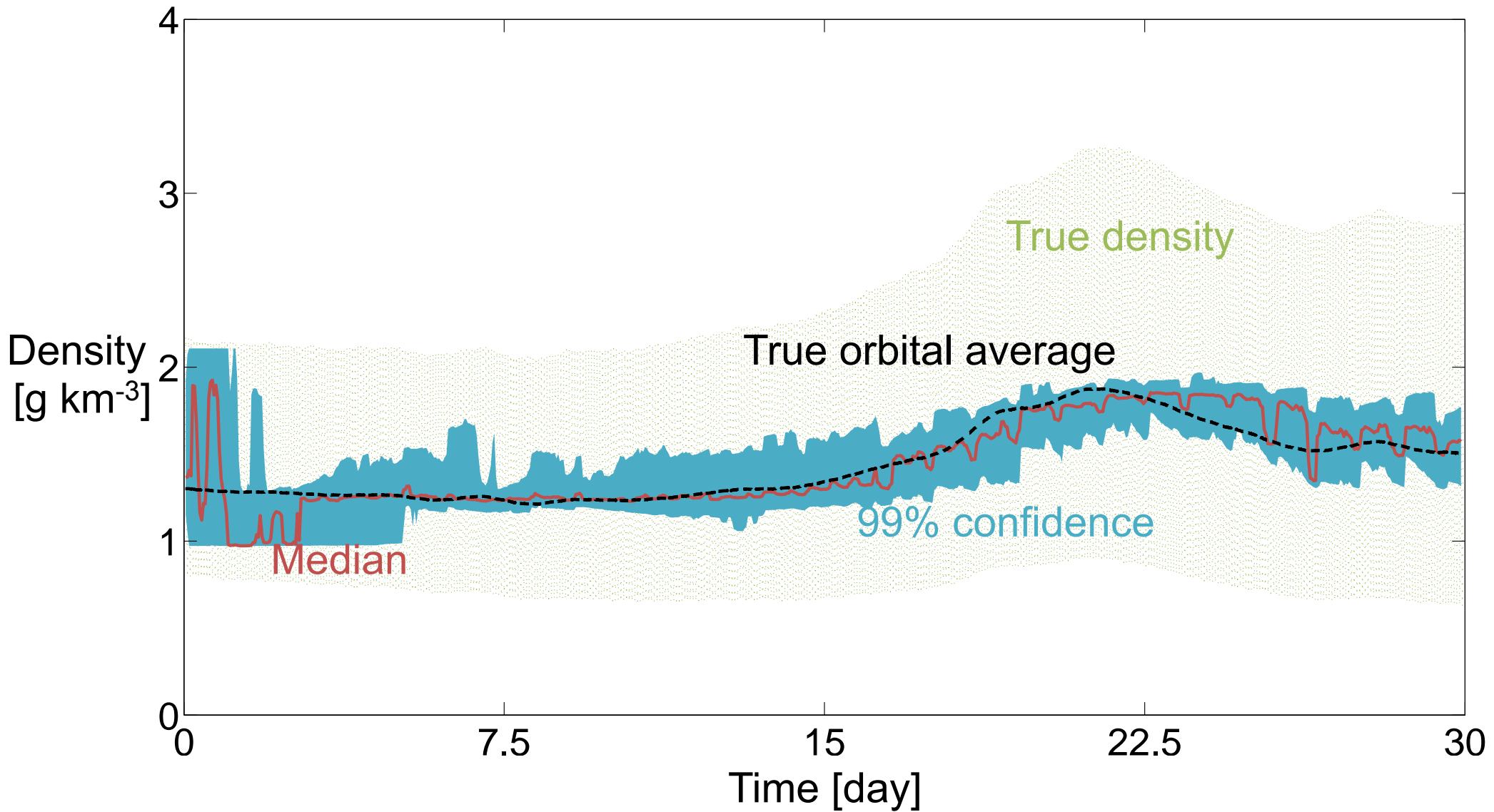
3. Comparing two implementations of the filter

| | Analytical | Numerical |
|-------------------------------------|------------------------------------|---|
| Modeled perturbations | $J_2 + \text{averaged drag}$ | $J_{4 \times 4}, \text{drag, SRP}$ |
| Force model | | |
| Ballistic coefficient | $\frac{(S_d(t) + S_{d,0})}{m} C_d$ | $\frac{(S_d(t) + S_{d,0})}{m} C_d$ |
| Density | $\bar{\rho}_0$ | $\bar{\rho}_0 \sum_i (c_i \cos(i L) + s_i \sin(i L))$ |
| Reflectivity coefficient | n/a | C_r |
| Time step | 1 hour | 3 min |
| Filter order, m | 4 | 1 |

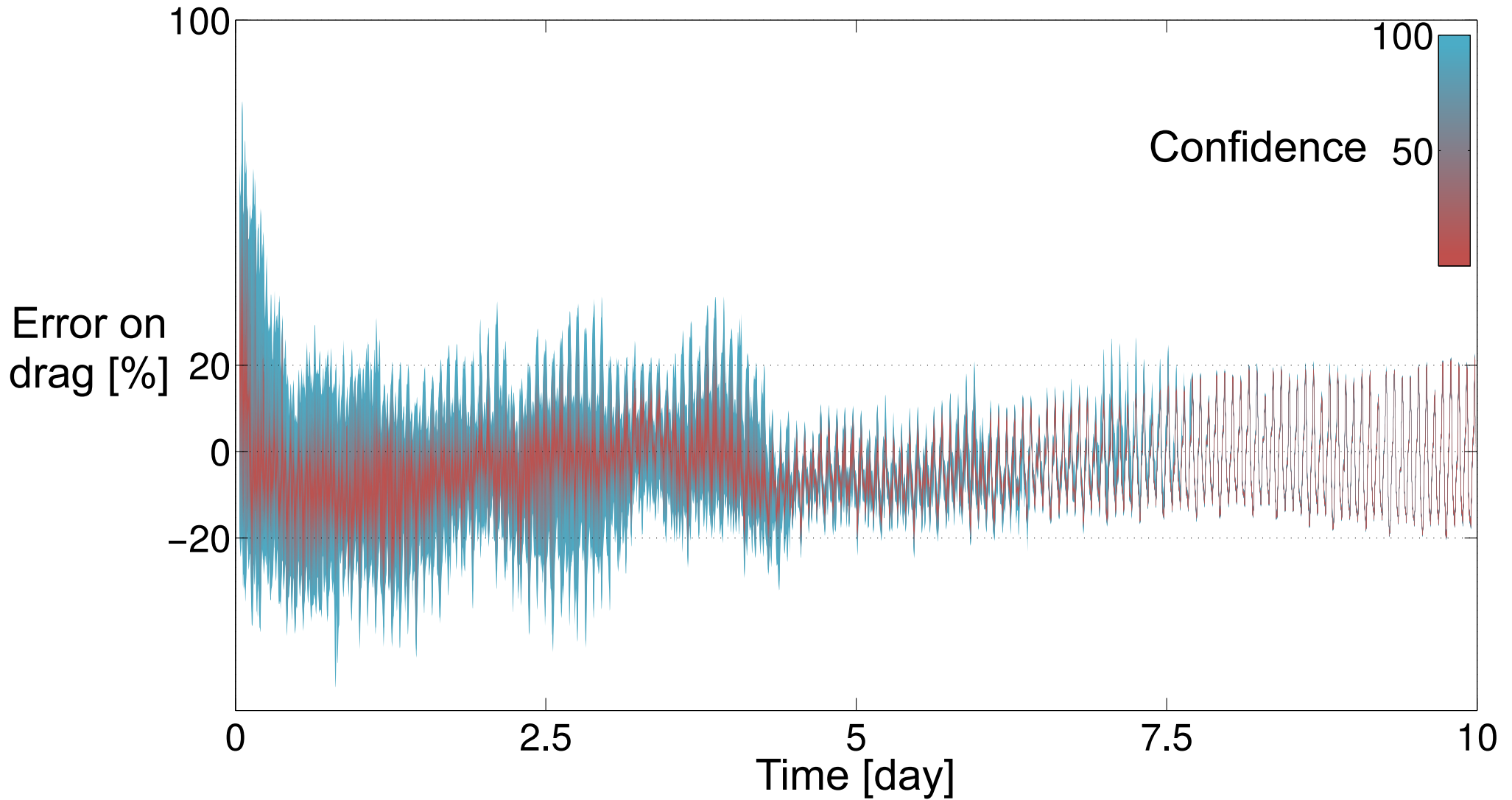
3. Estimation using the analytical propagator



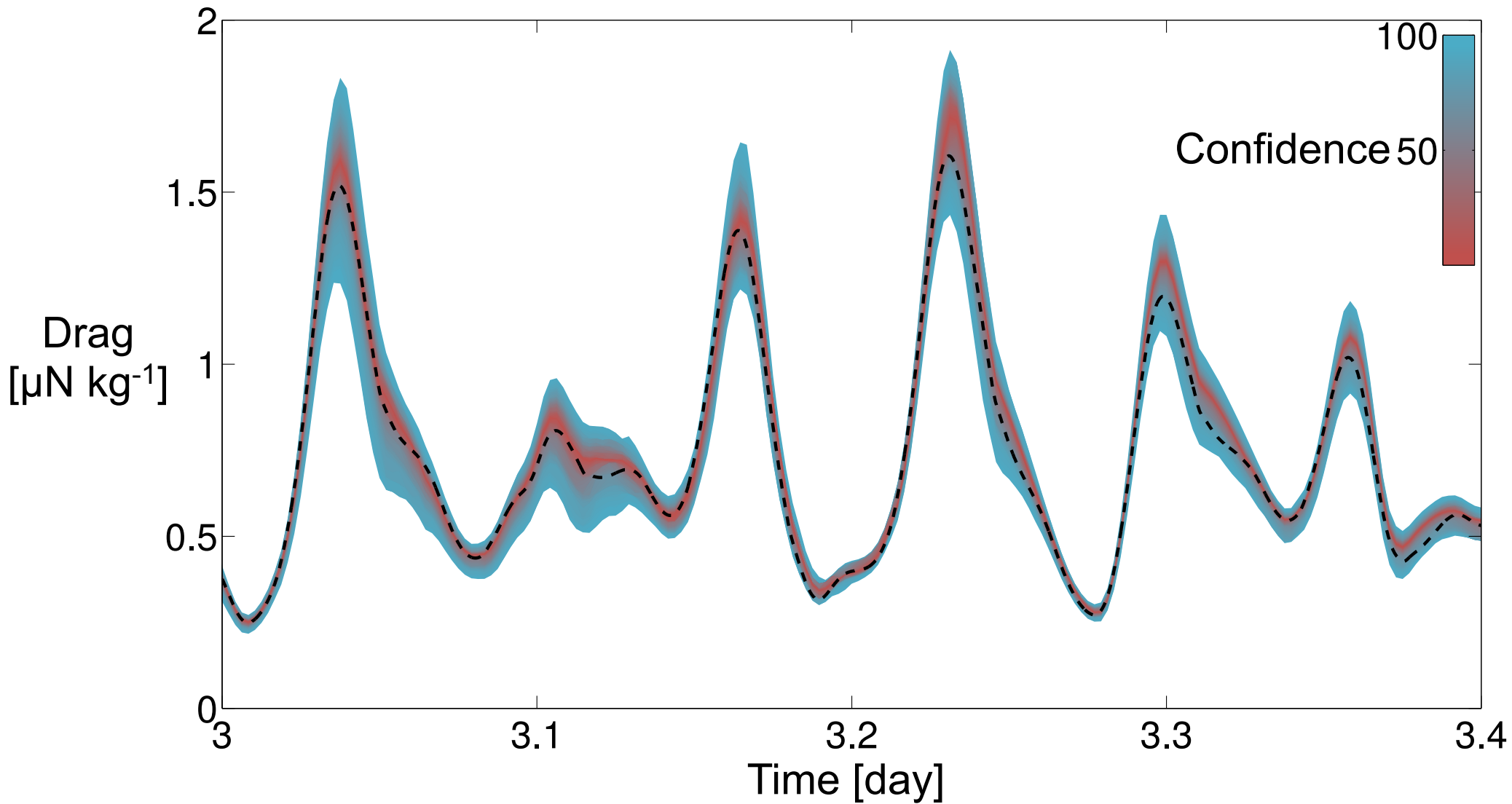
3. Solar activity variations are automatically detected



3. Estimation using the numerical propagator



3. Numerical propagators yield more flexibility



Conclusion

Particle filter enables extreme **flexibility** (nonlinear & non-Gaussian model)

Averaged elements are used to mitigate mis-modeled dynamics

Computational burden can be reduced by using **analytical techniques**

Good estimation from **sparse & fairly inaccurate measurements**

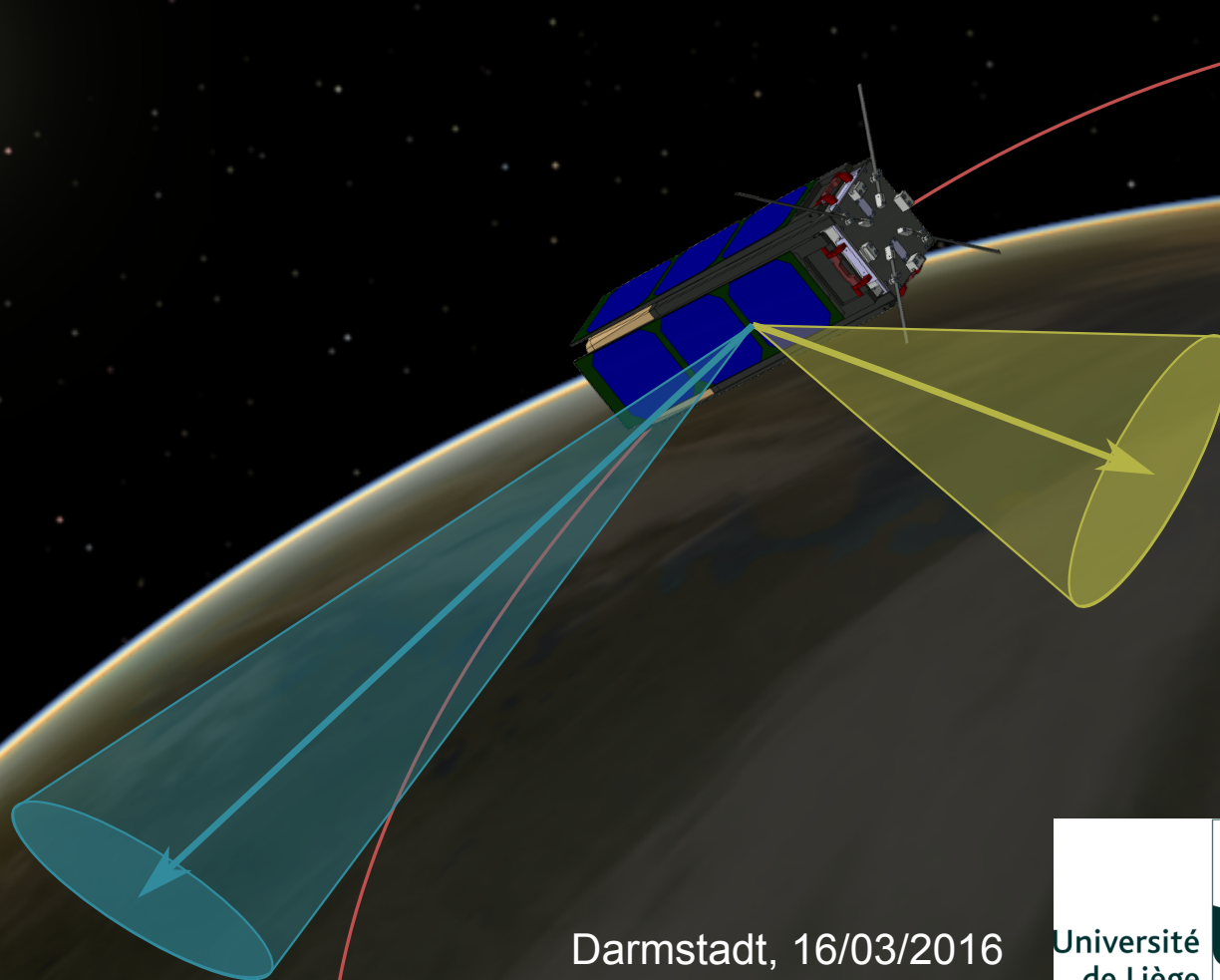
Adequate **tuning of filter's parameter** is mandatory
(recommendations are available in the paper)

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