

Performance Comparison of Particle Filter in Small Satellite Attitude Estimation

SSC21-WKV-03

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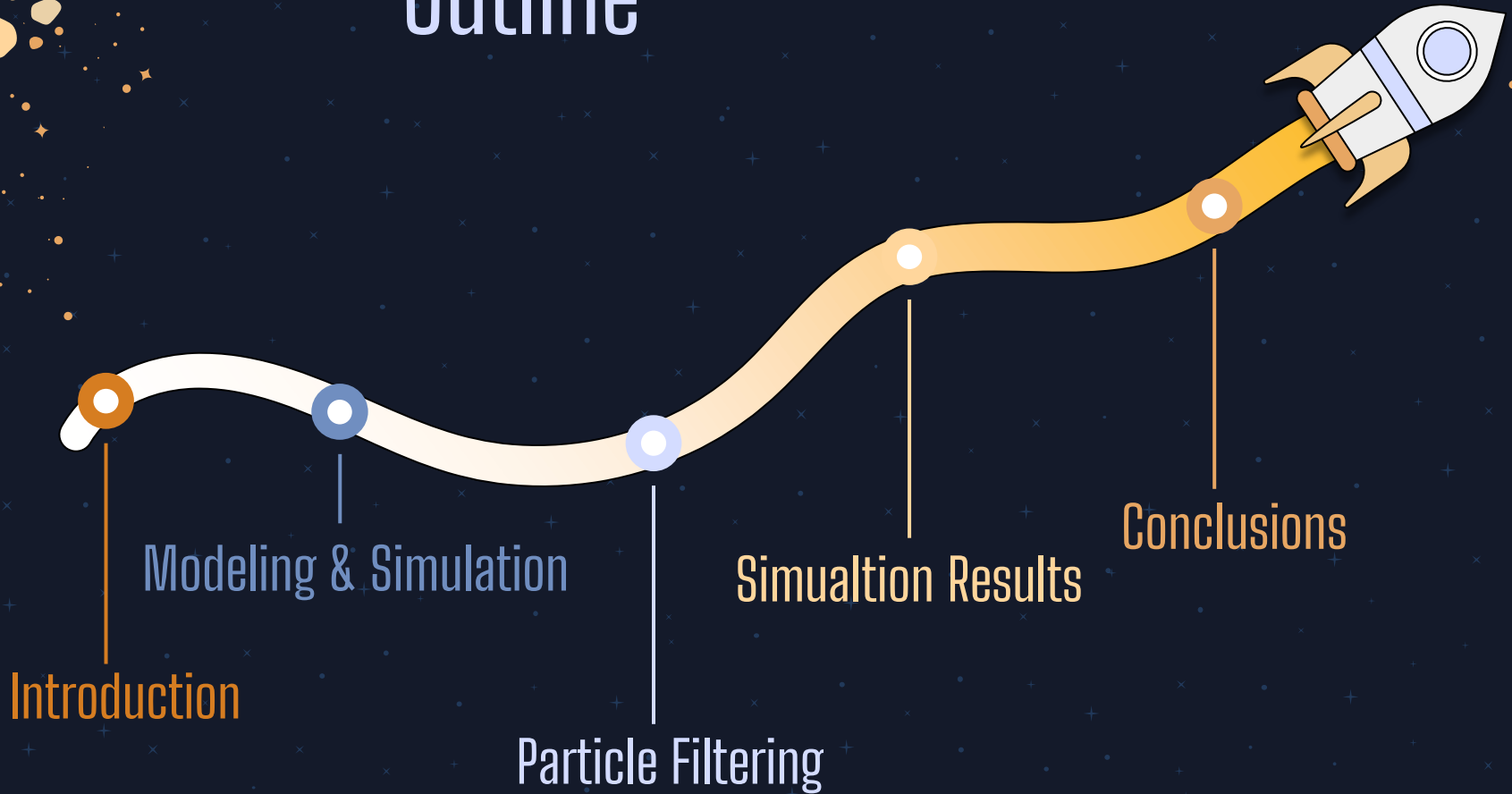


35th Annual Small Satellite Conference
August 7 – 12, 2021



Small satellite missions with higher complexities tend to demand more sophisticated requirements, which push the limits of classical attitude estimation methods.

Outline

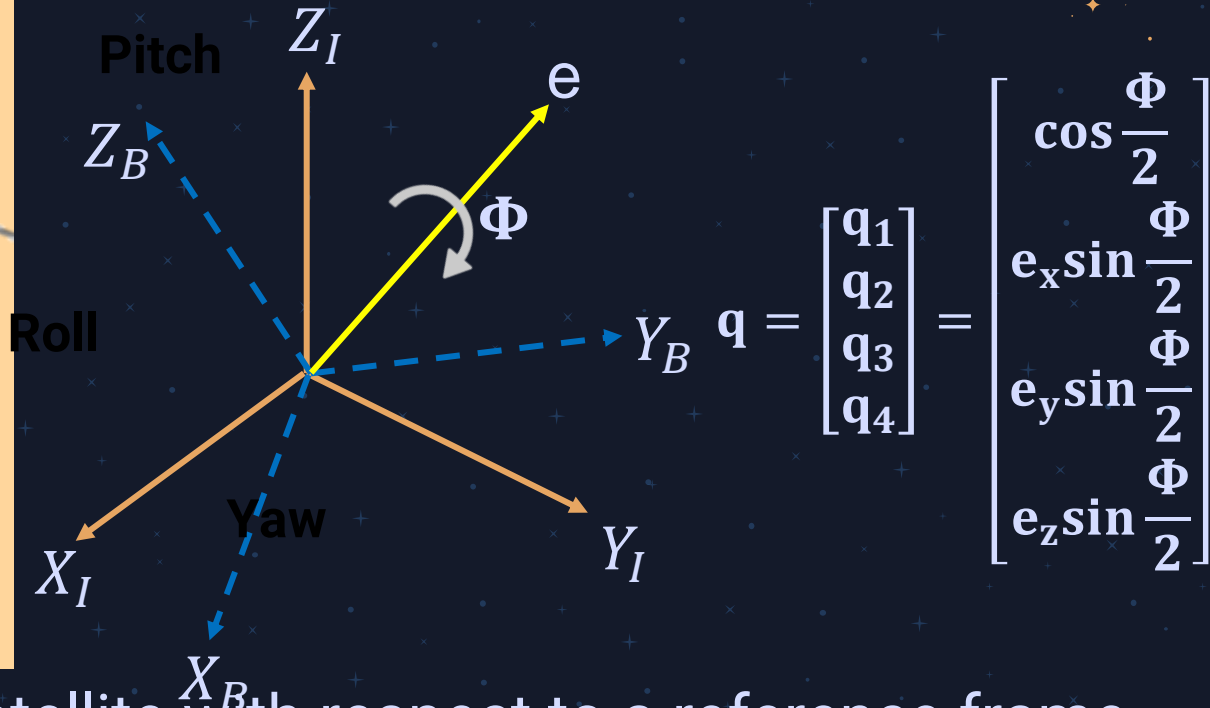
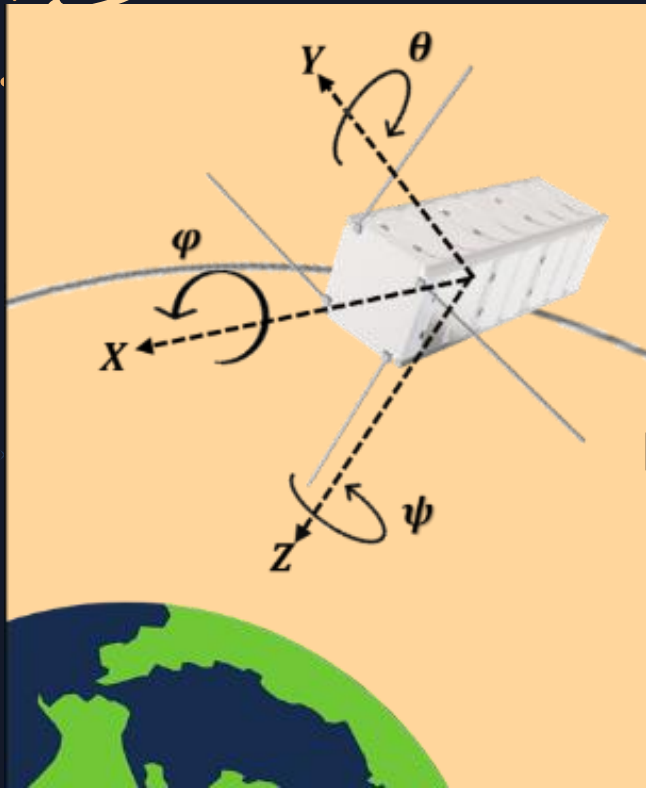


The background is a dark blue space scene filled with small white stars and yellow dots. On the left, a white rocket with a blue stripe is launching upwards, leaving a long, wavy yellow and white trail. On the right, a portion of the Earth is visible, showing blue oceans and green landmasses. Three satellite-like objects with purple solar panels are orbiting the Earth. In the upper right corner, a yellow and orange planet, resembling Jupiter, is visible.

1

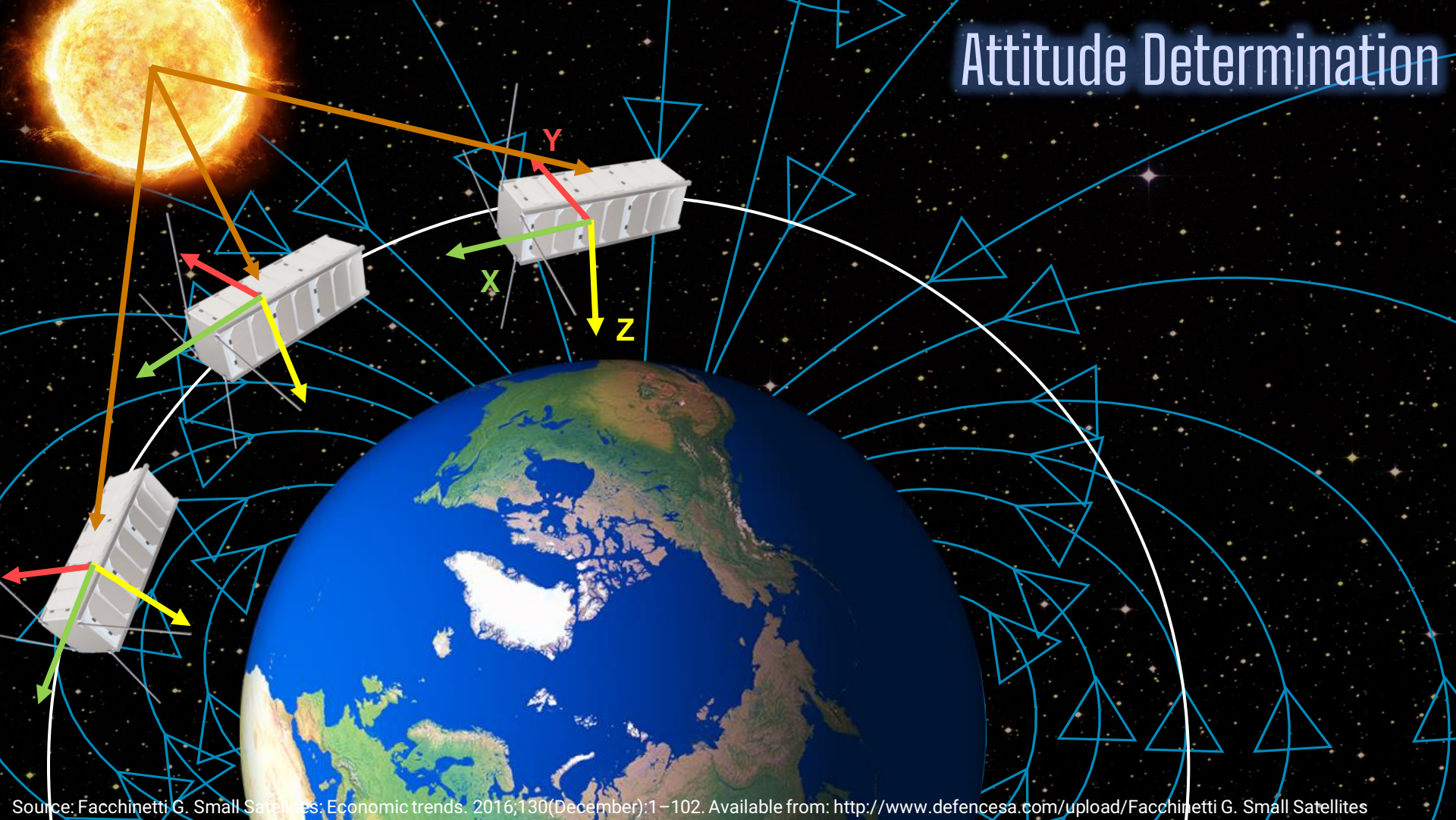
Introduction

Satellite Attitude



The orientation of a satellite with respect to a reference frame.

Attitude Determination



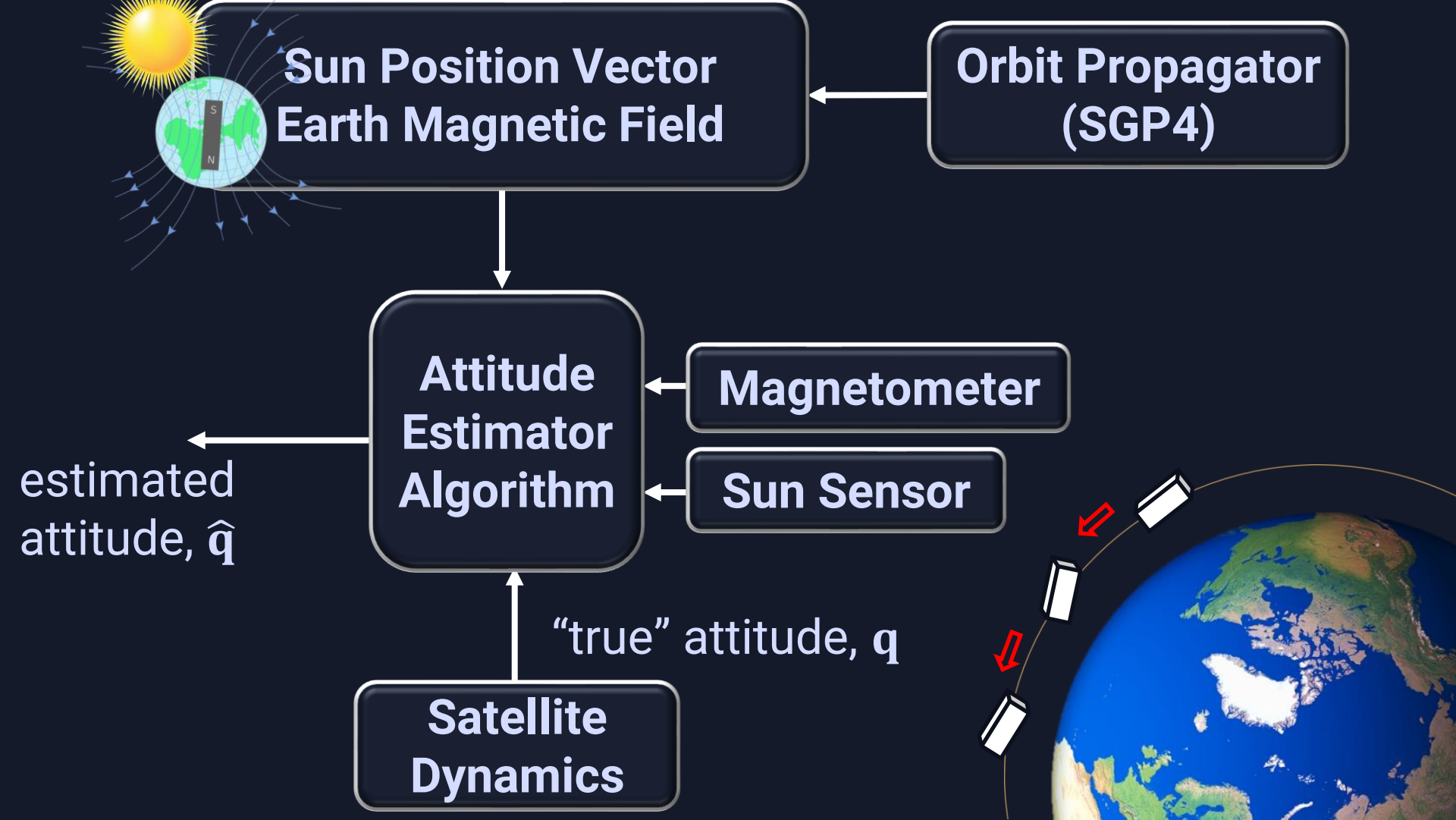
Attitude Estimation Algorithms

- TRIAD, Q-Method, Least Squares
 - Kalman Filter – Extended Kalman Filter (EKF)
 - Particle Filter
 - H-Infinity Filter
-
- The Particle Filter has been proposed for satellite attitude estimation and has shown improved performance compared with traditional approaches.



2

Modeling & Simulation





Satellite Dynamic & Kinematic Equations

Parameter	Value	
Small Satellite Properties	CubeSat Type	3U
	Mass	3.5 kg
	Dimensions	0.34 m × 0.10 m × 0.10 m
	Inertia matrix	$\begin{bmatrix} 0.0058 & 0 & 0 \\ 0 & 0.0366 & 0 \\ 0 & 0 & 0.0366 \end{bmatrix} \text{ kgm}^2$



Satellite Dynamic & Kinematic Equations

$$\dot{\omega} = \begin{bmatrix} -\left(\frac{I_z - I_y}{I_x}\right) \omega_y \omega_z + \frac{\tau_x}{I_x} \\ -\left(\frac{I_x - I_z}{I_y}\right) \omega_x \omega_z + \frac{\tau_y}{I_y} \\ -\left(\frac{I_y - I_x}{I_z}\right) \omega_x \omega_y + \frac{\tau_z}{I_z} \end{bmatrix}$$

Attitude Dynamics

$$\dot{q} = \frac{1}{2} \begin{bmatrix} 0 & \omega_z & -\omega_y & \omega_x \\ -\omega_z & 0 & \omega_x & \omega_y \\ \omega_y & -\omega_x & 0 & \omega_z \\ -\omega_x & -\omega_y & -\omega_z & 0 \end{bmatrix} \begin{bmatrix} q_1 \\ q_2 \\ q_3 \\ q_4 \end{bmatrix}$$

Attitude Kinematics

Sensors – Mathematical Models

Magnetometer $\mathbf{B}_{\text{meas}} = A(\mathbf{q})\mathbf{B}_{\text{ECI}} + \mathbf{v}_{\text{mag}}$

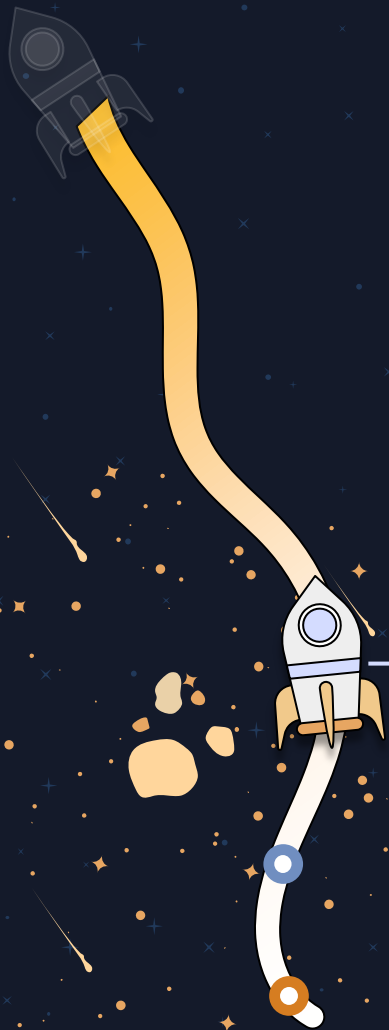
Sun Sensor $\mathbf{S}_{\text{meas}} = A(\mathbf{q})\mathbf{S}_{\text{ECI}} + \mathbf{v}_{\text{sun}}$

$$A(\mathbf{q}) = (q_4^2 - \mathbf{q}_{13}^T \mathbf{q}_{13})\mathbf{I}_{3 \times 3} - 2q_4[\mathbf{q}_{13} \times] + 2\mathbf{q}_{13}\mathbf{q}_{13}^T$$

3

Particle Filtering

The PF is a nonlinear estimation algorithm that approximates nonlinear functions using a set of random particles.



Models in PF Attitude Estimation



Attitude Dynamics Model

System Model

$$\mathbf{x}_{k+1} = f_k(\mathbf{x}_k, \mathbf{w}_k) \leftrightarrow p(\mathbf{x}_{k+1} | \mathbf{x}_k)$$



Simulated Sensors

Measurement Model

$$y_k = h_k(\mathbf{x}_k, \mathbf{v}_k) \leftrightarrow p(y_k | \mathbf{x}_{k+1})$$

Models in PF Attitude Estimation

$$\mathbf{x} = [\mathbf{q} \quad \boldsymbol{\omega}]^T = [q_1 \quad q_2 \quad q_3 \quad q_4 \quad \omega_x \quad \omega_y \quad \omega_z]^T$$

$$\dot{\mathbf{x}} = \begin{bmatrix} \dot{\mathbf{q}} \\ \dot{\boldsymbol{\omega}} \end{bmatrix} = \begin{bmatrix} \frac{1}{2} \boldsymbol{\Omega}(\mathbf{q}) \\ \mathbf{I}^{-1}(\boldsymbol{\tau} - \boldsymbol{\omega} \times \mathbf{I}\boldsymbol{\omega}) \end{bmatrix} \mathbf{x}$$

Particle Filter – Schematic

$i = 1, \dots, N$

$k = 0$ **Initialization**

Sample $x_0^i \sim p(x_0)$,
set $W_0^i = 1/N$

$k \rightarrow k + 1$

Prediction/ Importance Sampling

Sample $\tilde{x}_k^i \sim q(x_k | x_{k-1}^i)$

Measurement y_k

Update

Calculate the importance weights of particles

$$W_k^{(i)} = \frac{W_{k-1}^{(i)} p(y_k | x_k^{(i)}) p(x_k^{(i)} | x_{k-1}^{(i)})}{q(x_k^{(i)} | x_{k-1}^{(i)}, y_k)}$$

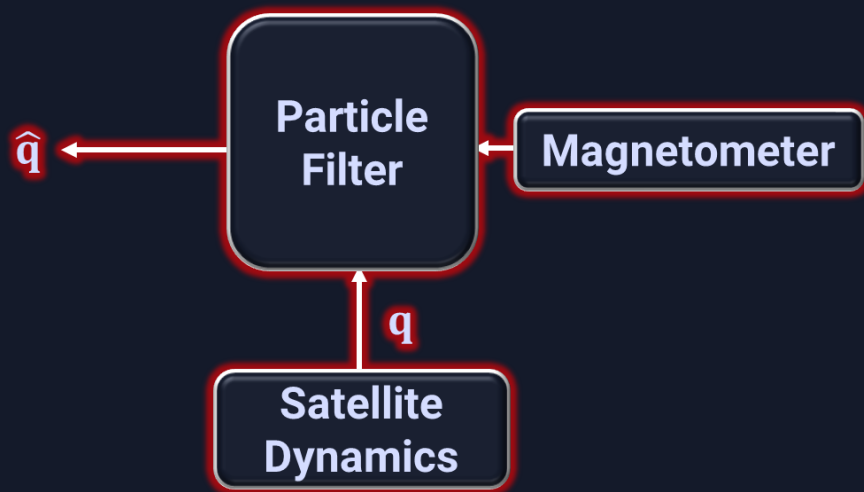
Normalize the weights $w_k^{(i)} = \frac{W_k^{(i)}}{\sum_{i=1}^N W_k^{(i)}}$

Resampling

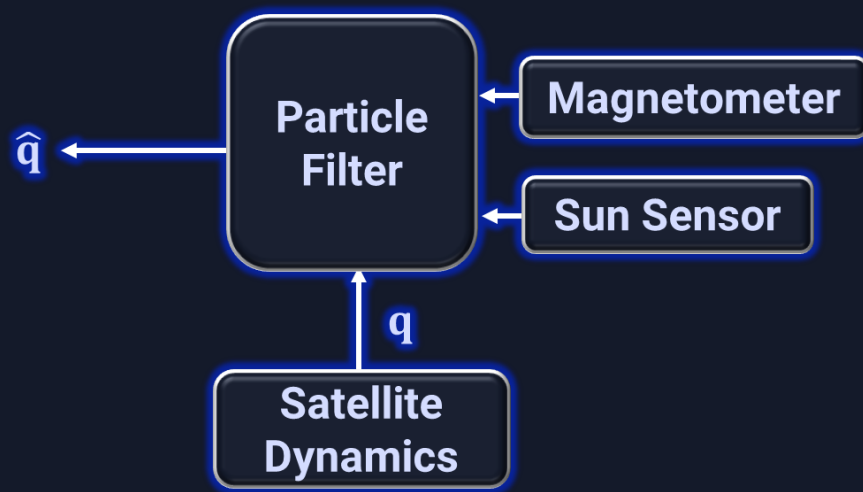
$$p(x_k | y_{1:k}) \approx \sum_{i=1}^N W_k^i x_{k,k-1}^i$$

Estimation \hat{x}_k

PF-MAG



PF-MAG+SUN



PF-MAG

PF-MAG+SUN

$k = 0$ Initialization
Sample $x_0^i \sim p(x_0)$,
set $W_0^i = 1/N$

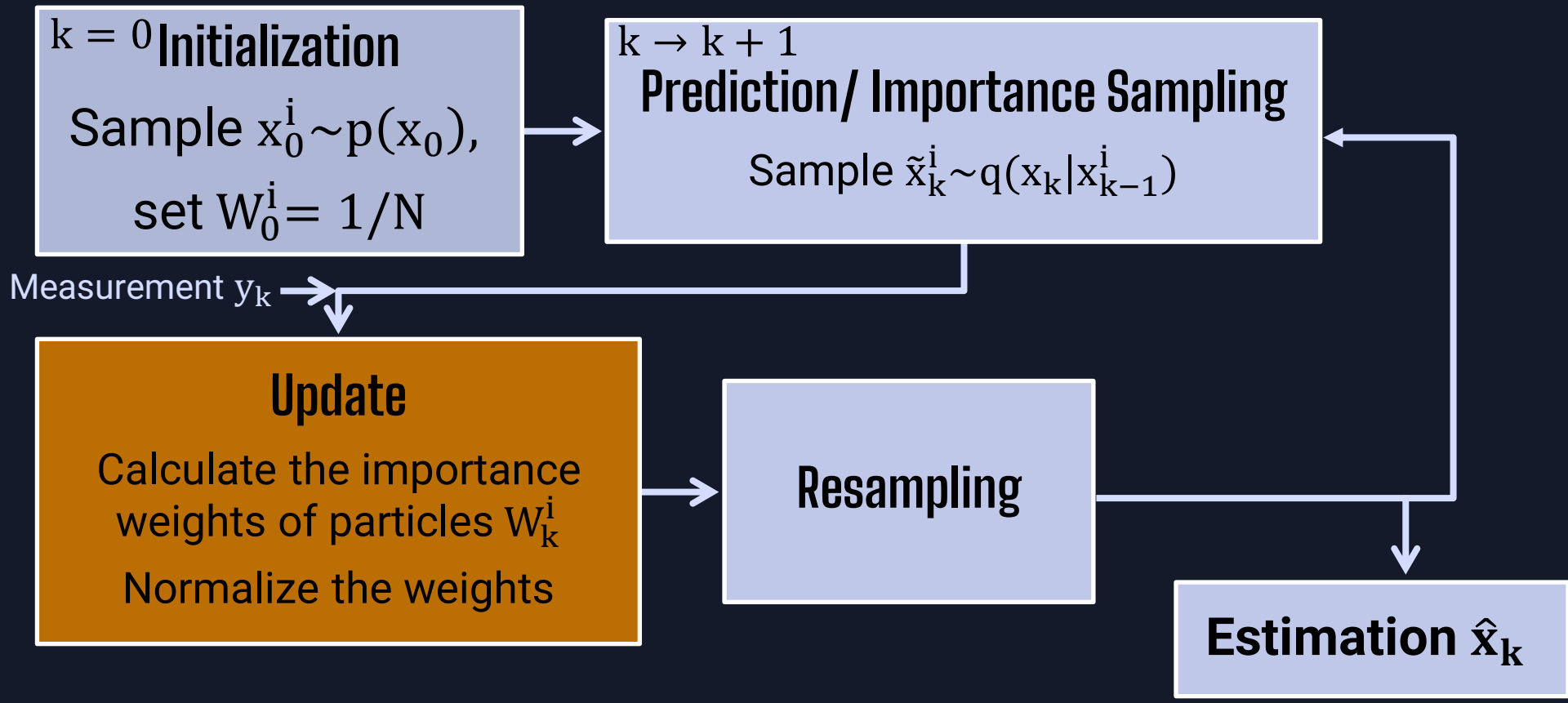
$k \rightarrow k + 1$
Prediction/ Importance Sampling
Sample $\tilde{x}_k^i \sim q(x_k | x_{k-1}^i)$

Measurement y_k →

Update
Calculate the importance weights of particles W_k^i
Normalize the weights

Resampling

Estimation \hat{x}_k



Kalman Gain

$$\mathbf{K} = \mathbf{P}_k^- \mathbf{H}_k^T (\mathbf{H}_k \mathbf{P}_k \mathbf{H}_k^T + \mathbf{R}_k)^{-1}$$

Extended Kalman Filter (EKF) EKF

$\hat{\mathbf{x}}_0, \mathbf{P}_0$

Prediction

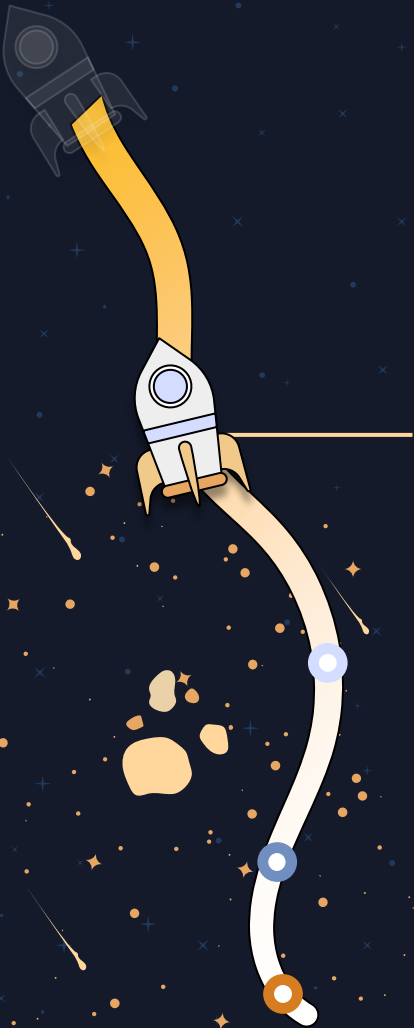
$$\hat{\mathbf{x}}_k^- = \mathbf{f}_{k-1}(\hat{\mathbf{x}}_{k-1}^+) + \mathbf{n}_k$$

$$\mathbf{P}_k^- = \mathbf{F}_{k-1} \mathbf{P}_{k-1}^+ \mathbf{F}_{k-1}^T + \mathbf{Q}_{k-1}$$

Update

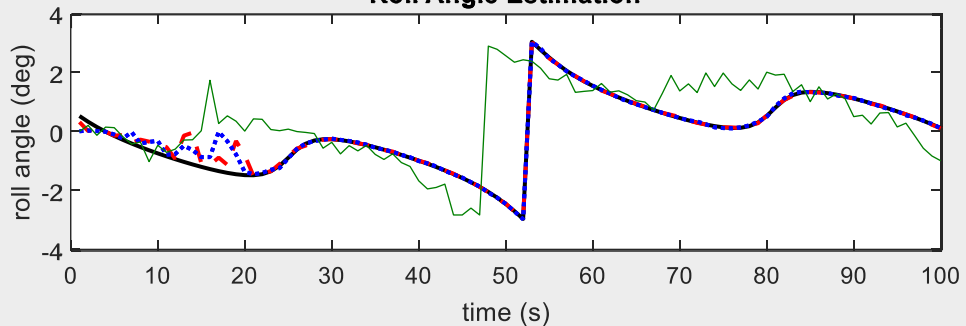
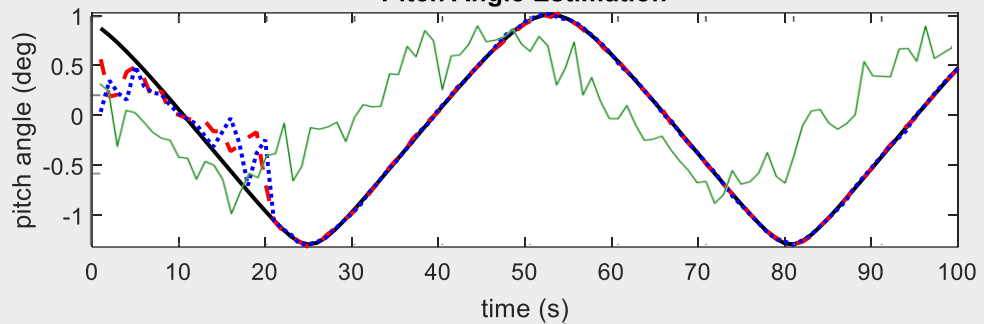
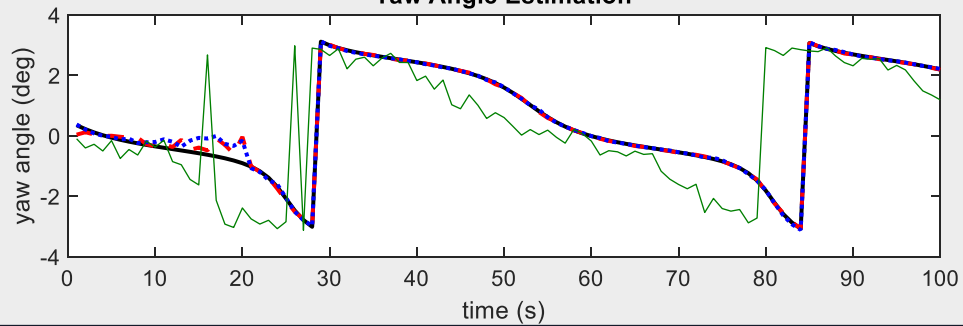
$$\hat{\mathbf{x}}_k^+ = \hat{\mathbf{x}}_k^- + \mathbf{K}(\mathbf{z}_k - \mathbf{H}_k \hat{\mathbf{x}}_k^-)$$

$$\mathbf{P}_k^+ = (\mathbf{I} - \mathbf{K} \mathbf{H}_k) \mathbf{P}_k^-$$

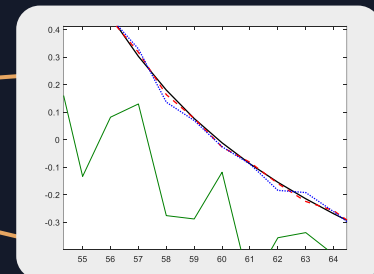
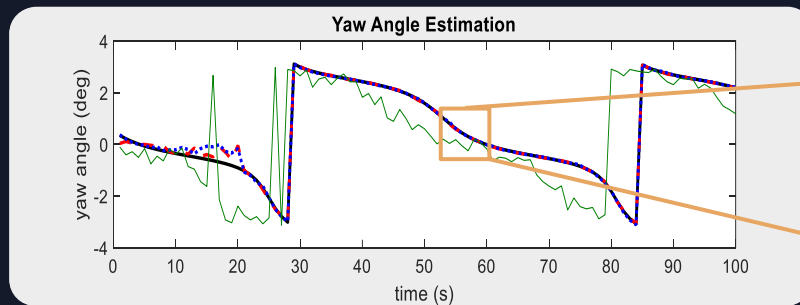
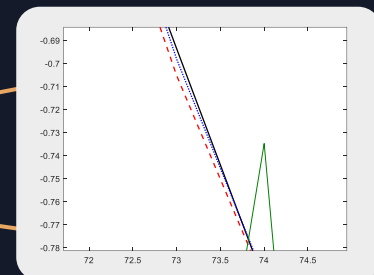
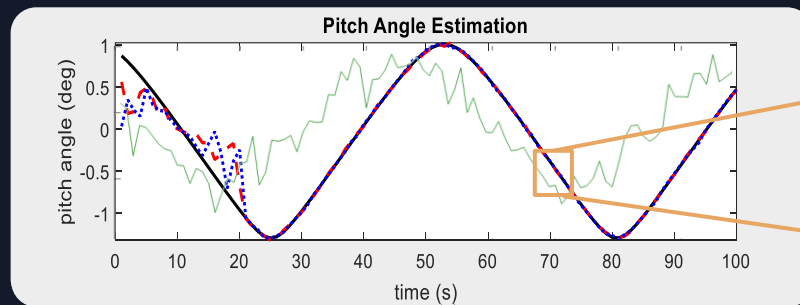
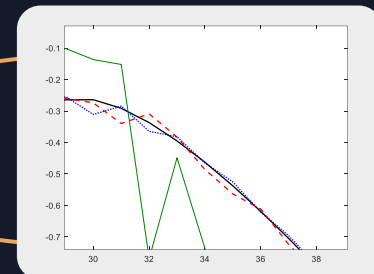
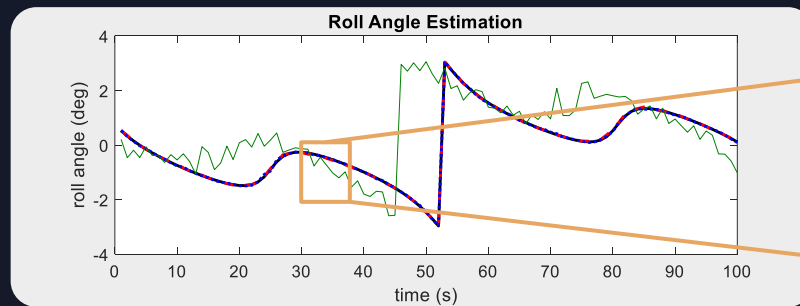


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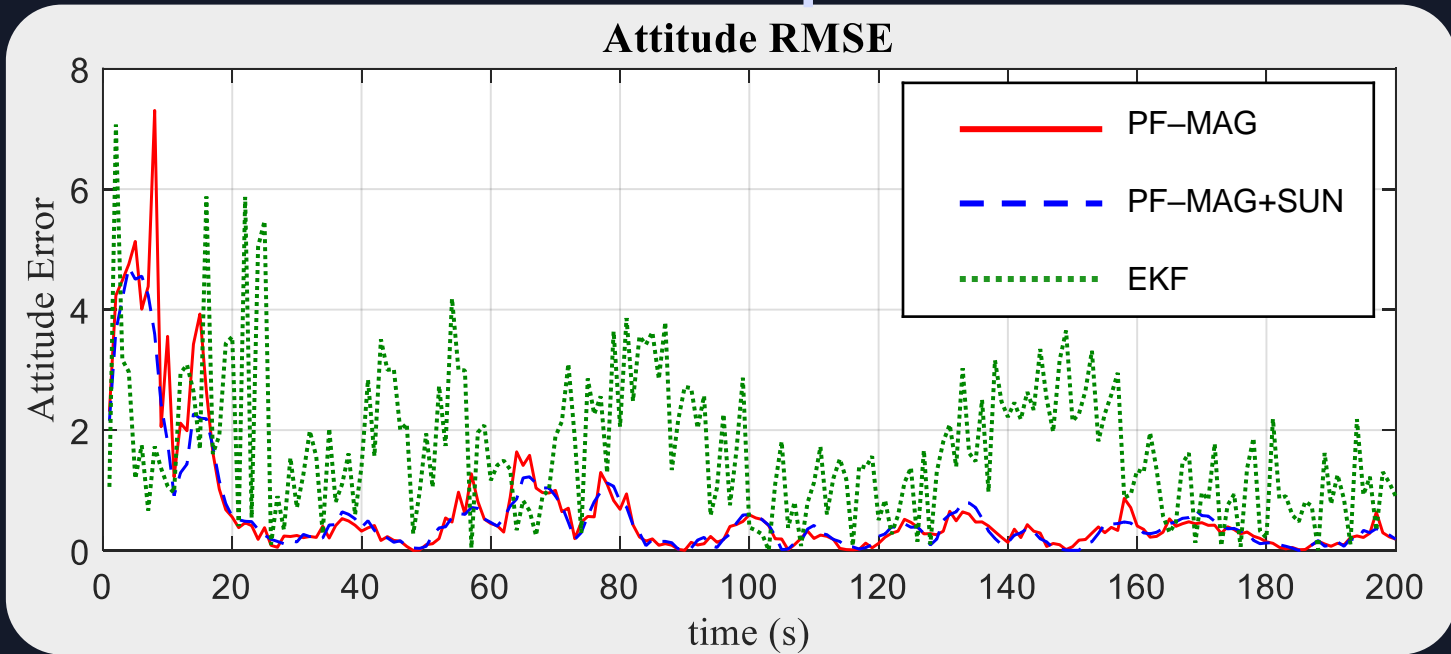
Simulation Results

Roll Angle Estimation**Pitch Angle Estimation****Yaw Angle Estimation**

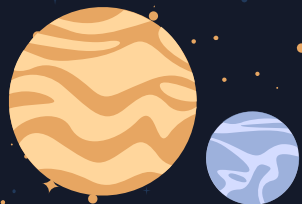
Error Analysis



Root Mean Square Error



- High accuracy attitude estimation achieved with PF ($\pm 0.01^\circ$), compared to EKF ($\pm 1^\circ$)
- Improvements in PF estimation with 2 sensors



Computation Burden



EKF

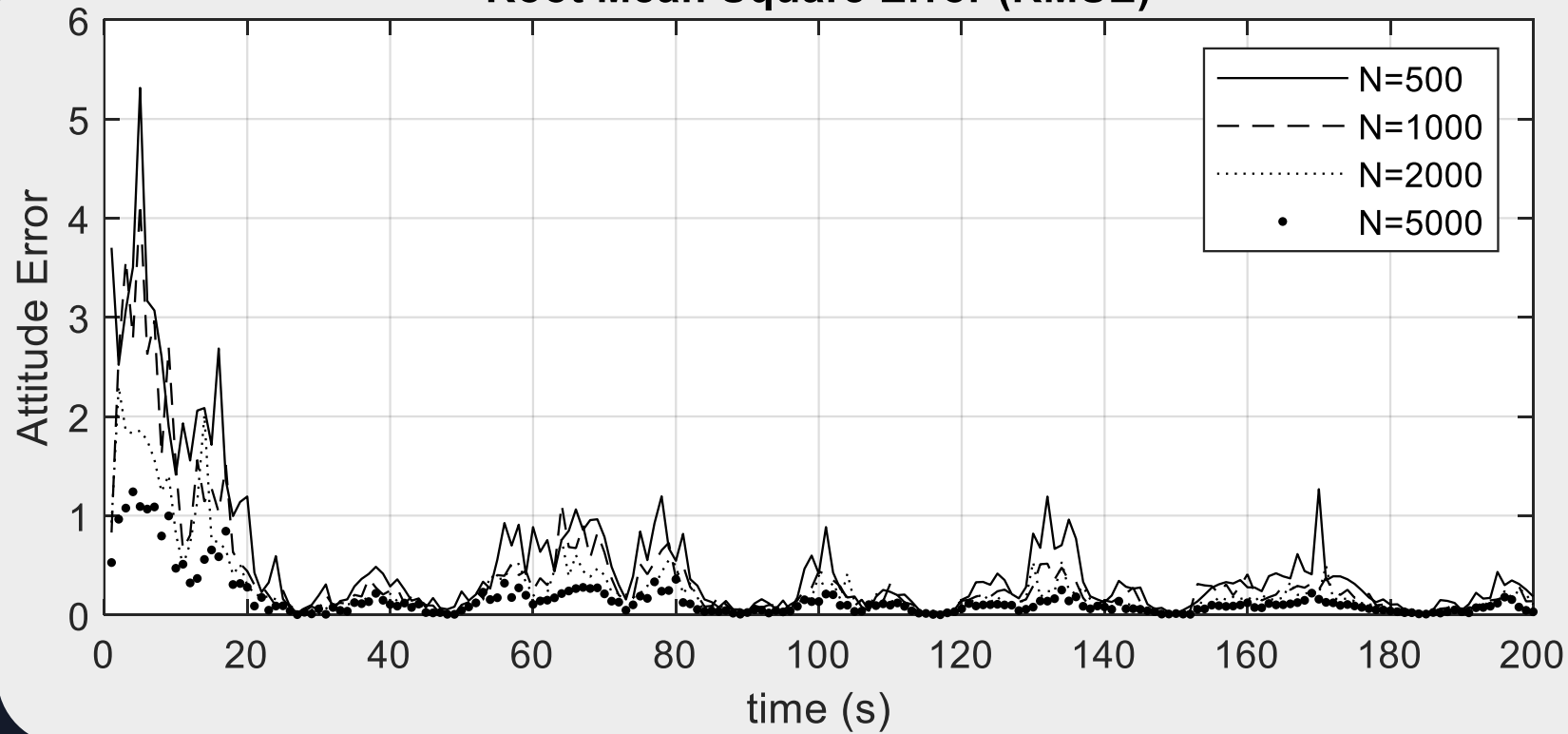


PF - (MAG)



PF - (MAG+SUN)

Root Mean Square Error (RMSE)

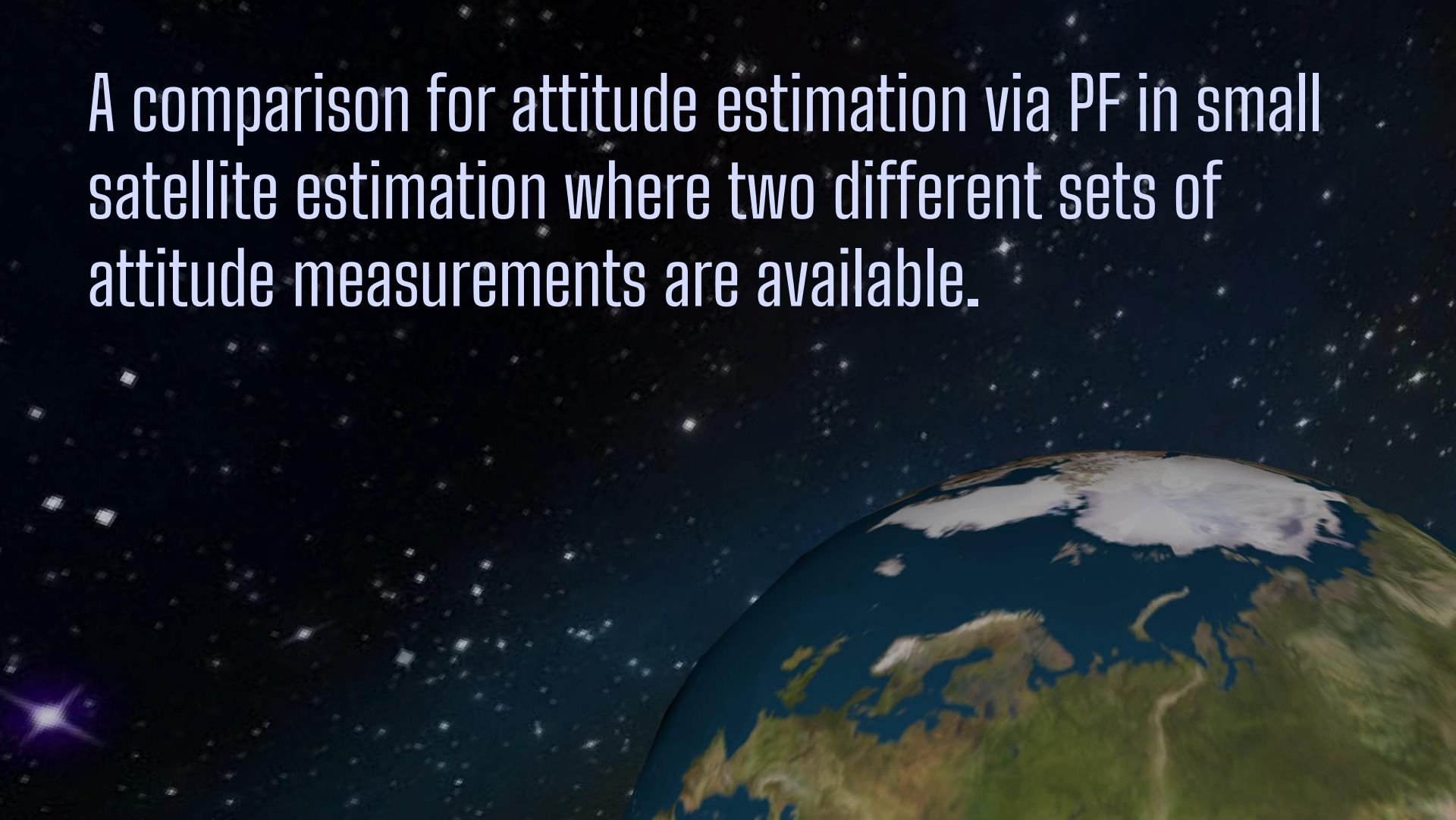


5

Conclusions

- High accuracy attitude estimation achieved with PF ($\pm 0.01^\circ$), compared to EKF ($\pm 1^\circ$).
- Two different attitude sensor configurations for PF were simulated.
- The inclusion of a sun sensor improved the PF attitude estimation, but the computational burden was higher.

A comparison for attitude estimation via PF in small satellite estimation where two different sets of attitude measurements are available.



Thank You!



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