Assessment of GPS-based accelerometry performance with adaptive filter settings

Jose van den IJssel, Christian Siemes, Pieter Visser Delft University of Technology



Swarm 10 Year Anniversary and Science Conference, 8-12 April 2024

## **Swarm GPS-derived densities**





Density signal varies over the mission with clear impact of solar cycle and altitude. GPS-based accelerometry settings are fixed and optimized for early mission.

# GPS-accelerometry performance assessment using GRACE data

- The GRACE satellites deliver high-accuracy accelerometer data → allows accurate comparison with GPS-derived results
- The GRACE satellites have experienced a large range of density signals from 10<sup>-13</sup> kg/m<sup>3</sup> to 10<sup>-11</sup> kg/m<sup>3</sup> → allows tuning of filter settings for various conditions









Selection taking into account: temperature control / data gaps / maneuvers

#### Implementation GRACE GPS-based accelerometry chain

Implementation based on Swarm L2 GPS-based accelerometry chain

- Extended Kalman filter approach with DLR GHOST s/w using 26hr arcs
- Undifferenced ionosphere-free GPS observations
- Final CODE ephemeris and 5s clock product
- Attitude information from star tracker data
- GRACE PCV map provided by Astronomical Institute of University of Bern
- Estimation of float carrier phase ambiguities
- State-of-the-art gravitational modelling
- SRP, Earth albedo and infrared modeling based on 8-plate macro model with surface properties for absorbed, diffusely or specular reflected radiation
- Instantaneous reemission thermal modeling for non-solar panel surfaces
- Smooth empirical accelerations recover the aerodynamic signal



## Tuning of emprical accelerations $\rightarrow$ adaptive $\sigma_P$

Empirical accelerations that absorb the aerodynamic signal are defined by a steady-state variance  $\sigma_a$ , process noise  $\sigma_P$  and correlation time  $\tau$ .

In radial and cross-track direction small aerodynamic signal  $\rightarrow \sigma_P = 5 \text{ nm/s}^2$ . In along-track direction large variation in aerodynamic signal  $\rightarrow$  adaptive  $\sigma_P$ . Adaptive  $\sigma_P$  is based on variance of NRLMSISE-00 model accelerations.



### Quality measures

Comparison GPS-derived aerodynamic accelerations with accelerometer data

- Remove RP signal from accelerometer data using the same panel model
- Rotate GPS-derived accelerations from orbit to satellite body frame

#### **Quality measures**

- RMS differences (accelerometer GPS)
- Relative recovery error (RMS<sub>diffs</sub>/RMS<sub>acc</sub>)
- Correlation
- PSD



## Impact of different process noise settings

elft



### Performance of low/medium/high signal cases





**ŤU**Delft

## Performance of low/medium/high signal cases

## Low signal example day

**ŤU**Delft



#### Medium signal example day



## High signal example day

**TU**Delft



#### Geomagnetic storm example day

**ŤU**Delft



### **Conclusions and outlook**

#### Conclusions

- Adaptive process noise improves the recovery of aerodynamic accelerations
- GPS-based accelerations show better agreement with accelerometer data than NRLMSISE-00 model accelerations
- Best performance with strong signal (0.995 correlation, 5.4% recovery error)
- Limited performance with low signal (0.867 correlation, 23.3% recovery error)

#### Outlook

- Apply adaptive process noise to Swarm GPS-based density retrieval
- Further improve SRP modelling (Poster 56 Natalia Hladczuk)
- Further improve GSI modelling (Poster 54 Sabin Anton)
- Add uncertainty information to our densities (Siemes et al., Uncertainty of thermosphere mass density observations derived from accelerometer and GNSS tracking data, ASR, 2023)
- Apply GPS-accelerometry to other missions: NanoMagSat, GDC...



