

An Efficient Algorithm for Refining Position and Velocity Outputs of Space borne GNSS Receivers

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Outline

- 1 Motivation
 - Background And Basic Problems
 - Previous Work
- 2 Proposed Algorithm
 - General Idea
 - Algorithm Details
 - Parameter Choose
- 3 Simulation Results
 - Correctness Check

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Background

- GNSS is widely used in various areas.
 - Precise and affordable
 - Orbit position and velocity sensor
- Four tracked navigational space vehicles in sight are needed to locate a GNSS receiver.

Basic Problems

- Lack of trackable GNSS space vehicles occasionally.
 - Swarm C, 530km, 2.6%
 - The number of trackable GNSS space vehicles will decrease as orbit altitude increases.
- Other undesirable factors.
 - Limited acquisition sensitivity
 - Erroneous bit synchronization
 - Space radiation effects

Previous Work

Short-arc Filtering Based on Runge-Kutta Algorithm

- Advantages:
 - High precision
 - Low short-term prediction error
- Disadvantages:
 - High computation complexity
 - Not robust enough

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General Idea

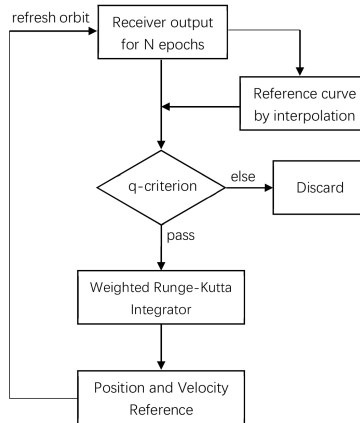


Figure: Flowchart of the proposed algorithm

Algorithm Details

● Exclude outliers

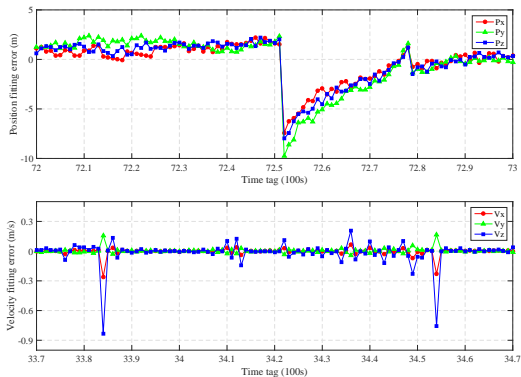


Figure: Third-order fitting error of position and velocity

Algorithm Details

- Gill's formula of the Runge-Kutta method

$$\mathbf{X}(t_0 + h) = \mathbf{X}_0 + \frac{1}{6}(k_1 + (2 - \sqrt{2})k_2 + (2 + \sqrt{2})k_3 + k_4) \quad (1)$$

where

$$\begin{cases} k_1 = hf(t_0, \mathbf{X}_0) \\ k_2 = hf(t_0 + \frac{h}{2}, \mathbf{X}_0 + \frac{k_1}{2}) \\ k_3 = hf(t_0 + \frac{h}{2}, \mathbf{X}_0 + \frac{\sqrt{2}-1}{2}k_1 + \frac{2-\sqrt{2}}{2}k_2) \\ k_4 = hf(t_0 + h, \mathbf{X}_0 - \frac{\sqrt{2}}{2}k_2 + \frac{2+\sqrt{2}}{2}k_3) \end{cases} \quad (2)$$

Algorithm Details

- Weighted sum

$$\mathbf{x}_{ref} = \sum_{n=1} \mathbf{x}_n / p \quad (3)$$

where

$$p = \sum_n \frac{1}{n}, \quad n \in \{n \leq N \mid q - \text{criterion pass}\} \quad (4)$$

Parameter Choose

- Better choice for epochs number N

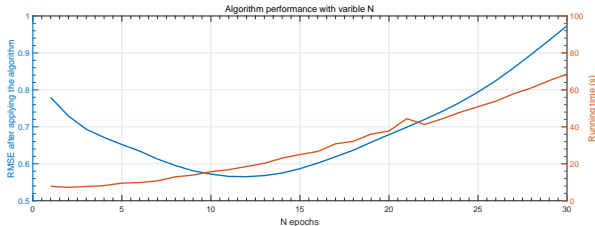


Figure: Relationship between the performance of the algorithm and N

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Correctness Check

- Correctness check on PVT data collected from GRACE B receiver

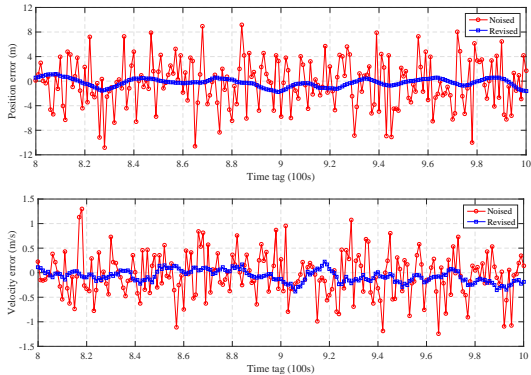


Figure: Comparison of position and velocity errors of x-axis

Simulation on LING QIAO

- Position errors of x, y and z axis before and after robustifying

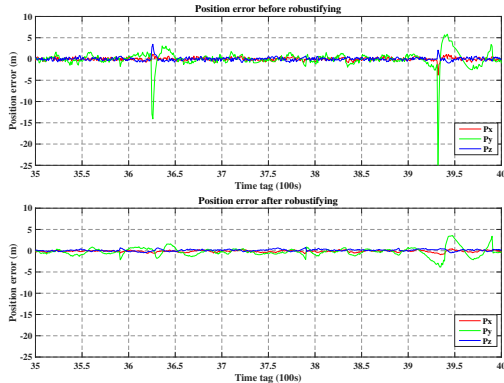


Figure: Comparison of position errors

Simulation on LING QIAO

- Velocity errors of x, y and z axis before and after robustifying

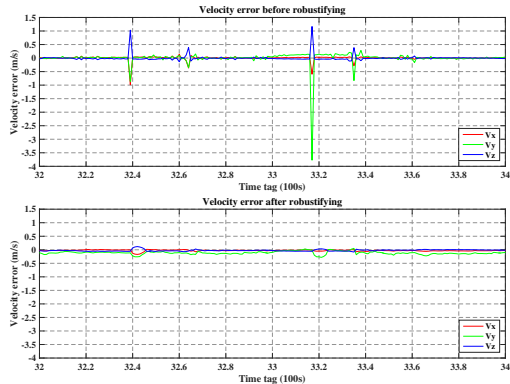


Figure: Comparison of velocity errors

Summary

- **Weighted Runge-Kutta** method guarantees accuracy.
- **Fit or interpolate** to obtain robustness.
- The algorithm effectively reduces RMSE of PVT data.
- Outlook
 - Different weighting parameters worth trying.
 - Can we **train** a group of weighting parameters?

Acknowledgement

Thank You For Your Listening!