

Intermittent GPS-aided VIO: Online Initialization and Calibration

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RPNG

Motivation: Why GPS-aided VIO?

Visual-Inertial-Odometry (VIO)

Pros	• High-quality ego-motion • Indoor/outdoor • Low-cost
Cons	• The estimation drifts • No global localization

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- Complementary sensor characteristics
- GPS-aided VIO: Low-cost, robust, drift-free, indoor/outdoor localization

Motivation: Parameters Needed

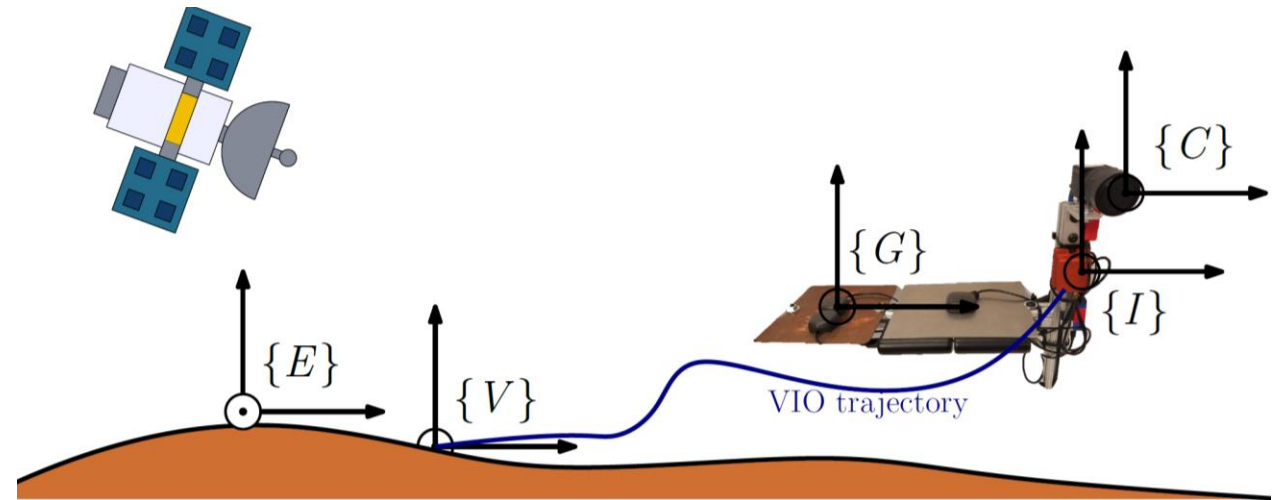
- GPS measurement $\mathbf{z}_{g_k}$ at time t_k

$$\mathbf{z}_{g_k} = {}^E\mathbf{p}_{G_k} = \underbrace{{}^E\mathbf{p}_V + {}^E\mathbf{R}^V}_{\text{Reference transformation between } \{V\} \text{ and } \{E\}} \mathbf{p}_{G_k} + \mathbf{n}_{g_k}$$

Reference transformation
between $\{V\}$ and $\{E\}$

$${}^V\mathbf{p}_{G_k} = \underbrace{{}^V\mathbf{p}_{I_k}}_{\text{IMU } \{I\} \text{ pose at } t_k \text{ in } \{V\}} + \underbrace{{}^V\mathbf{R}^I}_{\text{Extrinsic in } \{V\}} \mathbf{p}_G$$

$\{E\}$: ENU (East-North-Up) frame,
GPS reference frame
 $\{V\}$: VIO reference frame
 $\{G\}$: GPS sensor frame
 $\{I\}$: IMU sensor frame
 ${}^b\mathbf{p}_a$: Position of a in b frame
 ${}^b_a\mathbf{R}$: a to b frame rotation matrix

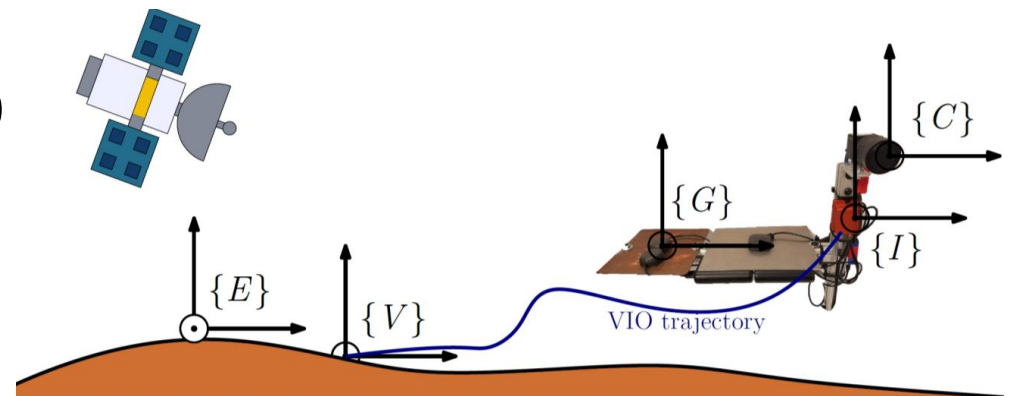


Challenges

- $\{ {}^E_V \mathbf{R}, {}^E \mathbf{p}_V \}$
 - No initial guess -> Need initialization
 - May not have GPS measurement at the beginning (e.g. start indoor)
 - Noisy GPS measurement -> Need robust initialization

- $\{ {}^I_k {}^V \mathbf{R}^\top, {}^V \mathbf{p}_{I_k} \} \{ {}^I \mathbf{p}_G \}$
 - Asynchronicity between GPS and VIO
 - Intermittent GPS measurement
 - Time offset calibration?

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Observability Analysis

- Question: Does using GPS make the state fully observable?
- State to analyze

$${}^V\mathbf{x} = \begin{bmatrix} {}^{I_k}\bar{q} & {}^V\mathbf{p}_{I_k} & {}^E_V\bar{q} & {}^E\mathbf{p}_V & {}^I\mathbf{p}_G \end{bmatrix} \quad {}^b_a\bar{q} : \text{Quaternion form of } {}^b_a\mathbf{R}$$

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- Found **4 unobservable directions** corresponding to:
 - Rotation along the axis of gravity between $\{V\}$ and $\{E\}$
 - Translation between $\{V\}$ and $\{E\}$

The system still retains 4 unobservable direction inherited from VIO system, due to estimation of the frame transformation

Proposed Fully Observable System

- Transform the state and its covariance to the ENU frame $\{E\}$

$$\mathbf{x} = g({}^V\mathbf{x}, \frac{E}{V}\bar{q}, {}^E\mathbf{p}_V) = \begin{bmatrix} I_k \bar{q} & {}^E\mathbf{p}_{I_k} & \frac{E}{V}\bar{q} & {}^E\mathbf{p}_V & I \mathbf{p}_G \end{bmatrix}$$

$$\tilde{\mathbf{x}} = \Psi^V \tilde{\mathbf{x}}, \quad \mathbf{P}_{\oplus} = \Psi \mathbf{P}_{\ominus} \Psi^{\top}$$

- Marginalize $\{\frac{E}{V}\bar{q}, {}^E\mathbf{p}_V\}$ term and get the following state

$$\text{marg}(\mathbf{x}) \rightarrow {}^E\mathbf{x} = \begin{bmatrix} I_k \bar{q} & {}^E\mathbf{p}_{I_k} & I \mathbf{p}_G \end{bmatrix}, \quad \text{marg}(\mathbf{P}_{\oplus}) \rightarrow {}^E\mathbf{P}$$

The system is now fully observable

GPS-VIO Initialization

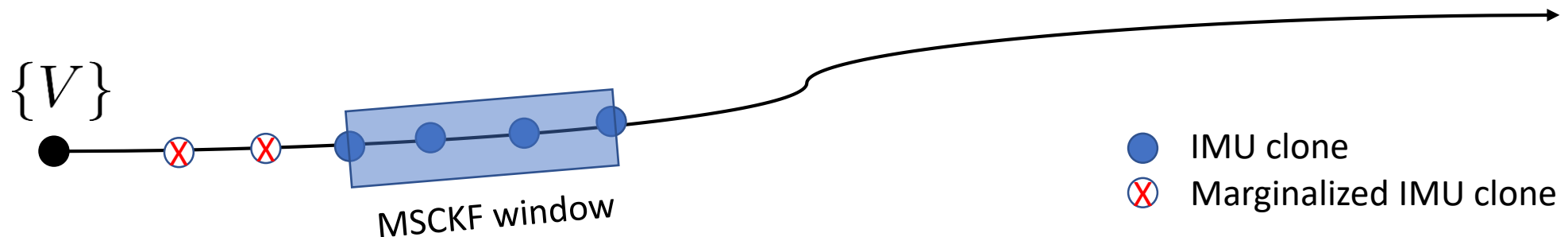
1) VIO can be independently initialized without GPS measurement

- Multi-State Constraint Kalman filter (MSCKF) state $\mathbf{x}_k$:

$$\mathbf{x}_k = \begin{bmatrix} \mathbf{x}_{I_k} & \mathbf{x}_{C_k} \end{bmatrix}$$

$$\mathbf{x}_{I_k} = \begin{bmatrix} I_k \bar{q} & {}^V \mathbf{p}_{I_k} & {}^V \mathbf{v}_{I_k} & \mathbf{b}_{\omega_k} & \mathbf{b}_{a_k} \end{bmatrix}$$

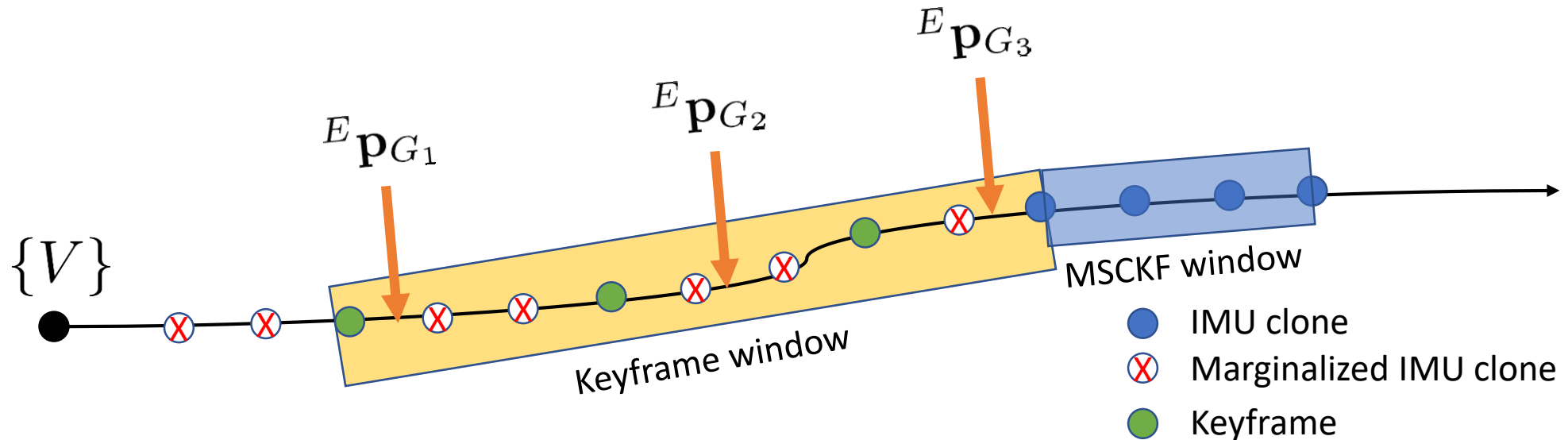
$$\mathbf{x}_{C_k} = \begin{bmatrix} I_{k-1} \bar{q} & {}^V \mathbf{p}_{I_{k-1}} & \cdots & I_{k-n} \bar{q} & {}^V \mathbf{p}_{I_{k-n}} \end{bmatrix}$$



GPS-VIO Initialization

2) Collect GPS measurements and sparse IMU keyframes

- Collect GPS measurements $\{^E \mathbf{p}_{G_1}, \dots, ^E \mathbf{p}_{G_n}\}$
- Sparse IMU key frames bound the GPS measurement $\{^V \mathbf{p}_{I_1}, \dots, ^V \mathbf{p}_{I_n}\}$



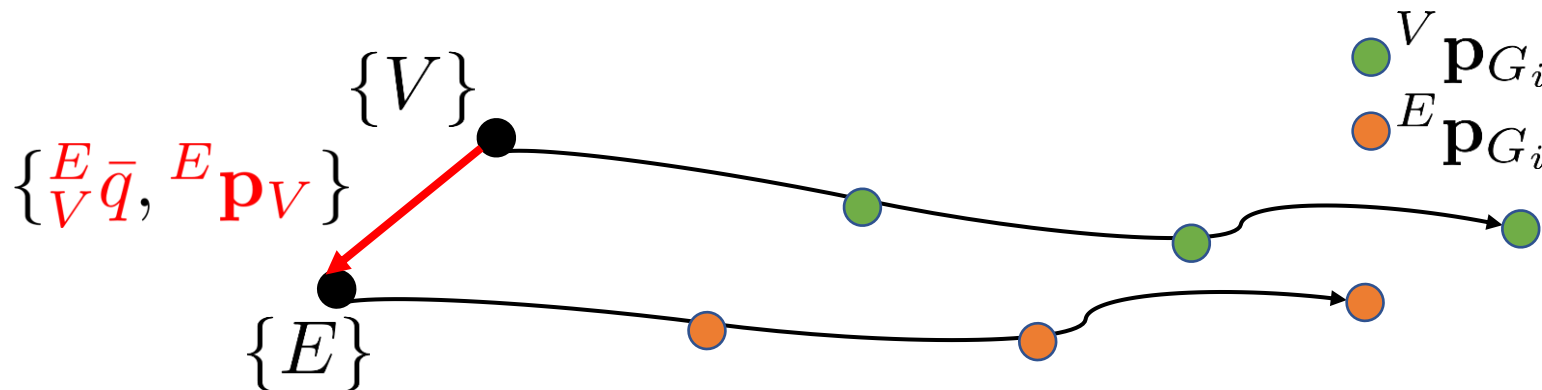
GPS-VIO Initialization

3) Compute 4 d.o.f. transformation between $\{V\}$ and $\{E\}$

- Use ${}^I\mathbf{p}_G$ and the IMU keyframes to compute GPS pose in $\{V\}$

$${}^V\mathbf{p}_{G_i} = {}^V\mathbf{p}_{I_i} + {}^I_i\mathbf{R}^\top {}^I\mathbf{p}_G \rightarrow \{{}^V\mathbf{p}_{G_1}, \dots, {}^V\mathbf{p}_{G_n}\}$$

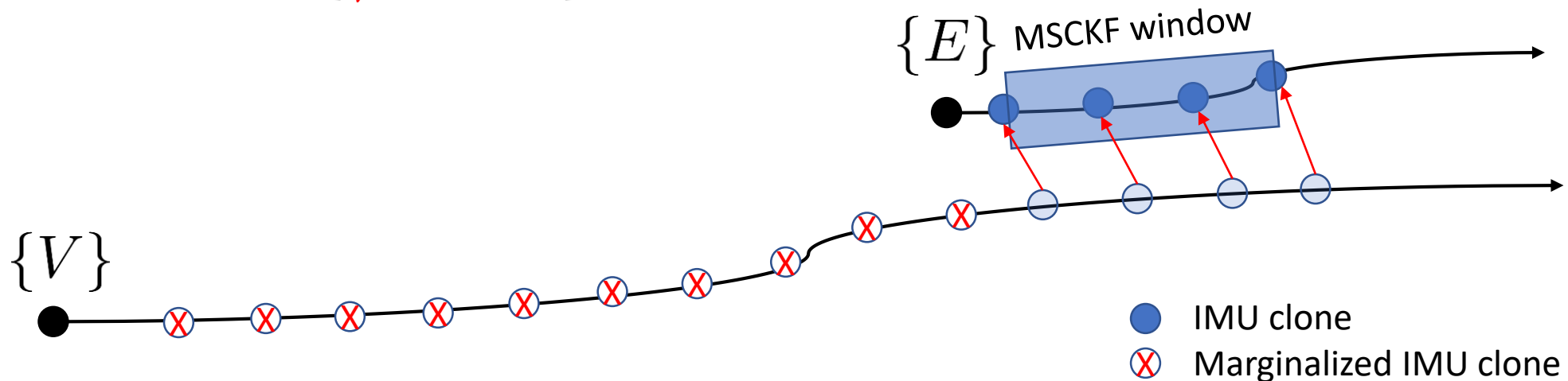
- Compute $\{{}^E_V\bar{\mathbf{q}}, {}^E\mathbf{p}_V\}$ that aligns $\{{}^E\mathbf{p}_{G_1}, \dots, {}^E\mathbf{p}_{G_n}\}$ and $\{{}^V\mathbf{p}_{G_1}, \dots, {}^V\mathbf{p}_{G_n}\}$ by solving quadratic constraint least-squares problem



GPS-VIO Initialization

4) Post process

- Perform delayed initialization to further correct $\{\frac{E}{V}\bar{q}, {}^E\mathbf{p}_V\}$
- Marginalize all the keyframes
- Transform the state to ENU $\{E\}$ reference frame
- Marginalize $\{\frac{E}{V}\bar{q}, {}^E\mathbf{p}_V\}$ from the state



GPS-VIO Initialization: Simulation Results

TABLE I: Average position and orientation errors over ten runs for different initialization distances and GPS noise values in units of meters/degree.

dist\ σ	0.1m	0.5m	1m	2m	5m
5m	1.59 / 0.65	7.08 / 3.20	14.32 / 6.56	29.37 / 69.84	69.17 / 92.37
10m	1.39 / 0.52	5.23 / 2.23	10.22 / 4.39	19.80 / 47.79	45.02 / 91.85
20m	0.90 / 0.29	2.68 / 1.08	5.02 / 2.07	9.78 / 4.08	25.49 / 49.75
50m	0.55 / 0.08	0.77 / 0.16	1.09 / 0.30	1.88 / 0.61	4.58 / 1.49
100m	0.51 / 0.09	0.49 / 0.06	0.55 / 0.12	0.85 / 0.24	2.18 / 0.63

The larger GPS noise magnitude,
less accurate initialization

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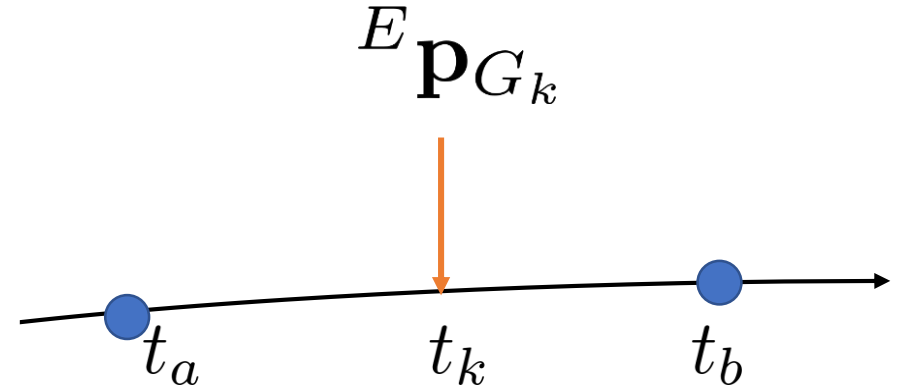
The larger GPS noise requires longer trajectory for the initialization

Initialization

GPS Measurement Update/Calibration

- GPS measurement function

$$\mathbf{z}_{g_k} := {}^E\mathbf{p}_{G_k} = {}^E\mathbf{p}_{I_k} + \frac{I_k}{E}\mathbf{R}^\top \mathbf{p}_G$$



- The measurement is asynchronous/intermittent to the VIO system:

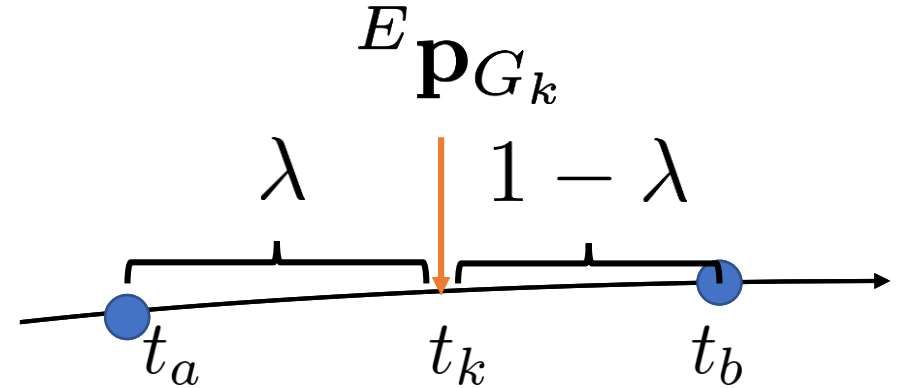
No $\{\frac{I_k}{E}\mathbf{R}, {}^E\mathbf{p}_{I_k}\}$ in the state

- GPS measurement can be delayed or sensor clocks mismatch:
Need time offset model/calibration

GPS Measurement Update/Calibration

- GPS measurement function

$$\mathbf{z}_{g_k} := {}^E\mathbf{p}_{G_k} = {}^E\mathbf{p}_{I_k} + {}^{I_k}\mathbf{R}^\top {}^I\mathbf{p}_G$$



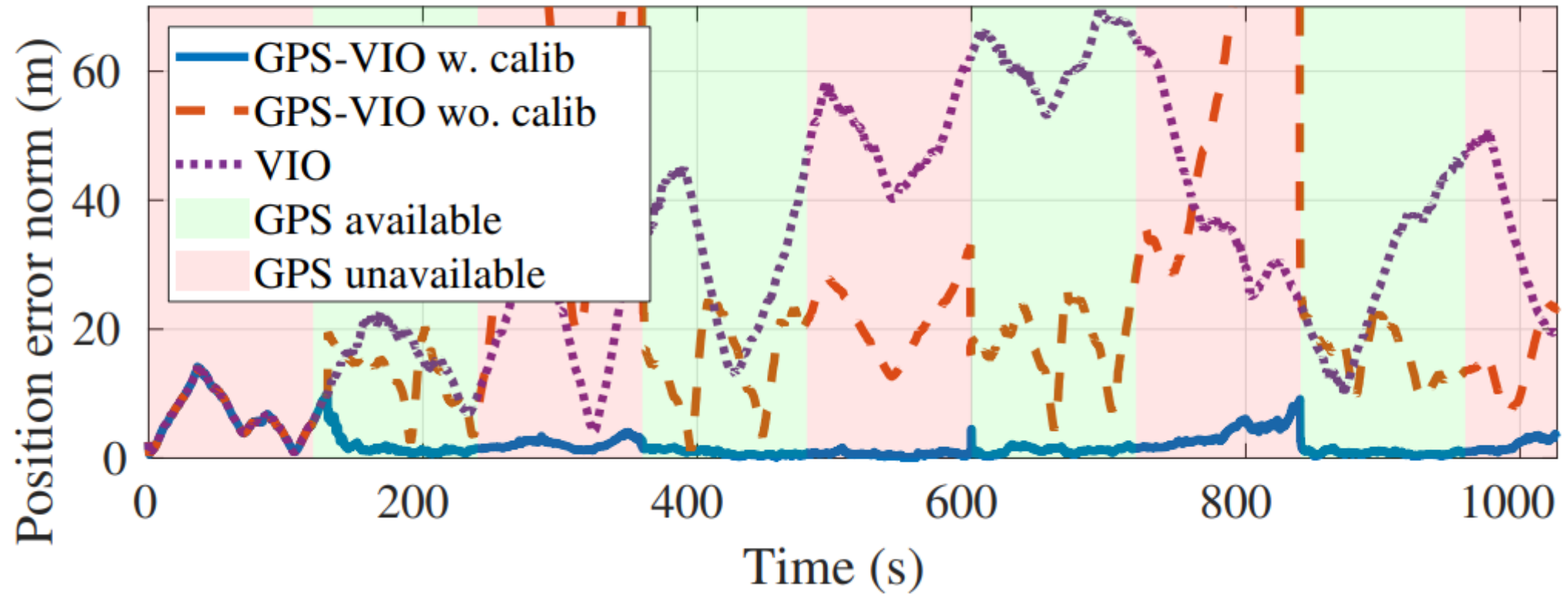
- Linear interpolation to get IMU pose at t_k & model the time offset
 - The bounding IMU poses at t_a and t_b ($t_a < t_k < t_b$)

$${}^{I_k}_E\mathbf{R} = \text{Exp} \left(\lambda \text{Log} \left({}^{I_b}_E\mathbf{R} {}^{I_a}_E\mathbf{R}^\top \right) \right) {}^{I_a}_E\mathbf{R}, \quad {}^E\mathbf{p}_{I_k} = (1 - \lambda) {}^E\mathbf{p}_{I_a} + \lambda {}^E\mathbf{p}_{I_b}$$

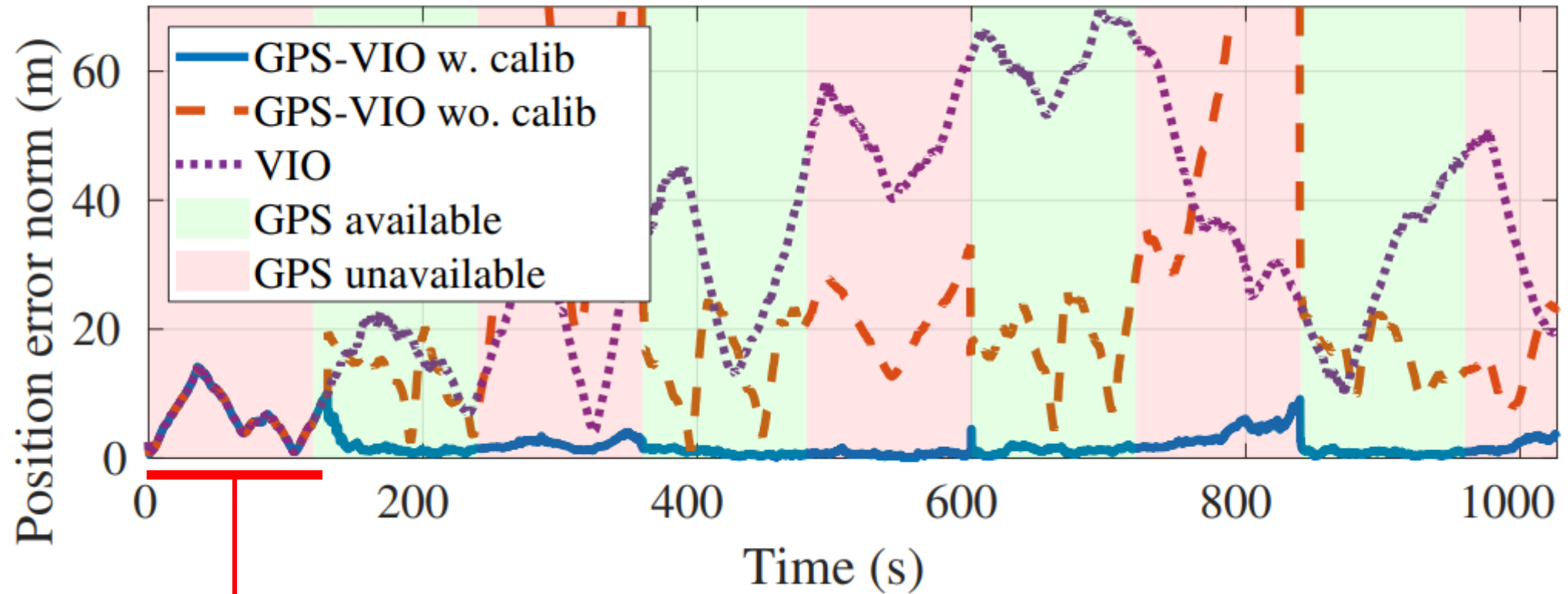
$$\lambda = \frac{(t_k + {}^I t_G - t_a)}{(t_b - t_a)}$$

Exp() and Log() are the SO(3) matrix exponential and logarithm functions

GPS-VIO Update/Cal. Simulation Results

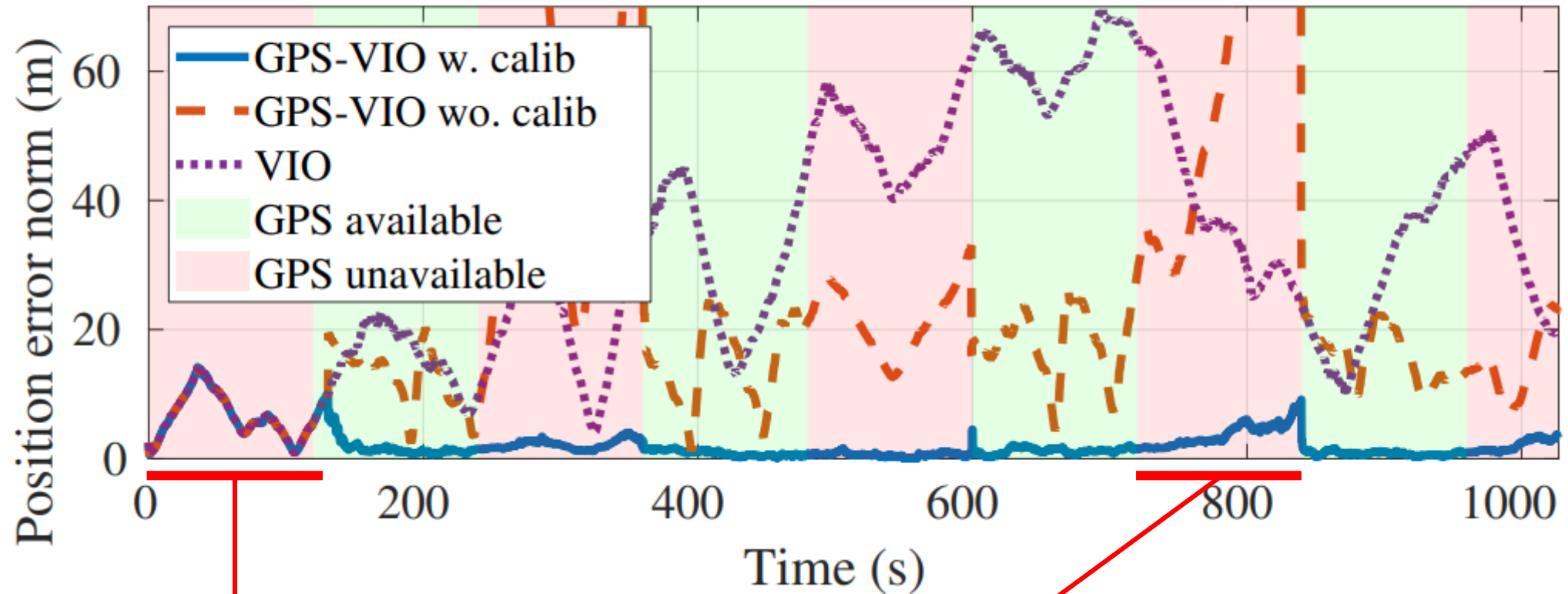


GPS-VIO Update/Cal. Simulation Results



System initialized without GPS

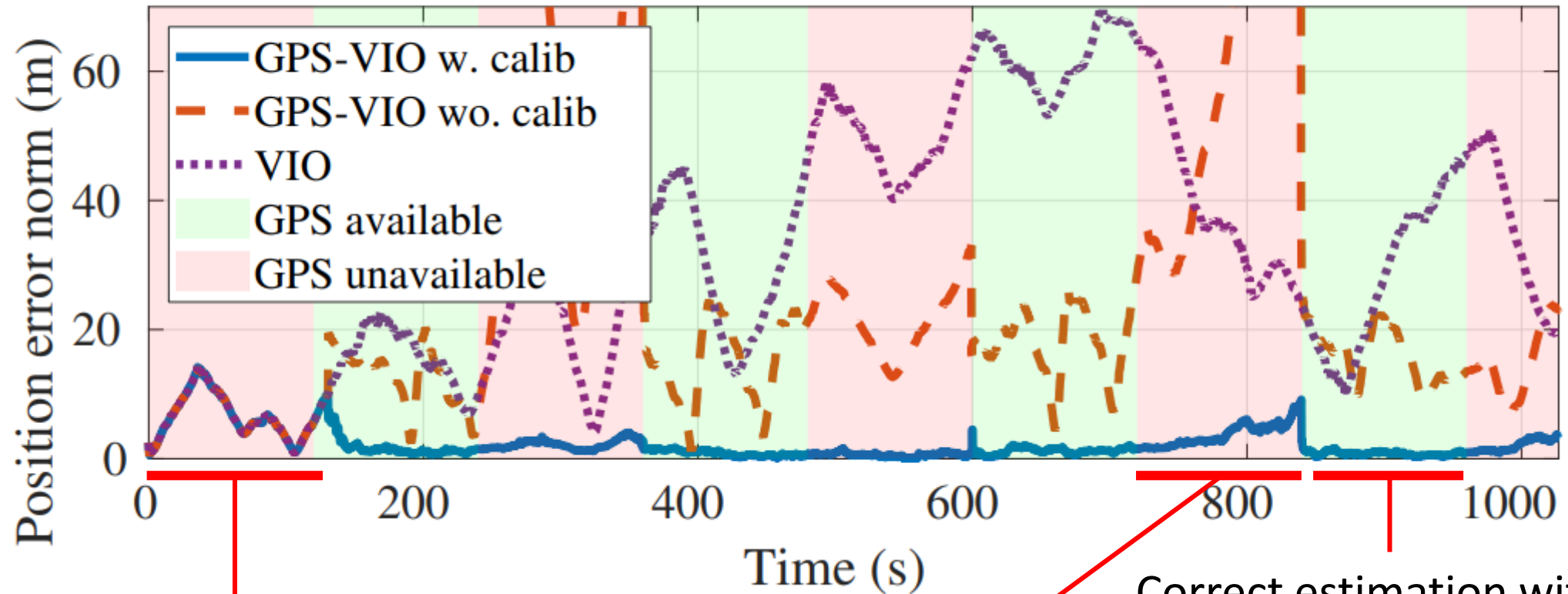
GPS-VIO Update/Cal. Simulation Results



System initialized without GPS

Perform pure VIO when no GPS

GPS-VIO Update/Cal. Simulation Results



System initialized without GPS

Perform pure VIO when no GPS

Correct estimation with GPS

GPS-VIO Update/Cal. Simulation Results

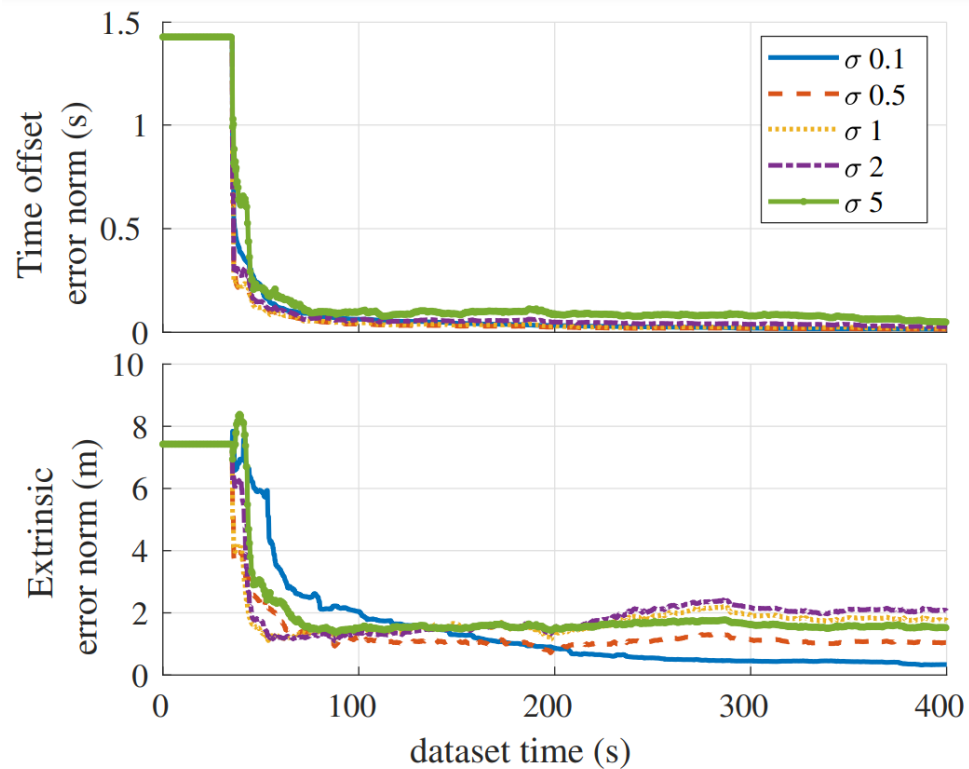
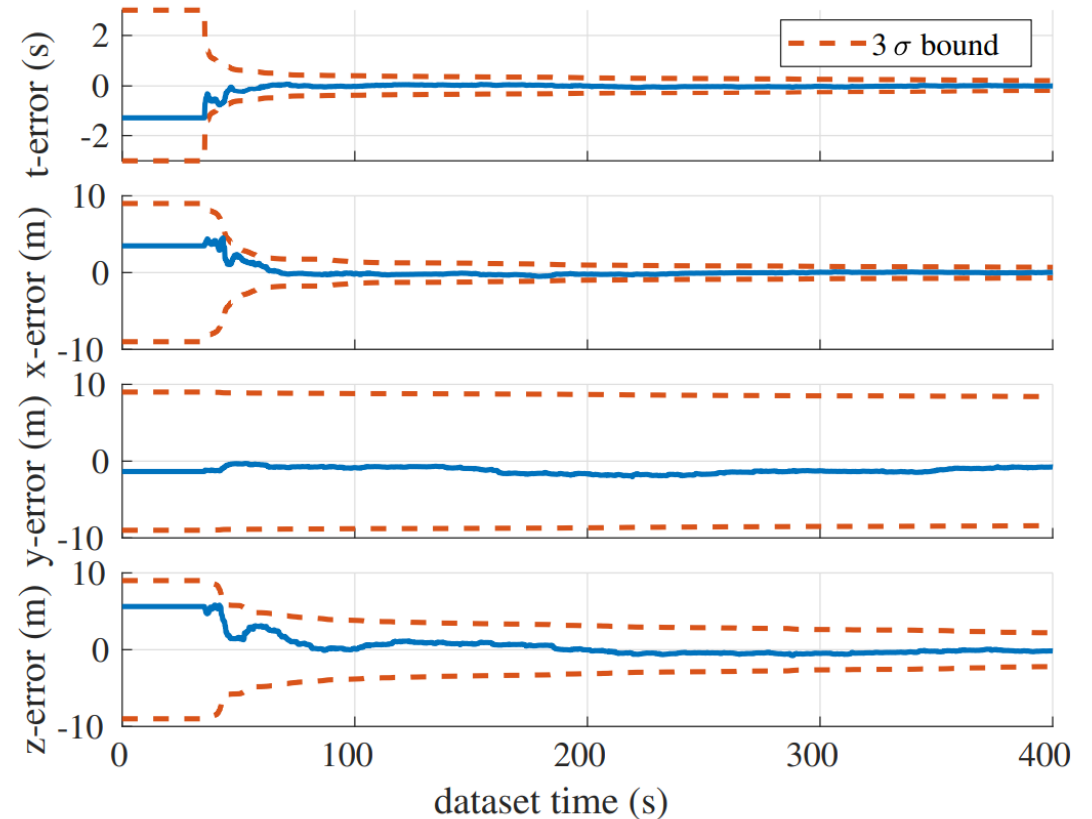


Fig. 4: The calibration errors respect to the size of GPS measurement noise.



- Can calibrate large initial errors
- Convergence affected by GPS noise magnitude

- Errors are bounded by 3 σ bound – **Consistent!**

Localization

Indoor to Outdoor

Conclusion

- Proved the state contains VIO to ENU reference frame transformation has **4 unobservable directions** by the observability analysis.
- Proposed the state transformation that removes the transformation parameters and become **fully observable**.
- Proposed the GPS-VIO initialization procedure that is **robust**, observable, and can handle intermittent GPS measurements
- Proposed asynchronous GPS measurement update method and **spatiotemporal calibration** method

Thank you

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