

Technical Note: INS State Error Model

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Abstract—Due to space limitations in [1] and [2], this Technical Note is supplied to explain why the dimension of the state vector $\mathbf{x} \in \mathbb{R}^{16}$ and the dimension of error state vector $\delta\mathbf{x} \in \mathbb{R}^{15}$. This Technical Note also describes the additive and multiplicative operations required to use $\delta\mathbf{x}$ to correct \mathbf{x} .

I. INTRODUCTION

Let $\mathbf{x} \in \mathbb{R}^{n_s}$ denote the rover state vector:

$$\mathbf{x}(t) = [\mathbf{p}^\top(t), \mathbf{v}^\top(t), \mathbf{q}^\top(t), \mathbf{b}_a^\top(t), \mathbf{b}_g^\top(t)]^\top \in \mathbb{R}^{n_s},$$

where \mathbf{p} , \mathbf{v} , \mathbf{b}_a , \mathbf{b}_g each in \mathbb{R}^3 represent the position, velocity, accelerometer bias and gyro bias vectors, respectively, and $\mathbf{q} \in \mathbb{R}^4$ represents the attitude quaternion ($n_s = 16$), each at time t . Let $\hat{\mathbf{x}} \in \mathbb{R}^{n_s}$ denote the estimate of the rover state vector:

$$\hat{\mathbf{x}}(t) = [\hat{\mathbf{p}}^\top(t), \hat{\mathbf{v}}^\top(t), \hat{\mathbf{q}}^\top(t), \hat{\mathbf{b}}_a^\top(t), \hat{\mathbf{b}}_g^\top(t)]^\top \in \mathbb{R}^{n_s}.$$

The error between $\mathbf{x}(t)$ and $\hat{\mathbf{x}}(t)$ is denoted as $\delta\mathbf{x}$. The error vector is

$$\delta\mathbf{x} = [\delta\mathbf{p}^\top, \delta\mathbf{v}^\top, \boldsymbol{\rho}^\top, \delta\mathbf{b}_a^\top, \delta\mathbf{b}_g^\top]^\top \in \mathbb{R}^{n_e},$$

where $\delta\mathbf{p}$ and $\delta\mathbf{v}$, each in \mathbb{R}^3 , represent the error between the true and computed position and velocity, respectively. The small-angle error state, denoted as $\boldsymbol{\rho} \in \mathbb{R}^{3 \times 1}$, is defined in Section 2.5.5 of [3], and discussed in Section II-B. The errors $\delta\mathbf{b}_a$ and $\delta\mathbf{b}_g$, each in \mathbb{R}^3 , represent the accelerometer bias, and gyro bias errors, respectively. Therefore $\delta\mathbf{x} \in \mathbb{R}^{15}$ (i.e. $n_e = 15$). The fact that $n_s = 16$ and $n_e = 15$ is discussed in Section II-B.

II. STATE CORRECTION

Let $\delta\hat{\mathbf{x}}$ denote an estimate of $\delta\mathbf{x}$. The state correction to the state vector $\hat{\mathbf{x}}$ is denoted as

$$\hat{\mathbf{x}}^+ = \hat{\mathbf{x}}^- \oplus \delta\hat{\mathbf{x}}.$$

The symbol $(-)$ denotes the prior estimate, whereas $(+)$ is the updated estimate. The symbol \oplus is discussed in Sections II-A and II-B.

A. Position, Velocity, and Bias Updates

Position, velocity, accelerometer bias and gyro bias, each have corrections which are additive. The state correction step is

$$\begin{aligned}\hat{\mathbf{p}}^+ &= \hat{\mathbf{p}}^- + \delta\mathbf{p} \\ \hat{\mathbf{v}}^+ &= \hat{\mathbf{v}}^- + \delta\mathbf{v} \\ \hat{\mathbf{b}}_a^+ &= \hat{\mathbf{b}}_a^- + \delta\mathbf{b}_a \\ \hat{\mathbf{b}}_g^+ &= \hat{\mathbf{b}}_g^- + \delta\mathbf{b}_g.\end{aligned}$$

B. Attitude Update

When the attitude error is sufficiently small (see Section 2.5.5 of [3]), the attitude can be represented as a set of small-angle planar rotations $\{\rho_x, \rho_y, \rho_z\}$ about three orthogonal axes $\{x, y, z\}$, thus the attitude error can be defined in \mathbb{R}^3 .

1) *Rotation Matrix*: Let $\mathbf{R}_b^n \in \mathbb{R}^{3 \times 3}$ represent the true rotation from body-frame (b) to navigation-frame (n) that is equivalent to $\mathbf{q}(t)$ (see eqn. D.13 in [3]). Let $\hat{\mathbf{R}}_b^n \in \mathbb{R}^{3 \times 3}$ represent the computed rotation that is equivalent to $\hat{\mathbf{q}}(t)$. The error between the true and computed rotation is

$$\mathbf{R}_n^n = (\mathbf{R}_b^n)(\hat{\mathbf{R}}_b^n),$$

where \mathbf{R}_n^n represents the rotation matrix from the computed to actual navigation frame. When the error between the true and computed rotation is zero, then $\mathbf{R}_n^n = \mathbf{I}$. Otherwise, as discussed in Section 2.6.1 of [3],

$$\mathbf{R}_n^n = [\mathbf{I} - \mathbf{P}]$$

where $\mathbf{P} = [\boldsymbol{\rho} \times]$, and $\boldsymbol{\rho} = [\rho_x, \rho_y, \rho_z]^\top \in \mathbb{R}^3$ (see eqn. 10.28 of [3]).

Using this notation, the attitude update (as defined in eqn. 10.29 of [3]) is

$$(\mathbf{R}_b^n)^+ = [\mathbf{I} - \mathbf{P}](\hat{\mathbf{R}}_b^n)^-.$$

Note that the attitude correction is multiplicative.

2) *Quaternion*: A similar approach to Section II-B.1 is valid when the attitude error is represented by a quaternion. Let \mathbf{q}_b^n represent the true quaternion from b -frame to n -frame. Let $\hat{\mathbf{q}}_b^n$ represent the computed quaternion. The error may be represented as

$$\mathbf{q}_n^n = \mathbf{q}_b^n \otimes \hat{\mathbf{q}}_b^n$$

where \mathbf{q}_n^n represents the quaternion from the computed to actual navigation frame. The symbol \otimes represents the quaternion multiplication operation defined in Section D of [3]. When the error between the true and computed rotation is zero, then $\mathbf{q}_n^n = [1, 0, 0, 0]^\top$, otherwise \mathbf{q}_n^n may be represented as

$$\mathbf{q}_n^n = \begin{bmatrix} \hat{\mathbf{q}}_s \\ \hat{\mathbf{q}}_v \end{bmatrix} = \begin{bmatrix} \sqrt{1 - \|\frac{1}{2}\boldsymbol{\rho}\|_2^2} \\ \frac{1}{2}\boldsymbol{\rho} \end{bmatrix} \approx \begin{bmatrix} 1 \\ \frac{1}{2}\boldsymbol{\rho} \end{bmatrix}, \quad (1)$$

where the scalar part of the quaternion is $\hat{\mathbf{q}}_s = 1$, and the vector part is $\hat{\mathbf{q}}_v = \boldsymbol{\rho}$. The approximation on the right-hand side of eqn. (1) is shown in Appendix I.

Using this notation, the multiplicative quaternion update is

$$\hat{\mathbf{q}}_b^n^+ = \mathbf{q}_b^n \otimes \hat{\mathbf{q}}_b^n^-.$$

Quaternion operations are defined in Section D of [3].

APPENDIX I
QUATERNION UPDATE APPROXIMATION

Let $\mathbf{f}(\boldsymbol{\rho}) = \sqrt{1 - \|\frac{1}{2}\boldsymbol{\rho}\|_2^2} \in \mathbb{R}^1$, and $\delta\boldsymbol{\rho} = \boldsymbol{\rho} - \mathbf{0} \in \mathbb{R}^{3 \times 1}$. By first-order Taylor series expansion of $\mathbf{f}(\boldsymbol{\rho})$, assuming small-angle $\boldsymbol{\rho}$, the quantity \mathbf{q}_n^n is

$$\begin{aligned} \mathbf{q}_n^n &= \begin{bmatrix} \mathbf{f}(\boldsymbol{\rho}) \\ \frac{1}{2}\boldsymbol{\rho} \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{f}(\mathbf{0}) + \left. \frac{\partial \mathbf{f}(\boldsymbol{\rho})}{\partial \boldsymbol{\rho}} \right|_{\boldsymbol{\rho}=\mathbf{0}} \delta\boldsymbol{\rho} + \delta\boldsymbol{\rho}^\top \left. \frac{\partial^2 \mathbf{f}(\boldsymbol{\rho})}{2 \partial \boldsymbol{\rho}^2} \right|_{\boldsymbol{\rho}=\mathbf{0}} \delta\boldsymbol{\rho} + \dots \\ \frac{1}{2}\boldsymbol{\rho} \end{bmatrix} \\ &= \begin{bmatrix} 1 + \left[\frac{-\boldsymbol{\rho}}{2\sqrt{1 - \|\frac{1}{2}\boldsymbol{\rho}\|_2^2}} \right] \bigg|_{\boldsymbol{\rho}=\mathbf{0}} \delta\boldsymbol{\rho} \\ \frac{1}{2}\boldsymbol{\rho} \end{bmatrix} \\ &= \begin{bmatrix} 1 \\ \frac{1}{2}\boldsymbol{\rho} \end{bmatrix}, \end{aligned}$$

where the derivation of the gradient and Hessian is provided in Appendix II. Note:

- $\mathbf{f}(\mathbf{0}) = \sqrt{1 - \|\frac{1}{2}\mathbf{0}\|_2^2} = 1$.
- For $\boldsymbol{\rho} \ll 1$, then $\delta\boldsymbol{\rho}^\top \left. \frac{\partial^2 \mathbf{f}(\boldsymbol{\rho})}{2 \partial \boldsymbol{\rho}^2} \right|_{\boldsymbol{\rho}=\mathbf{0}} \delta\boldsymbol{\rho} \approx 0$.
- $\hat{\mathbf{q}}_v$ is linear already.

APPENDIX II
GRADIENT AND HESSIAN DERIVATION

Let $\mathbf{h}(\mathbf{x}) = (\mathbf{x}^\top \mathbf{x})^{1/2}$ where $\mathbf{x} \in \mathbb{R}^{3 \times 1}$. The Jacobian of $\mathbf{h}(\mathbf{x})$ is

$$\frac{\partial \mathbf{h}(\mathbf{x})}{\partial \mathbf{x}} = \frac{\mathbf{x}^\top}{(\mathbf{x}^\top \mathbf{x})^{1/2}}.$$

The Hessian of $\mathbf{h}(\mathbf{x})$ is

$$\begin{aligned} \frac{\partial^2 \mathbf{h}(\mathbf{x})}{\partial \mathbf{x}^2} &= \frac{\mathbf{I}}{(\mathbf{x}^\top \mathbf{x})^{1/2}} - \left(\frac{1}{2} \right) (2) \frac{\mathbf{x} \mathbf{x}^\top}{(\mathbf{x}^\top \mathbf{x})^{3/2}} \\ &= \frac{\mathbf{x}^\top \mathbf{x} \mathbf{I} - \mathbf{x} \mathbf{x}^\top}{(\mathbf{x}^\top \mathbf{x})^{3/2}}. \end{aligned}$$

REFERENCES

- [1] P. F. Roysdon and J. A. Farrell, "GPS-INS Outlier Detection and Elimination using a Sliding Window Filter," *American Control Conference, In Press.*, 2017.
- [2] —, "Robust GPS-INS Outlier Accommodation using a Sliding Window Filter," *22th IFAC World Congress*, 2017.
- [3] J. A. Farrell, *Aided Navigation: GPS with High Rate Sensors*. McGraw Hill, 2008.