

## I. SUPPLEMENTAL MATERIAL

We only provide the formulation here. If you are interested in the detailed derivations, we recommend You can find all the relevant formula derivations in [1] and [2].

### A. Kinematics Function $\mathbf{f}()$

Suppose the IMU sample interval is  $\Delta t$ , the kinematics function is:

$$\mathbf{f}(\mathbf{x}_i, \mathbf{u}_i, \mathbf{w}_i) = \Delta t \begin{bmatrix} \boldsymbol{\omega}_{m_i} - \mathbf{b}_{\boldsymbol{\omega}_i} - \mathbf{n}_{\boldsymbol{\omega}_i} \\ {}^G \mathbf{v}_{I_i} \\ \mathbf{0}_{3 \times 1} \\ {}^G \mathbf{R}_{I_i}(\mathbf{a}_{m_i} - \mathbf{b}_{\mathbf{a}_i} - \mathbf{n}_{\mathbf{a}_i}) + {}^G \mathbf{g}_i \\ \mathbf{n}_{\mathbf{b}_{\boldsymbol{\omega}_i}} \\ \mathbf{n}_{\mathbf{b}_{\mathbf{a}_i}} \\ \mathbf{0}_{3 \times 1} \end{bmatrix}. \quad (1)$$

The system input and noise are:

$$\mathbf{u} \doteq [\boldsymbol{\omega}_m^T \quad \mathbf{a}_m^T]^T, \mathbf{w} \doteq [\mathbf{n}_{\boldsymbol{\omega}}^T \quad \mathbf{n}_{\mathbf{a}}^T \quad \mathbf{n}_{\mathbf{b}_{\boldsymbol{\omega}}}^T \quad \mathbf{n}_{\mathbf{b}_{\mathbf{a}}}^T] \quad (2)$$

### B. Formulation of $\mathbf{J}^n$

Recalling the definition of  $\mathbf{J}^n$ :

$$\mathbf{J}^n = \left( \frac{\partial (\hat{\mathbf{x}}_k^n \boxplus \delta \mathbf{x}_k^n) \boxminus \hat{\mathbf{x}}_k}{\partial \delta \mathbf{x}_k^n} \bigg|_{\delta \mathbf{x}_k^n = \mathbf{0}} \right)^{-1}. \quad (3)$$

Combining the definition of  $\delta \mathbf{x}$  and  $\mathbf{x}$ , we have:

$$\mathbf{J}^n = \begin{bmatrix} \mathbf{A}({}^G \mathbf{R}_{I_k}^n)^T & \mathbf{0}_{3 \times 15} \\ \mathbf{0}_{15 \times 3} & \mathbf{I}_{15 \times 15} \end{bmatrix} \quad (4)$$

where  $\mathbf{A}()$  is defined as:

$$\mathbf{A}(\mathbf{u}) = \mathbf{I} + \left( \frac{1 - \cos \|\mathbf{u}\|}{\|\mathbf{u}\|} \right) \frac{[\mathbf{u}]_{\times}}{\|\mathbf{u}\|} + \left( 1 - \frac{\sin(\|\mathbf{u}\|)}{\mathbf{u}} \right) \frac{[\mathbf{u}]_{\times}^2}{\|\mathbf{u}\|^2} \quad (5)$$

### C. Derivation of Jacobian Matrix

The Jacobian matrix is given by:

$$\mathbf{H}_j^n = \frac{\partial \mathbf{z}_j(\hat{\mathbf{x}}_k \boxplus \delta \mathbf{x}_k^n, {}^L \mathbf{p}_j)}{\partial \delta \mathbf{x}_k^n} \quad (6)$$

and the observation function is:

$$\mathbf{z}_j(\hat{\mathbf{x}}_k \boxplus \delta \mathbf{x}_k^n, {}^L \mathbf{p}_j) = s_{j,i}^n \mathbf{D}_j^n ({}^G \hat{\mathbf{p}}_j - {}^G \boldsymbol{\mu}_i) \quad (7)$$

The calculation of the derivative of  $s_{j,i}^n \mathbf{D}_j^n$  w.r.t.  $\delta \mathbf{x}_k^n$  is quite comprehensive and impractical. We assume  $s_{j,i}^n \mathbf{D}_j^n$  as constant for each iteration and update its value after the iteration in the implementation. Let's pre-process the observation function:

$$\begin{aligned} \mathbf{z}_j(\hat{\mathbf{x}}_k \boxplus \delta \mathbf{x}_k^n, {}^L \mathbf{p}_j) &= s_j \mathbf{D}_j^n \left( {}^G \hat{\mathbf{p}}_{I_k} + {}^G \delta \mathbf{p}_{I_k} - {}^G \boldsymbol{\mu}_j \right. \\ &\quad \left. + {}^G \hat{\mathbf{R}}_{I_k} \text{Exp}({}^G \delta \mathbf{r}_{I_k}) ({}^I \mathbf{R}_{L_k} {}^L \mathbf{p}_j + {}^I \mathbf{p}_{L_k}) \right). \end{aligned} \quad (8)$$

Due to the minute transformation between two LiDAR frames, we have:

$$\text{Exp}({}^G \delta \mathbf{r}_{I_k}) \approx \mathbf{I} + [{}^G \delta \mathbf{r}_{I_k}]_{\times}. \quad (9)$$

Combining (14) and (15) leads to:

$$\begin{aligned} \mathbf{z}_j(\hat{\mathbf{x}}_k \boxplus \delta \mathbf{x}_k^n, {}^L \mathbf{p}_j) &= s_j \mathbf{D}_j^n \left( {}^G \hat{\mathbf{p}}_{I_k} + {}^G \delta \mathbf{p}_{I_k} - {}^G \boldsymbol{\mu}_j \right. \\ &\quad \left. + {}^G \hat{\mathbf{R}}_{I_k} {}^I \mathbf{q}_j - {}^G \hat{\mathbf{R}}_{I_k} [{}^I \mathbf{q}_j]_{\times} {}^G \delta \mathbf{r}_{I_k} \right), \end{aligned} \quad (10)$$

where  ${}^I \mathbf{q}_j = {}^I \mathbf{R}_{L_k} {}^L \mathbf{p}_j + {}^I \mathbf{p}_{L_k}$ . Recalling (12), it is easy to derive the Jacobian matrix  $\mathbf{H}_j^n$  as following:

$$\mathbf{H}_j^n = \begin{bmatrix} -s_j \mathbf{D}_j^n [{}^I \mathbf{q}_j]_{\times} & s_j \mathbf{D}_j^n & \mathbf{0}_{3 \times 15} \end{bmatrix} \quad (11)$$

### D. Derivation of Jacobian Matrix

The Jacobian matrix is given by:

$$\mathbf{H}_j^n = \frac{\partial \mathbf{z}_j(\hat{\mathbf{x}}_k \boxplus \delta \mathbf{x}_k^n, {}^L \mathbf{p}_j)}{\partial \delta \mathbf{x}_k^n} \quad (12)$$

and the observation function is:

$$\mathbf{z}_j(\hat{\mathbf{x}}_k \boxplus \delta \mathbf{x}_k^n, {}^L \mathbf{p}_j) = s_{j,i}^n \mathbf{D}_j^n ({}^G \hat{\mathbf{p}}_j - {}^G \boldsymbol{\mu}_i) \quad (13)$$

The calculation of the derivative of  $s_{j,i}^n \mathbf{D}_j^n$  w.r.t.  $\delta \mathbf{x}_k^n$  is quite comprehensive and impractical. We assume  $s_{j,i}^n \mathbf{D}_j^n$  as constant for each iteration and update its value after the iteration in the implementation. Let's pre-process the observation function:

$$\begin{aligned} \mathbf{z}_j(\hat{\mathbf{x}}_k \boxplus \delta \mathbf{x}_k^n, {}^L \mathbf{p}_j) &= s_j \mathbf{D}_j^n \left( {}^G \hat{\mathbf{p}}_{I_k} + {}^G \delta \mathbf{p}_{I_k} - {}^G \boldsymbol{\mu}_j \right. \\ &\quad \left. + {}^G \hat{\mathbf{R}}_{I_k} \text{Exp}({}^G \delta \mathbf{r}_{I_k}) ({}^I \mathbf{R}_{L_k} {}^L \mathbf{p}_j + {}^I \mathbf{p}_{L_k}) \right). \end{aligned} \quad (14)$$

Due to the minute transformation between two LiDAR frames, we have:

$$\text{Exp}({}^G \delta \mathbf{r}_{I_k}) \approx \mathbf{I} + [{}^G \delta \mathbf{r}_{I_k}]_{\times}. \quad (15)$$

Combining (14) and (15) leads to:

$$\begin{aligned} \mathbf{z}_j(\hat{\mathbf{x}}_k \boxplus \delta \mathbf{x}_k^n, {}^L \mathbf{p}_j) &= s_j \mathbf{D}_j^n \left( {}^G \hat{\mathbf{p}}_{I_k} + {}^G \delta \mathbf{p}_{I_k} - {}^G \boldsymbol{\mu}_j \right. \\ &\quad \left. + {}^G \hat{\mathbf{R}}_{I_k} {}^I \mathbf{q}_j - {}^G \hat{\mathbf{R}}_{I_k} [{}^I \mathbf{q}_j]_{\times} {}^G \delta \mathbf{r}_{I_k} \right), \end{aligned} \quad (16)$$

where  ${}^I \mathbf{q}_j = {}^I \mathbf{R}_{L_k} {}^L \mathbf{p}_j + {}^I \mathbf{p}_{L_k}$ . Recalling (12), it is easy to derive the Jacobian matrix  $\mathbf{H}_j^n$  as following:

$$\mathbf{H}_j^n = \begin{bmatrix} -s_j \mathbf{D}_j^n [{}^I \mathbf{q}_j]_{\times} & s_j \mathbf{D}_j^n & \mathbf{0}_{3 \times 15} \end{bmatrix} \quad (17)$$

## REFERENCES

- [1] J. Sola, "Quaternion kinematics for the error-state kalman filter," *arXiv preprint arXiv:1711.02508*, 2017.
- [2] D. He, W. Xu, and F. Zhang, "Kalman filters on differentiable manifolds," *arXiv preprint arXiv:2102.03804*, 2021.