

Imu preintegration.

Predicting navigation state by IMU

Suppose we know the navigation state of the robot at time i , as well as the IMU measurements from the i time to j time. We want predict the state of robot at time j .

$$s_j^* = \mathcal{R}(s_i, \mathcal{D}(\xi(\zeta, b))) \quad (1)$$

The navigation state combined by attitude $R(\theta)$, position p and velocity v .

$$\begin{aligned} s_i &= (R_{nb}, p_{nb}, v_{nb}) \\ s_j &= (R_{nc}, p_{nc}, v_{nc}) \end{aligned} \quad (2)$$

- n denotes navigation state frame.
- b denotes body frame in time i .
- c denotes current frame in time j .
- θ is the lie algebra of R .

The retract action \mathcal{R} which defined on navigation state takes 2 parameters: s_i and \mathcal{D} to predict s_j .

The \mathcal{D} represents the difference between s_i and s_j .

$$d(\xi, s_i) = (R_{nc}, p_{nc}, v_{nc}) \quad (3)$$

ξ represents bias corrected preintegration measurement (PIM), which take 2 parameters, the PIM ζ and IMU bias b .

The Jacobian of s_i

$$J_{s_i}^{s_j^*} = J_{s_i}^{\mathcal{R}} + J_{\mathcal{D}}^{\mathcal{R}} J_{s_i}^{\mathcal{D}} \quad (4)$$

The Jacobian of b

$$J_b^{s_j^*} = J_{\mathcal{D}}^{\mathcal{R}} J_{\xi}^{\mathcal{D}} J_b^{\xi} \quad (5)$$

Preintegration measurement (PIM)

The PIM $\zeta(R(\theta), p, v)$ integrates all the IMU measurements without considering the IMU bias and the gravity.

ω_k^b, a_k^b are the acceleration and angular velocity measured by IMU (accelerometer + gyroscope) respectively.

$$\begin{aligned} R_{k+1} &= R_k \exp(\omega_k^b \Delta t) \\ p_{k+1} &= p_k + v_k \Delta t + R_k a_k^b \frac{\Delta t^2}{2} \\ v_{k+1} &= v_k + R_k a_k^b \Delta t \end{aligned} \quad (7)$$

n : navigation frame, b : body frame.

A: Derivative of old ζ

$$\begin{aligned} A &= \frac{\partial \zeta_{k+1}}{\partial \zeta_k} \\ &= \begin{bmatrix} \frac{\partial R_{k+1}}{\partial R_k} & \frac{\partial R_{k+1}}{\partial p_k} & \frac{\partial R_{k+1}}{\partial v_k} \\ \frac{\partial p_{k+1}}{\partial R_k} & \frac{\partial p_{k+1}}{\partial p_k} & \frac{\partial p_{k+1}}{\partial v_k} \\ \frac{\partial v_{k+1}}{\partial R_k} & \frac{\partial v_{k+1}}{\partial p_k} & \frac{\partial v_{k+1}}{\partial v_k} \end{bmatrix} \\ &= \begin{bmatrix} I_{3 \times 3} - \Delta t \widehat{\omega_k^b} & 0_{3 \times 3} & 0_{3 \times 3} \\ -R_k \widehat{a_k^b} \frac{\Delta t^2}{2} & I_{3 \times 3} & I_{3 \times 3} \Delta t \\ -R_k \widehat{a_k^b} \Delta t & 0_{3 \times 3} & I_{3 \times 3} \end{bmatrix} \end{aligned} \quad (8)$$

B: Derivative of input a

$$B = \frac{\partial \zeta_{k+1}}{\partial a_k^b} = \begin{bmatrix} \frac{\partial R_{k+1}}{\partial a_k^b} \\ \frac{\partial p_{k+1}}{\partial a_k^b} \\ \frac{\partial v_{k+1}}{\partial a_k^b} \end{bmatrix} = \begin{bmatrix} 0_{3 \times 3} \\ R_k \frac{\Delta t^2}{2} \\ R_k \Delta t \end{bmatrix} \quad (9)$$

C: Derivative of input ω

$$C = \frac{\partial \zeta_{k+1}}{\partial \omega_k^b} = \begin{bmatrix} \frac{\partial R_{k+1}}{\partial \omega_k^b} \\ \frac{\partial p_{k+1}}{\partial \omega_k^b} \\ \frac{\partial v_{k+1}}{\partial \omega_k^b} \end{bmatrix} = \begin{bmatrix} I_{3 \times 3} \Delta t \\ 0_{3 \times 3} \\ 0_{3 \times 3} \end{bmatrix} \quad (10)$$

Bias correct

We want correct ζ by a given accelerometer and gyroscope bias.

$$\xi(b + \Delta b) = \zeta \oplus (\Delta b_{acc} \frac{\partial \zeta}{\partial b_{acc}} + \Delta b_{\omega} \frac{\partial \zeta}{\partial b_{\omega}}) \quad (11)$$

- b_{acc} is bias for accelerometer.
- b_{ω} is bias for gyroscope.
- Because the parameter θ cannot be added directly, we define the combination of ζ with the symbol \oplus .

$$a \oplus b = [\log(\exp(\theta_a) \exp(\theta_b)), p_a + p_b, v_a + v_b] \quad (12)$$

The jacobian of bias for corrected PIM.

$$J_b^{\xi} = [\frac{\partial \zeta}{\partial b_{acc}}, \frac{\partial \zeta}{\partial b_{\omega}}] \quad (13)$$

Find the partial derivatives of accelerometer's bias

The bias model for accelerometer.

$$\tilde{a}_k^b = a_k^b - b_{acc} \quad (14)$$

$$\begin{aligned} \frac{\partial \zeta_{k+1}}{\partial b_{acc}} &= \frac{\partial \zeta_{k+1}}{\partial \zeta_k} \frac{\partial \zeta_k}{\partial b_{acc}} + \frac{\partial \zeta_{k+1}}{\partial \tilde{a}_k^b} \frac{\partial \tilde{a}_k^b}{\partial b_{acc}} \\ &= A \frac{\partial \zeta_k}{\partial b_{acc}} - B \end{aligned} \quad (15)$$

Find the partial derivatives of gyroscope's bias

$$\tilde{\omega}_k^b = \omega_k^b - b_{\omega} \quad (16)$$

$$\begin{aligned} \frac{\partial \zeta_{k+1}}{\partial b_{\omega}} &= \frac{\partial \zeta_{k+1}}{\partial \zeta_k} \frac{\partial \zeta_k}{\partial b_{\omega}} + \frac{\partial \zeta_{k+1}}{\partial \tilde{\omega}_k^b} \frac{\partial \tilde{\omega}_k^b}{\partial b_{\omega}} \\ &= A \frac{\partial \zeta_k}{\partial b_{\omega}} - C \end{aligned} \quad (17)$$

- \sim denotes the corrected measurement.

Delta between two states

The \mathcal{D} represents the difference between two s_i and s_j .

$$\mathcal{D} = (R_{bc}, p_{bc}, v_{bc}) \quad (18)$$

We can calculate \mathcal{D} from corrected PIM $\xi(R_{bc}^\xi, p_{bc}^\xi, v_{bc}^\xi)$ and velocity, which is included in s_i .

$$\mathcal{D}(\xi, s_i) = \begin{bmatrix} R_{bc}^\xi \\ p_{bc}^\xi + R_{nb}^{-1} v_{nb} \Delta t + R_{nb}^{-1} g \frac{\Delta t^2}{2} \\ v_{bc}^\xi + R_{nb}^{-1} g \Delta t \end{bmatrix} \quad (19)$$

- g is the gravity vector.
- $*$ denotes the predicted navigation state.

The jacobian matrix of navigation state

$$\begin{aligned} J_{s_i}^{\mathcal{D}} &= \begin{bmatrix} 0_{3 \times 3} & 0_{3 \times 3} & 0_{3 \times 3} \\ \frac{\partial p_{bc}}{\partial R_{nb}} & 0_{3 \times 3} & \frac{\partial p_{bc}}{\partial v_{nb}} \\ \frac{\partial v_{bc}}{\partial R_{nb}} & 0_{3 \times 3} & 0_{3 \times 3} \end{bmatrix} \\ &= \begin{bmatrix} 0_{3 \times 3} & 0_{3 \times 3} & 0_{3 \times 3} \\ \widehat{R_{nb}^{-1} v_{nb} \Delta t + R_{nb}^{-1} g \frac{\Delta t^2}{2}} & 0_{3 \times 3} & I_{3 \times 3} \Delta t \\ \widehat{R_{nb}^{-1} g \Delta t} & 0_{3 \times 3} & 0_{3 \times 3} \end{bmatrix} \end{aligned} \quad (20)$$

The jacobian matrix of ξ

$$J_{\xi}^{\mathcal{D}} = \begin{bmatrix} I_{3 \times 3} & 0_{3 \times 3} & 0_{3 \times 3} \\ 0_{3 \times 3} & I_{3 \times 3} & 0_{3 \times 3} \\ 0_{3 \times 3} & 0_{3 \times 3} & I_{3 \times 3} \end{bmatrix} \quad (21)$$

Retraction \mathcal{R}

The retract action \mathcal{R} which defined on navigation state takes 2 parameters: s_i and \mathcal{D} to predict s_j .

- s_j^* is the predicted s_j .

$$\begin{aligned} R_{nc}^* &= R_{nb} R_{bc} \\ p_{nc}^* &= p_{nb} + R_{nb} p_{bc} \\ v_{nc}^* &= v_{nb} + R_{nb} v_{bc} \end{aligned} \quad (22)$$

Derivative of s_i

$$J_{s_i}^{\mathcal{R}} = \begin{bmatrix} R_{bc}^{-1} & 0_{3 \times 3} & 0_{3 \times 3} \\ -R_{bc}^{-1} \widehat{p_{bc}} & R_{bc}^{-1} & 0_{3 \times 3} \\ -R_{bc}^{-1} \widehat{v_{bc}} & 0_{3 \times 3} & R_{bc}^{-1} \end{bmatrix} \quad (23)$$

Derivative of d

$$J_d^{\mathcal{R}} = \begin{bmatrix} I_{3 \times 3} & 0_{3 \times 3} & 0_{3 \times 3} \\ 0_{3 \times 3} & R_{bc}^{-1} & 0_{3 \times 3} \\ 0_{3 \times 3} & 0_{3 \times 3} & R_{bc}^{-1} \end{bmatrix} \quad (24)$$

Navigation state prediction error (residual function)

If navigation $state_j$ is measured by sensors, we can calculate the error between $state_j$ and $state_j^*$.

$$r_{jj^*} = \mathcal{L}(s_j, s_j^*) = \begin{bmatrix} \Delta R \\ \Delta p \\ \Delta v \end{bmatrix} = \begin{bmatrix} R_j^{-1} R_j^* \\ R_j^{-1} (p_j^* - p_j) \\ R_j^{-1} (v_j^* - v_j) \end{bmatrix} \quad (25)$$

Local \mathcal{L} is the inverse function of \mathcal{R} , which takes two navigation states, and get the delta between the two states in tangent vector space

Derivative of an s_j

$$J_{s_j}^{\mathcal{L}} = \begin{bmatrix} -\Delta R^{-1} & 0_{3 \times 3} & 0_{3 \times 3} \\ \widehat{\Delta p} & -I_{3 \times 3} & 0_{3 \times 3} \\ \widehat{\Delta v} & 0_{3 \times 3} & -I_{3 \times 3} \end{bmatrix} \quad (26)$$

Derivative of an s_j^*

$$J_{s_j^*}^{\mathcal{L}} = \begin{bmatrix} I_{3 \times 3} & 0_{3 \times 3} & 0_{3 \times 3} \\ 0_{3 \times 3} & \Delta R & 0_{3 \times 3} \\ 0_{3 \times 3} & 0_{3 \times 3} & \Delta R \end{bmatrix} \quad (27)$$

Overall Jaccobian for prediction error

To summarize, The prediction error r takes 3 parameters s_i , s_j and b . According to the chain rule, their Jaccobian can be written in the following form.

$$J_{s_j}^r = J_{s_j}^{\mathcal{L}} \quad (28)$$

$$J_{s_i}^r = J_{s_j^*}^{\mathcal{L}} J_{s_i}^{s_j^*} = J_{s_j^*}^{\mathcal{L}} (J_{s_i}^{\mathcal{R}} + J_{\mathcal{D}}^{\mathcal{R}} J_{s_i}^{\mathcal{D}}) \quad (29)$$

$$J_b^r = J_{s_j^*}^{\mathcal{L}} J_b^{s_j^*} = J_{s_j^*}^{\mathcal{L}} J_{\mathcal{D}}^{\mathcal{R}} J_{\xi}^{\mathcal{D}} J_b^{\xi} \quad (30)$$

Appendix

A-1. Proof of [Preintegration measurement (PIM)] (8)(9)(10)

A and B are the two lie groups: $\varphi(A, B) = AB$

$$\begin{aligned} \exp(\widehat{J_A \delta}) &= (AB)^{-1} (A \exp(\hat{\delta}) B) \\ &= B^{-1} A^{-1} A \exp(\hat{\delta}) B \\ &= \exp(B^{-1} \hat{\delta} B) \\ &= \exp(\widehat{B^{-1} \delta}) \end{aligned}$$

$$\begin{aligned} \exp(\widehat{J_B \delta}) &= (AB)^{-1} (AB \exp(\hat{\delta})) \\ &= B^{-1} A^{-1} AB \exp(\hat{\delta}) \\ &= \exp(\hat{\delta}) \end{aligned}$$

Hence:

$$J_A = B^{-1} \quad (\text{A1-1})$$

$$J_B = I \quad (\text{A1-2})$$

Proof of (8):

According to A1-1:

$$\begin{aligned} \frac{\partial R_{k+1}}{\partial R_k} &= \exp(-\omega_k^b \Delta t) \\ &= I_{3 \times 3} - \Delta t \widehat{\omega_k^b} \end{aligned}$$

A is a lie group, p is a vector: $\varphi(A, p) = Ap$

$$\begin{aligned}
J_A &= \frac{A \exp(\delta) p - Ap}{\delta} \\
&\cong \frac{Aa + A\hat{\delta}p - Ap}{\delta} \\
&= \frac{A\hat{\delta}p}{\delta} \\
&= -\frac{A\delta\hat{p}}{\delta} \\
&= -A\hat{p}
\end{aligned} \tag{A1-3}$$

$$\begin{aligned}
J_p &= \frac{A(p + \delta) - Ap}{\delta} \\
&= A
\end{aligned} \tag{A1-4}$$

According to A1-3:

$$\begin{aligned}
\frac{\partial p_{k+1}}{\partial R_k} &= -R_k \widehat{a_k^b} \frac{\Delta t^2}{2} \\
\frac{\partial v_{k+1}}{\partial R_k} &= -R_k \widehat{a_k^b} \Delta t
\end{aligned}$$

Proof of (9):

According to A1-4:

$$\begin{aligned}
\frac{\partial p_{k+1}}{\partial a_k^b} &= R_k \frac{\Delta t^2}{2} \\
\frac{\partial v_{k+1}}{\partial a_k^b} &= R_k \Delta t
\end{aligned}$$

Proof of (10):

According to A1-2:

$$\frac{\partial R_{k+1}}{\partial \omega_k^b} = I_{3 \times 3} \Delta t$$

A-2. Proof of [Delta between two states] (20)

A is a lie group, p is a vector: $\varphi(A, p) = A^{-1}p$

$$\begin{aligned}
J_A &= \frac{(A \exp(\hat{\delta}))^{-1} p - A^{-1} p}{\delta} \\
&= \frac{\exp(\widehat{-\delta}) A^{-1} p - A^{-1} p}{\delta} \\
&= \frac{(I - \hat{\delta}) A^{-1} p - A^{-1} p}{\delta} \\
&= \frac{-\hat{\delta} A^{-1} p}{\delta} \\
&= \frac{\delta x \widehat{A^{-1} p}}{\delta} \\
&= \widehat{A^{-1} p}
\end{aligned} \tag{A2-1}$$

$$\begin{aligned}
J_p &= \frac{T^{-1}(p + \delta) - T^{-1} p}{\delta} \\
&= \frac{T^{-1} \delta}{\delta} \\
&= T^{-1}
\end{aligned} \tag{A2-2}$$

Proof of (20)

The \mathcal{D} function:

$$\mathcal{D}(\xi, s_i) = \begin{bmatrix} R_{bc}^{\xi} \\ p_{bc}^{\xi} + R_{nb}^{-1} v_{nb} \Delta t + R_{nb}^{-1} g \frac{\Delta t^2}{2} \\ v_{bc}^{\xi} + R_{nb}^{-1} g \Delta t \end{bmatrix}$$

According to A2-1:

$$\begin{aligned}
\frac{\partial p_{bc}}{\partial R_{nb}} &= \widehat{R_{nb}^{-1} v_{nb}} \Delta t + \widehat{R_{nb}^{-1} g} \frac{\Delta t^2}{2} \\
\frac{\partial v_{bc}}{\partial R_{nb}} &= \widehat{R_{nb}^{-1} g} \Delta t
\end{aligned}$$

According to A2-2 and (22):

$$\begin{aligned}
\frac{\partial p_{bc}}{\partial v_{nb}} &= \frac{R_{nb}^{-1} (v_{nb} + R_{nb} \delta v_b) - R_{nb}^{-1} v_{nb}}{\delta v_b} \\
&= I_{3 \times 3} \Delta t
\end{aligned}$$

A-3. Proof of Retraction \mathcal{R} (23)(24)

The \mathcal{R} function:

$$\begin{aligned} R_{nc}^* &= R_{nb} R_{bc} \\ p_{nc}^* &= p_{nb} + R_{nb} p_{bc} \\ v_{nc}^* &= v_{nb} + R_{nb} v_{bc} \end{aligned}$$

The Jacobian of x for F :

$$J_x^F = \frac{\mathcal{L}(F(x), F(\mathcal{R}(x, \delta x)))}{\delta x} \quad (\text{A3-1})$$

Proof of $J_{s_i}^{\mathcal{R}}$ (23):

According to A1-1:

$$\frac{\partial R_{nc}^*}{\partial R_{nb}} = R_{bc}^{-1}$$

According to A2-2 and A3-1:

$$\begin{aligned} \frac{\partial p_{nc}^*}{\partial R_{nb}} &= \frac{R_{nc}^{-1} (R_{nb} \exp(\widehat{\delta\theta_b}) p_{bc} - R_{nb} p_{bc})}{\delta\theta_b} \\ &= -R_{bc}^{-1} \widehat{p_{bc}} \end{aligned}$$

$$\begin{aligned} \frac{\partial v_{nc}^*}{\partial R_{nb}} &= \frac{R_{nc}^{-1} (R_{nb} \exp(\widehat{\delta\theta_b}) v_{bc} - R_{nb} v_{bc})}{\delta\theta_b} \\ &= -R_{bc}^{-1} \widehat{v_{bc}} \end{aligned}$$

According to A1-3 and (22)(25):

$$\begin{aligned} \frac{\partial p_{bc}^*}{\partial p_{nb}} &= \frac{R_{nc}^{-1} (p_{nb} + R_{nb} \delta p_b - p_{nb})}{\delta p_b} \\ &= R_{bc}^{-1} \end{aligned}$$

$$\begin{aligned} \frac{\partial v_{bc}^*}{\partial v_{nb}} &= \frac{R_{nc}^{-1} (v_{nb} + R_{nb} \delta v_b - v_{nb})}{\delta v_b} \\ &= R_{bc}^{-1} \end{aligned}$$

Proof of $J_d^{\mathcal{R}}$ (24)

According to A2-2 and A3-1:

$$\frac{\partial p_{nc}^*}{\partial p_{bc}} = \frac{R_{nc}^{-1}(R_{nb}(p_{bc} + \delta p_b) - R_{nb}p_{bc})}{\delta p_b} = R_{bc}^{-1}$$

$$\frac{\partial v_{nc}^*}{\partial v_b} = \frac{R_{nc}^{-1}(R_{nb}(v_{bc} + \delta v_b) - R_{nb}v_{bc})}{\delta v_b} = R_{bc}^{-1}$$

A-4. Proof of Local \mathcal{L} (23)(24)

The \mathcal{L} function:

$$r_{jj^*} = \mathcal{L}(s_j, s_j^*) = \begin{bmatrix} \Delta R \\ \Delta p \\ \Delta v \end{bmatrix} = \begin{bmatrix} R_j^{-1} R_j^* \\ R_j^{-1}(p_j^* - p_j) \\ R_j^{-1}(v_j^* - v_j) \end{bmatrix}$$