

Let us define the state of the system as

$$\{\text{orientation } \mathbf{R} \in SO(3), \text{ velocity } \mathbf{v} \in \mathbb{R}^3, \text{ position } \mathbf{p} \in \mathbb{R}^3\}$$

and without bias for convenience. We call it an *extended pose*. Extended poses may be described by  $5 \times 5$  matrices in the following Lie group

$$SE_2(3) := \left\{ \mathbf{T} = \left[ \begin{array}{c|cc} \mathbf{R} & \mathbf{v} & \mathbf{p} \\ \hline \mathbf{0}_{2 \times 3} & \mathbf{I}_2 & \end{array} \right] \in \mathbb{R}^{5 \times 5} \middle| \begin{array}{l} \mathbf{R} \in SO(3) \\ \mathbf{v}, \mathbf{p} \in \mathbb{R}^3 \end{array} \right\}.$$

Our Kalman filter is an *error-state* filter. We define the state error as

$$\mathbf{T} := \exp(\boldsymbol{\xi}) \hat{\mathbf{T}}, \quad (1)$$

where  $\hat{\mathbf{T}}$  is a noise-free “mean”,  $\boldsymbol{\xi} \sim \mathcal{N}(\mathbf{0}_9, \boldsymbol{\Sigma})$  is a zero-mean multivariate Gaussian in  $\mathbb{R}^9$ , and  $\exp(\cdot)$  is the  $SE_2(3)$  exponential map.

Assume the filter obtains a measurement  $\mathbf{y} = h(\boldsymbol{\chi})$ , how computing the Jacobian

$$\mathbf{H} = \frac{\partial h}{\partial \boldsymbol{\xi}} \bigg|_{\hat{\boldsymbol{\chi}}} \quad (2)$$

to use this measurement in the filter?

First, developing the error, we have

$$\exp(\boldsymbol{\xi}) \hat{\mathbf{T}} \Leftrightarrow \begin{cases} \exp(\boldsymbol{\xi}^{\mathbf{R}}) = \mathbf{R} \hat{\mathbf{R}}^T \\ \boldsymbol{\xi}^{\mathbf{v}} = \mathbf{v} - \mathbf{R} \hat{\mathbf{R}}^T \hat{\mathbf{v}} \\ \boldsymbol{\xi}^{\mathbf{p}} = \mathbf{p} - \mathbf{R} \hat{\mathbf{R}}^T \hat{\mathbf{p}} \end{cases} \quad (3)$$

Linearizing the above results, i.e. applying

$$\exp(\boldsymbol{\xi}) \simeq \mathbf{I}_9 + \left[ \begin{array}{c|cc} \boldsymbol{\xi}_{\times}^{\mathbf{R}} & \boldsymbol{\xi}^{\mathbf{v}} & \boldsymbol{\xi}^{\mathbf{p}} \\ \hline \mathbf{0}_{2 \times 3} & \mathbf{0}_{2 \times 2} & \end{array} \right], \quad (4)$$

we obtain

$$\left( \mathbf{I}_9 + \left[ \begin{array}{c|cc} \boldsymbol{\xi}_{\times}^{\mathbf{R}} & \boldsymbol{\xi}^{\mathbf{v}} & \boldsymbol{\xi}^{\mathbf{p}} \\ \hline \mathbf{0}_{2 \times 3} & \mathbf{0}_{2 \times 2} & \end{array} \right] \right) \hat{\mathbf{T}} \Leftrightarrow \begin{cases} \boldsymbol{\xi}_{\times}^{\mathbf{R}} = \mathbf{R} \hat{\mathbf{R}}^T - \mathbf{I}_3 \\ \boldsymbol{\xi}^{\mathbf{v}} = \mathbf{v} - \hat{\mathbf{v}} - \boldsymbol{\xi}_{\times}^{\mathbf{R}} \hat{\mathbf{v}} \\ \boldsymbol{\xi}^{\mathbf{p}} = \mathbf{p} - \hat{\mathbf{p}} - \boldsymbol{\xi}_{\times}^{\mathbf{R}} \hat{\mathbf{p}} \end{cases} \quad (5)$$

Looking at the residual and Linearizing it as

$$\begin{aligned} \mathbf{r} &= \mathbf{y} - \hat{\mathbf{y}} \\ &= h(\boldsymbol{\chi}) - h(\hat{\boldsymbol{\chi}}) \\ &= h(\exp(\boldsymbol{\xi}) \hat{\boldsymbol{\chi}}) - h(\hat{\boldsymbol{\chi}}) \\ &\simeq h((\mathbf{I}_9 + \boldsymbol{\xi}^{\wedge}) \hat{\boldsymbol{\chi}}) - h(\hat{\boldsymbol{\chi}}) \end{aligned}$$

we can then identify the Jacobian term by term.

**Example** Assume we measure velocity in the vehicle frame, we have

$$\mathbf{y} = h(\boldsymbol{\chi}) = \mathbf{R}^T \mathbf{v}, \quad (6)$$

such that the residual is computed as

$$\begin{aligned} \mathbf{r} &= h(\boldsymbol{\chi}) - h(\hat{\boldsymbol{\chi}}) \\ &= \mathbf{R}^T \mathbf{v} - \hat{\mathbf{R}}^T \hat{\mathbf{v}} \\ &\simeq (\hat{\mathbf{R}} + \hat{\mathbf{R}} \boldsymbol{\xi}_{\times}^{\mathbf{R}})(\hat{\mathbf{v}} - \boldsymbol{\xi}_{\times}^{\mathbf{R}} \hat{\mathbf{v}} + \boldsymbol{\xi}^{\mathbf{v}}) - \hat{\mathbf{R}}^T \hat{\mathbf{v}} \\ &\simeq \hat{\mathbf{R}} \hat{\mathbf{v}} - \hat{\mathbf{R}} \hat{\mathbf{v}} - \hat{\mathbf{R}} \boldsymbol{\xi}_{\times}^{\mathbf{R}} \hat{\mathbf{v}} + \hat{\mathbf{R}} \boldsymbol{\xi}_{\times}^{\mathbf{R}} \hat{\mathbf{v}} + \hat{\mathbf{R}} \boldsymbol{\xi}^{\mathbf{v}} \\ &\simeq \hat{\mathbf{R}} \boldsymbol{\xi}^{\mathbf{v}} \end{aligned}$$

and the Jacobian is given as

$$\mathbf{H} = \begin{bmatrix} \mathbf{0}_{3 \times 3} & \hat{\mathbf{R}} & \mathbf{0}_{3 \times 3} \end{bmatrix}.$$