## Measurement Jacobians for Multi-Camera Visual-Inertial Navigation

Kevin Eckenhoff - keck@udel.edu Patrick Geneva - pgeneva@udel.edu Jesse Bloecker - jesseb@udel.edu Guoquan Huang - ghuang@udel.edu

Department of Mechanical Engineering University of Delaware, Delaware, USA

## RPNG

Robot Perception and Navigation Group (RPNG) Tech Report - RPNG-2018-MCVINS Last Updated - February 28, 2019 Consider a 3D feature captured in the image of the base camera b with timestamp  ${}^{b}t_{k}$ . The normalized feature measurement for this is given by:

$$\mathbf{z}_{k} = \begin{bmatrix} \frac{x}{z} \\ \frac{y}{z} \end{bmatrix} + \mathbf{n}_{k} \tag{1}$$

$$\begin{bmatrix} x\\ y\\ z \end{bmatrix} = {}^{C_b({}^bt_k)}\mathbf{p}_f = {}^{C_b}_{I}\mathbf{R}_G^{I({}^bt_k)}\mathbf{R} \left({}^{G}\mathbf{p}_f - {}^{G}\mathbf{p}_{I({}^bt_k)}\right) + {}^{C_b}\mathbf{p}_I$$
(2)

Here  $I({}^{b}t_{k})$  refers to the state of the IMU at true imaging time  ${}^{b}t_{k}$ , which gives simple Jacobians because the clone corresponding to  $I({}^{b}t_{k})$  is contained in our state vector.

The chain rule gives:

$$\frac{\partial \mathbf{z}_k}{\partial \delta \mathbf{x}} = \frac{\partial \mathbf{z}_k}{\partial^{C_b(b_{t_k})} \delta \mathbf{p}_f} \frac{\partial^{C_b(b_{t_k})} \delta \mathbf{p}_f}{\partial \delta \mathbf{x}}$$
(3)

$$\frac{\partial \mathbf{z}_k}{\partial C_b(^bt_k)\delta \mathbf{p}_f} = \begin{bmatrix} \frac{1}{z} & 0 & -\frac{x}{z^2} \\ 0 & \frac{1}{z} & -\frac{y}{z^2} \end{bmatrix}$$
(4)

$$\frac{\partial^{C_b({}^{b}t_k)}\delta\mathbf{p}_f}{\partial^{I(bt_k)}\delta\boldsymbol{\theta}_G} = {}_{I}^{C_b}\mathbf{R}\lfloor_G^{I(bt_k)}\mathbf{R}\left({}^{G}\mathbf{p}_f - {}^{G}\mathbf{p}_{I(bt_k)}\right)\rfloor$$
(5)

$$\frac{\partial^{C_b(^{b}t_k)}\delta\mathbf{p}_f}{\partial^C\delta\boldsymbol{\theta}_I} = \lfloor_I^{C_b}\mathbf{R}_G^{I(^{b}t_k)}\mathbf{R} \left({}^G\mathbf{p}_f - {}^G\mathbf{p}_{I(^{b}t_k)}\right) \rfloor$$
(6)

$$\frac{\partial^{C_b(^bt_k)}\delta\mathbf{p}_f}{\partial^G\delta\mathbf{p}_{I(^bt_k)}} = -_I^{C_b}\mathbf{R}_G^{I(^bt_k)}\mathbf{R}$$
(7)

$$\frac{\partial^{C_b({}^{b}t_k)}\delta\mathbf{p}_f}{\partial^G\delta\mathbf{p}_f} = {}_{I}^{C_b}\mathbf{R}_G^{I({}^{b}t_k)}\mathbf{R}$$
(8)

Now let us suppose that we we capture an image from another camera, cam i, and suppose that there exists an offset in the timestamps between cam b and cam i, such that if  ${}^{b}t_{k} = {}^{i}t_{k} + {}^{i}t_{b}$ , where  ${}^{b}t_{k}$  is the true time of the image expressed in the base camera's clock,  ${}^{i}t_{k}$  is the reported timestamp in the measuring camera's clock, and  ${}^{i}t_{b}$  is the unknown time offset.

For the measurement of a feature, we have:

$$C_{i}(^{i}t_{k}+^{i}t_{b})\mathbf{p}_{f} = {}_{I}^{C_{i}}\mathbf{R}_{G}^{I(^{i}t_{k}+^{i}t_{b})}\mathbf{R}\left({}^{G}\mathbf{p}_{f}-{}^{G}\mathbf{p}_{I(^{i}t_{k}+^{i}t_{b})}\right) + {}^{C_{i}}\mathbf{p}_{I}$$
(9)

We do not keep an estimate for the clone associated with this time, and instead rely on interpolation. Letting  ${}^{b}t_{1}$  and  ${}^{b}t_{2}$  denote the bounding IMU clones between which  ${}^{i}t_{k} + {}^{b}\hat{t}_{b}$  falls, we have:

$${}_{G}^{I(i_{t_{k}}+i_{t_{b}})}\mathbf{R} = \operatorname{Exp}\left(\lambda_{k}\operatorname{Log}\left({}_{G}^{I(b_{t_{2}})}\mathbf{R}{}_{I(b_{t_{1}})}^{G}\mathbf{R}\right)\right){}_{G}^{I(b_{t_{1}})}\mathbf{R}$$
(10)

$${}^{G}\mathbf{p}_{I(i_{t_{k}}+i_{t_{b}})} = (1-\lambda_{k})^{G}\mathbf{p}_{I(b_{t_{1}})} + \lambda_{k}{}^{G}\mathbf{p}_{I(b_{t_{2}})}$$
(11)

$$\lambda_k = \frac{{}^i t_k + {}^i t_b - {}^b t_1}{{}^b t_2 - {}^b t_1} \tag{12}$$

For this measurement, we can use the chain rule to compute the derivatives with respect to the bounding states and the unknown time offset:

## **RPNG-2018-MCVINS**

$$\frac{\partial^{C_i(it_k+it_b)}\delta\mathbf{p}_f}{\partial^{I(bt_1)}\delta\boldsymbol{\theta}_G} = \frac{\partial^{C_i(it_k+it_b)}\delta\mathbf{p}_f}{\partial^{I(it_k+it_b)}\delta\boldsymbol{\theta}_G} \frac{\partial^{I(it_k+it_b)}\delta\boldsymbol{\theta}_G}{\partial^{I(bt_2)}\delta\boldsymbol{\theta}_G}$$
(13)

$$\frac{\partial^{C_i(it_k+it_b)}\delta\mathbf{p}_f}{\partial^{I(bt_2)}\delta\boldsymbol{\theta}_G} = \frac{\partial^{C_i(it_k+it_b)}\delta\mathbf{p}_f}{\partial^{I(it_k+it_b)}\delta\boldsymbol{\theta}_G} \frac{\partial^{I(it_k+it_b)}\delta\boldsymbol{\theta}_G}{\partial^{I(bt_2)}\delta\boldsymbol{\theta}_G}$$
(14)

$$\frac{\partial^{C_i(i_{t_k}+i_{t_b})}\delta\mathbf{p}_f}{\partial^G\delta\mathbf{p}_{I(b_{t_1})}} = \frac{\partial^{C_i(i_{t_k}+i_{t_b})}\delta\mathbf{p}_f}{\partial^G\delta\mathbf{p}_{I(i_{t_k}+i_{t_b})}}\frac{\partial^G\mathbf{p}_{I(i_{t_k}+i_{t_b})}}{\partial^G\delta\mathbf{p}_{I(b_{t_1})}}$$
(15)

$$\frac{\partial^{C_i(i_{t_k}+i_{t_b})}\delta\mathbf{p}_f}{\partial^G\delta\mathbf{p}_{I(b_{t_2})}} = \frac{\delta^{C_i(i_{t_k}+i_{t_b})}\delta\mathbf{p}_f}{\partial^G\delta\mathbf{p}_{I(i_{t_k}+i_{t_b})}}\frac{\partial^G\delta\mathbf{p}_{I(i_{t_k}+i_{t_b})}}{\partial^G\delta\mathbf{p}_{I(b_{t_2})}}$$
(16)

$$\frac{\partial^{C_i(it_k+it_b)}\delta\mathbf{p}_f}{\partial\delta^i t_b} = \frac{\partial^{C_i(it_k+it_b)}\delta\mathbf{p}_f}{\partial^G\delta\mathbf{p}_{I(it_k+it_b)}} \frac{\partial^G\delta\mathbf{p}_{I(it_k+it_b)}}{\partial\delta^i t_b} + \frac{\partial^{C_i(it_k+it_b)}\delta\mathbf{p}_f}{\partial^{I(it_k+it_b)}\delta\boldsymbol{\theta}_G} \frac{\partial^{I(it_k+it_b)}\delta\boldsymbol{\theta}_G}{\partial\delta^i t_b}$$
(17)

The first terms in the chain rule are identical to those derived for the cam b case, but with the calibration between the IMU to cam i. The derivatives with respect to the position are computed as:

$$\frac{\partial^G \delta \mathbf{p}_{I(i_{t_k}+i_{t_b})}}{\partial^G \delta \mathbf{p}_{I(b_{t_1})}} = (1-\lambda_k) \mathbf{I}$$
(18)

$$\frac{\partial^G \delta \mathbf{p}_{I(i_{t_k}+i_{t_b})}}{\partial^G \delta \mathbf{p}_{I(b_{t_2})}} = \lambda_k \mathbf{I}$$
(19)

$$\frac{\partial^G \delta \mathbf{p}_{I(i_{t_k}+i_{t_b})}}{\partial \delta^i t_b} = \frac{1}{^{b}t_2 - ^{b}t_1} \left( {}^{G} \mathbf{p}_{I(b_{t_2})} - {}^{G} \mathbf{p}_{I(b_{t_1})} \right)$$
(20)

For the orientation derivatives, we use the following approximations for small angles  $\psi$ 

$$\operatorname{Exp}\left(\boldsymbol{\theta} + \boldsymbol{\psi}\right) \approx \operatorname{Exp}\left(\mathbf{J}_{l}\left(\boldsymbol{\theta}\right)\boldsymbol{\psi}\right)\operatorname{Exp}\left(\boldsymbol{\theta}\right)$$
(21)

$$\approx \operatorname{Exp}\left(\boldsymbol{\theta}\right) \operatorname{Exp}\left(\mathbf{J}_{r}\left(\boldsymbol{\theta}\right)\boldsymbol{\psi}\right) \tag{22}$$

where the definition of Jacobians of  $SO(3) \mathbf{J}_l(\cdot)$  and  $\mathbf{J}_r(\cdot)$  are the following:

$$\mathbf{J}_{l}(\phi) = \mathbf{I} + \frac{1 - \cos(\parallel \phi \parallel)}{\parallel \phi \parallel^{2}} \lfloor \phi \times \rfloor + \frac{\parallel \phi \parallel - \sin(\parallel \phi \parallel)}{\parallel \phi \parallel^{3}} \lfloor \phi \times \rfloor^{2}$$
(23)

$$\mathbf{J}_{r}(\phi) = \mathbf{I} - \frac{1 - \cos(\parallel \phi \parallel)}{\parallel \phi \parallel^{2}} \lfloor \phi \times \rfloor + \frac{\parallel \phi \parallel - \sin(\parallel \phi \parallel)}{\parallel \phi \parallel^{3}} \lfloor \phi \times \rfloor^{2}$$
(24)

We also define for convenience  ${}_{1}^{2}\boldsymbol{\theta} = \text{Log}\left({}_{I(^{b}t_{1})}^{I(^{b}t_{2})}\mathbf{R}\right)$ . In addition we will use the fact that  $\mathbf{R}\text{Exp}\left(\boldsymbol{\theta}\right) = \text{Exp}\left(\mathbf{R}\boldsymbol{\theta}\right)\mathbf{R}$ . We first expand the rotation interpolation equation with respect to a perturbation in the clone  $I(^{b}t_{1})$ :

$$\operatorname{Exp}\left(-{}^{I(i_{t_{k}}+i_{t_{b}})}\delta\boldsymbol{\theta}_{G}\right){}^{I(i_{t_{k}}+i_{t_{b}})}_{G}\mathbf{R}$$
(25)

$$= \operatorname{Exp}\left(\lambda_k \operatorname{Log}\left(\begin{smallmatrix} I^{(b_{t_2})} \mathbf{R}_{I(b_{t_1})}^G \mathbf{R} \operatorname{Exp}\left(I^{(b_{t_1})} \delta \boldsymbol{\theta}_G\right) \right)\right) \operatorname{Exp}\left(-I^{(b_{t_1})} \delta \boldsymbol{\theta}_G\right) I^{(b_{t_1})}_G \mathbf{R}$$
(26)

$$\approx \operatorname{Exp}\left(\lambda_{k}\left(_{1}^{2}\boldsymbol{\theta}+\mathbf{J}_{r}^{-1}\left(_{1}^{2}\boldsymbol{\theta}\right)^{I(^{b}t_{1})}\delta\boldsymbol{\theta}_{G}\right)\right)\operatorname{Exp}\left(-^{I(^{b}t_{1})}\delta\boldsymbol{\theta}_{G}\right)_{G}^{I(^{b}t_{1})}\mathbf{R}$$

$$(27)$$

$$\approx \operatorname{Exp}\left(\lambda_{k}\mathbf{J}_{l}\left(\lambda_{k1}^{2}\boldsymbol{\theta}\right)\mathbf{J}_{r}^{-1}\left(_{1}^{2}\boldsymbol{\theta}\right)^{I\left(^{b}t_{1}\right)}\delta\boldsymbol{\theta}_{G}\right)\operatorname{Exp}\left(\lambda_{k1}^{2}\boldsymbol{\theta}\right)\operatorname{Exp}\left(-^{I\left(^{b}t_{1}\right)}\delta\boldsymbol{\theta}_{G}\right)_{G}^{I\left(^{b}t_{1}\right)}\mathbf{R}$$

$$(28)$$

$$= \operatorname{Exp}\left(\lambda_{k}\mathbf{J}_{l}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)\mathbf{J}_{r}^{-1}\begin{pmatrix}^{2}\boldsymbol{\theta}\end{pmatrix}^{I(b_{1})}\delta\boldsymbol{\theta}_{G}\right)\operatorname{Exp}\left(-\operatorname{Exp}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)^{I(b_{1})}\delta\boldsymbol{\theta}_{G}\right)\operatorname{Exp}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\end{pmatrix}^{I(b_{1})}\mathbf{R} \quad (29)$$

$$= \left(\lambda_{k_{1}}\mathbf{J}_{k_{1}}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)\mathbf{J}_{r}^{-1}\begin{pmatrix}^{2}\boldsymbol{\theta}\end{pmatrix}^{I(b_{1})}\delta\boldsymbol{\theta}_{G}\right)\operatorname{Exp}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)^{I(b_{1})}\delta\boldsymbol{\theta}_{G}\right) = \left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)^{I(b_{1})}\delta\boldsymbol{\theta}_{G} \quad (29)$$

$$= \operatorname{Exp}\left(\lambda_{k}\mathbf{J}_{l}\left(\lambda_{k}_{1}^{2}\boldsymbol{\theta}\right)\mathbf{J}_{r}^{-1}\begin{pmatrix}2\\1\end{pmatrix} I^{(b_{1})}\delta\boldsymbol{\theta}_{G}\right)\operatorname{Exp}\left(-\operatorname{Exp}\left(\lambda_{k}_{1}^{2}\boldsymbol{\theta}\right)^{I(b_{1})}\delta\boldsymbol{\theta}_{G}\right)_{G}^{I(^{i}t_{k}+^{i}t_{b})}\mathbf{R}$$

$$(30)$$

$$\approx \operatorname{Exp}\left(\left(\lambda_{k}\mathbf{J}_{l}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)\mathbf{J}_{r}^{-1}\left(_{1}^{2}\boldsymbol{\theta}\right) - \operatorname{Exp}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)\right)^{I\left(^{b}t_{1}\right)}\delta\boldsymbol{\theta}_{G}\right)_{G}^{I\left(^{i}t_{k}+^{i}t_{b}\right)}\mathbf{R}$$

$$(31)$$

$$\Rightarrow \frac{\partial^{I(^{t}_{k}+^{t}_{b})}\delta\boldsymbol{\theta}_{G}}{\partial^{I(^{b}t_{1})}\delta\boldsymbol{\theta}_{G}} = -\left(\lambda_{k}\mathbf{J}_{l}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)\mathbf{J}_{r}^{-1}\left(_{1}^{2}\boldsymbol{\theta}\right) - \operatorname{Exp}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)\right)$$
(32)

Next we perturb  $I({}^{b}t_{2})$ 

$$\operatorname{Exp}\left(-{}^{I({}^{i}t_{k}+{}^{i}t_{b})}\delta\boldsymbol{\theta}_{G}\right){}^{I({}^{i}t_{k}+{}^{i}t_{b})}_{G}\mathbf{R}$$
(33)

$$= \operatorname{Exp}\left(\lambda_k \operatorname{Log}\left(\operatorname{Exp}\left(-{}^{I(^{b}t_2)}\delta\boldsymbol{\theta}_G\right)\right) {}^{I(^{b}t_2)}_G \mathbf{R}_{I(^{b}t_1)}^G \mathbf{R}\right) {}^{I(^{b}t_1)}_G \mathbf{R}$$
(34)

$$\approx \operatorname{Exp}\left(\lambda_{k}\left(_{1}^{2}\boldsymbol{\theta}-\mathbf{J}_{l}^{-1}\left(_{1}^{2}\boldsymbol{\theta}\right)^{I(^{b}t_{2})}\delta\boldsymbol{\theta}_{G}\right)\right)_{G}^{I(^{b}t_{1})}\mathbf{R}$$
(35)

$$\approx \operatorname{Exp}\left(-\lambda_{k}\mathbf{J}_{l}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)\mathbf{J}_{l}^{-1}\left(_{1}^{2}\boldsymbol{\theta}\right)^{I\left(^{b}t_{2}\right)}\delta\boldsymbol{\theta}_{G}\right)\operatorname{Exp}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)_{G}^{I\left(^{b}t_{1}\right)}\mathbf{R}$$
(36)

$$= \operatorname{Exp}\left(-\lambda_k \mathbf{J}_l\left(\lambda_{k_1}^2 \boldsymbol{\theta}\right) \mathbf{J}_l^{-1} \begin{pmatrix} 2\\ 1 \boldsymbol{\theta} \end{pmatrix}^{I(^{b}t_2)} \delta \boldsymbol{\theta}_G \right)_G^{I(^{i}t_k + ^{i}t_b)} \mathbf{R}$$
(37)

$$\Rightarrow \frac{\partial^{I(^{i}t_{k}+^{i}t_{b})}\delta\boldsymbol{\theta}_{G}}{\partial^{I(^{b}t_{2})}\delta\boldsymbol{\theta}_{G}} = \lambda_{k}\mathbf{J}_{l}\left(\lambda_{k}^{2}\boldsymbol{\theta}\right)\mathbf{J}_{l}^{-1}\left(_{1}^{2}\boldsymbol{\theta}\right)$$
(38)

Lastly we perturb  $\lambda_k$  by  $\delta \lambda_k = \frac{\delta^i t_b}{bt_2 - t_b}$ :

$$\operatorname{Exp}\left(-{}^{I(i_{t_{k}}+i_{t_{b}})}\delta\boldsymbol{\theta}_{G}\right){}^{I(i_{t_{k}}+i_{t_{b}})}_{G}\mathbf{R}$$
(39)

$$= \operatorname{Exp}\left( (\lambda_k + \delta \lambda_k) \operatorname{Log} \begin{pmatrix} I^{(b_{t_2})}_G \mathbf{R}_{I(b_{t_1})}^G \mathbf{R} \end{pmatrix} \right)_G^{I(b_{t_1})} \mathbf{R}$$
(40)

$$\approx \operatorname{Exp}\left(\mathbf{J}_{l}\left(\lambda_{k_{1}}^{2}\boldsymbol{\theta}\right)\right)_{1}^{2}\boldsymbol{\theta}\delta\lambda_{k}\right)_{G}^{I(i_{t_{k}}+i_{t_{b}})}\mathbf{R}$$
(41)

$$\Rightarrow \frac{\partial^{I(^{i}t_{k}+^{i}t_{b})}\delta\boldsymbol{\theta}_{G}}{\partial\delta\lambda_{k}} = -\mathbf{J}_{l}\left(\lambda_{k}^{2}\boldsymbol{\theta}\right)^{2}_{1}\boldsymbol{\theta}$$

$$(42)$$

$$\Rightarrow \frac{\partial^{I({}^{i}t_{k}+{}^{i}t_{b})}\delta\boldsymbol{\theta}_{G}}{\partial\delta^{i}t_{b}} = -\frac{1}{{}^{b}t_{2}-{}^{b}t_{1}}\mathbf{J}_{l}\left(\lambda_{k_{1}}{}^{2}\boldsymbol{\theta}\right){}_{1}^{2}\boldsymbol{\theta} = -\frac{1}{{}^{b}t_{2}-{}^{b}t_{1}}{}_{1}^{2}\boldsymbol{\theta}$$
(43)

## RPNG-2018-MCVINS