

公式推导：

$$\begin{bmatrix} r_p \\ r_q \\ r_v \\ r_{ba} \\ r_{bg} \end{bmatrix} = \begin{bmatrix} q_{wb_i}^* (p_{wb_j} - p_{wb_i} - v_i^w \Delta t + \frac{1}{2} g^w \Delta t^2) \alpha_{b_i b_j} \\ 2 [q_{b_i b_j}^* \otimes (q_{wb_i}^* \otimes q_{wb_j}^*)] \alpha_{yz} \\ q_{wb_i}^* (v_j^w - v_i^w + g^w \Delta t) - p_{b_i b_j} \\ b_{j^a}^a - b_{i^a}^a \\ b_{j^b}^b - b_{i^b}^b \end{bmatrix}$$

$$\frac{\partial r_{p_i}}{\partial p_{wb_i}} = \begin{bmatrix} -q_{wb_i}^* \end{bmatrix} \quad \frac{\partial r_{p_i}}{\partial p_{wb_j}} = \begin{bmatrix} q_{wb_i}^* \end{bmatrix}$$

$$\begin{aligned} \frac{\partial r_{p_i}}{\partial \theta_{bi}} &= \frac{\partial [q_{wb_i}^* \otimes \begin{bmatrix} 1 \\ \pm \delta \theta_{bi} \end{bmatrix}]^* (p_{wb_j} - p_{wb_i} - v_i^w \Delta t + \frac{1}{2} g^w \Delta t^2)}{\partial \delta \theta_{bi}} \\ &\stackrel{\text{第1个}}{=} \frac{\partial (R_{wb_i} \exp(\delta \theta_{bi}))^T (p_{wb_j} - p_{wb_i} - v_i^w \Delta t + \frac{1}{2} g^w \Delta t^2)}{\partial \delta \theta_{bi}} \\ &\stackrel{\text{第2个}}{=} \end{aligned}$$

$$= \frac{\partial \exp(\delta \theta_{bi}) R_{wb_i}^{-1} (p_{wb_j} - p_{wb_i} - v_i^w \Delta t + \frac{1}{2} g^w \Delta t^2)}{\partial \delta \theta_{bi}}$$

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$$= \frac{\partial (I - \delta \theta_{bi}) R_{wb_i}^{-1} (p_{wb_j} - p_{wb_i} - v_i^w \Delta t + \frac{1}{2} g^w \Delta t^2)}{\partial \delta \theta_{bi}}$$

$$= R_{wb_i}^{-1} (p_{wb_j} - p_{wb_i} - v_i^w \Delta t + \frac{1}{2} g^w \Delta t^2)$$

$$\frac{\partial r_{p_i}}{\partial v_i} = -q_{wb_i}^* \Delta t = -R_{wb_i}^{-1} \Delta t = -R_{b_i w} \Delta t$$

$$\begin{aligned} \frac{\partial r_{p_i}}{\partial b_{ai}} &= \frac{\partial (-\alpha_{b_i b_j})}{\partial b_{ai}} = \frac{\partial -(\alpha_{b_i b_j} + J_{b_i^a}^a \delta b_i^a + J_{b_i^b}^b \delta b_i^b)}{\partial b_{ai}} \\ &= -J_{b_i^a}^a \end{aligned}$$

$$\frac{\partial r_{p_i}}{\partial b_{gi}} \text{ 同上: } -J_{b_i^g}^g$$

$$\frac{\partial r_{q_i}}{\partial p_i} = 0$$

$$\frac{\partial r_{q_i}}{\partial q_i} = 2 \frac{\partial [q_{b_i b_i}^* \otimes (q_{b_i w}^* \otimes \begin{bmatrix} 1 \\ \pm \delta \theta_{b_i i} \end{bmatrix})^* \otimes q_{wb_j}^*]_{xyz}}{\partial \delta \theta_{b_i i}}$$

$$= -2 \frac{\partial [q_{bj}^* \otimes (q_{bi} \otimes [\pm \delta \theta_{bi}]) \otimes q_{bj}^*]_{\text{sys}}}{\partial \delta \theta_{bi}}$$

$$= -2 [0 \ I] [q_{bj}^* \otimes q_{bi}]_L [q_{bj}^*]_R \begin{bmatrix} 0 \\ \pm I \end{bmatrix}$$

$$\frac{\partial r_{qi}}{\partial v_i} = 0$$

$$\frac{\partial r_{qi}}{\partial \omega_{ai}} = 0$$

$$\begin{aligned} \frac{\partial r_{qi}}{\partial b_{ij}} &= 2 \frac{\partial \left[(q_{bi} \otimes [\pm J_{bi}^2 \delta b_i^g])^* \otimes (q_{bi}^* \otimes q_{bj}) \right]_{\text{sys}}}{\partial \delta b_i^g} \\ &= 2 [0 \ I] \left[\begin{bmatrix} 1 \\ \pm J_{bi}^2 \delta b_i^g \end{bmatrix}^* \otimes q_{bi}^* \otimes q_{bj} \right]^* \\ &\quad \frac{\partial \delta b_i^g}{\partial \delta b_i^g} \end{aligned}$$

$$= -2 [0 \ I] [q_{bj}^* \otimes q_{bi} \otimes [\pm J_{bi}^2]]_L \begin{bmatrix} 0 \\ \pm J_{bi}^2 \end{bmatrix}$$

$$\frac{\partial r_{vi}}{\partial p_i} = 0$$

$$\begin{aligned} \frac{\partial r_{vi}}{\partial q_i} &= \frac{\partial (R_{bi} \exp[\delta \theta_{bi}])^T (v_j^w - v_i^w + g^w \Delta t)}{\partial \delta \theta_{bi}} \\ &= \frac{(I - [\delta \theta_{bi}]^{\wedge}) R_{bi}^T (v_j^w - v_i^w + g^w \Delta t)}{\partial \delta \theta_{bi}} \end{aligned}$$

$$= [R_{bi}^T (v_j^w - v_i^w + g^w \Delta t)]^{\wedge}$$

$$\frac{\partial r_{vi}}{\partial v_i} = -R_{bi}^*$$

$$\frac{\partial r_{vi}}{\partial \omega_{ai}} = - \frac{\partial p_{bibi}}{\partial \omega_{ai}}$$

$$= - \frac{\partial p_{bibi} + J_{bi}^a \delta b_i^a + J_{bi}^g \delta b_i^g}{\partial \omega_{ai}}$$

$$= -J_{bi}^a$$

$$\begin{aligned}
 \frac{\partial r_i}{\partial b_i} &= - \frac{\partial P_i b_i}{\partial b_i} \\
 &= - \frac{\partial \bar{P}_i b_i + J_{b_i}^A \delta b_i^2 + J_{b_i}^G \delta b_i^3}{\partial b_i} \\
 &= - J_{b_i}^G
 \end{aligned}$$

$$\begin{aligned}
 \frac{\partial r_{bai}}{\partial p_i} &= 0, \quad \frac{\partial r_{bai}}{\partial q_i} = 0, \quad \frac{\partial r_{bai}}{\partial v_i} = 0, \quad \frac{\partial r_{bai}}{\partial b_i} = -I \\
 \frac{\partial r_{bai}}{\partial b_i} &= 0
 \end{aligned}$$

$$\frac{\partial f_{gi}}{\partial p_i} = 0, \quad \frac{\partial f_{gi}}{\partial q_i} = 0, \quad \frac{\partial f_{gi}}{\partial v_i} = 0, \quad \frac{\partial f_{gi}}{\partial b_{ai}} = 0.$$

$$\frac{\partial f_{gi}}{\partial b_{gi}} = -I$$

$$\frac{\partial f_{pi}}{\partial p_j} = R^{-1} w_{bi}, \quad \frac{\partial f_{pi}}{\partial q_j} = 0, \quad \frac{\partial f_{pi}}{\partial v_j} = 0.$$

$$\frac{\partial f_{pi}}{\partial b_{aj}} = 0, \quad \frac{\partial f_{pi}}{\partial b_{gj}} = 0$$

$$\frac{\partial f_{qi}}{\partial p_j} = 0, \quad \frac{\partial f_{qi}}{\partial q_j} = 2 \frac{\partial}{\partial q_{ij}} \left[q_{bij}^* \otimes q_{bi}^* \otimes q_{bj} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right]$$

$$= 2 \begin{bmatrix} 0 & I \end{bmatrix} \begin{bmatrix} q_{bij}^* \otimes q_{bi}^* \otimes q_{bj} \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\frac{\partial f_{qi}}{\partial v_j} = 0, \quad \frac{\partial f_{qi}}{\partial b_{aj}} = \frac{\partial f_{qi}}{\partial b_{gj}} = 0$$

$$\frac{\partial r_{vi}}{\partial p_j} = \frac{\partial r_{vi}}{\partial q_j} = \frac{\partial r_{vi}}{\partial b_{aj}} = \frac{\partial r_{vi}}{\partial b_{gj}} = 0.$$

$$\frac{\partial r_{vi}}{\partial v_j} = R_{wb_i}$$

$$\frac{\partial r_{Lvi}}{\partial p_j} = \frac{\partial r_{Lvi}}{\partial q_j} = \frac{\partial r_{Lvi}}{\partial v_j} = \frac{\partial r_{Lvi}}{\partial b_{gj}} = 0.$$

$$\frac{\partial r_{Lvi}}{\partial b_{aj}} = I$$

$$\frac{\partial r_{bgi}}{\partial p_j} = \frac{\partial r_{bgi}}{\partial q_j} = \frac{\partial r_{bgi}}{\partial v_j} = \frac{\partial r_{bgi}}{\partial b_{aj}} = 0.$$

$$\frac{\partial r_{bgi}}{\partial b_{gj}} = I$$

补全代码：

UpdateState: 中值法更新最优先验，然后F和B相对应的值补进去就行了

```
void IMUPreIntegrator::UpdateState(void) {
    ...
    //
    // TODO: a. update mean:
    //
    // 1. get w_mid:
    w_mid = 0.5 * (prev_w + curr_w);
    // 2. update relative orientation, so3:
    prev_theta_ij = state.theta_ij_;
    d_theta_ij = Sophus::SO3d::exp(w_mid*T);
    curr_theta_ij = prev_theta_ij * d_theta_ij;
    state.theta_ij_ = curr_theta_ij;
    // 3. get a_mid:
    a_mid = 0.5 * (prev_a + curr_a);
    // 4. update relative translation:
    state.alpha_ij_ += state.beta_ij_*T + 0.5*a_mid*T*T;
    // 5. update relative velocity:
    state.beta_ij_ += a_mid*T;
```

```

//
// TODO: b. update covariance:
//
// 1. intermediate results:
dR_inv = d_theta_ij.inverse().matrix();
prev_R = prev_theta_ij.matrix();
curr_R = curr_theta_ij.matrix();
prev_R_a_hat = prev_R * Sophus::S03d::hat(prev_a);
curr_R_a_hat = curr_R * Sophus::S03d::hat(curr_a);
//
// TODO: 2. set up F:
//
// F12 & F32:
F_ = MatrixF::Identity();
F_.block<3,3>(INDEX_ALPHA, INDEX_THETA) = -0.25*T*T * (prev_R_a_hat +
curr_R_a_hat*(Eigen::Matrix3d::Identity() - Sophus::S03d::hat(w_mid)*T));
F_.block<3,3>(INDEX_BETA, INDEX_THETA) = -0.5*T * (prev_R_a_hat +
curr_R_a_hat*(Eigen::Matrix3d::Identity() - Sophus::S03d::hat(w_mid)*T));
// F14 & F34:
F_.block<3,3>(INDEX_ALPHA, INDEX_B_A) = -0.25 * (prev_R + curr_R) * T*T;
F_.block<3,3>(INDEX_BETA, INDEX_B_A) = -0.5 * (prev_R + curr_R) * T;
// F15 & F35:
F_.block<3,3>(INDEX_ALPHA, INDEX_B_G) = 0.25*T*T*T * curr_R_a_hat;
F_.block<3,3>(INDEX_BETA, INDEX_B_G) = 0.5*T*T * curr_R_a_hat;
// F22:
F_.block<3,3>(INDEX_THETA, INDEX_THETA) = Eigen::Matrix3d::Identity() -
Sophus::S03d::hat(w_mid)*T;
//
// TODO: 3. set up G:
//
// G11 & G31:
B_ = MatrixB::Zero();
B_.block<3,3>(INDEX_ALPHA, INDEX_M_ACC_PREV) = 0.25*T*T * prev_R;
B_.block<3,3>(INDEX_BETA, INDEX_M_ACC_PREV) = 0.5*T * prev_R;
// G12 & G22 & G32:
B_.block<3,3>(INDEX_ALPHA, INDEX_M_GYR_PREV) = -0.125*T*T*T * curr_R_a_hat;
B_.block<3,3>(INDEX_THETA, INDEX_M_GYR_PREV) =
0.5*Eigen::Matrix3d::Identity()*T;
B_.block<3,3>(INDEX_BETA, INDEX_M_GYR_PREV) = -0.25*T*T * curr_R_a_hat;
// G13 & G33:
B_.block<3,3>(INDEX_ALPHA, INDEX_M_ACC_CURR) = 0.25 * curr_R * T*T;
B_.block<3,3>(INDEX_BETA, INDEX_M_ACC_CURR) = 0.5 * curr_R * T;
// G14 & G24 & G34:
B_.block<3,3>(INDEX_ALPHA, INDEX_M_GYR_CURR) = -0.125*T*T*T * curr_R_a_hat;
B_.block<3,3>(INDEX_THETA, INDEX_M_GYR_CURR) =
0.5*Eigen::Matrix3d::Identity()*T;
B_.block<3,3>(INDEX_BETA, INDEX_M_GYR_CURR) = -0.25*T*T * curr_R_a_hat;
// G45
B_.block<3,3>(INDEX_B_A, INDEX_R_ACC_PREV) = Eigen::Matrix3d::Identity() * T;
// G56
B_.block<3,3>(INDEX_B_G, INDEX_R_GYR_PREV) = Eigen::Matrix3d::Identity() * T;
// TODO: 4. update P_:
P_ = F_ * P_ * F_.transpose() + B_ * Q_ * B_.transpose();
//
// TODO: 5. update Jacobian:

```



```

//
J_ = F_ * J_;
}

```

oplusImpl : g2o更新状态量用的函数，除了姿态比较特殊外，其他直接用加法就行了。也要更新Vertex中对应的bias

```

virtual void oplusImpl(const double *update) override {
    //
    // TODO: do update
    //
    Eigen::Map<const Eigen::Matrix<double, 15, 1>> update_vec(update, 15);
    _estimate.pos += update_vec.block<3,1>(PRVAG::INDEX_POS, 0);
    _estimate.ori = _estimate.ori * Sophus::S03d::exp(update_vec.block<3,1>
(PRVAG::INDEX_ORI, 0));
    _estimate.vel += update_vec.block<3,1>(PRVAG::INDEX_VEL, 0);
    _estimate.b_a += update_vec.block<3,1>(PRVAG::INDEX_B_A, 0);
    _estimate.b_g += update_vec.block<3,1>(PRVAG::INDEX_B_G, 0);
    updateDeltaBiases(update_vec.block<3,1>(PRVAG::INDEX_B_A, 0),
update_vec.block<3,1>(PRVAG::INDEX_B_G, 0));
}

```

computeError : 在Edge里面计算误差值，基本上就是先用bias更新预积分之后吧误差公式添上就行了。

```

virtual void computeError() override {
    g2o::VertexPRVAG* v0 = dynamic_cast<g2o::VertexPRVAG*>(_vertices[0]);
    g2o::VertexPRVAG* v1 = dynamic_cast<g2o::VertexPRVAG*>(_vertices[1]);

    const Eigen::Vector3d &pos_i = v0->estimate().pos;
    const Sophus::S03d &ori_i = v0->estimate().ori;
    const Eigen::Vector3d &vel_i = v0->estimate().vel;
    const Eigen::Vector3d &b_a_i = v0->estimate().b_a;
    const Eigen::Vector3d &b_g_i = v0->estimate().b_g;

    const Eigen::Vector3d &pos_j = v1->estimate().pos;
    const Sophus::S03d &ori_j = v1->estimate().ori;
    const Eigen::Vector3d &vel_j = v1->estimate().vel;
    const Eigen::Vector3d &b_a_j = v1->estimate().b_a;
    const Eigen::Vector3d &b_g_j = v1->estimate().b_g;

    //
    // TODO: update pre-integration measurement caused by bias change:
    //

    if (v0 -> isUpdated() ) {
        Eigen::Vector3d d_b_a_i , d_b_g_i ;
        v0->getDeltaBiases(d_b_a_i, d_b_g_i);
        updateMeasurement(d_b_a_i,d_b_g_i);
    }

    //
    // TODO: compute error:
    //

```

```

        const Eigen::Vector3d &alpha_ij = _measurement.block<3, 1>(INDEX_P, 0);
        const Eigen::Vector3d &theta_ij = _measurement.block<3, 1>(INDEX_R, 0);
        const Eigen::Vector3d &beta_ij = _measurement.block<3, 1>(INDEX_V, 0);
        // const Sophus::S03d residual_ori_quat =
        (Sophus::S03d::exp(theta_ij).inverse() * ori_i.inverse() * ori_j);
        // const Eigen::Vector3d
        residual_ori(residual_ori_quat.unit_quaternion().x(),
        residual_ori_quat.unit_quaternion().y(),
        residual_ori_quat.unit_quaternion().z());
        _error.block<3, 1>(INDEX_P, 0) = ori_i.matrix() * (pos_j - pos_i -
        vel_i*T_ + 0.5*g_*T_*T_) - alpha_ij;
        _error.block<3, 1>(INDEX_R, 0) = (Sophus::S03d::exp(theta_ij).inverse() *
        ori_i.inverse() * ori_j).log();
        _error.block<3, 1>(INDEX_V, 0) = ori_i.inverse() * (vel_j - vel_i + g_ *
        T_) - beta_ij;
        _error.block<3, 1>(INDEX_A, 0) = b_a_j - b_a_i;
        _error.block<3, 1>(INDEX_G, 0) = b_g_j - b_g_i;
    }

```

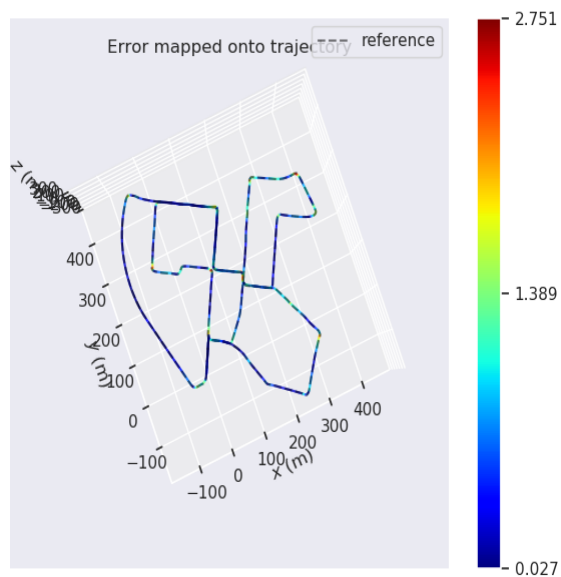
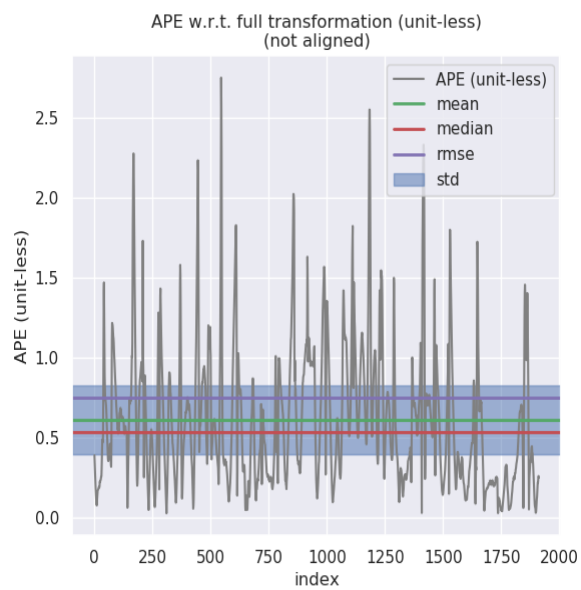
这一章有一点要注意就是pose和 T_{nb} 的区别。pose是当前的状态量(x)， T_{nb} 是观测量(y)，所以在计算的时候基本都要用pose_，只有在计算观测量与状态量之差才会用 T_{nb} 。

效果：



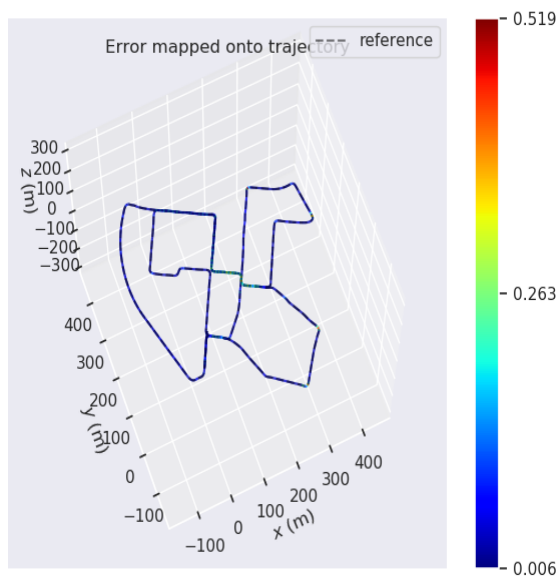
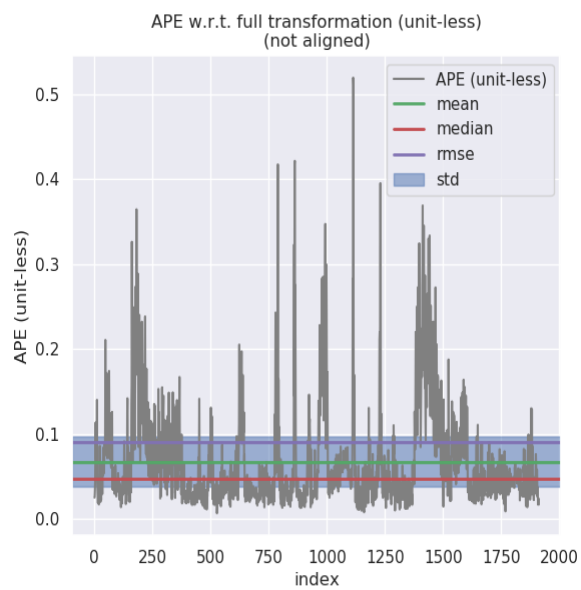
用IMU预积分

max	2.750943
mean	0.612079
median	0.531566
min	0.026736
rmse	0.747498
sse	1068.893565
std	0.429082



不用IMU预积分

max	0.519413
mean	0.067246
median	0.046663
min	0.006491
rmse	0.089805
sse	15.428082
std	0.059522



可以看出用IMU预积分反而效果变差了，应该是KITTI数据集的缘故

优秀部分：

编码器:

$$\begin{bmatrix} r_p \\ r_q \\ r_{pq} \end{bmatrix} = \begin{bmatrix} \hat{p}_{w_{ij}}^* (p_{w_{ij}} - p_{w_{ij}}) - \alpha_{b_{ij}} \dot{b}_j \\ 2 \left[\hat{b}_{ij}^* \otimes (\hat{q}_{w_{ij}}^* \otimes \hat{p}_{w_{ij}}) \right]_{xyz} \\ \dot{b}_j - \dot{b}_i \end{bmatrix}$$

由于文献以 α, θ, b_j 为状态量, 而非 p, \dot{b}, \dot{q} ,
我们需推导 α, θ, b_j 的状态更新公式, 以及 $\delta \alpha, \delta \theta, \delta b_j$.

$$\textcircled{1} \alpha_{b_{ijk}+1} = \alpha_{b_{ijk}} + \dot{b}_{b_{ijk}} \phi_k = \alpha_{b_{ijk}} + R_{b_{ijk}} \phi_k$$

$$\textcircled{2} \tilde{\alpha}_{b_{ijk}+1} = \tilde{\alpha}_{b_{ijk}} + \tilde{\dot{b}}_{b_{ijk}} \hat{\phi}_k, \quad \hat{\phi}_k = \frac{1}{\tilde{b}_{b_{ijk}} \phi_k + \tilde{b}_{b_{ijk}+1} \phi_{k+1}}$$

$$\textcircled{3} \tilde{\alpha}_{b_{ijk}+1} = \alpha_{b_{ijk}+1} + \delta \alpha_{b_{ijk}+1}$$

$$\tilde{\dot{b}}_{b_{ijk}} = \dot{b}_{b_{ijk}} \otimes \delta \dot{b}_{b_{ijk}} = R_{b_{ijk}} \exp(\delta \theta_k^{\wedge})$$

$$\hat{\phi}_k = \phi_k + \delta \phi_k.$$

$$\textcircled{4} \alpha_{b_{k+1}} + \delta \alpha_{b_{k+1}} = \alpha_{b_k} + \delta \alpha_{b_k} + R_{b_k} \exp(\delta \theta_k^{\wedge}) (\phi_k + \delta \phi_k)$$

$$\textcircled{5} \delta \alpha_{b_{k+1}} = \delta \alpha_{b_k} + R_{b_k} (\delta \phi_k + \delta \theta_k^{\wedge} \phi_k).$$

$$\textcircled{1} \dot{q}_{b_{k+1}} = \dot{q}_{b_k} \otimes \begin{bmatrix} 0 \\ \frac{1}{2} \dot{v}^b \delta t \end{bmatrix}, \quad \dot{w}^b = \frac{1}{2} \left[(\dot{w}^b - \frac{1}{2} \dot{g}) + (\dot{w}^{k+1} - \frac{1}{2} \dot{g}_{k+1}) \right]$$

$$\textcircled{2} \tilde{\dot{q}}_{b_{k+1}} = \tilde{\dot{q}}_{b_k} \otimes \begin{bmatrix} 0 \\ \frac{1}{2} \tilde{\dot{w}}^b \delta t \end{bmatrix}$$

$$\textcircled{3} \tilde{\dot{q}}_{b_{k+1}} = \dot{q}_{b_{k+1}} \otimes \delta \dot{q}_{b_{k+1}}$$

$$\tilde{\dot{w}}^b = \dot{w}^b + \delta \dot{w}^b$$

$$\textcircled{4} \dot{q}_{b_{k+1}} \otimes \delta \dot{q}_{b_{k+1}} = \dot{q}_{b_k} \otimes \delta \dot{q}_k \otimes \begin{bmatrix} 0 \\ \frac{1}{2} (\dot{w}^b + \delta \dot{w}^b) \delta t \end{bmatrix}$$

$$\textcircled{5} \dot{q}_{b_k} \otimes \begin{bmatrix} 0 \\ \frac{1}{2} \dot{w}^b \delta t \end{bmatrix} \otimes \delta \dot{q}_{b_{k+1}} = \dot{q}_{b_k} \otimes \delta \dot{q}_k \otimes \begin{bmatrix} 0 \\ \frac{1}{2} (\dot{w}^b + \delta \dot{w}^b) \delta t \end{bmatrix}$$

$$\begin{bmatrix} 0 \\ \frac{1}{2} \dot{w}^b \delta t \end{bmatrix} \otimes \delta \dot{q}_{b_{k+1}} = \delta \dot{q}_k \otimes \begin{bmatrix} 0 \\ \frac{1}{2} (\dot{w}^b + \delta \dot{w}^b) \delta t \end{bmatrix}$$

$$\delta \mathbf{z}_{k+1} = \begin{bmatrix} 0 & -\frac{1}{2} \mathbf{w}^T \delta t^T \\ \frac{1}{2} \mathbf{w}^T \delta t & (\frac{1}{2} \mathbf{w}^T \delta t)^T \end{bmatrix}^{-1} \begin{bmatrix} 0 & -\frac{1}{2} (\mathbf{w}^T + \delta \mathbf{w}^T) \delta t^T \\ \frac{1}{2} (\mathbf{w}^T + \delta \mathbf{w}^T) \delta t & (\frac{1}{2} (\mathbf{w}^T + \delta \mathbf{w}^T) \delta t)^T \end{bmatrix} \delta \mathbf{z}_k$$

$$\textcircled{1} \quad \theta_{k+1} = (\bar{\mathbf{w}}_k - \mathbf{b}_g)^T \theta_k \delta t + \theta_k, \quad \bar{\mathbf{w}}_k = \frac{1}{2} (\mathbf{w}_k + \mathbf{w}_{k+1})$$

$$\textcircled{2} \quad \tilde{\theta}_{k+1} = (\tilde{\bar{\mathbf{w}}}_k - \tilde{\mathbf{b}}_g)^T \tilde{\theta}_k \delta t + \tilde{\theta}_k$$

$$\textcircled{3} \quad \tilde{\theta}_k = \theta_k + \delta \theta_k$$

$$\tilde{\bar{\mathbf{w}}}_k = \bar{\mathbf{w}}_k + \delta \bar{\mathbf{w}}_k, \quad \delta \bar{\mathbf{w}}_k = \frac{1}{2} (\delta \mathbf{w}_k + \delta \mathbf{w}_{k+1})$$

$$\tilde{\mathbf{b}}_g = \mathbf{b}_g + \delta \mathbf{b}_g$$

$$\textcircled{4} \quad \theta_{k+1} + \delta \theta_{k+1} = (\bar{\mathbf{w}}_k + \delta \bar{\mathbf{w}}_k + \mathbf{b}_g + \delta \mathbf{b}_g)^T (\theta_k + \delta \theta_k) \delta t + \theta_k + \delta \theta_k$$

$$\textcircled{5} \quad \delta \theta_{k+1} = (I - \bar{\mathbf{w}}^T \delta t) \delta \theta_k + \delta t \delta \mathbf{b}_g + \delta \bar{\mathbf{w}}_k$$

$$\textcircled{1} \quad \mathbf{b}_{j+1}^g = \mathbf{b}_j^g + \delta \mathbf{b}_{j+1}^g \delta t$$

$$\textcircled{2} \quad \tilde{\mathbf{b}}_{j+1}^g = \tilde{\mathbf{b}}_j^g$$

$$\textcircled{3} \quad \tilde{\mathbf{b}}_{j+1}^g = \mathbf{b}_{j+1}^g + \delta \mathbf{b}_{j+1}^g + \delta \mathbf{b}_{j+1}^g \delta t$$

$$\textcircled{4} \quad \mathbf{b}_{j+1}^g + \delta \mathbf{b}_{j+1}^g = \mathbf{b}_j^g + \delta \mathbf{b}_j^g + \delta \mathbf{b}_{j+1}^g \delta t$$

$$\textcircled{5} \quad \delta \mathbf{b}_{j+1}^g = \delta \mathbf{b}_j^g + \delta \mathbf{b}_{j+1}^g \delta t$$

State Update Equation: $x = \begin{bmatrix} \delta \alpha_k \\ \delta \theta_k \\ \delta b_k \end{bmatrix}, u = \begin{bmatrix} \delta \phi_k \\ \delta w_k \\ \delta b_{k+1}^g \\ \delta b_k^g \end{bmatrix}$

$$\delta \alpha_{k+1} = \delta \alpha_k + R_k (\delta \phi_k + \delta \theta_k^T \phi_k)$$

$$\frac{\partial \alpha_{k+1}}{\partial \alpha_k} = I, \frac{\partial \alpha_{k+1}}{\partial \theta_k^T} = -R_k \phi_k, \frac{\partial \alpha_{k+1}}{\partial b_k} = 0.$$

$$\frac{\partial \alpha_{k+1}}{\partial \phi_k} = R_k, \frac{\partial \alpha_{k+1}}{\partial w_k} = \frac{\partial \alpha_{k+1}}{\partial w_{k+1}} = \frac{\partial \alpha_{k+1}}{\partial b_k^g} = 0.$$

$$\delta \theta_{k+1} = (I - \bar{w}^T \delta t) \delta \theta_k + \delta t \delta b_g + \delta w_k$$

$$\frac{\partial \theta_{k+1}}{\partial \alpha_k} = 0, \frac{\partial \theta_{k+1}}{\partial \theta_k} = I - \bar{w}^T \delta t, \frac{\partial \theta_{k+1}}{\partial b_g} = \delta t.$$

$$\frac{\partial \theta_{k+1}}{\partial \phi} = 0, \frac{\partial \theta_{k+1}}{\partial w_k} = \frac{1}{2} \delta t I = \frac{\partial \theta_{k+1}}{\partial w_{k+1}} \frac{\partial \theta_{k+1}}{\partial b_k^g} = 0$$

$$\delta b_{j+1}^g = \delta b_j^g + \delta b_{jj+1}^g + \delta t$$

$$\frac{\partial b_{j+1}^g}{\partial \alpha_k} = 0 = \frac{\partial b_{j+1}^g}{\partial \theta_k}, \frac{\partial b_{j+1}^g}{\partial b_k^g} = I.$$

$$\frac{\partial b_{j+1}^g}{\partial \phi} = 0 = \frac{\partial b_{j+1}^g}{\partial w_k} = \frac{\partial b_{j+1}^g}{\partial w_{k+1}} \frac{\partial b_{j+1}^g}{\partial b_k^g} = \delta t I$$

$$\therefore \bar{f}_k = \begin{bmatrix} I & -R_k [\phi_k]^T & 0 \\ 0 & I - [\bar{w}]^T \delta t & -I \delta t \\ 0 & 0 & I \end{bmatrix}$$

$$G_k = \begin{bmatrix} R_k & 0 & 0 & 0 \\ 0 & \frac{1}{2} \delta t I & \frac{1}{2} \delta t I & 0 \\ 0 & 0 & 0 & \delta t I \end{bmatrix}$$

Jacobian:

$$\frac{\partial r_p}{\partial p_i} = -R_{bi}^{-1}, \frac{\partial r_p}{\partial q_i} = R_{bi}^{-1}(p_{wj} - p_{wi})$$

$$\frac{\partial r_p}{\partial b_j^g} = \frac{\partial d_{bij}}{\partial b_j^g} = -J_{bij}^x$$

$$\frac{\partial r_e}{\partial p_i} = 0, \frac{\partial r_e}{\partial q_i} = -2 \begin{bmatrix} 0 & I \end{bmatrix} \begin{bmatrix} \mathbf{f}_{ij}^* \otimes \mathbf{b}_{bi} \end{bmatrix}_L \begin{bmatrix} \mathbf{f}_{ij} \end{bmatrix}_{R_{ij}^*} \begin{bmatrix} 0 \\ \frac{1}{2} J_{bij}^g \end{bmatrix}$$

$$\frac{\partial r_e}{\partial b_{ji}} = -2 \begin{bmatrix} 0 & I \end{bmatrix} \begin{bmatrix} \mathbf{f}_{ij}^* \otimes \mathbf{b}_{bi} \otimes \mathbf{b}_{ij} \end{bmatrix}_L \begin{bmatrix} 0 \\ \frac{1}{2} J_{bij}^g \end{bmatrix}$$

$$\frac{\partial r_{bji}}{\partial p_i} = 0, \frac{\partial r_{bji}}{\partial q_i} = 0, \frac{\partial r_{bji}}{\partial b_{ji}^g} = -I$$

$$\frac{\partial r_{pj}}{\partial p_j} = R_{bi}^{-1}, \frac{\partial r_{pj}}{\partial q_j} = 0, \frac{\partial r_{pj}}{\partial b_j^g} = 0$$

$$\frac{\partial r_{qv}}{\partial p_j} = 0, \frac{\partial r_{qv}}{\partial q_j} = 2 \begin{bmatrix} 0 & I \end{bmatrix} \begin{bmatrix} \mathbf{f}_{ij}^* \otimes \mathbf{f}_{bi}^* \otimes \mathbf{b}_{bi} \end{bmatrix}_L \begin{bmatrix} 0 \\ \frac{1}{2} I \end{bmatrix}$$

$$\frac{\partial r_{bi}^g}{\partial p_j} = 0, \frac{\partial r_{bi}^g}{\partial q_j} = 0, \frac{\partial r_{bi}^g}{\partial b_j^g} = I$$