

VINS online extrinsic calibration

Saturday, January 30, 2021

10:58 AM

extrinsics: from camera frame to IMU frame; bT_c

We construct least squares $Ax=0$, and use SVD to find the smallest singular value. Its corresponding eigenvector is the solution

To construct least squares:

between time $k, k+1$, we have IMU transformation $q_{b_k b_{k+1}}$, camera transformation $q_{c_k c_{k+1}}$

$$q_{b_{k+1}}^{b_k} \otimes q_{b_k}^{c_k} = q_{b_{k+1}}^{c_{k+1}} \otimes q_{c_{k+1}}^{c_k} = q_{b_{k+1}}^{c_k} \quad \downarrow \text{from } c_k \text{ to } b_{k+1} \text{ transformation.}$$

↘ equal ↙

$$\therefore ([q_{b_k b_{k+1}}]_L - [q_{c_k c_{k+1}}]_R) q_{bc} = Q_{k+1}^k q_{bc} = 0. \quad (6)$$

$[\cdot]_L, [\cdot]_R$ are left/right quaternion multiplication.

Stack the equations at multiple times, and use the robust kernel cost function:

$$\begin{bmatrix} w_1^0 \cdot Q_1^0 \\ w_2^1 \cdot Q_2^1 \\ \vdots \\ w_n^{n-1} \cdot Q_n^{n-1} \end{bmatrix} q_{bc} = Q_N \cdot q_{bc} = 0.$$

$$\text{where } w_{k+1}^k = \begin{cases} 1 & r_{k+1}^k < \text{threshold} \\ \frac{\text{threshold}}{r_{k+1}^k} & \text{otherwise} \end{cases}$$

for rotation residual r :

$$r_{k+1}^k = \arccos\left(\left(\text{tr}\left(\hat{R}_{bc}^{-1} R_{b_k b_{k+1}}^{-1} \hat{R}_{bc} R_{c_k c_{k+1}}\right) - 1\right) / 2\right)$$

In code:

if (initial_ex_rotation.CalibrationExRotation(corres, pre_integrations[frame_count] → delta_q, calib_ric))

corres: feature correspondence between two frames, normalized coordinates, a vector

delta_q: relative rotation from pre integration.

bool InitialExRotation::CalibrationExRotation(vector<pair<Vector3d, Vector3d>> corres,

Quaterniond delta_q_imu,

Matrix3d & calib_ric_result)

{

// record the time we enter this function

// each time construct one row in Eq (6)

frame_count++;

// use matched features to find essential matrix

// decompose it to get rotation.

Rc.push_back(solveRelativeR(corres));

Rimu.push_back(delta_q_imu.toRotationMatrix());

// convert $R_{b_{k+1}}^{b_k}$ to $R_{c_{k+1}}^{c_k}$: $R_{c_{k+1}}^{c_k} = R_c^b R_{b_{k+1}}^{b_k} R_b^c$

Rc_g.push_back(ric.inverse() * delta_q_imu * ric);

// construct A matrix

Eigen::MatrixXd A(frame_count * 4, 4);

A.setZero();

int sum_ok = 0;

for (int i=1; i<=frame_count; i++)

{

Quaterniond r1(Rc[i]);

Quaterniond r2(Rc_g[i]);

// angularDistance is the relative transformation theta

// used in angle-axis Qu.

// this is to compute weights for robust kernel.

double angular_distance = 180 / M_PI * r1.angularDistance(r2);

double huber = angular_distance > 5.0 ? 5.0 / angular_distance : 1.0;

++sum_ok;

Matrix4d L, R;

double w = Quaterniond(Rc[i]).w();

Vector3d q = Quaterniond(Rc[i]).vec();

L.block<3,3>(0,0) = w * Matrix3d::Identity(1) + Utility::skewSymmetric(q);

L.block<3,1>(0,3) = q;

L.block<1,3>(3,0) = -q.transpose();

L(3,3) = w;

Quaterniond R_ij(Rimu[i]);

w = R_ij.w();

q = R_ij.vec();

R.block<3,3>(0,0) = w * Matrix3d::Identity(1) - Utility::skewSymmetric(q);

R.block<3,1>(0,3) = q;

R.block<1,3>(3,0) = -q.transpose();

R(3,3) = w;

A.block<4,4>((i-1)*4,0) = huber * (L - R);

}

// use svd to solve the least squares system.

JacobiSVD<MatrixXd> svd(A, ComputeFullU | ComputeFullV);

// quaternion is $[x, y, z, w]^T$, i.e. $[q_v q_w]^T$

Matrix<double, 4, 1> x = svd.matrixV().col(3);

Quaterniond estimated_R(x);

ric = estimated_R.toRotationMatrix().inverse();

Vector3d ric_cov;

ric_cov = svd.singularValues().tail<3>();

// iterate WINDOW_SIZE times, and the second smallest singular value is larger than 0.25

if (frame_count >= WINDOW_SIZE && ric_cov(1) > 0.25)

{

calib_ric_result = ric;

return true;

}

else

return false;