

# Homework-05

## Homework-05

1. 推导雅可比
2. 完成新的内参模型下的标定, 且结果正确

修改思路

成果

使用解析式求导完成标定, 且结果正确

## 1. 推导雅可比

(这里的符号表示和PPT相比做了些改动。。比如零偏下标应该是a)

$$\boldsymbol{\theta}^{acc} = [S_{ayx} \quad S_{azx} \quad S_{azy} \quad K_{ax} \quad K_{ay} \quad K_{az} \quad b_{ax} \quad b_{ay} \quad b_{az}]^T \quad (1)$$

$$f(\boldsymbol{\theta}^{acc}) = \|g\|^2 - \|a\|^2 \quad (2)$$

$$a = (I - S_a)K'_a(A - b_a) \quad (3)$$

$$\begin{aligned} J &= \frac{\partial f}{\partial \boldsymbol{\theta}} \\ &= \frac{\partial(\|g\|^2 - \|a\|^2)}{\partial \boldsymbol{\theta}} \\ &= -\frac{\partial\|a\|^2}{\partial \boldsymbol{\theta}} \\ &= -\frac{\partial\|(I - S_a)K'_a(A - b_a)\|^2}{\partial \boldsymbol{\theta}} \end{aligned} \quad (4)$$

先计算a:

$$\begin{aligned} a &= (I - S_a)K'_a(A - b_a) \\ &= \begin{bmatrix} 1 & 0 & 0 \\ -S_{ayx} & 1 & 0 \\ -S_{azx} & -S_{azy} & 1 \end{bmatrix} \begin{bmatrix} 1/K_{ax} & 0 & 0 \\ 0 & 1/K_{ay} & 0 \\ 0 & 0 & 1/K_{az} \end{bmatrix} \begin{bmatrix} A_x - b_{ax} \\ A_y - b_{ay} \\ A_z - b_{az} \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & 0 \\ -S_{ayx} & 1 & 0 \\ -S_{azx} & -S_{azy} & 1 \end{bmatrix} \begin{bmatrix} (A_x - b_{ax})/K_{ax} \\ (A_y - b_{ay})/K_{ay} \\ (A_z - b_{az})/K_{az} \end{bmatrix} \\ &= \begin{bmatrix} (A_x - b_{ax})/K_{ax} \\ (A_y - b_{ay})/K_{ay} - (A_x - b_{ax})S_{ayx}/K_{ax} \\ (A_z - b_{az})/K_{az} - (A_x - b_{ax})S_{azx}/K_{ax} - (A_y - b_{ay})S_{azy}/K_{ay} \end{bmatrix} \end{aligned} \quad (5)$$

因此

$$\begin{aligned} \|a\|^2 &= a_x^2 + a_y^2 + a_z^2 \\ \boldsymbol{\theta}^{acc} &= [S_{ayx} \quad S_{azx} \quad S_{azy} \quad K_{ax} \quad K_{ay} \quad K_{az} \quad b_{ax} \quad b_{ay} \quad b_{az}]^T \end{aligned} \quad (6)$$

$$\begin{aligned}
-\frac{\partial \|a\|^2}{\partial \theta} &= -\frac{2a_x \partial a_x}{\partial \theta} - \frac{2a_y \partial a_y}{\partial \theta} - \frac{2a_z \partial a_z}{\partial \theta} \\
&= -2a_x \begin{bmatrix} 0 \\ 0 \\ 0 \\ -\frac{A_x - b_{ax}}{K_{ax}^2} \\ 0 \\ 0 \\ -\frac{1}{K_{ax}} \\ 0 \\ 0 \end{bmatrix} - 2a_y \begin{bmatrix} -\frac{A_x - b_{ax}}{K_{ax}} \\ 0 \\ 0 \\ \frac{(A_x - b_{ax})S_{ayx}}{K_{ax}^2} \\ -\frac{A_y - b_{ay}}{K_{ay}^2} \\ 0 \\ \frac{S_{ayx}}{K_{ax}} \\ -\frac{1}{K_{ay}} \end{bmatrix} - 2a_z \begin{bmatrix} 0 \\ -\frac{A_x - b_{ax}}{K_{ax}} \\ -\frac{A_y - b_{ay}}{K_{ay}} \\ \frac{(A_x - b_{ax})S_{azx}}{K_{ax}^2} \\ \frac{(A_y - b_{ay})S_{azy}}{K_{ay}^2} \\ -\frac{A_z - b_{az}}{K_{az}^2} \\ \frac{S_{azx}}{K_{ax}} \\ \frac{S_{azy}}{K_{ay}} \\ -\frac{1}{K_{az}} \end{bmatrix} \\
&= \begin{bmatrix} 2a_y \frac{A_x - b_{ax}}{K_{ax}} \\ 2a_z \frac{A_x - b_{ax}}{K_{ax}} \\ 2a_z \frac{A_y - b_{ay}}{K_{ay}} \\ 2a_x \frac{A_x - b_{ax}}{K_{ax}^2} - 2a_y \frac{(A_x - b_{ax})S_{ayx}}{K_{ax}^2} - 2a_z \frac{(A_x - b_{ax})S_{azx}}{K_{ax}^2} \\ 2a_y \frac{A_y - b_{ay}}{K_{ay}^2} - 2a_z \frac{(A_y - b_{ay})S_{azy}}{K_{ay}^2} \\ 2a_z \frac{A_z - b_{az}}{K_{az}^2} \\ 2a_x \frac{1}{K_{ax}} - 2a_z \frac{S_{azx}}{K_{ax}} \\ -2a_y \frac{S_{ayx}}{K_{ax}} - 2a_z \frac{S_{azy}}{K_{ay}} \\ 2a_y \frac{1}{K_{ay}} + 2a_z \frac{1}{K_{az}} \end{bmatrix} \tag{7}
\end{aligned}$$

其中

$$\begin{aligned}
-2a_x &= -\frac{2(A_x - b_{ax})}{K_{ax}} \\
-2a_y &= -\frac{2(A_y - b_{ay})}{K_{ay}} + \frac{2(A_x - b_{ax})S_{ayx}}{K_{ax}} \\
-2a_z &= -\frac{2(A_z - b_{az})}{K_{az}} + \frac{2(A_x - b_{ax})S_{azx}}{K_{ax}} + \frac{2(A_y - b_{ay})S_{azy}}{K_{ay}}
\end{aligned} \tag{8}$$

(9)

## 2. 完成新的内参模型下的标定, 且结果正确

# 修改思路

主要要看CalibratedTriad\_的结构：

```
/*
CalibratedTriad_( const _T &mis_yz = _T(0), const _T &mis_zy = _T(0), const _T &mis_zx = _T(0),
const _T &mis_xz = _T(0), const _T &mis_xy = _T(0), const _T &mis_yx = _T(0),
const _T &s_x = _T(1),    const _T &s_y = _T(1),    const _T &s_z = _T(1),
const _T &b_x = _T(0),    const _T &b_y = _T(0),    const _T &b_z = _T(0) );

~CalibratedTriad_();*/

*          OR A generic orthogonal sensor triad
*
* Triad model:
*
* -Misalignment matrix:
*
* general case:
*
*      [   1     -mis_yz     mis_zy   ]
* T = [ mis_xz       1     -mis_zx   ]
*      [ -mis_xy     mis_yx     1     ]
*
* "body" frame spacial case:
*
*      [   1     -mis_yz     mis_zy   ]
* T = [   0         1     -mis_zx   ]
*      [   0         0         1     ]
*
* Scale matrix:
*
*      [   s_x       0       0   ]
* K = [   0       s_y       0   ]
*      [   0       0       s_z   ]
*
```

构造函数中，依次将参数传入，前六个为mis\_yz , mis\_zy , mis\_zx , mis\_xz, mis\_xy, mis\_yx。然后可以构造出如上所示的矩阵，对应ppt中的安装误差矩阵。因此，我们需要将前三个参数置0，第4-6参数设置为需要优化的误差矩阵参数。

即如下所示，在残差中设置

```
80
81     * assume body frame same as accelerometer frame,
82     * so bottom left params in the misalignment matrix
83     CalibratedTriad_<_T2> calib_triad(
84         //
85         // TODO: implement lower triad model here
86         //
87         // // mis_yz, mis_zy, mis_zx:
88         // params[0], params[1], params[2],
89         // // mis_xz, mis_xy, mis_yx:
90         // _T2(0), _T2(0), _T2(0),
91         // mis_yz, mis_zy, mis_zx:
92         // _T2(0), _T2(0), _T2(0),
93         // mis_xz, mis_xy, mis_yx:
94         // params[0], params[1], params[2],
95         //     s_x,     s_y,     s_z:
96         // params[3], params[4], params[5],
97         //     b_x,     b_y,     b_z:
98         // params[6], params[7], params[8]
99     );
100
```

下图中，init\_acc\_calib\_是MultiPosCalibration\_中的成员变量，即初始化的calib参数，从主函数test\_imu\_calib.cpp可知，mis参数初始化均为0。因而我认为这里改不改，对运行没有影响。。改了之后，代码在逻辑上保持一致。

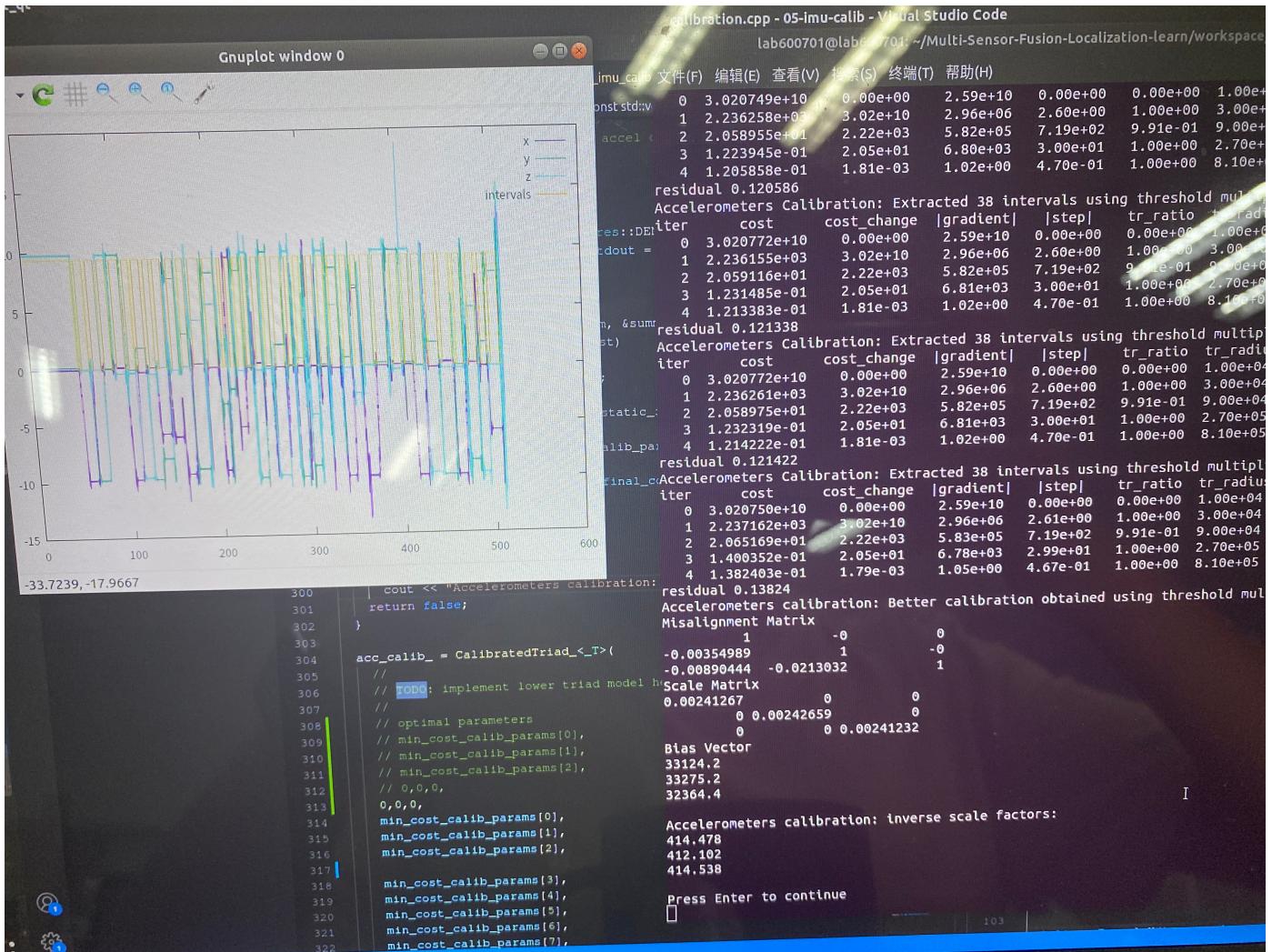
```
7   for (int th_mult = 2; th_mult <= 10; th_mult++)
8   {
9     std::vector< imu_tk::DataInterval > static_intervals;
10    std::vector< imu_tk::TriadData_<_T> > static_samples;
11    std::vector< double > acc_calib_params(9);
12
13    //
14    // TODO: implement lower triad model here
15    //
16    // CalibratedTriad_<_T> init_acc_calib_ is member variable
17    // acc_calib_params is parameters to be estimated
18    // init_acc_calib_.mis[] should be ZERO
19    // acc_calib_params[0] = init_acc_calib_.misYZ();
20    // acc_calib_params[1] = init_acc_calib_.misZY();
21    // acc_calib_params[2] = init_acc_calib_.misZX();
22    acc_calib_params[0] = init_acc_calib_.misXZ();
23    acc_calib_params[1] = init_acc_calib_.misXY();
24    acc_calib_params[2] = init_acc_calib_.misYX();
25
26    acc_calib_params[3] = init_acc_calib_.scaleX();
27    acc_calib_params[4] = init_acc_calib_.scaleY();
28    acc_calib_params[5] = init_acc_calib_.scaleZ();
29
30    acc_calib_params[6] = init_acc_calib_.biasX();
31    acc_calib_params[7] = init_acc_calib_.biasY();
32    acc_calib_params[8] = init_acc_calib_.biasZ();
33
34    std::vector< DataInterval > extracted_intervals;
35    staticIntervalsDetector ( acc_samples, th_mult*norm_
36      extractIntervalsSamples ( acc_samples, static_intervals,
37        static_samples, extracted_intervals,
38        interval_n_samples - acc_l
39      )
40    )
41  }
```

下面构造的CalibratedTriad\_矩阵，应该就是最后输出到终端用的。这里也要对应前面所说的结构，前三个参数值为0，第4-6个为优化算法算出来的最优参数。

这样，最终打印输出的就是下三角矩阵。

```
acc_calib_ = CalibratedTriad_<_T>(
    //
    // TODO: implement lower triad model here
    //
    // optimal parameters
    // min_cost_calib_params[0],
    // min_cost_calib_params[1],
    // min_cost_calib_params[2],
    // 0,0,0,
    0,0,0,
    min_cost_calib_params[0],
    min_cost_calib_params[1],
    min_cost_calib_params[2],
    min_cost_calib_params[3],
    min_cost_calib_params[4],
    min_cost_calib_params[5],
    min_cost_calib_params[6],
    min_cost_calib_params[7],
    min_cost_calib_params[8]
);
```

## 成果



可以看到 misalignment matrix 变成了下三角矩阵（原先是上三角）

## 使用解析式求导完成标定, 且结果正确

**TODO**