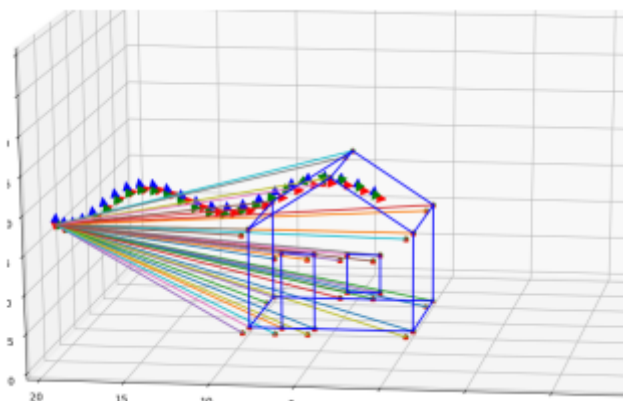


# 从零开始手写VIO 第七课作业

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## 1. 将第二讲的仿真数据集(视觉, IMU数据)接入我们的VINS代码, 并运行出轨迹结果.

- 仿真数据集无噪声
- 仿真数据集有噪声(不同噪声设定时, 需要配置 vins 中imu noise大小)



## 回答:

- **以下为第二版作业新增内容:**
  - 助教评语: 增大家对轨迹的量化分析, 及对同一噪声数据比较不同配置文件对定位精度的影响
  - 本次修改重点: 增加使用评估工具对不同配置文件对定位精度的量化评价
  - 本次评估工具为 `uzh-rpg/rpg-trajectory-evaluation`, 具体使用过程可以参考 github 上说明;
  - 本次分为两次评估:
    - 评估1: 仿真数据无噪声
    - 评估2: 构造三组有噪声的数据, 分别使用:
      - 原始噪声 IMU参数
      - x3.0的IMU参数
      - /3.0的IMU参数
  - 评估1: 针对仿真数据无噪声:
    - 将无噪声数据集imu-pose.txt 文件输入到vins 系统中, imu 噪声的yaml 参数如下:

```
#imu parameters The more accurate parameters you provide , the
better performance
acc_n : 0.019 # accelerometer measurement noise standard deviation
. #0.2 0.04
gyr_n : 0.015 # gyroscope measurement noise standard deviation .
#0.05 0.004
acc_w : 0.0001 # accelerometer bias random work noise standard
deviation . #0.02
gyr_w : 1.0e-5 # gyroscope bias random work noise standard
deviation . #4.0e-5
g_norm : 9.81007 # gravity magnitude
```

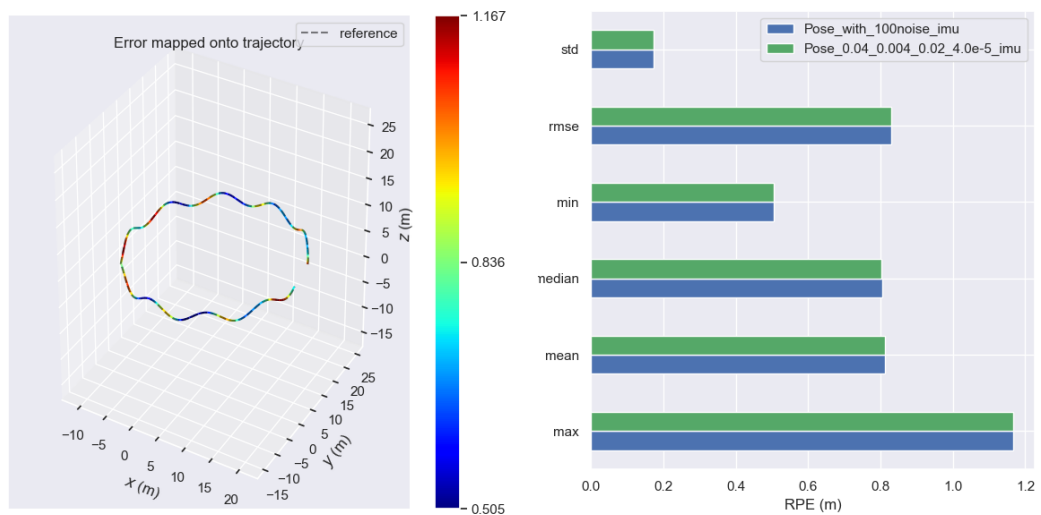
运行程序, 得到轨迹:

- 针对无噪声imu 数据集进行分析, 修改yaml 配置文件中的imu 标定噪声及bias 参数(扩大), 并分析其对结果的影响:

```
#imu parameters The more accurate parameters you provide , the
better performance
acc_n : 1.90 # accelerometer measurement noise standard deviation .
#0.2 0.04
gyr_n : 1.50 # gyroscope measurement noise standard deviation .
#0.05 0.004
acc_w : 0.01 # accelerometer bias random work noise standard
deviation . #0.02
gyr_w : 1.0e-3 # gyroscope bias random work noise standard
deviation . #4.0e-3
g_norm : 9.81007 # gravity magnitude
```

运行程序, 得到轨迹:

- 两次的标定参数结果与真值误差对比结果:



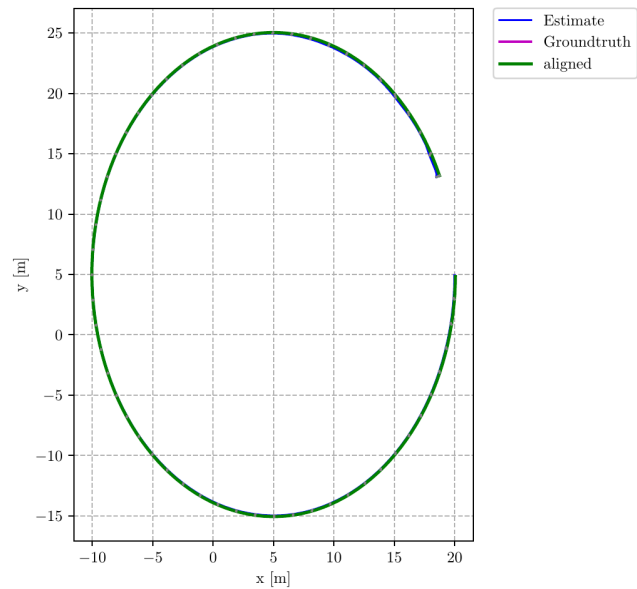
(a) 原始yaml的imu标定参数结果与真值误差对比结果

(b) 将 imu 标定参数扩大100倍, 两者与真值误差结果对比

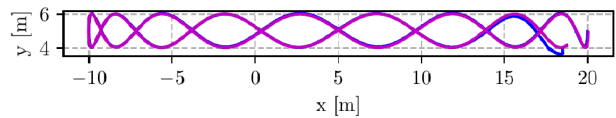
不带噪声IMU数据两种情况下与真值误差数值结果如下:

name	max/m	mean/m	rmse/m	sse	轨迹总长/m
原始 yaml	1.167762	0.812515	0.830518	128.985	110.266

name	max/m	mean/m	rmse/m	sse	轨迹总长/m
imu参数100倍	1.167276	0.812294	0.830292	128.915	110.239



无噪声轨迹 top-psyaw



无噪声轨迹 side-psyaw

结果分析: 针对无噪声的IMU数据集, 由于IMU信息和视觉信息两者接近真值, IMU标定噪声参数对结果影响不大.

- 评估2: 针对仿真数据有噪声:
  - 将有噪声数据集imu-pose\_noise.txt 文件输入到vins 系统中, 原始的有噪声的imu数据如下:

```

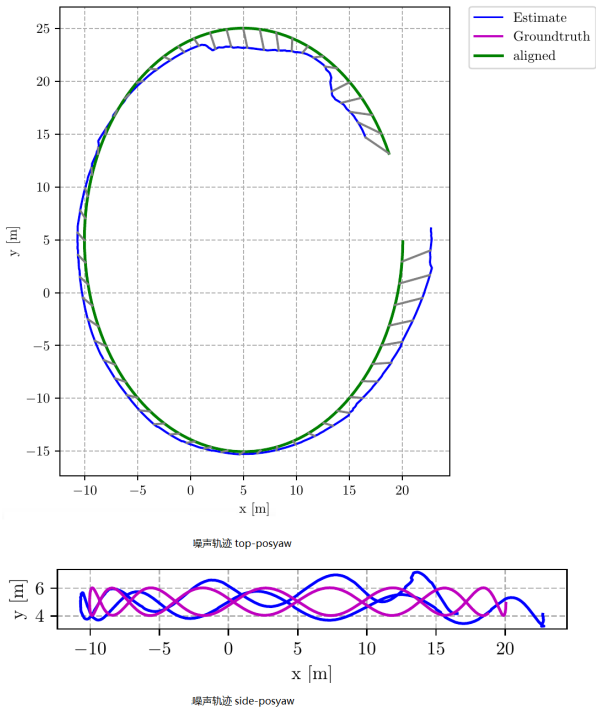
acc_n : 0.019 # accelerometer measurement noise standard deviation .
#0.2 0.04
gyr_n : 0.015 # gyroscope measurement noise standard deviation . #0.05
0.004
acc_w : 0.0001 # accelerometer bias random work noise standard
deviation . #0.02
gyr_w : 1.0e-5 # gyroscope bias random work noise standard deviation .
#4.0e-5
g_norm : 9.81007 # gravity magnitude

```

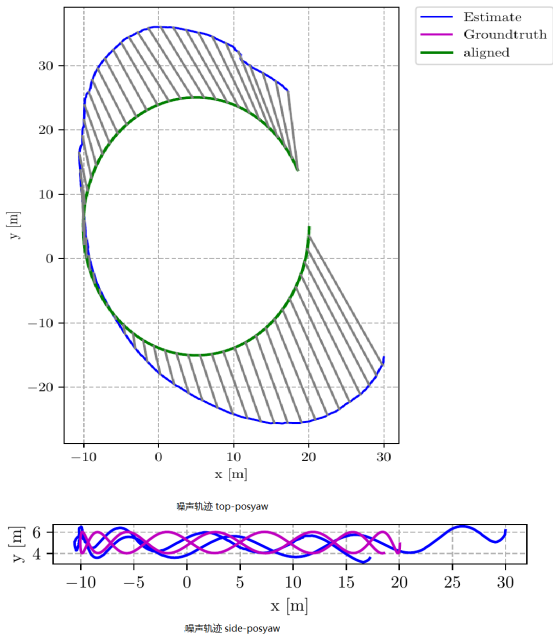
- 修改如上噪声数据, 多次绘制轨迹, 对多次轨迹数据的比较如下表所示:

name	max/m	mean/m	rmse/m	sse	轨迹总长/m
A 有噪声 (原始)	3.203756	1.340572	0.468402	1.510936	110.179
B 有噪声 (x3)	22.513000	9.931972	0.458918	11.429269	126.996
C 有噪声 (/3.0)	1.15201	0.345822	0.074846	0.401324	112.373

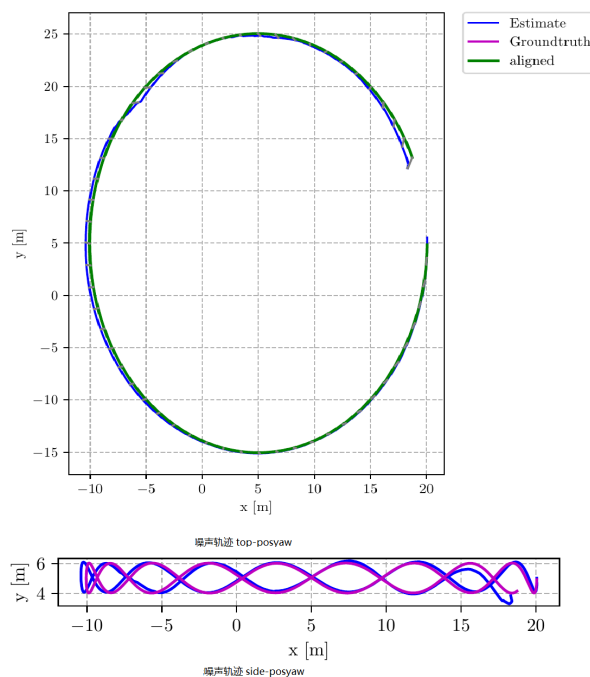
■ A 有噪声 (原始): acc-n: 0.019, gyr-n: 0.015, acc-w: 0.0001, gyr-w: 1.0e-5



■ B 有噪声 (x3): acc-n: 0.019x3, gyr-n: 0.015x3, acc-w: 0.0001x3, gyr-w: 1.0e-5x3



- C 有噪声 (/3): acc-n: 0.019/3.0, gyr-n: 0.015/3.0, acc-w: 0.0001/3.0, gyr-w: 1.0e-5/3.0



#### ■ 结果分析:

- 需要注意的是, yaml 参数和仿真程序的噪声设置一致, 并不随上述IMU噪声变化而变化
- 经过量化分析可以看到, 当噪声为x3时, 轨迹漂移严重, 误差较大; 而当噪声为/3.0时, 轨迹接近真实值;

#### ● 以下为第一版既有内容:

#### ● 整体思路过程:

- **数据文件准备:** 整个作业要用第二课代码生成数据文件, 第二课程序运行过程参见第二课文档, 下面重点讲解第二课生成后的数据文件中, 需要用到的三类文件:

- **Image文件:** Image数据都在 bin/keyframe 下, all\_lines\_nnn.txt 文件(nnn代表数字), 根据第二课代码 utilities.cpp 中的 save\_features() 函数看, 每个文件中都存放着image的points的坐标, 格式为 x y z 1 u v , 其中 (x, y, z) 为Image的 pose, (u, v) 为归一化平面上的像素坐标(请参见第二课代码, 在输出时, 像素坐标已经做了归一化);
- **IMU文件:** IMU数据在bin目录下的imu\_pose.txt 和 imu\_pose\_noise.txt, 一个是不带噪声的, 一个是带噪声的; 根据第二课代码utilities.cpp中的save\_Pose()函数看, 每个文件中保存的是IMU的测量数据, 格式为 time q.w q.x q.y q.z t0 t1 t2 gyro0 gyro1 gyro2 acc0 acc1 acc2 , 分别是 时间戳 旋转 平移 gyro acc;
- **Image文件对应的时间戳文件:** bin/keyframe 中的文件中的 points 行, 并不带有时间戳信息, 第二课的代码是仿照TUM的方式, 使用 bin/cam\_pose.txt 这个Image位姿文件来记录时间戳, 它的每一行都对应到 bin/keyframe 目录中的一个文件, 所以这个文件也需要读取, 读取它的第一列时间戳数据作为 bin/keyframe 中对应文件的时间戳;

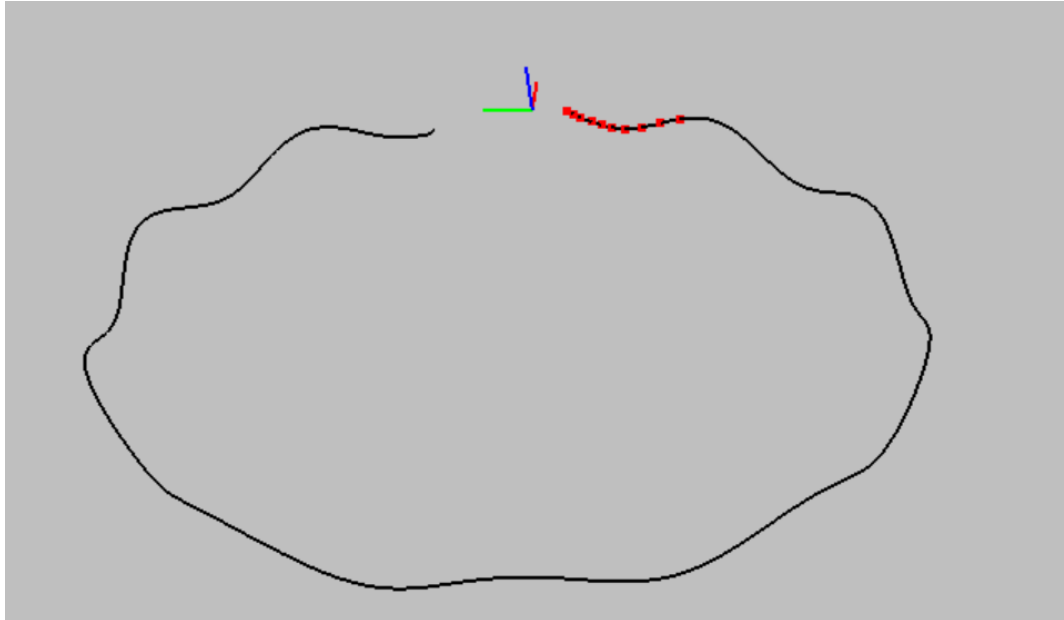
- **数据文件读取编程:** 文件读取好之后, 需要读取并解析文件内容, 分为IMU和Image两类:

2. 第七课的代码是读取EuRoC的数据集的IMU数据以及Image数据的, 要改成读取上述的文件;
3. 第七课代码中是读取Image图片, 然后提取角点进行跟踪, 但第二课生成的已经是特征点点了, 因此要改这部分代码, 直接读取特征点;

- **配置文件准备:** 要把第二课的参数设置, 变为第七课的 yaml 格式, 具体说明如下:

4. 第二课的一些外参的配置数据是定义在 `param.h` 文件中的, 例如IMU频率, noise等等
  5. 第七课是使用 `yaml` 文件保存参数的;
  6. 因此要根据第二课的参数写一份 `yaml` 文件, 以保证第七课代码按照这些外参来处理数据.
- **执行结果:** 先展示无噪声和有噪声的执行结果 (针对有噪声列举了三组参数的情况), 然后再详细讲解开发过程;

- **无噪声**情况下, 执行截图如下:

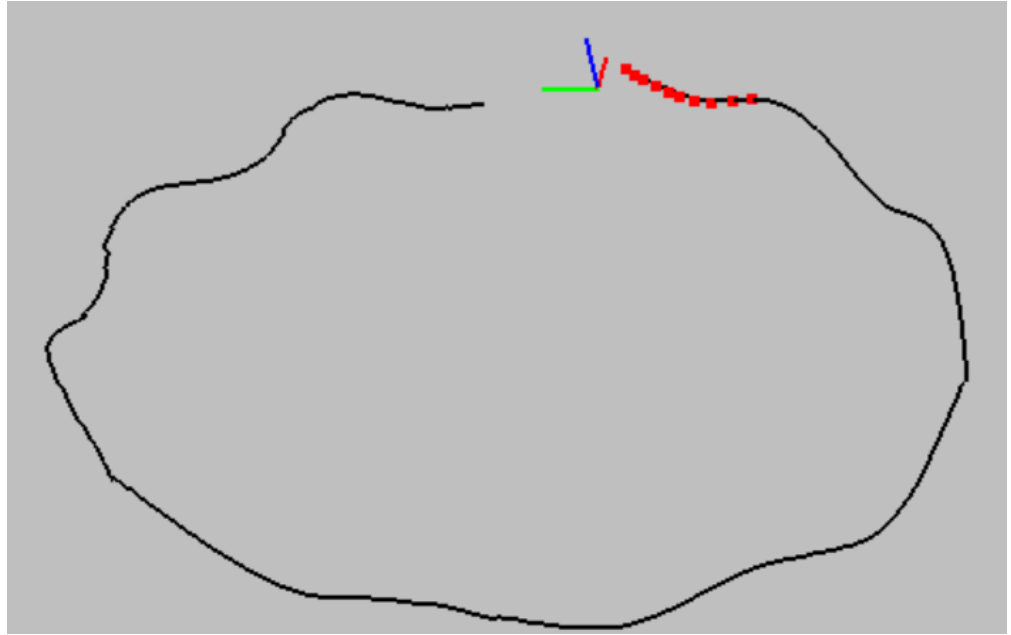


```
iter: 0 , chi= 6.140108 , Lambda= 500000.000000
iter: 1 , chi= 5.228841 , Lambda= 304965.555250
iter: 2 , chi= 5.222351 , Lambda= 101655.185083
iter: 3 , chi= 5.215390 , Lambda= 33885.061694
iter: 4 , chi= 5.203370 , Lambda= 11295.020565
iter: 5 , chi= 5.188376 , Lambda= 3765.006855
iter: 6 , chi= 5.177777 , Lambda= 1255.002285
iter: 7 , chi= 5.172609 , Lambda= 418.334095
iter: 8 , chi= 5.169391 , Lambda= 139.444698
iter: 9 , chi= 5.167681 , Lambda= 46.481566
problem solve cost: 85.824247 ms
makeHessian cost: 58.720183 ms
----- update bprior -----
      before: 6446.604701
              4.005215
      after: 6643.459320
              3.989154
edge factor cnt: 0
1 BackEnd processImage dt: 95.481682 stamp: 19.966705 p_wi: -0.021015
-2.066730 -0.089363
```

- **有噪声**情况(1): 采用第二课默认的噪声参数
  - 噪声:

```
acc_n: 0.019
gyr_n: 0.015
acc_w: 0.0001
gyr_w: 1.0e-5
```

- 截图如下:



```
iter: 0 , chi= 473.873120 , Lambda= 500000.000000
iter: 1 , chi= 451.735331 , Lambda= 166666.666667
iter: 2 , chi= 447.013107 , Lambda= 55555.555556
iter: 3 , chi= 445.226435 , Lambda= 18518.518519
iter: 4 , chi= 443.881027 , Lambda= 7181.348854
iter: 5 , chi= 442.622470 , Lambda= 4787.565902
iter: 6 , chi= 441.588852 , Lambda= 3191.710602
iter: 7 , chi= 440.589629 , Lambda= 2127.807068
iter: 8 , chi= 439.489866 , Lambda= 1418.538045
iter: 9 , chi= 438.199689 , Lambda= 945.692030
problem solve cost: 104.742582 ms
      makeHessian cost: 77.043999 ms
----- update bprior -----
      before: 34350.226414
              43.123175
      after: 39869.763292
              42.930783
edge factor cnt: 0
1 BackEnd processImage dt: 121.957470 stamp: 19.966705 p_wi:
  1.072711 -1.385537  0.398769
```

- 有噪声情况(2): 相比第二课参数, gyr\_w 变大一个数量级, gyr\_n从0.015增大为0.025
  - 噪声:

```
acc_n: 0.019
gyr_n: 0.025
acc_w: 0.0001
gyr_w: 1.0e-4
```

○ 截图如下:



```
iter: 0 , chi= 847.428429 , Lambda= 500000.000000
  iter: 1 , chi= 672.488878 , Lambda= 185305.432647
  iter: 2 , chi= 671.576133 , Lambda= 988295.640783
  iter: 3 , chi= 671.377867 , Lambda= 5270910.084175
  iter: 4 , chi= 671.359124 , Lambda= 7027880.112234
  iter: 5 , chi= 671.346347 , Lambda= 37482027.265247
  iter: 6 , chi= 671.207238 , Lambda= 24988018.176831
  iter: 7 , chi= 671.195249 , Lambda= 33317357.569109
  iter: 8 , chi= 671.188364 , Lambda= 44423143.425478
  iter: 9 , chi= 671.183721 , Lambda= 59230857.900638
problem solve cost: 132.615362 ms
  makeHessian cost: 72.506770 ms
  ----- update bprior -----
    before: 32320.186723
           66.143780
    after: 27124.072584
           65.288490
  edge factor cnt: 0
  1 BackEnd processImage dt: 143.417885 stamp: 19.966705 p_wi:
-3.198176 -7.918088 1.
```

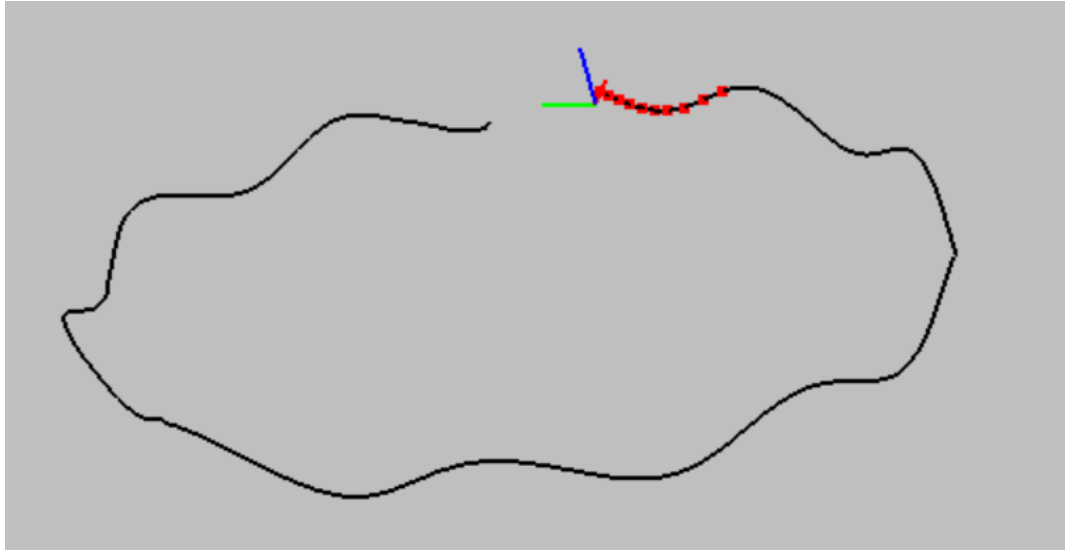
- 有噪声情况(3): 相比第二课参数, gyr\_w 变大一个数量级, gyr\_n从0.015减小为0.005, acc\_n从0.019减小为 0.001, acc\_w从 0.0001 增大为 0.0002;

○ 噪声:

```
acc_n: 0.001
gyr_n: 0.005
acc_w: 0.0002
gyr_w: 1.0e-5
```



- 截图如下:



```

iter: 0 , chi= 415.169866 , Lambda= 500000.000000
  iter: 1 , chi= 408.441531 , Lambda= 333333.333333
  iter: 2 , chi= 408.440760 , Lambda= 728177777.777779
  iter: 3 , chi= 408.439266 , Lambda= 310689185185.185181
  iter: 4 , chi= 408.439053 , Lambda= 1657008987654.320801
  iter: 5 , chi= 408.439020 , Lambda= 8837381267489.710938
sqrt(currentChi_) <= stopThresholdLM_
problem solve cost: 85.169250 ms
  makeHessian cost: 37.999698 ms
----- update bprior -----
      before: 99876.230406
              34.877407
      after: 95316.902088
              33.176471
edge factor cnt: 0
1 BackEnd processImage dt: 97.560266 stamp: 19.966705 p_wi: 0.290740
-0.361437 0.530511

```

- 详细过程讲解:

- 数据文件准备

- 把第二课数据(即bin下的数据文件)拷贝到第七课的 bin/data 下, 拷贝后如下:

```

hadoop@ubuntu:~/Documents/vins_sys_code/bin/data$ ls
all_points.txt      data_gen            imu_int_pose.txt
keyframe
cam_pose_tum.txt    demo.png            imu_pose_noise.txt
cam_pose.txt        imu_int_pose_noise.txt  imu_pose.txt

```

- 设置配置文件

- 根据第二课的 param.h 中的参数的值, 参考第七课中 config/euroc\_config.yaml 的格式和内容, 编写自己的 simulation\_config.yaml 文件, 内容如下, 刷新的有 图片尺寸 image\_width , image\_height , 相机内参 fx, fy, cx, cy, 偏差与噪声 acc\_n, gyr\_n, acc\_w, gyr\_w, 其他数值基本可以复用 euroc\_config.yaml 的;

```
%YAML:1.0
```

```

#common parameters
imu_topic: "/imu0"
image_topic: "/cam0/image_raw"
output_path: "/home/hadoop/Documents/vins_sys_code/result"

#camera calibration
model_type: PINHOLE
camera_name: camera
image_width: 640
image_height: 640
distortion_parameters:
  k1: 0
  k2: 0
  p1: 0
  p2: 0
projection_parameters:
  fx: 460
  fy: 460
  cx: 255
  cy: 255

# Extrinsic parameter between IMU and Camera.
estimate_extrinsic: 0  # 0 Have an accurate extrinsic parameters.
                        # 1 Have an initial guess about extrinsic
                        # 2 Don't know anything about extrinsic
                        # parameters. We will optimize around your initial guess.
                        # parameters. You don't need to give R,T. We will try to calibrate
                        # it. Do some rotation movement at beginning.
#If you choose 0 or 1, you should write down the following matrix.
#Rotation from camera frame to imu frame, imu^R_cam
extrinsicRotation: !!opencv-matrix
  rows: 3
  cols: 3
  dt: d
  data: [0, 0, -1,
        -1, 0, 0,
        0, 1, 0]
#Translation from camera frame to imu frame, imu^T_cam
extrinsicTranslation: !!opencv-matrix
  rows: 3
  cols: 1
  dt: d
  data: [0.05, 0.04, 0.03]

#feature traker paprameters
max_cnt: 150          # max feature number in feature tracking
min_dist: 30          # min distance between two features
freq: 30              # frequence (Hz) of publish tracking
result. At least 10Hz for good estimation. If set 0, the frequence
will be same as raw image
F_threshold: 1.0      # ransac threshold (pixel)
show_track: 1         # publish tracking image as topic
equalize: 1           # if image is too dark or light, trun on
equalize to find enough features
fisheye: 0            # if using fisheye, trun on it. A circle
mask will be loaded to remove edge noisy points

```

```

#optimization parameters
max_solver_time: 0.04 # max solver iteration time (ms), to
guarantee real time
max_num_iterations: 8 # max solver iterations, to guarantee real
time
keyframe_parallax: 10.0 # keyframe selection threshold (pixel)

#imu parameters          The more accurate parameters you provide, the
better performance
acc_n: 0.019             # accelerometer measurement noise standard
deviation. #0.2 0.04
gyr_n: 0.015             # gyroscope measurement noise standard
deviation. #0.05 0.004
acc_w: 0.0001            # accelerometer bias random work noise
standard deviation. #0.02
gyr_w: 1.0e-5            # gyroscope bias random work noise standard
deviation. #4.0e-5
g_norm: 9.81007          # gravity magnitude

#loop closure parameters
loop_closure: 0           # start loop closure
load_previous_pose_graph: 0 # load and reuse previous pose
graph; load from 'pose_graph_save_path'
fast_relocalization: 0    # useful in real-time and large
project
pose_graph_save_path: "/home/weihao/Desktop/From Zero to One
VIO/homework/vins_sys_code/result/" # save and load path

#unsynchronization parameters
estimate_td: 0             # online estimate time offset
between camera and imu
td: 0.0                    # initial value of time offset.
unit: s. readed image clock + td = real image clock (IMU clock)

#rolling shutter parameters
rolling_shutter: 0          # 0: global shutter camera, 1:
rolling shutter camera
rolling_shutter_tr: 0        # unit: s. rolling shutter read
out time per frame (from data sheet).

#visualization parameters
save_image: 0              # save image in pose graph for
visualization prupose; you can close this function by setting 0
visualize_imu_forward: 0    # output imu forward propogation to
achieve low latency and high frequency results
visualize_camera_size: 0.4  # size of camera marker in RVIZ

```

#### ○ 数据文件读取编程:

- 整个程序编写涉及新增或修改如下几个文件:
  - test/run\_simulation.cpp : 新增, 程序入口, 仿照 run\_euroc.cpp 编写, 用于触发读取imu,image数据, 并启动多个线程进行优化和图形绘制;
  - src/System.cpp: 修改, 重载其中的 PubImageData() 函数, 读取 keyframe 下的文件中的数据, 做适当转换后, 保存到变量 feature\_points 中;
- test/run\_simulation.cpp 代码如下:

```

//
// Author: Arthur <shihezichen@live.cn>
// Date: 2019.08.10
// Description:
//     Read the IMU and Image data from data files and do the
// optimization of them , then draw the final trajectory.

#include <unistd.h>
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <iostream>
#include <thread>
#include <iomanip>

#include <cv.h>
#include <opencv2/opencv.hpp>
#include <highgui.h>
#include <eigen3/Eigen/Dense>
#include "System.h"

using namespace std;
using namespace cv;
using namespace Eigen;

const int nDelayTimes = 2;

std::shared_ptr<System> pSystem;//宏定义整个VIO系统类

// 获取IMU数据
void PubImuData(string sData_path)
{
    string sImu_data_file = sData_path + "imu_pose.txt";
    cout << "1 PubImuData start sImu_data_file: " <<
sImu_data_file << endl;
    ifstream fsImu;
    fsImu.open(sImu_data_file.c_str());
    if (!fsImu.is_open())
    {
        cerr << "Failed to open imu file! " << sImu_data_file <<
endl;
        return;
    }//读取文件

    std::string sImu_line;
    double dStampNSec = 0.0;//时间戳
    Vector3d vAcc;//加速度
    Vector3d vGyr;//陀螺仪
    Eigen::Quaterniond q;
    Eigen::Vector3d t;
    while (std::getline(fsImu, sImu_line) && !sImu_line.empty()) //
read imu data
    {
        std::istringstream ssImuData(sImu_line);

```

```

        ssImuData >> dStampNSec >> q.w() >> q.x() >> q.y() >> q.z()
>> t(0) >> t(1) >> t(2) >> vGyr.x() >> vGyr.y() >> vGyr.z() >>
vAcc.x() >> vAcc.y() >> vAcc.z();
        //cout << "Imu t: " << fixed << dStampNSec << " gyr: " <<
vGyr.transpose() << " acc: " << vAcc.transpose() << endl;
        pSystem->PubImuData(dStampNSec, vGyr, vAcc);
        usleep(5000*nDelayTimes); //将调用usleep函数的线程挂起一段时间,单
位是微秒
    }
    fsImu.close();
}

// 获取图像数据
void PubImageData(string sData_path)
{
    // 由于 keyframe 文件夹下的文件中每一行的image的point没有时间戳,
    // 因此通过读取图像位姿文件cam_pose_tum.txt, 它对应行的第一列时间戳可作
    为point的时间戳
    string sImage_file = sData_path + "cam_pose_tum.txt";
    cout << "1 PubImageData start sImage_file image timestamp: " <<
sImage_file << endl;
    ifstream fsImage; //图像位姿文件,这里主要是需要图像的时间戳
    fsImage.open(sImage_file.c_str());
    if (!fsImage.is_open())
    {
        cerr << "Failed to open image file! " << sImage_file <<
endl;
        return;
    }

    std::string sImage_line;
    double dStampNSec;
    int file_count = 0; //标记keyframe下对应的feature文件

    // cv::namedWindow("SOURCE IMAGE", CV_WINDOW_AUTOSIZE);
    while (std::getline(fsImage, sImage_line) &&
!sImage_line.empty())
    {
        std::istringstream ssImuData(sImage_line); //读取图像位姿文件,
        为了得到时间戳
        ssImuData >> dStampNSec;
        //cout << "Image t : " << fixed << dStampNSec << endl;
        std::stringstream featurePath;
        featurePath<<sData_path<<"keyframe/all_points_"
<<file_count<<".txt";
        string sfeaturePath = featurePath.str();
        //cout<<sfeaturePath<<endl;
        pSystem->PubImageData(dStampNSec, sfeaturePath);
        usleep(50000*nDelayTimes);
        file_count++;
    }
    fsImage.close();
}

int main(int argc, char **argv)
{
    if(argc != 3)
    {

```

```

        cerr << "./run_simulation PATH_TO_FOLDER/bin/
PATH_TO_CONFIG/config \n"
        << "For example: ./run_simulation ./ ../config/"<<
endl;
        return -1;
    }
    // 保存数据的文件夹, 例如 bin/data/
    string sData_path = argv[1];
    string sConfig_path = argv[2];
    pSystem.reset(new System(sConfig_path)); //核心类
    // std::thread 初始化为函数, 表示直接执行函数, 初始化为类函数, 则后边应该跟着类
    std::thread thd_BackEnd(&System::ProcessBackEnd, pSystem); //后端
    优化最重要的

    // sleep(5);
    std::thread thd_PubImuData(PubImuData, sData_path); //获取IMU数据
    线程

    std::thread thd_PubImageData(PubImageData, sData_path); //获取图像
    数据线程

    std::thread thd_Draw(&System::Draw, pSystem); //显示轨迹线程

    // 数据运行完就把线程停掉, 但画图和主线程仍然在运行
    thd_PubImuData.join();
    thd_PubImageData.join();

    thd_BackEnd.join();
    thd_Draw.join();

    cout << "main end... see you ... <Ctrl>+C to exit." << endl;
    return 0;
}

```

- src/System.cpp 中重载的函数 PubImageData 代码如下:

```

// 重载图像获取函数, 直接获取特征点, 并赋值给 feature_points
void System::PubImageData(double dStampSec, string &feature_name)
{
    // 特征点文件操作并判断文件是否能够正常打开
    std::ifstream f;
    f.open(feature_name.c_str());
    if(!f.is_open())
    {
        std::cerr << "can not open image feature file:"
        <<feature_name<<std::endl;
        return;
    }

    //flag赋值
    if (!init_feature)
    {
        // 跳过第一张图片, 因为不能计算光流速度
        cout << "1 PubImageData skip the first detected feature,
        which doesn't contain optical flow speed" << endl;
    }
}

```

```

        init_feature = 1;
        return;
    }

    if (first_image_flag) //第一张图片
    {
        cout << "2 PubImageData first_image_flag" << endl;
        first_image_flag = false;
        first_image_time = dStampSec;
        last_image_time = dStampSec;
        return;
    }
    // detect unstable camera stream
    // 检查当前帧和上一帧图像的时间戳,假如太大则lost
    if (dStampSec - last_image_time > 1.0 || dStampSec <
last_image_time)
    {
        cerr << "3 PubImageData image discontinue! reset the
feature tracker!" << endl;
        first_image_flag = true;
        last_image_time = 0;
        pub_count = 1;
        return;
    }
    last_image_time = dStampSec;

    PUB_THIS_FRAME = true;
    TicToc t_r;
    // 新建一帧图像特征来保存模拟数据保存好的特征点
    if(PUB_THIS_FRAME)
    {
        pub_count++;
        shared_ptr<IMG_MSG> feature_points(new IMG_MSG());
        feature_points->header = dStampSec;
        int pts_id = 0;
        // 逐行读取特征点
        while(!f.eof())
        {
            std::string s;
            std::getline(f,s);
            if(! s.empty())
            {
                //将读取到的一行数据赋值给ss(x,y,z,1,u,v)
                std::stringstream ss;
                ss<<s;
                Eigen::Vector4d point_3d;
                double u, v;
                ss >> point_3d(0) >> point_3d(1) >> point_3d(2) >>
point_3d(3) >> u >> v;
                // 其中(u,v)已经是归一化平面上的坐标,且非畸变
                double x = u;
                double y = v;
                double z = 1.0;
                // 通过畸变公式,求得畸变下的坐标(即成像平面上的坐标)
                double image_u = FOCAL_LENGTH * u + COL/2.0;
                double image_v = FOCAL_LENGTH * v + ROW/2.0;
                // 保存归一化平面上的坐标

```

```

        feature_points->points.push_back(Vector3d(x, y,
z));

        // pts_id 在本例中暂时不涉及, 被置为0
        feature_points->id_of_point.push_back(pts_id);
        // 保存成像平面上的坐标
        feature_points->u_of_point.push_back(image_u);
        feature_points->v_of_point.push_back(image_v);
        // 速度在本例中暂时不涉及, 被置为0
        feature_points->velocity_x_of_point.push_back(0);
        feature_points->velocity_y_of_point.push_back(0);
    }
    pts_id++;
}

if (!init_pub)
{
    cout << "4 PubImage init_pub skip the first image!" <<
endl;

    init_pub = 1;
}
else
{
    m_buf.lock();
    feature_buf.push(feature_points); //图像特征点的buffer
    // cout << "5 PubImage t : " << fixed <<
feature_points->header
    //      << " feature_buf size: " << feature_buf.size()
<< endl;

    m_buf.unlock();
    con.notify_one();
}
}
}

```

#### ■ 编译运行:

- 由于需要得到带有噪声和不带噪声的两个图形, 因此需要针对带有噪声和不带噪声两种情况, 修改 `run_simulation.cpp` 代码中 `PubImuData()` 函数中的 `string sImu_data_file = sData_path + "imu_pose_noise.txt";` 使之分别读取 `imu_pose.txt` 和 `imu_pose_noise.txt`, 然后分别编译和执行;
- 整个编译执行过程如下:

```

mkdir build
cd build
cmake ..
make -j4
cd ../bin
./run_simulation ./data/ ../config/

```