

INERTIAL MEASUREMENT UNIT CALIBRATION AND ORIENTATION TRACKING



Presented to

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Outline

Introduction

- Motivations
- Objective
- Related work

Methods

- IMU Calibration (LSM6DSM, iPhone 7)
- Orientation Tracking
 - Extended Kalman filtering
 - High & low pass filtering
 - Mahony filter

Experiment Result

Conclusion and Further Work

Reference



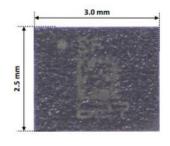
Introduction

Motivation

- Orientation tracking is one of most fundamental problems for many problem like navigation and mapping
- IMU is the most popular solution for the attitude estimation

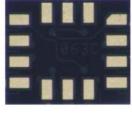
Objectives

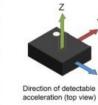
- Explore the calibration process without external equipment
- Compare three basic sensor fusion algorithm



Package Top View

Package Side View





Package Bottom View



LSM6DSM (Credit:

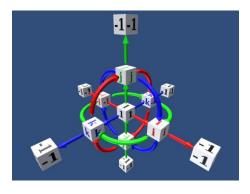
http://www.mems.me/mems/device_analysis_201612/3801.h tml)

Introduction

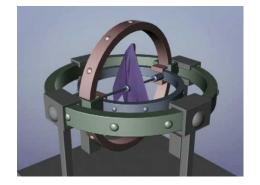
Related work

Rotation representation Quaternion

- Less computational overhead
- No gimbal lock



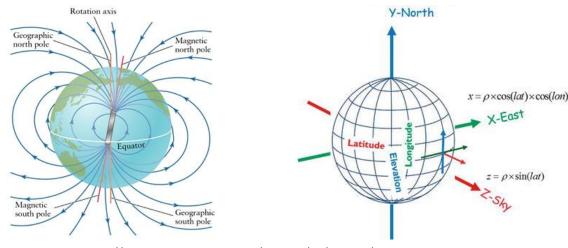
Credit: https://www.reddit.com/r/math/comm ents/42yc0i/visualizing_quaternions/



Credit: https://www.youtube.com/watch?v=zc 8b2Jo7mno

Magnetometer

Vm: Calibration for north direction



Credit: https://www.uavnavigation.com/support/kb/general/general-system-info/introduction-magnetometers

Calibration

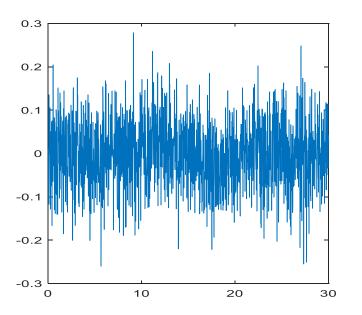
- Types of error (Bias)
 - Non-accurate scaling
 - Axis misalignments
 - Bias instability (non zero biases)

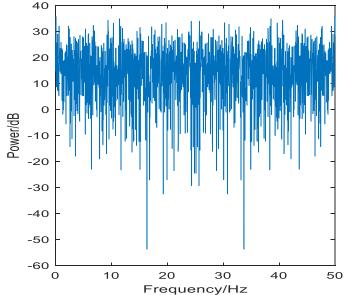
- Methods to measure the IMU errors & biases
 - The averaged gyro signal = Bias (static)
 - Sum of (bias + projection of G) = Local precise gravity (rotating)
 - Magnetic off-center circle = Hard & soft iron distortions

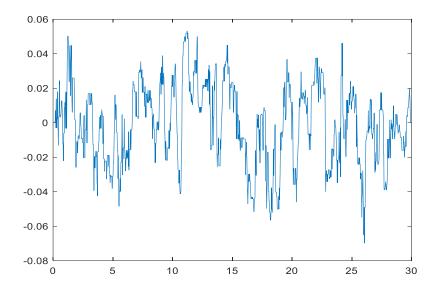


Calibration (Gyroscope)

- Bias instabilities (low-cost embedded IMU sensor)
- Calibration procedure for the Gyroscope (static)
 - I. Gaussian filtering (identify the noise type)
 - II. Median filtering for nonlinear signal (non-zero, un-Gaussian noises)
 - III. Obtain the error compensation function (parameters)





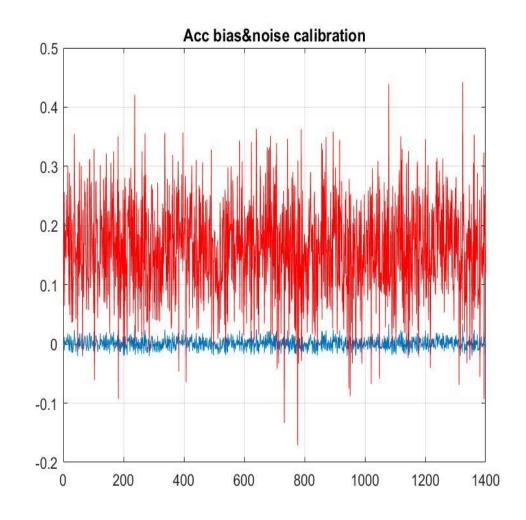


Calibration (Accelerometer)

- Variable biases and measurement noise
- Misalignments
- Scale factor errors

Basic calibration framework for accelerometer

- I. Find the exact value of local gravity
- II. Cost function used to estimate accelerometers'parameters (sum of (Bias + projection g local g))
- III. Employ the Levenberg-Marquardt (LM) algorithm to minimize the cost function
- IV. Obtain the offset parameters

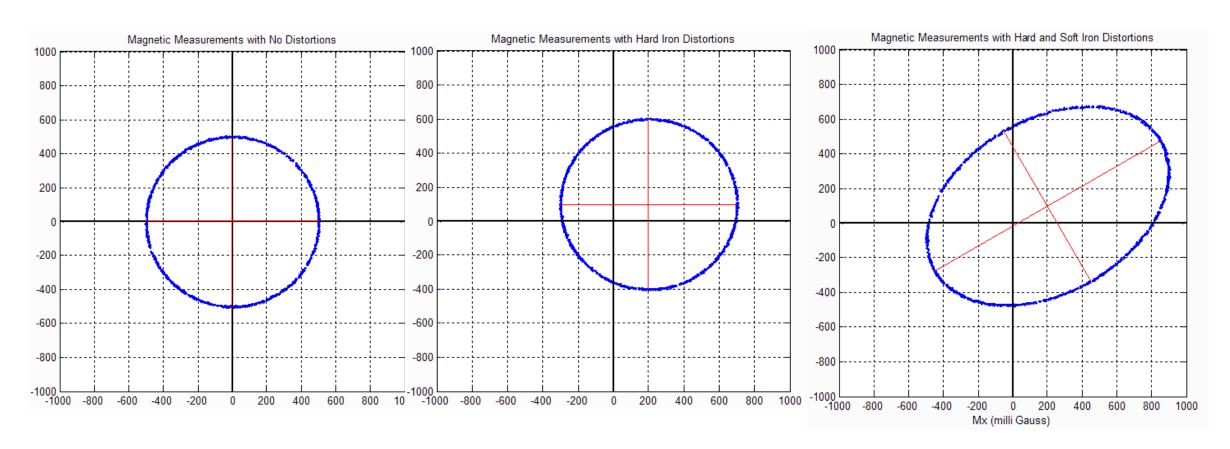


Calibration (Magnetometer)



Hard iron distortions

Soft & Hard iron distortions



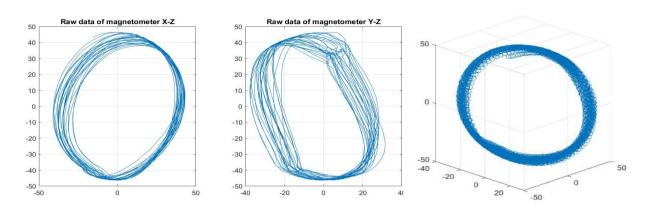
Hard iron distortions have a much larger contribution to the total uncorrected error than soft iron distortions

Calibration (Magnetometer)

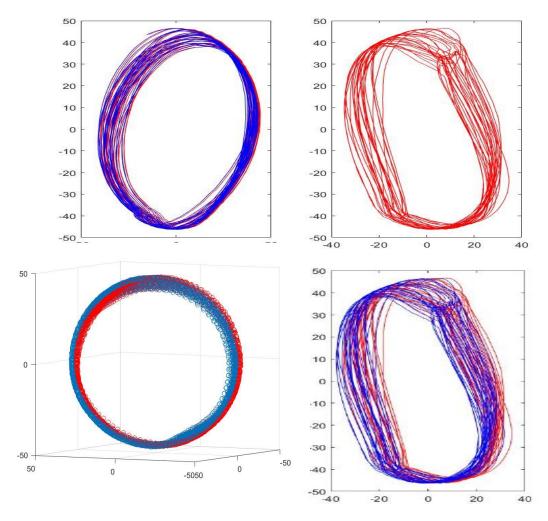
Calibration procedure for the magnetometer

- I. Sweep the device through eight figure pattern
- II. Draw the magnetic measurements of raw data
- III. Built the error model and obtain the offset distance
- IV. Optimize the cost function
- V. Obtain the final triaxial compensation parameters

Magnetic measurement of raw data ('8' rotating)

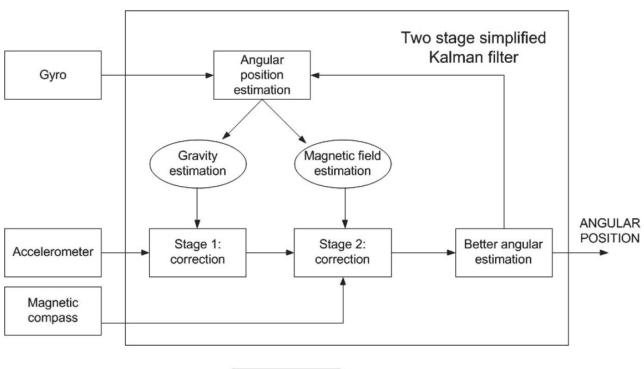


Magnetic measurement after calibration



Double Stage Kalman Filter

Double Stage Kalman Filter



$$\begin{split} \hat{q}_{k}^{-} &= A_{k} \hat{q}_{k-1} \\ P_{k}^{-} &= A_{k} P_{k-1} A_{k}^{T} + Q_{k-1} \\ K_{k} &= P_{k}^{-} H_{k}^{T} \big(H_{k} P_{k}^{-} H_{k}^{T} + V_{k} R_{k} V_{k}^{T} \big)^{-1} \\ \hat{q}_{k} &= \hat{q}_{k}^{-} + K_{k} \big(z_{k} - h(\hat{q}_{k}^{-}, 0) \big) \\ P_{k} &= (I - K_{k} H_{k}) P_{k}^{-} \end{split}$$

$$h_1(q_k) = \hat{g} = R_n^b \begin{bmatrix} 0 \\ 0 \\ |g| \end{bmatrix} = |g| \begin{bmatrix} 2q_1q_3 - 2q_0q_2 \\ 2q_0q_1 + 2q_2q_3 \\ q_0^2 - q_1^2 - q_2^2 + q_3^2 \end{bmatrix}$$

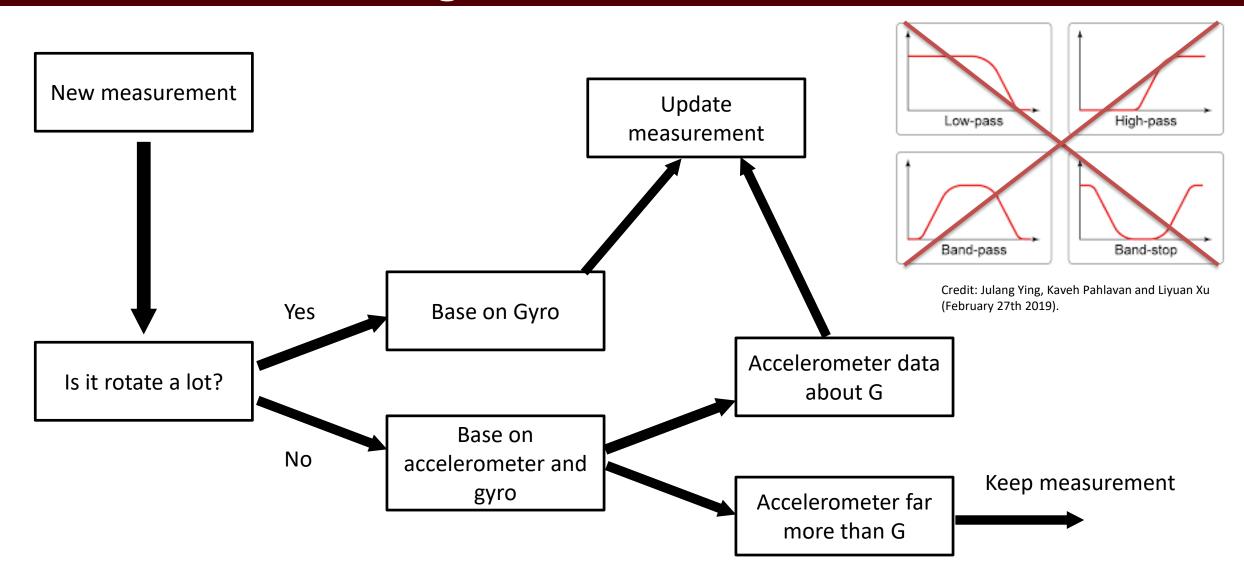
$$R_n^b = \begin{bmatrix} q_0^2 + q_1^2 - q_2^2 - q_3^2 & 2q_1q_2 + 2q_0q_3 & 2q_1q_3 - 2q_0q_2 \\ 2q_1q_2 - 2q_0q_3 & q_0^2 - q_1^2 + q_2^2 - q_3^2 & 2q_2q_3 + 2q_0q_1 \\ 2q_1q_3 + 2q_0q_2 & 2q_2q_3 - 2q_0q_1 & q_0^2 - q_1^2 - q_2^2 + q_3^2 \end{bmatrix}$$

$$H_{k1} = \frac{\partial h_{1[i]}}{\partial q_{[j]}} = \begin{bmatrix} -2q_2 & 2q_3 & -2q_0 & 2q_1\\ 2q_1 & 2q_0 & 2q_3 & 2q_2\\ 2q_0 & -2q_1 & -2q_2 & 2q_3 \end{bmatrix}$$

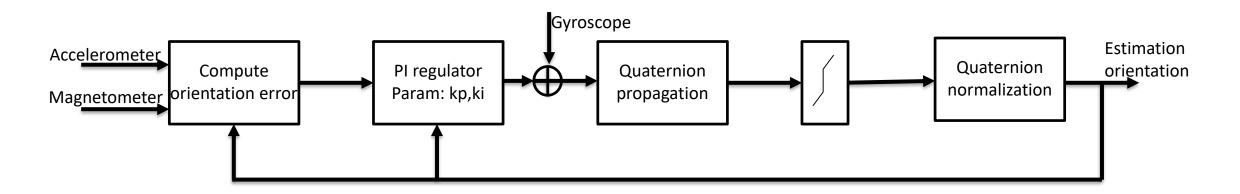
$$h_2(q_k) = \widehat{m} = R_n^b \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 2q_1q_2 + 2q_0q_3 \\ q_0^2 - q_1^2 - q_2^2 - q_3^2 \\ 2q_2q_3 - 2q_0q_1 \end{bmatrix}$$

$$H_{k2} = \frac{\partial h_{2[i]}}{\partial q_{[j]}} = \begin{bmatrix} 2q_3 & 2q_2 & 2q_1 & 2q_0 \\ 2q_0 & -2q_1 & -2q_2 & -2q_3 \\ -2q_1 & -2q_0 & 2q_3 & 2q_2 \end{bmatrix}.$$

High Low Pass Filter



Mahony Filter



Correction to the gyroscope

$$k_p * e_m + k_i * e_i$$

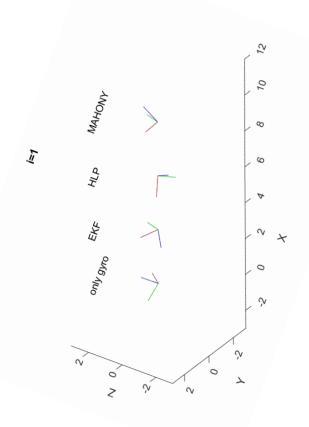
e_m: sensor error of the gyroscope

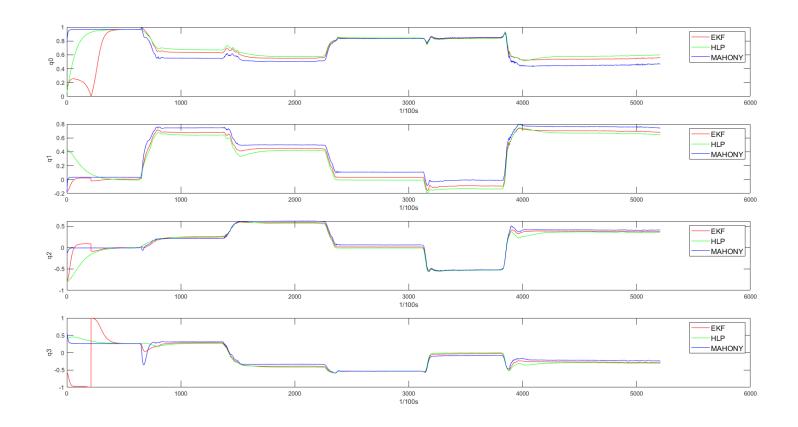
e_i: integral error

Kp: proportional gain

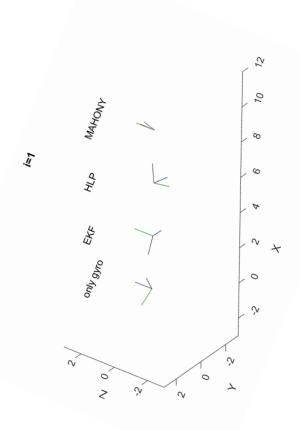
Ki: integral gain

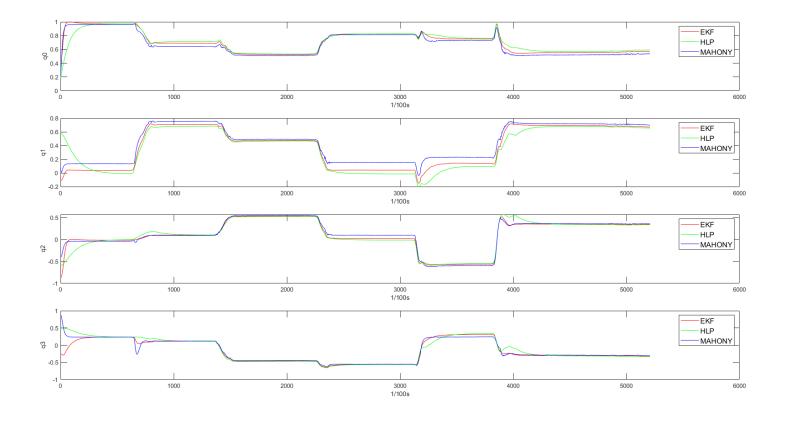
Mild orientation changing without calibration



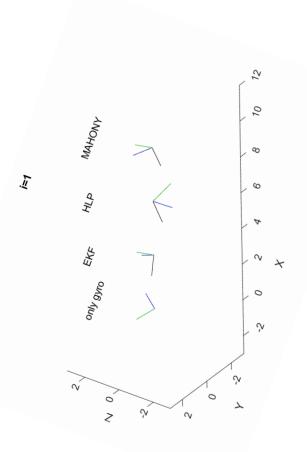


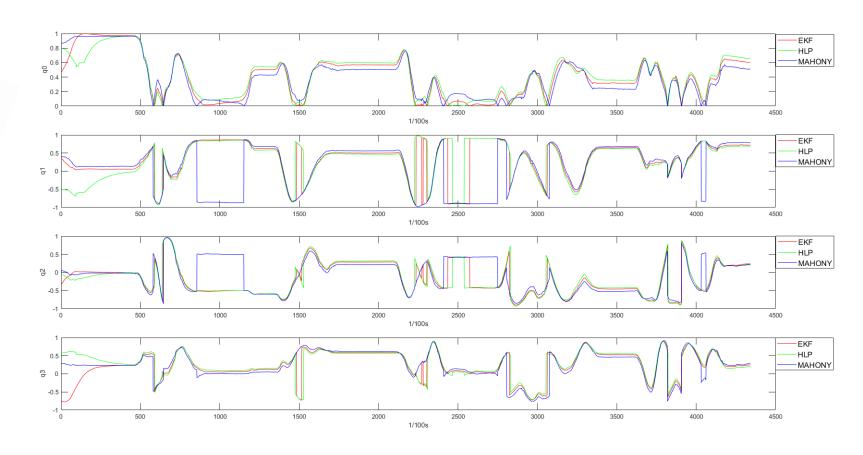
Mild orientation changing with calibration



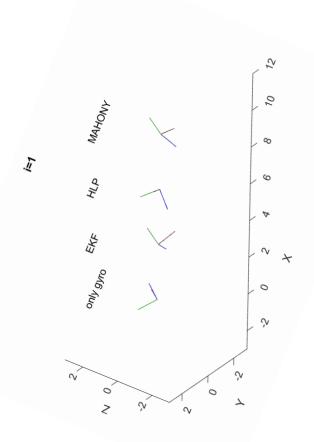


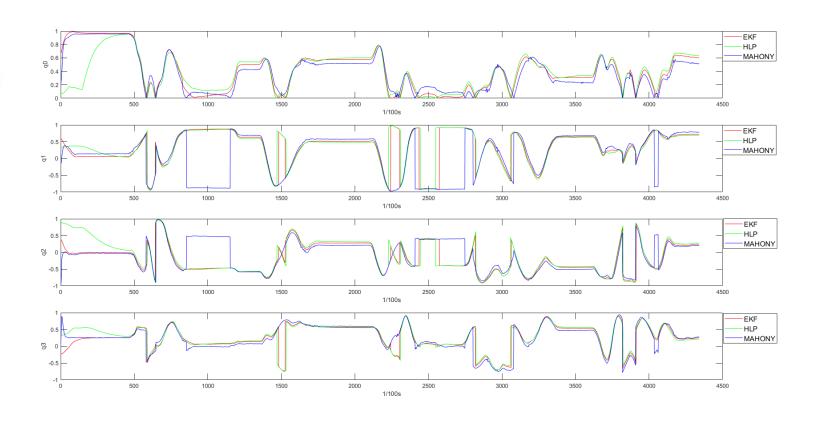
Violent orientation changing without calibration





Violent orientation changing with calibration





Conclusion and Future Work

Conclusion

- Calibration use in this project can get a better result in mild rotation changing case
- If the orientation changing rapidly, orientation tracking could be less accuracy while calibration for system error could be less effective.

Future work

- Dynamic calibration for gyro and accelerametor
- Explore the more calibration method
- Compare more different calibration methods and justify their difference (pros and cons)
- Explore the state of the art sensor fusion algorithms
- Compare new fusion algorithm with classic algorithm

References

- [1] Tedaldi, David, Alberto Pretto, and Emanuele Menegatti. "A robust and easy to implement method for IMU calibration without external equipments." 2014 IEEE International Conference on Robotics and Automation (ICRA). IEEE, 2014.
- [2] Sabatelli, Simone, et al. "A double-stage Kalman filter for orientation tracking with an integrated processor in 9-D IMU." *IEEE Transactions on Instrumentation and Measurement* 62.3 (2012): 590-598.
- [3] Mahony, Robert, Tarek Hamel, and Jean-Michel Pflimlin. "Nonlinear complementary filters on the special orthogonal group." *IEEE Transactions on automatic control* 53.5 (2008): 1203-1218.
- [4] Shike Shen: 9 Axis IMU Calibration [https://blog.csdn.net/shenshikexmu/article/details/80013444]

Thank you!

Any Questions?

