## Attitude Estimation using low-cost MEMS Sensors EE617 Sensors in Instrumentation (Instructor: Prof. Siddharth Tallur)

Mihir Kavishwar (17D070004)

Electrical Engineering, IIT Bombay

Mihir Kavishwar (EE, IIT Bombay)

EE617 Sensors in Instrumentation

December 11, 2020 1 / 14

## Why Attitude Matters

- Attitude (not to be confused with disposition or state of mind) of any rigid body describes how it is oriented in space with respect to some frame of reference
- In various applications such as in spacecrafts, aeroplanes, navigation systems, land vehicles, smartphones, VR gadgets, fitness trackers, etc, estimating the attitude accurately is of utmost importance



Source: Wikipedia







#### Source:

https://www.awwwards.com/vr-is-readyfor-consumers-are-you-ready-for-it.html

< ロ > < 同 > < 回 > < 回 >

#### Mihir Kavishwar (EE, IIT Bombay)

### EE617 Sensors in Instrumentation

## Mathematically Describing Attitude

- There are 2 ways to describe attitude which are commonly used:
  - Euler Angles  $(\alpha, \beta, \gamma)$  or  $(\psi, \theta, \phi)$  or (yaw, pitch, roll)
  - Quaternions a + bi + cj + dk
- Euler angles represent the orientation in terms of angles by which body underwent sequence of rotations along principal axes
- Quaternions have 2 parts scalar 'a' which represents amount of rotation and vector (b, c, d) which represents axis of rotation
- There are other ways to describe orientation such as rotation matrices

・ 何 ト ・ ヨ ト ・ ヨ ト

## Sensors for measuring Attitude

- Today, in almost all applications where we want to measure attitude, we use a class of sensors called Microelectromechanical Systems (MEMS)
- MEMS Accelerometers measure the proper acceleration while MEMS Gyroscopes measure the rate of change of orientation along primary axes in body frame
- MEMS Inertial Measurement Units (IMUs) use a combination of accelerometers, gyroscopes and sometimes magnetometer, to report the same quantities



Source: https://www.fierceelectronics.com/components/roger-grace-authors-eight-part-series-sensor-mems

## Sensors for measuring Attitude

- Typically, the performance of gyroscopes is limited by drift while performance of accelerometers is limited by high frequency noise due to vibrations
- To get more precision we need high grade sensors which are costly



Source: https://semiengineering.com/are-todays-mems-gyros-good-enough/

## Getting high accuracy with low-cost sensors

- Over the years, lot of research has been done in the field of State Estimation and Control
- We borrow some of these ideas which allow us to combine information from multiple noisy sensors and produce estimates whose accuracy is comparable to high grade sensors
- The literature that I reviewed used 4 kinds of filtering algorithms:
  - Kalman Filter (KF)
  - Extended Kalman Filter (EKF)
  - Unscented Kalman Filter (UKF)
  - Digital Complementary Filter

| Initial estimates for $\hat{x}_k$ and $P_{k-1}$ |   |  |  |  |
|---|---|--|--|--|
|   |   |  |  |  |
|   | Time Update ("Predict")   |  |  |  |
|   | (1) Project the state ahead   |  |  |  |
|   | $\hat{x}_k^- = A\hat{x}_{k-1} + Bu_k$   |  |  |  |
|   | (2) Project the error covariance ahead  |  |  |  |
|   | $P_k^- = AP_{k-1}A^T + Q$   |  |  |  |
| Ľ   |   |  |  |  |
| Н   | Measurement Update ("Correct")  |  |  |  |
| Ĺ   | (1) Compute the Kalman gain<br>$K_{k} = P_{k}^{-}H^{T}(HP_{k}^{-}H^{T} + R)^{-1}$ |  |  |  |
|   | $ \begin{array}{llllllllllllllllllllllllllllllllllll$                             |  |  |  |
|   | $P_k = (I - K_k H) P_k^-$   |  |  |  |

Source: Carmen M. N. Brigante, Nunzio Abbate, Adriano Basile, Alessandro Carmelo Faulisi, and Salvatore Sessa, "Towards Miniaturization of a MEMS-Based Wearable Motion Capture System", *IEEE Transactions on Industrial Electronics*, vol. 58, no. 8, August 2011.

# Case Study 1 - Wearable Motion Capture System

- This work presents a modular architecture to develop a wearable system for real-time human motion capture
- A quaternion based EKF algorithm is used for fusing the information from three-axis accelerometer, a three-axis magnetometer, and a three-axis gyroscope
- Compared to commercial systems, the cost of the proposed system was reduced by a factor of about eight due to an embedded design based on MEMS sensors





Ref: Carmen M. N. Brigante, Nunzio Abbate, Adriano Basile, Alessandro Carmelo Faulisi, Salvatore Sessa, "Towards Miniaturization of a MEMS-Based Wearable Motion Capture System", *IEEE Transactions on Industrial Electronics*, 2011, O. C.

Mihir Kavishwar (EE, IIT Bombay)

EE617 Sensors in Instrumentation

December 11, 2020 7 / 14

## Case Study 2 - Moving Land Vehicle

- Typically, an accelerometer is used to correct gyroscope drift by measuring the attitude from gravitational acceleration
- This is difficult in vehicular applications as accelerometer measurements are corrupted by external accelerations produced due to vehicle movements
- Authors show that vehicle kinematics allow the removal of external accelerations from the lateral and vertical axis accelerometer measurements



Ref: Hamad Ahmed and Muhammad Tahir, "Accurate Attitude Estimation of a Moving Land Vehicle Using Low-Cost MEMS IMU Sensors", IEEE Transactions on Intelligent Transportation Systems, vol. 18, no. 7, July 2017.

- In order to compare how different algorithms perform for simple attitude stimation task, I implemented KF, EKF and UKF in MATLAB
- The input data was taken from my smartphone's accelerometer and gyroscope when it was in motion
- True states were simulated by adding gaussian noise with specified covariance matrices using MATLAB functions
- All the estimates were plotted on the same graph to see how well they track true states



Source: Heikki Hyyti, Arto Visala, "A DCM Based Attitude Estimation Algorithm for Low-Cost MEMS IMUs", Hindawi Publishing Corporation International Journal of Navigation and Observation, Volume 2015, Article ID 503814

for k=1:999 A\_EKF = delf\_delx(x\_hat\_EKF(:,k),u(:,k),freq); 194 x\_hat\_predict\_EKF(:,k) = process(x\_hat\_EKF(:,k),u(:,k), freq); % Non Linear P\_predict\_EKF(:,:,k) = A\_EKF\*P\_EKF(:,:,k)\*A\_EKF' + (Gamma^2).\*Q; % Kalman Gain Computation R\_EKF = meas\_noise\_cov(f(k+1),x\_hat\_predict\_EKF(1:3,k),var\_a, var\_f, g); L\_EKF(:,:,k) = P\_predict\_EKF(:,:,k)\*H'/(H\*P\_predict\_EKF(:,:,k)\*H' + R\_EKF); 288 201 % Update 282 e EKF(:,k) = v(:,k+1)-H\*x hat predict EKF(:,k); 203 x\_hat\_EKF(:,k+1) = x\_hat\_EKF(:,k) + L\_EKF(:,:,k)\*e\_EKF(:,k); 284 P\_EKF(:,:,k+1) = (eye(6)-L\_EKF(:,:,k)\*H)\*P\_predict\_EKF(:,:,k); % Normalization norm mat = cat(1, cat(2, eye(3)./norm(x hat EKF(1:3,k+1)),zeros(3,3)),cat(2,zeros(3,3),eye(3))); 285 J = norm jacobian(x hat EKF(:,k+1)); 289 x bat FKF(:.k+1) = norm mat\*x bat FKF(:.k+1): P\_EKF(:,:,k+1)=J\*P\_EKF(:,:,k+1)\*J'; %Estimated Angles theta\_hat\_EKF(k+1)=asin(-x\_hat\_EKF(1,k+1)); phi\_hat\_EKF(k+1)=atan2(x\_hat\_EKF(2,k+1),x\_hat\_EKF(3,k+1));

Source: MATLAB code written by me

イロト イヨト イヨト イヨト



Figure: Estimating Roll ( $\phi$ )

Mihir Kavishwar (EE, IIT Bombay)

EE617 Sensors in Instrumentation

December 11, 2020 11 / 14

イロト イ部ト イヨト イヨト 一日



Figure: Estimating Pitch  $(\theta)$ 

Mihir Kavishwar (EE, IIT Bombay)

EE617 Sensors in Instrumentation

December 11, 2020 12 / 14

▲□▶ ▲圖▶ ▲国▶ ▲国▶ 二百

### Table: Root Mean Squared Error comparison

|                             | KF     | EKF    | UKF    |
|-----------------------------|--------|--------|--------|
| RMSE for $\theta$ (radians) | 0.0241 | 0.0189 | 0.0321 |
| RMSE for $\phi$ (radians)   | 0.0231 | 0.0155 | 0.0428 |

For this system it is evident that EKF performs better than both KF and UKF. Among KF and UKF, KF performs better.

- 4 回 ト 4 ヨ ト 4 ヨ ト

### References

Nguyen Ho Quoc Phuong, Hee-Jun Kang, Young-Soo Suh, Young-Sik Ro, "A DCM Based Orientation Estimation Algorithm with an Inertial Measurement Unit and a Magnetic Compass", *Journal of Universal Computer Science*, vol. 15, no. 4 (2009), 859-876.



Carmen M. N. Brigante, Nunzio Abbate, Adriano Basile, Alessandro Carmelo Faulisi, and Salvatore Sessa, "Towards Miniaturization of a MEMS-Based Wearable Motion Capture System", *IEEE Transactions on Industrial Electronics*, vol. 58, no. 8, August 2011.



Hector Garcia de Marina, Felipe Espinosa and Carlos Santos, "Adaptive UAV Attitude Estimation Employing Unscented Kalman Filter, FOAM and Low-Cost MEMS Sensors", *Sensors 2012, 12.* 

Heikki Hyyti, Arto Visala, "A DCM Based Attitude Estimation Algorithm for Low-Cost MEMS IMUs", Hindawi Publishing Corporation International Journal of Navigation and Observation, Volume 2015, Article ID 503814.



Nur Hazliza Ariffin, Badariah Bais, Norhana Arsad, "Low Cost MEMS Gyroscope and Accelerometer Implementation Without Kalman Filter For Angle Estimation", 2016 International Conference on Advances in Electrical, Electronic and System Engineering, 14-16 Nov 2016, Putrajaya, Malaysia.



Hamad Ahmed and Muhammad Tahir, "Accurate Attitude Estimation of a Moving Land Vehicle Using Low-Cost MEMS IMU Sensors", *IEEE Transactions on Intelligent Transportation Systems*, vol. 18, no. 7, July 2017.

Yizhai Zhang, Kehao Song, Jingang Yi, Panfeng Huang, Zhansheng Duan and Qijie Zhao, "Absolute Attitude Estimation of Rigid Body on Moving Platform Using Only Two Gyroscopes and Relative Measurements", *IEEE/ASME Transactions on Mechatronics*, vol. 23, no. 3, June 2018.

< □ > < 同 > < 回 > < 回 > < 回 >