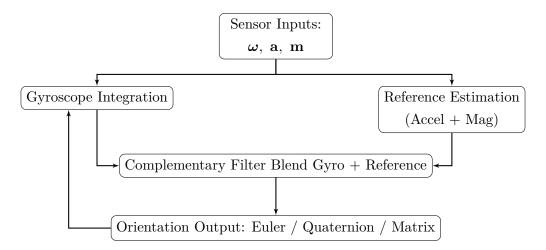
# Complementary Filter for 3D Attitude Estimation in AHRS Systems

## Introduction

Attitude and Heading Reference Systems (AHRS) estimate the orientation of a body in 3D space by combining data from inertial sensors. The **complementary filter** is one of the simplest and most efficient algorithms for estimating roll, pitch, and yaw by fusing gyroscope, accelerometer, and magnetometer data.

# 1 Complementary Filter Architecture

The following diagram illustrates the data flow in a complementary filter for orientation estimation:



This flow diagram represents how the complementary filter uses both high-frequency gyroscope data and low-frequency corrections from the accelerometer and magnetometer to estimate robust orientation.

# 2 Basic Principle of the Complementary Filter

A complementary filter takes advantage of the frequency characteristics of different sensors:

- Gyroscopes are accurate for short-term angular velocity but drift over time.
- Accelerometers provide a long-term reference for roll and pitch via gravity.
- Magnetometers provide a long-term reference for yaw via the Earth's magnetic field.

The idea is to combine these sources using a high-pass filter (for the gyroscope) and a low-pass filter (for the accelerometer and magnetometer), resulting in a stable and drift-compensated orientation estimate.

All measurements from the gyroscope, accelerometer, and magnetometer are assumed to be expressed in the **body-fixed frame**, and orientation is computed relative to the **NED** (North-East-Down) frame.

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#### 4 Mathematical Formulation

#### 4.1 Estimation of Roll, Pitch, and Yaw

Let  $\phi$ ,  $\theta$ , and  $\psi$  be roll, pitch, and yaw, respectively. Let  $\phi_g$ ,  $\theta_g$ , and  $\psi_g$  be the angles obtained by integrating gyroscope data:

$$\phi_g(t) = \phi(t - \Delta t) + \omega_x(t) \cdot \Delta t$$

$$\theta_g(t) = \theta(t - \Delta t) + \omega_y(t) \cdot \Delta t$$

$$\psi_q(t) = \psi(t - \Delta t) + \omega_z(t) \cdot \Delta t$$

Let  $\phi_a$ ,  $\theta_a$  be the roll and pitch derived from the accelerometer (gravity vector):

$$\phi_a = \arctan 2(a_y, a_z), \quad \theta_a = \arctan 2(-a_x, \sqrt{a_y^2 + a_z^2})$$

Let  $\psi_m$  be the yaw angle derived from the magnetometer, using **tilt compensation** to correct for current roll and pitch. First, the magnetometer vector  $\mathbf{m} = [m_x, m_y, m_z]^T$  is transformed as:

$$m_x^c = m_x \cos \theta + m_z \sin \theta$$
  

$$m_y^c = m_x \sin \phi \sin \theta + m_y \cos \phi - m_z \sin \phi \cos \theta$$
  

$$\psi_m = \arctan 2(-m_y^c, m_x^c)$$

Final complementary filter:

$$\phi = \alpha \cdot \phi_g + (1 - \alpha) \cdot \phi_a$$

$$\theta = \alpha \cdot \theta_g + (1 - \alpha) \cdot \theta_a$$

$$\psi = \alpha \cdot \psi_g + (1 - \alpha) \cdot \psi_m$$

Where  $\alpha \in [0, 1]$  is the filter coefficient.

## 4.2 Direct Estimation of the Orientation Quaternion

The orientation quaternion  $q = [q_0, q_1, q_2, q_3]^T$  can be directly estimated by combining the gyroscope-integrated quaternion  $q_g$  with a reference quaternion  $q_a$  constructed from the ac-

celerometer and magnetometer readings.

First, integrate angular velocity from the gyroscope using:

$$\dot{q}_g = \frac{1}{2} q_g \otimes \begin{bmatrix} 0 \\ \omega_x \\ \omega_y \\ \omega_z \end{bmatrix}, \quad q_g(t) = q_g(t - \Delta t) + \dot{q}_g(t) \cdot \Delta t$$

To estimate  $q_a$ , we use the accelerometer and magnetometer as reference vectors. The gravity vector in the NED frame is  $\mathbf{g}_n = [0, 0, 1]^T$  and the magnetic field vector is  $\mathbf{m}_n = [1, 0, 0]^T$  (assuming north-aligned field).

Normalize the accelerometer and magnetometer measurements:

$$\mathbf{g}_b = rac{\mathbf{a}}{\|\mathbf{a}\|}, \quad \mathbf{m}_b = rac{\mathbf{m}}{\|\mathbf{m}\|}$$

Apply the TRIAD algorithm to find the orientation quaternion  $q_a$  that aligns  $\mathbf{g}_b \to \mathbf{g}_n$  and  $\mathbf{m}_b \to \mathbf{m}_n$ :

1. Define orthonormal bases:

$$\mathbf{v}_1^b = \mathbf{g}_b, \qquad \qquad \mathbf{v}_2^b = \frac{\mathbf{g}_b \times \mathbf{m}_b}{\|\mathbf{g}_b \times \mathbf{m}_b\|}, \qquad \qquad \mathbf{v}_3^b = \mathbf{v}_1^b \times \mathbf{v}_2^b$$
 $\mathbf{v}_1^n = \mathbf{g}_n, \qquad \qquad \mathbf{v}_2^n = \frac{\mathbf{g}_n \times \mathbf{m}_n}{\|\mathbf{g}_n \times \mathbf{m}_n\|}, \qquad \qquad \mathbf{v}_3^n = \mathbf{v}_1^n \times \mathbf{v}_2^n$ 

2. Construct direction cosine matrices:

$$R_b = [\mathbf{v}_1^b \ \mathbf{v}_2^b \ \mathbf{v}_3^b], \quad R_n = [\mathbf{v}_1^n \ \mathbf{v}_2^n \ \mathbf{v}_3^n]$$

3. Compute the rotation matrix:

$$R = R_n R_b^T$$

4. Extract the quaternion  $q_a$  from R using the standard DCM-to-quaternion conversion:

$$q_0 = \frac{1}{2}\sqrt{1 + R_{11} + R_{22} + R_{33}}$$

$$q_1 = \frac{R_{32} - R_{23}}{4q_0}, \quad q_2 \qquad \qquad = \frac{R_{13} - R_{31}}{4q_0}, \quad q_3 = \frac{R_{21} - R_{12}}{4q_0}$$

The complementary filter then blends the gyroscope and reference quaternions:

$$q(t) = \text{normalize}(\alpha q_q(t) + (1 - \alpha)q_a(t))$$

#### 4.3 Direct Estimation of the Rotation Matrix

The orientation matrix  $R(t) \in SO(3)$  can be computed by integrating the angular velocity and correcting using the accelerometer and magnetometer.

From gyroscope integration:

$$\dot{R}_g = R_g[\boldsymbol{\omega}]_{\times}, \quad R_g(t) = R_g(t - \Delta t) + \dot{R}_g(t) \cdot \Delta t$$

where  $[\boldsymbol{\omega}]_{\times}$  is the skew-symmetric matrix:

$$[oldsymbol{\omega}]_{ imes} = egin{bmatrix} 0 & -\omega_z & \omega_y \ \omega_z & 0 & -\omega_x \ -\omega_y & \omega_x & 0 \end{bmatrix}$$

Construct a reference rotation matrix  $R_a$  from body-frame measurements of gravity  $\mathbf{g}_b$  and magnetic field  $\mathbf{m}_b$  using the TRIAD basis:

$$\mathbf{v}_1^b = \mathbf{g}_b, \ \mathbf{v}_2^b = \frac{\mathbf{g}_b \times \mathbf{m}_b}{\|\mathbf{g}_b \times \mathbf{m}_b\|}, \ \mathbf{v}_3^b = \mathbf{v}_1^b \times \mathbf{v}_2^b$$

Then:

$$R_a = [\mathbf{v}_1^b \ \mathbf{v}_2^b \ \mathbf{v}_3^b]$$

The complementary filter updates the rotation matrix as:

$$R(t) = \text{orthonormalize}(\alpha R_g(t) + (1 - \alpha)R_a(t))$$

where orthonormalize ensures that R(t) remains a valid rotation matrix.

# 5 Filter Tuning

The coefficient  $\alpha$  controls the balance between fast gyroscope response and slow correction from the accelerometer and magnetometer:

- A typical value is  $\alpha = 0.98$ .
- Higher values trust the gyroscope more (less correction).
- Lower values increase sensitivity to drift correction.

# 6 Advantages and Limitations

## Advantages

- Simple and computationally efficient.
- Real-time performance suitable for embedded systems.
- Full 3D orientation coverage with all three sensors.

#### Limitations

- Sensitive to magnetometer disturbances (e.g., motors, metal).
- Accuracy degrades in highly dynamic conditions.
- Cannot fully represent orientation using quaternions or rotation matrices without correction.