"State Estimation with Kalman Filter"

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Abstract

Estimate Euler angles using extended Kalman filter and IMU measurements

1 Equations

Kalman filter involves two steps: prediction and update.

1.1 Prediction

Predict states (x) using **discrete-time model** and update covariance (P). Note that, Q is process covariance (σ^2) . In other word, how good is the model.

$$x' = Fx + Bu \tag{1}$$

$$P = FPF^T + Q (2)$$

1.2 Update

Update states (x) and covariance (P) based on predicted state (x'), sensor measurements (z), and measurement matrix (H). Note that, R is measurement covariance (σ^2) , which is provided by sensor manufacturer or by experiment.

$$z_{est} = Hx' \tag{3}$$

$$y = z - z_{est} \tag{4}$$

$$S = HPH^T + R \tag{5}$$

$$K = PH^T S^{-1} \tag{6}$$

$$x = x' + Ky \tag{7}$$

$$P = (I - KH)P \tag{8}$$

1.3 Extended Kalman Filter

Extended Kalman filter is similar to Kalman filter. However, the nonlinear model is used for prediction and its Jacobian matrix is calculated for F and H.

Eq (1) becomes
$$x' = f(x, u) \tag{9}$$

F and H in Eq (19), (23), (24), and (26) becomes Jacobian matrices that are evaluated at $x^{\prime}.$

2 Implementation

The filter estimates Euler angles using body angular rate (p, q, r) and body acceleration (a_x, a_y, a_z) . This section provides the derivation of the nonlinear model, Jacobian matrix, and initialization of the filter.

2.1 Model

The model is based on kinematic that relates Euler angles, body angular rate, and body acceleration. This section begins with frame rotation, and then the nonlinear kinematic equations. And it will ends with the Jacobian matrices.

2.1.1 Frame rotation

Rotate acceleration vector from inertia to body frame

$$\begin{bmatrix} \dot{u} \\ \dot{v} \\ \dot{w} \end{bmatrix} = R_i^b \begin{bmatrix} a_x \\ a_y \\ a_z \end{bmatrix} \tag{10}$$

where

$$R_i^b = \begin{bmatrix} c_{\theta}c_{\psi} & c_{\theta}s_{\psi} & -s_{\theta} \\ (-c_{\phi}s_{\psi} + s_{\phi}s_{\theta}c_{\psi}) & (c_{\phi}c_{\psi} + s_{\phi}s_{\theta}s_{\psi}) & s_{\phi}c_{\theta} \\ (s_{\phi}s_{\psi} + c_{\phi}s_{\theta}c_{\psi}) & (-s_{\phi}c_{\psi} + c_{\phi}s_{\theta}s_{\psi}) & c_{\phi}c_{\theta} \end{bmatrix}$$
(11)

Rotate body angular rate vector to Euler frame

$$\begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix} = R_b^e \begin{bmatrix} p \\ q \\ r \end{bmatrix} \tag{12}$$

where

$$R_b^e = \begin{bmatrix} 1 & s_{\phi}t_{\theta} & c_{\phi}t_{\theta} \\ 0 & c_{\phi} & -s_{\phi} \\ 0 & s_{\phi}/c_{\theta} & c_{\phi}/c_{\theta} \end{bmatrix}$$
(13)

2.1.2 Kinematic

Given the velocity, v, of an object over time, the position, p, becomes

$$p = \int_0^t v dt \tag{14}$$

In discrete-time, it is represented as

$$p' = p + v\Delta T \tag{15}$$

where ΔT is the time step.

Discrete-time model By combining the equations above, the equations become:

$$\begin{bmatrix} \phi' \\ \theta' \\ \psi' \end{bmatrix} = I_{3x3} \begin{bmatrix} \phi \\ \theta \\ \psi \end{bmatrix} + R_b^e(\phi, \theta, \psi) \Delta T \begin{bmatrix} p \\ q \\ r \end{bmatrix}$$
 (16)

Accelerometer Measurement Express measure in terms of filter's states

$$\begin{bmatrix} \dot{u} \\ \dot{v} \\ \dot{w} \end{bmatrix} = R_i^b(\phi, \theta, \psi) \begin{bmatrix} 0 \\ 0 \\ g \end{bmatrix} = g \begin{bmatrix} -s_{\theta} \\ s_{\phi}c_{\theta} \\ c_{\phi}c_{\theta} \end{bmatrix}$$
(17)

where g is gravitational acceleration

2.2 Summary

We want to estimate Euler angles, ϕ , θ , and ψ using Extended Kalman filter. u is body angular rate, p, q, and r. x is Euler angles, ϕ , θ , and ψ . z is body acceleration.

Prediction

$$\begin{bmatrix} \phi' \\ \theta' \\ \psi' \end{bmatrix} = I_{3x3} \begin{bmatrix} \phi \\ \theta \\ \psi \end{bmatrix} + R_b^e(\phi, \theta, \psi) \Delta T \begin{bmatrix} p \\ q \\ r \end{bmatrix}$$
 (18)

where ΔT is sample time.

$$P = F_j P F_j^T + Q (19)$$

where $F_j = I_{3x3}$

Update

$$z_{est} = H_j x' \tag{20}$$

where measurement matrix, H_j , is a Jacobian matrix.

$$H_{j} = \begin{bmatrix} \frac{\partial \dot{u}}{\partial \phi} & \frac{\partial \dot{u}}{\partial \theta} & \frac{\partial \dot{u}}{\partial \psi} \\ \frac{\partial \dot{v}}{\partial \phi} & \frac{\partial \dot{v}}{\partial \theta} & \frac{\partial \dot{v}}{\partial \psi} \\ \frac{\partial \dot{w}}{\partial \phi} & \frac{\partial \dot{w}}{\partial \theta} & \frac{\partial \dot{w}}{\partial \psi} \end{bmatrix} = g \begin{bmatrix} 0 & -c_{\theta} & 0 \\ c_{\phi}c_{\theta} & -s_{\phi}s_{\theta} & 0 \\ -s_{\phi}c_{\theta} & -c_{\phi}s_{\theta} & 0 \end{bmatrix}$$
(21)

$$y = z - z_{est} \tag{22}$$

$$S = H_j P H_j^T + R (23)$$

$$K = PH_j^T S^{-1} (24)$$

$$x = x' + Ky \tag{25}$$

$$P = (I - KH)P \tag{26}$$

2.3 Initialization

Initialize filter's states, ϕ , θ , ψ using accelerometer measurement, which is body acceleration, a_{bx} , a_{by} , a_{bz} .

Using Equation 17, we have

$$\begin{bmatrix} a_{bx} \\ a_{by} \\ a_{bz} \end{bmatrix} = g \begin{bmatrix} -s_{\theta} \\ s_{\phi}c_{\theta} \\ c_{\phi}c_{\theta} \end{bmatrix}$$
 (27)

We can express ϕ and θ as

$$\theta = -asin(\frac{a_{bx}}{q})\tag{28}$$

$$\phi = asin(\frac{a_{by}}{gc_{\theta}}) \tag{29}$$

Since there is no magnetometer, we assume

$$\psi = 0 \tag{30}$$

Assume that the accelerometer is at rest, it returns gravitational acceleration only.

$$g = \sqrt{a_{bx}^2 + a_{by}^2 + a_{bz}^2} \tag{31}$$