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```
function [xhat, meas] = filterTemplate(calAcc, calGyr, calMag)

% FILTERTEMPLATE Filter template
%
% This is a template function for how to collect and filter data
% sent from a smartphone live. Calibration data for the
% accelerometer, gyroscope and magnetometer assumed available as
% structs with fields m (mean) and R (variance).
%
% The function returns xhat as an array of structs comprising t
% (timestamp), x (state), and P (state covariance) for each
% timestamp, and meas an array of structs comprising t (timestamp),
% acc (accelerometer measurements), gyr (gyroscope measurements),
% mag (magnetometer measurements), and orint (orientation quaternions
% from the phone). Measurements not available are marked with NaNs.
%
% As you implement your own orientation estimate, it will be
% visualized in a simple illustration. If the orientation estimate
% is checked in the Sensor Fusion app, it will be displayed in a
% separate view.
%
% Note that it is not necessary to provide inputs (calAcc, calGyr,
calMag).
```

Setup necessary infrastructure

```
import('com.liu.sensordata.*'); % Used to receive data.

% Filters used
magFilt = true;
accFilt = true;
```

Filter settings

```
t0 = []; % Initial time (initialize on first data received)
nx = 4; % Assuming that you use q as state variable.
% Add your filter settings here.
T = 1/25;
```

```

% Gyroscope covariances
sigma_w = 1/10; % was 0.01
Rw = diag([sigma_w^2 sigma_w^2 sigma_w^2]);
sigma_v = 0.0014;
Rv = diag([sigma_v^2 sigma_v^2 sigma_v^2]);
gyr_old = [0; 0; 0];

% Accelerometer parameters and covariances
g_abs = 9.82; % 9.8908 from estimate using stationary data.
g_lim = [0.8 1.2]*g_abs; % +- 20% marginal
sigma_a = 0.3; % From variance estimate using stationary data.
Ra = diag([sigma_a^2 sigma_a^2 sigma_a^2]);

% Magnetometer parameters and covariances
m_abs = 39.1103; % 39.1103 from estimate using stationary data.
m_lim = [0.6 1.4]*m_abs; % +- 20% marginal
sigma_m = 0.9; % From variance estimate using stationary data.s
Rm = diag([sigma_m^2 sigma_m^2 sigma_m^2]);

% Current filter state.
x = [1; 0; 0; 0];
P = eye(nx, nx);

% Saved filter states.
xhat = struct('t', zeros(1, 0),...
             'x', zeros(nx, 0),...
             'P', zeros(nx, nx, 0));

meas = struct('t', zeros(1, 0),...
             'acc', zeros(3, 0),...
             'gyr', zeros(3, 0),...
             'mag', zeros(3, 0),...
             'orient', zeros(4, 0));

try

```

Create data link

```

server = StreamSensorDataReader(3400);
% Makes sure to resources are returned.
sentinel = onCleanup(@() server.stop());

server.start(); % Start data reception.

% Used for visualization.
figure(1);
subplot(1, 2, 1);
ownView = OrientationView('Own filter', gca); % Used for
visualization.
googleView = [];
counter = 0; % Used to throttle the displayed frame rate.

*****

```

```
* SensorDataReader started on 10.0.161.113:3400
*****
ERROR: Failed to open port or no connection attempted.
Unsuccessful connecting to client!
Make sure to start streaming from the phone *after*running this
function!
```

Filter loop

```
while server.status() % Repeat while data is available
    % Get the next measurement set, assume all measurements
    % within the next 5 ms are concurrent (suitable for sampling
    % in 100Hz).
    data = server.getNext(5);

    if isnan(data(1)) % No new data received
        continue; % Skips the rest of the look
    end
    t = data(1)/1000; % Extract current time

    if isempty(t0) % Initialize t0
        t0 = t;
    end

    gyr = data(1, 5:7)';
    % Gyroscope
    if ~any(isnan(gyr))
        w = gyr;
        x_tmp = x;
        P_tmp = P;
        [x, P] = tu_qw(x, P, gyr, T, Rw);
        [x, P] = mu_normalizeQ(x, P);
        x_kmin1 = x_tmp;
        P_kmin1 = P_tmp;
    else
        % x_tmp = x;
        % P_tmp = P;
        % [w, x, P] = tu_qw_pred(x, P, x_kmin1, P_kmin1, w, T, Rw);
        % [x, P] = mu_normalizeQ(x, P);
        % x_kmin1 = x_tmp;
        % P_kmin1 = P_tmp;
    end

    acc = data(1, 2:4)';
    if accFilt
        if ~any(isnan(acc)) % Acc measurements are available.
            % Approximate g, skip update if "outlier"
            a_abs = sqrt(sum(acc.^2));
            if (a_abs > g_lim(1)) && (a_abs < g_lim(2))
                g0 = [0; 0; a_abs];
                [x, P] = mu_g(x, P, acc, Ra, g0);
            end
        end
    end
end
```

```

        [x, P] = mu_normalizeQ(x, P);
    end
end
end

mag = data(1, 8:10)';
if magFilt
    if ~any(isnan(mag)) % Mag measurements are available.
        % Approximate m, skip update if "outlier"
        m_abs = sqrt(sum(mag.^2));
        if (m_abs > m_lim(1)) && (m_abs < m_lim(2))
            m0 = [0; 12; -36.09]; m0 = m0/norm(m0);
            [x, P] = mu_m(x, P, mag, Rm, m0);
            [x, P] = mu_normalizeQ(x, P);
        end
    end
end

orientation = data(1, 18:21)'; % Google's orientation estimate.

% Visualize result
if rem(counter, 10) == 0
    setOrientation(ownView, x(1:4));
    title(ownView, 'OWN', 'FontSize', 16);
    if ~any(isnan(orientation))
        if isempty(googleView)
            subplot(1, 2, 2);
            % Used for visualization.
            googleView = OrientationView('Google filter', gca);
        end
        setOrientation(googleView, orientation);
        title(googleView, 'GOOGLE', 'FontSize', 16);
    end
end
counter = counter + 1;

% Save estimates
xhat.x(:, end+1) = x;
xhat.P(:, :, end+1) = P;
xhat.t(end+1) = t - t0;

meas.t(end+1) = t - t0;
meas.acc(:, end+1) = acc;
meas.gyr(:, end+1) = gyr;
meas.mag(:, end+1) = mag;
meas.orient(:, end+1) = orientation;

end

catch e
    fprintf(['Unsuccessful connecting to client!\n' ...

```

```

        'Make sure to start streaming from the phone *after*'...
        'running this function!']]);
    end

end

ans =

    struct with fields:

        t: [1x0 double]
        x: [4x0 double]
        P: [4x4x0 double]

```

"tu_qw"

```

function [x, P] = tu_qw(x, P, omega, T, Rw)
% EKF time update step

% Gyroscope measurement noise covariance (estimated from stationary
data)
sigma_v = 0.0014;
Rv = diag([sigma_v^2 sigma_v^2 sigma_v^2]);

% Process matrices
F = T/2*Somega(omega) + eye(4);
F_tilde = F;
G = T/2*Sq(x);
G_tilde = G;

% Prediction step
xp = F*x;
Pp = F_tilde*P*F_tilde' + G_tilde*Rw*G_tilde';

% Update step
% Equation:  $y_k = H*x_k + B*x_{kmin1\_kmin1} + v_k$ 
H = 2/T*pinv(Sq(x));
B = -H;
Bx = B*x;
yhat = H*xp + Bx;

S = H*Pp*H' + B*P*B' + Rv;
K = Pp*H'*(S^-1);

x = xp + K*(omega - yhat);
P = Pp - K*S*K';

end

```

"tu_qw_pred"

```
function [w, x, P] = tu_qw_pred(x, P, x_kmin1, P_kmin1, omega, T, Rw)
% EKF time update step

% Estimate omega
H = 2/T*pinv(Sq(x_kmin1));
B = -H;
Bx = B*x_kmin1;
w = H*x + Bx;

% recalculate Process matrices
F = T/2*Somega(w) + eye(4);
F_tilde = F;
G = T/2*Sq(x);
G_tilde = G;

% Prediction step
x = F*x;
P = F_tilde*P*F_tilde' + G_tilde*Rw*G_tilde';

end
```

"mu_g"

```
function [x, P] = mu_g(x, P, y, Ra, g0)
% EKF accelerometer measurement update

% Measurement matrices
hx = Qq(x)'*g0;
[dQ0, dQ1, dQ2, dQ3] = dQqdq(x);
Jhx = [dQ0'*g0 dQ1'*g0 dQ2'*g0 dQ3'*g0];

% Measurement update
S = Jhx*P*Jhx' + Ra;
K = P*Jhx'*(S^-1);

x = x + K*(y - hx);
P = P - K*S*K';

end
```

"mu_m"

```
function [x, P] = mu_m(x, P, y, Rm, m0)
```

```

% EKF accelerometer measurement update

% normalize
ynorm = y/norm(y);
m0 = m0/norm(m0);
Rm = Rm/norm(y);

% Measurement matrices
hx = Qq(x)'*m0;
[dQ0, dQ1, dQ2, dQ3] = dQq dq(x);
Jhx = [dQ0'*m0 dQ1'*m0 dQ2'*m0 dQ3'*m0];

% Measurement update
S = Jhx*P*Jhx' + Rm;
K = P*Jhx'*(S^-1);

x = x + K*(y - hx);
P = P - K*S*K';

end

```

"Project_Implementation"

```

% SSY345 Project Implementation

% SECTION 4.1
% Task 2) See file Project_test.m. Measurement data loaded below.

load('data_FrassePhone_stationary.mat');

%% SECTION 4.2
% Task 3) Design the EKF time update step
% Implemented in file "tu_qw.m".

% Time between samples
T = 1/fs;

% Sizes
K = length(t);
n = 4;

% Gyroscope covariances
sigma_w = 0.01;
Rw = diag([sigma_w^2 sigma_w^2 sigma_w^2]);
sigma_v = 0.0014;
Rv = diag([sigma_v^2 sigma_v^2 sigma_v^2]);

% Accelerometer parameters and covariances
g_abs = 9.82; % 9.8908 from estimate using stationary data.

```

```

g_lim = [0.9 1.1]*g_abs;    % +- 10% marginal
sigma_a = 0.03; % From variance estimate using stationary data.
Ra = diag([sigma_a^2 sigma_a^2 sigma_a^2]);

% Magnetometer parameters and covariances
m_abs = 39.1103;    % 39.1103 from estimate using stationary data.
m_lim = [0.9 1.1]*m_abs;    % +- 10% marginal
sigma_m = 0.5; % From variance estimate using stationary data.s
Rm = diag([sigma_m^2 sigma_m^2 sigma_m^2]);

% Initial estimates and covariances
x_0 = [0; 0; 1; 0];
P_0 = diag([1 1 1 1]);

% Visualisation stuff
figure(1)
ownView = OrientationView('Own filter', gca);

xw = zeros(4,K);
Pw = zeros(4,4,K);
xw(:,1) = x_0;
Pw(:, :, 1) = P_0;

xa = zeros(4,K);
Pa = zeros(4,4,K);
xm = zeros(4,K);
Pm = zeros(4,4,K);

x = zeros(4,K);
P = zeros(4,4,K);
NaN_gyr = 1;
NaN_acc = 1;
for k = 2:K

    % Gyroscope measurement computations
    % Prediction:
    [xw(:,k), Pw(:, :, k)] = tu_qw(xw(:,k-1), Pw(:, :, k-1), y_gyr(:,k),
    T, Rw);

    % Include normalization of quaternion before using... where?
    [xw(:,k), Pw(:, :, k)] = mu_normalizeQ(xw(:,k), Pw(:, :, k));

    % Accelerometer computations
    % Measurement update:
    if ~isnan(y_acc(:,k))
        % Approximate g, skip update if "outlier"
        a_abs = sqrt(sum(y_acc(:,k).^2));
        if (a_abs > g_lim(1)) && (a_abs < g_lim(2))
            g0 = [0; 0; a_abs];
            [xa(:,k), Pa(:, :, k)] = mu_g(xw(:,k), Pw(:, :, k),
            y_acc(:,k), Ra, g0);
            [xa(:,k), Pa(:, :, k)] = mu_normalizeQ(xa(:,k), Pa(:, :, k));
        else
            xa(:,k) = xw(:,k);

```

```

        Pa(:, :, k) = Pw(:, :, k);
    end
else
    xa(:, k) = xw(:, k);
    Pa(:, :, k) = Pw(:, :, k);
end

% Magnetometer computations
% Measurement update:
if ~isnan(y_mag(:, k))
    % Approximate m, skip update if "outlier"
    m_abs = sqrt(sum(y_mag(:, k+1).^2));
    if (m_abs > m_lim(1)) && (m_abs < m_lim(2))
        m0 = [0; 19.18; -34.09];
        [xm(:, k), Pm(:, :, k)] = mu_m(xa(:, k), Pa(:, :, k),
y_mag(:, k), Rm, m0);
        [xm(:, k), Pm(:, :, k)] = mu_normalizeQ(xm(:, k), Pm(:, :, k));
    elseif ~isnan(y_acc(:, k))
        xm(:, k) = xa(:, k);
        Pm(:, :, k) = Pa(:, :, k);
    else
        xm(:, k) = xw(:, k);
        Pm(:, :, k) = Pw(:, :, k);
    end
elseif ~isnan(y_acc(:, k))
    xm(:, k) = xa(:, k);
    Pm(:, :, k) = Pa(:, :, k);
else
    xm(:, k) = xw(:, k);
    Pm(:, :, k) = Pw(:, :, k);
end

% Combine accelerometer and magnetometer information

% Update:
% Temporary for only gyroscope (Task 5)
%   x(:, k+1) = xwp(:, k+1);
%   P(:, :, k+1) = Pwp(:, :, k+1);

%   x(2:4, k+1) = xa(:, k+1);
%   P(2:4, 2:4, k+1) = Pa(:, :, k+1);
%   [x(:, k+1), P(:, :, k+1)] = mu_normalizeQ(xwp(:, k+1), Pwp(:, :, k+1));

setOrientation(ownView, xm(:, k));
title(ownView, 'OWN', 'FontSize', 16);
pause(0.01)
end

```

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