

SSY191 - Sensor Fusion and Nonlinear Filtering

Implementation of Home Assignment 04

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```

function [xs, Ps, xf, Pf, xp, Pp] = nonLinRTSSmoothen(Y, x_0, P_0, f, Q, h, R, sigmaPoints, type)
% NONLINRTSSMOOTHER Filters measurement sequence Y using a
% non-linear Kalman filter.
%
% Input:
%   Y           [m x N] Measurement sequence for times 1,...,N
%   x_0         [n x 1] Prior mean for time 0
%   P_0         [n x n] Prior covariance
%   f           Motion model function handle
%   T           Sampling time
%   Q           [n x n] Process noise covariance
%   S           [n x N] Sensor position vector sequence
%   h           Measurement model function handle
%   R           [n x n] Measurement noise covariance
%   sigmaPoints Handle to function that generates sigma points.
%   type        String that specifies type of non-linear filter/smoothen
%
% Output:
%   xf          [n x N] Filtered estimates for times 1,...,N
%   Pf          [n x n x N] Filter error covariance
%   xp          [n x N] Predicted estimates for times 1,...,N
%   Pp          [n x n x N] Filter error covariance
%   xs          [n x N] Smoothed estimates for times 1,...,N
%   Ps          [n x n x N] Smoothing error covariance

%% Parameters
N = size(Y,2);
n = length(x_0);
m = size(Y,1);

%% Forward filtering

%   f2 = @(x) f(x,T);
%   h2 = @(x) h(x,S(:,1));
[xf, Pf, xp, Pp] = nonLinearKalmanFilter(Y, x_0, P_0, f, Q, h, R, type);

%% Backward filtering

% initialize outputs
xs(:,N) = xf(:,N);
Ps(:, :, N) = Pf(:, :, N);

% backward recursion - last filter output is already smoothed
for k=N-1:-1:1
    [xs(:,k), Ps(:, :, k)] = nonLinRTSSupdate(xs(:,k+1), Ps(:, :, k+1), xf(:,k), Pf(:, :, k), xp(:,k+1), Pp(:, :, k+1), f, sigmaPoints, type);
end

end

```

```

function [xs, Ps] = nonLinRTSSupdate(xs_kplus1, ...
                                    Ps_kplus1, ...
                                    xf_k, ...
                                    Pf_k, ...
                                    xp_kplus1, ...
                                    Pp_kplus1, ...
                                    f, ...
                                    sigmaPoints, ...
                                    type)

% NONLINRTSSUPDATE Calculates mean and covariance of smoothed state
% density, using a non-linear Gaussian model.
%
% Input:
%   xs_kplus1   Smoothing estimate for state at time k+1
%   Ps_kplus1   Smoothing error covariance for state at time k+1
%   xf_k        Filter estimate for state at time k
%   Pf_k        Filter error covariance for state at time k
%   xp_kplus1   Prediction estimate for state at time k+1
%   Pp_kplus1   Prediction error covariance for state at time k+1
%   f           Motion model function handle
%   T           Sampling time
%   sigmaPoints Handle to function that generates sigma points.
%   type        String that specifies type of non-linear filter/smoother
%
% Output:
%   xs          Smoothed estimate of state at time k
%   Ps          Smoothed error covariance for state at time k

%% calculate Pkkip1 = P_{k,k+1|k}

if strcmp( type, 'EKF' )
    % evaluate motion model at  $\hat{x}_{k|k}$ 
    [~, df_xk] = f(xf_k);
    % approximate  $P_{k,k+1|k}$ 
    Pkkip1 = Pf_k * df_xk';

else % UKF or CKF
    % calculate sigma points of  $p(x_{k|k})$ 
    [SP,W] = sigmaPoints(xf_k, Pf_k, type);
    % predict covariance  $P_{k,k+1|k}$ 
    Pkkip1 = zeros(size(xf_k,1));
    for i=1: numel(W)
        Pkkip1 = Pkkip1 + (SP(:,i)-xf_k)*(f(SP(:,i))-xp_kplus1).' * W(i);
    end
end

%% backward recursion

% calculate smoothing gain
Gk = Pkkip1 / Pp_kplus1;
% calculate Smoothed estimate and covariance at time k
xs = xf_k + Gk * ( xs_kplus1 - xp_kplus1 );
Ps = Pf_k - Gk * ( Pp_kplus1 - Ps_kplus1 ) * Gk';

end

```

```

function [Xr, Wr, j] = resampl(X, W)
% RESAMPLE Resample particles and output new particles and weights.
% resampled particles.
%
% if old particle vector is x, new particles x.new is computed as x(:,j)
%
% Input:
% X [n x N] Particles, each column is a particle.
% W [1 x N] Weights, corresponding to the samples
%
% Output:
% Xr [n x N] Resampled particles, each corresponding to some particle
% from old weights.
% Wr [1 x N] New weights for the resampled particles.
% j [1 x N] vector of indices refering to vector of old particles

N=size(X,2);

% Generates the segmented numberline from 0 to 1 (upper edge not included since it will never be
segment = [0 cumsum(W)];

% draw samples from uniform distribution on [0,1]
samples = rand([1 N]);

j=zeros(1,N);
for i=1:N
    j(i) = find(samples(i) >= segment,1,'last');
end

Wr = 1/N*ones(1,N);
Xr = X(:,j);
end

```

```

function [X_k, W_k] = pfFilterStep(X_kmin1, W_kmin1, y_k, proc_f, proc_Q, meas_h, meas_R)
% PFFILTERSTEP Compute one filter step of a SIS/SIR particle filter.
%
% Input:
% X_kmin1 [n x N] Particles for state x in time k-1
% W_kmin1 [1 x N] Weights for state x in time k-1
% y_k [m x 1] Measurement vector for time k
% proc_f Handle for process function f(x.k-1)
% proc_Q [n x n] process noise covariance
% meas_h Handle for measurement model function h(x.k)
% meas_R [m x m] measurement noise covariance
%
% Output:
% X_k [n x N] Particles for state x in time k
% W_k [1 x N] Weights for state x in time k

% calculate p(x(i)_k|x(i)_{k-1})
X_k = mvnrnd( proc_f(X_kmin1)',proc_Q ');

% calculate p(y_k|x(i)_k)
Wy = mvnpdf(y_k',meas_h(X_k)',meas_R)';

% compute weights
W_k = W_kmin1 .* Wy;
W_k = W_k / sum(W_k);
end

```

```

function [xfp, Pfp, Xp, Wp] = pfFilter(x_0, P_0, Y, proc_f, proc_Q, meas_h, meas_R, N, bResample, pl
    % PFFILTER Filters measurements Y using the SIS or SIR algorithms and a
    % state-space model.
    %
    % Input:pfFilter(x_0
    %   x_0          [n x 1] Prior mean
    %   P_0          [n x n] Prior covariance
    %   Y            [m x K] Measurement sequence to be filtered
    %   proc_f       Handle for process function f(x.k-1)
    %   proc_Q       [n x n] process noise covariance
    %   meas_h       Handle for measurement model function h(x.k)
    %   meas_R       [m x m] measurement noise covariance
    %   N            Number of particles
    %   bResample    boolean false - no resampling, true - resampling
    %   plotFunc     Handle for plot function that is called when a filter
    %                recursion has finished.
    % Output:
    %   xfp          [n x K] Posterior means of particle filter
    %   Pfp          [n x n x K] Posterior error covariances of particle filter
    %   Xp           [n x N x K] Particles for posterior state distribution in times 1:K
    %   Wp           [N x K] Non-resampled weights for posterior state x in times 1:K

    n = size(x_0,1);
    K = size(Y,2);

    % allocate memory
    xfp = zeros(n,K);
    Pfp = zeros(n,n,K);
    Xp = zeros(n,N,K);
    Wp = zeros(N,K);

    % sample initial particles around prior distribution
    Xp(:, :, 1) = mvnrnd(x_0, P_0, N)';
    Wp(:, 1) = 1/N * ones(1, N);

    j = 1:N;
    for k=2:K+1
        % perform a particle filter step for the next measurement
        [Xp(:, :, k), Wp(:, k)] = pfFilterStep(Xp(:, :, k-1), Wp(:, k-1)', Y(:, k-1), proc_f, proc_Q, meas_h,
        plotFunc(k-1, Xp(:, :, k), Xp(:, :, k-1), Wp(:, k)', j);
        % resample
        if bResample
            [Xp(:, :, k), Wp(:, k), j] = resample(Xp(:, :, k), Wp(:, k)');
        end
        % estimate mean and covariance given the particles
        xfp(:, k) = sum( Xp(:, :, k) .* Wp(:, k)' , 2 );
        Pfp(:, :, k) = Wp(:, k)' .* (Xp(:, :, k) - xfp(:, k)) * (Xp(:, :, k) - xfp(:, k))';
    end

    % remove prior from vector
    xfp = xfp(:, 2:end);
    Pfp = Pfp(:, :, 2:end);
    Xp = Xp(:, :, 2:end);
    Wp = Wp(:, 2:end);
end

```