

$$\left\{ \begin{array}{l} \square \\ r_x = v_x \\ \square \\ r_y = v_y \\ \square \\ v_x = 0 \\ \square \\ v_y = -g + \frac{\rho v_y^2}{2m\beta} \\ \square \\ \Delta h = 0 \\ \square \\ \Delta \beta = 0 \\ \square \\ \Delta \rho_0 = 0 \\ \square \\ \Delta k_p = 0 \end{array} \right.$$

$$F = \left[\begin{array}{cccccccc} 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & -\frac{\alpha}{\hat{k}_p} & 0 & \frac{2\alpha}{\hat{v}_y} & 0 & -\frac{\alpha}{\beta} & \frac{\alpha}{\rho_0} & \frac{\hat{\gamma}_y}{\hat{k}_p^2} \alpha \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

$$\alpha = \frac{\rho \hat{v}_y^2}{2m\beta}$$

$$\phi_{k-1} = I + F \Delta t = \left[\begin{array}{cccccccc} 1 & 0 & \Delta t & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & \Delta t & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & -\frac{\alpha}{k_p} & 0 & 1 + \frac{2\alpha}{v_y} \Delta t & 0 & -\frac{\alpha}{\beta} \Delta t & \frac{\alpha}{\rho_0} \Delta t & \frac{\gamma_y}{k_p^2} \alpha \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{array} \right]$$

$$u_{k-1} = \int_{t_{k-1}}^{t_k} \phi(t_k, \tau) G(\tau) u(\tau) d\tau = -g \begin{bmatrix} 0 \\ \frac{1}{2} \Delta t \\ 0 \\ \Delta t + \frac{\alpha}{v_y} \Delta t^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\mathcal{Q}_{k-1} = \int_{t_{k-1}}^{t_k} \phi(t_k, \tau) \mathcal{Q}_s(\tau) \phi^T(t_k, \tau) d\tau$$

$$\mathcal{Q}_s(\tau) = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & q_{s1} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & q_{s2} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & q_{s3} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & q_{s4} & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & q_{s5} \end{bmatrix}$$

$$\mathcal{Q}_{k-1} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & \mathcal{Q}_{42} & 0 & \mathcal{Q}_{44} & 0 & \mathcal{Q}_{46} & \mathcal{Q}_{47} & \mathcal{Q}_{48} \\ 0 & 0 & 0 & 0 & \mathcal{Q}_{55} & 0 & 0 & 0 \\ 0 & 0 & 0 & \mathcal{Q}_{64} & 0 & \mathcal{Q}_{66} & 0 & 0 \\ 0 & 0 & 0 & \mathcal{Q}_{74} & 0 & 0 & \mathcal{Q}_{77} & 0 \\ 0 & 0 & 0 & \mathcal{Q}_{84} & 0 & 0 & 0 & \mathcal{Q}_{88} \end{bmatrix}$$

$$Q_{42} = \left(\frac{1}{2} \square t^2 + \frac{2}{3} \frac{\alpha}{v_y} \Delta t^3 \right) q_{s1}$$

$$Q_{44} = \left(1 + \frac{2\alpha}{v_y} \Delta t \right) \Delta t \cdot q_{s1} + \dots$$

$$Q_{46} = -\frac{\alpha}{2\beta} q_{s3} \Delta t^2$$

$$Q_{47} = -\frac{\alpha}{2\rho_0} q_{s4} \Delta t^2$$

$$Q_{48} = -\frac{\gamma_y \alpha}{2k_p^2} \Delta t^2$$

$$Q_{55} = q_{s2} \Delta t$$

$$Q_{64} = -\frac{\alpha}{2\beta} \Delta t^2$$

$$q_{66} = q_{s3} \Delta t$$

$$q_{74} = \frac{\alpha}{2\rho_0} q_{s4} \Delta t^2$$

$$q_{77} = q_{s4} \Delta t$$

$$q_{84} = \frac{\gamma_y \alpha}{k_p^2} q_{s5} \Delta t^2$$

$$q_{88} = q_{s5} \Delta t$$

Matlab code for Environment

```
function [stateout, meas] = Envrionment(state0, wk, m)

% Initialization
deltat = 0.1;
g=9.836;
persistent state; %variable to save states of environment

if isempty(state)
    state = state0; %initialize the state using state0 from input.
end

h_ = 2;
bet_ = 150;
rho0_ = 1.225;
kp_ = 9200;
%
ry = state(2);
vy = state(4);
deltah = state(5);
deltabet = state(6);
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deltarho = state(7);
deltakp = state(8);
%
h=h_+deltah;
bet=bet_+deltabet;
rho0=rho0_+deltarho;
kp=kp_+deltakp;
%
alpha = rho0*exp(-ry/kp)*vy*vy/(2*m*bet);
% calculate the state transverse matrix. This is calculated using equation
above and also used in EKF.
phi = [1 0 deltat 0 0 0 0 0; 0 1 0 deltat 0 0 0 0; 0 0 1 0 0 0 0 0; 0 -
alpha*deltat/kp 0 1+rho0*exp(-ry/kp)*vy/(m*bet)*deltat 0 -alpha/bet*deltat
alpha/rho0*deltat ry*alpha*deltat/(kp*kp);
0 0 0 0 1 0 0 0; 0 0 0 0 0 1 0 0; 0 0 0 0 0 0 1 0; 0 0 0 0 0 0 0 1];
u = [0; -g*deltat*deltat/2; 0; -g*deltat*(1+rho0*exp(-ry/kp)*vy/(2*m*bet)); 0;
0; 0; 0];
%update state and mix with process noise wk which is given by input.You can
set the covariance for process noise in the Simulink.
state = phi*state + u + wk;

% Measurement matrix to be used to output measurements.
H = [1 0 0 0 0 0 0 0; 0 1 0 0 0 0 0 0];
b = [0; -h];
%calculate the measurements.
y = H*state + b;

meas = y;
%output the updated state
stateout = state;

```

Matlab code for EKF

```

function [residual,xhatPreOut, xhatOut] = EXTKALMAN(meas)

persistent P;
persistent xhat

if isempty(P)
    xhat = [0; 1000; 0; 0; 0; 0; 0; 0];
    P = zeros(8);
    P(2)=10;
    P(4)=1;
end

g=9.836;
delta_t=0.5;
m = 1;
h_ = 2;
bet_ = 150;
rho0_ = 1.225;
kp_ = 9200;
%get the states of last step to calculate the state transverse matrix.
ry = xhat(2);
vy = xhat(4);
deltah = xhat(5);

```

```

deltabet = xhat(6);
deltarho = xhat(7);
deltakp = xhat(8);
%
h=h_+deltah;
bet=bet_+deltabet;
rho0=rho0_+deltarho;
kp=kp_+deltakp;
%
alpha = rho0*exp(-ry/kp)*vy*vy/(2*m*bet);
%
% 1. Compute Phi, Q, and R
% calculate phi from the equation above. It is same used in environment.
Phi = [1 0 delta_t 0 0 0 0 0; 0 1 0 delta_t 0 0 0 0; 0 0 1 0 0 0 0 0; 0 -
alpha*delta_t/kp 0 1+rho0*exp(-ry/kp)*vy/(m*bet)*delta_t 0 -alpha/bet*delta_t
alpha/rho0*delta_t ry*alpha*delta_t/(kp*kp);
0 0 0 0 1 0 0 0; 0 0 0 0 0 1 0 0; 0 0 0 0 0 0 1 0; 0 0 0 0 0 0 0 1];
Q = diag([0 10 0 1 0 0 0 0]);
R = diag([1 4]);

% 2. Propagate the covariance matrix:

$$\mathbf{P}_k^- = \Phi_{k-1} \mathbf{P}_{k-1}^+ \Phi_{k-1}^T + \mathbf{Q}_{k-1}$$

P = Phi*P*Phi' + Q;

% 3. Propagate the track estimate::
u = [0; -g*delta_t*delta_t/2; 0; -g*delta_t*(1+rho0*exp(-ry/kp)*vy/(2*m*bet));
0; 0; 0; 0];

xhat = Phi*xhat+u;
%output the prio estimates.
xhatPreOut = xhat;
% 4 a). Compute observation estimates:
H = [1 0 0 0 0 0 0 0; 0 1 0 0 0 0 0 0];
b = [0; -h];

$$\mathbf{h}_k(\hat{\mathbf{x}}_k^-)$$

yhat = H*xhat + b;

% 4 c). Compute residual (Estimation Error)

$$\mathbf{y}_k - \mathbf{h}_k(\hat{\mathbf{x}}_k^-)$$

residual = meas - yhat;

% 5. Compute Kalman Gain:

$$\mathbf{K}_k = \mathbf{P}_k^- \mathbf{H}_k^T (\hat{\mathbf{x}}_k^-) \left[ \mathbf{H}_k(\hat{\mathbf{x}}_k^-) \mathbf{P}_k^- \mathbf{H}_k^T (\hat{\mathbf{x}}_k^-) + \mathbf{R}_k \right]^{-1}$$

W = P*H'*inv(H*P*H'+ R);
%
% % 6. Update post estimates

$$\hat{\mathbf{x}}_k^+ = \hat{\mathbf{x}}_k^- + \mathbf{K}_k (\mathbf{y}_k - \mathbf{h}_k(\hat{\mathbf{x}}_k^-))$$

xhat = xhat + W*residual;
%
% % 7. Update Covariance Matrix

$$\mathbf{P}_k^+ = [\mathbf{I} - \mathbf{K}_k \mathbf{H}_k(\hat{\mathbf{x}}_k^-)] \mathbf{P}_k^-$$

P = (eye(8)-W*H)*P*(eye(8)-W*H)' + W*R*W';

```

```
%output the post estimates.  
xhatOut = xhat;
```

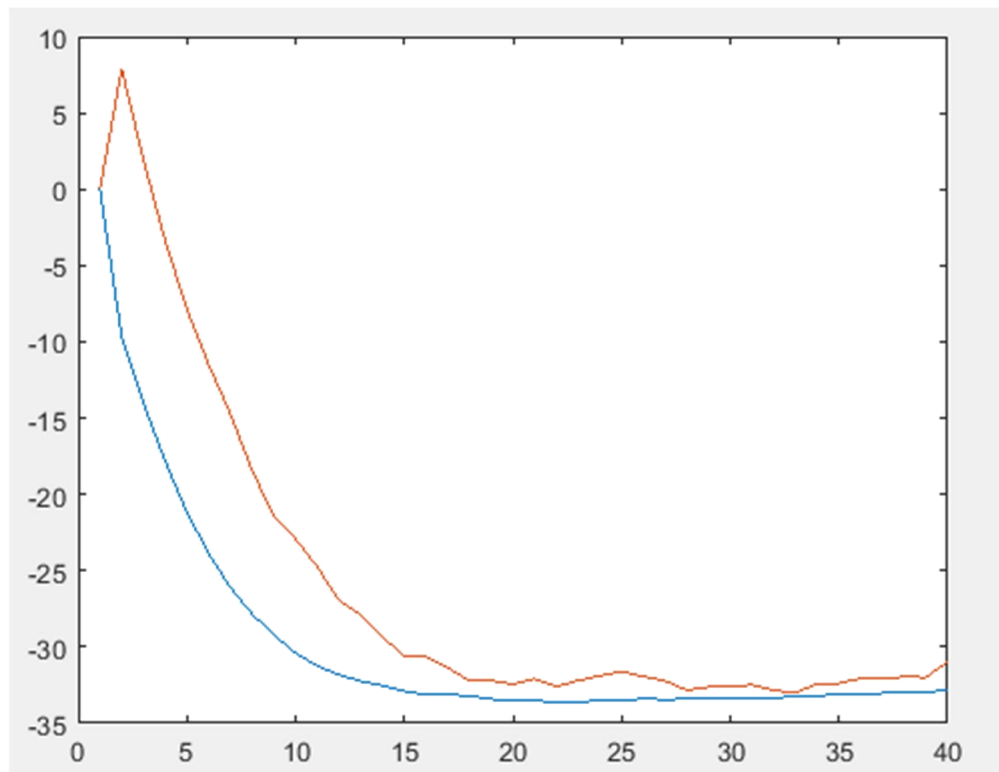


Fig1. \pm standard deviation for v_y

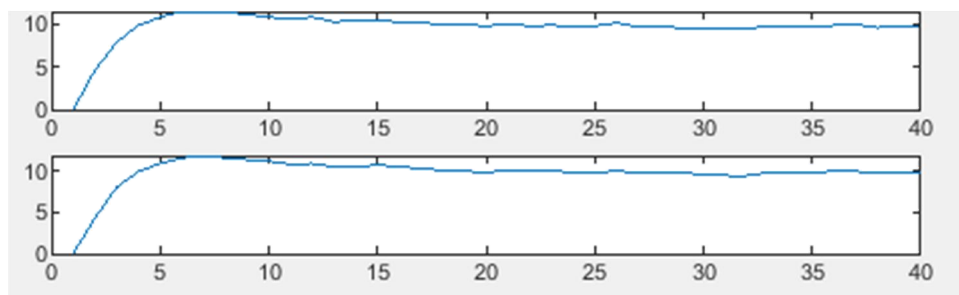


Fig2.covariance for post and prio estimates for v_y

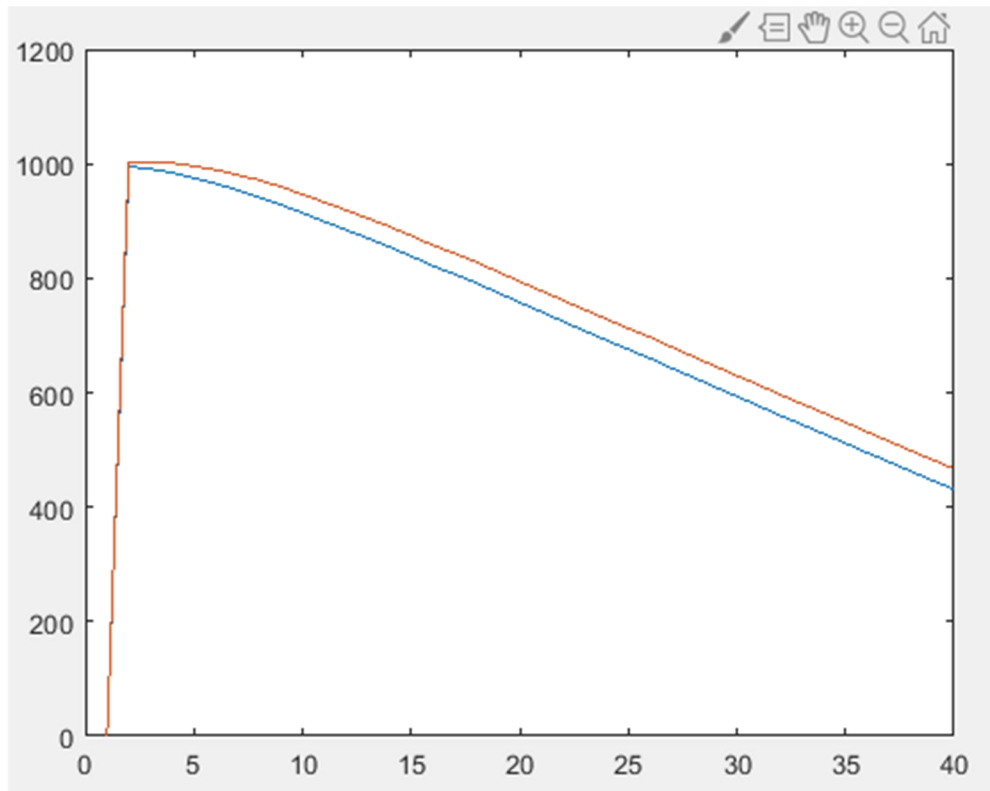


Fig3 \pm standard deviation for r_y

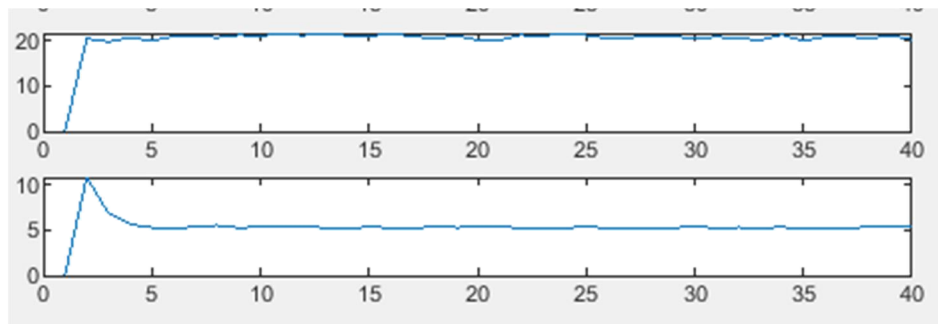


Fig4.covariance for post and prio estimates for r_y