

# Homework 4 Report

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## Problem 1 Results

The solution and derivation for Problem 1 can be found below the analysis of Problem 2 (past the appendix).

## Problem 2 Results

To solve this problem, a Kalman filter was implemented in MATLAB, the code for which can be found in Appendix A. The plots showing  $\pm 1\sigma$  of the position and the velocity covariance at each time are shown in Figure 1 prior to and after each measurement. For the position, we can see that at each second (i.e. at each measurement), the covariance decreases, and then increases until the next measurement. This makes sense because our uncertainty in our position will increase during the propagation step, and then will be reduced once we use the information that the measurement provides. For the velocity plot, the uncertainty decreases between measurements, and then drops even more after the update step. At  $t = 0$  we receive a measurement, which immediately allows us to clamp down on the position uncertainty, but has no effect on the initial standard deviation of velocity. This makes sense because we only have position measurements and have not propagated the states at  $t = 0$ . The final standard deviation of the position is  $0.030582 [m]$  and the final standard deviation of the velocity is  $0.031183 [\frac{m}{s}]$ . The result therefore, is that we are able to achieve a standard deviation for our position that is tighter (lower) than the standard deviation of the measurement noise (which is  $0.1 [m]$ ). This makes sense because the filter is conditioned on all the previous measurements, rather than one individual measurement.

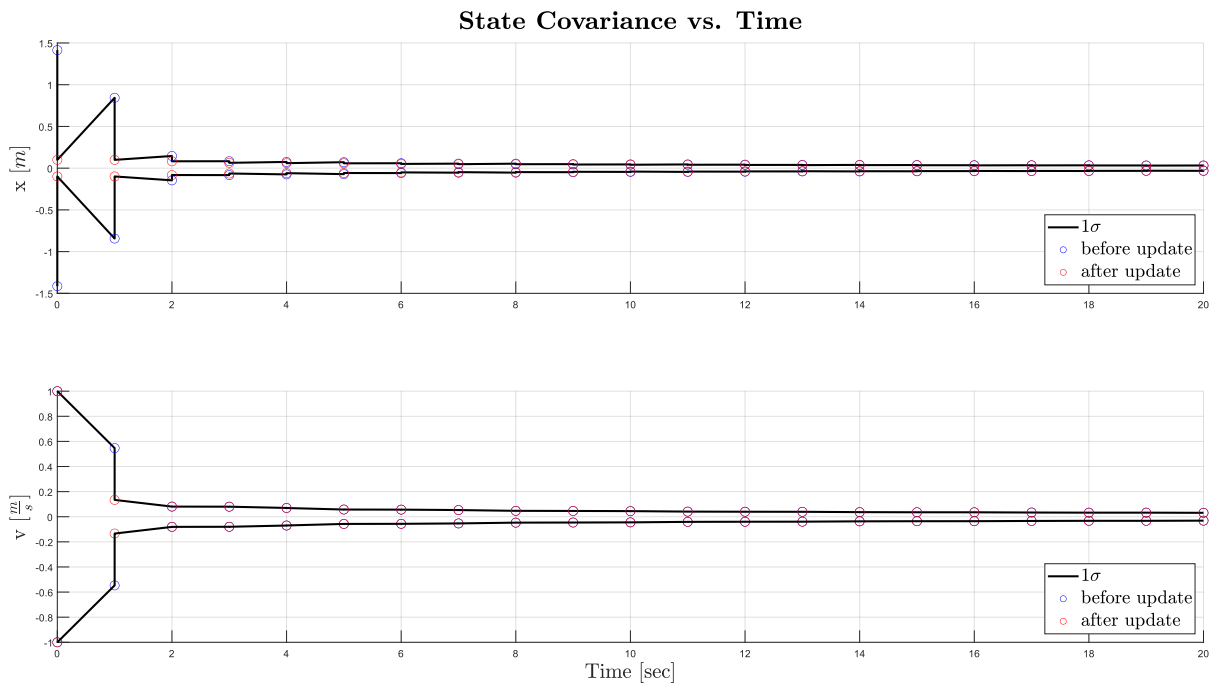
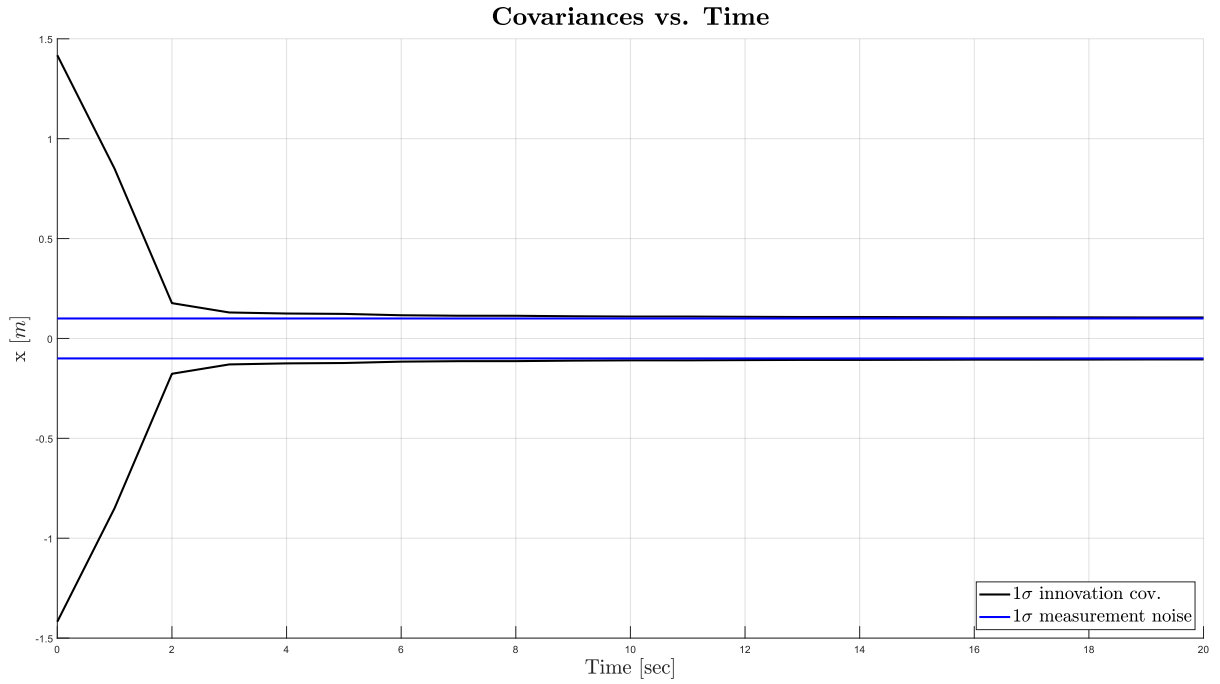


Figure 1: Position and Velocity Covariance versus Time.

The  $\pm 1\sigma$  of the measurement noise covariance (constant for this system) and the innovation covariance is plotted in Figure 2. We can observe that the  $\pm 1\sigma$  of innovation covariance approaches the  $\pm 1\sigma$  of

the measurement noise covariance ( $\sqrt{P_{vv}}$ ) as time increases. At the final time of  $t = 20$ , the  $\pm 1\sigma$  of the innovation covariance is  $0.10503 [m]$ , as compared with the constant  $\pm 1\sigma$  of measurement noise of  $.1 [m]$ . This makes sense because the  $\pm 1\sigma$  of the innovation shouldn't be lower than  $\pm 1\sigma$  of the measurement noise covariance, but will approach this value as more and more measurements are received.



**Figure 2: Measurement noise and Innovation Covariance versus Time.**

## A Matlab Code

```
%% AERO 626 Homework 4
%
% Texas A&M University
% Aerospace Engineering
% van Wijk, David

%% Problem 2

plot_flag = true;

rng(100) % Seed to reproduce results
opts = odeset('AbsTol',1e-6,'RelTol',1e-6);

mx0 = [1; 0];           %[m], [m/s]
Pxx0 = diag([2,1]);     %[m]^2, [m/s]^2
Pvv = .1^2;             %[m]^2 constant covariance

N = 21;
tspan = linspace(0,20,N);
Fx = [0 1; -1 0];
Hx = [1 0];
Hv = 1;
x0 = mx0 + chol(Pxx0)'*randn(2,1);

% Store data
std_x_full = zeros(N*2,2);
std_innov_full = zeros(N,2);

count = 0;
for i = 1:N
    % Propagate the truth state
    if i == 1
        xk = x0;
    else
        [~,X] = ode45(@(t,x) prop(t,x,Fx),[tspan(i-1),tspan(i)],xkm1,opts);
        xk = X(end,1:2)';
    end

    % Generate measurement
    zk = Hx*xk + chol(Pvv)'*randn;

    % Propagate step of Kalman filter
    if i == 1
        mxkm = mx0;
        Pxxkm = Pxx0;
    else
        [~,X] = ode45(@(t,x) prop(t,x,Fx),[tspan(i-1),tspan(i)],[mxkm1;
            Pxxkm1(:)],opts);
        mxkm = X(end,1:2)';
        Pxxkm = reshape(X(end,3:end)',2,2);
    end
end
```

```

end

% Store stuff
count = count + 1;
std_x_full(count,:) = sqrt(diag(Pxxkm))';

mzkm = Hx*mxkm;
Pxzkm = Pxxkm*Hx';
Pzzkm = Hx*Pxxkm*Hx' + Hv*Pvv*Hv';

% Kalman Gain
Kk = Pxzkm/Pzzkm;

% Update mean and covariance
mxkp = mxkm + Kk*(zk - mzkm);
Pxxkp = Pxxkm - Pxzkm*Kk' - Kk*(Pxzkm)' + Kk*(Pzzkm)*Kk';

% Re-initialize for next loop
xkm1 = xk;
mxkm1 = mxkp;
Pxxkm1 = Pxxkp;

% Store stuff
count = count + 1;
std_x_full(count,:) = sqrt(diag(Pxxkp))';
std_innov_full(i,:) = sqrt(diag(Pzzkm))';
end

disp(['The final standard deviation of the position is: ', num2str(
    std_x_full(end,1)), ' [m]'])
disp(['The final standard deviation of the velocity is: ', num2str(
    std_x_full(end,2)), ' [m/s]'])
disp(['The final standard deviation of the innovation covariance is: ',
    num2str(std_innov_full(end,1)), ' [m]'])

if plot_flag
    axis_sz = 20; yaxis_sz = 20; legend_sz = 18;

    % Figure 1
    txt = '1$\sigma$';
    opts_pts1 = {80,'b',"o"};
    opts_pts2 = {80,'r',"o"};

    figure; grid on; set(gcf, 'WindowState', 'maximized');
    subplot(2,1,1); hold on; grid on;
    title('\textbf{State Covariance vs. Time}','FontSize',25,'interpreter',
        '','latex')
    t_mod = sort([tspan tspan]);
    a1 = plot(t_mod,std_x_full(:,1),'-','Color','k','LineWidth',2,'
        MarkerSize',20);
    b1 = scatter(tspan,std_x_full(1:2:end,1),opts_pts1{:});
    b2 = scatter(tspan,std_x_full(2:2:end,1),opts_pts2{:});
    plot(t_mod,-std_x_full(:,1),'-','Color','k','LineWidth',2,'MarkerSize

```

```

    ',20)
scatter(tspan,-std_x_full(1:2:end,1),opts_pts1{:})
scatter(tspan,-std_x_full(2:2:end,1),opts_pts2{:})
ylabel('x [$m$]', 'FontSize',yaxis_sz, 'interpreter','latex')
legendtxt = {txt, 'before update', 'after update'};
legend([a1 b1 b2],legendtxt, 'FontSize',legend_sz, 'interpreter','latex'
    ', 'location','southeast')

subplot(2,1,2); hold on; grid on;
a2 = plot(t_mod,std_x_full(:,2),'-','Color','k','LineWidth',2, '
    MarkerSize',20);
b3 = scatter(tspan,std_x_full(1:2:end,2),opts_pts1{:});
b4 = scatter(tspan,std_x_full(2:2:end,2),opts_pts2{:});
scatter(tspan,-std_x_full(1:2:end,2),opts_pts1{:})
scatter(tspan,-std_x_full(2:2:end,2),opts_pts2{:})
plot(t_mod,-std_x_full(:,2),'-','Color','k','LineWidth',2,'MarkerSize
    ',20)
xlabel('Time [sec]', 'FontSize',xaxis_sz, 'interpreter','latex')
ylabel('v [$\frac{m}{s}$]', 'FontSize',yaxis_sz, 'interpreter','latex')
legend([a2 b3 b4],legendtxt, 'FontSize',legend_sz, 'interpreter','latex'
    ', 'location','southeast')

% Figure 2
figure; grid on; set(gcf, 'WindowState', 'maximized'); hold on;
title('\textbf{Covariances vs. Time}', 'FontSize',25, 'interpreter','
    latex')
a1 = plot(tspan,std_innov_full(:,1),'-','Color','k','LineWidth',2, '
    MarkerSize',20);
plot(tspan,-std_innov_full(:,1),'-','Color','k','LineWidth',2, '
    MarkerSize',20)
a2 = plot(tspan,-ones(N,1)*sqrt(Pvv),'-','Color','b','LineWidth',2, '
    MarkerSize',20);
plot(tspan,ones(N,1)*sqrt(Pvv),'-','Color','b','LineWidth',2, '
    MarkerSize',20)
ylabel('x [$m$]', 'FontSize',yaxis_sz, 'interpreter','latex')
xlabel('Time [sec]', 'FontSize',xaxis_sz, 'interpreter','latex')
legend([a1, a2],{'1$\sigma$ innovation cov.','1$\sigma$ measurement
    noise'}, 'FontSize',legend_sz, 'interpreter','latex', 'location', '
    southeast')

```

end

%% Functions

```

function dx = prop(~,x,Fx)
% Propagate state or mean
dx = Fx*x(1:2);

if length(x) > 2
% Reshape cov. matrix
P = reshape(x(3:end)',2,2);
% Prop cov.
dP = Fx*P + P*Fx';

```

```
dx = [dx; dP(:)];  
end  
end
```