

Kalman Filtering Example

```
clear all

% Set default text interpreter for plots
set(0, 'DefaultTextInterpreter', 'none')
set(0, 'DefaultLegendInterpreter', 'none')
set(0, 'DefaultAxesTickLabelInterpreter', 'none')
```

Define a process model

```
% Sampling period
Ts = 0.5;

% Discrete-time transfer function
Gpd = tf(0.3, [1 -0.7], Ts);
```

Generate some data by simulating the system

```
% Number of sample periods
nT = 20;

% Time vector
t = Ts*(0:nT)';

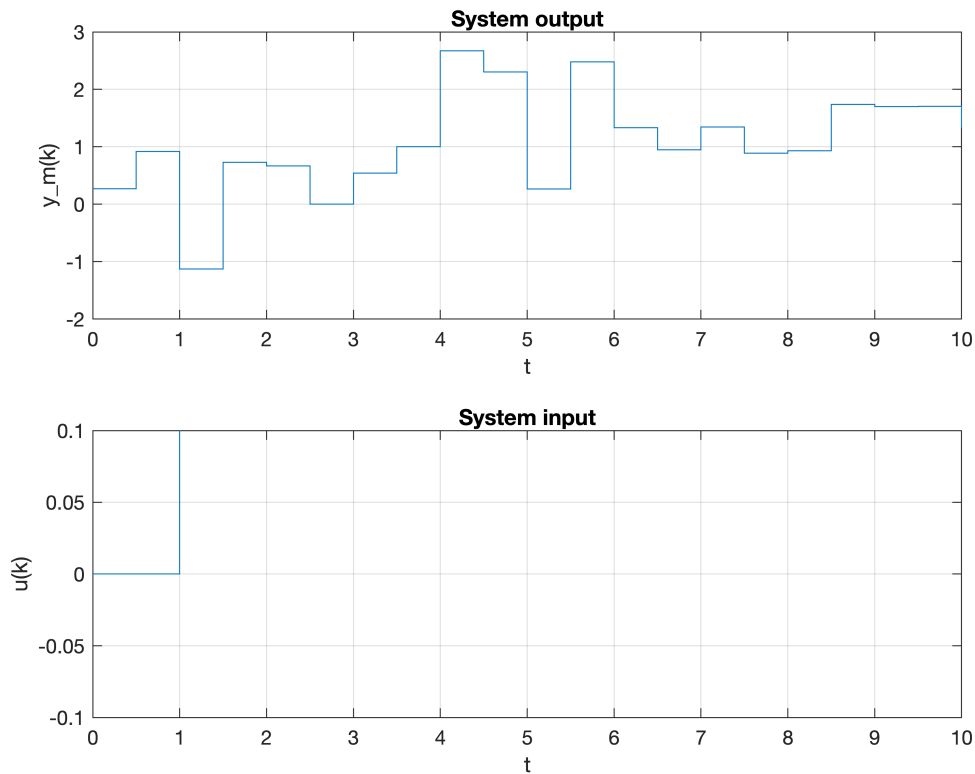
% Input signal
u = zeros(nT+1,1);
u(t>=1) = 1;

[y, t] = lsim(Gpd,u,t);

% Add measurement noise
rng(0);
v = 0.5*randn(nT+1,1);
y_m = y + v;
```

Plot simulation results

```
figure(1); clf
subplot(2,1,1)
stairs(t,y_m)
xlabel('t')
ylabel('y_m(k)')
title('System output')
grid on
subplot(2,1,2)
stairs(t,u)
ylim([-0.1 0.1])
xlabel('t')
ylabel('u(k)')
title('System input')
grid on
```



Define observer

```
% Process model for use by observer
A = 0.7;
B = 1;
C = 0.3;
D = 0;

% Kalman filter parameters
P0 = 1; % initial variance of the estimate
Q = 0.01; % estimated process noise variance
R = 0.5^2; % measurement noise variance
KF1 = KalmanFilter(A,B,C,Ts,P0,Q,R,'KF1');
```

Simulate observer

```
% Array to store simulation results
y_est = nan(nT+1,1);
obs = KF1;
% Initial estimate (at t=0)
y_est(1,:) = obs.ykp1_est;
for i = 1:nT

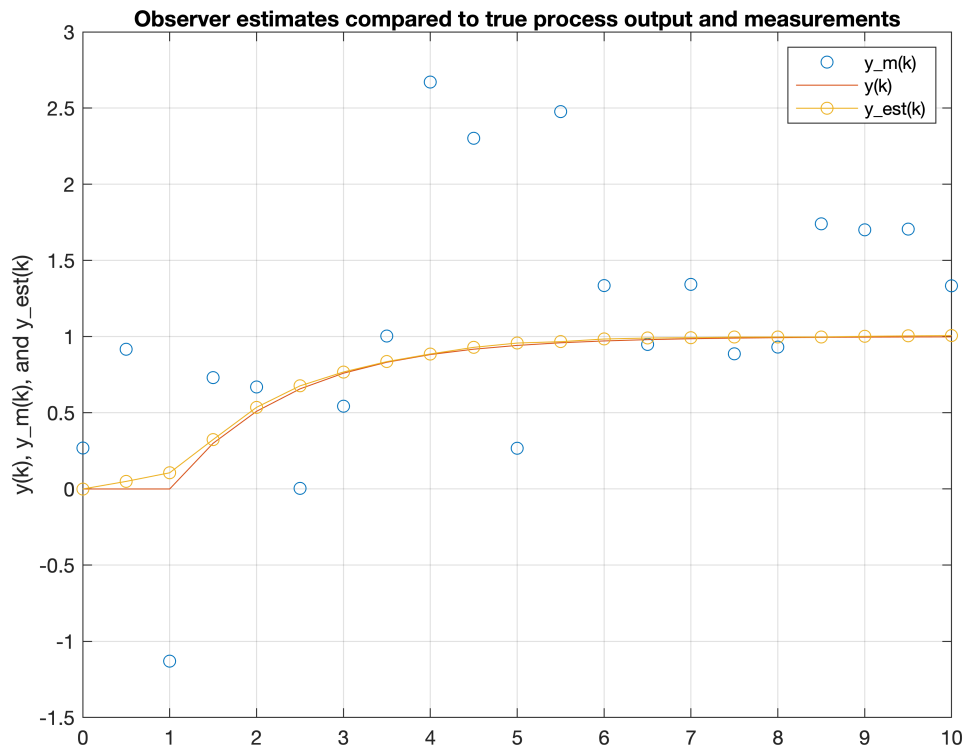
    % update observer
    obs.update(y_m(i), u(i));

    % get estimate of output at next sample time
    y_est(i+1,:) = obs.ykp1_est;
```

```
end
```

Compare observer estimates to plant data

```
figure(2); clf
plot(t,y_m,'o',t,y,'-',t,y_est,'o-')
grid on
ylabel('y(k), y_m(k), and y_est(k)')
legend('y_m(k)', 'y(k)', 'y_est(k)')
title("Observer estimates compared to true process output and measurements")
```



```
% Display simulation results
table(t,u,y,v,y_m,y_est)
```

```
ans = 21×6 table
```

	t	u	y	v	y_m	y_est
1	0	0	0	0.2688	0.2688	0
2	0.5000	0	0	0.9169	0.9169	0.0498
3	1	1	0	-1.1294	-1.1294	0.1063
4	1.5000	1	0.3000	0.4311	0.7311	0.3245
5	2	1	0.5100	0.1594	0.6694	0.5359
6	2.5000	1	0.6570	-0.6538	0.0032	0.6769
7	3	1	0.7599	-0.2168	0.5431	0.7679

	t	u	y	v	y_m	y_est
8	3.5000	1	0.8319	0.1713	1.0032	0.8360
9	4	1	0.8824	1.7892	2.6715	0.8862
10	4.5000	1	0.9176	1.3847	2.3024	0.9298
11	5	1	0.9424	-0.6749	0.2674	0.9578
12	5.5000	1	0.9596	1.5175	2.4771	0.9671
13	6	1	0.9718	0.3627	1.3345	0.9844
14	6.5000	1	0.9802	-0.0315	0.9487	0.9908
15	7	1	0.9862	0.3574	1.3435	0.9933
16	7.5000	1	0.9903	-0.1025	0.8878	0.9970
17	8	1	0.9932	-0.0621	0.9311	0.9974
18	8.5000	1	0.9953	0.7448	1.7401	0.9979
19	9	1	0.9967	0.7045	1.7012	1.0021
20	9.5000	1	0.9977	0.7086	1.7063	1.0049
21	10	1	0.9984	0.3357	1.3341	1.0068