



## General Rules

- Due date is **30 May 2011, 23:55 (sharp)**.
- Late submission is **not** allowed.
- There is **no** teaming up. The homework has to be done/turned in individually.
- In case of cheating, all parts involved (source(s) and receiver(s)) get zero.
- **Use MATLAB installed on department machines** for programming. A tutorial for MATLAB is available at [http://www.mathworks.com/help/pdf\\_doc/matlab/getstart.pdf](http://www.mathworks.com/help/pdf_doc/matlab/getstart.pdf).
- Your report must be prepared with a word processor (latex is recommended). Presentation of your reports will affect your grade.
- You are supposed to submit your MATLAB codes and reports (pdf) electronically as one zip file which is named as “studentid.zip” through COW (<https://cow.ceng.metu.edu.tr>). (The files name should be main.m, particle.m, k\_predict.m, k\_update.m and hw3.pdf (for report) ).
- Also submit hard copy of your report.
- **Warning: Hard copy of report will also be graded.**

## Homework

Tracking an object is difficult because of noise. Noise can arise from the movement of the object, called process noise, and from sensors, called measurement noise. Kalman filtering is a well-known method to estimate true state values from noisy data. Kalman filter can be seen as a recursive estimator as it only uses the estimated state from the previous time step and the current measurement to estimate the current state.

Suppose you have a linear system model as described in homework 2. This object (particle) is tracked with a sensor, such as a radar, that only measures the position of the object. By using the measurements, you are asked to estimate the true state of the object with a Kalman filter.

The `particle` function in homework 2 is to be used to create the trajectory data with process noise. The system matrix in homework 2 should also be used in the Kalman filter.

Create two files to be named `k_predict.m` and `k_update.m`. These files must hold the functions whose prototypes are given below:

```
function [Zpredict,Ppredict] = k_predict(Zestimate,A,P,Q)
....
end
```

```
function [Zupdate,Pupdate] = k_update(Zpredict,Ppredict,ZN,H,R)
```

```
....  
end
```

For the `k_predict` function,  $Z_{estimate}$  is the estimated state at time  $k - 1$ ,  $A$  is the system matrix,  $P$  is error covariance matrix at time  $k - 1$ ,  $Q$  is the covariance matrix of the process noise. The forces in homework 2 should be process noise obtained from the Gaussian distribution with zero mean and covariance  $Q$ .  $Z_{predict}$  is predicted state estimate at time  $k$  and  $P_{predict}$  is predicted estimate error covariance at time  $k$ .

For the `k_update` function,  $ZN$  is the measurement data at time  $k$ ,  $H$  is the measurement model,  $R$  is the covariance matrix of measurement noise,  $Z_{update}$  is updated state estimate at time  $k$ ,  $P_{update}$  is updated estimate error covariance at time  $k$ .

Measurement data (at time  $k$ ) can be obtained as follows:

$$ZN^{\rightarrow}(k)_{2 \times 1} = H_{2 \times 4} * z^{\rightarrow}(k)_{4 \times 1} + V^{\rightarrow}(k)_{2 \times 1}$$

where  $z$  is state vector,  $H$  is the measurement model,  $V$  is the measurement noise obtained from the Gaussian distribution with zero mean and covariance  $R$ .

Evaluate this system over 100 time steps and estimate the position of the object from the (noisy) measurement data at each time step.

Modify the `main.m` in homework 2 so that it now uses the `particle.m`, `k_predict.m` and `k_update.m`.

Assume that the covariance matrix  $Q$  of process noise is the diagonal matrix with 0.02 along the diagonal, the covariance matrix  $R$  of measurement noise is the diagonal matrix with 5,  $T$  (time interval) is 1,  $m = 0.5$ . The initial position is given by  $x = 0$  and  $y = 0$ , and initial velocities in both directions are 1. We are assuming proper units.

Plot the  $x$  positions of the true state ( $z$ ), estimated state ( $Z_{estimate}$ ) and the measurement of true state ( $ZN$ ) at the same figure with different colors over the whole duration. For example:

```
plot(z(1,:), 'b')  
hold on  
plot(Zestimate(1,:), 'r')  
hold on  
plot(ZN(1,:), 'g')
```

Plot the  $y$  positions of the true state ( $z$ ), estimated state ( $Z_{estimate}$ ) and the observation of true state ( $ZN$ ) at the same figure with different colors over the whole duration. For example:

```
figure  
plot(z(2,:), 'b')
```

```
hold on
plot(Zestimate (2,:), 'r')
hold on
plot(ZN (2,:), 'g')
```

In the report describe your system matrix, distribution matrix, input vector, state vector briefly and explain the steps of the Kalman filter clearly. Produce the plots with the given parameters: Two different plots, the first one is for x position and the second one is for y position.