

# Hand in problem 3

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December 17, 2022

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The secret key is 'Pyukumuku'.

## Task 1

- For the state-space model of the form

$$s(k+1) = As(k) + w(k) \quad (1)$$

$$z(k) = Cs(k) + v(k) \quad (2)$$

- The matrices A, C and w(k) are given as follows:

$$A = \begin{bmatrix} 1 & T & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & T \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad (3)$$

$$C = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \quad (4)$$

- The process noise  $w(k)$  and measurement noise  $v(k)$  are zero mean random variables with a normal distribution and covariance matrices  $Q$  and  $R$  independent of the state variable  $s(k)$ .

$$\begin{bmatrix} w(k) \\ v(k) \end{bmatrix} \sim N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} Q & 0 \\ 0 & R \end{bmatrix}\right) \quad (5)$$

## Task 2

- The measured and noise free positions are plotted as follows:

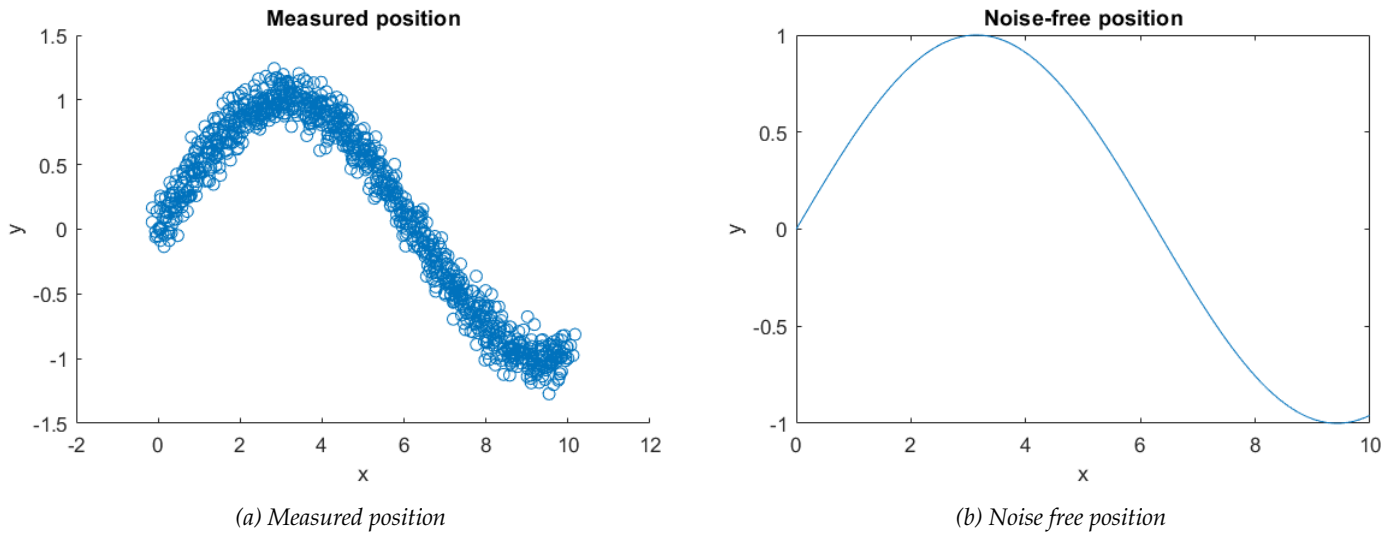


Figure 1: Position data

### Task 3

- The completed code for the Kalman filter is as follows in the figure 2.

```

for t=1:N
    % Filter update based on measurement
    % Xfilt(:,t) = Xpred(:,t) + ...
    Xfilt(:,t) = Xpred(:,t) + P*C'*inv(C*P*C'+R)*(Y(:,t)-C*Xpred(:,t));

    % Uncertainty update
    Pplus = P - P*C'*inv(C*P*C'+R)*C*P; %TODO: This line is missing some code!

    % Prediction
    Xpred(:,t+1) = A * Xfilt(:,t); %TODO: This line is missing some code!

    % Uncertainty propagation
    P = A*Pplus*A' + Q; %TODO: This line is missing some code!
end

```

Figure 2: Kalman Filter

### Task 4

- The covariance matrix  $R$  is diagonal in nature and only affects the  $x$  and  $y$  position of the object. By calculating the variance of  $x$  and  $y$  positions, we get

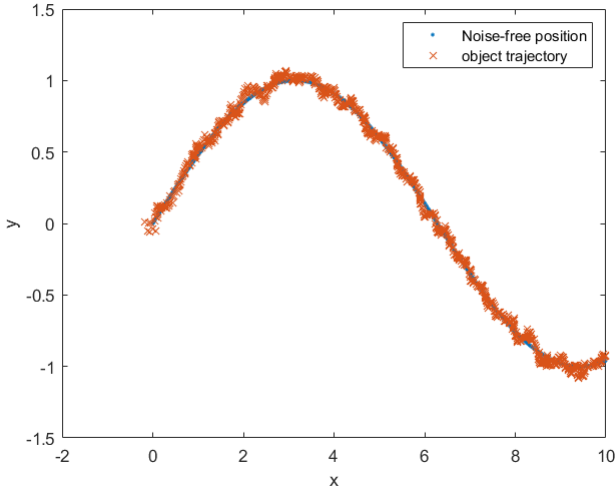
$$R \approx \begin{bmatrix} \sigma_x^2 & 0 \\ 0 & \sigma_y^2 \end{bmatrix} = \begin{bmatrix} 9.6054e^{-3} & 0 \\ 0 & 9.4060e^{-3} \end{bmatrix} \quad (6)$$

- Since the  $Q$  matrix should be selected to model possible speed changes, it is of the following form.

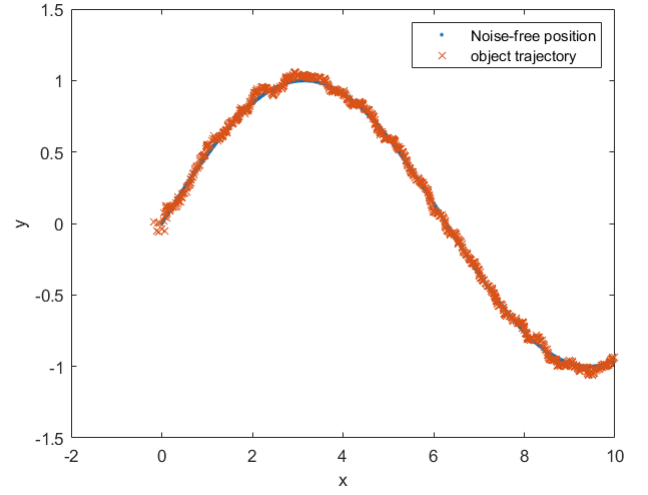
$$Q = \gamma * \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad (7)$$

- By changing the value of  $\gamma$ , different behaviour of Kalman filter was observed as shown in figure 3.
- With  $\gamma = 1e^{-4}$ , the best results were found. Therefore the covariance matrix  $Q$  was chosen as

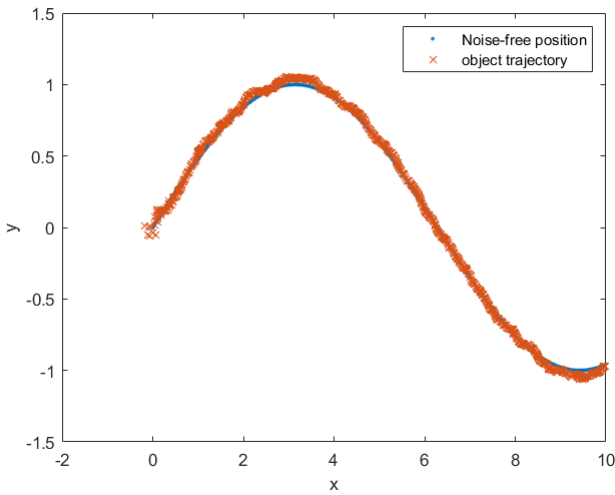
$$Q = 1e^{-4} * \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad (8)$$



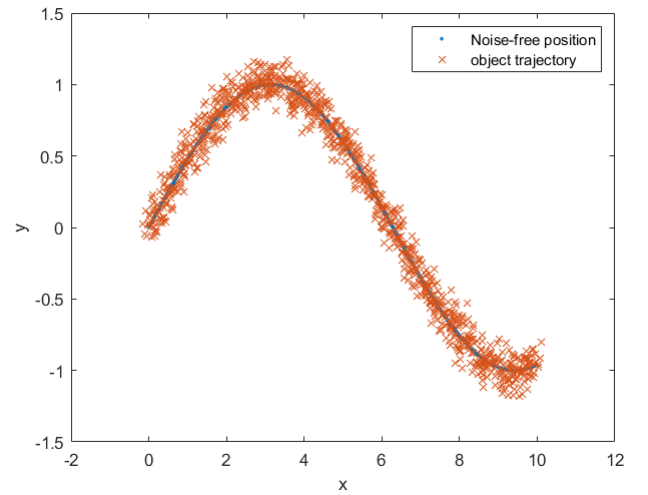
(a) Filter behaviour with  $\gamma = 1e^{-2}$



(b) Filter behaviour with  $\gamma = 1e^{-3}$



(c) Filter behaviour with  $\gamma = 1e^{-4}$



(d) Filter behaviour with  $\gamma = 1e^{+2}$

Figure 3: Kalman Filter behaviour

- The estimated speed of the object in  $x$  and  $y$  direction is plotted in figure 4.

- The value of  $\gamma$  chosen in the previous section also had an impact of how fast the speed was estimated. Best compromise on speed versus accuracy was arrived with  $\gamma = 1e^{-4}$ .

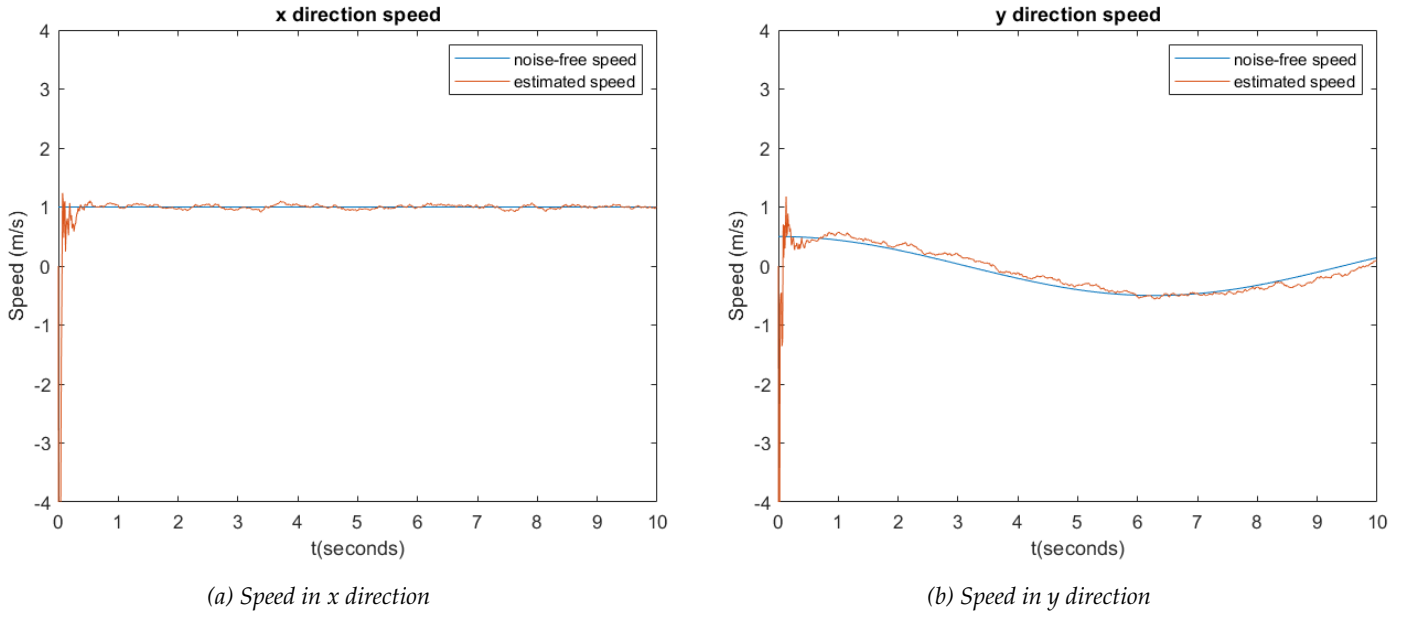


Figure 4: Speed of the object