

2) KALMAN FILTER 2D IMPLEMENTATION:

$\hat{x}_k^- = A\hat{x}_{k-1}^- + B\hat{u}_{k-1}^-$: (Time Update Stage \rightarrow Project the state ahead)

$$\rightarrow \hat{x}_k^- = \begin{pmatrix} 1 & \Delta T \\ 0 & 1 \end{pmatrix} \hat{x}_{k-1}^- + \begin{pmatrix} 0 & \frac{\Delta T^2}{2} \\ 0 & \Delta T \end{pmatrix} \hat{u}_{k-1}^-$$

- We can guess any points for this 2*1 matrix: \hat{x}_{k-1}^- and for this question \hat{u}_{k-1}^- is given.

$P_k^- = AP_{k-1}A^T + Q$: (Time Update Stage \rightarrow Project the error covariance ahead, initiate estimates for \hat{x}_{k-1}^- and P_{k-1})

$\rightarrow P_{k-1}$ will be scalar * I (Identity matrix) $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$. We have to guess the scalar.

$\rightarrow \begin{pmatrix} 1 & \Delta T \\ 0 & 1 \end{pmatrix} P_{k-1} \begin{pmatrix} 1 & 0 \\ \Delta T & 1 \end{pmatrix} + Q$. Q is given in this problem.

After getting P_k^- using above equation we now will find Kalman Gain.

$$K_k = \frac{P_k^- H^T}{H P_k^- H^T + R}$$

(Measurement Update (“Correct”) Stage \rightarrow Computing the Kalman Gain)

\rightarrow At this point, we will know the value of P_k^- using above equations and **H** will be Identity Matrix $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ and value of **R** is given in this problem.

After finding Kalman gain, we will find new \hat{x}_k

$\text{New } \hat{x}_k = \hat{x}_k^- + K_k(z_k - H\hat{x}_k^-)$: (Measurement Update (“Correct”) Stage \rightarrow Updating estimate with measurement z_k), Value of z_k is given in this problem.

Now, we will find new P_k^- : ((Measurement Update (“Correct”) Stage \rightarrow Updating the error covariance)

$$\text{New } P_k = (I - K_k H) P_k^-$$

\rightarrow Now, for the next iteration our new \hat{x}_k and P_k value will be used for \hat{x}_k^- and P_k^- respectively. Continue this process to find further values. Store the data of \hat{x}_k and plot \hat{x}_k and z_k using **matplotlib**.