

# Machine Problem 5

## 2D Kalman Filter: An Application

Kislaya Joshi (**kj276**) and Shreya Bajpai (**sb1437**)

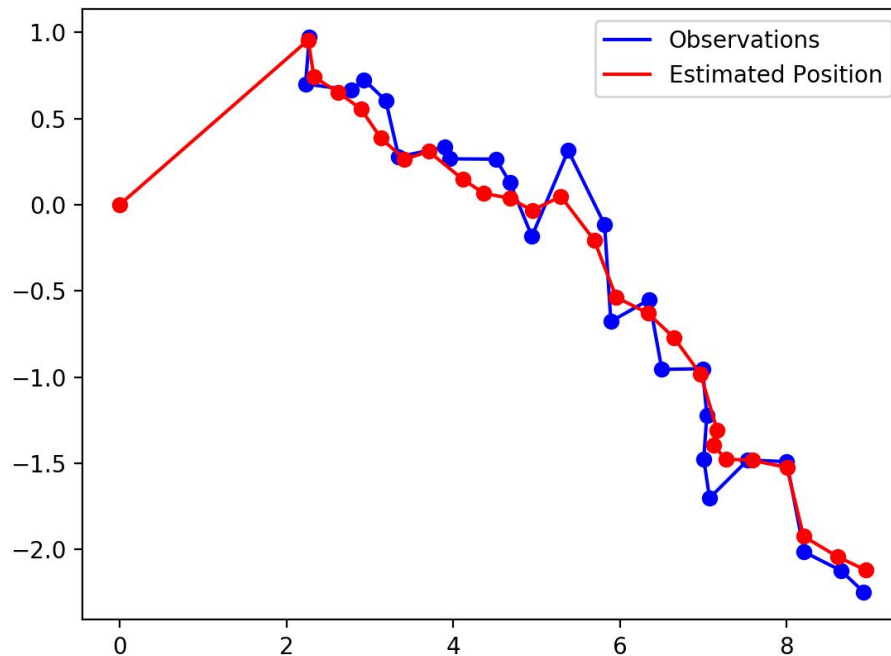
---

### A1.

1. For calculating the updated equations we use the equations for iterative computation.
2. For the time update we use the equations:  
$$X'(k) = A * X'(k-1) + B * U(k-1)$$
$$P'(k) = A * P(k-1) * \text{transpose}(A) + Q$$
3. For the measurement update we first calculate the kalman gain with the equation  
$$\text{Kalman\_gain} = P'(k) * \text{transpose}(H) / (H * P'(k) * \text{transpose}(H) + R)$$
4. Now using the kalman gain we can calculate the updated measurements, ie:  
$$X(k) = X'(k) + \text{Kalman\_gain} * (z(k) - H * X'(k))$$
$$P(k) = (I - \text{Kalman\_gain} * H) P'(K)$$

Where A, B and H are identity matrices of size 2 and Q and R are the given covariance matrices.  
Our initial P is Lambda\*Identity where lambda is 2.

**A3.** The plot asked in A4 is included below.



#### A4.

The value of  $P$  used is  $\text{Lambda} \times \text{Identity matrix}$  where our  $\text{lambda}$  value is 2 so the  $P$  value we used is  $2 \times I$ .

#### B2.

Since the Kalman Filter is a predictor-corrector type estimator, we shoot only when it is optimal, which is when it minimizes the **estimated error covariance**. This is when the estimated error in  $x$  is  $< 0.0015$  and the estimated error in  $y < 0.014$ .

#### B3.

Not implemented.