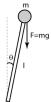
CMPE 252, Artificial Intelligence and Data Engineering, Spring 2023

Home Assignment 2
Probabilistic Reasoning in Dynamical Systems, Kalman Filter, due: March, 15, 2023, 11:59pm.

Setting



Equation-Of-Motion of the pendulum:
$$\begin{bmatrix} \dot{\theta}_1(t) \\ \dot{\theta}_2(t) \end{bmatrix} = \begin{bmatrix} \dot{\theta}_2(t) \\ -\frac{g}{\ell} \sin \theta_1(t) \end{bmatrix}$$
 where $\theta_1(t)$ and $\theta_2(t)$ are the angle and the angular velocity.

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EOM linearised around the bottom with $x(t) = [\theta_1(t); \theta_2(t)]$:

$$\dot{x}(t) = egin{bmatrix} 0 & 1 \ -rac{g}{\ell} & 0 \end{bmatrix} x(t) + \eta, \text{ where } \eta \text{ is Gaussian noise}$$

EOM in discrete time with the linear observation model. The sensor reading, y[k], provides the angular velocity with the Gaussian noise ν .

Dynamics model:
$$x[k+1] = A_Dx[k] + \mu[k]$$

Observation model:
$$y[k] = Cx[k] + \nu[k]$$



Parameters

$$A_D = \begin{pmatrix} 0.99951 & 0.00999837 \\ -0.097984 & 0.99951 \end{pmatrix}, C = [0, 1]$$

▶ The covariance matrix of η :

$$W = \begin{pmatrix} 1.00207e^{-7} & 4.31094e^{-8} \\ 4.31094e^{-8} & 9.6011e^{-6} \end{pmatrix}$$

▶ The mean vector of η :

$$\mu_{\eta} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

▶ The variance of ν :

$$V = (0.01)$$

▶ The mean of ν : $\mu_{\nu} = 0$

Questions

- 1 Starting from x[0]=[0.1;0.0], simulate the pendulum dynamics for T=499 steps, where $\nu[k]$ is sampled from the two dimensional Gaussian with the mean μ_{η} and the covariance W, and $\nu[k]$ is sampled from the one dimensional Gaussian with the mean μ_{ν} , and variance, V. The simulated state and observation trajectories are arrays with dimensions 2×500 and 1×500 , respectively. (In practice, the true state is unknown.)
- 2 Plot on the same figure the true angle, x[1,:], the true angular velocity, x[2,:], and the measured noisy angular velocity, y[:], as a function of time. Add the legend.

Questions

- 3 State reconstruction by Kalman filter:
 - 3.1 Calculate recursively the state estimation covariance matrix, $\Sigma_{t+1|t}$, for $t \in [1,500]$. Assume $\Sigma_{0|-1} = 1e^{-3} * \text{'IdentityMartrix'}_{2\times 2}$. See Slide 8-18 at https://stanford.edu/class/ee363/lectures/kf.pdf
 - 3.2 Reconstruct the full state, $\hat{x}_{t+1|t}$, for $t \in [1,500]$ from the sensor readings, y[:], only. Assume $\hat{x}_{0|-1} = [0.1;0]$. See Slide 8-20. (x[1,:] and x[2,:] are unknown/hidden.)
- 4 Add to the previous plot $\hat{x}_{t+1|t}[1,:]$, $\hat{x}_{t+1|t}[2,:]$, and $\hat{y}_{t+1|t}[:]$. Add legend. Explain your observations.
- 5 Repeat 10 times the questions 1, 3.1, and 3.2. Plots all the curves on the same plot as a function of time, and explain the meaning of the variance of the curves for $\hat{x}_{t+1|t}[1,:]$, $\hat{x}_{t+1|t}[2,:]$, and $\hat{y}_{t+1|t}[:]$.

What to submit

- ▶ Your code in ipynb, which enables to reconstruct your plots.
- ► A pdf file with the plot for Q2+Q4, the plot for Q5, the explanations, and your full code.
- ► Submit these two files separately (not in a zip file) in Canvas.