

4.5. a. Using Eq 4.27

$$Q_{k+1} = \int_{t_{k+1}}^{t_k} e^{A(t_k - \tau)} Q_c(\tau) e^{A^T(t_k - \tau)} d\tau$$

$$A = \begin{bmatrix} -F_0/2v_0 & 0 \\ 0 & -F_0/v_0 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix}$$

$$B = \begin{bmatrix} 1 & 1 \\ \frac{c_1 - c_0}{v_0} & \frac{c_2 - c_0}{v_0} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}$$

$$Q_c = B \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} B^T = \begin{bmatrix} 2 & 3 \\ 3 & 5 \end{bmatrix}$$

We know that matrix exponential e^M of a diagonal matrix

$$M = \begin{bmatrix} m_{11} & 0 \\ 0 & m_{22} \end{bmatrix} \text{ is of the form } e^M = \begin{bmatrix} e^{m_{11}} & 0 \\ 0 & e^{m_{22}} \end{bmatrix}$$

$$\therefore e^{A(t_k - \tau)} = \begin{bmatrix} e^{\tau - t_k} & 0 \\ 0 & e^{2\tau - 2t_k} \end{bmatrix} = e^{A^T(t_k - \tau)}$$

$$\begin{aligned} e^{A(t_k - \tau)} Q_c(\tau) e^{A(t_k - \tau)} &= \begin{bmatrix} e^{\tau - t_k} & 0 \\ 0 & e^{2\tau - 2t_k} \end{bmatrix} \begin{bmatrix} 2 & 3 \\ 3 & 5 \end{bmatrix} \begin{bmatrix} e^{\tau - t_k} & 0 \\ 0 & e^{2\tau - 2t_k} \end{bmatrix} \\ &= \begin{bmatrix} 2e^{\tau - t_k} & 3e^{\tau - t_k} \\ 3e^{2\tau - 2t_k} & 5e^{2\tau - 2t_k} \end{bmatrix} \begin{bmatrix} e^{\tau - t_k} & 0 \\ 0 & e^{2\tau - 2t_k} \end{bmatrix} \\ &= \begin{bmatrix} 2e^{2\tau - 2t_k} & 3e^{3\tau - 3t_k} \\ 3e^{3\tau - 3t_k} & 5e^{4\tau - 4t_k} \end{bmatrix} \end{aligned}$$

$$\therefore Q_{k+1} = \int_{t_{k+1}}^{t_k} \begin{bmatrix} 2e^{2\tau - 2t_k} & 3e^{3\tau - 3t_k} \\ 3e^{3\tau - 3t_k} & 5e^{4\tau - 4t_k} \end{bmatrix} d\tau = \begin{bmatrix} 1 - e^{-2t_k + 2t_{k+1}} & 1 - e^{-3t_k + 3t_{k+1}} \\ 1 - e^{-3t_k + 3t_{k+1}} & \underline{\underline{\frac{5}{4}(1 - e^{4t_k - 4t_{k+1}})}} \end{bmatrix}$$

b. Using Eq 4.28

$$Q_{k-1} = Q_c(t_k) \Delta t = \begin{bmatrix} 2 & 3 \\ 3 & 5 \end{bmatrix} (t_k - t_{k-1}) = \begin{bmatrix} 2(t_k - t_{k-1}) & 3(t_k - t_{k-1}) \\ 3(t_k - t_{k-1}) & 5(t_k - t_{k-1}) \end{bmatrix}$$

c. We can expand the solution of (a) using a Taylor series as follows:

$$Q_{k-1} = \begin{bmatrix} 1 - (1 - 2\Delta t + (2\Delta t)^2/2! + (2\Delta t)^3/3! + \dots) & 1 - (1 - 3\Delta t + (3\Delta t)^2/2! + (3\Delta t)^3/3! + \dots) \\ 1 - (1 - 3\Delta t + (3\Delta t)^2/2! + (3\Delta t)^3/3! + \dots) & \frac{5}{4} - \frac{5}{4}(1 - 4\Delta t + (4\Delta t)^2/2! + \dots) \end{bmatrix}$$

where $\Delta t = t_k - t_{k-1}$

From this representation, we can see that for small Δt higher order terms can be ignored, making our result approximately equal to the result from part (b).

\therefore Answer to part (a) \approx Answer to part (b)

$$4.8. F = \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \end{bmatrix}$$

$$Q = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

$$i=0 \Rightarrow F^0 Q (F^T)^0 = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

$$i=1 \Rightarrow \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \end{bmatrix} = \begin{bmatrix} \frac{1}{4} & \frac{1}{4} \\ 0 & \frac{1}{4} \end{bmatrix}$$

$$i=2 \Rightarrow \begin{bmatrix} \frac{1}{2}^2 & 0 \\ 0 & \frac{1}{2}^2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2}^2 & 0 \\ 0 & \frac{1}{2}^2 \end{bmatrix} = \begin{bmatrix} \frac{1}{16} & \frac{1}{16} \\ 0 & \frac{1}{16} \end{bmatrix}$$

$$i=3 \Rightarrow \begin{bmatrix} \frac{1}{2}^3 & 0 \\ 0 & \frac{1}{2}^3 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2}^3 & 0 \\ 0 & \frac{1}{2}^3 \end{bmatrix} = \begin{bmatrix} \frac{1}{64} & \frac{1}{64} \\ 0 & \frac{1}{64} \end{bmatrix}$$

$$i=4 \Rightarrow \begin{bmatrix} \frac{1}{2}^4 & 0 \\ 0 & \frac{1}{2}^4 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2}^4 & 0 \\ 0 & \frac{1}{2}^4 \end{bmatrix} = \begin{bmatrix} \frac{1}{256} & \frac{1}{256} \\ 0 & \frac{1}{256} \end{bmatrix}$$

$$i=5 \Rightarrow \begin{bmatrix} \frac{1}{2}^5 & 0 \\ 0 & \frac{1}{2}^5 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2}^5 & 0 \\ 0 & \frac{1}{2}^5 \end{bmatrix} = \begin{bmatrix} \frac{1}{1024} & \frac{1}{1024} \\ 0 & \frac{1}{1024} \end{bmatrix}$$

$$i=6 \Rightarrow \begin{bmatrix} \frac{1}{2}^6 & 0 \\ 0 & \frac{1}{2}^6 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2}^6 & 0 \\ 0 & \frac{1}{2}^6 \end{bmatrix} = \begin{bmatrix} \frac{1}{4096} & \frac{1}{4096} \\ 0 & \frac{1}{4096} \end{bmatrix}$$

$$i=6 \Rightarrow \begin{bmatrix} \frac{1}{2^6} & 0 \\ 0 & \frac{1}{2^6} \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2^6} & 0 \\ 0 & \frac{1}{2^6} \end{bmatrix} = \begin{bmatrix} \frac{1}{4096} & \frac{1}{4096} \\ 0 & \frac{1}{4096} \end{bmatrix}$$

$$i=7 \Rightarrow \begin{bmatrix} \frac{1}{2^7} & 0 \\ 0 & \frac{1}{2^7} \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2^7} & 0 \\ 0 & \frac{1}{2^7} \end{bmatrix} = \begin{bmatrix} \frac{1}{16384} & \frac{1}{16384} \\ 0 & \frac{1}{16384} \end{bmatrix}$$

⋮

P begins to converge by $i=7$.

Adding the matrices, we find that

$$P = \sum_{i=0}^{\infty} F^i Q (F^T)^i \approx \begin{bmatrix} 1.3333 & 0 \\ 0 & 1.3333 \end{bmatrix}$$

4.11. Using $\bar{x}_0 = 1$

$$P_0 = 2$$

$$f = -0.5$$

$$q_c = 1$$

$$\Delta t = 0.01s$$

$$t_f = 5s$$

and using Eqn 4.32

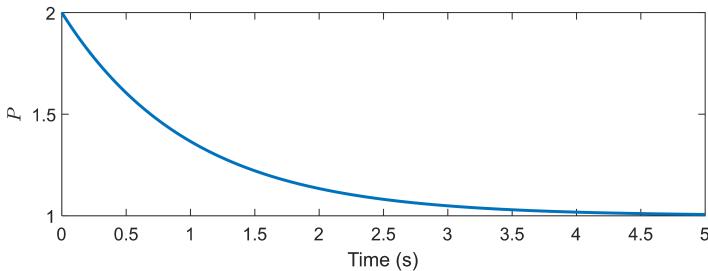
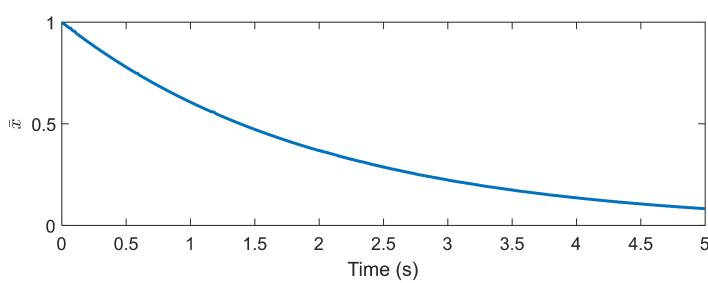
$$\bar{x}_k = \exp(kf\Delta t) \bar{x}_0$$

and Eqn 4.33

$$P_k = (2fP_{k-1} + q_c)\Delta t + P_{k-1}$$

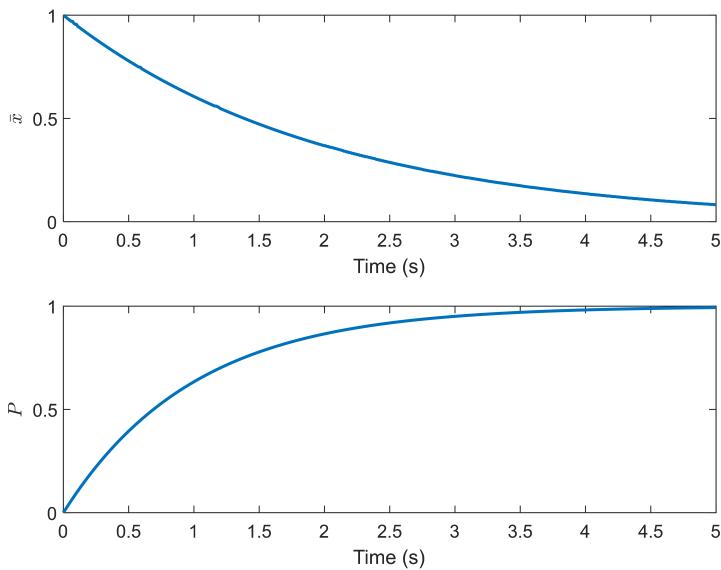
the following plots were found for \bar{x} and P :

\bar{x} and P when $P_0 = 2$



With $P_0 = 0$ the following plots were found:

\bar{x} and P when $P_0 = 0$



In both plots, the steady-state value of the variance appears to be 1, which is consistent with the result of the analytically determined steady-state variance using the equation from the text:

$$P_{k-1} = -\frac{q_c}{2f} = -\frac{1}{2 * -0.5} = 1$$

4.13.a. Using Eq 4.49

$$\dot{P} = AP + PA^T + Q_C$$

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1/C \\ -R/L & -R/L \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1/L \end{bmatrix} u$$

Using $R=3$, $L=1$, $C=0.5$

$$A = \begin{bmatrix} 0 & 2 \\ -1 & -3 \end{bmatrix}$$

$$Q_C = B q_c B^T = \begin{bmatrix} 0 \\ 1 \end{bmatrix} (1) \begin{bmatrix} 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

$$AP + PA^T + Q_C = 0$$

$$\begin{bmatrix} 0 & 2 \\ -1 & -3 \end{bmatrix} P + P \begin{bmatrix} 0 & -1 \\ 2 & -3 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & -1 \end{bmatrix}$$

*Solved using MATLAB

$$P = \begin{bmatrix} \frac{1}{3} & 0 \\ 0 & \frac{1}{6} \end{bmatrix}$$

b. $A = \begin{bmatrix} 0 & 2 \\ -1 & -3 \end{bmatrix}$

$$\det(A - \lambda I) = \det \begin{bmatrix} -\lambda & 2 \\ -1 & -3-\lambda \end{bmatrix}$$

$$= (-\lambda)(-3-\lambda) + 2$$

$$= \lambda^2 + 3\lambda + 2$$

$$= (\lambda+1)(\lambda+2)$$

$$\ker(A + I) = \ker \begin{bmatrix} 1 & 2 \\ -1 & -2 \end{bmatrix} = \begin{bmatrix} -0.8944 \\ 0.4472 \end{bmatrix}$$

$$\ker(A + 2I) = \ker \begin{bmatrix} 2 & 2 \\ -1 & -1 \end{bmatrix} = \begin{bmatrix} -0.7071 \\ 0.7071 \end{bmatrix}$$

$$\therefore S = \begin{bmatrix} -0.8944 & -0.7071 \\ 0.4472 & 0.7071 \end{bmatrix}$$

$$\hat{A} = S^{-1} A S = \begin{bmatrix} -0.8944 & -0.7071 \\ 0.4472 & 0.7071 \end{bmatrix}^{-1} \begin{bmatrix} 0 & 2 \\ -1 & -3 \end{bmatrix} \begin{bmatrix} -0.8944 & -0.7071 \\ 0.4472 & 0.7071 \end{bmatrix}$$
$$= \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix}$$

Using Eq 1.71

$$e^{AT} = Q e^{\hat{A}T} Q^{-1}$$
$$= \begin{bmatrix} -0.8944 & -0.7071 \\ 0.4472 & -0.7071 \end{bmatrix} \begin{bmatrix} e^{-T} & 0 \\ 0 & e^{-2T} \end{bmatrix} \begin{bmatrix} -0.8944 & -0.7071 \\ 0.4472 & -0.7071 \end{bmatrix}^{-1}$$
$$= \begin{bmatrix} 2e^{-T} - e^{-2T} & 2e^{-T} - 2e^{-2T} \\ e^{-2T} - e^{-T} & 2e^{-2T} - e^{-T} \end{bmatrix}$$

∴ Proven

2. Estimate X_0 and V_0

$$v^* = \begin{bmatrix} 3.9 \end{bmatrix}$$

2. Estimate X_0 and V_0

$$X_0^* = \begin{bmatrix} 3.9 \\ 0.5 \end{bmatrix}$$

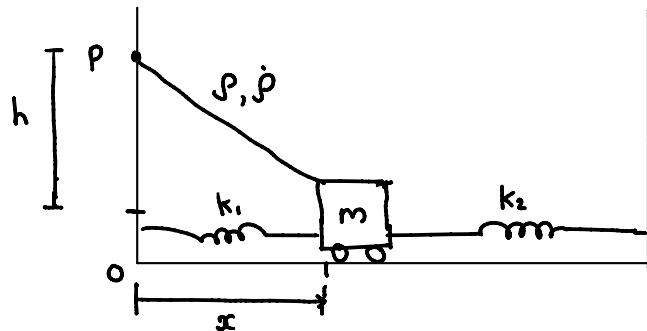
$$\bar{P}_0 = \begin{bmatrix} 1000 & 0 \\ 0 & 100 \end{bmatrix}$$

$$R = \begin{bmatrix} 0.0625 & 0 \\ 0 & 0.01 \end{bmatrix}$$

$$k_1 = 2.6, k_2 = 3.5, m = 18, h = 6.4$$

$$t_{\text{span}} = 0 : 1 : 11$$

Let the set up be as follows



$x(t)$ = position \bar{x} = static equilibrium position

$v(t)$ = velocity

$\dot{x}(t) = v(t)$

$$P(t) = (x(t)^2 + h^2)^{1/2}$$

$$\dot{P}(t) = \frac{d}{dt} (x(t)^2 + h^2)^{1/2} = \frac{1}{2} (x^2 + h^2)^{-1/2} 2x \dot{x} = \frac{xv}{P}$$

We know that for a mass spring system

$$\ddot{x} = -(k_1 + k_2)(x - \bar{x})/m = \dot{v}$$

$$\text{let } \omega^2 = (k_1 + k_2)/m$$

Assembling state vector \bar{X} :

$$\bar{X} = \begin{bmatrix} x \\ v \end{bmatrix}$$

$$\dot{\bar{X}} = \begin{bmatrix} \dot{x} \\ \dot{v} \end{bmatrix} = \begin{bmatrix} v \\ \dot{v} \end{bmatrix} = F$$

$$\dot{\underline{X}} = \begin{bmatrix} \dot{x} \\ \dot{v} \end{bmatrix} = \begin{bmatrix} v \\ -\omega^2(x - \bar{x}) \end{bmatrix} = F$$

Assembling observations \underline{Y} :

$$\underline{Y} = \begin{bmatrix} s \\ \dot{s} \end{bmatrix} + \begin{bmatrix} \epsilon_s \\ \epsilon_{\dot{s}} \end{bmatrix} = C(\underline{X}(t), t) + \epsilon(t)$$

$$\text{where } C = \begin{bmatrix} s \\ \dot{s} \end{bmatrix} = \begin{bmatrix} [x^2 + h^2]^{1/2} \\ xv/\rho \end{bmatrix}$$

Linearizing dynamics:

$$A = \frac{\partial}{\partial \underline{X}} F = \begin{bmatrix} \frac{\partial f_1}{\partial x}, & \frac{\partial f_1}{\partial \dot{x}}, \\ \frac{\partial f_2}{\partial x}, & \frac{\partial f_2}{\partial \dot{x}} \end{bmatrix} = \begin{bmatrix} \frac{\partial v}{\partial x} & \frac{\partial v}{\partial v} \\ \frac{\partial}{\partial x}(-\omega^2(x - \bar{x})) & \frac{\partial}{\partial v}(-\omega^2(x - \bar{x})) \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 \\ -\omega^2 & 0 \end{bmatrix}$$

Linearizing measurement model:

$$\tilde{H} = \frac{\partial C}{\partial \underline{X}} = \begin{bmatrix} \frac{\partial s}{\partial x}, & \frac{\partial s}{\partial \dot{x}}, \\ \frac{\partial \dot{s}}{\partial x}, & \frac{\partial \dot{s}}{\partial \dot{x}} \end{bmatrix} = \begin{bmatrix} \frac{\partial s}{\partial x} & \frac{\partial s}{\partial v} \\ \frac{\partial \dot{s}}{\partial x} & \frac{\partial \dot{s}}{\partial v} \end{bmatrix}$$

* Computing partials in MATLAB

$$\tilde{H} = \begin{bmatrix} x/\rho & 0 \\ v/\rho - x^2 v / \rho^3 & x/\rho \end{bmatrix}$$

State transition matrix Φ :

$$\dot{\Phi} = A \Phi(t, t_0) \quad \Phi(t_0, t_0) = I$$

$$\begin{bmatrix} \dot{\Phi}_{11} & \dot{\Phi}_{12} \\ \dot{\Phi}_{21} & \dot{\Phi}_{22} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\omega^2 & 0 \end{bmatrix} \begin{bmatrix} \Phi_{11} & \Phi_{12} \\ \Phi_{21} & \Phi_{22} \end{bmatrix} = \begin{bmatrix} \Phi_{21} & \Phi_{22} \\ -\omega^2 \Phi_{11} & -\omega^2 \Phi_{12} \end{bmatrix}$$

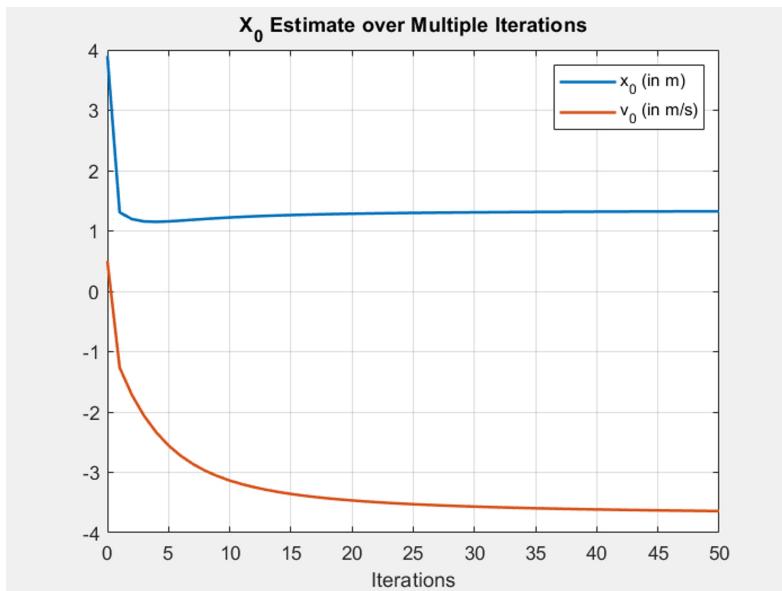
* Solving using dsolve in MATLAB

$$\Phi(t, t_0) = \begin{bmatrix} \cos(\omega t) & \sin(\omega t)/\omega \\ -\omega \sin(\omega t) & \cos(\omega t) \end{bmatrix}$$

After implementing this setup in MATLAB, X_0 after 3 iterations is

$$\begin{bmatrix} 1.157 \\ -2.065 \end{bmatrix}$$

When running more iterations, the values for X_0 changed as follows:



3. Assembling state vector \bar{X} :

$$\begin{aligned} \bar{X} &= \begin{bmatrix} x & y & z & v_x & v_y & v_z \end{bmatrix}^T, \\ \dot{\bar{X}} &= \begin{bmatrix} \dot{x} & \dot{y} & \dot{z} & \dot{v}_x & \dot{v}_y & \dot{v}_z \end{bmatrix}^T \\ &= \begin{bmatrix} v_x & v_y & v_z & g & 0 & 0 \end{bmatrix} = F \end{aligned}$$

Assembling observations \bar{Y} :

$$\bar{Y} = \begin{bmatrix} p \\ \alpha \\ \beta \end{bmatrix} + \begin{bmatrix} \epsilon_p \\ \epsilon_\alpha \\ \epsilon_\beta \end{bmatrix} = G(\bar{X}(t), t) + \epsilon(t)$$

$$\text{where } G = \begin{bmatrix} p \\ \alpha \\ \beta \end{bmatrix} = \begin{bmatrix} (x^2 + y^2 + z^2)^{1/2} \\ \tan^{-1}(y/x) \\ \tan^{-1}(z/\sqrt{x^2 + y^2}) \end{bmatrix}$$

Answer for part (b)

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Linearizing dynamics:

$$A = \frac{\partial}{\partial \dot{x}} F = \begin{bmatrix} \frac{\partial f_1}{\partial \dot{x}_1} & \dots & \frac{\partial f_1}{\partial \dot{x}_6} \\ \frac{\partial f_2}{\partial \dot{x}_1} & \dots & \vdots \\ \frac{\partial f_3}{\partial \dot{x}_1} & \dots & \vdots \\ \frac{\partial f_4}{\partial \dot{x}_1} & \dots & \vdots \\ \frac{\partial f_5}{\partial \dot{x}_1} & \dots & \vdots \\ \frac{\partial f_6}{\partial \dot{x}_1} & \dots & \frac{\partial f_6}{\partial \dot{x}_6} \end{bmatrix}$$

* Computing in MATLAB

$$A = \begin{bmatrix} 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Linearizing measurement model:

$$\tilde{H} = \frac{\partial g}{\partial \dot{x}} = \begin{bmatrix} \frac{\partial g}{\partial \dot{x}_1} & \dots & \frac{\partial g}{\partial \dot{x}_6} \\ \frac{\partial g}{\partial \dot{x}_1} & \dots & \frac{\partial g}{\partial \dot{x}_6} \\ \frac{\partial g}{\partial \dot{x}_1} & \dots & \frac{\partial g}{\partial \dot{x}_6} \end{bmatrix}$$

* Computing partials in MATLAB:

$$= \begin{bmatrix} x/\rho & y/\rho & z/\rho & 0 & 0 & 0 \\ -y/(y^2+x^2) & x/(x^2+y^2) & 0 & 0 & 0 & 0 \\ -xz/\rho^2\sqrt{x^2+y^2} & -y^2/\rho^2\sqrt{x^2+y^2} & \frac{\sqrt{x^2+y^2}}{\rho^2} & 0 & 0 & 0 \end{bmatrix}$$

Answer for part (c)

State transition matrix Φ :

$$\dot{\Phi}(t, t_0) = A(t, t_0) \quad \Phi(t_0, t_0) = I$$

$$\begin{bmatrix} \dot{\Phi}_{11} & \dots & \dot{\Phi}_{16} \\ \vdots & \ddots & \vdots \\ \dot{\Phi}_{61} & \dots & \dot{\Phi}_{66} \end{bmatrix} = \begin{bmatrix} A_{11} & \dots & A_{16} \\ \vdots & \ddots & \vdots \\ A_{61} & \dots & A_{66} \end{bmatrix} \begin{bmatrix} \Phi_{11} & \dots & \Phi_{16} \\ \vdots & \ddots & \vdots \\ \Phi_{61} & \dots & \Phi_{66} \end{bmatrix}$$

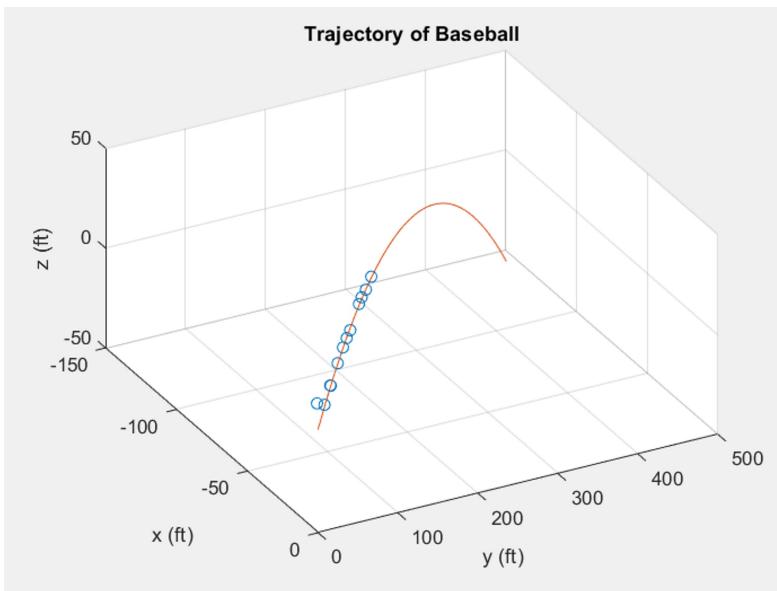
$$\begin{bmatrix} \vdots & \ddots & \vdots \\ \Phi_{61} & \cdots & \Phi_{66} \end{bmatrix} = \begin{bmatrix} \vdots & \ddots & \vdots \\ A_{61} & \cdots & A_{66} \end{bmatrix} \quad \begin{bmatrix} \vdots & \ddots & \vdots \\ \Phi_{61} & \cdots & \Phi_{66} \end{bmatrix}$$

* Solving using `dsolve` in MATLAB

$$\Phi(t, t_0) = \begin{bmatrix} 1 & 0 & 0 & t & 0 & 0 \\ 0 & 1 & 0 & 0 & t & 0 \\ 0 & 0 & 1 & 0 & 0 & t \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Answer for part (a)

- d. The process was repeated 4 times (as gone over in class), and the following trajectory was obtained:



Using the distance formula, it was found that the baseball landed 405.71 ft from the origin.

- e. To find the uncertainty of the estimate, the last P_0 term was propagated to the final time using $\Phi(t, 0)$.
 $t_f = 3.41s$

$$P = \underline{\Phi} P_0 \underline{\Phi}^T = * \text{Evaluated in MATLAB}$$

The Frobenius norm of this matrix was found

$$\sqrt{\sum_{j=1}^6 \sum_{i=1}^6 a_{ij}^2} = \underline{1.3289}$$