

- 1) Let w be the displacement, the Equation of Motion is

$$m\ddot{w} + kw = u$$

$$\ddot{w} + 100w = u$$

The equation above can be converted into the State-Space form of

$$x' = \begin{bmatrix} x_1' \\ x_2' \end{bmatrix} = \begin{bmatrix} \dot{w} \\ \ddot{w} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\frac{k}{m} & 0 \end{bmatrix} \begin{bmatrix} w \\ \dot{w} \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{1}{m} \end{bmatrix} \cdot u$$

because $\ddot{w} = \frac{u}{m} - \frac{k}{m} \cdot w$

$$y = \ddot{w} = \begin{bmatrix} -\frac{k}{m} & 0 \end{bmatrix} \begin{bmatrix} w \\ \dot{w} \end{bmatrix} + \frac{1}{m} \cdot u$$

so $A_c = \begin{bmatrix} 0 & 1 \\ -100 & 0 \end{bmatrix}, B_c = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, C = [-100 \ 0], D = 1$

- 2) With the sampling interval $\Delta t = 0.01 \text{ sec}$, applying the analytical power series expressions we obtain the discrete-time matrices A and B :

$$A = e^{A_c \Delta t} = I + A_c \Delta t + \frac{1}{2!} (A_c \Delta t)^2 + \frac{1}{3!} (A_c \Delta t)^3 + \frac{1}{4!} (A_c \Delta t)^4 + \dots$$

$$B = (I \Delta t + \frac{1}{2!} A_c \Delta t^2 + \frac{1}{3!} A_c^2 \Delta t^3 + \frac{1}{4!} A_c^3 \Delta t^4 + \frac{1}{5!} A_c^4 \Delta t^5 + \dots) \cdot B_c$$

preserving the first 5 terms, by MATLAB, the solutions are:

$$A = \begin{bmatrix} 0.9950 & 0.01 \\ -0.9983 & 0.9950 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 0.01 \end{bmatrix}, C = [-100 \ 0], D = 1$$

The *c2d* function generates the same matrices.

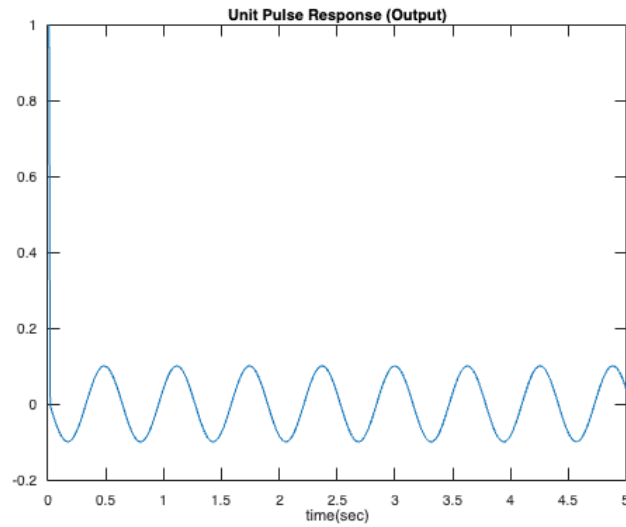
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a = 2x2
    0.9950    0.0100
   -0.9983    0.9950

b = 2x1
    0.0000
    0.0100

c = 1x2
   -100     0

d = 1
```

Outputs of the first 5 seconds under the unit pulse are plotted as below

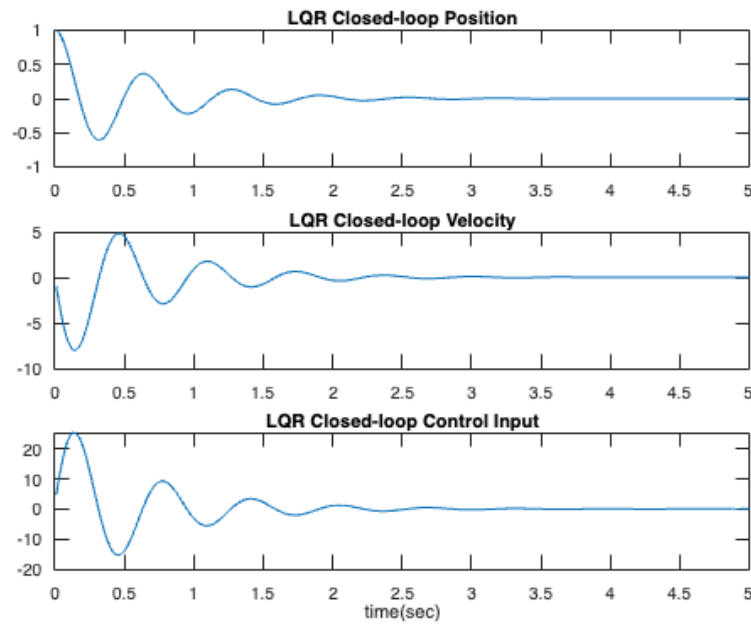


3) The closed-loop state-space model is

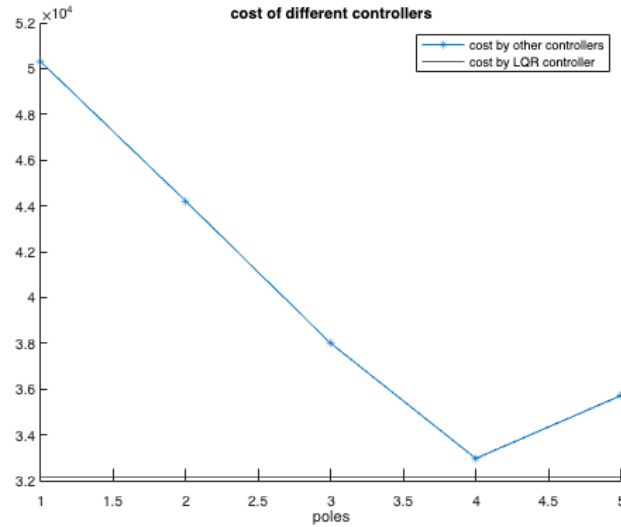
$$x(k+1) = (A + BF)x(k)$$

$$y(k) = (C + DF)x(k)$$

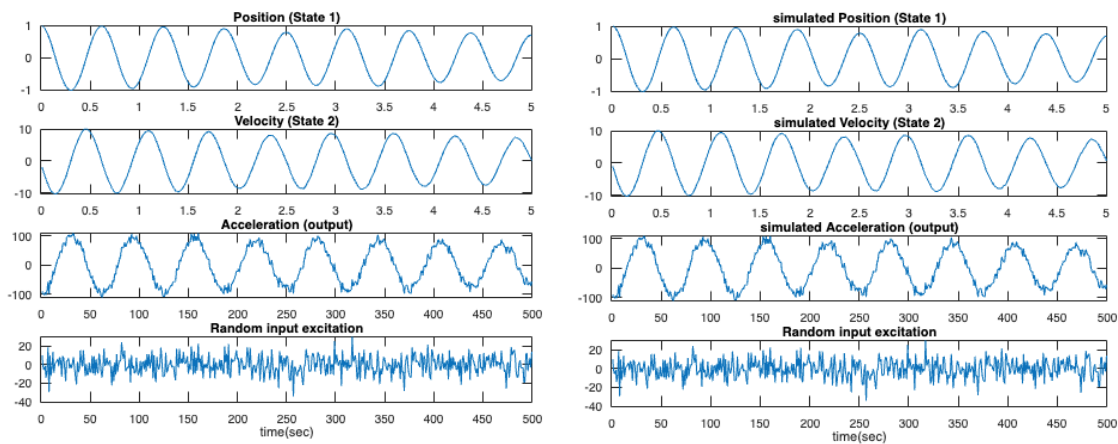
where $u(k) = Fx(k)$ and F is computed such that the cost J is minimized. The states and controlled inputs are plotted as below:



By comparing the cost associated with the LQR optimal controller and other controllers by placing poles inside the unit complex circle, we found **the LQR indeed returns the lowest cost.**



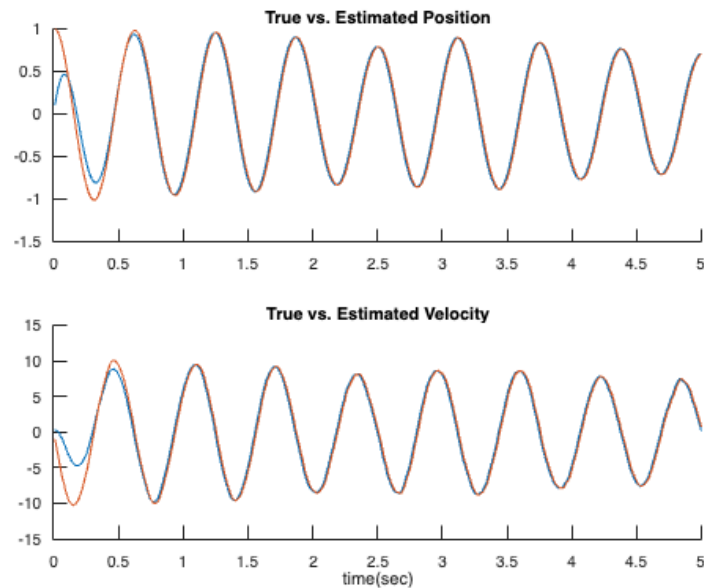
- 4) With the same state-space model and random generated inputs, **manual coding and simulated methods generate the sample plots.**



- 5) The one-equation version of the Luenberger observer with gain M is defined as:

$$\hat{x}(k+1) = (A + MC)\hat{x}(k) + (B + MD)u(k) - My(k)$$

the true versus estimated states comparison is plotted as below and **the observe correctly estimates the system states.**



6) The model for observed-based state-feedback control system is

$$\begin{bmatrix} x(k+1) \\ \hat{x}(k+1) \end{bmatrix} = \begin{bmatrix} A & BF \\ -MC & A + MC + BF \end{bmatrix} \begin{bmatrix} x(k) \\ \hat{x}(k) \end{bmatrix}$$

The results for the first 5 seconds are

