

Practice 1

Linear Quadratic Regulator

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Variant:

k	Block mass m , kg	Viscous damping coefficient, c	Initial conditions $x(0)$
2	3	0.2	$\begin{bmatrix} \pi/2 \\ -1 \end{bmatrix}$

Simulink Model:

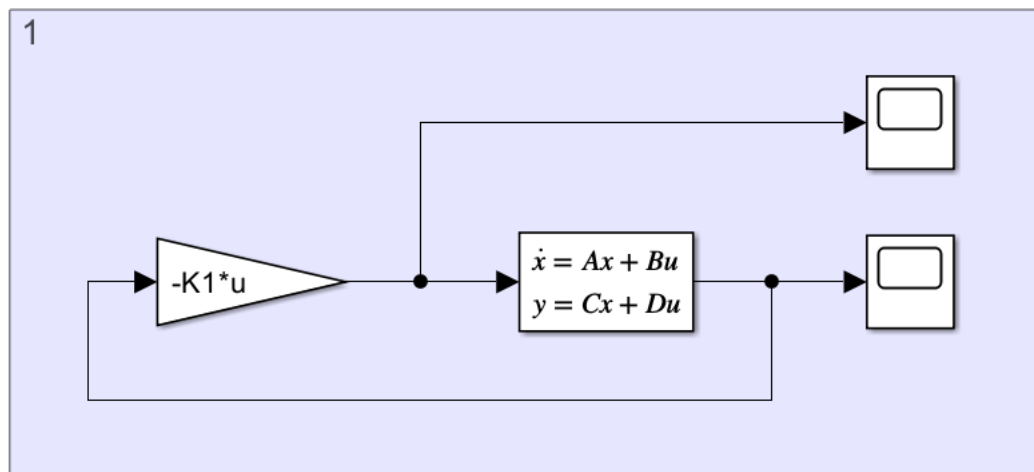


Figure 1. State-Space Model in Simulink.

1) Cheap control: $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, R[0.01]$

$$Q1 = [1 \ 0; 0 \ 1];$$

$$R1 = 0.01;$$

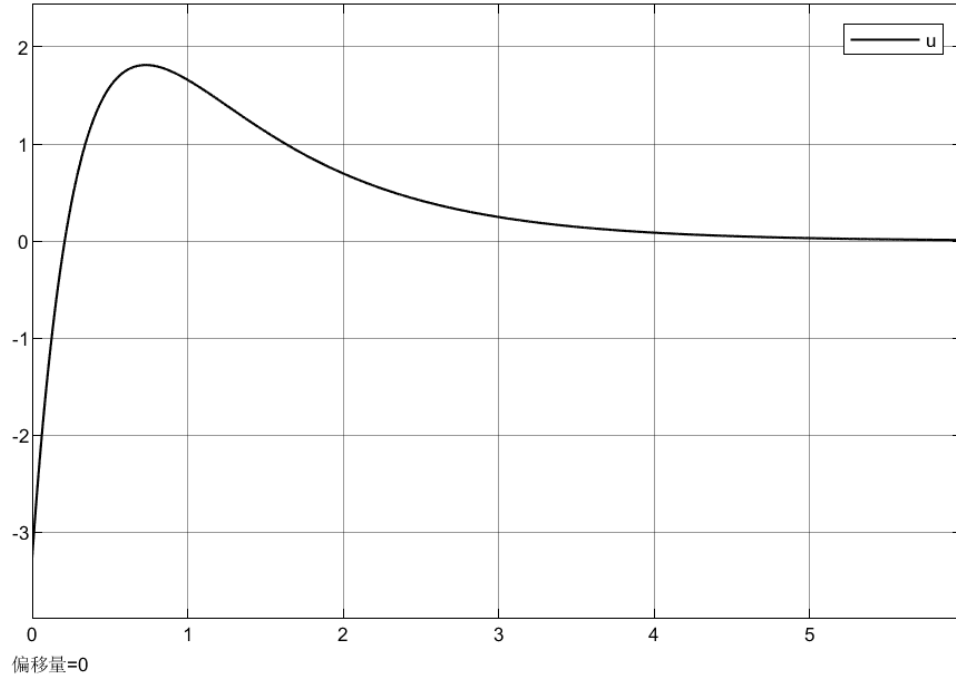


Figure 2. diagram of $u(t)$.

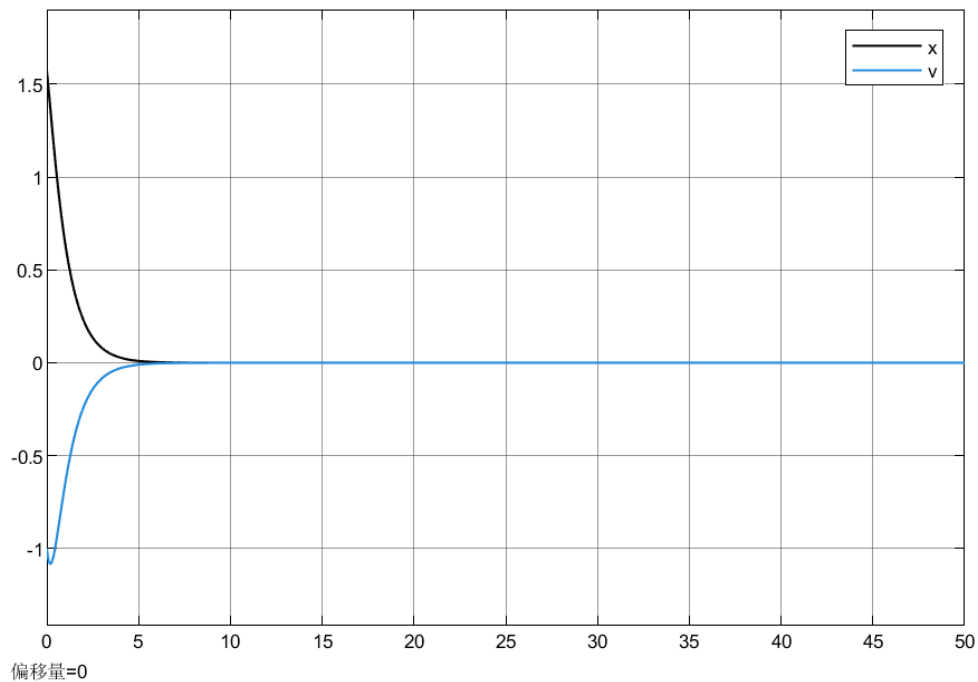


Figure 3. diagram of position and speed.

Comment:

In this case, it took about 5 seconds for $u(t)$ to decrease to 0. And The position and speed are correct and compatible.

2) Expensive control: $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, R[100]$

$Q2 = [1 \ 0; 0 \ 1];$
 $R2 = 100;$

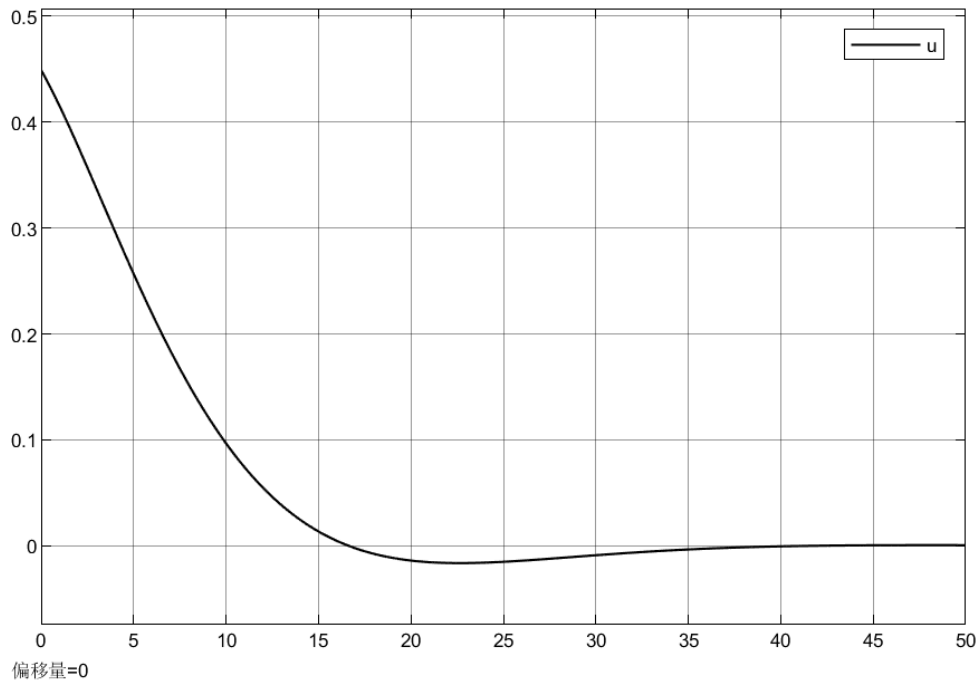


Figure 2. diagram of $u(t)$.

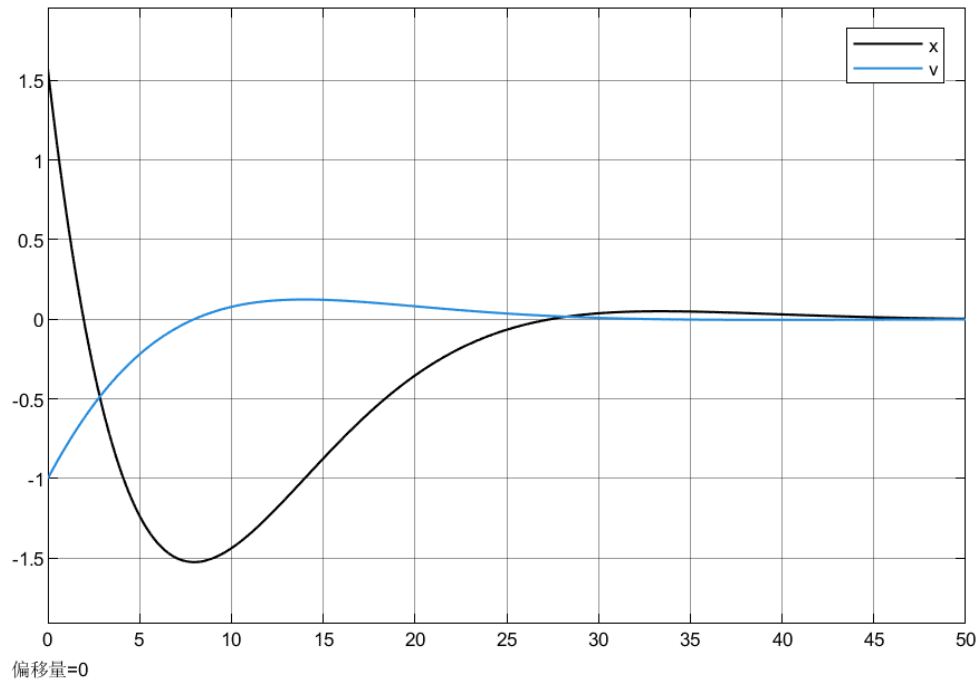


Figure 3. diagram of position and speed.

Comment:

In this case, it took about 50 seconds for $u(t)$ to decrease to 0. And The position and speed are correct and compatible. Looks like it took longer.

3) Only penalize the velocity state: $Q = \begin{bmatrix} 0.001 & 0 \\ 0 & 10 \end{bmatrix}, R[1]$

$$Q3 = [0.001 \ 0; 0 \ 10];$$

$$R3 = 1;$$

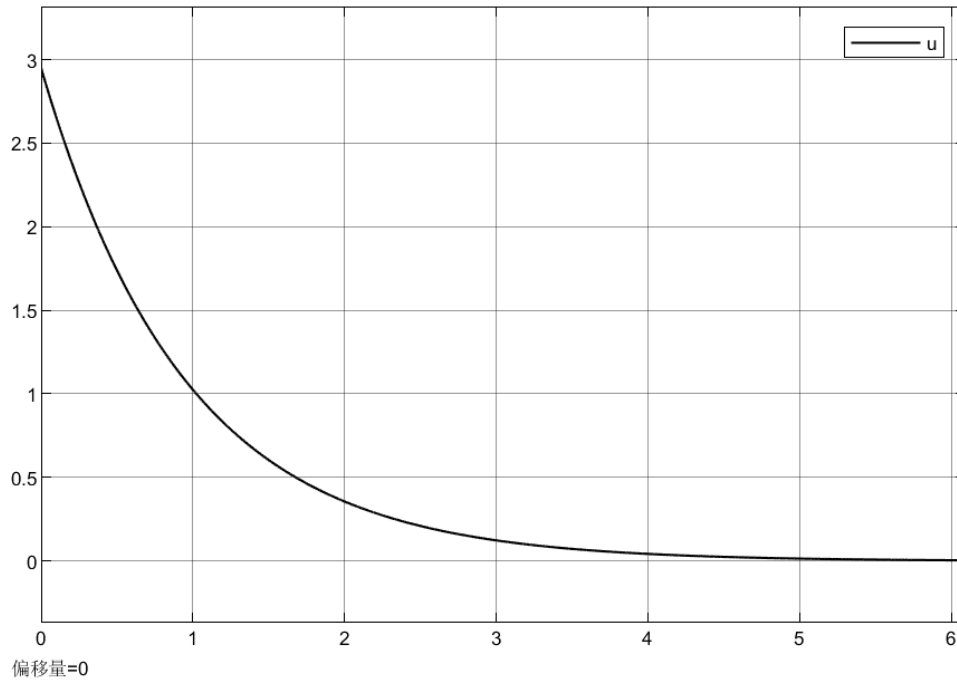


Figure 2. diagram of $u(t)$.

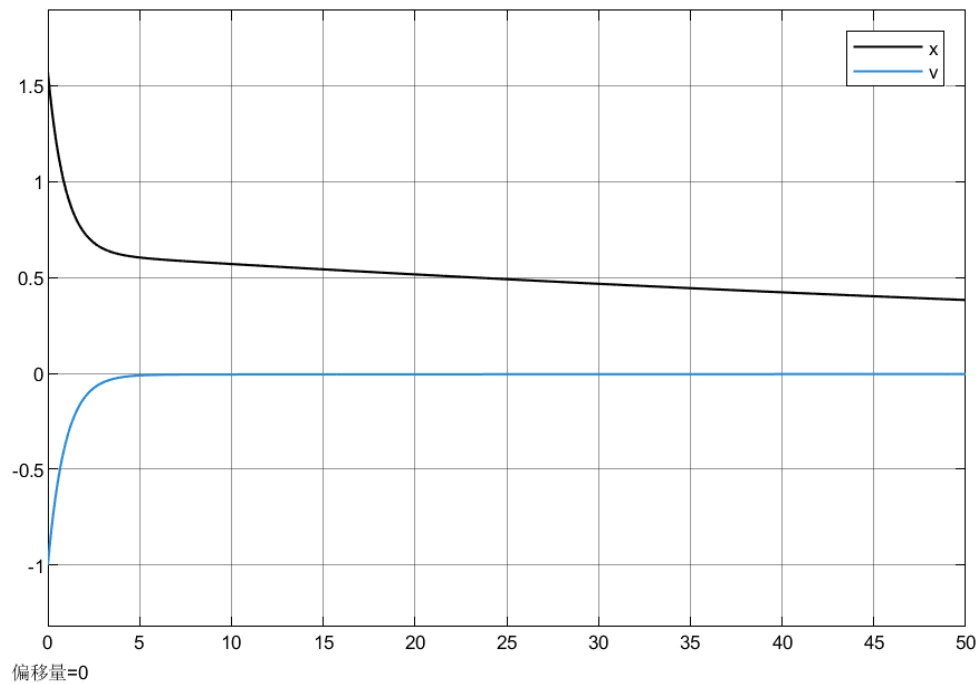


Figure 3. diagram of position and speed.

Comment:

In this case, it took about 5 seconds for $u(t)$ to decrease to 0. And position x and velocity v do not converge to 0 at the same time. Looks like it took longer to get the position attenuation to 0.

Conclusions:

Through the three sets of experiments using different Q and R matrices in the Linear Quadratic Regulator (LQR) design, we observed how the choice of these parameters significantly affects the system's control performance. A smaller R value led to a faster response with more control effort, while a larger R caused slower response with reduced control action. Additionally, the weighting of state variables in Q influenced the convergence behavior of position and velocity. This exercise demonstrated the importance of appropriately tuning Q and R to balance performance and energy cost in optimal control systems.

Source Code:

```
m = 3;  
c = 0.2;  
x0 = [pi/2; -1];
```

1) Cheap control

```
A = [0 1; 0 -c/m];  
B = [0; 1/m];  
C = eye(2);  
D = zeros(2,1);  
  
Q1 = [1 0; 0 1];  
R1 = 0.01;  
  
[K1, P1] = lqr(A, B, Q1, R1);
```

2) Expensive control

```
Q2 = [1 0; 0 1];  
R2 = 100;  
  
[K2, P2] = lqr(A, B, Q2, R2);
```

3) Only penalize the velocity state

```
Q3 = [0.001 0; 0 10];  
R3 = 1;  
  
[K3, P3] = lqr(A, B, Q3, R3);
```