

# ASYMPTOTIC STABILIZATION



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# CONTROL ENGINEERING WITH PYTHON

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# SYMBOLS



Code



Worked Example



Graph



Exercise



Definition



Numerical Method



Theorem



Analytical Method



Remark



Theory



Information



Hint



Warning



Solution



# IMPORTS

```
from numpy import *  
from numpy.linalg import *  
from numpy.testing import *  
from scipy.integrate import *  
from scipy.linalg import *  
from matplotlib.pyplot import *
```



# ASYMPTOTIC STABILIZATION

When the system

$$\dot{x} = Ax, \quad x \in \mathbb{R}^n$$

is not asymptotically stable at the origin, maybe there are some inputs  $u \in \mathbb{R}^m$  such that

$$\dot{x} = Ax + Bu$$

that we can use to stabilize asymptotically the system?



# LINEAR FEEDBACK

We search for  $u$  as a linear feedback:




$$u(t) = -Kx(t)$$

for some  $K \in \mathbb{R}^{m \times n}$ .



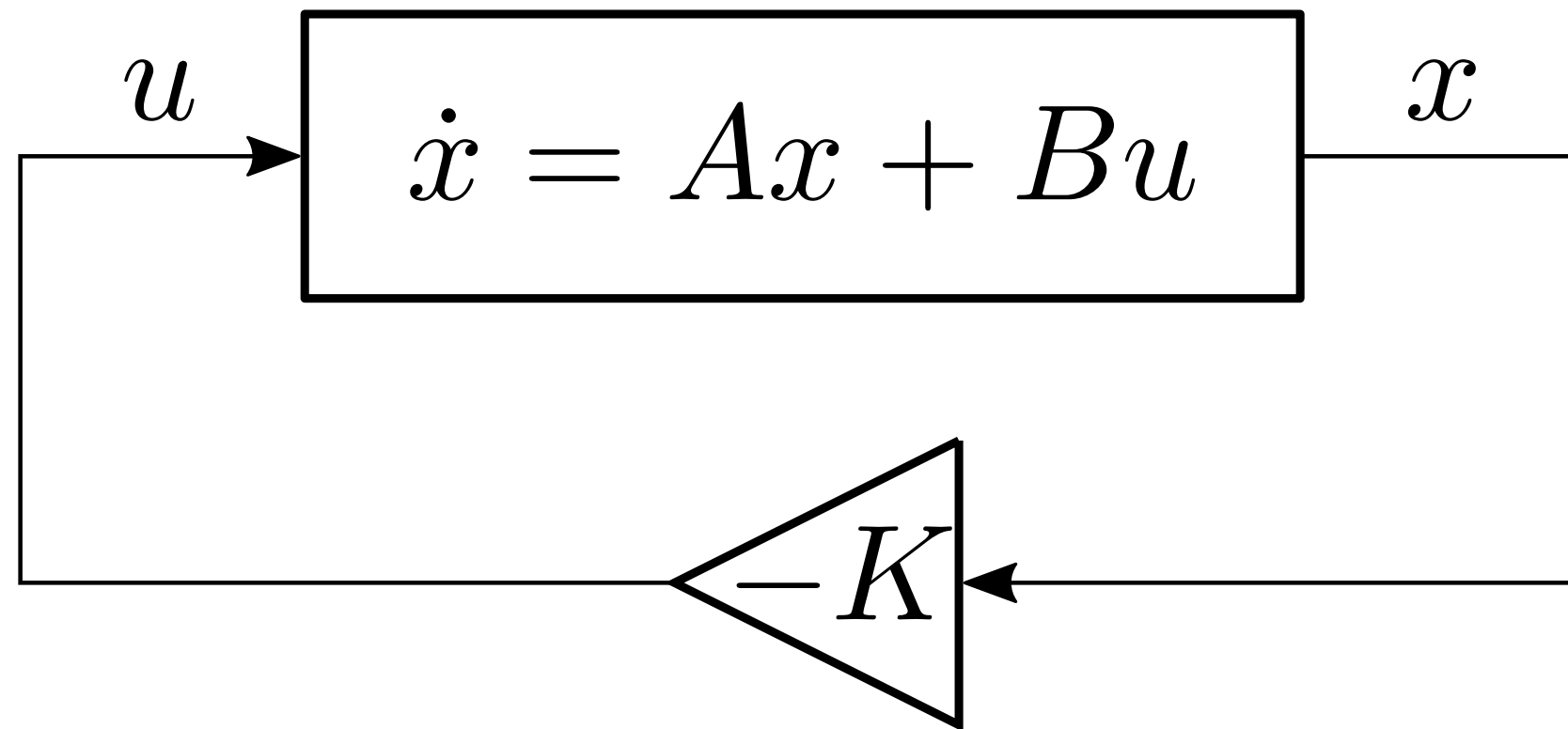
## NOTE

In this scheme

-  The full system state  $x(t)$  **must be measured**.
-  This information is then **fed back** into the system.
-  The feedback **closes the loop**.



# CLOSED-LOOP DIAGRAM







# CLOSED-LOOP DYNAMICS

When

$$\dot{x} = Ax + Bu$$

$$u = -Kx$$

the state  $x \in \mathbb{R}^n$  evolves according to:

$$\dot{x} = (A - BK)x$$



## REMINDER

The closed-loop system is asymptotically stable iff every eigenvalue of the matrix

$$A - BK$$

is in the open left-hand plane.



# SPECTRUM AS A MULTISSET

Multisets remember the multiplicity of their elements.  
It's convenient to describe the spectrum of matrices:

$$A := \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix} \Rightarrow \sigma(A) = \{1, 1, 2\}$$

$$0 \notin \sigma(A), 1 \in \sigma(A), 1 \in^2 \sigma(A), 1 \notin^3 \sigma(A)$$

$$2 \in \sigma(A), 2 \notin^2 \sigma(A)$$



# POLE ASSIGNMENT

## ASSUMPTIONS

- The system

$$\dot{x} = Ax + Bu, \quad x \in \mathbb{R}^n, \quad u \in \mathbb{R}^p$$

is controllable.

- $\Lambda$  is a symmetric multiset of  $n$  complex numbers:

$$\Lambda = \{\lambda_1, \dots, \lambda_n\} \subset \mathbb{C} \quad \text{and} \quad \lambda \in^k \Lambda \Rightarrow \bar{\lambda} \in^k \Lambda.$$



# POLE ASSIGNMENT

## CONCLUSION

$\Rightarrow$  There is a matrix  $K \in \mathbb{R}^{n \times m}$  such that

$$\sigma(A - BK) = \Lambda.$$



# POLE ASSIGNMENT

Consider the double integrator  $\ddot{x} = u$

$$\frac{d}{dt} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$$

(in standard form)



```
from scipy.signal import place_poles  
A = array([[0, 1], [0, 0]])  
B = array([[0], [1]])  
poles = [-1, -2]  
K = place_poles(A, B, poles).gain_matrix
```



```
assert_almost_equal(K, [[2.0, 3.0]])  
eigenvalues, _ = eig(A - B @ K)  
assert_almost_equal(eigenvalues, [-1, -2])
```

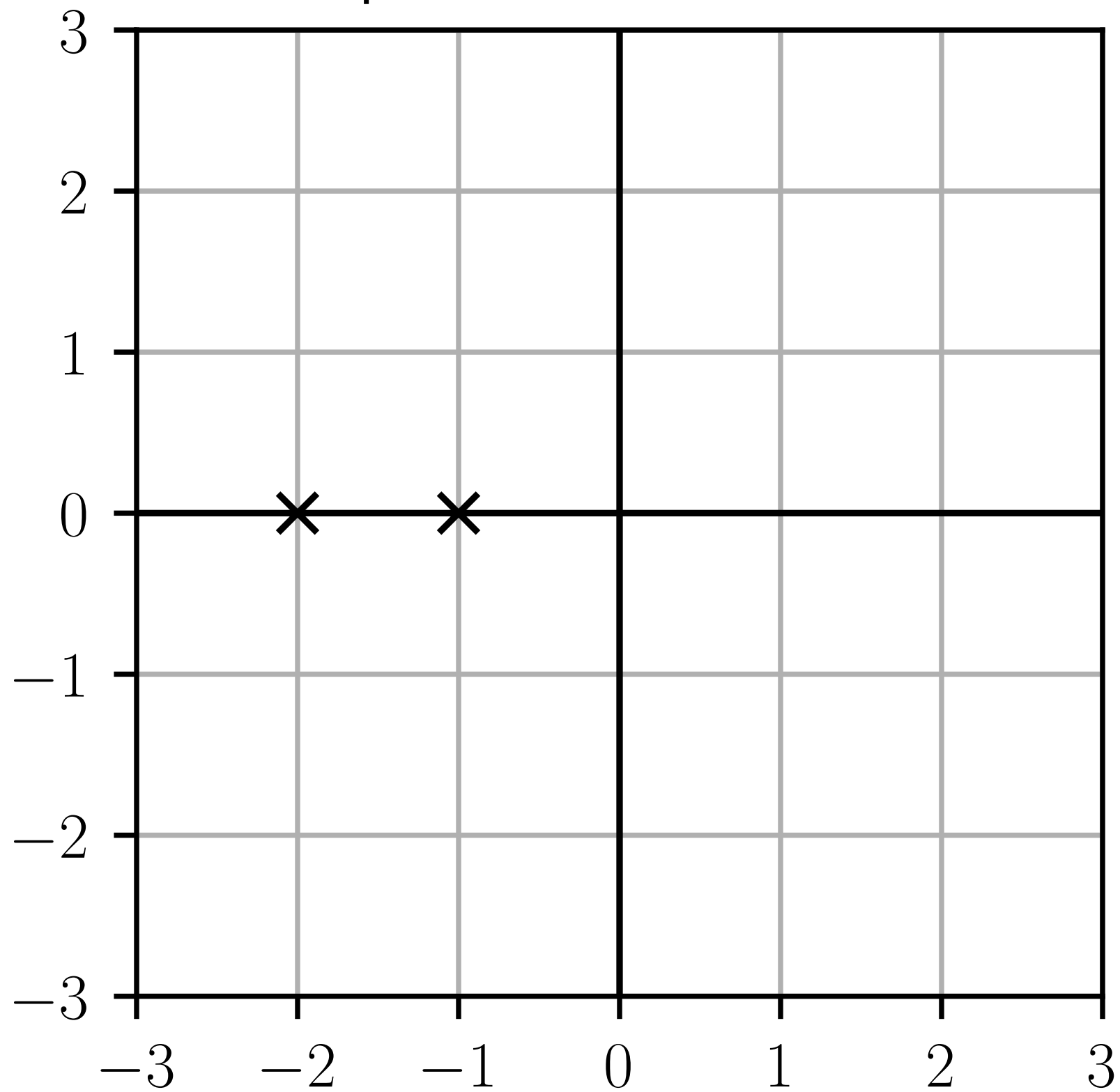




# SPECTRUM

```
figure()
x = [real(s) for s in eigenvalues]
y = [imag(s) for s in eigenvalues]
plot(x, y, "kx")
xticks([-3, -2, -1, 0, 1, 2, 3])
yticks([-3, -2, -1, 0, 1, 2, 3])
plot([0, 0], [-3, 3], "k")
plot([-3, 3], [0, 0], "k")
title("Spectrum of  $A-BK$ "); grid(True)
```



# Spectrum of $A - BK$



## LIMITATIONS

— The `place_poles` function rejects eigenvalues whose multiplicity is higher than the rank of  $B$ .

In the previous example,  $\text{rank } B = 1$ , so

-  `poles = [-1, -1]` won't work.
-  `poles = [-1, -2]` will.



# POLE ASSIGNMENT

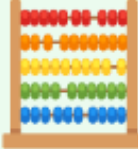
Consider the system with dynamics

$$\dot{x}_1 = x_1 - x_2 + u$$

$$\dot{x}_2 = -x_1 + x_2 + u$$

We apply the control law

$$u = -k_1x_1 - k_2x_2.$$

1. 

Can we assign the poles of the closed-loop system freely by a suitable choice of  $k_1$  and  $k_2$ ?

2. 🧠

Explain this result.



# POLE ASSIGNMENT

1. 

$$\begin{aligned} A - BK &= \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix} - \begin{bmatrix} 1 \\ 1 \end{bmatrix} \begin{bmatrix} k_1 & k_2 \end{bmatrix} \\ &= \begin{bmatrix} 1 - k_1 & -1 - k_2 \\ -1 - k_1 & 1 - k_2 \end{bmatrix} \end{aligned}$$



$$\begin{aligned}
 \det A - BK &= \det \begin{pmatrix} s - 1 + k_1 & 1 + k_2 \\ 1 + k_1 & s - 1 + k_2 \end{pmatrix} \\
 &= (s - 1 + k_1)(s - 1 + k_2) - (1 + k_1)(1 + k_2) \\
 &= s^2 + (k_1 + k_2)s - 2(k_1 + k_2)
 \end{aligned}$$

$$\begin{aligned}\sigma(A - BK) &= \{\lambda_1, \lambda_2\} \\ &= \{\lambda \in \mathbb{C} \mid s^2 + (k_1 + k_2)s - 2(k_1 + k_2) = 0\}\end{aligned}$$

Since the characteristic polynomial is also

$$(s - \lambda_1)(s - \lambda_2)$$

we get


$$k_1 + k_2 = -\lambda_1 - \lambda_2, \quad -2(k_1 + k_2) = \lambda_1 \lambda_2$$

Thus we have

$$\lambda_1 \lambda_2 = 2(\lambda_1 + \lambda_2) \Rightarrow \lambda_2 = \frac{2\lambda_1}{\lambda_1 - 2}$$

and both poles cannot be assigned freely; for example if we select  $\lambda_1 = 1$ , we end up with  $\lambda_2 = -2$ .

## 2.

We have not checked the assumptions of  Pole Assignment yet.

The commandability matrix is

$$[B, AB] = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$$

whose rank is  $1 < 2$ .

Since the system is not controllable, pole assignment may fail and it does here.



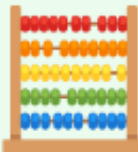
# PENDULUM

Consider the pendulum with dynamics:

$$m\ell^2\ddot{\theta} + b\dot{\theta} + mg\ell \sin \theta = u$$

Numerical Values:

$$m = 1.0, \ell = 1.0, b = 0.1, g = 9.81$$

1. 

Compute the linearized dynamics of the system  
around the equilibrium  $\theta = \pi$  and  $\dot{\theta} = 0$  ( $u = 0$ ).

## 2.

Design a control law

$$u = -k_1(\theta - \pi) - k_2\dot{\theta}$$

such that the closed-loop linear system is asymptotically stable, with a time constant equal to 10 sec.

### 3.

Simulate this control law on the nonlinear systems  
when  $\theta(0) = 0.9\pi$  and  $\dot{\theta}(0) = 0$ .





**PENDULUM**

1. 

Let  $\Delta\theta = \theta - \pi$ ,  $\omega = \dot{\theta}$ ,  $\Delta\omega = \omega$ ,  $\Delta u = u$ .

We notice that

$$\begin{aligned}\sin \theta &= \sin(\pi + \Delta\theta) \\ &= -\sin \Delta\theta \\ &\approx -\Delta\theta\end{aligned}$$

The system dynamics can be approximated around  $(\theta, \omega) = (\pi, 0)$  by

$$(d/dt)\Delta\theta = \Delta\omega$$

and

$$m\ell^2(d/dt)\Delta\omega + b\Delta\omega - mg\ell\Delta\theta = \Delta u.$$

or in standard form

$$\frac{d}{dt} \begin{bmatrix} \Delta\theta \\ \Delta\omega \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ g/\ell & -b/(m\ell^2) \end{bmatrix} \begin{bmatrix} \Delta\theta \\ \Delta\omega \end{bmatrix} + \begin{bmatrix} 0 \\ 1/(m\ell^2) \end{bmatrix} \Delta u$$

## 2.

$m = 1.0$

$l = 1.0$

$b = 0.1$

$g = 9.81$

```
A = array([[ 0,          1],
           [g/l , - b/(m*1*1)])
B = array([[ 0],
           [1/(m*1*1)])
t1, t2 = 10.0, 5.0
poles = [-1/t1, -1/t2]
K = place_poles(A, B, poles).gain_matrix
```

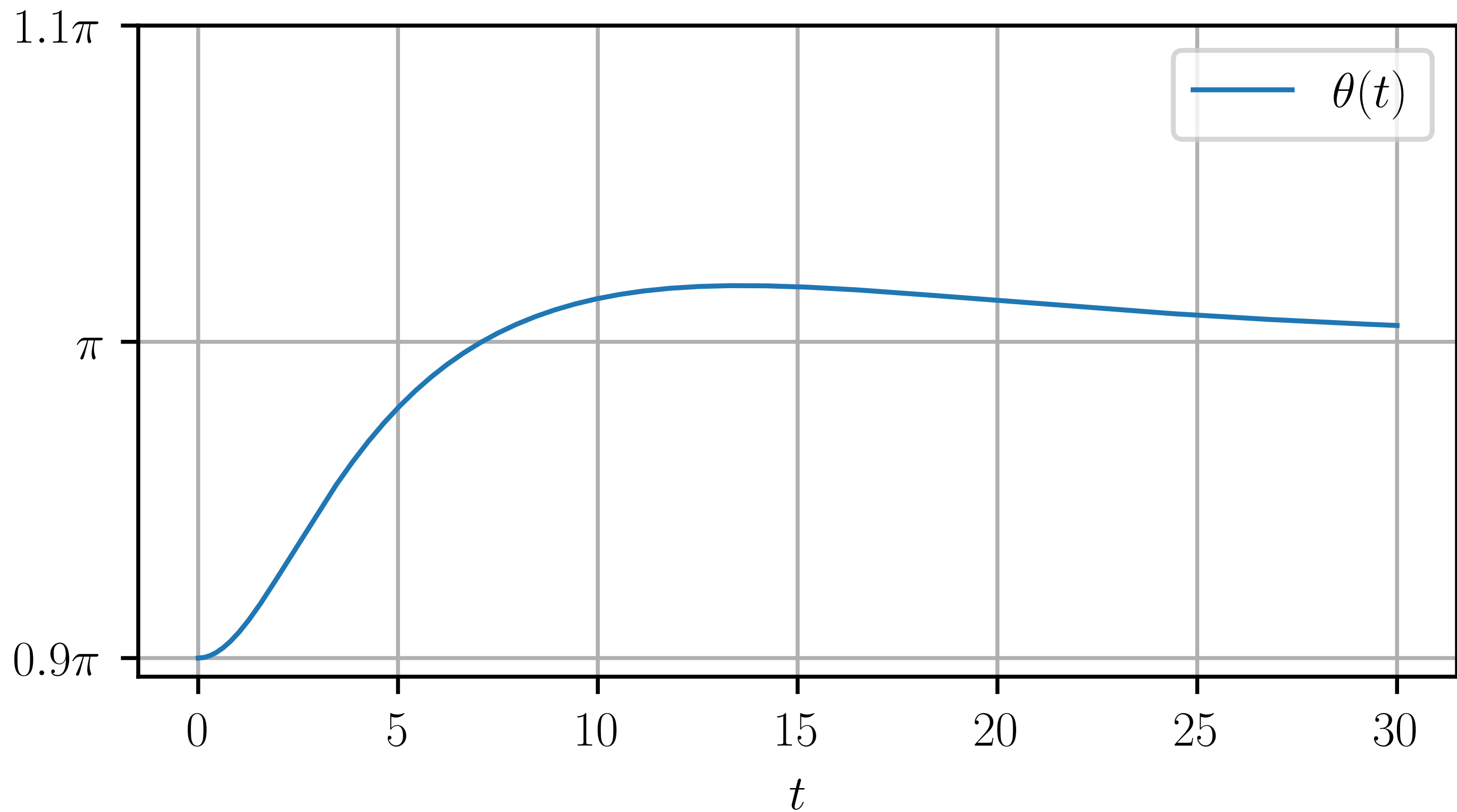
### 3.

```
def fun(t, theta_omega):  
    theta, omega = theta_omega  
     $\Delta$ theta,  $\Delta$ omega = theta - pi, omega  
     $\Delta$ u = - K @ [ $\Delta$ theta,  $\Delta$ omega]  
    u =  $\Delta$ u[0] #  $\Delta$ u has a (1,) shape  
    dtheta = omega  
    domega = - (g/l)*sin(theta) - b/(m*l*l)*omega \  
              + 1.0/(m*l*l)*u  
    return array([dtheta, domega])
```

```
t_span = [0.0, 30.0]
y0 = [0.9*pi, 0.0]
r = solve_ivp(fun, t_span, y0, dense_output=True)
t = linspace(t_span[0], t_span[-1], 1000)
thetat, omega_t = r["sol"](t)
```



```
figure()
plot(t, thetat, label=r"$\theta(t)$")
xlabel("$t$")
yticks([0.9*pi, pi, 1.1*pi],
        [r"$0.9\pi$", r"$\pi$", r"$1.1\pi$"])
grid(True); legend()
```





# DOUBLE SPRING

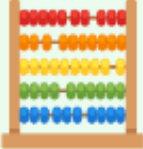

Consider the dynamics:

$$m_1 \ddot{x}_1 = -k_1 x_1 - k_2(x_1 - x_2) - b_1 \dot{x}_1$$

$$m_2 \ddot{x}_2 = -k_2(x_2 - x_1) - b_2 \dot{x}_2 + u$$

Numerical values:

$$m_1 = m_2 = 1, \quad k_1 = 1, \quad k_2 = 100, \quad b_1 = 2, \quad b_2 = 20$$

1.  

Compute the poles of the system.

Is the origin asymptotically stable?

## 2.

Use a linear feedback to:

- kill the oscillatory behavior of the solutions,
- “speed up” the dynamics.



# DOUBLE SPRING SYSTEM

# 1.

Let  $v_1 = \dot{x}_1$ ,  $v_2 = \dot{x}_2$ . With the state  $(x_1, v_1, x_2, v_2)$ :

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -(k_1 + k_2)/m_1 & -b_1/m_1 & k_2/m_1 & 0 \\ 0 & 0 & 0 & 1 \\ k_2/m_2 & 0 & -k_2/m_2 & -b_2/m_2 \end{bmatrix}$$

$$B = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1/m_2 \end{bmatrix}$$



$m1 = m2 = 1$

$k1 = 1; k2 = 100$

$b1 = 2; b2 = 20$

```
A = array([
    [      0,      1,      0,      0],
    [-(k1+k2)/m1, -b1/m1, k2/m1,      0],
    [      0,      0,      0,      1],
    [      k2/m2,      0, -k2/m2, -b2/m2]
])
B = array([[0.0], [0.0], [0.0], [1/m2]])
```

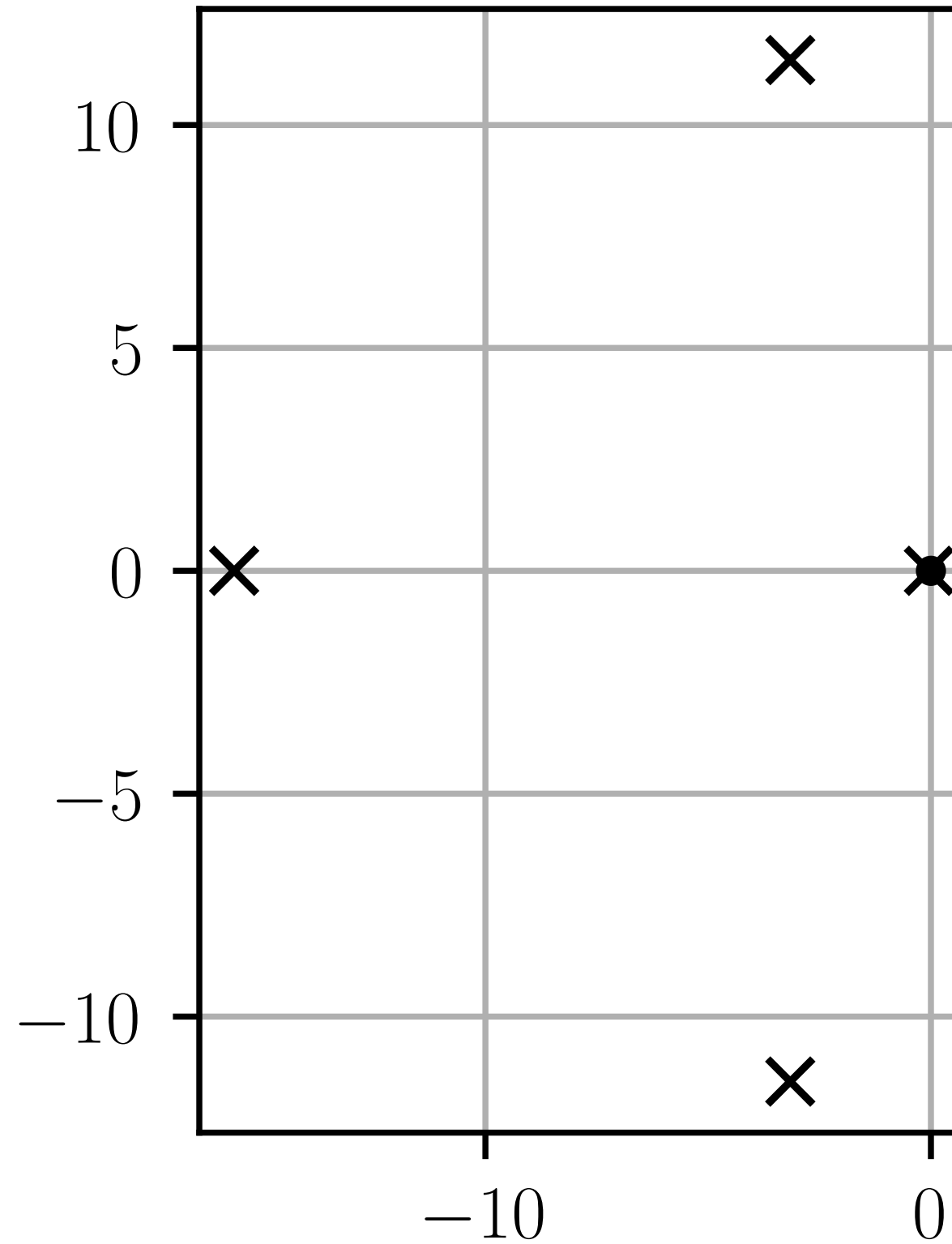
```
eigenvalues, _ = eig(A)
```

```
>>> eigenvalues  
array([-15.64029062 +0.j           ,  
       -3.15722141+11.45767938j,  
       -3.15722141-11.45767938j,  
       -0.04526657 +0.j           ])
```

Since all eigenvalues have a negative real part, the double-spring system is asymptotically stable.

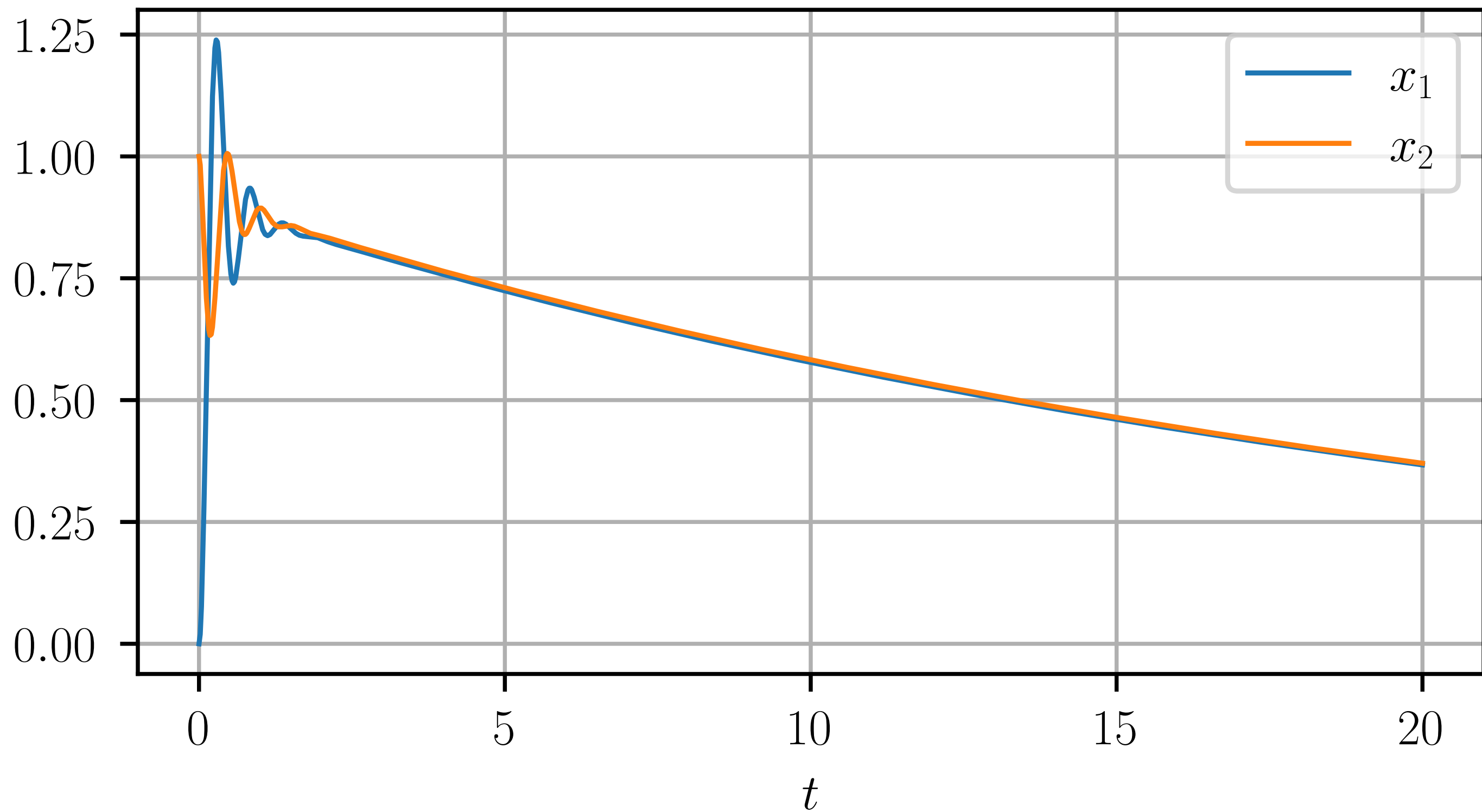
```
figure()  
x = [real(s) for s in eigenvalues]  
y = [imag(s) for s in eigenvalues]  
plot(x, y, "kx")  
plot(0.0, 0.0, "k.")  
gca().set_aspect(1.0)  
title("Spectrum of $A$"); grid(True)
```

Spectrum of  $A$



```
y0 = [0.0, 0.0, 1.0, 0.0]
t = linspace(0.0, 20.0, 1000)
yt = array([expm(A * t_) for t_ in t]) @ y0
x1t, x2t = yt[:, 0], yt[:, 2]
```

```
figure()  
plot(t, x1t, label="$x_1$")  
plot(t, x2t, label="$x_2$")  
xlabel("$t$")  
grid(True); legend()
```





## 2.

```
eigenvalues[3] = - 1 / 0.1  
K = place_poles(A, B, eigenvalues).gain_matrix  
print(repr(eig(A - B @ K)[0]))
```

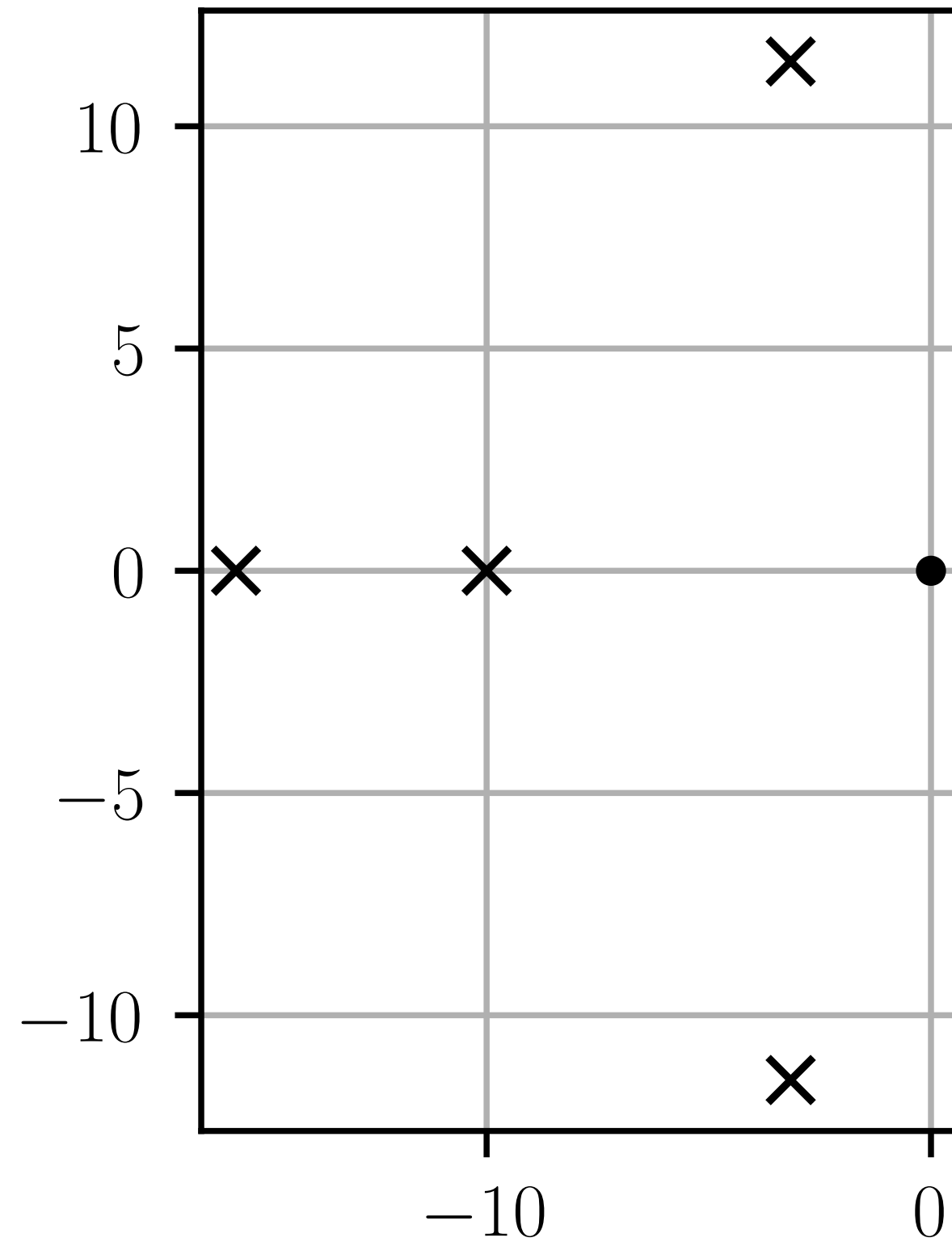
```
eigenvalues, _ = eig(A - B @ K)
```

```
>>> eigenvalues
```

```
array([-15.64029062 +0.j           ,  
       -3.15722141+11.45767938j,  
       -3.15722141-11.45767938j,  
       -1.          +0.j           ])
```

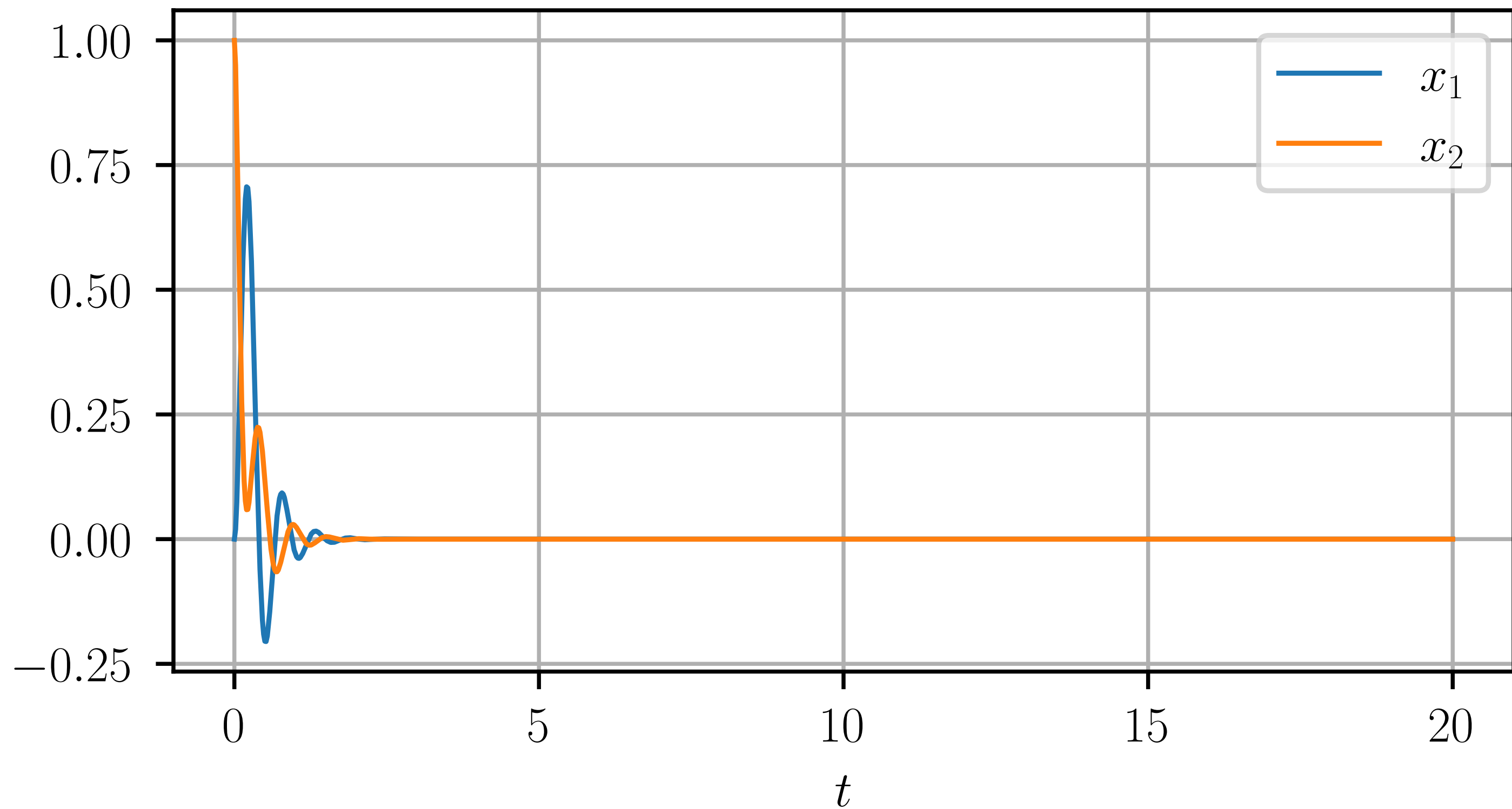
```
figure()  
x = [real(s) for s in eigenvalues]  
y = [imag(s) for s in eigenvalues]  
plot(x, y, "kx")  
plot(0.0, 0.0, "k.")  
gca().set_aspect(1.0)  
title("Spectrum of  $A - B K$ "); grid(True)
```

Spectrum of  $A - BK$



```
y0 = [0.0, 0.0, 1.0, 0.0]
t = linspace(0.0, 20.0, 1000)
yt = array([expm((A-B@K) * t_) for t_ in t]) @ y0
x1t, x2t = yt[:, 0], yt[:, 2]
```

```
figure()  
plot(t, x1t, label="$x_1$")  
plot(t, x2t, label="$x_2$")  
xlabel("$t$")  
grid(True); legend()
```



```
eigenvalues[0] = - 1 / 0.09  
eigenvalues[1] = - 1 / 0.08  
K = place_poles(A, B, eigenvalues).gain_matrix  
print(repr(eig(A - B @ K)[0]))
```

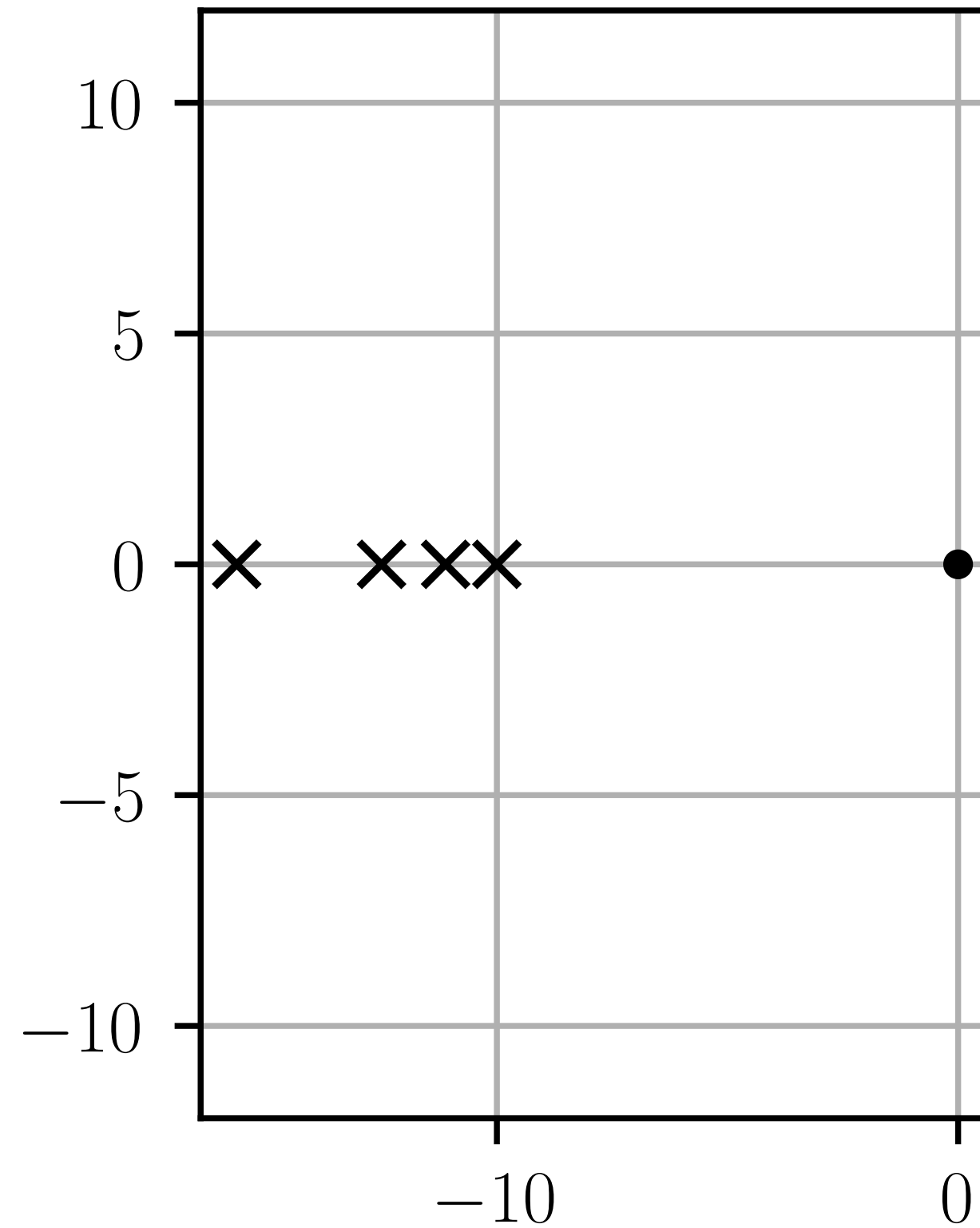


```
eigenvalues, _ = eig(A - B @ K)
```

```
>>> eigenvalues  
array([-15.64029062+0.j,  
       -12.5          +0.j,  
       -11.11111111+0.j,  
       -10.           +0.j])
```

```
figure()
x = [real(s) for s in eigenvalues]
y = [imag(s) for s in eigenvalues]
plot(x, y, "kx")
plot(0.0, 0.0, "k.")
ylim(-12, 12)
gca().set_aspect(1.0)
title("Spectrum of $A - B K$"); grid(True)
```

Spectrum of  $A - BK$



```
y0 = [0.0, 0.0, 1.0, 0.0]
t = linspace(0.0, 20.0, 1000)
yt = array([expm((A-B@K) * t_) for t_ in t]) @ y0
x1t, x2t = yt[:, 0], yt[:, 2]
```

```
figure()  
plot(t, x1t, label="$x_1$")  
plot(t, x2t, label="$x_2$")  
xlabel("$t$")  
grid(True); legend()
```

