

INTERNAL DYNAMICS



Sébastien Boisgérault

CONTROL ENGINEERING WITH PYTHON

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-  [Mines ParisTech, PSL University](#)

SYMBOLS



Code



Worked Example



Graph



Exercise



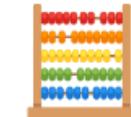
Definition



Numerical Method



Theorem



Analytical Method



Remark



Theory



Information



Hint



Warning



Solution



IMPORTS

```
from numpy import *
from numpy.linalg import *
from scipy.linalg import *
from matplotlib.pyplot import *
from mpl_toolkits.mplot3d import *
from scipy.integrate import solve_ivp
```



STREAMPLOT HELPER

```
def Q(f, xs, ys):
    X, Y = meshgrid(xs, ys)
    v = vectorize
    fx = v(lambda x, y: f([x, y])[0])
    fy = v(lambda x, y: f([x, y])[1])
    return X, Y, fx(X, Y), fy(X, Y)
```



We are interested in the behavior of the solution to

$$\dot{x} = Ax, \quad x(0) = x_0 \in \mathbb{R}^n$$

First, we study some elementary systems in this class.

SCALAR CASE, REAL-VALUED

$$\dot{x} = ax$$

$$a \in \mathbb{R}, \quad x(0) = x_0 \in \mathbb{R}.$$



Solution:

$$x(t) = e^{at}x_0$$



Proof:

$$\frac{d}{dt}e^{at}x_0 = ae^{at}x_0 = ax(t)$$

and

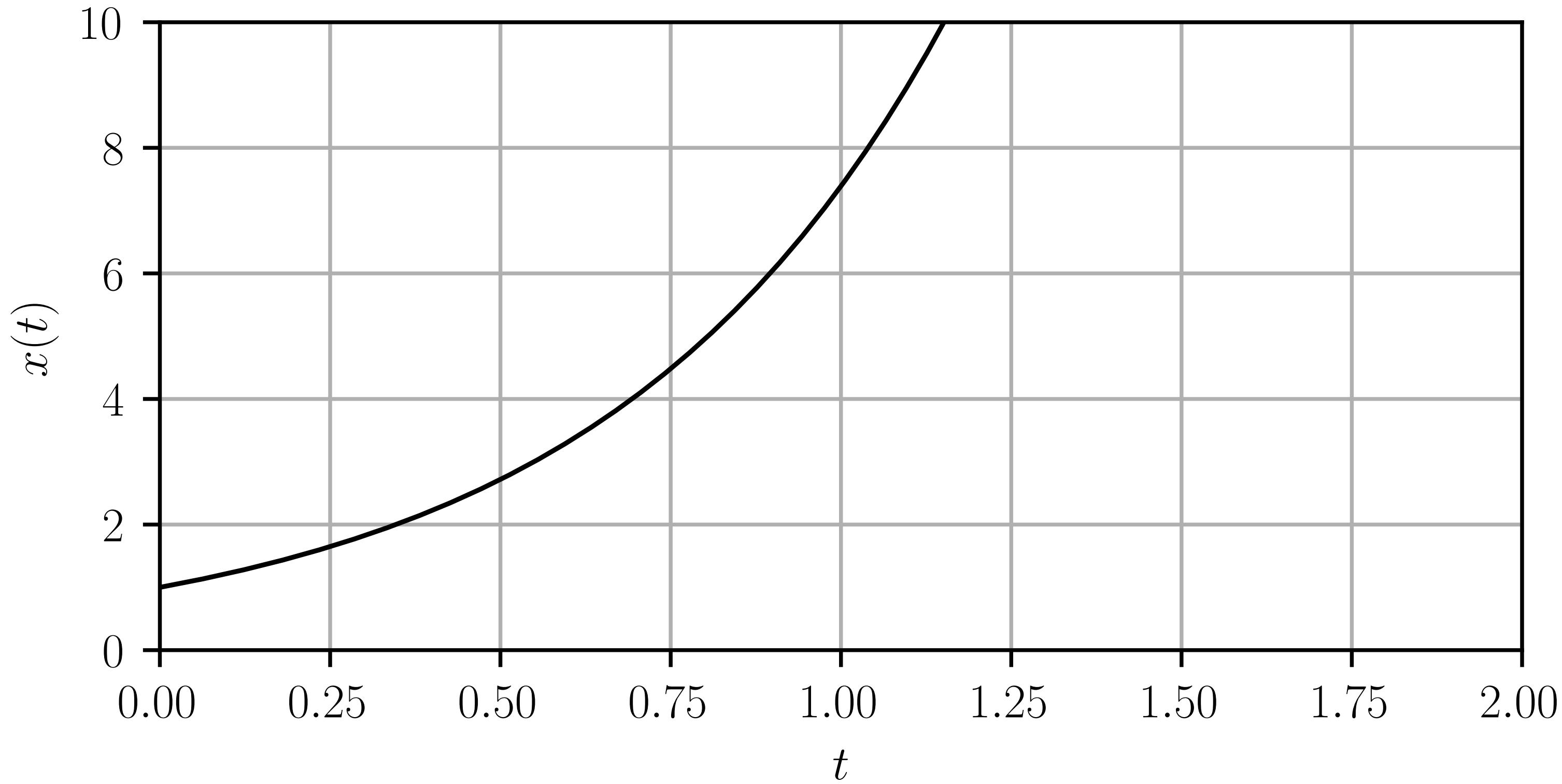
$$x(0) = e^{a \times 0}x_0 = x_0.$$



TRAJECTORY

```
a = 2.0; x0 = 1.0
figure()
t = linspace(0.0, 3.0, 1000)
plot(t, exp(a*t)*x0, "k")
xlabel("$t$"); ylabel("$x(t)$"); title(f"$a={a}$")
grid(); axis([0.0, 2.0, 0.0, 10.0])
```

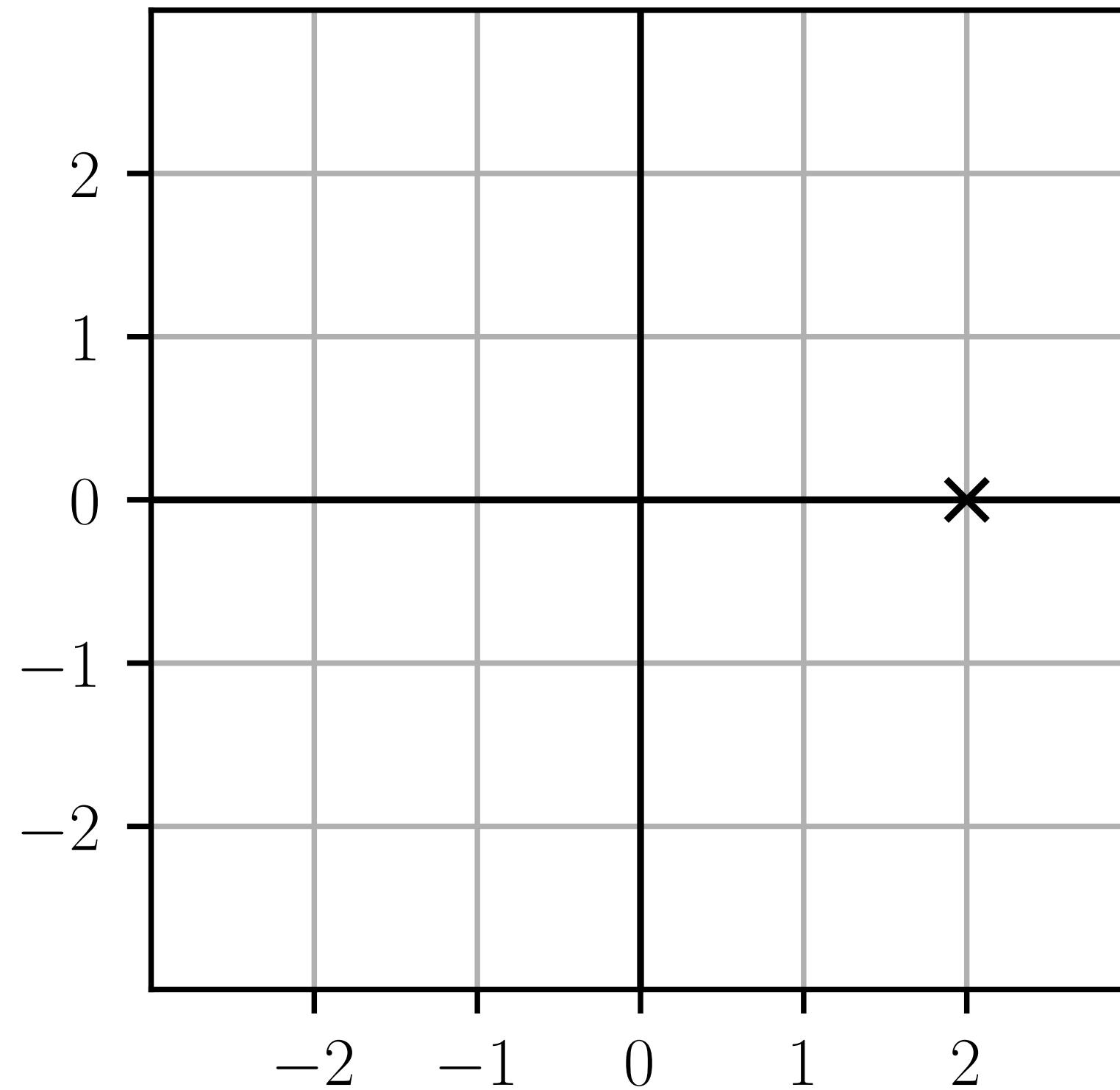
$$a = 2.0$$





```
figure()  
plot(real(a), imag(a), "x", color="k")  
gca().set_aspect(1.0)  
xlim(-3,3); ylim(-3,3);  
plot([-3,3], [0,0], "k")  
plot([0, 0], [-3, 3], "k")  
xticks([-2,-1,0,1,2]); yticks([-2,-1,0,1,2])  
title(f"$a={a}$"); grid(True)
```

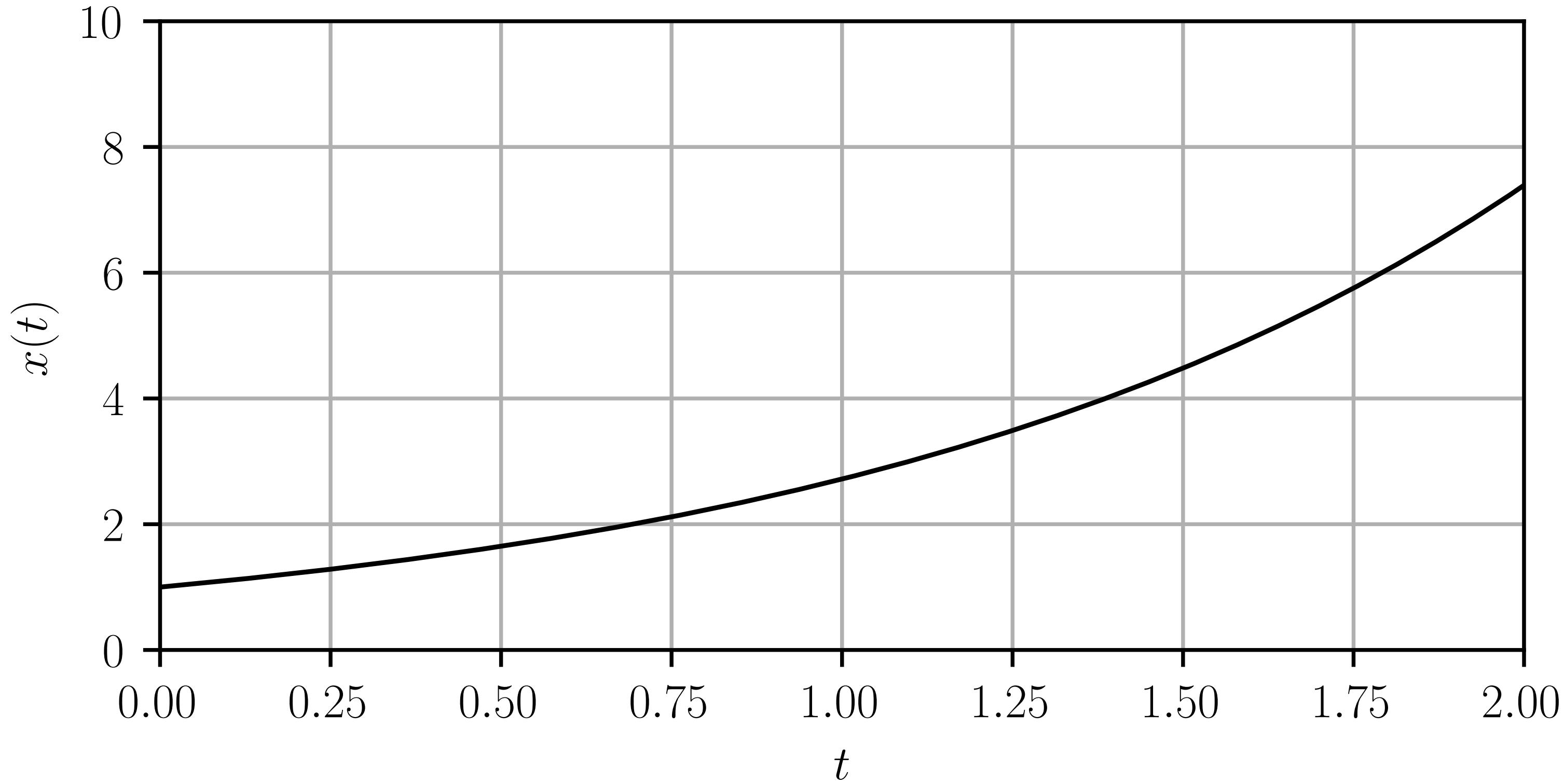
$$a = 2.0$$





```
a = 1.0; x0 = 1.0
figure()
t = linspace(0.0, 3.0, 1000)
plot(t, exp(a*t)*x0, "k")
xlabel("$t$"); ylabel("$x(t)$"); title(f"$a={a}$")
grid(); axis([0.0, 2.0, 0.0, 10.0])
```

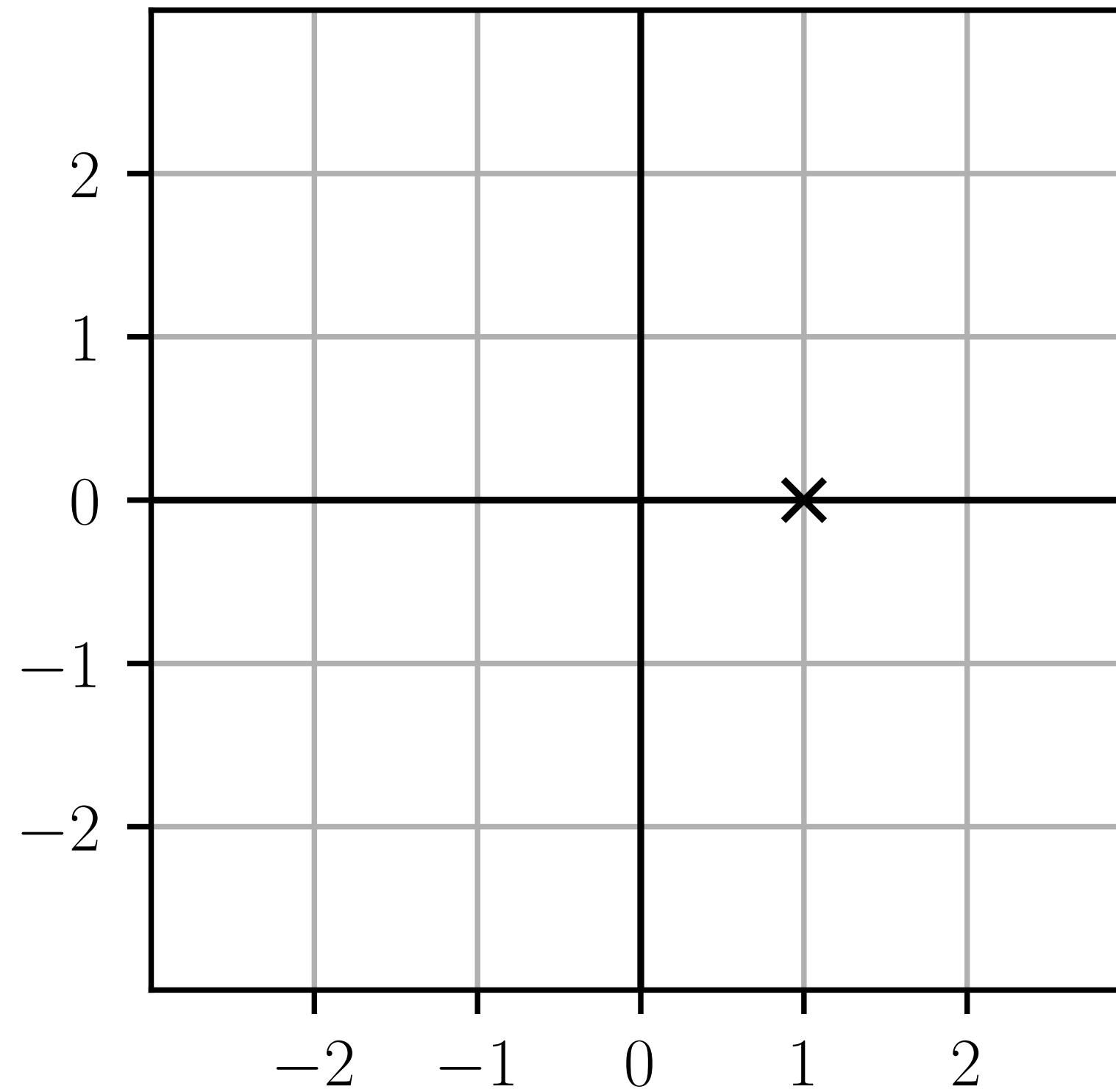
$$a = 1.0$$





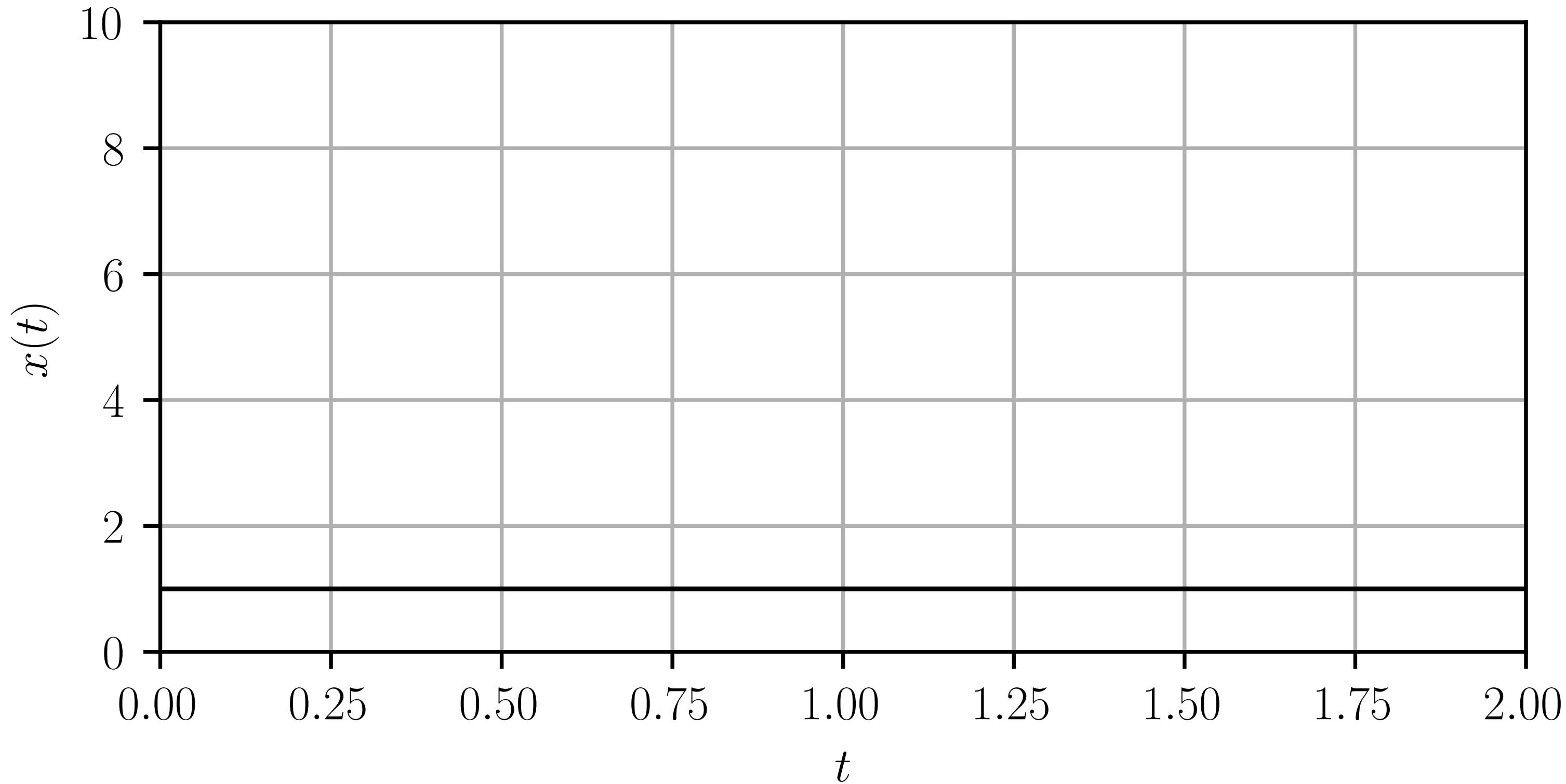
```
figure()  
plot(real(a), imag(a), "x", color="k")  
gca().set_aspect(1.0)  
xlim(-3,3); ylim(-3,3);  
plot([-3,3], [0,0], "k")  
plot([0, 0], [-3, 3], "k")  
xticks([-2,-1,0,1,2]); yticks([-2,-1,0,1,2])  
title(f"$a={a}$"); grid(True)
```

$$a = 1.0$$





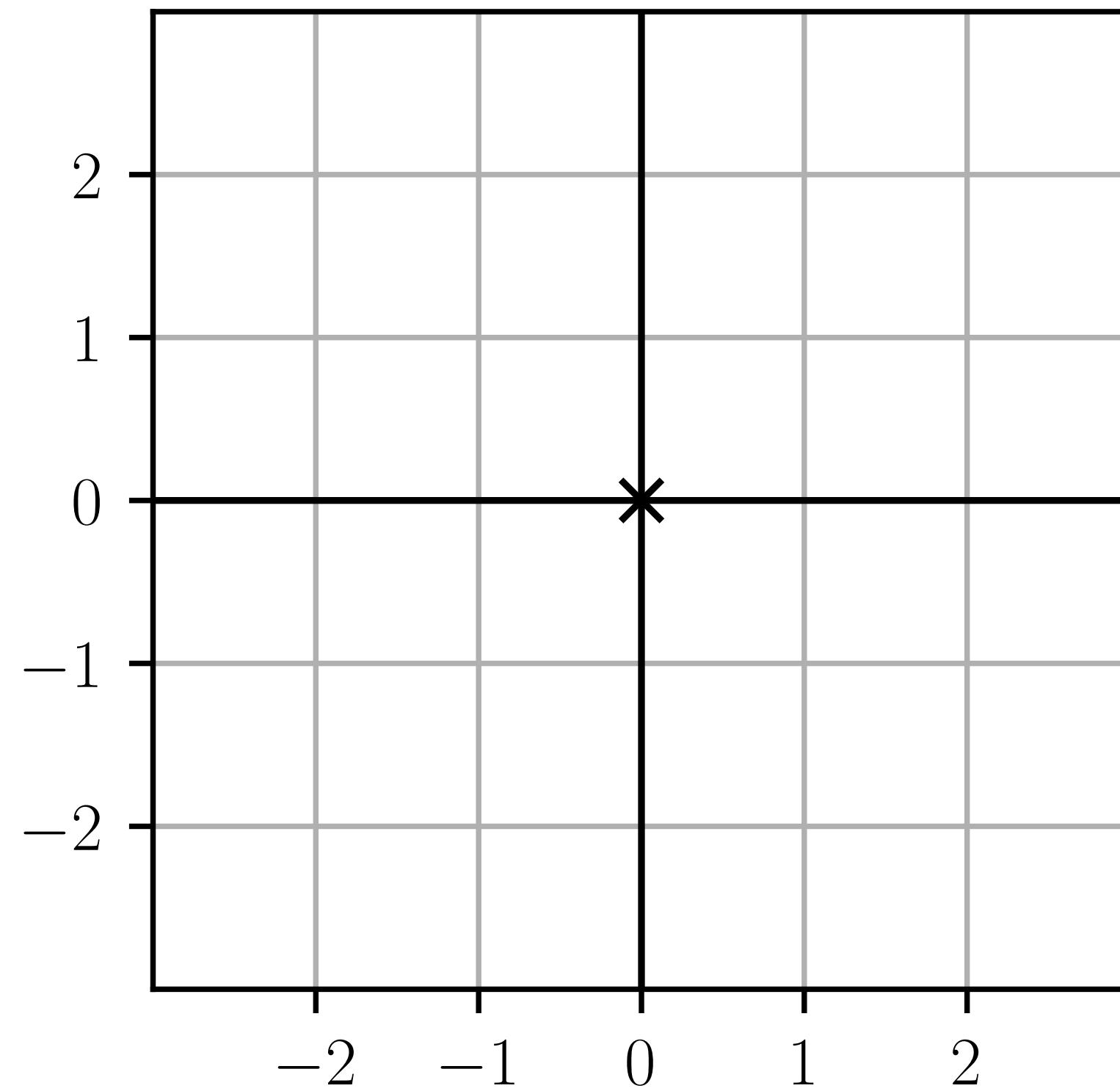
```
a = 0.0; x0 = 1.0
figure()
t = linspace(0.0, 3.0, 1000)
plot(t, exp(a*t)*x0, "k")
xlabel("$t$"); ylabel("$x(t)$"); title(f"$a={a}$")
grid(); axis([0.0, 2.0, 0.0, 10.0])
```

$a = 0.0$ 



```
figure()  
plot(real(a), imag(a), "x", color="k")  
gca().set_aspect(1.0)  
xlim(-3,3); ylim(-3,3);  
plot([-3,3], [0,0], "k")  
plot([0, 0], [-3, 3], "k")  
xticks([-2,-1,0,1,2]); yticks([-2,-1,0,1,2])  
title(f"$a={a}$"); grid(True)
```

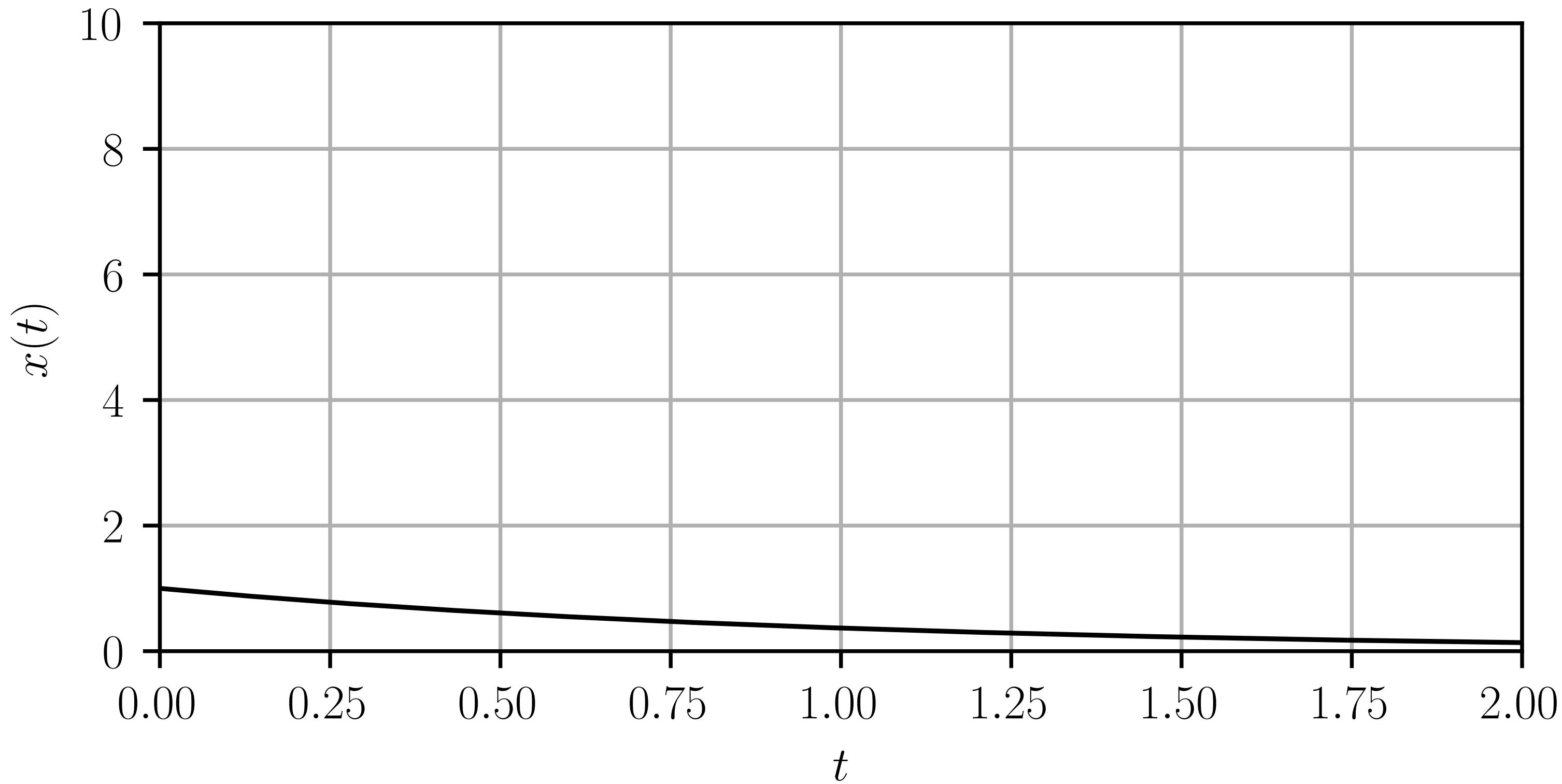
$$a = 0.0$$





```
a = -1.0; x0 = 1.0
figure()
t = linspace(0.0, 3.0, 1000)
plot(t, exp(a*t)*x0, "k")
xlabel("$t$"); ylabel("$x(t)$"); title(f"$a={a}$")
grid(); axis([0.0, 2.0, 0.0, 10.0])
```

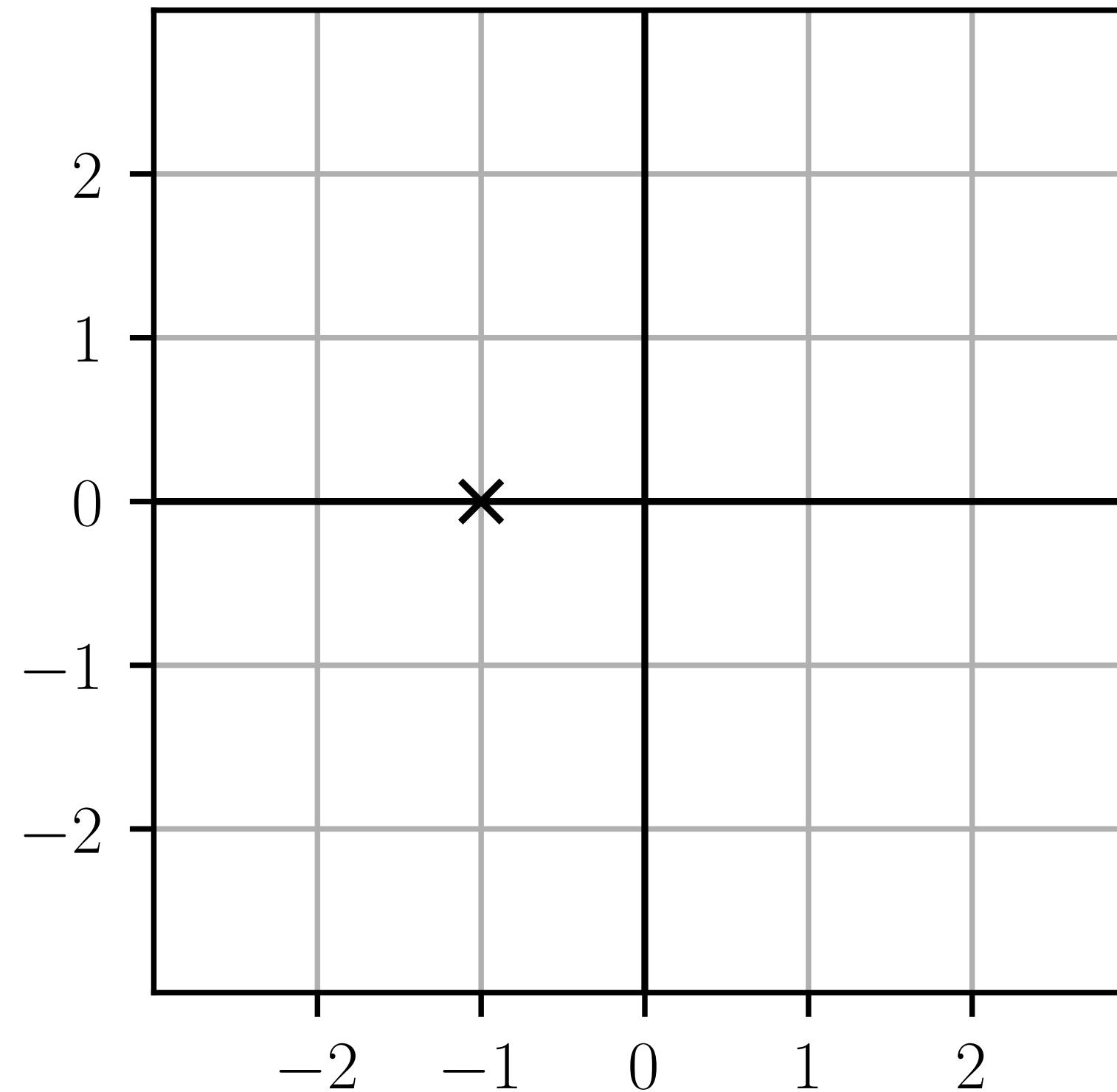
$$a = -1.0$$





```
figure()  
plot(real(a), imag(a), "x", color="k")  
gca().set_aspect(1.0)  
xlim(-3,3); ylim(-3,3);  
plot([-3,3], [0,0], "k")  
plot([0, 0], [-3, 3], "k")  
xticks([-2,-1,0,1,2]); yticks([-2,-1,0,1,2])  
title(f"$a={a}$"); grid(True)
```

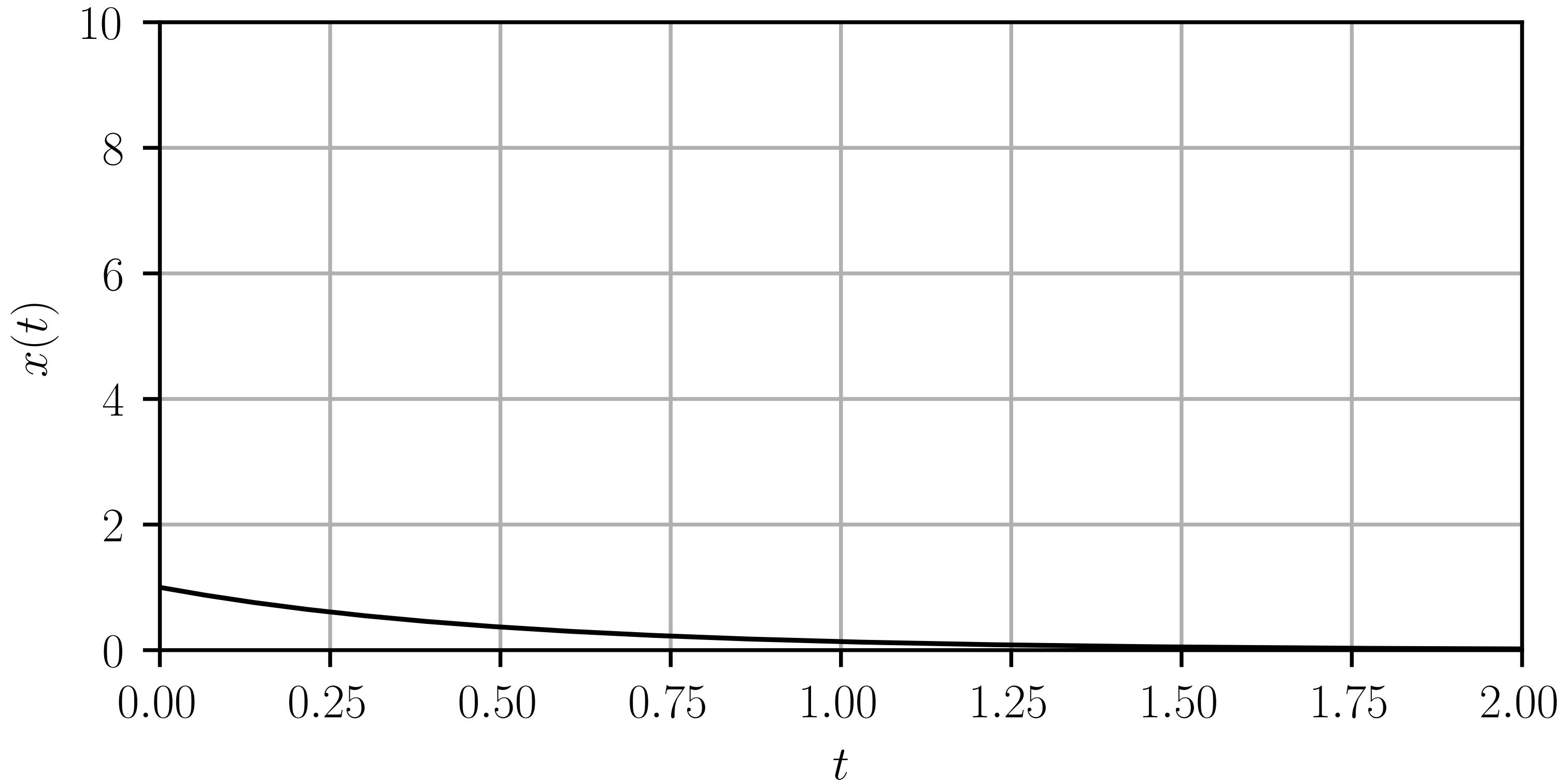
$$a = -1.0$$





```
a = -2.0; x0 = 1.0
figure()
t = linspace(0.0, 3.0, 1000)
plot(t, exp(a*t)*x0, "k")
xlabel("$t$"); ylabel("$x(t)$"); title(f"$a={a}$")
grid(); axis([0.0, 2.0, 0.0, 10.0])
```

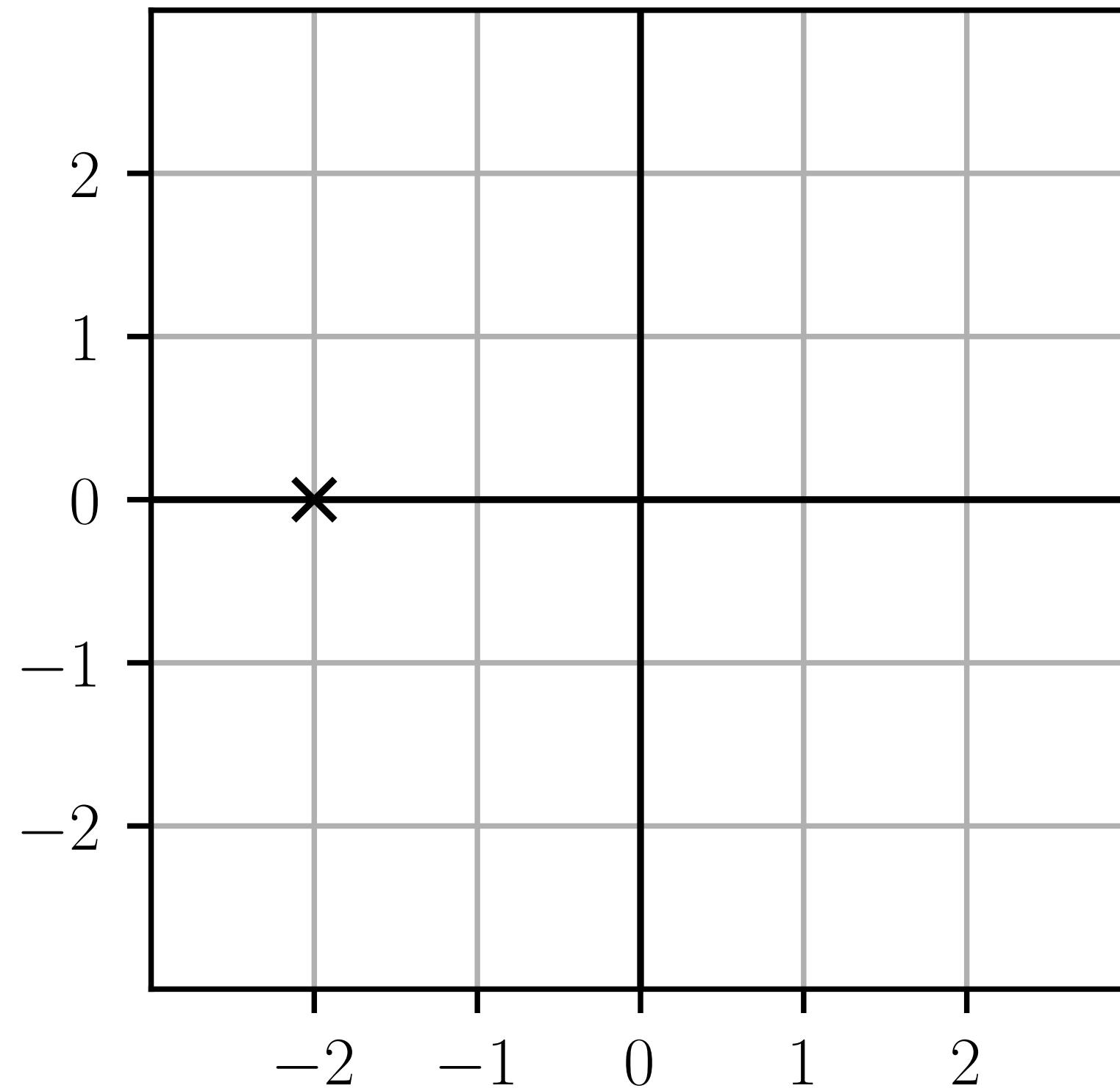
$$a = -2.0$$





```
figure()
plot(real(a), imag(a), "x", color="k")
gca().set_aspect(1.0)
xlim(-3,3); ylim(-3,3);
plot([-3,3], [0,0], "k")
plot([0, 0], [-3, 3], "k")
xticks([-2,-1,0,1,2]); yticks([-2,-1,0,1,2])
title(f"$a={a}$"); grid(True)
```

$$a = -2.0$$





ANALYSIS

The origin is globally asymptotically stable when

$$a < 0.0$$

i.e. a is in the open left-hand plane.

Let the **time constant** τ be

$$\tau := 1/|a|.$$

When the system is asymptotically stable,

$$x(t) = e^{-t/\tau} x_0.$$

QUANTITATIVE CONVERGENCE

τ controls the speed of convergence to the origin:

time t	distance to the origin $ x(t) $
0	$ x(0) $
τ	$\simeq (1/3) x(0) $
3τ	$\simeq (5/100) x(0) $
\vdots	\vdots
$+\infty$	0

VECTOR CASE, DIAGONAL, REAL- VALUED

$$\dot{x}_1 = a_1 x_1, \quad x_1(0) = x_{10}$$

$$\dot{x}_2 = a_2 x_2, \quad x_2(0) = x_{20}$$

i.e.

$$A = \begin{bmatrix} a_1 & 0 \\ 0 & a_2 \end{bmatrix}$$



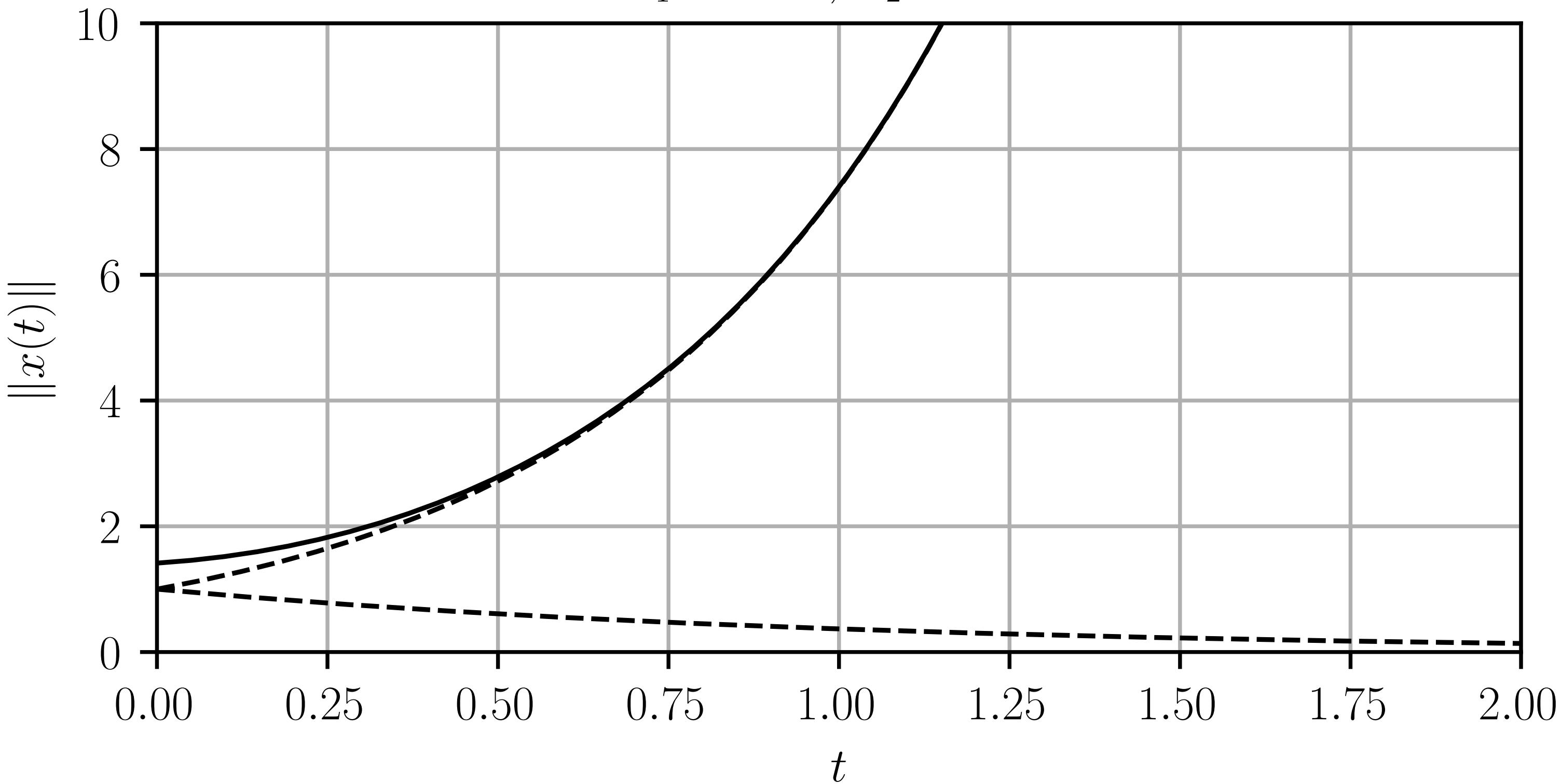
Solution: by linearity

$$x(t) = e^{a_1 t} \begin{bmatrix} x_{10} \\ 0 \end{bmatrix} + e^{a_2 t} \begin{bmatrix} 0 \\ x_{20} \end{bmatrix}$$



```
a1 = -1.0; a2 = 2.0; x10 = x20 = 1.0
figure()
t = linspace(0.0, 3.0, 1000)
x1 = exp(a1*t)*x10; x2 = exp(a2*t)*x20
xn = sqrt(x1**2 + x2**2)
plot(t, xn , "k")
plot(t, x1, "k--")
plot(t, x2 , "k--")
xlabel("$t$"); ylabel("$|x(t)|$"); title(f"$a_1={a1}, \\"; a_2={a2}$")
grid(); axis([0.0, 2.0, 0.0, 10.0])
```

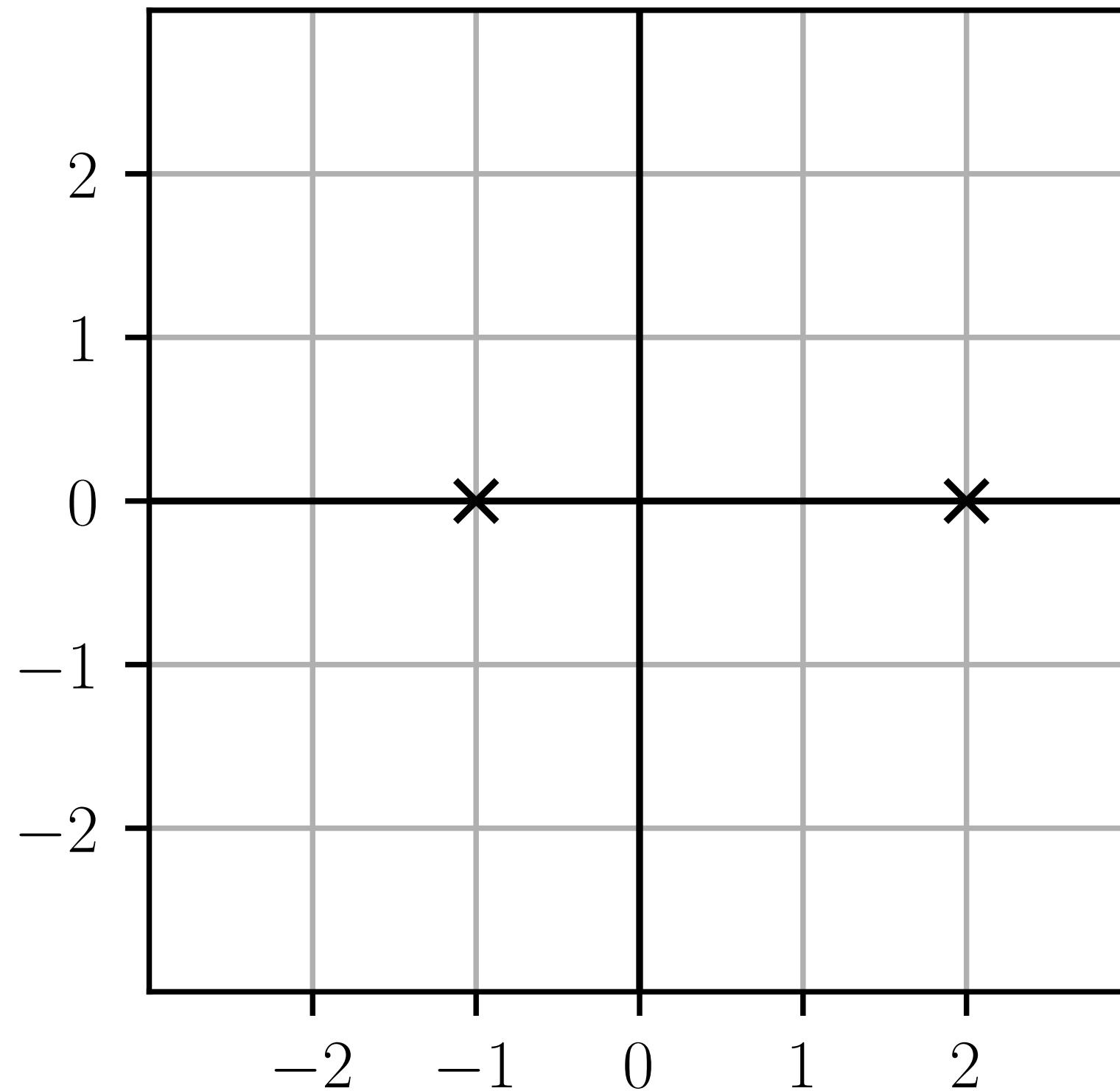
$$a_1 = -1.0, \quad a_2 = 2.0$$





```
figure()
plot(real(a1), imag(a1), "x", color="k")
plot(real(a2), imag(a2), "x", color="k")
gca().set_aspect(1.0)
xlim(-3,3); ylim(-3,3);
plot([-3,3], [0,0], "k")
plot([0, 0], [-3, 3], "k")
xticks([-2,-1,0,1,2]); yticks([-2,-1,0,1,2])
title(f"$a_1={a1}, \\"; a_2={a2}\$")
grid(True)
```

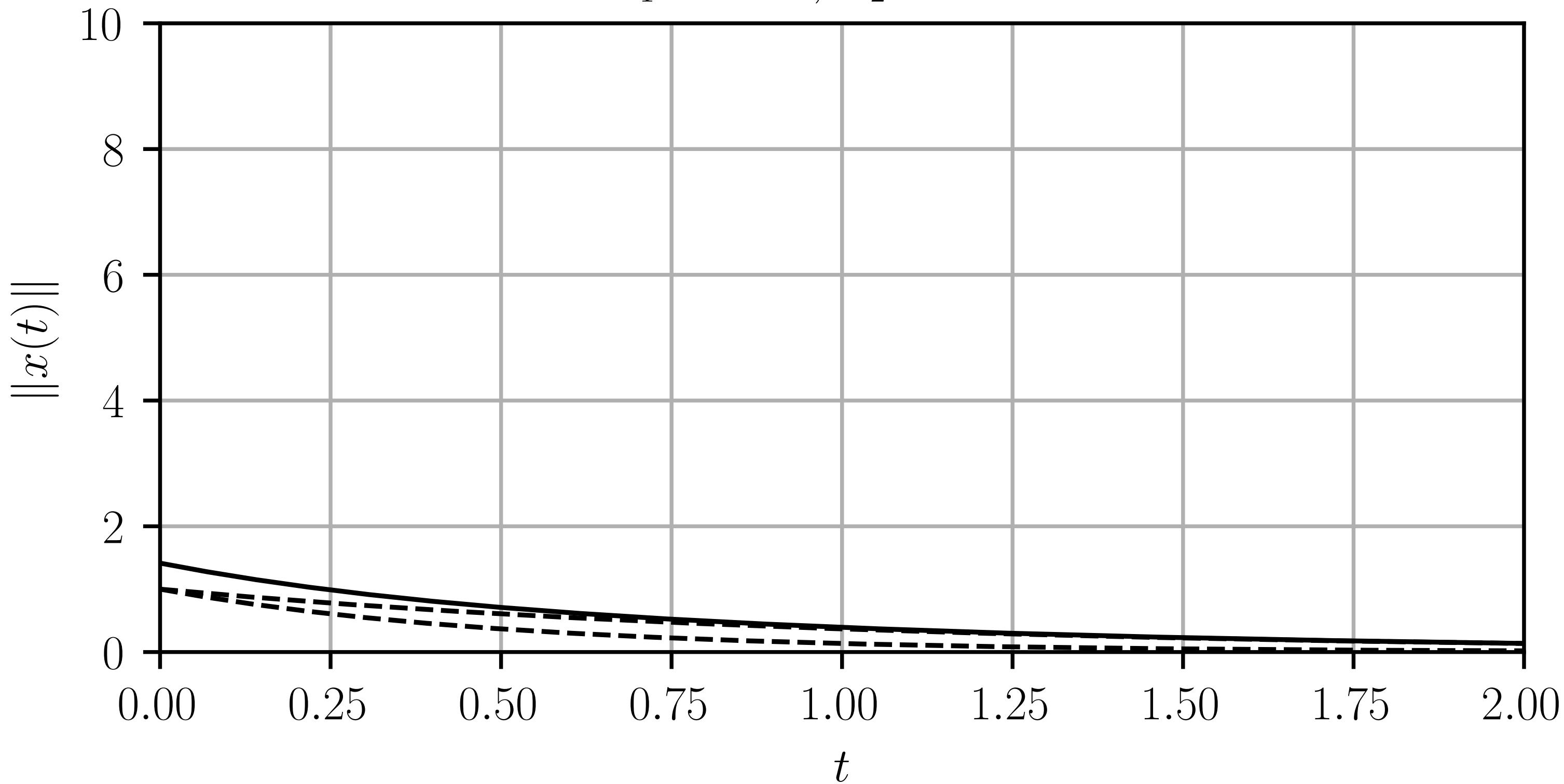
$$a_1 = -1.0, \ a_2 = 2.0$$





```
a1 = -1.0; a2 = -2.0; x10 = x20 = 1.0
figure()
t = linspace(0.0, 3.0, 1000)
x1 = exp(a1*t)*x10; x2 = exp(a2*t)*x20
xn = sqrt(x1**2 + x2**2)
plot(t, xn , "k")
plot(t, x1, "k--")
plot(t, x2 , "k--")
xlabel("$t$"); ylabel("$|x(t)|$"); title(f"$a_1={a1}, \\"; a_2={a2}$")
grid(); axis([0.0, 2.0, 0.0, 10.0])
```

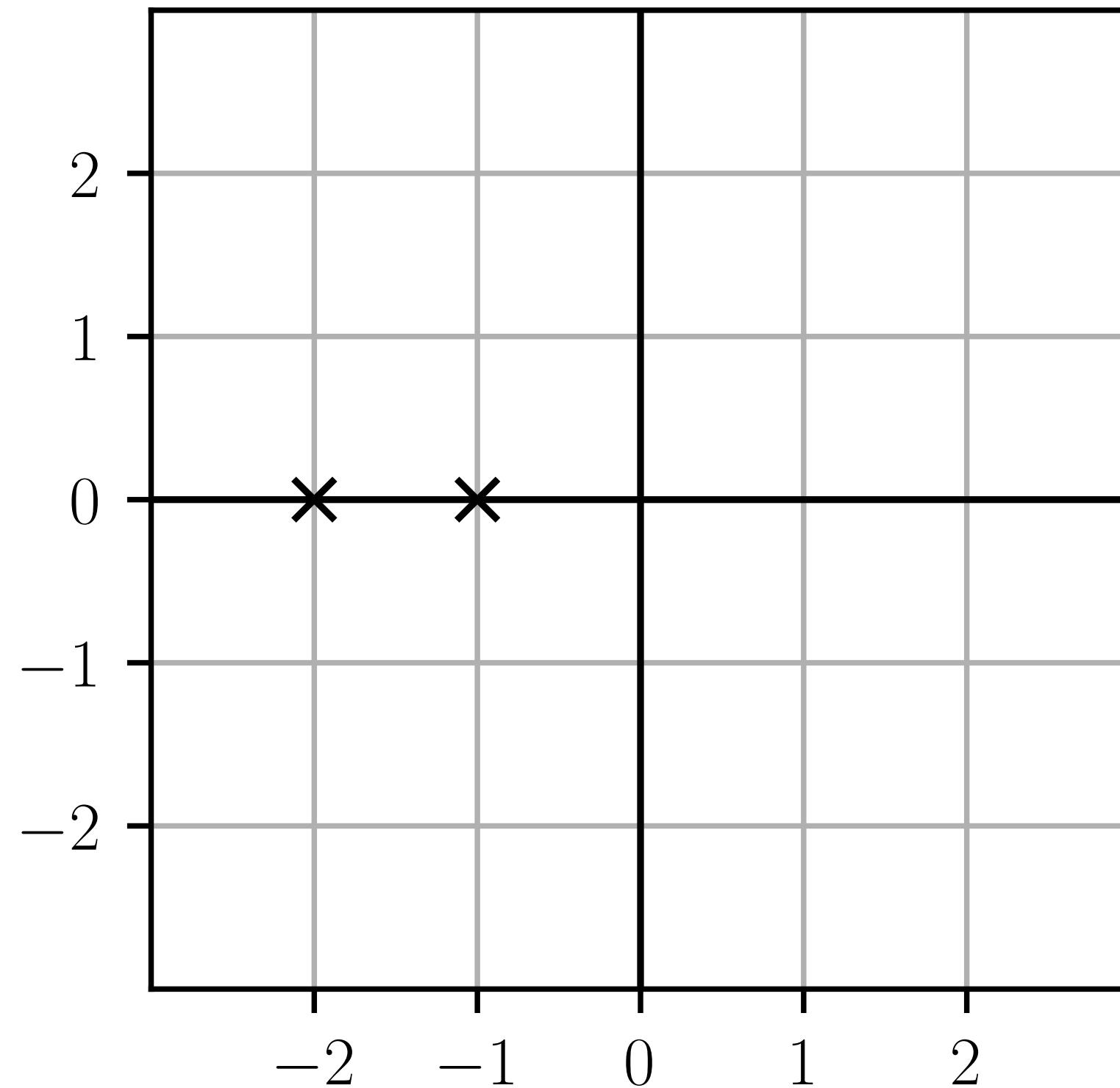
$$a_1 = -1.0, \quad a_2 = -2.0$$





```
figure()
plot(real(a1), imag(a1), "x", color="k")
plot(real(a2), imag(a2), "x", color="k")
gca().set_aspect(1.0)
xlim(-3,3); ylim(-3,3);
plot([-3,3], [0,0], "k")
plot([0, 0], [-3, 3], "k")
xticks([-2,-1,0,1,2]); yticks([-2,-1,0,1,2])
title(f"$a_1={a1}, \\"; a_2={a2}\$")
grid(True)
```

$$a_1 = -1.0, \quad a_2 = -2.0$$





ANALYSIS

- 💎 The rightmost a_i determines the asymptotic behavior,
- 💎 The origin is globally asymptotically stable if and only if
every a_i is in the open left-hand plane.

SCALAR CASE, COMPLEX-VALUED

$$\dot{x} = ax$$

$$a \in \mathbb{C}, x(0) = x_0 \in \mathbb{C}.$$



Solution: formally, the same old solution

$$x(t) = e^{at} x_0$$

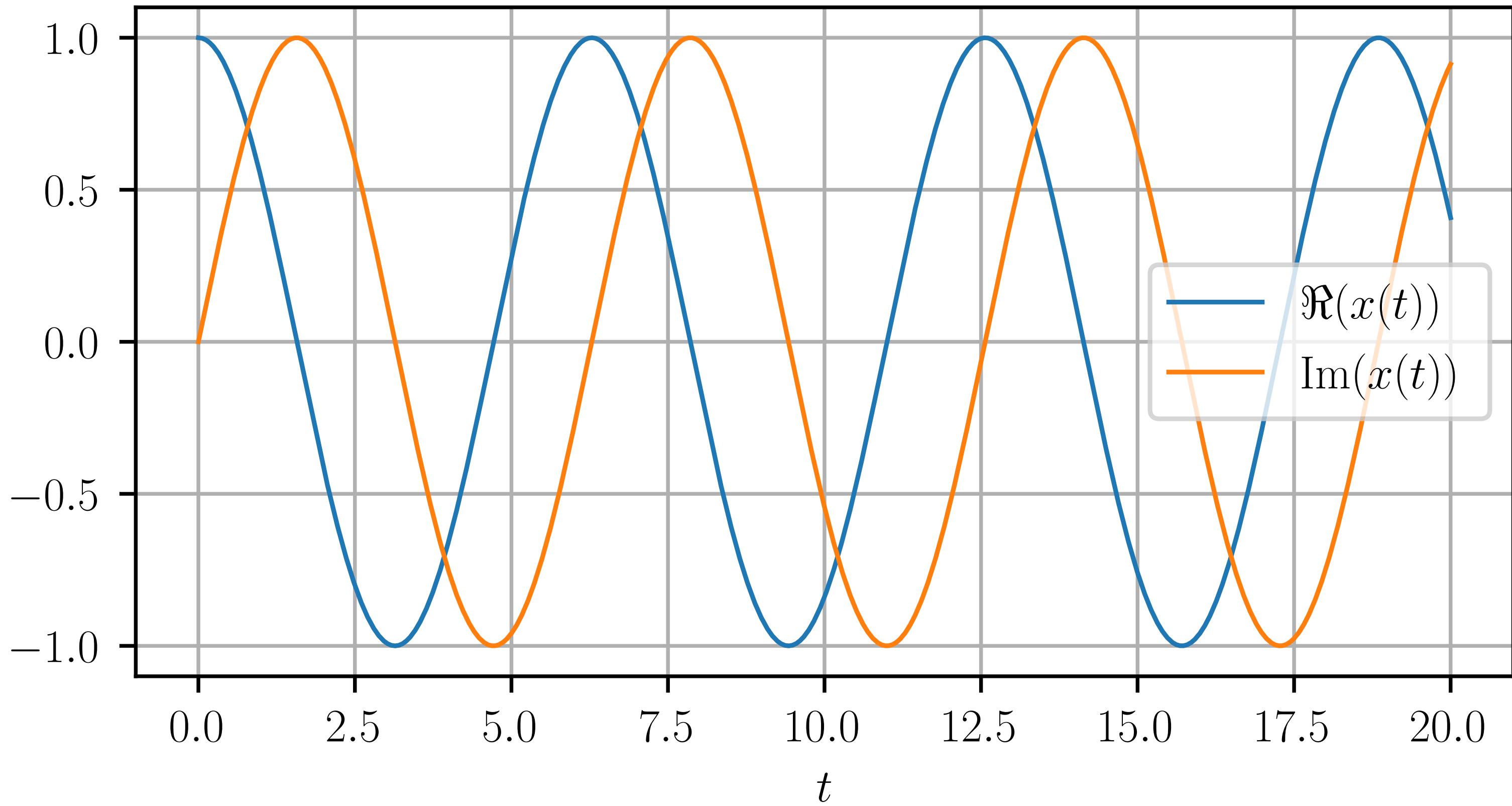
But now, $x(t) \in \mathbb{C}$:

if $a = \sigma + i\omega$ and $x_0 = |x_0|e^{i\angle x_0}$

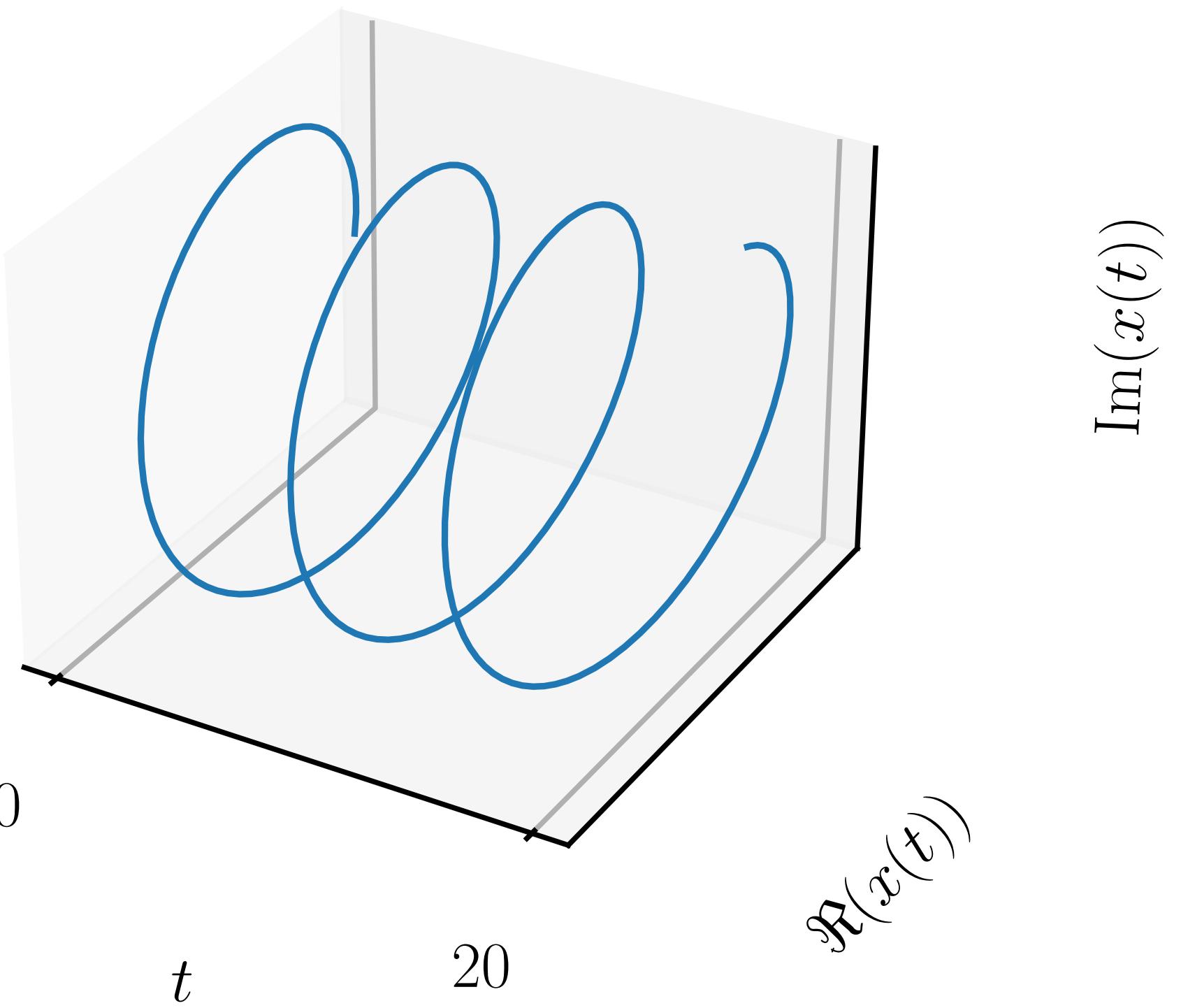
$|x(t)| = |x_0|e^{\sigma t}$ and $\angle x(t) = \angle x_0 + \omega t$.



```
a = 1.0j; x0=1.0
figure()
t = linspace(0.0, 20.0, 1000)
plot(t, real(exp(a*t)*x0), label="$\mathbb{R}\mathrm{e}(x(t))$")
plot(t, imag(exp(a*t)*x0), label="$\mathbb{I}\mathrm{m}(x(t))$")
xlabel("$t$")
legend(); grid()
```



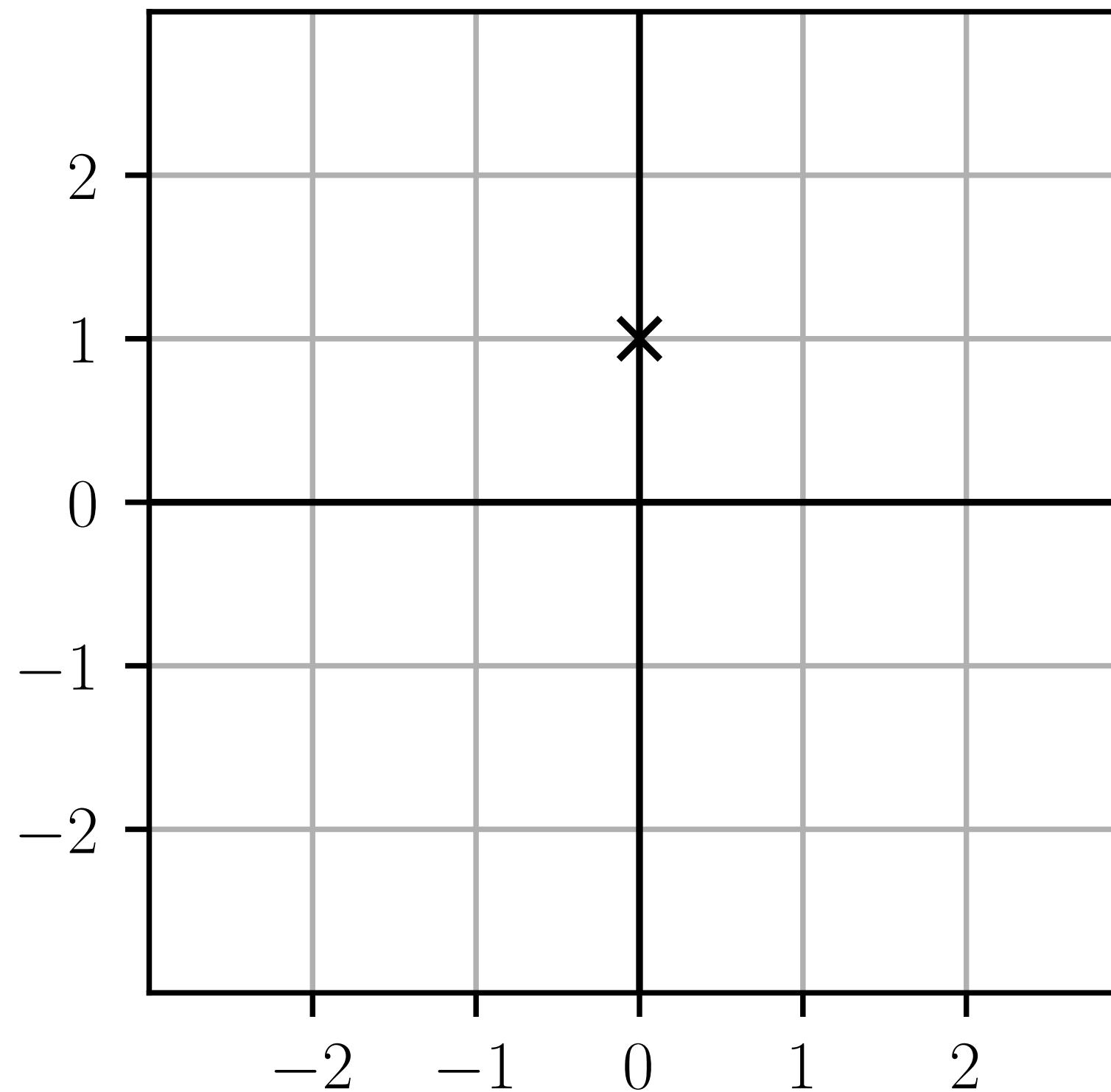
```
fig = figure()
ax = fig.add_subplot(111, projection="3d")
zticks = ax.set_zticks
ax.plot(t, real(exp(a*t)*x0), imag(exp(a*t)*x0))
xticks([0.0, 20.0]); yticks([]); zticks([])
ax.set_xlabel("$t$")
ax.set_ylabel("$\Re(x(t))$")
ax.set_zlabel("$\mathrm{Im}(x(t))$")
```





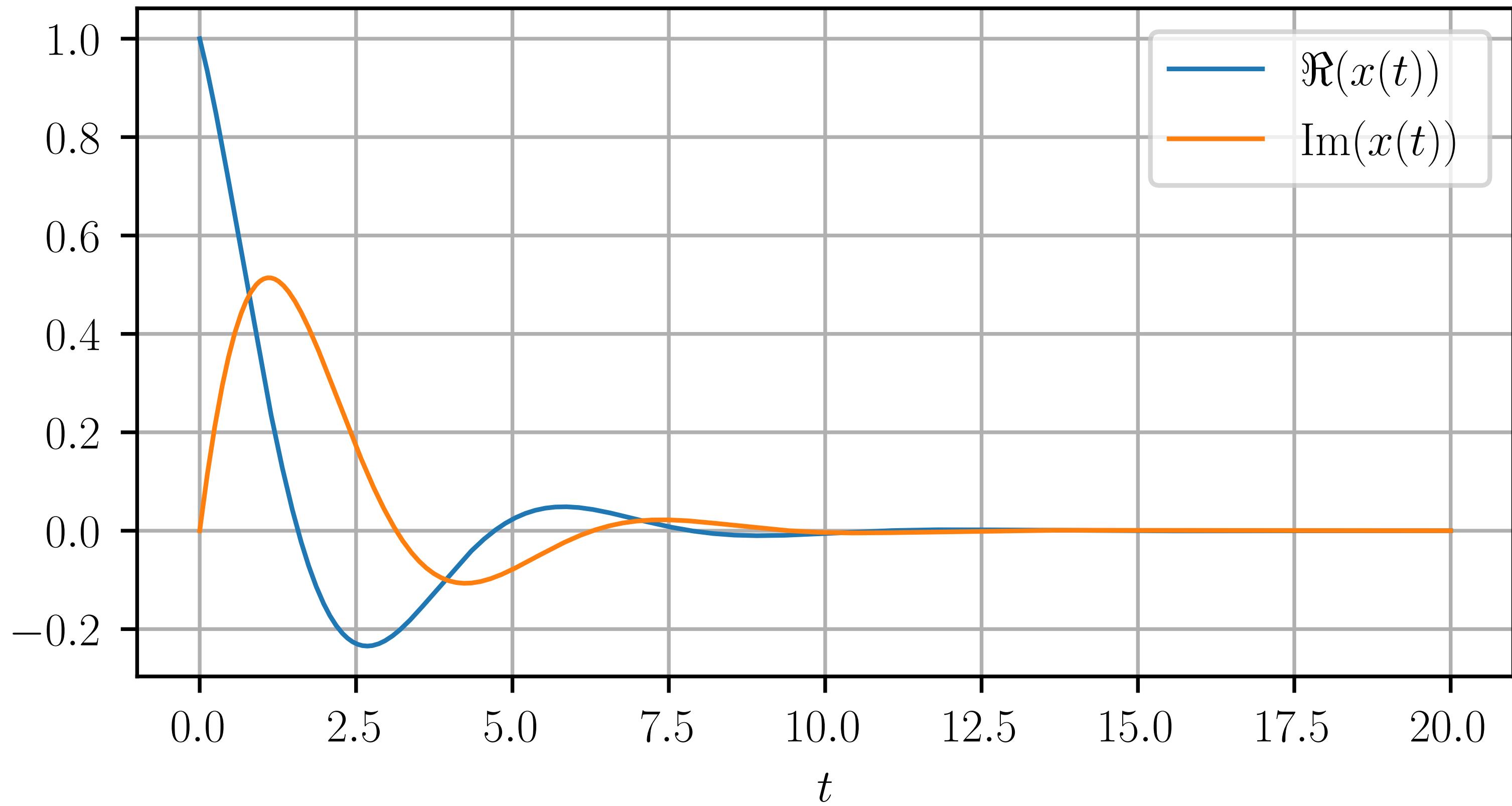
```
figure()  
plot(real(a), imag(a), "x", color="k")  
gca().set_aspect(1.0)  
xlim(-3,3); ylim(-3,3);  
plot([-3,3], [0,0], "k")  
plot([0, 0], [-3, 3], "k")  
xticks([-2,-1,0,1,2]); yticks([-2,-1,0,1,2])  
title(f"$a={a}$"); grid(True)
```

$$a = 1j$$



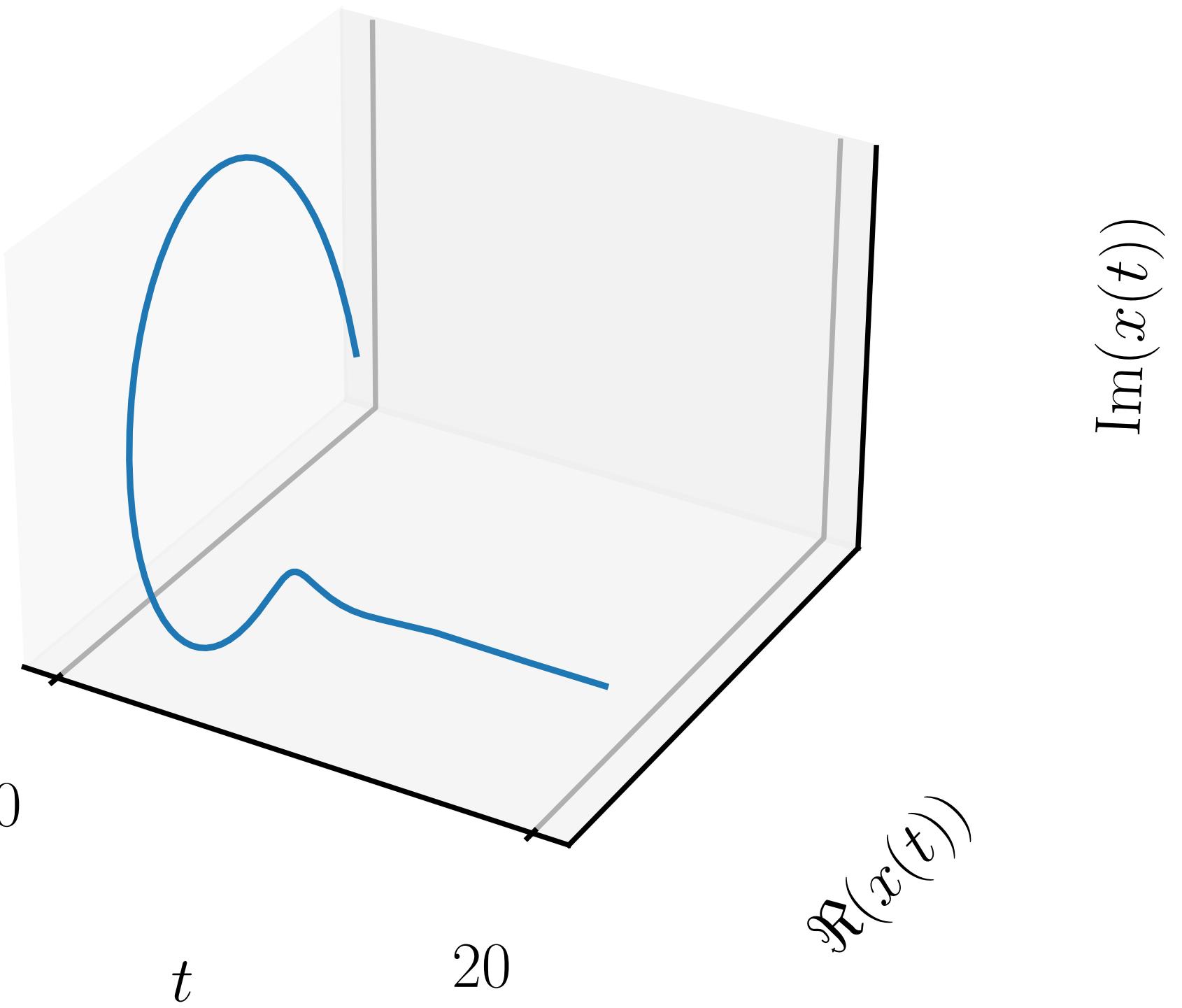


```
a = -0.5 + 1.0j; x0=1.0
figure()
t = linspace(0.0, 20.0, 1000)
plot(t, real(exp(a*t)*x0), label="$\mathbb{R}\mathrm{e}(x(t))$")
plot(t, imag(exp(a*t)*x0), label="$\mathbb{I}\mathrm{m}(x(t))$")
xlabel("$t$")
legend(); grid()
```





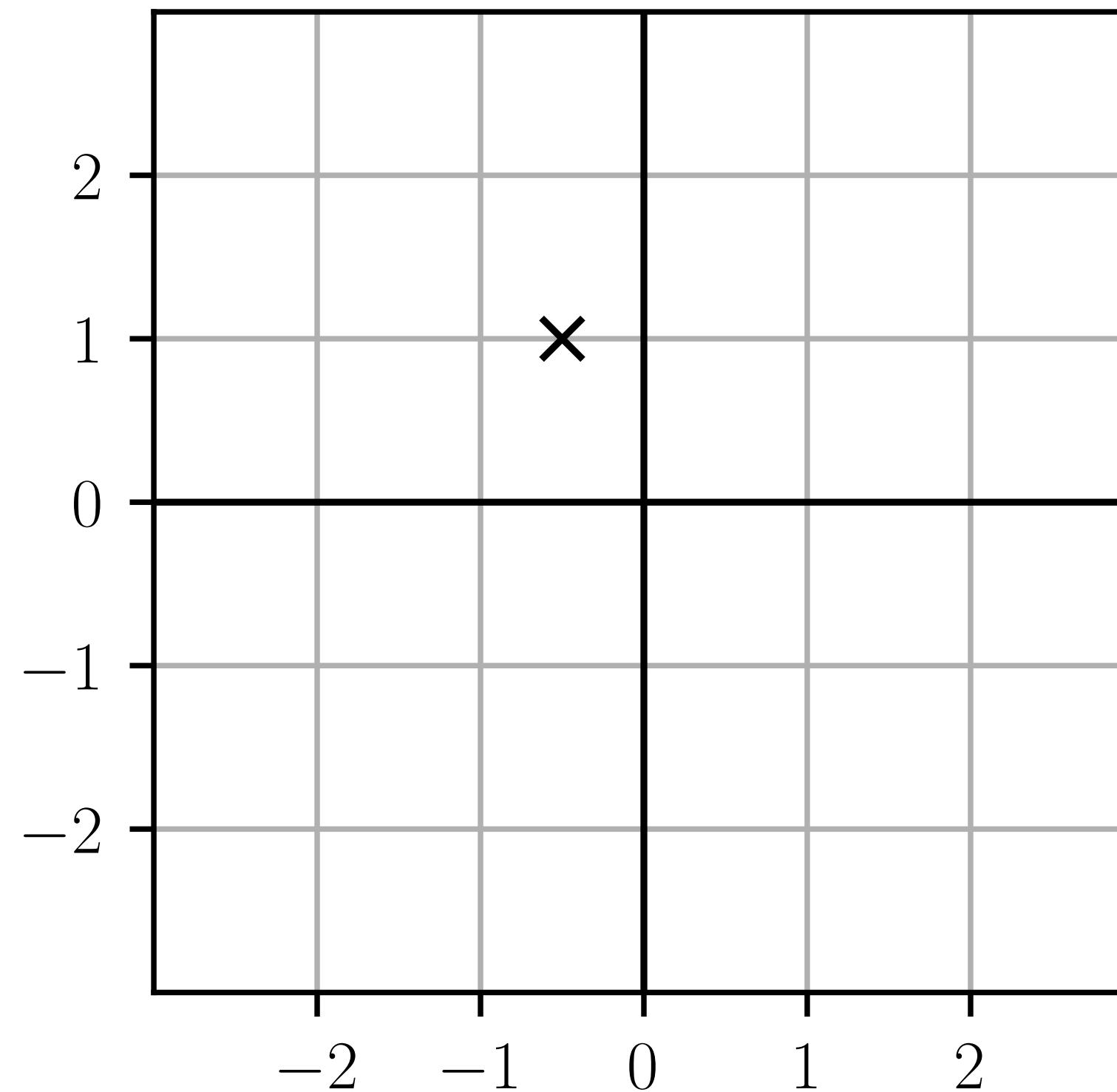
```
fig = figure()
ax = fig.add_subplot(111, projection="3d")
zticks = ax.set_zticks
ax.plot(t, real(exp(a*t)*x0), imag(exp(a*t)*x0))
xticks([0.0, 20.0]); yticks([]); zticks([])
ax.set_xlabel("$t$")
ax.set_ylabel("$\Re(x(t))$")
ax.set_zlabel("$\mathrm{Im}(x(t))$")
```





```
figure()
plot(real(a), imag(a), "x", color="k")
gca().set_aspect(1.0)
xlim(-3,3); ylim(-3,3);
plot([-3,3], [0,0], "k")
plot([0, 0], [-3, 3], "k")
xticks([-2,-1,0,1,2]); yticks([-2,-1,0,1,2])
title(f"$a={a}$")
grid(True)
```

$$a = (-0.5 + 1j)$$





ANALYSIS

- 💎 The origin is globally asymptotically stable iff a is in the open left-hand plane: $\Re(a) < 0$.
- 💎 If $a =: \sigma + i\omega$,
 - 🔑 $\tau = 1/|\sigma|$ is the **time constant**.
 - 🔑 ω the **rotational frequency** of the oscillations.



EXPONENTIAL MATRIX

If $M \in \mathbb{C}^{n \times n}$, its **exponential** is defined as:

$$e^M = \sum_{k=0}^{+\infty} \frac{M^k}{k!} \in \mathbb{C}^{n \times n}$$



The exponential of a matrix M is **not** the matrix with elements $e^{M_{ij}}$ (the elementwise exponential).

- 🐍 elementwise exponential: `exp` (numpy module),
- 🐍 exponential: `expm` (scipy.linalg module).

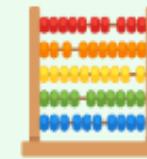


EXPONENTIAL MATRIX

Let

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

1.



Compute the exponential of M .



Hint:

$$\cosh x := \frac{e^x + e^{-x}}{2}, \quad \sinh x := \frac{e^x - e^{-x}}{2}.$$

2.



Compute numerically:

- `exp(M)` (`numpy`)
- `expm(M)` (`scipy.linalg`)

and check the results consistency.



EXPONENTIAL MATRIX

1. 

We have

$$M = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, M^2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$$

and hence for any $j \in \mathbb{N}$,

$$M^{2j+1} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, M^{2j} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I.$$

$$\begin{aligned}
e^M &= \sum_{k=0}^{+\infty} \frac{M^k}{k!} \\
&= \left(\sum_{j=0}^{+\infty} \frac{1}{(2j)!} \right) I + \left(\sum_{j=0}^{+\infty} \frac{1}{(2j+1)!} \right) M \\
&= \left(\sum_{k=0}^{+\infty} \frac{1^k + (-1)^k}{2(k!)} \right) I + \left(\sum_{k=0}^{+\infty} \frac{1^k - (-1)^k}{2(k!)} \right) M \\
&= (\cosh 1)I + (\sinh 1)M
\end{aligned}$$

Thus,

$$e^M = \begin{bmatrix} \cosh 1 & \sinh 1 \\ \sinh 1 & \cosh 1 \end{bmatrix}.$$

2.

```
>>> M = [[0.0, 1.0], [1.0, 0.0]]
```

```
>>> exp(M)
array([[1.          , 2.71828183],
       [2.71828183, 1.          ]])
```

```
>>> expm(M)
array([[1.54308063, 1.17520119],
       [1.17520119, 1.54308063]])
```

These results are consistent:

```
>>> array([[exp(0.0), exp(1.0)],  
...           [exp(1.0), exp(0.0)]])  
  
array([[1.          , 2.71828183],  
       [2.71828183, 1.          ]])
```

```
>>> array([[cosh(1.0), sinh(1.0)],  
...           [sinh(1.0), cosh(1.0)]])  
  
array([[1.54308063, 1.17520119],  
       [1.17520119, 1.54308063]])
```



Note that

$$\begin{aligned}\frac{d}{dt} e^{At} &= \frac{d}{dt} \sum_{n=0}^{+\infty} \frac{A^n}{n!} t^n \\&= \sum_{n=1}^{+\infty} \frac{A^n}{(n-1)!} t^{n-1} \\&= A \sum_{n=1}^{+\infty} \frac{A^{n-1}}{(n-1)!} t^{n-1} = Ae^{At}\end{aligned}$$

Thus, for any $A \in \mathbb{C}^{n \times n}$ and $x_0 \in \mathbb{C}^n$,

$$\frac{d}{dt}(e^{At}x_0) = A(e^{At}x_0)$$



INTERNAL DYNAMICS

The solution of

$$\dot{x} = Ax \text{ and } x(0) = x_0$$

is

$$x(t) = e^{At}x_0.$$



G.A.S. \Leftrightarrow L.A.

 For any dynamical system, if the origin is a globally asymptotically stable equilibrium, then it is a locally attractive equilibrium.

 For linear systems, the converse result also holds.

 Let's prove this!

1. 

Show that for any linear system $\dot{x} = Ax$, if the origin is locally attractive, then it is also globally attractive.

2.

Show that linear system $\dot{x} = Ax$, if the origin is globally attractive, then it is also globally asymptotically stable.



Hint: Consider the solutions $e_k(t) := e^{At}e_k$ associated to $e_k(0) = e_k$ where (e_1, \dots, e_n) is the canonical basis of the state space.



G.A.S. \leftrightarrow L.A.

1. 

If the origin is locally attractive, then there is a $\varepsilon > 0$ such that for any $x_0 \in \mathbb{R}^n$ such that $\|x_0\| \leq \varepsilon$,

$$\lim_{t \rightarrow +\infty} e^{At} x_0 = 0.$$

Now, let any $x_0 \in \mathbb{R}^n$. Since the norm of $\varepsilon x_0 / \|x_0\|$ is ε , and by linearity of e^{At} , we obtain

$$\begin{aligned}\lim_{t \rightarrow +\infty} e^{At} x_0 &= \lim_{t \rightarrow +\infty} e^{At} \left(\frac{\|x_0\|}{\varepsilon} \varepsilon \frac{x_0}{\|x_0\|} \right) \\ &= \frac{\|x_0\|}{\varepsilon} \lim_{t \rightarrow +\infty} e^{At} \left(\varepsilon \frac{x_0}{\|x_0\|} \right) \\ &= 0.\end{aligned}$$

Thus the origin is globally attractive.

2. 

Let X_0 be a bounded set of \mathbb{R}^n . Since

$$x_0 = \sum_{k=1}^n x_{0k} e_k,$$

the solution $x(t)$ of $\dot{x} = Ax, x(0) = x_0$ satisfies

$$x(t) = e^{At} x_0 = e^{At} \left(\sum_{k=1}^n x_{0k} e_k \right) = \sum_{k=1}^n x_{0k} e^{At} e_k.$$

$$\begin{aligned}
\|x(t)\| &= \left\| \sum_{k=1}^n x_{0k} e^{At} e_k \right\| \\
&\leq \sum_{k=1}^n |x_{0k}| \|e^{At} e_k\| \\
&= \sum_{k=1}^n |x_{0k}| \|e_k(t)\| \\
&\leq \left(\sum_{k=1}^n |x_{0k}| \right) \max_{k=1, \dots, n} \|e_k(t)\|
\end{aligned}$$

Since X_0 is bounded, there is a $\alpha > 0$ such that for any $x_0 = (x_{01}, \dots, x_{0n})$ in X_0 ,

$$\|x_0\|_1 := \sum_{k=1}^n |x_{0k}| \leq \alpha.$$

Since for every $k = 1, \dots, n$, $\lim_{t \rightarrow +\infty} \|e_k(t)\| = 0$,

$$\lim_{t \rightarrow +\infty} \max_{k=1, \dots, n} \|e_k(t)\| = 0.$$

Finally

$$\begin{aligned}\|x(t, x_0)\| &\leq \left(\sum_{k=1}^n |x_{0k}| \right) \max_{k=1, \dots, n} \|e_k(t)\| \\ &\leq \alpha \max_{k=1, \dots, n} \|e_k(t)\|\end{aligned}$$

Thus $\|x(t, x_0)\| \rightarrow 0$ when $t \rightarrow \infty$, uniformly w.r.t. $x_0 \in X_0$. In other words, the origin is globally asymptotically stable.



EIGENVALUE & EIGENVECTOR

Let $A \in \mathbb{C}^n$. If $x \neq 0 \in \mathbb{C}^n$, $s \in \mathbb{C}$ and

$$Ax = sx$$

x is an eigenvector of A , s is an eigenvalue of A .

The **spectrum** of A is the set of its eigenvalues.

It is characterized by:

$$\sigma(A) := \{s \in \mathbb{C} \mid \det(sI - A) = 0\}.$$



MODES & POLES

Consider the system $\dot{x} = Ax$.

- a **mode** of the system is an eigenvector of A ,
- a **pole** of the system is an eigenvalue of A .



STABILITY CRITERIA

Let $A \in \mathbb{C}^{n \times n}$.

The origin of $\dot{x} = Ax$ is globally asymptotically stable

$$\iff$$

all eigenvalues of A have a negative real part.

$$\iff$$

$$\max\{\Re s \mid s \in \sigma(A)\} < 0.$$

WHY DOES THIS CRITERIA WORK?

Assume that:

- A is diagonalizable.

( very likely unless A has some special structure.)

Let $\sigma(A) = \{\lambda_1, \dots, \lambda_n\}$.

There is an invertible matrix $P \in \mathbb{C}^{n \times n}$ such that

$$P^{-1}AP = \text{diag}(\lambda_1, \dots, \lambda_n) = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & \cdots & \lambda_n \end{bmatrix}$$

Thus, if $y = P^{-1}x$, $\dot{x} = Ax$ is equivalent to

$$\begin{cases} \dot{y}_1 = \lambda_1 y_1 \\ \dot{y}_2 = \lambda_2 y_2 \\ \vdots = \vdots \\ \dot{y}_n = \lambda_n y_n \end{cases}$$

The system is G.A.S. iff each component of the system is, which holds iff $\Re \lambda_i < 0$ for each i .

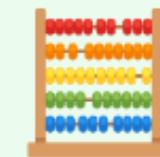


SPRING-MASS SYSTEM

Consider the scalar ODE

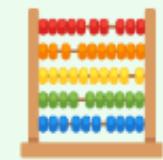
$$\ddot{x} + kx = 0, \text{ with } k > 0$$

1.



Represent this system as a first-order ODE.

2.



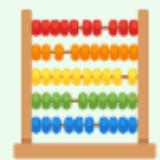
Is this system asymptotically stable?



Do the solutions have oscillatory components?

Find the set of associated rotational frequencies.

4.



Same set of questions (1., 2., 3.) for

$$\ddot{x} + b\dot{x} + kx = 0$$

when $b > 0$.



SPRING-MASS SYSTEM

1. 

$$\frac{d}{dt} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -k & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} = A \begin{bmatrix} x \\ \dot{x} \end{bmatrix}$$

2. 

We have

$$\max\{\Re s \mid s \in \sigma(A)\} = 0,$$

hence the system is not globally asymptotically stable.

3. 

Since

$$\det(sI - A) = \det \begin{pmatrix} s & -1 \\ k & s \end{pmatrix} = s^2 + k,$$

the spectrum of A is

$$\sigma(A) = \{s \in \mathbb{C} \mid \det(sI - A) = 0\} = \{i\sqrt{k}, -i\sqrt{k}\}.$$

The system poles are $\pm i\sqrt{k}$.

The general solution $x(t)$ can be decomposed as

$$x(t) = x_+ e^{i\sqrt{k}t} + x_- e^{-i\sqrt{k}t}.$$

Thus the components of $x(t)$ oscillate at the rotational frequency

$$\omega = \sqrt{k}.$$

4. 

$$\frac{d}{dt} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -k & -b \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} = A \begin{bmatrix} x \\ \dot{x} \end{bmatrix}$$

$$\det(sI - A) = \det \begin{pmatrix} s & -1 \\ k & s+b \end{pmatrix} = s^2 + bs + k,$$

Let $\Delta := b^2 - 4k$. If $b \geq 2\sqrt{k}$, then $\Delta \geq 0$ and

$$\sigma(A) = \left\{ \frac{-b + \sqrt{\Delta}}{2}, \frac{-b - \sqrt{\Delta}}{2} \right\}.$$

Otherwise,

$$\sigma(A) = \left\{ \frac{-b + i\sqrt{-\Delta}}{2}, \frac{-b - i\sqrt{-\Delta}}{2} \right\}.$$

Thus, if $b \geq 2\sqrt{k}$,

$$\max\{\Re s \mid s \in \sigma(A)\} = \frac{-b + \sqrt{b^2 - 4k}}{2} < 0$$

and otherwise

$$\max\{\Re s \mid s \in \sigma(A)\} = -\frac{b}{2} < 0.$$

In each case, the system is globally asymptotically stable.

If $b \geq 2\sqrt{k}$, the poles are real-valued; the components of the solution do not oscillate.

If $0 < b < 2\sqrt{k}$, the imaginary part of the poles is

$$\pm \frac{\sqrt{4k - b^2}}{2} = \pm \sqrt{k - (b/2)^2},$$

thus the solution components oscillate at the rotational frequency

$$\omega = \sqrt{k - (b/2)^2}.$$

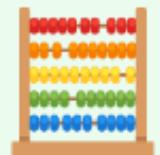


INTEGRATOR CHAIN

Consider the system

$$\dot{x} = Jx \text{ with } J = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ 0 & 0 & 0 & \cdots & 0 \end{bmatrix}$$

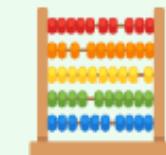
1.



Compute the solution $x(t)$ when

$$x(0) = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}.$$

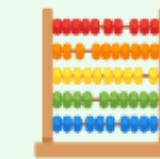
2.



Compute the solution for an arbitrary $x(0)$

$$x(0) = \begin{bmatrix} x_1(0) \\ \vdots \\ x_n(0) \end{bmatrix}.$$

3.



Same questions for the system

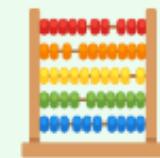
$$\dot{x} = (\lambda I + J)x$$

for some $\lambda \in \mathbb{C}$.



Hint: Find the ODE satisfied by $y(t) := x(t)e^{-\lambda t}$.

4.



Is the system asymptotically stable ?

5. 

Why does the stability analysis of this system matter ?



INTEGRATOR CHAIN

1.



Let $x = (x_1, \dots, x_n)$.

The ODE $\dot{x} = Jx$ is equivalent to:

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = x_3$$

$$\begin{matrix} \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots \end{matrix}$$

$$\dot{x}_{n-1} = x_n$$

$$\dot{x}_n = 0.$$

When $x(0) = (0, \dots, 0, 1)$,

- $\dot{x}_n = 0$ yields $x_n(t) = 1$, then
- $\dot{x}_{n-1} = x_n$ yields $x_{n-1}(t) = t$,
- ...
- $\dot{x}_k = x_{k+1}$ yields

$$x_k(t) = \frac{t^{n-k}}{(n - k)!}.$$

To summarize:

$$x(t) = \begin{bmatrix} t^{n-1}/(n-1)! \\ \vdots \\ t^{n-1-k}/(n-1-k)! \\ \vdots \\ t \\ 1 \end{bmatrix}$$

2. 

We note that

$$x(0) = x_1(0) \begin{bmatrix} 1 \\ \vdots \\ 0 \\ 0 \end{bmatrix} + \cdots + x_{n-1}(0) \begin{bmatrix} 0 \\ \vdots \\ 1 \\ 0 \end{bmatrix} + x_n(0) \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}.$$

Similarly to the previous question, we find that:

$$x(0) = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \end{bmatrix} \rightarrow x(t) = \begin{bmatrix} t^{n-2}/(n-2)! \\ \vdots \\ t \\ 1 \\ 0 \end{bmatrix}$$

$$x(0) = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ 0 \\ 0 \end{bmatrix} \rightarrow x(t) = \begin{bmatrix} t^{n-3}/(n-3)! \\ \vdots \\ 1 \\ 0 \\ 0 \end{bmatrix}$$

And more generally, by linearity:

$$x(t) = \begin{bmatrix} x_1(0) + \cdots + x_{n-1}(0) \frac{t^{n-2}}{(n-2)!} + x_n(0) \frac{t^{n-1}}{(n-1)!} \\ \vdots \\ x_{n-2}(0) + x_{n-1}(0)t + x_n(0) \frac{t^2}{2} \\ x_{n-1}(0) + x_n(0)t \\ x_n(0) \end{bmatrix}$$

3. 

If $\dot{x}(t) = (\lambda I + J)x(t)$ and $y(t) = x(t)e^{-\lambda t}$, then

$$\begin{aligned}\dot{y}(t) &= \dot{x}(t)e^{-\lambda t} + x(t)(-\lambda e^{-\lambda t}) \\ &= (\lambda I + J)x(t)e^{-\lambda t} - \lambda Ix(t)e^{-\lambda t} \\ &= Jx(t)e^{-\lambda t} \\ &= Jy(t).\end{aligned}$$

Since $y(0) = x(0)e^{-\lambda 0} = x(0)$ we get

$$x(t) = \begin{bmatrix} x_1(0) + \cdots + x_n(0) \frac{t^{n-1}}{(n-1)!} \\ \vdots \\ x_{n-1}(0) + x_n(0)t \\ x_n(0) \end{bmatrix} e^{\lambda t}.$$

4.

The structure of $x(t)$ shows that

- If $\Re\lambda < 0$, then the system is asymptotically stable.
- If $\Re\lambda \geq 0$, then the system is not.

For example when $x(0) = (1, 0, \dots, 0)$, we have

$$x(t) = (1, 0, \dots, 0).$$

5.



Every square complex matrix A , even if it is not diagonalizable, can be decomposed into a block-diagonal matrix where each block has the structure $\lambda I + J$.

Thus, the result of the previous question allows to prove the  Stability Criteria in the general case.