

Real, distinct eigenvalues (Sect. 5.7)

- ▶ Review: $n \times n$ linear differential systems.
- ▶ The case of diagonalizable matrices.
- ▶ Second order equations and first order systems.
- ▶ Examples: 2×2 linear systems.
- ▶ Classification of 2×2 diagonalizable systems.
- ▶ Phase portraits for 2×2 systems.

Review: $n \times n$ linear differential systems.

Recall:

- ▶ Given an $n \times n$ matrix $A(t)$, n -vector $\mathbf{b}(t)$, find $\mathbf{x}(t)$ solution

$$\mathbf{x}'(t) = A(t)\mathbf{x}(t) + \mathbf{b}(t).$$

- ▶ The system is *homogeneous* iff $\mathbf{b} = 0$, that is,

$$\mathbf{x}'(t) = A(t)\mathbf{x}(t).$$

- ▶ The system has *constant coefficients* iff matrix A does not depend on t , that is,

$$\mathbf{x}'(t) = A\mathbf{x}(t) + \mathbf{b}(t).$$

- ▶ We study homogeneous, constant coefficient systems, that is,

$$\mathbf{x}'(t) = A\mathbf{x}(t).$$

Review: $n \times n$ linear differential systems.

Recall:

- ▶ Given continuous functions A, \mathbf{b} on $(t_1, t_2) \subset \mathbb{R}$, a constant $t_0 \in (t_1, t_2)$ and a vector \mathbf{x}_0 , there exists a unique function \mathbf{x} solution of the IVP

$$\mathbf{x}'(t) = A(t)\mathbf{x}(t) + \mathbf{b}(t), \quad \mathbf{x}(t_0) = \mathbf{x}_0.$$

- ▶ Today we learn to find such solution in the case of homogeneous, constant coefficients, $n \times n$ linear systems,

$$\mathbf{x}'(t) = A\mathbf{x}(t).$$

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The case of diagonalizable matrices.

Theorem (Diagonalizable matrix)

If $n \times n$ matrix A is diagonalizable, with a linearly independent eigenvectors set $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ and corresponding eigenvalues $\{\lambda_1, \dots, \lambda_n\}$, then the general solution \mathbf{x} to the homogeneous, constant coefficients, linear system

$$\mathbf{x}'(t) = A\mathbf{x}(t)$$

is given by the expression below, where $c_1, \dots, c_n \in \mathbb{R}$,

$$\mathbf{x}(t) = c_1\mathbf{v}_1 e^{\lambda_1 t} + \dots + c_n\mathbf{v}_n e^{\lambda_n t}.$$

Remark:

- ▶ The differential system for the variable \mathbf{x} is coupled, that is, A is not diagonal.
- ▶ We transform the system into a system for a variable \mathbf{y} such that the system for \mathbf{y} is decoupled, that is, $\mathbf{y}'(t) = D\mathbf{y}(t)$, where D is a diagonal matrix.
- ▶ We solve for $\mathbf{y}(t)$ and we transform back to $\mathbf{x}(t)$.

The case of diagonalizable matrices.

Proof: Since A is diagonalizable, we know that $A = PDP^{-1}$, with

$$P = [\mathbf{v}_1, \dots, \mathbf{v}_n], \quad D = \text{diag}[\lambda_1, \dots, \lambda_n].$$

Equivalently, $P^{-1}AP = D$. Multiply $\mathbf{x}' = A\mathbf{x}$ by P^{-1} on the left

$$P^{-1}\mathbf{x}'(t) = P^{-1}A\mathbf{x}(t) \Leftrightarrow (P^{-1}\mathbf{x})' = (P^{-1}AP)(P^{-1}\mathbf{x}).$$

Introduce the new unknown $\mathbf{y}(t) = P^{-1}\mathbf{x}(t)$, then

$$\mathbf{y}'(t) = D\mathbf{y}(t) \Leftrightarrow \begin{cases} y_1'(t) = \lambda_1 y_1(t), \\ \vdots \\ y_n'(t) = \lambda_n y_n(t), \end{cases} \Rightarrow \mathbf{y}(t) = \begin{bmatrix} c_1 e^{\lambda_1 t} \\ \vdots \\ c_n e^{\lambda_n t} \end{bmatrix}.$$

The case of diagonalizable matrices.

Proof: Recall: $\mathbf{y}(t) = P^{-1}\mathbf{x}(t)$, and $\mathbf{y}(t) = \begin{bmatrix} c_1 e^{\lambda_1 t} \\ \vdots \\ c_n e^{\lambda_n t} \end{bmatrix}$.

Transform back to $\mathbf{x}(t)$, that is,

$$\mathbf{x}(t) = P\mathbf{y}(t) = [\mathbf{v}_1, \dots, \mathbf{v}_n] \begin{bmatrix} c_1 e^{\lambda_1 t} \\ \vdots \\ c_n e^{\lambda_n t} \end{bmatrix}$$

We conclude: $\mathbf{x}(t) = c_1\mathbf{v}_1 e^{\lambda_1 t} + \dots + c_n\mathbf{v}_n e^{\lambda_n t}$. □

Remark:

- ▶ $A\mathbf{v}_i = \lambda_i\mathbf{v}_i$.
- ▶ The eigenvalues and eigenvectors of A are crucial to solve the differential linear system $\mathbf{x}'(t) = A\mathbf{x}(t)$.

The case of diagonalizable matrices.

Remark: Here is another argument useful to understand why the vector $\mathbf{x}(t) = c_1\mathbf{v}_1 e^{\lambda_1 t} + \dots + c_n\mathbf{v}_n e^{\lambda_n t}$ is solution of the linear system $\mathbf{x}'(t) = A\mathbf{x}(t)$. On the one hand, derivate \mathbf{x} ,

$$\mathbf{x}'(t) = c_1\lambda_1\mathbf{v}_1 e^{\lambda_1 t} + \dots + c_n\lambda_n\mathbf{v}_n e^{\lambda_n t}.$$

On the other hand, compute $A\mathbf{x}(t)$,

$$A\mathbf{x}(t) = c_1(A\mathbf{v}_1) e^{\lambda_1 t} + \dots + c_n(A\mathbf{v}_n) e^{\lambda_n t},$$

$$A\mathbf{x}(t) = c_1\lambda_1\mathbf{v}_1 e^{\lambda_1 t} + \dots + c_n\lambda_n\mathbf{v}_n e^{\lambda_n t}.$$

We conclude: $\mathbf{x}'(t) = A\mathbf{x}(t)$.

Remark: Unlike the proof of the Theorem, this second argument does not show that $\mathbf{x}(t) = c_1\mathbf{v}_1 e^{\lambda_1 t} + \dots + c_n\mathbf{v}_n e^{\lambda_n t}$ are all possible solutions to the system.

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Second order equations and first order systems.

Theorem (Reduction to first order)

Every solution y to the second order linear equation

$$y'' + p(t)y' + q(t)y = g(t), \quad (1)$$

defines a solution $x_1 = y$ and $x_2 = y'$ of the 2×2 first order linear differential system

$$x_1' = x_2, \quad (2)$$

$$x_2' = -q(t)x_1 - p(t)x_2 + g(t). \quad (3)$$

Conversely, every solution x_1, x_2 of the 2×2 first order linear system in Eqs. (2)-(3) defines a solution $y = x_1$ of the second order differential equation in (1).

Second order equations and first order systems.

Proof:

(\Rightarrow) Given y solution of $y'' + p(t)y' + q(t)y = g(t)$,
introduce $x_1 = y$ and $x_2 = y'$, hence $x_1' = y' = x_2$, that is,

$$x_1' = x_2.$$

Then, $x_2' = y'' = -q(t)y - p(t)y' + g(t)$. That is,

$$x_2' = -q(t)x_1 - p(t)x_2 + g(t).$$

(\Leftarrow) Introduce $x_2 = x_1'$ into $x_2' = -q(t)x_1 - p(t)x_2 + g(t)$.

$$x_1'' = -q(t)x_1 - p(t)x_1' + g(t),$$

that is

$$x_1'' + p(t)x_1' + q(t)x_1 = g(t).$$

□

Second order equations and first order systems.

Example

Express as a first order system the equation

$$y'' + 2y' + 2y = \sin(at).$$

Solution: Introduce the new unknowns

$$x_1 = y, \quad x_2 = y' \quad \Rightarrow \quad x_1' = x_2.$$

Then, the differential equation can be written as

$$x_2' + 2x_2 + 2x_1 = \sin(at).$$

We conclude that

$$x_1' = x_2.$$

$$x_2' = -2x_1 - 2x_2 + \sin(at).$$

◁

Second order equations and first order systems.

Remark: Systems of first order equations can, sometimes, be transformed into a second order single equation.

Example

Express as a single second order equation the 2×2 system and solve it,

$$x_1' = -x_1 + 3x_2,$$

$$x_2' = x_1 - x_2.$$

Solution: Compute x_1 from the second equation: $x_1 = x_2' + x_2$.
Introduce this expression into the first equation,

$$(x_2' + x_2)' = -(x_2' + x_2) + 3x_2,$$

$$x_2'' + x_2' = -x_2' - x_2 + 3x_2,$$

$$x_2'' + 2x_2' - 2x_2 = 0.$$

Second order equations and first order systems.

Example

Express as a single second order equation the 2×2 system and solve it,

$$x_1' = -x_1 + 3x_2,$$

$$x_2' = x_1 - x_2.$$

Solution: Recall: $x_2'' + 2x_2' - 2x_2 = 0$.

$$r^2 + 2r - 2 = 0 \Rightarrow r_{\pm} = \frac{1}{2} [-2 \pm \sqrt{4 + 8}] \Rightarrow r_{\pm} = -1 \pm \sqrt{3}.$$

Therefore, $x_2 = c_1 e^{r_+ t} + c_2 e^{r_- t}$. Since $x_1 = x_2' + x_2$,

$$x_1 = (c_1 r_+ e^{r_+ t} + c_2 r_- e^{r_- t}) + (c_1 e^{r_+ t} + c_2 e^{r_- t}),$$

We conclude: $x_1 = c_1(1 + r_+) e^{r_+ t} + c_2(1 + r_-) e^{r_- t}$. \triangleleft

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Examples: 2×2 linear systems.

Example

Find the general solution to $\mathbf{x}' = A\mathbf{x}$, with $A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$.

Solution: Find eigenvalues and eigenvectors of A . We found that:

$$\lambda_1 = 4, \quad \mathbf{v}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad \text{and} \quad \lambda_2 = -2, \quad \mathbf{v}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}.$$

Fundamental solutions are

$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

The general solution is $\mathbf{x}(t) = c_1 \mathbf{x}^{(1)}(t) + c_2 \mathbf{x}^{(2)}(t)$, that is,

$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}, \quad c_1, c_2 \in \mathbb{R}. \quad \triangleleft$$

Examples: 2×2 linear systems.

Remark:

Re-writing the solution vector $\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}$ in

components $\mathbf{x}(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$, then

$$x_1(t) = c_1 e^{4t} - c_2 e^{-2t}, \quad x_2(t) = c_1 e^{4t} + c_2 e^{-2t}.$$

Introducing the fundamental matrix $X(t) = [\mathbf{x}^{(1)}(t), \mathbf{x}^{(2)}(t)]$ and the vector \mathbf{c} ,

$$X(t) = \begin{bmatrix} e^{4t} & -e^{-2t} \\ e^{4t} & e^{-2t} \end{bmatrix}, \quad \mathbf{c} = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix},$$

then the general solution above can be expressed as follows

$$\mathbf{x}(t) = X(t)\mathbf{c} \Leftrightarrow \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = \begin{bmatrix} e^{4t} & -e^{-2t} \\ e^{4t} & e^{-2t} \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}.$$

Examples: 2×2 linear systems.

Example

Solve the IVP $\mathbf{x}' = A\mathbf{x}$, where $\mathbf{x}(0) = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$, and $A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$.

Solution: The general solution: $\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}$.

The initial condition is,

$$\mathbf{x}(0) = \begin{bmatrix} 2 \\ 4 \end{bmatrix} = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix}.$$

We need to solve the linear system

$$\begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \end{bmatrix} \Rightarrow \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 4 \end{bmatrix}.$$

Therefore, $\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$, hence $\mathbf{x}(t) = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t} + \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}$. \triangleleft

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Classification of 2×2 diagonalizable systems.

Remark:

Diagonalizable 2×2 matrices A with real coefficients are classified according to their eigenvalues.

- (a) Matrix A has two different, real eigenvalues $\lambda_1 \neq \lambda_2$, so it has two non-proportional eigenvectors $\mathbf{v}_1, \mathbf{v}_2$ (eigen-directions). (Section 5.7)
- (b) Matrix A has two different, complex eigenvalues $\lambda_1 = \bar{\lambda}_2$, so it has two non-proportional eigenvectors $\mathbf{v}_1, \mathbf{v}_2$. (Section 5.8)
- (c-1) Matrix A has repeated, real eigenvalues, $\lambda_1 = \lambda_2 \in \mathbb{R}$ with two non-proportional eigenvectors $\mathbf{v}_1, \mathbf{v}_2$. (Section 5.9)

Remark:

- (c-2) We will also study in Section 5.9 how to obtain solutions to a 2×2 system $\mathbf{x}' = A\mathbf{x}$ in the case that A is not diagonalizable and A has only one eigen-direction.

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Phase portraits for 2×2 systems.

Remark:

- ▶ There are two main types of graphs for solutions of 2×2 linear systems:
 - (i) The graphs of the vector components;
 - (ii) The phase portrait.
- ▶ Case (i): Express the solution in vector components $\mathbf{x}(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$, and graph x_1 and x_2 as functions of t .
(Recall the solution in the IVP of the previous Example: $x_1(t) = 3e^{4t} - e^{-2t}$ and $x_2(t) = 3e^{4t} + e^{-2t}$.)
- ▶ Case (ii): Express the solution as a vector-valued function,
$$\mathbf{x}(t) = c_1 \mathbf{v}_1 e^{\lambda_1 t} + c_2 \mathbf{v}_2 e^{\lambda_2 t},$$
and plot the vector $\mathbf{x}(t)$ for different values of t .
- ▶ Case (ii) is called a *phase portrait*.

Phase portraits for 2×2 systems.

Example

Plot the phase portrait of several linear combinations of the fundamental solutions found above,

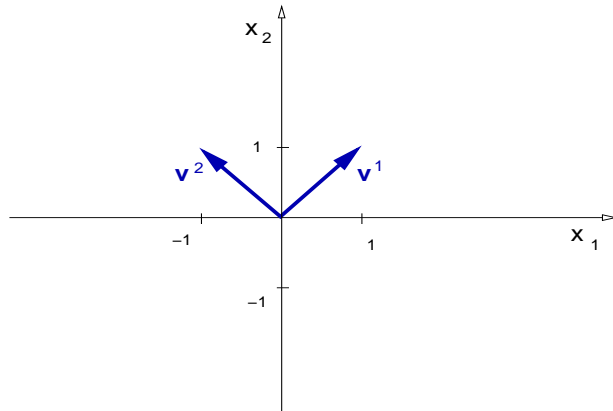
$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

Solution:

We start plotting the vectors

$$\mathbf{v}^1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix},$$

$$\mathbf{v}^2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}.$$



Phase portraits for 2×2 systems.

Example

Plot the phase portrait of several linear combinations of the fundamental solutions found above,

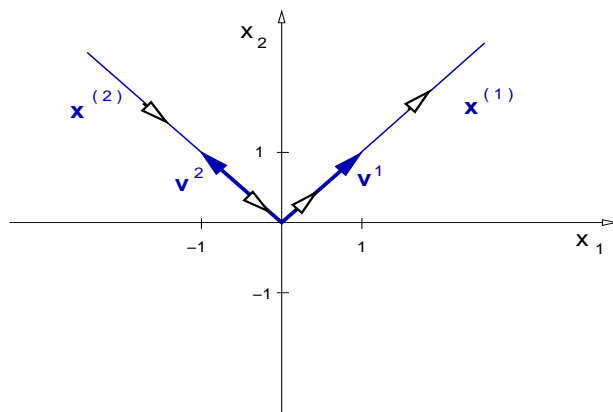
$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

Solution:

We now plot the functions

$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t},$$

$$\mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$



Phase portraits for 2×2 systems.

Example

Plot the phase portrait of several linear combinations of the fundamental solutions found above,

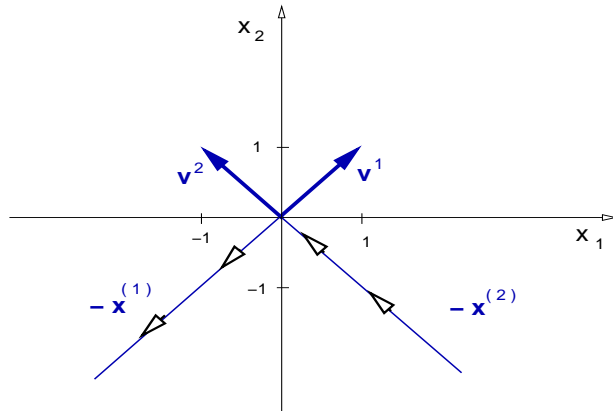
$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

Solution:

We now plot the functions

$$-\mathbf{x}^{(1)} = -\begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t},$$

$$-\mathbf{x}^{(2)} = -\begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$



Phase portraits for 2×2 systems.

Example

Plot the phase portrait of several linear combinations of the fundamental solutions found above,

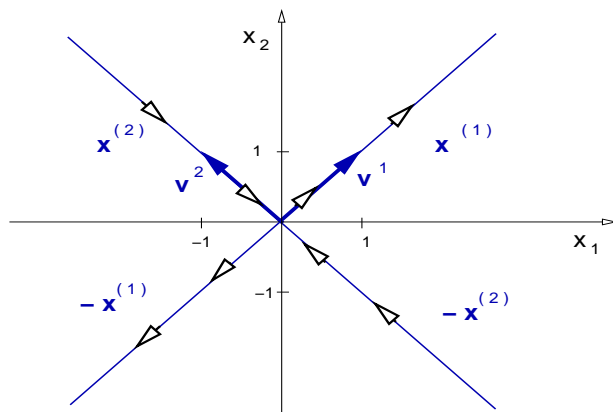
$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

Solution:

We now plot the four functions

$$\mathbf{x}^{(1)}, \quad -\mathbf{x}^{(1)},$$

$$\mathbf{x}^{(2)}, \quad -\mathbf{x}^{(2)}.$$



Phase portraits for 2×2 systems.

Example

Plot the phase portrait of several linear combinations of the fundamental solutions found above,

$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

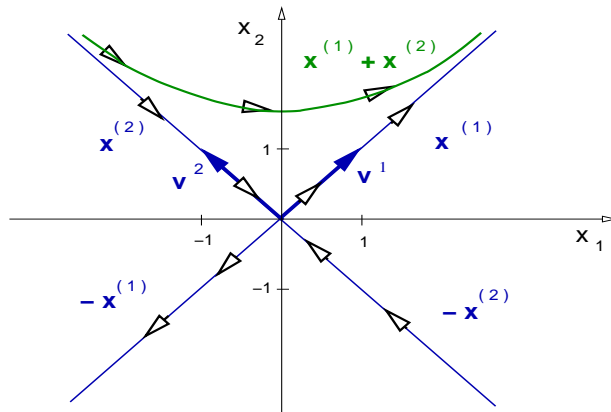
Solution:

We now plot the four functions

$$\mathbf{x}^{(1)}, \quad -\mathbf{x}^{(1)}, \quad \mathbf{x}^{(2)}, \quad -\mathbf{x}^{(2)},$$

and $\mathbf{x}^{(1)} + \mathbf{x}^{(2)}$,

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t} + \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$



Phase portraits for 2×2 systems.

Example

Plot the phase portrait of several linear combinations of the fundamental solutions found above,

$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

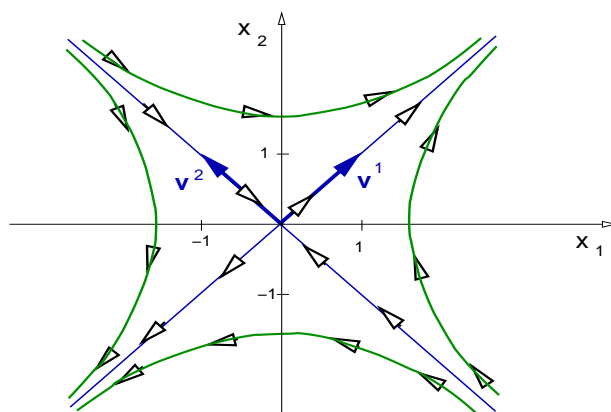
Solution:

We now plot the eight functions

$$\mathbf{x}^{(1)}, \quad -\mathbf{x}^{(1)}, \quad \mathbf{x}^{(2)}, \quad -\mathbf{x}^{(2)},$$

$$\mathbf{x}^{(1)} + \mathbf{x}^{(2)}, \quad -\mathbf{x}^{(1)} + \mathbf{x}^{(2)},$$

$$\mathbf{x}^{(1)} - \mathbf{x}^{(2)}, \quad -\mathbf{x}^{(1)} - \mathbf{x}^{(2)}.$$



Phase portraits for 2×2 systems.

Problem:

Case (a): Consider a 2×2 matrix A having two different, real eigenvalues $\lambda_1 \neq \lambda_2$, so A has two non-proportional eigenvectors $\mathbf{v}_1, \mathbf{v}_2$ (eigen-directions).

Given a solution $\mathbf{x}(t) = c_1 \mathbf{v}_1 e^{\lambda_1 t} + c_2 \mathbf{v}_2 e^{\lambda_2 t}$, to $\mathbf{x}'(t) = A\mathbf{x}(t)$, plot different solution vectors $\mathbf{x}(t)$ on the plane as function of t for different choices of the constants c_1 and c_2 .

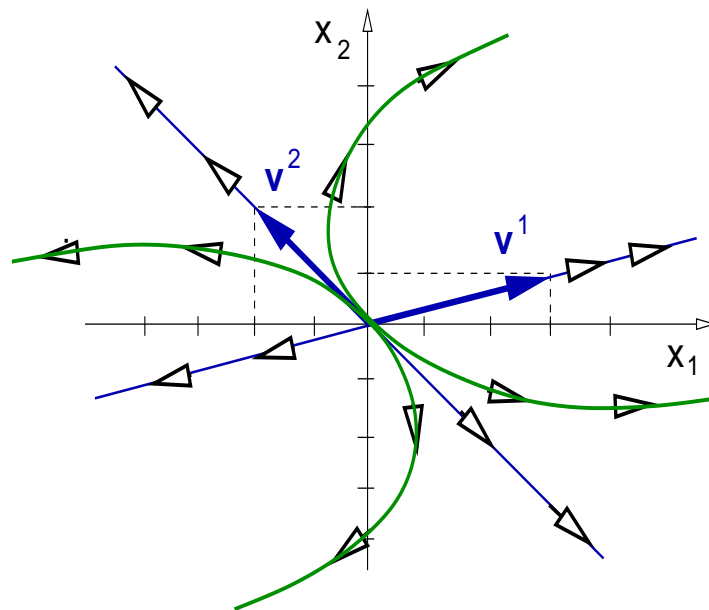
The plots are different depending on the eigenvalues signs.

We have the following three sub-cases:

- (i) $0 < \lambda_2 < \lambda_1$, both positive;
- (ii) $\lambda_2 < 0 < \lambda_1$, one positive the other negative;
- (iii) $\lambda_2 < \lambda_1 < 0$, both negative.

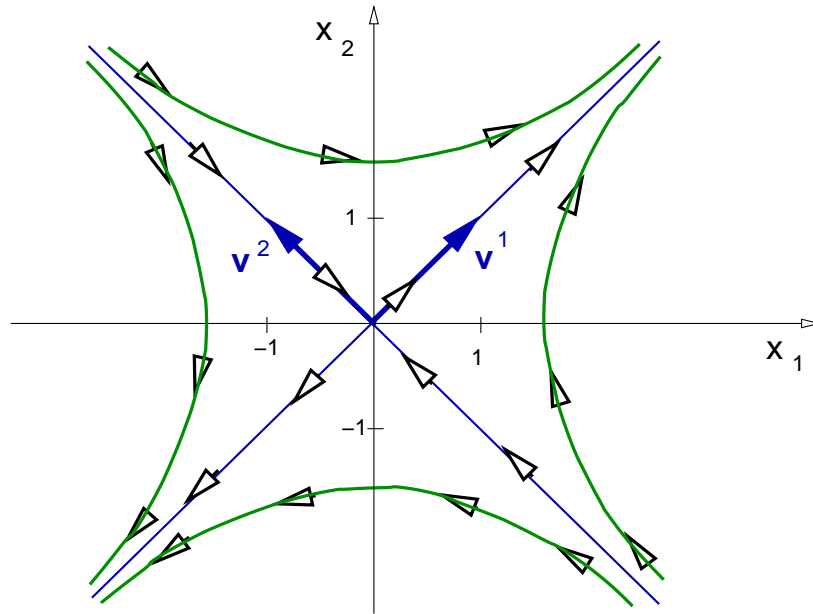
Phase portraits for 2×2 systems.

Phase portrait: Case (a), two different, real eigenvalues $\lambda_1 \neq \lambda_2$, sub-case $0 < \lambda_2 < \lambda_1$, both eigenvalue positive.



Phase portraits for 2×2 systems.

Phase portrait: Case (a), two different, real eigenvalues $\lambda_1 \neq \lambda_2$, sub-case $\lambda_2 < 0 < \lambda_1$, one eigenvalue positive the other negative.



Phase portraits for 2×2 systems.

Phase portrait: Case (a), two different, real eigenvalues $\lambda_1 \neq \lambda_2$, sub-case $\lambda_2 < \lambda_1 < 0$, both eigenvalues negative.

