

## Real, distinct eigenvalues (Sect. 5.7)

- ▶ Review:  $n \times n$  linear differential systems.
- ▶ The case of diagonalizable matrices.
- ▶ Examples:  $2 \times 2$  linear systems.
- ▶ Classification of  $2 \times 2$  diagonalizable systems.
- ▶ Phase portraits for  $2 \times 2$  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.

## Real, distinct eigenvalues (Sect. 5.7)

- ▶ Review:  $n \times n$  linear differential systems.
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- ▶ **Examples:  $2 \times 2$  linear systems.**
- ▶ Classification of  $2 \times 2$  diagonalizable systems.
- ▶ Phase portraits for  $2 \times 2$  systems.

## Examples: $2 \times 2$ 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: $2 \times 2$ 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: $2 \times 2$ 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  $\mathbf{x}(t) = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t} + \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}$ .  $\triangleleft$

## Real, distinct eigenvalues (Sect. 5.7)

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- ▶ Phase portraits for  $2 \times 2$  systems.

## Classification of $2 \times 2$ diagonalizable systems.

### Remark:

Diagonalizable  $2 \times 2$  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  $2 \times 2$  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 $2 \times 2$ systems.

Remark:

- ▶ There are two main types of graphs for solutions of  $2 \times 2$  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 $2 \times 2$ systems.

### Example

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

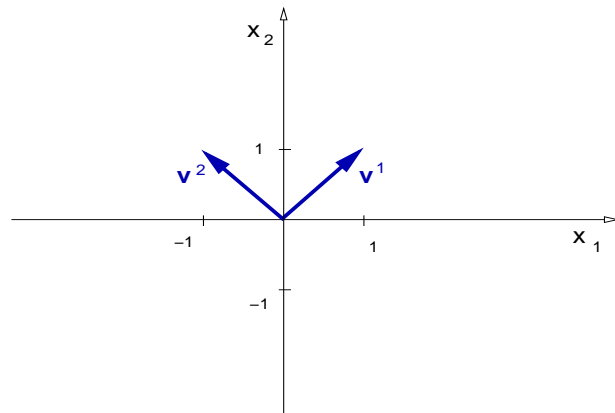
$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

### 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 $2 \times 2$ systems.

### Example

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

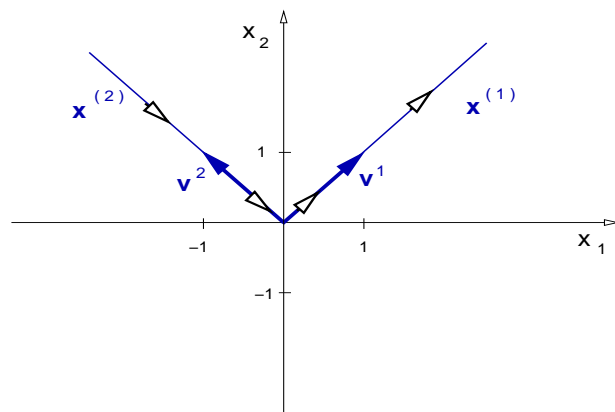
$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

### 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 $2 \times 2$ systems.

### Example

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

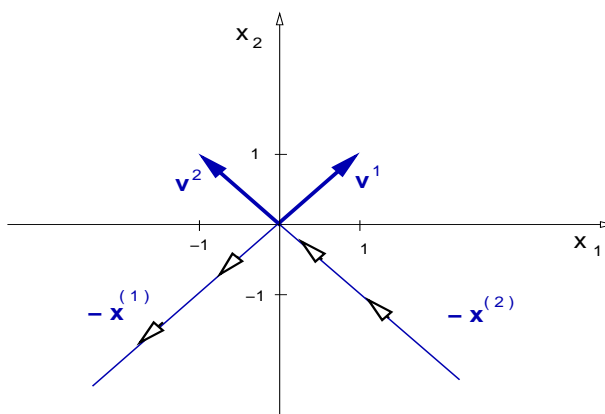
$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

### 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 $2 \times 2$ systems.

### Example

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

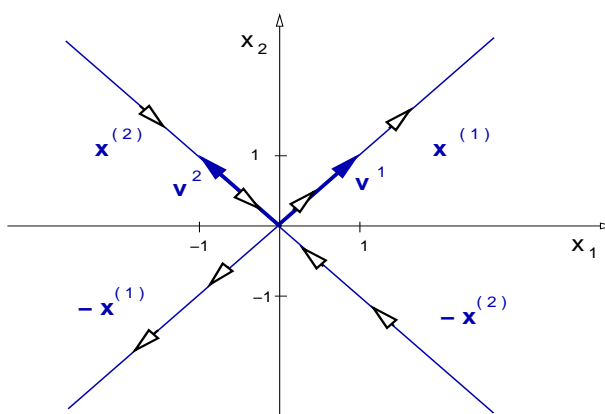
$$\mathbf{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t}, \quad \mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}.$$

### Solution:

We now plot the four functions

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

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



## Phase portraits for $2 \times 2$ systems.

### Example

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

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

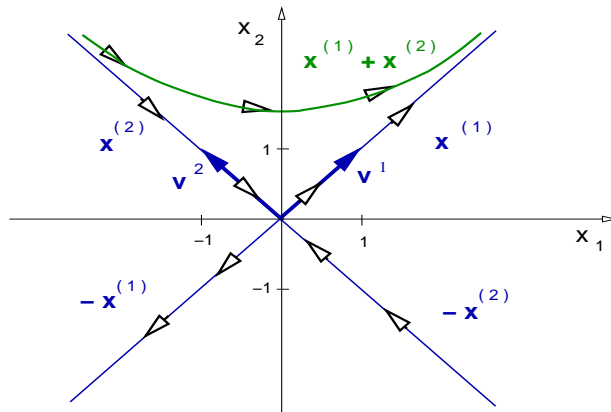
### Solution:

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 $2 \times 2$ systems.

### Example

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

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

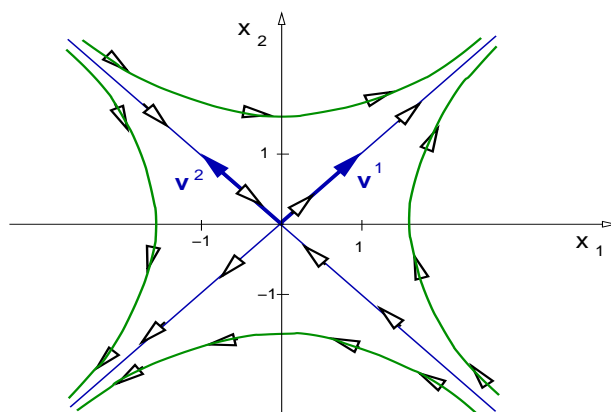
### Solution:

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 $2 \times 2$ systems.

**Problem:**

**Case (a):** Consider a  $2 \times 2$  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  $c_2$ .

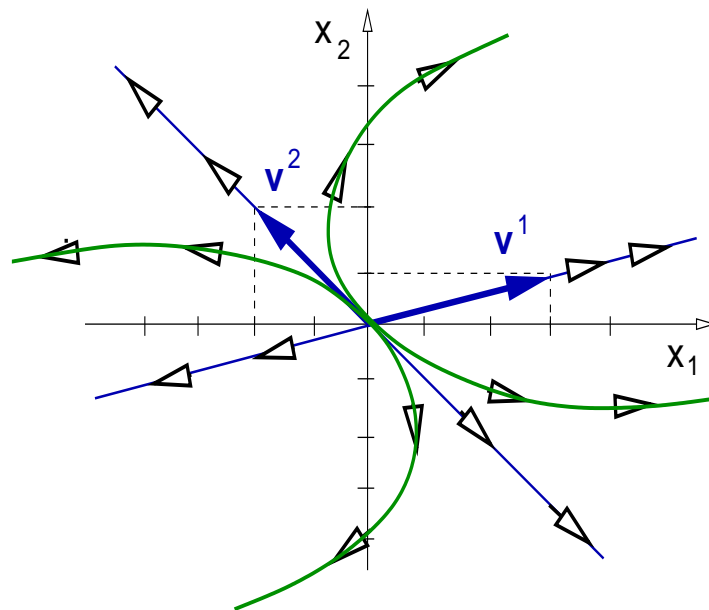
The plots are different depending on the eigenvalues signs.

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.

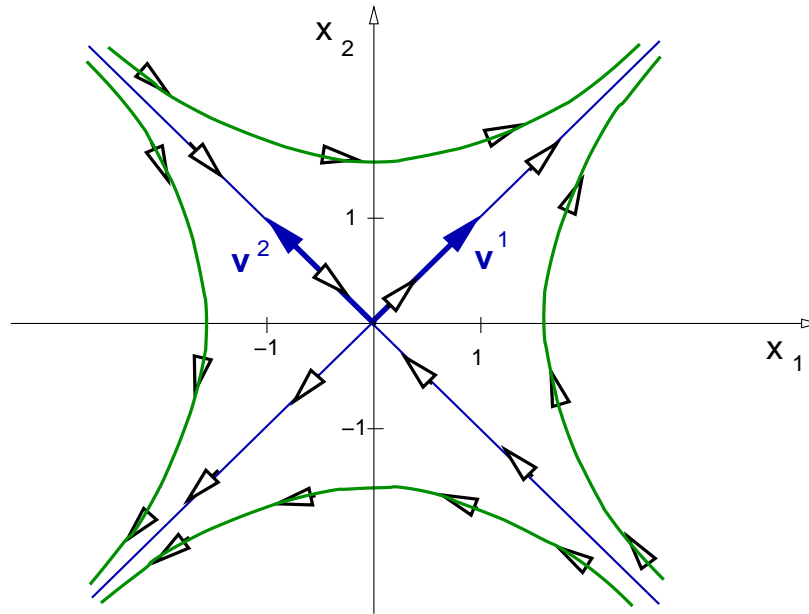
## Phase portraits for $2 \times 2$ systems.

**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 $2 \times 2$ 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 $2 \times 2$ systems.

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

