

math 235 L31

- Plan:
- \*  $n \times n$  linear differential systems
  - \* Homogeneous, constant coefficients systems
  - \* Real, non-zero, distinct eigenvalues
  - \*  $2 \times 2$  systems: phase portraits.

(7.5)

\* Review:  $n \times n$  linear diff. systems

- Given an  $n \times n$   $A(t)$ ,  $n$ -vector  $b(t)$ , find  $x(t)$  sol. of

$$\boxed{x'(t) = A(t)x(t) + b(t)} \quad (1)$$

- The system (1) is **homogeneous** iff  $b = 0$ , that is,

$$\boxed{x'(t) = A(t)x(t).} \quad (2)$$

- The system (1) has **constant coefficients** iff matrix  $A$  does not depend on  $t$ , that is,

$$\boxed{x'(t) = A x(t) + b(t)} \quad (3)$$

- We study homogeneous, constant coeff. systems, that is,

$$\boxed{x'(t) = A x(t)} \quad (4)$$

- Main result: (previous class)

Given continuous functions  $A, b$  on  $[0, T] \subset \mathbb{R}$ , and  $\underline{x}_0$ , there exists a unique function  $\underline{x}$  sol. of IVP

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

$$\underline{x}(0) = \underline{x}_0$$

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

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

4

\* Eigen-pairs provide sols. to diff. sys..

Propos.

If  $\lambda, \underline{v}$  are eigenvalue and eigenvector of an  $n \times n$  matrix  $A$ , then the  $n$ -vector-valued function

$$\underline{x}(t) = \underline{v} e^{\lambda t}, \quad t \in \mathbb{R}$$

is a solution of the  $n \times n$  linear homogeneous, constant coeff. diff. system

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

Proof:

$$\begin{aligned} \underline{x}'(t) &= \underline{v} (e^{\lambda t})' \\ &= \lambda \underline{v} e^{\lambda t} \end{aligned}$$

$$\underline{x}'(t) = \lambda \underline{x}(t)$$

$$\begin{aligned} A \underline{x}(t) &= A \underline{v} e^{\lambda t} \\ &= \lambda \underline{v} e^{\lambda t} \end{aligned}$$

$$A \underline{x}(t) = \lambda \underline{x}(t)$$

so:  $\underline{x}'(t) = A \underline{x}(t)$  □

## Remarks:

- To find all solutions of  $\dot{x}(t) = A x(t)$  is important to find all eigenvalues - eigenvectors of matrix  $A$ .
- **Not** every  $n \times n$  matrix  $A$  has a l.i. set of  $n$  eigenvectors.
- We will study systems  $\dot{x}(t) = A x(t)$  where  $n \times n$  matrix  $A$  satisfies:

(a)  $\det(A) \neq 0$ . (invertible).

(b)  $A$ : real-valued matrix.

Remark: - From Linear Algebra we know that such  $n \times n$  matrix must satisfy:

$n \times n$  A

$n$  distinct eigenvalues

Repeated eigenvalues.

l.i. set of  $n$  eigenvectors

l.i. set of  $n$  eigenvectors

l.i. set of  $m < n$  eigenvectors

Real case

Complex case

(7.8)

(7.5)

(7.6)

7

\*  $n \times n$  A having a d.i. set of n eigenvectors  
is important because:

Thm:

If  $n \times n$  matrix  $A$  has a d.i. set of  $n$  eigenvectors  $\{ \underline{v}^{(1)}, \dots, \underline{v}^{(n)} \}$  with corresponding eigenvalues  $\{ \lambda_1, \dots, \lambda_n \}$ ,

Then, a fundamental set of solutions to the diff. system

$$\underline{x}'(t) = A \underline{x}(t) \quad (4)$$

is given by

$$\{ \underline{x}^{(1)}(t) = \underline{v}^{(1)} e^{\lambda_1 t}, \dots, \underline{x}^{(n)}(t) = \underline{v}^{(n)} e^{\lambda_n t} \}$$

(5)

Proof:  $\underline{x}^{(i)} = \underline{v}^{(i)} e^{\lambda_i t}$   $i=1, \dots, n$  is sol. of (4).

Since:

$$\underline{x}^{(i)'}(t) = \lambda_i \underline{v}^{(i)} e^{\lambda_i t} = A \underline{v}^{(i)} e^{\lambda_i t} = A \underline{x}^{(i)}(t)$$

The set in (5) is a fundamental set,  
since

$$X(t) = [ \underline{x}^{(1)}(t), \dots, \underline{x}^{(n)}(t) ]$$

has non-zero Wronskian

$$W(t) = \det(X(t)) \neq 0.$$

For the proof, recall Abel's Theorem in  
this case:

$$\left[ \begin{array}{l} \text{If } W(t_0) \neq 0, \quad t_0 \in \mathbb{R}, \\ \text{then } W(t) \neq 0 \quad \text{for all } t \in \mathbb{R}. \end{array} \right]$$

In our case:

$$\begin{aligned} W(0) &= \det(X(0)) \\ &= \det([ \underline{v}^{(1)}, \dots, \underline{v}^{(n)} ]) \neq 0 \end{aligned}$$

Since

$$\{ \underline{v}^{(1)}, \dots, \underline{v}^{(n)} \} \text{ l.i.}$$

□



\* Class Today : We study the case:

-  $\left[ \begin{array}{l} n \times n \text{ } A \text{ has real, non-zero,} \\ n \text{ distinct eigenvalues } \lambda_1, \dots, \lambda_n. \end{array} \right]$

-  $\left[ \begin{array}{l} \text{From Linear Algebra: } A \text{ has a} \\ \text{d.i. set of } n \text{ eigenvectors} \\ \{ \underline{v}^{(1)}, \dots, \underline{v}^{(n)} \}. \end{array} \right]$

- From Thm above:  $\underline{x}'(t) = A \underline{x}(t)$  has a fundamental set given by

$$\left\{ \underline{x}^{(i)}(t) = \underline{v}^{(i)} e^{\lambda_i t} \right\}$$

$i = 1, \dots, n.$

\* Example : [ Find the general sol. to  $x'(t) = A x(t)$ ,  $A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$  ]

Sol:

- Find all eigen values of A.

$$P(\lambda) = \begin{vmatrix} 1-\lambda & 3 \\ 3 & 1-\lambda \end{vmatrix} = (1-\lambda)^2 - 9$$

$$(\lambda-1)^2 - 9 = 0 \Rightarrow \lambda_{\pm} = 1 \pm 3$$

$$\lambda_{\pm} = 1 \pm 3$$

$$\boxed{\lambda_+ = 4}, \quad \boxed{\lambda_- = -2}$$

A 2x2 has two distinct eigenvalues.

so A has a l.i. set of two eigenvectors.

- Find all eigenvectors of A.

$$\boxed{\lambda_+ = 4}$$

$$A - 4I = \begin{bmatrix} -3 & 3 \\ 3 & -3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix} \Rightarrow v_1 = v_2 \Rightarrow \boxed{v^{(+)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}}$$

$$\lambda_- = -2$$

$$A + 2I = \begin{bmatrix} 3 & 3 \\ 3 & 3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \Rightarrow v_1 = -v_2 \Rightarrow \underline{v}^{(-)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

We verify that  $\{ \underline{v}^{(+)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \underline{v}^{(-)} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \}$  is l.i.

So, a fundamental sol. of  $\underline{x}' = A \underline{x}$  is

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

So the general sol. of  $\underline{x}' = A \underline{x}$  is

$$\underline{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}$$

in components:  $\underline{x}(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$

$$x_1(t) = c_1 e^{4t} - c_2 e^{-2t}$$

$$x_2(t) = c_1 e^{4t} + c_2 e^{-2t}$$

$$\lambda_{(-)} < 0 < \lambda_{(+)}$$

Example : Find the sol. to the IVP

$$x'(t) = A x(t) \quad , \quad A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$$

$$x(0) = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$

Sol:

(1) Find the general sol. of  $x' = Ax$ .

$$x(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t} \quad (6)$$

(2) Impose the initial cond. on (6).

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

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

$$X(0) = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \Rightarrow [X(0)]^{-1} = \frac{1}{1+1} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$$

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

$$= \frac{1}{2} \begin{bmatrix} 6 \\ 2 \end{bmatrix} \Rightarrow \boxed{\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}}$$

$$\underline{X}(t) = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{4t} + \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t} \quad (7)$$

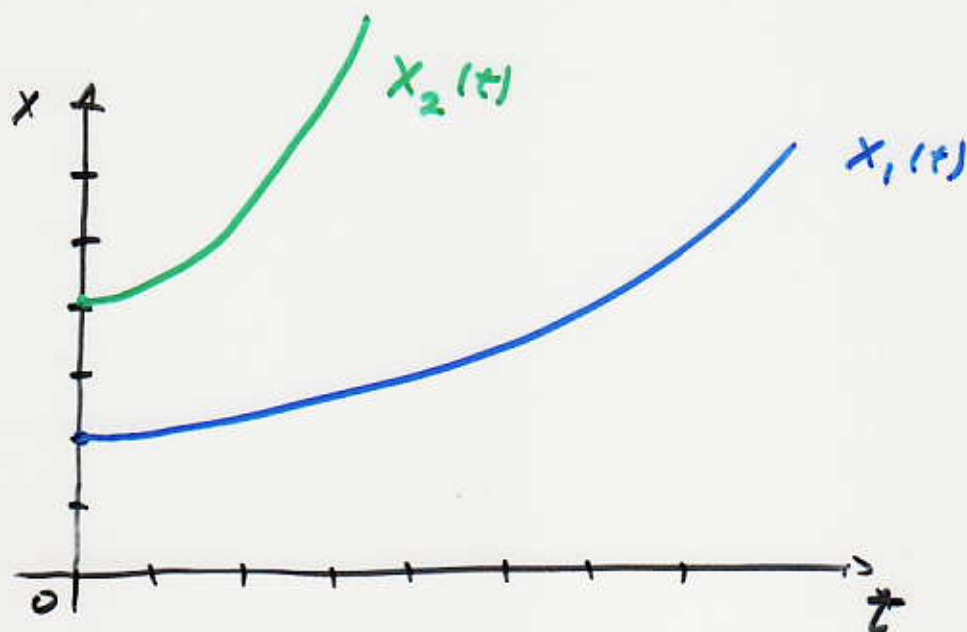
In components:  $\underline{X}(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$ ,

$$x_1(t) = 3 e^{4t} - e^{-2t} \quad (8)$$

$$x_2(t) = 3 e^{4t} + e^{-2t} \quad (9)$$

\* Remarks:

(1) We can always plot the graph of each component,  $x_1(t)$ ,  $x_2(t)$ , as function of  $t$ .



(2) We can plot the whole vector  $x(t)$  on the plane.

This is called:

|   |
|---|
| <p>phase portrait<br/>or<br/>phase diagram.</p> |
|---|

\* Phase portraits : 2x2 systems.

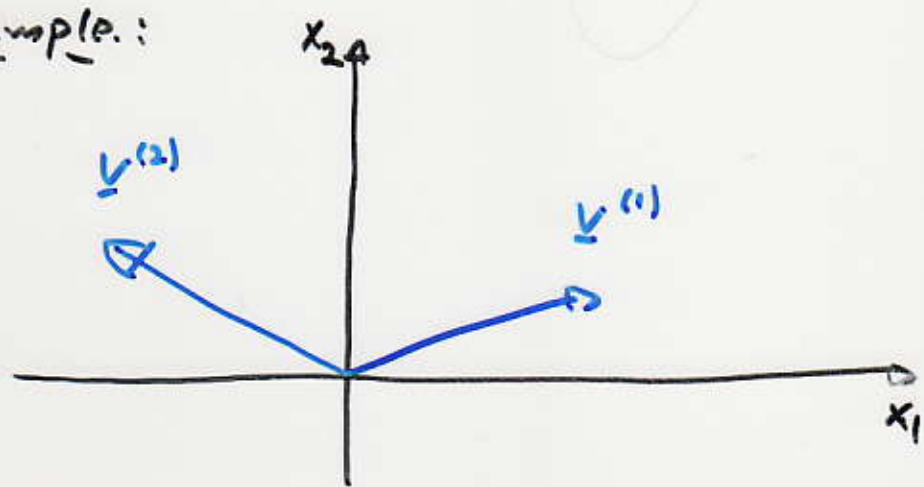
- Let  $A$  be  $2 \times 2$ , with distinct, non-zero eigenvalues  $\lambda_1 \neq \lambda_2$ , and l.i. set  $\{\underline{v}^{(1)}, \underline{v}^{(2)}\}$  of eigenvectors.
- So, the general sol. of  $\underline{x}'(t) = A \underline{x}(t)$  is:

$$\underline{x}(t) = c_1 \underline{v}^{(1)} e^{\lambda_1 t} + c_2 \underline{v}^{(2)} e^{\lambda_2 t}$$

- Phase portrait: plot  $\underline{x}(t)$  as function of  $t$  for all possible choices of  $c_1, c_2$ .

First: Plot  $\underline{v}^{(1)}, \underline{v}^{(2)}$ .

For example:



(No time in this plot.)

Second: Plot the simplest solutions:

|  |
|--|
| $\underline{x}^{(1)}(t) = \underline{v}^{(1)} e^{\lambda_1 t}$ |
| $\underline{x}^{(2)}(t) = \underline{v}^{(2)} e^{\lambda_2 t}$ |

$(c_1=1, c_2=0)$

$(c_1=0, c_2=1)$

Third: Plot the general sol.

|   |
|---|
| $\underline{x}(t) = c_1 \underline{x}^{(1)}(t) + c_2 \underline{x}^{(2)}(t).$ |
|---|

-o-

Back to step two: We need to study different cases:

- (1) 

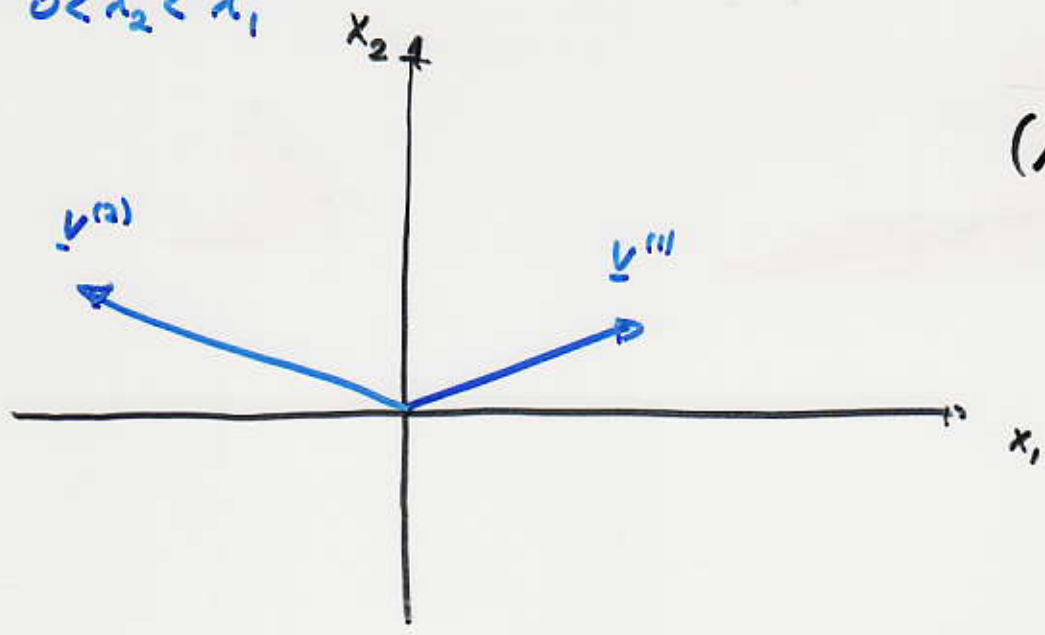
|                             |
|-----------------------------|
| $0 < \lambda_2 < \lambda_1$ |
|-----------------------------|
- (2) 

|                             |
|-----------------------------|
| $\lambda_2 < 0 < \lambda_1$ |
|-----------------------------|
- (3) 

|                             |
|-----------------------------|
| $\lambda_2 < \lambda_1 < 0$ |
|-----------------------------|

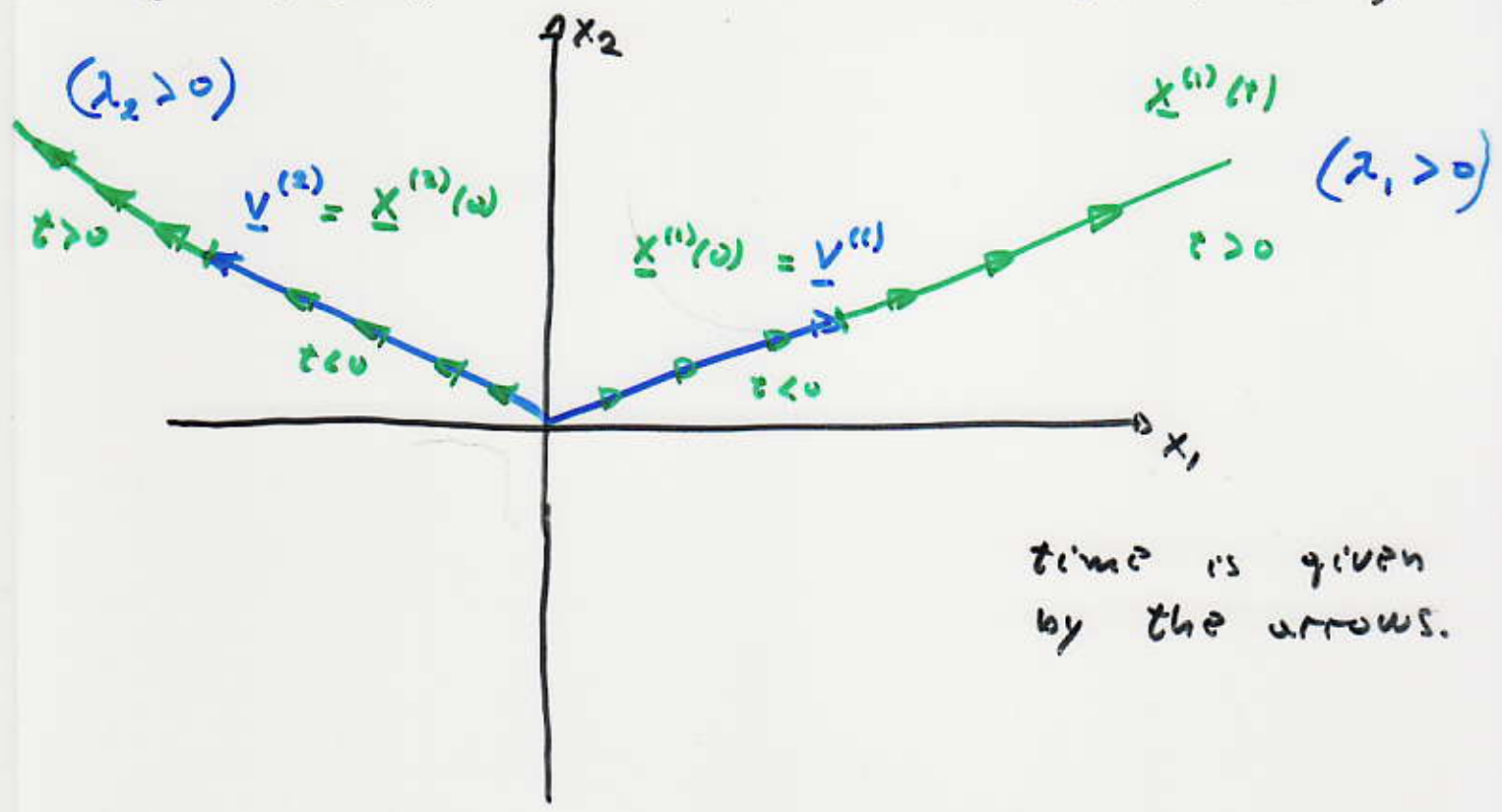


(1)  $0 < \lambda_2 < \lambda_1$



$\underline{x}^{(1)}(t) = v^{(1)} e^{\lambda_1 t}$  ,  $(c_1=1, c_2=0)$

$\underline{x}^{(2)}(t) = v^{(2)} e^{\lambda_2 t}$  ,  $(c_1=0, c_2=1)$



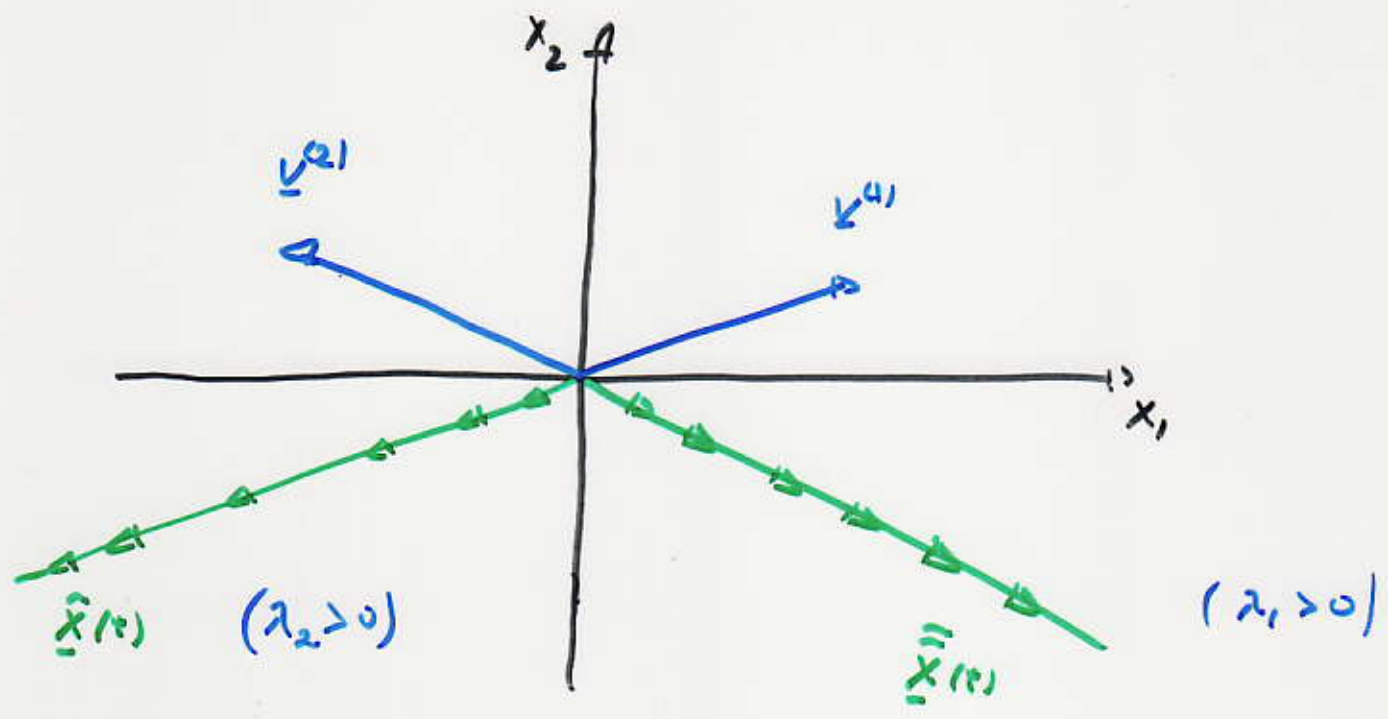
time is given by the arrows.

$$\tilde{x}^{(1)}(t) = -\underline{v}^{(1)} e^{\lambda_1 t}$$

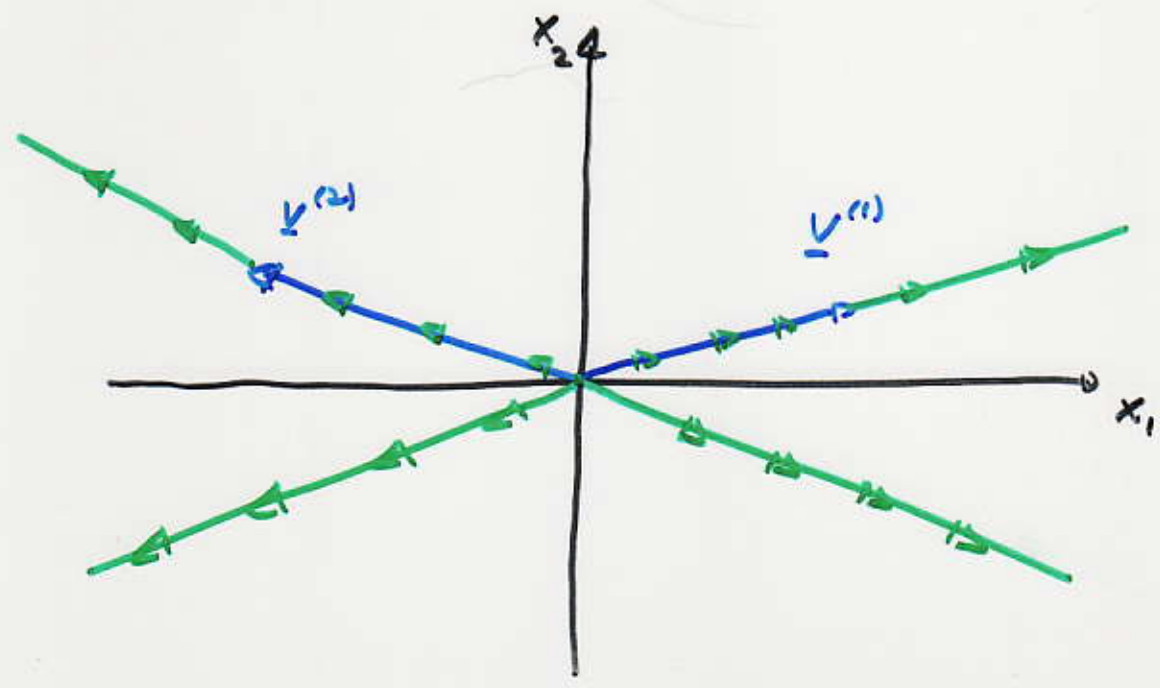
$$(c_1 = -1, c_2 = 0)$$

$$\tilde{x}^{(2)}(t) = -\underline{v}^{(2)} e^{\lambda_2 t}$$

$$(c_1 = 0, c_2 = -1)$$

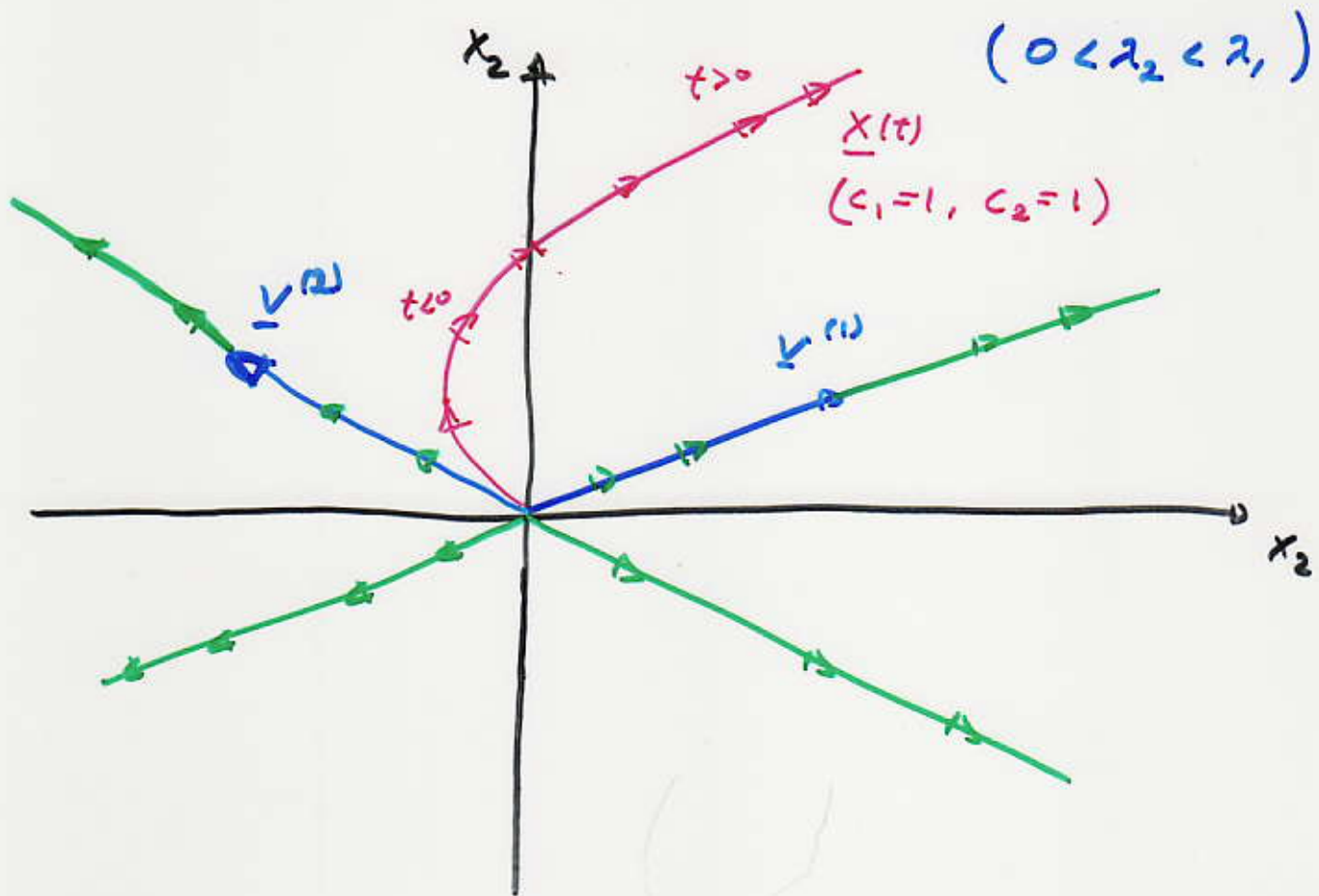


So, four different solutions are:

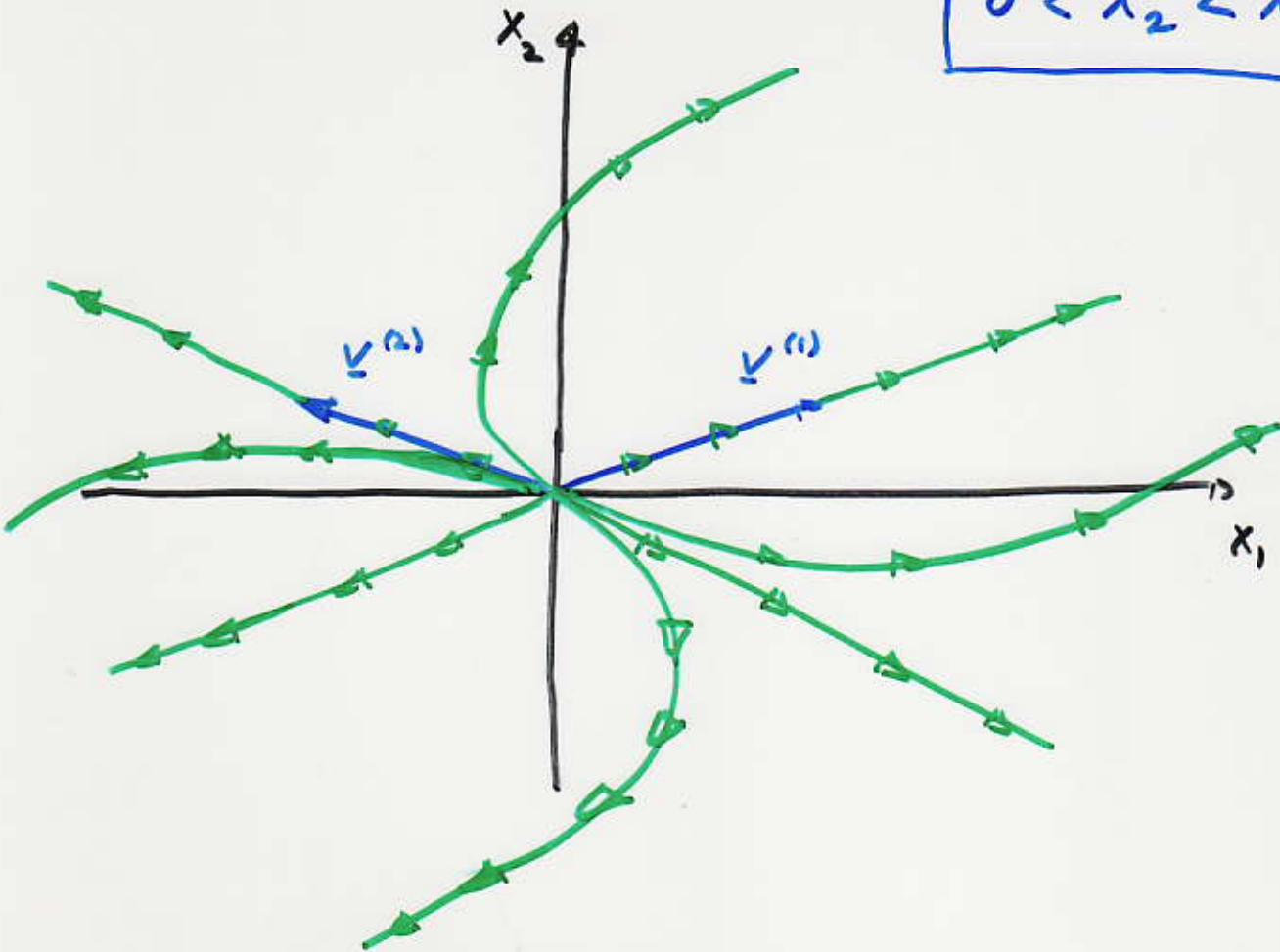


one more solution:  $c_1 = 1, c_2 = 1$

$$\underline{x}(t) = \underline{v}^{(1)} e^{\lambda_1 t} + \underline{v}^{(2)} e^{\lambda_2 t}$$



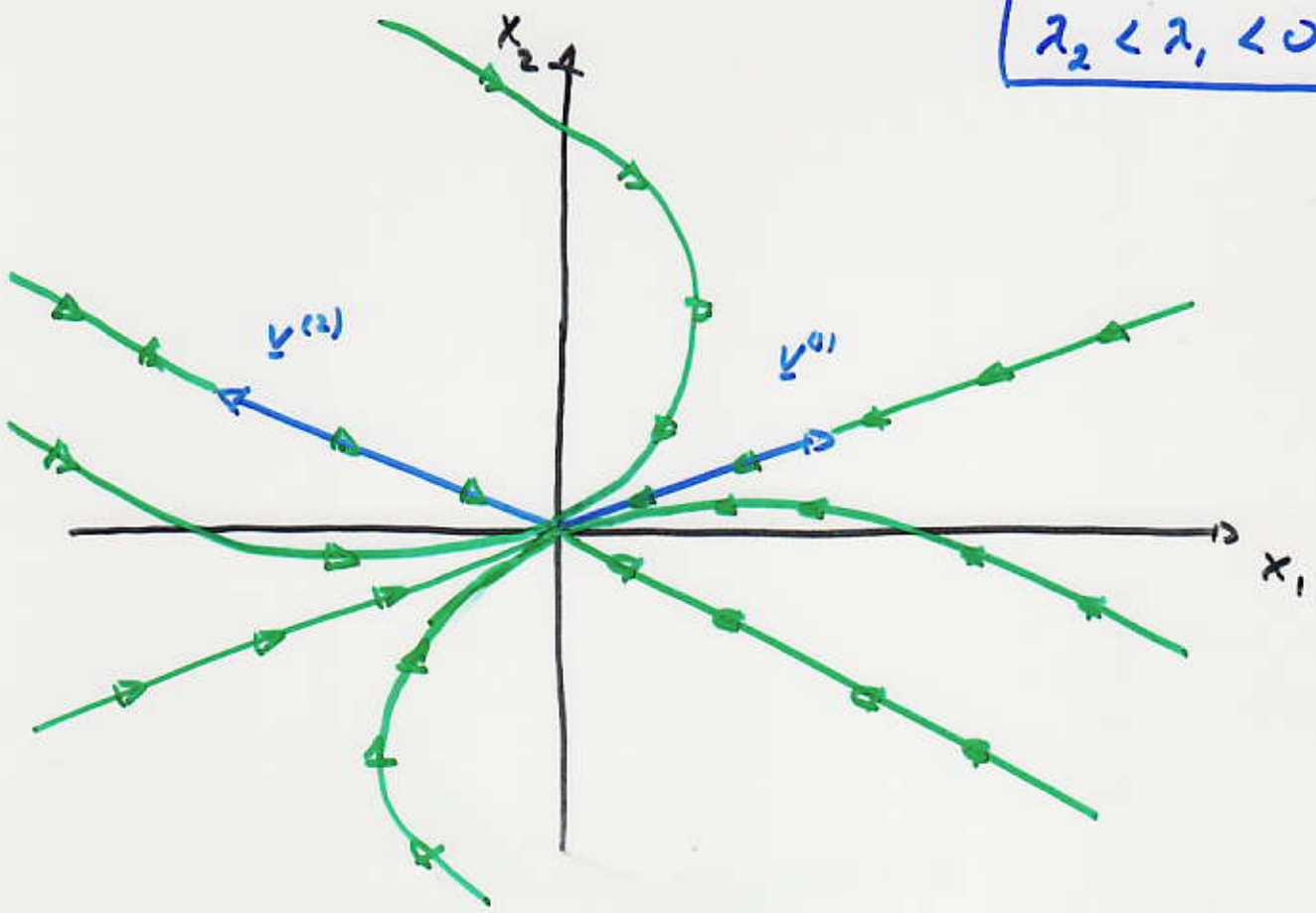
$$0 < \lambda_2 < \lambda_1$$



- eight solutions plotted.
- Arrows indicate time evolution (t)

$\underline{x} = \underline{0}$  : unstable solution

$\lambda_2 < \lambda_1 < 0$



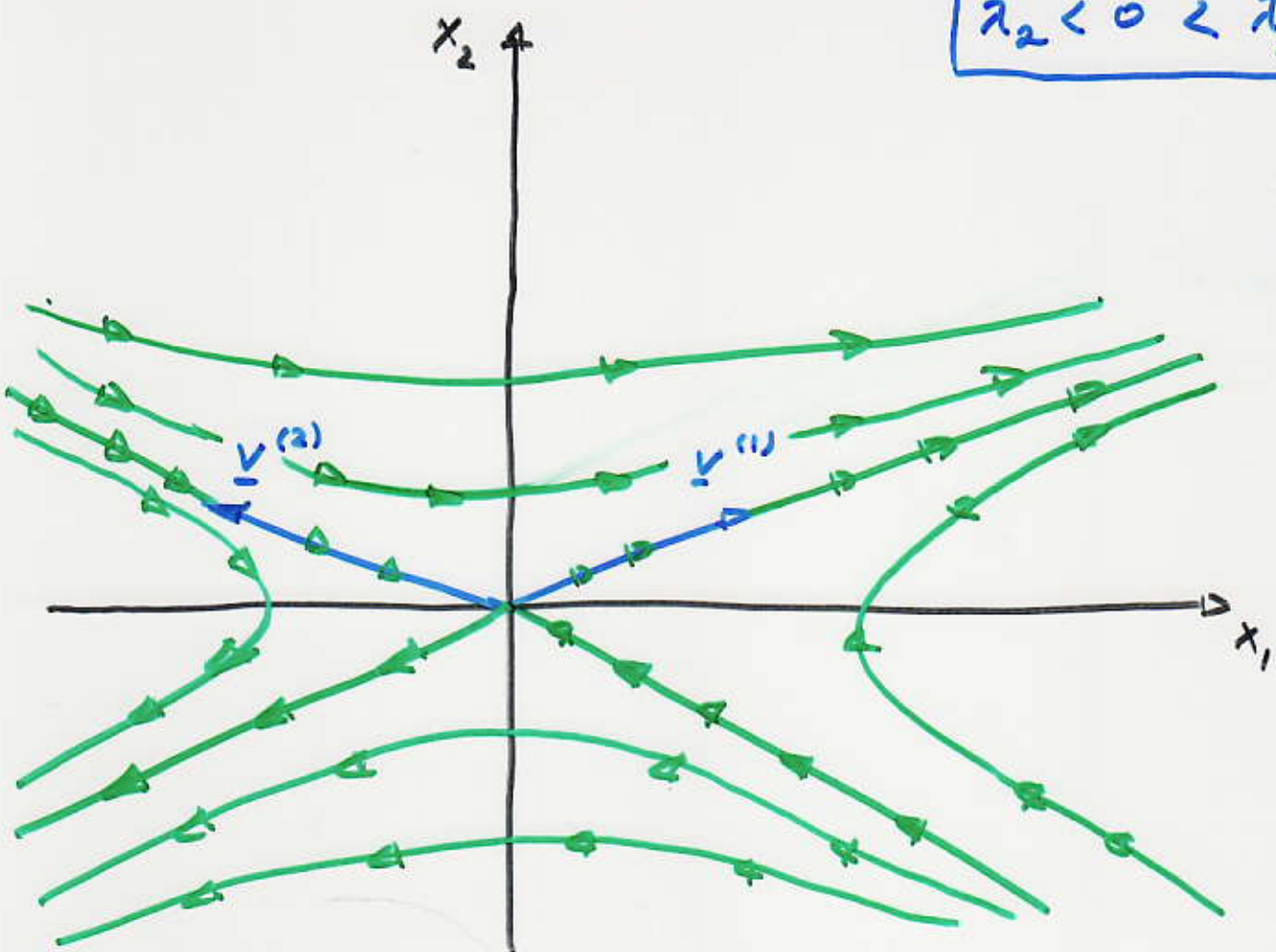
-  $\underline{x}(t) = c_1 \underline{v}^{(1)} e^{-|\lambda_1|t} + c_2 \underline{v}^{(2)} e^{-|\lambda_2|t}$

- eight solutions plotted.

- arrows indicate time evolution (t)

$\underline{x} = \underline{0}$  Stable solution

$\lambda_2 < 0 < \lambda_1$



-  $x(t) = c_1 \underline{v}^{(1)} e^{\lambda_1 t} + c_2 \underline{v}^{(2)} e^{-|\lambda_2| t}$

- ten solutions plotted
- arrows indicate time evolution (t)

$x = 0$  unstable solution

(saddle point.)