

## Real, repeated eigenvalues (Sect. 5.9)

- ▶ Review: Classification of  $2 \times 2$  diagonalizable systems.
- ▶ Repeated eigenvalue diagonalizable  $2 \times 2$  system.
- ▶ Repeated eigenvalue non-diagonalizable  $2 \times 2$  system.
- ▶ Phase portraits for  $2 \times 2$  systems.

## Review: Classification of $2 \times 2$ systems.

### Remark:

Diagonalizable  $2 \times 2$  matrices  $A$  with real coefficients are classified according to their eigenvalues.

- (a)  $\lambda_1 \neq \lambda_2$ , real-valued. Hence,  $A$  has two non-proportional eigenvectors  $\mathbf{v}_1, \mathbf{v}_2$  (eigen-directions), (Section 5.7).
- (b)  $\lambda_1 = \bar{\lambda}_2$ , complex-valued. Hence,  $A$  has two non-proportional eigenvectors  $\mathbf{v}_1 = \bar{\mathbf{v}}_2$ , (Section 5.8).
- (c-1)  $\lambda_1 = \lambda_2$  real-valued with two non-proportional eigenvectors  $\mathbf{v}_1, \mathbf{v}_2$ , (Section 5.9).

### Remark:

- (c-2)  $\lambda_1 = \lambda_2$  real-valued with only one eigen-direction. Hence,  $A$  is not diagonalizable, (Section 5.9).

## Real, repeated eigenvalues (Sect. 5.9)

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## Repeated eigenvalue diagonalizable $2 \times 2$ system.

**Remark:** For  $2 \times 2$  systems the situation is fairly simple.

### Theorem

*Every  $2 \times 2$  diagonalizable matrix  $A$  with repeated eigenvalue  $\lambda$  has the form  $A = \lambda I$ .*

**Proof:** Since  $A$  is diagonalizable, exists  $P$  invertible such that

$$A = P \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} P^{-1} = P \lambda I P^{-1} = \lambda P P^{-1} = \lambda I.$$

**Remark:** The  $\mathbf{x}$  general solution for  $\mathbf{x}' = \lambda I \mathbf{x}$  is simple □

$$\mathbf{x}(t) = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} e^{\lambda t} \Leftrightarrow \mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} e^{\lambda t} + c_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix} e^{\lambda t}.$$

**Remark:** The solution phase portraits are always straight lines passing through the origin.

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## Repeated eigenvalue non-diagonalizable $2 \times 2$ system.

### Remark:

Diagonalizable  $2 \times 2$  matrices  $A$  with real coefficients are classified according to their eigenvalues.

- (a)  $\lambda_1 \neq \lambda_2$ , real-valued. Hence,  $A$  has two non-proportional eigenvectors  $\mathbf{v}_1, \mathbf{v}_2$  (eigen-directions), (Section 5.7).
- (b)  $\lambda_1 = \bar{\lambda}_2$ , complex-valued. Hence,  $A$  has two non-proportional eigenvectors  $\mathbf{v}_1 = \bar{\mathbf{v}}_2$ , (Section 5.8).
- (c-1)  $\lambda_1 = \lambda_2$  real-valued with two non-proportional eigenvectors  $\mathbf{v}_1, \mathbf{v}_2$ , (Section 5.9).

### Remark:

- (c-2)  $\lambda_1 = \lambda_2$  real-valued with only one eigen-direction. Hence,  $A$  is not diagonalizable, (Section 5.9). Next Class.

## Repeated eigenvalue non-diagonalizable $2 \times 2$ system.

### Example

Show that matrix  $B = \frac{1}{2} \begin{bmatrix} 3 & 1 \\ -1 & 5 \end{bmatrix}$  is not diagonalizable.

**Solution:** We need to show that all eigenvectors of matrix  $B$  are proportional to each other. We start computing the eigenvalues,

$$p(\lambda) = \det(B - \lambda I) = \begin{vmatrix} \frac{3}{2} - \lambda & \frac{1}{2} \\ -\frac{1}{2} & \frac{5}{2} - \lambda \end{vmatrix} = \left(\frac{3}{2} - \lambda\right)\left(\frac{5}{2} - \lambda\right) + \frac{1}{4}$$

$$p(\lambda) = \lambda^2 - 4\lambda + 4 = 0 \quad \Rightarrow \quad \lambda_{\pm} = 2.$$

We now compute the corresponding eigenvectors,

$$(B - 2I) = \begin{bmatrix} \frac{3}{2} - 2 & \frac{1}{2} \\ -\frac{1}{2} & \frac{5}{2} - 2 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} & \frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix}.$$

Hence all eigenvectors are proportional to  $\mathbf{v} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$   $\triangleleft$

## Repeated eigenvalue non-diagonalizable $2 \times 2$ system.

### Theorem (Repeated eigenvalue)

If  $\lambda$  is an eigenvalue of an  $n \times n$  matrix  $A$  having algebraic multiplicity  $r = 2$  and only one associated eigen-direction, then the differential equation

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

has a linearly independent set of solutions given by

$$\{\mathbf{x}^{(1)}(t) = \mathbf{v} e^{\lambda t}, \quad \mathbf{x}^{(2)}(t) = (\mathbf{v} t + \mathbf{w}) e^{\lambda t}\}.$$

where the vector  $\mathbf{w}$  is solution of

$$(A - \lambda I)\mathbf{w} = \mathbf{v}$$

which always has a solution  $\mathbf{w}$ .

## Repeated eigenvalue non-diagonalizable $2 \times 2$ system.

Recall: The case of a single second order equation

$$y'' + a_1 y' + a_0 y = 0$$

with characteristic polynomial

$$p(r) = r^2 + a_1 r + a_0 = (r - r_1)^2.$$

In this case a fundamental set of solutions is

$$\{y_1(t) = e^{r_1 t}, \quad y_2(t) = t e^{r_1 t}\}.$$

This is not the case with systems of first order linear equations,

$$\{\mathbf{x}^{(1)}(t) = \mathbf{v} e^{\lambda t}, \quad \mathbf{x}^{(2)}(t) = (\mathbf{v} t + \mathbf{w}) e^{\lambda t}\}.$$

In general,  $\mathbf{w} \neq \mathbf{0}$ .

## Repeated eigenvalue non-diagonalizable $2 \times 2$ system.

### Example

Find fundamental solutions of  $\mathbf{x}' = A\mathbf{x}$ , with  $A = \frac{1}{4} \begin{bmatrix} -6 & 4 \\ -1 & -2 \end{bmatrix}$ .

**Solution:** Find the eigenvalues of  $A$ . Its characteristic polynomial is

$$p(\lambda) = \begin{vmatrix} \left(-\frac{3}{2} - \lambda\right) & 1 \\ -\frac{1}{4} & \left(-\frac{1}{2} - \lambda\right) \end{vmatrix} = \left(\lambda + \frac{3}{2}\right)\left(\lambda + \frac{1}{2}\right) + \frac{1}{4}.$$

So  $p(\lambda) = \lambda^2 + 2\lambda + 1 = (\lambda + 1)^2$ . The roots and multiplicity are

$$\lambda = -1, \quad r = 2.$$

The corresponding eigenvectors are the solutions of  $(A + I)\mathbf{v} = \mathbf{0}$ ,

$$\begin{bmatrix} \left(-\frac{3}{2} + 1\right) & 1 \\ -\frac{1}{4} & \left(-\frac{1}{2} + 1\right) \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} & 1 \\ -\frac{1}{4} & \frac{1}{2} \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -2 \\ 1 & -2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix}$$

## Repeated eigenvalue non-diagonalizable $2 \times 2$ system.

### Example

Find fundamental solutions of  $\mathbf{x}' = A\mathbf{x}$ , with  $A = \frac{1}{4} \begin{bmatrix} -6 & 4 \\ -1 & -2 \end{bmatrix}$ .

**Solution:** Recall:  $\lambda = -1$ , with  $r = 2$ , and  $(A + I) \rightarrow \begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix}$ .

The eigenvector components satisfy:  $v_1 = 2v_2$ . We obtain,

$$\lambda = -1, \quad \mathbf{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix} v_2.$$

We conclude that this eigenvalue has only one eigen-direction.

**Matrix  $A$  is not diagonalizable.**

Theorem above says we need to find  $\mathbf{w}$  solution of  $(A + I)\mathbf{w} = \mathbf{v}$ .

$$\left[ \begin{array}{cc|c} -\frac{1}{2} & 1 & 2 \\ -\frac{1}{4} & \frac{1}{2} & 1 \end{array} \right] \rightarrow \left[ \begin{array}{cc|c} 1 & -2 & -4 \\ 1 & -2 & -4 \end{array} \right] \rightarrow \left[ \begin{array}{cc|c} 1 & -2 & -4 \\ 0 & 0 & 0 \end{array} \right]$$

## Repeated eigenvalue non-diagonalizable $2 \times 2$ system.

### Example

Find fundamental solutions of  $\mathbf{x}' = A\mathbf{x}$ , with  $A = \frac{1}{4} \begin{bmatrix} -6 & 4 \\ -1 & -2 \end{bmatrix}$ .

**Solution:** Recall that:

$$\lambda = -1, \quad \mathbf{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix} v_2, \quad \text{and} \quad (A + I)\mathbf{w} = \mathbf{v} \Rightarrow \left[ \begin{array}{cc|c} 1 & -2 & -4 \\ 0 & 0 & 0 \end{array} \right].$$

We obtain  $w_1 = 2w_2 - 4$ . That is,  $\mathbf{w} = \begin{bmatrix} 2 \\ 1 \end{bmatrix} w_2 + \begin{bmatrix} -4 \\ 0 \end{bmatrix}$ .

Given a solution  $\mathbf{w}$ , then  $c\mathbf{v} + \mathbf{w}$  is also a solution,  $c \in \mathbb{R}$ .

We choose the simplest solution,  $\mathbf{w} = \begin{bmatrix} -4 \\ 0 \end{bmatrix}$ . We conclude,

$$\mathbf{x}^{(1)}(t) = \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{-t}, \quad \mathbf{x}^{(2)}(t) = \left( \begin{bmatrix} 2 \\ 1 \end{bmatrix} t + \begin{bmatrix} -4 \\ 0 \end{bmatrix} \right) e^{-t}. \quad \triangleleft$$

## Repeated eigenvalue non-diagonalizable $2 \times 2$ system.

### Example

Find the solution  $\mathbf{x}$  to the IVP

$$\mathbf{x}' = A\mathbf{x}, \quad \mathbf{x}(0) = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad A = \frac{1}{4} \begin{bmatrix} -6 & 4 \\ -1 & -2 \end{bmatrix}.$$

**Solution:** The general solution is

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

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

$$\begin{bmatrix} 2 & -4 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \Rightarrow \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \frac{1}{4} \begin{bmatrix} 0 & 4 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1/4 \end{bmatrix}.$$

We conclude:  $\mathbf{x}(t) = \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{-t} + \frac{1}{4} \left( \begin{bmatrix} 2 \\ 1 \end{bmatrix} t + \begin{bmatrix} -4 \\ 0 \end{bmatrix} \right) e^{-t}$ .  $\triangleleft$

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

### Example

Sketch a phase portrait for solutions of

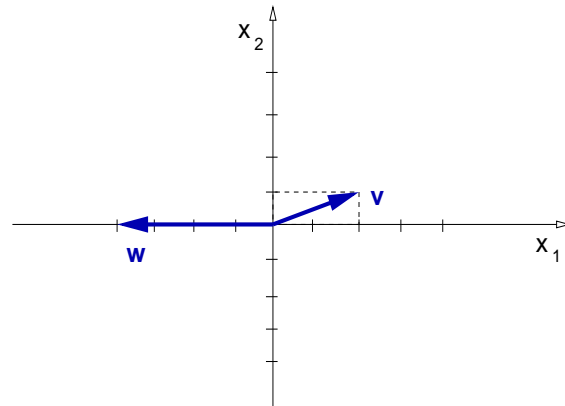
$$\mathbf{x}' = A\mathbf{x}, \quad A = \frac{1}{4} \begin{bmatrix} -6 & 4 \\ -1 & -2 \end{bmatrix}.$$

Solution:

We start plotting the vectors

$$\mathbf{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix},$$

$$\mathbf{w} = \begin{bmatrix} -4 \\ 0 \end{bmatrix}.$$



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

### Example

Sketch a phase portrait for solutions of

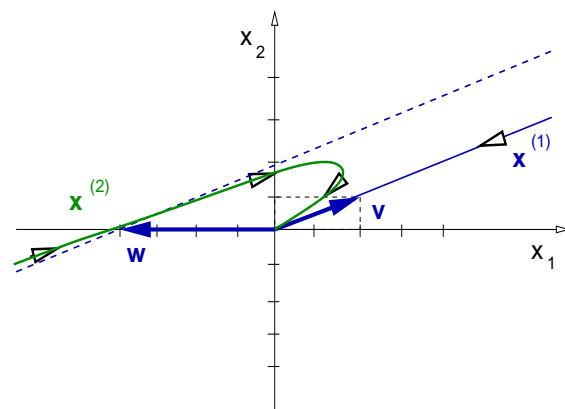
$$\mathbf{x}' = A\mathbf{x}, \quad A = \frac{1}{4} \begin{bmatrix} -6 & 4 \\ -1 & -2 \end{bmatrix}.$$

Solution:

Now plot the solutions

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

$$\mathbf{x}^{(2)} = \left( \begin{bmatrix} 2 \\ 1 \end{bmatrix} t + \begin{bmatrix} -4 \\ 0 \end{bmatrix} \right) e^{-t}.$$





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

### Example

Sketch a phase portrait for solutions of

$$\mathbf{x}' = A\mathbf{x}, \quad A = \frac{1}{4} \begin{bmatrix} -6 & 4 \\ -1 & -2 \end{bmatrix}.$$

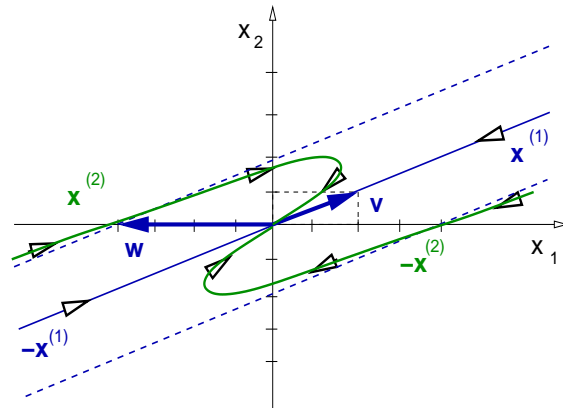
**Solution:**

Now plot the solutions

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

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

This is the case  $\lambda < 0$ .



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

### Example

Given any vectors  $\mathbf{v}$  and  $\mathbf{w}$ , and any constant  $\lambda$ , plot the phase portraits of the functions

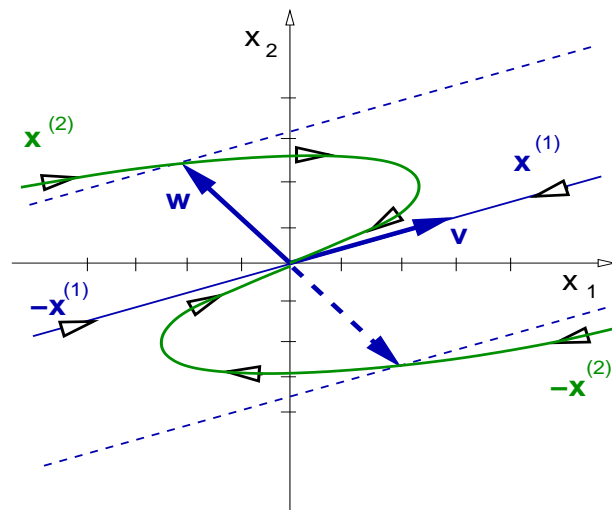
$$\mathbf{x}^{(1)}(t) = \mathbf{v} e^{\lambda t}, \quad \mathbf{x}^{(2)}(t) = (\mathbf{v} t + \mathbf{w}) e^{\lambda t},$$

**Solution:**

The case  $\lambda < 0$ . We plot the functions

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

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



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

### Example

Given any vectors  $\mathbf{v}$  and  $\mathbf{w}$ , and any constant  $\lambda$ , plot the phase portraits of the functions

$$\mathbf{x}^{(1)}(t) = \mathbf{v} e^{\lambda t}, \quad \mathbf{x}^{(2)}(t) = (\mathbf{v} t + \mathbf{w}) e^{\lambda t},$$

### Solution:

The case  $\lambda > 0$ . We plot the functions

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

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

