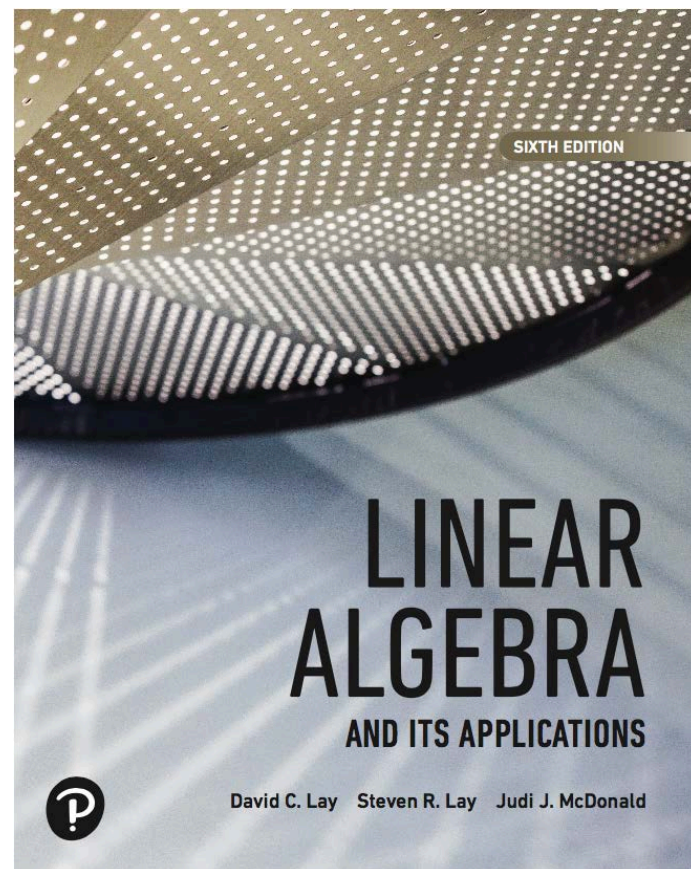


# 5

## Eigenvalues and Eigenvectors

### 5.7

#### Applications To Differential Equations



# DYNAMICAL SYSTEMS

- A system of differential equations is given by the equations:

$$x_1' = a_{11}x_1 + \cdots + a_{1n}x_n$$

$$x_2' = a_{21}x_1 + \cdots + a_{2n}x_n$$

$$\vdots$$

$$x_n' = a_{n1}x_1 + \cdots + a_{nn}x_n$$

- where  $x_1, \dots, x_n$  are differentiable functions of  $t$ , with derivatives  $x_1', \dots, x_n'$ , and the  $a_{ij}$  are constants.

# DYNAMICAL SYSTEMS

- Written in matrix form, the system can be expressed as:

$$\mathbf{x}' = A\mathbf{x}$$

- This equation is *linear* because both differentiation of functions and multiplication of vectors by a matrix are linear transformations.

# DYNAMICAL SYSTEMS

- A fundamental set of solutions is a *basis* for the set of all solutions, an  $n$ -dimensional vector space of functions.
- If a vector  $\mathbf{x}_0$  is specified, then the **initial value problem** is to construct the (unique) function  $\mathbf{x}$  such that  $\mathbf{x}' = A\mathbf{x}$  and  $\mathbf{x}(0) = \mathbf{x}_0$ .

# DYNAMICAL SYSTEMS

- Notice if  $\mathbf{x}(t) = \mathbf{v}e^{\lambda t}$  for some vector  $\mathbf{v}$ , then  $\mathbf{x}'(t) = \lambda \mathbf{v}e^{\lambda t}$ .
- For this choice of  $\mathbf{x}(t)$ , notice
$$\mathbf{x}'(t) = A\mathbf{x}(t)$$
if and only if
$$A\mathbf{v} = \lambda \mathbf{v}.$$
- Thus each eigenvalue–eigenvector pair of  $A$  provides a solution of  $\mathbf{x}'(t) = A\mathbf{x}(t)$ .

# Dynamical Systems

- **Example:** Consider  $A = \begin{bmatrix} -1.5 & 0.5 \\ 1 & -1 \end{bmatrix}$ .

Find all solutions to  $\mathbf{x}' = A\mathbf{x}$  with  $\mathbf{x}(0) = \begin{bmatrix} 5 \\ 4 \end{bmatrix}$ .

- **Solution:** The eigenvalues of  $A$  are  $\lambda_1 = -0.5$  and  $\lambda_2 = -2$ ,  
with corresponding eigenvectors  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$  and  $\mathbf{v}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ .

# Dynamical Systems

- Set  $\mathbf{x}(t) = c_1 \mathbf{v}_1 e^{-0.5t} + c_2 \mathbf{v}_2 e^{-2t}$  and  $\mathbf{x}(0) = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2$ .

- Solving  $\begin{bmatrix} 5 \\ 4 \end{bmatrix} = c_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix}$  results in  $c_1 = 3$  and  $c_2 = -2$ .

- The solution to the initial boundary problem is

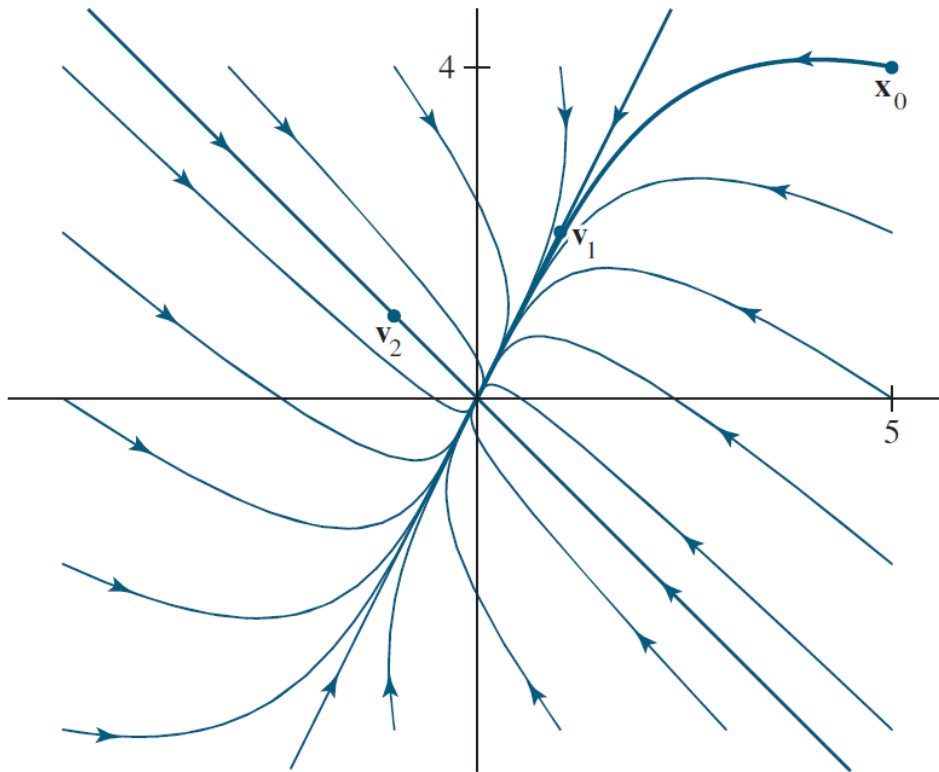
$$\mathbf{x}(t) = 3 \begin{bmatrix} 1 \\ 2 \end{bmatrix} e^{-0.5t} - 2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-2t}$$

# Dynamical Systems

- The origin is called an **attractor**, or **sink**, of this dynamical system because all trajectories are drawn into the origin.
- The direction of greatest attraction is along the line through **0** and  $\mathbf{v}_2$  corresponding to the more negative eigenvalue,  $\lambda = -2$ .



# Dynamical Systems



**FIGURE 2** The origin as an attractor.

# Dynamical Systems

- If the eigenvalues are positive instead of negative, the corresponding trajectories would be similar in shape, but the trajectories would be traversed **away** from the origin.
- In such a case, the origin is called a **repeller**, or **source**, of the dynamical system.
- The direction of greatest repulsion is the line containing the trajectory of the eigenfunction corresponding to the more positive eigenvalue.

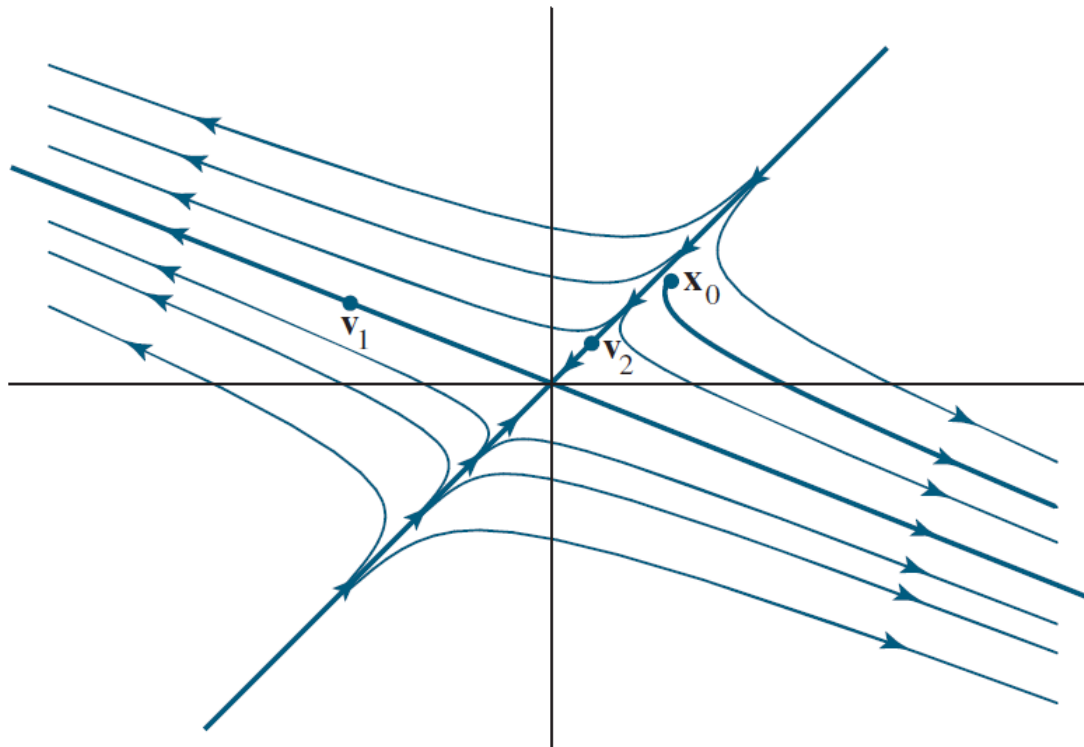
# Dynamical Systems

- A saddle point arises whenever the matrix  $A$  has both positive and negative eigenvalues.
- The origin is called a **saddle point** of the dynamical system because some trajectories approach the origin at first and then change direction and move away from the origin

# Dynamical Systems

- The direction of greatest repulsion is the line through  $\mathbf{v}_1$  and  $\mathbf{0}$ , corresponding to the positive eigenvalue.
- The direction of greatest attraction is the line through  $\mathbf{v}_2$  and  $\mathbf{0}$ , corresponding to the negative eigenvalue.

# Dynamical Systems (12 of 12)



**FIGURE 3** The origin as a saddle point.

# Decoupling a Dynamical System

- Suppose the eigenfunctions for  $A$  are  $\mathbf{v}_1 e^{\lambda_1 t}, \mathbf{v}_2 e^{\lambda_2 t}, \dots, \mathbf{v}_n e^{\lambda_n t}$  with  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  linearly independent eigenvectors.
- Let  $P = [\mathbf{v}_1 \cdots \mathbf{v}_n]$ , and let  $D$  be the diagonal matrix with entries  $\lambda_1 \dots \lambda_n$ , so that  $A = PDP^{-1}$ .
- Now make a **change of variable**, defining a new function  $\mathbf{y}$  by  $\mathbf{y}(t) = P^{-1}\mathbf{x}(t)$ , or, equivalently,  $\mathbf{x}(t) = P\mathbf{y}(t)$

# DECOUPLING A DYNAMICAL SYSTEM

- Substituting  $\mathbf{x}(t) = P\mathbf{y}(t)$  into  $\dot{\mathbf{x}}(t) = A\mathbf{x}(t)$  results in  $P\dot{\mathbf{y}}(t) = PD P^{-1}\mathbf{x}(t) = PD\mathbf{y}(t)$ .
- Multiplying both sides of the equation by  $P^{-1}$  creates  $\dot{\mathbf{y}}(t) = D\mathbf{y}(t)$ .
- The change of variable from  $\mathbf{x}$  to  $\mathbf{y}$  has *decoupled* the system of differential equations, because the derivative of each scalar function  $y_k$  depends only on  $y_k$ .

# DECOUPLING A DYNAMICAL SYSTEM

- Substituting  $\mathbf{x}(t)=P\mathbf{y}(t)$  into  $\mathbf{x}'(t) = A\mathbf{x}(t)$  results in  $P\mathbf{y}'(t) = PDP^{-1}\mathbf{x}(t) = PD\mathbf{y}(t)$ .
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