5.1 Matrices and Linear Systems

Review of Matrix Notation and Terminology

An $m \times n$ matrix **A** is a rectangular array of mn numbers (or elements) arranged in m (horizontal) rows and n (vertical) columns:

$$\mathbf{A} = egin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \ a_{21} & a_{22} & \cdots & a_{2n} \ dots & dots & dots & dots \ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}$$

Two m imes n matrices $\mathbf{A} = \begin{bmatrix} a_{ij} \end{bmatrix}$ and $\mathbf{B} = \begin{bmatrix} b_{ij} \end{bmatrix}$ are said to be equal if corresponding elements are equal. We have

$$egin{aligned} \mathbf{A} + \mathbf{B} &= ig[a_{ij}ig] + ig[b_{ij}ig] = ig[a_{ij} + b_{ij}ig] \ & c\mathbf{A} = \mathbf{A}c = ig[ca_{ij}ig] \end{aligned}$$

We have

$$A+B = B+A$$

$$(A+B)+C = A+(B+C)$$

$$c(A+B) = CA+CB$$

$$(C+d)A = CA+dA$$

The **transpose** \mathbf{A}^T of the $m \times n$ matrix $\mathbf{A} = \begin{bmatrix} a_{ij} \end{bmatrix}$ is the $n \times m$ matrix whose j th column is the j th row of \mathbf{A}

Example:
$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix}$$
 then $A^{T} = \begin{bmatrix} 1 & 3 & 5 \\ 2 & 4 & 6 \end{bmatrix}$

Matrix Multiplication

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$$\mathbf{a} = egin{bmatrix} a_1 \, a_2 \, \cdots \, a_p \end{bmatrix} \hspace{1mm} ext{and} \hspace{1mm} \mathbf{b} = egin{bmatrix} b_1 \, b_2 \, \cdots \, b_p \end{bmatrix}^T$$

then the **scalar product** $\mathbf{a} \cdot \mathbf{b}$ is defined as follows:

$$\mathbf{a}\cdot\mathbf{b}=\sum_{k=1}^pa_kb_k=a_1b_1+a_2b_2+\dots+a_pb_p$$

The product **AB** of two matrices is defined only if the number of columns of **A** is equal to the number of rows of **B**. If **A** is an $m \times p$ matrix and **B** is a $p \times n$ matrix, then their product **AB** is the $m \times n$ matrix

 $\mathbf{C} = egin{bmatrix} c_{ij} \end{bmatrix}$

where c_{ij} is the scalar product of the *i*th row vector \mathbf{a}_i of \mathbf{A} and the *j*th column vector \mathbf{b}_j of \mathbf{B} . Thus

$$\mathbf{C} = \mathbf{A}\mathbf{B} = ig[\mathbf{a}_i\cdot\mathbf{b}_jig]$$

If $\mathbf{A} = ig[a_{ij}ig]$ and $\mathbf{B} = ig[b_{ij}ig]$, then we have

$$c_{ij} = \sum_{k=1}^p a_{ik} b_{kj}$$

For the computation by hand, it is easy to remember by visualizing the picture

$$\mathbf{a}_{i} \rightarrow \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1p} \\ a_{21} & a_{22} & \cdots & a_{2p} \\ \vdots & \vdots & & \vdots \\ a_{i1} & a_{i2} & \cdots & a_{ip} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mp} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1j} & \cdots & b_{1n} \\ b_{21} & b_{22} & \cdots & b_{2j} & \cdots & b_{2n} \\ \vdots & \vdots & & \vdots \\ b_{p1} & b_{p2} & \cdots & b_{pj} & \cdots & b_{pn} \end{bmatrix},$$

which shows that one forms the dot product of the row vector \mathbf{a}_i with the column vector \mathbf{b}_j to obtain the element c_{ij} in the *i*th row and the *j* th column of \mathbf{AB} .

Inverse Matrices

The **identity** matrix of order *n* is the square matrix

$$a \in \mathbb{R}, \ a \cdot 1 = 1 \cdot a = a$$

$$a \cdot b = b \cdot a = 1$$

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We have

AI = IA = A

If ${f A}$ is a square matrix, then an inverse of ${f A}$ is a square matrix ${f B}$ of the same order as ${f A}$ such that both

 $\mathbf{I} = \begin{pmatrix} 0\\0\\\vdots\\\vdots\\ 0\end{pmatrix}$

$$AB = I and BA = I$$

We denote such \mathbf{B} by \mathbf{A}^{-1} .

RmL: Note B may not exist! E.g. $A = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$

When A⁻¹ exists?

In linear algebra it is proved that \mathbf{A}^{-1} exists if and only if the determinant $\det(\mathbf{A})$ of the square matrix \mathbf{A} is nonzero. If so, the matrix **A** is said to be **nonsingular**; if det $(\mathbf{A}) = 0$, then **A** is called a **singular** matrix.

Example 1 Find \mathbf{AB} and \mathbf{BA} given

$$A = \begin{pmatrix} 5 & 3 & 4 \\ 3 & -2 & 1 \end{pmatrix} \text{ and } B = \begin{pmatrix} 1 & 2 \\ 4 & 5 \\ 2 & 3 \end{pmatrix}$$

$$ANS: AB = \begin{pmatrix} 5 & 3 & 4 \\ 3 & -2 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 4 & 5 \\ -2 & 3 \end{pmatrix} = \begin{pmatrix} 5 \times |+3 \times 4 + 2 \times 4 \\ -3 \times |+3 \times 5 + 4 \times 3 \end{pmatrix}$$

$$ANS: AB = \begin{pmatrix} 5 & 3 & 4 \\ -3 & -2 & -1 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ -4 & 5 \\ -2 & 3 \end{pmatrix} = \begin{pmatrix} 5 \times |+3 \times 4 + 2 \times 4 \\ -3 \times |-2 \times 4 + | \times 2 \end{pmatrix}$$

$$= \begin{pmatrix} 25 & 37 \\ -3 & -1 \end{pmatrix}$$

$$BA = \begin{pmatrix} 1 & 2 \\ 4 & 5 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} 5 & 3 & 4 \\ 3 & -2 & 1 \end{pmatrix} = \begin{pmatrix} 5t6 & 3-4 & 4+2 \\ 20+15 & 12-10 & 16+5 \\ 10t & 9 & 0 & 8+3 \end{pmatrix}$$

$$= \begin{pmatrix} 11 & -1 & 6 \\ 35 & 2 & 21 \\ 19 & 0 & 11 \end{pmatrix}$$

Note $AB \neq BA$

Matrix-Valued Functions

A **matrix-valued function** is a matrix such as

$$\mathbf{A}(t) = egin{pmatrix} a_{11}(t) & a_{12}(t) & \cdots & a_{1n}(t) \ a_{21}(t) & a_{22}(t) & \cdots & a_{2n}(t) \ dots & dots & dots & dots \ a_{n1}(t) & a_{n2}(t) & \cdots & a_{nn}(t) \end{pmatrix}$$

in which each entry is a function of t.

We say that the matrix function $\mathbf{A}(t)$ is **continuous** (or **differentiable**) at a point (or on an interval) if each of its elements has the same property. The **derivative** of a differentiable matrix function is defined by elementwise differentiation:

$$\mathbf{A}'(t) = \frac{d\mathbf{A}}{dt} = \left[\frac{da_{ij}}{dt}\right]$$

Example 2 Let A and B be the matrices as in Example 1. Let

$$\mathbf{x} = \begin{pmatrix} e^{-2t} \\ 3t \end{pmatrix}_{\mathbf{x}} \text{ and } \mathbf{y} = \begin{pmatrix} t^3 \\ \tan t \\ \sin t \end{pmatrix}_{\mathbf{x}}$$

Find Ay and Bx. Are the products Ax and By well-defined?

ANS

AWS:
$$A\vec{y} = \begin{pmatrix} s & 3 & 4 \\ 3 & -2 & 1 \end{pmatrix}_{23} \begin{pmatrix} t^3 \\ tant \\ sint \end{pmatrix}_{3\times 1} = \begin{pmatrix} st^3 + 3tant + 4sint \\ 3t^3 - 2tant + sint \end{pmatrix}_{2\times 1}$$

 $B\vec{z} = \begin{pmatrix} 1 & 2 \\ + & 5 \\ 2 & 3 \end{pmatrix}_{3\times 2} \begin{pmatrix} e^{-2t} \\ 3t \end{pmatrix}_{3\times 1} = \begin{pmatrix} e^{-2t} + 6t \\ 4e^{-2t} + 1st \\ 2e^{-3t} + 9t \end{pmatrix}_{3\times 1}$
The product $A \neq and B\vec{z}$ are not well-defined. since A
is a 2×3 matrix but χ is a 2×1 matrix, and,
 B is a 3×2 matrix, but χ is a 3×1 matrix
Example 3 Find A' if

$$\mathbf{A}(t) = egin{pmatrix} 3t & t^2 \ t^3 & 3+t^4 \end{pmatrix}$$

$$A' = \begin{pmatrix} (3t)' & (t^2)' \\ (t^3)' & (3+t^q)' \end{pmatrix} = \begin{pmatrix} 3 & 2t \\ \partial t^2 & 4t^3 \end{pmatrix}$$

First-Order Linear Systems

We discuss here the general system of n first-order linear equations

$$\begin{aligned} x_{1}' &= p_{11}x_{1} + p_{12}x_{2} + \cdots + p_{1n}x_{n} + f_{1}(t), \\ x_{2}' &= p_{21}x_{1} + p_{22}x_{2} + \cdots + p_{2n}x_{n} + f_{2}(t), \\ x_{3}' &= p_{31}x_{1} + p_{32}x_{3} + \cdots + p_{3n}x_{n} + f_{3}(t), \\ \vdots \\ x_{n}' &= p_{n1}x_{1} + p_{n2}x_{2} + \cdots + p_{nn}x_{n} + f_{n}(t), \end{aligned} \\ \begin{aligned} x_{1}' \\ x_{2}' \\ x_{3}' \\ \vdots \\ x_{n}' \\ z_{n}' \\$$

If we introduce the coefficient matrix

$$\mathbf{P}(t) = \Big[p_{ij}(t) \Big]$$

and the column vectors

$$\mathbf{x} = egin{bmatrix} x_i \end{bmatrix} \quad ext{and} \quad \mathbf{f}(t) = egin{bmatrix} f_i(t) \end{bmatrix}$$

Then the above system takes the form of a single matrix equation

$$\frac{d\mathbf{x}}{dt} = \mathbf{P}(t)\mathbf{x} + \mathbf{f}(t) \tag{1}$$

A solution of Eq. (1) on the open interval I is a column vector function $\mathbf{x}(t) = \lfloor x_i(t) \rfloor$ such that the component functions of \mathbf{x} satisfy the above system identically on I.

Example 4 Write the given system in the for $\mathbf{x}' = \mathbf{P}(t)\mathbf{x} + \mathbf{f}(t)$.

$$\begin{array}{cccc} & x' = x + 3y + 2e^{t}, \\ & y' = 4x - y - t^{2} \end{array}$$

$$\begin{array}{ccccc} \text{ANS:} & \text{We have} \\ & \overrightarrow{x} = \begin{bmatrix} x(t) \\ y(t) \end{bmatrix}, & \text{Ptt} = \begin{bmatrix} 1 & 3 \\ 4 & -1 \end{bmatrix}, & \overrightarrow{f}(t) = \begin{bmatrix} 2e^{t} \\ -t^{2} \end{bmatrix} \end{array}$$

$$\begin{array}{ccccc} \text{Then} \\ & \overbrace{y} \end{bmatrix} = \begin{bmatrix} 1 & 3 \\ 4 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} 2e^{t} \\ -t^{2} \end{bmatrix}$$

$$\begin{array}{ccccc} \overrightarrow{T} \\ \overrightarrow{T} \\ \overrightarrow{T} \\ \overrightarrow{X}'(t) \end{pmatrix} = \begin{array}{ccccc} 1 & 3 \\ \overrightarrow{T} \\ \overrightarrow{T} \\ \overrightarrow{T} \\ \overrightarrow{T} \\ \overrightarrow{T} \\ \overrightarrow{T} \end{array}$$

$$(2) x' = 2x - 3y = 2x - 3y + 0 \cdot Z$$

$$y' = x + y + 2z,$$

$$z' = 5y - 7z = 0 \cdot x + 5y - 7Z$$
ANS: We have $\vec{x} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$, $P = \begin{bmatrix} 2 & -3 & 0 \\ 1 & 1 & 2 \\ 0 & 5 & -7 \end{bmatrix}$, $\vec{f}(t) = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$
So $\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} 2 & -3 & 0 \\ 1 & 1 & 2 \\ 0 & 5 & -7 \end{bmatrix} \begin{bmatrix} x \\ y \\ Z \end{bmatrix}$

To solve the Eq. (1) in general, we consider first the the **associated homogeneous equation**

$$\frac{d\mathbf{x}}{dt} = \mathbf{P}(t)\mathbf{x} \tag{2}$$

We expect it to have *n* solutions $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ that are independent in some appropriate sense, and such that every solution of Eq. (2) is a linear combination of these *n* particular solutions.

Given n solutions $\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_n$ of Eq. (2), we write

$$\mathbf{x}_{j}(t) = \begin{pmatrix} x_{1j}(t) \\ \vdots \\ x_{ij}(t) \\ \vdots \\ x_{nj}(t) \end{pmatrix}$$
(3)

Theorem 1 Principle of Superposition

Let $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ be *n* solutions of the homogeneous linear equation in (2) on the open interval I. If c_1, c_2, \dots, c_n are constants, then the linear combination

$$\mathbf{x}(t) = c_1 \mathbf{x}_1(t) + c_2 \mathbf{x}_2(t) + \dots + c_n \mathbf{x}_n(t)$$

is also a solution of Eq. (2) on I.

Independence and General Solutions

The vector-valued functions $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ are **linearly dependent** on the interval I provided that there exist constants c_1, c_2, \dots, c_n not all zero such that

$$c_1\mathbf{x}_1(t) + c_2\mathbf{x}_2(t) + \dots + c_n\mathbf{x}_n(t) = \mathbf{0}$$

for all *t* in *I*. Otherwise, they are **linearly independent**.

Just as in the case of a single nth-order equation, there is a Wronskian determinant that tells us whether or not n given solutions of the homogeneous equation in (1) are linearly dependent. If $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ are such solutions, then their Wronskian is the $n \times n$ determinant

$$W(t) = egin{bmatrix} x_{11}(t) & x_{12}(t) & \cdots x_{1n}(t) \ x_{21}(t) & x_{22}(t) & \cdots x_{2n}(t) \ dots & dots & dots \ x_{n1}(t) & x_{n2}(t) & \cdots x_{nn}(t) \end{bmatrix}$$

using the notation in (3) for the components of the solutions.

Theorem 2 Wronskians of Solutions

Suppose that $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ are *n* solutions of the homogeneous linear equation $\mathbf{x}' = \mathbf{P}(t)\mathbf{x}$ on an open interval *I*. Suppose also that $\mathbf{P}(t)$ is continuous on *I*. Let

$$W=W(\mathbf{x}_1,\mathbf{x}_2,\cdots,\mathbf{x}_n)$$

Then

- If $\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_n$ are linearly dependent on I, then W = 0 at every point of I.
- If $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ are linearly independent on I, then $W \neq 0$ at each point of I. Thus there are only two possibilities for solutions of homogeneous systems: Either W = 0 at every point of I, or $W \neq 0$ at no point of I.

Theorem 3 General Solutions of Homogeneous Systems

Let $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ be n linearly independent solutions of the homogeneous linear equation $\mathbf{x}' = \mathbf{P}(t)\mathbf{x}$ on an open interval I, where $\mathbf{P}(t)$ is continuous. If $\mathbf{x}(t)$ is any solution whatsoever of the equation $\mathbf{x}' = \mathbf{P}(t)\mathbf{x}$ on I, then there exist numbers $c_1, c_2, \dots c_n$ such that

$$\mathbf{x}(t) = c_1 \mathbf{x}_1(t) + c_2 \mathbf{x}_2(t) + \dots + c_n \mathbf{x}_n(t)$$

for all t in I.

Example 5 In the following question, first verify that the given vectors are solutions of the given system. Then use the Wronskian to show that they are linearly independent. Finally, write the general solution of the system Step 2 $\mathbf{x}' = egin{pmatrix} 4 & -3 \ 6 & -7 \end{pmatrix} \mathbf{x}; \quad \mathbf{x}_1 = e^{2t} egin{pmatrix} 3 \ 2 \end{pmatrix}, \quad \mathbf{x}_2 = e^{-5t} egin{pmatrix} 1 \ 3 \end{pmatrix}$ ANS: Step 1 For $\vec{x}_1, \ LHS = \vec{x}_1' = \begin{pmatrix} 6 \cdot e^{zt} \\ 4 \cdot e^{zt} \end{pmatrix} \stackrel{\checkmark}{=} RHS = \begin{pmatrix} 4 & -3 \\ 6 & -7 \end{pmatrix} \begin{pmatrix} 3e \\ 2e^{2t} \end{pmatrix} \stackrel{=}{=} \begin{pmatrix} 6 \cdot e^{zt} \\ 4 \cdot e^{zt} \end{pmatrix}$ Step 2. We compute the Wronskian of Ritt) Ritt): $W(t) = \begin{vmatrix} 3e^{3t} & e^{5t} \\ 2e^{3t} & 3e^{-5t} \end{vmatrix} = 9e^{-3t} - 2e^{-3t} = 7e^{-3t} \neq 0$ By Thm 2, we know $\vec{x}_1(t) \quad \vec{x}_2(t)$ are linearly independent. Step 3. The general solution is (by Thm 3) $\begin{pmatrix} x | t \\ y | t \\ y | t \\ z \end{pmatrix} = \vec{x} | t \\ = c_{1} \vec{x}_{1} | t \\$ $\Rightarrow \begin{cases} x_{1}(t) = 3C_{1}e^{2t} + C_{2}e^{-5t} \\ y_{1}(t) = 2C_{1}e^{2t} + 2C_{2}e^{-5t} \end{cases}$

Step 1

Example 6 Find a particular solution of the linear system that satisfies the given initial conditions.

$$\mathbf{x}' = \begin{pmatrix} 4 & -3 \\ 6 & -7 \end{pmatrix} \mathbf{x}; \quad \mathbf{x}_{1} = e^{2t} \begin{pmatrix} 3 \\ 2 \end{pmatrix}, \quad \mathbf{x}_{2} = e^{-5t} \begin{pmatrix} 1 \\ 3 \end{pmatrix}, \quad (0) = 8, \quad (0) = 0$$

$$ANS: By Example S, We hnow
$$\begin{pmatrix} \mathbf{x}(t) \\ \mathbf{y}(t) \end{pmatrix} = \mathbf{x}'(t) = \begin{pmatrix} 3c_{1} e^{\mathbf{x}t} + c e^{-\mathbf{x}t} \\ 2c_{1} e^{\mathbf{x}t} + 3c_{2} e^{-\mathbf{x}t} \end{pmatrix}$$

$$Since \begin{pmatrix} \mathbf{x}(0) \\ \mathbf{y}(0) \end{pmatrix} = \begin{pmatrix} 3c_{1} e^{\mathbf{x}t} + c_{2} e^{-\mathbf{x}t} \\ 2c_{1} e^{\mathbf{x}0} + 3c_{2} e^{-\mathbf{x}0} \end{pmatrix} = \begin{pmatrix} 8 \\ 0 \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} 8 \\ 0 \end{pmatrix} = \begin{pmatrix} 3c_{1} + c_{3} \\ 2c_{1} + 3c_{3} \end{pmatrix} \Rightarrow \begin{pmatrix} 3c_{1} + c_{4} = 8 \\ 2c_{1} + 3c_{3} = 0 \end{pmatrix}$$

$$\Rightarrow \begin{cases} c_{1} = \frac{2^{\mathbf{x}t}}{7} \\ c_{3} = -\frac{16}{7} \\ Thus \\ \mathbf{x}'(t) = \begin{pmatrix} \frac{7t}{7} e^{\mathbf{x}t} - \frac{16}{7} e^{-\mathbf{x}t} \\ \frac{48}{7} e^{\mathbf{x}t} - \frac{48}{7} e^{-\mathbf{x}t} \end{bmatrix} = \begin{pmatrix} \kappa(t) \\ \mathbf{y}(t) \end{pmatrix}$$$$