

Applied Linear Algebra

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Chapter 2

Linear Transformation

Chia-Hui Chang

Email: chia@csie.ncu.edu.tw

National Central University, Taiwan

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2.1 Introduction to Linear Transformations and Their Inverse

See Figure 1

Encryption of a coordinate $\vec{x} = \begin{bmatrix} 5 \\ 42 \end{bmatrix}$ to \vec{y} by the following code:

$$\begin{aligned} y_1 &= x_1 + 3x_2 = 131 \\ y_2 &= 2x_1 + 5x_2 = 220 \end{aligned}$$

At the headquarter, $\vec{y} = \begin{bmatrix} 131 \\ 220 \end{bmatrix}$ is received. We need to determine the actual \vec{x} by solve the linear system.

$$A\vec{x} = \vec{b}$$

i.e.
$$\begin{aligned} x_1 + 3x_2 &= 131 \\ 2x_1 + 5x_2 &= 220 \end{aligned}$$

If $\vec{y} = \begin{bmatrix} 133 \\ 223 \end{bmatrix}$ We need to solve it again by:

$$\begin{aligned}x_1 + 3x_2 &= 133 \\ 2x_1 + 5x_2 &= 223\end{aligned}$$

For a general formula, we need solve the system

$$\begin{aligned}x_1 + 3x_2 &= y_1 \\ 2x_1 + 5x_2 &= y_2\end{aligned}$$

for arbitrary constants y_1 and y_2 .

For sender: $\vec{x} \rightarrow \vec{y}$ (encoding)

For receiver: $\vec{y} \rightarrow \vec{x}$ (decoding)

$$\left| \begin{array}{l} x_1 + 3x_2 = y_1 \\ 2x_1 + 5x_2 = y_2 \end{array} \right| \xrightarrow{-2(I)}$$

$$\left| \begin{array}{l} x_1 + 3x_2 = y_1 \\ -x_2 = -2y_1 + y_2 \end{array} \right| \xrightarrow{\div(-1)}$$

$$\left| \begin{array}{l} x_1 + 3x_2 = y_1 \\ x_2 = 2y_1 - y_2 \end{array} \right| \xrightarrow{-3(II)}$$

$$\left| \begin{array}{l} x_1 = -5y_1 + 3y_2 \\ x_2 = 2y_1 - y_2 \end{array} \right|$$

The decoding formula is:

$$\begin{aligned} x_1 &= -5y_1 + 3y_2 \\ x_2 &= 2y_1 - y_2 \end{aligned}$$

or $\vec{x} = B\vec{y}$, where $B = \begin{bmatrix} -5 & 3 \\ 2 & -1 \end{bmatrix}$

Definition. We say that the matrix B is the inverse of the matrix A and write $B = A^{-1}$.

$$\begin{array}{ccc} & \vec{y} = A\vec{x}, A = \begin{bmatrix} 1 & 3 \\ 2 & 5 \end{bmatrix} & \\ \vec{x} & \xleftrightarrow{\hspace{10em}} & \vec{y} \\ & \vec{x} = B\vec{y}, B = \begin{bmatrix} -5 & 3 \\ 2 & -1 \end{bmatrix} & \end{array}$$

The coding transformation is represented as

$$\underbrace{\begin{bmatrix} y_1 \\ y_2 \end{bmatrix}}_{\vec{y}} = \begin{bmatrix} x_1 + 3x_2 \\ 2x_1 + 5x_2 \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 3 \\ 2 & 5 \end{bmatrix}}_A \underbrace{\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}}_{\vec{x}}$$

or succinctly, as $\boxed{\vec{y} = A\vec{x}}$.

A transformation of the form $\vec{y} = A\vec{x}$ is called a **linear transformation**.

Function: Consider two sets X and Y . A function $T : X \rightarrow Y$ is a rule that associates with each element $x \in X$ a unique element $y \in Y$.

The set X is called the *domain* and Y is called its *codomain*.

Definition. A function T from R^n to R^m is called a **linear transformation** if there is an $m \times n$ matrix A such that

$$T(\vec{x}) = A\vec{x}, \text{ for all } \vec{x} \text{ in } R^n.$$

Example. The linear transformation system

$$\begin{aligned} y_1 &= 7x_1 + 3x_2 - 9x_3 + 8x_4 \\ y_2 &= 6x_1 + 2x_2 - 8x_3 + 7x_4 \\ y_3 &= 8x_1 + 4x_2 + 7x_4 \end{aligned}$$

(a function from R^4 to R^3) can be represented by the 3×4 matrix

$$A = \begin{bmatrix} 7 & 3 & -9 & 8 \\ 6 & 2 & -8 & 7 \\ 8 & 4 & 0 & 7 \end{bmatrix}$$

Example. *The identity transformation system*

$$\begin{aligned}y_1 &= x_1 \\y_2 &= x_2 \\&\vdots \\y_n &= x_n\end{aligned}$$

(a linear transformation from \mathbb{R}^n to \mathbb{R}^n whose output equals its input) is represented by $n \times n$ matrix

$$A = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$$

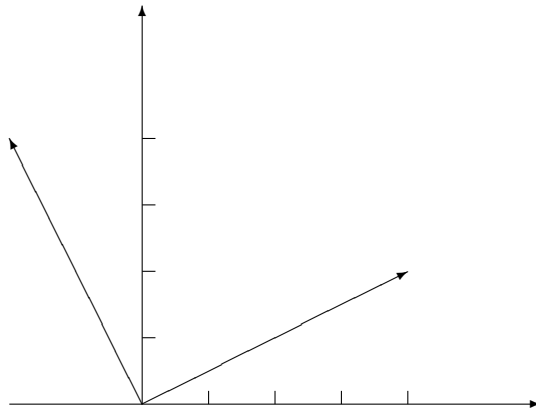
This matrix is called the **identity matrix** and is denoted by I_n :

$$I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad \text{etc.}$$

Example. Give a geometric interpretation of the linear transformation

$$\vec{y} = A\vec{x}, \text{ where } A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -x_2 \\ x_1 \end{bmatrix}.$$



See Figure 4 (pp.45).

Fact 2.1.2 Consider a linear transformation T

from R^n to R^m . Let $\vec{e}_i = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \leftarrow \text{ith}$

The matrix of T can be represented as

$$A = \left[\begin{array}{c|c|c|c} T(\vec{e}_1) & T(\vec{e}_2) & \cdots & T(\vec{e}_n) \\ \hline \hline \hline \hline \end{array} \right]$$

Since

$$T(\vec{e}_i) = A\vec{e}_i = \left[\begin{array}{c|c|c|c} \vec{v}_1 & \vec{v}_2 & \cdots & \vec{v}_n \\ \hline \hline \hline \hline \end{array} \right] \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} = \vec{v}_i$$

Example. Find the inverse for the following matrix:

$$\begin{bmatrix} 1 & 2 \\ 3 & 9 \end{bmatrix}$$

Solution

$$\left[\begin{array}{cc|c} 1 & 2 & y_1 \\ 3 & 9 & y_2 \end{array} \right] -3(I)$$

$$\longrightarrow \left[\begin{array}{cc|c} 1 & 2 & y_1 \\ 0 & 3 & -3y_1 + y_2 \end{array} \right] \div 3$$

$$\longrightarrow \left[\begin{array}{cc|c} 1 & 2 & y_1 \\ 0 & 1 & -y_1 + \frac{1}{3}y_2 \end{array} \right] -2(II)$$

$$\longrightarrow \left[\begin{array}{cc|c} 1 & 0 & 3y_1 - \frac{2}{3}y_2 \\ 0 & 1 & -y_1 + \frac{1}{3}y_2 \end{array} \right]$$

$$\Rightarrow \begin{bmatrix} 1 & 2 \\ 3 & 9 \end{bmatrix}^{-1} = \begin{bmatrix} 3 & -\frac{2}{3} \\ -1 & \frac{1}{3} \end{bmatrix}$$

Example. Find the inverse for the following matrix: $\begin{bmatrix} 3 & -\frac{2}{3} \\ -1 & \frac{1}{3} \end{bmatrix}$

Solution

$$\left[\begin{array}{cc|c} 3 & -\frac{2}{3} & y_1 \\ -1 & \frac{1}{3} & y_2 \end{array} \right] \div 3$$

$$\longrightarrow \left[\begin{array}{cc|c} 1 & -\frac{2}{9} & \frac{1}{3}y_1 \\ -1 & \frac{1}{3} & y_2 \end{array} \right] + (I)$$

$$\longrightarrow \left[\begin{array}{cc|c} 1 & -\frac{2}{9} & \frac{1}{3}y_1 \\ 0 & \frac{1}{9} & \frac{1}{3}y_1 + y_2 \end{array} \right] \times 9$$

$$\longrightarrow \left[\begin{array}{cc|c} 1 & -\frac{2}{9} & \frac{1}{3}y_1 \\ 0 & 1 & 3y_1 + 9y_2 \end{array} \right] + \frac{2}{9}(II)$$

$$\longrightarrow \left[\begin{array}{cc|c} 1 & 0 & y_1 + 2y_2 \\ 0 & 1 & 3y_1 + 9y_2 \end{array} \right]$$

$$\Rightarrow \left[\begin{array}{cc} 3 & -\frac{2}{3} \\ -1 & \frac{1}{3} \end{array} \right]^{-1} = \left[\begin{array}{cc} 1 & 2 \\ 3 & 9 \end{array} \right]$$

Example. *Not all linear transformations are invertible. Consider the matrix $A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$*

If $\vec{y} = \begin{bmatrix} 89 \\ 178 \end{bmatrix}$, to solve the system

$$\left| \begin{array}{rcl} x_1 & +2x_2 & = 89 \\ 2x_1 & +4x_2 & = 178 \end{array} \right|$$

We discover there are infinitely many solutions

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 89 - 2t \\ t \end{bmatrix}$$

We say that the coding matrix A are *noninvertible*.

Homework. *Exercises 2.1: 4, 5, 7, 10, 12, 15*

2.2 Linear Transformation in Geometry

Example. 1 Consider a linear transformation system $T(\vec{x}) = A\vec{x}$ from R^n to R^m .

a. $T(\vec{v} + \vec{w}) = T(\vec{v}) + T(\vec{w})$

In words, the transformation of the sum of two vectors equals the sum of the transformation.

b. $T(k\vec{v}) = kT(\vec{v})$

In words, the transformation of a scalar multiple of a vector is the scalar multiple of the transform.

See Figure 1 (pp.50).

Fact A transformation T from R^n to R^m is linear iff

a. $T(\vec{v} + \vec{w}) = T(\vec{v}) + T(\vec{w})$, for all \vec{v}, \vec{w} in R^n , and

b. $T(k\vec{v}) = kT(\vec{v})$, for all \vec{v} in R^n and all scalars k .

Proof

Idea: To prove the converse, we must show a matrix A such that $T(\vec{x}) = A\vec{x}$. Consider a transformation T from R^n to R^m that satisfy (a) and (b), find A .

Example. 2 Consider a linear transformation T from \mathbb{R}^2 to \mathbb{R}^2 . The vectors $T\vec{e}_1$ and $T\vec{e}_2$ are sketched in Figure 2. Sketch the **image** of the unit square under this transformation.

See Figure 2. (pp. 51)

Example. 3 Consider a linear transformation T from \mathbb{R}^2 to \mathbb{R}^2 such that $T(\vec{v}_1) = \frac{1}{2}\vec{v}_1$ and $T(\vec{v}_2) = 2\vec{v}_2$, for the vectors \vec{v}_1 and \vec{v}_2 in Figure 5. On the same axes, sketch $T(\vec{x})$, for the given vector \vec{x} .

See Figure 5. (pp. 52)

[Rotation]

Example. 4 Let T be the counterclockwise rotation through an angle α .

a. Draw sketches to illustrate that T is a linear transformation.

b. Find the matrix of T .

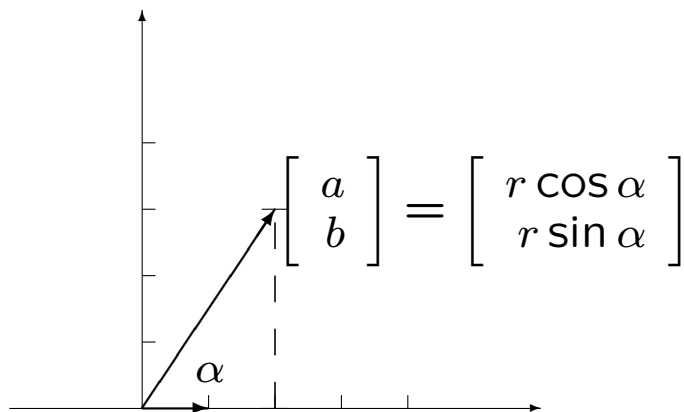
Example. 5 Give a geometric interpretation of the linear transformation.

$$T(\vec{x}) = \begin{bmatrix} a & -b \\ b & a \end{bmatrix} \vec{x}$$

Rotation-dilations A matrix with this form

$$\begin{bmatrix} a & -b \\ b & a \end{bmatrix}$$

denotes a counterclockwise rotation through the angle α followed by a dilation by the factor r where $\tan(\alpha) = \frac{b}{a}$ and $r = \sqrt{a^2 + b^2}$. Geometrically,



[Shears]

Example. 6 Consider the linear transformation

$$T(\vec{x}) = \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix} \vec{x}$$

To understand this transformation, sketch the image of the **unit square**.

Solution The transformation $T(\vec{x}) = \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix} \vec{x}$ is called a *shear* parallel to the x_1 -axis.

Definition. Shear Let L be a line in \mathbb{R}^2 . A linear transformation T from \mathbb{R}^2 to \mathbb{R}^2 is called a **shear parallel to L** if

a. $T(\vec{v}) = \vec{v}$, for all vectors \vec{v} on L , and

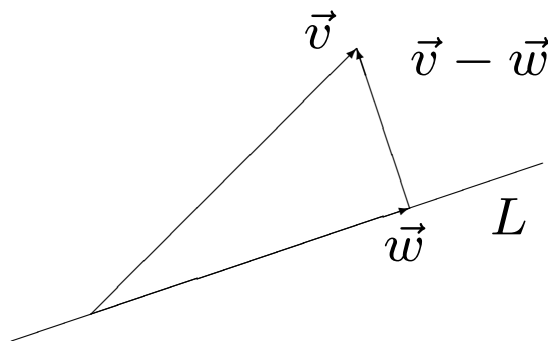
b. $T(\vec{x}) - \vec{x}$ is parallel to L for all vectors $\vec{x} \in \mathbb{R}^2$.

Example. 7 Consider two perpendicular vectors \vec{u} and \vec{w} in \mathbb{R}^2 . Show that the transformation

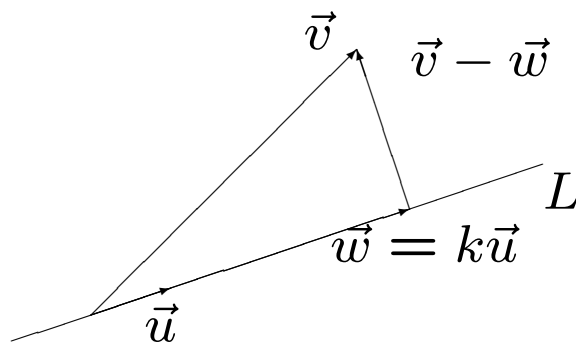
$$T(\vec{x}) = \vec{x} + (\vec{u} \cdot \vec{x})\vec{w}$$

is a shear parallel to the line L spanned by \vec{w} .

Consider a line L in R^2 . For any vector \vec{v} in R^2 , there is a unique vector \vec{w} on L such that $\vec{v} - \vec{w}$ is perpendicular to L .



How can we generalize the idea of an orthogonal projection to lines in R^n ?



Definition. orthogonal projection Let L be a line in R^n consisting of all scalar multiples of some unit vector \vec{u} . For any vector \vec{v} in R^n there is a unique vector \vec{w} on L such that $\vec{v} - \vec{w}$ is perpendicular to L , namely, $\vec{w} = (\vec{u} \cdot \vec{v})\vec{u}$. This vector \vec{w} is called the **orthogonal projection** of \vec{v} onto L :

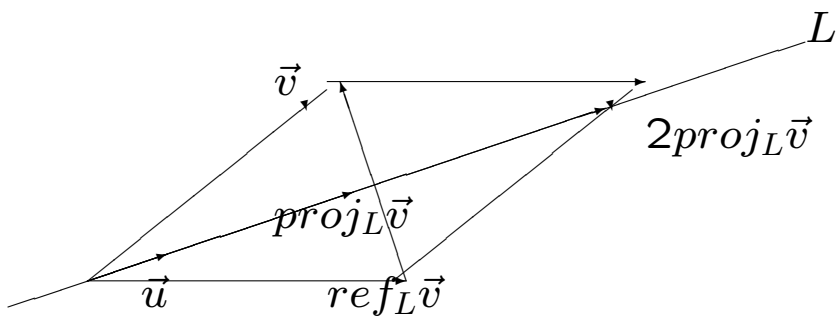
$$\text{proj}_L(\vec{v}) = (\vec{u} \cdot \vec{v})\vec{u}$$

The transformation proj_L from R^n to R^n is linear.

Definition. Let L be a line in \mathbb{R}^n , the vector $2(\text{proj}_L \vec{v}) - \vec{v}$ is called the **reflection** of \vec{v} in L :

$$\text{ref}_L(\vec{v}) = 2(\text{proj}_L \vec{v}) - \vec{v} = 2(\vec{u} \cdot \vec{v})\vec{u} - \vec{v}$$

where \vec{u} is a unit vector on L .



Homework. Exercise 2.2: 1, 9, 13, 17, 27

2.3 The Inverse Of a Linear Transformation

Definition. *A function T from X to Y is called invertible if the equation $T(x)=y$ has a **unique solution x** in X for each y in Y .*

Denote the inverse of T as T^{-1} from Y to X , and write

$$T^{-1}(y)=(\text{the unique } x \text{ in } X \text{ such that } T(x) = y)$$

Note

$$T^{-1}(T(x)) = x, \text{ for all } x \text{ in } X, \text{ and}$$

$$T(T^{-1}(y)) = y, \text{ for all } y \text{ in } Y.$$

If a function T is invertible, then so is T^{-1} ,

$$(T^{-1})^{-1} = T$$

Consider the case of a *linear transformation* from R^n to R^m given by $\vec{y} = A\vec{x}$ where A is an $m \times n$ matrix, the transformation is invertible if the linear system $A\vec{x} = \vec{y}$ has a unique solution.

1. **Case 1:** $m < n$ The system $A\vec{x} = \vec{y}$ has either no solutions or infinitely many solutions, for any \vec{y} in R^m . Therefore $\vec{y} = A\vec{x}$ is noninvertible.
2. **Case 2:** $m = n$ The system $A\vec{x} = \vec{y}$ has a unique solution iff $rref(A) = I_n$, or equivalently, if $rank(A) = n$.
3. **Case 3:** $m > n$ The transformation $\vec{y} = A\vec{x}$ is noninvertible, because we can find a vector \vec{y} in R^m such that the system $A\vec{x} = \vec{y}$ is inconsistent.

Definition. Invertible Matrix A matrix A is called invertible if the linear transformation $\vec{y} = A\vec{x}$ is invertible. The matrix of inverse transformation is denoted by A^{-1} . If the transformation $\vec{y} = A\vec{x}$ is invertible. its inverse is $\vec{x} = A^{-1}\vec{y}$.

Fact

An $m \times n$ matrix A is invertible if and only if

1. A is a square matrix (i.e., $m=n$), and
2. $rref(A) = I_n$.

Example. *Is the matrix A invertible?*

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}.$$

Solution

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \xrightarrow[-7(I)]{-4(I)} \begin{bmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 0 & -6 & -12 \end{bmatrix} \xrightarrow{\div(-3)}$$

$$\begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & -6 & -12 \end{bmatrix} \xrightarrow[-6(II)]{-2(II)} \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{bmatrix}.$$

A fails to be invertible, since $rref(A) \neq I_3$.

Fact Let A be an $n \times n$ matrix.

1. Consider a vector \vec{b} in R^n . If A is invertible, then the system $A\vec{x} = \vec{b}$ has the unique solution $\vec{x} = A^{-1}\vec{b}$. If A is noninvertible, then the system $A\vec{x} = \vec{b}$ has infinitely many solutions or none.
2. Consider the special case when $\vec{b} = \vec{0}$. The system $A\vec{x} = \vec{0}$ has $\vec{x} = \vec{0}$ as a solution. If A is invertible, then this is the only solution. If A is noninvertible, then there are infinitely many other solutions.

If a matrix A is invertible, how can we find the inverse matrix A^{-1} ?

Consider the matrix

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 3 & 2 \\ 3 & 8 & 2 \end{bmatrix}.$$

or, equivalently, the linear transformation

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} x_1 + x_2 + x_3 \\ 2x_1 + 3x_2 + 2x_3 \\ 3x_1 + 8x_2 + 2x_3 \end{bmatrix}.$$

To find the inverse transformation, we solve this system for input variables x_1, x_2, x_3 :

$$\left| \begin{array}{cccc} x_1 + x_2 + x_3 = & y_1 & & \\ 2x_1 + 3x_2 + 2x_3 = & & y_2 & \\ 3x_1 + 8x_2 + 2x_3 = & & & y_3 \end{array} \right| \begin{array}{l} \longrightarrow \\ -2(I) \\ -3(I) \end{array}$$

$$\left| \begin{array}{cccc} x_1 + x_2 + x_3 = & y_1 & & \\ & x_2 = & -2y_1 + y_2 & \\ & 5x_2 - 3x_3 = & -3y_1 + y_3 & \end{array} \right| \begin{array}{l} -(II) \\ \longrightarrow \\ -5(II) \end{array}$$

$$\left| \begin{array}{cccc} x_1 + x_3 = & 3y_1 - y_2 & & \\ x_2 = & -2y_1 + y_2 & & \\ -x_3 = & 7y_1 - 5y_2 + y_3 & & \end{array} \right| \begin{array}{l} \longrightarrow \\ \\ \div(-1) \end{array}$$

$$\left| \begin{array}{cccc} x_1 + x_3 = & 3y_1 - y_2 & & \\ x_2 = & -2y_1 + y_2 & & \\ x_3 = & -7y_1 + 5y_2 - y_3 & & \end{array} \right| \begin{array}{l} -(III) \\ \longrightarrow \\ \end{array}$$

$$\left| \begin{array}{cccc} x_1 = & 10y_1 - 6y_2 + y_3 & & \\ x_2 = & -2y_1 + y_2 & & \\ x_3 = & -7y_1 + 5y_2 - y_3 & & \end{array} \right|.$$

We have found the inverse transformation; its matrix is

$$B = A^{-1} = \begin{bmatrix} 10 & -6 & 1 \\ -2 & 1 & 0 \\ -7 & 5 & -1 \end{bmatrix}.$$

We can write the preceding computations in matrix form:

$$\left[\begin{array}{ccc|ccc} 1 & 1 & 1 & : & 1 & 0 & 0 \\ 2 & 3 & 2 & : & 0 & 1 & 0 \\ 3 & 8 & 2 & : & 0 & 0 & 1 \end{array} \right] \begin{array}{l} \longrightarrow \\ -2(I) \\ -3(I) \end{array}$$

$$\left[\begin{array}{ccc|ccc} 1 & 1 & 1 & : & 1 & 0 & 0 \\ 0 & 1 & 0 & : & -2 & 1 & 0 \\ 0 & 5 & -1 & : & -3 & 0 & 1 \end{array} \right] \begin{array}{l} -(II) \\ \longrightarrow \\ -5(II) \end{array}$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 1 & : & 3 & -1 & 0 \\ 0 & 1 & 0 & : & -2 & 1 & 0 \\ 0 & 0 & -1 & : & 7 & -5 & 1 \end{array} \right] \begin{array}{l} \longrightarrow \\ \div(-1) \end{array}$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 1 & : & 3 & -1 & 0 \\ 0 & 1 & 0 & : & -2 & 1 & 0 \\ 0 & 0 & 1 & : & -7 & 5 & -1 \end{array} \right] \begin{array}{l} -(III) \\ \longrightarrow \end{array}$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & : & 10 & -6 & 1 \\ 0 & 1 & 0 & : & -2 & 1 & 0 \\ 0 & 0 & 1 & : & -7 & 5 & -1 \end{array} \right].$$

This process can be described succinctly as follows:

Find the inverse of a matrix

To find the inverse of an $n \times n$ matrix A , form the $n \times (2n)$ matrix $\left[A \ : \ I_n \right]$ and compute $\text{rref} \left[A \ : \ I_n \right]$.

- If $\text{rref} [A:I_n]$ is of the form $[I_n:B]$, then A is invertible, and $A^{-1} = B$.
- If $\text{rref} [A:I_n]$ is of another form (*i.e.*, *its left half fails to be I_n*), then A is not invertible. (Note that the left half of $\text{rref} [A:I_n]$ is $\text{rref}(A)$.)

The inverse of a 2×2 matrix is particularly easy to find.

Inverse and determinant of a 2×2 matrix

1. The 2×2 matrix $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$

is invertible if (and only if) $ad - bc \neq 0$. Quantity $\boxed{ad - bc}$ is called the determinant of A , written $\det(A)$:

$$\det(A) = \det \begin{bmatrix} a & b \\ c & d \end{bmatrix} = ad - bc.$$

2. If $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ is invertible, then

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = \frac{1}{\det(A)} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}.$$

Compare this with Exercise 2.1.13.

Homework. *Exercise 2.3 21–27, 41*

2.4 MATRIX PRODUCTS

The *composite* of two functions: $y = \sin(x)$ and $z = \cos(y)$ is $z = \cos(\sin(x))$.

Consider two transformation systems:

$$\vec{y} = A\vec{x}, \text{ with } A = \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix}$$

$$\vec{z} = B\vec{y}, \text{ with } B = \begin{bmatrix} 6 & 7 \\ 8 & 9 \end{bmatrix}$$

The composite of the two transformation systems is

$$\vec{z} = B(A\vec{x})$$

Question: Is $\vec{z} = T(\vec{x})$ linear? If so, what's the matrix?

(a) Find the matrix for the composite:

$$\begin{aligned} z_1 &= 6y_1 + 7y_2 & \text{and} & & y_1 &= x_1 + 2x_2 \\ z_2 &= 8y_1 + 9y_2 & & & y_2 &= 3x_1 + 5x_2 \end{aligned}$$

$$\begin{aligned} z_1 &= 6(x_1 + 2x_2) + 7(3x_1 + 5x_2) \\ &= (6 \cdot 1 + 7 \cdot 3)x_1 + (6 \cdot 2 + 7 \cdot 5)x_2 \\ &= 27x_1 + 47x_2 \end{aligned}$$

$$\begin{aligned} z_2 &= 8(x_1 + 2x_2) + 9(3x_1 + 5x_2) \\ &= (8 \cdot 1 + 9 \cdot 3)x_1 + (8 \cdot 2 + 9 \cdot 5)x_2 \\ &= 35x_1 + 61x_2 \end{aligned}$$

This shows the composite is linear with matrix

$$\begin{bmatrix} 6 \cdot 1 + 7 \cdot 3 & 6 \cdot 2 + 7 \cdot 5 \\ 8 \cdot 1 + 9 \cdot 3 & 8 \cdot 2 + 9 \cdot 5 \end{bmatrix} = \begin{bmatrix} 27 & 47 \\ 35 & 61 \end{bmatrix}$$

(b) Use Fact to show the transformation $T(\vec{x}) = B(A\vec{x})$ is linear:

$$T(\vec{v} + \vec{w}) = B(A(\vec{v} + \vec{w})) = B(A\vec{v} + A\vec{w}) = B(A\vec{v}) + B(A\vec{w}) = T(\vec{v}) + T(\vec{w})$$

$$T(k\vec{v}) = B(A(k\vec{v})) = B(k(A\vec{v})) = k(B(A\vec{v})) = kT(\vec{v})$$

Once we know that T is linear, we can find its matrix by computing the vectors: $T(\vec{e}_1)$ and $T(\vec{e}_2)$:

$$T(\vec{e}_1) = B(A(\vec{e}_1)) = B(\text{first column of } A) = \begin{bmatrix} 6 & 7 \\ 8 & 9 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 27 \\ 35 \end{bmatrix}$$

$$T(\vec{e}_2) = B(A(\vec{e}_2)) = B(\text{second column of } A) = \begin{bmatrix} 6 & 7 \\ 8 & 9 \end{bmatrix} \begin{bmatrix} 2 \\ 5 \end{bmatrix} = \begin{bmatrix} 47 \\ 61 \end{bmatrix}$$

The matrix of $T(\vec{x}) = B(A\vec{x}) = BA(\vec{x})$:

$$= \begin{bmatrix} \left. \begin{array}{c} | \\ T(\vec{e}_1) \\ | \end{array} \right. & \left. \begin{array}{c} | \\ T(\vec{e}_2) \\ | \end{array} \right. \end{bmatrix} = \begin{bmatrix} 27 & 47 \\ 35 & 61 \end{bmatrix}$$

Definition. Matrix multiplication

1. Let B be an $m \times n$ matrix and A a $q \times p$ matrix. The product BA is defined if (and only if) $n = q$.
2. If B is an $m \times n$ matrix and A an $n \times p$ matrix, then the product BA is defined as the matrix of the linear transformation $T(\vec{x}) = B(A\vec{x})$. This means that $T(\vec{x}) = B(A\vec{x}) = (BA)\vec{x}$, for all \vec{x} in R^p . The product BA is an $m \times p$ matrix.

Let B be an $m \times n$ matrix and A an $n \times p$ matrix. Let's think about the columns of the matrix BA :

$$\begin{aligned}
 & (\textit{i} \text{th columns of } BA) \\
 &= (BA)\vec{e}_i \\
 &= B(A\vec{e}_i) \\
 &= B(\textit{i} \text{th column of } A).
 \end{aligned}$$

If we denote the columns of A by $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_p$, we can write

$$BA = B \underbrace{\begin{bmatrix} | & | & \dots & | \\ \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_p \\ | & | & & | \end{bmatrix}}_A = \begin{bmatrix} | & | & \dots & | \\ B\vec{v}_1 & B\vec{v}_2 & \dots & B\vec{v}_p \\ | & | & & | \end{bmatrix}.$$

The matrix product, column by column

Let B be an $m \times n$ matrix and A an $n \times p$ matrix with columns $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_p$. Then, the product BA is

$$BA = B \begin{bmatrix} | & | & \dots & | \\ \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_p \\ | & | & \dots & | \end{bmatrix} = \begin{bmatrix} | & | & \dots & | \\ B\vec{v}_1 & B\vec{v}_2 & \dots & B\vec{v}_p \\ | & | & \dots & | \end{bmatrix}.$$

To find BA , we can multiply B with the columns of A and combine the resulting vectors.

Fact Matrix multiplication is noncommutative: $AB \neq BA$, in general. However, at times it does happen that $AB = BA$; then, we say that the matrices A and B commute.

The matrix product, entry by entry

Let B be an $m \times n$ matrix and A an $n \times p$ matrix. The ij th entry of BA is the dot product of the i th row of B and the j th column of A .

$$\begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ b_{21} & b_{22} & \cdots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{i1} & b_{i2} & \cdots & b_{in} \\ \vdots & \vdots & \ddots & \vdots \\ b_{m1} & b_{m2} & \cdots & b_{mn} \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1j} & \cdots & a_{1p} \\ a_{21} & a_{22} & \cdots & a_{2j} & \cdots & a_{2p} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nj} & \cdots & a_{np} \end{bmatrix}$$

is the $m \times p$ matrix whose ij th entry is

$$b_{i1}a_{1j} + b_{i2}a_{2j} + \cdots + b_{in}a_{nj} = \sum_{k=1}^n b_{ik}a_{kj}.$$

Example. $\begin{bmatrix} 6 & 7 \\ 8 & 9 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix} =$

$$\begin{bmatrix} 6 \cdot 1 + 7 \cdot 3 & 6 \cdot 2 + 7 \cdot 5 \\ 8 \cdot 1 + 9 \cdot 3 & 8 \cdot 2 + 9 \cdot 5 \end{bmatrix} = \begin{bmatrix} 27 & 47 \\ 35 & 61 \end{bmatrix}.$$

We have done these computations before. (where?)

Matrix Algebra

Fact For an invertible $n \times n$ matrix A .

$$AA^{-1} = I_n \text{ and } A^{-1}A = I_n.$$

Fact For an $m \times n$ matrix A .

$$AI_n = I_m A = A.$$

Fact Matrix multiplication is associative

$$(AB)C = A(BC).$$

We can write simply ABC for the product $(AB)C = A(BC)$.

Proof (a) $(AB)C = (AB)[\vec{v}_1 \ \vec{v}_2 \ \cdots \ \vec{v}_q]$
 $= [(AB)\vec{v}_1 \ (AB)\vec{v}_2 \ \cdots \ (AB)\vec{v}_q]$

and

$$A(BC) = A[B\vec{v}_1 \ B\vec{v}_2 \ \cdots \ B\vec{v}_q]$$

$$= [A(B\vec{v}_1) \ A(B\vec{v}_2) \ \cdots \ A(B\vec{v}_q)]$$

Since $(AB)\vec{v}_i = A(B\vec{v}_i)$, by definition of the matrix product, we find that $(AB)C = A(BC)$.

Proof (b) Consider two linear transformations

$$T(\vec{x}) = ((AB)C)\vec{x}$$

and

$$L(\vec{x}) = (A(BC))\vec{x}$$

are identical because,

$$T(\vec{x}) = ((AB)C)\vec{x} = (AB)(C\vec{x}) = A(B(C\vec{x}))$$

and

$$L(\vec{x}) = (A(BC))\vec{x} = A((BC)\vec{x}) = A(B(C\vec{x}))$$

If A and B are invertible $n \times n$ matrices, is BA invertible?

$$\vec{y} = BA\vec{x}$$

multiply both sides by B^{-1}

$$B^{-1}\vec{y} = B^{-1}BA\vec{x} = I_n A\vec{x} = A\vec{x}$$

next, multiply both sides by A^{-1}

$$A^{-1}B^{-1}\vec{y} = A^{-1}A\vec{x} = I_n \vec{x} = \vec{x}$$

This computation shows that the linear transformation is invertible since

$$\vec{x} = A^{-1}B^{-1}\vec{y}$$

Fact The inverse of a product of matrices

If A and B are invertible $n \times n$ matrices, then BA is invertible as well, and

$$(BA)^{-1} = A^{-1}B^{-1}.$$

Pay attention to the order of the matrices.

Proof Verify it by yourself.

Fact Let A and B be two $n \times n$ matrices such that

$$BA = I_n.$$

Then,

- a. A and B are both invertible.
- b. $A^{-1} = B$ and $B^{-1} = A$, and
- c. $AB = I_n$.

Proof (a) To demonstrate A is invertible it suffices to show that the linear system $A\vec{x} = \vec{0}$ has only the solution $\vec{x} = \vec{0}$.

$$BA\vec{x} = B\vec{0} = \vec{0}$$

(b) $B = A^{-1}$ since

$$(BA)A^{-1} = (I_n)A^{-1} = A^{-1}$$

and

$$B^{-1} = (A^{-1})^{-1} = A$$

(c) $AB = AA^{-1} = I_n$

Example. $B = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$ is the inverse of $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$.

it suffices to verify that $BA = I_2$:

$$\begin{aligned} BA &= \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} \\ &= \frac{1}{ad-bc} \begin{bmatrix} ad - bc & bd - bd \\ ac - ac & ad - bc \end{bmatrix} = I_2. \end{aligned}$$

Example.

Suppose A , B and C are three $n \times n$ matrices and $ABC = I_n$. Show that B is invertible, and express B^{-1} in term of A and C .

Solution

Write $ABC = (AB)C = I_n$. We have $C(AB) = I_n$. Since matrix multiplication is associative, we can write $(CA)B = I_n$. We conclude that B is invertible, and $B^{-1} = CA$.

Distributive property for matrices

Fact If A, B are $n \times n$, and C, D are $n \times p$ matrices, then

$$A(C + D) = AC + AD$$

and

$$(A + B)C = AC + BC.$$

Fact If A is an $m \times n$ matrix, B an $n \times p$ matrix, and k a scalar, then

$$(kA)B = A(kB) = k(AB).$$

Partitioned Matrices

It is sometimes useful to break a large matrix down into smaller submatrices by slicing it up with horizontal or vertical lines that go all the way through the matrix.

For example, we can think of the 4×4 matrix

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 8 & 7 & 6 \\ 5 & 4 & 3 & 2 \end{bmatrix}$$

as a 2×2 matrix whose "entries" are four 2×2 matrices:

$$A = \left[\begin{array}{cc|cc} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ \hline 9 & 8 & 7 & 6 \\ 5 & 4 & 3 & 2 \end{array} \right] = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix},$$

with $A_{11} = \begin{bmatrix} 1 & 2 \\ 5 & 7 \end{bmatrix}$, $A_{12} = \begin{bmatrix} 3 & 4 \\ 7 & 8 \end{bmatrix}$, etc.

The submatrices in such a partition need not be of equal size; for example, we could have

$$B = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} = \left[\begin{array}{cc|c} 1 & 2 & 3 \\ 4 & 5 & 6 \\ \hline 7 & 8 & 9 \end{array} \right] = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}.$$

A useful property of partitioned matrices is the following:

Multiplying partitioned matrices Partitioned matrices can be multiplied as though the submatrices were scalars:

$$AB =$$

$$\begin{bmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ A_{21} & A_{22} & \dots & A_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ A_{i1} & A_{i2} & \dots & A_{in} \\ \vdots & \vdots & \ddots & \vdots \\ A_{m1} & A_{m2} & \dots & A_{mn} \end{bmatrix} \begin{bmatrix} B_{11} & B_{12} & \dots & B_{1j} & \dots & B_{1p} \\ B_{21} & B_{22} & \dots & B_{2j} & \dots & B_{2p} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ B_{n1} & B_{n2} & \dots & B_{nj} & \dots & B_{np} \end{bmatrix}$$

is the partitioned matrix whose ij th "entry" is the matrix

$$A_{i1}B_{1j} + A_{i2}B_{2j} + \dots + A_{in}B_{nj} = \sum_{k=1}^n A_{ik}B_{kj},$$

provided that all the products $A_{ik}B_{kj}$ are defined.

Example.

$$\begin{aligned} & \left[\begin{array}{cc|c} 0 & 1 & -1 \\ 1 & 0 & 1 \end{array} \right] \left[\begin{array}{cc|c} 1 & 2 & 3 \\ 4 & 5 & 6 \\ \hline 7 & 8 & 9 \end{array} \right] \\ &= \left[\left[\begin{array}{cc} 0 & 1 \\ 1 & 0 \end{array} \right] \left[\begin{array}{cc} 1 & 2 \\ 4 & 5 \end{array} \right] + \left[\begin{array}{c} -1 \\ 1 \end{array} \right] [7 \ 8] \mid \left[\begin{array}{cc} 0 & 1 \\ 1 & 0 \end{array} \right] \left[\begin{array}{c} 3 \\ 6 \end{array} \right] + \left[\begin{array}{c} -1 \\ 1 \end{array} \right] [9] \right] \\ &= \left[\begin{array}{cc|c} -3 & -3 & -3 \\ 8 & 10 & 12 \end{array} \right]. \end{aligned}$$

Compute this product without using a partition, and see whether you find the same result.

Example.

$$A = \begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix},$$

where A_{11} is an $n \times n$ matrix, A_{22} is an $m \times m$ matrix, and A_{12} is an $n \times m$ matrix.

- a. For which choices of A_{11} , A_{12} , and A_{22} is A invertible ?
- b. If A is invertible, what is A^{-1} (in terms of A_{11} , A_{12} , A_{22})?

Solution

We are looking for an $(n + m) \times (n + m)$ matrix B such that

$$BA = I_{n+m} = \begin{bmatrix} I_n & 0 \\ 0 & I_m \end{bmatrix}.$$

Let us partition B in the same way as A :

$$B = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix},$$

where B_{11} is $n \times n$, B_{22} is $m \times m$, etc. The fact that B is the inverse of A means that

$$\begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} \begin{bmatrix} A_{11} & A_{12} \\ 0 & B_{22} \end{bmatrix} = \begin{bmatrix} I_n & 0 \\ 0 & I_m \end{bmatrix},$$

or using

$$\begin{vmatrix} B_{11}A_{11} & = & I_n \\ B_{11}A_{12} + B_{12}A_{22} & = & 0 \\ B_{21}A_{11} & = & 0 \\ B_{21}A_{12} + B_{22}A_{22} & = & I_m \end{vmatrix}.$$

We have to solve this system for the submatrices B_{ij} .

1. By Equation 1, A_{11} must be invertible, and $B_{11} = A_{11}^{-1}$.
2. By Equation 3, $B_{21} = 0$ (Multiply by A_{11}^{-1} from the right)
3. Equation 4 now simplifies to $B_{22}A_{22} = I_m$. Therefore, A_{22} must be invertible, and $B_{22} = A_{22}^{-1}$.
4. Lastly, Solve for B_{12} by Equation 2

$$A_{11}^{-1}A_{12} + B_{12}A_{22} = 0$$

$$\Rightarrow B_{12}A_{22} = -A_{11}^{-1}A_{12}$$

$$\Rightarrow B_{12} = -A_{11}^{-1}A_{12}A_{22}^{-1}$$

So

a. A is invertible if (and only if) both A_{11} and A_{22} are invertible (no condition is imposed on A_{12}).

b. If A is invertible, then its inverse is

$$A^{-1} = \begin{bmatrix} A_{11}^{-1} & -A_{11}^{-1}A_{12}A_{22}^{-1} \\ 0 & A_{22}^{-1} \end{bmatrix}.$$

Verify this result for the following example:

Example. 5

$$\left[\begin{array}{cc|ccc} 1 & 1 & 1 & 2 & 3 \\ 1 & 2 & 4 & 5 & 6 \\ \hline 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{array} \right]^{-1} = \left[\begin{array}{cc|ccc} 2 & -1 & 2 & 1 & 0 \\ -1 & 1 & -3 & -3 & -3 \\ \hline 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{array} \right].$$

Homework.

Exercise 2.4: 5, 13, 17, 23, 27, 35