MAT 260 LINEAR ALGEBRA LECTURE 22

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1.4 — Algebraic properties of matrix operations

Assuming that the sizes of the matrices are such that the indicated operations can be performed, we have the following properties in matrix arithmetic.

- 1. A + B = B + A (commutative law for addition)
- 2. A + (B + C) = (A + B) + C (associative law for addition)
- 3. A(BC) = (AB)C (associative law for multiplication)
- 4. A(B+C) = AB + AC (left distributive law for multiplication)
- 5. (A+B)C = AC + BC (right distributive law for multiplication)
- 6. k(AB) = (kA)B = A(kB) (properties of scalar multiplication)
- 7. $k(\ell A) = (k\ell)A$ (property of scalar multiplication)
- 8. k(A+B) = kA + kB (distributive law for scalar multiplication and matrix addition)
- 9. $(k+\ell)A = kA + \ell A$ (distributive law for scalar multiplication and scalar addition)

The zero matrix, i.e. a matrix with all the entries 0, is often denoted by O. It is the additive identity for matrix addition. By that, we mean

$$A + O = A$$
 and $O + A = A$

for all A such that A and O are of the same size. Note that 0A = O (0 is a scalar) since scalar multiplication is entry-wise. In fact, if kA = O, then either k = 0 or A = O.

Warning: If AB = O, it does not mean that A = O or B = O. (Recall the example from last lecture.) As a result, the cancellation law does not work, i.e. AB = AC and $A \neq O$ may not imply that B = C, since A(B - C) = O and $A \neq O$ does not imply that B - C = O.

The square matrices

$$(1), \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}, \dots$$

are multiplicative identities for matrix multiplication, denoted by I_n and are called the **identity matrix** of order n. This means that for all matrix A of dimensions $r \times c$,

$$AI_c = I_r A = A$$
.

Theorem 1. If A is a square matrix of order n, then the reduced row echelon form of A is either I_n or has a zero row.

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Given a square matrix A of order n, B is the **multiplicative inverse** of A if

$$AB = BA = I_n$$
.

This is an analogue of how we deal with real numbers. 1 is the multiplicative identity among real numbers, and a multiplicative inverse of a nonzero number k is $\frac{1}{k}$ since $k \cdot \frac{1}{k} = \frac{1}{k} \cdot k = 1$.

Theorem 2. If B is a multiplicative inverse of A, then it is unique.

Proof. Assume that we have another matrix C such that $AC = CA = I_n$, then

$$B = BI_n = B(AC) = (BA)C = I_nC = C.$$

By uniqueness, we can safely write that $B = A^{-1}$ if B is a multiplicative inverse of A.

Let A be a square matrix of order n. If A has a multiplicative inverse, then A is called **invertible** or **nonsingular**; otherwise, A is called **singular**. Note that if A is invertible, then A^{-1} is also invertible, and the inverse of A^{-1} is A. Hence, A and A^{-1} are inverses of each other.

If A is a square matrix with a zero row, then it is always singular. This is because AB also has a zero row in the same position for all matrices B of the same size. Similarly, if A is a square matrix with a zero column, then it is singular.

Recall that when we have a system of linear equations, it can be written as the matrix equation

$$A \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix}.$$

If A is an invertible square matrix, then

$$\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = A^{-1} \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix}.$$

In general, we would always like to detect which square matrix is invertible or not.

Theorem 3. The matrix

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

is invertible if and only if $ad - bc \neq 0$. Furthermore, if A is invertible, then

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

Theorem 4. If both A and B are invertible matrices of order n, then the matrix AB is also invertible, and $(AB)^{-1} = B^{-1}A^{-1}$.

Question. How about $(ABC)^{-1}$?

Warning: A + B may not be invertible even if A and B are invertible matrices of order n.

If A is a square matrix of order n, then we define $A^0 = I_n$, and $A^s = A \cdot A \cdot \cdots \cdot A$ (s copies). Moreover, if A is invertible, by Theorem 4, we have $(A^2)^{-1} = (A^{-1})^2$. Hence, we can define $A^{-s} = A^{-1} \cdot A^{-1} \cdot \cdots \cdot A^{-1}$ (s copies). We also have the index law $A^s A^t = A^{s+t}$ and $(A^s)^t = A^{st}$. Finally, $(kA)^{-1} = k^{-1}A^{-1}$ if $k \neq 0$.

With all these exponents of A defined, we can have a **matrix polynomial** in A. For example, if $p(x) = a_0 + a_1x + a_2x^2 + \cdots + a_sx^s$ is an s degree polynomial in x, then $p(A) = a_0I_n + a_1A + a_2A^2 + \cdots + a_sA^s$. Since we are only dealing with exponents of A, we have $p_1(A)p_2(A) = p_2(A)p_1(A)$, where $p_1(x)$ and $p_2(x)$ are polynomials in x.

However, be very careful when we deal with polynomials involving two or more matrices. For instance, $(A + B)^2 \neq A^2 + 2AB + B^2$ in general. Rather, it is $A^2 + AB + BA + B^2$.

Theorem 5. If A is an invertible matrix, then $(A^{-1})^{\top} = (A^{\top})^{-1}$.