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2.8 Consequences of Invertibility

"TFAE"

Memorize!



Theorem — The Really Big Theorem on Invertibility:

The following conditions are equivalent for a linear operator $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$, with standard matrix $[T] = A$:

1. T is an invertible operator.
2. A is an invertible matrix.
3. The rref of A is I_n .
4. A is the product of elementary matrices.
5. T is one-to-one.
6. $\ker(T) = \text{nullspace}(A) = \{\vec{0}_n\}$.
7. $\text{nullity}(T) = \text{nullity}(A) = 0$.
8. T is onto.
9. $\text{range}(T) = \mathbb{R}^n$.
10. $\text{rank}(T) = n$.

Dim Th

$$T : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$\text{rank}(T) + \text{nullity}(T) = n$$

$$\dim(\text{range}(T)) + \dim(\ker(T)) = n$$

$\rightarrow \text{range}(T)$

11. $\text{colspace}(A) = \mathbb{R}^n$.
12. The columns of A are linearly independent.
13. The columns of A Span \mathbb{R}^n .
14. The columns of A form a basis for \mathbb{R}^n .
15. $\text{rowspan}(A) = \mathbb{R}^n$.
16. The rows of A are linearly independent.
17. The rows of A Span \mathbb{R}^n .
18. The rows of A form a basis for \mathbb{R}^n .
19. The homogeneous equation $A\vec{x} = \vec{0}_n$ has only the trivial solution.
20. *For every $n \times 1$ matrix \vec{b} , the system $A\vec{x} = \vec{b}$ is **consistent**.*
21. *For every $n \times 1$ matrix \vec{b} , the system $A\vec{x} = \vec{b}$ has **exactly one solution**.*
22. *There exists an $n \times 1$ matrix \vec{b} , such that the system $A\vec{x} = \vec{b}$ has **exactly one solution**.*

Read and Study this proof from the book! Some of the directions are big results and some of them are definitions. Some of them are easy directions. I showed a few of them in class: (7) \Leftrightarrow (10)

One Sided Inverses

Given A , we must find B so that:

$$AB = I_n \text{ and } BA = I_n$$

If B only satisfies the first equation, we call B a “right” inverse for A .

If B only satisfies the second equation we call B a “left” inverse for A .

Luckily, there's no need for this nonsense:

Theorem: An $n \times n$ matrix A is invertible if and only if we can find an $n \times n$ matrix B such that $AB = I_n$ or $BA = I_n$. Thus, a “right” inverse is also a “left” inverse, and vice versa.

This is subtle and deep result. It has a nice proof too! See the extra notes.

Proof: think of BA as a matrix representing the composition of two operators.

The Inverse of a Composition and Matrix Product

Theorem: If $T_1, T_2 : \mathbb{R}^n \rightarrow \mathbb{R}^n$ are both invertible operators, then $T_2 \circ T_1$ is also invertible, and furthermore:

$$[T_2 \circ T_1]^{-1} = [T_1]^{-1}[T_2]^{-1}.$$

Analogously, if A and B are invertible $n \times n$ matrices, then AB is also invertible, and furthermore:

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

Another key result in Linear Algebra!

The converse is also true!

By the "Equivalence of LT & Mat Thm" we immediately get:

Theorem: If $T_1, T_2 : \mathbb{R}^n \rightarrow \mathbb{R}^n$ are operators and the *composition* $T_2 \circ T_1$ is *invertible*, then *both* T_2 and T_1 are also invertible. Analogously if A and B are two $n \times n$ matrices and the *product* AB is *invertible*, then *both* A and B are invertible.