

6.3. KERNEL AND RANGE

Kernel of a linear transformation

- Kernel tells you how much is eliminated.

Definition 6.3.1 If $T : R^n \rightarrow R^m$ is a linear transformation, then the set of vectors in R^n that T maps into $\mathbf{0}$ is called the *kernel* of T and is denoted by $\ker(T)$.

- Example:
 - O-operator: Then R^n is the kernel.
 - Identity operator: $\{0\}$ is the kernel.
 - Orthogonal projection to a plane: the perpendicular line through the origin.

Theorem 6.3.2 *If $T : R^n \rightarrow R^m$ is a linear transformation, then the kernel of T is a subspace of R^n .*

- ⊙ Proof: We can do scalar multiplications and vector additions in the kernel.
- ⊙ The kernel of a matrix transformation T_A is the set of x such that $Ax=0$.

Theorem 6.3.3 *If A is an $m \times n$ matrix, then the kernel of the corresponding linear transformation is the solution space of $Ax = 0$.*

Definition 6.3.4 *If A is an $m \times n$ matrix, then the solution space of the linear system $Ax = 0$, or, equivalently, the kernel of the transformation T_A , is called the **null space** of the matrix A and is denoted by $\text{null}(A)$.*

Theorem 6.3.5 *If $T : R^n \rightarrow R^m$ is a linear transformation, then T maps subspaces of R^n into subspaces of R^m .*

- ⦿ Proof: This follows from the fact that T preserves additions and scalar multiplications.

Range of a linear transformation

Definition 6.3.6 If $T : R^n \rightarrow R^m$ is a linear transformation, then the *range* of T , denoted by $\text{ran}(T)$, is the set of all vectors in R^m that are images of at least one vector in R^n . Stated another way, $\text{ran}(T)$ is the image of the domain R^n under the transformation T .

⦿ Examples:

- For 0-operator: Range is $\{O\}$.
- For Id: the range is R^m .
- For orthogonal projections to a plane P : the range is the plane P .

Theorem 6.3.7 If $T : R^n \rightarrow R^m$ is a linear transformation, then $\text{ran}(T)$ is a subspace of R^m .

Range of a matrix transformation

- ⊙ A $m \times n$ matrix
- ⊙ $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$.
- ⊙ $T_A(x) = Ax$.

Theorem 6.3.8 *If A is an $m \times n$ matrix, then the range of the corresponding linear transformation is the column space of A .*

- ⊙ See Example 5.
- ⊙ Example 6: To check whether some vector is in the range.

Existence and Uniqueness

- ⦿ Existence question: Is every vector in the codomain of T in the range? (If not, which subspace is the range.)
- ⦿ Uniqueness question: Can two vectors map to a same vector under T ?

Definition 6.3.9 A transformation $T : R^n \rightarrow R^m$ is said to be *onto* if its range is the entire codomain R^m ; that is, every vector in R^m is the image of at least one vector in R^n .

Definition 6.3.10 A transformation $T : R^n \rightarrow R^m$ is said to be *one-to-one* (sometimes written 1-1) if T maps distinct vectors in R^n into distinct vectors in R^m .

- ⦿ Example: A rotation in \mathbb{R}^2 .
 - This is one-to-one since it has a nonsingular matrix.
 - This is also onto since the matrix has an inverse.
- ⦿ Example: An orthogonal projection to a plane.
 - This is not one-to-one since many vectors go to O .
 - This is not onto since P is not all of the codomain.
- ⦿ See Examples 9 and 10.

Theorem 6.3.11 *If $T : R^n \rightarrow R^m$ is a linear transformation, then the following statements are equivalent.*

(a) *T is one-to-one.*

(b) $\ker(T) = \{\mathbf{0}\}$.

- ⦿ Proof: (a) \rightarrow (b). $T(\mathbf{0})=\mathbf{0}$. If $T(x)=\mathbf{0}$, then $x=\mathbf{0}$ since T is one-to-one. Thus $\text{Ker}(T)=\{\mathbf{0}\}$.
- ⦿ (b) \rightarrow (a). Suppose x_1 is not x_2 . If $T(x_1)=T(x_2)$, then $T(x_1-x_2)=\mathbf{0}$. Thus, $x_1-x_2=\mathbf{0}$ as $\ker(T)=\{\mathbf{0}\}$. Therefore $x_1=x_2$.

One to one and onto from linear systems.

- ⊙ $T_A(x)=0 \leftrightarrow Ax=0.$
- ⊙ $T_A(x)=b \leftrightarrow Ax = b.$

Theorem 6.3.12 *If A is an $m \times n$ matrix, then the corresponding linear transformation $T_A: R^n \rightarrow R^m$ is one-to-one if and only if the linear system $A\mathbf{x} = \mathbf{0}$ has only the trivial solution.*

Theorem 6.3.13 *If A is an $m \times n$ matrix, then the corresponding linear transformation $T_A: R^n \rightarrow R^m$ is onto if and only if the linear system $A\mathbf{x} = \mathbf{b}$ is consistent for every \mathbf{b} in R^m .*

- ⊙ These are solvable questions.

Theorem 6.3.14 *If $T : R^n \rightarrow R^n$ is a linear operator on R^n , then T is one-to-one if and only if it is onto.*

- ⊙ Proof: Theorem 4.4.7 (d) and (e) are equivalent. (d) \leftrightarrow one-to-one
- ⊙ (e) \leftrightarrow onto.

Theorem 6.3.15 *If A is an $n \times n$ matrix, and if T_A is the linear operator on R^n with standard matrix A , then the following statements are equivalent.*

- (a) *The reduced row echelon form of A is I_n .*
- (b) *A is expressible as a product of elementary matrices.*
- (c) *A is invertible.*
- (d) *$A\mathbf{x} = \mathbf{0}$ has only the trivial solution.*
- (e) *$A\mathbf{x} = \mathbf{b}$ is consistent for every vector \mathbf{b} in R^n .*
- (f) *$A\mathbf{x} = \mathbf{b}$ has exactly one solution for every vector \mathbf{b} in R^n .*
- (g) *The column vectors of A are linearly independent.*
- (h) *The row vectors of A are linearly independent.*
- (i) *$\det(A) \neq 0$.*
- (j) *$\lambda = 0$ is not an eigenvalue of A .*
- (k) *T_A is one-to-one.*
- (l) *T_A is onto.*