

## 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:  $\{O\}$  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  $A\mathbf{x} = \mathbf{0}$ .*

**Definition 6.3.4** *If  $A$  is an  $m \times n$  matrix, then the solution space of the linear system  $A\mathbf{x} = \mathbf{0}$ , 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  $T(\mathbf{x})=\mathbf{0}$ , then  $\mathbf{x}=\mathbf{0}$  since  $T$  is one-to-one. Thus  $\text{Ker}(T)=\{\mathbf{0}\}$ .
- ⦿ (b) $\rightarrow$ (a). Suppose  $\mathbf{x}_1$  is not  $\mathbf{x}_2$ . If  $T(\mathbf{x}_1)=T(\mathbf{x}_2)$ , then  $T(\mathbf{x}_1-\mathbf{x}_2)=\mathbf{0}$ . Thus,  $\mathbf{x}_1-\mathbf{x}_2=\mathbf{0}$  as  $\ker(T)=\{\mathbf{0}\}$ . Therefore  $\mathbf{x}_1=\mathbf{x}_2$ .

# 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.*