

Lecture Summary: Image and Kernel of Linear Transformations

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Key Points

- **Definition of Kernel:**

- For a linear transformation $f : V \rightarrow W$, the **kernel** of f is defined as:

$$\ker(f) = \{v \in V \mid f(v) = 0\}.$$

- The kernel is a subspace of V :

- * Closure under addition: If $v_1, v_2 \in \ker(f)$, then $v_1 + v_2 \in \ker(f)$.
- * Closure under scalar multiplication: If $v \in \ker(f)$ and $\alpha \in \mathbb{R}$, then $\alpha v \in \ker(f)$.

- **Definition of Image:**

- For $f : V \rightarrow W$, the **image** of f (also called the range) is:

$$\text{Im}(f) = \{w \in W \mid \exists v \in V, w = f(v)\}.$$

- The image is a subspace of W :

- * Closure under addition: If $w_1, w_2 \in \text{Im}(f)$, then $w_1 + w_2 \in \text{Im}(f)$.
- * Closure under scalar multiplication: If $w \in \text{Im}(f)$ and $\alpha \in \mathbb{R}$, then $\alpha w \in \text{Im}(f)$.

- **Kernel and Injectivity:**

- A linear transformation f is injective if and only if $\ker(f) = \{0\}$ (the zero subspace).

- **Image and Surjectivity:**

- A linear transformation f is surjective if and only if $\text{Im}(f) = W$.

- **Relation to Matrices:**

- For $f : V \rightarrow W$, let $\beta = \{v_1, \dots, v_n\}$ and $\gamma = \{w_1, \dots, w_m\}$ be ordered bases for V and W , respectively.
- The corresponding matrix A is defined by expressing $f(v_j)$ as a linear combination of $\{w_1, \dots, w_m\}$:

$$f(v_j) = \sum_{i=1}^m a_{ij} w_i.$$

- The j th column of A consists of the coefficients a_{ij} .

- **Kernel and Null Space:**

- $\ker(f)$ corresponds to the null space of A , which is:

$$\text{Null}(A) = \{x \in \mathbb{R}^n \mid Ax = 0\}.$$

- The kernel is isomorphic to the null space of the associated matrix A .
- **Image and Column Space:**
 - $\text{Im}(f)$ corresponds to the column space of A , which is the span of the columns of A .
 - Finding a basis for $\text{Im}(f)$ involves identifying pivot columns of A in row-reduced form.
- **Summary of Isomorphisms:**
 - $\ker(f)$ is isomorphic to the null space of A .
 - $\text{Im}(f)$ is isomorphic to the column space of A .

Simplified Explanation

Example 1: Linear Transformation on \mathbb{R}^2 Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be defined as:

$$f(x, y) = (2x, y).$$

- Kernel: Solve $f(x, y) = (0, 0)$:

$$(2x, y) = (0, 0) \implies x = 0, y = 0.$$

- $\ker(f) = \{(0, 0)\}$ (the zero subspace).
- Image: Check if every $(u, v) \in \mathbb{R}^2$ can be written as:

$$(2x, y) = (u, v).$$

- Solve:

$$x = \frac{u}{2}, y = v.$$

- $\text{Im}(f) = \mathbb{R}^2$ (entire space).

Example 2: Linear Transformation on \mathbb{R}^2 Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be defined as:

$$f(x, y) = (2x, 0).$$

- Kernel: Solve $f(x, y) = (0, 0)$:

$$(2x, 0) = (0, 0) \implies x = 0.$$

- $\ker(f) = \{(0, y) \mid y \in \mathbb{R}\}$ (the y -axis).
- Image: Check if every $(u, v) \in \mathbb{R}^2$ can be written as:

$$(2x, 0) = (u, v).$$

- $v = 0, u = 2x \implies x = \frac{u}{2}.$
- $\text{Im}(f) = \{(u, 0) \mid u \in \mathbb{R}\}$ (the x -axis).

Conclusion

In this lecture, we:

- Defined the kernel and image of linear transformations.
- Connected kernels to null spaces and images to column spaces.
- Established isomorphisms between these subspaces and their matrix representations.
- Demonstrated examples to compute kernel and image using matrices.

This lays the foundation for understanding the structure of linear transformations and their interplay with matrices.