

Recitation 1.2: Linear Transformations

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- (ii) For all $c \in \mathbb{F}$ and $v \in V$: $f(c \cdot v) = c \cdot f(v)$.

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- If $V = \mathbb{F}^n$ and $W = \mathbb{F}^m$ for some $m, n \geq 1$, then we can represent a linear transformation $f : V \rightarrow W$ by an $m \times n$ matrix T , with entries in \mathbb{F} , such that for all $v \in V$, $f(v) = Tv$.

Linear Transformation Examples

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- Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be given by $f(v) = v + 1$. Then f is not a linear transformation, as $f(0) = 1$, but $f(0) = f(0 + 0) = f(0) + f(0) = 2$, and $1 \neq 2$.

Injective Linear Transformations

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- $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ when $n \geq m$, with
$$f((x_1, x_2, \dots, x_n)) = (x_1, x_2, \dots, x_m).$$

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- $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$ given by $f(v) = c \cdot v$, for any $c \in \mathbb{R}$ with $c \neq 0$ is an isomorphism.
- Exercise: if n and m are finite, then $\mathbb{F}^n \cong \mathbb{F}^m$ if and only if $n = m$.
- Exercise: if V is an n -dimensional vector space over \mathbb{F} , then $V \cong \mathbb{F}^n$.

Kernel of a Linear Transformation

Definition

Let $f : V \rightarrow W$ be a linear transformation. The *kernel* of f , denoted $\ker(f)$ is given by:

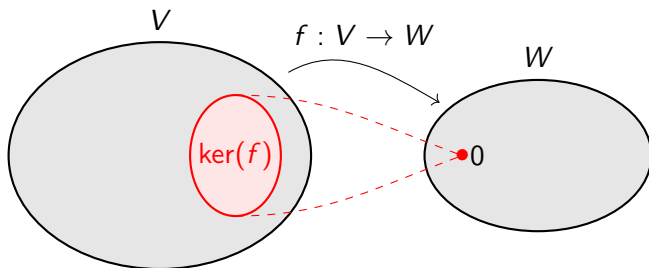
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- Suppose $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ is given by $f((x, y)) = x + y$. $\ker(f)$ is the one-dimensional subspace consisting of all elements $(x, y) \in \mathbb{R}^2$ with $x = -y$. This subspace has basis $\{(1, -1)\}$.

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- Suppose $n > m$ and $f : \mathbb{F}^n \rightarrow \mathbb{F}^m$ is projection onto the first m coordinates. Then the basis for $\ker(f)$ is $\{e_i\}_{m+1 \leq i \leq n}$, where e_i is the standard i -th basis element for \mathbb{F}^n .

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 - Step 4: find solution to system of equations.
 - Step 5: parameterize the solution to form a basis for the kernel.

Kernel Calculation Example

Suppose $f : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ is given by $T = \begin{bmatrix} 3 & 5 & 1 \\ 4 & 1 & 7 \\ 2 & 3 & 1 \end{bmatrix}$.

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Step 2: We convert T to $\text{rref}(T)$:

$$\begin{aligned} \begin{bmatrix} 3 & 5 & 1 \\ 4 & 1 & 7 \\ 2 & 3 & 1 \end{bmatrix} &\xrightarrow{r1 \rightarrow r1 - r3} \begin{bmatrix} 1 & 2 & 0 \\ 4 & 1 & 7 \\ 2 & 3 & 1 \end{bmatrix} \xrightarrow{\substack{r2 \mapsto r2 - 4r1 \\ r3 \mapsto r3 - 2r1}} \begin{bmatrix} 1 & 2 & 0 \\ 0 & -7 & 7 \\ 0 & -1 & 1 \end{bmatrix} \\ &\xrightarrow{r2 \mapsto r2 - 8r3} \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & -1 \\ 0 & -1 & 1 \end{bmatrix} \xrightarrow{\substack{r3 \mapsto r3 + r2 \\ r1 \mapsto r1 - 2r2}} \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{bmatrix} \end{aligned}$$

Kernel Calculation Example

Step 3: This rref matrix gives the following system of equations:

$$1x + 0y + 2z = 0$$

$$0x + 1y - 1z = 0$$

Step 4: We can solve, yielding $y = z$ and $x = -2z$.

Step 5: We can let z be the free variable. Setting $z = 1$, we get a basis for $\ker(T)$, hence $\ker(f)$, is given by $\{(-2, 1, 1)\}$.

Image of a Linear Transformation

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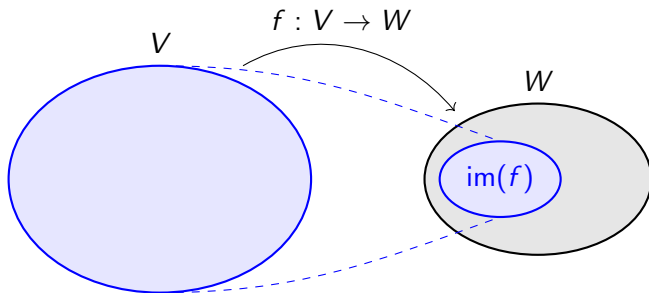


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 - Step 3: use pivot columns to extract basis vectors from columns of the original matrix.

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Step 1,2: From before, we computed the rref for T as:

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Step 3: The pivot columns are columns 1 and 2. These columns in the original matrix give a basis for T . Hence, a basis for $\text{im}(f)$ is given by $\{(3, 4, 2), (5, 1, 3)\}$.