

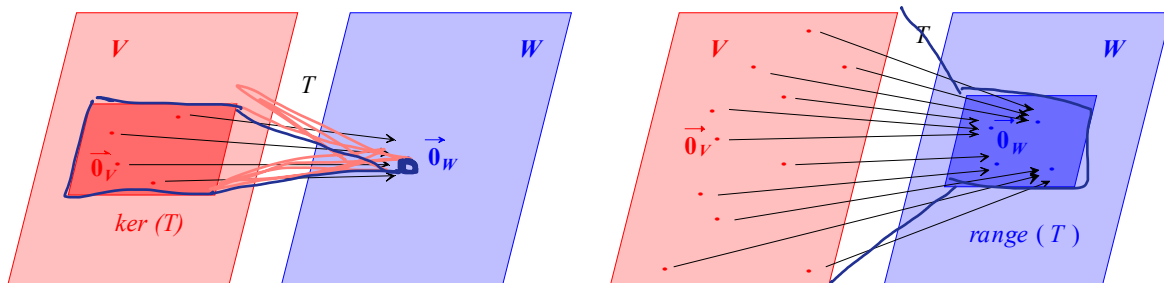
3.7 One-to-One and Onto

Linear Transformations ;

Compositions of

Linear Transformations

Review:



$$\ker(T) = \left\{ \vec{v} \in V \mid \underline{T(\vec{v})} = \vec{0}_W \right\} \quad \checkmark$$

$$\text{range}(T) = \left\{ \vec{w} \in W \mid \underline{\vec{w}} = T(\vec{v}) \text{ for some } \vec{v} \in V \right\} \quad \checkmark$$

One-to-One Transformations

\mathbb{R}^n \mathbb{R}^m

Definition: We say that a linear transformation $T: V \rightarrow W$ is one-to-one or *injective* if the image of different vectors from the domain are different vectors from the codomain:

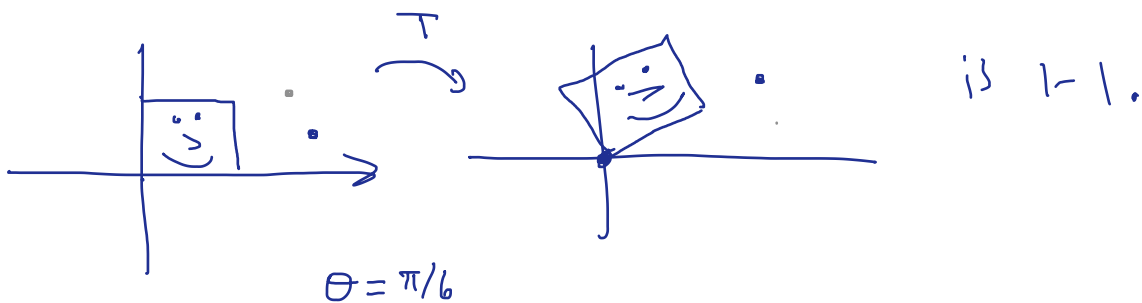
$$\vec{v}_1, \vec{v}_2 \in V$$

$$\text{If } \vec{v}_1 \neq \vec{v}_2 \text{ then } T(\vec{v}_1) \neq T(\vec{v}_2).$$

[exactly same as before]

We again say that T is an ~~injection~~ or an ~~embedding~~.

$$\text{Ex } T: \mathbb{R}^2 \rightarrow \mathbb{R}^2, [T] = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$



Theorem: A linear transformation $T : V \rightarrow W$ is one-to-one if and only if the only way two vectors from the domain have the same image in the codomain is for them to be the same vector to begin with:

[contrapositive] \star If $T(\vec{v}_1) = T(\vec{v}_2)$ then $\vec{v}_1 = \vec{v}_2$. (Useful one)

In other words, the only solution to $T(\vec{v}_1) = T(\vec{v}_2)$ is $\vec{v}_1 = \vec{v}_2$.

Theorem: A linear transformation $T : V \rightarrow W$ is one-to-one *if and only if* $\ker(T) = \{\vec{0}_V\}$.

Same exact proof as in $V = \mathbb{R}^n, W = \mathbb{R}^m$.

(\Leftarrow) Assume $\ker(T) = \{\vec{0}_V\}$. WTS T is 1-1.

Let $\vec{v}_1, \vec{v}_2 \in V$. Let $T(\vec{v}_1) = T(\vec{v}_2)$. Then:

$$\vec{0}_W = T(\vec{v}_1) - T(\vec{v}_2) \quad (\text{by assumption})$$

$$= T(\vec{v}_1 - \vec{v}_2) \quad (\text{both properties: Homog. \& Add Prop})$$

So $\vec{v}_1 - \vec{v}_2 \in \ker(T) = \{\vec{0}_V\}$. So: $\vec{v}_1 - \vec{v}_2 = \vec{0}_V$. So: $\vec{v}_1 = \vec{v}_2$. \square

(\Rightarrow) exercise. \square

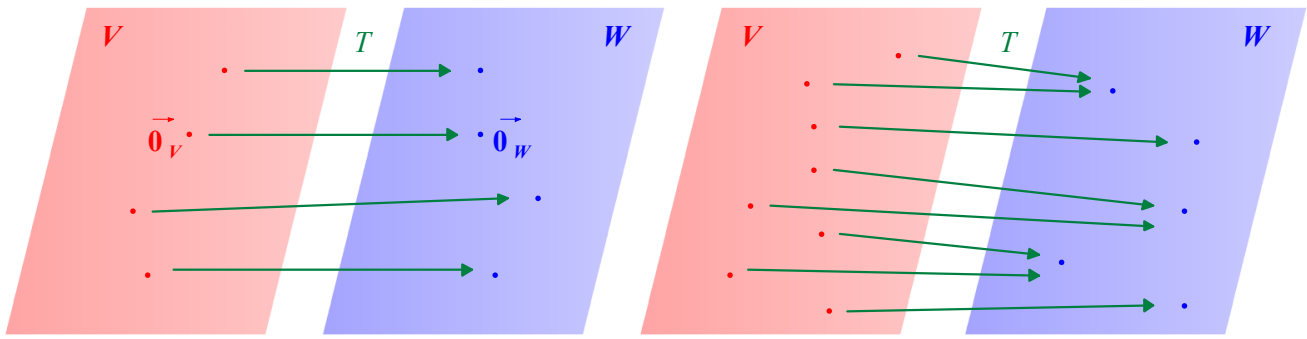
Onto Transformations

\mathbb{R}^n \mathbb{R}^m

Definition: We say that a linear transformation $T : V \rightarrow W$ is onto or *surjective* if the range of T is *all* of W :

def $range(T) = W.$

Since $rank(T) = dim(range(T))$, we can also say that T is onto if and only if $rank(T) = dim(W)$, in the case when W is finite dimensional.



one-to-one \Leftrightarrow
 $ker(T) = \{ \vec{0}_V \}$

onto \Leftrightarrow
 $range(T) = W$

Finding the Kernel and Range Using $[T]_{B,B'}$

The information provided by $[T]_{B,B'}$ and its rref simply needs to be *decoded* with respect to the appropriate basis: we use B for $\ker(T)$ and B' for $\text{range}(T)$.

↙ matrix

↙ GJR

Theorem: Suppose that $T : V \rightarrow W$ is a linear transformation, with $\dim(V) = n$ and $\dim(W) = m$, both finite-dimensional vector spaces. Let $B = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$ be a basis for V , and let $B' = \{\vec{w}_1, \vec{w}_2, \dots, \vec{w}_m\}$ be a basis for W . Let us construct the $m \times n$ matrix $[T]_{B,B'}$ as we did in the previous Section, and let R be the rref of $[T]_{B,B'}$. Suppose that:

$$\{\vec{z}_1, \vec{z}_2, \dots, \vec{z}_k\} \subset \mathbb{R}^n = NS(R) = NS(A)$$

is the basis that we obtain for $\text{nullspace}([T]_{B,B'})$ using R , as we did in Chapter 2.

By the Uniqueness of Representation Property, we know that there exists $\vec{u}_i \in V$ so that $\langle \vec{u}_i \rangle_B = \vec{z}_i$ for every $i = 1 \dots k$.

We conclude that the set $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\} \subset V$ is a **basis** for $\ker(T)$.

As usual, if there are no free variables in R , then $\text{nullspace}([T]_{B,B'}) = \{\vec{0}_n\}$, and consequently, $\ker(T) = \{\vec{0}_V\}$.

So: $\ker(T)$ has basis $\{\vec{u}_1, \dots, \vec{u}_k\} \in V$.

free variables = $\dim(\ker(T)) = \text{nullity}(T)$.

Similarly, the set of original columns:

$$\{\vec{c}_{i_1}, \vec{c}_{i_2}, \dots, \vec{c}_{i_r}\} \subset \mathbb{R}^m$$

from $[T]$

from $[T]_{B, B'}$ corresponding to the leading 1's of R form a basis for $\text{columnspace}([T]_{B, B'})$ as we found in Chapter 2, and there exists $\vec{d}_j \in W$ so that $\langle \vec{d}_j \rangle_{B'} = \vec{c}_{i_j}$ for every $j = 1 \dots r$.

We conclude that the set $\{\vec{d}_1, \vec{d}_2, \dots, \vec{d}_r\} \subset W$ is a **basis** for $\text{range}(T)$.

If T is the zero transformation, then $\text{range}(T) = \{\vec{0}_W\}$.

$\{\vec{c}_{i_1}, \dots, \vec{c}_{i_r}\}$ ^{basis} col in $[T]$ w/ leading 1
(decode to W (b/c $\text{range}(T) \cong W$))

No proof.

The Dimension Theorem for Abstract Vector Spaces

key result in LA!!!

★ The Dimension Theorem: ★

→ general: doesn't need to be 1-1 or onto

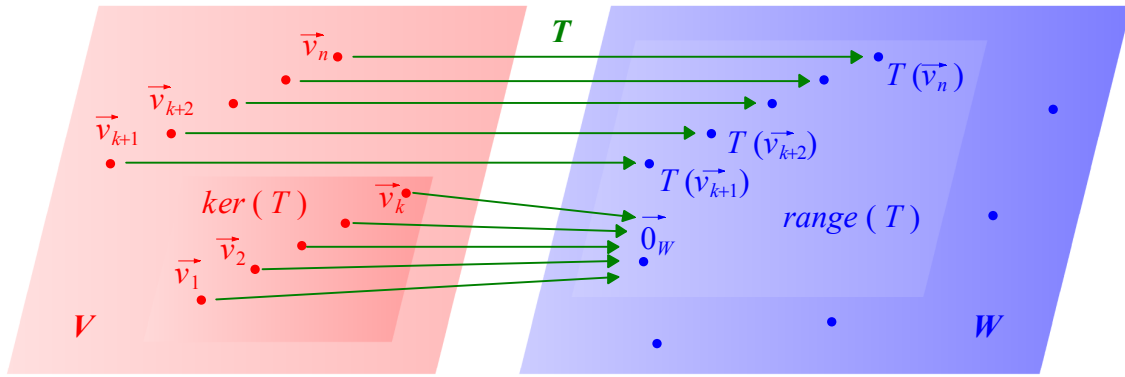
Let $T : V \rightarrow W$ be a linear transformation, and suppose that V is finite dimensional with $\dim(V) = n$. Then, both $\ker(T)$ and $\text{range}(T)$ are finite dimensional, and we can define:

- $\text{rank}(T) = \dim(\text{range}(T))$, and $\text{card}(B_1)$
- $\text{nullity}(T) = \dim(\ker(T))$. $\text{card}(B_2)$

Furthermore, these quantities are related by the equation:

$$\begin{aligned} & \text{rank}(T) + \text{nullity}(T) \\ &= n = \dim(V) = \dim(\text{domain of } T). \end{aligned}$$

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



Idea of the Proof

Let's say $\text{nullity}(T) = k$.

$\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k\}$ a basis for $\ker(T)$

Use the *Extension Theorem*:

$B = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k, \vec{v}_{k+1}, \vec{v}_{k+2}, \dots, \vec{v}_n\}$ a basis for V .

Examine $\text{range}(T)$:

$$\begin{aligned}\vec{v} &= c_1\vec{v}_1 + c_2\vec{v}_2 + \cdots + c_k\vec{v}_k + \\ &\quad c_{k+1}\vec{v}_{k+1} + c_{k+2}\vec{v}_{k+2} + \cdots + c_n\vec{v}_n.\end{aligned}$$

$$\begin{aligned}T(\vec{v}) &= c_1T(\vec{v}_1) + c_2T(\vec{v}_2) + \cdots + c_kT(\vec{v}_k) + \\ &\quad c_{k+1}T(\vec{v}_{k+1}) + c_{k+2}T(\vec{v}_{k+2}) + \cdots + c_nT(\vec{v}_n) \\ &= c_{k+1}T(\vec{v}_{k+1}) + c_{k+2}T(\vec{v}_{k+2}) + \cdots + c_nT(\vec{v}_n)\end{aligned}$$

This tells us that every vector in $\text{range}(T)$ is

Show that $\{T(\vec{v}_{k+1}), T(\vec{v}_{k+2}), \dots, T(\vec{v}_n)\}$ is linearly independent:

$$d_{k+1}T(\vec{v}_{k+1}) + d_{k+2}T(\vec{v}_{k+2}) + \dots + d_nT(\vec{v}_n) = \vec{0}_W.$$

$$T(d_{k+1}\vec{v}_{k+1} + d_{k+2}\vec{v}_{k+2} + \dots + d_n\vec{v}_n) = \vec{0}_W.$$

Conclusion: the vector inside the parenthesis is in . . .

$$\begin{aligned} & d_{k+1}\vec{v}_{k+1} + d_{k+2}\vec{v}_{k+2} + \cdots + d_n\vec{v}_n \\ &= d_1\vec{v}_1 + d_2\vec{v}_2 + \cdots + d_k\vec{v}_k, \end{aligned}$$

$$\begin{aligned} & -d_1\vec{v}_1 - d_2\vec{v}_2 - \cdots - d_k\vec{v}_k + \\ & d_{k+1}\vec{v}_{k+1} + d_{k+2}\vec{v}_{k+2} + \cdots + d_n\vec{v}_n = \vec{0}_W \end{aligned}$$

Conclusion:

Comparing Dimensions

Theorem: Suppose $T : V \rightarrow W$ is a linear transformation of finite dimensional vector spaces. Then:

a) if $\dim(V) < \dim(W)$, then T cannot be onto. ✓ ✓

b) if $\dim(V) > \dim(W)$, then T cannot be one-to-one. ✓ ✓

Compositions of Linear Transformations

eg: $\mathbb{R}^n \xrightarrow{\mathbb{R}^{1 \times n}} \mathbb{R}^k \xrightarrow{\mathbb{R}^{1 \times k}} \mathbb{R}^m$

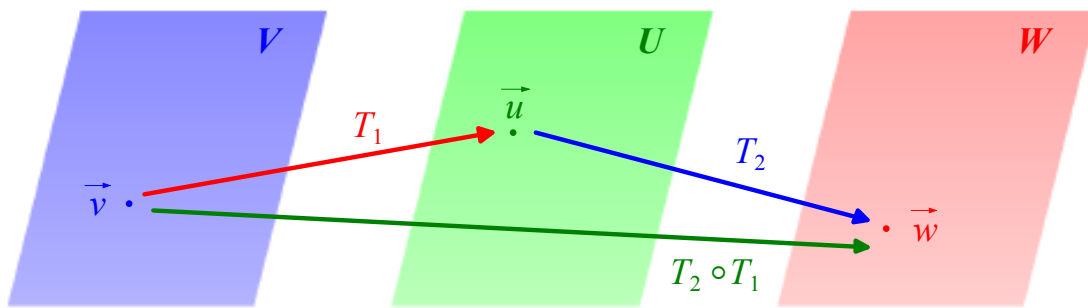
Definition/Theorem: Suppose $T_1 : V \rightarrow U$, and $T_2 : U \rightarrow W$ are linear transformations. The **composition**:

$$T_2 \circ T_1 : V \rightarrow W$$

is again a linear transformation, with action given as follows:

Suppose $\vec{v} \in V$, $T_1(\vec{v}) = \vec{u} \in U$, and $T_2(\vec{u}) = \vec{w} \in W$. Then:

$$(T_2 \circ T_1)(\vec{v}) = T_2(T_1(\vec{v})) = T_2(\vec{u}) = \vec{w}.$$



The Composition of Two Linear Transformations

Note these can all be ∞ -dim'l

NIS • Add Prop: ✓
• Homogeneity Prop: ✓

Do it here. □

The Matrix of a Composition

Theorem: Let $T_1 : V \rightarrow U$ and $T_2 : U \rightarrow W$ be linear transformations of finite dimensional vector spaces. Let B be a basis for V , B' a basis for U , and B'' a basis for W . Then:

$$[T_2 \circ T_1]_{B, B''} = [T_2]_{B', B''} * [T_1]_{B, B'}$$

Proof:
 Set-up. $[T_2 \circ T_1]_{B, B''} = Z_{m \times n}$ (by Thm in §3.6)

$B = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$
 basis V

$$[[T_2 \circ T_1(\vec{v}_1)]_{B''} \mid \dots \mid [T_2 \circ T_1(\vec{v}_n)]_{B''}]$$

$B' = \{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$
 basis U

$$[T_1]_{B, B'} = [[T_1(\vec{v}_1)]_{B'} \mid \dots \mid [T_1(\vec{v}_n)]_{B'}] = \prod_{k \times n}$$

$B'' = \{\vec{w}_1, \vec{w}_2, \dots, \vec{w}_m\}$
 basis W

$$[T_2]_{B', B''} = [[T_2(\vec{u}_1)]_{B''} \mid \dots \mid [T_2(\vec{u}_k)]_{B''}] = \prod_{m \times k}$$

SO WTS $Z_{m \times n} = \prod_{m \times k} * \prod_{k \times n}$

• First column of Z : $[T_2 \circ T_1(\vec{v}_1)]_{B''} = [T_2(T_1(\vec{v}_1))]_{B''}$
 $= [T_2]_{B', B''} * [T_1(\vec{v}_1)]_{B'}$
 $= \prod * \left([T_1]_{B, B'} * [\vec{v}_1]_B \right) = \prod * \prod * [\vec{v}_1]$

• First Column of $\overline{X \circ Y}$: $\overline{X \circ Y} (1^{\text{st}} \text{ col } \overline{Y}) = \overline{X} * [T_1(\vec{v}_1)]_{B'}$
 $= \overline{X} * \left([T_1]_{A, B'} * [\vec{v}_1]_B \right)$

$$[T_2 \circ T_1]_{B, B''}$$

• So works for all columns too!

$$= Z$$



$$= \left[[(T_2 \circ T_1)(\vec{v}_1)]_{B''} \mid [(T_2 \circ T_1)(\vec{v}_2)]_{B''} \mid \dots \mid [(T_2 \circ T_1)(\vec{v}_n)]_{B''} \right]$$

$$[T_2]_{B', B''}$$

$$= X$$

$$= \left[[T_2(\vec{u}_1)]_{B''} \mid [T_2(\vec{u}_2)]_{B''} \mid \dots \mid [T_2(\vec{u}_k)]_{B''} \right]$$

$$[T_1]_{B, B'}$$

$$= Y$$

$$= \left[[T_1(\vec{v}_1)]_{B'} \mid [T_1(\vec{v}_2)]_{B'} \mid \dots \mid [T_1(\vec{v}_n)]_{B'} \right]$$

Function Spaces Preserved by the Derivative

Example: Find the matrix of the *second derivative*, D^2 , applied to the function space:

$$V = \text{Span}(\{x^2 e^{4x}, xe^{4x}, e^{4x}\})$$