

## 2.8 Composition and Invertibility of Linear Transformations

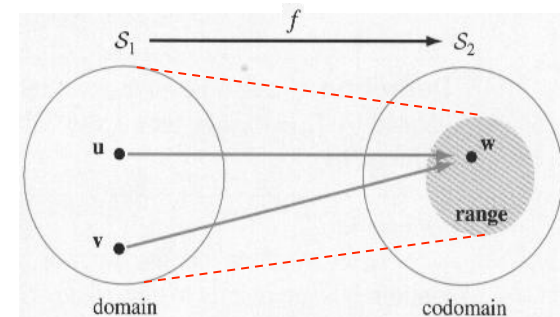
The standard matrix of a linear transformation  $T$  can be used to find a **generating set** for the **range of  $T$** .

Example:  $T : \mathcal{R}^3 \rightarrow \mathcal{R}^2$ ,

$$T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} 3x_1 - 4x_2 \\ 2x_1 + x_3 \end{bmatrix}$$

$$A = \begin{bmatrix} T(\mathbf{e}_1) & T(\mathbf{e}_2) & T(\mathbf{e}_3) \end{bmatrix} = \begin{bmatrix} 3 & -4 & 0 \\ 2 & 0 & 1 \end{bmatrix}$$

Now,  $\mathbf{w} \in \text{range}(T) \Leftrightarrow \mathbf{w} = T(\mathbf{v})$  for some  $\mathbf{v} \in \mathcal{R}^2$ .



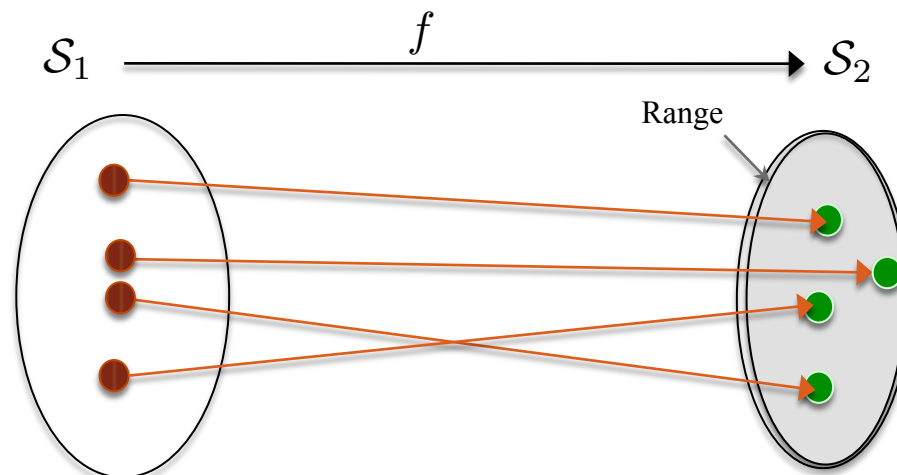
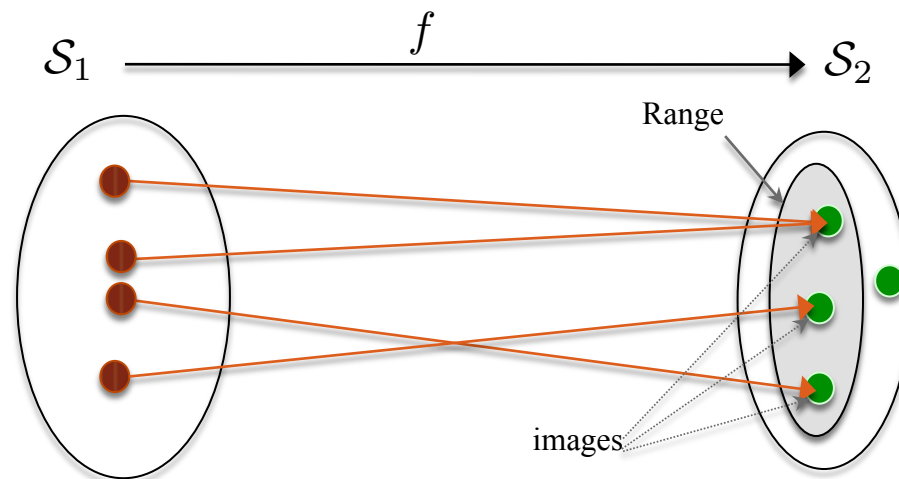
## Property

The range of a linear transformation equals the span of the columns of its standard matrix.

## *Proof*

## Definition

A function  $f : \mathcal{R}^n \rightarrow \mathcal{R}^m$  is said to be **onto** if its range is all of  $\mathcal{R}^m$ ; that is, if every vector in  $\mathcal{R}^m$  is an image.



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## (Review) Theorem 1.6

The following statements about an  $m \times n$  matrix  $A$  are equivalent:

- (a) The span of the columns of  $A$  is  $\mathcal{R}^m$ .
- (c) the rank of  $A$  is  $m$ .

## Theorem 2.10

Let  $T : \mathcal{R}^n \rightarrow \mathcal{R}^m$  be a linear transformation with standard matrix  $A$ . The following conditions are equivalent:

- (a)  $T$  is onto; that is, the range of  $T$  is  $\mathcal{R}^m$ .
- (b) The columns of  $A$  form a generating set for  $\mathcal{R}^m$ .
- (c)  $\text{rank}(A) = m$ .

## *Proof*

Example: Is  $T$  onto?

$$T : \mathcal{R}^3 \rightarrow \mathcal{R}^3 \text{ with } T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} x_1 + 2x_2 + 4x_3 \\ x_1 + 3x_2 + 6x_3 \\ 2x_1 + 5x_2 + 10x_3 \end{bmatrix}$$

$$\Rightarrow A = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix} \text{ with the reduced row echelon form}$$

$$R = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

The system of linear equations  $A\mathbf{x} = \mathbf{b}$  may be written as  $T_A(\mathbf{x}) = \mathbf{b}$ .  
 $A\mathbf{x} = \mathbf{b}$  has a solution if and only if  $\mathbf{b}$  is in the range of  $T_A$ .

$T_A$  is **not onto** if and only if  $\exists \mathbf{b}$  such that  $A\mathbf{x} = \mathbf{b}$  is **inconsistent**.

## Definition

A function  $f : \mathcal{R}^n \rightarrow \mathcal{R}^m$  is said to be **onto** if its range is all of  $\mathcal{R}^m$ ; that is, if every vector in  $\mathcal{R}^m$  is an image.

## (Review) Theorem 1.6

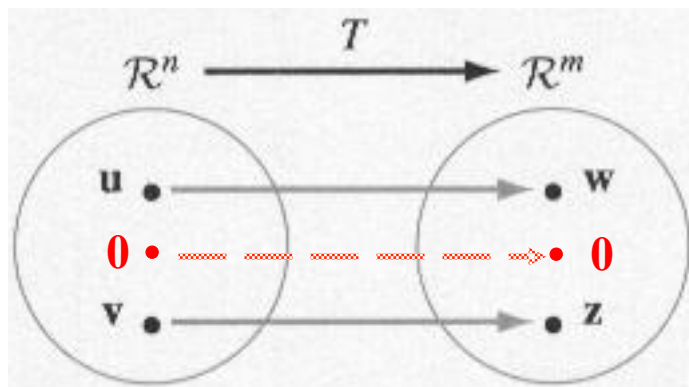
The following statements about an  $m \times n$  matrix  $A$  are equivalent:

- (a) The span of the columns of  $A$  is  $\mathcal{R}^m$ .
- (c) the rank of  $A$  is  $m$ .

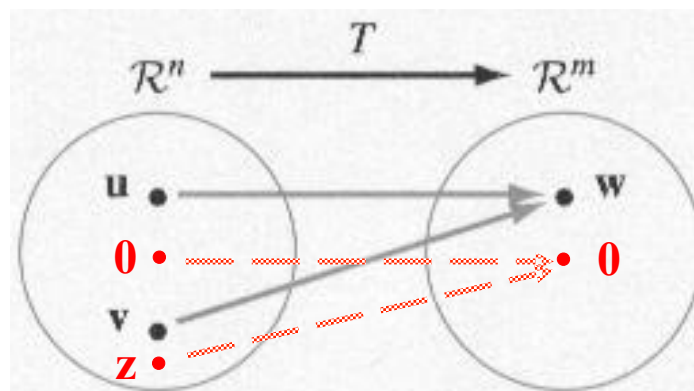
***Proof***

## Definition

A function  $f : \mathcal{R}^n \rightarrow \mathcal{R}^m$  is said to be **one-to-one** if every pair of distinct vectors in  $\mathcal{R}^n$  has distinct images. That is, if  $\mathbf{u}$  and  $\mathbf{v}$  are distinct vectors in  $\mathcal{R}^n$ , then  $f(\mathbf{u})$  and  $f(\mathbf{v})$  are distinct vectors in  $\mathcal{R}^m$ .



one-to-one

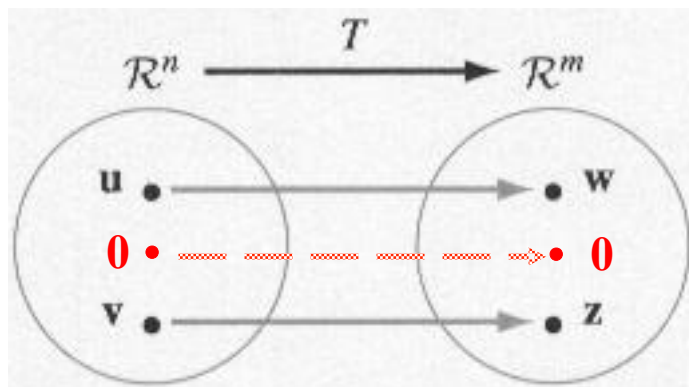


not one-to-one

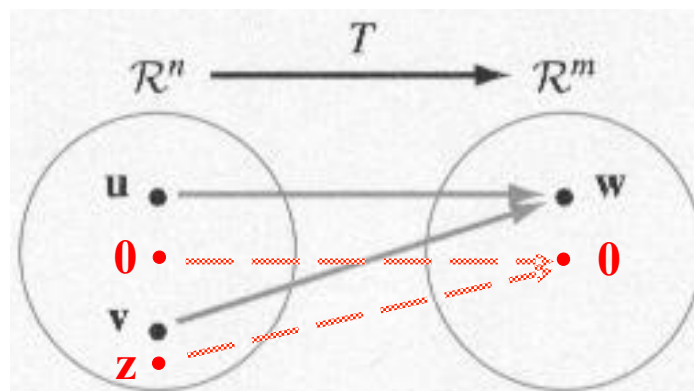
An equivalent condition for a function  $f$  to be one-to-one is that  $f(\mathbf{u}) = f(\mathbf{v})$  implies  $\mathbf{u} = \mathbf{v}$ .

## Definition

A function  $f : \mathcal{R}^n \rightarrow \mathcal{R}^m$  is said to be **one-to-one** if every pair of distinct vectors in  $\mathcal{R}^n$  has distinct images. That is, if  $\mathbf{u}$  and  $\mathbf{v}$  are distinct vectors in  $\mathcal{R}^n$ , then  $f(\mathbf{u})$  and  $f(\mathbf{v})$  are distinct vectors in  $\mathcal{R}^m$ .



one-to-one



not one-to-one

## Definition

Let  $T : \mathcal{R}^n \rightarrow \mathcal{R}^m$  be linear. The **null space** of  $T$  is the set of all  $\mathbf{v} \in \mathcal{R}^n$  such that  $T(\mathbf{v}) = \mathbf{0}$ .

## Property

A linear transformation is one-to-one if and only if its null space contains only  $\mathbf{0}$ .



Example:

$$T : \mathcal{R}^3 \rightarrow \mathcal{R}^2 \text{ with } T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} x_1 - x_2 + 2x_3 \\ -x_1 + x_2 - 3x_3 \end{bmatrix}$$

Find a generating set for the null space of  $T$ .

the null space of  $T$  is the set of solutions to  $A\mathbf{x} = \mathbf{0}$ , where

$$A = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

is the standard matrix of  $T$ , and has the reduced row echelon form

$$R = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

The reduced row echelon form corresponds to the linear equations

$$\begin{array}{rcl} x_1 - x_2 & & = 0 \\ & x_3 & = 0 \end{array}$$

so a generating set for the null space of  $T$  is  $\{ [1 \ 1 \ 0]^T \}$

## Theorem 2.11

Let  $T : \mathcal{R}^n \rightarrow \mathcal{R}^m$  be a linear transformation with standard matrix  $A$ . The following conditions are equivalent:

- (a)  $T$  is one-to-one.
- (b) The null space of  $T$  consists only of the zero vector.
- (c) The columns of  $A$  are linearly independent.
- (d)  $\text{rank}(A) = n$ .

## (Review) Theorem 1.8

The following statements about an  $m \times n$  matrix  $A$  are equivalent:

- (a) The columns of  $A$  are linearly independent.
- (d)  $\text{rank}(A) = n$ .

***Proof***

Example:

$$T : \mathcal{R}^3 \rightarrow \mathcal{R}^3 \text{ with } T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} x_1 + 2x_2 + 4x_3 \\ x_1 + 3x_2 + 6x_3 \\ 2x_1 + 5x_2 + 10x_3 \end{bmatrix} = A\mathbf{x}$$

The standard matrix  $A$  has rank 2 (nullity 1), so  $T$  is not one-to-one.

If  $T_A$  is **one-to-one** and the solution of  $A\mathbf{x} = \mathbf{b}$  exists, then the solution is **unique**. Conversely, if there is at most one solution to  $A\mathbf{x} = \mathbf{b}$  for every  $\mathbf{b}$ , then  $T_A$  is **one-to-one**.

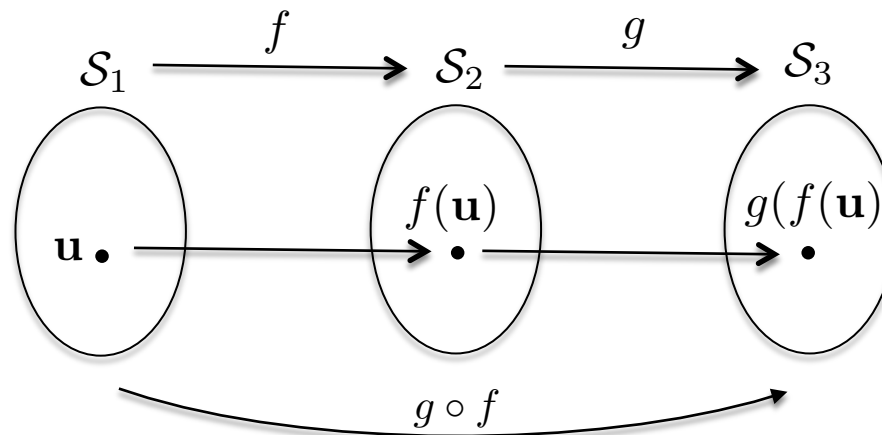
Example:

$$A = \begin{bmatrix} 0 & 0 & 1 & 3 & 3 \\ 2 & 3 & 1 & 5 & 2 \\ 4 & 6 & 1 & 6 & 2 \\ 4 & 6 & 1 & 7 & 1 \end{bmatrix} \xrightarrow[\text{echelon form}]{\text{row reduced}} R = \begin{bmatrix} 1 & 1.5 & 0 & 1 & 0 \\ 0 & 0 & 1 & 3 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

rank  $A = 3$  and nullity  $A = 2. \Rightarrow T_A$  is not onto and not one-to-one.  
 $\Rightarrow A\mathbf{x} = \mathbf{b}$  is inconsistent for  $\mathbf{b}$  not in the range of  $T_A$ , and the solutions of  $A\mathbf{x} = \mathbf{b}$  is never unique.

## Definition.

Given two functions  $f : S_1 \rightarrow S_2$  and  $g : S_2 \rightarrow S_3$ , the **composite**  $g \circ f : S_1 \rightarrow S_3$  is the function defined by  $(g \circ f)(\mathbf{u}) = g(f(\mathbf{u}))$  for all  $\mathbf{u}$  in  $S_1$ . Sometimes the “ $\circ$ ” is dropped for brevity.



Example:  $f\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = \begin{bmatrix} x_1^2 \\ x_1 x_2 \\ x_1 + x_2 \end{bmatrix} \quad g\left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}\right) = \begin{bmatrix} x_1 - x_3 \\ 3x_2 \end{bmatrix}$

$$\Rightarrow (g \circ f)\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = \begin{bmatrix} x_1^2 - (x_1 + x_2) \\ 3x_1 x_2 \end{bmatrix}$$

Given  $A \in \mathcal{R}^{m \times n}$  and  $B \in \mathcal{R}^{p \times m}$ , we have  $T_{BA}(\mathbf{v}) = (BA)\mathbf{v} = B(A\mathbf{v})$   
 $= B(T_A(\mathbf{v})) = T_B(T_A(\mathbf{v}))$  for all  $\mathbf{v} \in \mathcal{R}^n$ . Thus  $T_{BA} = T_B \circ T_A = T_B T_A$ .

## Theorem 2.12

If  $T : \mathcal{R}^n \rightarrow \mathcal{R}^m$  and  $U : \mathcal{R}^m \rightarrow \mathcal{R}^p$  are linear transformations with standard matrices  $A$  and  $B$ , respectively, then the composition  $UT : \mathcal{R}^n \rightarrow \mathcal{R}^p$  is also linear, and its standard matrix is  $BA$ .

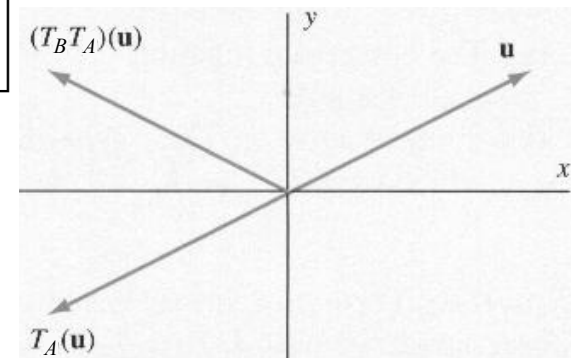
### *Proof*

Example:  $A = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \Rightarrow T_A : \text{rotation by } 180^\circ \text{ in } \mathbb{R}^2.$

$B = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \Rightarrow T_B : \text{reflection about the } x\text{-axis in } \mathbb{R}^2.$

$$T_{BA} \left( \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \right) = (BA) \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} -u_1 \\ u_2 \end{bmatrix}$$

$\Rightarrow T_{BA} : \text{reflection about the } y\text{-axis in } \mathbb{R}^2$   
 $= \text{rotation by } 180^\circ \text{ followed by}$   
 $\text{reflection about the } x\text{-axis in } \mathbb{R}^2.$



## Definition.

A function  $f : S_1 \rightarrow S_2$  is **invertible** if there exists a function  $g : S_2 \rightarrow S_1$  such that  $(g \circ f)(\mathbf{v}) = g(f(\mathbf{v})) = \mathbf{v}$  for all  $\mathbf{v}$  in  $S_1$  and  $(f \circ g)(\mathbf{u}) = f(g(\mathbf{u})) = \mathbf{u}$  for all  $\mathbf{u}$  in  $S_2$ . If  $f$  is invertible, then the function  $g$  is unique, called the **inverse** of  $f$  and denoted as  $f^{-1}$ .

If  $A \in \mathcal{R}^{n \times n}$  is invertible, then for all  $\mathbf{v} \in \mathbb{R}^n$  we have  $T_A(T_{A^{-1}}(\mathbf{v})) = (T_A T_{A^{-1}})(\mathbf{v}) = T_{AA^{-1}}(\mathbf{v}) = T_{I_n}(\mathbf{v}) = I_n \mathbf{v} = \mathbf{v}$ , and  $T_{A^{-1}}(T_A(\mathbf{v})) = \mathbf{v}$ . Thus  $T_A^{-1} = T_{A^{-1}}$ .

Example:

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix} \quad A^{-1} = \begin{bmatrix} -5 & 2 \\ 3 & -1 \end{bmatrix}$$

$$T_A \left( \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right) = \begin{bmatrix} v_1 + 2v_2 \\ 3v_1 + 5v_2 \end{bmatrix}$$

$$T_A^{-1} \left( \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right) = T_{A^{-1}} \left( \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right) = \begin{bmatrix} -5 & 2 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} -5v_1 + 2v_2 \\ 3v_1 - v_2 \end{bmatrix}$$

## Theorem 2.13

Let  $T : \mathcal{R}^n \rightarrow \mathcal{R}^n$  be a linear transformation with standard matrix  $A$ . Then  $T$  is invertible if and only if  $A$  is invertible, in which case  $T^{-1} = T_{A^{-1}}$ . Thus  $T^{-1}$  is linear, and its standard matrix is  $A^{-1}$ .

*Proof*

Property of $T$	The number of solutions of $A\mathbf{x}=\mathbf{b}$	Property of the columns of $A$	Property of the rank of $A$	Theorem
$T$ is onto	$A\mathbf{x} = \mathbf{b}$ has at least one solution for every $\mathbf{b}$ in $\mathcal{R}^m$ .	The columns of $A$ are a generating set for $\mathcal{R}^m$ .	$\text{rank } A = m$	<b>2.10</b>
$T$ is one-to-one	$A\mathbf{x} = \mathbf{b}$ has at most one solution for every $\mathbf{b}$ in $\mathcal{R}^m$ .	The columns of $A$ are linearly independent.	$\text{rank } A = n$	<b>2.11</b>
$T$ is invertible	$A\mathbf{x} = \mathbf{b}$ has a unique solution for every $\mathbf{b}$ in $\mathcal{R}^m$	The columns of $A$ are a linearly independent generating set for $\mathcal{R}^m$ .	$\text{rank } A = m = n$	<b>2.13</b>



## Homework Set for Section 2.8

Section 2.8: Problems 1, 5, 13, 17, 19, 25, 29, 33, 37