



NTNU
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TMA4115 - Calculus 3
Lecture 11, Feb 20

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Norwegian University of Science and Technology
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Review of last week's lecture



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- introduced and solved *homogeneous* and *nonhomogeneous* matrix equations,
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- looked at applications of linear systems,
- introduced and studied *linear dependence* and *linear independence* of vectors.



Today's lecture



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We shall introduce and study



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- *onto* linear transformations,
- *one-to-one* linear transformations.



Transformations



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Transformations

- A *transformation* (or *function* or *mapping*) T from \mathbb{R}^n to \mathbb{R}^m is a rule that assigns to each vector \mathbf{x} in \mathbb{R}^n a vector $T(\mathbf{x})$ in \mathbb{R}^m .



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- For \mathbf{x} in \mathbb{R}^n , the vector $T(\mathbf{x})$ in \mathbb{R}^m is called the image of \mathbf{x} (under the action of T).
- The set of all images $T(\mathbf{x})$ is called the range (or image) of T .



Example



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Example

For each $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \in \mathbb{R}^3$, let

$$T \left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} \sqrt{x_1^2 + x_2^2 + x_3^2} \\ \cos(x_1 x_2 x_3 \pi) \end{bmatrix}$$



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Then T is a transformation from \mathbb{R}^3 to \mathbb{R}^2 .

The image of $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ is $\begin{bmatrix} \sqrt{1^2 + 1^2 + 1^2} \\ \cos(\pi) \end{bmatrix} = \begin{bmatrix} \sqrt{3} \\ -1 \end{bmatrix}$.



Example (cont.)

The range of T is $\left\{ \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} : y \geq 0, y_2 \in [-1, 1] \right\}$ (this is not completely obvious).



Example (cont.)

The range of T is $\left\{ \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} : y_1 \geq 0, y_2 \in [-1, 1] \right\}$ (this is not completely obvious).

If $f_1(x_1, x_2, x_3) = \sqrt{x_1^2 + x_2^2 + x_3^2}$ and

$f_2(x_1, x_2, x_3) = \cos(x_1 x_2 x_3 \pi)$, then f_1 and f_2 are real-valued functions on \mathbb{R}^3 and

$$T \left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} f_1(x_1, x_2, x_3) \\ f_2(x_1, x_2, x_3) \end{bmatrix}$$

for all $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \in \mathbb{R}^3$.



Transformations and real-valued functions of several variables



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Transformations and real-valued functions of several variables

If f_1, f_2, \dots, f_m are real-valued functions on \mathbb{R}^n , then

$$T \left(\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \right) = \begin{bmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \vdots \\ f_m(x_1, x_2, \dots, x_n) \end{bmatrix}$$

defines a transformation from \mathbb{R}^n to \mathbb{R}^m .



Transformations and real-valued functions of several variables

Conversely, if T is a transformation from \mathbb{R}^n to \mathbb{R}^m , then there exist real-valued functions f_1, f_2, \dots, f_m on \mathbb{R}^n such that

$$T \left(\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \right) = \begin{bmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \vdots \\ f_m(x_1, x_2, \dots, x_n) \end{bmatrix}$$

for all $\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n$.



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Example



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Example

For each $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \in \mathbb{R}^2$, let

$$S\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = \begin{bmatrix} 2x_1 - x_2 \\ x_1 \\ 3x_2 \end{bmatrix}$$



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Then S is a transformation from \mathbb{R}^2 to \mathbb{R}^3 .

Notice that

$$S\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = x_1 \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} -1 \\ 0 \\ 3 \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ 1 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$



Example (cont.)

It follows that the range of S is $\text{Span} \left\{ \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 3 \end{bmatrix} \right\}$.



Matrix transformations



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Matrix transformations

If A is an $m \times n$ matrix and we for every \mathbf{x} in \mathbb{R}^n let $T(\mathbf{x}) = A\mathbf{x}$, then T is a transformation from \mathbb{R}^n to \mathbb{R}^m .



Matrix transformations

If A is an $m \times n$ matrix and we for every \mathbf{x} in \mathbb{R}^n let $T(\mathbf{x}) = A\mathbf{x}$, then T is a transformation from \mathbb{R}^n to \mathbb{R}^m .
The range of A is $\text{Span}\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\}$ where $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ are the columns of A .



Linear transformations



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A transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is *linear* if:



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① $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$ for all \mathbf{u}, \mathbf{v} in \mathbb{R}^n ;



Linear transformations

A transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is *linear* if:

- 1 $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$ for all \mathbf{u}, \mathbf{v} in \mathbb{R}^n ;
- 2 $T(c\mathbf{u}) = cT(\mathbf{u})$ for all scalars c and all \mathbf{u} in \mathbb{R}^n .



Matrix transformations are linear



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Matrix transformations are linear

Every matrix transformation is linear because if A is an $m \times n$ matrix, then $A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v}$ for all \mathbf{u}, \mathbf{v} in \mathbb{R}^n and $A(c\mathbf{u}) = cA\mathbf{u}$ for all scalars c and all \mathbf{u} in \mathbb{R}^n .



Properties of linear transformations



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If $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a linear transformation, then:

- 1 $T(\mathbf{0}) = \mathbf{0}$ because $T(\mathbf{0}) = T(0\mathbf{u}) = 0T(\mathbf{u}) = \mathbf{0}$ for any vector \mathbf{u} in \mathbb{R}^n .



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- 1 $T(\mathbf{0}) = \mathbf{0}$ because $T(\mathbf{0}) = T(0\mathbf{u}) = 0T(\mathbf{u}) = \mathbf{0}$ for any vector \mathbf{u} in \mathbb{R}^n .
- 2 $T(c\mathbf{u} + d\mathbf{v}) = cT(\mathbf{u}) + dT(\mathbf{v})$ for all scalars c and d and all vectors \mathbf{u} and \mathbf{v} in \mathbb{R}^n , because $T(c\mathbf{u} + d\mathbf{v}) = T(c\mathbf{u}) + T(d\mathbf{v}) = cT(\mathbf{u}) + dT(\mathbf{v})$.



Example



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Example

A company manufactures two products. For \$1.00 worth of product B, the company spends \$0.45 on material, \$0.25 on labor, and \$0.15 on overhead. For \$1.00 worth of product C, the company spends \$0.40 on material, \$0.30 on labor, and \$0.15 on overhead.



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Let x_1 denote the value of product B and x_2 the value of product C that the company manufactures.

Let y_1 , y_2 and y_3 denote the costs of material, labor and overhead the company spends on this.



Example (cont.)

Then

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 0.45x_1 + 0.40x_2 \\ 0.25x_1 + 0.30x_2 \\ 0.15x_1 + 0.15x_2 \end{bmatrix}$$



Example (cont.)

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So if we define $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$ by

$$T(\mathbf{x}) = \begin{bmatrix} 0.45 & 0.40 \\ 0.25 & 0.30 \\ 0.15 & 0.15 \end{bmatrix} \mathbf{x}$$

then $\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = T(\mathbf{x})$.



Example



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Example

Let $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$ be a linear transformation such that

$$T\left(\begin{bmatrix} 1 \\ 0 \end{bmatrix}\right) = \begin{bmatrix} 1 \\ 3 \\ -1 \end{bmatrix} \text{ and } T\left(\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right) = \begin{bmatrix} -3 \\ 5 \\ 7 \end{bmatrix}.$$


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- 1 Let $\mathbf{u} = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$. Let us find the image $T(\mathbf{u})$ of \mathbf{u} under T .



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1 Let $\mathbf{u} = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$. Let us find the image $T(\mathbf{u})$ of \mathbf{u} under T .

$$\begin{aligned} T(\mathbf{u}) &= T\left(2 \begin{bmatrix} 1 \\ 0 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \end{bmatrix}\right) = 2T\left(\begin{bmatrix} 1 \\ 0 \end{bmatrix}\right) - T\left(\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right) \\ &= 2 \begin{bmatrix} 1 \\ 3 \\ -1 \end{bmatrix} - \begin{bmatrix} -3 \\ 5 \\ 7 \end{bmatrix} = \begin{bmatrix} 5 \\ 1 \\ -9 \end{bmatrix} \end{aligned}$$



Example (cont.)

- 2 Let us find the image under T of $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$.



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$$\begin{aligned} T(\mathbf{x}) &= T\left(x_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}\right) = x_1 T\left(\begin{bmatrix} 1 \\ 0 \end{bmatrix}\right) + x_2 T\left(\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right) \\ &= x_1 \begin{bmatrix} 1 \\ 3 \\ -1 \end{bmatrix} + x_2 \begin{bmatrix} -3 \\ 5 \\ 7 \end{bmatrix} = \begin{bmatrix} x_1 - 3x_2 \\ 3x_1 + 5x_2 \\ -x_1 + 7x_2 \end{bmatrix} \end{aligned}$$



Example (cont.)

- 2 Let us find the image under T of $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$.

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It follows that $T(\mathbf{x}) = A\mathbf{x}$ for all $\mathbf{x} \in \mathbb{R}^2$ where

$$A = \begin{bmatrix} 1 & -3 \\ 3 & 5 \\ -1 & 7 \end{bmatrix}.$$



Example (cont.)

- 3 Is there more than one \mathbf{x} whose image under T is

$$\begin{bmatrix} 5 \\ 1 \\ -9 \end{bmatrix} ?$$



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The question is equivalent to the question: Does the equation $A\mathbf{x} = \begin{bmatrix} 5 \\ 1 \\ -9 \end{bmatrix}$ have more than one solution?



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- 3 Is there more than one \mathbf{x} whose image under T is $\begin{bmatrix} 5 \\ 1 \\ -9 \end{bmatrix}$?

The question is equivalent to the question: Does the equation $A\mathbf{x} = \begin{bmatrix} 5 \\ 1 \\ -9 \end{bmatrix}$ have more than one solution?

To answer that question we reduce the augmented matrix of the equation to an echelon form.



Example (cont.)

$$\begin{bmatrix} 1 & -3 & 5 \\ 3 & 5 & 1 \\ -1 & 7 & -9 \end{bmatrix}$$



Example (cont.)

$$\begin{bmatrix} 1 & -3 & 5 \\ 3 & 5 & 1 \\ -1 & 7 & -9 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -3 & 5 \\ 0 & 14 & -14 \\ 0 & 4 & -4 \end{bmatrix}$$



Example (cont.)

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We see that there is no free variable, so there is only one \mathbf{x} such that $T(\mathbf{x}) = \begin{bmatrix} 5 \\ 1 \\ -9 \end{bmatrix}$.



Example (cont.)

4 Let $\mathbf{c} = \begin{bmatrix} 3 \\ 2 \\ 5 \end{bmatrix}$. Let us determine if \mathbf{c} is in the range of T .



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\mathbf{c} is in the range of T if and only if the equation $A\mathbf{x} = \mathbf{c}$ is consistent.



Example (cont.)

- 4 Let $\mathbf{c} = \begin{bmatrix} 3 \\ 2 \\ 5 \end{bmatrix}$. Let us determine if \mathbf{c} is in the range of T .

\mathbf{c} is in the range of T if and only if the equation $A\mathbf{x} = \mathbf{c}$ is consistent.

In order to determine whether the equation $A\mathbf{x} = \mathbf{c}$ is consistent or not, we reduce the augmented matrix $[A \ \mathbf{c}]$ to an echelon form.



Example (cont.)

$$\begin{bmatrix} 1 & -3 & 3 \\ 3 & 5 & 2 \\ -1 & 7 & 5 \end{bmatrix}$$



Example (cont.)

$$\begin{bmatrix} 1 & -3 & 3 \\ 3 & 5 & 2 \\ -1 & 7 & 5 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -3 & 3 \\ 0 & 14 & -7 \\ 0 & 4 & 8 \end{bmatrix}$$



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Example (cont.)

$$\begin{aligned} \begin{bmatrix} 1 & -3 & 3 \\ 3 & 5 & 2 \\ -1 & 7 & 5 \end{bmatrix} &\rightarrow \begin{bmatrix} 1 & -3 & 3 \\ 0 & 14 & -7 \\ 0 & 4 & 8 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -3 & 3 \\ 0 & 2 & -1 \\ 0 & 1 & 2 \end{bmatrix} \\ &\rightarrow \begin{bmatrix} 1 & -3 & 3 \\ 0 & 1 & 2 \\ 0 & 2 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -3 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & -5 \end{bmatrix} \end{aligned}$$



Example (cont.)

$$\begin{aligned} \begin{bmatrix} 1 & -3 & 3 \\ 3 & 5 & 2 \\ -1 & 7 & 5 \end{bmatrix} &\rightarrow \begin{bmatrix} 1 & -3 & 3 \\ 0 & 14 & -7 \\ 0 & 4 & 8 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -3 & 3 \\ 0 & 2 & -1 \\ 0 & 1 & 2 \end{bmatrix} \\ &\rightarrow \begin{bmatrix} 1 & -3 & 3 \\ 0 & 1 & 2 \\ 0 & 2 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -3 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & -5 \end{bmatrix} \end{aligned}$$

We see that the system is inconsistent, so \mathbf{c} is not in the range of T .



The matrix of a linear transformation



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The matrix of a linear transformation

Theorem 10

Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation. Then there exists a unique matrix A such that $T(\mathbf{x}) = A\mathbf{x}$ for all \mathbf{x} in \mathbb{R}^n .



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The matrix A is called the *standard matrix* of T .



Proof



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Proof

Let $A = [T(\mathbf{e}_1) \ \dots \ T(\mathbf{e}_n)]$.



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Let $A = [T(\mathbf{e}_1) \ \dots \ T(\mathbf{e}_n)]$. We must show that $T(\mathbf{x}) = A\mathbf{x}$ for all \mathbf{x} in \mathbb{R}^n .



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Let $A = [T(\mathbf{e}_1) \ \dots \ T(\mathbf{e}_n)]$. We must show that $T(\mathbf{x}) = A\mathbf{x}$ for all \mathbf{x} in \mathbb{R}^n .

Let $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$ be a vector in \mathbb{R}^n .



Proof (cont.)

Then

$$\begin{aligned}T(\mathbf{x}) &= T\left(\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}\right) = T\left(x_1 \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix} + \cdots + x_n \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}\right) \\&= T(x_1 \mathbf{e}_1 + x_2 \mathbf{e}_2 + \cdots + x_n \mathbf{e}_n) \\&= x_1 T(\mathbf{e}_1) + x_2 T(\mathbf{e}_2) + \cdots + x_n T(\mathbf{e}_n) \\&= A \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = A\mathbf{x}.\end{aligned}$$



Example



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Example

Let $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ be a linear transformation such that $T(x_1, x_2, x_3) = (x_1 - 3x_2 - x_3, x_2 + 4x_3, 2x_1 + 9x_2 + 5x_3)$.
Let us find the standard matrix of T .



Solution



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Solution

$$T(\mathbf{e}_1) = T(1, 0, 0) = (1, 0, 2) = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix},$$



Solution

$$T(\mathbf{e}_1) = T(1, 0, 0) = (1, 0, 2) = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix},$$

$$T(\mathbf{e}_2) = T(0, 1, 0) = (-3, 1, 9) = \begin{bmatrix} -3 \\ 1 \\ 9 \end{bmatrix},$$



Solution

$$T(\mathbf{e}_1) = T(1, 0, 0) = (1, 0, 2) = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix},$$

$$T(\mathbf{e}_2) = T(0, 1, 0) = (-3, 1, 9) = \begin{bmatrix} -3 \\ 1 \\ 9 \end{bmatrix},$$

$$T(\mathbf{e}_3) = T(0, 0, 1) = (-1, 4, 5) = \begin{bmatrix} -1 \\ 4 \\ 5 \end{bmatrix},$$



Solution

$$T(\mathbf{e}_1) = T(1, 0, 0) = (1, 0, 2) = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix},$$

$$T(\mathbf{e}_2) = T(0, 1, 0) = (-3, 1, 9) = \begin{bmatrix} -3 \\ 1 \\ 9 \end{bmatrix},$$

$$T(\mathbf{e}_3) = T(0, 0, 1) = (-1, 4, 5) = \begin{bmatrix} -1 \\ 4 \\ 5 \end{bmatrix}, \text{ so the standard}$$

$$\text{matrix of } T \text{ is } \begin{bmatrix} 1 & -3 & -1 \\ 0 & 1 & 4 \\ 2 & 9 & 5 \end{bmatrix}.$$



Example



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Example

Let $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be the function given by $T(\mathbf{x}) = \mathbf{y}$ where \mathbf{y} is the vector we obtain from rotating \mathbf{x} by an angle of θ around zero.

Let us show that T is a linear transformation and find its standard matrix.



Solution



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Solution

Let $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ be a vector in \mathbb{R}^2 , and let $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = T(\mathbf{x})$.



Solution

Let $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ be a vector in \mathbb{R}^2 , and let $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = T(\mathbf{x})$.

Let $r = |x_1 + ix_2| = \sqrt{x_1^2 + x_2^2}$ and $\phi = \text{Arg}(x_1 + ix_2)$.



Solution

Let $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ be a vector in \mathbb{R}^2 , and let $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = T(\mathbf{x})$.

Let $r = |x_1 + ix_2| = \sqrt{x_1^2 + x_2^2}$ and $\phi = \text{Arg}(x_1 + ix_2)$. Then $x_1 = r \cos \phi$ and $x_2 = r \sin \phi$,



Solution

Let $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ be a vector in \mathbb{R}^2 , and let $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = T(\mathbf{x})$.

Let $r = |x_1 + ix_2| = \sqrt{x_1^2 + x_2^2}$ and $\phi = \text{Arg}(x_1 + ix_2)$. Then $x_1 = r \cos \phi$ and $x_2 = r \sin \phi$, so $y_1 = r \cos(\phi + \theta) = r \cos \phi \cos \theta - r \sin \phi \sin \theta = x_1 \cos \theta - x_2 \sin \theta$



Solution

Let $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ be a vector in \mathbb{R}^2 , and let $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = T(\mathbf{x})$.

Let $r = |x_1 + ix_2| = \sqrt{x_1^2 + x_2^2}$ and $\phi = \text{Arg}(x_1 + ix_2)$. Then $x_1 = r \cos \phi$ and $x_2 = r \sin \phi$, so $y_1 = r \cos(\phi + \theta) = r \cos \phi \cos \theta - r \sin \phi \sin \theta = x_1 \cos \theta - x_2 \sin \theta$ and $y_2 = r \sin(\phi + \theta) = r \sin \phi \cos \theta + r \cos \phi \sin \theta = x_2 \cos \theta + x_1 \sin \theta$.



Solution

Let $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ be a vector in \mathbb{R}^2 , and let $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = T(\mathbf{x})$.

Let $r = |x_1 + ix_2| = \sqrt{x_1^2 + x_2^2}$ and $\phi = \text{Arg}(x_1 + ix_2)$. Then $x_1 = r \cos \phi$ and $x_2 = r \sin \phi$, so $y_1 = r \cos(\phi + \theta) = r \cos \phi \cos \theta - r \sin \phi \sin \theta = x_1 \cos \theta - x_2 \sin \theta$ and $y_2 = r \sin(\phi + \theta) = r \sin \phi \cos \theta + r \cos \phi \sin \theta = x_2 \cos \theta + x_1 \sin \theta$.

So if we let $A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$, then $T(\mathbf{x}) = A\mathbf{x}$ for every \mathbf{x} in \mathbb{R}^2 .



Solution

Let $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ be a vector in \mathbb{R}^2 , and let $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = T(\mathbf{x})$.

Let $r = |x_1 + ix_2| = \sqrt{x_1^2 + x_2^2}$ and $\phi = \text{Arg}(x_1 + ix_2)$. Then $x_1 = r \cos \phi$ and $x_2 = r \sin \phi$, so $y_1 = r \cos(\phi + \theta) = r \cos \phi \cos \theta - r \sin \phi \sin \theta = x_1 \cos \theta - x_2 \sin \theta$ and $y_2 = r \sin(\phi + \theta) = r \sin \phi \cos \theta + r \cos \phi \sin \theta = x_2 \cos \theta + x_1 \sin \theta$.

So if we let $A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$, then $T(\mathbf{x}) = A\mathbf{x}$ for every \mathbf{x} in \mathbb{R}^2 .

It follows that T is linear, and that A is the standard matrix of T .



One-to-one transformations



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One-to-one transformations

Definition of one-to-one transformations

A transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be *one-to-one* (or *injective*) if each \mathbf{b} in \mathbb{R}^m is the image of at most one \mathbf{x} in \mathbb{R}^n .



One-to-one transformations

Definition of one-to-one transformations

A transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be *one-to-one* (or *injective*) if each \mathbf{b} in \mathbb{R}^m is the image of at most one \mathbf{x} in \mathbb{R}^n .

Notice that each \mathbf{b} in \mathbb{R}^m does not have to be in the image of T in order for T to be one-to-one.



One-to-one transformations

Theorem 11

A linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is one-to-one if and only if the equation $T(\mathbf{x}) = \mathbf{0}$ has only the trivial solution.



Proof



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Proof

We have that $T(\mathbf{0}) = \mathbf{0}$ since T is linear.



Proof

We have that $T(\mathbf{0}) = \mathbf{0}$ since T is linear.

If T is one-to-one, then there is at most one \mathbf{x} such that $T(\mathbf{x}) = \mathbf{0}$.



Proof

We have that $T(\mathbf{0}) = \mathbf{0}$ since T is linear.

If T is one-to-one, then there is at most one \mathbf{x} such that $T(\mathbf{x}) = \mathbf{0}$. So the equation $T(\mathbf{x}) = \mathbf{0}$ has only the trivial solution.



Proof

We have that $T(\mathbf{0}) = \mathbf{0}$ since T is linear.

If T is one-to-one, then there is at most one \mathbf{x} such that $T(\mathbf{x}) = \mathbf{0}$. So the equation $T(\mathbf{x}) = \mathbf{0}$ has only the trivial solution.

If T is not one-to-one, then there are two different vectors \mathbf{u} and \mathbf{v} in \mathbb{R}^n such that $T(\mathbf{u}) = T(\mathbf{v})$.



Proof

We have that $T(\mathbf{0}) = \mathbf{0}$ since T is linear.

If T is one-to-one, then there is at most one \mathbf{x} such that $T(\mathbf{x}) = \mathbf{0}$. So the equation $T(\mathbf{x}) = \mathbf{0}$ has only the trivial solution.

If T is not one-to-one, then there are two different vectors \mathbf{u} and \mathbf{v} in \mathbb{R}^n such that $T(\mathbf{u}) = T(\mathbf{v})$. We then have that $T(\mathbf{u} - \mathbf{v}) = T(\mathbf{u}) - T(\mathbf{v}) = \mathbf{0}$, so $\mathbf{x} = \mathbf{u} - \mathbf{v}$ is a nontrivial solution to the equation $T(\mathbf{x}) = \mathbf{0}$.



Example



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Example

Let T be the linear transformation whose standard matrix is

$$A = \begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ -1 & 2 & 4 & 5 \end{bmatrix}$$

Let us determine if T is one-to-one.



Solution



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Solution

T is one-to-one if and only if the equation $A\mathbf{x} = \mathbf{0}$ only has the trivial solution.



Solution

T is one-to-one if and only if the equation $A\mathbf{x} = \mathbf{0}$ only has the trivial solution. So we reduce A to an echelon form and check if A has a pivot position in every column.



Solution

T is one-to-one if and only if the equation $A\mathbf{x} = \mathbf{0}$ only has the trivial solution. So we reduce A to an echelon form and check if A has a pivot position in every column.

$$\begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ -1 & 2 & 4 & 5 \end{bmatrix}$$



Solution

T is one-to-one if and only if the equation $A\mathbf{x} = \mathbf{0}$ only has the trivial solution. So we reduce A to an echelon form and check if A has a pivot position in every column.

$$\begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ -1 & 2 & 4 & 5 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ 0 & 0 & 0 & 12 \end{bmatrix}$$



Solution

T is one-to-one if and only if the equation $A\mathbf{x} = \mathbf{0}$ only has the trivial solution. So we reduce A to an echelon form and check if A has a pivot position in every column.

$$\begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ -1 & 2 & 4 & 5 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ 0 & 0 & 0 & 12 \end{bmatrix}$$

Since A does not have a pivot position in every column, the equation $A\mathbf{x} = \mathbf{0}$ has a free position and therefore a nontrivial solution, so T is not one-to-one.



How to determine if a linear transformation is one-to-one



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How to determine if a linear transformation is one-to-one

The following procedure outlines how to determine if a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is one-to-one:



How to determine if a linear transformation is one-to-one

The following procedure outlines how to determine if a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is one-to-one:

- 1 Find the standard matrix A of T .



How to determine if a linear transformation is one-to-one

The following procedure outlines how to determine if a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is one-to-one:

- 1 Find the standard matrix A of T .
- 2 Reduce A to an echelon form.



How to determine if a linear transformation is one-to-one

The following procedure outlines how to determine if a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is one-to-one:

- 1 Find the standard matrix A of T .
- 2 Reduce A to an echelon form.
- 3 If A has a pivot position in every column, then the equation $A\mathbf{x} = \mathbf{0}$ has no free variable, so the equation $A\mathbf{x} = \mathbf{0}$ only has the trivial solution, and T is one-to-one.



Onto transformations



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Onto transformations

Definition of onto transformations

A transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be *onto* (or *surjective*) if each \mathbf{b} in \mathbb{R}^m is the image of at least one \mathbf{x} in \mathbb{R}^n .



Onto transformations

Definition of onto transformations

A transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be *onto* (or *surjective*) if each \mathbf{b} in \mathbb{R}^m is the image of at least one \mathbf{x} in \mathbb{R}^n .

Notice that a transformation T is onto if and only if the image of T is all of \mathbb{R}^m .



Example



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Example

Let T be the linear transformation whose standard matrix is

$$A = \begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ -1 & 2 & 4 & 5 \end{bmatrix}$$

Let us determine if T is onto.



Solution



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Solution

T is onto if and only if the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 .



Solution

T is onto if and only if the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 . We know that the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 if and only if A has a pivot position in every row.



Solution

T is onto if and only if the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 . We know that the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 if and only if A has a pivot position in every row. So we reduce A to an echelon form and check if A has a pivot position in every row.



Solution

T is onto if and only if the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 . We know that the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 if and only if A has a pivot position in every row. So we reduce A to an echelon form and check if A has a pivot position in every row.

$$\begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ -1 & 2 & 4 & 5 \end{bmatrix}$$



Solution

T is onto if and only if the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 . We know that the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 if and only if A has a pivot position in every row. So we reduce A to an echelon form and check if A has a pivot position in every row.

$$\begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ -1 & 2 & 4 & 5 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ 0 & 0 & 0 & 12 \end{bmatrix}$$



Solution

T is onto if and only if the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 . We know that the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^3 if and only if A has a pivot position in every row. So we reduce A to an echelon form and check if A has a pivot position in every row.

$$\begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ -1 & 2 & 4 & 5 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -2 & -4 & 7 \\ 0 & 2 & 6 & -3 \\ 0 & 0 & 0 & 12 \end{bmatrix}$$

We see that A has a pivot position in every row, so T is onto.



How to determine if a linear transformation is onto



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How to determine if a linear transformation is onto

The following procedure outlines how to determine if a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is onto:



How to determine if a linear transformation is onto

The following procedure outlines how to determine if a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is onto:

- 1 Find the standard matrix A of T .



How to determine if a linear transformation is onto

The following procedure outlines how to determine if a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is onto:

- 1 Find the standard matrix A of T .
- 2 Reduce A to an echelon form.



How to determine if a linear transformation is onto

The following procedure outlines how to determine if a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is onto:

- 1 Find the standard matrix A of T .
- 2 Reduce A to an echelon form.
- 3 If A has a pivot position in every row, then the equation $A\mathbf{x} = \mathbf{b}$ is consistent for all \mathbf{b} in \mathbb{R}^m and T is onto.



Onto and one-to-one linear transformations



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Onto and one-to-one linear transformations

Theorem 12

Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation and let A be the standard matrix for T .



Onto and one-to-one linear transformations

Theorem 12

Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation and let A be the standard matrix for T . Then:



Onto and one-to-one linear transformations

Theorem 12

Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation and let A be the standard matrix for T . Then:

- 1 T maps \mathbb{R}^n onto \mathbb{R}^m if and only if the columns of A span \mathbb{R}^m .



Onto and one-to-one linear transformations

Theorem 12

Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation and let A be the standard matrix for T . Then:

- 1 T maps \mathbb{R}^n onto \mathbb{R}^m if and only if the columns of A span \mathbb{R}^m .
- 2 T is one-to-one if and only if the columns of A are linearly independent.



Proof



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Proof

T is onto



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Proof

T is onto \iff the equation $A\mathbf{x} = \mathbf{b}$ has a solution
for every $\mathbf{b} \in \mathbb{R}^m$



Proof

T is onto \iff the equation $A\mathbf{x} = \mathbf{b}$ has a solution
for every $\mathbf{b} \in \mathbb{R}^m$
 \iff the columns of A span \mathbb{R}^m .



Proof

T is onto \iff the equation $A\mathbf{x} = \mathbf{b}$ has a solution
for every $\mathbf{b} \in \mathbb{R}^m$
 \iff the columns of A span \mathbb{R}^m .

T is one-to-one



Proof

T is onto \iff the equation $A\mathbf{x} = \mathbf{b}$ has a solution
for every $\mathbf{b} \in \mathbb{R}^m$
 \iff the columns of A span \mathbb{R}^m .

T is one-to-one \iff the equation $A\mathbf{x} = \mathbf{0}$ has a
nontrivial solution



Proof

T is onto \iff the equation $A\mathbf{x} = \mathbf{b}$ has a solution
for every $\mathbf{b} \in \mathbb{R}^m$
 \iff the columns of A span \mathbb{R}^m .

T is one-to-one \iff the equation $A\mathbf{x} = \mathbf{0}$ has a
nontrivial solution
 \iff the columns of A are
linearly independent.



Example



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Example

Let $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$ be a linear transformation such that $T(x_1, x_2) = (x_1 + 2x_2, 2x_1 + x_2, 0)$. Let us determine if T is onto and if it is one-to-one.



Solution

Every linear combination of $\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$ has a zero as its third entry, so the columns of A does not span \mathbb{R}^3 .



Solution

Every linear combination of $\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$ has a zero as its third entry, so the columns of A does not span \mathbb{R}^3 . It follows that T is not onto.



Solution

Every linear combination of $\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$ has a zero as its third entry, so the columns of A does not span \mathbb{R}^3 . It follows that T is not onto.

Since $\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$ is not multiple of $\begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$, and $\begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$ is not a multiple of $\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$, the columns of A are linearly independent.



Solution

Every linear combination of $\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$ has a zero as its third entry, so the columns of A does not span \mathbb{R}^3 . It follows that T is not onto.

Since $\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$ is not multiple of $\begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$, and $\begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$ is not a multiple of

$\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$, the columns of A are linearly independent. It follows that T is one-to-one.



Tomorrow's lecture

Tomorrow we shall

- look at *applications* of linear models,
- look at the use of *Maple* and *WolframAlpha*.

Section 1.10 in “Linear Algebras and Its Applications” (pages 80-90).



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