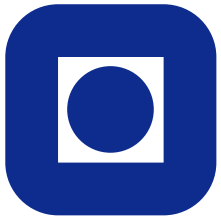

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TMA4115 - Calculus 3

Lecture 13, Feb 27

Toke Meier Carlsen
Norwegian University of Science and Technology

Spring 2013

Review of last week's lecture

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Review of last week's lecture

Last week we looked at

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- *linear transformations*,

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- *linear transformations*,
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Review of last week's lecture

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- *linear transformations*,
- the *standard matrix* of a linear transformation,
- *onto* linear transformations,
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- *applications* of linear models,

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Review of last week's lecture

Last week we looked at

- *linear transformations*,
- the *standard matrix* of a linear transformation,
- *onto* linear transformations,
- *one-to-one* linear transformations,
- *applications* of linear models,
- *Maple* and *WolframAlpha*.

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Today's lecture

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Today's lecture

We shall

Today's lecture

We shall

- see how we can *add* and *multiply* matrices,

Today's lecture

We shall

- see how we can *add* and *multiply* matrices,
- define and study *invertible* matrices.

Matrix addition and scalar multiplication

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Matrix addition and scalar multiplication

- If A and B are $m \times n$ matrices, then the *sum* $A + B$ is the $m \times n$ matrix whose entries are the sums of the corresponding entries in A and B.

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Matrix addition and scalar multiplication

- If A and B are $m \times n$ matrices, then the *sum* $A + B$ is the $m \times n$ matrix whose entries are the sums of the corresponding entries in A and B.
- The sum $A + B$ is defined only when A and B are the same size.

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Matrix addition and scalar multiplication

- If A and B are $m \times n$ matrices, then the *sum* $A + B$ is the $m \times n$ matrix whose entries are the sums of the corresponding entries in A and B.
- The sum $A + B$ is defined only when A and B are the same size.
- If r is a scalar and A is an $m \times n$ matrix, then the *scalar multiple* rA is the $m \times n$ matrix whose entries are r times the corresponding entries in A.

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Matrix addition and scalar multiplication

- If A and B are $m \times n$ matrices, then the *sum* $A + B$ is the $m \times n$ matrix whose entries are the sums of the corresponding entries in A and B.
- The sum $A + B$ is defined only when A and B are the same size.
- If r is a scalar and A is an $m \times n$ matrix, then the *scalar multiple* rA is the $m \times n$ matrix whose entries are r times the corresponding entries in A.
- We will use $-A$ to denote the matrix $(-1)A$.

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Examples

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Examples

Let $A = \begin{bmatrix} 1 & 2 & 3 \\ -1 & 4 & 0 \end{bmatrix}$, $B = \begin{bmatrix} 3 & 2 & -6 \\ 0 & 7 & 1 \end{bmatrix}$, and $C = \begin{bmatrix} 2 & 1 \\ -4 & 3 \end{bmatrix}$.

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Examples

Let $A = \begin{bmatrix} 1 & 2 & 3 \\ -1 & 4 & 0 \end{bmatrix}$, $B = \begin{bmatrix} 3 & 2 & -6 \\ 0 & 7 & 1 \end{bmatrix}$, and $C = \begin{bmatrix} 2 & 1 \\ -4 & 3 \end{bmatrix}$.

Then $A + B = \begin{bmatrix} 4 & 4 & -3 \\ -1 & 11 & 1 \end{bmatrix}$

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Examples

Let $A = \begin{bmatrix} 1 & 2 & 3 \\ -1 & 4 & 0 \end{bmatrix}$, $B = \begin{bmatrix} 3 & 2 & -6 \\ 0 & 7 & 1 \end{bmatrix}$, and $C = \begin{bmatrix} 2 & 1 \\ -4 & 3 \end{bmatrix}$.

Then $A + B = \begin{bmatrix} 4 & 4 & -3 \\ -1 & 11 & 1 \end{bmatrix}$ and $-2C = \begin{bmatrix} -4 & -2 \\ 8 & -6 \end{bmatrix}$.

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Let $A = \begin{bmatrix} 1 & 2 & 3 \\ -1 & 4 & 0 \end{bmatrix}$, $B = \begin{bmatrix} 3 & 2 & -6 \\ 0 & 7 & 1 \end{bmatrix}$, and $C = \begin{bmatrix} 2 & 1 \\ -4 & 3 \end{bmatrix}$.

Then $A + B = \begin{bmatrix} 4 & 4 & -3 \\ -1 & 11 & 1 \end{bmatrix}$ and $-2C = \begin{bmatrix} -4 & -2 \\ 8 & -6 \end{bmatrix}$.

$A + C$ is not defined because A and C have different sizes.

Properties of matrix addition and scalar multiplication

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Properties of matrix addition and scalar multiplication

Theorem 1

Let A, B, and C be matrices of the same size, and let r and s be scalars.

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Properties of matrix addition and scalar multiplication

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Properties of matrix addition and scalar multiplication

Theorem 1

Let A, B, and C be matrices of the same size, and let r and s be scalars. Then:

1. $A + B = B + A$

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Properties of matrix addition and scalar multiplication

Theorem 1

Let A, B, and C be matrices of the same size, and let r and s be scalars. Then:

1. $A + B = B + A$
2. $(A + B) + C = A + (B + C)$

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Properties of matrix addition and scalar multiplication

Theorem 1

Let A, B, and C be matrices of the same size, and let r and s be scalars. Then:

1. $A + B = B + A$
2. $(A + B) + C = A + (B + C)$
3. $A + 0 = A$ where 0 is the matrix of the same size of A whose entries are all zero.

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Properties of matrix addition and scalar multiplication

Theorem 1

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1. $A + B = B + A$
2. $(A + B) + C = A + (B + C)$
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4. $r(A + B) = rA + rB$

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5. $(r + s)A = rA + sA$

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Properties of matrix addition and scalar multiplication

Theorem 1

Let A, B, and C be matrices of the same size, and let r and s be scalars. Then:

1. $A + B = B + A$
2. $(A + B) + C = A + (B + C)$
3. $A + 0 = A$ where 0 is the matrix of the same size of A whose entries are all zero.
4. $r(A + B) = rA + rB$
5. $(r + s)A = rA + sA$
6. $r(sA) = (rs)A$

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Composition of linear transformations

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Composition of linear transformations

Let $T : \mathbb{R}^p \to \mathbb{R}^n$ and $S : \mathbb{R}^n \to \mathbb{R}^m$ be linear transformations.

Composition of linear transformations

Let $T : \mathbb{R}^p \to \mathbb{R}^n$ and $S : \mathbb{R}^n \to \mathbb{R}^m$ be linear transformations. Then $S \circ T$ is the transformation from $\mathbb{R}^p \to \mathbb{R}^m$ such that $S \circ T(\mathbf{x}) = S(T(\mathbf{x}))$ for $\mathbf{x} \in \mathbb{R}^p$.

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$S \circ T(\mathbf{x}) = S(T(\mathbf{x}))$ for $\mathbf{x} \in \mathbb{R}^p$.
If $\mathbf{u}$ and $\mathbf{v}$ are vectors in $\mathbb{R}^p$ and c is a scalar, then

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If $\mathbf{u}$ and $\mathbf{v}$ are vectors in $\mathbb{R}^p$ and c is a scalar, then

- $S \circ T(\mathbf{u} + \mathbf{v}) = S(T(\mathbf{u} + \mathbf{v})) = S(T(\mathbf{u}) + T(\mathbf{v})) = S(T(\mathbf{u})) + S(T(\mathbf{v})) = S \circ T(\mathbf{u}) + S \circ T(\mathbf{v})$,

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- $S \circ T(c\mathbf{u}) = S(T(c\mathbf{u})) = S(cT(\mathbf{u})) = cS(T(\mathbf{u})) = cS \circ T(\mathbf{u})$.

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So $S \circ T$ is linear.

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Composition of linear transformations

Let $T : \mathbb{R}^p \to \mathbb{R}^n$ and $S : \mathbb{R}^n \to \mathbb{R}^m$ be linear transformations. Then $S \circ T$ is the transformation from $\mathbb{R}^p \to \mathbb{R}^m$ such that $S \circ T(\mathbf{x}) = S(T(\mathbf{x}))$ for $\mathbf{x} \in \mathbb{R}^p$.
If $\mathbf{u}$ and $\mathbf{v}$ are vectors in $\mathbb{R}^p$ and c is a scalar, then

- $S \circ T(\mathbf{u} + \mathbf{v}) = S(T(\mathbf{u} + \mathbf{v})) = S(T(\mathbf{u}) + T(\mathbf{v})) = S(T(\mathbf{u})) + S(T(\mathbf{v})) = S \circ T(\mathbf{u}) + S \circ T(\mathbf{v})$, and
- $S \circ T(c\mathbf{u}) = S(T(c\mathbf{u})) = S(cT(\mathbf{u})) = cS(T(\mathbf{u})) = cS \circ T(\mathbf{u})$.

So $S \circ T$ is linear. What is the standard matrix of $S \circ T$?

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The standard matrix of $S \circ T$

The standard matrix of $S \circ T$

Let A be the standard matrix of S, let B be the standard matrix of T, and let $\mathbf{b}_1, \mathbf{b}_2, \ldots, \mathbf{b}_p$ be the columns of B.

The standard matrix of $S \circ T$

Let A be the standard matrix of S, let B be the standard matrix of T, and let $\mathbf{b}_1, \mathbf{b}_2, \ldots, \mathbf{b}_p$ be the columns of B.

Then we have for $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_p \end{bmatrix}$ in $\mathbb{R}^p$ that

$$\begin{aligned} S \circ T(\mathbf{x}) &= S(T(\mathbf{x})) = S(B\mathbf{x}) \\ &= A(B\mathbf{x}) = A(x_1\mathbf{b}_1 + x_2\mathbf{b}_2 + \cdots + x_p\mathbf{b}_p) \\ &= x_1 A\mathbf{b}_1 + x_2 A\mathbf{b}_2 + \ldots x_p A\mathbf{b}_p. \end{aligned}$$

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The standard matrix of $S \circ T$

Let A be the standard matrix of S, let B be the standard matrix of T, and let $\mathbf{b}_1, \mathbf{b}_2, \ldots, \mathbf{b}_p$ be the columns of B.

Then we have for $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_p \end{bmatrix}$ in $\mathbb{R}^p$ that

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It follows that the standard matrix of $S \circ T$ is the $m \times p$ matrix $[A\mathbf{b}_1 \ A\mathbf{b}_2 \ \ldots \ A\mathbf{b}_p]$.

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Matrix multiplication

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Matrix multiplication

Definition

If A is an $m \times n$ matrix, and if B is an $n \times p$ matrix with columns $\mathbf{b}_1, \ldots, \mathbf{b}_p$, then the product AB is the $m \times p$ matrix whose columns are $A\mathbf{b}_1, \ldots, A\mathbf{b}_p$.

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Matrix multiplication

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If A is an $m \times n$ matrix, and if B is an $n \times p$ matrix with columns $\mathbf{b}_1, \ldots, \mathbf{b}_p$, then the product AB is the $m \times p$ matrix whose columns are $A\mathbf{b}_1, \ldots, A\mathbf{b}_p$.
That is,

$$AB = A[\mathbf{b}_1 \ \mathbf{b}_2 \ \ldots \ \mathbf{b}_p] = [A\mathbf{b}_1 \ A\mathbf{b}_2 \ \ldots \ A\mathbf{b}_p].$$

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If A is an $m \times n$ matrix, and if B is an $n \times p$ matrix with columns $\mathbf{b}_1, \ldots, \mathbf{b}_p$, then the product AB is the $m \times p$ matrix whose columns are $A\mathbf{b}_1, \ldots, A\mathbf{b}_p$.
That is,

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Each column of AB is a linear combination of the columns of A using weights from the corresponding column of B.

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The standard matrix of the composition of 2 linear transformation

Let $T : \mathbb{R}^p \to \mathbb{R}^n$ and $S : \mathbb{R}^n \to \mathbb{R}^m$ be linear transformations. If A is the standard matrix of S, and B is the standard matrix of T, then AB is the standard matrix of $S \circ T$.

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Example

Let $A = \begin{bmatrix} 2 & 1 \\ -1 & 3 \end{bmatrix}$ and $B = \begin{bmatrix} 2 & 1 & -3 \\ 2 & 4 & 0 \end{bmatrix}$. Let us compute AB.

Solution

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Solution

$$A\mathbf{b}_1 = \begin{bmatrix} 2 & 1 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$$

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Solution

$$A\mathbf{b}_1 = \begin{bmatrix} 2 & 1 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$$

$$A\mathbf{b}_2 = \begin{bmatrix} 2 & 1 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 4 \end{bmatrix} = \begin{bmatrix} 6 \\ 11 \end{bmatrix}$$

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$$A\mathbf{b}_1 = \begin{bmatrix} 2 & 1 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$$

$$A\mathbf{b}_2 = \begin{bmatrix} 2 & 1 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 4 \end{bmatrix} = \begin{bmatrix} 6 \\ 11 \end{bmatrix}$$

$$A\mathbf{b}_3 = \begin{bmatrix} 2 & 1 \\ -3 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = \begin{bmatrix} -6 \\ 3 \end{bmatrix}$$

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$$\text{So } AB = \begin{bmatrix} 6 & 6 & -6 \\ 4 & 11 & 3 \end{bmatrix}.$$

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Row—column rule for computing AB

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Row—column rule for computing AB

If a product AB is defined, then the entry in row i and column j of AB is the sum of the products of corresponding entries from row i of A and column j of B.

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Row—column rule for computing AB

If a product AB is defined, then the entry in row i and column j of AB is the sum of the products of corresponding entries from row i of A and column j of B.
If A is an $m \times n$ matrix, B is an $n \times p$ matrix, a_{ik} denotes the (i,k)-entry in A, b_{kj} denotes the (k,j)-entry in B, and $(AB)_{ij}$ denotes the (i,j)-entry in A, then

$$(AB)_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{in}b_{nj}.$$

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Example

Let $A = \begin{bmatrix} 2 & -2 \\ 1 & -1 \\ 3 & 0 \end{bmatrix}$ and $B = \begin{bmatrix} -2 & 1 \\ 4 & 0 \end{bmatrix}$. Let us compute AB.

Solution

$$\begin{bmatrix} 2 & -2 \\ 1 & -1 \\ 3 & 0 \end{bmatrix} \begin{bmatrix} -2 & 1 \\ 4 & 0 \end{bmatrix} = \begin{bmatrix} 2\cdot(-2)+(-2)\cdot 4 & 2\cdot 1+(-2)\cdot 0 \\ 1\cdot(-2)+(-1)\cdot 4 & 1\cdot 1+(-1)\cdot 0 \\ 3\cdot(-2)+0\cdot 4 & 3\cdot 1+0\cdot 0 \end{bmatrix}$$

$$= \begin{bmatrix} -12 & 2 \\ -6 & 1 \\ -6 & 3 \end{bmatrix}$$

Properties of matrix multiplication

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Properties of matrix multiplication

Theorem 2

Let r be a scalar, let A be an $m \times n$ matrix, and let B and C have sizes for which the indicated sums and products are defined.

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Properties of matrix multiplication

Theorem 2

Let r be a scalar, let A be an $m \times n$ matrix, and let B and C have sizes for which the indicated sums and products are defined. Then:

1. $A(BC) = (AB)C$

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Properties of matrix multiplication

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Let r be a scalar, let A be an $m \times n$ matrix, and let B and C have sizes for which the indicated sums and products are defined. Then:

1. $A(BC) = (AB)C$
2. $A(B + C) = AB + AC$

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1. $A(BC) = (AB)C$
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3. $(B + C)A = BA + CA$

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1. $A(BC) = (AB)C$
2. $A(B + C) = AB + AC$
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4. $r(AB) = (rA)B = A(rB)$

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2. $A(B + C) = AB + AC$
3. $(B + C)A = BA + CA$
4. $r(AB) = (rA)B = A(rB)$
5. $I_m A = A I_n = A$

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AB and BA might be different

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AB and BA might be different

- BA might not be defined, even if AB is defined

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AB and BA might be different

- BA might not be defined, even if AB is defined (if, for example A is a 2×2 matrix and B is a 2×3 matrix, then AB is defined, but BA is not).

AB and BA might be different

- BA might not be defined, even if AB is defined (if, for example A is a 2×2 matrix and B is a 2×3 matrix, then AB is defined, but BA is not).
- Even when AB and BA are both defined, then AB and BA might not have the same size

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- BA might not be defined, even if AB is defined (if, for example A is a 2×2 matrix and B is a 2×3 matrix, then AB is defined, but BA is not).
- Even when AB and BA are both defined, then AB and BA might not have the same size (if, for example A is a 3×2 matrix and B is a 2×3 matrix, then AB is a 3×3 matrix and BA is a 2×2 matrix).

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- BA might not be defined, even if AB is defined (if, for example A is a 2×2 matrix and B is a 2×3 matrix, then AB is defined, but BA is not).
- Even when AB and BA are both defined, then AB and BA might not have the same size (if, for example A is a 3×2 matrix and B is a 2×3 matrix, then AB is a 3×3 matrix and BA is a 2×2 matrix).
- Even when AB and BA are both defined, and AB and BA have the same size, then AB and BA might be different.

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Warnings

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Warnings

1. In general, $AB \neq BA$

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1. In general, $AB \neq BA$ (for instance, $\begin{bmatrix} 1 & 2 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 3 & 1 \end{bmatrix} \neq \begin{bmatrix} 2 & 0 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ -1 & 0 \end{bmatrix}$).
2. The cancellation laws do not hold for matrix multiplication. That is, if $AB = AC$, then it is not true in general that $B = C$

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1. In general, $AB \neq BA$ (for instance, $\begin{bmatrix} 1 & 2 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 3 & 1 \end{bmatrix} \neq \begin{bmatrix} 2 & 0 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ -1 & 0 \end{bmatrix}$).
2. The cancellation laws do not hold for matrix multiplication. That is, if $AB = AC$, then it is not true in general that $B = C$ (for instance, $\begin{bmatrix} 3 & -6 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 3 & -6 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} -3 & -5 \\ 2 & 1 \end{bmatrix}$).

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Warnings

1. In general, $AB \neq BA$ (for instance, $\begin{bmatrix} 1 & 2 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 3 & 1 \end{bmatrix} \neq \begin{bmatrix} 2 & 0 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ -1 & 0 \end{bmatrix}$).
2. The cancellation laws do not hold for matrix multiplication. That is, if $AB = AC$, then it is not true in general that $B = C$ (for instance, $\begin{bmatrix} 3 & -6 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 3 & -6 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} -3 & -5 \\ 2 & 1 \end{bmatrix}$).
3. If a product AB is the zero matrix, you cannot conclude in general that either $A = 0$ or $B = 0$

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Commuting matrices

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Commuting matrices

- If $AB = BA$, then we say that A and B *commute* (*kommutere* in Norwegian).

Commuting matrices

- If $AB = BA$, then we say that A and B *commute* (*kommutere* in Norwegian).
- A necessary (but not sufficient) condition for A and B to commute is that there are square matrices of the same size.

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Powers of a matrix

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Powers of a matrix

- If A is a square matrix and k is a positive integer, then A^k denotes the product of k copies of A. That is,

$$A^k = \underbrace{A \dots A}_{k \text{ times}}.$$

Example

In a certain region, 5% of a city's population moves to the surrounding suburbs each year, and 3% of the suburban population moves into the city.

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In a certain region, 5% of a city's population moves to the surrounding suburbs each year, and 3% of the suburban population moves into the city.
We saw last week that if r_i denotes the population in the city in year $2000 + i$, s_i denotes the population in the suburbs in year $2000 + i$ and $\mathbf{x}_i = \begin{bmatrix} r_i \\ s_i \end{bmatrix}$, then $\mathbf{x}_{i+1} = \begin{bmatrix} 0.95 & 0.03 \\ 0.05 & 0.97 \end{bmatrix} \mathbf{x}_i$.

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So $\mathbf{x}_n = \begin{bmatrix} 0.95 & 0.03 \\ 0.05 & 0.97 \end{bmatrix}^n \mathbf{x}_0$.

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The transpose of a matrix

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The transpose of a matrix

Definition

Given an $m \times n$ matrix A, the *transpose* of A is the $n \times m$, denoted by A^T, whose columns are formed from the corresponding rows of A.

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Examples

Let $A = \begin{bmatrix} 2 & 2 \\ -4 & 5 \end{bmatrix}$ and $B = \begin{bmatrix} 3 & 2 & -7 \\ -1 & -1 & 3 \end{bmatrix}$.

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Examples

Let $A = \begin{bmatrix} 2 & 2 \\ -4 & 5 \end{bmatrix}$ and $B = \begin{bmatrix} 3 & 2 & -7 \\ -1 & -1 & 3 \end{bmatrix}$.

Then $A^T = \begin{bmatrix} 2 & -4 \\ 2 & 5 \end{bmatrix}$ and $B^T = \begin{bmatrix} 3 & -1 \\ 2 & -1 \\ -7 & 3 \end{bmatrix}$.

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Properties of the transpose

Properties of the transpose

Theorem 3

Let A and B denote matrices whose sizes are appropriate for the following sums and products.

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3. $(rA)^T = rA^T$ for any scalar r.

Properties of the transpose

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Let A and B denote matrices whose sizes are appropriate for the following sums and products. Then:

1. $(A^T)^T = A$.
2. $(A + B)^T = A^T + B^T$.
3. $(rA)^T = rA^T$ for any scalar r.
4. $(AB)^T = B^T A^T$.

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Invertible transformations

Invertible transformations

Let $T : \mathbb{R}^n \to \mathbb{R}^m$ be a transformation.

Invertible transformations

Let $T : \mathbb{R}^n \to \mathbb{R}^m$ be a transformation.

1. T is said to be *invertible* if there exists a transformation $S : \mathbb{R}^m \to \mathbb{R}^n$ such that $S \circ T(\mathbf{x}) = \mathbf{x}$ for all $\mathbf{x}$ in $\mathbb{R}^n$ and $T \circ S(\mathbf{y}) = \mathbf{y}$ for all $\mathbf{y}$ in $\mathbb{R}^m$.

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   Proof: Assume that T is invertible and let $S : \mathbb{R}^m \to \mathbb{R}^n$ be a transformation such that $S \circ T(\mathbf{x}) = \mathbf{x}$ for all $\mathbf{x}$ in $\mathbb{R}^n$ and $T \circ S(\mathbf{y}) = \mathbf{y}$ for all $\mathbf{y}$ in $\mathbb{R}^m$.

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Invertible transformations (cont.)

If $\mathbf{u}$ and $\mathbf{v}$ are vectors in $\mathbb{R}^n$ and $T(\mathbf{u}) = T(\mathbf{v})$, then $\mathbf{u} = S(T(\mathbf{u})) = S(T(\mathbf{v})) = \mathbf{v}$.

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Invertible transformations (cont.)

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Invertible transformations (cont.)

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Assume on the other hand that T is onto and one-to-one.

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Invertible transformations (cont.)

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Assume on the other hand that T is onto and one-to-one. Then there is for each $\mathbf{y}$ in $\mathbb{R}^m$ a unique $\mathbf{x}$ in $\mathbb{R}^n$ such that $T(\mathbf{x}) = \mathbf{y}$.

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Invertible transformations (cont.)

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Assume on the other hand that T is onto and one-to-one. Then there is for each $\mathbf{y}$ in $\mathbb{R}^m$ a unique $\mathbf{x}$ in $\mathbb{R}^n$ such that $T(\mathbf{x}) = \mathbf{y}$. Let $S(\mathbf{y}) = \mathbf{x}$. Then S is a transformation from $\mathbb{R}^m$ to $\mathbb{R}^n$ and $T(S(\mathbf{y})) = \mathbf{y}$ for all $\mathbf{y}$ in $\mathbb{R}^m$.

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Invertible transformations (cont.)

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Assume on the other hand that T is onto and one-to-one. Then there is for each $\mathbf{y}$ in $\mathbb{R}^m$ a unique $\mathbf{x}$ in $\mathbb{R}^n$ such that $T(\mathbf{x}) = \mathbf{y}$. Let $S(\mathbf{y}) = \mathbf{x}$. Then S is a transformation from $\mathbb{R}^m$ to $\mathbb{R}^n$ and $T(S(\mathbf{y})) = \mathbf{y}$ for all $\mathbf{y}$ in $\mathbb{R}^m$. Let $\mathbf{x}$ be a vector in $\mathbb{R}^n$. Then $T(S(T(\mathbf{x}))) = T(\mathbf{x})$, and since T is one-to-one, it follows that $S(T(\mathbf{x})) = \mathbf{x}$. Thus, T is invertible.

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Invertible transformations (cont.)

3. There can at most be one transformation $S : \mathbb{R}^m \to \mathbb{R}^n$ such that $S \circ T(\mathbf{x}) = \mathbf{x}$ for all $\mathbf{x}$ in $\mathbb{R}^n$ and $T \circ S(\mathbf{y}) = \mathbf{y}$ for all $\mathbf{y}$ in $\mathbb{R}^m$. If such an S exists, then it is called the *inverse* of T and is denoted by T^{-1}.

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Invertible transformations (cont.)

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Proof: Assume that S_1 and S_2 are transformations from $\mathbb{R}^m$ to $\mathbb{R}^n$ such that $S_1 \circ T(\mathbf{x}) = S_2 \circ T(\mathbf{x}) = \mathbf{x}$ for all $\mathbf{x}$ in $\mathbb{R}^n$ and $T \circ S_1(\mathbf{y}) = T \circ S_2(\mathbf{y}) = \mathbf{y}$ for all $\mathbf{y}$ in $\mathbb{R}^m$.

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Invertible transformations (cont.)

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Invertible transformations (cont.)

3. There can at most be one transformation $S : \mathbb{R}^m \to \mathbb{R}^n$ such that $S \circ T(\mathbf{x}) = \mathbf{x}$ for all $\mathbf{x}$ in $\mathbb{R}^n$ and $T \circ S(\mathbf{y}) = \mathbf{y}$ for all $\mathbf{y}$ in $\mathbb{R}^m$. If such an S exists, then it is called the *inverse* of T and is denoted by T^{-1}.
Proof: Assume that S_1 and S_2 are transformations from $\mathbb{R}^m$ to $\mathbb{R}^n$ such that $S_1 \circ T(\mathbf{x}) = S_2 \circ T(\mathbf{x}) = \mathbf{x}$ for all $\mathbf{x}$ in $\mathbb{R}^n$ and $T \circ S_1(\mathbf{y}) = T \circ S_2(\mathbf{y}) = \mathbf{y}$ for all $\mathbf{y}$ in $\mathbb{R}^m$. If y is a vector in $\mathbb{R}^m$, then $S_1(\mathbf{y}) = S_1(T(S_2(\mathbf{y}))) = S_2(\mathbf{y})$. This shows that $S_1 = S_2$.

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Invertible linear transformations

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Invertible linear transformations

Let $T : \mathbb{R}^n \to \mathbb{R}^m$ be a linear transformation.

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Invertible linear transformations

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   Proof: Let A be the standard matrix of T.

Invertible linear transformations

Let $T : \mathbb{R}^n \to \mathbb{R}^m$ be a linear transformation.

1. A necessary (but not sufficient) condition for T to be onto is that $n \geq m$.
   Proof: Let A be the standard matrix of T. Then T is onto if and only if A has a pivot position in every row.

Invertible linear transformations

Let $T : \mathbb{R}^n \to \mathbb{R}^m$ be a linear transformation.

1. A necessary (but not sufficient) condition for T to be onto is that $n \geq m$.
   Proof: Let A be the standard matrix of T. Then T is onto if and only if A has a pivot position in every row. If $n < m$, then A has fewer rows than columns, so A cannot have a pivot position in every row, and T is therefore not onto.

Invertible linear transformations (cont.)

2. A necessary (but not sufficient) condition for T to be one-to-one is that $n \leq m$.

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   Proof: T is one-to-one if and only if A has a pivot position in every column. If $n > m$, then A has fewer columns than rows, so A cannot have a pivot position in every column, and T is therefore not one-to-one.

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Invertible linear transformations (cont.)

2. A necessary (but not sufficient) condition for T to be one-to-one is that $n \leq m$.
   Proof: T is one-to-one if and only if A has a pivot position in every column. If $n > m$, then A has fewer columns than rows, so A cannot have a pivot position in every column, and T is therefore not one-to-one.
3. It follows that a necessary (but not sufficient) condition for T to be invertible is that $n = m$.

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Invertible linear transformations (cont.)

4. If T is invertible, then T^{-1} is a linear transformation.

Invertible linear transformations (cont.)

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Proof: Suppose that T is invertible. Let **u** and **v** be vectors in $\mathbb{R}^n$ and let c be a scalar.

Invertible linear transformations (cont.)

4. If T is invertible, then T^{-1} is a linear transformation.
Proof: Suppose that T is invertible. Let $\mathbf{u}$ and $\mathbf{v}$ be vectors in $\mathbb{R}^n$ and let c be a scalar. Then

$$\begin{aligned} T^{-1}(\mathbf{u}+\mathbf{v}) &= T^{-1}(T(T^{-1}(\mathbf{u}))+T(T^{-1}(\mathbf{v}))) \\ &= T^{-1}(T(T^{-1}(\mathbf{u})+T^{-1}(\mathbf{v}))) \\ &= T^{-1}(\mathbf{u})+T^{-1}(\mathbf{v}) \end{aligned}$$

and

$$\begin{aligned} T^{-1}(c\mathbf{u}) = T^{-1}(cT(T^{-1}(\mathbf{u}))) &= T^{-1}(T(cT^{-1}(\mathbf{u}))) \\ &= cT^{-1}(\mathbf{u}), \end{aligned}$$

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Invertible linear transformations (cont.)

so T^{-1} is linear.

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Invertible linear transformations (cont.)

so T^{-1} is linear.

5. If $n = m$ and A is the standard matrix of T, then T is invertible if and only if there exists a matrix C such that $AC = CA = I_n$.

Invertible linear transformations (cont.)

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   Proof: Suppose that T is invertible, and let C be the standard matrix of T^{-1}.

Invertible linear transformations (cont.)

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   Proof: Suppose that T is invertible, and let C be the standard matrix of T^{-1}. If $\mathbf{x}$ is a vector in $\mathbb{R}^n$, then

$$AC\mathbf{x} = T(C\mathbf{x}) = T(T^{-1}\mathbf{x}) = \mathbf{x}$$

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   and

$$CA\mathbf{x} = T^{-1}(A\mathbf{x}) = T^{-1}(T\mathbf{x}) = \mathbf{x}.$$

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Invertible linear transformations (cont.)

so T^{-1} is linear.

5. If $n = m$ and A is the standard matrix of T, then T is invertible if and only if there exists a matrix C such that $AC = CA = I_n$.
   Proof: Suppose that T is invertible, and let C be the standard matrix of T^{-1}. If $\mathbf{x}$ is a vector in $\mathbb{R}^n$, then
   $$AC\mathbf{x} = T(C\mathbf{x}) = T(T^{-1}\mathbf{x}) = \mathbf{x}$$
   and
   $$CA\mathbf{x} = T^{-1}(A\mathbf{x}) = T^{-1}(T\mathbf{x}) = \mathbf{x}.$$
   It follows that $AC = CA = I_n$.

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Invertible linear transformations (cont.)

Suppose on the other hand that C is a matrix such that $AC = CA = I_n$.

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Invertible linear transformations (cont.)

Suppose on the other hand that C is a matrix such that $AC = CA = I_n$. Let $S(\mathbf{x}) = C\mathbf{x}$ for each $\mathbf{x}$ in $\mathbb{R}^n$.

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Invertible linear transformations (cont.)

Suppose on the other hand that C is a matrix such that $AC = CA = I_n$. Let $S(\mathbf{x}) = C\mathbf{x}$ for each $\mathbf{x}$ in $\mathbb{R}^n$. Then S is a linear transformation from $\mathbb{R}^n$ to $\mathbb{R}^n$, and

$$S \circ T(\mathbf{x}) = C(T(\mathbf{x})) = C(A\mathbf{x}) = (CA)\mathbf{x} = I_n\mathbf{x} = \mathbf{x}$$

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Invertible linear transformations (cont.)

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Invertible linear transformations (cont.)

Suppose on the other hand that C is a matrix such that $AC = CA = I_n$. Let $S(\mathbf{x}) = C\mathbf{x}$ for each $\mathbf{x}$ in $\mathbb{R}^n$. Then S is a linear transformation from $\mathbb{R}^n$ to $\mathbb{R}^n$, and

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Thus, T is invertible.

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The inverse of a matrix

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The inverse of a matrix

- An $n \times n$ matrix A is said to be *invertible* if there exists an $n \times n$ matrix C such that $AC = CA = I_n$.

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The inverse of a matrix

- An $n \times n$ matrix A is said to be *invertible* if there exists an $n \times n$ matrix C such that $AC = CA = I_n$.
- There can at most be one $n \times n$ matrix C such that $AC = CA = I_n$. If such a C exists, then it is called the *inverse* of A and is denoted by A^{-1}.

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- If $T : \mathbb{R}^n \to \mathbb{R}^n$ is a linear transformation and A is the standard matrix of T, then T is invertible if and only if A is.

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- If T and A are invertible, then A^{-1} is the standard matrix of T^{-1}.

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- If $T : \mathbb{R}^n \to \mathbb{R}^n$ is a linear transformation and A is the standard matrix of T, then T is invertible if and only if A is.
- If T and A are invertible, then A^{-1} is the standard matrix of T^{-1}.
- A matrix that is not invertible is sometimes called a *singular* matrix, and an invertible matrix is called a *nonsingular* matrix.

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Tomorrow's lecture

Tomorrow we shall

- further study *invertible* matrices,
- look at *the invertible matrix theorem*.

Section 2.3 in "Linear Algebras and Its Applications" (pages 111–116).

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