

Second Recitation Class

Linear Algebra

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Matrix Algebra

Definition

We define the sum of matrices and scalar multiples of matrices,

$$\begin{bmatrix} a_{11} & \cdots & a_{1m} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nm} \end{bmatrix} + \begin{bmatrix} b_{11} & \cdots & b_{1m} \\ \vdots & \ddots & \vdots \\ b_{n1} & \cdots & b_{nm} \end{bmatrix} = \begin{bmatrix} a_{11} + b_{11} & \cdots & a_{1m} + b_{1m} \\ \vdots & \ddots & \vdots \\ a_{n1} + b_{n1} & \cdots & a_{nm} + b_{nm} \end{bmatrix}$$

and scalar multiples of matrices,

$$k \begin{bmatrix} a_{11} & \cdots & a_{1m} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nm} \end{bmatrix} = \begin{bmatrix} ka_{11} & \cdots & ka_{1m} \\ \vdots & \ddots & \vdots \\ ka_{n1} & \cdots & ka_{nm} \end{bmatrix}$$

Question:

Can you verify the set of all $n \times m$ matrices is a vector space?

Calculate $A\bar{x}$

Definition

Dot product between each *row* vector in A and \bar{x} .

Theorem

Linear combination of *column* vectors in A .

Theorem

Algebraic rule for $A\bar{x}$,

1. $A(\bar{x} + \bar{y}) = A\bar{x} + A\bar{y}$, and
2. $A(k\bar{x}) = k(A\bar{x})$.

Linear Transformations

Definition

A function T from \mathbb{R}^m to \mathbb{R}^n is called a **linear transformation** if there exists an $n \times m$ matrix A such that

$$T(\bar{x}) = A\bar{x},$$

for all \bar{x} in the vector space \mathbb{R}^m

Question:

Why we can always find a matrix A ?

Example

Find the matrix I of identity map such that $Ix = x$ for $x \in \mathbb{R}^m$.

Find the matrix of linear transformation

Theorem

The matrix of a linear transformation T from \mathbb{R}^m to \mathbb{R}^n is

$$A = \begin{bmatrix} | & | & \cdots & | \\ T(\bar{e}_1) & T(\bar{e}_1) & \cdots & T(\bar{e}_m) \\ | & | & \cdots & | \end{bmatrix},$$

here we have \bar{e}_i be the i th standard basis in \mathbb{R}^m .

Question:

There is a subtle problem. This theorem is to find a matrix A for T , but we need a A to make it a linear transformation first.

Redefine Linear Transformation

Theorem

A transformation T from \mathbb{R}^m to \mathbb{R}^n is linear if and only if

1. $T(\bar{v} + \bar{w}) = T(\bar{v}) + T(\bar{w})$, for all vectors \bar{v} and \bar{w} in \mathbb{R}^m , and
2. $T(k\bar{v}) = kT(\bar{v})$, for all vectors $\bar{v} \in \mathbb{R}^m$ and all scalars k .

Proof.

With these two properties, we can expand $T(\bar{x})$ and find A using previous theorem.

Comment:

We notice that there are two equivalent definition of a linear transformation.

1. There is an $n \times m$ matrix A which satisfies $T(\bar{x}) = A\bar{x}$.
2. The transformation T is linear.

Linear Transformations in Geometry

Summary

In this part, we focus on linear transformations defined on \mathbb{R}^2 or \mathbb{R}^3 . The main task is to find the matrix of a linear transformation in geometry.

Procedure

1. Verify this linear transformation in geometry in linear.
2. Find the output of standard basis e_i , $i = \overline{1, m}$.
3. Write down the matrix as

$$A = \begin{bmatrix} | & | & \cdots & | \\ T(\overline{e_1}) & T(\overline{e_1}) & \cdots & T(\overline{e_m}) \\ | & | & \cdots & | \end{bmatrix},$$

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Scaling

Definiton

Scaling of \bar{x} with factor k is the map such that

$$\bar{x} \rightarrow k\bar{x}.$$

Solution

This transformation is linear from geometry.

For input e_1 and e_2 , we have

$$e_1 \rightarrow ke_1, \quad e_2 \rightarrow ke_2,$$

here we suppose the scalar is k .

Then the matrix is

$$A = \begin{bmatrix} k & 0 \\ 0 & k \end{bmatrix}.$$

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Orthogonal Projections

Definition

Orthogonal projection of \bar{x} on L is $proj_L(\bar{x})$ along L such that

$$(\bar{x} - proj_L(\bar{x})) \perp L.$$

Solution

Take a vector \bar{w} along line L , the orthogonal projection of \bar{x} on L is

$$proj_L(\bar{x}) = \left(\frac{\bar{x} \cdot \bar{w}}{\bar{w} \cdot \bar{w}} \right) \bar{w}.$$

If we take $\bar{u} = \bar{w} / \|\bar{w}\|$, then

$$proj_L(\bar{x}) = (\bar{x} \cdot \bar{u}) \bar{u}.$$

This transformation is linear.

Orthogonal Projections

Solution(Contd).

For input \overline{e}_1 and \overline{e}_2 , we have

$$\overline{e}_1 \rightarrow \begin{pmatrix} u_1^2 \\ u_1 u_2 \end{pmatrix}, \quad \overline{e}_2 \rightarrow \begin{pmatrix} u_2 u_1 \\ u_2^2 \end{pmatrix}.$$

Then the matrix is

$$A = \begin{bmatrix} u_1^2 & u_2 u_1 \\ u_1 u_2 & u_2^2 \end{bmatrix}$$

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Definiotn

The reflection of \bar{x} with respect to line L is $ref_L(\bar{x})$ such that

$$\bar{x} - proj_L(\bar{x}) = -(ref_L(\bar{x}) - proj_L(\bar{x})).$$

Solution

We can find $ref_L(\bar{x})$ using the orthogonal projection,

$$ref_L(\bar{x}) = 2proj_L(\bar{x}) - \bar{x}.$$

This is a linear transformation.

Solution(Contd.)

For input e_1 and e_2 ,

$$e_1 \rightarrow \begin{pmatrix} 2u_1^2 - 1 \\ 2u_1 u_2 \end{pmatrix}, \quad e_2 \rightarrow \begin{pmatrix} 2u_2 u_1 \\ 2u_2^2 - 1 \end{pmatrix}.$$

Then

$$A = \begin{bmatrix} 2u_1^2 - 1 & 2u_2 u_1 \\ 2u_1 u_2 & 2u_2^2 - 1 \end{bmatrix},$$

and we can write A as

$$A = \begin{bmatrix} a & b \\ b & -a \end{bmatrix}, \quad a^2 + b^2 = 1.$$

Question:

Why we have this form? (Hint: Consider $T(\overline{e_1})$ and $T(\overline{e_2})$ directly.)

Projection and Reflection

Example

Find a nonzero matrix A such that

$$A^2 = A,$$

and a nonzero matrix B such that

$$B^3 = B.$$

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Orthogonal Projections and Reflections in Space

Definition

In \mathbb{R}^3 , orthogonal projections and reflections are defined similar to \mathbb{R}^2 .

Example

Find the matrix of reflection with respect to plane
 $2x_1 + x_2 - 2x_3 = 0$.

Example

Find the matrix of orthogonal projection on plane
 $2x_1 + x_2 - 2x_3 = 0$.

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Rotation

Definition

The rotation in \mathbb{R}^2 with angle θ is a rotation centered at the origin in the counterclockwise direction.

Theorem

The matrix of a counterclockwise rotation in \mathbb{R}^2 through an angle θ is

$$\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}.$$

Rotation

Example

Check that if $A(\theta_1), A(\theta_2)$ is the matrix of counterclockwise rotation with angle θ_1, θ_2 , then

$$A(\theta_1)A(\theta_2) = A(\theta_1 + \theta_2).$$

Rotations Combined with a Scaling

For a scaling with scalar k and a counterclockwise rotation with angle θ , the combination of these two transformation is

$$\begin{bmatrix} k \cos \theta & -k \sin \theta \\ k \sin \theta & k \cos \theta \end{bmatrix}.$$

Note that these two transformations are commutative.

Theorem

A matrix of the form

$$\begin{bmatrix} a & -b \\ b & a \end{bmatrix}$$

represents a rotation combined with a scaling.

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Shears

Definition

A horizontal shear keep x_2 unchanged but change x_1 to $x_1 + kx_2$.

A vertical shear keep x_1 unchanged but change x_2 to $kx_1 + x_2$.

Theorem

The matrix of a horizontal shear is of the form

$$\begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix},$$

and the matrix of a vertical shear is of the form

$$\begin{bmatrix} 1 & 0 \\ k & 1 \end{bmatrix},$$

where k is an arbitrary constant.

Shears

Example

If A is a shear matrix, what is $(A - I_2)^2$? Consider $A^2\bar{x} - 2A\bar{x} + \bar{x}$.

Matrix Products

Motivation

The product of matrices is motivated by the composition of linear transformation. For linear transformations $T_2 : \mathbb{R}^m \rightarrow \mathbb{R}^k$ and $T_1 : \mathbb{R}^k \rightarrow \mathbb{R}^n$, we want to find the matrix of linear transformation $T_1 \circ T_2 : \mathbb{R}^m \rightarrow \mathbb{R}^n$, defined as

$$(T_1 \circ T_2)(\cdot) = T_1(T_2(\cdot)).$$

Question:

The composition of linear transformations is still linear transformation. Why?

Matrix Products

Solution

If the matrices of T_1, T_2 are A, B , we then have

$$T_1(T_2(e_j)) = A(Be_j) = Ab_j,$$

here e_j is the j th standard basis of \mathbb{R}^m , and the matrix of $T_1(T_2(\cdot))$ is

$$\begin{bmatrix} | & | & \cdots & | \\ Ab_1 & Ab_2 & \cdots & Ab_k \\ | & | & \cdots & | \end{bmatrix},$$

here b_j is the j th column of matrix B , $j = \overline{1, m}$.

Matrix Product

Example

Find the matrix product

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

Alternative Ways to Calculate Matrix Product

We can expand Ab_j in two ways.

Expand by Row (1)

If the i th row of A is a'_i , then

$$Ab_j = \begin{bmatrix} a'_1 \cdot b_j \\ \vdots \\ a'_n \cdot b_j \end{bmatrix}.$$

Then the product of A and B is

$$AB = \begin{bmatrix} a'_1 \cdot b_1 & \cdots & a'_1 \cdot b_m \\ \vdots & \ddots & \vdots \\ a'_n \cdot b_1 & \cdots & a'_n \cdot b_m \end{bmatrix}.$$

Alternative Ways to Calculate Matrix Product

Expand by Row (2)

Consider the product of a row vector and a matrix

$$\begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix} \begin{bmatrix} b_{11} & \cdots & b_{1m} \\ \vdots & \ddots & \vdots \\ b_{k1} & \cdots & b_{km} \end{bmatrix} = \begin{bmatrix} x_1 b_{11} & x_1 b_{1m} \\ + & + \\ \vdots & \vdots \\ + & + \\ x_k b_{k1} & x_k b_{km} \end{bmatrix} .$$

If we denote the i th row of B as b'_i , then

$$xB = \sum_{i=1}^k x_i b'_i.$$

This inspired us to use row vectors to calculate matrix product,

$$AB = \begin{bmatrix} \sum_{i=1}^k a_{1i} b'_i \\ \vdots \\ \sum_{i=1}^k a_{ni} b'_i \end{bmatrix} = \begin{bmatrix} a'_1 B \\ \vdots \\ a'_n B \end{bmatrix} .$$

The result is individual **row** of A times B .

Alternative Ways to Calculate Matrix Product

Exmample

Find the matrix product

$$\begin{bmatrix} 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & k \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

Alternative Ways to Calculate Matrix Product

Expand by Column (1)

If the i th column of A is a_i , then

$$Ab_j = \sum_{i=1}^k a_i b_{ij}.$$

Expand by Column (2)

Inspired by expansion by column, we can split the value that only relate to a_i ,

$$AB = \sum_{i=1}^k \begin{bmatrix} | & | & \cdots & | \\ a_i b_{i1} & a_i b_{i2} & \cdots & a_i b_{im} \\ | & | & \cdots & | \end{bmatrix} = \sum_{i=1}^k a_i b'_i.$$

This gives us a way to consider the matrix as the sum of product of **column vector in A** and **row vector in B** .

Properties of Matrix Product

Row Times Column

We have defined the product of a row vector and a column vector as

$$\begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = x_1 y_1 + \cdots + x_n y_n$$

Column Times Row

For a column vector times a row vector, we use the definition of matrix product,

$$\begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \begin{bmatrix} x_1 & \cdots & x_m \end{bmatrix} = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} x_1, \cdots, \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} x_m = \begin{bmatrix} x_1 y_1 & \cdots & x_m y_1 \\ \vdots & \ddots & \vdots \\ x_1 y_n & \cdots & x_m y_n \end{bmatrix}.$$

Properties of Matrix Product

Associativity

The product between matrices is associative, i.e.,

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

Commutative

The product between matrices is generally not commutative,

$$AB \neq BA.$$

Question: Can you find a matrix D that is commutative to any matrix?

Block Matrix

Theorem

If we separate A and B into blocks with appropriate sizes,

$$\begin{aligned} AB &= \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} \\ &= \begin{bmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{bmatrix} \end{aligned}$$

Gauss-Jordan in Matrix Product

Motivation

Each operation we perform in Gauss-Jordan elimination is to times a matrix on the original system.

$$E_n \cdots E_2 E_1 A = I$$

Example

Find all E to perform Gauss-Jordan on

$$A = \begin{bmatrix} 2 & 1 \\ 6 & 7 \end{bmatrix}.$$

Inverse of Matrix

Definition

For a linear transformation T , T^{-1} is its inverse if

$$T(T^{-1}(\cdot)) = id(\cdot), \quad \text{and} \quad T^{-1}(T(\cdot)) = id(\cdot),$$

here $id(\cdot)$ is the identity map.

Definition

The inverse of matrix A is the matrix of the inverse of its linear transformation, we denote it A^{-1} .

Theorem

The inverse of matrix A has properties

$$A^{-1}A = AA^{-1} = I.$$

Here I is the identity matrix.

Find the Inverse of a Matrix

Motivation

Recall in Gauss-Jordan, we find a series of matrices E_n, \dots, E_1 such that

$$E_n \cdots E_2 E_1 A = I,$$

then the product of $E_n \cdots E_1$ is the inverse of A .

Comment:

Note that this process may not always work.

Inverse of Matrix

Solution

To calculate easily, we write down the matrices A and I in an augmented matrix,

$$\left[\begin{array}{c|c} A & I \end{array} \right]$$

when we times do row manipulations on A , the same operation is performed on I , so

$$\begin{aligned} & \left[\begin{array}{c|c} E_n \cdots E_1 A & E_n \cdots E_1 I \end{array} \right] \\ &= \left[\begin{array}{c|c} I & A^{-1} \end{array} \right] \end{aligned}$$

Invertibility

Note that we may not always find the inverse of a matrix.

Theorem

A $n \times n$ matrix has inverse if and only if

- ▶ $\text{rank} A = n$,
- ▶ $\text{rref}(A) = I_n$.

Comment:

Note that the inverse exists if Gauss-Jordan elimination works.

Theorem

If B and A are $n \times n$ matrices such that

$$BA = I_n,$$

then B and A are both invertible and $A^{-1} = B, B^{-1} = A$.

Properties of Invertible Matrices

Theorem

If BA is invertible and A, B are $n \times n$ matrices, we have

$$(BA)^{-1} = A^{-1}B^{-1}.$$