Calculus and linear algebra for biomedical engineering Week 4: Inverse matrices and determinants

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Overview

- Matrix-by-matrix multiplication
- 2 Inverse matrices
- Computing inverses
- Determinants

Motivation

Suppose that we have variables $x_1, \ldots, x_m, y_1, \ldots, y_n, z_1, \ldots, z_k$, which are related by the following linear equations:

$$x = Ay$$
, $y = Bz$,

with suitable matrices A, B.

Is there a way to directly express the dependence of \mathbf{x} on \mathbf{z} as $\mathbf{x} = C\mathbf{z}$, in terms of a matrix C?

The answer is yes, and the matrix C is the matrix product of A and B.

Matrix multiplication

Definition.

Let $A \in \mathbb{R}^{m \times n}$ and $B \in \mathbb{R}^{n \times k}$. Denote the columns of B by $\mathbf{b}_1, \dots, \mathbf{b}_k$. Then the matrix $AB \in \mathbb{R}^{m \times k}$ is defined as

$$AB = (A\mathbf{b}_1|A\mathbf{b}_2|\dots|A\mathbf{b}_k) .$$

Hence the jth column of AB is obtained by multiplying A with the jth column of B.

Properties of matrix multiplication

Theorem. Let $A, A' \in \mathbb{R}^{m \times n}$ and $B, B' \in \mathbb{R}^{n \times k}$.

1 If $\mathbf{0}_{m,n}$ denotes the zero $m \times n$ matrix, then

$$\mathbf{0}_{k,m}A = \mathbf{0}_{k,n} , A\mathbf{0}_{n,k} = \mathbf{0}_{m,k} .$$

- ② Distributive laws: A(B + B') = AB + AB' and (A + A')B = AB + A'B.
- **3** Associativity: For all $C \in \mathbb{R}^{k \times \ell}$: A(BC) = (AB)C.
- Note: Matrix multiplication is not commutative!

If we regard vectors $\mathbf{x} \in \mathbb{R}^n$ as $n \times 1$ -matrices, matrix-by-vector multiplication is a special case of matrix-by-matrix multiplication. Recall motivation: If $\mathbf{x} = A\mathbf{y}$ and $\mathbf{y} = B\mathbf{z}$, how does \mathbf{x} depend on \mathbf{z} ? By associativity:

$$x = A(Bz) = (AB)z$$

Identity matrix

Definition. For $m \in \mathbb{N}$ let the $(m \times m)$ identity matrix l_m be defined by

Then $I_m \mathbf{x} = \mathbf{x}$, for all $\mathbf{x} \in \mathbb{R}^m$.

As a consequence, for arbitrary $A \in \mathbb{R}^{n \times m}$:

$$A = I_n A = A I_m$$
.

Inverse matrix

Definition. Let $A \in \mathbb{R}^{n \times n}$. Then $B \in \mathbb{R}^{n \times n}$ is called inverse of A. denoted $B = A^{-1}$. if

$$AB = BA = I_n$$
.

If A has an inverse, A is called invertible or regular.

Theorem. Let $A, B \in \mathbb{R}^{n \times n}$.

- (a) If $AB = I_n$ or $BA = I_n$, then $B = A^{-1}$.
- (b) If A and B are invertible, then AB is invertible with $(AB)^{-1} = B^{-1}A^{-1}$

Theorem. Let $A \in \mathbb{R}^{n \times n}$. Then the following are equivalent:

- (a) A is invertible.
- (b) For every vector $\mathbf{y} \in \mathbb{R}^n$, the system $A\mathbf{x} = \mathbf{y}$ has a unique solution.
- (c) The linear system Ax = 0 has rank n.

Sketch of proof:

- (a) \Rightarrow (b): Define $\mathbf{x} = A^{-1}\mathbf{y}$. Then $A\mathbf{x} = A(A^{-1}\mathbf{y}) = I_n\mathbf{y} = \mathbf{y}$.
- (b) \Rightarrow (a): For i = 1, ..., n, find the solution \mathbf{x}_i of

$$Ax_i = e_i$$
, with $e_i = (0, ..., 0, 1, 0, ..., 0)^T$

(entry 1 at *i*th position), and let $X = (\mathbf{x}_1 | \dots | \mathbf{x}_n)$. Then

$$AX = (Ax_1 | \dots | Ax_n) = (e_1 | \dots | e_n) = I_n.$$

Matrix inversion and systems of linear equations

Recall:

- Inverting a matrix $A \in \mathbb{R}^{n \times n}$ amounts to simultaneously solving n linear systems of equations $A\mathbf{x}_i = \mathbf{e}_i$. Here the coefficient matrix is always the same, only the right hand side changes.
- Linear systems are systematically solved by Gauss elimination. The steps in the Gauss elimination only depend on the coefficient matrix.
- ⇒ The Gauss algorithm should be useful for the computation of inverse matrices.

Matrix inversion via Gauss algorithm

Let $A \in \mathbb{R}^{n \times n}$ be given.

Introduce the extended matrix

$$\tilde{A} = (A|I_n) \in \mathbb{R}^{n \times 2n}$$
.

② Using Gauss elimination, transform the extended matrix to

$$\tilde{A}_1 = (B|C)$$
,

where $B \in \mathbb{R}^{n \times n}$ has upper triangular form. Make sure to apply all basic transformations to the right hand side also.

- 3 If B has rank < n, the matrix A is not invertible.
- If B has rank n, apply basic transformations to \tilde{A}_1 to obtain

$$\tilde{A}_2 = (I_n | C') .$$

Then $C' = A^{-1}$.

Inverting a 4×4 matrix

We wish to invert A via the Gauss algorithm, where

$$A = \left(\begin{array}{rrrr} 1 & -1 & -1 & 1 \\ 3 & -2 & -3 & 4 \\ 4 & -1 & -3 & 10 \\ 2 & 0 & -5 & -4 \end{array}\right) .$$

Following the usual scheme, we get

$$\tilde{A} = \begin{pmatrix} 1 & -1 & -1 & 1 & 1 & 0 & 0 & 0 \\ 3 & -2 & -3 & 4 & 0 & 1 & 0 & 0 \\ 4 & -1 & -3 & 10 & 0 & 0 & 1 & 0 \\ 2 & 0 & -5 & -4 & 0 & 0 & 0 & 1 \end{pmatrix}$$

$$\sim \begin{pmatrix} 1 & -1 & -1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & -3 & 1 & 0 & 0 \\ 0 & 3 & 1 & 6 & -4 & 0 & 1 & 0 \\ 0 & 2 & -3 & -6 & -2 & 0 & 0 & 1 \end{pmatrix}$$

Inverting a 4×4 matrix

Now the entries above the diagonal are also removed using basic transformations

Inverting a 4×4 matrix

Hence:

$$A^{-1} = \begin{pmatrix} -92 & 53 & -14 & -5 \\ -22 & 12 & -3 & -1 \\ -52 & 30 & -8 & -3 \\ 19 & -11 & 3 & 1 \end{pmatrix}$$

General definition of determinant

Definition. Let $A \in \mathbb{R}^{(n+1)\times (n+1)}$, and $1 \leq i \leq n+1$. Then the submatrix $A_i \in \mathbb{R}^{n \times n}$ is obtained by deleting the first column and ith row of A.

Definition. Let $A = (a_{i,j})_{i,j}$ be a matrix.

- For $A = (a) \in \mathbb{R}^{1 \times 1}$, we define $\det(A) = a$.
- 2 Let $\det : \mathbb{R}^{n \times n} \to \mathbb{R}$ be already defined. We then define $\det: \mathbb{R}^{(n+1)\times (n+1)} \to \mathbb{R}$ as follows:

$$\det(A) = a_{1,1}\det(A_1) - a_{2,1}\det(A_2) \pm \ldots + (-1)^n a_{n+1,1}\det(A_{n+1})$$
$$= \sum_{j=1}^{n+1} (-1)^{j-1} a_{j,1}\det(A_j)$$

Example: Computing a 2×2 -determinant

Let

$$A = \left(\begin{array}{cc} a & b \\ c & d \end{array}\right) \in \mathbb{R}^{2 \times 2}$$

Then, in the notation of the definition, the submatrices A_1 and A_2 are given as

$$A_1 = (d) \in \mathbb{R}^{1 \times 1}, \ A_2 = (b) \in \mathbb{R}^{1 \times 1}$$

and thus

$$\det(A) = \det\begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - cb.$$

Example: Computing a 3×3 determinant

Let

$$A = \left(\begin{array}{rrr} 1 & 7 & 4 \\ 3 & 3 & 2 \\ -1 & 6 & 5 \end{array}\right) .$$

Using the definition, and the formula for 2×2 determinants,

$$\det(A) = 1 \cdot \det\begin{pmatrix} 3 & 2 \\ 6 & 5 \end{pmatrix} - 3 \cdot \begin{pmatrix} 7 & 4 \\ 6 & 5 \end{pmatrix}$$

$$+ (-1) \cdot \begin{pmatrix} 7 & 4 \\ 3 & 2 \end{pmatrix}$$

$$= 1 \cdot (15 - 12) - 3(35 - 24) - 1(14 - 12)$$

$$= -32.$$

Determinants and invertibility

Theorem. (Cramer's rule)

Let

$$A = (\mathbf{a}_1 | \dots | \mathbf{a}_n) \in \mathbb{R}^{n \times n}$$
,

with $\det(A) \neq 0$. Let $\mathbf{y} \in \mathbb{R}^n$. For i = 1, ..., n, define B_i by replacing the *i*th column of A by \mathbf{y} ,

$$B_i = (\mathbf{a}_1 | \dots | \mathbf{a}_{i-1} | \mathbf{y} | \mathbf{a}_{i+1} | \mathbf{a}_n) \in \mathbb{R}^{n \times n}$$
.

Then the vector $\mathbf{x} = (x_1, \dots, x_n)^T$, with

$$x_i = \frac{\det(B_i)}{\det(A)} ,$$

is the unique solution of the linear equation $A\mathbf{x} = \mathbf{y}$. Consequence. Let $A \in \mathbb{R}^{n \times n}$. Then A is invertible if and only if $\det(A) \neq 0$.

Example: Solving a 2×2 system with Cramer's rule

We want to solve the equation $\begin{pmatrix} 2 & 3 \\ 4 & 8 \end{pmatrix} \mathbf{x} = \begin{pmatrix} 8 \\ -4 \end{pmatrix}$.

The coefficient matrix has determinant 4, thus there is a unique solution $\mathbf{x} = (x_1, x_2)^T$, with

$$x_1 = \frac{\det \begin{pmatrix} 8 & 3 \\ -4 & 8 \end{pmatrix}}{4} = \frac{76}{4} = 19$$

and

$$x_2 = \frac{\det \begin{pmatrix} 2 & 8 \\ 4 & -4 \end{pmatrix}}{4} = \frac{-40}{4} = -10$$

Note: For high dimensions, Cramer's rule is not an efficient way to solve linear systems!

Computational rules for determinants

Theorem. Let $A \in \mathbb{R}^{n \times n}$, $A = (a_{i,j})_{i,j=1...,n}$

- (a) Let $A^T = (b_{i,j})_{i,j=1...n}$, where $b_{i,j} = a_{j,i}$. Then $\det(A^T) = \det(A)$.
- (b) If A = BC for suitable matrices $B, C \in \mathbb{R}^{n \times n}$, then $\det(A) = \det(B)\det(C)$.
- (c) Let A be of the form

$$A = \left(\begin{array}{cc} A_1 & B \\ \mathbf{0}_{n-k,k} & A_2 \end{array}\right) ,$$

with $A_1 \in \mathbb{R}^{k \times k}$, $A_2 \in \mathbb{R}^{(n-k) \times (n-k)}$, and $C \in \mathbb{R}^{k \times (n-k)}$. Then

$$\det(A) = \det(A_1)\det(A_2) .$$

(d) If B is upper triangular with diagonal entries d_1, \ldots, d_n , then $\det(B) = d_1 d_2 \ldots d_n$.

Determinants via basic transformations

Theorem. Let $A, B \in \mathbb{R}^n$ be such that B is obtained from A by a basic transformation.

- (a) If B is obtained by swapping two rows, then det(B) = -det(A).
- (b) If B is obtained by multiplication of a row with a constant c, then det(B) = c det(A).
- (c) If B is obtained by adding a multiple of a column to another column, then det(B) = det(A).

The same rules apply if B is obtained by applying a basic transformation to columns of A.

Example: Using basic transformations

Recall the matrix A from above

$$A = \left(\begin{array}{rrr} 1 & 7 & 4 \\ 3 & 3 & 2 \\ -1 & 6 & 5 \end{array}\right) .$$

Using basic transformations, we can compute $\det(A)$ as

$$\det(A) = \det\begin{pmatrix} 1 & 7 & 4 \\ 0 & -18 & -10 \\ 0 & 13 & 9 \end{pmatrix}$$
$$= \det\begin{pmatrix} -18 & -10 \\ 13 & 9 \end{pmatrix}$$
$$= -162 + 130 = -32.$$

Example: 4 × 4 determinant

Consider the matrix

$$A = \left(\begin{array}{cccc} 1 & 2 & 3 & 1 \\ 7 & 3 & 1 & 3 \\ 5 & 0 & 0 & 4 \\ 2 & 0 & 0 & 0 \end{array}\right)$$

Then, after swapping rows

$$\det(A) = -\det\begin{pmatrix} 2 & 0 & 0 & 0 \\ 7 & 3 & 1 & 3 \\ 5 & 0 & 0 & 4 \\ 1 & 2 & 3 & 1 \end{pmatrix}$$

$$= -\det\begin{pmatrix} 2 & 7 & 5 & 1 \\ 0 & 3 & 0 & 2 \\ 0 & 1 & 0 & 3 \\ 0 & 2 & 4 & 1 \end{pmatrix} \quad (\text{using } \det(B) = \det(B^T))$$

Example: 4×4 determinant

$$\dots = -2\det\begin{pmatrix} 3 & 0 & 2 \\ 1 & 0 & 3 \\ 3 & 4 & 1 \end{pmatrix}$$
 (definition of det)
$$= 2\det\begin{pmatrix} 0 & 3 & 2 \\ 0 & 1 & 3 \\ 4 & 3 & 1 \end{pmatrix}$$
 (swapping columns)
$$= 8\det\begin{pmatrix} 3 & 2 \\ 1 & 3 \end{pmatrix}$$
 (definition of det)
$$= 8(9-2) = 56$$
.

Example: 4×4 -determinant using block matrices

Again consider the matrix

$$A = \left(\begin{array}{cccc} 1 & 2 & 3 & 1 \\ 7 & 3 & 1 & 3 \\ 5 & 0 & 0 & 4 \\ 2 & 0 & 0 & 0 \end{array}\right)$$

Swapping columns transforms A into a block matrix:

$$\det(A) = -\det\begin{pmatrix} 3 & 2 & 1 & 1 \\ 1 & 3 & 7 & 3 \\ 0 & 0 & 5 & 4 \\ 0 & 0 & 2 & 0 \end{pmatrix}$$
$$= -\det\begin{pmatrix} 3 & 2 \\ 1 & 3 \end{pmatrix} \det\begin{pmatrix} 5 & 4 \\ 2 & 0 \end{pmatrix}$$
$$= -(9-2) \cdot (-8) = 56$$

Summary

- Matrix-by-matrix multiplication and its properties.
- Identity matrix
- Invertible matrices
- Computing inverse matrices by the Gauss algorithm
- Definition and basic properties of determinants
- Determinants and invertibility. Cramer's rule.
- Computing determinants: Gauss algorithm, block matrices, etc.