

Linear Algebra

HSLU, Semester 4

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1 Vectors

1.1 Linear combination

A sum of scalings of vectors is called a linear combination of the vectors.

Let \vec{u} , \vec{v} be vectors, and a , b be scalars, $a, b \in \mathbb{R}$, then:

$$a \cdot \vec{u} + b \cdot \vec{v}$$

Generalizing this to a set of vectors $\vec{u}_1, \dots, \vec{u}_n$, and scalars a_1, \dots, a_n , we have:

$$\sum_{i=1}^n a_i \cdot \vec{u}_i$$

1.2 Cross product (vector product)

The cross product of two vectors \vec{u} and \vec{v} is a vector \vec{w} that is perpendicular to both \vec{u} and \vec{v} , and has a magnitude equal to the area of the parallelogram formed by \vec{u} and \vec{v} .

Let $\vec{u} = (x, y, z)$ and $\vec{v} = (t, s, q)$

$$\vec{u} \times \vec{v} = (yq - sz, -(xq - tz), xs - yt)$$

1.3 Unit vectors

Components of the unit vectors of the vector basis are as follow:

$$\vec{i} = \vec{e}_x = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \quad \vec{j} = \vec{e}_y = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \quad \vec{k} = \vec{e}_z = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

with their norms being $\|\vec{i}\| = \|\vec{j}\| = \|\vec{k}\| = 1$

2 Gaussian elimination

We want to archieve a stair of zeros on the bottom left of the matrix:

$$\left[\begin{array}{ccc|c} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & * & * \end{array} \right]$$

where $*$ is a non-zero number.

This also applies to higher dimensions, for example:

$$\left[\begin{array}{cccccc|c} * & * & * & * & * & * & * \\ 0 & * & * & * & * & * & * \\ 0 & 0 & * & * & * & * & * \\ 0 & 0 & 0 & * & * & * & * \\ 0 & 0 & 0 & 0 & * & * & * \\ 0 & 0 & 0 & 0 & 0 & * & * \\ 0 & 0 & 0 & 0 & 0 & 0 & * \end{array} \right]$$

2.1 Solution cases

2.1.1 1 possible solution

$$\left[\begin{array}{ccc|c} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & * & * \end{array} \right] \quad \begin{cases} x = * \\ y = * \\ z = * \end{cases}$$

2.1.2 Infinitely many solutions

$$\left[\begin{array}{ccc|c} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & 0 & 0 \end{array} \right] \quad \begin{cases} x = * \\ y = * \\ z = t \in \mathbb{R} \end{cases}$$

2.1.3 No solution

$$\left[\begin{array}{ccc|c} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & 0 & * \end{array} \right] \quad \begin{cases} x = * \\ y = * \\ 0 = * \rightarrow z \in \emptyset \end{cases}$$

3 dunno I forgot to take notes

4 Inverse and Determinant

Let A be a square matrix. The inverse of A , denoted by A^{-1} , satisfies the following:

$$A \cdot A^{-1} = I_n$$

where I_n is the identity matrix $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$

4.1 Determinant

For $M \in \mathbb{R}^{2 \times 2}$, $\det M = ad - bc$

For $M \in \mathbb{R}^{3 \times 3}$, use Sarrus rule:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & | & a_{11} & a_{12} \\ a_{21} & a_{22} & a_{23} & | & a_{21} & a_{22} \\ a_{31} & a_{32} & a_{33} & | & a_{31} & a_{32} \end{bmatrix}$$

$$\det M = a_{11}a_{22}a_{33} + a_{12}a_{23}a_{31} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31} - a_{11}a_{23}a_{32} - a_{12}a_{21}a_{33}$$

For $M \in \mathbb{R}^{n \times n}$, $n > 3$, apply row echelon and multiply the diagonal pivots.

4.2 Inverse

Let M be a 2×2 matrix and $\begin{matrix} a_{1 \leq i \leq m} \\ 1 \leq j \leq n \end{matrix}$

$$M^{-1} = \frac{1}{\det M} \operatorname{adj}(M), \quad \det M \neq 0$$

where:

$$A \in \mathbb{R}^{2 \times 2}, \operatorname{adj}(A) = \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

For $B \in \mathbb{R}^{3 \times 3}$, apply Gauss - Jordan elimination $[A|I_n] \Leftrightarrow [I_n|A^{-1}]$

4.3 Linear systems

$$Ax = b \iff x = A^{-1} \bullet b$$

5 Add something here

6 Linear independency

6.1 with 1 vector

If we only have one vector \vec{v}_k at any \mathbb{R}^k , then it is independent.

7 Linearità

$$f(\alpha v + \beta w) = \alpha f(v) + \beta f(w)$$

7.1 Rotation

7.1.1 Clockwise

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

7.1.2 Counter-clockwise

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

8 Inverse map

$L(v) = w$ is invertible if and only if v is a basis (linear independent) and w is also linear independent.

$$w = L(v) \iff v = L^{-1}(w)$$

If $L : \mathbb{R}^n \mapsto \mathbb{R}^n$ has a matrix A , then

$$L^{-1}(x) = A^{-1}x$$

then L is invertible iff $\det A \neq 0$

Follows that, for $L(e_1) = v$ and $L(e_2) = w$, L is invertible iff v, w are linearly independent, as e_1, e_2 are linearly independent.

To calculate $L^{-1} \in \mathbb{R}^2$:

$$L^{-1} = \frac{1}{\det L} \operatorname{adj} L = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

8.1 Kernel and image

Let A be a matrix, then:

$$\begin{aligned} \ker A &= \{x \in \mathbb{R}^n \mid Ax = 0\} \\ \operatorname{im} A &= \{Ax \mid x \in \mathbb{R}^n\} \implies \operatorname{im} A = \operatorname{span}\{a_1, \dots, a_n\} \end{aligned}$$