

Vector space

Definition: A *vector space* $(V, +, \cdot)$ over a field $(\mathbb{K}, +, \cdot)$ is a set V with a binary operation $+$ on V and a binary operation $\cdot : \mathbb{K} \times V \rightarrow V$ called *scalar multiplication* s.t.:

- ▶ $(V, +)$ is an Abelian group
- ▶ $\forall a, b \in \mathbb{K}, \forall \mathbf{v} \in V : (a \cdot b) \cdot \mathbf{v} = a \cdot (b \cdot \mathbf{v})$
- ▶ $\forall \mathbf{v} \in V : 1 \cdot \mathbf{v} = \mathbf{v}$
- ▶ $\forall a, b \in \mathbb{K}, \forall \mathbf{v} \in V : (a + b) \cdot \mathbf{v} = (a \cdot \mathbf{v}) + (b \cdot \mathbf{v})$
- ▶ $\forall a \in \mathbb{K}, \forall \mathbf{u}, \mathbf{v} \in V : a \cdot (\mathbf{u} + \mathbf{v}) = (a \cdot \mathbf{u}) + (a \cdot \mathbf{v})$

Elements of \mathbb{K} are called *scalars*, elements of V are called *vectors*.

Distinguish the zero scalar $0 \in \mathbb{K}$ and the zero vector $\mathbf{0} \in V$.

We have opposite scalars $-a \in \mathbb{K}$ and opposite vectors $-\mathbf{v} \in V$.

There are inverse scalars $a^{-1} \in \mathbb{K}$ but *no inverse vectors* \mathbf{v}^{-1} !

Products $\mathbf{v} \cdot a$ and $\mathbf{v} \cdot \mathbf{u}$ are not defined and are *formally wrong!*

The product symbol \cdot is often omitted and has priority to $+$.

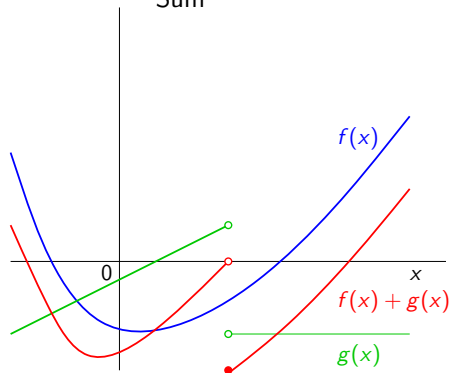
Examples

- ▶ The *arithmetic* vector space \mathbb{K}^n of dimension n over \mathbb{K} .
Vectors are ordered n -tuples of elements of \mathbb{K} .
Additions and scalar multiplications are done componentwise.
Any field \mathbb{K} yields the vector space \mathbb{K}^1 of the same cardinality.
- ▶ $\mathbb{K}^{m \times n}$... matrices of order $m \times n$ over \mathbb{K} .
- ▶ $V = \{0\}$ the *trivial* vector space over any field \mathbb{K} .
- ▶ Polynomials with coefficients in \mathbb{K} .
- ▶ Polynomials of bounded degree.

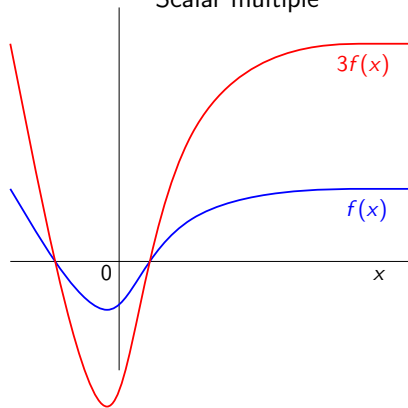
The vector space of real functions over the field \mathbb{R}

Vectors f, g are the real functions of real variable

Sum



Scalar multiple



Set systems as vector spaces

Let \mathcal{X} be a system of subsets of a set X that is closed under taking symmetric difference Δ .

Then $(\mathcal{X}, \Delta, \cdot)$ is a vector space over \mathbb{Z}_2 , where \cdot is defined as: $0 \cdot A = \emptyset$, the neutral element of Δ , and $1 \cdot A = A$ for all $A \in \mathcal{X}$.

Observe that $\forall A, B, C \in \mathcal{X} : (A \Delta B) \Delta C = A \Delta (B \Delta C)$,

since the result contains those elements of X that belong to an odd number of sets A, B and C .

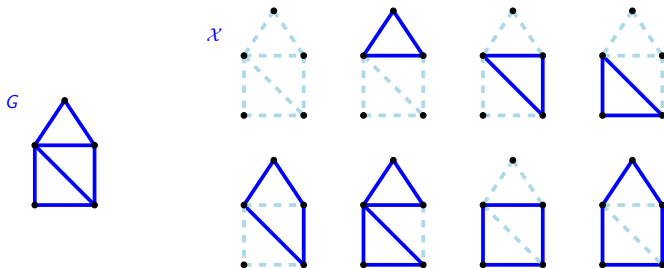
We may represent any $A \in \mathcal{X}$ by its *characteristic function*

$$\chi_A : X \rightarrow \mathbb{Z}_2 \text{ defined as } \chi_A(x) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}$$

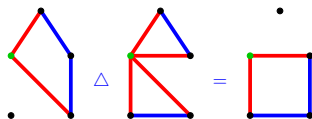
Observation: The set $A \Delta B$ has the characteristic function $\chi_{A \Delta B} = \chi_A + \chi_B$, since $1 + 1 = 0$ in \mathbb{Z}_2 .

A vector space on graphs

Let $X = E_G$ and \mathcal{X} contain those edge sets A such that every vertex of G is incident with an even number of edges in A . Such sets A induce the so called *even subgraphs* of G .



Observe that Δ preserves the even degree, since the symmetric difference of two sets of *even* cardinality, namely the *edges* incident to a *vertex*, has also an *even* cardinality.



Properties of vector spaces

Since $(V, +)$ is a group, we have already proved:

- ▶ uniqueness of the vector $\mathbf{0}$,
- ▶ uniqueness of $-\mathbf{u}$,
- ▶ correctness of equiv. transforms: $\mathbf{u} = \mathbf{v} \Leftrightarrow \mathbf{u} + \mathbf{w} = \mathbf{v} + \mathbf{w}$,
- ▶ solubility of equations: $\mathbf{x} + \mathbf{u} = \mathbf{v} \Leftrightarrow \mathbf{x} = \mathbf{v} - \mathbf{u}$.

Observation: For any $\mathbf{v} \in V$ and $a \in \mathbb{K}$ holds: $0\mathbf{v} = a\mathbf{0} = \mathbf{0}$.

Proof:

$$0\mathbf{v} = 0\mathbf{v} + \mathbf{0} = 0\mathbf{v} + 0\mathbf{v} - 0\mathbf{v} = (0 + 0)\mathbf{v} - 0\mathbf{v} = 0\mathbf{v} - 0\mathbf{v} = \mathbf{0}$$

$$a\mathbf{0} = a\mathbf{0} + \mathbf{0} = a\mathbf{0} + a\mathbf{0} - a\mathbf{0} = a(\mathbf{0} + \mathbf{0}) - a\mathbf{0} = a\mathbf{0} - a\mathbf{0} = \mathbf{0}$$

Observation: For any $\mathbf{v} \in V$ holds $(-1)\mathbf{v} = -\mathbf{v}$.

Proof: $(-1)\mathbf{v} + \mathbf{v} = (-1)\mathbf{v} + 1\mathbf{v} = (-1 + 1)\mathbf{v} = 0\mathbf{v} = \mathbf{0}$

Observation: If $a\mathbf{v} = \mathbf{0}$ then $a = 0$ or $\mathbf{v} = \mathbf{0}$.

Proof: If $a \neq 0$ then $\mathbf{v} = 1\mathbf{v} = a^{-1}a\mathbf{v} = a^{-1}\mathbf{0} = \mathbf{0}$.

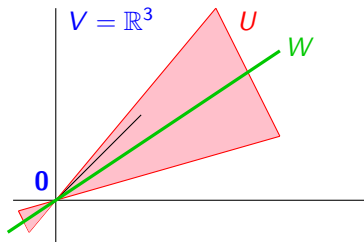
Subspace

Definition: Let V be a vector space over \mathbb{K} then a **subspace** U is a nonempty subset of V satisfying:

- ▶ $\forall \mathbf{u}, \mathbf{v} \in U : \mathbf{u} + \mathbf{v} \in U$
- ▶ $\forall \mathbf{v} \in U, \forall a \in \mathbb{K} : a\mathbf{v} \in U$

Example: A plane U through the origin $\mathbf{0}$ is a subspace of $V = \mathbb{R}^3$.

A line $W \subset U$ through the origin is a subspace of U and of V as well.



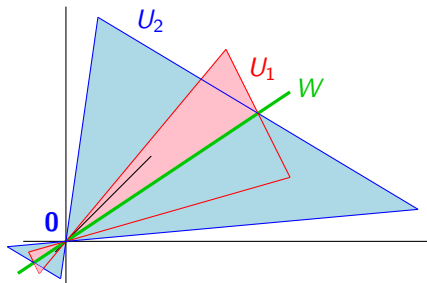
Polynomials of degree ≤ 5 form a subspace of the function space.

Observation: Any subspace is also a vector space, because the existential axioms follow the closure: $\mathbf{0} = 0\mathbf{v} \in U$ and $-\mathbf{v} = (-1)\mathbf{v} \in U$. Other axioms hold already on V .

Intersection of subspaces

Theorem: Let $(U_i, i \in I)$ be any system of subspaces of a space V . The intersection of this system $\bigcap_{i \in I} U_i$ is also a subspace of V .

Example: Planes U_1 and U_2 are subspaces of the vector space \mathbb{R}^3 . Their intersection is the line W . It is also a subspace of \mathbb{R}^3 .



Intersection of subspaces

Theorem: Let $(U_i, i \in I)$ be any system of subspaces of a space V . The intersection of this system $\bigcap_{i \in I} U_i$ is also a subspace of V .

Proof: Let $W = \bigcap_{i \in I} U_i$. We show that W is closed on $+$ and \cdot .

$\forall \mathbf{u}, \mathbf{v} \in W :$

$$\mathbf{u}, \mathbf{v} \in W \Rightarrow \forall i \in I : \mathbf{u}, \mathbf{v} \in U_i \Rightarrow \forall i \in I : \mathbf{u} + \mathbf{v} \in U_i \Rightarrow \mathbf{u} + \mathbf{v} \in W$$

$\forall a \in \mathbb{K}, \mathbf{v} \in W :$

$$\mathbf{v} \in W \Rightarrow \forall i \in I : \mathbf{v} \in U_i \Rightarrow \forall i \in I : a\mathbf{v} \in U_i \Rightarrow a\mathbf{v} \in W$$

Note that when $I = \emptyset$, then the empty intersection

$$W = \bigcap_{i \in \emptyset} U_i = V \text{ is a subspace of } V \text{ itself.}$$

Linear combination, linear hull

Definition: The *linear hull* $\mathcal{L}(X)$ of a subset X of a vector space V is the intersection of all subspaces U of V that contain X .

Formally $\mathcal{L}(X) = \bigcap \{U : X \subseteq U, U \text{ is a subspace of } V\}$

It is also called *the subspace generated by X*

or the *linear span* and may be denoted by $\text{span}(X)$.

Examples: For $V = \mathbb{R}^3$, $\mathcal{L}(\{(2, 2, 2)^T\}) = \{(a, a, a)^T, a \in \mathbb{R}\}$

... the line containing points having all three coordinates identical

$\mathcal{L}(\{(1, 0, 0)^T, (0, 1, 0)^T\}) = \{(a, b, 0)^T, a, b \in \mathbb{R}\}$

... the plane determined by the first two axes

Definition: A *linear combination* of vectors $\mathbf{v}_1, \dots, \mathbf{v}_k \in V$ over \mathbb{K} is any vector $\mathbf{u} = a_1 \mathbf{v}_1 + \dots + a_k \mathbf{v}_k$ where $a_1, \dots, a_k \in \mathbb{K}$.

Theorem: Let V be a vector space over \mathbb{K} and X be a subset of V . Then $\mathcal{L}(X)$ is the set of all linear combinations of vectors from X .

Example for $W = \{\mathbf{x} : \mathbf{Ax} = \mathbf{0}\}$

$$\text{For } \mathbf{A} = \begin{pmatrix} 1 & -2 & 4 & 0 & 3 \\ -2 & 4 & -6 & 6 & 0 \\ 4 & -8 & 13 & -9 & -4 \\ 3 & -6 & 9 & -9 & 1 \end{pmatrix} \sim \sim \begin{pmatrix} 1 & -2 & 4 & 0 & 3 \\ 0 & 0 & 1 & 3 & 2 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

we solve $W = \{p_1(2, 1, 0, 0, 0)^T + p_2(12, 0, -3, 1, 0)^T : p_1, p_2 \in \mathbb{R}\}$

1. W could be viewed the *intersection* of the following subspaces:

$$U_1 = \{\mathbf{x} : (1, -2, 4, 0, 3) \cdot \mathbf{x} = 0\}$$

$$U_2 = \{\mathbf{x} : (-2, 4, -6, 6, 0) \cdot \mathbf{x} = 0\}$$

$$U_3 = \{\mathbf{x} : (4, -8, 13, -9, -4) \cdot \mathbf{x} = 0\}$$

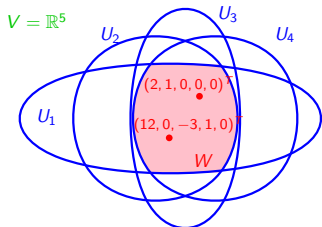
$$U_4 = \{\mathbf{x} : (3, -6, 9, -9, 1) \cdot \mathbf{x} = 0\} \quad \mathbb{R}^5.$$

Each U_i corresponds to one equation of the system.

(Each is a *hyperplane* in

2. The same subspace W was already described as a *set of linear combinations*
 $W = \mathcal{L}((2, 1, 0, 0, 0)^T, (12, 0, -3, 1, 0)^T)$

Correctness of Gaussian elimination and backward substitution yields that both ways describe the same space.



Proof of the theorem

Theorem: Let V be a vector space over \mathbb{K} and X be a subset of V . Then $\mathcal{L}(X)$ is the set of all linear combinations of vectors from X .

For $W_1 = \bigcap_{X \subseteq U_i \subseteq V} U_i$, $W_2 = \left\{ \sum_{i=1}^k a_i \mathbf{v}_i : k \in \mathbb{N}, a_i \in \mathbb{K}, \mathbf{v}_i \in X \right\}$.

we want to show $W_1 = \mathcal{L}(X) = W_2$.

W_2 is a subspace, because it is closed on scalar multiples

$$\mathbf{u} \in W_2 \Rightarrow \mathbf{u} = \sum_{i=1}^k a_i \mathbf{v}_i \Rightarrow \alpha \mathbf{u} = \alpha \sum_{i=1}^k a_i \mathbf{v}_i = \sum_{i=1}^k (\alpha a_i) \mathbf{v}_i \Rightarrow \alpha \mathbf{u} \in W_2$$

and analogously also on addition

$$\mathbf{u}, \mathbf{u}' \in W_2 \Rightarrow \dots \Rightarrow \mathbf{u} + \mathbf{u}' \in W_2$$

Since $X \subseteq W_2$, we have W_2 among the intersecting subspaces U_i .

Hence $W_1 \subseteq W_2$.

Every U_i contains X and is closed on addition and scalar multiples.

Thus every U_i contains all linear combinations of vectors of X .

Hence $\forall U_i : W_2 \subseteq U_i \Rightarrow W_2 \subseteq W_1$.

Spaces determined by a matrix

Definition: The *kernel* of $\mathbf{A} \in \mathbb{K}^{m \times n}$ is $\ker(\mathbf{A}) = \{\mathbf{x} \in \mathbb{K}^n : \mathbf{A}\mathbf{x} = \mathbf{0}\}$, the *row space* $\mathcal{R}(\mathbf{A}) \subseteq \mathbb{K}^n$ is the linear hull of the rows of \mathbf{A} , the *column space* $\mathcal{C}(\mathbf{A}) \subseteq \mathbb{K}^m$ is the linear hull of the columns of \mathbf{A} .

Example: For the matrix $\mathbf{A} = \begin{pmatrix} 1 & 2 & 0 & 1 \\ 2 & 0 & 2 & 1 \\ 1 & 1 & 2 & 0 \end{pmatrix} \in \mathbb{Z}_3^{3 \times 4}$ we get

The row space:

$$\mathcal{R}(\mathbf{A}) = \left\{ \begin{array}{lll} (0, 0, 0, 0)^T, & (1, 2, 0, 1)^T, & (2, 1, 0, 2)^T, \\ (2, 0, 2, 1)^T, & (0, 2, 2, 2)^T, & (1, 1, 2, 0)^T, \\ (1, 0, 1, 2)^T, & (2, 2, 1, 0)^T, & (0, 1, 1, 1)^T \end{array} \right\} \subseteq \mathbb{Z}_3^4$$

The column space:

$$\mathcal{C}(\mathbf{A}) = \left\{ \begin{array}{lll} (0, 0, 0)^T, & (1, 2, 1)^T, & (2, 1, 2)^T, \\ (2, 0, 1)^T, & (0, 2, 2)^T, & (1, 1, 0)^T, \\ (1, 0, 2)^T, & (2, 2, 0)^T, & (0, 1, 1)^T \end{array} \right\} \subseteq \mathbb{Z}_3^3$$

Spaces determined by a matrix

Definition: The *kernel* of $\mathbf{A} \in \mathbb{K}^{m \times n}$ is $\ker(\mathbf{A}) = \{\mathbf{x} \in \mathbb{K}^n : \mathbf{A}\mathbf{x} = \mathbf{0}\}$, the *row space* $\mathcal{R}(\mathbf{A}) \subseteq \mathbb{K}^n$ is the linear hull of the rows of \mathbf{A} , the *column space* $\mathcal{C}(\mathbf{A}) \subseteq \mathbb{K}^m$ is the linear hull of the columns of \mathbf{A} .

Formally: $\mathcal{C}(\mathbf{A}) = \{\mathbf{u} \in \mathbb{K}^m : \mathbf{u} = \mathbf{A}\mathbf{x}, \mathbf{x} \in \mathbb{K}^n\}$
 $\mathcal{R}(\mathbf{A}) = \{\mathbf{v} \in \mathbb{K}^n : \mathbf{v} = \mathbf{A}^T \mathbf{y}, \mathbf{y} \in \mathbb{K}^m\}$

Observation: The kernel $\ker(\mathbf{A})$ is a subspace of \mathbb{K}^n .

Observation: Elementary transforms do not alter $\mathcal{R}(\mathbf{A})$ nor $\ker(\mathbf{A})$.

Observation: $\forall \mathbf{v} \in \mathcal{R}(\mathbf{A}), \forall \mathbf{x} \in \ker(\mathbf{A}) : \mathbf{v}^T \mathbf{x} = 0$.

Proof: $\mathbf{v}^T \mathbf{x} = (\mathbf{A}^T \mathbf{y})^T \mathbf{x} = \mathbf{y}^T \mathbf{A}\mathbf{x} = \mathbf{y}^T \mathbf{0} = 0$ for some $\mathbf{y} \in \mathbb{K}^m$.

Corollary: Each subspace of an arithmetic vector space is a set of solutions of some homogeneous system of linear equations.

Proof: Take a basis of V as rows of \mathbf{B} , then take a basis of $\ker(\mathbf{B})$ as rows of \mathbf{A} . Now: $\ker(\mathbf{A}) = \ker(\ker(\mathbf{B})) = \mathcal{R}(\mathbf{B}) = V$.