General vector spaces

7.1 Vector space in its full glory

$$\begin{array}{c}
 \mathbb{R} & \left(\begin{array}{c} addddson \\ scaling \end{array} \right), \quad \mathbb{R}^{n\times n}, \quad \bigcup \\ & \left(\begin{array}{c} scaling \end{array} \right), \quad \mathbb{R}^{n\times n}, \quad \bigcup \\ & \left(\begin{array}{c} scaling \end{array} \right), \quad \mathbb{R}^{n\times n}, \quad \bigcup \\ & \left(\begin{array}{c} scaling \end{array} \right), \quad \mathbb{F} \in \{\mathbb{R}, \mathbb{C}\} \\ \hline \\ & \mathbb{R}, \mathbb{C} \rightarrow field \quad \left(\begin{array}{c} calcalistical for numbers \end{array} \right), \quad \mathbb{F} \in \{\mathbb{R}, \mathbb{C}\} \\ \hline \\ & \mathbb{R}, \mathbb{C} \rightarrow field \quad \left(\begin{array}{c} calcalistical for numbers \end{array} \right), \quad \mathbb{F} \in \{\mathbb{R}, \mathbb{C}\} \\ \hline \\ & \mathbb{P} = \{\mathbb{P} = \mathbb{P} = \{\mathbb{R}, \mathbb{R}\} \\ \hline \\ & \mathbb{P} = \{\mathbb{P} = \mathbb{P} = \{\mathbb{R}, \mathbb{R}\} \\ \hline \\ & \mathbb{P} = \{\mathbb{P} = \mathbb{P} = \{\mathbb{P} = \mathbb{P} \\ \\ & \mathbb{P} = \{\mathbb{P} = \mathbb{P} = \{\mathbb{P} = \mathbb{P} \\ \\ & \mathbb{P} = \{\mathbb{P} = \mathbb{P} \\ \\ \\ & \mathbb{P} = \{\mathbb{P} = \mathbb{P} \\ \\ & \mathbb{P} = \{\mathbb{P} = \mathbb{P} \\ \\ \\ & \mathbb{P} = \{\mathbb{P} = \mathbb{P} \\ \\ & \mathbb{P}$$

Example 7.2. \mathbb{R}^n and \mathbb{C}^n . At this point, we are very familiar with the space \mathbb{F}^n , where the vectors have *n* components consisting of numbers from \mathbb{F} and the addition and scalar multiplication is done componentwise:

$$\lambda \in \mathbb{F}, \ \mathbf{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \Rightarrow \lambda \mathbf{v} = \lambda \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} := \begin{pmatrix} \lambda v_1 \\ \vdots \\ \lambda v_n \end{pmatrix}$$
$$\mathbf{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}, \ \mathbf{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \Rightarrow \mathbf{u} + \mathbf{v} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix} + \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} := \begin{pmatrix} u_1 + v_1 \\ \vdots \\ u_n + v_n \end{pmatrix}$$

Now, this is now just a special case of an \mathbb{F} -vector space.

Example 7.3. Matrices. The set of matrices $V := \mathbb{F}^{m \times n}$ together with the matrix addition and scalar multiplication

$$\underbrace{\begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix}}_{A} + \underbrace{\begin{pmatrix} b_{11} & \cdots & b_{1n} \\ \vdots & \vdots \\ b_{m1} & \cdots & b_{mn} \end{pmatrix}}_{B} := \underbrace{\begin{pmatrix} a_{11} + b_{11} & \cdots & a_{1n} + b_{1n} \\ \vdots & \vdots \\ a_{m1} + b_{m1} & \cdots & a_{mn} + b_{mn} \end{pmatrix}}_{A+B}.$$

$$\underbrace{\lambda \cdot \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix}}_{A} := \underbrace{\begin{pmatrix} \lambda a_{11} & \cdots & \lambda a_{1n} \\ \vdots & \vdots \\ \lambda a_{m1} & \cdots & \lambda a_{mn} \end{pmatrix}}_{\lambda \cdot A}.$$
defines also an F-vector space.
$$hatrices \quad are \quad abstruct vectors$$

Example 7.4. Functions. Let $\mathcal{F}(\mathbb{R})$ be the set of functions $\mathbf{f} : \mathbb{R} \to \mathbb{R}$. For all $\alpha \in \mathbb{R}$ and $\mathbf{f}, \mathbf{g} \in \mathcal{F}(\mathbb{R})$ define $\alpha \cdot \mathbf{f}$ and $\mathbf{f} + \mathbf{g}$ by

$$\begin{array}{rcl} (\alpha \cdot \mathbf{f})(x) & := & \alpha \cdot \mathbf{f}(x) \\ (\mathbf{f} + \mathbf{g})(x) & := & \mathbf{f}(x) + \mathbf{g}(x) \end{array}$$

for all $x \in \mathbb{R}$.



This is a natural definition for the α -multiple of a function and the sum of two functions.

$$\begin{aligned} & \int f: R \to R \quad (for an formal and colored) \\ & \int f: R \to R \quad (for an formal and colored) \\ & \int f: R \to R \quad (for an formal for an formal for an equation for an equation for a formal f$$

7 General vector spaces

To show: $f,g \in P(\mathbb{R}), \alpha \in \mathbb{R} \implies f+g \in P(\mathbb{R})$ $\alpha \cdot f \in P(\mathbb{R})$ (*) (closed under + and .)

Are (1) - (8) satisfied?

4

Now checking (1)-(8) is very fast because:



Proposition & Definition 7.7. Linear subspace \mathcal{U} is again a vech
SpaceLet V be an \mathbb{F} -vector space and let U be a non-empty subset of V, which is closed
under vector addition and scalar multiplication of V, which means(a) for all $\mathbf{u}, \mathbf{v} \in U$, we have $\mathbf{u} + \mathbf{v} \in U$ and(b) for all $\alpha \in \mathbb{F}$ and $\mathbf{u} \in U$, we have $\alpha \cdot \mathbf{u} \in U$.Then U is also an \mathbb{F} -vector space. In this case, U is called a linear subspace of V
or in short a subspace of V $\forall A \in \mathbb{F}$ $\forall A \in \mathbb{F}$ <

7.2 Linear subspaces

Example 7.8. Quadratic polynomials. Let $\mathcal{P}_2(\mathbb{R})$ be the set of all polynomials with degree ≤ 2 , which means

all functions $\mathbf{p}: \mathbb{R} \to \mathbb{R}$, $\mathbf{p}(x) = a_2 x^2 + a_1 x + a_0$ with $a_2, a_1, a_0 \in \mathbb{R}$.

Is $\mathcal{P}_2(\mathbb{R})$ with the vector addition + and \cdot from $\mathcal{F}(\mathbb{R})$ a vector space?

Obviously, $\mathcal{P}_2(\mathbb{R}) \subset \mathcal{F}(\mathbb{R})$ and $\mathcal{P}_2(\mathbb{R}) \neq \emptyset$. Using Proposition 7.7 we only have to check that $\mathcal{P}_2(\mathbb{R})$ is closed under + and \cdot , which means that we have to check (a) and (b):

Let $\mathbf{p}, \mathbf{q} \in \mathcal{P}_2(\mathbb{R})$ and $\alpha \in \mathbb{R}$. Then, there are $a_2, a_1, a_0, b_2, b_1, b_0 \in \mathbb{R}$ such that

 $\mathbf{p}(x) = a_2 x^2 + a_1 x + a_0$ and $\mathbf{q}(x) = b_2 x^2 + b_1 x + b_0$.

Hence:

$$(\mathbf{p} + \mathbf{q})(x) = \mathbf{p}(x) + \mathbf{q}(x) = (a_2 x^2 + a_1 x + a_0) + (b_2 x^2 + b_1 x + b_0) = (a_2 + b_2) x^2 + (a_1 + b_1) x + (a_0 + b_0), \quad \in \mathcal{F}_2(\mathcal{R}) = (\alpha \cdot \mathbf{p})(x) = \alpha \cdot \mathbf{p}(x) = \alpha \cdot (a_2 x^2 + a_1 x + a_0) = (\alpha a_2) x^2 + (\alpha a_1) x + (\alpha a_0)$$

We conclude that $\mathbf{p} + \mathbf{q} \in \mathcal{P}_2(\mathbb{R})$ and $\alpha \cdot \mathbf{p} \in \mathcal{P}_2(\mathbb{R})$. The set $\mathcal{P}_2(\mathbb{R})$ is a subspace of $\mathcal{F}(\mathbb{R})$ and a vector space for its own.

Analogously, for $n \in \mathbb{N}_0$, we define $\mathcal{P}_n(\mathbb{R})$ as the set of all polynomials with degree $\leq n$. It forms again a vector space with the operations + and \cdot from \mathcal{F} .

Same proof!

Example 7.9. Upper triangular matrices Let $n \in \mathbb{N}$ and $\mathbb{R}^{n \times n}$ the set of all upper triangular matrices $A \in \mathbb{R}^{n \times n}$. The operations + and \cdot are the same as before for all matrices.

Check:
$$\nabla IR^{n \times n}$$
 hon-empty and (a), (b).
 $C = \nabla R^{n \times n}$

Example 7.10. The set of all matrices U in the following form:

dimension Foresight: dim (Pn) = n+1

7.3 Recollection: basis, dimension and other stuff

Let $\mathbb{F} \in \{\mathbb{R}, \mathbb{C}\}$ and V be an \mathbb{F} -vector space with vector addition + and scalar multiplication \cdot .

As we did for \mathbb{R}^n and later for \mathbb{C}^n , we introduce notions like *linear independence*, *basis*, *dimension* and related definitions. In spite of considering abstract vector spaces, the notions still work exactly the same.

Definition 7.11. Same as before: Basis, dimension, and so on Let V be an \mathbb{F} -vector space with operations + and \cdot . • For $k \in \mathbb{N}$, vectors $\mathbf{v}_1, \ldots, \mathbf{v}_k \in V$ and scalars $\alpha_1, \ldots, \alpha_k \in \mathbb{F}$ the vector $\alpha_1 \mathbf{v}_1 + \ldots + \alpha_k \mathbf{v}_k = \sum_{i=1}^k \alpha_i \mathbf{v}_i \in V$ is called a linear combination of the vectors $\mathbf{v}_1, \ldots, \mathbf{v}_k$. • The set of all possible linear combinations for the vectors of a subset $M \subset V$ is called the linear hull or span of M: $\operatorname{Span}(M) := \{\lambda_1 \mathbf{u}_1 + \dots + \lambda_k \mathbf{u}_k : \mathbf{u}_1, \dots, \mathbf{u}_k \in M, \lambda_1, \dots, \lambda_k \in \mathbb{F}, k \in \mathbb{N}\}.$ • A family $\mathcal{E} = (\mathbf{v}_1, \dots, \mathbf{v}_k)$ consisting of k vectors from V is called a generating system for the subspace $U \subset V$, if $U = \text{Span}(\mathbf{v}_1, \ldots, \mathbf{v}_k)$. • A family $\mathcal{E} = (\mathbf{v}_1, \dots, \mathbf{v}_k)$ consisting of k vectors from V is called linearly dependent if o can be represented by a non-trivial linear combination of vectors from \mathcal{E} . If there is no such non-trivial linear combination, the family is called linearly independent. • A family \mathcal{E} that is a generating system for $U \subset V$ and linearly independent is called a basis of U. • The number of elements for a basis of U is called the dimension of U. We just write $\dim(U)$. Rule of thumb: Basis, dimension and similar things • A generating family $\mathcal{E} = (\mathbf{v}_1, \dots, \mathbf{v}_k)$ of U is called this way because we can reach each point in U with linear combinations of vector from $\mathcal E$ and no other points. • A family \mathcal{E} is linear independent if we need all "family members" to span (or generate) the subspace $\text{Span}(\mathcal{E})$. • A basis \mathcal{B} of U is a smallest generating set U. (We cannot omit a vector from B.)• The dimension of a subspace U = the number of elements of a basis of U. (All bases have the same number of elements, just redo the proof in Proposition 3.25.) = the smallest possible size for a generating system of U. (With less vectors) it is not possible to span the whole space U.) = the maximal number of vectors from U that form a linearly independent family. (If you choose more vectors, there are always linearly dependent.)

Example 7.12. – Matrix vector spaces

(a) The vector space $\mathbb{C}^{2\times 3}$ of all complex 2 × 3-matrices can be written in the following way:

$$\mathbb{C}^{2\times3} = \left\{ \begin{pmatrix} \alpha & \beta & \gamma \\ \delta & \varepsilon & \varphi \end{pmatrix} : \alpha, \beta, \gamma, \delta, \varepsilon, \varphi \in \mathbb{C} \right\}$$

= $\left\{ \alpha \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} + \beta \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} + \gamma \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} + \delta \begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix} + \varepsilon \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} + \varphi \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} : \alpha, \beta, \gamma, \delta, \varepsilon, \varphi \in \mathbb{C} \right\}$
= $\operatorname{Span} \left(\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \right\}$

Hence

$$\mathcal{B} = \left(\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \right)$$

is a generating system for $\mathbb{C}^{2\times 3}$. \mathcal{B} is also linearly independent: From

$$\underbrace{\alpha \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_{\begin{pmatrix} \alpha & \beta & \gamma \\ \delta & \varepsilon & \varphi \end{pmatrix}} \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} + \underbrace{\gamma \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}}_{\begin{pmatrix} \alpha & \beta & \gamma \\ \delta & \varepsilon & \varphi \end{pmatrix}} + \underbrace{\delta \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}}_{\begin{pmatrix} \alpha & \beta & \gamma \\ \delta & \varepsilon & \varphi \end{pmatrix}} = \mathbf{o}, \boldsymbol{\varepsilon} \begin{pmatrix} \boldsymbol{0} & 0 & 0 \\ \boldsymbol{0} & 0 & 1 \end{pmatrix}$$

we conclude $\alpha = \beta = \gamma = \delta = \varepsilon = \varphi = 0$. Hence, \mathcal{B} is a basis of $\mathbb{C}^{2\times 3}$ and the dimension of $\mathbb{C}^{2\times 3}$ is $|\mathcal{B}| = 6$. Analogously, one can prove: dim $(\mathbb{F}^{m\times n}) = m \cdot n$.

(b) In a similar way, we can prove that

$$\mathcal{B} = \left(\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right)$$
(7.1)

forms a basis of $\mathbb{R}^{2 \times 2}$. Hence:

$$\left\{ \begin{pmatrix} \alpha & \beta \\ 0 & \gamma \end{pmatrix} : \alpha, \beta, \gamma \in \mathbb{R} \right\} = \left\{ \alpha \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \beta \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + \gamma \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} : \alpha, \beta, \gamma \in \mathbb{R} \right\}$$

We conclude: dim $(\mathbb{R}^{2\times 2}) = 2 + 1 = 3$. Analogously for given $n \in \mathbb{N}$, one can prove dim $(\mathbb{R}^{n\times n}) = n + (n-1) + \ldots + 1 = \frac{n(n+1)}{2}$.

(c) As a special vector space, we look at:

$$U = \left\{ \begin{pmatrix} \alpha & 0 & \alpha \\ 0 & \beta & 0 \\ \alpha & 0 & \alpha \end{pmatrix} : \alpha, \beta \in \mathbb{C} \right\} = \left\{ \alpha \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{pmatrix} + \beta \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} : \alpha, \beta \in \mathbb{C} \right\}$$

$$= \operatorname{Span}\left(\underbrace{\begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{pmatrix}}_{=:A}, \underbrace{\begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_{=:B}\right).$$
(7.2)

Hence, $\mathcal{B} := (A, B)$ is a generating system for U. Again, we show that \mathcal{B} is also linearly independent. From $\alpha A + \beta B = \mathbf{0}$, one gets

$$\begin{pmatrix} \alpha & 0 & \alpha \\ 0 & \beta & 0 \\ \alpha & 0 & \alpha \end{pmatrix} = \alpha A + \beta B = \mathbf{o} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix},$$

and concludes $\alpha = \beta = 0$. Therefore, \mathcal{B} is a basis of U and dim(U) = 2.

Example 7.13. – **Polynomial space** $\mathcal{P}_2(\mathbb{R})$. We define the special polynomials $\mathbf{m}_0, \mathbf{m}_1, \mathbf{m}_2 \in \mathcal{P}_2$ by

$$\mathbf{m}_0(x) := 1$$
, $\mathbf{m}_1(x) := x$, and $\mathbf{m}_2(x) := x^2$ for all $x \in \mathbb{R}$

and see:

$$\mathcal{P}_{2}(\mathbb{R}) = \{ x \mapsto a_{2}x^{2} + a_{1}x + a_{0} : a_{2}, a_{1}, a_{0} \in \mathbb{R} \} = \{ a_{2}\mathbf{m}_{2} + a_{1}\mathbf{m}_{1} + a_{0}\mathbf{m}_{0} : a_{2}, a_{1}, a_{0} \in \mathbb{R} \}$$

= Span($\mathbf{m}_{0}, \mathbf{m}_{1}, \mathbf{m}_{2}$)

Hence, $\mathcal{B} := (\mathbf{m}_0, \mathbf{m}_1, \mathbf{m}_2)$ is a generating system for $\mathcal{P}_2(\mathbb{R})$.

Remains to show:
$$(m_0, m_1, m_1)$$
 are fine eiding.

$$K h_0 + \beta h_0 + \gamma m_1 = 0 \qquad (tare vector 0: R \rightarrow R) \\ K \rightarrow 0 \qquad (tare vector 0: R \rightarrow R) \\ K \rightarrow 0 \qquad (tare vector 0: R \rightarrow R) \\ K \rightarrow 0 \qquad (tare 0: R \rightarrow R) \\ K \rightarrow 0 \qquad (tare R) \\ K \rightarrow$$

Proposition & Definition 7.14. Monomial basis of $\mathcal{P}_n(\mathbb{R})$ Let $n \in \mathbb{N}_0$. The particular polynomials $\mathbf{m}_0, \mathbf{m}_1, \dots, \mathbf{m}_n \in \mathcal{P}_n(\mathbb{R})$ defined by $\mathbf{m}_0(x) = 1$, $\mathbf{m}_1(x) = x$, \dots , $\mathbf{m}_{n-1}(x) = x^{n-1}$, $\mathbf{m}_n(x) = x^n$ for all $x \in \mathbb{R}$ are called monomials. The family $\mathcal{B} = (\mathbf{m}_0, \mathbf{m}_1, \dots, \mathbf{m}_n)$ forms a basis of $\mathcal{P}_n(\mathbb{R})$ and is called the monomial basis. Hence $\dim(\mathcal{P}_n(\mathbb{R})) = n + 1$.

Corollary 7.15. The method of equating the coefficients

Let **p** and **q** be two real polynomials with degree $n \in \mathbb{N}$, which means

$$\mathbf{p}(x) = a_n x^n + \ldots + a_1 x + a_0$$
 and $\mathbf{q}(x) = b_n x^n + \ldots + b_1 x + b_0$

for some coefficients $a_n, \ldots, a_1, a_0, b_n, \ldots, b_1, b_0 \in \mathbb{R}$. If we have the equality $\mathbf{p} = \mathbf{q}$, which means

$$a_n x^n + \ldots + a_1 x + a_0 = b_n x^n + \ldots + b_1 x + b_0, \tag{7.3}$$

for all $x \in \mathbb{R}$, then we can conclude $a_n = b_n, \ldots, a_1 = b_1$ and $a_0 = b_0$.

Remark:

Since dim $(\mathcal{P}_n(\mathbb{R})) = n + 1$ and we have the inclusions

$$\mathcal{P}_0(\mathbb{R}) \subset \mathcal{P}_1(\mathbb{R}) \subset \mathcal{P}_2(\mathbb{R}) \subset \cdots \subset \mathcal{P}(\mathbb{R}) \subset \mathcal{F}(\mathbb{R}),$$

we conclude that $\dim(\mathcal{P}(\mathbb{R}))$ and $\dim(\mathcal{F}(\mathbb{R}))$ cannot be finite natural numbers. Symbolically, we write $\dim(\mathcal{P}(\mathbb{R})) = \infty$ in such a case.