Course 5

Linear independence, bases

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Chapter 2. Vector Spaces

1. Basic properties

2. Subspaces

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Following [Klein], we present a way of achieving lossy compression of images.

Definition

Let V be a vector space over K. We say that the vectors $v_1, \ldots, v_n \in V$ are (or the set of vectors $\{v_1, \ldots, v_n\}$ is):

(1) *linearly independent* in V if for every $k_1, \ldots, k_n \in K$,

$$k_1 v_1 + \cdots + k_n v_n = 0 \implies k_1 = \cdots = k_n = 0.$$

(2) *linearly dependent* in V if they are not linearly independent, that is, $\exists k_1, \ldots, k_n \in K$ not all zero such that

$$k_1 v_1 + \cdots + k_n v_n = 0.$$

(1) A set consisting of a single vector v is linearly dependent $\iff$ $v = 0$.

(2) As an immediate consequence of the definition, we notice that if V is a vector space over K and $X, Y \subseteq V$ such that $X \subseteq Y$, then:

(i) If Y is linearly independent, then X is linearly independent.

(ii) If X is linearly dependent, then Y is linearly dependent. Thus, every set of vectors containing the zero vector is linearly dependent.

(3) More generally, an infinite set of vectors of V is called *linearly independent* if any finite subset is linearly independent, and *linearly dependent* if there exists a finite subset which is linearly dependent.

Linear dependence

Theorem

Let V be a vector space over K. Then the vectors $v_1, \ldots, v_n \in V$ are linearly dependent if and only if one of the vectors is a linear combination of the others, that is, $\exists j \in \{1, \ldots, n\}$ such that

$$v_j = \sum_{\substack{i=1 \\ i \neq j}}^{n} \alpha_i v_i$$

for some $\alpha_i \in K$, where $i \in \{1, \ldots, n\}$ and $i \neq j$.

Proof. [...]

Examples I

(a) Let V_2 be the real vector space of all vectors (in the classical sense) in the plane with a fixed origin O. Recall that the addition is the usual addition of two vectors by the parallelogram rule and the external operation is the usual scalar multiplication of vectors by real scalars. Then:

(i) one vector v is linearly dependent in $V_2 \iff v = 0$;

(ii) two vectors are linearly dependent in $V_2 \iff$ they are collinear;

(iii) three vectors (or more) are always linearly dependent in V_2.

Now let V_3 be the real vector space of all vectors (in the classical sense) in the space with a fixed origin O. Then:

(i) one vector v is linearly dependent in $V_3 \iff v = 0$;

(ii) two vectors are linearly dependent in $V_3 \iff$ they are collinear;

(iii) three vectors are linearly dependent in $V_3 \iff$ they are coplanar;

(iv) four vectors (or more) are always linearly dependent in V_3.

Examples II

(b) If K is a field and $n \in \mathbb{N}^*$, then the vectors $e_1 = (1, 0, 0, \ldots, 0)$, $e_2 = (0, 1, 0, \ldots, 0)$, $\ldots$, $e_n = (0, 0, 0, \ldots, 1) \in K^n$ are linearly independent in the canonical vector space K^n over K [...].

(c) Let K be a field and $n \in \mathbb{N}$. Then the vectors

$$1, X, X^2, \ldots, X^n$$

are linearly independent in the vector space $K_n[X] = \{f \in K[X] \mid \text{degree}(f) \leq n\}$ over K. More generally, the vectors

$$1, X, X^2, \ldots, X^n, \ldots \quad (n \in \mathbb{N})$$

form an infinite linearly independent set in the vector space $K[X]$ over K.

Theorem

Let $n \in \mathbb{N}$, $n \geq 2$.
(i) Two vectors in the canonical vector space K^n are linearly dependent $\iff$ their components are respectively proportional.
(ii) n vectors in the canonical vector space K^n are linearly dependent $\iff$ the determinant consisting of their components is zero.

Proof. [...]

Basis

For the sake of simplicity and because of our limited needs, til the end of the chapter, *by a vector space we will understand a finitely generated vector space*.

Definition

Let V be a vector space over K. A list of vectors $B = (v_1, \ldots, v_n) \in V^n$ is called a *basis* of V if:
(i) B is linearly independent in V;
(ii) B is a system of generators for V, that is, $\langle B \rangle = V$.

Existence of basis I

Theorem

Every vector space V over K has a basis.

Proof. If $V = \{0\}$, then it has the basis $\emptyset$.

Now let $V = \langle B \rangle \neq \{0\}$, where $B = (v_1, \ldots, v_n)$.

• If B is linearly independent, then B is a basis and we are done. Suppose that the list B is linearly dependent. Then $\exists j_1 \in \{1, \ldots, n\}$ such that $v_{j_1} = \sum_{\substack{i=1 \\ i \neq j_1}}^{n} k_i v_i$ for some $k_i \in K$. It follows that $V = \langle B \setminus \{v_{j_1}\} \rangle$, because every vector of V can be written as a linear combination of the vectors of $B \setminus \{v_{j_1}\}$.

• If $B \setminus \{v_{j_1}\}$ is linearly independent, it is a basis and we are done. Otherwise, $\exists j_2 \in \{1, \ldots, n\} \setminus \{j_1\}$ such that $v_{j_2} = \sum_{\substack{i=1 \\ i \neq j_1, j_2}}^{n} k_i' v_i$ for some $k_i' \in K$. Then $V = \langle B \setminus \{v_{j_1}, v_{j_2}\} \rangle$, because every vector of V can be written as a linear combination of the vectors of $B \setminus \{v_{j_1}, v_{j_2}\}$.

• If $B \setminus \{v_{j_1}, v_{j_2}\}$ is linearly independent, then it is a basis and we are done. Otherwise, we continue the procedure. If all the previous intermediate subsets are linearly dependent, we get to

$$V = \langle B \setminus \{v_{j_1}, \ldots, v_{j_{n-1}}\}\rangle = \langle v_{j_n}\rangle.$$

• If v_{j_n} were linearly dependent, then $v_{j_n} = 0$, hence we have $V = \langle v_{j_n}\rangle = \{0\}$, contradiction. Hence v_{j_n} is linearly independent and thus forms a single element basis of V. $\qquad \square$

We shall see that a vector space may have more than one basis.

Theorem

Let V be a vector space over K. A list $B = (v_1, \ldots, v_n)$ of vectors in V is a basis of V if and only if every vector $v \in V$ can be uniquely written as a linear combination of the vectors $v_1, \ldots, v_n$, that is,

$$v = k_1 v_1 + \cdots + k_n v_n$$

for some unique $k_1, \ldots, k_n \in K$.

Proof. [...]

Coordinates of a vector in a basis

Definition

Let V be a vector space over K, $B = (v_1, \ldots, v_n)$ a basis of V and $v \in V$. Then the scalars $k_1, \ldots, k_n \in K$ appearing in the unique writing of v as a linear combination

$$v = k_1 v_1 + \cdots + k_n v_n$$

of the vectors of B are called the *coordinates of v in the basis B*.

Examples I

(a) If K is a field and $n \in \mathbb{N}^*$, then the list $E = (e_1, \ldots, e_n)$ of vectors of K^n, where

$$\begin{cases} e_1 = (1, 0, 0, \ldots, 0) \\ e_2 = (0, 1, 0, \ldots, 0) \\ \ldots\ldots\ldots \\ e_n = (0, 0, 0, \ldots, 1) \end{cases}$$

is a basis of the canonical vector space K^n over K, called the *canonical basis* (or *standard basis*). Indeed, each vector $v = (x_1, \ldots, x_n) \in K^n$ has a unique writing $v = x_1 e_1 + \cdots + x_n e_n$ as a linear combination of the vectors of E, hence E is a basis of V. Notice that the coordinates of a vector in the canonical basis are just the components of that vector, fact that is not true in general.

In particular, the canonical vector space $\mathbb{Z}_2^n$ over $\mathbb{Z}_2$ has the above canonical basis $E = (e_1, \ldots, e_n)$, where 0 and 1 are just the elements $\widehat{0}$ and $\widehat{1}$ of $\mathbb{Z}_2$.

Also, if $n = 1$, the set $\{1\}$ is a basis of the canonical vector space K over K. For instance, $\{1\}$ is a basis of the vector space $\mathbb{C}$ over $\mathbb{C}$.

(b) Consider the canonical real vector space $\mathbb{R}^2$. We already know a basis of $\mathbb{R}^2$, namely the canonical basis $((1, 0), (0, 1))$. But it is easy to show that the list $((1, 1), (0, 1))$ is also a basis of $\mathbb{R}^2$. Therefore, a vector space may have more than one basis.

Also, note that $\{e_1\}$ is linearly independent, but not a system of generators, while the list $(e_1, e_2, e_1 + e_2)$ is a system of generators, but not linearly independent. Hence none of the two lists is a basis of the canonical real vector space $\mathbb{R}^2$.

(c) Let V_3 be the real vector space of all vectors (in the classical sense) in the space with a fixed origin O. Then a basis of V_3 consists of the three pairwise orthogonal *unit vectors* $\overrightarrow{i}$, $\overrightarrow{j}$, $\overrightarrow{k}$.

(d) Let K be a field and $n \in \mathbb{N}$. Then the list

$$E = (1, X, X^2, \ldots, X^n)$$

is a basis of the vector space $K_n[X] = \{f \in K[X] \mid \mathrm{degree}(f) \leq n\}$ over K, because every vector (polynomial) $f \in K_n[X]$ can be uniquely written as a linear combination $a_0 \cdot 1 + a_1 \cdot X + \cdots + a_n \cdot X^n$ $(a_0, \ldots, a_n \in K)$ of the vectors of E. In this case, the coordinates of a vector $f \in K_n[X]$ in the basis B are just its coefficients as a polynomial.

(e) Consider the real vector space
$\mathbb{R}_2[X] = \{f \in \mathbb{R}[X] \mid \mathrm{degree}(f) \leq 2\}$. We have seen that the list
$E = (1, X, X^2)$ is a basis of $\mathbb{R}_2[X]$. Let us show that the list

$$B = (1, X - 1, (X - 1)^2)$$

is also a basis of $\mathbb{R}_2[X]$. Let $g = a_0 + a_1 X + a_2 X^2 \in \mathbb{R}_2[X]$ [...].
It turns out that the coordinates of a vector
$g = a_0 + a_1 X + a_2 X^2 \in \mathbb{R}_2[X]$ in the basis B are $a_0 + a_1 + a_2$,
$a_1 + 2a_2$, a_2.

(f) Let K be a field. The list

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

is a basis of the vector space $M_2(K)$ over K.

More generally, let $m, n \in \mathbb{N}$, $m, n \geq 2$ and consider the matrices $E_{ij} = (a_{kl})$, where

$$a_{kl} = \begin{cases} 1 & \text{if } k = i \text{ and } l = j \\ 0 & \text{otherwise} \end{cases}.$$

Then the list consisting of all matrices E_{ij} is a basis of the vector space $M_{m,n}(K)$ over K.

In this case, the coordinates of a vector $A \in M_{m,n}(K)$ in the above basis are just the entries of that matrix.

(g) Consider the real vector space $M_2(\mathbb{R})$. We have seen that

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

is a basis of $M_2(\mathbb{R})$. Let us show that the list

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

is also a basis of $M_2(\mathbb{R})$ [...].

It turns out that the coordinates of a vector $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \in M_2(\mathbb{R})$

in the basis B are $a - d + b - c$, $d - b$, c, b.

(h) Since $\forall z \in \mathbb{C}$, $\exists! x, y \in \mathbb{R}$ such that $z = x \cdot 1 + y \cdot i$, the list $B = (1, i)$ is a basis of the vector space $\mathbb{C}$ over $\mathbb{R}$.
The coordinates of a vector $z \in \mathbb{C}$ in the basis B are just its real and its imaginary part.

Bases and linear maps

Theorem

Let $f : V \to V'$ be a K-linear map and let $B = (v_1, \ldots, v_n)$ be a basis of V. Then f is determined by its values on the vectors of the basis B.

Proof. [...]

Corollary

Let $f, g : V \to V'$ be K-linear maps and let $B = (v_1, \ldots, v_n)$ be a basis of V. If $f(v_i) = g(v_i)$, $\forall i \in \{1, \ldots, n\}$, then $f = g$.

Proof. [...]

Bases and linear maps II

Theorem

Let $f : V \to V'$ be a K-linear map, and let $X = (v_1, \ldots, v_n)$ be a list of vectors in V.

(i) If f is injective and X is linearly independent in V, then $f(X)$ is linearly independent in V'.

(ii) If f is surjective and X is a system of generators for V, then $f(X)$ is a system of generators for V'.

(iii) If f is bijective and X is a basis of V, then $f(X)$ is a basis of V'.

Proof. [...]

Definition

Let $k, n \in \mathbb{N}^*$ be such that $k < n$, and let u be a vector of the canonical vector space K^n over K. Then the *closest k-sparse* vector associated to u is defined as the vector obtained from u by replacing all but its k largest magnitude components by zero.

Example

Consider an image consisting of a single row of four pixels with intensities 200, 50, 200 and 75 respectively. We know that such an image can be viewed as a vector $u = (200, 50, 200, 75)$ in the real canonical vector space $\mathbb{R}^4$. The closest 2-sparse vector associated to u is the vector $\tilde{u} = (200, 0, 200, 0)$.

Suppose that we need to store a grayscale image of (say) $n = 2000 \times 1000$ pixels more compactly. We can view it as a vector v in the real canonical vector space $\mathbb{R}^n$. If we just store its

associated closest k-sparse vector, then the compressed image may be far from the original.

One may use the following *lossy compression algorithm*:

- **Step 1.** *Consider a suitable basis $B = (v_1, \ldots, v_n)$ of the real canonical vector space $\mathbb{R}^n$.*
- **Step 2.** *Determine the n-tuple u (which is desired to have as many zeros as possible) of the coordinates of v in the basis B.*
- **Step 3.** *Replace u by the closest k-sparse n-tuple $\tilde{u}$ for a suitable k, and store $\tilde{u}$.*
- **Step 4.** *In order to recover an image from $\tilde{u}$, compute the corresponding linear combination of the vectors of B with scalars the components of $\tilde{u}$.*

Consider the following image:

First, use the closest sparse vector which supresses all but 10% of the components of v, and secondly, use the lossy compression algorithm which supresses all but 10% of the components of u in order to get the following images respectively: