

# Linear Algebra 1

## Lecture #11

### Linear functions (part II)

Irena Penev

December 20-21, 2023

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  - ① The effect of a linear function on linearly independent and spanning sets

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  - ② Computing bases of the images and preimages of subspaces under linear functions
  - ③ Linear functions and bases
  - ④ Isomorphisms

- 1 The effect of a linear function on linearly independent and spanning sets

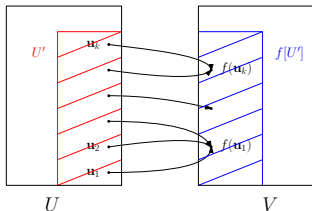
- 1 The effect of a linear function on linearly independent and spanning sets

### Theorem 4.2.11

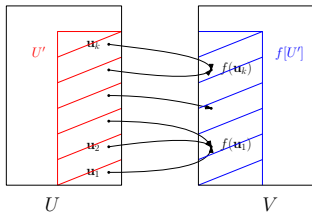
Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Let  $\mathbf{u}_1, \dots, \mathbf{u}_k \in U$ , and set

$U' := \text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ . Then all the following hold:

- $U'$  is a subspace of  $U$ , and  $f[U']$  is a subspace of  $V$ ;
- $f[U'] = f[\text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)] = \text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k))$ , i.e. vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  span  $f[U'] = f[\text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)]$ ;
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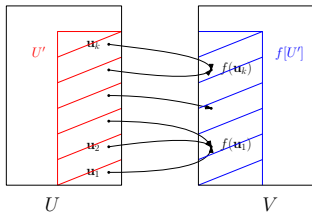


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*Proof of (a).*



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- Ⓐ  $U'$  is a subspace of  $U$ , and  $f[U']$  is a subspace of  $V$ ;

*Proof of (a).* The fact that  $U'$  is a subspace of  $U$  follows immediately from Theorem 3.1.11, and the fact that  $f[U']$  is a subspace of  $V$  follows from 4.2.3(a). This proves (a).

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*Proof of (b).*

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*Proof of (b).*

$$\begin{aligned} \text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)) &= \{ \alpha_1 f(\mathbf{u}_1) + \dots + \alpha_k f(\mathbf{u}_k) \mid \alpha_1, \dots, \alpha_k \in \mathbb{F} \} \\ &\stackrel{(*)}{=} \{ f(\alpha_1 \mathbf{u}_1 + \dots + \alpha_k \mathbf{u}_k) \mid \alpha_1, \dots, \alpha_k \in \mathbb{F} \} \\ &\stackrel{(**)}{=} \{ f(\mathbf{u}) \mid \mathbf{u} \in \text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k) \} \\ &= f[\text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)] = f[U'], \end{aligned}$$

where (\*) follows from the linearity of the  $f$  (and more precisely, from Prop. 4.1.5), and (\*\*) follows from the definition of span.

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- Ⓒ  $\dim(f[U']) \leq \dim(U') \leq k$ .

*Proof of (c).*

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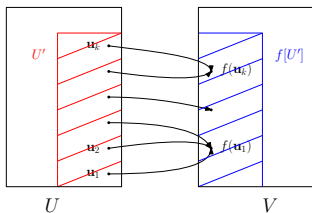
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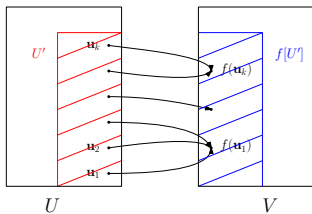
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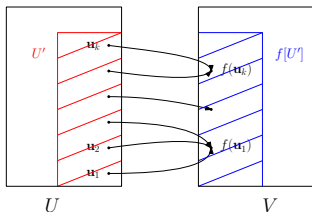




### Corollary 4.2.12

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , let  $f : U \rightarrow V$  be a linear function, and let  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  be a spanning set of  $U$ . Then  $\text{Im}(f) = \text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k))$  and  $\text{rank}(f) = \dim(\text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k))) \leq k$ .

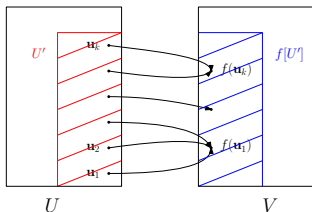
*Proof.*



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*Proof.* By hypothesis,  $U = \text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ .

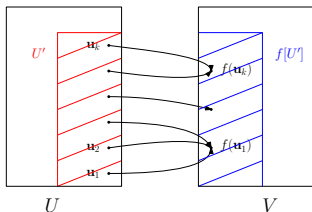


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*Proof.* By hypothesis,  $U = \text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ . So, by Theorem 4.2.11(b), we have that  $\text{Im}(f) = f[U] = \text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k))$ ,





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Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , let  $f : U \rightarrow V$  be a linear function, and let  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  be a spanning set of  $U$ . Then  $\text{Im}(f) = \text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k))$  and  $\text{rank}(f) = \dim(\text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k))) \leq k$ .

*Proof.* By hypothesis,  $U = \text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ . So, by Theorem 4.2.11(b), we have that  $\text{Im}(f) = f[U] = \text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k))$ , and by Theorem 4.2.11(c), we have that  $\text{rank}(f) = \dim(\text{Im}(f)) = \dim(f[U]) \leq k$ .  $\square$

### Theorem 4.2.13

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , let  $f : U \rightarrow V$  be a linear function, and let  $\mathbf{u}_1, \dots, \mathbf{u}_k \in U$ . Then all the following hold:

- (a) if  $f$  is one-to-one and vectors  $\mathbf{u}_1, \dots, \mathbf{u}_k$  are linearly independent in  $U$ , then vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  are linearly independent in  $V$ ;
- (b) if vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  are linearly independent in  $V$ , then vectors  $\mathbf{u}_1, \dots, \mathbf{u}_k$  are linearly independent in  $U$ ;
- (c) if  $f$  is onto and vectors  $\mathbf{u}_1, \dots, \mathbf{u}_k$  span  $U$ , then vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  span  $V$ ;
- (d) if  $f$  is one-to-one and vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  span  $V$ , then vectors  $\mathbf{u}_1, \dots, \mathbf{u}_k$  span  $U$ .

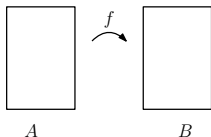
- Proof: Lecture Notes.
- Informal summary: next slide.

- Theorem 4.2.13 (schematically and informally):

$$f : U \xrightarrow{\text{linear}} V$$

(a)-(b)	$\mathbf{u}_1, \dots, \mathbf{u}_k$ are linearly independent	$\xRightarrow{\text{if } f \text{ is 1-1}}$ $\xleftarrow{\text{always}}$	$f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$ are linearly independent
(c)-(d)	$\mathbf{u}_1, \dots, \mathbf{u}_k$ span $U$	$\xRightarrow{\text{if } f \text{ is onto}}$ $\xleftarrow{\text{if } f \text{ is 1-1}}$	$f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$ span $V$

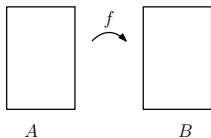
- **Dimension considerations:**



- As we know, for any function  $f : A \rightarrow B$ , where  $A$  and  $B$  are finite sets, the following hold:
  - if  $f$  is one-to-one, then  $|A| \leq |B|$ ;
  - if  $f$  is onto, then  $|A| \geq |B|$ ;
  - if  $f$  is a bijection, then  $|A| = |B|$ .

(Actually, the above is true even if we allow  $A$  and  $B$  to be infinite, but to make sense of the statement, we would need infinite cardinals. We omit the details.)

- **Dimension considerations:**



- As we know, for any function  $f : A \rightarrow B$ , where  $A$  and  $B$  are finite sets, the following hold:
  - if  $f$  is one-to-one, then  $|A| \leq |B|$ ;
  - if  $f$  is onto, then  $|A| \geq |B|$ ;
  - if  $f$  is a bijection, then  $|A| = |B|$ .

(Actually, the above is true even if we allow  $A$  and  $B$  to be infinite, but to make sense of the statement, we would need infinite cardinals. We omit the details.)

- In the case of **linear** functions, Theorem 4.2.14 (next slide) gives us a very similar statement, only involving dimension (rather than cardinality) of the domain and codomain.

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- Ⓐ if  $f$  is one-to-one, then  $\dim(U) \leq \dim(V)$ ;
- Ⓑ if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;
- Ⓒ if  $f$  is an isomorphism, then  $\dim(U) = \dim(V)$ .

*Proof.*

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- Ⓐ if  $f$  is one-to-one, then  $\dim(U) \leq \dim(V)$ ;
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- Ⓒ if  $f$  is an isomorphism, then  $\dim(U) = \dim(V)$ .

*Proof.* Obviously, (a) and (b) together imply (c).

### Theorem 4.2.14

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- Ⓒ if  $f$  is an isomorphism, then  $\dim(U) = \dim(V)$ .

*Proof.* Obviously, (a) and (b) together imply (c). So, it is enough to prove (a) and (b).



### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- (a) if  $f$  is one-to-one, then  $\dim(U) \leq \dim(V)$ ;

*Proof (continued).* (a) We prove the contrapositive: we assume that  $\dim(U) > \dim(V)$  (and in particular,  $\dim(V)$  is finite), and we prove that  $f$  is **not** one-to-one.

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

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Set  $n := \dim(V)$ .

### Theorem 4.2.14

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*Proof (continued).* (a) We prove the contrapositive: we assume that  $\dim(U) > \dim(V)$  (and in particular,  $\dim(V)$  is finite), and we prove that  $f$  is **not** one-to-one.

Set  $n := \dim(V)$ . Since  $\dim(U) > \dim(V)$ , we know that  $U$  has a linearly independent set of size greater than  $n$ .

- Indeed, if  $U$  is finite-dimensional, then any one of its bases is a linearly independent set of size  $\dim(U) > n$ , and if  $U$  is infinite-dimensional, then Proposition 3.2.18 guarantees that  $U$  has linearly independent sets of any finite size.

So, fix a linearly independent set  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  of  $U$ , with  $k > n$ .

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- Ⓐ if  $f$  is one-to-one, then  $\dim(U) \leq \dim(V)$ ;

*Proof (continued).* (a) We prove the contrapositive: we assume that  $\dim(U) > \dim(V)$  (and in particular,  $\dim(V)$  is finite), and we prove that  $f$  is **not** one-to-one.

Set  $n := \dim(V)$ . Since  $\dim(U) > \dim(V)$ , we know that  $U$  has a linearly independent set of size greater than  $n$ .

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So, fix a linearly independent set  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  of  $U$ , with  $k > n$ . Since  $\dim(V) = n$ , Theorem 3.2.17(a) guarantees that the set  $\{f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)\}$  is linearly dependent. But now Theorem 4.2.13(a) guarantees that  $f$  is not one-to-one.

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- (b) if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;

*Proof (continued).* (b) Assume that  $f$  is onto; we must show that  $\dim(U) \geq \dim(V)$ .

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

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*Proof (continued).* (b) Assume that  $f$  is onto; we must show that  $\dim(U) \geq \dim(V)$ . We may assume that  $n := \dim(U)$  is finite, for otherwise, we are done.

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- (b) if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;

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### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- ⓑ if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;

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Fix any basis  $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  of  $U$ .



### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- (b) if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;

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Fix any basis  $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  of  $U$ . In particular, vectors  $\mathbf{u}_1, \dots, \mathbf{u}_n$  span  $U$ , and so since  $f$  is onto, Theorem 4.2.13(c) guarantees that vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_n)$  span  $V$ .

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- (b) if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;

*Proof (continued).* (b) Assume that  $f$  is onto; we must show that  $\dim(U) \geq \dim(V)$ . We may assume that  $n := \dim(U)$  is finite, for otherwise, we are done. We must show that  $\dim(V) \leq n$ .

Fix any basis  $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  of  $U$ . In particular, vectors  $\mathbf{u}_1, \dots, \mathbf{u}_n$  span  $U$ , and so since  $f$  is onto, Theorem 4.2.13(c) guarantees that vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_n)$  span  $V$ . But then by Theorem 3.2.14, some subset of  $\{f(\mathbf{u}_1), \dots, f(\mathbf{u}_n)\}$  is a basis of  $V$ , and it follows that  $\dim(V) \leq n$ .  $\square$

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- Ⓐ if  $f$  is one-to-one, then  $\dim(U) \leq \dim(V)$ ;
- Ⓑ if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;
- Ⓒ if  $f$  is an isomorphism, then  $\dim(U) = \dim(V)$ .

- 2 Computing bases of the images and preimages of subspaces under linear functions

### Proposition 4.2.15

Let  $\mathbb{F}$  be a field, let  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  be a linear function, let  $A \in \mathbb{F}^{n \times m}$  be the standard matrix of  $f$ , let  $\mathbf{u}_1, \dots, \mathbf{u}_k \in \mathbb{F}^m$  ( $k \geq 1$ ), and set  $U := \text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ . Then

$$f[U] = \text{Col}\left(A \begin{bmatrix} \mathbf{u}_1 & \dots & \mathbf{u}_k \end{bmatrix}\right),$$

and moreover, the pivot columns of the matrix  $A \begin{bmatrix} \mathbf{u}_1 & \dots & \mathbf{u}_k \end{bmatrix}$  form a basis of  $f[U]$ .

- First an example, then a proof.

### Example 4.2.16

Let  $f : \mathbb{Z}_2^5 \rightarrow \mathbb{Z}_2^4$  be the linear function whose standard matrix is

$$A = \begin{bmatrix} 1 & 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 1 \end{bmatrix},$$

and consider the vectors

$$\mathbf{u}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \quad \mathbf{u}_2 = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \quad \mathbf{u}_3 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \quad \mathbf{u}_4 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

in  $\mathbb{Z}_2^5$ . Set  $U := \text{Span}(\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3, \mathbf{u}_4)$ . Find a basis for  $f[U]$ .

*Solution.*

*Solution.* Our goal is to find the pivot columns of the matrix  $A \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 & \mathbf{u}_4 \end{bmatrix}$ , since by Proposition 4.2.15, those columns form a basis of  $f[U]$ . First, by multiplying matrices, we obtain

$$\begin{aligned} A \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 & \mathbf{u}_4 \end{bmatrix} &= \begin{bmatrix} 1 & 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}. \end{aligned}$$

*Solution (continued).* By row reducing, we obtain

$$\text{RREF}\left(A \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 & \mathbf{u}_4 \end{bmatrix}\right) = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$



*Solution (continued).* By row reducing, we obtain

$$\text{RREF}\left(A \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 & \mathbf{u}_4 \end{bmatrix}\right) = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

As we can see, the pivot columns of  $A \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 & \mathbf{u}_4 \end{bmatrix}$  are its first and fourth column.

*Solution (continued).* By row reducing, we obtain

$$\text{RREF}\left(A \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 & \mathbf{u}_4 \end{bmatrix}\right) = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

As we can see, the pivot columns of  $A \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 & \mathbf{u}_4 \end{bmatrix}$  are its first and fourth column. Therefore,

$$\left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix} \right\}$$

is a basis of  $f[U]$ .  $\square$

### Proposition 4.2.15

Let  $\mathbb{F}$  be a field, let  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  be a linear function, let  $A \in \mathbb{F}^{n \times m}$  be the standard matrix of  $f$ , let  $\mathbf{u}_1, \dots, \mathbf{u}_k \in \mathbb{F}^m$  ( $k \geq 1$ ), and set  $U := \text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ . Then

$$f[U] = \text{Col}\left(A \begin{bmatrix} \mathbf{u}_1 & \dots & \mathbf{u}_k \end{bmatrix}\right),$$

and moreover, the pivot columns of the matrix  $A \begin{bmatrix} \mathbf{u}_1 & \dots & \mathbf{u}_k \end{bmatrix}$  form a basis of  $f[U]$ .

*Proof.*

### Proposition 4.2.15

Let  $\mathbb{F}$  be a field, let  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  be a linear function, let  $A \in \mathbb{F}^{n \times m}$  be the standard matrix of  $f$ , let  $\mathbf{u}_1, \dots, \mathbf{u}_k \in \mathbb{F}^m$  ( $k \geq 1$ ), and set  $U := \text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ . Then

$$f[U] = \text{Col}\left(A \begin{bmatrix} \mathbf{u}_1 & \dots & \mathbf{u}_k \end{bmatrix}\right),$$

and moreover, the pivot columns of the matrix  $A \begin{bmatrix} \mathbf{u}_1 & \dots & \mathbf{u}_k \end{bmatrix}$  form a basis of  $f[U]$ .

*Proof.* First, we compute (next slide):

*Proof (continued).*

$$\begin{aligned} f[U] &= f[\text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)] \\ &\stackrel{(*)}{=} \text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)) \\ &\stackrel{(**)}{=} \text{Col}\left( \begin{bmatrix} f(\mathbf{u}_1) & \dots & f(\mathbf{u}_k) \end{bmatrix} \right) \\ &\stackrel{(***)}{=} \text{Col}\left( \begin{bmatrix} A\mathbf{u}_1 & \dots & A\mathbf{u}_k \end{bmatrix} \right) \\ &\stackrel{(***)}{=} \text{Col}\left( A \begin{bmatrix} \mathbf{u}_1 & \dots & \mathbf{u}_k \end{bmatrix} \right), \end{aligned}$$

where (\*) follows from Theorem 4.2.11(b), (\*\*) follows from the definition of the column space, and (\*\*\*) follows from the fact that  $A$  is the standard matrix of  $f$ , and (\*\*\*\*) follows from the definition of matrix multiplication.

*Proof (continued).*

$$\begin{aligned} f[U] &= f[\text{Span}(\mathbf{u}_1, \dots, \mathbf{u}_k)] \\ &\stackrel{(*)}{=} \text{Span}(f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)) \\ &\stackrel{(**)}{=} \text{Col}\left( \begin{bmatrix} f(\mathbf{u}_1) & \dots & f(\mathbf{u}_k) \end{bmatrix} \right) \\ &\stackrel{(***)}{=} \text{Col}\left( \begin{bmatrix} A\mathbf{u}_1 & \dots & A\mathbf{u}_k \end{bmatrix} \right) \\ &\stackrel{(***)}{=} \text{Col}\left( A \begin{bmatrix} \mathbf{u}_1 & \dots & \mathbf{u}_k \end{bmatrix} \right), \end{aligned}$$

where (\*) follows from Theorem 4.2.11(b), (\*\*) follows from the definition of the column space, and (\*\*\*) follows from the fact that  $A$  is the standard matrix of  $f$ , and (\*\*\*\*) follows from the definition of matrix multiplication. By Theorem 3.3.4, the pivot columns of a matrix form a basis of the column space of that matrix, and the result follows.  $\square$

### Proposition 4.2.18

Let  $\mathbb{F}$  be a field, let  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  be a linear function, let  $A \in \mathbb{F}^{n \times m}$  be the standard matrix of  $f$ , let  $\mathbf{v}_1, \dots, \mathbf{v}_k \in \mathbb{F}^n$  ( $k \geq 1$ ), and set  $V := \text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_k)$ . Then

$$\begin{aligned} f^{-1}[V] &= \left\{ \mathbf{x} \in \mathbb{F}^m \mid \exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \begin{bmatrix} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} = \mathbf{0} \right\} \\ &= \left\{ \mathbf{x} \in \mathbb{F}^m \mid \exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} \in \text{Nul} \left( \begin{bmatrix} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{bmatrix} \right) \right\}. \end{aligned}$$

- First an example, then a proof.

### Example 4.2.19

Consider the linear function  $f : \mathbb{R}^4 \rightarrow \mathbb{R}^5$  whose standard matrix is

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & -2 & -4 & 0 \\ -2 & -3 & -6 & 1 \\ 4 & 0 & 0 & 0 \\ 2 & -1 & -2 & 0 \end{bmatrix},$$

and consider the following vectors in  $\mathbb{R}^5$ :

- $\mathbf{v}_1 = [-1 \quad 6 \quad 9 \quad -4 \quad 1]^T$ ;
- $\mathbf{v}_2 = [2 \quad 2 \quad -2 \quad 8 \quad 5]^T$ ;
- $\mathbf{v}_3 = [0 \quad 0 \quad 0 \quad -1 \quad 0]^T$ ;
- $\mathbf{v}_4 = [0 \quad -2 \quad -3 \quad -1 \quad -1]^T$ ;
- $\mathbf{v}_5 = [0 \quad -1 \quad -2 \quad 1 \quad 0]^T$ ;
- $\mathbf{v}_6 = [-3 \quad -1 \quad 2 \quad -11 \quad -6]^T$ .

Set  $V := \text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_6)$ . Find a basis of  $f^{-1}[V]$ .



*Solution.* We apply Proposition 4.2.18.

*Solution.* We apply Proposition 4.2.18. We first form the matrix

$$\begin{aligned} C &:= \left[ A \mid \mathbf{v}_1 \quad \mathbf{v}_2 \quad \mathbf{v}_3 \quad \mathbf{v}_4 \quad \mathbf{v}_5 \quad \mathbf{v}_6 \right] \\ &= \left[ \begin{array}{cccc|cccccc} 1 & 0 & 0 & 0 & -1 & 2 & 0 & 0 & 0 & -3 \\ 0 & -2 & -4 & 0 & 6 & 2 & 0 & -2 & -1 & -1 \\ -2 & -3 & -6 & 1 & 9 & -2 & 0 & -3 & -2 & 2 \\ 4 & 0 & 0 & 0 & -4 & 8 & -1 & -1 & 1 & -11 \\ 2 & -1 & -2 & 0 & 1 & 5 & 0 & -1 & 0 & -6 \end{array} \right], \end{aligned}$$

and we find the general solution of the matrix-vector equation

$$\underbrace{\left[ A \mid \mathbf{v}_1 \quad \mathbf{v}_2 \quad \mathbf{v}_3 \quad \mathbf{v}_4 \quad \mathbf{v}_5 \quad \mathbf{v}_6 \right]}_{=C} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} = \mathbf{0},$$

where the vector  $\mathbf{x}$  has four entries (because  $A$  has four columns) and the vector  $\mathbf{y}$  has six entries (because we have six vectors  $\mathbf{v}_1, \dots, \mathbf{v}_6$ ).



*Solution (continued).* By row reducing, we obtain

$$\text{RREF}(C) = \left[ \begin{array}{cccc|cccccc} 1 & 0 & 0 & 0 & -1 & 2 & 0 & 0 & 0 & -3 \\ 0 & 1 & 2 & 0 & -3 & -1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -2 & -1 & 0 & 0 & 0 & -2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 \end{array} \right].$$

So, the general solution of our matrix-vector equation is

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} = \begin{bmatrix} q - 2r + 3t \\ -2p + 3q + r - s \\ p \\ 2q + r + 2t \\ \hline q \\ r \\ -s \\ s \\ -t \\ t \end{bmatrix}, \quad \text{where } p, q, r, s, t \in \mathbb{R}.$$

*Solution (continued).* By row reducing, we obtain

$$\text{RREF}(C) = \left[ \begin{array}{cccc|cccccc} 1 & 0 & 0 & 0 & -1 & 2 & 0 & 0 & 0 & -3 \\ 0 & 1 & 2 & 0 & -3 & -1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -2 & -1 & 0 & 0 & 0 & -2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 \end{array} \right].$$

So, the general solution of our matrix-vector equation is

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} = \begin{bmatrix} q - 2r + 3t \\ -2p + 3q + r - s \\ p \\ 2q + r + 2t \\ \hline q \\ r \\ -s \\ s \\ -t \\ t \end{bmatrix}, \quad \text{where } p, q, r, s, t \in \mathbb{R}.$$

But as per Proposition 4.2.18, we only need  $\mathbf{x}$ !

*Solution.* So, we simply ignore the part below the horizontal dotted line, and we obtain:

$$\mathbf{x} = \begin{bmatrix} q - 2r + 3t \\ -2p + 3q + r - s \\ p \\ 2q + r + 2t \end{bmatrix}, \quad \text{where } p, q, r, s, t \in \mathbb{R}.$$

*Solution.* So, we simply ignore the part below the horizontal dotted line, and we obtain:

$$\mathbf{x} = \begin{bmatrix} q - 2r + 3t \\ -2p + 3q + r - s \\ p \\ 2q + r + 2t \end{bmatrix}, \quad \text{where } p, q, r, s, t \in \mathbb{R}.$$

By separating parameters, we obtain

$$\mathbf{x} = p \begin{bmatrix} 0 \\ -2 \\ 1 \\ 0 \end{bmatrix} + q \begin{bmatrix} 1 \\ 3 \\ 0 \\ 2 \end{bmatrix} + r \begin{bmatrix} -2 \\ 1 \\ 0 \\ 1 \end{bmatrix} + s \begin{bmatrix} 0 \\ -1 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} 3 \\ 0 \\ 0 \\ 2 \end{bmatrix},$$

where  $p, q, r, s, t \in \mathbb{R}$ .

*Solution.* So, we simply ignore the part below the horizontal dotted line, and we obtain:

$$\mathbf{x} = \begin{bmatrix} q - 2r + 3t \\ -2p + 3q + r - s \\ p \\ 2q + r + 2t \end{bmatrix}, \quad \text{where } p, q, r, s, t \in \mathbb{R}.$$

By separating parameters, we obtain

$$\mathbf{x} = p \begin{bmatrix} 0 \\ -2 \\ 1 \\ 0 \end{bmatrix} + q \begin{bmatrix} 1 \\ 3 \\ 0 \\ 2 \end{bmatrix} + r \begin{bmatrix} -2 \\ 1 \\ 0 \\ 1 \end{bmatrix} + s \begin{bmatrix} 0 \\ -1 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} 3 \\ 0 \\ 0 \\ 2 \end{bmatrix},$$

where  $p, q, r, s, t \in \mathbb{R}$ .

In view of Proposition 4.2.18, we now have that (next slide):



*Solution.*

$$f^{-1}[V] = \left\{ p \begin{bmatrix} 0 \\ -2 \\ 1 \\ 0 \end{bmatrix} + q \begin{bmatrix} 1 \\ 3 \\ 0 \\ 2 \end{bmatrix} + r \begin{bmatrix} -2 \\ 1 \\ 0 \\ 1 \end{bmatrix} + s \begin{bmatrix} 0 \\ -1 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} 3 \\ 0 \\ 0 \\ 2 \end{bmatrix} \mid p, q, r, s, t \in \mathbb{R} \right\}$$

$$= \text{Span} \left( \begin{bmatrix} 0 \\ -2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 3 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} -2 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 0 \\ 0 \\ 2 \end{bmatrix} \right)$$

$$= \text{Col} \left( \underbrace{\begin{bmatrix} 0 & 1 & -2 & 0 & 3 \\ -2 & 3 & 1 & -1 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 1 & 0 & 2 \end{bmatrix}}_{=:B} \right).$$

*Solution (continued).* Reminder:

$$f^{-1}[V] = \text{Col}\left(\underbrace{\begin{bmatrix} 0 & 1 & -2 & 0 & 3 \\ -2 & 3 & 1 & -1 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 1 & 0 & 2 \end{bmatrix}}_{=:B}\right).$$

*Solution (continued).* Reminder:

$$f^{-1}[V] = \text{Col}\left(\underbrace{\begin{bmatrix} 0 & 1 & -2 & 0 & 3 \\ -2 & 3 & 1 & -1 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 1 & 0 & 2 \end{bmatrix}}_{=:B}\right).$$

We note that the five vectors that we obtained in the second-to-last line above are not necessarily linearly independent, and so to find an actual basis of  $f^{-1}[V]$ , we row reduce the matrix  $B$  and use Theorem 3.3.4. Indeed, Theorem 3.3.4 guarantees that the pivot columns of  $B$  form a basis of  $\text{Col}(B) = f^{-1}[V]$ .

- In fact, we can immediately see that they are not linearly independent: no five vectors in  $\mathbb{R}^4$  are linearly independent (by Theorem 3.2.17(a)).
- More generally, though, the reason our computation does not necessarily yield linearly independent vectors is because we “cut off” the entries below the vertical dotted line.

*Solution (continued).* Reminder:

$$f^{-1}[V] = \text{Col}\left(\underbrace{\begin{bmatrix} 0 & 1 & -2 & 0 & 3 \\ -2 & 3 & 1 & -1 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 1 & 0 & 2 \end{bmatrix}}_{=:B}\right).$$

*Solution (continued).* Reminder:

$$f^{-1}[V] = \text{Col}\left(\underbrace{\begin{bmatrix} 0 & 1 & -2 & 0 & 3 \\ -2 & 3 & 1 & -1 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 1 & 0 & 2 \end{bmatrix}}_{=:B}\right).$$

By row reducing, we obtain

$$\text{RREF}(B) = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 7/5 \\ 0 & 0 & 1 & 0 & -4/5 \\ 0 & 0 & 0 & 1 & 17/5 \end{bmatrix}.$$

Thus, the pivot columns of  $B$  are its leftmost four columns, and those four columns form a basis of  $f^{-1}[V]$ .

*Solution (continued).* So, our final answer is that

$$\left\{ \begin{bmatrix} 0 \\ -2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 3 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} -2 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \\ 0 \\ 0 \end{bmatrix} \right\}$$

is a basis of  $f^{-1}[V]$ .  $\square$

### Proposition 4.2.18

Let  $\mathbb{F}$  be a field, let  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  be a linear function, let  $A \in \mathbb{F}^{n \times m}$  be the standard matrix of  $f$ , let  $\mathbf{v}_1, \dots, \mathbf{v}_k \in \mathbb{F}^n$  ( $k \geq 1$ ), and set  $V := \text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_k)$ . Then

$$\begin{aligned} f^{-1}[V] &= \left\{ \mathbf{x} \in \mathbb{F}^m \mid \exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \begin{bmatrix} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{bmatrix} \begin{bmatrix} -\mathbf{x} \\ \mathbf{y} \end{bmatrix} = \mathbf{0} \right\} \\ &= \left\{ \mathbf{x} \in \mathbb{F}^m \mid \exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \begin{bmatrix} -\mathbf{x} \\ \mathbf{y} \end{bmatrix} \in \text{Nul} \left( \begin{bmatrix} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{bmatrix} \right) \right\}. \end{aligned}$$

*Proof.*

### Proposition 4.2.18

Let  $\mathbb{F}$  be a field, let  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  be a linear function, let  $A \in \mathbb{F}^{n \times m}$  be the standard matrix of  $f$ , let  $\mathbf{v}_1, \dots, \mathbf{v}_k \in \mathbb{F}^n$  ( $k \geq 1$ ), and set  $V := \text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_k)$ . Then

$$\begin{aligned} f^{-1}[V] &= \left\{ \mathbf{x} \in \mathbb{F}^m \mid \exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \begin{bmatrix} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{bmatrix} \begin{bmatrix} -\mathbf{x} \\ \mathbf{y} \end{bmatrix} = \mathbf{0} \right\} \\ &= \left\{ \mathbf{x} \in \mathbb{F}^m \mid \exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \begin{bmatrix} -\mathbf{x} \\ \mathbf{y} \end{bmatrix} \in \text{Nul} \left( \begin{bmatrix} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{bmatrix} \right) \right\}. \end{aligned}$$

*Proof.* Set  $A = \begin{bmatrix} \mathbf{a}_1 & \dots & \mathbf{a}_m \end{bmatrix}$ .



### Proposition 4.2.18

Let  $\mathbb{F}$  be a field, let  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  be a linear function, let  $A \in \mathbb{F}^{n \times m}$  be the standard matrix of  $f$ , let  $\mathbf{v}_1, \dots, \mathbf{v}_k \in \mathbb{F}^n$  ( $k \geq 1$ ), and set  $V := \text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_k)$ . Then

$$\begin{aligned} f^{-1}[V] &= \left\{ \mathbf{x} \in \mathbb{F}^m \mid \exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \begin{bmatrix} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{bmatrix} \begin{bmatrix} -\mathbf{x} \\ \mathbf{y} \end{bmatrix} = \mathbf{0} \right\} \\ &= \left\{ \mathbf{x} \in \mathbb{F}^m \mid \exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \begin{bmatrix} -\mathbf{x} \\ \mathbf{y} \end{bmatrix} \in \text{Nul} \left( \begin{bmatrix} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{bmatrix} \right) \right\}. \end{aligned}$$

*Proof.* Set  $A = \begin{bmatrix} \mathbf{a}_1 & \dots & \mathbf{a}_m \end{bmatrix}$ . Then for all vectors

$\mathbf{x} = \begin{bmatrix} x_1 & \dots & x_m \end{bmatrix}^T$  in  $\mathbb{F}^m$ , we have the following sequence of equivalent statements (next slide):

*Proof (continued).*

$$\mathbf{x} \in f^{-1}[V]$$

$$\iff f(\mathbf{x}) \in \underbrace{\text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_k)}_{=V}$$

$$\stackrel{(*)}{\iff} A\mathbf{x} \in \text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_k)$$

$$\iff \underbrace{x_1\mathbf{a}_1 + \dots + x_m\mathbf{a}_m}_{=A\mathbf{x}} \in \text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_k)$$

$$\stackrel{(**)}{\iff} \exists \alpha_1, \dots, \alpha_k \in \mathbb{F} \text{ s.t. } x_1\mathbf{a}_1 + \dots + x_m\mathbf{a}_m = \alpha_1\mathbf{v}_1 + \dots + \alpha_k\mathbf{v}_k$$

$$\iff \exists \alpha_1, \dots, \alpha_k \in \mathbb{F} \text{ s.t. } x_1\mathbf{a}_1 + \dots + x_m\mathbf{a}_m - \alpha_1\mathbf{v}_1 - \dots - \alpha_k\mathbf{v}_k = \mathbf{0},$$

where (\*) follows from the fact that  $A$  is the standard matrix of  $f$ , and (\*\*) follows from the definition of span.

*Proof (continued).*

$$\mathbf{x} \in f^{-1}[V]$$

previous

slide  
 $\Leftrightarrow$

$$\exists \alpha_1, \dots, \alpha_k \in \mathbb{F} \text{ s.t. } x_1 \mathbf{a}_1 + \dots + x_m \mathbf{a}_m - \alpha_1 \mathbf{v}_1 - \dots - \alpha_k \mathbf{v}_k = \mathbf{0}$$

(\*\*\*)  
 $\Leftrightarrow$

$$\exists y_1, \dots, y_k \in \mathbb{F} \text{ s.t. } x_1 \mathbf{a}_1 + \dots + x_m \mathbf{a}_m + y_1 \mathbf{v}_1 + \dots + y_k \mathbf{v}_k = \mathbf{0}$$

$\Leftrightarrow$

$$\exists y_1, \dots, y_k \in \mathbb{F} \text{ s.t. } \left[ \begin{array}{cccc} \mathbf{a}_1 & \dots & \mathbf{a}_m & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{array} \right] \begin{bmatrix} x_1 \\ \vdots \\ x_m \\ y_1 \\ \vdots \\ y_k \end{bmatrix} = \mathbf{0}$$

$\Leftrightarrow$

$$\exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \left[ \begin{array}{c} A \\ \mathbf{v}_1 \quad \dots \quad \mathbf{v}_k \end{array} \right] \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} = \mathbf{0},$$

where (\*\*\*) follows via substitution  $y_i := -\alpha_i \forall i \in \{1, \dots, k\}$ .

*Proof (continued).*

$$\mathbf{x} \in f^{-1}[V]$$

previous  
slide  
 $\Leftrightarrow$

$$\exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \left[ \begin{array}{c|ccc} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{array} \right] \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} = \mathbf{0}$$

(\*\*\*\*)  
 $\Leftrightarrow$

$$\exists \mathbf{y} \in \mathbb{F}^k \text{ s.t. } \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} \in \text{Nul} \left( \left[ \begin{array}{c|ccc} A & \mathbf{v}_1 & \dots & \mathbf{v}_k \end{array} \right] \right),$$

where (\*\*\*\*) follows from the definition of the null space. The result is now immediate.  $\square$

### ③ Linear functions and bases

### 3 Linear functions and bases

- Reminder:

#### Theorem 1.10.5

Let  $\mathbb{F}$  be a field, and let  $\mathbf{a}_1, \dots, \mathbf{a}_m$  be any vectors in  $\mathbb{F}^n$ . Then there exists a **unique** linear function  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  that satisfies  $f(\mathbf{e}_1) = \mathbf{a}_1, \dots, f(\mathbf{e}_m) = \mathbf{a}_m$ , where  $\mathbf{e}_1, \dots, \mathbf{e}_m$  are the standard basis vectors of  $\mathbb{F}^m$ . Moreover, this linear function  $f$  is given by  $f(\mathbf{x}) = A\mathbf{x}$  for all  $\mathbf{x} \in \mathbb{F}^m$ , where  $A = \begin{bmatrix} \mathbf{a}_1 & \dots & \mathbf{a}_m \end{bmatrix}$ .

### 3 Linear functions and bases

- Reminder:

#### Theorem 1.10.5

Let  $\mathbb{F}$  be a field, and let  $\mathbf{a}_1, \dots, \mathbf{a}_m$  be any vectors in  $\mathbb{F}^n$ . Then there exists a **unique** linear function  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  that satisfies  $f(\mathbf{e}_1) = \mathbf{a}_1, \dots, f(\mathbf{e}_m) = \mathbf{a}_m$ , where  $\mathbf{e}_1, \dots, \mathbf{e}_m$  are the standard basis vectors of  $\mathbb{F}^m$ . Moreover, this linear function  $f$  is given by  $f(\mathbf{x}) = A\mathbf{x}$  for all  $\mathbf{x} \in \mathbb{F}^m$ , where  $A = \begin{bmatrix} \mathbf{a}_1 & \dots & \mathbf{a}_m \end{bmatrix}$ .

- Our next goal is to generalize Theorem 1.10.5 to linear functions  $f : U \rightarrow V$ , where  $U$  and  $V$  are vector spaces over a field  $\mathbb{F}$ , and  $U$  is **finite-dimensional**.
  - Instead of using the standard basis  $\mathcal{E}_m = \{\mathbf{e}_1, \dots, \mathbf{e}_m\}$ , we will use an arbitrary basis of  $U$ .

- Suppose that  $V$  is a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and that  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ .



- Suppose that  $V$  is a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and that  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ .
- By Theorem 3.2.7, every vector of  $V$  can be written as linear combination of the vectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$  in a unique way, that is,  $\forall \mathbf{v} \in V \exists! \alpha_1, \dots, \alpha_n \in \mathbb{F}$  s.t.

$$\mathbf{v} := \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n,$$

and the *coordinate vector* of  $\mathbf{v}$  with respect to the basis  $\mathcal{B}$  is defined to be

$$[\mathbf{v}]_{\mathcal{B}} := \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix}.$$

- Suppose that  $V$  is a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and that  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ .
- By Theorem 3.2.7, every vector of  $V$  can be written as linear combination of the vectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$  in a unique way, that is,  $\forall \mathbf{v} \in V \exists! \alpha_1, \dots, \alpha_n \in \mathbb{F}$  s.t.

$$\mathbf{v} := \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n,$$

and the *coordinate vector* of  $\mathbf{v}$  with respect to the basis  $\mathcal{B}$  is defined to be

$$[\mathbf{v}]_{\mathcal{B}} := \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix}.$$

- As our next proposition shows,  $[\cdot]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.
  - It essentially allows us to “translate” vectors of an  $n$ -dimensional vector space ( $n \neq 0$ ) into vectors in  $\mathbb{F}^n$ .

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof.*

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof.* We start by proving that  $\left[ \cdot \right]_{\mathcal{B}}$  is linear.

1. Fix  $\mathbf{x}, \mathbf{y} \in V$ . WTS  $\left[ \mathbf{x} + \mathbf{y} \right]_{\mathcal{B}} = \left[ \mathbf{x} \right]_{\mathcal{B}} + \left[ \mathbf{y} \right]_{\mathcal{B}}$ .

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof.* We start by proving that  $\left[ \cdot \right]_{\mathcal{B}}$  is linear.

1. Fix  $\mathbf{x}, \mathbf{y} \in V$ . WTS  $\left[ \mathbf{x} + \mathbf{y} \right]_{\mathcal{B}} = \left[ \mathbf{x} \right]_{\mathcal{B}} + \left[ \mathbf{y} \right]_{\mathcal{B}}$ . Set  $\left[ \mathbf{x} \right]_{\mathcal{B}} = \left[ \alpha_1 \ \dots \ \alpha_n \right]^T$  and  $\left[ \mathbf{y} \right]_{\mathcal{B}} = \left[ \beta_1 \ \dots \ \beta_n \right]^T$ .

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

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1. Fix  $\mathbf{x}, \mathbf{y} \in V$ . WTS  $\left[ \mathbf{x} + \mathbf{y} \right]_{\mathcal{B}} = \left[ \mathbf{x} \right]_{\mathcal{B}} + \left[ \mathbf{y} \right]_{\mathcal{B}}$ . Set

$$\left[ \mathbf{x} \right]_{\mathcal{B}} = \begin{bmatrix} \alpha_1 & \dots & \alpha_n \end{bmatrix}^T \text{ and } \left[ \mathbf{y} \right]_{\mathcal{B}} = \begin{bmatrix} \beta_1 & \dots & \beta_n \end{bmatrix}^T.$$

Then  $\mathbf{x} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$  and  $\mathbf{y} = \beta_1 \mathbf{v}_1 + \dots + \beta_n \mathbf{v}_n$ ;

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $[\cdot]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof.* We start by proving that  $[\cdot]_{\mathcal{B}}$  is linear.

1. Fix  $\mathbf{x}, \mathbf{y} \in V$ . WTS  $[\mathbf{x} + \mathbf{y}]_{\mathcal{B}} = [\mathbf{x}]_{\mathcal{B}} + [\mathbf{y}]_{\mathcal{B}}$ . Set  $[\mathbf{x}]_{\mathcal{B}} = [\alpha_1 \ \dots \ \alpha_n]^T$  and  $[\mathbf{y}]_{\mathcal{B}} = [\beta_1 \ \dots \ \beta_n]^T$ . Then  $\mathbf{x} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$  and  $\mathbf{y} = \beta_1 \mathbf{v}_1 + \dots + \beta_n \mathbf{v}_n$ ; consequently,

$$\mathbf{x} + \mathbf{y} = (\alpha_1 + \beta_1) \mathbf{v}_1 + \dots + (\alpha_n + \beta_n) \mathbf{v}_n,$$

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof.* We start by proving that  $\left[ \cdot \right]_{\mathcal{B}}$  is linear.

1. Fix  $\mathbf{x}, \mathbf{y} \in V$ . WTS  $\left[ \mathbf{x} + \mathbf{y} \right]_{\mathcal{B}} = \left[ \mathbf{x} \right]_{\mathcal{B}} + \left[ \mathbf{y} \right]_{\mathcal{B}}$ . Set  $\left[ \mathbf{x} \right]_{\mathcal{B}} = \left[ \alpha_1 \ \dots \ \alpha_n \right]^T$  and  $\left[ \mathbf{y} \right]_{\mathcal{B}} = \left[ \beta_1 \ \dots \ \beta_n \right]^T$ . Then  $\mathbf{x} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$  and  $\mathbf{y} = \beta_1 \mathbf{v}_1 + \dots + \beta_n \mathbf{v}_n$ ; consequently,

$$\mathbf{x} + \mathbf{y} = (\alpha_1 + \beta_1) \mathbf{v}_1 + \dots + (\alpha_n + \beta_n) \mathbf{v}_n,$$

$$\text{and so } \left[ \mathbf{x} + \mathbf{y} \right]_{\mathcal{B}} = \left[ \alpha_1 + \beta_1 \ \dots \ \alpha_n + \beta_n \right]^T.$$



### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof.* We start by proving that  $\left[ \cdot \right]_{\mathcal{B}}$  is linear.

1. Fix  $\mathbf{x}, \mathbf{y} \in V$ . WTS  $\left[ \mathbf{x} + \mathbf{y} \right]_{\mathcal{B}} = \left[ \mathbf{x} \right]_{\mathcal{B}} + \left[ \mathbf{y} \right]_{\mathcal{B}}$ . Set  $\left[ \mathbf{x} \right]_{\mathcal{B}} = \left[ \alpha_1 \ \dots \ \alpha_n \right]^T$  and  $\left[ \mathbf{y} \right]_{\mathcal{B}} = \left[ \beta_1 \ \dots \ \beta_n \right]^T$ . Then  $\mathbf{x} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$  and  $\mathbf{y} = \beta_1 \mathbf{v}_1 + \dots + \beta_n \mathbf{v}_n$ ; consequently,

$$\mathbf{x} + \mathbf{y} = (\alpha_1 + \beta_1) \mathbf{v}_1 + \dots + (\alpha_n + \beta_n) \mathbf{v}_n,$$

and so  $\left[ \mathbf{x} + \mathbf{y} \right]_{\mathcal{B}} = \left[ \alpha_1 + \beta_1 \ \dots \ \alpha_n + \beta_n \right]^T$ . We now have that (next slide):

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $[\cdot]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof (continued).*

$$\begin{aligned} [\mathbf{x} + \mathbf{y}]_{\mathcal{B}} &= \begin{bmatrix} \alpha_1 + \beta_1 \\ \vdots \\ \alpha_n + \beta_n \end{bmatrix} \\ &= \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix} + \begin{bmatrix} \beta_1 \\ \vdots \\ \beta_n \end{bmatrix} \\ &= [\mathbf{x}]_{\mathcal{B}} + [\mathbf{y}]_{\mathcal{B}}. \end{aligned}$$

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof (continued).* Similarly (details: Lecture Notes):

$$2. \forall \mathbf{x} \in V, \alpha \in \mathbb{F}: \left[ \alpha \mathbf{x} \right]_{\mathcal{B}} = \alpha \left[ \mathbf{x} \right]_{\mathcal{B}}.$$

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof (continued).* Similarly (details: Lecture Notes):

$$2. \forall \mathbf{x} \in V, \alpha \in \mathbb{F}: \left[ \alpha \mathbf{x} \right]_{\mathcal{B}} = \alpha \left[ \mathbf{x} \right]_{\mathcal{B}}.$$

So,  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is linear.

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof (continued).* It remains to show that  $\left[ \cdot \right]_{\mathcal{B}}$  is a bijection, i.e. that it is one-to-one and onto  $\mathbb{F}^n$ .

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof (continued).* It remains to show that  $\left[ \cdot \right]_{\mathcal{B}}$  is a bijection, i.e. that it is one-to-one and onto  $\mathbb{F}^n$ .

Since  $V$  and  $\mathbb{F}^n$  are both  $n$  dimensional, Corollary 4.2.10 guarantees that  $f$  is one-to-one iff  $f$  is onto  $\mathbb{F}^n$ .

### Proposition 4.3.1

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $V$ . Then  $\left[ \cdot \right]_{\mathcal{B}} : V \rightarrow \mathbb{F}^n$  is an isomorphism.

*Proof (continued).* It remains to show that  $\left[ \cdot \right]_{\mathcal{B}}$  is a bijection, i.e. that it is one-to-one and onto  $\mathbb{F}^n$ .

Since  $V$  and  $\mathbb{F}^n$  are both  $n$  dimensional, Corollary 4.2.10 guarantees that  $f$  is one-to-one iff  $f$  is onto  $\mathbb{F}^n$ . So, it is enough to show that  $f$  is onto  $\mathbb{F}^n$ .

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$$\left[ \mathbf{v} \right]_{\mathcal{B}} = \left[ \alpha_1 \ \dots \ \alpha_n \right]^T.$$

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- Reminder:

### Theorem 1.10.5

Let  $\mathbb{F}$  be a field, and let  $\mathbf{a}_1, \dots, \mathbf{a}_m$  be any vectors in  $\mathbb{F}^n$ . Then there exists a **unique** linear function  $f : \mathbb{F}^m \rightarrow \mathbb{F}^n$  that satisfies  $f(\mathbf{e}_1) = \mathbf{a}_1, \dots, f(\mathbf{e}_m) = \mathbf{a}_m$ , where  $\mathbf{e}_1, \dots, \mathbf{e}_m$  are the standard basis vectors of  $\mathbb{F}^m$ . Moreover, this linear function  $f$  is given by  $f(\mathbf{x}) = A\mathbf{x}$  for all  $\mathbf{x} \in \mathbb{F}^m$ , where  $A = \begin{bmatrix} \mathbf{a}_1 & \dots & \mathbf{a}_m \end{bmatrix}$ .

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- Let's generalize this!

### Theorem 4.3.2

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and assume that  $U$  is finite-dimensional. Let  $\mathcal{B} = \{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be a basis of  $U$ , and let  $\mathbf{v}_1, \dots, \mathbf{v}_n \in V$ .<sup>a</sup> Then there exists a unique linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . Moreover, if the vector space  $U$  is non-trivial (i.e.  $n \neq 0$ ), then this unique linear function  $f : U \rightarrow V$  satisfies the following: for all  $\mathbf{u} \in U$ , we have that

$$f(\mathbf{u}) = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n,$$

where  $\begin{bmatrix} \mathbf{u} \end{bmatrix}_{\mathcal{B}} = \begin{bmatrix} \alpha_1 & \dots & \alpha_n \end{bmatrix}^T$ . On the other hand, if  $U$  is trivial (i.e.  $U = \{\mathbf{0}\}$ ),<sup>b</sup> then  $f : U \rightarrow V$  is given by  $f(\mathbf{0}) = \mathbf{0}$ .

---

<sup>a</sup>Here,  $\mathbf{v}_1, \dots, \mathbf{v}_n$  are arbitrary vectors in  $V$ . They are not necessarily pairwise distinct.

<sup>b</sup>Note that in this case, we have that  $n = 0$  and  $\mathcal{B} = \emptyset$ .

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*Proof.* Suppose first that the vector space  $U$  is trivial, i.e.  $n = 0$  and  $U = \{\mathbf{0}\}$ . Then the function  $f : U \rightarrow V$  given by  $f(\mathbf{0}) = \mathbf{0}$  is obviously linear, and moreover, it vacuously satisfies  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$  (because  $n = 0$ , and so both  $\mathbf{u}_1, \dots, \mathbf{u}_n$  and  $\mathbf{v}_1, \dots, \mathbf{v}_n$  are empty lists of vectors).

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From now on, we assume that the vector space  $U$  is non-trivial, i.e. that  $n \neq 0$ .

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From now on, we assume that the vector space  $U$  is non-trivial, i.e. that  $n \neq 0$ . We must prove the existence and the uniqueness of the linear function  $f$  satisfying the required properties.

*Proof (continued).* **Existence.** Let  $f : U \rightarrow V$  be defined as in the statement of the theorem, i.e. for all  $\mathbf{u} \in U$ , we set

$$f(\mathbf{u}) = \alpha_1 \mathbf{v}_1 + \cdots + \alpha_n \mathbf{v}_n,$$

where  $\left[ \mathbf{u} \right]_{\mathcal{B}} = \left[ \alpha_1 \quad \cdots \quad \alpha_n \right]^T$ .

*Proof (continued).* **Existence.** Let  $f : U \rightarrow V$  be defined as in the statement of the theorem, i.e. for all  $\mathbf{u} \in U$ , we set

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where  $\left[ \mathbf{u} \right]_{\mathcal{B}} = \left[ \alpha_1 \ \cdots \ \alpha_n \right]^T$ . Note that this means that for all  $\alpha_1, \dots, \alpha_n \in \mathbb{F}$ , we have that

$$f(\alpha_1 \mathbf{u}_1 + \cdots + \alpha_n \mathbf{u}_n) = \alpha_1 \mathbf{v}_1 + \cdots + \alpha_n \mathbf{v}_n.$$

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Let us show that  $f$  is linear and satisfies

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Let us show that  $f$  is linear and satisfies

$f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . For the latter, we note that for all  $i \in \{1, \dots, n\}$ , we have that

$$\begin{aligned} f(\mathbf{u}_i) &= f(0\mathbf{u}_1 + \cdots + 0\mathbf{u}_{i-1} + 1\mathbf{u}_i + 0\mathbf{u}_{i+1} + \cdots + 0\mathbf{u}_n) \\ &= 0\mathbf{v}_1 + \cdots + 0\mathbf{v}_{i-1} + 1\mathbf{v}_i + 0\mathbf{v}_{i+1} + \cdots + 0\mathbf{v}_n \\ &= \mathbf{v}_i. \end{aligned}$$

This proves that  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ .

*Proof (continued).* Let us now show that  $f$  is linear. We verify that  $f$  satisfies the two axioms from the definition of a linear function.

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$$\left[ \mathbf{x} \right]_{\mathcal{B}} = \left[ \alpha_1 \quad \dots \quad \alpha_n \right]^T \quad \text{and} \quad \mathbf{y} = \left[ \beta_1 \quad \dots \quad \beta_n \right]^T.$$

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$$\begin{aligned} f(\mathbf{x} + \mathbf{y}) &\stackrel{(*)}{=} (\alpha_1 + \beta_1)\mathbf{v}_1 + \dots + (\alpha_n + \beta_n)\mathbf{v}_n \\ &= (\alpha_1\mathbf{v}_1 + \dots + \alpha_n\mathbf{v}_n) + (\beta_1\mathbf{v}_1 + \dots + \beta_n\mathbf{v}_n) \\ &\stackrel{(**)}{=} f(\mathbf{x}) + f(\mathbf{y}), \end{aligned}$$

where both  $(*)$  and  $(**)$  follow from the construction of  $f$ .

*Proof (continued).* 2. Fix  $\mathbf{u} \in U$  and  $\alpha \in \mathbb{F}$ . WTS  $f(\alpha\mathbf{u}) = \alpha f(\mathbf{u})$ .

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Set  $\begin{bmatrix} \mathbf{u} \end{bmatrix}_{\mathcal{B}} = \begin{bmatrix} \alpha_1 & \dots & \alpha_n \end{bmatrix}^T$ . Then

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$\left[ \alpha\mathbf{u} \right]_{\mathcal{B}} = \left[ \alpha\alpha_1 \ \dots \ \alpha\alpha_n \right]^T$ , and we see that

$$\begin{aligned} f(\alpha\mathbf{u}) &\stackrel{(*)}{=} (\alpha\alpha_1)\mathbf{v}_1 + \dots + (\alpha\alpha_n)\mathbf{v}_n \\ &= \alpha(\alpha_1\mathbf{v}_1 + \dots + \alpha_n\mathbf{v}_n) \\ &\stackrel{(**)}{=} \alpha f(\mathbf{u}), \end{aligned}$$

where both (\*) and (\*\*) follow from the construction of  $f$ .

*Proof (continued).* 2. Fix  $\mathbf{u} \in U$  and  $\alpha \in \mathbb{F}$ . WTS  $f(\alpha\mathbf{u}) = \alpha f(\mathbf{u})$ .

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where both (\*) and (\*\*) follow from the construction of  $f$ .

By 1. and 2., we see that  $f$  is linear. This completes the proof of existence.

*Proof (continued).* **Uniqueness.** Let  $f_1, f_2 : U \rightarrow V$  be linear functions that satisfy  $f_1(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_1(\mathbf{u}_n) = \mathbf{v}_n$  and  $f_2(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_2(\mathbf{u}_n) = \mathbf{v}_n$ . WTS  $f_1 = f_2$ .

*Proof (continued).* **Uniqueness.** Let  $f_1, f_2 : U \rightarrow V$  be linear functions that satisfy  $f_1(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_1(\mathbf{u}_n) = \mathbf{v}_n$  and  $f_2(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_2(\mathbf{u}_n) = \mathbf{v}_n$ . WTS  $f_1 = f_2$ . Fix  $\mathbf{u} \in U$ . WTS  $f_1(\mathbf{u}) = f_2(\mathbf{u})$ .

*Proof (continued).* **Uniqueness.** Let  $f_1, f_2 : U \rightarrow V$  be linear functions that satisfy  $f_1(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_1(\mathbf{u}_n) = \mathbf{v}_n$  and  $f_2(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_2(\mathbf{u}_n) = \mathbf{v}_n$ . WTS  $f_1 = f_2$ . Fix  $\mathbf{u} \in U$ . WTS  $f_1(\mathbf{u}) = f_2(\mathbf{u})$ . Set  $\begin{bmatrix} \mathbf{u} \end{bmatrix}_{\mathcal{B}} = \begin{bmatrix} \alpha_1 & \dots & \alpha_n \end{bmatrix}^T$ .

*Proof (continued).* **Uniqueness.** Let  $f_1, f_2 : U \rightarrow V$  be linear functions that satisfy  $f_1(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_1(\mathbf{u}_n) = \mathbf{v}_n$  and  $f_2(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_2(\mathbf{u}_n) = \mathbf{v}_n$ . WTS  $f_1 = f_2$ . Fix  $\mathbf{u} \in U$ . WTS  $f_1(\mathbf{u}) = f_2(\mathbf{u})$ . Set  $\left[ \mathbf{u} \right]_{\mathcal{B}} = \left[ \alpha_1 \quad \dots \quad \alpha_n \right]^T$ . Then

$$\begin{aligned}
 f_1(\mathbf{u}) &= f_1(\alpha_1 \mathbf{u}_1 + \dots + \alpha_n \mathbf{u}_n) && \text{by the linearity of } f_1 \\
 &= \alpha_1 f_1(\mathbf{u}_1) + \dots + \alpha_n f_1(\mathbf{u}_n) && \text{(and more precisely,} \\
 & && \text{by Proposition 4.1.5)} \\
 &= \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n && \text{because} \\
 & && \underline{f_1(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_1(\mathbf{u}_n) = \mathbf{v}_n} \\
 &= \alpha_1 f_2(\mathbf{u}_1) + \dots + \alpha_n f_2(\mathbf{u}_n) && \text{because} \\
 & && \underline{f_2(\mathbf{u}_1) = \mathbf{v}_1, \dots, f_2(\mathbf{u}_n) = \mathbf{v}_n} \\
 &= f_2(\alpha_1 \mathbf{u}_1 + \dots + \alpha_n \mathbf{u}_n) && \text{by the linearity of } f_2 \\
 & && \text{(and more precisely,} \\
 & && \text{by Proposition 4.1.5)} \\
 &= f_2(\mathbf{u}).
 \end{aligned}$$

Thus,  $f_1 = f_2$ . This proves uniqueness.  $\square$

### Theorem 4.3.2

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and assume that  $U$  is finite-dimensional. Let  $\mathcal{B} = \{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be a basis of  $U$ , and let  $\mathbf{v}_1, \dots, \mathbf{v}_n \in V$ .<sup>a</sup> Then there exists a unique linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . Moreover, if the vector space  $U$  is non-trivial (i.e.  $n \neq 0$ ), then this unique linear function  $f : U \rightarrow V$  satisfies the following: for all  $\mathbf{u} \in U$ , we have that

$$f(\mathbf{u}) = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n,$$

where  $\begin{bmatrix} \mathbf{u} \end{bmatrix}_{\mathcal{B}} = \begin{bmatrix} \alpha_1 & \dots & \alpha_n \end{bmatrix}^T$ . On the other hand, if  $U$  is trivial (i.e.  $U = \{\mathbf{0}\}$ ),<sup>b</sup> then  $f : U \rightarrow V$  is given by  $f(\mathbf{0}) = \mathbf{0}$ .

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<sup>a</sup>Here,  $\mathbf{v}_1, \dots, \mathbf{v}_n$  are arbitrary vectors in  $V$ . They are not necessarily pairwise distinct.

<sup>b</sup>Note that in this case, we have that  $n = 0$  and  $\mathcal{B} = \emptyset$ .



### Corollary 4.3.3

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and assume that  $U$  is finite-dimensional. Let  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  be a linearly independent set of vectors in  $U$ , and let  $\mathbf{v}_1, \dots, \mathbf{v}_k \in V$ .<sup>a</sup> Then there exists a linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_k) = \mathbf{v}_k$ . Moreover, if  $V$  is non-trivial, then this linear function  $f$  is unique iff  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  is a basis of  $U$ .

---

<sup>a</sup>Here,  $\mathbf{v}_1, \dots, \mathbf{v}_k$  are arbitrary vectors in  $V$ . They are not necessarily pairwise distinct.

- **Remark:** If  $V$  is trivial (i.e.  $V = \{\mathbf{0}\}$ , and consequently  $\mathbf{v}_1 = \dots = \mathbf{v}_k = \mathbf{0}$ ), then there exists exactly one **function** from  $U$  to  $V$ , this function maps all elements of  $U$  to  $\mathbf{0}$ , and obviously, it is linear.

### Corollary 4.3.3

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and assume that  $U$  is finite-dimensional. Let  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  be a linearly independent set of vectors in  $U$ , and let  $\mathbf{v}_1, \dots, \mathbf{v}_k \in V$ .<sup>a</sup> Then there exists a linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_k) = \mathbf{v}_k$ . Moreover, if  $V$  is non-trivial, then this linear function  $f$  is unique iff  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  is a basis of  $U$ .

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*Proof (outline).*

### Corollary 4.3.3

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and assume that  $U$  is finite-dimensional. Let  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  be a linearly independent set of vectors in  $U$ , and let  $\mathbf{v}_1, \dots, \mathbf{v}_k \in V$ .<sup>a</sup> Then there exists a linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_k) = \mathbf{v}_k$ . Moreover, if  $V$  is non-trivial, then this linear function  $f$  is unique iff  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  is a basis of  $U$ .

---

<sup>a</sup>Here,  $\mathbf{v}_1, \dots, \mathbf{v}_k$  are arbitrary vectors in  $V$ . They are not necessarily pairwise distinct.

*Proof (outline).* Using Theorem 3.2.19, we extend  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  to a basis of  $U$ , and then we apply Theorem 4.3.2. The details are left as an exercise.  $\square$

## ④ Isomorphisms

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- Recall that, for vector spaces  $U$  and  $V$  over a field  $\mathbb{F}$ , a function  $f : U \rightarrow V$  is an *isomorphism* if it is linear and a bijection.

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- Recall that, for vector spaces  $U$  and  $V$  over a field  $\mathbb{F}$ , a function  $f : U \rightarrow V$  is an *isomorphism* if it is linear and a bijection.
- Vector spaces  $U$  and  $V$  (over the same field  $\mathbb{F}$ ) are *isomorphic*, and we write  $U \cong V$ , if there exists an isomorphism  $f : U \rightarrow V$ .

### Proposition 4.4.1

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be an isomorphism. Then  $f^{-1} : V \rightarrow U$  is also an isomorphism.

$$U \begin{array}{c} \xrightarrow{f} \\ \xleftarrow{f^{-1}} \end{array} V$$

*Proof.* The same as for isomorphisms  $f : \mathbb{F}^n \rightarrow \mathbb{F}^n$  (details: Lecture Notes).  $\square$

### Proposition 4.4.2

Let  $U$ ,  $V$ , and  $W$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  and  $g : V \rightarrow W$  be isomorphisms. Then  $g \circ f : U \rightarrow W$  is an isomorphism.

$$\begin{array}{ccccc} & & g \circ f & & \\ & \frown & & \smile & \\ U & \xrightarrow{f} & V & \xrightarrow{g} & W \end{array}$$

*Proof.*



### Proposition 4.4.2

Let  $U$ ,  $V$ , and  $W$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  and  $g : V \rightarrow W$  be isomorphisms. Then  $g \circ f : U \rightarrow W$  is an isomorphism.

$$\begin{array}{ccccc} & & g \circ f & & \\ & \frown & & \smile & \\ & f & & g & \\ U & \longrightarrow & V & \longrightarrow & W \end{array}$$

*Proof.* Since  $f : U \rightarrow V$  and  $g : V \rightarrow W$  are linear functions (because they are isomorphisms), Proposition 4.1.7 guarantees that their composition  $g \circ f : U \rightarrow W$  is also linear.

### Proposition 4.4.2

Let  $U$ ,  $V$ , and  $W$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  and  $g : V \rightarrow W$  be isomorphisms. Then  $g \circ f : U \rightarrow W$  is an isomorphism.

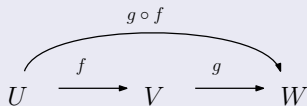
$$\begin{array}{ccccc} & & g \circ f & & \\ & \frown & & \smile & \\ U & \xrightarrow{f} & V & \xrightarrow{g} & W \end{array}$$

*Proof.* Since  $f : U \rightarrow V$  and  $g : V \rightarrow W$  are linear functions (because they are isomorphisms), Proposition 4.1.7 guarantees that their composition  $g \circ f : U \rightarrow W$  is also linear.

Since  $f : U \rightarrow V$  and  $g : V \rightarrow W$  are bijections, Proposition 1.10.17 guarantees that  $g \circ f : U \rightarrow W$  is also a bijection.

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*Proof.* Since  $f : U \rightarrow V$  and  $g : V \rightarrow W$  are linear functions (because they are isomorphisms), Proposition 4.1.7 guarantees that their composition  $g \circ f : U \rightarrow W$  is also linear.

Since  $f : U \rightarrow V$  and  $g : V \rightarrow W$  are bijections, Proposition 1.10.17 guarantees that  $g \circ f : U \rightarrow W$  is also a bijection.

So,  $g \circ f : U \rightarrow W$  is linear and a bijection, i.e. it is an isomorphism.  $\square$

### Theorem 4.4.3

Let  $U$ ,  $V$ , and  $W$  be vector spaces over a field  $\mathbb{F}$ . Then all the following hold:

- Ⓐ  $U \cong U$ ;
- Ⓑ if  $U \cong V$ , then  $V \cong U$ ;
- Ⓒ if  $U \cong V$  and  $V \cong W$ , then  $U \cong W$ .

*Proof.*

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*Proof.* (a) Clearly,  $\text{Id}_U : U \rightarrow U$  (the identity function on  $U$ ) is an isomorphism. So,  $U \cong U$ .

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- (b) if  $U \cong V$ , then  $V \cong U$ ;
- (c) if  $U \cong V$  and  $V \cong W$ , then  $U \cong W$ .

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(b) Suppose that  $U \cong V$ . Then there exists an isomorphism  $f : U \rightarrow V$ . But then by Proposition 4.4.2,  $f^{-1} : V \rightarrow U$  is also an isomorphism.



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*Proof.* (a) Clearly,  $\text{Id}_U : U \rightarrow U$  (the identity function on  $U$ ) is an isomorphism. So,  $U \cong U$ .

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(c) Suppose that  $U \cong V$  and  $V \cong W$ .

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(c) Suppose that  $U \cong V$  and  $V \cong W$ . Then there exist isomorphisms  $f : U \rightarrow V$  and  $g : V \rightarrow W$ .

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- Reminder: Theorem 4.2.13 (schematically and informally):

$$f : U \xrightarrow{\text{linear}} V$$

(a)-(b)	$\mathbf{u}_1, \dots, \mathbf{u}_k$ are linearly independent	$\xRightarrow{\text{if } f \text{ is 1-1}}$ $\xleftarrow{\text{always}}$	$f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$ are linearly independent
(c)-(d)	$\mathbf{u}_1, \dots, \mathbf{u}_k$ span $U$	$\xRightarrow{\text{if } f \text{ is onto}}$ $\xleftarrow{\text{if } f \text{ is 1-1}}$	$f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$ span $V$

- Reminder: Theorem 4.2.13 (schematically and informally):

$$f : U \xrightarrow{\text{linear}} V$$

(a)-(b)       $\mathbf{u}_1, \dots, \mathbf{u}_k$  are linearly independent       $\begin{array}{c} \text{if } f \text{ is 1-1} \\ \Rightarrow \\ \leftarrow \\ \text{always} \end{array}$        $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  are linearly independent

(c)-(d)       $\mathbf{u}_1, \dots, \mathbf{u}_k$  span  $U$        $\begin{array}{c} \text{if } f \text{ is onto} \\ \Rightarrow \\ \leftarrow \\ \text{if } f \text{ is 1-1} \end{array}$        $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  span  $V$

### Theorem 4.4.4

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , let  $f : U \rightarrow V$  be an isomorphism, and let  $\mathbf{u}_1, \dots, \mathbf{u}_k \in U$ . Then all the following hold:

- Ⓐ vectors  $\mathbf{u}_1, \dots, \mathbf{u}_k$  are linearly independent in  $U$  iff vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  are linearly independent in  $V$ ;
- Ⓑ vectors  $\mathbf{u}_1, \dots, \mathbf{u}_k$  span  $U$  iff vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  span  $V$ ;
- Ⓒ  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  is a basis of  $U$  iff  $\{f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)\}$  is a basis of  $V$ .

*Proof.* This follows from Theorem 4.2.13 (details: Lecture Notes).

### Theorem 4.4.4

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , let  $f : U \rightarrow V$  be an isomorphism, and let  $\mathbf{u}_1, \dots, \mathbf{u}_k \in U$ . Then all the following hold:

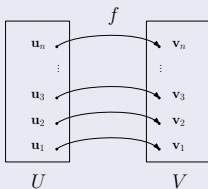
- Ⓐ vectors  $\mathbf{u}_1, \dots, \mathbf{u}_k$  are linearly independent in  $U$  iff vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  are linearly independent in  $V$ ;
- Ⓑ vectors  $\mathbf{u}_1, \dots, \mathbf{u}_k$  span  $U$  iff vectors  $f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)$  span  $V$ ;
- Ⓒ  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  is a basis of  $U$  iff  $\{f(\mathbf{u}_1), \dots, f(\mathbf{u}_k)\}$  is a basis of  $V$ .

- Proposition 4.4.5 (next slide) is a converse of sorts of Theorem 4.4.4(c).
  - It essentially states that any linear function that (injectively) maps a basis onto a basis is an isomorphism.



### Proposition 4.4.5

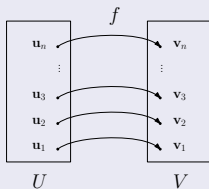
Let  $U$  and  $V$  be finite-dimensional vector spaces over a field  $\mathbb{F}$ . Assume that  $\dim(U) = \dim(V) =: n$ . Let  $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be a basis for  $U$ , and let  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  be a basis for  $V$ . Then there exists a unique linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . Moreover, this linear function  $f$  is an isomorphism.



*Proof.*

### Proposition 4.4.5

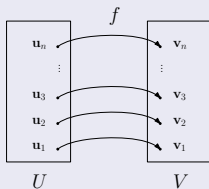
Let  $U$  and  $V$  be finite-dimensional vector spaces over a field  $\mathbb{F}$ . Assume that  $\dim(U) = \dim(V) =: n$ . Let  $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be a basis for  $U$ , and let  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  be a basis for  $V$ . Then there exists a unique linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . Moreover, this linear function  $f$  is an isomorphism.



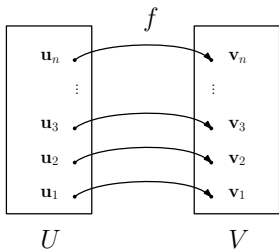
*Proof.* The existence and uniqueness of the linear function  $f$  follows from Theorem 4.3.2.

### Proposition 4.4.5

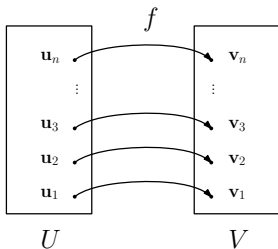
Let  $U$  and  $V$  be finite-dimensional vector spaces over a field  $\mathbb{F}$ . Assume that  $\dim(U) = \dim(V) =: n$ . Let  $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be a basis for  $U$ , and let  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  be a basis for  $V$ . Then there exists a unique linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . Moreover, this linear function  $f$  is an isomorphism.



*Proof.* The existence and uniqueness of the linear function  $f$  follows from Theorem 4.3.2. But by hypothesis,  $U$  and  $V$  are finite-dimensional vector spaces satisfying  $\dim(U) = \dim(V)$ , and so by Corollary 4.2.10, it is enough to show that  $f$  is onto.

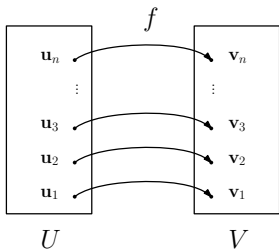


*Proof (continued).* Fix  $\mathbf{v} \in V$ .



*Proof (continued).* Fix  $\mathbf{v} \in V$ . Since  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis for  $V$ , we know that there exist scalars  $\alpha_1, \dots, \alpha_n \in \mathbb{F}$  s.t.

$$\mathbf{v} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n.$$



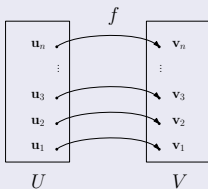
*Proof (continued).* Fix  $\mathbf{v} \in V$ . Since  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis for  $V$ , we know that there exist scalars  $\alpha_1, \dots, \alpha_n \in \mathbb{F}$  s.t.  
 $\mathbf{v} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$ . But now

$$\begin{aligned}
 f(\alpha_1 \mathbf{u}_1 + \dots + \alpha_n \mathbf{u}_n) &\stackrel{(*)}{=} \alpha_1 f(\mathbf{u}_1) + \dots + \alpha_n f(\mathbf{u}_n) \\
 &= \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n \\
 &= \mathbf{v},
 \end{aligned}$$

where  $(*)$  follows from the linearity of  $f$  (and more precisely, from Proposition 4.1.5). So,  $f$  is onto, and we are done.  $\square$

### Proposition 4.4.5

Let  $U$  and  $V$  be finite-dimensional vector spaces over a field  $\mathbb{F}$ . Assume that  $\dim(U) = \dim(V) =: n$ . Let  $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be a basis for  $U$ , and let  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  be a basis for  $V$ . Then there exists a unique linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . Moreover, this linear function  $f$  is an isomorphism.



- Reminder:

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- Ⓐ if  $f$  is one-to-one, then  $\dim(U) \leq \dim(V)$ ;
- Ⓑ if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;
- Ⓒ if  $f$  is an isomorphism, then  $\dim(U) = \dim(V)$ .



- Reminder:

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- Ⓐ if  $f$  is one-to-one, then  $\dim(U) \leq \dim(V)$ ;
  - Ⓑ if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;
  - Ⓒ if  $f$  is an isomorphism, then  $\dim(U) = \dim(V)$ .
- By Theorem 4.2.14(c), any two isomorphic vector spaces have the same dimension.

- Reminder:

### Theorem 4.2.14

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be a linear function. Then all the following hold:

- (a) if  $f$  is one-to-one, then  $\dim(U) \leq \dim(V)$ ;
- (b) if  $f$  is onto, then  $\dim(U) \geq \dim(V)$ ;
- (c) if  $f$  is an isomorphism, then  $\dim(U) = \dim(V)$ .

- By Theorem 4.2.14(c), any two isomorphic vector spaces have the same dimension.
- Theorem 4.4.6 (next slide) guarantees that, in the case of **finite-dimensional** vector spaces, the converse is also true: any two vector spaces (over the same field) that have the same finite dimension are isomorphic.
  - We give two proofs of Theorem 4.4.6!

### Theorem 4.4.6

Let  $U$  and  $V$  be **finite-dimensional** vector spaces over a field  $\mathbb{F}$ . Then  $U$  and  $V$  are isomorphic iff  $\dim(U) = \dim(V)$ .

- **Warning:** This theorem is only true for finite-dimensional vector spaces, and it becomes false for infinite-dimensional ones.

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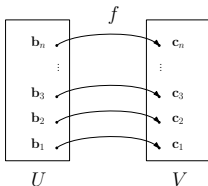
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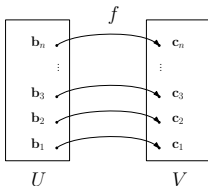
$f(\mathbf{b}_1) = \mathbf{c}_1, \dots, f(\mathbf{b}_n) = \mathbf{c}_n$ , and moreover, this linear function  $f$  is an isomorphism.



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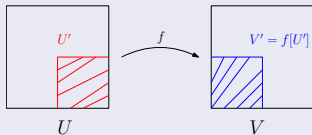
So,  $U$  and  $V$  are isomorphic.  $\square$



### Proposition 4.4.7

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and let  $f : U \rightarrow V$  be an isomorphism, and let  $U' \subseteq U$ . Then  $U'$  is a subspace of  $U$  iff  $V' := f[U']$  is a subspace of  $V$ . Moreover, in this case, all the following hold:

- (a) the function  $f' : U' \rightarrow V'$  given by  $f'(\mathbf{u}) = f(\mathbf{u})$  for all  $\mathbf{u} \in U'$  is an isomorphism;
- (b)  $U' \cong V'$ ;
- (c)  $\dim(U') = \dim(V')$ .



*Proof.* Lecture Notes.  $\square$