

# Lecture 11:

## Isomorphisms

### Dot product

### Matrices as linear functions

---

#### Warmup.

1. Let  $f : X \rightarrow Y$  be a function. Recall/prove each of the following statements.
  - (a) The function  $f$  is injective if and only if it has a left inverse, i.e. a function  $g : Y \rightarrow X$  such that  $g \circ f = \text{id}_X$ .
  - (b) The function  $f$  is surjective if and only if it has a right inverse, i.e. a function  $h : Y \rightarrow X$  such that  $f \circ h = \text{id}_Y$ .
  - (c) If  $f$  has both a left inverse  $g : Y \rightarrow X$  and a right inverse  $h : Y \rightarrow X$ , then  $g = h$ . (In this case, we say  $f$  is **invertible** and write  $g = h = f^{-1}$ .  
From (a) and (b), we know that  $f$  is invertible if and only if it's bijective.)
  - (d) Suppose  $f : X \rightarrow Y$  and  $g : Y \rightarrow Z$  are both bijective functions. Then  $g \circ f$  is also bijective.
2. Let  $f : U \rightarrow V$  and  $g : V \rightarrow W$  be linear functions (where  $U, V, W$  are all vector spaces over  $F$ ). Prove that  $g \circ f : U \rightarrow W$  is also linear.

## Injective linear functions (Let $U$ and $V$ denote vector spaces over $F$ .)

### Very Useful Theorem 1.

A linear function  $h : U \rightarrow V$  is injective if and only if  $\mathcal{N}(h) = 0$ .

**Proof.**

( $\Rightarrow$ ) Suppose  $h$  is injective. Compute  $\mathcal{N}(h)$ .

( $\Leftarrow$ ) Suppose  $\mathcal{N}(h) = 0$ .

Suppose  $h(\mathbf{x}) = h(\mathbf{y})$  for some  $\mathbf{x}, \mathbf{y} \in U$ .

**Corollary 2.** If  $h : U \rightarrow V$  is linear and  $V$  is finite-dimensional, then the following are equivalent:

1.  $h$  is injective;
  2.  $\text{nullity}(h) = 0$ ;
  3.  $\text{rank}(h) = \dim(U)$ ;
  4. If  $\mathcal{B}$  is a basis for  $V$ , then  $h(\mathcal{B})$  is a basis for  $\mathcal{R}(h)$  (as a multiset).
- (See our proof of Rank-Nullity/homework.)

## Isomorphisms (Let $U$ and $V$ denote vector spaces over $F$ .)

We call a bijective linear function an **isomorphism**.

**Example.** Given an ordered basis  $B = \langle \mathbf{b}_1, \dots, \mathbf{b}_n \rangle$  of a vector space  $V$ , the representation  $\text{Rep}_B : V \rightarrow F^n$ , given by

$$c_1 \mathbf{b}_1 + \dots + c_n \mathbf{b}_n \mapsto (c_1, \dots, c_n),$$

is an isomorphism.

For example, using the standard ordered bases, we have the isomorphisms

$$\mathcal{P}_n(F) \rightarrow F^{n+1} \quad \text{defined by} \quad c_0 + c_1 x + \dots + c_n x^n \mapsto (c_0, c_1, \dots, c_n);$$

and

$$M_2(F) \rightarrow F^4 \quad \text{defined by} \quad \begin{pmatrix} a & b \\ c & d \end{pmatrix} \mapsto (a, b, c, d).$$

### Lemma 3.

If  $h : U \rightarrow V$  is an isomorphism, then  $h^{-1} : V \rightarrow U$  is also an isomorphism.

**Proof:** Exercise! *Hint:* Similarly to the warmup,

$$f^{-1}(\mathbf{v} + \mathbf{w}) = f^{-1}(f(f^{-1}(\mathbf{v})) + f(f^{-1}(\mathbf{w}))) \dots$$

## Isomorphisms

(Let  $U$  and  $V$  denote vector spaces over  $F$ .)

We say that  $U$  is **isomorphic** to  $V$  if there exists an isomorphism  $h : U \rightarrow V$ .

If so, we write  $U \cong V$ . (There might be lots of iso's!)

**Examples:** We just saw that  $P_n(F) \cong F^{n+1}$  and  $M_2(F) \cong F^4$ .

**Theorem 4.** Isomorphism gives an equivalence relation on the set of vector spaces over a field  $F$ . Namely, the relation

$$U \sim V \quad \text{whenever} \quad U \cong V,$$

is reflexive, symmetric, and transitive. (See warmup & Lemma 3.)

**AMAZING Thm 5.** For vector spaces  $U$  and  $V$  over a field  $F$ ,

$$U \cong V \quad \text{if and only if} \quad \dim(U) = \dim(V).$$

**Proof.** ( $\Rightarrow$ ) Suppose there is an isomorphism  $h : U \rightarrow V$ . Let  $\mathcal{B}$  be a basis of  $U$ . Then by Cor. 2 #4,  $h(\mathcal{B})$  is a basis of  $V$  (as a multiset), and

$$\dim(U) = |\mathcal{B}| = |h(\mathcal{B})| = \dim(V).$$

( $\Leftarrow$ ) Suppose  $\dim(U) = \dim(V)$ . Let  $\mathcal{A}$  and  $\mathcal{B}$  be bases of  $U$  and  $V$ , respectively. Namely,  $|\mathcal{A}| = |\mathcal{B}|$  (i.e. there's a bijective function  $f : \mathcal{A} \rightarrow \mathcal{B}$ ).

**Pf 1:** If  $|\mathcal{A}| = |\mathcal{B}| = n$ , then both  $U$  and  $V$  are isomorphic to  $F^n$  by  $\text{Rep}_{\mathcal{A}}$  and  $\text{Rep}_{\mathcal{B}}$ , respectively. Hence  $U \cong V$  (by Thm. 4).

**Pf 2:** Since  $\mathcal{B} \subseteq V$ , we can linearly extend  $f : \mathcal{A} \rightarrow \mathcal{B} \subseteq V$  to a linear function  $h : U \rightarrow V$  that has  $h(\mathbf{a}) = f(\mathbf{a}) \in \mathcal{B}$  for all  $\mathbf{a} \in \mathcal{A}$ . By Cor. 2,  $h$  is an isomorphism. Hence  $U \cong V$ .  $\square$

## Isomorphisms

(Let  $U$  and  $V$  denote vector spaces over  $F$ .)

**Corollary 6.** If  $V$  is a **finite-dimensional** vector space, and  $h : V \rightarrow V$  is linear map from  $V$  **to itself**, then the following are equivalent:

1.  $h$  is injective;
2.  $h$  is surjective;
3.  $h$  is an isomorphism.

We call a homomorphism of the form  $h : V \rightarrow V$  an **endomorphism**.

**Proof.** Use dimension! Three BIG facts:

- (I) The null space  $\mathcal{N}(h)$  and the range  $\mathcal{R}(h)$  are both subspaces of  $V$ .
- (II) If  $Y$  is a vector space and  $X \subseteq Y$  is a subspace, then  $\dim(X) \leq \dim(Y)$ . And if  $Y$  is *finite dimensional*, then

$$\dim(X) = \dim(Y) \quad \text{if and only if} \quad X = Y.$$

- (III) **Thm. 5:**  $U \cong V$  if and only if  $\dim(U) = \dim(V)$ .

---

**Caution:** If  $V$  is infinite-dimensional, there are linear maps that are injective but not surjective, and vice versa.

**Examples:**

$$f : \mathbb{R}[x] \rightarrow \mathbb{R}[x] \text{ by } p(x) \mapsto p(x^2)$$

$$d : \mathbb{R}[x] \rightarrow \mathbb{R}[x] \text{ by } p(x) \mapsto \frac{d}{dx}p(x)$$

$$\iota : \mathbb{R}[x] \rightarrow \mathbb{R}[x] \text{ by } p(x) \mapsto \int_0^x p(t) dt$$

## Back to matrices!

The augmented matrix

$$\left( \begin{array}{cccc|c} 3 & 2 & -1 & 0 & b_1 \\ 4 & 0 & -5 & 1 & b_2 \\ 0 & 1 & 2 & 3 & b_3 \end{array} \right)$$

encodes the linear system

$$\begin{pmatrix} 3x_1 + 2x_2 + (-1)x_3 + 0x_4 \\ 4x_1 + 0x_2 + (-5)x_3 + (1)x_4 \\ 0x_1 + (1)x_2 + 2x_3 + 3x_4 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}.$$

Define the **dot product** of two vectors in  $F^n$  by  $\cdot : F^n \times F^n \rightarrow F$

$$(c_1, c_2, \dots, c_n) \cdot (d_1, d_2, \dots, d_n) := c_1d_1 + c_2d_2 + \dots + c_nd_n.$$

For example, for any  $x_1, \dots, x_n \in F$ ,

$$(3, 2, -1, 0) \cdot (x_1, x_2, x_3, x_4) = 3x_1 + 2x_2 + (-1)x_3 + 0x_4,$$

$$(4, 0, -5, 1) \cdot (x_1, x_2, x_3, x_4) = 4x_1 + 0x_2 + (-5)x_3 + (1)x_4, \text{ and}$$

$$(0, 1, 2, 3) \cdot (x_1, x_2, x_3, x_4) = 0x_1 + (1)x_2 + 2x_3 + 3x_4.$$

So the above linear system is *also* encoded by

$$\begin{pmatrix} (3, 2, -1, 0) \cdot (x_1, x_2, x_3, x_4) \\ (4, 0, -5, 1) \cdot (x_1, x_2, x_3, x_4) \\ (0, 1, 2, 3) \cdot (x_1, x_2, x_3, x_4) \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$

Notice

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} \mapsto \begin{pmatrix} (3, 2, -1, 0) \cdot (x_1, x_2, x_3, x_4) \\ (4, 0, -5, 1) \cdot (x_1, x_2, x_3, x_4) \\ (0, 1, 2, 3) \cdot (x_1, x_2, x_3, x_4) \end{pmatrix}$$

defines a function  $h : F^4 \mapsto F^3$ ! For example,

$$\begin{aligned} h \left( \begin{pmatrix} 1 \\ 5 \\ -2 \\ 0 \end{pmatrix} \right) &= \begin{pmatrix} (3, 2, -1, 0) \cdot (1, 5, -2, 0) \\ (4, 0, -5, 1) \cdot (1, 5, -2, 0) \\ (0, 1, 2, 3) \cdot (1, 5, -2, 0) \end{pmatrix} \\ &= \begin{pmatrix} 3(1) + 2(5) + (-1)(-2) + 0(0) \\ 4(1) + 0(5) + (-5)(-2) + (1)(0) \\ 0(1) + (1)(5) + 2(-2) + 3(0) \end{pmatrix} = \begin{pmatrix} 15 \\ 14 \\ 1 \end{pmatrix} \end{aligned}$$

It turns out that this is a linear function! [See HW 6 for properties of dot product.] Specifically,  $h$  is the linear extension of the function

$$\begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} \mapsto \begin{pmatrix} 3 \\ 4 \\ 0 \end{pmatrix}, \quad \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \end{pmatrix} \mapsto \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix}, \quad \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix} \mapsto \begin{pmatrix} -1 \\ -5 \\ 2 \end{pmatrix}, \quad \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \mapsto \begin{pmatrix} 0 \\ 1 \\ 3 \end{pmatrix}.$$

To a matrix  $A \in M_{k,\ell}(F)$ ,

$$A = \begin{pmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,\ell} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,\ell} \\ \vdots & \vdots & \ddots & \vdots \\ a_{k,1} & a_{k,2} & \cdots & a_{k,\ell} \end{pmatrix},$$

we associate a function  $A : F^\ell \rightarrow F^k$  given by

$$A : \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_\ell \end{pmatrix} \mapsto \begin{pmatrix} (a_{1,1}, \dots, a_{1,\ell}) \cdot (v_1, \dots, v_\ell) \\ (a_{2,1}, \dots, a_{2,\ell}) \cdot (v_1, \dots, v_\ell) \\ \vdots \\ (a_{k,1}, \dots, a_{k,\ell}) \cdot (v_1, \dots, v_\ell) \end{pmatrix} \quad \text{i.e.} \quad A : \mathbf{v} \mapsto \begin{pmatrix} \mathbf{r}_1 \cdot \mathbf{v}^T \\ \mathbf{r}_2 \cdot \mathbf{v}^T \\ \vdots \\ \mathbf{r}_k \cdot \mathbf{v}^T \end{pmatrix},$$

where  $\mathbf{r}_i$  is the  $i$ th row vector of  $A$ , and  $\mathbf{v}^T$  is the **transpose** of  $\mathbf{v}$  (from Homework 4). *Notation:* write  $A\mathbf{v}$  to mean  $A(\mathbf{v})$ .

**Example.** Back to our example from before,

$$\left( \begin{array}{cccc|c} 3 & 2 & -1 & 0 & b_1 \\ 4 & 0 & -5 & 1 & b_2 \\ 0 & 1 & 2 & 3 & b_3 \end{array} \right) \quad \text{and} \quad \begin{pmatrix} 3 & 2 & -1 & 0 \\ 4 & 0 & -5 & 1 \\ 0 & 1 & 2 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$

mean the same thing!

**You try:** Let

$$A = \begin{pmatrix} 0 & 1 & 3 \\ -1 & 2 & -1 \end{pmatrix} : F^3 \mapsto F^2.$$

(a) For  $\mathbf{v} = \begin{pmatrix} x \\ y \\ z \end{pmatrix} \in F^3$ , compute  $A\mathbf{v}$ .

(b) Recall that the standard (ordered) basis of  $F^3$  is

$$\mathcal{E} = \langle \mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3 \rangle, \quad \text{where} \quad \mathbf{e}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{e}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \quad \mathbf{e}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}.$$

Compute  $A\mathbf{e}$  for each  $\mathbf{e} \in \mathcal{E}$ .

(c) Compute  $\mathcal{N}(A)$  (the vectors  $\mathbf{v} \in F^3$  for which  $A\mathbf{v} = \mathbf{0}_{F^2}$ ).

(d) What did you notice about your answer to part (b)? Can you prove your answer for a general matrix?

(e) How does the range of the function  $A$  relate to spaces we've studied before having to do with matrices?

(f) How does the rank of the function  $A$  relate to statistics we've studied before having to do with matrices?

(g) How does the nullspace of  $A$  relate spaces we've studied before having to do with matrices? nullity?

## Solutions to warmup

1. Let  $f : X \rightarrow Y$  be a function. Recall/prove each of the following statements.
- (a) The function  $f$  is injective if and only if it has a left inverse, i.e. a function  $g : Y \rightarrow X$  such that  $g \circ f = \text{id}_X$ .  
( $\Rightarrow$ ) Build  $g$ ! For each  $y \in f(X)$ , there is exactly one  $x_y \in X$  w/  $f(x_y) = y$ ; define  $g(y) = x_y$ . And for  $y \in Y - f(X)$ , define  $g(y)$  arbitrarily.  
( $\Leftarrow$ ) If  $f(x) = f(x')$ , then  $x = g(f(x)) = g(f(x')) = x'$ .
- (b) The function  $f$  is surjective if and only if it has a right inverse, i.e. a function  $h : Y \rightarrow X$  such that  $f \circ h = \text{id}_Y$ .  
( $\Rightarrow$ ) Build  $h$ ! For each  $y \in Y$ , choose one  $x_y \in f^{-1}(y)$ , and define  $h(y) = x_y$ .  
( $\Leftarrow$ ) If  $y \in Y$ , then  $f(h(y)) = y$ , so  $y \in f(X)$ .
- (c) If  $f$  has both a left inverse  $g : Y \rightarrow X$  and a right inverse  $h : Y \rightarrow X$ , then  $g = h$ . (In this case, we say  $f$  is **invertible** and write  $g = h = f^{-1}$ . From (a) and (b), we know that  $f$  is invertible if and only if it's bijective.)  
Pf. We have  $h = \text{id}_Y \circ h = (g \circ f) \circ h = g \circ (f \circ h) = g \circ \text{id}_Y = g$ .
- (d) Suppose  $f : X \rightarrow Y$  and  $g : Y \rightarrow Z$  are both bijective functions. Then  $g \circ f$  is also bijective.  
Pf. Since  $f$  and  $g$  are bijective, they have two-sided inverses  $f^{-1}$  and  $g^{-1}$ . Then  $f^{-1} \circ g^{-1}$  is a two-sided inverse for  $g \circ f$ .

2. Let  $f : U \rightarrow V$  and  $g : V \rightarrow W$  be linear functions (where  $U, V, W$  are all vector spaces over  $F$ ). Prove that  $g \circ f : U \rightarrow W$  is also linear.

Pf. For all  $\mathbf{u}, \mathbf{v} \in U$  and  $\lambda \in F$ , we have

$$\begin{aligned}(g \circ f)(\mathbf{u} + \mathbf{v}) &= g(f(\mathbf{u} + \mathbf{v})) \\ &= g(f(\mathbf{u}) + f(\mathbf{v})) \\ &= g(f(\mathbf{u})) + g(f(\mathbf{v})) \\ &= (g \circ f)(\mathbf{u}) + (g \circ f)(\mathbf{v})\end{aligned}$$

and

$$\begin{aligned}(g \circ f)(\lambda \mathbf{u}) &= g(f(\lambda \mathbf{u})) \\ &= g(\lambda f(\mathbf{u})) \\ &= \lambda g(f(\mathbf{u})) \\ &= \lambda (g \circ f)(\mathbf{u}).\end{aligned}$$