#### **18.700 - LINEAR ALGEBRA, DAY 9 INVERTIBILITY AND ISOMORPHISMS**

SAM SCHIAVONE

#### **CONTENTS**



#### I. PRE-CLASS PLANNING

### I.1. **Goals for lesson.**

- (1) Students will learn the definition of matrix multiplication.
- (2) Students will learn that the row rank = the column rank of a matrix.
- (3) Students will learn the definition of invertibility and isomorphism.
- (4) Students will learn that an *n*-dimensional vector space is isomorphic to  $\mathbb{F}^n$ .
- (5) Students will learn that if dim(*V*) = *n* and dim(*W*) = *m*, then  $\mathcal{L}(\bar{V}, W) \cong M_{m \times n}(\mathbb{F})$ .
- (6) Students will learn that  $[T(v)]_C = c[T]_B[v]_B$ .
- (7) Students will learn the change of basis formula.

#### I.2. **Methods of assessment.**

- (1) Student responses to questions posed during lecture
- (2) Student responses to worksheet

# I.3. **Materials to bring.** (1) Laptop + adapter (2) Worksheets

# II. LESSON <sup>P</sup>LAN **(0:00)**

Announcements: • Exam 1: Wednesday, October 9th in class. No pset this week; instead review packet. • TA office hours: Tuesday, Oct 8th, 7:00 - 9:00pm, 2-361

#### II.1. **Last time.**

- Defined the image of a linear map.
- Rank-Nullity Theorem: If  $T: V \to W$  is linear, then  $\dim(V) = \dim(\ker(T)) +$  $dim(img(T)).$
- Defined coordinate vector  $[v]_{\mathcal{B}} \in \mathbb{F}^n$  for  $v \in V$ .
- Defined the matrix [*T*] of a linear map with respect to a choice of bases.

**Remark 1.** Linear vs affine. The function  $f : \mathbb{R} \to \mathbb{R}$ ,  $f(x) = mx + b$  is linear iff  $b = 0$ . (In general, these translates of linear maps are called *affine maps*.)

II.2. **Matrix multiplication.** Suppose that *U*, *V*, *W* are finite-dimensional vector spaces with bases

$$
\mathcal{B} := (u_1, \dots, u_p)
$$
  
\n
$$
\mathcal{C} := (v_1, \dots, v_n)
$$
  
\n
$$
\mathcal{D} := (w_1, \dots, w_m).
$$

Suppose  $T: U \rightarrow V$  and  $S: V \rightarrow W$  are linear maps. We previously saw that the composition  $ST: U \rightarrow W$  is linear. We now define matrix multiplication in such a way that

$$
[ST] = [S][T].
$$

Let  $A := [S]$  and  $B := [T]$ . Then for each  $j = 1, \ldots, p$  we have

$$
(ST)(u_j) = S\left(\sum_{k=1}^n B_{kj}v_k\right) = \sum_{k=1}^n B_{kj}S(v_k) = \sum_{k=1}^n B_{kj}\sum_{i=1}^m A_{ik}w_k = \sum_{i=1}^m \sum_{k=1}^n (A_{ik}B_{kj}) w_k.
$$

Thus  $\left[ST\right]$  is the  $m \times p$  matrix whose  $i, j$  entry is  $\sum$ *k*=1  $(A_{ik}B_{kj}).$ 

**Definition 2.** Given an  $m \times n$  matrix *A* and a  $n \times p$  matrix *B*, their product *AB* is defined to be the  $m \times p$  matrix whose *i*, *j* entry is *n* ∑ *k*=1  $(A_{ik}B_{kj}).$ 

So we multiply the entries of row *j* of *A* by those of column *k* of *B*, then add these together.

#### **Example 3.**

$$
\begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix} \begin{pmatrix} 6 & 5 & 0 & 0 \\ 2 & 1 & 0 & -1 \end{pmatrix} = \cdots
$$

2

[Ask students about the other order *BA*.]

**Proposition 4.** If  $T \in \mathcal{L}(U, V)$  and  $S \in \mathcal{L}(V, W)$ , then  $\left[ ST \right] = [S][T]$ .

*Proof.* This is true by the definition of matrix multiplication and the earlier calculation done as motivation.  $\Box$ 

Let *A* be an  $m \times n$  matrix.

- For  $i = 1, \ldots, m$ , let  $A_{i,j}$  denote row *i* of  $A$ , which is a  $1 \times n$  matrix.
- For  $j = 1, ..., n$ , let  $A_{.j}$  denote column *j* of  $A$ , which is an  $m \times 1$  matrix.

The next few results give different interpretations of matrix multiplication. Let *A* be an *m*  $\times$  *n* matrix and *B* be an *n*  $\times$  *p* matrix.

### **Lemma 5.**

$$
(AB)_{ij} = A_{i,.}B_{\cdot,j}
$$

*for all i* = 1, ..., *m* and all *j* = 1, ..., *p*. [Draw picture of row and column.]

*Proof.* True by formula defining matrix multiplication. □

### **Lemma 6.**

$$
(AB)_{\cdot,j} = A(B_{\cdot,j})
$$

*for all*  $j = 1, \ldots, p$ .

*Proof.* Exercise. Both are  $m \times 1$  matrices. Check that their  $i^{\text{th}}$  entries are equal using the formula. [Draw picture applying *A* to each of the columns of *B*.]  $\Box$ 

**Lemma 7.** Suppose A is 
$$
m \times n
$$
 and  $x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$  is  $n \times 1$ . Then  
\n $Ax = x_1 A_{\cdot 1} + \dots + x_n A_{\cdot n}$ .

*I.e., Ax is the linear combination of the columns of A with coefficients given by the entries of x.*

*Proof.* Exercise. □

### **Lemma 8.**

- *(a)* For  $j = 1, \ldots, p$ ,  $(AB)_{\cdot,j}$  *(column j) is a linear combination of the columns of A with coefficients from B*·,*<sup>j</sup> (column j).*
- *(b)* For  $i = 1, \ldots, m$ ,  $(AB)_{i}$ , (row i) is a linear combination of the rows of B with coefficients *from Ai*,· *(row i).*

*Proof.* Exercise. [Draw picture of second part.] □

### **Definition 9.**

- The *column space* of *A*, denoted Col(*A*), is the span of the columns of *A*. The *column rank* is the dimension of Col(*A*).
- The *row space* of *A*, denoted Row(*A*), is the span of the rows of *A*. The *row rank* is the dimension of Row(*A*).

We'll see that these two quantities are actually equal!

**Definition 10.** The *transpose* of a matrix *A*, denoted *A t* , is obtained from *A* by interchanging rows and columns. I.e.,

$$
(A^t)_{ij} = A_{ji}.
$$

**Lemma 11** (Column-row factorization). *Suppose A is m*  $\times$  *n and has column rank*  $c \in \mathbb{Z}_{\geq 1}$ *. Then there exist an m*  $\times$  *c matrix C and a c*  $\times$  *n matrix R such that A* = *CR. [Details left as an exercise.]*

*Proof.* The columns  $A_{\cdot,1}, \ldots, A_{\cdot,n}$  each an  $m \times 1$  matrix, span Col(A). By a previous result, this list can be reduced to a basis  $v_1, \ldots, v_c$  of Col(A), which by definition must have length *c*. Use these as the columns of a *m* × *c* matrix *C*.

For  $k = 1, \ldots, n$ , column *k* of *A* is a linear combination of the columns of *C* (since these are a basis), so there exist scalars  $R_{1k}, \ldots, R_{ck} \in \mathbb{F}$  such that

$$
A_k = R_{1k}v_1 + \cdots + R_{ck}v_c.
$$

Use the coefficients  $R_{1k}, \ldots, R_{ck}$  as the entries of the  $k^\text{th}$  column of a  $c \times n$  matrix  $R$ . Then  $A = CR$ .

**Theorem 12** (Column rank = row rank). *Suppose*  $A \in M_{m \times n}(\mathbb{F})$ . *Then the column rank and row rank of A are equal.*

*Proof.* Let *c* be the column rank of *A*. Let *A* = *CR* be the column-row factorization of *A* given by the previous lemma, where *C* is  $m \times c$  and *R* is  $c \times n$ . Since every row of *A* can be written as a linear combination of the rows of *R*, and *R* has *c* rows, then the row rank of *A* is  $\leq c$ , which is the column rank of *A*.

We obtain the reverse inequality by applying the same argument to  $A<sup>t</sup>$ , which yields

column rank of  $A =$  row rank of  $A^t \leq$  column rank of  $A^t =$  row rank of  $A$ .

**Definition 13.** The *rank* of a matrix is its column rank (= its row rank).

# II.3. **Invertibility and Isomorphisms.**

#### **Definition 14.**

- $T \in \mathcal{L}(V, W)$  is *invertible* if there exists a linear map  $S \in \mathcal{L}(W, V)$  such that  $ST = I_V$ and  $TS = I_W$ .
- With the above notation,  $S \in \mathcal{L}(W, V)$  is called an *inverse* of *T*.

**Lemma 15.** *An invertible linear map has a unique inverse.*

*Proof idea.* Given inverses *S*1, *S*2, then

$$
\cdots = S_1 T S_2 = \cdots
$$

□

□

If  $T$  is invertible, we denote its inverse by  $T^{-1}.$ 

#### **Example 16.**

• Let

$$
T: \mathbb{F}^2 \to \mathbb{F}^2
$$

$$
\begin{pmatrix} x \\ y \end{pmatrix} \mapsto \begin{pmatrix} x+y \\ y \end{pmatrix}.
$$

Then  $T^{-1}$  is given by

$$
\begin{pmatrix} x \\ y \end{pmatrix} \mapsto \begin{pmatrix} x - y \\ y \end{pmatrix}.
$$

[Write out at least one composition.]

• Let

$$
R: \mathbb{R}^2 \to \mathbb{R}^2
$$

$$
\begin{pmatrix} x \\ y \end{pmatrix} \mapsto \begin{pmatrix} \frac{1}{\sqrt{2}}x - \frac{1}{\sqrt{2}}y \\ \frac{1}{\sqrt{2}}x + \frac{1}{\sqrt{2}}y \end{pmatrix}.
$$

(Rotation counterclockwise by  $\pi/4$ .) <u>Claim</u>:  $R^{-1}$  is given by

$$
\begin{pmatrix} x \\ y \end{pmatrix} \mapsto \begin{pmatrix} \frac{1}{\sqrt{2}}x + \frac{1}{\sqrt{2}}y \\ -\frac{1}{\sqrt{2}}x + \frac{1}{\sqrt{2}}y \end{pmatrix}.
$$

**Lemma 17.** *A linear map*  $T \in \mathcal{L}(V, W)$  *is invertible iff it is injective and surjective.* 

*Proof.* ( $\Rightarrow$ ): Assume *T* is invertible. <u>One-to-one</u>: Suppose  $T(u) = T(v)$  for some  $u, v \in V$ . Applying  $T^{-1}$  to both sides, then

$$
u = T^{-1}(T(u)) = T^{-1}(T(v)) = v.
$$

 $Onto: Given *w* ∈ *W*, then  $T^{-1}(w) ∈ V$  and  $T(T^{-1}(w)) = w$ , so  $w ∈ \text{img}(T)$ .$ </u>

(∈): Assume *T* is injective and surjective. Given  $w \in W$ , since *T* is surjective then there exists *v*  $\in$  *V* such that *T*(*v*) = *w*. Suppose *v*<sub>1</sub>, *v*<sub>2</sub>  $\in$  *V* are both such preimages. Then

$$
T(v_1)=w=T(v_2)
$$

and since *T* is injective, then  $v_1 = v_2$ . Thus there is a *unique*  $v \in V$  such that  $T(v) = w$ . Define the map  $S: W \to V$  as follows: given  $w \in W$ , let  $v \in V$  be the unique element such that  $T(v) = w$ . Defined  $S(w) = v$ . Then by definition we have  $T(S(w)) = T(v) = w$ , so  $TS = I_W$ . It remains to show  $ST = I_V$ .

Given  $v \in V$ , then

$$
T((ST)(v)) = (TS)(T(v)) = I_W(T(v)) = T(v).
$$

Since *T* is one-to-one, then  $(ST)(v) = v$ . Thus  $ST = I_V$ .

It remains to show that *S* is linear. Suppose  $w_1, w_2 \in W$ . Then

$$
T(S(w_1) + S(w_2)) = T(S(w_1)) + T(S(w_2)) = w_1 + w_2.
$$

Now by definition,  $S(w_1 + w_2)$  is the unique element that maps to  $w_1 + w_2$  under *T*. Thus

$$
S(w_1) + S(w_2) = S(w_1 + w_2).
$$

The proof that *S* respects scalar multiplication is similar. □

**Theorem 18.** *Suppose V and W are finite-dimensional vector spaces with*  $dim(V) = dim(W)$ . *For any*  $T \in \mathcal{L}(V, W)$ , the following are equivalent.

*(i) T is invertible.*

*(ii) T is injective.*

*(iii) T is surjective.*

*Proof.* (ii)  $\implies$  (iii): Suppose *T* is injective. Then dim(ker(*T*)) = 0. By the Rank-Nullity Theorem, then

$$
\dim(V) = \dim(\ker(T)) + \dim(\text{img}(T)),
$$

so dim(img(*T*)) = dim(*V*) = dim(*W*). Then img(*T*) = *W*, so *W* is surjective.

 $(iii) \implies (ii)$ : Similar.

By previous result, (i)  $\iff$  (ii) and (iii), so they are all equivalent.

**Remark 19.** Warning! Finite-dimensionality is necessary in the above theorem. Consider the left-shift map

$$
L: \mathbb{F}^{\infty} \to \mathbb{F}^{\infty}
$$

$$
(x_1, x_2, x_3, \ldots) \mapsto (x_2, x_3, \ldots).
$$

This map is surjective but not injective.

**Proposition 20.** *Suppose V and W are finite-dimensional vector spaces with*  $dim(V) = dim(W)$ *. Given*  $T \in \mathcal{L}(V, W)$  *and*  $S \in \mathcal{L}(W, V)$ *, then*  $ST = I \iff TS = I$ .

*Proof.* ( $\Rightarrow$ ): Assume *ST* = *I*. Given  $v \in \text{ker}(T)$ , then  $T(v) = 0$ , so applying *S*, we have [start in middle]

$$
v = I(v) = S(T(v)) = S(0) = 0.
$$

Thus ker(*T*) = {0}, so *T* is injective. Since *V* and *W* have the same dimension, then *T* is invertible by the previous result. Thus  $T^{-1}$  exists. Applying  $T^{-1}$  on the right to both sides of  $I = ST$ , we have [start in middle]

$$
T^{-1} = IT^{-1} = STT^{-1} = S.
$$

Thus  $TS = TT^{-1} = I$ , as desired.

( $\Leftarrow$ ): Swap the roles of *S* and *T*.

II.3.1. *Isomorphic vector spaces.* The notion of isomorphism describes when two vector spaces are essentially "the same."

**Definition 21.** An *isomorphism* (of vector spaces) is an invertible linear map. Two vector spaces *V* and *W* are *isomorphic*, denoted  $V \cong W$ , if there is an isomorphism  $V \to W$ .

An isomorphism  $T: V \to W$  is essentially just a relabeling:  $v \in V$  is instead relabeled as  $T(v) \in W$ .

Q: How can we tell when two vector spaces are isomorphic?

**Theorem 22** (Dimension determines isomorphism)**.** *Two finite-dimensional vector spaces over* **F** *are isomorphic iff they have the same dimension.*

*Proof.* Suppose *V* and *W* are finite-dimensional vector spaces.

( $\Rightarrow$ ): Assume *V* and *W* are isomorphic. Then there exists an isomorphism *T* : *V* → *W*. Then *T* is injective and surjective so

$$
ker(T) = \{0\} \qquad \text{and} \qquad img(T) = W.
$$

By Rank-Nullity, then [ask students]

$$
\dim(V) = \dim(\ker(T)) + \dim(\text{img}(T)) = \dim(W).
$$

(⇐): Assume dim(*V*) = dim(*W*). Let  $v_1, \ldots, v_n$  be a basis for *V* and  $w_1, \ldots, w_n$  be a basis for *W*. By a previous result, there is a unique linear map  $T: V \rightarrow W$  such that  $T(v_i) = w_i$  for all  $i = 1, ..., n$ . Since  $w_1, ..., w_n$  span *W*, then *T* is surjective. Either by Rank-Nullity, or by using the fact that  $w_1, \ldots, w_n$  are linearly independent, *T* is injective. (Details left as exercise.) Thus *T* is injective and surjective, hence an isomorphism.  $\Box$ 

**Corollary 23.** Let V be an n-dimensional vector space. Then V is isomorphic to  $\mathbb{F}^n$ .

*Proof.* Both have dimension *n*. □

**Remark 24.** We can also give an explicit isomorphism. Choose a basis  $\mathcal{B} = (v_1, \ldots, v_n)$ for *V* and consider the coordinate vector map

$$
\varphi_{\mathcal{B}}: V \to \mathbb{F}^n
$$

$$
v \mapsto [v]_{\mathcal{B}}
$$

and the linear map

$$
S: \mathbb{F}^n \to V
$$
  
 $(a_1, \ldots, a_n) \mapsto a_1v_1 + \cdots + a_nv_n.$ 

Exercise: show these maps are mutually inverse isomorphisms.

**Example 25.**  $\mathcal{P}_m(\mathbb{F})$  has dimension [ask students]  $m+1$ , hence is isomorphic to  $\mathbb{F}^{m+1}$ .

**Proposition 26.** *Suppose*  $\mathcal{B} := (v_1, \ldots, v_n)$  *is a basis of V and*  $\mathcal{C} := (w_1, \ldots, w_m)$  *is a basis of W* (so dim(*V*) = *n* and dim(*W*) = *m*). Then the map

$$
\mathcal{L}(V, W) \to M_{m \times n}(\mathbb{F})
$$

$$
T \mapsto c[T]_{\mathcal{B}}
$$

*is an isomorphism.*

*Proof.* Exercise. □

**Corollary 27.** *Suppose V and W are finite-dimensional. Then*  $dim(\mathcal{L}(V, W)) = dim(V)$  dim(*W*).

II.3.2. *Linear maps as matrices.*

**Proposition 28** (Multiplication by a matrix is linear). Let  $A \in M_{m \times n}(\mathbb{F})$ . The left multipli*cation map*

$$
L_A: \mathbb{F}^n \to \mathbb{F}^m
$$

$$
v \mapsto Av
$$

*is linear.*

*Proof.* Considering *v* as an  $n \times 1$  matrix, this follows by properties of matrix multiplication.  $\Box$ 

Let *V* and *W* be vector spaces with bases  $\mathcal{B} := (v_1, \ldots, v_n)$  and  $\mathcal{C} := (w_1, \ldots, w_m)$ , respectively. Recall, for  $T: V \to W$  linear, the matrix of T with respect to B and C is given by

$$
c[T]B = \begin{pmatrix} | & | & | \\ [T(v_1)]_C & \cdots & [T(v_n)]_C \\ | & | & | \end{pmatrix}
$$

**Proposition 29.** *With notation as above,*

$$
[T(v)]_{\mathcal{C}} = c[T]_{\mathcal{B}}[v]_{\mathcal{B}}
$$

*for all*  $v \in V$ .

*Proof.* Given  $v \in V$ , there exist unique scalars  $a_1, \ldots, a_n \in \mathbb{F}$  such that  $v = a_1v_1 + \cdots + a_nv_n$ *anvn*. Since *T* is linear, then

$$
T(v) = T(a_1v_1 + \cdots + a_nv_n) = a_1T(v_1) + \cdots + a_nT(v_n).
$$

Since the coordinate vector map is linear, then

$$
[T(v)]_{\mathcal{C}} = [a_1 T(v_1) + \cdots + a_n T(v_n)]_{\mathcal{C}} = a_1 [T(v_1)]_{\mathcal{C}} + \cdots + a_n [T(v_n)]_{\mathcal{C}}
$$
  
= 
$$
\begin{pmatrix} | & & | \\ [T(v_1)]_{\mathcal{C}} & \cdots & [T(v_n)]_{\mathcal{C}} \\ | & & | \end{pmatrix} \begin{pmatrix} a_1 \\ \vdots \\ a_n \end{pmatrix} = c [T]_{\mathcal{B}} [v]_{\mathcal{B}}.
$$

The equality  $[T(v)]_C = c[T]_B[v]_B$  can be stated by saying the following diagram "commutes."

$$
V \xrightarrow{T} W
$$
  
\n
$$
\varphi_B \downarrow \qquad \qquad \downarrow \varphi_C
$$
  
\n
$$
\mathbb{F}^n \xrightarrow{L_{\mathcal{C}}[T]_B} \mathbb{F}^m
$$

[Draw image of *v* traveling both directions.]

**Proposition 30.** *Suppose V and W are finite-dimensional and*  $T \in \mathcal{L}(V, W)$ *. Then the rank of T* (*i.e.*,  $dim(img(T))$  *is equal to the (column) rank of* [*T*].

*Proof.* Exercise. □

II.3.3. *Change of basis.* Q: How does the natrix  $c[T]_B$  change if we change the bases  $B$  and  $\mathcal{C}$ ?

**Definition 31.** Let  $n \in \mathbb{Z}_{\geq 0}$ . The  $n \times n$  *identity matrix I* is the  $n \times n$  matrix with 1s on the diagonal and 0s elsewhere:

$$
I = \begin{pmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{pmatrix}.
$$

□

**Remark 32.** We use *I* for both the identity operator and the identity matrix. With respect to *any* basis, the matrix of the identity operator  $I_V$  is  $I$ .

**Definition 33.** An  $n \times n$  matrix *A* is *invertible* if there is a  $n \times n$  matrix *B* such that  $AB =$  $BA = I$ . We call *B* the *inverse* of *A* and denote it  $A^{-1}$ .

**Lemma 34.** *The inverse of a matrix is unique.*

*Proof.* Same as for linear maps. □

**Theorem 35.** *Let U*, *V, and W be vector spaces with bases* B, C*, and* D*, respectively. Given*  $T \in \mathcal{L}(U, V)$  and  $S \in \mathcal{L}(V, W)$ , then

$$
D[ST]_{\mathcal{B}} = D[S]_{\mathcal{C}} C[T]_{\mathcal{B}}.
$$

*Proof.* Follows by the definition of matrix multiplication. □

**Corollary 36** (Change of basis matrix)**.** *Suppose* B *and* C *are both bases for V. Then*

$$
B[I]_{\mathcal{C}} = c[I]_{\mathcal{B}}^{-1}
$$

*Proof.*

$$
I = g[I]_{\mathcal{B}} = g[I]_{\mathcal{C}} c[I]_{\mathcal{B}}.
$$

.

.

**Proposition 37** (Change of basis formula)**.** *Suppose* B *and* C *are both bases of V. Given*  $T \in \mathcal{L}(V)$ , let  $A := [T]_{\mathcal{B}}$ ,  $B := [T]_{\mathcal{C}}$ , and  $C = \mathcal{B}[I]_{\mathcal{C}}$ . Then

$$
A = CBC^{-1}
$$

*Proof.*

$$
B[T]B = B[I]c c[T]c c[I]B = (c[I]B)^{-1}c[T]c c[I]B.
$$

**Definition 38.** Two  $n \times n$  matrices A and B are *similar* or *conjugate* if there is an invertible matrix *P* such that  $B = PAP^{-1}$ .

□