## 1. INVERTING LINEAR TRANSFORMATIONS AND MATRICES

Let  $id_X : X \to X$  denote the 'identify function' on a set X, given by  $id_X(x) = x$  for all elements  $x \in X$ .

**Definition 1.1.** A linear transformation  $T : \mathbf{R}^n \to \mathbf{R}^m$  is called invertible if there exists another linear transformation  $S : \mathbf{R}^m \to \mathbf{R}^n$  such that  $T \circ S = \mathrm{id}_{\mathbf{R}^m}$  and  $S \circ T = \mathrm{id}_{\mathbf{R}^n}$ . We call S the inverse of T, and we write  $T^{-1} = S$ .

**Theorem 1.2.** Let  $T : \mathbf{R}^n \to \mathbf{R}^m$  be a linear transformation with standard matrix  $A \in \mathcal{M}_{m \times n}$ . Then the following are equivalent

- (1) T is invertible.
- (2) There is a matrix  $B \in \mathcal{M}_{n \times m}$  such that  $AB = I_{m \times m}$  and  $BA = I_{n \times n}$ .
- (3) The equation  $A\mathbf{x} = \mathbf{b}$  has a unique solution for every  $\mathbf{b} \in \mathbf{R}^m$ .
- (4) A is square and non-singular.

*Proof.* My strategy will be to show that (1)  $\iff$  (2) and (3)  $\iff$  (4) and then finally (2)  $\iff$  (3).

That (1) and (2) are equivalent is more or less immediate: i.e. if  $S : \mathbf{R}^m \to \mathbf{R}^n$  if some other linear transformation and B be its matrix. From class we know that the matrix for  $T \circ S$ is AB and the matrix for  $\mathrm{id}_{\mathbf{R}^m}$  is  $I_{m \times m}$ . Hence  $T \circ S = \mathrm{id}_{\mathbf{R}^m}$  is equivalent to  $AB = I_{m \times m}$ . Similarly,  $S \circ T = \mathrm{id}_{\mathbf{R}^n}$  is the same as  $BA = I_{n \times n}$ .

To see that (3) and (4) are equivalent, note first that if (3) holds, then A must be row equivalent to a matrix  $\tilde{A}$  in reduced echelon form with a pivot in every column (because  $A\mathbf{x} = \mathbf{b}$  has at most one solution) and a pivot in every row (because  $A\mathbf{x} = \mathbf{b}$  has at *least* one solution, no matter what  $\mathbf{b}$  is). Since there is at most one pivot in each row and column, it follows that the number of rows and columns of  $\tilde{A}$  are the same. That is,  $\tilde{A}$  and therefore also A are square matrices. From here, the equivalence between (3) and (4) is part of Proposition 4.1.6 in Shifrin (which I stated in class, too).

To see that (2)  $\implies$  (3), let  $B \in \mathcal{M}_{n \times m}$  be the matrix in (2) and  $\mathbf{b} \in \mathbf{R}^m$  be any given vector. If  $\mathbf{x} \in \mathbf{R}^m$  solves  $A\mathbf{x} = \mathbf{b}$ , then

$$B\mathbf{b} = B(A\mathbf{x}) = (BA)\mathbf{x} = I\mathbf{x} = \mathbf{x}.$$

That is,  $\mathbf{x} = B\mathbf{b}$  is the only possible solution of  $A\mathbf{x} = \mathbf{b}$ . On the other hand, I can plug back in to check that it really is a solution:

$$A\mathbf{x} = A(B\mathbf{b}) = (AB)\mathbf{b} = I\mathbf{b} = \mathbf{b}.$$

In short  $A\mathbf{x} = \mathbf{b}$  has the unique solution  $\mathbf{x} := B\mathbf{b}$ , i.e. (3) holds.

To see that (3)  $\implies$  (2), recall from above that (3) implies that m = n, so that  $A \in \mathcal{M}_{n \times n}$ . I let  $\mathbf{e}_j \in \mathbf{R}^n$  be the *j*th standard basis vector and apply (3) to get a vector  $\mathbf{b}_j \in \mathbf{R}^n$  satisfying  $A\mathbf{b}_j = \mathbf{e}_j$ . Then I define  $B := \begin{bmatrix} \mathbf{b}_1 & \dots & \mathbf{b}_n \end{bmatrix} \in \mathcal{M}_{n \times n}$ . It follows that

$$AB = \begin{bmatrix} A\mathbf{b}_1 & \dots & A\mathbf{b}_n \end{bmatrix} = \begin{bmatrix} \mathbf{e}_1 & \dots & \mathbf{e}_n \end{bmatrix} = I.$$

That BA = I, too, follows from the lemma below.

Given the last item in this theorem, we can restrict out discussion of invertibility to linear transformations with the same source and target.

**Definition 1.3.** A square matrix  $A \in \mathcal{M}_{n \times n}$  is invertible if and only if there exists  $B \in \mathcal{M}_{n \times n}$  such that AB = BA = I. We then call B the inverse of A and write  $A^{-1} := B$ .

**Lemma 1.4.** Given two square matrices  $A, B \in \mathcal{M}_{n \times n}$ , we have AB = I if and only if BA = I.

I follow the argument given in Shifrin, which is rather clever.

Proof. Since AB = I and I is a non-singular square matrix, it follows from our homework problem 4.2.17b that A and B are both non-singular. Hence by Proposition 4.1.6 again the linear system  $B\mathbf{x} = \mathbf{b}$  has a solutions for any  $\mathbf{b} \in \mathbf{R}^n$ . In particular, for each standard basis vector  $\mathbf{e}_j \in \mathbf{R}^n$  there is a vectors  $\mathbf{c}_j \in \mathbf{R}^n$  such that  $B\mathbf{c}_j = \mathbf{e}_j$ . And as in the proof of the Theorem, if I set  $C = [\mathbf{c}_1 \dots \mathbf{c}_n]$ , then it follows that BC = I. This allows me to compute the product ABC in two different ways. On the one hand ABC = A(BC) = AI = A. On the other hand ABC = (AB)C = IC = C. Comparing the answers, I see that A = C. So I = BC = BA, which is what I aimed to show.  $\Box$