# Linear Algebra with Application (LAWA 2021)Lecture 3



Francesco Dolce francesco.dolce@fjfi.cvut.cz

March 3rd, 2021

As in the previous lecture, let us consider matrices over the field  $\mathbb{R}$ .

## 1 Transposition

Let  $A = (a_{ij}) \in \mathcal{M}_{m,n}(\mathbb{R})$  be a matrix. The *transpose* of A is the matrix  $A^T \in \mathcal{M}_{n,m}(\mathbb{R})$  defined as

 $A^T = (b_{ij})$  with  $b_{ij} = a_{ji}$  for all i, j.

A matrix A is called *symmetric* if  $A^T = A$ .

#### Example 1 Let

$$A = \begin{pmatrix} 3 & 2 \\ 1 & 0 \end{pmatrix}, \quad B = \begin{pmatrix} 1 & 0 & -2 \\ 5 & -1 & 0 \end{pmatrix} \text{ and } C = \begin{pmatrix} 2 & 1 \\ 1 & 0 \end{pmatrix}.$$

Then we have

$$A^{T} = \begin{pmatrix} 3 & 1 \\ 2 & 0 \end{pmatrix}, \quad B^{T} = \begin{pmatrix} 1 & 5 \\ 0 & -1 \\ -2 & 0 \end{pmatrix} \text{ and } C^{T} = \begin{pmatrix} 2 & 1 \\ 1 & 0 \end{pmatrix}$$

Moreover, the matrix C is symmetric, while A and B are not.

**Proposition 2** Let A, B be two matrices over  $\mathbb{R}$  having the right size and let  $\lambda \in \mathbb{R}$ . Then

1. If A is symmetric then m = n; 2.  $(A^T)^T = A$ ; 3.  $(\lambda A)^T = \lambda A^T$ ; 4.  $(A + B)^T = A^T + B^T$ ; 5.  $(AB)^T = B^T A^T$ .

Proof. (Exercise)

## 2 Matrix inverse

Let us consider a square matrix  $A \in \mathcal{M}_{n,n}(\mathbb{R})$ . A matrix *B* is called an *inverse* of *A* if AB = I and BA = I. Note that if such a matrix *B* exists, then it has the same size as *A*.

A square matrix having an inverse is called an *invertible matrix*.

**Example 3** Let us consider the matrix  $A = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix} \in \mathcal{M}_{2,2}(\mathbb{R})$ . The matrix A is invertible and the matrix  $B = \begin{pmatrix} -1 & 1 \\ 1 & 0 \end{pmatrix}$  one of its inverses. Indeed,

$$AB = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} -1 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = I$$

and

$$BA = \begin{pmatrix} -1 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = I.$$

Similarly to what we have done in Lecture 1 (Proposition 9) we can prove that the inverse of a matrix, when it exists, is unique.

**Proposition 4** Let  $A \in \mathcal{M}_{n,n}(\mathbb{R})$  an invertible matrix. Then its inverse is unique.

*Proof.* Let  $B, C \in \mathcal{M}_{n,n}(\mathbb{R})$  and let us suppose that both matrices are inverses of A. Then

$$B = BI = B(AC) = (BA)C = IC = C.$$

Whenever a matrix A is invertible we denote by  $A^{-1}$  its unique inverse. Note that not all matrices have inverses. **Example 5** Let us conside the matrix  $A = \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}$ . This matrix is not invertible. Indeed, if we suppose by contradiction that there exists a matrix  $B = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  such that AB = I, then we would have

$$\begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} a & b \\ a & b \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}.$$

which implies at the same time that a = 1 and a = 0, a contradiction.

Also, non-square matrices do not have inverses and the zero matrix  $O_n$  is not invertible neither (Exercise).

Let us now give some properties about inverses.

**Theorem 6** The following properties hold for square matrices.

- 1. The identity matrix I is invertible and its inverse is I itself.
- 2. If A is invertible, then so is  $A^{-1}$ , and  $(A^{-1})^{-1} = A$ .
- 3. If A and B are invertible, then so is AB, and  $(AB)^{-1} = B^{-1}A^{-1}$ .
- 4. Let  $A_1, A_2, \ldots A_k$  be invertible matrices, then their product  $A_1 A_2 \cdots A_k$  is also invertible, and  $(A_1 A_2 \cdots A_k)^{-1} = A_k^{-1} A_{k-1}^{-1} \cdots A_1^{-1}$
- 5. If A is invertible, then for all  $k \ge 1$  the matrix  $A^k$  is invertible as well and  $(A^k)^{-1} = (A^{-1})^k$ .
- 6. If A is invertible, then so is  $A^T$ , and  $(A^T)^{-1} = (A^{-1})^T$ .
- 7. If  $A^T$  is invertible, then so is A and  $A^{-1} = ((A^T)^{-1})^T$ .
- 8. If A is invertible and  $\lambda \neq 0$  is a real number, then  $\lambda A$  is also invertible and  $(\lambda A)^{-1} = \frac{1}{\lambda} A^{-1}$ .

#### Proof.

- 1. This easily follows from the fact that II = I.
- 2. The second item also follows from the fact that  $A^{-1}A = I$  and  $AA^{-1} = I$ .
- 3. It holds because

$$(AB)(B^{-1}A^{-1}) = A(BB^{-1})A^{-1} = AIA^{-1} = AA^{-1} = I$$

and

$$(B^{-1}A^{-1})(AB) = B^{-1}(A^{-1}A)B = B^{-1}IB = B^{-1}B = I.$$

4. Let us prove it by induction on k. The case k = 1 is trivial and the case k = 2 follows from point 3. So, let us suppose that the property holds for k - 1, that is that  $A_1 A_2 \cdots A_{k-1}$  is invertible and that its inverse is  $A_{k-1}^{-1} \cdots A_2^{-1} A_1^{-1}$ . Then

$$(A_1A_2\cdots A_{k-1}A_k) = (A_1A_2\cdots A_{k-1})A_k$$

is a product of two invertible matrices and, by the previous point, is invertible itself. Moreover, its inverse is exactly

$$A_k^{-1}(A_1A_2\cdots A_{k-1})^{-1} = A_k^{-1}A_{k-1}^{-1}\cdots A_2^{-1}A_1^{-1}.$$

- 5. This easily follows from the previous item by choosing  $A_i = A$  for all  $1 \le i \le k$ .
- 6. Using the last point of Proposition 2 we have

$$A^{T}(A^{-1})^{T} = (A^{-1}A)^{T} = I^{T} = I$$

and

$$(A^{-1})^T A^T = (AA^{-1})^T = I^T = I.$$

7. Using both the fact that  $(A^T)^{-1} = (A^{-1})^T$ , proved in the previous point, and that  $(A^T)^T = A$ , proved in Proposition 2, we have

$$A((A^T)^{-1})^T = A((A^{-1})^T)^T = AA^{-1} = I$$

and

$$((A^T)^{-1})^T A = ((A^{-1})^T)^T A = A^{-1}A = I$$

8. To prove the last point we use Proposition 17 in Lecture 2 and show that

$$(\lambda A)\left(\frac{1}{\lambda}A^{-1}\right) = \left(\lambda\frac{1}{\lambda}\right)\left(AA^{-1}\right) = 1I = I$$

and

$$\left(\frac{1}{\lambda}A^{-1}\right)(\lambda A) = \left(\frac{1}{\lambda}\lambda\right)(A^{-1}A) = 1I = I.$$

**Corollary 7** Let  $A, B \in \mathcal{M}_{n,n}(\mathbb{R})$ . If A and AB are both invertible, then B is also invertible.

Proof. (Exercise)

-

### 3 Diagonal and triangular matrices

A square matrix  $A = (a_{ij}) \in \mathcal{M}_{n,n}(\mathbb{R})$  is called *diagonal* if every entry not in the main diagonal is 0, that is if for every i, j with  $1 \leq i, j \leq n$  and  $i \neq j$  one has  $(a_{ij}) = 0$ .

Example 8 The matrices

$$A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & -5 \end{pmatrix}, \quad B = \begin{pmatrix} -1 & 0 & 0 & 0 \\ 0 & 4 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}, \quad C = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

are diagonal.

When a matrix  $A = (a_{ij})$  is diagonal, we can also denote it simply as  $A = \text{diag}(a_{11}, a_{22}, \ldots, a_{nn})$ .

Example 9 Let us consider again the three matrices of Example 8. We have

 $A = \operatorname{diag}(1, 2, -5), \quad B = \operatorname{diag}(-1, 4, 0, 1) \quad \text{and} \quad C = \operatorname{diag}(0, 0).$ 

**Proposition 10** Let  $A, B \in \mathcal{M}_{n,n}(\mathbb{R})$  be two diagonal matrices. Then

- 1. A + B is diagonal;
- 2. AB is diagonal.

Proof. (Exercise)

A square matrix  $A = (a_{ij})$  is called *upper triangular* if every entry below the main diagonal is zero, that if for every i, j with i > j one has  $a_{ij} = 0$ . An upper triangular matrix is called *strictly upper triangular* if the entries on the main diagonal are zero as well.

In a symmetric way we define *lower triangular* and *strictly lower triangular* matrices.

**Example 11** Let us consider the four matrices

$$A = \begin{pmatrix} 6 & 9 & 1 \\ 0 & 0 & 2 \\ 0 & 0 & 1 \end{pmatrix}, \quad B = \begin{pmatrix} 0 & 2 & 2 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}, \quad C = \begin{pmatrix} 3 & 0 & 0 \\ 0 & 0 & 0 \\ 2 & 1 & 0 \end{pmatrix} \text{ and } D = \begin{pmatrix} 0 & 0 & 0 \\ 5 & 0 & 0 \\ 7 & 6 & 0 \end{pmatrix}$$

The matrices A and B are upper triangular. Moreover the matrix B is strictly upper triangular. Similarly C and D are lower triangular with D being strictly lower triangular.

**Proposition 12** Let  $A, B \in \mathcal{M}_{n,n}(\mathbb{R})$  be two upper triangular matrices. Then

1. A + B is upper triangular.

2. AB is upper triangular.

Proof. (Exercise)

A similar result also holds by replacing the condition "upper triangular" with "strictly upper triangular", "lower triangular" or "strictly lower triangular".