# First Read on Linear Algebra, Lecture 4

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First Version: September 22, 2025 Last Updated: September 22, 2025

## 4 Matrices

#### 4.1 Basic Definitions

**Definition 4.1.1.** Let m nand n be nonnegative integers and  $\mathbb{F}$  be a field. A m-byn matrix A is a rectangular array of elements of  $\mathbb{F}$  with m rows and n columns, written

$$A = \begin{pmatrix} A_{1,1} & \cdots & A_{1,n} \\ \vdots & & \vdots \\ A_{m,1} & \cdots & A_{m,n} \end{pmatrix}$$

where we write  $A_{i,j}$  to denote the entry in row i, column j of A. When it is not clear, we write  $A = (a_{i,j})$  to represent that A is a matrix with entries given by  $a_{i,j}$ . The set of all  $m \times n$  matrices given by entries in  $\mathbb{F}$  is given by  $\mathcal{M}_{m \times n}(\mathbb{F})$ .

**Definition 4.1.2.** Given  $A, B \in \mathcal{M}_{m \times n}(\mathbb{F})$ , we define A + B to be the matrix whose entries are given by  $(A + B)_{i,j} = A_{i,j} + B_{i,j}$ .

**Proposition 4.1.3.** Matrices addition is commutative and associative. That is, given matrices L, M, N where addition is well-defined, we have

$$M + N = N + M$$

and

$$L + (M+N) = (L+M) + N$$

*Proof.* This is a consequence of the fact that addition over a field is commutative and associative.  $\Box$ 

**Remark 4.1.4.** The  $m \times n$  **zero matrix** is the matrix with m rows and n columns with entries all 0. We write 0 to denote this matrix or  $0_{m,n}$  when the dimension is not clear. Then, the zero matrix is the additive identity on matrices.

**Definition 4.1.5.** Given  $A \in \mathcal{M}_{m \times n}(\mathbb{F})$  and  $\lambda \in \mathbb{F}$ , we define  $\lambda A$  to be the matrix whose entries are given by  $(\lambda A)_{i,j} = \lambda A_{i,j}$ .

**Proposition 4.1.6.** Given a matrix A and  $\lambda, \mu \in \mathbb{F}$ , We have the following identities:

$$0A = 0_{m,n} \qquad A + (-A) = 0_{m,n} \qquad 1A = A$$
$$(\lambda + \mu)A = \lambda A + \mu A \qquad \lambda (A + B) = \lambda A + \lambda B \qquad \lambda (\mu A) = (\lambda \mu)A$$

*Proof.* Straightforward by checking what happens to each entry.

**Remark 4.1.7.** The above proposition shows that  $\mathcal{M}_{m,n}(\mathbb{F})$  is a vector space over  $\mathbb{F}$ .

**Definition 4.1.8.** Given  $A \in \mathcal{M}_{m \times k}(\mathbb{F})$  and  $B \in \mathcal{M}_{k \times n}(\mathbb{F})$ , we define AB to be the matrix whose entries are given by  $(AB)_{i,j} = \sum_{\ell=1}^k a_{i,\ell} b_{\ell,j}$ .

**Definition 4.1.9.** The  $n \times n$  identity matrix  $I_n$  is the matrix with entries

$$\delta_{i,j} := \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$$

We write I to denote the identity matrix when the dimension is clear. The entries  $\delta_{i,j}$  is also called the **Kronecker delta**.

**Proposition 4.1.10.** Let A be an  $m \times n$  matrix. Then,

$$A0_{n,p} = 0_{n,p}$$
  $0_{l,m}A = 0_{l,n}$   $AI_n = A$   $I_MA = A$ 

*Proof.* Is a straightforward check by seeing what the entries are.

**Proposition 4.1.11.** Matrix multiplication is associative and distributive. That is, for matrix A, B, C where the operations are defined, we have

$$A(BC) = (AB)C$$

and

$$A(B+C) = AB + AC$$
  $(A+B)C = AC + BC$ 

*Proof.* Is a straightforward check by seeing what the entries are. For associativity we note that finite sums can be swapped.  $\Box$ 

**Remark 4.1.12.** Matrix multiplication is in general not commutative, and cannot deduce from AB = 0 that A = 0 or B = 0. For instance,

$$\begin{pmatrix} 1 & 2 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 1 & -1 \end{pmatrix} = \begin{pmatrix} 3 & -3 \\ -1 & 1 \end{pmatrix} \qquad \begin{pmatrix} 1 & -1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ -1 & 0 \end{pmatrix} = \begin{pmatrix} 2 & 2 \\ 2 & 2 \end{pmatrix}$$

and

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

**Definition 4.1.13.** For a square matrix A, we write  $A^2$  for the product AA and  $A^n$  for

$$\underbrace{AA\cdots A}_{n \ times}$$

where  $A^0 = I$ . Then, given a polynomial  $p(x) = a_k x^k + \cdots + a_0$ , define

$$p(A) = a_k A^k + \dots + a_0 I$$

**Definition 4.1.14.** Let A be a square matrix. We say that B is an **inverse** of A if BA = AB = I, and we write  $A^{-1}$  to represent such inverse. A matrix with an inverse is **invertible** and otherwise is called **singular**.

**Proposition 4.1.15.** We have the following properties about invertible matrices:

1. Matrix inverses are unique

- 2. If A, B are invertible matrices, then AB is invertible with  $(AB)^{-1} = B^{-1}A^{-1}$
- 3. If A is invertible, then so is  $A^{-1}$  with  $(A^{-1})^{-1} = A$

*Proof.* (i) Suppose B and C are inverses for A. Then,

$$C = I_n C = (BA)C = B(AC) = BI_n = B$$

(ii) We have

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

(iii) We note that

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

and so  $(A^{-1})^{-1} = A$  by uniqueness.

**Definition 4.1.16.** Given  $A \in \mathcal{M}_{m \times n}(\mathbb{F})$ , we define the **transpose** of A, written  $A^T$  to be the  $n \times m$  matrix given by  $(A^T)_{i,j} = A_{j,i}$ .

**Proposition 4.1.17.** Given matrices A, B and  $\lambda \in \mathbb{F}$ , we have the following properties (when operations make sense):

- $(A+B)^T = A^T + B^T$  and  $(\lambda A)^T = \lambda A^T$
- $(AB)^T = B^T A^T$
- $\bullet \ (A^T)^T = A$
- A square matrix A is invertible if and only if  $A^T$  is invertible. Then, we have  $(A^T)^1 = (A^{-1})^T$ .

### 4.2 ERO Decomposition

**Definition 4.2.1.** A **elementary row operation** or **ERO** is an operation that is of the following form:

- $S_{i,j}$ : swapping rows i and j
- $M_i(\lambda)$ : multiplies row i by  $\lambda \neq 0$
- For  $i \neq j$ ,  $A_{i,j}(\lambda)$ : adds  $\lambda$  times row i to row j.

**Proposition 4.2.2.** On a  $m \times n$  matrix A, applying EROs is equivalent to premultiplying A by certain matrices, which we give by  $S_{I,J}$ ,  $M_I(\lambda)$ , and  $A_{I,J}(\lambda)$  by:

the 
$$(i, j)$$
th entry of  $S_{I,J} = \begin{cases} 1 & i = j \neq I, J, \\ 1 & i = J, j = I, \\ 1 & i = I, j = J, \\ 0 & otherwise. \end{cases}$ 

the 
$$(i,j)$$
th entry of  $M_I(\lambda) = \begin{cases} 1 & i = j \neq I, \\ \lambda & i = j = I, \\ 0 & otherwise. \end{cases}$ 

the 
$$(i, j)$$
th entry of  $A_{I,J}(\lambda) = \begin{cases} 1 & i = j, \\ \lambda & i = J, \ j = I, \\ 0 & otherwise. \end{cases}$ 

These matrices are known as elementary matrices.

*Proof.* By explicitly calculating what happens to each entry after the premultiplication.

Proposition 4.2.3. Elementary matrices are invertible.

*Proof.* Follows from the fact that

$$(S_{i,j})^{-1} = S_{j,i} = S_{i,j}$$
  $(A_{i,j}(\lambda))^{-1} = A_{i,j}(-\lambda)$   $(M_i(\lambda))^{-1} = M_i(\lambda^{-1})$