

Linear Algebra Organised

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Preface

I hope that this set of notes will provide a comprehensive summary of some of the most fundamental aspects of linear algebra. At the very least, because I wrote these notes principally FOR understanding, they should be a much easier read than Hoffman and Kunze etc, even if they **by design do not offer much in the way of motivation**.

These notes assume a familiarity with high school mathematics and a little bit of group theory, because those concepts can easily be googled or thought through within a few minutes. Aside from that, for theorems and lemmas in general I will offer very brief exposition before explaining the broad structure of the proof (colored in green), before finally painstakingly providing a detailed proof (provided in orange).

I don't really have much else to say so let's just get into it.

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Chapter 1

The Basics

This chapter is responsible for laying the foundations of whatever you are going to read in these notes, that is, the fundamental language with which linear algebra is expressed.

I will be using a **field** in most of my definitions because they cover the full range of real and complex arithmetic.

1.1 Systems of Linear Equations

Among the core objects of study in linear algebra are linear equations and systems thereof, which frequently appear in science and mathematics.

While there is obviously far more to studying linear algebra, of particular interest in this section are the solution(s) to these systems which we will explore in detail.

We first get the pesky definitions out of the way:

Definition: Linear Equation

Where F is any field, $(c_1, c_2, \dots, c_{n+1})$ are constants in F and (x_1, x_2, \dots, x_n) are variables in F , a **linear equation in n variables** is any equation of the form

$$c_1x_1 + c_2x_2 + \dots + c_nx_n = c_{n+1} \quad (1.1)$$

Remark. In academic texts, the left hand side of 1.1 is often called a **linear combination** of the variables (x_1, x_2, \dots, x_n) .

Definition: System of Linear Equations

A **system of m linear equations in n unknowns** is a collection of m linear equations involving n variables. (An example is provided below):

$$\begin{cases} c_{1,1}x_1 + c_{1,2}x_2 + \dots + c_{1,n}x_n = c_{1,n+1} \\ c_{2,1}x_1 + c_{2,2}x_2 + \dots + c_{2,n}x_n = c_{2,n+1} \\ \vdots \\ c_{m,1}x_1 + c_{m,2}x_2 + \dots + c_{m,n}x_n = c_{m,n+1} \end{cases} \quad (1.2)$$

Remark. If $c_{1,n+1} = c_{2,n+1} = \dots = c_{m,n+1} = 0$, our system of linear equations is **homogeneous**.

Definition: Solutions of a System

For any system of linear equations (1.2), if a collection of n **values** (also known as an n-tuple) $D = (d_1, d_2, \dots, d_n)$ all in F are such that the system is satisfied by setting $x_1 = d_1, x_2 = d_2, \dots, x_n = d_n$ is called a **solution of the system of linear equations**.

The collection of all such n-tuples that satisfy the system of equations is also known as a **solution set** of the system. We will mathematically describe solution sets in a later section.

By playing around with the definitions above (perhaps by substituting the constant values for numbers), a few facts about the solutions of systems should become apparent. Unfortunately these same observations were made by Gauss centuries earlier so you will not become a famous mathematician for your endeavours :(.

Theorem 1.1: Gaussian Elimination

If one system of linear equations is manipulated into another by the following actions:

1. Rescaling (Multiplication of both sides of one equation by a non-zero constant.)
2. Row Combination (Addition of a non-zero multiple of equation i to both sides of another equation j)
3. Swapping (Swapping of two equations)

The solution set of the new system of linear equations is the same as the original system, that is, these two systems of linear equations are **equivalent**.

Proof

Let a system of n linear equations be such that it has a solution $D = (d_1, d_2, \dots, d_n) \in S$ where d_1, d_2, \dots, d_n are all in F and S is the solution set of the original system of linear equations. Similarly, let S' be the solution set of the new system of linear equations constructed by multiplying the i^{th} equation in the system by $g \in F \setminus \{0\}$.

1. If D is a solution of the system then it satisfies the i^{th} equation, i.e.

$$\sum_{k=1}^n c_{i,k} d_k = c_{i,n+1}$$

Hence if we multiplied both sides of this equation by a constant $g \in F \setminus \{0\}$, obviously

$$g \sum_{k=1}^n c_{i,k} d_k = g \cdot c_{i,n+1}$$

This implies that $S \subseteq S'$.

Now let $D' = (d'_1, d'_2, \dots, d'_n)$ be a solution of the new system. By multiplying the above equation by $\frac{1}{g}$ in F , we get

$$g \sum_{k=1}^n c_{i,k} d'_k = g \cdot c_{i,n+1} \implies \sum_{k=1}^n c_{i,k} d'_k = c_{i,n+1}$$

so we see that every solution $D' \in S'$ satisfying the new system of equations also satisfies the original system of equations. So $S' \subseteq S \implies S = S'$.

2. If D satisfies the system of equations then it should satisfy

$$g \cdot \sum_{k=1}^n c_{i,k} x_k + \sum_{k=1}^n c_{j,k} x_k = g \cdot c_{i,n+1} + c_{j,n+1} \quad (1.3)$$

Conversely if D satisfies 1.3 then it should satisfy

$$\begin{aligned} g \cdot \sum_{k=1}^n c_{i,k} x_k + \sum_{k=1}^n c_{j,k} x_k - g \cdot \sum_{k=1}^n c_{i,k} x_k &= g \cdot c_{i,n+1} + c_{j,n+1} \\ &\quad - g \cdot c_{i,n+1} \\ \implies \sum_{k=1}^n c_{j,k} x_k &= c_{j,n+1} \end{aligned}$$

3. Nothing fundamentally changes about the system of equations except the order in which the equations are written. If D satisfies all equations in the system, it will still do so after two equations are swapped. ■

Remark. Gaussian Elimination is why we can solve systems of linear equations manually via an 'eliminate-and-substitute' strategy, the solving technique taught in secondary/high schools that you should be familiar with.

Remark. The proof given above elides an explicit statement of the invertibility of each of these operations. This will be clarified in the next section.

1.2 Matrices

Another building block of the mathematical language of linear algebra is the matrix, which among various other functions provides a neat shorthand with which we can describe systems of linear equations.

Definition: Matrix

Generally, where $i, j, m, n \in \mathbb{Z}^+$, an $m \times n$ **matrix** A is an array of *numerical entries* with m rows and n columns. We denote the entry in the i^{th} row and j^{th} column as A_{ij} .

However, an alternative, more formal definition is possible:

(Hoffman and Kunze) An $m \times n$ **matrix** A is a function mapping the ordered pairs (i, j) , $1 \leq i \leq m$, $1 \leq j \leq n$ to numbers in a field F .

We should also note that matrices come equipped with the following operations:

1. Addition:

$$(A + B)_{ij} = A_{ij} + B_{ij}$$

where A and B are any $m \times n$ matrices.

2. Multiplication:

$$(AB)_{ij} = \sum_{k=1}^r A_{ik} B_{kj}$$

Note that A must have r columns while B must have r rows. If A is a $q \times r$ matrix while B is a $r \times s$ matrix, then AB is a $q \times s$ matrix.

3. Transposition:

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

where A and B are any $m \times n$ matrices.

There also exist special cases of matrices important enough to have their own names (though we will only define the relevant ones now):

Definition: Column/Row Vector

A **column vector** is an $n \times 1$ matrix.

A **row vector** is a $1 \times n$ matrix.

Remark. This is what most people think of when they hear/say a vector. Naturally, a vector is more general than that.

Definition: Zero Matrix

A **zero matrix** $0^{m \times n}$ is any matrix whose entries are **all** zero.

Definition: Square Matrix

A **square matrix** is an $n \times n$ matrix.

Definition: Identity Matrix

A **identity matrix** is an $n \times n$ matrix whose entries on the main diagonal are 1 and all other entries are zero.

(Using **Kronecker Delta**) An **identity matrix** $I^{n \times n}$ is such that $I_{ij} = \delta_{ij}$.

Remarks. It is also possible to (informally) define a matrix recursively in terms of smaller matrices. A **block matrix** is any $n \times m$ matrix which can be thought of as comprising of several smaller matrices such that matrices in the same sectioned-off row or column have the same height or width. Usually, a block matrix can be denoted in the manner below:

$$\begin{pmatrix} M_{1,1} & M_{1,2} & \dots & M_{1,r} \\ M_{2,1} & M_{2,2} & \dots & M_{2,r} \\ \vdots & \vdots & \ddots & \vdots \\ M_{s,1} & M_{s,2} & \dots & M_{s,r} \end{pmatrix}$$

where $M_{i,j}$ is an $h_i \times w_j$ matrix s.t. $\sum_{i=1}^{h_i} = s$ and $\sum_{j=1}^{w_j} = r$.

1.3 Elementary Row Operations

Using **Matrix** and **Column/Row Vector**, we can rehash the definition in **System of Linear Equations**.

Definition: System of Linear Equations (Matrix Multiplication Form)

(See **System of Linear Equations**) A **system of m linear equations in n unknowns** can be written as the matrix product

$$AX = Y.$$

where A is an $m \times n$ matrix and X and Y are $m \times 1$ column vectors.

This analogue of **System of Linear Equations** is also equipped with operations that behave similarly to those in **Gaussian Elimination**, except they act on matrices instead of equations. These Elementary Row Operations are defined below:

1. Rescaling (e_r):

$$e_r(A, g, k)_{ij} = \begin{cases} g \cdot A_{ij} & \text{if } i = k \\ A_{ij} & \text{otherwise} \end{cases}$$

where A is any $m \times n$ matrix.

2. Row Combination (e_{rc}):

$$e_{rc}(A, g, r, s)_{ij} = \begin{cases} g \cdot A_{rj} + A_{sj} & \text{if } i = s \\ A_{ij} & \text{otherwise} \end{cases}$$

where A is any $m \times n$ matrix.

3. Swapping (e_s):

$$e_s(A, r, s)_{ij} = \begin{cases} A_{sj} & \text{if } i = r \\ A_{rj} & \text{if } i = s \\ A_{ij} & \text{otherwise} \end{cases}$$

where A is any $m \times n$ matrix.

Of particular note is the invertibility of these operations.

Theorem 1.2: Invertibility of Elementary Row Operations

The **elementary row operations** e_r, e_{rc}, e_s are invertible.

Proof

Sketch

Since they are easy enough to see, we will simply exhibit the inverse functions themselves.

- The inverse of e_r .

Consider the operation

$$e'_r(A)_{ij} = \begin{cases} \frac{1}{g} \cdot A_{ij} & \text{if } i = k \\ A_{ij} & \text{otherwise} \end{cases}$$

Quite clearly,

$$\begin{aligned} e'_r(e_r(A), g, k)_{ij} &= \begin{cases} \frac{1}{g} \cdot g \cdot A_{ij} & \text{if } i = k \\ A_{ij} & \text{otherwise} \end{cases} \\ &= A_{ij} \end{aligned}$$

- The inverse of e_{rc} .

Consider the operation

$$e'_{rc}(A)_{ij} = \begin{cases} -g \cdot A_{rj} + A_{sj} & \text{if } i = s \\ A_{ij} & \text{otherwise} \end{cases}$$

Quite clearly,

$$\begin{aligned} e'_{rc}(e_{rc}(A))_{ij} &= \begin{cases} -g \cdot A_{rj} + g \cdot A_{rj} + A_{sj} & \text{if } i = s \\ A_{ij} & \text{otherwise} \end{cases} \\ &= A_{ij} \end{aligned}$$

Lastly, we note that e_s is its own inverse. Let our matrix A be

represented by $\begin{pmatrix} \overline{a_1} \\ \overline{a_2} \\ \vdots \\ \overline{a_m} \end{pmatrix}$ where a_i is a row vector denoting the i^{th} row of A.

Thus, suppose WLOG that $r < s$.

$$\begin{aligned}
 e_s(e_s(A)) &= e_s \left(e_s \left(\left(\begin{array}{c} \frac{a_1}{} \\ \frac{a_2}{} \\ \vdots \\ \frac{a_m}{} \end{array} \right) \right) \right) \\
 &= e_s \left(\left(\begin{array}{c} \frac{a_1}{} \\ \frac{a_2}{} \\ \vdots \\ \frac{a_s}{} \\ \vdots \\ \frac{a_r}{} \\ \vdots \\ \frac{a_m}{} \end{array} \right) \right) \\
 &= \left(\begin{array}{c} \frac{a_1}{} \\ \frac{a_2}{} \\ \vdots \\ \frac{a_r}{} \\ \vdots \\ \frac{a_s}{} \\ \vdots \\ \frac{a_m}{} \end{array} \right) \\
 &= A
 \end{aligned}$$

Due to the **Invertibility of Elementary Row Operations**, if a matrix B may be derived from the matrix A via a series of elementary row operations, the converse is also true. This gives us the following definition.

Definition: Row Equivalence

If a matrix B may be derived from the matrix A via a series of elementary row operations, the matrices A and B are said to be **row equivalent**.