

3 Lecture 3: Linear Algebra I

Most, if not all, algebra learned throughout K-12 educations deals with *scalar* algebra. Each variable only represents a single number. But, a variable can represent more than one element. Instead:

- *Scalar*: one element, x
- *Vector*: n elements
 - \mathbf{x} includes the set of scalars $x_1, x_2, \dots, x_i, \dots, x_n$
- *Matrix*: $n \times m$ elements
 - \mathbf{X} include the set of scalars $x_{11}, \dots, x_{1m}, x_{21}, \dots, x_{2m}, \dots, x_{n1}, \dots, x_{nm}$

There are considerable gains to be made through this notation. Beyond a more efficient notation, we're able to easily carry out all sorts of useful manipulations. We'll work up to these benefits in this lecture through various concepts.

3.1 Systems of equations

- *System of equations*: two or more equations with the same variables
 - To find a unique solution, we need as many equations as variables. E.g.,

$$6x_1 - 3x_2 + 4x_3 = -13$$

$$6x_1 = -13 + 3x_2 - 4x_3$$

$$x_1 = -\frac{13}{6} + \frac{1}{2}x_2 - \frac{2}{3}x_3$$

- We can use this single equation and solution to create a series of solutions, i.e. if $x_2 = 2$ and $x_3 = 3$, then $x_1 = -\frac{13}{6}$ and so on.
- Consider the following system of (linear) equations:

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \dots + a_{2n}x_n = b_2$$

$$\vdots$$

$$a_{m1}x_1 + a_{m2}x_2 + a_{m3}x_3 + \dots + a_{mn}x_n = b_m$$

Drawing on yesterday's lecture, we can also write this all as: $\sum_{i=1}^m \sum_{j=1}^n a_{ij}x_j = b_i$ Or, bringing matrix notation in, as: $\mathbf{Ax} = \mathbf{b}$, where \mathbf{A} is an $n \times m$ matrix.

3.1.1 Solving for a system of equation

Consider the following system of equations. Throughout r references row number. By setting all variables but one per equation, we are solving by *elimination*. We'll later introduce Gauss-Jordan elimination, which is the same process, but with an augmented matrix.

- First, the equations:

$$\begin{aligned} 1x_1 + 1x_2 &= 100,000 \\ 0.05x_1 + 0.09x_2 &= 7,800 \end{aligned}$$

- Add $-0.05r_1$ to r_2

$$\begin{aligned} 1x_1 + 1x_2 &= 100,000 \\ 0x_1 + 0.04x_2 &= 2,800 \end{aligned}$$

- $25r_2$

$$\begin{aligned} 1x_1 + 1x_2 &= 100,000 \\ 0x_1 + 1x_2 &= 70,000 \end{aligned}$$

- Subtract r_2 from r_1

$$\begin{aligned} 1x_1 + 0x_2 &= 30,000 \\ 0x_1 + 1x_2 &= 70,000 \end{aligned}$$

- This gives us: $x_1 = 30,000$ and $x_2 = 70,000$

Now, let's try a lengthier example. Again, we're not substituting equations into each variable. We're eliminating all variables but one from each equation.

- First, the equations:

$$\begin{aligned} 1x_1 + 2x_2 + 3x_3 &= 6 \\ 2x_1 - 3x_2 + 2x_3 &= 14 \\ 3x_1 + 1x_2 - 1x_3 &= -2 \end{aligned}$$

- Subtract $-2r_1$ from r_2 and subtract $-3r_1$ from r_3

$$\begin{aligned} 1x_1 + 2x_2 + 3x_3 &= 6 \\ 0x_1 - 7x_2 - 4x_3 &= 2 \\ 0x_1 - 5x_2 - 10x_3 &= -20 \end{aligned}$$

- Multiply r_3 by $-\frac{1}{5}$ and flip r_2 and r_3

$$\begin{aligned} 1x_1 + 2x_2 + 3x_3 &= 6 \\ 0x_1 + 1x_2 + 2x_3 &= 4 \\ 0x_1 - 7x_2 - 4x_3 &= 2 \end{aligned}$$

- Subtract $2r_2$ from r_1 and add $7r_2$ to r_3

$$\begin{aligned} 1x_1 + 0x_2 - 1x_3 &= -2 \\ 0x_1 + 1x_2 + 2x_3 &= 4 \\ 0x_1 + 0x_2 + 10x_3 &= 30 \end{aligned}$$

- Divide r_3 by 10

$$\begin{aligned} 1x_1 + 0x_2 - 1x_3 &= -2 \\ 0x_1 + 1x_2 + 2x_3 &= 4 \\ 0x_1 + 0x_2 + 1x_3 &= 3 \end{aligned}$$

- Add r_3 to r_1 and subtract $2r_3$ from r_2

$$\begin{aligned} 1x_1 + 0x_2 - 0x_3 &= 1 \\ 0x_1 + 1x_2 + 0x_3 &= -2 \\ 0x_1 + 0x_2 + 1x_3 &= 3 \end{aligned}$$

- $x_1 = 1, x_2 = -2, x_3 = 3$

Writing all of the x 's gets tedious though. So let's introduce vectors.

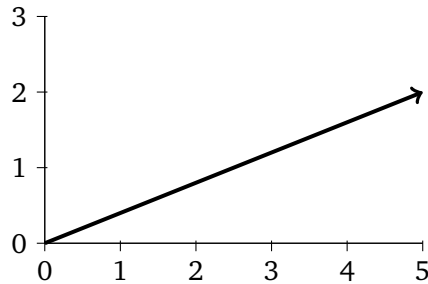
3.2 Vectors

- Examples:

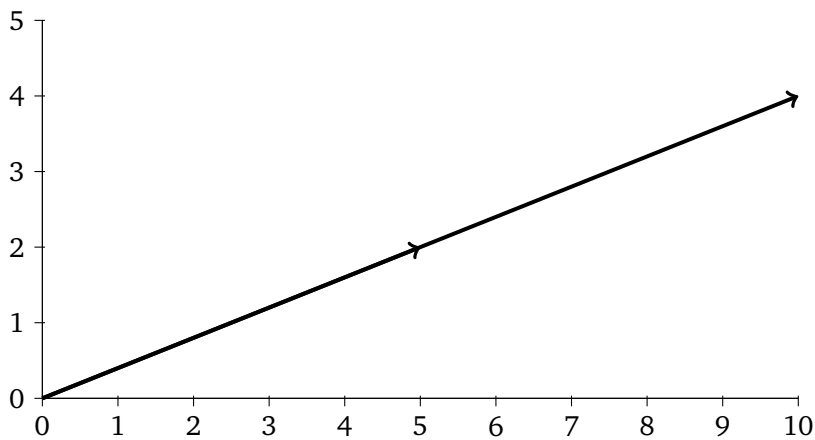
$$\circ \mathbf{x} = [x_1, x_2, x_3, \dots, x_n]$$

○ $\mathbf{x} = (5, 2)$

- ◆ If, like in this case, there are two dimensions (number of components), then we can visually understand the vector as:



○ If we multiply by a scalar: $a = 2$, $a\mathbf{x} = [10, 4]$



3.2.1 Vector length

- Also called the *norm*, length is not the same as *dimensions*. The formula is:

$$\|\mathbf{x}\| = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

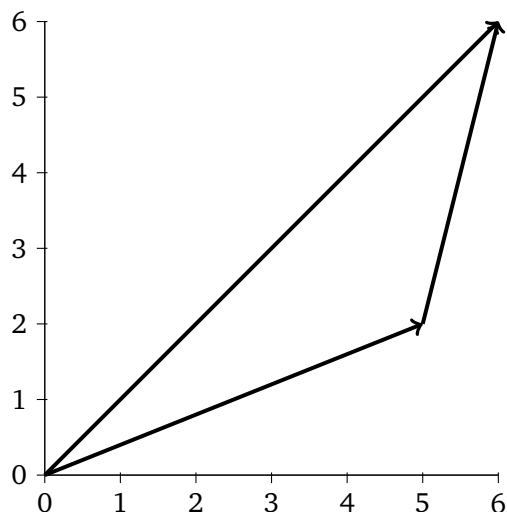
- Considering the visuals above, in two dimensions we are just using the Pythagorean theorem. But the formula is extendable to k-dimensions. For $\mathbf{x} = (5, 2)$, $\|\mathbf{x}\| = \sqrt{25 + 4} = \sqrt{29}$.

- Another example

○ $\mathbf{x} = (5, 2), \mathbf{y} = (1, 4)$

- ◆ $\mathbf{x} + \mathbf{y} = (6, 6)$

- ◆ $\|\mathbf{x} + \mathbf{y}\| = \sqrt{36 + 36} = \sqrt{72}$



3.2.2 Vector multiplication

- If c is a *scalar* and we multiply $a(x_1, x_2, \dots, x_n)$, then we get $(ax_1, ax_2, \dots, ax_n)$. Dividing by a scalar works the same way.
- But what about multiplying one vector by another vector? We use the **dot product**: $\mathbf{a} \cdot \mathbf{b}$. Another name for this operation is the **inner product**.²

- If \mathbf{a} and \mathbf{b} are both n -dimensional, then $\mathbf{a} \cdot \mathbf{b} = a_1b_1 + a_2b_2 + \dots + a_nb_n = \sum_{i=1}^n a_ib_i$

$$\text{Ex: } \begin{bmatrix} 5 \\ 2 \end{bmatrix} \cdot [1 \ 4] = (5 \cdot 1) + (2 \cdot 4) = 5 + 8 = 13$$

- Note: the result of the dot product of vectors is a *scalar*.
- The **outer product** of two vectors instead produces a matrix:

$$\begin{bmatrix} a_1 \\ a_2 \end{bmatrix} \cdot \begin{bmatrix} b_1 & b_2 \end{bmatrix} = \begin{bmatrix} a_1b_1 & a_1b_2 \\ a_2b_1 & a_2b_2 \end{bmatrix}$$

- The dimensions of this matrix are the two outer dimensions of the vectors multiplied together:

$$\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}_{3 \times 1} [1 \ 2 \ 3]_{1 \times 3} = \begin{bmatrix} 3 & 6 & 9 \\ 2 & 4 & 6 \\ 1 & 2 & 3 \end{bmatrix}_{3 \times 3}$$

²For a nice review of vector manipulation, see <https://people.cs.clemson.edu/~dhouse/courses/401/notes/vectors.pdf>

- But the inner dimensions must match up. See 1 and 1 above. If the first matrix's number of columns is not equal to the second matrix's number of rows, then cannot multiply.

3.3 Matrices

- A **matrix** is a rectangular table of numbers or variables arranged in a specific order in rows and columns. We express dimensions by rows, n , and columns, m . The dimensions of a matrix $A_{n \times m}$ are pronounced 'n by m'.

$$X = \begin{bmatrix} x_{11} & x_{12} & x_{13} \\ x_{21} & x_{22} & x_{23} \\ x_{31} & x_{32} & x_{33} \end{bmatrix} \in \mathbb{R}^{nm} = \mathbb{R}^{3 \times 3}$$

- If $m = n$, then the matrix is square.
- **Types of matrices:**

$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} = \text{zero matrix} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix} = \text{diagonal matrix} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \text{identity matrix}$$

3.4 Matrix operators

- **Addition**
 - Must have the same number of elements

$$\begin{bmatrix} 1 & 2 & 3 \\ 0 & 0 & 1 \\ 1 & 2 & 1 \end{bmatrix} + \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix} = \text{can't do}$$

$$\begin{bmatrix} 1 & 2 \\ 0 & 0 \\ 1 & 2 \end{bmatrix} + \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix} = \begin{bmatrix} 2 & 4 \\ 3 & 4 \\ 6 & 8 \end{bmatrix}$$

- **Transposition**
 - Rotate so that the first column becomes the first row:

$$X = \begin{bmatrix} 1 & 2 \\ 0 & 0 \\ 1 & 2 \end{bmatrix}_{3 \times 2}, \quad X^T = \begin{bmatrix} 1 & 0 & 1 \\ 2 & 0 & 2 \end{bmatrix}_{2 \times 3}$$

- **Multiplication**

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} 7 & 0 & -1 \\ 1 & 3 & 1 \end{bmatrix}$$

- Because \mathbf{A} is 2×2 and \mathbf{B} is 2×3 , we can multiply. But \mathbf{AB}^T is undefined because \mathbf{B} is 3×2 .

$$\begin{aligned} \mathbf{AB} &= \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 7 & 0 & -1 \\ 1 & 3 & 1 \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{A}_{11}\mathbf{B}_{11} + \mathbf{A}_{12}\mathbf{B}_{21} & \mathbf{A}_{11}\mathbf{B}_{12} + \mathbf{A}_{12}\mathbf{B}_{22} & \mathbf{A}_{11}\mathbf{B}_{13} + \mathbf{A}_{12}\mathbf{B}_{23} \\ \mathbf{A}_{21}\mathbf{B}_{11} + \mathbf{A}_{22}\mathbf{B}_{21} & \mathbf{A}_{21}\mathbf{B}_{12} + \mathbf{A}_{22}\mathbf{B}_{22} & \mathbf{A}_{21}\mathbf{B}_{13} + \mathbf{A}_{22}\mathbf{B}_{23} \end{bmatrix} \\ &= \begin{bmatrix} (1 \cdot 7) + (2 \cdot 1) & (1 \cdot 0) + (2 \cdot 3) & (1 \cdot -1) + (2 \cdot 1) \\ (3 \cdot 7) + (4 \cdot 1) & (3 \cdot 0) + (4 \cdot 3) & (3 \cdot -1) + (4 \cdot 1) \end{bmatrix} = \begin{bmatrix} 9 & 6 & 1 \\ 25 & 12 & 1 \end{bmatrix} \end{aligned}$$

- Ex with identity matrix:

$$\begin{aligned} \mathbf{I}_{2 \times 2} \mathbf{X} &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \\ &\text{b/c } \begin{bmatrix} (1 \cdot 1) + (0 \cdot 3) & (0 \cdot 1) + (1 \cdot 2) \\ (0 \cdot 1) + (1 \cdot 3) & (0 \cdot 3) + (1 \cdot 4) \end{bmatrix} \end{aligned}$$