A. Linear Systems. We illustrate the process of solving a linear system:

$$\begin{cases} x + 2y + 3z = 0 \\ -3x - 5y - 8z = 8 \\ 2x + 6y + z = 2 \end{cases}$$

First we convert to an **augmented matrix** with variable coefficients on the left and constants on the right.

Next we apply **Gaussian elimination** to solve.

It involves the following row operations:

swap:

scale:

linearly combine:

The first nonzero entry in each row is called a **pivot**.

**Step 1.** Use row operations to make all entries below the top pivot equal 0.

A **linear system** is a system of equations in which variables appear linearly. In other words: each side of the equation is a linear combination of variables and constants. If you forgot: a linear combination of variables x, y, z for example has form ax + by + cz where a, b, and c are constant.

A **matrix** is a rectangular array of numbers.

This has the effect of **eliminating** x from the bottom equations.

**Step 2.** Use row operations to make all entries below the next highest pivot equal **0**, without upsetting the first column.

**Step 3.** Finish solving using back-substitution.

In other words, solve the bottom equation, then substitute the solution into the next–from–bottom equation, and so on.

B. Free Variables. In the previous example if we had made the adjustment:

$$\begin{cases} x + 2y + 3z = 0 \\ -3x - 5y - 8z = 8 \end{cases} \mapsto \begin{cases} x + 2y + 3z = 0 \\ -3x - 5y - 8z = 8 \\ 2x + 6y + \boxed{z} = \boxed{2} \end{cases}$$

The end of Gaussian elimination would have lead to:

$$\left(\begin{array}{ccc|c}
1 & 2 & 3 & 0 \\
0 & 1 & 1 & 8 \\
0 & 0 & & 0
\end{array}\right)$$

which is **inconsistent** because it has:

On the other hand if we had instead made the adjustment:

$$\begin{cases} x + 2y + 3z = 0 \\ -3x - 5y - 8z = 8 \end{cases} \mapsto \begin{cases} x + 2y + 3z = 0 \\ -3x - 5y - 8z = 8 \\ 2x + 6y + \boxed{z} = \boxed{2} \end{cases}$$

The end of Gaussian elimination would have lead to:

$$\left(\begin{array}{ccc|c}
1 & 2 & 3 & 0 \\
0 & 1 & 1 & 8 \\
0 & 0 & 8
\end{array}\right)$$

The system is **consistent** but:

In this case: assign every variable whose column has no pivot as a **free variable**, and use back—substitution to solve for the others in terms of free variables.

The idea is that the equation 0 = 1 is

unsatisfiable.

It is consistent because there are no unsatisfiable equations at the end of our Gaussian elimination.

A free variable can take on any value, meaning there will be  $\infty$  solutions to the system.

The goal of Gaussian elimination is to use row–operations to convert an augmented matrix to **row–echelon form**. This form is characterized by:

- 1. Each pivot is:
- 2. All rows of all zeroes are:

C. **Matrix Operations.** In general a **matrix** is a rectangular array of numbers.

Of special note are (column) vectors which contain only one column:

For us, a vector will always refer to a column vector. But it in many contexts it is perfectly reasonable to talk about row vectors too.

There are number of linear operations we can execute with matrices.

scalar multiplication: 
$$5\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} =$$

addition: 
$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} + \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} =$$

**transpose**: 
$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^T =$$

zero matrix: 0 =

**Properties.** Let A and B be matrices of the same size. Let k be a scalar.

commutativity of addition: A + B =

distributivity of scalar multiplication: k(A + B) =

zero matrix as additive identity: A + 0 =

A real number is referred to as a **scalar**. In scalar multiplication, each entry is multiplied (scaled) by that scalar.

Only matrices of the same **size** (same number of rows, same number of columns) can be added. We simply add the entries located in the same positions.

Taking the transpose swaps rows and columns.

## D. Matrix Product. We define (matrix)(column vector) so we can convert:

$$\begin{cases} x + 2y + 3z = 0 \\ -3x - 5y - 8z = 8 \end{cases} \mapsto$$
$$2x + 6y + z = 2$$

In other words:

$$\begin{pmatrix} row & 1 \\ \vdots \\ row & m \end{pmatrix} (column vector) =$$

And then we define product (matrix) (matrix) by:

$$\begin{pmatrix} \text{row 1} \\ \vdots \\ \text{row m} \end{pmatrix} \begin{pmatrix} \text{column 1} & \cdots & \text{column n} \end{pmatrix} =$$

or equivalently:

(entry in row i and column j of product) =

For example:

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} =$$

The identity matrix I has 1s along the main diagonal and 0s elsewhere:

I =

and has the property that:

$$IA =$$

$$AI =$$

The expression on the right is referred to as the **matrix form** of the system: Ax = b.

Here ⋅ is the **dot product** which you learned in Calculus III. If you forgot:

$$\langle a_1, \dots, a_n \rangle \cdot \langle b_1, \dots, b_n \rangle$$
  
=  $a_1b_1 + \dots + a_nb_n$ 

The matrix product is not always defined. It requires that we take the dot product of rows of the first matrix with columns of the second. But a dot product can only be taken of vectors with the same number of entries.

The **main diagonal** is the diagonal from the upper–left to the bottom–right.

Here A is a matrix so that the multiplication is defined.

**Example 1.** Consider the matrices.

$$D = \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \text{ and } N = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$$

Compute the following, which model pecularities of the matrix product.

DN =

ND =

The matrix product is **not** generally commutative. That is, it is possible that:

AB

 $N^2 =$ 

The matrix product does **not** have the cancellation property. That is:

AB = 0 and  $A \neq 0$  does not imply:

A matrix N is **nilpotent** if there is a positive integer k so:

Certainly there is no nonzero real number that, when raised to a power, yields 0. Matrices are weird.

Nonetheless, the matrix product does have some reasonable properties.

**Matrix Product Properties.** Let A, B, C be matrices and k be a scalar.

commutativity with scalar multiplication: k(AB) =

distributivity: A(B + C) =

$$(A + B)C =$$

associativity: A(BC) =

multiplication with zero matrix: A0 =