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## Unit I: $Ax = b$ and the Four Subspaces

### Session 1.1: The Geometry of Linear Equations

We have a system of equations:

$$\begin{cases} 2x - y = 0 \\ -x + 2y = 3 \end{cases}$$

#### Row Picture

Line  $2x - y = 0$  and line  $-x + 2y = 0$  intersects at the point  $(1, 2)$ , so  $(1, 2)$  is the solution of the system of equations.

Maybe I should draw a X-Y coordinates here...

#### Column Picture

We rewrite the system of linear equations as a single equation:

$$x \begin{bmatrix} 2 \\ -1 \end{bmatrix} + y \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix}$$

We see  $x$  and  $y$  as scalars of column vectors:  $v_1 = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$  and  $v_2 = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$ , and the sum  $xv_1 + yv_2$  is called a *linear combination* of  $v_1$  and  $v_2$ .

Geometrically, we can find one copy of  $v_1$  added to two copies of  $v_2$  just equals the vector  $\begin{bmatrix} 0 \\ 3 \end{bmatrix}$ . Then the solution should be  $x = 1, y = 2$ .

I will add a figure when time is available >\_>

## Matrix Picture

We rewrite the equations in our example as a compact form,

$$Ax = b,$$

that is

$$\begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix}$$

## Matrix Multiplication

$$\begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 1 \begin{bmatrix} 2 \\ -1 \end{bmatrix} + 2 \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix}$$

A matrix times by a vector is just **a linear combination of the column vectors of the matrix**.

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## Session 1.2: An Overview of Key Ideas

### Vectors

Let us take linear combinations of vectors.

### Matrices

The product of a matrix and a vector is a combination of the columns of the matrix.

### Subspaces

All combinations of column vectors creates a subspace. The subspaces of  $\mathbb{R}^3$  are:

- the origin,
- a line through the origin,
- a plane through the origin,
- all of  $\mathbb{R}^3$ .

## Conclusion

- $A$  is invertible
  - $\Leftrightarrow Ax = b$  has the unique solution  $x$  for each  $b$
  - $\Leftrightarrow Ax = 0$  has no non-zero solution  $x$
  - $\Leftrightarrow$  The columns of  $A$  are *independent*
  - $\Leftrightarrow$  All vectors  $Ax$  cover the whole vector space

Example:  $A = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix}$

- $A$  is not invertible
  - $\Leftrightarrow Ax = b$  has a solution  $x$  only for some of  $b$  in the vector space
  - $\Leftrightarrow Ax = 0$  has non-zero solutions  $x$
  - $\Leftrightarrow$  The columns of  $A$  are *dependent*
  - $\Leftrightarrow$  All vectors  $Ax$  lies in only a subspace of the vector space

Example:  $A = \begin{bmatrix} 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix}$

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## Session 1.3: Elimination with Matrices

### Method of Elimination

We have an example  $Ax = b$ ,

$$A = \begin{bmatrix} 1 & 2 & 1 \\ 3 & 8 & 1 \\ 0 & 4 & 1 \end{bmatrix} \text{ and } b = \begin{bmatrix} 2 \\ 12 \\ 2 \end{bmatrix}.$$

Steps of Elimination:

- Step 1: subtract 3 times row 1 from row 2;
- Step 2: subtract 2 times row 2 from row 3.

$$A = \begin{bmatrix} 1 & 2 & 1 \\ 3 & 8 & 1 \\ 0 & 4 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 1 \\ 0 & 2 & -2 \\ 0 & 4 & 1 \end{bmatrix} \rightarrow U = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 1 & -2 \\ 0 & 0 & 5 \end{bmatrix}$$

$$b = \begin{bmatrix} 2 \\ 12 \\ 2 \end{bmatrix} \rightarrow \cdots \rightarrow \begin{bmatrix} 2 \\ 6 \\ -10 \end{bmatrix}$$

Thus, we can easily solve the systems of equations, 
$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \\ -2 \end{bmatrix}.$$

### Elimination Matrices

The product of a matrix (3x3) and a column vector (3x1) is a column vector (3x1) that is a linear combination of the columns of the matrix.

The product of a row vector (1x3) and a matrix (3x3) is a row vector (1x3) that is a linear combination of the rows of the matrix.

For example,

$$\begin{bmatrix} 1 & 0 & 0 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 1 \\ 3 & 8 & 1 \\ 0 & 4 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 2 & -2 \\ 0 & 4 & 1 \end{bmatrix}.$$

Multiplying on the left by a permutation matrix exchanges the rows of a matrix, while multiplying on the right exchanges the columns. For example,

$$P = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

$P$  is a *permutation matrix* and the first and second rows of the matrix  $PA$  are the second and first rows of the matrix  $A$ .

Note, matrix multiplication is *associative* but *not commutative*.

## Inverses

We have a matrix:

$$E_{21} = \begin{bmatrix} 1 & 0 & 0 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

which subtracts 3 times row 1 from row 2. To “**undo**” this operation we must add 3 times row 1 to row 2 using the inverse matrix:

$$E_{21}^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

In fact,  $E_{21}^{-1}E_{21} = I$ .

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## Session 1.4: Multiplication and Inverse Matrices

### Four and a half ways we see matrix multiplication

We have  $AB = C$ .  $A$  is an  $m \times n$  matrix and  $B$  is an  $n \times p$  matrix, then  $C$  is an  $m \times p$  matrix. We use  $c_{ij}$  to denote the entry in row  $i$  and column  $j$  of matrix  $C$  and the same denotation applies to  $a_{ij}$  and  $b_{ij}$ .

#### 1. Row times column

$$c_{ij} = \sum_{k=1}^n a_{ik}b_{kj}$$

#### 2. Columns

The product of matrix  $A$  and column  $j$  of matrix  $B$  equals column  $j$  of matrix  $C$ . This tells us that the columns of  $C$  are combinations of columns of  $A$ .

$$A \begin{bmatrix} | & | & | \\ \text{column1} & \text{column2} & \text{column3} \\ | & | & | \end{bmatrix} = \begin{bmatrix} | & | & | \\ A(\text{column1}) & A(\text{column2}) & A(\text{column3}) \\ | & | & | \end{bmatrix}$$

### 3.Rows

The product of row  $i$  of matrix  $A$  and matrix  $B$  equals row  $i$  of matrix  $C$ . So the rows of  $C$  are combinations of rows of  $B$ .

$$\begin{bmatrix} \text{---} & \text{row1} & \text{---} \\ \text{---} & \text{row2} & \text{---} \\ \text{---} & \text{row3} & \text{---} \end{bmatrix} B = \begin{bmatrix} \text{---} & (\text{row1})B & \text{---} \\ \text{---} & (\text{row2})B & \text{---} \\ \text{---} & (\text{row3})B & \text{---} \end{bmatrix}$$

### 4. Column times row

$$AB = \sum_{k=1}^n \begin{bmatrix} a_{1k} \\ \vdots \\ a_{mk} \end{bmatrix} [b_{k1} \quad \cdots \quad b_{kp}]$$

note: a typo in the MIT's lecture summary here:  $b_{kp}$ , not  $b_{kn}$ .

### 5. Blocks

$$\begin{bmatrix} A_1 & A_2 \\ A_3 & A_4 \end{bmatrix} \begin{bmatrix} B_1 & B_2 \\ B_3 & B_4 \end{bmatrix} = \begin{bmatrix} A_1B_1 + A_2B_3 & A_1B_2 + A_2B_4 \\ A_3B_1 + A_4B_3 & A_3B_2 + A_4B_4 \end{bmatrix}$$

### Inverses

If  $A$  is *singular* or *not invertible*,

then  $A$  does not have an inverse,

and we can find some non-zero vector  $x$  for which  $Ax = 0$

### Gauss-Jordan Elimination

$$E [ A \mid I ] = [ I \mid E ]$$

If  $EA = I$ , then  $E = A^{-1}$ .

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### Session 1.5: Factorization into $A = LU$

#### Inverse of a product

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

#### Transpose of a product

$$(AB)^T = B^T A^T, \quad (A^T)^{-1} = (A^{-1})^T$$

$$A = LU$$

We can use elimination to convert  $A$  into an upper triangular matrix  $U$ , that is  $EA = U$ , and further we can also convert this to a factorization  $A = LU$  in which  $L = E^{-1}$ .

For example, in a three dimensional case, if  $E_{32}E_{31}E_{21}A = U$  then  $A = E_{21}^{-1}E_{31}^{-1}E_{32}^{-1}U = LU$ . Suppose  $E_{31}$  is the identity matrix and  $E_{32}$  and  $E_{21}$  are as shown below:

$$\begin{array}{ccc} E_{32} & E_{21} & E \\ \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -5 & 1 \end{bmatrix} & \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} & = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 10 & -5 & 1 \end{bmatrix} \end{array}.$$

Here  $L = E^{-1} = E_{21}^{-1}E_{32}^{-1}$ .

$$\begin{array}{ccc}
E_{21}^{-1} & E_{32}^{-1} & L \\
\begin{bmatrix} 1 & 0 & 0 \\ \underline{2} & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} & \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & \underline{5} & 1 \end{bmatrix} & = \begin{bmatrix} 1 & 0 & 0 \\ \underline{2} & 1 & 0 \\ 0 & \underline{5} & 1 \end{bmatrix}
\end{array}$$

Notice the 0 in row three column one of  $L$ , where  $E$  had a 10. The factorization  $A = LU$  is preferable to the statement  $EA = U$  because the combination of row subtractions does not have the effect on  $L$  that it did on  $E$ .

**If there are no row exchanges, the multipliers from the elimination matrices are copied directly into  $L$ .**

### Cost of elimination

If we define a typical operation is to multiply one row and then subtract it from another, then the total number of operations needed to factor  $n \times n$   $A$  into  $LU$  is on the order of  $n^3$ :

$$1^2 + 2^2 + \cdots + (n-1)^2 + n^2 = \sum_{i=1}^n i^2 \approx \int_0^n x \, dx = \frac{1}{3}n^3.$$

While we're factoring  $A$  we're also operating on  $b$ . That costs about  $n^2$  operations, which is hardly worth counting compared to  $1/3n^3$ .

### Row exchanges

- The inverse of any permutation matrix  $P$  is  $P^{-1} = P^T$ .
- There are  $n!$  different ways to permute the rows of an  $n \times n$  matrix (including the permutation that leaves all rows unfixed) so there are  $n!$  permutation matrices. These matrices form a *multiplicative group*.