

Linear Algebra Recitation

Week 4

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October 1, 2021

Gaussian elimination

$$PA = \begin{bmatrix} P_1 \\ P_2 \\ P_3 \end{bmatrix} A = \begin{bmatrix} P_1 A \\ P_2 A \\ P_3 A \end{bmatrix} = \begin{bmatrix} a_2 \\ a_1 \\ a_3 \end{bmatrix} \Rightarrow \begin{matrix} P_1 = [0 & 1 & 0] \\ P_2 = [1 & 0 & 0] \\ P_3 = [0 & 0 & 1] \end{matrix}$$

- Elementary row operations do not change the solution of a linear system of equations.
- (i) Row permutations (ii) multiplying a row by a scalar (iii) adding a scalar multiple of a row to another row

$$A = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}_{3 \times 3}, PA = \begin{bmatrix} a_2 \\ a_1 \\ a_3 \end{bmatrix}_{3 \times 3}, P = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$MA = \begin{bmatrix} a_1 \\ 2a_2 \\ a_3 \end{bmatrix}_{3 \times 3}, M = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- Basic example: Let $A = \begin{bmatrix} 1 & 2 & 1 \\ 1 & 1 & 2 \\ 2 & 1 & 3 \end{bmatrix}$, $b = \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix}$. Solve $Ax = b$.

Find pivot columns.

$$RA = \begin{bmatrix} a_1 \\ 2a_1 + a_2 \\ a_2 + 3a_1 \end{bmatrix}_{3 \times 3}$$

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

$$\left[\begin{array}{ccc|c} 1 & 2 & 1 & 1 \\ 1 & 1 & 2 & -1 \\ 2 & 1 & 3 & 2 \end{array} \right] \sim \left[\begin{array}{ccc|c} 1 & 2 & 1 & 1 \\ 0 & -1 & 1 & -2 \\ 0 & -3 & 1 & 0 \end{array} \right] \sim \left[\begin{array}{ccc|c} 1 & 2 & 1 & 1 \\ 0 & -1 & 1 & -2 \\ 0 & 0 & 1 & -3 \end{array} \right] \text{ REF}$$

$$\sim \left[\begin{array}{ccc|c} 1 & 2 & 0 & 4 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & -3 \end{array} \right] \sim \left[\begin{array}{ccc|c} 1 & 0 & 0 & 6 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & -3 \end{array} \right] \text{ RREF}$$

Jordan

⊛ # of Pivot Columns
= # L.I. Cols of A

⊛ Pivot cols span $\text{Col}(A)$

Now that we know Gaussian elimination,

- Let's revisit some old topics that we can now fully understand.

- Determine whether $\text{Span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 2 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ -1 \\ 1 \end{bmatrix} \right\}$ is a linearly independent set.

$$A = \begin{bmatrix} 1 & 2 & 0 & -1 \\ 2 & 3 & 2 & 1 \\ 0 & 1 & 0 & -1 \\ -1 & -1 & 3 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 0 & -1 \\ 0 & -1 & 2 & 3 \\ 0 & 1 & 0 & -1 \\ 0 & 1 & 3 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 0 & -1 \\ 0 & -1 & 2 & 3 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 5 & 3 \end{bmatrix}$$

$$\sim \begin{bmatrix} \textcircled{1} & 2 & 0 & -1 \\ 0 & \textcircled{-1} & 2 & 3 \\ 0 & 0 & \textcircled{1} & 1 \\ 0 & 0 & 0 & -2 \end{bmatrix}$$

REF

Now that we know Gaussian elimination,

- Let $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$ be a linear transformation such that $T(1, 1, 2) = (3, -1, 1)$, $T(2, 1, 1) = (0, 1, -2)$, and $T(1, 2, 3) = (-1, 0, 1)$. Find $T(1, -1, 2)$.

$$u, v \in V, \lambda \in \mathbb{R}, T(\lambda u + v) = \lambda T(u) + T(v)$$

$$T\left(\begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}\right) = \begin{pmatrix} 3 \\ -1 \\ 1 \end{pmatrix}, \quad T\left(\begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix}\right) = \begin{pmatrix} 0 \\ 1 \\ -2 \end{pmatrix}, \quad T\left(\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}\right) = \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}$$

$$\text{Find } T\left(\begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix}\right).$$

$$b = \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} + x_2 \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix} + x_3 \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}, \quad x_1, x_2, x_3 \in \mathbb{R}$$

$$T\left(\begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix}\right) = T\left(x_1 \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} + x_2 \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix} + x_3 \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}\right) = x_1 T\left(\begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}\right) + x_2 T\left(\begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix}\right) + x_3 T\left(\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}\right)$$

$$\begin{aligned} \text{A} \quad \begin{bmatrix} 1 & 2 & 1 \\ 1 & 1 & 2 \\ 2 & 1 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} &= b \quad \begin{bmatrix} 1 & 2 & 1 & | & 1 \\ 1 & 1 & 2 & | & -1 \\ 2 & 1 & 3 & | & 2 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 1 & | & 1 \\ 0 & -1 & 1 & | & -2 \\ 0 & -3 & 1 & | & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 1 & | & 1 \\ 0 & -1 & 1 & | & -2 \\ 0 & 0 & -2 & | & 6 \end{bmatrix} \\ &\sim \begin{bmatrix} 1 & 2 & 1 & | & 1 \\ 0 & -1 & 1 & | & -2 \\ 0 & 0 & 1 & | & -3 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & | & 2 \\ 0 & -1 & 0 & | & -1 \\ 0 & 0 & 1 & | & -3 \end{bmatrix} \end{aligned}$$

$$\underbrace{\begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix}} = 1 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} - 1 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + 2 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$\int = x_1 \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} + x_2 \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix} + x_3 \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}_B$$

Fundamental Theorem of Linear Algebra

$$u \perp v \Leftrightarrow u \cdot v = 0$$

- Recall the definition of orthogonal ^{complement} ~~subspaces~~ ("perp")

$$W \subseteq V \quad W^\perp = \{v \in V : v \cdot w = 0, w \in W\}$$

- Show that $\mathcal{N}(A) \stackrel{?}{=} \text{col}(A^T)^\perp$

$$W \cap W^\perp = \{\vec{0}\}$$

$$u \neq \vec{0} \quad u \cdot u = 0 \Leftrightarrow u = \vec{0}$$

$$x \in \mathcal{N}(A)$$

$$\Rightarrow Ax = \vec{0}$$

$$\begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} x = \begin{bmatrix} a_1 \cdot x \\ a_2 \cdot x \\ \vdots \\ a_n \cdot x \end{bmatrix} = \vec{0}$$

$$\begin{aligned} a_1 \cdot x &= 0 \\ a_2 \cdot x &= 0 \\ &\vdots \\ a_n \cdot x &= 0 \end{aligned}$$

$$x \in (\text{col}(A^T))^\perp$$

$$x \in (\text{col}(A^T))^\perp$$

$$Ax = \begin{bmatrix} a_1 \cdot x \\ a_2 \cdot x \\ \vdots \\ a_n \cdot x \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \vec{0}$$

$$\Rightarrow x \in \mathcal{N}(A)$$

- Show that for a subspace $S \subseteq V$, $\dim S + \dim S^\perp = \dim V$.
(Try it at home. Here is a proof: <https://textbooks.math.gatech.edu/ila/orthogonal-complements.html>.)

Fundamental Theorem of Linear Algebra $A: \mathbb{R}^n \rightarrow \mathbb{R}^m$

- Obtain the fundamental theorem of linear algebra from the above results

$$\dim(\text{Col}(A^T)) + \overset{(\text{Col}(A^T))^{\perp}}{\dim(N(A))} = n = \dim(\mathbb{R}^n) \\ (= \text{Rank}(A))$$

- The four fundamental subspaces: \rightarrow Next week

Solution of $Ax = b$

- Always think in terms of the FTLA. Also, always view a matrix-vector product Ax as a linear combination of the columns of A .
- If a solution exists, then $x = x_p + x_h$, where $x_h \in \mathcal{N}(A)$. $Ax_h = 0$
- Thus the solution set is $X = x_p + \mathcal{N}(A)$.
- x unique \iff solution set X contains only one element $\iff \mathcal{N}(A)$ is trivial $\xLeftrightarrow{\text{FTLA}} \dim(\text{col}(A)) = \text{rank}(A) = n$.
- Thus, you can tell if the system has a unique solution by checking the number of linearly independent columns/rows ($= \text{rank}(A)$). *also $b \in \text{col}(A)$*
- When does $Ax = b$ have infinitely many solutions?

$$\begin{cases} \mathcal{N}(A) \text{ is not trivial} & \iff \text{rank}(A) < n \\ \text{and } b \in \text{col}(A) \end{cases}$$

- When does $Ax = b$ have no solutions? $A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$
 $b \notin \text{col}(A)$