

1.3: On the Solutions of Linear Systems; matrix algebra.

I Prelim.

RREF: After Gauss - Jordan Elimination, we say a matrix is in reduced row echelon form, or RREF.

so given matrix A , we can find $\text{rref}(A)$.

Rank: We define the rank of matrix A , or $\text{rank}(A)$, as the number of leading ones in $\text{rref}(A)$.

read carefully thru ex 1-5 and chms 1.3.3 & 1.3.4 on your own.

II Matrix algebra.

— adding matrices (by element)

$$\begin{bmatrix} a_{11} & \dots & a_{1m} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nm} \end{bmatrix}_{n \times m} + \begin{bmatrix} b_{11} & \dots & b_{1m} \\ \vdots & & \vdots \\ b_{n1} & \dots & b_{nm} \end{bmatrix}_{n \times m} = \begin{bmatrix} a_{11}+b_{11} & \dots & a_{1m}+b_{1m} \\ \vdots & & \vdots \\ a_{n1}+b_{n1} & \dots & a_{nm}+b_{nm} \end{bmatrix}_{n \times m}$$

— multiplying a matrix by a scalar. (by element/entry)

$$k \begin{bmatrix} a_{11} & \dots & a_{1m} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nm} \end{bmatrix} = \begin{bmatrix} ka_{11} & \dots & ka_{1m} \\ \vdots & & \vdots \\ ka_{n1} & \dots & ka_{nm} \end{bmatrix}$$

matrix multiplication. (row by column).

ex1: Find $A \cdot B =$

$$\begin{matrix} 3 \times 2 & 2 \times 4 \\ \searrow & \swarrow \\ & \text{Same} \\ & \text{dim of product.} \end{matrix}$$

$$\begin{bmatrix} 2 & \vec{w}_1 & 6 \\ 0 & \vec{w}_2 & 4 \\ 1 & \vec{w}_3 & 2 \end{bmatrix}_{3 \times 2} \cdot \begin{bmatrix} \vec{v}_1 & \vec{v}_2 & \vec{v}_3 & \vec{v}_4 \\ 4 & 5 & 2 & 3 \end{bmatrix}_{2 \times 4}$$

We are using \vec{w}_i 's for rows and \vec{v}_j 's for cols.

Notice our old friend the dot product...

ex2: $\vec{w}_1 \cdot \vec{v}_1 = [2 \ 6] \cdot \begin{bmatrix} 1 \\ 4 \end{bmatrix} = 26$

(the dot product of a row & col vector is a scalar).

more generally.

$$A \cdot B = \begin{bmatrix} \vec{w}_1 \\ \vdots \\ \vec{w}_n \end{bmatrix}_{n \times m} \cdot \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_p \end{bmatrix}_{m \times p}$$

$$= \begin{bmatrix} \vec{w}_1 \cdot \vec{v}_1 & \dots & \vec{w}_1 \cdot \vec{v}_p \\ \vdots & & \vdots \\ \vec{w}_n \cdot \vec{v}_1 & \dots & \vec{w}_n \cdot \vec{v}_p \end{bmatrix}_{n \times p}$$

and if like \vec{v}_1

$$\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix}, \text{ then } A\vec{x} = \begin{bmatrix} \vec{w}_1 \cdot \vec{x} \\ \vdots \\ \vec{w}_n \cdot \vec{x} \end{bmatrix}_{n \times 1}$$

We want another expression for $A\vec{x}$

ex 3:

$$A\vec{x} = \begin{bmatrix} 2 & 6 \\ 0 & 4 \\ 1 & 2 \end{bmatrix}_{3 \times 2} \begin{bmatrix} 1 \\ 4 \end{bmatrix}_{2 \times 1}$$

$$= \begin{bmatrix} 2(1) + 6(4) \\ 0(1) + 4(4) \\ 1(1) + 2(4) \end{bmatrix}$$

$$= 1 \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} + 4 \begin{bmatrix} 6 \\ 4 \\ 2 \end{bmatrix}$$

and

$$A\vec{x} = \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_m \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix} = x_1 \vec{v}_1 + \dots + x_m \vec{v}_m$$

Def: A vector $\vec{b} \in \mathbb{R}^N$ is called a linear combination of $\vec{v}_1, \dots, \vec{v}_m \in \mathbb{R}^N$ if there exist scalars x_1, \dots, x_m s.t. $\vec{b} = x_1 \vec{v}_1 + \dots + x_m \vec{v}_m$.

so $A\vec{x}$ is a linear combo of $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m$

This will be used extensively in the theoretical areas. We can see ...

$$A\vec{x} = x_1 \vec{v}_1 + \dots + x_m \vec{v}_m \quad (\text{you can break apart } A\vec{x})$$

and

$$x_1 \vec{v}_1 + \dots + x_m \vec{v}_m = A\vec{x} \quad (\text{you can combine a lin. combo into } A\vec{x}).$$

Two rules for $A\vec{x}$: If A is an $n \times m$ matrix and $\vec{x}, \vec{y} \in \mathbb{R}^m$, and k is a scalar

$$(a) \quad A(\vec{x} + \vec{y}) = A\vec{x} + A\vec{y}$$

$$(b) \quad A(k\vec{x}) = kA\vec{x}$$

□ proof of (b).

$$\begin{aligned} A(k\vec{x}) &= \begin{bmatrix} a_{11} & \dots & a_{1m} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nm} \end{bmatrix} \begin{bmatrix} kx_1 \\ \vdots \\ kx_m \end{bmatrix} = \begin{bmatrix} a_{11}kx_1 + \dots + a_{1m}kx_m \\ \vdots \\ a_{n1}kx_1 + \dots + a_{nm}kx_m \end{bmatrix} \\ &= k \begin{bmatrix} a_{11}x_1 + \dots + a_{1m}x_m \\ \vdots \\ a_{n1}x_1 + \dots + a_{nm}x_m \end{bmatrix} \\ &= k \begin{bmatrix} a_{11} & \dots & a_{1m} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nm} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix} \\ &= kA\vec{x} \quad \blacksquare \end{aligned}$$

1.3
5/5

Linear System:
$$\begin{cases} x_1 + 2x_2 = 8 \\ 3x_1 - x_2 = 3 \end{cases}$$

w/ augmented matrix $[A | b]$
$$\left[\begin{array}{cc|c} 1 & 2 & 8 \\ 3 & -1 & 3 \end{array} \right]$$

$$\begin{bmatrix} 1 & 2 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 8 \\ 3 \end{bmatrix} \Rightarrow x_1 \begin{bmatrix} 1 \\ 3 \end{bmatrix} + x_2 \begin{bmatrix} 2 \\ -1 \end{bmatrix} = \begin{bmatrix} 8 \\ 3 \end{bmatrix}$$

