

Approximating Solutions to CDE's via Finite Differences (Continued)

In our last lecture we derived

$$\begin{cases} \underline{x}_{i+1} = \underline{x}_i + hA \underline{x}_i, & i=0,1,2,\dots \\ \underline{x}_0 = \underline{x}_0, & \text{known.} \end{cases}$$

$$\text{So } \underline{x}_{i+1} = \underline{I}_n \underline{x}_i + hA \underline{x}_i = (\underline{I}_n + hA) \underline{x}_i$$

$$(\dagger) \quad \underline{x}_{i+1} = M \underline{x}_i \quad \begin{array}{l} M = \underline{I}_n + hA \\ \text{a constant matrix.} \end{array}$$

$$\text{Note: } \underline{x}_i = M \underline{x}_{i-1}$$

$$\underline{x}_{i-1} = M \underline{x}_{i-2}$$

etc

$$\text{So } (\dagger) \quad \underline{x}_{i+1} = M(M \underline{x}_{i-1}) = M(M(M \underline{x}_{i-2})) \dots$$

By induction:

$$(\S) \quad \underset{\sim}{x}_n = M^n \underset{\sim}{x}_0$$

$n=1, 2, \dots$

Note that $(\S)$ is familiar to us:

if we call $\underset{\sim}{x}_n = \underset{\sim}{z}$, $B = M^n$, $\underset{\sim}{x}_0 = \underset{\sim}{c}$

then $(\S)$ is of the form

$$\underset{\sim}{z} = B \underset{\sim}{c},$$

A linear transformation, with known B and $\underset{\sim}{c}$.

MATRIX OPERATIONS

Thm: Let A, B, C be same-sized matrices $\mathbb{R}^{m \times n}$

- (a) $A+B = B+A$
- (b) $(A+B)+C = A+(B+C)$
- (c) $A+\phi = A$
- (d) $\alpha(A+B) = \alpha A + \alpha B$
- (e) $(\alpha+\beta)A = \alpha A + \beta A$
- (f) $\alpha(\beta A) = (\alpha\beta)A$

There are many ways to define the operation of multiplication in vector spaces. Here we focus on the simplest "matrix multiplication", as described below:

$$\text{If } A \in \mathbb{R}^{m \times n} \quad B \in \mathbb{R}^{n \times p} \quad A = [a_{ij}], B = [b_{ij}]$$

$$AB = \sum_{j=1}^n a_{ij} b_{jk}$$

$$\text{ex) } A = \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \quad B = \begin{bmatrix} 4 & 3 & 6 \\ 1 & -2 & 3 \end{bmatrix}$$

$$A \in \mathbb{R}^{2 \times 2} \quad B \in \mathbb{R}^{2 \times 3}$$

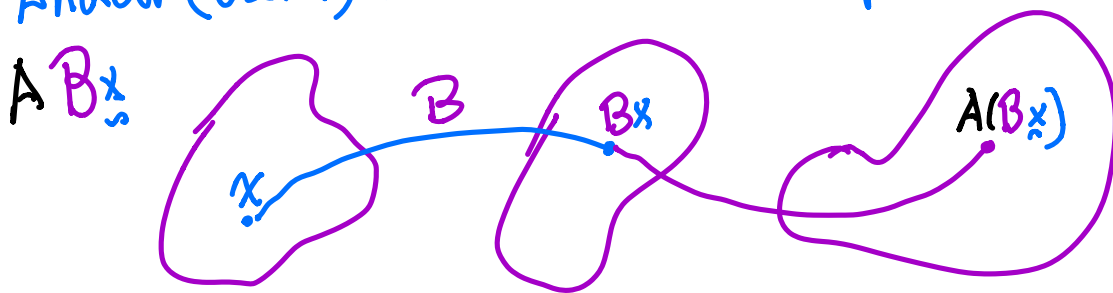
$$\therefore AB \in \mathbb{R}^{2 \times 3}$$

$$B = \begin{bmatrix} 4 & 3 & 6 \\ 1 & -2 & 3 \end{bmatrix}$$

$$A \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 11 & 0 & 21 \\ -1 & 13 & -9 \end{bmatrix} \quad \underline{AB}$$

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Another (useful) Take on Matrix Multiplication:



$$\text{So } (AB)x = Cx$$

$$= \begin{bmatrix} \overset{1}{Ab_1} & \overset{1}{Ab_2} & \dots & \overset{1}{Ab_p} \end{bmatrix} x$$

columns of C

"Multiplication AB is a composition of 2 linear transformations"

Going back to the example

$$\text{ex) } A = \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \quad B = \begin{bmatrix} 4 & 3 & 6 \\ 1 & -2 & 3 \end{bmatrix}$$

$$A \in \mathbb{R}^{2 \times 2} \quad B \in \mathbb{R}^{2 \times 3}$$

$$\therefore AB \in \mathbb{R}^{2 \times 3}$$

$$B = \begin{bmatrix} 4 & 3 & 6 \\ 1 & -2 & 3 \end{bmatrix}$$

$$A \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 11 & 0 & 21 \\ -1 & 13 & -9 \end{bmatrix} = \begin{bmatrix} Ab_1 & Ab_2 & Ab_3 \end{bmatrix}$$

$$B = \begin{bmatrix} \overset{1}{b_1} & \overset{1}{b_2} & \overset{1}{b_3} \end{bmatrix}$$

$$Ab_1 = \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 4 \\ 1 \end{bmatrix} = \begin{bmatrix} 11 \\ -1 \end{bmatrix} \quad Ab_2 = \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 3 \\ -2 \end{bmatrix} = \begin{bmatrix} 0 \\ 13 \end{bmatrix}$$

$$Ab_3 = \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 6 \\ 3 \end{bmatrix} = \begin{bmatrix} 21 \\ -9 \end{bmatrix}$$