

MORE PROOFS

"Without loss of generality" WLOG, where the essence of the original problem is not lost when using a simpler example.

ex. 3 numbers a, b, c

WLOG, let a be the largest, c be the smallest
This will normally be in relabeling/reordering cases.

THEOREM: If $A\vec{x} = \vec{b}$ has 2 or more solutions, then the columns of A are linearly dependent.

PROOF: Let $\vec{x}_1 \neq \vec{x}_2$ be 2 distinct solutions.

$$A\vec{x}_1 = \vec{b} \quad \text{and} \quad A\vec{x}_2 = \vec{b}$$

SHOW: For some i , $\vec{a}_i = \sum_{j=1, j \neq i}^n c_j \cdot \vec{a}_j$

$$A\vec{x}_1 = A\vec{x}_2 \rightarrow A(\vec{x}_1 - \vec{x}_2) = 0$$

$$A \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = 0 \rightarrow y_1 \cdot \vec{a}_1 + y_2 \cdot \vec{a}_2 + \dots + y_n \cdot \vec{a}_n = 0$$

At least one y_i has to be nonzero

$$\text{WLOG, let } y_1 \neq 0 \rightarrow \vec{a}_1 = -\frac{y_2}{y_1} \cdot \vec{a}_2 + \dots + -\frac{y_n}{y_1} \cdot \vec{a}_n$$

QED.

LINEAR DEPENDENCE

Second Definition: $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n\}$ are linearly dependent if there exist c_1, c_2, c_n , not all zero, such that $c_1 \cdot \vec{a}_1 + c_2 \cdot \vec{a}_2 + \dots + c_n \cdot \vec{a}_n = 0$.

LINEAR TRANSFORMATIONS

Matrices are linear operators.

$$\begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$$

Rotation Matrix

$$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

Reflection
($\vec{x} \rightarrow -\vec{x}$)

Matrix is a linear trans.

f is a linear trans. if:

$$f(x, y) = f(x) + f(y)$$

$$f(\alpha x) = \alpha f(x)$$

eg. $f(2\vec{x}) = 2\vec{x}$

$$f(\vec{x} + \vec{y}) = 2\vec{x} + 2\vec{y} = f(\vec{x}) + f(\vec{y})$$

$$f(\alpha \vec{x}) = 2\alpha \vec{x} = \alpha f(\vec{x})$$

STATE TRANSFORMATIONS

Vectors often represent the state of a system.

eg. state of car $\rightarrow x(t), y(t), v(t) \rightarrow \vec{s}(t) = \begin{bmatrix} x(t) \\ y(t) \\ v(t) \end{bmatrix}$

EECS 16B dives into this with Control Theory

eg. System of Pumps & Reservoirs

$x_A(t)$ = water in tank A @ time t.
 $x_B(t)$ = water in tank B @ time t.
 $x_C(t)$ = water in tank C @ time t.

(run pump)

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \quad \xrightarrow{\qquad Q \qquad}$$

$\vec{x}(t+1)$ $\vec{x}(t)$ \rightarrow State Transition Matrix

What about $\vec{x}(t+2)$ in terms of $\vec{x}(t)$? $\vec{x}(t+2000)$?

$$Q \vec{x}(t) = \vec{x}(t+1) \quad Q\vec{x}(t+1) = \vec{x}(t+2)$$

$$(Q \cdot Q) \cdot \vec{x}(t) = \vec{x}(t+2)$$

See next page for information

$$Q \cdot Q = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

identity trans.

MATRIX-MATRIX MULTIPLICATION

$$A \cdot B = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = A \begin{bmatrix} \vec{b}_1 & \vec{b}_2 \end{bmatrix} = \text{stacked matrix-vector mult!}$$

$$= \begin{bmatrix} a_{11}b_{11} + a_{12}b_{21} & a_{11}b_{12} + a_{12}b_{22} \\ a_{21}b_{11} + a_{22}b_{21} & a_{21}b_{12} + a_{22}b_{22} \end{bmatrix}$$

$$A \cdot B = A [\vec{b}_1 \ \vec{b}_2 \ \dots \ \vec{b}_n] = [A\vec{b}_1 \ A\vec{b}_2 \ \dots \ A\vec{b}_n]$$

e.g.

$$\begin{array}{ccc} & \overset{1/3 \rightarrow}{\curvearrowleft} & O_A \\ \overset{1/3}{\curvearrowright} & & \overset{1/3}{\curvearrowright} \\ & \overset{1/2 \rightarrow}{\curvearrowleft} & O_B \\ & \overset{3/4}{\curvearrowright} & \overset{1/4}{\curvearrowright} \end{array}$$

$$Q = \begin{bmatrix} 1/3 & 0 & 0 \\ 1/3 & 1/2 & 3/4 \\ 1/3 & 1/2 & 1/4 \end{bmatrix} \begin{array}{l} \text{inflow to } A \\ \text{B} \\ \text{C} \end{array}$$

outflow of A B C

$Q \cdot Q \cdot Q \dots Q$ to see water levels at the end of time
 np. matmul(Q, Q) np.linalg.matrix_power(Q, 20)

PROPERTIES

① Matrix-Matrix Multiplication does not commute.

$AB \neq BA$, usually
is associative.

$$C(B(Ax)) = ((CB)(Ax)) \neq ABC$$