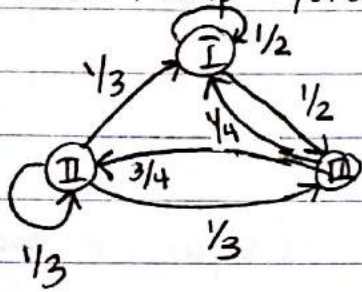


Water Pump System:



State:

$$\vec{s}[n] = \begin{bmatrix} s_1[n] \\ s_2[n] \\ s_3[n] \end{bmatrix}$$

amt of water in I

state vector @
time n

$$s[n]: \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \quad s[n+1]: \begin{bmatrix} 1/2 \\ 0 \\ 1/2 \end{bmatrix} \quad \vec{s}[n+1] = A s[n]$$

↑ state transition matrix

$\vec{e}_1 \rightarrow$

unit vector

$$A: \begin{bmatrix} 1/2 & 1/3 & 1/4 \\ 0 & 1/3 & 3/4 \\ 1/2 & 1/3 & 0 \end{bmatrix}$$

← probability of moving to I
" " II
" " III

↑ ↑ ↑
from I from II from III

$$[\vec{a}_1, \vec{a}_2, \vec{a}_3] \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = [\vec{a}_1]$$

$$\vec{s}[n] = \begin{bmatrix} \alpha \\ \beta \\ \gamma \end{bmatrix} \quad \vec{s}[n+1] = \begin{bmatrix} \frac{1}{2}\alpha + \frac{1}{3}\beta + \frac{1}{4}\gamma \\ 0\alpha + \frac{1}{3}\beta + \frac{3}{4}\gamma \\ \frac{1}{2}\alpha + \frac{1}{3}\beta + 0\gamma \end{bmatrix} = A s[n]$$

(state evolution equation)

Q: If given $\vec{s}[n+1]$, can you determine $s[n]$?

A: Depends on A.

Dfn: A is said to be invertible if there exists a matrix, denoted by A^{-1} s.t. $A \cdot A^{-1} = I$

$$I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \text{ (of any dimension)}$$

think of I as 1. $IA = A$. (in world of square matrices)

Suppose A^{-1} exists.

$$A^{-1}[A]s = A^{-1}A s = s$$

$$\rightarrow A^{-1}s = s$$

Matrix Inverses:

* SQUARE MATRICES!

$$\cdot A^{-1}A = AA^{-1} = I$$

• If A has an inverse, it is unique

Assume B & C are inverses of A :

$$BA = I$$

$$AC = I \rightarrow B(AC) = BI = B$$

$$(BA)C = B \rightarrow C = B$$

• If A is invertible, $A\vec{x} = \vec{b}$ has a unique solution

$$A^{-1}A\vec{x} = A^{-1}\vec{b}$$

$$\vec{x} = A^{-1}\vec{b} \checkmark$$

• If A has linearly dependent columns, it cannot have an inverse

$$A = [\vec{a}_1, \dots, \vec{a}_N] \rightarrow \vec{\alpha} \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_N \end{bmatrix} \in \mathbb{R}^N \text{ s.t. } A\vec{\alpha} = \vec{0}, \vec{\alpha} \neq \vec{0}$$

• Assume an inverse exists:

$$A\vec{\alpha} = \vec{0}$$

$$A^{-1}A\vec{\alpha} = A^{-1}\vec{0} \rightarrow \vec{\alpha} = \vec{0} \text{ contradicts linear dependence}$$

3 Reservoir Example:

$$A = \begin{bmatrix} 1/2 & 1/3 & 1/4 \\ 0 & 1/3 & 3/4 \\ 1/2 & 1/3 & 0 \end{bmatrix}$$

$$A^{-1} = [\vec{x}_1 \quad \vec{x}_2 \quad \vec{x}_3]$$

$$\begin{matrix} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} & \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} & \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \\ \uparrow & \uparrow & \uparrow \\ \vec{e}_1 & \vec{e}_2 & \vec{e}_3 \end{matrix}$$

$$AA^{-1} = I = A[\vec{x}_1 \quad \vec{x}_2 \quad \vec{x}_3] = [\vec{e}_1 \quad \vec{e}_2 \quad \vec{e}_3]$$

$$[A\vec{x}_1 \quad A\vec{x}_2 \quad A\vec{x}_3] = [\vec{e}_1 \quad \vec{e}_2 \quad \vec{e}_3]$$

$$\hookrightarrow A\vec{x}_1 = \vec{e}_1$$

$$A\vec{x}_2 = \vec{e}_2$$

$$A\vec{x}_3 = \vec{e}_3$$

$$[A \mid I] \quad (\text{want to get } A \text{ to look like } I)$$

$$\left[\begin{array}{ccc|ccc} 1/2 & 1/3 & 1/4 & 1 & 0 & 0 \\ 0 & 1/3 & 3/4 & 0 & 1 & 0 \\ 1/2 & 1/3 & 0 & 0 & 0 & 1 \end{array} \right]$$

This method is called Gauss-Jordan elimination

$$= \left[\begin{array}{ccc|ccc} 1/2 & 1/3 & 1/4 & 1 & 0 & 0 \\ 0 & 1/3 & 3/4 & 0 & 1 & 0 \\ 0 & 0 & -1/4 & -1 & 0 & 1 \end{array} \right]$$

$$= \left[\begin{array}{ccc|ccc} 1/2 & 0 & 0 & 3 & -1 & -2 \\ 0 & 1/3 & 0 & -3 & 1 & 3 \\ 0 & 0 & -1/4 & -1 & 0 & 1 \end{array} \right]$$

$$= \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & 6 & -2 & -4 \\ 0 & 1 & 0 & -9 & 3 & 9 \\ 0 & 0 & 1 & 4 & 0 & -4 \end{array} \right]$$

$$A^{-1} = \begin{bmatrix} 6 & -2 & -4 \\ -9 & 3 & 9 \\ 4 & 0 & -4 \end{bmatrix}$$

↑
YAY!

2x2 case

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \rightarrow A^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

$\Delta = ad-bc$ determ. of A

Rank of a matrix $A \in \mathbb{R}^{n \times n}$

$\text{rank}(A) = \#$ linearly independent columns of A
(pivots)

$\#$ linearly independent rows of A for
square matrices

A is rank deficient if it has fewer than
 N lin. indep cols (rows)

Range of A : $\text{span}\{\vec{a}_1, \dots, \vec{a}_n\}$

If A is full rank, $\text{span}\{\vec{a}_1, \dots, \vec{a}_n\} = \mathbb{R}^n$

$$A\vec{x} = \vec{b}$$