

Welcome to EECS 16A!

Designing Information Devices and Systems I



Ana Arias and Miki Lustig
Sp 2022

Lecture 3A
Matrix xForms



Announcements

- Last time:
 - Proofs
 - Span
- Today:
 - Linear (in)dependence
 - Matrix Transformations

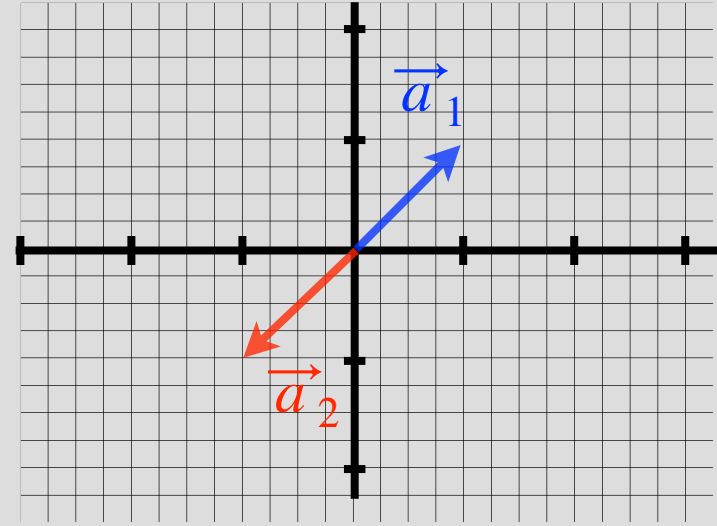


Linear Dependence

Recall:

$$A = \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}$$

\downarrow \downarrow
 \vec{a}_1 \vec{a}_2



\vec{a}_1 and \vec{a}_2 are linearly dependent

$$\vec{a}_1 = -\vec{a}_2$$



Linear Dependence

- Definition 1:

A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_N\}$ are linearly dependent if

$\exists \{\alpha_1, \alpha_2, \dots, \alpha_N\} \in \mathbb{R}$, such that:

$$\vec{a}_i = \sum_{j \neq i} \alpha_j \vec{a}_j \quad 1 \leq i, j \leq M$$

For example: if $\vec{a}_2 = 3\vec{a}_1 - 2\vec{a}_5 + 6\vec{a}_7$

↓

\vec{a}_i in the span of all \vec{a}_j s

Linear Dependence

Are these linearly dependent?

$$\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 3 \\ 1 \end{bmatrix} \right\}$$

Need to solve:

Linear Dependence

Are these linearly dependent?

$$\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 3 \\ 1 \end{bmatrix} \right\}$$

Are linearly dependent

Need to solve:

$$\begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

but we showed that....

$$\frac{b_1 + b_2}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \frac{b_1 - b_2}{2} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

So....

$$\frac{3+1}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \frac{3-1}{2} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

Linear dependence / independence

$$\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 3 \\ 1 \end{bmatrix} \right\} \Rightarrow 2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 1 \\ -1 \end{bmatrix} - \begin{bmatrix} 3 \\ 1 \end{bmatrix} = 0$$

- Definition 2:

A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_N\}$ are linearly dependent if $\exists \{\alpha_1, \alpha_2, \dots, \alpha_N\} \in \mathbb{R}$, such that:

$$\sum_{i=1}^N \alpha_i \vec{a}_i = 0$$

As long as not all $a_i = 0$

- Definition:

A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_N\}$ are linearly independent if they are not dependent

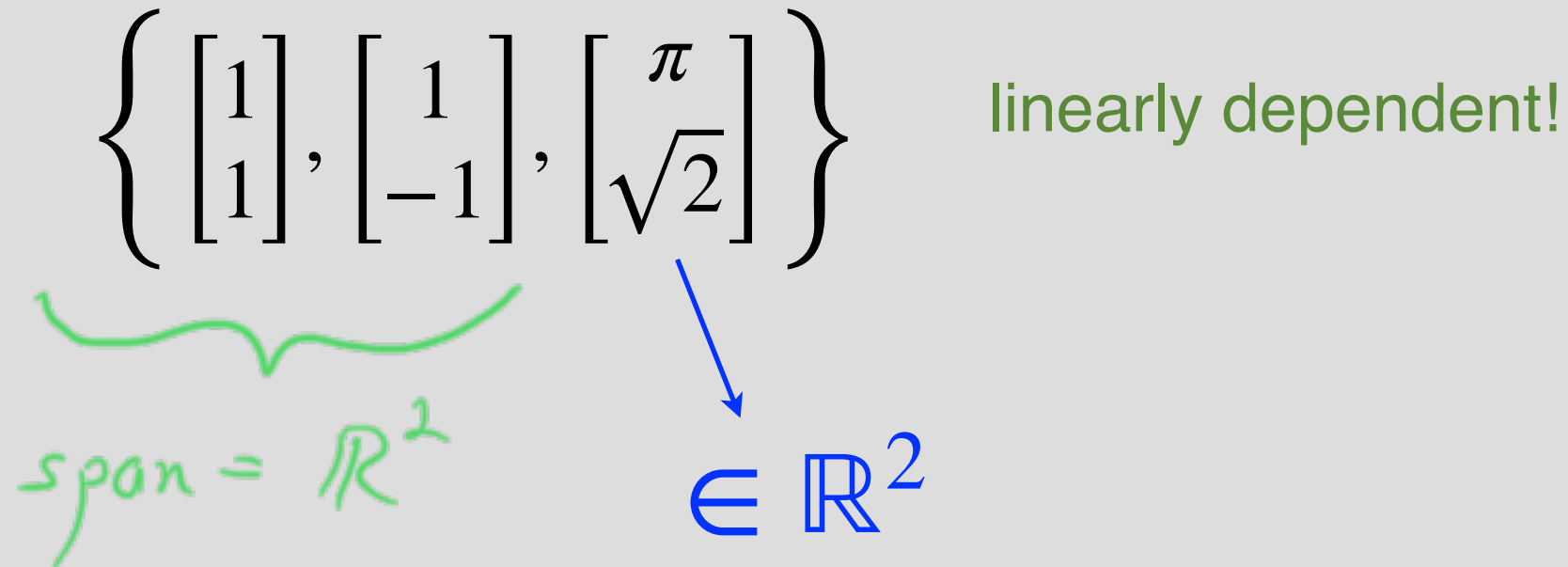
Linear dependence / independence

Are these linearly dependent?

$$\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} \pi \\ \sqrt{2} \end{bmatrix} \right\} \quad \text{linearly dependent!}$$

span = \mathbb{R}^2

$\in \mathbb{R}^2$

A green curly brace under the first two vectors in the set is labeled "span = R^2". A blue arrow points from the third vector in the set down to the text "in R^2".

Solutions for linear equations

- Theorem: if the columns of the matrix A are linearly dependent then, $A\vec{x} = \vec{b}$ does not have a unique solution

PROOF Consider the counter-example $S \triangleq \{0, \bullet\}$, $\tau \triangleq \{(\bullet, \bullet), (\bullet, 0), (0, 0)\}$ so that $\mathcal{M}_\tau = \{(i, \lambda \ell \cdot \bullet), (j, \lambda \ell \cdot 0), (k, \lambda \ell \cdot (\ell < m ? \bullet \dot{\iota} 0))\}$. We let $\mathcal{X} \triangleq \{(i, \sigma) \mid \forall j < i : \sigma_j = \bullet\}$ so that $\neg FD(\mathcal{X})$. We have $\mathcal{M}_\tau \downarrow_\bullet = \{(i, \lambda \ell \cdot \bullet), (k, \lambda \ell \cdot (\ell < m ? \bullet \dot{\iota} 0)) \mid k < m\}$, $\mathcal{M}_\tau \downarrow_0 = \{(j, \lambda \ell \cdot 0), (k, \lambda \ell \cdot (\ell < m ? \bullet \dot{\iota} 0)) \mid k \geq m\}$ and $\oplus \llbracket \mathcal{X} \rrbracket = \{(i, \sigma) \mid \forall j \leq i : \sigma_j = \bullet\}$. We have $\alpha_{\mathcal{M}_\tau}^{\vee}(\oplus \llbracket \mathcal{X} \rrbracket) = \{s \mid \mathcal{M}_\tau \downarrow_s \subseteq \oplus \llbracket \mathcal{X} \rrbracket\} = \{\bullet\}$ whereas $\widetilde{pre}[\tau](\alpha_{\mathcal{M}_\tau}^{\vee}(\mathcal{X})) = \widetilde{pre}[\tau](\{s \mid \mathcal{M}_\tau \downarrow_s \subseteq \mathcal{X}\}) = \widetilde{pre}[\tau](\{\bullet\}) = \{s \mid \forall s' : t(s, s') \Rightarrow s' = \bullet\} = \emptyset$ since $t(s, \bullet)$ implies $s = \bullet$ and $t(\bullet, 0)$ holds. ■

Solutions for linear equations

- Theorem: if the columns of the matrix A are linearly dependent then, $A\vec{x} = \vec{b}$ does not have a unique solution

Proof for $A \in \mathbb{R}^{3 \times 3}$

know: columns are linearly ~~independent~~

show: more than 1 solution

Concept: pick some specific solution \vec{x}^* , and show that there's another one

Let: $A\vec{x}^* = \vec{b}$ and $A = [\vec{a}_1 \ \vec{a}_2 \ \vec{a}_3]$

From linear dependence Def 2:

$$\alpha_1 \vec{a}_1 + \alpha_2 \vec{a}_2 + \alpha_3 \vec{a}_3 = 0$$

Solutions for linear equations

- Theorem: if the columns of the matrix A are linearly dependent then, $A\vec{x} = \vec{b}$ does not have a unique solution

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From linear dependence Def 2:

$$\alpha_1 \vec{a}_1 + \alpha_2 \vec{a}_2 + \alpha_3 \vec{a}_3 = 0 \longrightarrow \begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \vec{a}_3 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{bmatrix} = \vec{0} \quad \Rightarrow A\vec{\alpha} = 0$$

Set $\vec{x}^\dagger = \vec{x}^* + \vec{\alpha}$

$$\Rightarrow A\vec{x}^\dagger = A(\vec{x}^* + \vec{\alpha}) = A\vec{x}^* + A\vec{\alpha} = \vec{b} + 0$$

So \vec{x}^\dagger is another solution!

Matrix Transformations

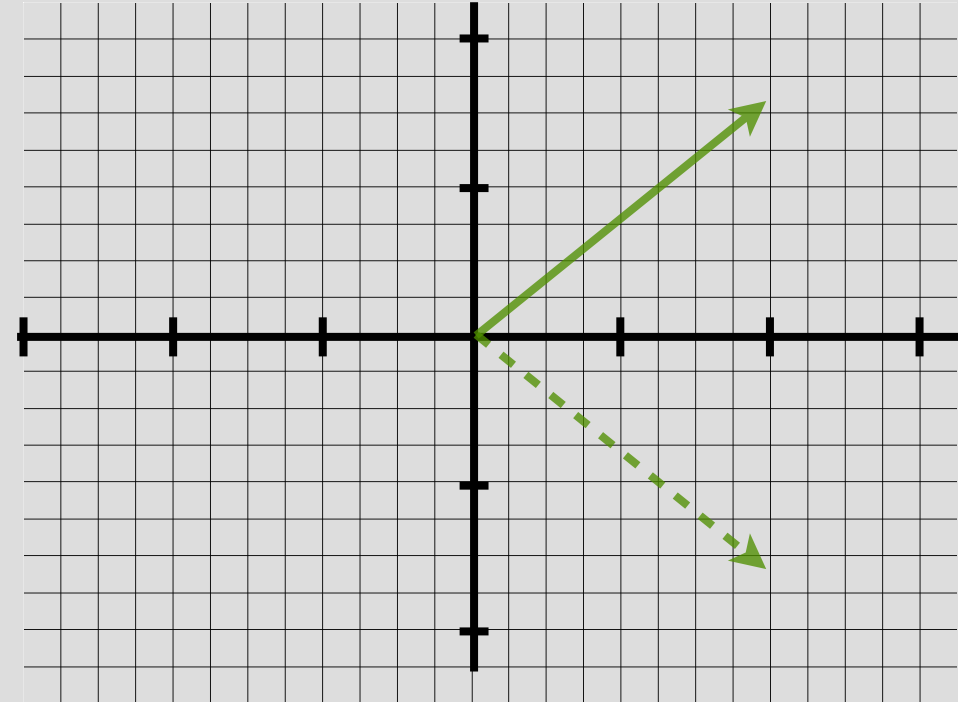
$$\begin{bmatrix} \cos 90^\circ & \sin 90^\circ \\ -\sin 90^\circ & \cos 90^\circ \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 0 \\ a_1 \end{bmatrix}$$

Matrices are operators that transform vectors

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

Example:

$$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ -x_2 \end{bmatrix}$$



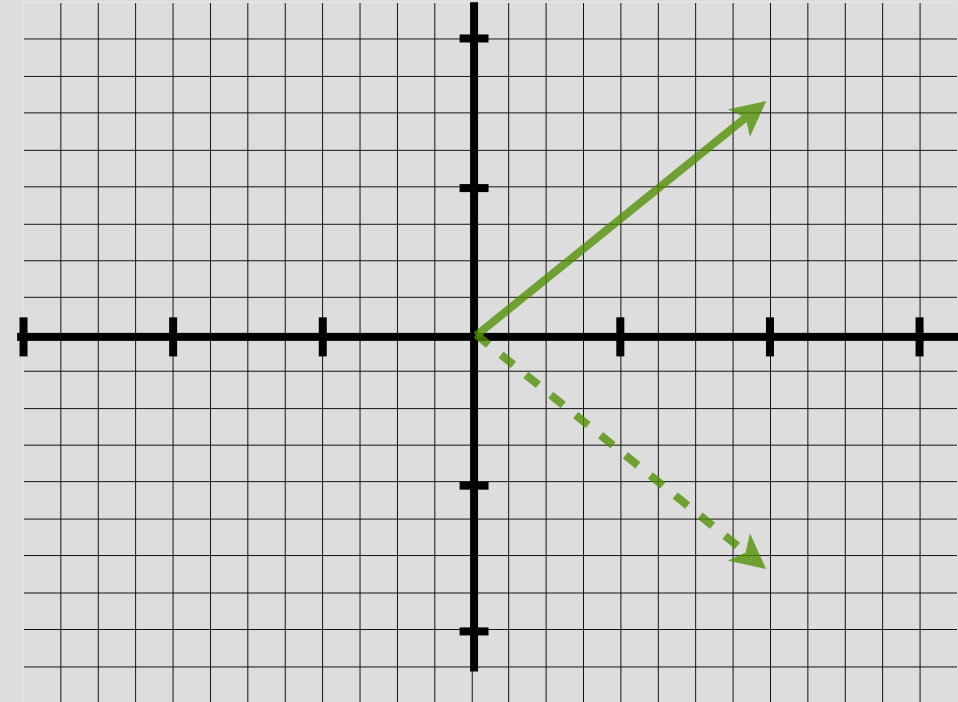
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https://www.youtube.com/watch?v=LhF_56SxrGk



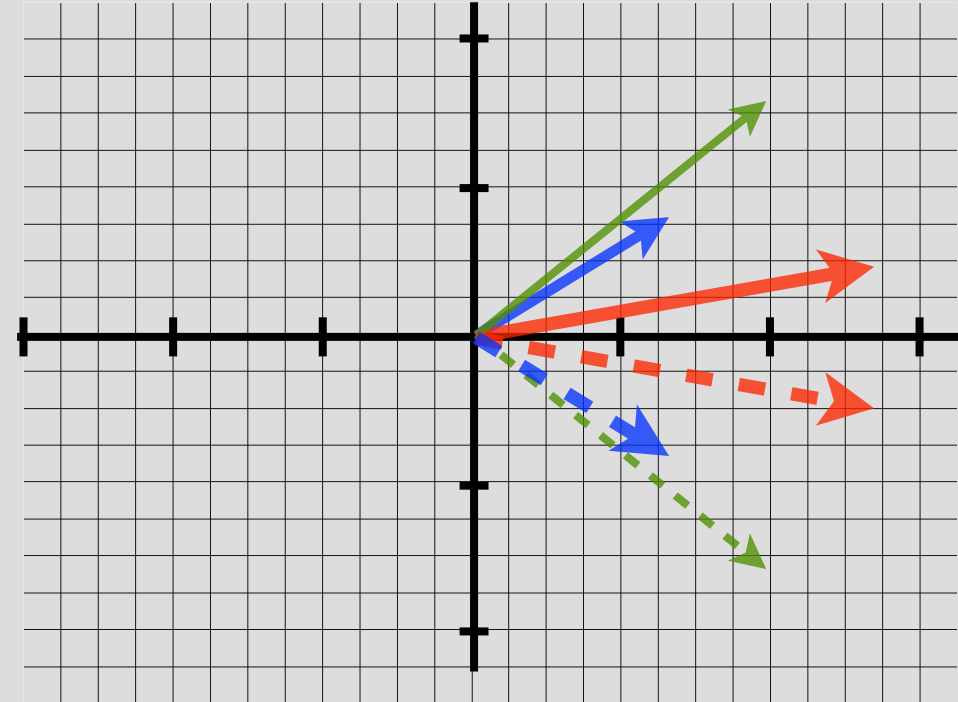
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Reflection Matrix!



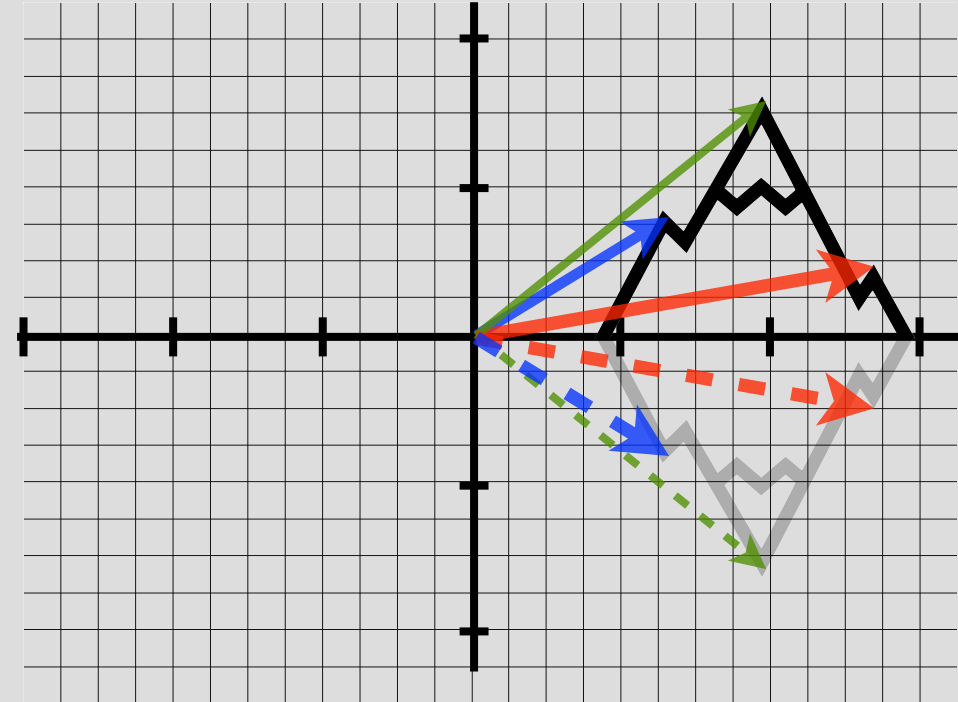
Matrices are operators that transform vectors

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

Example:

$$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ -x_2 \end{bmatrix}$$

Reflection Matrix!



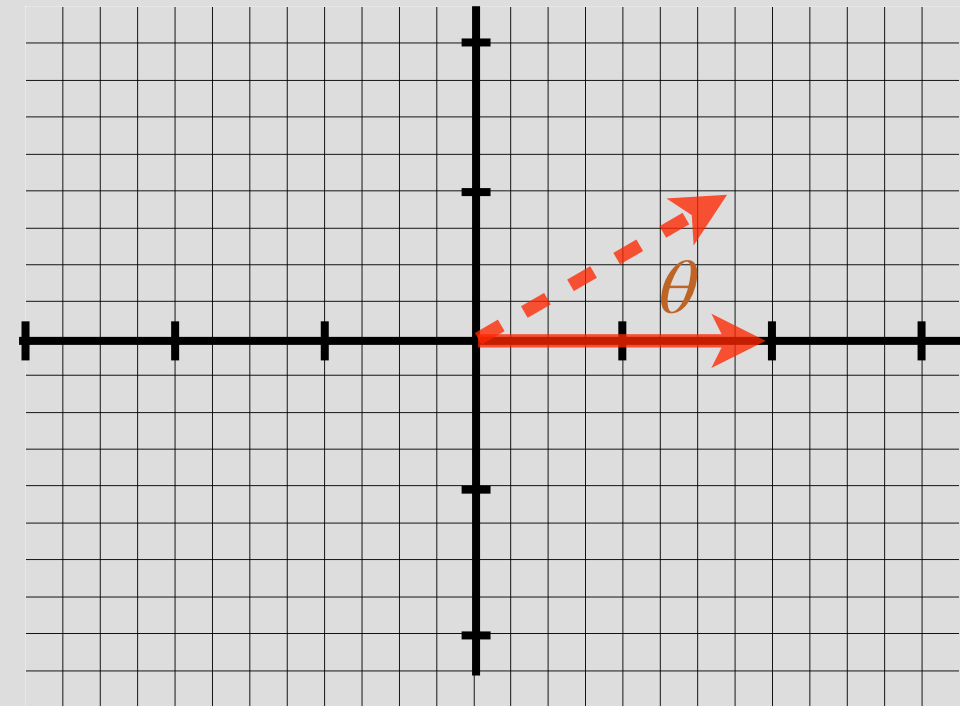
Matrices are operators that transform vectors

Example 2:
$$\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \cos(\theta)x_1 - \sin(\theta)x_2 \\ \sin(\theta)x_1 + \cos(\theta)x_2 \end{bmatrix}$$

$$\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \cos(\theta) \\ \sin(\theta) \end{bmatrix}$$

Rotation Matrix!

$$\begin{bmatrix} \cos 90^\circ & \sin 90^\circ \\ -\sin 90^\circ & \cos 90^\circ \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 0 \\ a_1 \end{bmatrix}$$



Linear Transformation of vectors

f : is a linear transformation if:

$$f(\alpha \vec{x}) = \alpha f(\vec{x}) \quad \alpha \in \mathbb{R}$$

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

Claim: Matrix-vector multiplications satisfy linear transformation

$$A \cdot (\alpha \vec{x}) = \alpha A \vec{x}$$

Proof via explicitly writing the elements

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

Vectors as states, Matrices as state transition

Vectors can represent states of a system

Example: The state of a car at time = t

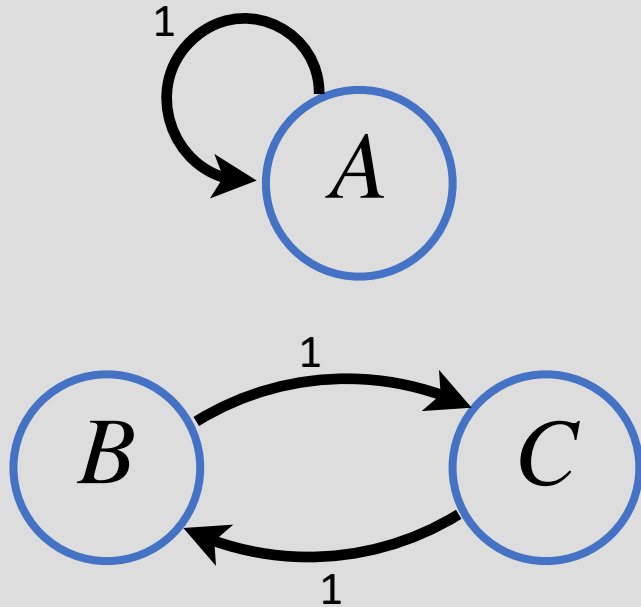
$$\vec{S}(t) = \begin{bmatrix} x(t) \\ y(t) \\ v(t) \\ \theta(t) \end{bmatrix} \left. \begin{array}{l} \} \text{position} \\ \} \text{velocity} \end{array} \right\}$$

Q: Is that enough?

A: need orientation or $v_x(t), v_y(t)$

Graph Transition Matrices

Example: Reservoirs and Pumps



Q: What is the state?

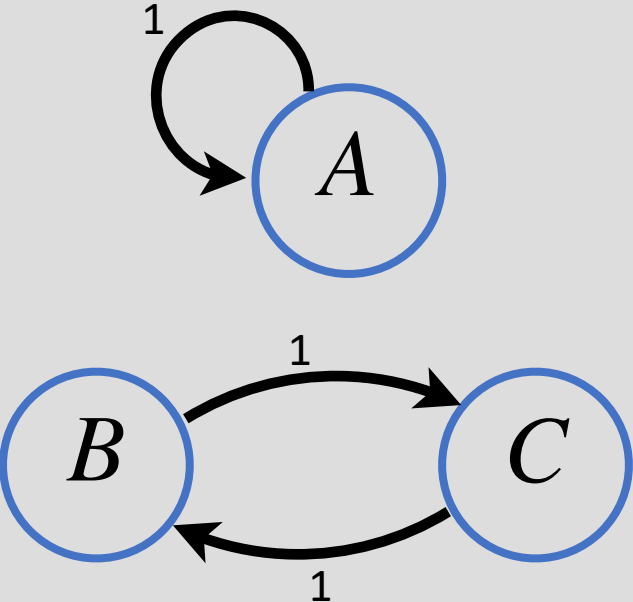
A: Water in each reservoir

$$\vec{x}(t) = \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

Pumps move water...

What would the state be tomorrow?

State Transition Matrices



State Transition Matrices

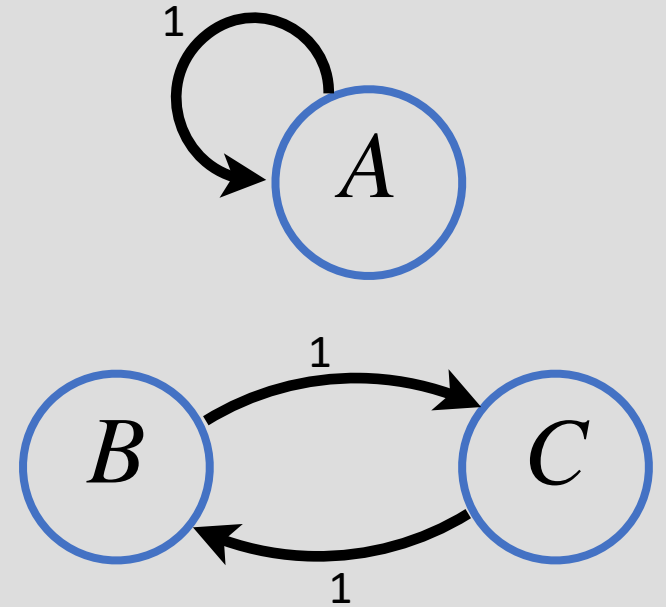
$$x_A(t + 1) = x_A(t)$$

$$x_B(t + 1) = x_C(t)$$

$$x_C(t + 1) = x_B(t)$$

Write as a matrix-vector multiplication:

$$\begin{bmatrix} x_A(t + 1) \\ x_B(t + 1) \\ x_C(t + 1) \end{bmatrix} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$



State Transition Matrices

$$x_A(t + 1) = x_A(t)$$

$$x_B(t + 1) = x_C(t)$$

$$x_C(t + 1) = x_B(t)$$

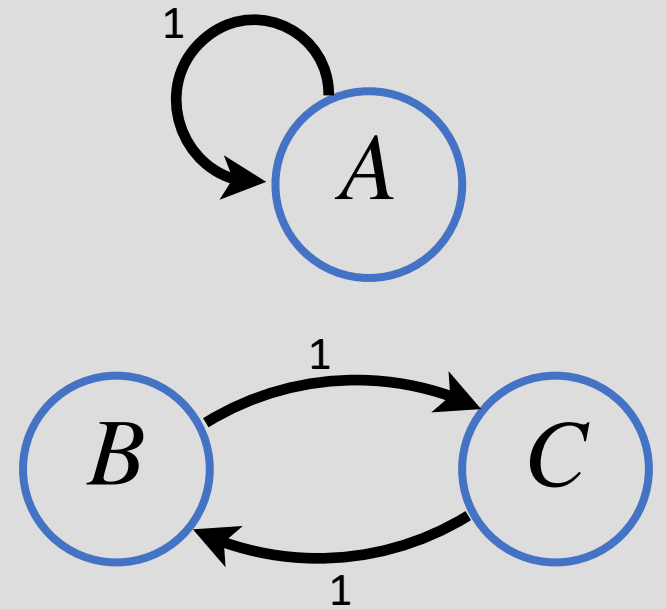
Write as a matrix-vector multiplication:

$$\begin{bmatrix} x_A(t + 1) \\ x_B(t + 1) \\ x_C(t + 1) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

$$\text{or } \vec{x}(t + 1) = Q \vec{x}(t)$$

What is the state after 2 times?

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

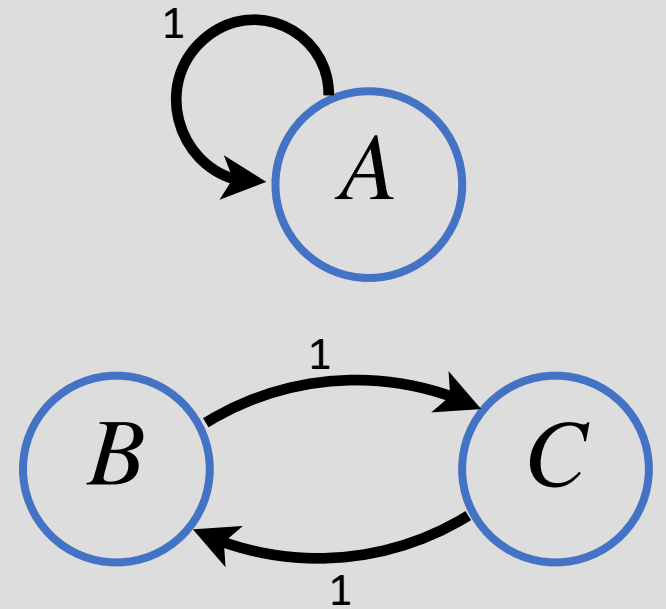


State Transition Matrices

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

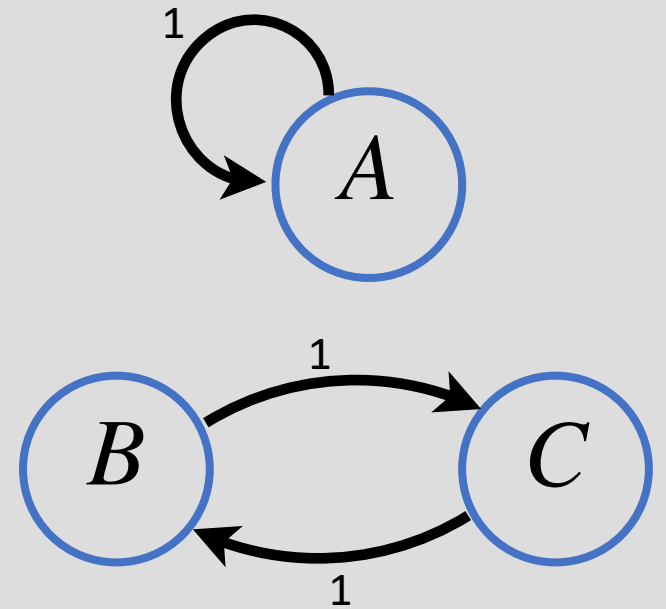
$$\vec{x}(0) = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

What is the state after at $t=1, 2$?



State Transition Matrices

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$



$$\vec{x}(0) = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

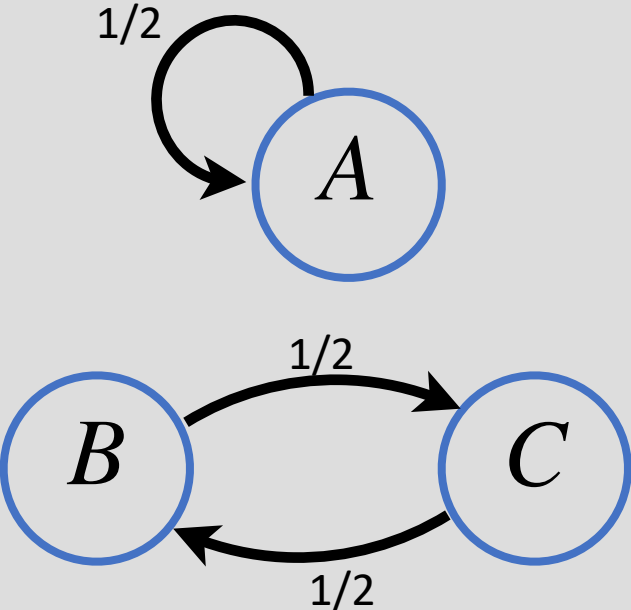
What is the state after at $t=1, 2$?

$$\textcircled{1} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

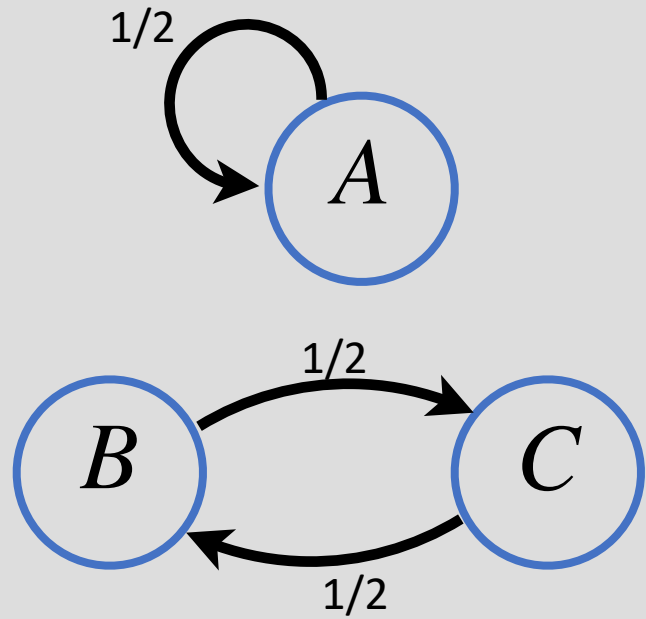
$$\textcircled{2} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

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

State Transition Matrices



State Transition Matrices



$$x[t+1] = \begin{bmatrix} 1/2 & 0 & 0 \\ 0 & 1/2 & 0 \\ 0 & 1/2 & 0 \end{bmatrix} x(t)$$

Non-conservative!

$$Q^2 = \begin{bmatrix} 1/4 & 0 & 0 \\ 0 & 1/4 & 0 \\ 0 & 1/4 & 0 \end{bmatrix} \begin{bmatrix} 1/2 & 0 & 0 \\ 0 & 1/2 & 0 \\ 0 & 1/2 & 0 \end{bmatrix} = \begin{bmatrix} 1/8 & 0 & 0 \\ 0 & 1/8 & 0 \\ 0 & 1/8 & 0 \end{bmatrix}$$

Q) What will happen if we keep going?

A) Numbers will diminish to zero

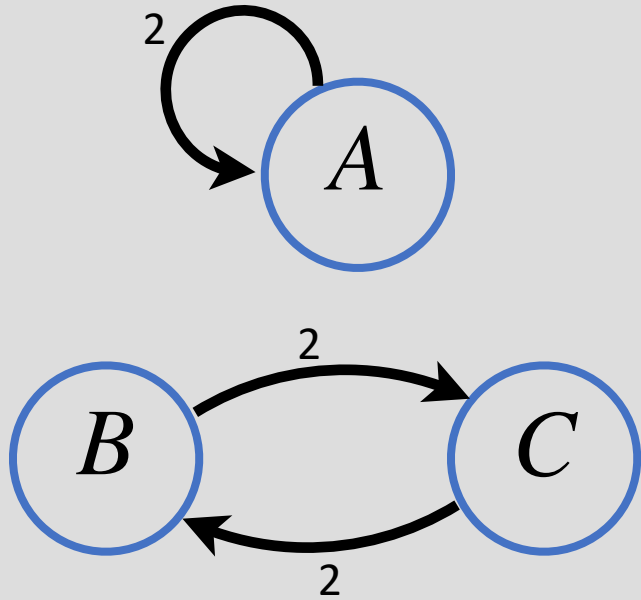
Google

DEAD SEA SOUTH
1984





State Transition Matrices



$$x(t+1) = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 0 & 2 \\ 0 & 2 & 0 \end{bmatrix} x(t)$$

$$\uparrow^2 \Rightarrow \begin{bmatrix} 2 & 0 & 0 \\ 0 & 0 & 2 \\ 0 & 2 & 0 \end{bmatrix} \begin{bmatrix} 2 & 0 & 0 \\ 0 & 0 & 2 \\ 0 & 2 & 0 \end{bmatrix} = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 4 \end{bmatrix}$$

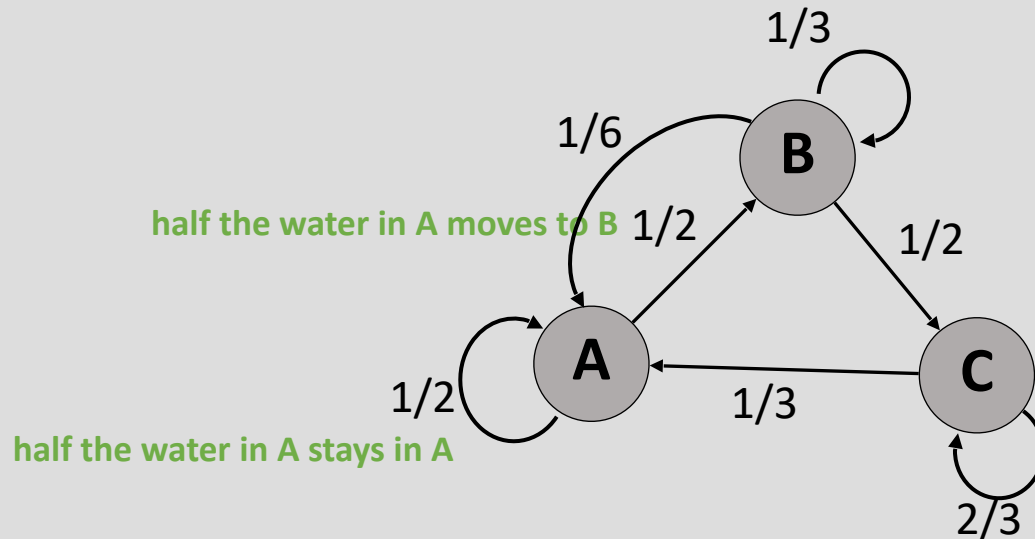
Q) What will happen if we keep going?

A) Numbers will explode to infinity



Graph Representation

Ex: Reservoirs and Pumps



Nodes

I have 3 reservoirs: A,B,C
and I want to keep track of how
much water is in each

When I turn on some pumps, water
moves between the reservoirs.

Where the water moves and what
fraction is represented by arrows.

Edge weights

Edges

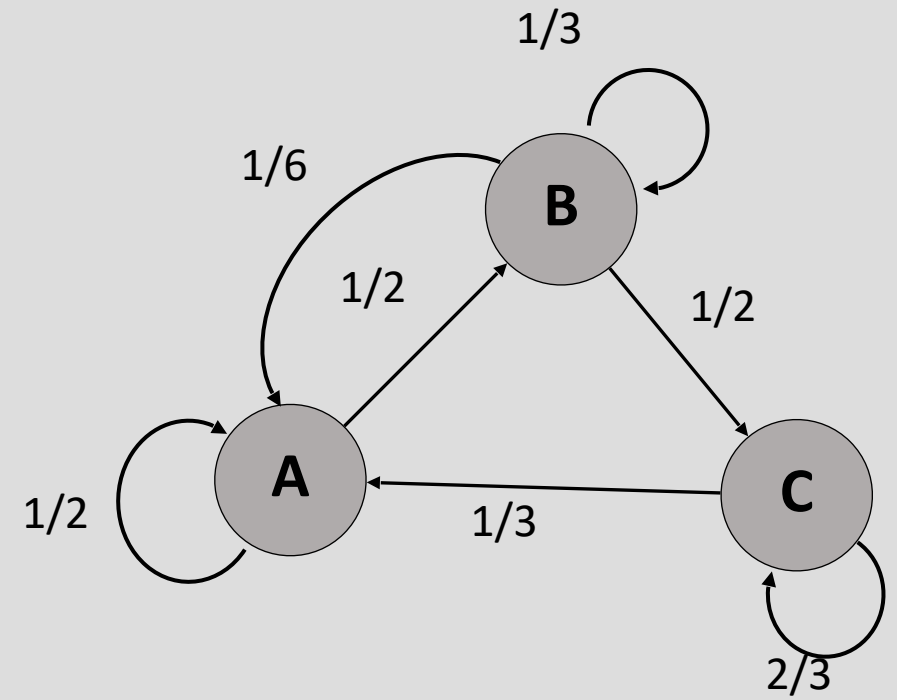
“directed” graph because
arrows have a direction

Where does the rest of the water in A go? Need to label that too...

Can you tell me how much water in each after pumps start? Need to know initial amounts

Exercise:

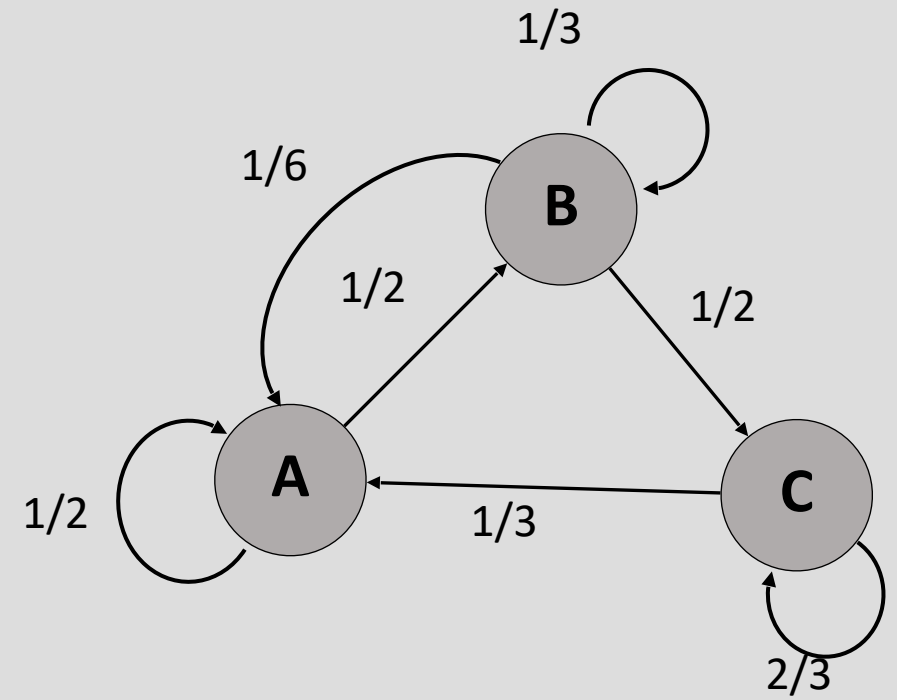
$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} A \rightarrow A & B \rightarrow A & C \rightarrow A \\ A \rightarrow B & B \rightarrow B & C \rightarrow B \\ A \rightarrow C & B \rightarrow C & C \rightarrow C \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$



Exercise:

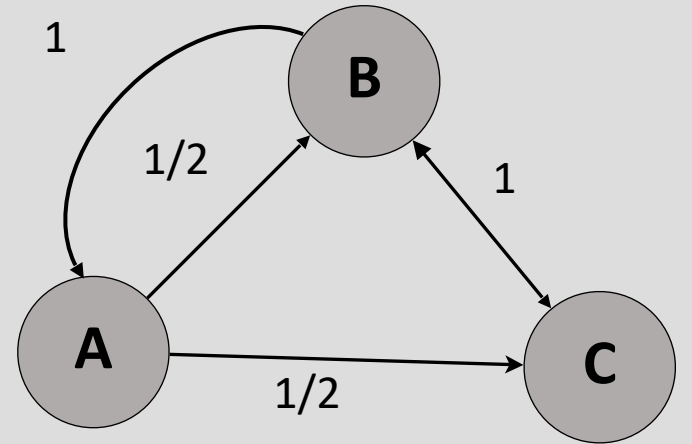
$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & \frac{1}{6} & \frac{1}{3} \\ \frac{1}{2} & \frac{1}{3} & 0 \\ 0 & \frac{1}{2} & \frac{2}{3} \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

The matrix above is a transition matrix with rows representing the next state and columns representing the current state. The entries are: $\frac{1}{2}$ (A→A), $\frac{1}{6}$ (B→A), $\frac{1}{3}$ (C→A) in the first row; $\frac{1}{2}$ (A→B), $\frac{1}{3}$ (B→B), 0 (C→B) in the second row; 0 (A→C), $\frac{1}{2}$ (B→C), $\frac{2}{3}$ (C→C) in the third row.



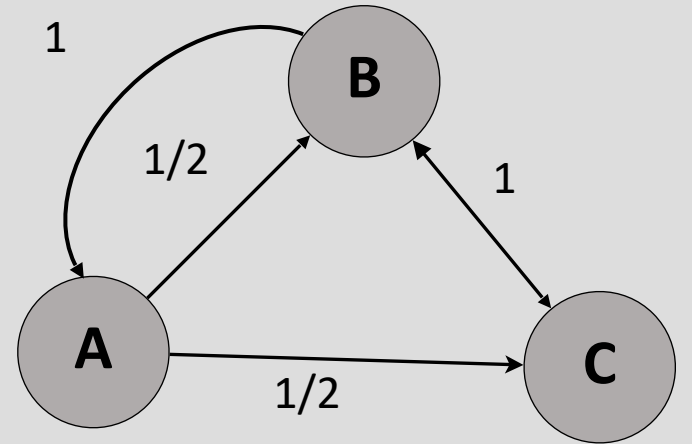
Example 2:

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} A \rightarrow A & B \rightarrow A & C \rightarrow A \\ A \rightarrow B & B \rightarrow B & C \rightarrow B \\ A \rightarrow C & B \rightarrow C & C \rightarrow C \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$



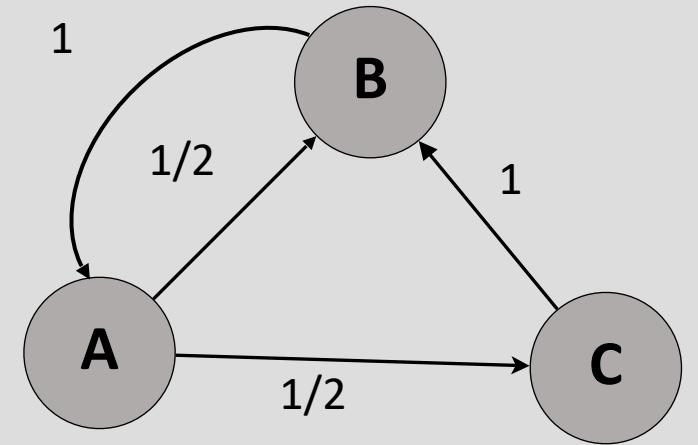
Example 2:

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} \overset{A \rightarrow A}{0} & \overset{B \rightarrow A}{1} & \overset{C \rightarrow A}{0} \\ \overset{A \rightarrow B}{\frac{1}{2}} & \overset{B \rightarrow B}{0} & \overset{C \rightarrow B}{1} \\ \overset{A \rightarrow C}{\frac{1}{2}} & \overset{B \rightarrow C}{0} & \overset{C \rightarrow C}{0} \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$



What about the reverse?

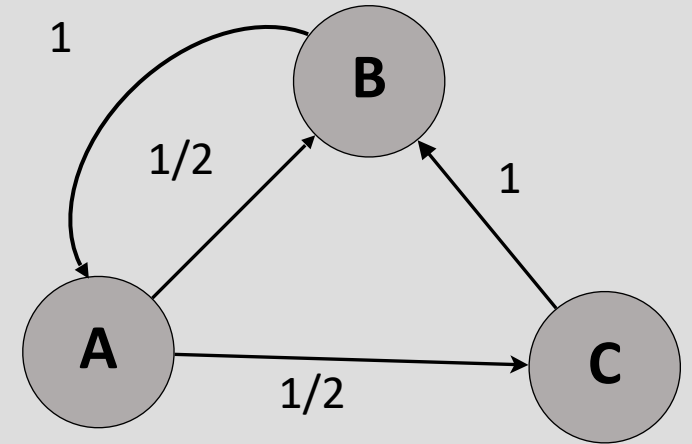
$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & 1 \\ \frac{1}{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$



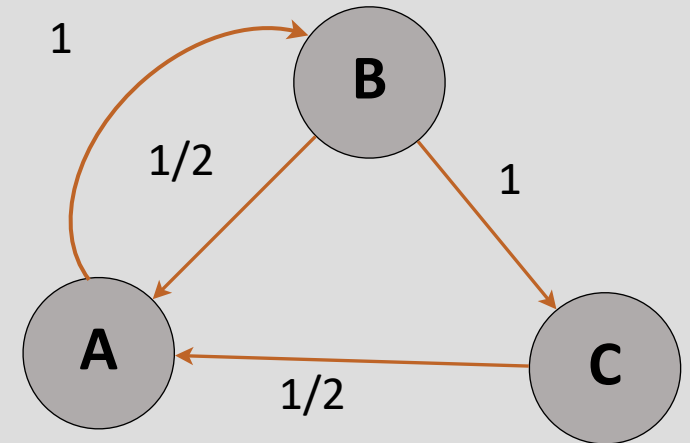
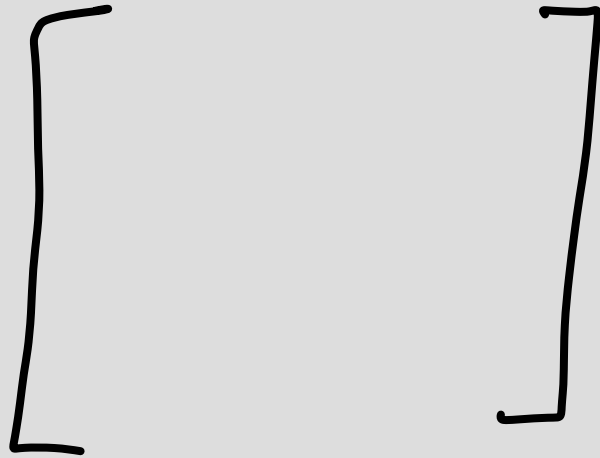
Q) Will flipping the arrows make us go back in time?

What about the reverse?

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & 1 \\ \frac{1}{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

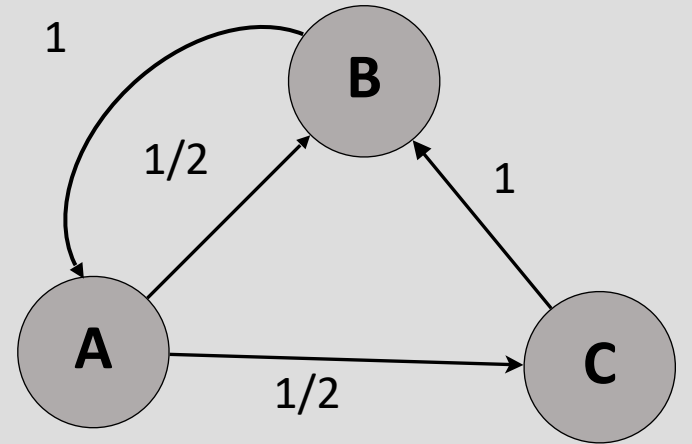


Q) Will flipping the arrows make us go back in time?



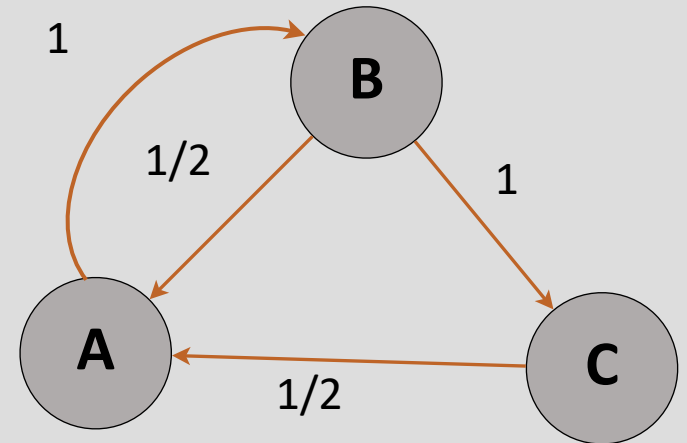
What about the reverse?

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & 1 \\ \frac{1}{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$



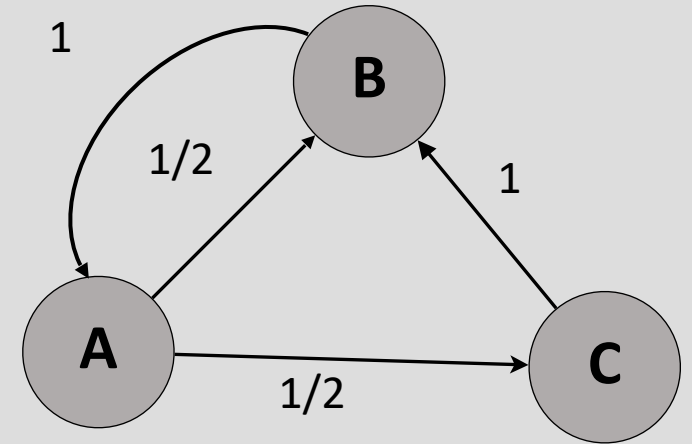
Q) Will flipping the arrows make us go back in time?

$$\begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$



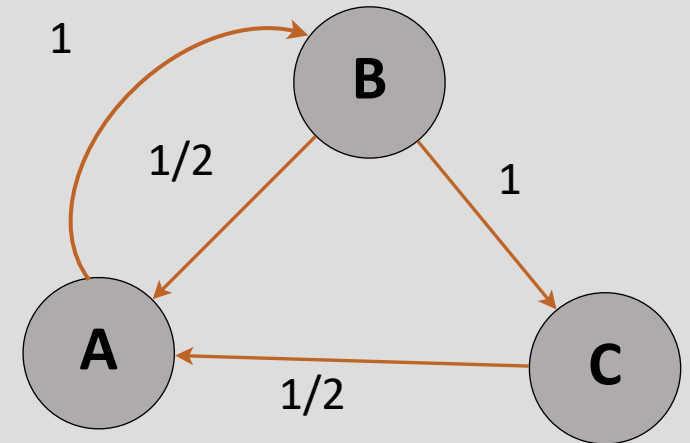
What about the reverse?

$$\begin{array}{l}
 6 \\
 10 \\
 2
 \end{array}
 \begin{bmatrix}
 x_A(t+1) \\
 x_B(t+1) \\
 x_C(t+1)
 \end{bmatrix}
 =
 \begin{bmatrix}
 0 & 1 & 0 \\
 \frac{1}{2} & 0 & 1 \\
 \frac{1}{2} & 0 & 0
 \end{bmatrix}
 \begin{bmatrix}
 x_A(t) \\
 x_B(t) \\
 x_C(t)
 \end{bmatrix}
 \begin{array}{l}
 4 \\
 6 \\
 8
 \end{array}$$



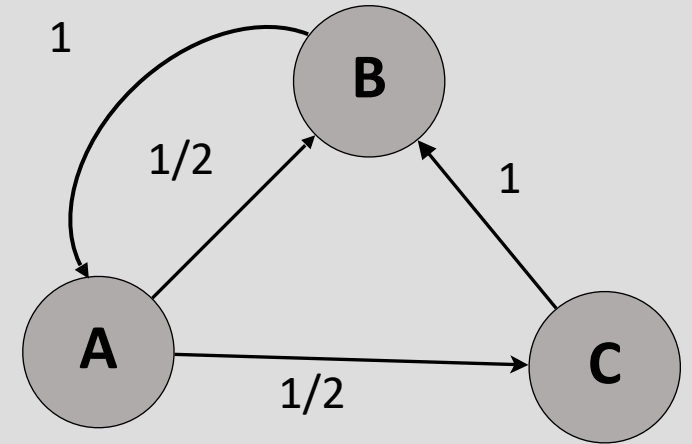
Q) Will flipping the arrows make us go back in time?

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



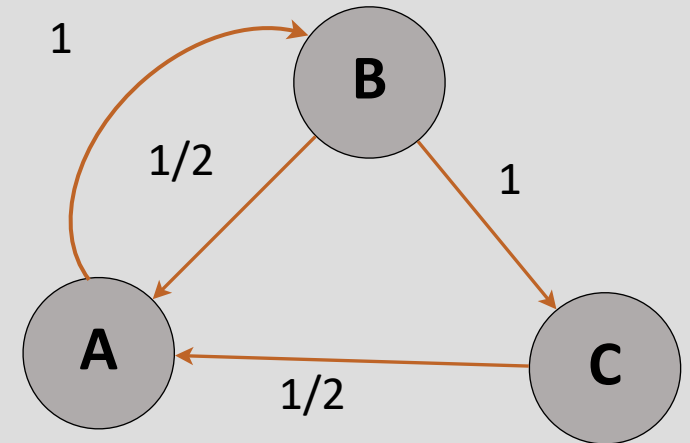
What about the reverse?

$$\begin{matrix} 6 \\ 10 \\ 2 \end{matrix} \begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & 1 \\ \frac{1}{2} & 0 & 0 \end{bmatrix} \begin{matrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{matrix} \begin{matrix} 4 \\ 6 \\ 8 \end{matrix}$$



Q) Will flipping the arrows make us go back in time?

$$\begin{bmatrix} 7 \\ 6 \\ 10 \end{bmatrix} = \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 6 \\ 10 \\ 2 \end{bmatrix}$$



A) In general, no!

Matrix Transpose

If the elements of the matrix $A \in \mathbb{R}^{N \times M}$ are a_{ij}

The elements of $A^T \in \mathbb{R}^{M \times N}$ are a_{ji}

Matrix transpose is not (generally) an inverse!

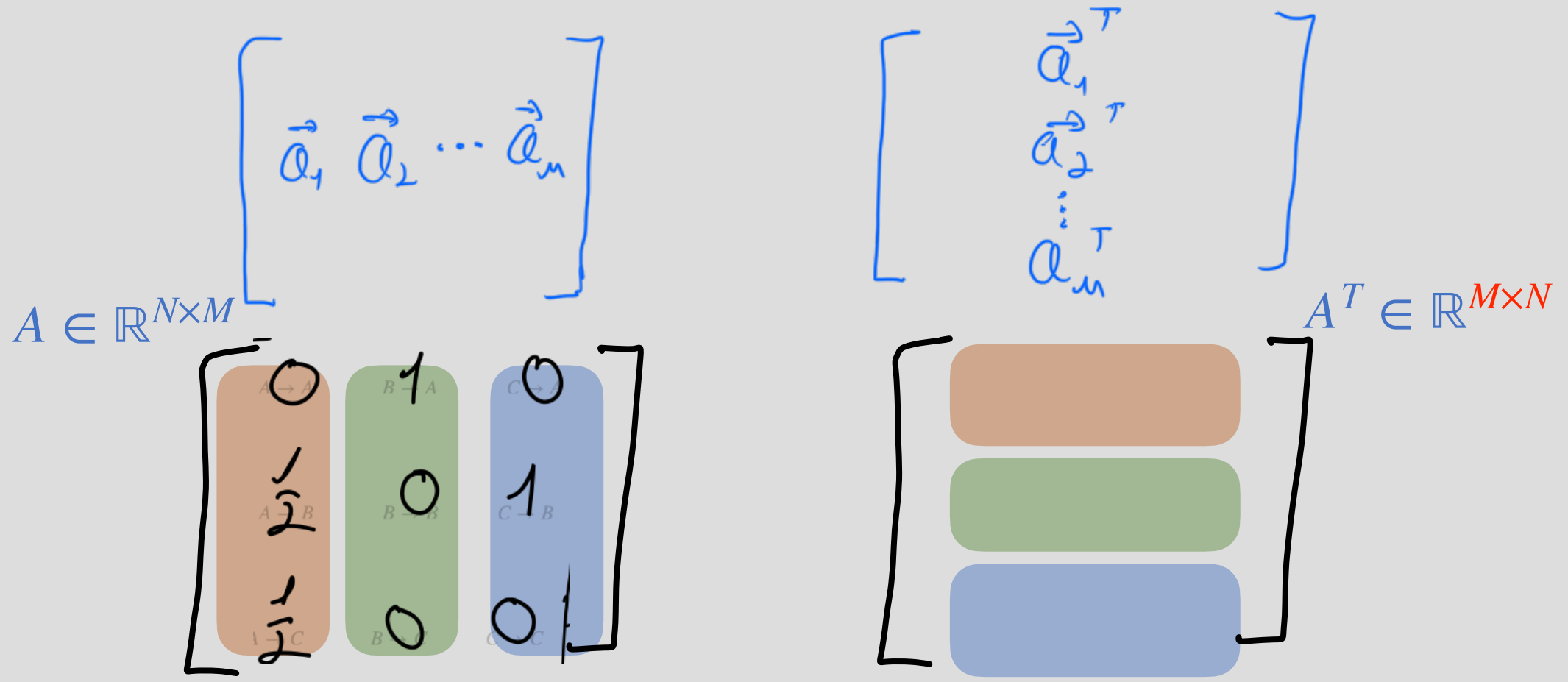
$$A \in \mathbb{R}^{N \times M} \left[\begin{array}{c} \vec{a}_1 \quad \vec{a}_2 \quad \dots \quad \vec{a}_m \end{array} \right] \quad \left[\begin{array}{c} \vec{a}_1^T \\ \vec{a}_2^T \\ \vdots \\ \vec{a}_m^T \end{array} \right] \quad A^T \in \mathbb{R}^{M \times N}$$

Matrix Transpose

If the elements of the matrix $A \in \mathbb{R}^{N \times M}$ are a_{ij}

The elements of $A^T \in \mathbb{R}^{M \times N}$ are a_{ji}

Matrix transpose is not (generally) an inverse!

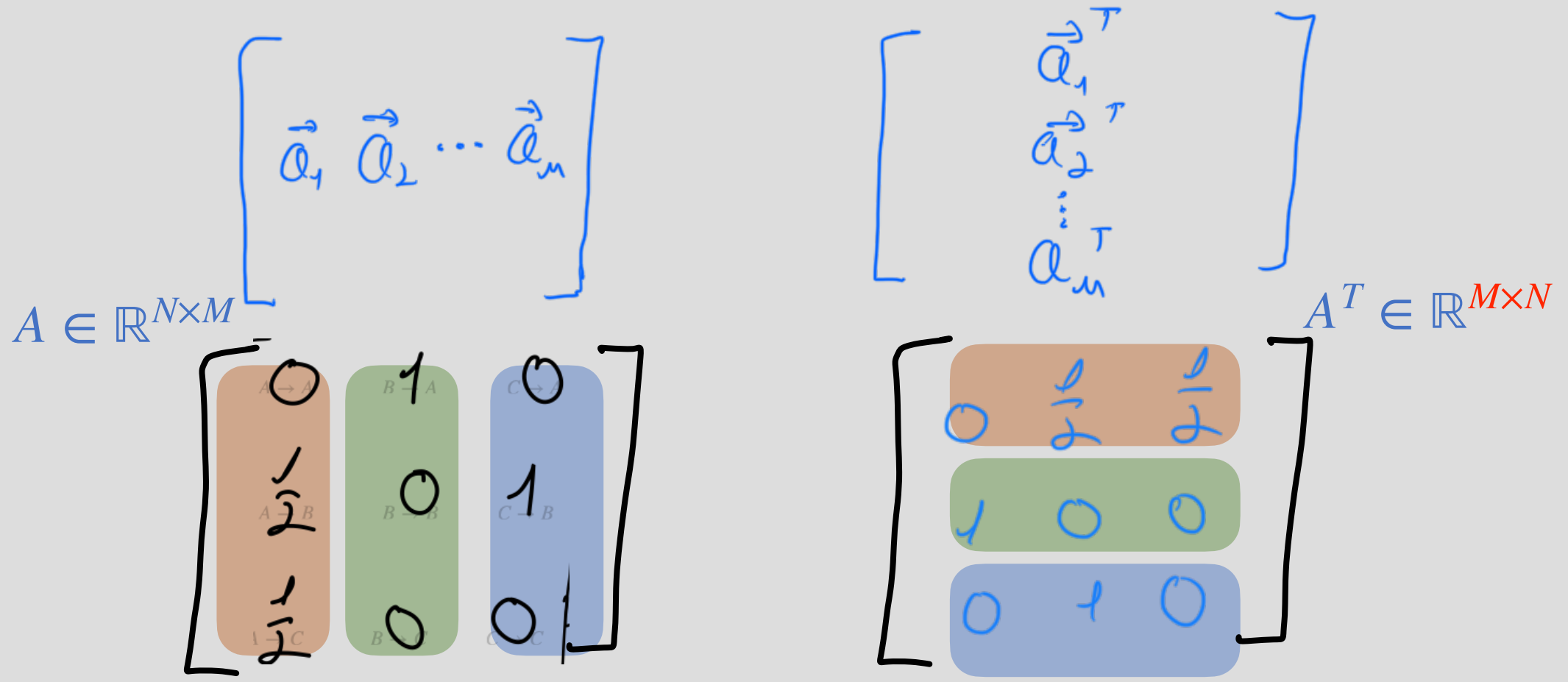


Matrix Transpose

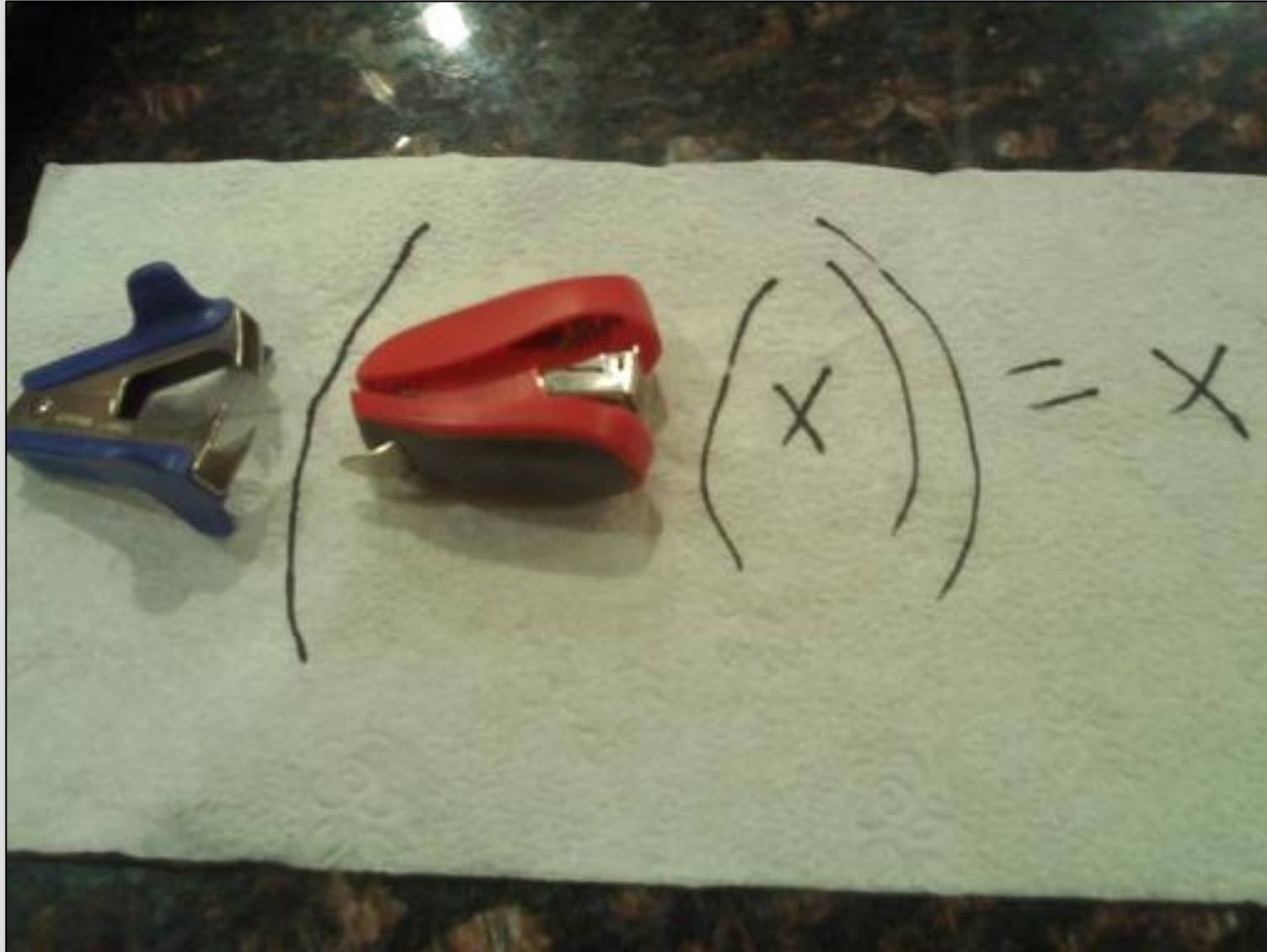
If the elements of the matrix $A \in \mathbb{R}^{N \times M}$ are a_{ij}

The elements of $A^T \in \mathbb{R}^{M \times N}$ are a_{ji}

Matrix transpose is not (generally) an inverse!



Matrix Inversion



Matrix Inverse

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

Is there a square matrix P such that we can go back in time?

$$\vec{x}(t) = P\vec{x}(t+1)$$

Yes, if : $PQ = I$

As consequence : $QP = I$

$$P\vec{x}(t+1) = PQ\vec{x}(t)$$

$$P\vec{x}(t+1) = I\vec{x}(t)$$

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

$$\vec{x}(t+1) = QP\vec{x}(t+1)$$

$$\vec{x}(t+1) = I\vec{x}(t+1)$$

Matrix Inverse - Formal definition

- Definition: Let $P, Q \in \mathbb{R}^{N \times N}$ be square matrices.
 - P is the inverse of Q if $PQ = QP = I$

We say that $P = Q^{-1}$ and $Q = P^{-1}$

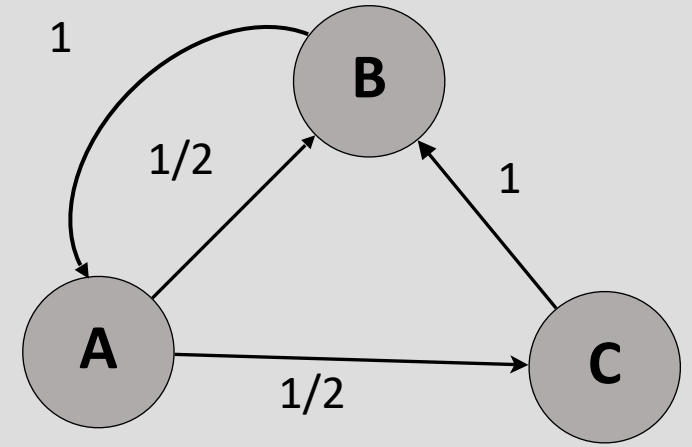
Q: What about non-square matrices?

A: EECS16B!

Computing the Matrix Inverse

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & 1 \\ \frac{1}{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

Q



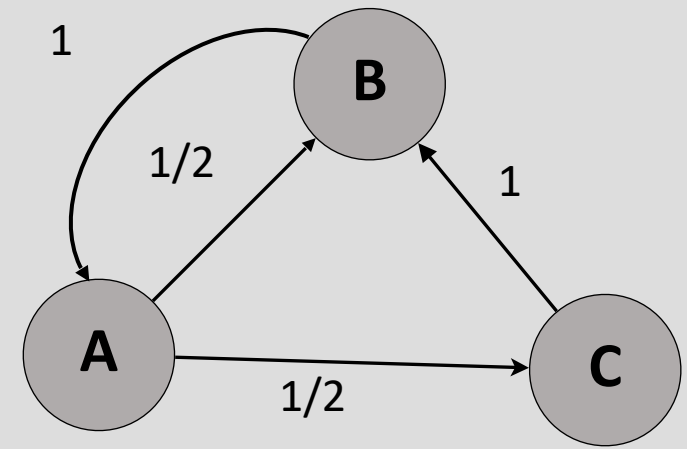
- Want $P = Q^{-1}$ such that $\vec{x}(t) = P\vec{x}(t + 1)$
 - Need: $QP = I$

Computing the Matrix Inverse

Need: $QP = I$

Pose as a linear set of equations.

Solve with Gaussian Elimination



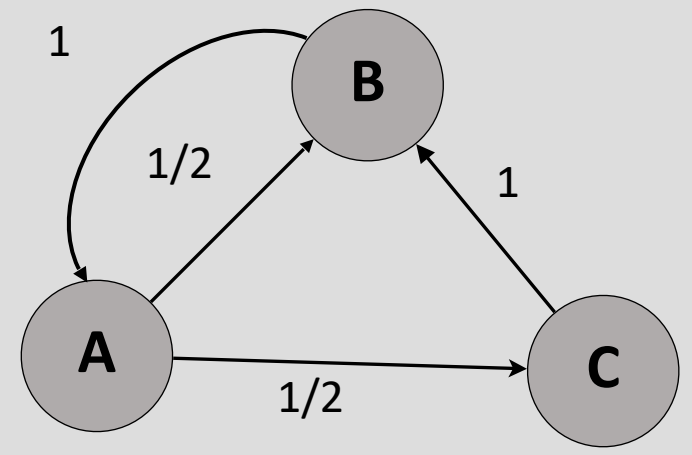
Computing the Matrix Inverse

Need: $QP = I$

Pose as a linear set of equations.

Solve with Gaussian Elimination

$$\begin{matrix} Q & & P \\ \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & 1 \\ \frac{1}{2} & 0 & 0 \end{bmatrix} & \begin{bmatrix} p_{11} & p_{12} & p_{13} \\ p_{21} & p_{22} & p_{23} \\ p_{31} & p_{32} & p_{33} \end{bmatrix} & = & \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ & \vec{p}_1 & & \vec{b}_1 \\ & & & I \end{matrix}$$



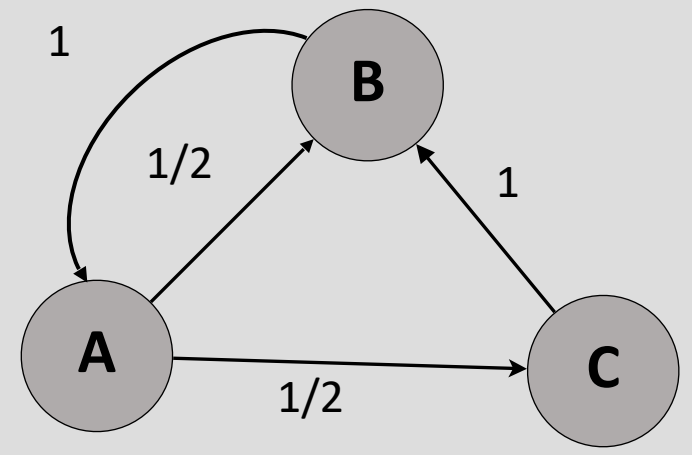
Computing the Matrix Inverse

Need: $QP = I$

Pose as a linear set of equations.

Solve with Gaussian Elimination

$$\begin{matrix} Q & P & \\ \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & 1 \\ \frac{1}{2} & 0 & 0 \end{bmatrix} & \begin{bmatrix} p_{11} & p_{12} & p_{13} \\ p_{21} & p_{22} & p_{23} \\ p_{31} & p_{32} & p_{33} \end{bmatrix} & = & \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ & \begin{matrix} \vec{p}_1 & \vec{p}_2 & \vec{p}_3 \end{matrix} & & \begin{matrix} \vec{b}_1 & \vec{b}_2 & \vec{b}_3 \end{matrix} \end{matrix}$$



Matrix Inverse via Gaussian Elimination

$$\begin{array}{c} \mathbf{Q} \\ \left[\begin{array}{ccc|ccc} 0 & 1 & 0 & 1 & 0 & 0 \\ \frac{1}{2} & 0 & 1 & 0 & 1 & 0 \\ \frac{1}{2} & 0 & 0 & 0 & 0 & 1 \end{array} \right] \end{array}$$

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

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

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

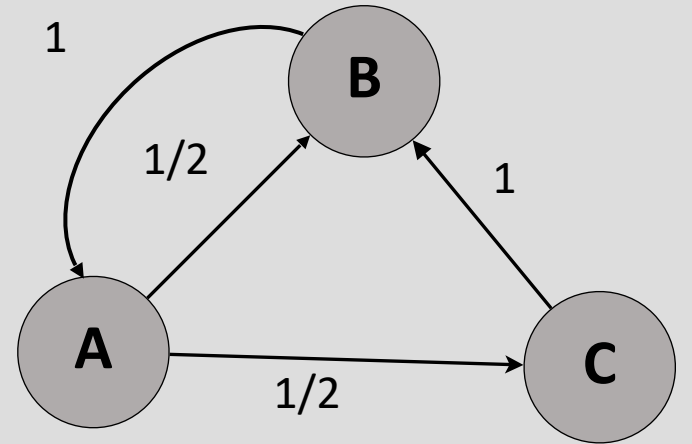
$$\left[\begin{array}{ccc|ccc} 1 & 0 & 2 & 0 & 2 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ \frac{1}{2} & 0 & 0 & 0 & 0 & 1 \end{array} \right]$$

$$\mathbf{I} \quad \mathbf{P}$$

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

Let's check

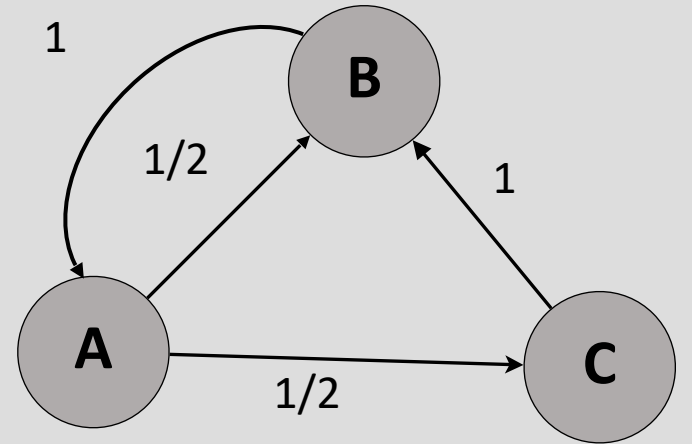
$$\begin{array}{l}
 6 \\
 10 \\
 2
 \end{array}
 \begin{bmatrix}
 x_A(t+1) \\
 x_B(t+1) \\
 x_C(t+1)
 \end{bmatrix}
 =
 \begin{bmatrix}
 0 & 1 & 0 \\
 1/2 & 0 & 1 \\
 1/2 & 0 & 0
 \end{bmatrix}
 \begin{bmatrix}
 x_A(t) \\
 x_B(t) \\
 x_C(t)
 \end{bmatrix}
 \begin{array}{l}
 4 \\
 6 \\
 8
 \end{array}$$



$$\begin{bmatrix}
 \\
 \\
 \\
 \end{bmatrix}
 =
 \begin{bmatrix}
 0 & 0 & 2 \\
 1 & 0 & 0 \\
 0 & 1 & -1
 \end{bmatrix}
 \begin{bmatrix}
 6 \\
 10 \\
 2
 \end{bmatrix}$$

Let's check

$$\begin{matrix} 6 \\ 10 \\ 2 \end{matrix} \begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & 1 \\ \frac{1}{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix} \begin{matrix} 4 \\ 6 \\ 8 \end{matrix}$$



$$\begin{bmatrix} 4 \\ 6 \\ 8 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 2 \\ 1 & 0 & 0 \\ 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} 6 \\ 10 \\ 2 \end{bmatrix}$$

And now we can take any number of steps backwards!

Can we always invert a function?

• Can we always invert a function $f^{-1}(f(\vec{x})) = \vec{x}$?

- $f(x) = x^2$?

- $f(x) = ax$?

- $f(x) = Ax$?

Invertibility of Linear Transformations

- Theorem: A is invertible, if and only if (iff) the columns of A are linearly independent.
 1. If columns of A are lin. dep. then A^{-1} does not exist
 2. If A^{-1} exists, then the cols. of A are linearly independent

Proof concept: Assume linear dependence and invertibility and show that it is a contradiction

From linear independence: $\exists \vec{\alpha} \neq 0$ such that $A\vec{\alpha} = 0$

Assume A^{-1} exists

$$\begin{aligned} A\vec{\alpha} &= 0 \\ A^{-1}A\vec{\alpha} &= A^{-1}0 \\ I\vec{\alpha} &= 0 \end{aligned}$$

But $\vec{\alpha} \neq 0$! Hence A^{-1} does not exist

Inverse of a 2x2 matrix

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

1. Flip a and d
2. Negate b and c
3. Divide by $ad - bc$

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

Derive via Gauss Elimination!