

# Welcome to EECS 16A!

## Designing Information Devices and Systems I



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Sp 2022

Lecture 3A  
Matrix xForms



# Announcements

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

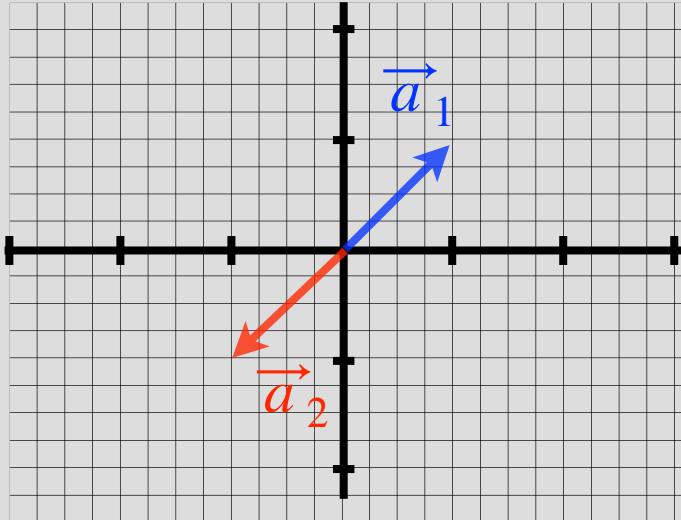


# Linear Dependence

Recall:

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

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



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

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

Department of  
Redundancy  
Department

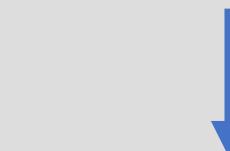
# Linear Dependence

- 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?

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

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  $\alpha_i = 0$
- 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?

$$\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} \pi \\ \sqrt{2} \end{bmatrix} \right\}$$

span =  $\mathbb{R}^2$

linearly dependent!

$$\in \mathbb{R}^2$$

# 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  $\mathbf{S} \triangleq \{\circ, \bullet\}$ ,  $\tau \triangleq \{\langle \bullet, \bullet \rangle, \langle \bullet, \circ \rangle, \langle \circ, \circ \rangle\}$  so that  $\mathcal{M}_\tau = \{\langle i, \lambda \ell \cdot \bullet \rangle, \langle j, \lambda \ell \cdot \circ \rangle, \langle k, \lambda \ell \cdot (\ell < m ? \bullet \& \circ) \rangle\}$ . We let  $\mathcal{X} \triangleq \{\langle i, \sigma \rangle \mid \forall j < i : \sigma_j = \bullet\}$  so that  $\neg FD(\mathcal{X})$ . We have  $\mathcal{M}_{\tau \downarrow \bullet} = \{\langle i, \lambda \ell \cdot \bullet \rangle, \langle k, \lambda \ell \cdot (\ell < m ? \bullet \& \circ) \rangle \mid k < m\}$ ,  $\mathcal{M}_{\tau \downarrow \circ} = \{\langle j, \lambda \ell \cdot \circ \rangle, \langle k, \lambda \ell \cdot (\ell < m ? \bullet \& \circ) \rangle \mid k \geq m\}$  and  $\oplus\{\mathcal{X}\} = \{\langle i, \sigma \rangle \mid \forall j \leq i : \sigma_j = \bullet\}$ . We have  $\alpha_{\mathcal{M}_\tau}^*(\oplus\{\mathcal{X}\}) = \{s \mid \mathcal{M}_{\tau \downarrow s} \subseteq \oplus\{\mathcal{X}\}\} = \{\bullet\}$  whereas  $\widetilde{pre}[\tau](\alpha_{\mathcal{M}_\tau}^*(\mathcal{X})) = \widetilde{pre}[\tau](\{s \mid \mathcal{M}_{\tau \downarrow s} \subseteq \mathcal{X}\}) = \widetilde{pre}[\tau](\{\bullet\}) = \{s \mid \forall s' : t(s, s') \Rightarrow s' = \bullet\} = \emptyset$  since  $t(s, \bullet)$  implies  $s = \bullet$  and  $t(\bullet, \circ)$  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

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

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From linear dependence Def 2:

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

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

$$\Rightarrow A \vec{x}^\dagger = A(\vec{x}^* + \vec{\alpha}) = A \vec{x}^* + A \vec{\alpha} = \vec{b} + 0 \quad \text{So } \vec{x}^\dagger \text{ is another solution!}$$

# Matrix Transformations

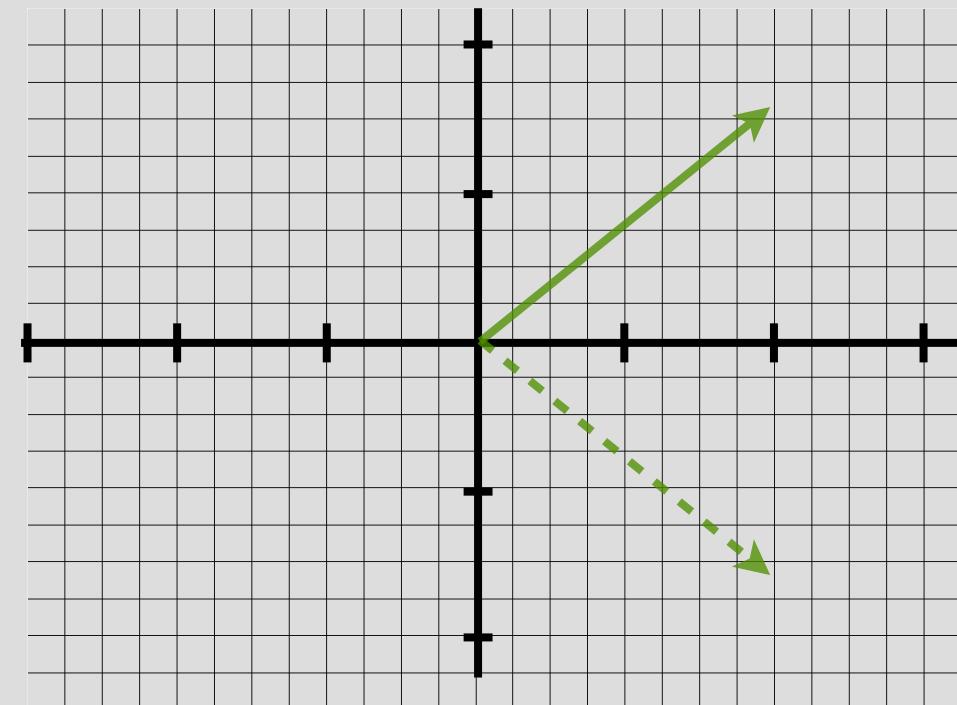
$$\begin{bmatrix} \cos 90^\circ & \sin 90^\circ \\ -\sin 90^\circ & \cos 90^\circ \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

# Matrices are operators that transform vectors

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

Example:

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



# Matrices are operators that transform vectors

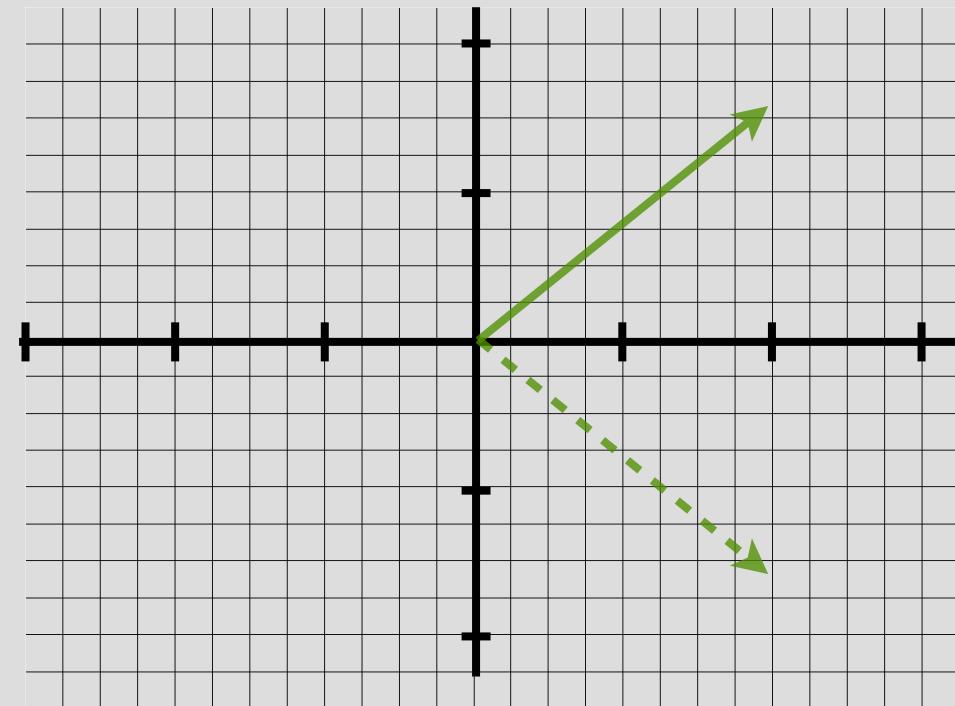
$$A \xrightarrow{\quad} \vec{x} = b$$

Example:

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



[https://www.youtube.com/watch?v=LhF\\_56SxrGk](https://www.youtube.com/watch?v=LhF_56SxrGk)



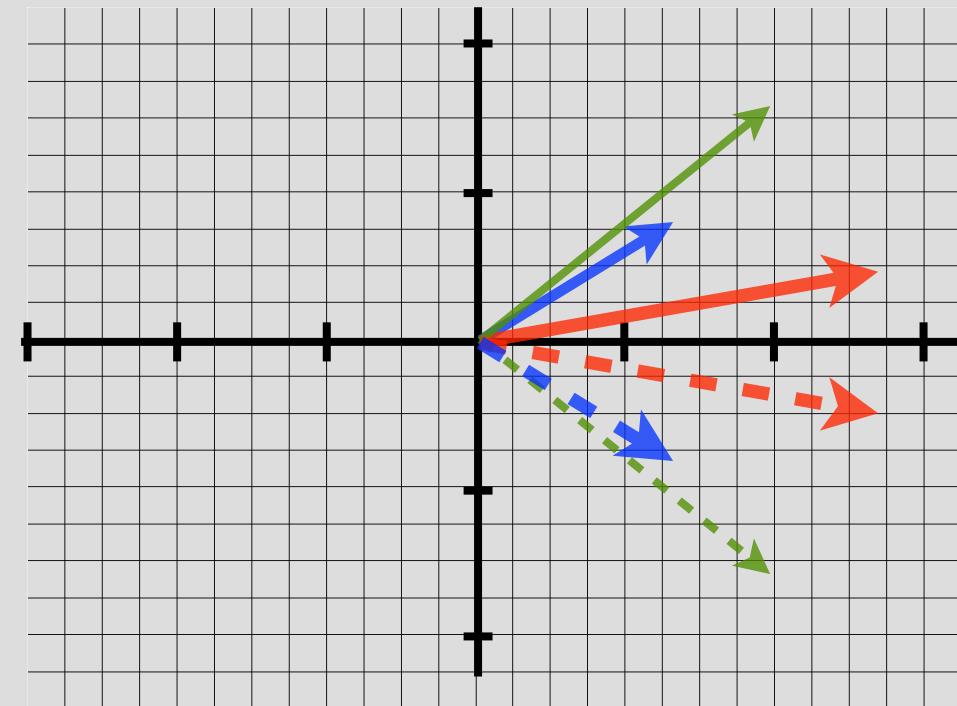
# Matrices are operators that transform vectors

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

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!



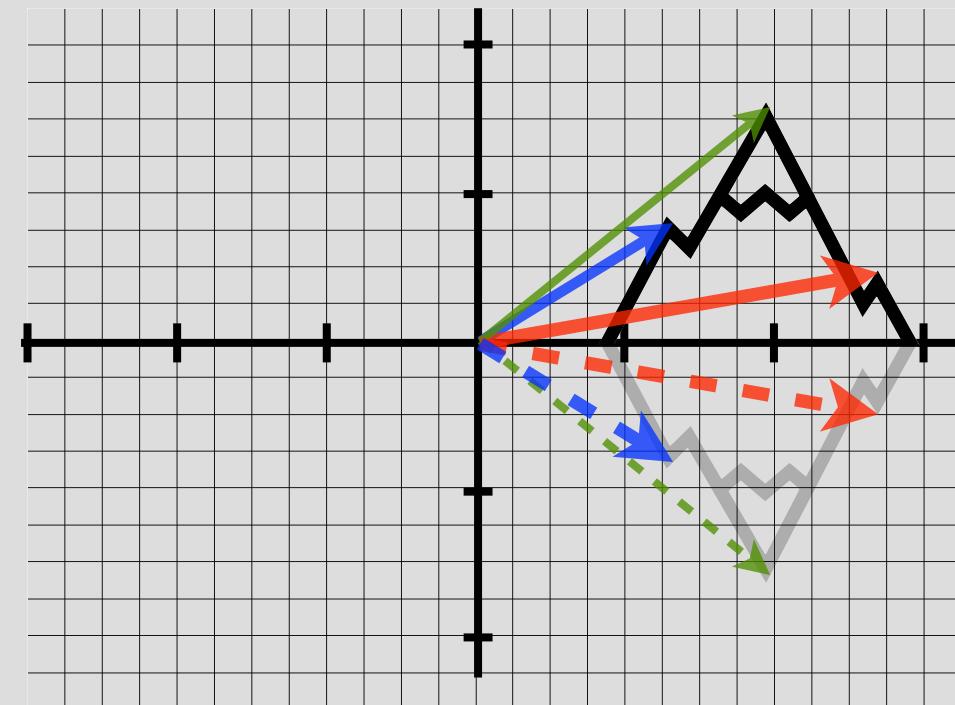
# Matrices are operators that transform vectors

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

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

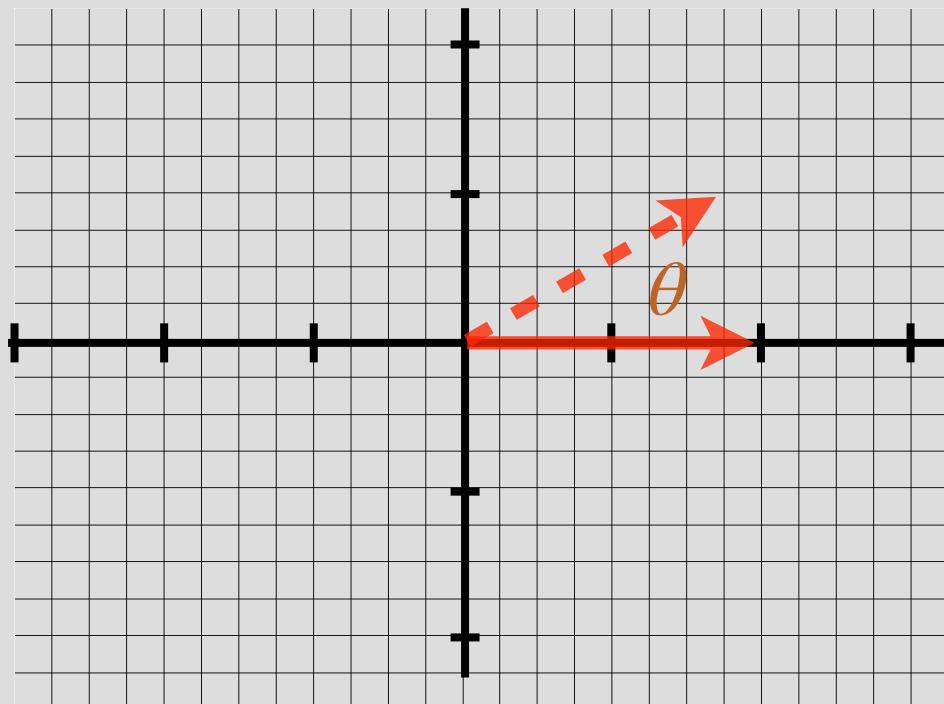
Example 2:

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

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

Rotation Matrix!

$$\begin{bmatrix} \cos 90^\circ & \sin 90^\circ \\ -\sin 90^\circ & \cos 90^\circ \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$



# 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

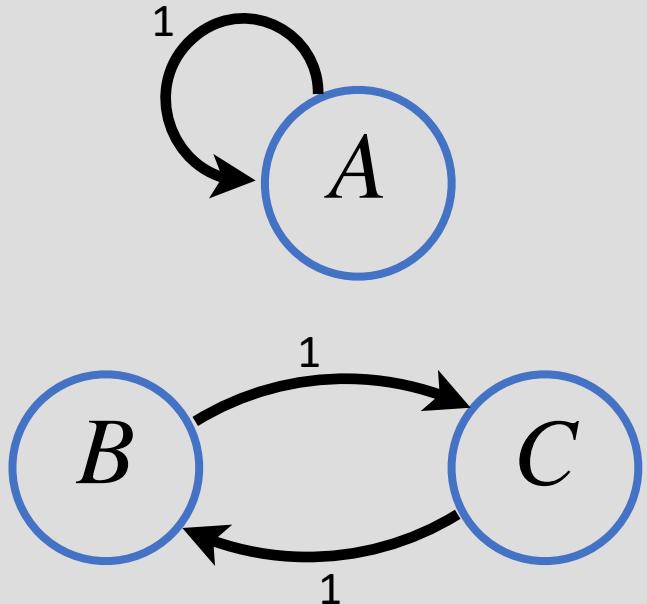
$$\vec{s}(t) = \begin{bmatrix} x(t) \\ y(t) \\ v(t) \\ \theta(t) \end{bmatrix} \quad \left. \begin{array}{l} \text{position} \\ \text{velocity} \end{array} \right\}$$

Q: Is that enough?

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

# Graph Transition Matrices

Example: Reservoirs and Pumps



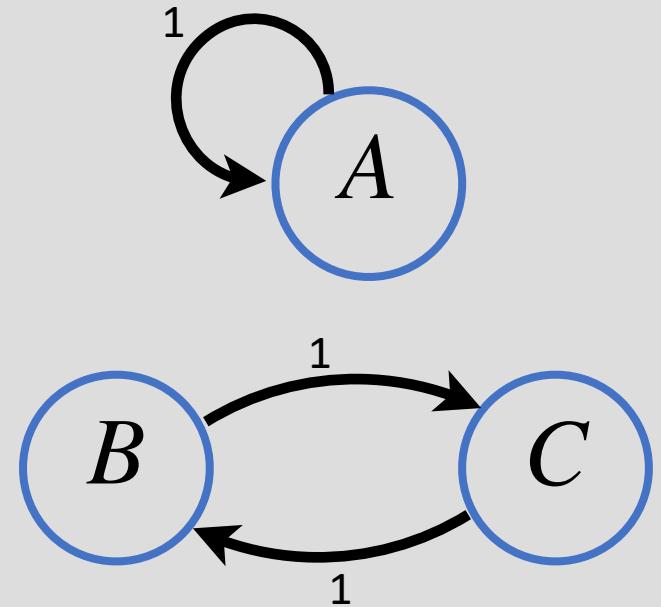
Q: What is the state?

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

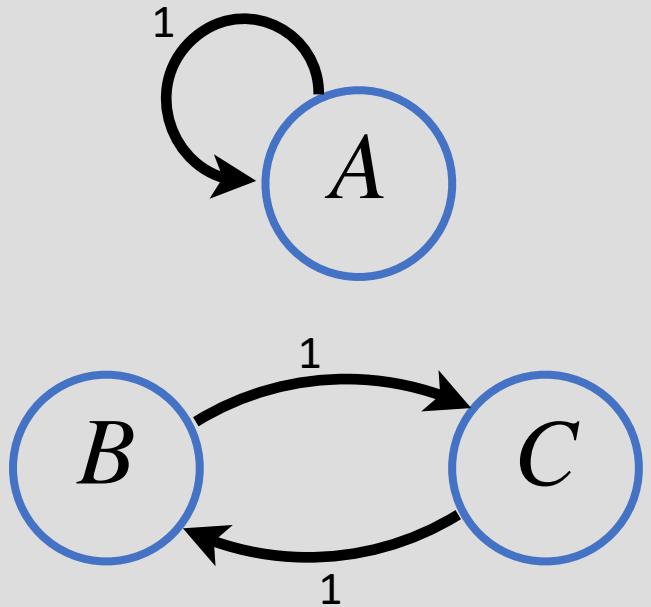
$$x_A(t+1) = x_A(t)$$

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

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

Write as a matrix-vector multiplication:

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



# State Transition Matrices

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

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

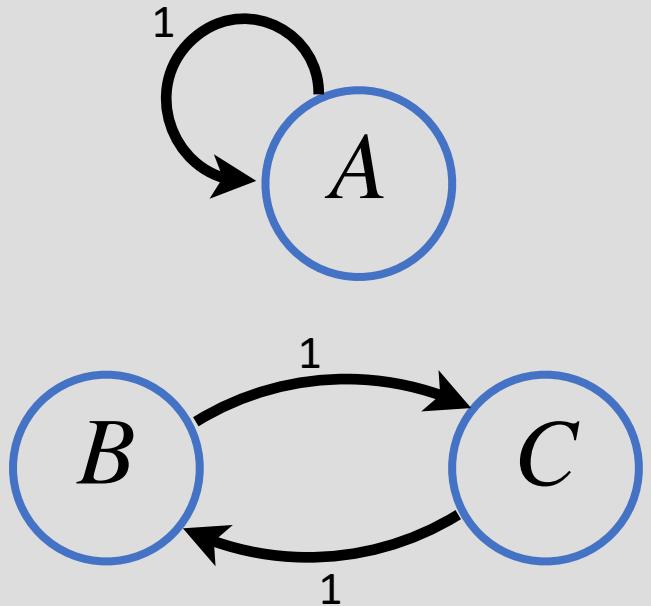
$$x_C(t+1) = x_B(t)$$

Write as a matrix-vector multiplication:

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix} \quad \text{or } \vec{x}(t+1) = Q \vec{x}(t)$$

What is the state after 2 times?

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

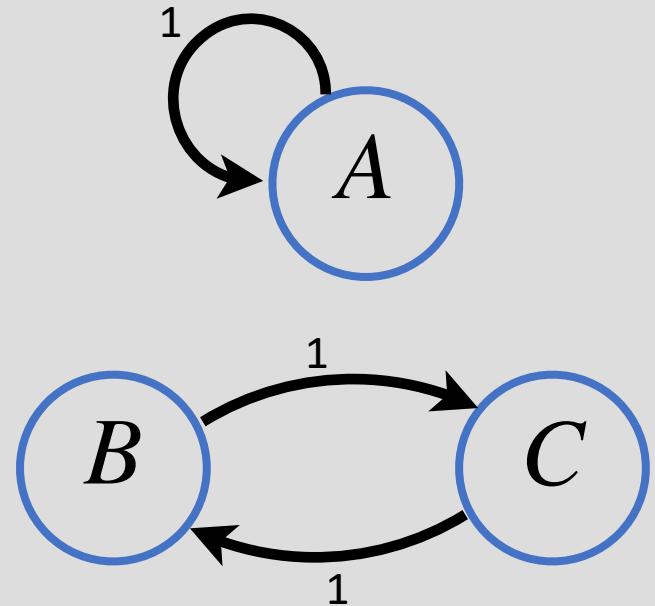


# State Transition Matrices

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

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

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

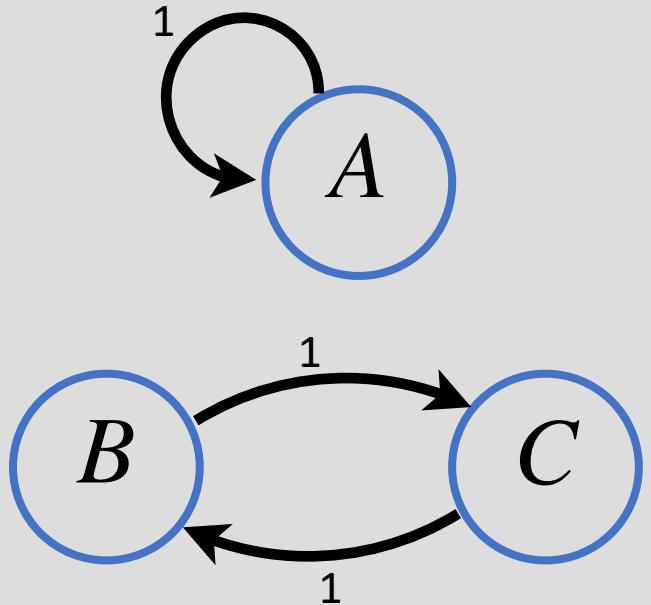


# State Transition Matrices

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

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

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

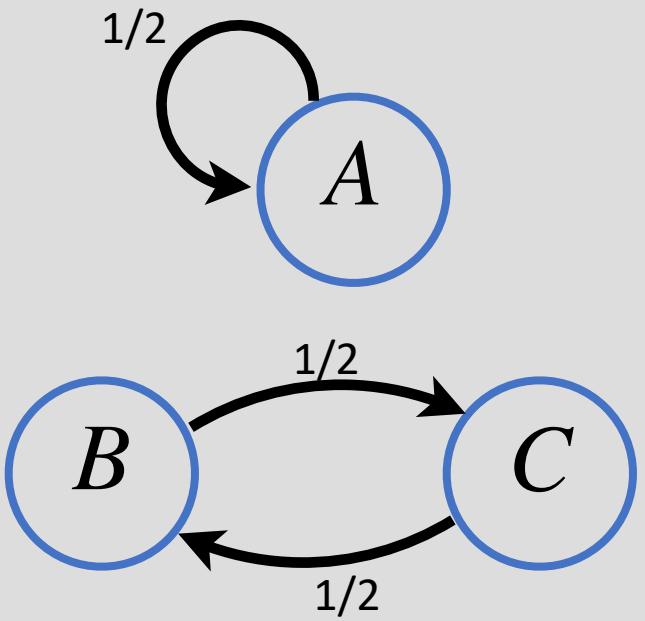


①  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$

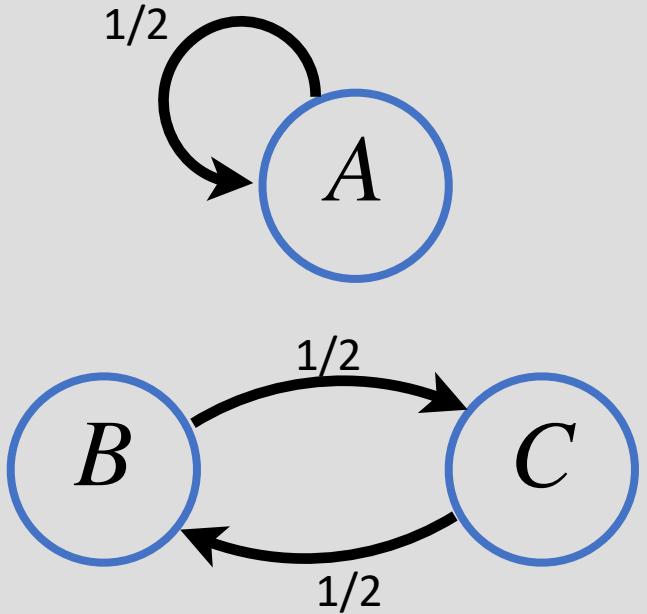
②  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$

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

# State Transition Matrices



# State Transition Matrices



$$\mathcal{X}[t+1] = \begin{bmatrix} \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} \mathcal{X}(t)$$

Non-conservative!

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

Q) What will happen if we keep going?

A) Numbers will diminish to zero

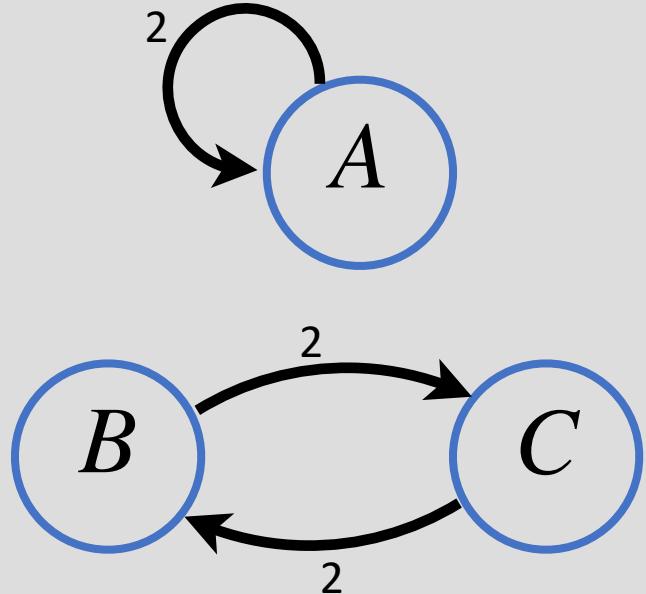
Google

DEAD SEA SOUTH  
1984





# State Transition Matrices



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

$$\tilde{x} = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 0 & 2 \\ 0 & 2 & 0 \end{bmatrix} \begin{bmatrix} 2 & 0 & 0 \\ 0 & 0 & 2 \\ 0 & 2 & 0 \end{bmatrix} = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 4 \end{bmatrix}$$

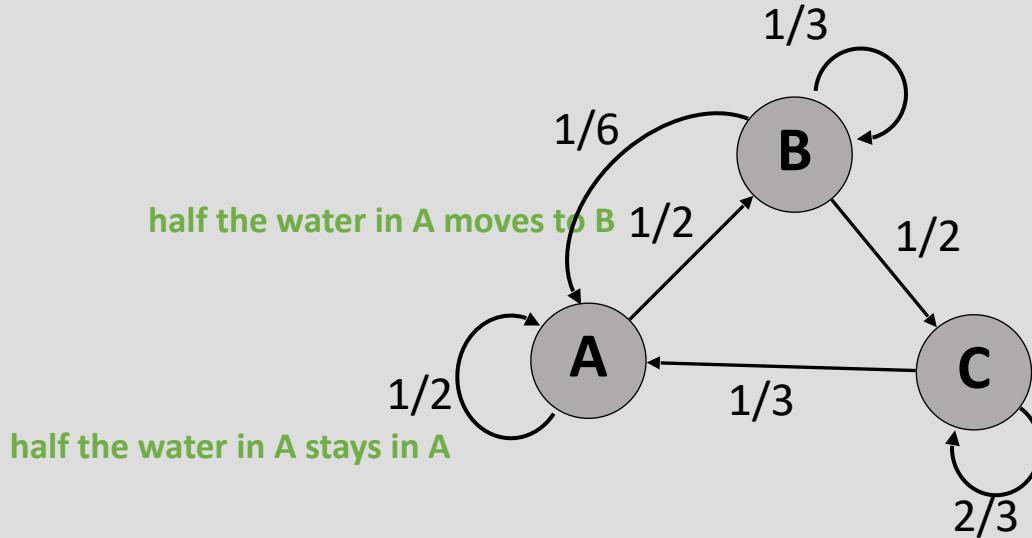
Q) What will happen if we keep going?

A) Numbers will explode to infinity



# Graph Representation

Ex: Reservoirs and Pumps



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

## 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.

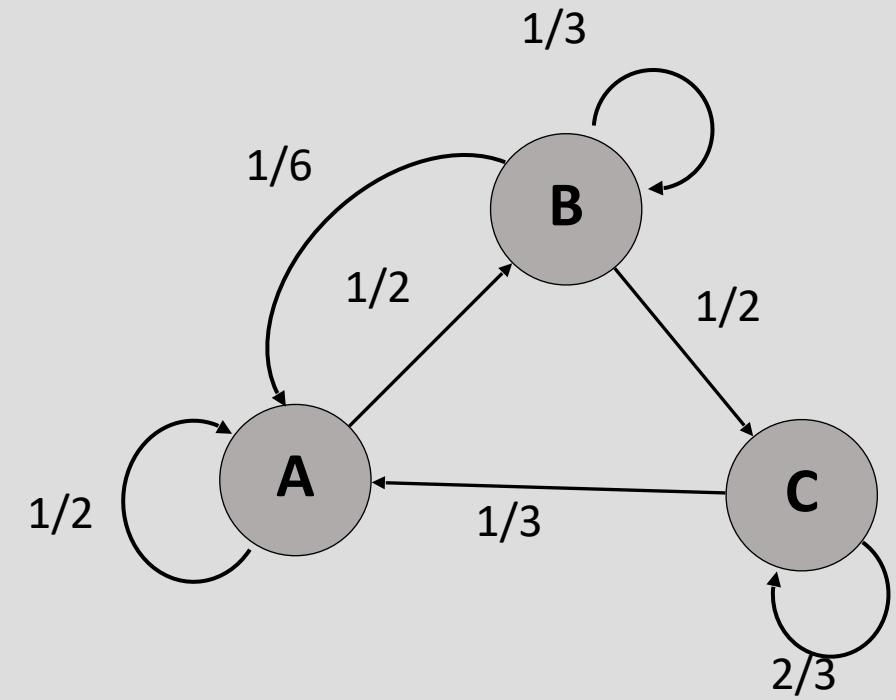
## Edge weights

## Edges

“directed” graph because  
arrows have a direction

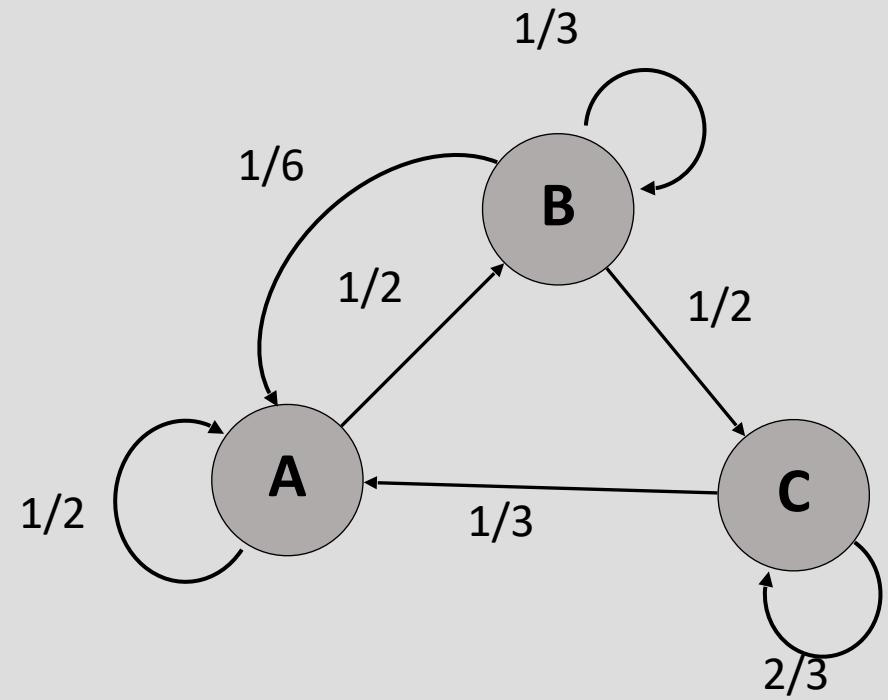
## Exercise:

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



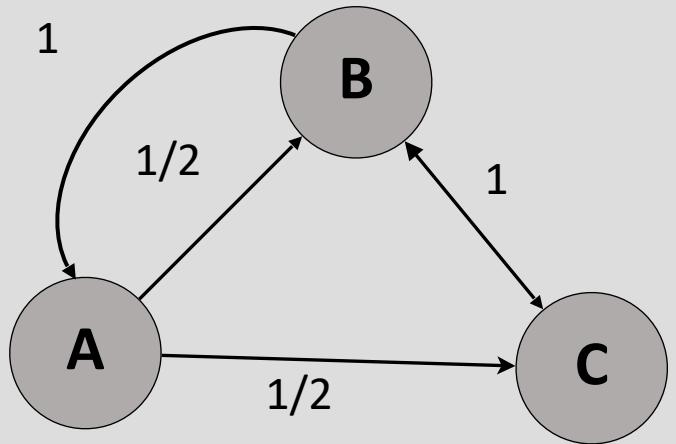
## Exercise:

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



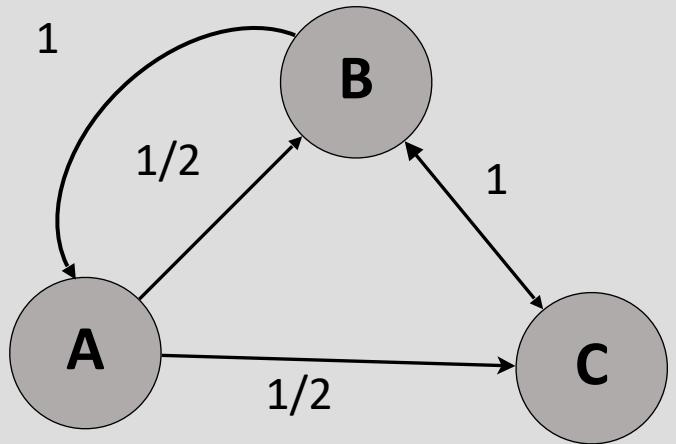
## Example 2:

$$\begin{bmatrix} \chi_A(t+1) \\ \chi_B(t+1) \\ \chi_C(t+1) \end{bmatrix} = \begin{bmatrix} A \rightarrow A & B \rightarrow A & C \rightarrow A \\ A \rightarrow B & B \rightarrow B & C \rightarrow B \\ A \rightarrow C & B \rightarrow C & C \rightarrow C \end{bmatrix} \begin{bmatrix} \chi_A(t) \\ \chi_B(t) \\ \chi_C(t) \end{bmatrix}$$



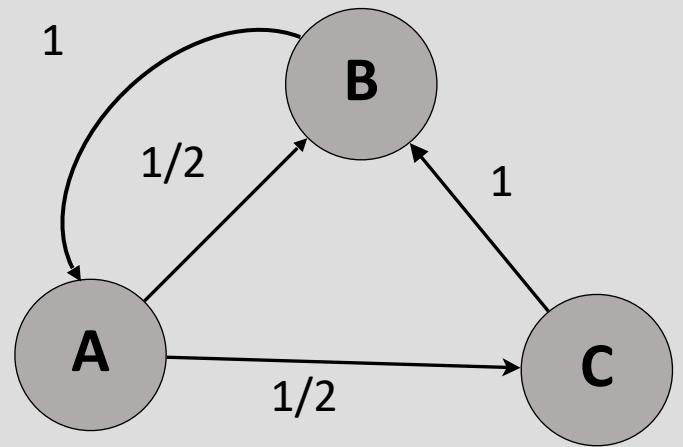
## Example 2:

$$\begin{bmatrix} x_{c_A}(t+1) \\ x_{c_B}(t+1) \\ x_{c_C}(t+1) \end{bmatrix} = \begin{bmatrix} A \xrightarrow{1} A & B \xrightarrow{1} A & C \xrightarrow{1} A \\ A \xrightarrow{1/2} B & B \xrightarrow{1} B & C \xrightarrow{1} B \\ A \xrightarrow{1/2} C & B \xrightarrow{1} C & C \xrightarrow{1} C \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$



# What about the reverse?

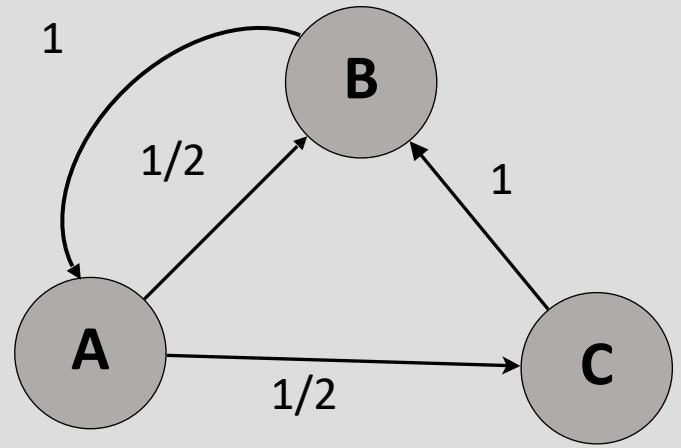
$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} A \rightarrow A & B \rightarrow A & C \rightarrow A \\ A \rightarrow B & B \rightarrow B & C \rightarrow B \\ A \rightarrow C & B \rightarrow C & C \rightarrow C \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t)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



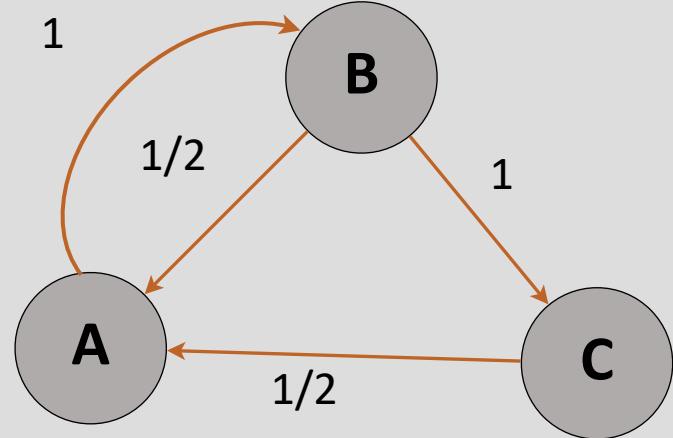
Q) Will flipping the arrows make us go back in time?

# What about the reverse?

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

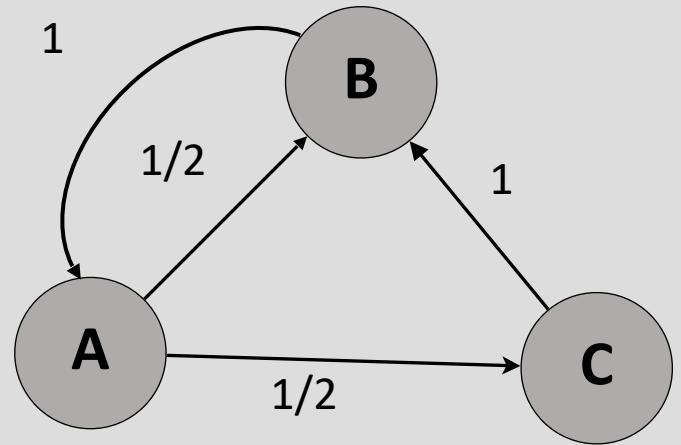


Q) Will flipping the arrows make us go back in time?



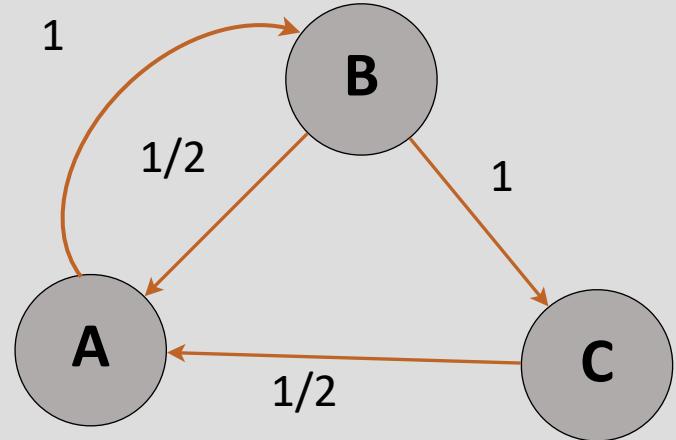
# What about the reverse?

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



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

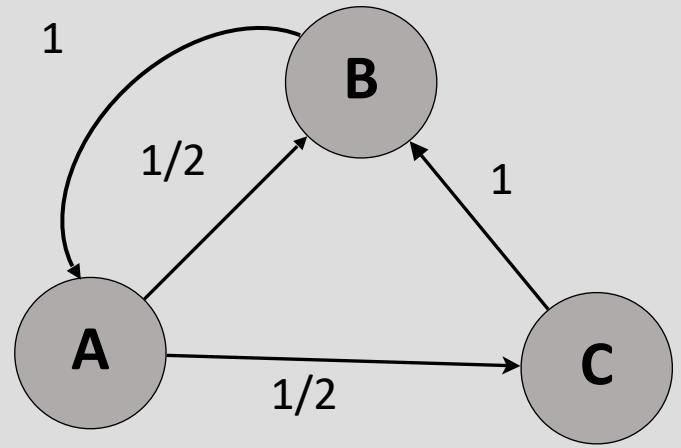
$$\begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$



# What about the reverse?

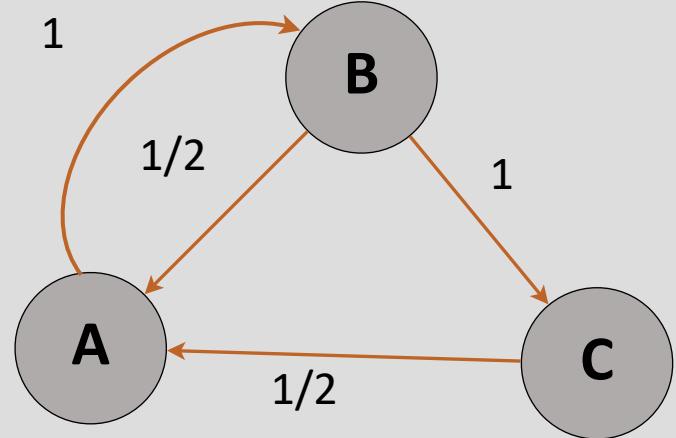
$$\begin{bmatrix} 6 & x_A(t+1) \\ 10 & x_B(t+1) \\ 2 & x_C(t+1) \end{bmatrix} = \begin{bmatrix} A \rightarrow A & B \rightarrow A & C \rightarrow A \\ A \rightarrow B & B \rightarrow B & C \rightarrow B \\ A \rightarrow C & B \rightarrow C & C \rightarrow C \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

4  
6  
8



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

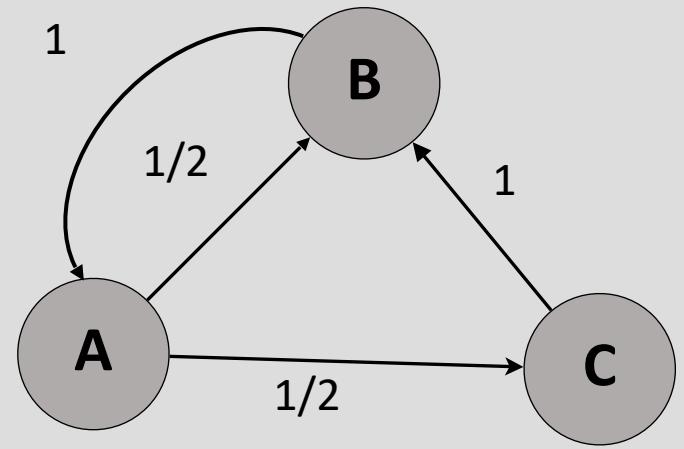
$$\begin{bmatrix} ] \\ ] \end{bmatrix} = \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} ] \\ ] \end{bmatrix}$$



# What about the reverse?

$$\begin{bmatrix} 6 \\ 10 \\ 2 \end{bmatrix} \begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} A \rightarrow A & B \rightarrow A & C \rightarrow A \\ A \rightarrow B & B \rightarrow B & C \rightarrow B \\ A \rightarrow C & B \rightarrow C & C \rightarrow C \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

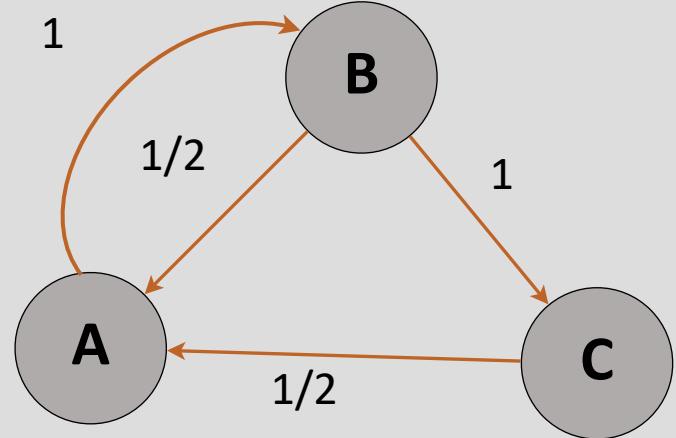
4  
6  
8



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

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

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!

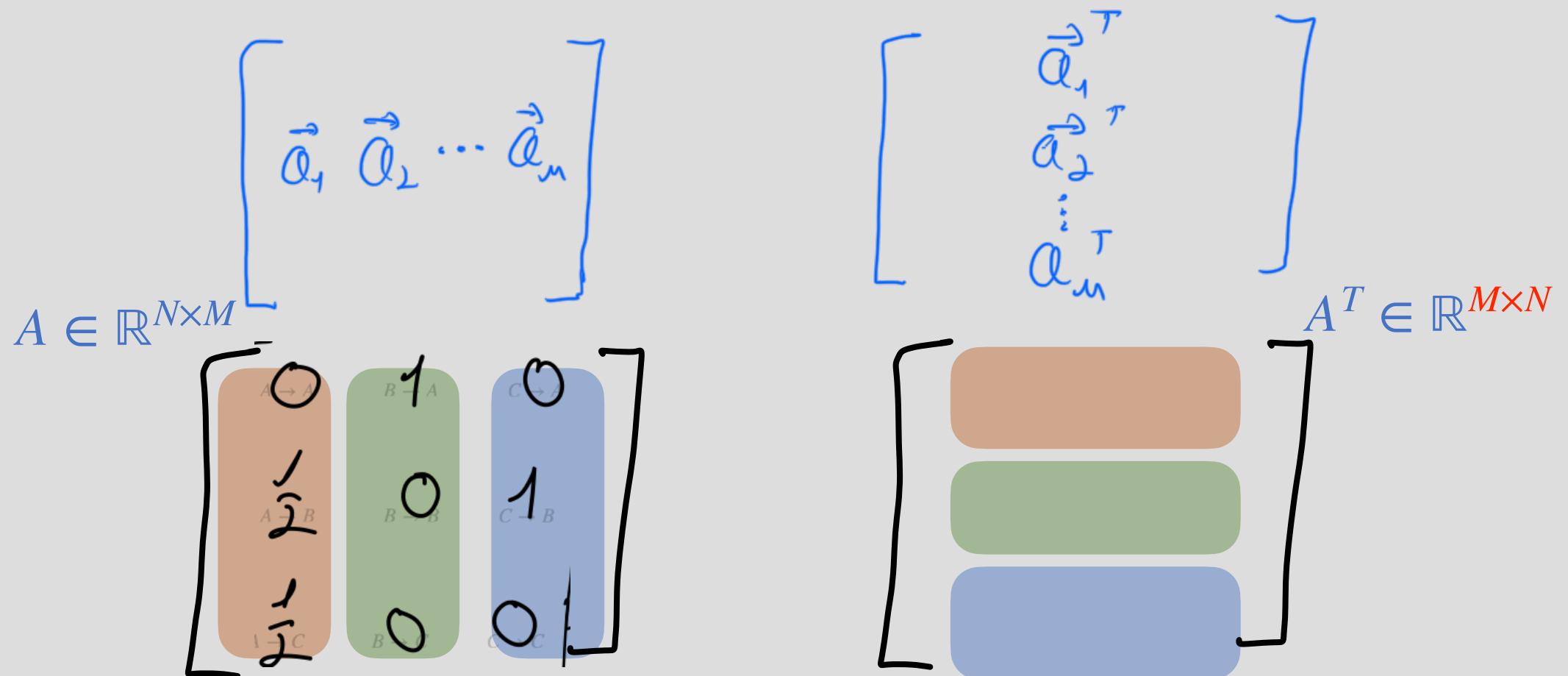
$$A \in \mathbb{R}^{N \times M} \quad \left[ \begin{array}{c} \vec{a}_1 \vec{a}_2 \cdots \vec{a}_m \\ \vdots \end{array} \right] \quad A^T \in \mathbb{R}^{M \times N} \quad \left[ \begin{array}{c} \vec{a}_1^T \\ \vec{a}_2^T \\ \vdots \\ \vec{a}_m^T \end{array} \right]$$

# 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!

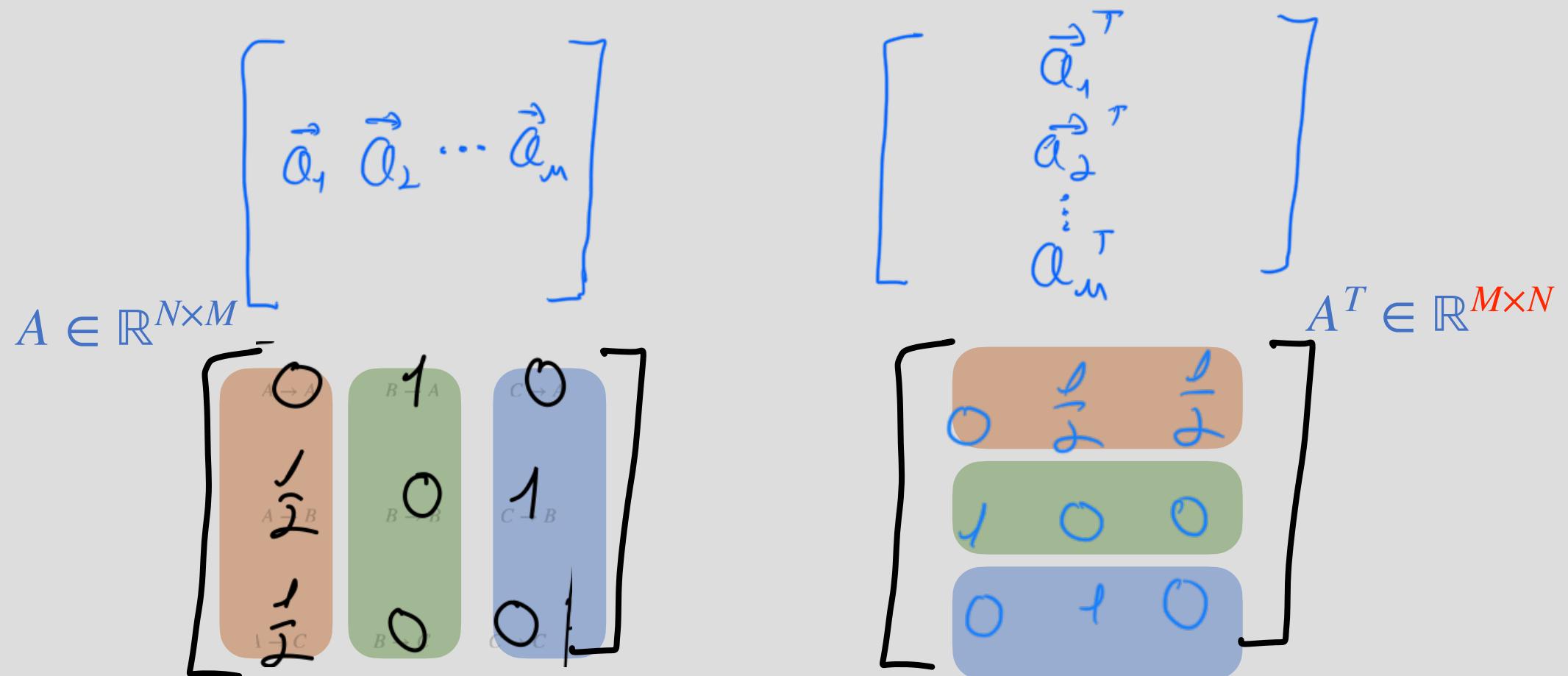


# Matrix Transpose

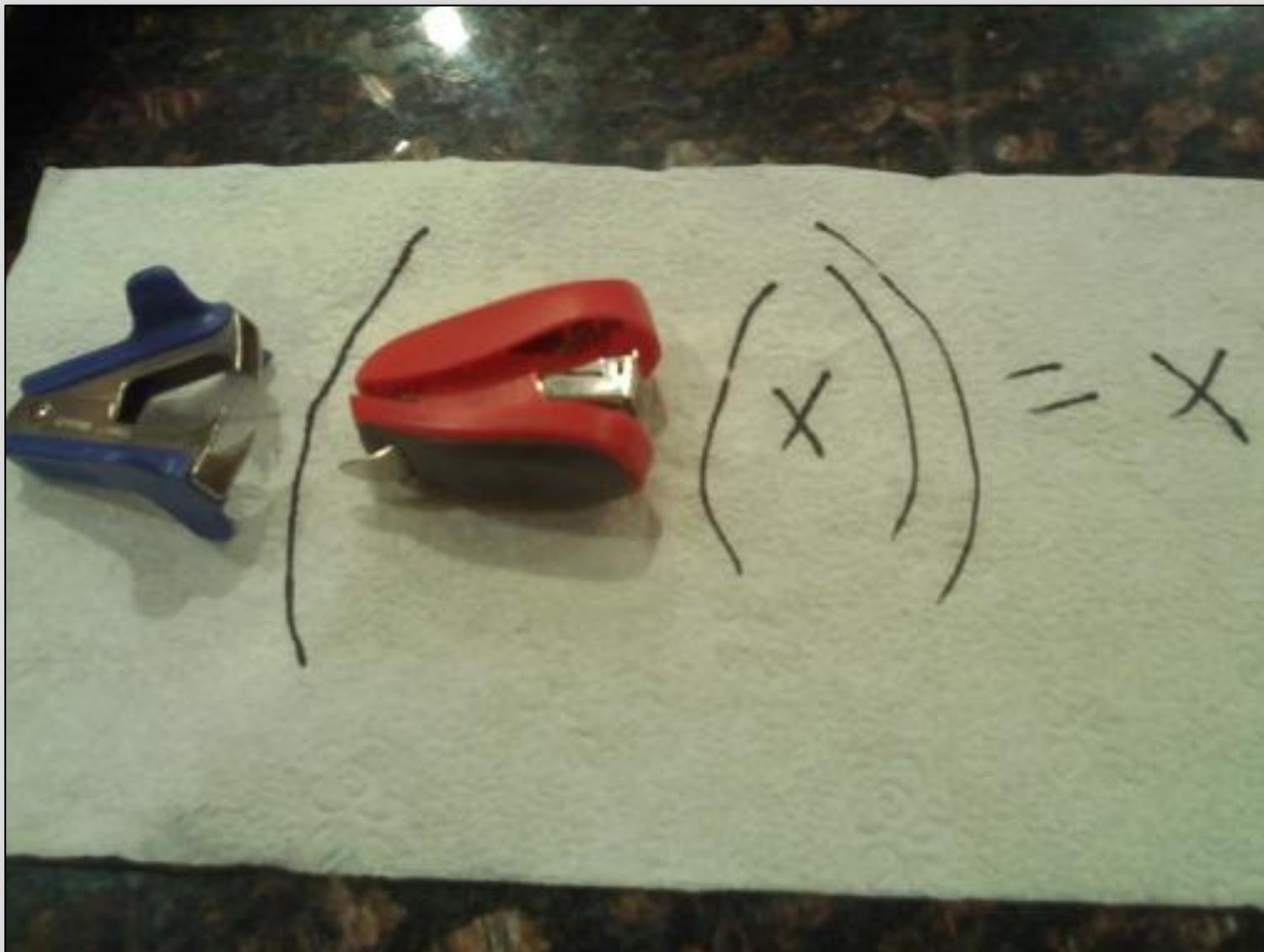
If the elements of the matrix  $A \in \mathbb{R}^{N \times M}$  are  $a_{ij}$

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 :  $QP = I$

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

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

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

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

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

# 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?

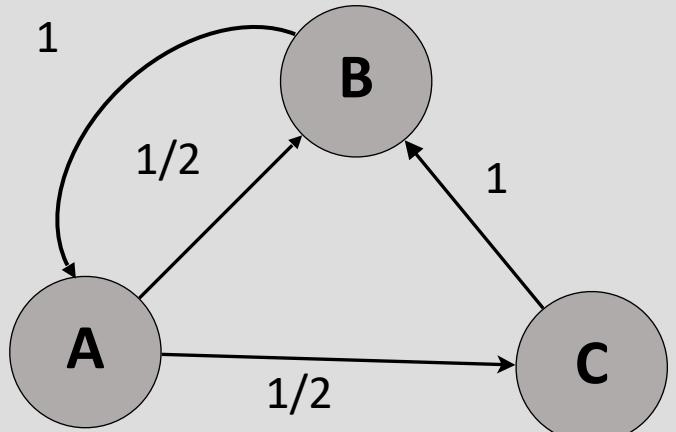
A: EECS16B!

# Computing the Matrix Inverse

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

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

$A \rightarrow A \quad B \rightarrow A \quad C \rightarrow A$   
 $A \rightarrow B \quad B \rightarrow B \quad C \rightarrow B$   
 $A \rightarrow C \quad B \rightarrow C \quad C \rightarrow C$



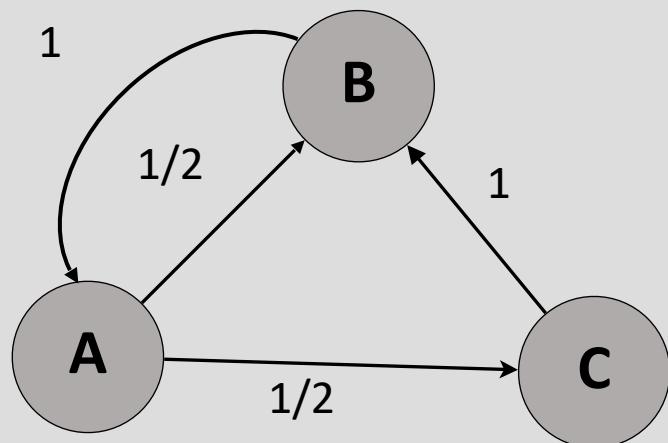
- Want  $P = Q^{-1}$  such that  $\vec{x}(t) = P\vec{x}(t+1)$ 
  - Need:  $QP = I$

# Computing the Matrix Inverse

Need:  $QP = I$

Pose as a linear set of equations.

Solve with Gaussian Elimination



# Computing the Matrix Inverse

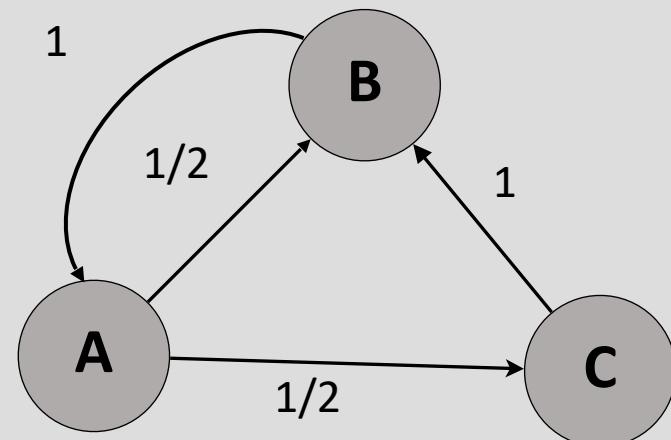
Need:  $QP = I$

Pose as a linear set of equations.

Solve with Gaussian Elimination

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

$\vec{p}_1$      $\vec{b}_1$



# Computing the Matrix Inverse

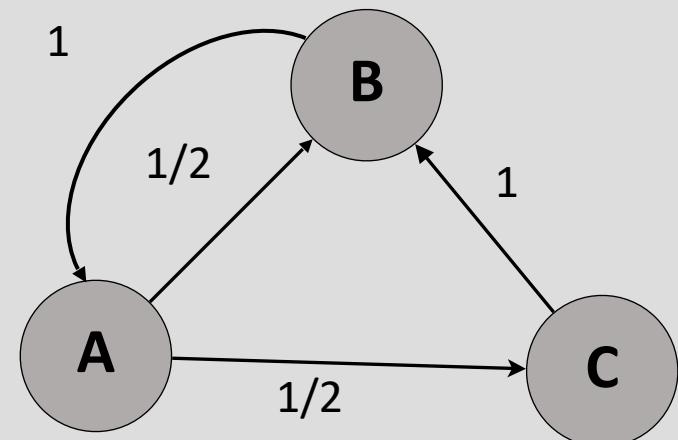
Need:  $QP = I$

Pose as a linear set of equations.

Solve with Gaussian Elimination

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

$\vec{p}_1 \quad \vec{p}_2 \quad \vec{p}_3$        $\vec{b}_1 \quad \vec{b}_2 \quad \vec{b}_3$



# Matrix Inverse via Gaussian Elimination

*Q*

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

*I*

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

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

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

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

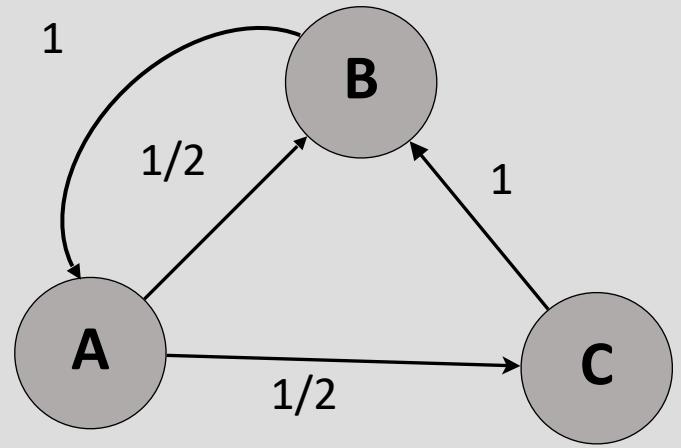
*I*      *P*

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

# Let's check

$$\begin{bmatrix} 6 \\ 10 \\ 2 \end{bmatrix} \begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} A \rightarrow A & B \rightarrow A & C \rightarrow A \\ A \rightarrow B & B \rightarrow B & C \rightarrow B \\ A \rightarrow C & B \rightarrow C & C \rightarrow C \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

4      6      8

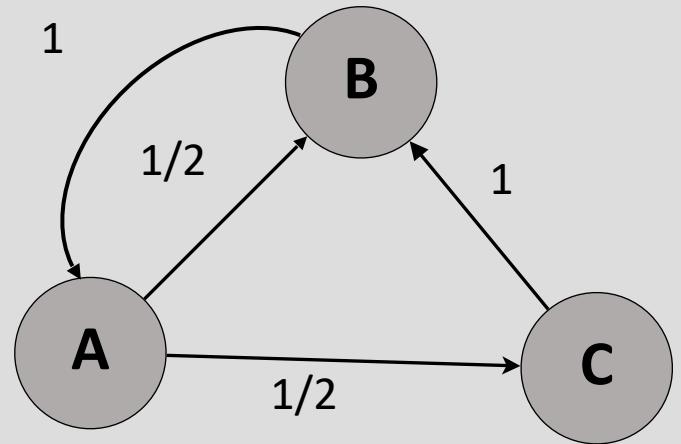


$$\begin{bmatrix} & & \\ & & \\ & & \end{bmatrix} = \begin{bmatrix} 0 & 0 & 2 \\ 1 & 0 & 0 \\ 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} 6 \\ 10 \\ 2 \end{bmatrix}$$

# Let's check

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

4      6      10      2



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

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.
  - If columns of  $A$  are lin. dep. then  $A^{-1}$  does not exist
  - 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

$$A\vec{\alpha} = 0$$
$$A^{-1}A\vec{\alpha} = A^{-1}0$$
$$I\vec{\alpha} = 0$$

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!