

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

EECS 16A

Matrix Transformations

Last time: Linear combination of vectors

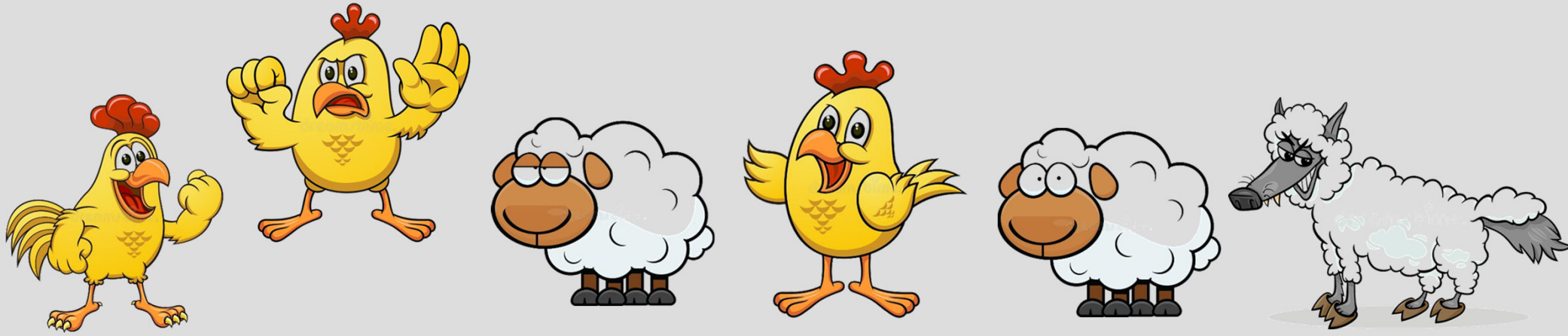
- Given set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \in \mathbb{R}^N$, and coefficients $\{\alpha_1, \alpha_2, \dots, \alpha_M\} \in \mathbb{R}$
- A linear combination of vectors is defined as: $\vec{b} \triangleq \alpha_1 \vec{a}_1 + \alpha_2 \vec{a}_2 + \dots + \alpha_M \vec{a}_M$ } scale and add the vectors
↑ defined as

Example:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \vec{a}_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$
$$= x_1 \vec{a}_1 + x_2 \vec{a}_2 + x_3 \vec{a}_3$$

Matrix-vector multiplication is a linear combination of the columns of A!

Example: write as a linear combination of kinds of feet



e: # eyes
b: # beaks
f: # of feet
c: chicken
s: sheep
w: wolf

$$f = 2c + 4s + 4w$$

$$b = 1c$$

$$e = 2c + 2s + 2w$$

$$\begin{bmatrix} f \\ b \\ e \end{bmatrix} = \begin{bmatrix} 2 & 4 & 4 \\ 1 & 0 & 0 \\ 2 & 2 & 2 \end{bmatrix} \begin{bmatrix} c \\ s \\ w \end{bmatrix}$$

Is there unique sol'n?

No! col 2 & 3 same \rightarrow Lin. Dep.!

I cannot tell sheep
from wolf! 😞

Last time: Span / Column Space / Range

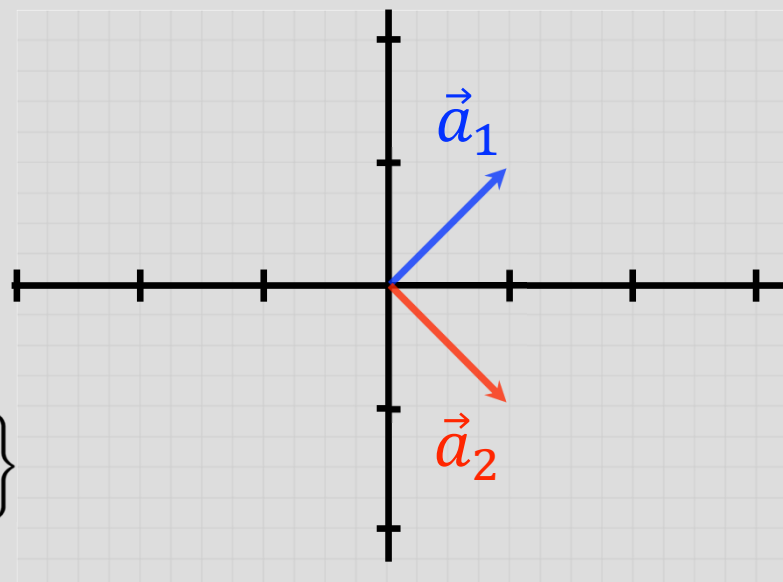
Span of the columns of A: the set of all vectors \vec{b} s.t. $A\vec{x} = \vec{b}$ has a solution

- the set of all vectors that can be reached by all possible linear combinations of the columns of A

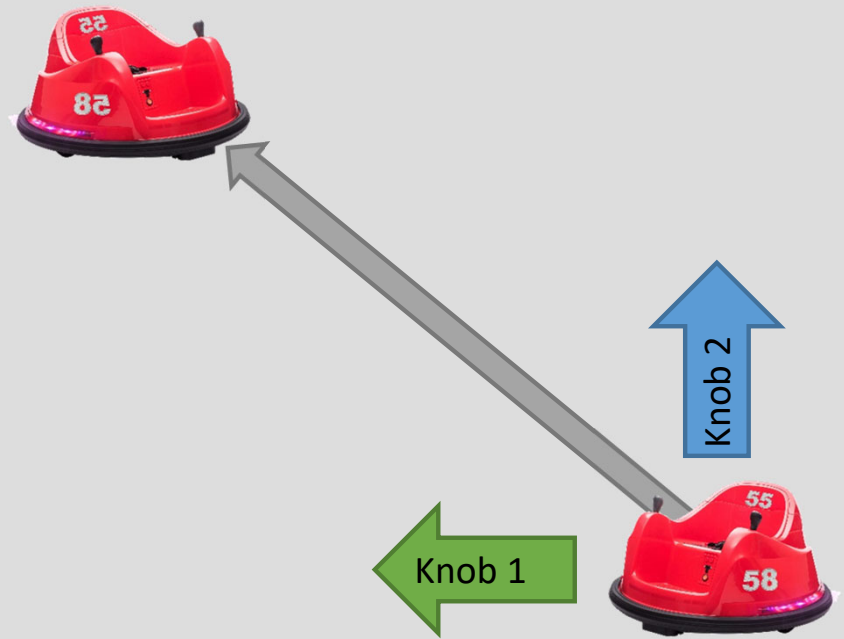
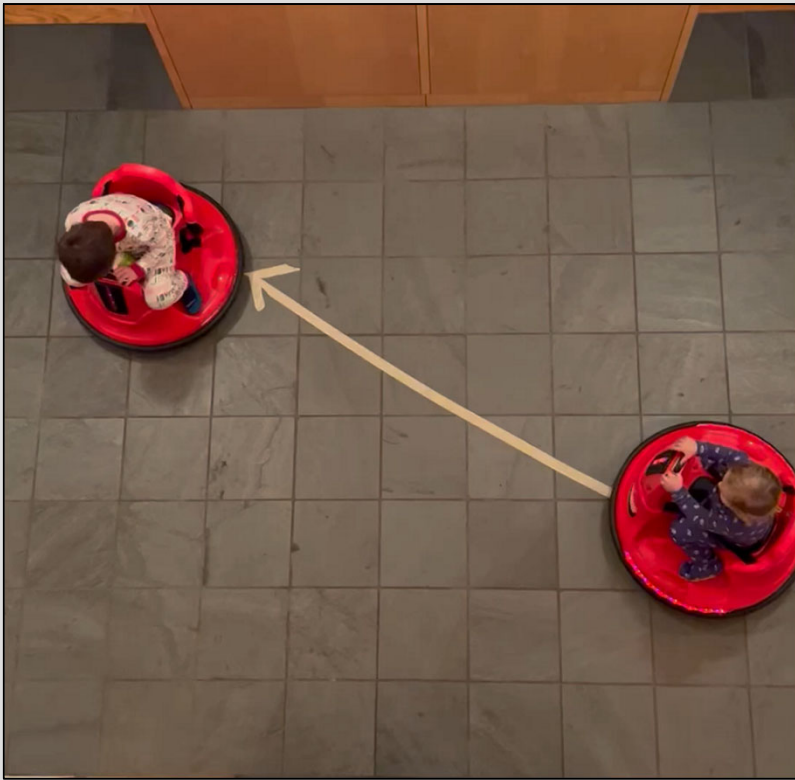
Example: span of the cols of A is \mathbb{R}^2 !

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

$$\text{span}(\text{cols of } A) = \left\{ \vec{v} \mid \vec{v} = \alpha \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \beta \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad \alpha, \beta \in \mathbb{R} \right\}$$



Last time: Solutions to $Ax=b$ are in the span of $\text{cols}(A)$



Linear Dependence

Definition 1:

A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \in \mathcal{R}^N$ are linearly dependent if $\exists \{\alpha_1, \alpha_2, \dots, \alpha_M\} \in \mathbb{R}$, such that: $\vec{a}_i = \sum_{j \neq i} \alpha_j \vec{a}_j \quad 1 \leq i, j \leq M$

↑ there exists coeffs. one vector is lin. combo. of others

Definition 2:

A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \in \mathcal{R}^N$ are linearly dependent if $\exists \{\alpha_1, \alpha_2, \dots, \alpha_M\} \in \mathbb{R}$, such that: $\sum_{i=1}^M \alpha_i \vec{a}_i = 0$ As long as not all $\alpha_i = 0$

Definition:

A set of vectors are linearly independent if they are not dependent

Linear Dependence

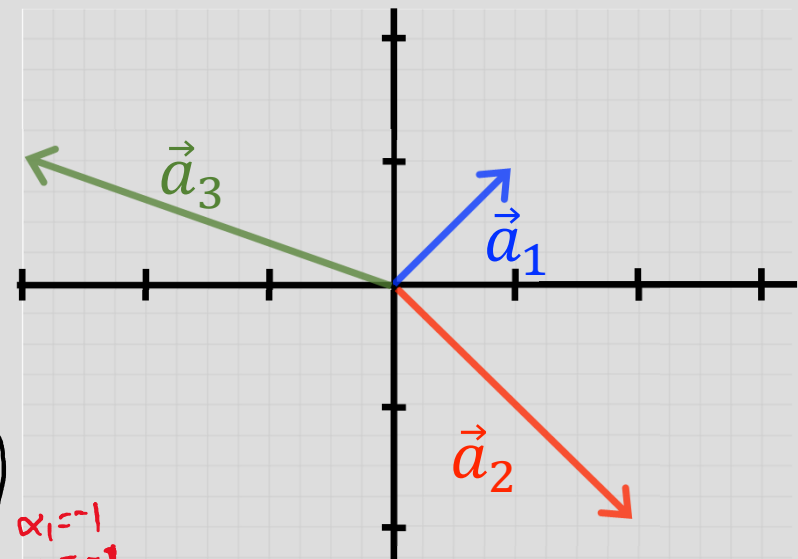
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$$A = \begin{bmatrix} 1 & 2 & -3 \\ 1 & -2 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} = \alpha_1 \begin{bmatrix} 2 \\ -2 \end{bmatrix} + \alpha_2 \begin{bmatrix} -3 \\ 1 \end{bmatrix}$$

$$\left[\begin{array}{cc|c} 2 & -3 & 1 \\ -2 & 1 & 1 \end{array} \right] \xrightarrow{R_2+R_1} \left[\begin{array}{cc|c} 2 & -3 & 1 \\ 0 & -2 & 2 \end{array} \right] \xrightarrow{R_2/2} \left[\begin{array}{cc|c} 2 & -3 & 1 \\ 0 & 1 & -1 \end{array} \right] \xrightarrow{R_1/2} \left[\begin{array}{cc|c} 1 & -\frac{3}{2} & \frac{1}{2} \\ 0 & 1 & -1 \end{array} \right] \xrightarrow{R_1 + \frac{3R_2}{2}} \left[\begin{array}{cc|c} 1 & 0 & -\frac{1}{2} \\ 0 & 1 & -1 \end{array} \right]$$



$$\alpha_1 = -1 \\ \alpha_2 = -1$$

Linear Dependence

Definition 1:

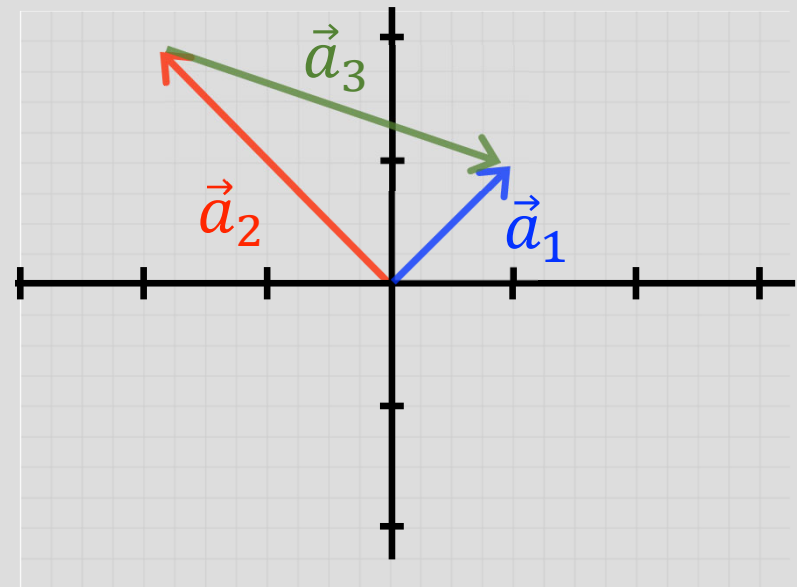
A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \in \mathcal{R}^N$ are linearly dependent if $\exists \{\alpha_1, \alpha_2, \dots, \alpha_M\} \in \mathbb{R}$, such that: $\vec{a}_i = \sum_{j \neq i} \alpha_j \vec{a}_j \quad 1 \leq i, j \leq M$

$$A = \begin{bmatrix} 1 & 2 & -3 \\ 1 & -2 & 1 \end{bmatrix}$$

check: $-1 \begin{bmatrix} 2 \\ -2 \end{bmatrix} - 1 \begin{bmatrix} -3 \\ 1 \end{bmatrix} = \begin{bmatrix} -2+3 \\ 2-1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \checkmark$

α_1 α_2 col 1

$$-\vec{a}_2 - \vec{a}_3 = \vec{a}_1$$



Linear Dependence

Definition 1:

A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \in \mathcal{R}^N$ are linearly dependent if $\exists \{\alpha_1, \alpha_2, \dots, \alpha_M\} \in \mathbb{R}$, such that: $\vec{a}_i = \sum_{j \neq i} \alpha_j \vec{a}_j \quad 1 \leq i, j \leq M$

Definition 2:

A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \in \mathcal{R}^N$ are linearly dependent if $\exists \{\alpha_1, \alpha_2, \dots, \alpha_M\} \in \mathbb{R}$, such that: $\sum_{i=1}^M \alpha_i \vec{a}_i = \vec{0}$ As long as not all $\alpha_i = 0$

Definition:

A set of vectors are linearly independent if they are not dependent

Linear Dependence

Definition 2:

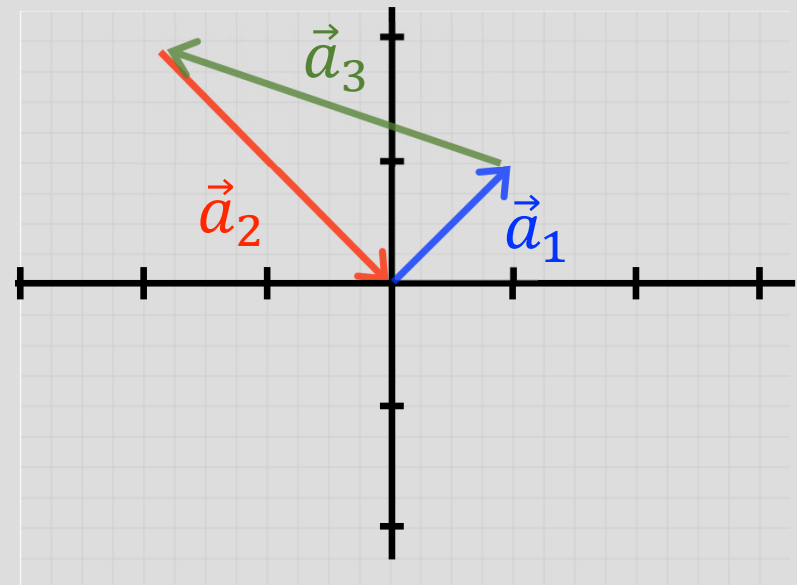
A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \in \mathcal{R}^N$ are linearly dependent if

$\exists \{\alpha_1, \alpha_2, \dots, \alpha_M\} \in \mathbb{R}$, such that: $\sum_{i=1}^M \alpha_i \vec{a}_i = \vec{0}$ As long as not all $\alpha_i = 0$

$$A = \begin{bmatrix} 1 & 2 & -3 \\ 1 & -2 & 1 \end{bmatrix}$$

$$\vec{a}_1 = -\vec{a}_2 - \vec{a}_3$$

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



Linear Dependence

Definition 1:

A set of vectors $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \in \mathcal{R}^N$ are linearly dependent if $\exists \{\alpha_1, \alpha_2, \dots, \alpha_M\} \in \mathbb{R}$, such that: $\vec{a}_i = \sum_{j \neq i} \alpha_j \vec{a}_j \quad 1 \leq i, j \leq M$

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Definition:

A set of vectors are linearly independent if they are not dependent

Examples

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

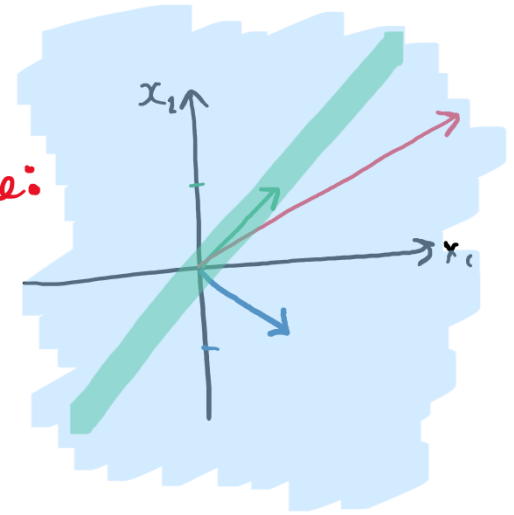
not rational #s, but real

these two vectors already span \mathbb{R}^2

This vector is in \mathbb{R}^2 so doesn't change span

linearly dependent!

3 vectors in 2D space:



$$\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\} \rightarrow \text{span } \mathbb{R}^2$$

linearly independent!

$$\text{cols} \left(\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 5 \\ 0 & 0 & 2 \end{bmatrix} \right)$$

span \mathbb{R}^2

linearly independent!

does this get to anywhere in \mathbb{R}^3 ? yes

$$\alpha_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \alpha_2 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 5 \\ 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & | & 0 \\ 0 & 1 & | & 5 \\ 0 & 0 & | & 2 \end{bmatrix} \rightarrow 0 \neq 2$$

No sol'n!

Prove it!

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

Prove it!

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

Let's prove for a 3×3 A mtrx:

What we know:

cols are lin. dep. $A = [\vec{a}_1 \ \vec{a}_2 \ \vec{a}_3]$

Want to show: more than one sol'n

concept: Pick some sol'n, \vec{x}^* , and show that there's another one

since \vec{x}^* is a sol'n, then $A\vec{x}^* = \vec{b}$

From lin. indep. def'n #2:

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

Anything in direction $\vec{\alpha}$ is "invisible" to this mtrx (Null space)

Now try $\vec{x}' = \vec{x}^* + \vec{\alpha}$

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

because linear

$\uparrow A\vec{x}' = \vec{b}$, so \vec{x}' is also a sol'n!

Prove it!

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 dependent

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:

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

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

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

Pop Quiz



[Responses](#)

After doing Gaussian Elimination on a system of linear equations $A\vec{x}=\vec{b}$, the augmented matrix looks like below. Choose the most accurate statement:

$$\left[\begin{array}{cccc|c} 1 & * & 0 & 0 & 5 \\ 0 & 0 & 0 & 0 & 4 \\ 0 & 0 & 1 & 0 & 3 \\ 0 & 0 & 0 & 1 & 2 \end{array} \right]$$

- $Ax=b$ has no solution
- $Ax=0$ has infinite solutions
- The columns of A are linearly dependent
- All of the above

Matrix Transformations

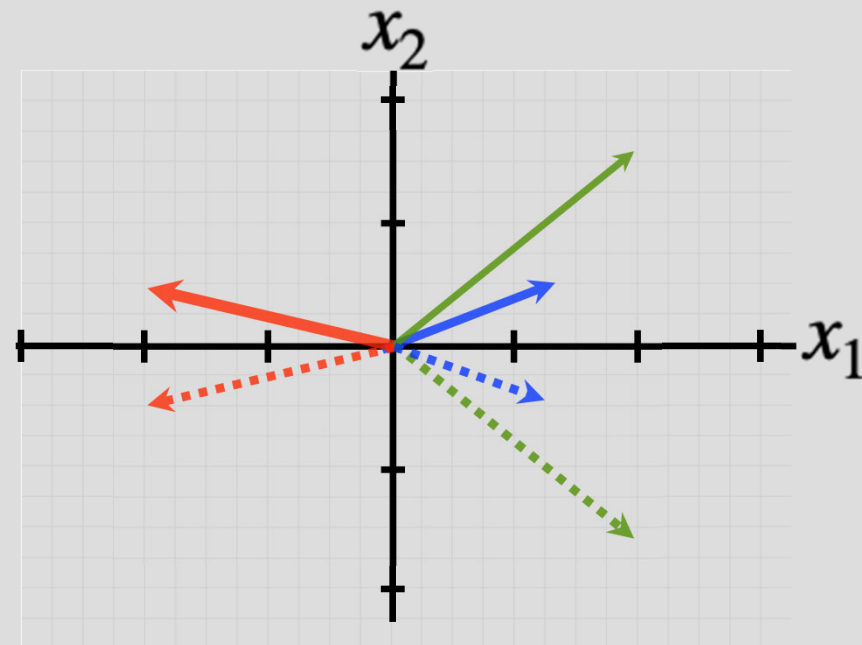
$$\begin{bmatrix} \cos 90^\circ & \sin 90^\circ \\ -\sin 90^\circ & \cos 90^\circ \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 0 \\ a_1 \end{bmatrix}$$

Matrices are operators that transform vectors

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

Reflection
Matrix!

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



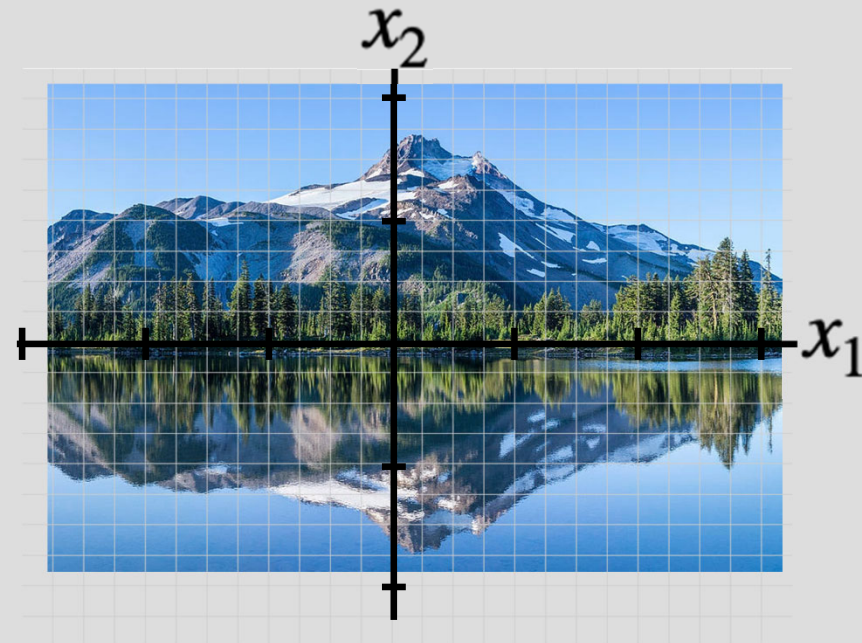
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How would I design a matrix to reflect about x_2 axis?

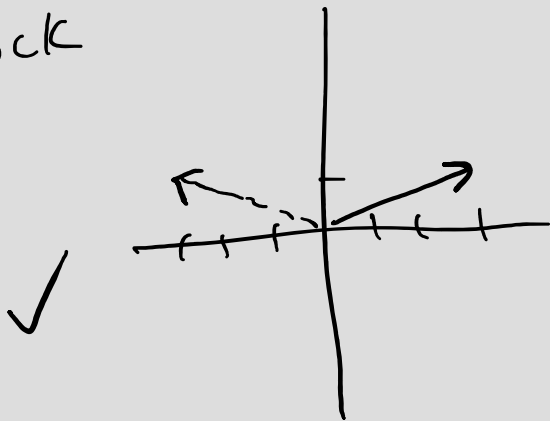


How would I design a matrix to reflect about x_2 axis?

Want

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

check



Reflection
Matrix!

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

try $\vec{x} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$

$$\begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 1 \end{bmatrix} = \begin{bmatrix} -3 \\ 1 \end{bmatrix}$$

Matrices are operators that transform vectors

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

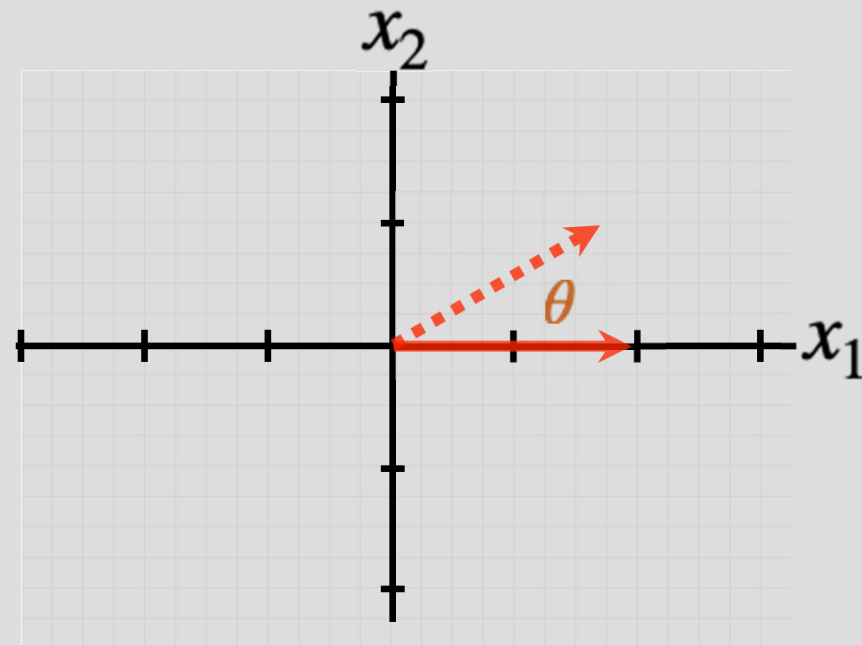
What does it do?

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

Rotation Matrix!

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

But they're different?!?



What if I rotate twice?

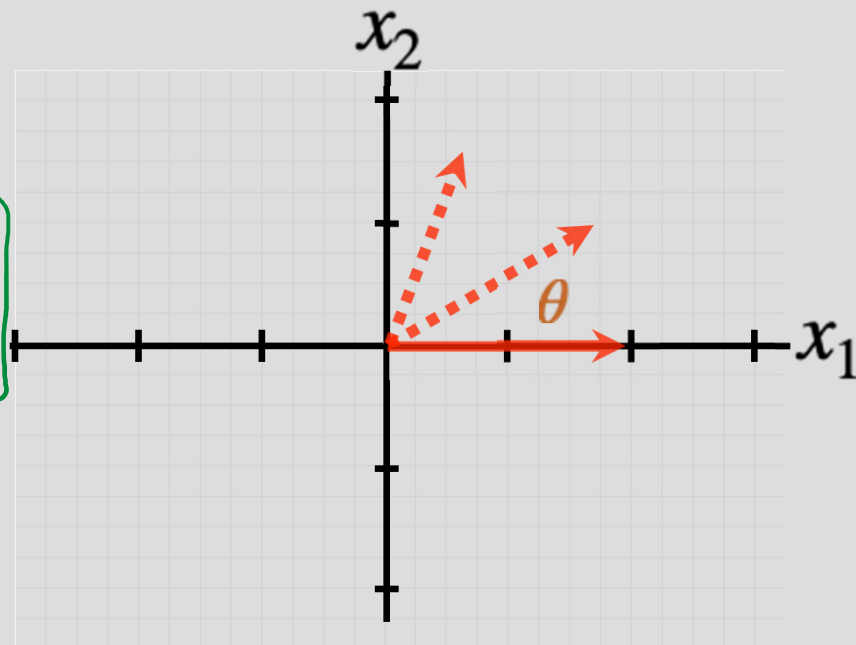
$$\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \cos(\theta) \\ \sin(\theta) \end{bmatrix}$$

Rotation Matrix!

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

$$\begin{bmatrix} (\cos\theta\cos\theta - \sin\theta\sin\theta) \\ \sin\theta\cos\theta + \cos\theta\sin\theta \end{bmatrix} = \begin{matrix} \theta = 90^\circ \\ \begin{bmatrix} 0^2 - 1^2 \\ 1 \cdot 0 + 0 \cdot 1 \end{bmatrix} = \begin{bmatrix} -1 \\ 0 \end{bmatrix} \end{matrix}$$

rotated 180° ✓



What's the matrix transform?



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

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