

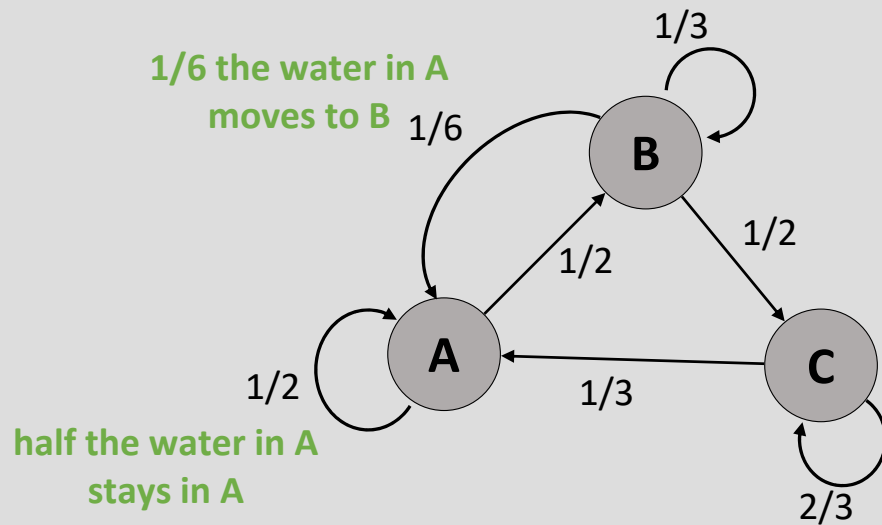
	vectors, i, j adding and multiplying vectors
	Equations for straight lines and planes
	subspace, linear combinations

EECS 16A

Vector Spaces:
Null spaces and Columnspaces

Last time: Graph Representation



“directed” graph because arrows have a direction

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.

Edges

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

Edge weights

Last time: Matrix inverses

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

- We can use Gaussian Elimination (**Gauss-Jordan method**) to find the inverse of a square matrix
- Once we have the inverse, we can use it to solve system of equations

So matrix inverse is like division? Sort of, but matrix division doesn't technically exist

What if $Ax=b$ has infinite solutions? No way to predict x from b , so A is not invertible

**The right tool can make all
the difference**



Calculating matrix inverses: Gauss-Jordan method

Pose as a linear set of equations. Solve with Gaussian Elimination

Augmented
mtx form: $\left[A \mid I \right] \xrightarrow[\text{Elim.}]{\text{Gauss.}} \left[I \mid A^{-1} \right]$

what if G.E.
doesn't work?
↳ There is no inverse!
(or you made a mistake)

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 $ad - bc$

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

Can we always invert a function?

- Can we always invert a function $f^{-1}(f(\vec{x})) = \vec{x}$?
 - $f(x) = x^2$?
 - $f(x) = ax$?
 - $f(x) = Ax$?

Proof: Invertibility of Linear Transformations

Theorem: A is invertible, if and only if (iff) the columns of A are linearly independent. *(unique sol'n)*

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

What we know:

cols of A are lin. dep.



$$\exists \vec{\alpha} \neq \vec{0}, \text{ s.t. } A\vec{\alpha} = \vec{0}$$

↑
there exists

↑
some non-trivial combo of cols(A) $\rightarrow \vec{0}$

Proof by contradiction: Assume A^{-1} exists

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

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

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

To show:

A^{-1} does not exist

Equivalent Statements

- Matrix A is **invertible**
- $A\vec{x} = \vec{b}$ has a unique solution
- A has linearly independent columns (A is **full rank**)
- A has a **trivial nullspace**
- The **determinant** of A is not zero

Today's Jargon

- **Rank** of a matrix A is the number of linearly independent columns
- **Nullspace** of a matrix A is the set of solutions to $A\vec{x} = \vec{0}$
- A **vector space** is a set of vectors connected by two operators $(+, \cdot)$
- A vector **subspace** is a subset of vectors that have “nice properties”
- A **basis** for a vector space is a minimum set of vectors needed to represent all vectors in the space
- **Dimension** of a vector space is the number of basis vectors
- **Column space** is the span (range) of the columns of a matrix
- **Row space** is the span of the rows of a matrix

Vector Space

A vector space is a set of vectors and scalars ($\mathbb{V} \in \mathbb{R}^N, \mathbb{F} \in \mathbb{R}$) and two operators $\cdot, +$ that satisfy the following:

set of vectors
set of scalars

multiply

for all $\vec{x}, \vec{y}, \vec{z} \in \mathbb{V}$ and $\alpha, \beta \in \mathbb{R}$

Axioms of closure
Properties

1) $\alpha \vec{x} \in \mathbb{V}$

2) $\vec{x} + \vec{y} \in \mathbb{V}$

3) $\vec{x} + (\vec{y} + \vec{z}) = (\vec{x} + \vec{y}) + \vec{z}$ (associativity)

4) $\vec{x} + \vec{y} = \vec{y} + \vec{x}$ (commutativity)

Axioms of addition
(+)

5) $\exists \vec{0} \in \mathbb{V}$ s.t. $\vec{x} + \vec{0} = \vec{x}$ (additive identity)

6) $\exists (-\vec{x}) \in \mathbb{V}$ s.t. $\vec{x} + (-\vec{x}) = \vec{0}$ (additive inverse)

7) $\alpha(\vec{x} + \vec{y}) = \alpha\vec{x} + \alpha\vec{y}$ (distributivity)

Axioms of scaling
(\cdot)

8) $\alpha \cdot (\beta \vec{x}) = (\alpha\beta) \cdot \vec{x}$

9) $(\alpha + \beta)\vec{x} = \alpha\vec{x} + \beta\vec{x}$

10) $1 \cdot \vec{x} = \vec{x}$

Pop Quiz: Is it a vector space? (including operations \cdot and $+$)

The set of all 2×2 matrices:

✓ $\begin{bmatrix} a & b \\ c & d \end{bmatrix} \in \mathbb{R}^{2 \times 2} ?$

A vector space, is a set of vectors and scalars ($\mathbb{V} \in \mathbb{R}^N, \mathbb{F} \in \mathbb{R}$)

and two operators $\cdot, +$ that satisfy the following:

1) $\alpha \vec{x} \in \mathbb{V}$

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

still in set ✓

2) $\vec{x} + \vec{y} \in \mathbb{V}$

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} + \begin{bmatrix} e & f \\ g & h \end{bmatrix} = \begin{bmatrix} a+e & b+f \\ c+g & d+h \end{bmatrix}$$

still in set ✓

3) $\vec{x} + (\vec{y} + \vec{z}) = (\vec{x} + \vec{y}) + \vec{z}$

Mtx addition is is associative ✓

4) $\vec{x} + \vec{y} = \vec{y} + \vec{x}$

Also commutative ✓

5) $\exists \vec{0} \in \mathbb{V}$ s.t. $\vec{x} + \vec{0} = \vec{x}$

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \checkmark$$

6) $\exists (-\vec{x}) \in \mathbb{V}$ s.t. $\vec{x} + (-\vec{x}) = \vec{0}$

$$\begin{bmatrix} -a & -b \\ -c & -d \end{bmatrix} + \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \checkmark$$

7) $\alpha(\vec{x} + \vec{y}) = \alpha\vec{x} + \alpha\vec{y}$

8) $\alpha \cdot (\beta\vec{x}) = (\alpha\beta) \cdot \vec{x} \checkmark$

9) $(\alpha + \beta)\vec{x} = \alpha\vec{x} + \beta\vec{x} \checkmark$

10) $1 \cdot \vec{x} = \vec{x} \checkmark$

$$(\alpha + \beta) \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \alpha \begin{bmatrix} a & b \\ c & d \end{bmatrix} + \beta \begin{bmatrix} a & b \\ c & d \end{bmatrix} \checkmark$$

YES!
it's a vector space

Pop Quiz: Is it a vector space? (including operations \cdot and $+$)

A vector space, is a set of vectors and scalars ($\mathbb{V} \in \mathbb{R}^N, \mathbb{F} \in \mathbb{R}$) and two operators $\cdot, +$ that satisfy the following:

- 1) $\alpha \vec{x} \in \mathbb{V}$
- 2) $\vec{x} + \vec{y} \in \mathbb{V}$
- 3) $\vec{x} + (\vec{y} + \vec{z}) = (\vec{x} + \vec{y}) + \vec{z}$
- 4) $\vec{x} + \vec{y} = \vec{y} + \vec{x}$
- 5) $\exists \vec{0} \in \mathbb{V}$ s.t. $\vec{x} + \vec{0} = \vec{x}$
- 6) $\exists (-\vec{x}) \in \mathbb{V}$ s.t. $\vec{x} + (-\vec{x}) = \vec{0}$
- 7) $\alpha(\vec{x} + \vec{y}) = \alpha\vec{x} + \alpha\vec{y}$
- 8) $\alpha \cdot (\beta\vec{x}) = (\alpha\beta) \cdot \vec{x}$
- 9) $(\alpha + \beta)\vec{x} = \alpha\vec{x} + \beta\vec{x}$
- 10) $1 \cdot \vec{x} = \vec{x}$



[Responses](#)

✓ Is \mathbb{R}^2 a vector space?
Also $\mathbb{R}, \mathbb{R}^3, \mathbb{R}^4 \dots$

✓ $\begin{bmatrix} a & b \\ c & d \end{bmatrix} \in \mathbb{R}^{2 \times 2}$?

✗ $\alpha \in \mathbb{R}, \alpha \geq 0$?
doesn't satisfy ⑥

✓ $\text{span} \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$?

✗ $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$?

✓ 0 ?

Subspaces

- A subspace \mathbb{U} consists of a subset of \mathbb{V} in vector space $(\mathbb{V}, \mathbb{F}, +, \cdot)$

$\mathbb{U} \subset \mathbb{V}$ and have 3 properties:

1. Contains $\vec{0}$, i.e., $\vec{0} \in \mathbb{U}$

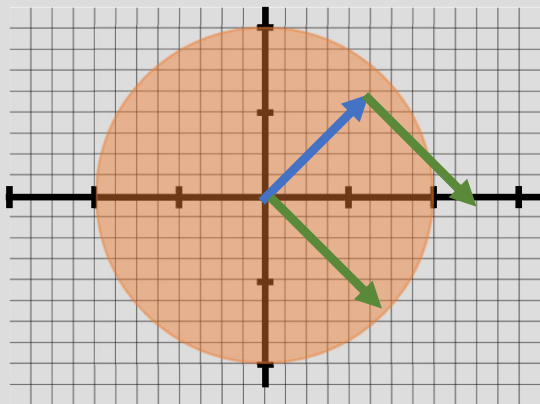
2. Closed under vector addition: $\vec{v}_1, \vec{v}_2 \in \mathbb{U} \Rightarrow \vec{v}_1 + \vec{v}_2 \in \mathbb{U}$

3. Closed under scalar multiplication: $\vec{v}_1 \in \mathbb{U}, \alpha \in \mathbb{F} \Rightarrow \alpha \vec{v}_1 \in \mathbb{U}$

can't
escape
set with
 $+, \cdot$

set of vectors scalars operations

Q: Consider all vectors \vec{v} who's length < 1 . Is this a subspace?



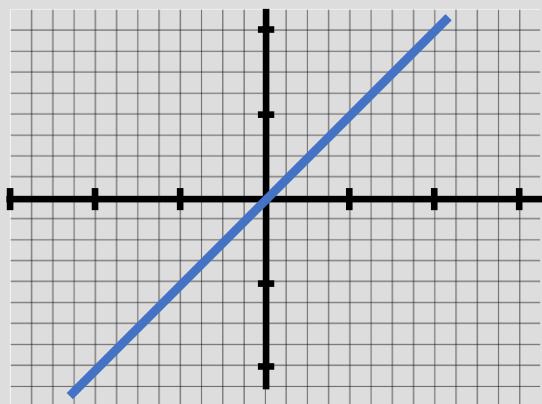
A: not closed under addition,
nor scalar multiplication

Subspaces

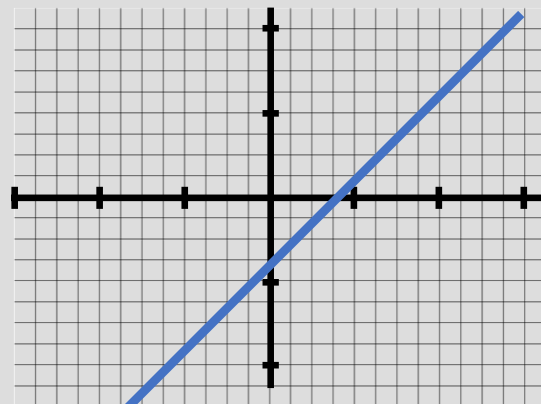
- A subspace \mathbb{U} consists of a subset of \mathbb{V} in vector space $(\mathbb{V}, \mathbb{F}, +, \cdot)$
 - $\mathbb{U} \subset \mathbb{V}$ and have 3 properties
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 3. Closed under scalar multiplication: $\vec{v}_1 \in \mathbb{U}, \alpha \in \mathbb{F} \Rightarrow \alpha \vec{v}_1 \in \mathbb{U}$

Q: Is $\text{Span} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$ a subspace?

Q: What about this?



A: Yes!

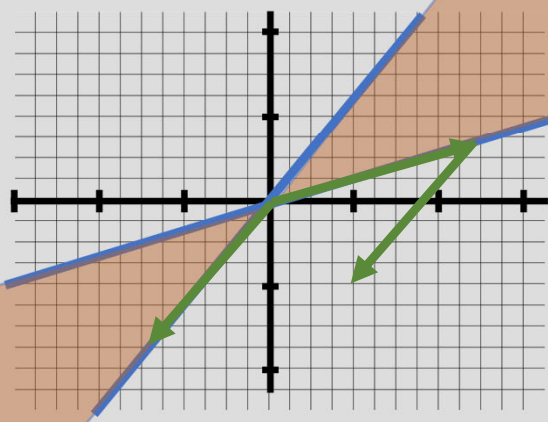


A: $\vec{0} \notin \mathbb{U}$
No!

Subspaces

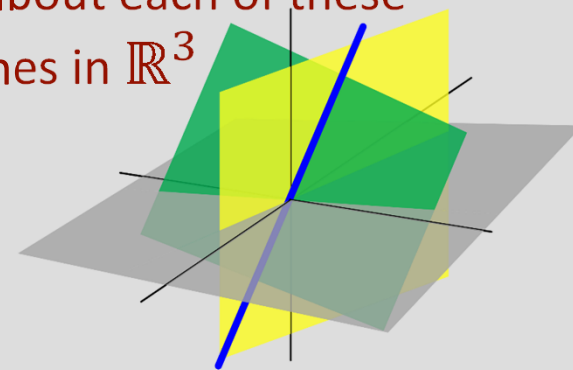
- A subspace \mathbb{U} consists of a subset of \mathbb{V} in vector space $(\mathbb{V}, \mathbb{F}, +, \cdot)$
 - $\mathbb{U} \subset \mathbb{V}$ and have 3 properties
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 3. Closed under scalar multiplication: $\vec{v}_1 \in \mathbb{U}, \alpha \in \mathbb{F} \Rightarrow \alpha \vec{v}_1 \in \mathbb{U}$

Q: What about this?



A: Not closed under addition!

Q: What about each of these 2D planes in \mathbb{R}^3



A: yes, as long as passing through $\vec{0}$

Example: set of all upper triangular 2x2 matrices

$$W = \left\{ \begin{bmatrix} a & b \\ 0 & d \end{bmatrix} \mid a, b, d \in \mathbb{R} \right\}, V = \mathbb{R}^{2 \times 2}$$

Is W a subspace of V ?

✓ 1. Zero vector?

✓ 2. Closed under addition?

✓ 3. Closed under scalar multiplication?

$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ is in W ✓

$$\begin{bmatrix} a_1 & b_1 \\ 0 & d_1 \end{bmatrix} + \begin{bmatrix} a_2 & b_2 \\ 0 & d_2 \end{bmatrix} = \begin{bmatrix} a_1+a_2 & b_1+b_2 \\ 0 & d_1+d_2 \end{bmatrix}$$

still upper tri!

$$\alpha \begin{bmatrix} a_1 & b_1 \\ 0 & d_1 \end{bmatrix} = \begin{bmatrix} \alpha a_1 & \alpha b_1 \\ 0 & \alpha d_1 \end{bmatrix} \text{ still in } W! \checkmark$$

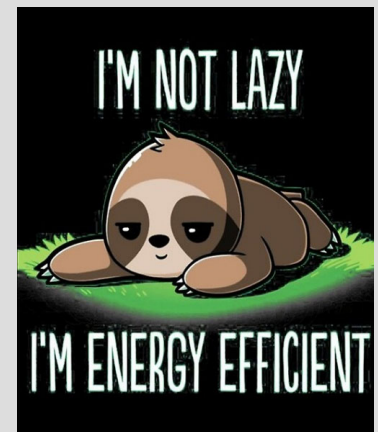
YES upper tri. mtx is a subspace of $\mathbb{R}^{2 \times 2}$

Basis → the minimum set of vectors that spans a vector space

What is the **most efficient** representation of the vector space?

$$\left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 0.5 \\ -0.7 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} e \\ \pi \end{bmatrix}, \begin{bmatrix} 6 \\ 4 \end{bmatrix}, \begin{bmatrix} 4 \\ 6 \end{bmatrix} \right\}$$

can only span \mathbb{R}^2 ,
so only need 2 lin. ind. vectors
↳ Pick any **two!**



Basis → the minimum set of vectors that spans a vector space

Definition: given \mathbb{V} , a set of vectors $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_N\}$ is a **basis** of the vector space, if it satisfies:

- $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_N\}$ are linearly independent
- $\forall \vec{v} \in \mathbb{V}, \exists \alpha_1, \alpha_2, \dots, \alpha_N \in \mathbb{R}$ such that $\vec{v} = \alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2 + \dots + \alpha_N \vec{v}_N$

Examples: which are a basis for $V = \mathbb{R}^3$?

✓ $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$

✓ $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 10 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$

✓ $\left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right\}$

✗ $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\}$

✗ $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 10 \\ 0 \end{bmatrix} \right\}$

Column Space

Definition: The range/span/column space of a set of vectors is the set of all possible linear combinations:

$$\text{span}\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \triangleq \left\{ \sum_{m=1}^M \alpha_m \vec{a}_m \mid \alpha_1, \alpha_2, \dots, \alpha_M \in \mathbb{R} \right\}$$

Example:

$$A = \begin{bmatrix} a & 0 \\ 0 & b \\ 0 & 0 \end{bmatrix}$$

$$\vec{v}_1 = A\vec{u}_1, \vec{v}_2 = A\vec{u}_2$$

Q: Are the columns of A a basis? ❌

Q: Is the column space of A a subspace? ✅

1. Zero vector?
2. Closed under addition?
3. Closed under scalar multiplication?

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

$$\vec{v}_1 + \vec{v}_2 = A\vec{u}_1 + A\vec{u}_2 = A(\vec{u}_1 + \vec{u}_2)$$

$$\alpha\vec{v}_1 = \alpha A\vec{u}_1 = A(\alpha\vec{u}_1)$$

Today's Jargon

- **Rank** of a matrix A is the number of linearly independent columns
- **Nullspace** of a matrix A is the set of solutions to $A\vec{x} = \vec{0}$
- A **vector space** is a set of vectors connected by two operators $(+, \cdot)$
- A vector **subspace** is a subset of vectors that have “nice properties”
- A **basis** for a vector space is a minimum set of vectors needed to represent all vectors in the space
- **Dimension** of a vector space is the number of basis vectors
- **Column space** is the span (range) of the columns of a matrix
- **Row space** is the span of the rows of a matrix

Rank

USA Today University Ranking for Cal:

- #1 in Computer Systems
- #3 in Electrical/Electronic/Communications
- #3 in Computer Engineering

Rank

- $A \in \mathbb{R}^{N \times M}$, $\text{Rank}\{A\} = \dim\{\text{Span}\{\text{cols}(A)\}\}$

$$A = \begin{bmatrix} a & 0 \\ 0 & b \\ 0 & 0 \end{bmatrix}$$

2

$$A = \begin{bmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

2

$$A = \begin{bmatrix} a & 0 & 0 \\ b & 0 & 0 \\ c & 0 & 0 \end{bmatrix}$$

1

- $\text{Rank}\{A\} = \dim\{\text{Span}\{\text{cols}(A)\}\} \leq \min(M, N)$

Can rank be larger than input $\dim(A)$? No!

Where do the rest of the dimensions go? To the null space

Null Space

- Definition: The null-space of $A \in \mathbb{R}^{N \times M}$ is the set of all vectors $\vec{x} \in \mathbb{R}^M$ such that: $A\vec{x} = \vec{0}$

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

How many solutions for \vec{x} satisfy the above?

Example: what is the null space?

$$\begin{bmatrix} 1 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Linearly
independent!

$$\vec{x} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$\vec{0}$ is always in the null space — trivial Null space

Example: what is the null space?

$$\begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Linearly
dependent!

$$\vec{x} = \alpha \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\left[\begin{array}{cc|c} 1 & -2 & 0 \\ 0 & 0 & 0 \end{array} \right] \rightarrow x_1 = 2x_2$$
$$\hookrightarrow \vec{x} = \begin{bmatrix} 2\alpha \\ \alpha \end{bmatrix}$$

A has a non-trivial null-space, $\text{span} \left\{ \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right\}$

Example

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

We know that $\vec{v}_0 \in \text{Null}(A)$

$$\rightarrow A\vec{v}_0 = \vec{0}$$

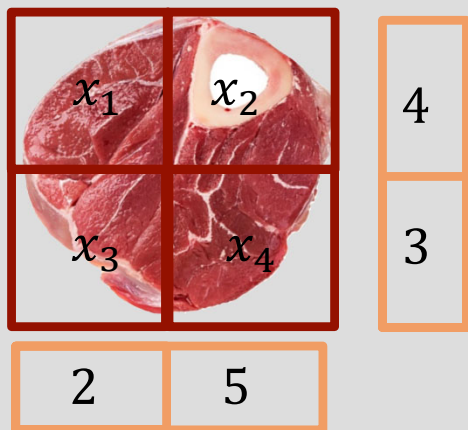
We know 1 solution: \vec{x}_0

$$\rightarrow A\vec{x}_0 = \vec{b}$$

Then: $\vec{x}_0 + \alpha\vec{v}_0$ is also a solution

$$\begin{aligned}\rightarrow A(\vec{x}_0 + \alpha\vec{v}_0) &= A\vec{x}_0 + A(\alpha\vec{v}_0) \\ &= \vec{b} + \alpha A\vec{v}_0 \\ &= \vec{b}\end{aligned}$$

Back to Tomography!



$$\left[\begin{array}{cccc|c} 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 \end{array} \right]$$

Gaussian Elimination

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

x_4 is a free variable

$$x_1 - x_4 = 0 \rightarrow x_1 = x_4$$

$$x_2 + x_4 = 0 \rightarrow x_2 = -x_4$$

$$x_3 + x_4 = 0 \rightarrow x_3 = -x_4$$

pick $x_4 = \alpha$

$$\Rightarrow \vec{x} = \alpha \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}$$

Possible reconstruction

$$\begin{bmatrix} x_1 & x_2 \\ x_3 & x_4 \end{bmatrix} = \begin{bmatrix} 1 & 3 \\ 1 & 2 \end{bmatrix} + \alpha \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$$

Rank

- $A \in \mathbb{R}^{N \times M}$, $\text{Rank}\{A\} = \dim\{\text{Span}\{\text{cols}\{A\}\}\}$
- $\text{Rank}\{A\} = \dim\{\text{Span}\{A\}\} \leq \min(M, N)$
- Rank = L , mean the matrix $A \in \mathbb{R}^{N \times M}$ has L independent rows & columns

$$\text{Rank}\{A\} + \dim\{\text{Null}\{A\}\} = M$$

Rank-Nullity Theorem

"Full rank"
means rank is max
possible ($\min(M, N)$)