

**Theorem:**  
a cat has nine tails.

**Proof:**  
No cat has eight tails. A  
cat has one tail more  
than no cat  
therefore, a cat has  
nine tails.

**Proof that  $\$1 = 1 \text{ cent}$**

$$\begin{aligned}\$1 &= 100 \text{ cents} \\ &= (10 \text{ cents})^2 \\ &= (\$0.1)^2 \\ &= \$0.01 \\ &= 1 \text{c}\end{aligned}$$

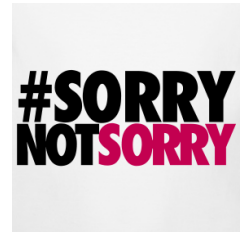
**WRONG!**

EE16A

Introduction to Proofs, Span, Linear Dependence and Independence

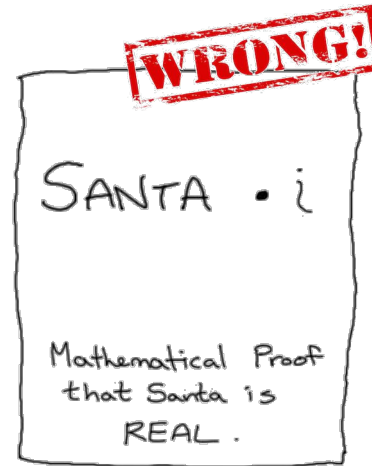
# Admin

- Warning: if you don't have Python experience, the lab/bootcamp will be long and hard!
  - Allocate the time, work together, don't get too discouraged



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- Today:
  - Span
  - Proofs!
  - Linear (in)dependence



# Last time: Multiplying matrices/vectors

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ a_{21} & \cdots & a_{2n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} b_{11} & \cdots & b_{1p} \\ b_{21} & \cdots & b_{2p} \\ \vdots & \ddots & \vdots \\ b_{n1} & \cdots & b_{np} \end{bmatrix} = \begin{bmatrix} c_{11} & \cdots & c_{1p} \\ c_{21} & \cdots & c_{2p} \\ \vdots & \ddots & \vdots \\ c_{m1} & \cdots & c_{mp} \end{bmatrix}$$

$m \times n$        $n \times p$        $m \times p$

Must be same!

# Systems of equations $A\vec{x} = \vec{b}$

**'system' matrix** **solve for me!** **measurements**

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ a_{21} & & a_{2n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

$m \times n$   $n \times 1$   $m \times 1$

**# equations/  
measurements** **# unknowns**

The diagram illustrates the system of equations  $A\vec{x} = \vec{b}$ . The matrix  $A$  is labeled as the 'system' matrix and has dimensions  $m \times n$ . The vector  $\vec{x}$  is labeled as the unknowns and has dimensions  $n \times 1$ . The vector  $\vec{b}$  is labeled as the measurements and has dimensions  $m \times 1$ . Red arrows point from the dimensions  $m$  and  $n$  to the text '# equations/measurements' and '# unknowns' respectively. The matrix  $A$  is shaded gray and contains a large 'A' watermark. The vector  $\vec{x}$  is shaded gray and contains a large 'x' watermark. The vector  $\vec{b}$  is shaded gray and contains a large 'b' watermark. The text 'solve for me!' is written in green above the vector  $\vec{x}$ .

# Last time: Row view

Rows represent how much the variables affect a particular measurement.

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ a_{21} & & a_{2n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix} \rightarrow \begin{array}{l} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{array}$$

# Last time: Column view

Columns represent how much a particular variable affects all measurements.

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ a_{21} & & a_{2n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$
$$x_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} + x_2 \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} + \cdots + x_n \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$
$$x_1 \vec{a}_1 + x_2 \vec{a}_2 + \cdots + x_n \vec{a}_n = \vec{b}$$

Linear Combination of  $\vec{a}$  vectors  
weighted by the unknowns!

# Linear combinations

Any scaling or addition of vectors

## Definition:

For a set of vectors  $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\} \in \mathbb{R}^N$  where  $\{\alpha_1, \alpha_2, \dots, \alpha_M\} \in \mathbb{R}$

then  $\vec{w} \triangleq \alpha_1 \vec{a}_1 + \alpha_2 \vec{a}_2 + \dots + \alpha_M \vec{a}_M$  is called a **linear combination** of the a-vectors



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Example: what are some linear combinations of  $\vec{x}_1 = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$   $\vec{x}_2 = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$  ?

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*Handwritten notes:*  
- "Indicates a set of vectors" points to the set notation.  
- "where all elements are real" points to  $\mathbb{R}$ .  
- "each vector has dimension N" points to  $\mathbb{R}^N$ .  
- "some coefficients" points to  $\alpha_1, \alpha_2, \dots, \alpha_M$ .  
- "are scalars" points to  $\mathbb{R}$ .  
- "is defined as" points to the definition symbol  $\triangleq$ .  
- "some new vector" points to  $\vec{w}$ .

Example: what are some linear combinations of  $\vec{x}_1 = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$   $\vec{x}_2 = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$  ?

$$\begin{aligned} \text{ex: } & 5\vec{x}_1 + 2\vec{x}_2 \\ & -6\vec{x}_1 - \vec{x}_2 \\ & 0\vec{x}_1 + 5\vec{x}_2 \end{aligned}$$

What does it mean to solve  $A\vec{x} = \vec{b}$  w.r.t. lin. combos. <sup>linear combinations</sup>  
with respect to

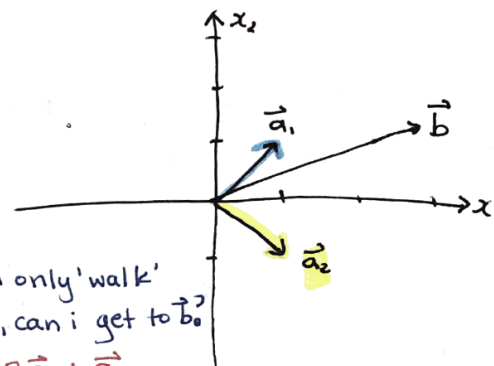
Before: find  $x_1, x_2, \dots, x_n$  that satisfy eqns (rows)

Now: What lin. combo. of cols of  $A$  give  $\vec{b}$ ? column view!

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

$$\begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

$\vec{a}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$   $\vec{a}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$



say i can only 'walk' along  $\vec{a}_1, \vec{a}_2$ , can i get to  $\vec{b}$ ?  
yes! with  $2\vec{a}_1 + \vec{a}_2$

Try G.E. instead:

$$\left[ \begin{array}{cc|c} 1 & -1 & 3 \\ 1 & 1 & 1 \end{array} \right] \rightarrow \left[ \begin{array}{cc|c} 1 & -1 & 3 \\ 0 & 2 & -2 \end{array} \right] \rightarrow \left[ \begin{array}{cc|c} 1 & 0 & 2 \\ 0 & 1 & 1 \end{array} \right] \rightarrow x_1 = 2, x_2 = 1$$

✓ yay!

Can i reach any other  $\vec{b}$  vector i want with combos of  $\vec{a}_1, \vec{a}_2$ ?

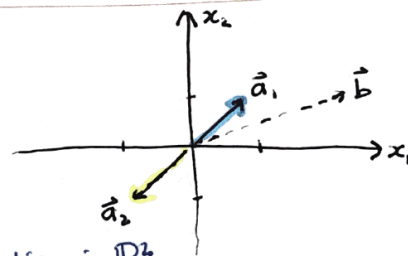
(Note: can go backwards)

Yes! And we'll prove it later.  $\leftarrow$  as long as in  $\mathbb{R}^2$

Example:

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

$\vec{a}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$   $\vec{a}_2 = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$



Now can i reach any  $\vec{b}$  vector?

No, not unless along a particular line in  $\mathbb{R}^2$

**SPAN**

of the cols of  $A$  is the set of all vectors  $\vec{b}$  such that

$A\vec{x} = \vec{b}$  has a solution (no need to be unique)

jargon alert!  
aka "Range"  
"col space of  $A$ "

i.e. the set of ~~possible~~ <sup>all possible</sup> vectors that can be reached  
by <sup>all possible</sup> lin. combo. of cols of  $A$

Example: What is the span of  $A = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$ ?

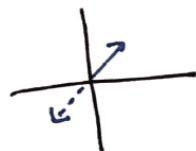


↳ entire 2D plane! Because, with  $A\vec{x} = \vec{b}$  we can

choose  $\alpha_1, \alpha_2$  such that we reach any point in  $\mathbb{R}^2$

set notation:  $\text{span}(A) = \left\{ \vec{v} \mid \vec{v} = \alpha_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \alpha_2 \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \alpha_1, \alpha_2 \in \mathbb{R} \right\}$

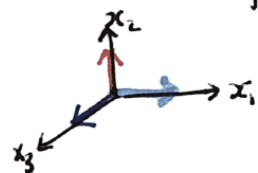
what is span of  $A = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$ ?



↳ A line! Cannot choose  $\alpha_1, \alpha_2$  to escape line

Why not? because  $\vec{a}_1, \vec{a}_2$  are linearly dependent

Example: What is span of  $\vec{x}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \vec{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \vec{x}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ ?



The entire 3D space  $\mathbb{R}^3$ !

Example: What is the span of  $\vec{0}$ ?  $\vec{0}$

This is the one vector you can always reach (always in span)

Fancy  
jargon

If  $\exists \vec{x}$  s.t.  $A\vec{x} = \vec{b}$  then we say  $\vec{b} \in \text{span}(\text{cols of } A)$


there exists a vector such that

are in

all the places you can get to

What if  $\vec{b}$  is NOT in  $\text{span}(\text{cols of } A)$ ?

↳ if meas.  $\vec{b}$  are not reachable by system  $A$ , then they are wrong (inconsistent) → No solution!

Let's go back to  $A = \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}$  case. 

↳ we had 2 cols, but could only 'reach' (span)  
a 1D line, because cols were linearly dependent

i.e.  $\vec{a}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$   $\vec{a}_2 = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$  and  $\boxed{\vec{a}_2 = -\vec{a}_1}$  ~~one to~~

no new info!  
Redundant redundant

Linear Dependence of a set of vectors means at least one is redundant (useless)  
depends only on A

↳ let's us check whether system will work (has a unique sol'n) before we take measurements (doesn't need  $\vec{b}$ )

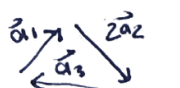
Definition 1: A set of vectors  $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\}$  is linearly dependent if  $\vec{a}_j = \sum_{m \neq j} \alpha_m \vec{a}_m$  for some  $\{\alpha_m \in \mathbb{R}\}_{m \neq j}$   
and some  $1 \leq j \leq M$

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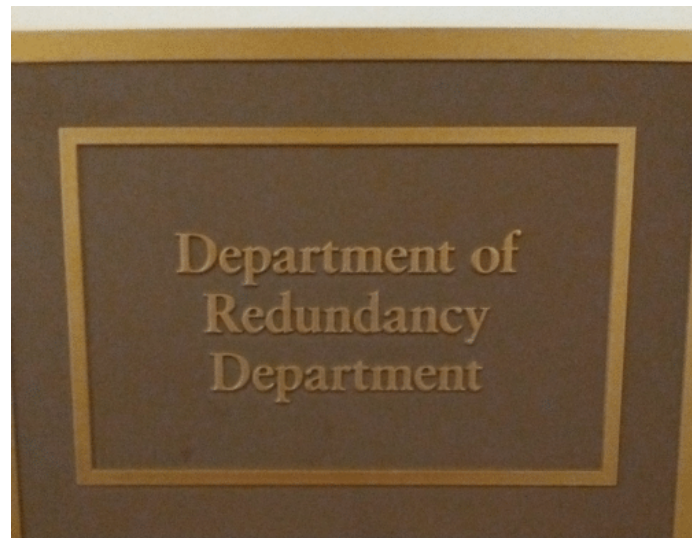
Example:  $\vec{a}_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$   $\vec{a}_2 = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$   $\vec{a}_3 = \begin{bmatrix} 3 \\ 4 \\ 1 \end{bmatrix}$  Can any vector in set be written as lin. combo. of

yes!  $\begin{bmatrix} 3 \\ 4 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + 2 \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$  ← others?

$\vec{a}_3 = \vec{a}_1 + 2\vec{a}_2$  ✓ so the set is lin. dep.!

OR  $\vec{0} = \vec{a}_1 + 2\vec{a}_2 - \vec{a}_3$  

Definition 2: A set of vectors  $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_M\}$  all in  $\mathbb{R}^N$  is linearly dependent if there are  $\alpha_1, \alpha_2, \dots, \alpha_M \in \mathbb{R}$  such that  $\sum_{m=1}^M \alpha_m \vec{a}_m = \vec{0}$    
 scalar coeffs? + exclude trivial case not all  $\alpha_i = 0$   
 ← lin. combos of all vectors =  $\vec{0}$



Are these two defs equivalent? yes!

② What is definition of Linear Independence? NOT lin. dep.

③ True or False: can  $\vec{0}$  be in linearly independent set of vectors?

No! eg.  $\{\vec{0}, \vec{a}_1, \vec{a}_2, \dots, \vec{a}_n\}$

↳ can always write  $\vec{0} = 0\vec{a}_1 + 0\vec{a}_2 + \dots + 0\vec{a}_n$   
and from definition 1 it is lin. dep!

④ Give an example of 3 linearly independent vectors:

e.g.  $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$  \* should be 3D vectors

⑤ Example: Is  $\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ 1 \end{bmatrix} \right\}$  linearly (in)dependent?

3 vectors in 2D space!

one will surely be a lin. combo. of others.

↳ span of first two is  $\mathbb{R}^2$

### Properties of $A\vec{x} = \vec{b}$ systems of Equations

①  $A\vec{x} = \vec{b}$  has a solution iff  $\vec{b} \in \text{span}(\text{cols of } A)$

if and only if

②  $A\vec{x}$  is a linear combination of the columns of  $A$

(column view)

③ unique solution  $\equiv$  unique  $\vec{x}$  that satisfies  $A\vec{x} = \vec{b}$   
is defined as

\* NEW ④ Every  $\vec{b} \in \text{span}(\text{cols of } A)$  is uniquely spanned iff the cols of  $A$  are Linearly Independent

Possible cases:

$\vec{b} \notin \text{span}(\text{cols of } A)$  → no solution  
not in

↳ if (cols  $A$ ) lin. indep. → 1 sol'n

$\vec{b} \in \text{span}(\text{cols of } A)$  → if (cols  $A$ ) lin. dep. →  $\infty$  sol'n



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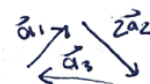
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