

Lecture 5: Vector spaces (continued)

Thursday, September 3, 2015

9:30 AM

Admin:

Aside:

Matrix multiplication in practice: Strassen's algorithm

http://en.wikipedia.org/wiki/Strassen_algorithm

$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \quad B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}$$

$$\begin{aligned} \text{Compute } M_1 &= (A_{11} + A_{22})(B_{11} + B_{22}), & M_5 &= (A_{11} + A_{12})B_{22} \\ M_2 &= (A_{21} + A_{22})B_{11}, & M_6 &= (A_{21} - A_{11})(B_{11} + B_{12}) \\ M_3 &= A_{11}(B_{12} - B_{22}), & M_7 &= (A_{12} - A_{22})(B_{21} + B_{22}) \\ M_4 &= A_{22}(B_{21} - B_{11}) \end{aligned}$$

using 7 matrix multiplications (instead of the 8 obvious ones $A_{11}B_{11} + A_{12}B_{21}, \dots, A_{21}B_{12} + A_{22}B_{22}$).

$$\Rightarrow AB = \begin{pmatrix} M_1 + M_4 - M_5 + M_7 & M_3 + M_5 \\ M_2 + M_4 & M_1 - M_2 + M_3 + M_6 \end{pmatrix}$$

Asymptotic complexity $O(n^{\log_2 7} = 2.81\dots)$ for $n \times n$ matrices.

Method 2: Matrix inversion

• Definition • Existence • Properties

Definition: If A is an $n \times n$ square matrix,
then A^{-1} is the matrix satisfying

$$A \cdot A^{-1} = I,$$

if such a matrix exists.

Examples:

$$I^{-1} = I$$

$$\begin{pmatrix} 1 & 2 \\ 0 & 1 \end{pmatrix}^{-1} = \begin{pmatrix} 1 & -2 \\ 0 & 1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}^{-1} = \text{does not exist!}$$

Exercise: Prove that for a general 2×2 matrix

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

then A^{-1} exists $\iff ad - bc \neq 0$.

If $ad - bc \neq 0$,

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

Properties of the matrix inverse:

- $AA^{-1} = I = A^{-1}A$
Equivalently, $(A^{-1})^{-1} = A$
- If it exists, then A^{-1} is unique.

Proof:

Say X and Y are both inverses of A .

$$\begin{aligned} X &= XI = X(AY) \\ &= (XA)Y \\ &= IY \\ &= Y \quad \checkmark \end{aligned}$$

- If A and B are invertible, so is AB :
 $(AB)^{-1} = B^{-1}A^{-1}$

$$\text{Since } (B^{-1}A^{-1})AB = B^{-1}IB = I \quad \checkmark$$

- If A, B, C are invertible, so is ABC :
 $(ABC)^{-1} = C^{-1}B^{-1}A^{-1}$.

- **Not all matrices are invertible!**

How to invert a matrix

Goal: Solve for the matrix X in

$$AX = I$$

$$\Leftrightarrow A \underbrace{\begin{pmatrix} x_1 & y_1 & z_1 \\ x_2 & y_2 & z_2 \\ x_3 & y_3 & z_3 \end{pmatrix}}_X = \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}}_I$$

$$\Leftrightarrow A \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, A \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, A \begin{pmatrix} z_1 \\ z_2 \\ z_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

Use Gaussian elimination!

Example: $A = \begin{pmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 1 \end{pmatrix}$

$$\left(\begin{array}{ccc|ccc} 2 & -1 & 0 & 1 & 0 & 0 \\ -1 & 2 & -1 & 0 & 1 & 0 \\ 0 & -1 & 1 & 0 & 0 & 1 \end{array} \right) \xrightarrow{-\frac{1}{2}} \left(\begin{array}{ccc|ccc} 2 & -1 & 0 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & -1 & \frac{1}{2} & 1 & 0 \\ 0 & -1 & 1 & 0 & 0 & 1 \end{array} \right) \xrightarrow{\frac{2}{3}}$$

$\underbrace{\quad}_{\text{r.h.s. for 1st system}} \quad \underbrace{\quad}_{\text{2nd system}} \quad \underbrace{\quad}_{\text{3rd system}}$

means $AX = I$

$$\rightarrow \left(\begin{array}{ccc|ccc} 2 & -1 & 0 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & -1 & \frac{1}{2} & 1 & 0 \\ 0 & 0 & \frac{1}{3} & \frac{1}{3} & \frac{2}{3} & 1 \end{array} \right) \uparrow_3$$

$$\rightarrow \left(\begin{array}{ccc|ccc} 2 & -1 & 0 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & 0 & \frac{3}{2} & 3 & 3 \\ 0 & 0 & 1 & 1 & 2 & 3 \end{array} \right) \uparrow_{\frac{2}{3}}$$

$$\rightarrow \left(\begin{array}{ccc|ccc} 2 & 0 & 0 & 2 & 2 & 2 \\ 0 & 1 & 0 & 1 & 2 & 2 \\ 0 & 0 & 1 & 1 & 2 & 3 \end{array} \right) \xrightarrow{\text{divide by 2}}$$

$$\rightarrow \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 1 & 1 \\ 0 & 1 & 0 & 1 & 2 & 2 \\ 0 & 0 & 1 & 1 & 2 & 3 \end{array} \right)$$

$\underbrace{\quad}_x \quad \underbrace{\quad}_y \quad \underbrace{\quad}_z$

means $IX = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 2 \\ 1 & 2 & 3 \end{pmatrix}$

$$\Rightarrow X = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 2 \\ 1 & 2 & 3 \end{pmatrix} \checkmark$$

But: Matrix inversion is not the answer we want.

In practice, we almost **never** invert matrices!

- ① Too slow.
- ② Might not even exist.
- ③ Numerically unstable

Example:

```
>> A = [.01 0; 0 1];
inv(A)
```

ans =

```
100    0
    0    1
```

```
>> A = [.02 0; 0 1];
inv(A)
```

ans =

```
50    0
    0    1
```

- ④ Can take sparse matrices to dense matrices
(\Rightarrow uses too much memory!)

Example:

```
>> n = 10;
e = ones(n-1,1);
A = -2*eye(n) + diag(ones(n-1,1), -1) + diag(ones(n-1,1), 1)
```

A =

```
-2    1    0    0    0    0    0    0    0    0
 1   -2    1    0    0    0    0    0    0    0
 0    1   -2    1    0    0    0    0    0    0
 0    0    1   -2    1    0    0    0    0    0
 0    0    0    1   -2    1    0    0    0    0
 0    0    0    0    1   -2    1    0    0    0
 0    0    0    0    0    1   -2    1    0    0
 0    0    0    0    0    0    1   -2    1    0
 0    0    0    0    0    0    0    1   -2    1
 0    0    0    0    0    0    0    0    1   -2
```

```
>> inv(A)
```

ans =

```
-0.9091 -0.8182 -0.7273 -0.6364 -0.5455 -0.4545 -0.3636 -0.2727 -0.1818 -0.0909
-0.8182 -1.6364 -1.4545 -1.2727 -1.0909 -0.9091 -0.7273 -0.5455 -0.3636 -0.1818
-0.7273 -1.4545 -2.1818 -1.9091 -1.6364 -1.3636 -1.0909 -0.8182 -0.5455 -0.2727
-0.6364 -1.2727 -1.9091 -2.5455 -2.1818 -1.8182 -1.4545 -1.0909 -0.7273 -0.3636
-0.5455 -1.0909 -1.6364 -2.1818 -2.7273 -2.2727 -1.8182 -1.3636 -0.9091 -0.4545
-0.4545 -0.9091 -1.3636 -1.8182 -2.2727 -2.7273 -2.1818 -1.6364 -1.0909 -0.5455
-0.3636 -0.7273 -1.0909 -1.4545 -1.8182 -2.1818 -2.5455 -1.9091 -1.2727 -0.6364
-0.2727 -0.5455 -0.8182 -1.0909 -1.3636 -1.6364 -1.9091 -2.1818 -1.4545 -0.7273
-0.1818 -0.3636 -0.5455 -0.7273 -0.9091 -1.0909 -1.2727 -1.4545 -1.6364 -0.8182
-0.0909 -0.1818 -0.2727 -0.3636 -0.4545 -0.5455 -0.6364 -0.7273 -0.8182 -0.9091
```

Recall:

closed under addition (all vertices)

Vector space = closed under addition (of vectors)
and multiplication (by scalars)
(plus a few other properties)

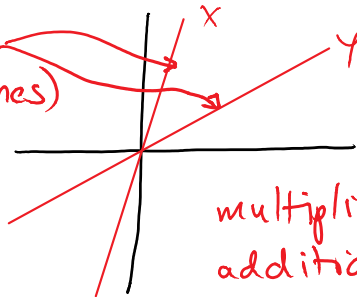
Examples:

- $\mathbb{Z} = \{\text{integers}\}$

$\dots -2 -1 0 1 2 \dots$

NOT a vector space
(closed under addition,
but not multiplication)

- this set
(union of 2 lines)



NOT a vector space

- $\mathbb{Q} = \{\text{rationals } \frac{a}{b}\}$

addition ✓

multiplication X

$(\pi \cdot 1 = \pi)$
scalar \mathbb{Q} \mathbb{Q}

NOT a vector space

Examples: Vector spaces are everywhere!

① vectors

$$\mathbb{R}^3 = \{(a, b, c) \mid a, b, c \in \mathbb{R}\}$$

\mathbb{C}^3 is a vector space over \mathbb{C}
(over \mathbb{R} too)

② matrices

$$\mathbb{R}^{m \times n}$$

③ sets

But these are NOT vector spaces:

$$\{1\}, \{(1, 0, 0)\}$$

$$\{(0, 0), (1, 0)\}$$

the interval $[0, 1]$

④ function spaces

{all functions from $\underset{\mathbb{R}}{S}$ to \mathbb{R} }

f, g

$$(f+g)(x) \equiv f(x) + g(x)$$

$$(\alpha g)(x) = \alpha \cdot g(x)$$

- polynomials of degree $\leq k = 3$

- differentiable functions $\mathbb{R} \rightarrow \mathbb{R}$

⑤ Subspaces!

Exercise: What are ALL subspaces of \mathbb{R}^2 ?

$\{0\}$, lines through 0
 $\{(x,y) : ax+by=0\}$, \mathbb{R}^2 itself

NOT subspaces: other lines, curves

Important: Lines/planes/hyperplanes that don't go through the origin (0) are NOT subspaces!!

⑥ Examples over other fields

\mathbb{F}_p = field of numbers mod p (for a prime p)

$\mathbb{F}_2 = \{0, 1\}$

Bit strings of length n form a vector space:

$n=3$: $(0,0,0), (0,0,1), (0,1,0), (0,1,1)$
 $(1,0,0), (1,0,1), (1,1,0), (1,1,1)$
 addition is coordinate-wise, mod 2
 $(0,0,1) + (0,1,1) = (0,1,0)$

subspace, eg.,

$$\text{Span}(\{(0,0,1), (1,0,1)\}) \\ = \{(0,0,0), (0,0,1), (1,0,1), (1,0,0)\}$$

Problems:

1. How many subspaces are there of \mathbb{R}^2 ?

Answer: Infinitely many!
 (lines through the origin)

2. How many subspaces are there of \mathbb{R} ?

Answer: Two! $\{0\}$ and \mathbb{R} itself.

3. How many subspaces are there of $\{0,1\}^2$?

$\{(0,0)\}$, everything $\{0,1\}^2$

$\{(0,0), (0,1)\}$, $\{(0,0), (1,0)\}$

$\{(0,0), (1,0)\}$, $\{(0,0), (1,1)\}$

and that's it!

(if a subspace contains two of the nonzero points, then it also includes their sum, which is the last nonzero point: $(1,0) + (0,1) + (1,1) = (0,0)$ means that any two sum to the third)

$$1 + 4 + 1 = 6$$

4. How many subspaces are there of $\{0,1\}^n$?

We'll answer this later! Definitely $< \infty$ though

⑧ The **sum** of two subspaces is a subspace.

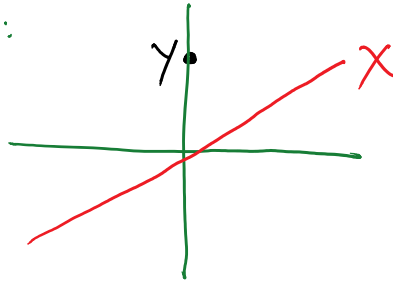
Definition: For two subsets X and Y in a vector space V , let

space V , let

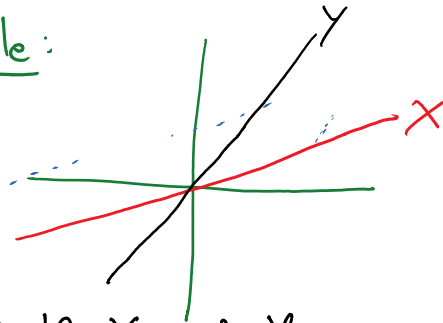
$$X + Y = \{x + y \mid x \in X, y \in Y\}$$

(In English: all sums of a vector in X and a vector in Y .)

Example:



Example:



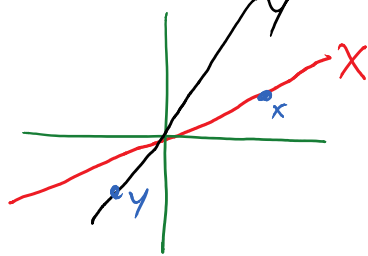
Claim 1: If X and Y are subspaces, then $X + Y$ is a subspace.

Claim 2: For subsets S and T of a vector space V ,

$$S \cup T = \emptyset \quad S \cap T = \emptyset \quad S + T = S \cup T$$

$$\text{Span}(S) + \text{Span}(T) = \text{Span}(S \cup T)$$

Example: From before



$$\begin{aligned} X &= \text{Span}(\{x\}) \\ Y &= \text{Span}(\{y\}) \\ \text{Span}(\{x, y\}) &= \mathbb{R}^2 = X + Y \end{aligned}$$

Definition: **Affine subspace** = translated subspace

ie., a set $\vec{u} + V$ for a vector $\vec{u} \neq \vec{0}$
and subspace V .

③ The **SPAN** of any (finite or infinite) set of points S

Span(S) is defined to be the set of all finite linear combinations of elements from S

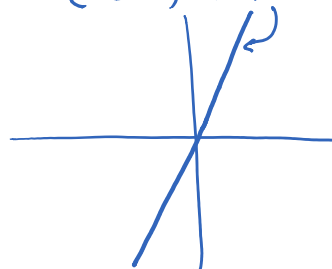
ie., all sums $\alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2 + \dots + \alpha_r \vec{v}_r$
for scalars α_j and $\vec{v}_j \in S$

By definition, this is closed under $+$, \cdot ,
and hence is a vector space.

Examples: What are:

• $\text{Span}\{(1, 2)\} = \text{the line } \{(x, 2x) | x \in \mathbb{R}\}$

• $\text{Span}\{(1, 2), (-1, -2)\}$
= the same line



= the same line $\quad \bigg|$

- $\text{Span}\{(0,1), (1,0)\} = \text{the plane } \mathbb{R}^2$
- $\text{Span}\{1, x, x^2, x^3, \dots\} = \text{all polynomials}$

Note: The infinite sum $\sum_{i=0}^{\infty} x^i$
is not in this span.

- $\text{Span}\left\{\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}\right\} = \text{the } xz\text{-plane in } \mathbb{R}^3$
(since $\begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$,
it doesn't increase the span)

- $\text{Span}\left\{\begin{pmatrix} 1 \\ 2 \\ 1 \\ 1 \\ 5 \end{pmatrix}, \begin{pmatrix} -2 \\ -4 \\ 0 \\ 4 \\ -2 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \\ 2 \\ 4 \\ 9 \end{pmatrix}\right\}$

-this is a subspace of \mathbb{R}^5 — it must be
either a line, a plane, or a 3D hyperplane

Two approaches to find out:

① Add vectors one at a time

$\text{Span}\{(1, 2, 1, 1, 5)\}$ is a line.

Does $(-2, -4, 0, 4, -2)$ give something new,
or is it already in that line?

—something new, since it is not a multiple of $(1, 2, 1, 1, 5)$

Does $(1, 2, 2, 4, 9)$ give something new,
or is it already in the plane $\text{Span}\left\{\begin{pmatrix} 1 \\ 2 \\ 1 \\ 1 \\ 5 \end{pmatrix}, \begin{pmatrix} -2 \\ -4 \\ 0 \\ 4 \\ -2 \end{pmatrix}\right\}$?

It lies in the plane \iff there is a solution to

$$\begin{pmatrix} 1 \\ 2 \\ 2 \\ 4 \\ 9 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 1 \\ 1 \\ 5 \end{pmatrix}x + \begin{pmatrix} -2 \\ -4 \\ 0 \\ 4 \\ -2 \end{pmatrix}y$$

$$(1, 2, 2, 4, 9) = (1, 2, 1, 1, 5)x + (-2, -4, 0, 4, -2)y$$

$$\Leftrightarrow \begin{pmatrix} 1 & -2 \\ 2 & -4 \\ 1 & 0 \\ 1 & 4 \\ 5 & -2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 2 \\ 4 \\ 9 \end{pmatrix} \text{ has a solution}$$

There is a solution: $(x, y) = (2, 1/2)$

Therefore the span is the 2D plane

$$\text{Span} \left\{ \begin{pmatrix} 1 \\ 2 \\ 1 \\ 1 \\ 5 \end{pmatrix}, \begin{pmatrix} -2 \\ -4 \\ 0 \\ 4 \\ -2 \end{pmatrix} \right\} \checkmark$$

② Start with all the vectors, and try to simplify them
This is what Gaussian elimination does:

$$\text{Let } M = \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ -2 & -4 & 0 & 4 & -2 \\ 1 & 2 & 2 & 4 & 9 \end{pmatrix}$$

$$M \xrightarrow{GE} \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ 0 & 0 & 2 & 6 & 8 \\ 0 & 0 & 1 & 3 & 4 \end{pmatrix} \xrightarrow{GE} \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ 0 & 0 & 1 & 3 & 4 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\xrightarrow{GE} \begin{pmatrix} 1 & 2 & 0 & -2 & 1 \\ 0 & 0 & 1 & 3 & 4 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

Note: Adding a multiple of one row to another does not change the span of the rows.

$$\text{Span} \{ \vec{v}_1, \vec{v}_2, \vec{v}_3, \dots \}$$

$$= \text{Span} \{ \vec{v}_1, \vec{v}_2 + \beta \vec{v}_1, \vec{v}_3, \dots \}$$

since anything you can reach with a linear combination $\alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2 + \alpha_3 \vec{v}_3 + \dots$

can also be reached with a linear combination of the new vectors

$$(\alpha_1 - \beta \alpha_2) \vec{v}_1 + \alpha_2 (\vec{v}_2 + \beta \vec{v}_1) + \alpha_3 \vec{v}_3 + \dots,$$

and vice versa.

\Rightarrow The nonzero rows left over after Gaussian elimination (are a minimal set of vectors) that span

the same set as the original rows.

Exercise: Prove (by contradiction) that $\text{Span}(S)$ is the smallest vector space containing S .

Claim: $\text{Span}(S)$ is the **smallest** vector space that contains all the points in S .

Proof:

- Let T be a vector space containing all of S .
 - Let $\vec{v} \in \text{Span}(S)$.
 - $\Rightarrow \vec{v} = \alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2 + \dots + \alpha_r \vec{v}_r$, with all $\vec{v}_i \in S$
 - \Rightarrow all $\vec{v}_i \in T$
 - \Rightarrow all $\alpha_i \vec{v}_i \in T$ (closure under mult.)
 - $\Rightarrow \vec{v} = \sum_i \alpha_i \vec{v}_i \in T$ (closure under addition)
- $\Rightarrow \text{Span}(S) \subseteq T$. \checkmark \square

⑦ Polynomials = $\text{Span}\{1, x, x^2, x^3, \dots\}$

Continuous functions

Differentiable functions

Functions f with $f(1) = 0$, $f'(2) = 0$

More examples of vector spaces

Space	Closed under addition?	Closed under multiplication?	Vector space?
$V_1 = \left\{ (b_1, b_2, b_3) \in \mathbb{R}^3 \mid \text{s.t. } b_1 - 2b_2 + 3b_3 = 0 \right\}$	\checkmark	\checkmark	\checkmark yes
$V_2 = \left\{ (b_1, b_2, b_3) \in \mathbb{R}^3 \mid \text{s.t. } b_2 b_3 = 0 \right\}$	\times	\checkmark	\times no
$\{b \in \mathbb{R}^3 \mid b_1 + b_2 = 1\}$	\times	\times	\times
$\text{Span}(\{(1, 1, 0), (2, 0, 1)\})$	\checkmark	\checkmark	\checkmark

$\text{affine subspace} = \{(1, 0)\} + V_1$

$\text{Span}(\{(1,1,0), (2,0,1)\})$	✓	✓	✓
$\left\{ \begin{array}{l} \text{upper-triangular} \\ m \times n \text{ matrices} \end{array} \right\}$	✓	✓	✓
$\{\text{Diagonal } n \times n \text{ matrices}\}$	✓	✓	✓

$$V_7 = \left\{ 10 \times 10 \text{ matrices } A \right. \\ \left. \text{with } \text{Trace}(A) = \sum_{j=1}^{10} a_{jj} = 0 \right\}$$

✓

✓

✓

$$\left\{ 10 \times 10 \text{ matrices } A \right. \\ \left. \text{with } \text{Tr}(A) = 1 \right\}$$

X

X

affine subspace
 $= V_7 + \left\{ \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \right\}$

$$\left\{ 3 \times 3 \text{ matrices } A \right. \\ \left. \text{with } A \begin{pmatrix} 1 \\ 1 \end{pmatrix} = 0 \right\}$$

✓

✓

✓

$$\left\{ \text{symmetric } n \times n \text{ matrices (ie. } A = A^T) \right\}$$

✓

✓

✓

$$\left\{ \begin{array}{l} \text{arithmetic progressions,} \\ \text{ie., sequences } (x_1, x_2, x_3, \dots) \\ \text{with } x_j - x_{j-1} \text{ constant} \end{array} \right\}$$

(eg., $(0, 2, 4, 6, 8, 10, \dots)$)

✓

✓

✓

$$\left\{ \begin{array}{l} \text{differentiable functions} \\ f: \mathbb{R} \rightarrow \mathbb{R} \text{ with } f'(2) = 3 \end{array} \right\}$$

X

X

affine subspace
 $= 3x + \{f \text{ with } f'(2) = 0\}$

$$\left\{ (0, 0), (2, 1), (4, 2), (6, 3), \dots, (94, 47), (96, 48) \right\}$$

X

X

X

$$\subset (\mathbb{Z}/97\mathbb{Z}) \times (\mathbb{Z}/97\mathbb{Z})$$

(97 is prime)



$$\left\{ \begin{array}{l} \text{all matrices that commute} \\ \text{with a given matrix } A \end{array} \right\}$$

✓

✓

✓

Proof. To prove (4.1.1), demonstrate that the two closure properties **(A1)** and **(M1)** hold for $\mathcal{S} = \mathcal{X} + \mathcal{Y}$. To show **(A1)** is valid, observe that if $\mathbf{u}, \mathbf{v} \in \mathcal{S}$, then $\mathbf{u} = \mathbf{x}_1 + \mathbf{y}_1$ and $\mathbf{v} = \mathbf{x}_2 + \mathbf{y}_2$, where $\mathbf{x}_1, \mathbf{x}_2 \in \mathcal{X}$ and $\mathbf{y}_1, \mathbf{y}_2 \in \mathcal{Y}$. Because \mathcal{X} and \mathcal{Y} are closed with respect to addition, it follows that $\mathbf{x}_1 + \mathbf{x}_2 \in \mathcal{X}$ and $\mathbf{y}_1 + \mathbf{y}_2 \in \mathcal{Y}$, and therefore $\mathbf{u} + \mathbf{v} = (\mathbf{x}_1 + \mathbf{x}_2) + (\mathbf{y}_1 + \mathbf{y}_2) \in \mathcal{S}$. To verify **(M1)**, observe that \mathcal{X} and \mathcal{Y} are both closed with respect to scalar multiplication so that $\alpha \mathbf{x}_1 \in \mathcal{X}$ and $\alpha \mathbf{y}_1 \in \mathcal{Y}$ for all α , and consequently $\alpha \mathbf{u} = \alpha \mathbf{x}_1 + \alpha \mathbf{y}_1 \in \mathcal{S}$ for all α . To prove (4.1.2), suppose $\mathcal{S}_X = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_r\}$ and $\mathcal{S}_Y = \{\mathbf{y}_1, \mathbf{y}_2, \dots, \mathbf{y}_t\}$, and write

$$\begin{aligned} \mathbf{z} \in \text{span}(\mathcal{S}_X \cup \mathcal{S}_Y) &\iff \mathbf{z} = \sum_{i=1}^r \alpha_i \mathbf{x}_i + \sum_{i=1}^t \beta_i \mathbf{y}_i = \mathbf{x} + \mathbf{y} \text{ with } \mathbf{x} \in \mathcal{X}, \mathbf{y} \in \mathcal{Y} \\ &\iff \mathbf{z} \in \mathcal{X} + \mathcal{Y}. \quad \blacksquare \end{aligned}$$

Example 4.1.8

If $\mathcal{X} \subseteq \mathbb{R}^2$ and $\mathcal{Y} \subseteq \mathbb{R}^2$ are subspaces defined by two different lines through the origin, then $\mathcal{X} + \mathcal{Y} = \mathbb{R}^2$. This follows from the parallelogram law—sketch a picture for yourself.

Exercises for section 4.1

4.1.1. Determine which of the following subsets of \mathbb{R}^n are in fact subspaces of \mathbb{R}^n ($n > 2$).

- (a) $\{\mathbf{x} \mid x_i \geq 0\}$, (b) $\{\mathbf{x} \mid x_1 = 0\}$, (c) $\{\mathbf{x} \mid x_1 x_2 = 0\}$,
 (d) $\left\{ \mathbf{x} \mid \sum_{j=1}^n x_j = 0 \right\}$, (e) $\left\{ \mathbf{x} \mid \sum_{j=1}^n x_j = 1 \right\}$,
 (f) $\{\mathbf{x} \mid \mathbf{A}\mathbf{x} = \mathbf{b}, \text{ where } \mathbf{A}_{m \times n} \neq \mathbf{0} \text{ and } \mathbf{b}_{m \times 1} \neq \mathbf{0}\}$.



4.1.2. Determine which of the following subsets of $\mathbb{R}^{n \times n}$ are in fact subspaces of $\mathbb{R}^{n \times n}$.

- (a) The symmetric matrices. (b) The diagonal matrices.
 (c) ~~The nonsingular matrices.~~ (d) ~~The singular matrices.~~
 (e) The triangular matrices. (f) The upper-triangular matrices.
 (g) All matrices that commute with a given matrix \mathbf{A} .
 (h) All matrices such that $\mathbf{A}^2 = \mathbf{A}$.
 (i) All matrices such that $\text{trace}(\mathbf{A}) = 0$.

4.1.3. If \mathcal{X} is a plane passing through the origin in \mathbb{R}^3 and \mathcal{Y} is the line through the origin that is perpendicular to \mathcal{X} , what is $\mathcal{X} + \mathcal{Y}$?

4.1.4. Why must a real or complex nonzero vector space contain an infinite number of vectors?

4.1.5. Sketch a picture in \mathbb{R}^3 of the subspace spanned by each of the following.



(a) $\left\{ \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix}, \begin{pmatrix} 2 \\ 6 \\ 4 \end{pmatrix}, \begin{pmatrix} -3 \\ -9 \\ -6 \end{pmatrix} \right\}$, (b) $\left\{ \begin{pmatrix} -4 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 5 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \right\}$,

(c) $\left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\}$.

4.1.6. Which of the following are spanning sets for \mathbb{R}^3 ?

- (a) $\{(1 \ 1 \ 1)\}$ (b) $\{(1 \ 0 \ 0), (0 \ 0 \ 1)\}$,
 (c) $\{(1 \ 0 \ 0), (0 \ 1 \ 0), (0 \ 0 \ 1), (1 \ 1 \ 1)\}$,
 (d) $\{(1 \ 2 \ 1), (2 \ 0 \ -1), (4 \ 4 \ 1)\}$,
 (e) $\{(1 \ 2 \ 1), (2 \ 0 \ -1), (4 \ 4 \ 0)\}$.

4.1.7. For a vector space \mathcal{V} , and for $\mathcal{M}, \mathcal{N} \subseteq \mathcal{V}$, explain why $\text{span}(\mathcal{M} \cup \mathcal{N}) = \text{span}(\mathcal{M}) + \text{span}(\mathcal{N})$.

skip

4.1.8. Let \mathcal{X} and \mathcal{Y} be two subspaces of a vector space \mathcal{V} .

- (a) Prove that the intersection $\mathcal{X} \cap \mathcal{Y}$ is also a subspace of \mathcal{V} .
 (b) Show that the union $\mathcal{X} \cup \mathcal{Y}$ need not be a subspace of \mathcal{V} .

4.1.9. For $\mathbf{A} \in \mathbb{R}^{m \times n}$ and $\mathcal{S} \subseteq \mathbb{R}^{n \times 1}$, the set $\mathbf{A}(\mathcal{S}) = \{\mathbf{Ax} \mid \mathbf{x} \in \mathcal{S}\}$ contains all possible products of \mathbf{A} with vectors from \mathcal{S} . We refer to $\mathbf{A}(\mathcal{S})$ as the set of *images* of \mathcal{S} under \mathbf{A} .



- (a) If \mathcal{S} is a subspace of \mathbb{R}^n , prove $\mathbf{A}(\mathcal{S})$ is a subspace of \mathbb{R}^m .
 (b) If $\mathbf{s}_1, \mathbf{s}_2, \dots, \mathbf{s}_k$ spans \mathcal{S} , show $\mathbf{As}_1, \mathbf{As}_2, \dots, \mathbf{As}_k$ spans $\mathbf{A}(\mathcal{S})$.



4.1.10. With the usual addition and multiplication, determine whether or not the following sets are vector spaces over the real numbers.

- (a) \mathbb{R} , (b) \mathbb{C} , (c) The rational numbers.

4.1.11. Let $\mathcal{M} = \{\mathbf{m}_1, \mathbf{m}_2, \dots, \mathbf{m}_r\}$ and $\mathcal{N} = \{\mathbf{m}_1, \mathbf{m}_2, \dots, \mathbf{m}_r, \mathbf{v}\}$ be two sets of vectors from the same vector space. Prove that $\text{span}(\mathcal{M}) = \text{span}(\mathcal{N})$ if and only if $\mathbf{v} \in \text{span}(\mathcal{M})$.

4.1.12. For a set of vectors $\mathcal{S} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$, prove that $\text{span}(\mathcal{S})$ is the intersection of all subspaces that contain \mathcal{S} . **Hint:** For $\mathcal{M} = \bigcap_{\mathcal{S} \subseteq \mathcal{V}} \mathcal{V}$, prove that $\text{span}(\mathcal{S}) \subseteq \mathcal{M}$ and $\mathcal{M} \subseteq \text{span}(\mathcal{S})$.

Nullspace is a line
$$\begin{bmatrix} 1 & 0 & 1 \\ 5 & 4 & 9 \\ 2 & 4 & 6 \end{bmatrix} \begin{bmatrix} c & c & -c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}.$$

The nullspace of B is the line of all points $x = c, y = c, z = -c$. (The line goes through the origin, as any subspace must.) We want to be able, for any system $Ax = b$, to find

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The nullspace of B is the line of all points $x = c, y = c, z = -c$. (The line goes through the origin, as any subspace must.) We want to be able, for any system $Ax = b$, to find $C(A)$ and $N(A)$: all attainable right-hand sides b and all solutions to $Ax = 0$.

The vectors b are in the column space and the vectors x are in the nullspace. We shall compute the dimensions of those subspaces and a convenient set of vectors to generate them. We hope to end up by understanding all *four* of the subspaces that are intimately related to each other and to A —the column space of A , the nullspace of A , and their two perpendicular spaces.

Problem Set 2.1

1. Construct a subset of the x - y plane \mathbf{R}^2 that is
 - (a) closed under vector addition and subtraction, but not scalar multiplication.
 - (b) closed under scalar multiplication but not under vector addition.

Hint: Starting with u and v , add and subtract for (a). Try cu and cv for (b).
2. Which of the following subsets of \mathbf{R}^3 are actually subspaces?
 - (a) The plane of vectors (b_1, b_2, b_3) with first component $b_1 = 0$.
 - (b) The plane of vectors b with $b_1 = 1$.
 - (c) The vectors b with $b_2 b_3 = 0$ (this is the union of two subspaces, the plane $b_2 = 0$ and the plane $b_3 = 0$).
 - (d) All combinations of two given vectors $(1, 1, 0)$ and $(2, 0, 1)$.
 - (e) The plane of vectors (b_1, b_2, b_3) that satisfy $b_3 - b_2 + 3b_1 = 0$.
3. Describe the column space and the nullspace of the matrices

$$A = \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 0 & 0 & 3 \\ 1 & 2 & 3 \end{bmatrix} \quad \text{and} \quad C = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

4. What is the smallest subspace of 3 by 3 matrices that contains all symmetric matrices *and* all lower triangular matrices? What is the largest subspace that is contained in both of those subspaces?
5. Addition and scalar multiplication are required to satisfy these eight rules:

1. $x + y = y + x$.
 2. $x + (y + z) = (x + y) + z$.
 3. There is a unique “zero vector” such that $x + 0 = x$ for all x .
 4. For each x there is a unique vector $-x$ such that $x + (-x) = 0$.
 5. $1x = x$.
 6. $(c_1 c_2)x = c_1(c_2 x)$.
 7. $c(x + y) = cx + cy$.
 8. $(c_1 + c_2)x = c_1 x + c_2 x$.
- (a) Suppose addition in \mathbf{R}^2 adds an extra 1 to each component, so that $(3, 1) + (5, 0)$ equals $(9, 2)$ instead of $(8, 1)$. With scalar multiplication unchanged, which rules are broken?
- (b) Show that the set of all positive real numbers, with $x + y$ and cx redefined to equal the usual xy and x^c , is a vector space. What is the “zero vector”?
- (c) Suppose $(x_1, x_2) + (y_1, y_2)$ is defined to be $(x_1 + y_2, x_2 + y_1)$. With the usual $cx = (cx_1, cx_2)$, which of the eight conditions are not satisfied?
6. Let \mathbf{P} be the plane in 3-space with equation $x + 2y + z = 6$. What is the equation of the plane \mathbf{P}_0 through the origin parallel to \mathbf{P} ? Are \mathbf{P} and \mathbf{P}_0 subspaces of \mathbf{R}^3 ?
7. Which of the following are subspaces of \mathbf{R}^∞ ?
- (a) All sequences like $(1, 0, 1, 0, \dots)$ that include infinitely many zeros.
 - (b) All sequences (x_1, x_2, \dots) with $x_j = 0$ from some point onward.
 - (c) All decreasing sequences: $x_{j+1} \leq x_j$ for each j .
 - (d) All convergent sequences: the x_j have a limit as $j \rightarrow \infty$.
 - (e) All arithmetic progressions: $x_{j+1} - x_j$ is the same for all j .
 - (f) All geometric progressions $(x_1, kx_1, k^2 x_1, \dots)$ allowing all k and x_1 .
8. Which of the following descriptions are correct? The solutions x of

$$Ax = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

form

- (a) a plane.
- (b) a line.
- (c) a point.
- (d) a subspace.

- (e) the nullspace of A .

- (e) the nullspace of A .
 (f) the column space of A .
9. Show that the set of nonsingular 2 by 2 matrices is not a vector space. Show also that the set of *singular* 2 by 2 matrices is not a vector space.
10. The matrix $A = \begin{bmatrix} 2 & -2 \\ 2 & -2 \end{bmatrix}$ is a “vector” in the space \mathbf{M} of all 2 by 2 matrices. Write the zero vector in this space, the vector $\frac{1}{2}A$, and the vector $-A$. What matrices are in the smallest subspace containing A ?
11. (a) Describe a subspace of \mathbf{M} that contains $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ but not $B = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$.
 (b) If a subspace of \mathbf{M} contains A and B , must it contain I ?
 (c) Describe a subspace of \mathbf{M} that contains no nonzero diagonal matrices.
12. The functions $f(x) = x^2$ and $g(x) = 5x$ are “vectors” in the vector space \mathbf{F} of all real functions. The combination $3f(x) - 4g(x)$ is the function $h(x) = \underline{\hspace{1cm}}$. Which rule is broken if multiplying $f(x)$ by c gives the function $f(cx)$?
13. If the sum of the “vectors” $f(x)$ and $g(x)$ in \mathbf{F} is defined to be $f(g(x))$, then the “zero vector” is $g(x) = x$. Keep the usual scalar multiplication $cf(x)$, and find two rules that are broken.
14. Describe the smallest subspace of the 2 by 2 matrix space \mathbf{M} that contains
- (a) $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ and $\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$. (b) $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ and $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$.
- (c) $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$. (d) $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$, $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $\begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix}$.
15. Let \mathbf{P} be the plane in \mathbf{R}^3 with equation $x + y - 2z = 4$. The origin $(0, 0, 0)$ is not in \mathbf{P} ! Find two vectors in \mathbf{P} and check that their sum is not in \mathbf{P} .
16. \mathbf{P}_0 is the plane through $(0, 0, 0)$ parallel to the plane \mathbf{P} in Problem 15. What is the equation for \mathbf{P}_0 ? Find two vectors in \mathbf{P}_0 and check that their sum is in \mathbf{P}_0 .
17. The four types of subspaces of \mathbf{R}^3 are planes, lines, \mathbf{R}^3 itself, or \mathbf{Z} containing only $(0, 0, 0)$.
- (a) Describe the three types of subspaces of \mathbf{R}^2 .
 (b) Describe the five types of subspaces of \mathbf{R}^4 .
18. (a) The intersection of two planes through $(0, 0, 0)$ is probably a $\underline{\hspace{1cm}}$ but it could be a $\underline{\hspace{1cm}}$. It can't be the zero vector \mathbf{Z} !
 (b) The intersection of a plane through $(0, 0, 0)$ with a line through $(0, 0, 0)$ is probably a $\underline{\hspace{1cm}}$ but it could be a $\underline{\hspace{1cm}}$.

(c) If \mathbf{S} and \mathbf{T} are subspaces of \mathbf{R}^5 , their intersection $\mathbf{S} \cap \mathbf{T}$ (vectors in both subspaces) is a subspace of \mathbf{R}^5 . Check the requirements on $x + y$ and cx .

19. Suppose \mathbf{P} is a plane through $(0,0,0)$ and \mathbf{L} is a line through $(0,0,0)$. The smallest vector space containing both \mathbf{P} and \mathbf{L} is either ____ or ____.
20. True or false for \mathbf{M} = all 3 by 3 matrices (check addition using an example)?
- (a) The skew-symmetric matrices in \mathbf{M} (with $A^T = -A$) form a subspace.
 - (b) The unsymmetric matrices in \mathbf{M} (with $A^T \neq A$) form a subspace.
 - (c) The matrices that have $(1, 1, 1)$ in their nullspace form a subspace.

Problems 21–30 are about column spaces $C(A)$ and the equation $Ax = b$.

21. Describe the column spaces (lines or planes) of these particular matrices:

$$A = \begin{bmatrix} 1 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 1 & 0 \\ 0 & 2 \\ 0 & 0 \end{bmatrix} \quad \text{and} \quad C = \begin{bmatrix} 1 & 0 \\ 2 & 0 \\ 0 & 0 \end{bmatrix}.$$

22. For which right-hand sides (find a condition on b_1, b_2, b_3) are these systems solvable?

$$(a) \quad \begin{bmatrix} 1 & 4 & 2 \\ 2 & 8 & 4 \\ -1 & -4 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}, \quad (b) \quad \begin{bmatrix} 1 & 4 \\ 2 & 9 \\ -1 & -4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}.$$

23. Adding row 1 of A to row 2 produces B . Adding column 1 to column 2 produces C . A combination of the columns of ____ is also a combination of the columns of A . Which two matrices have the same column ____?

$$A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix} \quad \text{and} \quad C = \begin{bmatrix} 1 & 3 \\ 2 & 6 \end{bmatrix}.$$

24. For which vectors (b_1, b_2, b_3) do these systems have a solution?

$$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}.$$

25. (Recommended) If we add an extra column b to a matrix A , then the column space gets larger unless _____. Give an example in which the column space gets larger and an example in which it doesn't. Why is $Ax = b$ solvable exactly when the column space *doesn't* get larger by including b ?
26. The columns of AB are combinations of the columns of A . This means: *The column space of AB is contained in (possibly equal to) the column space of A .* Give an example where the column spaces of A and AB are not equal.

27. If A is any 8 by 8 invertible matrix, then its column space is _____. Why?

27. If A is any 8 by 8 invertible matrix, then its column space is _____. Why?
28. True or false (with a counterexample if false)?
- (a) The vectors b that are not in the column space $C(A)$ form a subspace.
 - (b) If $C(A)$ contains only the zero vector, then A is the zero matrix.
 - (c) The column space of $2A$ equals the column space of A .
 - (d) The column space of $A - I$ equals the column space of A .
29. Construct a 3 by 3 matrix whose column space contains $(1, 1, 0)$ and $(1, 0, 1)$ but not $(1, 1, 1)$. Construct a 3 by 3 matrix whose column space is only a line.
30. If the 9 by 12 system $Ax = b$ is solvable for every b , then $C(A) = \underline{\hspace{2cm}}$.
31. Why isn't \mathbf{R}^2 a subspace of \mathbf{R}^3 ?

2.2 Solving $Ax = 0$ and $Ax = b$

Chapter 1 concentrated on square invertible matrices. There was one solution to $Ax = b$ and it was $x = -A^{-1}b$. That solution was found by elimination (not by computing A^{-1}). A rectangular matrix brings new possibilities— U may not have a full set of pivots. This section goes onward from U to a reduced form R —**the simplest matrix that elimination can give**. R reveals all solutions immediately.

For an invertible matrix, the nullspace contains only $x = 0$ (multiply $Ax = 0$ by A^{-1}). The column space is the whole space ($Ax = b$ has a solution for every b). The new questions appear when the nullspace contains *more than the zero vector* and/or the column space contains *less than all vectors*:

- Any vector x_n in the nullspace can be added to a particular solution x_p . The solutions to all linear equations have this form, $x = x_p + x_n$:

Complete solution $Ax_p = b$ **and** $Ax_n = 0$ **produce** $A(x_p + x_n) = b$.

- When the column space doesn't contain every b in \mathbf{R}^m , we need the conditions on b that make $Ax = b$ solvable.

A 3 by 4 example will be a good size. We will write down all solutions to $Ax = 0$. We will find the conditions for b to lie in the column space (so that $Ax = b$ is solvable). The 1 by 1 system $0x = b$, one equation and one unknown, shows two possibilities:

$0x = b$ has *no solution* unless $b = 0$. The column space of the 1 by 1 zero matrix contains only $b = 0$.

$0x = 0$ has *infinitely many solutions*. The nullspace contains *all* x . A particular solution is $x_p = 0$, and the complete solution is $x = x_p + x_n = 0 + (\text{any } x)$.

More important examples:
SUBSPACES OF A MATRIX

Let $A \in \mathbb{R}^{m \times n}$ be an $m \times n$ real-valued matrix.

- $\text{Range}(A)$

Observe: $\vec{b} \in \text{Range}(A) \iff A\vec{x} = \vec{b}$ has a solution

- $\text{Range}(A^T)$
- $\text{Kernel}(A)$