

Vector spaces

ROB 501

Necmiye Ozay

- **Vector space over a field**
 - **Theorem relating bases and dimension for finite dimensional vector spaces**
 - **Representations of vectors**
 - **Change of basis matrix**
 - **Linear operators (if time)**
 - **Matrix representations of linear operators**

Course Announcements

- Midterm exam is on October 25, from 6:30pm-9pm (in-person exam)
- Final exam is on December 19 (+/- 12 hours). Take home exam. Details (duration, etc.) are being sort out. Make sure you have good internet connection on that date.

Recap

- Last week we defined:
 - Linear combinations
 - Linear independence (of finite and infinite sets)
 - Subspaces
 - Span
 - Basis
 - Dimension

Dimension

The maximal number of elements in any linearly independent set of vector in $(\mathcal{X}, \mathcal{F})$, is called the **dimension** of $(\mathcal{X}, \mathcal{F})$.

Basis

A set of vectors B in $(\mathcal{X}, \mathcal{F})$ is a basis if

1. B is linearly independent
2. $\text{span}\{B\} = \mathcal{X}$

Theorem: In an n – dimensional vector space **ANY** set of n linearly independent vectors is a basis.

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want to show $\{v^1, \dots, v^n\}$ is a basis.

Proof: Let $(\mathcal{X}, \mathcal{F})$ be n -dimensional and let $\{v^1, \dots, v^n\}$ be a linearly independent set.

To Show: $\forall x \in \mathcal{X}, \exists \alpha_1, \dots, \alpha_n \in \mathcal{F}$ such that $x = \alpha_1 v^1 + \dots + \alpha_n v^n$, i.e., $\mathcal{X} = \text{span}\{v^1, \dots, v^n\}$

How: Because $(\mathcal{X}, \mathcal{F})$ is n -dimensional, $\{x, v^1, \dots, v^n\}$ is linearly dependent. ^① Otherwise, the $\dim \mathcal{X} > n$ which it isn't. Hence, $\exists \beta_0, \beta_1, \dots, \beta_n \in \mathcal{F}$, NOT ALL ZERO, such that $\beta_0 x + \beta_1 v^1 + \dots + \beta_n v^n = 0$. ^②

① by def'n. of dimension

② by def'n of being linearly dependent.

Claim: $\beta_0 \neq 0$

Proof of the claim: Assume by contradiction that $\beta_0 = 0$.

Then, 1) At least one of β_1, \dots, β_n is non-zero.

2) $\beta_1 v^1 + \dots + \beta_n v^n = 0$. (this is ②)

1) and 2) together imply $\{v^1, \dots, v^n\}$ is linearly dependent

\rightarrow contradiction $\therefore \beta_0 \neq 0$

Theorem: In an n -dimensional vector space ANY set of n linearly independent vectors is a basis.

Proof: Let $(\mathcal{X}, \mathcal{F})$ be n -dimensional and let $\{v^1, \dots, v^n\}$ be a linearly independent set.

To Show: $\forall x \in \mathcal{X}, \exists \alpha_1, \dots, \alpha_n \in \mathcal{F}$ such that $x = \alpha_1 v^1 + \dots + \alpha_n v^n$, i.e., $\mathcal{X} = \text{span}\{v^1, \dots, v^n\}$

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Claim: $\beta_0 \neq 0$

Proof: Suppose that $\beta_0 = 0$. Then

1. At least one of β_1, \dots, β_n is non-zero
2. $\beta_1 v^1 + \dots + \beta_n v^n = 0$

1 and 2 above, imply that $\{v^1, \dots, v^n\}$ is linearly dependent, which is a contradiction. Hence $\beta_0 = 0$ cannot hold. Completing the proof, we write

$$\boxed{\beta_0 x = -\beta_1 v^1 - \dots - \beta_n v^n}$$

$$x = \underbrace{\left(\frac{-\beta_1}{\beta_0}\right)}_{\alpha_1} v^1 + \dots + \underbrace{\left(\frac{-\beta_n}{\beta_0}\right)}_{\alpha_n} v^n$$

$$\therefore \alpha_1 = \frac{-\beta_1}{\beta_0}, \dots, \alpha_n = \frac{-\beta_n}{\beta_0}$$

Side question
why $\frac{\beta_i}{\beta_0} \in \mathcal{F}$
 $\frac{1}{\beta_0}$ exists (multiplicative inverse)
 $\frac{1}{\beta_0} \cdot \beta_i \in \mathcal{F}$ (closedness under multi.)

Proposition Let $(\mathcal{X}, \mathcal{F})$ be a vector space and suppose that $B = \{b^1, b^2, \dots\}$ is a basis for $(\mathcal{X}, \mathcal{F})$. Let $x \in \mathcal{X}$ and suppose that

$$x = \alpha_1 b^1 + \dots + \alpha_k b^k \quad (1)$$

and

$$x = \bar{\alpha}_1 b^1 + \dots + \bar{\alpha}_k b^k \quad (2)$$

Then, $\alpha_i = \bar{\alpha}_i$ for all $1 \leq i \leq k$.

$$0 = \underset{(1)}{x} - \underset{(2)}{x} = (\alpha_1 b^1 + \dots + \alpha_k b^k) - (\bar{\alpha}_1 b^1 + \dots + \bar{\alpha}_k b^k)$$

$$0 = (\alpha_1 - \bar{\alpha}_1) b^1 + \dots + (\alpha_k - \bar{\alpha}_k) b^k$$

Note that b^1, \dots, b^k are linearly independent
(because they are in B and B is a basis).
 $\left. \begin{array}{l} \downarrow \\ 1, 0 \end{array} \right\} \alpha_i = \bar{\alpha}_i$

Proposition Let $(\mathcal{X}, \mathcal{F})$ be a vector space and suppose that $B = \{b^1, b^2, \dots\}$ is a basis for $(\mathcal{X}, \mathcal{F})$. Let $x \in \mathcal{X}$ and suppose that

$$x = \alpha_1 b^1 + \dots + \alpha_k b^k$$

and

$$x = \bar{\alpha}_1 b^1 + \dots + \bar{\alpha}_k b^k$$

Then, $\alpha_i = \bar{\alpha}_i$ for all $1 \leq i \leq k$.

Proof:

$$\begin{aligned} 0 = x - x &= (\alpha_1 b^1 + \dots + \alpha_k b^k) - (\bar{\alpha}_1 b^1 + \dots + \bar{\alpha}_k b^k) \\ &= (\alpha_1 - \bar{\alpha}_1) b^1 + \dots + (\alpha_k - \bar{\alpha}_k) b^k \end{aligned}$$

Because $\{b^1, \dots, b^k\} \subset B$ implies that $\{b^1, \dots, b^k\}$ is linearly independent, we deduce that $\alpha_i - \bar{\alpha}_i = 0$ for all $1 \leq i \leq k$.

Representations of Vectors

Let (X, \mathbb{F}) be an n -dimensional vector space.
Let $\mathbf{v} = \{v^1, \dots, v^n\}$ be a basis. And let $x \in X$

$$\underline{x \in X} \iff x = \alpha_1 v^1 + \dots + \alpha_n v^n$$

$$[x]_{\mathbf{v}}$$

notation

for representation
of x wrt basis \mathbf{v}

$$= \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix} \in \mathbb{F}^n$$

unique!
(by the previous
proposition)

$$[x]_{\mathbf{v}} = \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix} \in \mathbb{F}^n$$

Representations of Vectors

Example: $\mathcal{F} = \mathbb{R}$, $\mathcal{X} = \{2 \times 2 \text{ matrices with real coefficients}\}$

“natural”

$v = \{v^1, v^2, v^3, v^4\}$

Basis 1: $v^1 = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$, $v^2 = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$, $v^3 = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$, $v^4 = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$

Basis 2: $w^1 = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$, $w^2 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$, $w^3 = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$, $w^4 = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$

$w = \{w^1, w^2, w^3, w^4\}$

Consider:

$$x = \begin{bmatrix} 5 & 3 \\ 1 & 4 \end{bmatrix}$$

$$[x]_v = ?$$

$$[x]_w = ?$$

$$x = \begin{bmatrix} 5 & 3 \\ 1 & 4 \end{bmatrix} = 5v^1 + 3v^2 + 1 \cdot v^3 + 4 \cdot v^4$$

$$[x]_v = \begin{bmatrix} 5 \\ 3 \\ 1 \\ 4 \end{bmatrix} \in \mathbb{R}^4$$

$$x = \begin{bmatrix} 5 & 3 \\ 1 & 4 \end{bmatrix} = 5w^1 + 2w^2 + 1w^3 + 4w^4$$

$$[x]_w = \begin{bmatrix} 5 \\ 2 \\ 1 \\ 4 \end{bmatrix} \in \mathbb{R}^4$$

Facts:

1. Addition of vectors in $(X, \mathcal{F}) \iff$
Addition of representations in $(\mathcal{F}^n, \mathcal{F})$

$$\forall x, y \in X \quad [x+y]_v = [x]_v + [y]_v$$

2. Same for scalar multiplication:

$$\forall x \in X, \forall \alpha \in \mathcal{F} \quad [\alpha x]_v = \alpha [x]_v$$

3. Once you fix a basis, any n -dimensional vector space (X, \mathcal{F}) "looks like" $(\mathcal{F}^n, \mathcal{F})$

Ex: $X = \mathcal{P}_3(t) = \{ \text{all polynomials w/ real coefficients w/ degree} \leq 3 \}$

$$\mathcal{F} = \mathbb{R}$$

$u = \{ \underset{u^1}{1}, \underset{u^2}{t}, \underset{u^3}{t^2}, \underset{u^4}{t^3} \}$ is a basis for $(\mathcal{P}_3(t), \mathbb{R})$

Consider $p_1(t) = t + 2t^2 + 3t^3 \in \mathcal{P}_3(t)$

$$p_1(t) = t + 2t^2 + 3t^3 = 0 \cdot u^1 + 1 \cdot u^2 + 2 \cdot u^3 + 3 \cdot u^4$$

$$[p_1]_u = \begin{bmatrix} 0 \\ 1 \\ 2 \\ 3 \end{bmatrix} \in \mathbb{R}^4$$

$$\bar{u} = \{ \bar{u}^1 = 1, \bar{u}^2 = t, \bar{u}^3 = t+2t^2, \bar{u}^4 = \frac{t^3}{2} \}$$
 is also a basis for $(\mathcal{P}_3(t), \mathbb{R})$

$$[p_1]_{\bar{u}} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 6 \end{bmatrix} \in \mathbb{R}^4$$

$$p_1(t) = t + 2t^2 + 3t^3 = \bar{\alpha}_1 \cdot \bar{u}^1 + \bar{\alpha}_2 \cdot \bar{u}^2 + \bar{\alpha}_3 \cdot \bar{u}^3 + \bar{\alpha}_4 \cdot \bar{u}^4$$

Change of Basis Matrix

- We are given a finite dimensional vector space $(\mathcal{X}, \mathcal{F})$ and two bases $\{u\} = \{u^1, u^2, \dots, u^m\}$ and $\{\bar{u}\} = \{\bar{u}^1, \bar{u}^2, \dots, \bar{u}^m\}$.

$$\alpha = [x]_u = \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_m \end{bmatrix} \longleftrightarrow x = \alpha_1 u^1 + \dots + \alpha_m u^m \quad \bar{\alpha} = [x]_{\bar{u}} = \begin{bmatrix} \bar{\alpha}_1 \\ \vdots \\ \bar{\alpha}_m \end{bmatrix} \longleftrightarrow x = \bar{\alpha}_1 \bar{u}^1 + \dots + \bar{\alpha}_m \bar{u}^m$$

Theorem: There exists an invertible $n \times n$ matrix P with coefficients in \mathcal{F} such that $\forall x \in \mathcal{X}$

$$[x]_{\bar{u}} = P \cdot [x]_u.$$

Moreover, $P = [P_1 \mid P_2 \mid \dots \mid P_n]$ with $P_i = [u^i]_{\bar{u}}$

Proof: $x = \sum_{i=1}^n \alpha_i u^i = \sum_{i=1}^n \bar{\alpha}_i \bar{u}^i$

$$\therefore [x]_u = \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix} \doteq \alpha \quad [x]_{\bar{u}} = \begin{bmatrix} \bar{\alpha}_1 \\ \vdots \\ \bar{\alpha}_n \end{bmatrix} \doteq \bar{\alpha}$$

$$\bar{\alpha} = [x]_{\bar{u}} = \left[\sum_{i=1}^n \alpha_i u^i \right]_{\bar{u}} \quad \text{by Fact 1 and 2 of representations.}$$

$$= \sum_{i=1}^n \alpha_i \underbrace{[u^i]_{\bar{u}}}_{P_i}$$

$$= \sum_{i=1}^n \alpha_i P_i = \underbrace{[P_1 | \dots | P_n]}_P \underbrace{\begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix}}_{\alpha}$$

$$\therefore \bar{\alpha} = P \cdot \alpha, \text{ where } P_i = [u^i]_{\bar{u}}$$

By the same reasoning, we can find \bar{P} s.t. $\alpha = \bar{P} \bar{\alpha}$,
 where $\bar{P}_i = [\bar{u}^i]_u$.

$$\text{Hence, } \bar{\alpha} = P \underbrace{\alpha}_{\bar{P} \cdot \bar{\alpha}} = P \bar{P} \bar{\alpha} \Rightarrow P \bar{P} = I$$

$$\text{Similarly } \alpha = \underbrace{\bar{P} \bar{\alpha}}_{P \alpha} = \bar{P} P \alpha \Rightarrow \bar{P} P = I$$

$$\bar{P} = P^{-1}$$

Example: $\mathcal{X} = \{2 \times 2 \text{ matrices with real coefficients}\}$, $\mathcal{F} = \mathbb{R}$.

$$\boxed{u} = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\} \leftarrow$$

$$\bar{u} = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$$

We have the following relations:

$$\bar{\alpha} = P\alpha, P_i = [u^i]_{\bar{u}}, \quad \alpha = \bar{P}\bar{\alpha}, \bar{P}_i = [\bar{u}^i]_u, \quad \bar{P}^{-1} = P, \quad P^{-1} = \bar{P}$$

Typically, compute the easier of P or \bar{P} , and compute the other by inversion. For this example, we choose to compute \bar{P}

$$\bar{P}_1 = [\bar{u}^1]_u = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\bar{P}_2 = [\bar{u}^2]_u = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix}$$

$$\bar{P}_3 = [\bar{u}^3]_u = \begin{bmatrix} 0 \\ 1 \\ -1 \\ 0 \end{bmatrix}$$

$$\bar{P}_4 = [\bar{u}^4]_u = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\bar{u}^1 = 1 \cdot u^1 + 0 \cdot u^2 + 0 \cdot u^3 + 0 \cdot u^4$$

$$\bar{u}^2 = 0 \cdot u^1 + 1 \cdot u^2 + 1 \cdot u^3 + 0 \cdot u^4$$

\vdots

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

Therefore, $\bar{P} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$ and $P = \bar{P}^{-1} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & .5 & .5 & 0 \\ 0 & .5 & -.5 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$

What if we did it the other direction?

$$\begin{aligned}
 P_1 = [u^1]_{\bar{u}} &= \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \leftrightarrow \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}}_{u^1} = 1 \cdot \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}}_{\bar{u}^1} + 0 \cdot \underbrace{\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}}_{\bar{u}^2} + 0 \cdot \underbrace{\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}}_{\bar{u}^3} + 0 \cdot \underbrace{\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}}_{\bar{u}^4} \\
 P_2 = [u^2]_{\bar{u}} &= \begin{bmatrix} 0 \\ .5 \\ .5 \\ 0 \end{bmatrix} \leftrightarrow \underbrace{\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}}_{u^2} = 0 \cdot \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + 0.5 \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} + .5 \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} + 0 \cdot \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \\
 P_3 = [u^3]_{\bar{u}} &= \begin{bmatrix} 0 \\ .5 \\ -.5 \\ 0 \end{bmatrix} \leftrightarrow \underbrace{\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}}_{u^3} = 0 \cdot \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + 0.5 \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} - .5 \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} + 0 \cdot \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \\
 P_4 = [u^4]_{\bar{u}} &= \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \leftrightarrow \underbrace{\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}}_{u^4} = 0 \cdot \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + 0 \cdot \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} + 0 \cdot \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} + 1 \cdot \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}
 \end{aligned}$$

↙

Therefore, $P = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & .5 & .5 & 0 \\ 0 & .5 & -.5 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$ and $\bar{P} = P^{-1} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$

Exercise

earlier this

Representation example from ~~last~~ lecture:

$$x = \begin{bmatrix} 5 & 3 \\ 1 & 4 \end{bmatrix}$$

$$[x]_u = \begin{bmatrix} 5 \\ 3 \\ 1 \\ 4 \end{bmatrix}$$

$$[x]_{\bar{u}} = \begin{bmatrix} 5 \\ 2 \\ 1 \\ 4 \end{bmatrix}$$

$$[x]_{\bar{u}} = P \cdot [x]_u$$
$$[x]_u = \bar{P} \cdot [x]_{\bar{u}}$$

Check that the P we computed works in relating the two representations (it works for any x)

OFFICE HOURS.

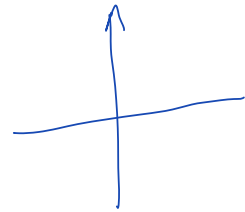
Fact: For any $S \subset X$,

$\text{span}(S)$ is a subspace of X .

And it is the smallest subspace that contains S .

$$\begin{aligned} [x]_{\bar{u}} &= \left[\sum_{i=1}^n \bar{\alpha}_i \bar{u}^i \right]_{\bar{u}} \\ &= \sum_{i=1}^n \bar{\alpha}_i \underbrace{[\bar{u}^i]_{\bar{u}}}_{e^i} \\ &= \sum_{i=1}^n \bar{\alpha}_i e^i = \begin{bmatrix} \bar{\alpha}_1 \\ \vdots \\ \bar{\alpha}_n \end{bmatrix} \end{aligned}$$

$$e^i = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \text{ with } i\text{th element}$$



In Q1: $x \in \mathbb{R}^n$
 $x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$ $x_i \in \mathbb{R}$

$$x^{(1)} = \begin{bmatrix} x_1^{(1)} \\ x_2^{(1)} \\ \vdots \\ x_n^{(1)} \end{bmatrix} \in S_c \quad x^{(1)}, x_2^{(1)} = 0$$

$$x^{(2)} \in S_c \quad x^{(2)} = \begin{bmatrix} x_1^{(2)} \\ x_2^{(2)} \\ \vdots \\ x_n^{(2)} \end{bmatrix} \quad x_1^{(2)}, x_2^{(2)} = 0$$

$$\alpha x^{(1)} + x^{(2)} = x^{(3)} = \begin{bmatrix} x_1^{(3)} \\ x_2^{(3)} \\ \vdots \\ x_n^{(3)} \end{bmatrix}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \in S_{(d)} \quad \left(\text{since } x \in S_{(d)}, \text{ we know } x_1 + \dots + x_n = 0 \right)$$

scalar multiplication

$$\alpha \in \mathbb{R} \\ \bar{x} = \alpha x = \begin{bmatrix} \alpha x_1 \\ \vdots \\ \alpha x_n \end{bmatrix} = \begin{bmatrix} \bar{x}_1 \\ \vdots \\ \bar{x}_n \end{bmatrix} \quad \begin{aligned} \sum \bar{x}_i &= \alpha \sum x_i \\ &= 0 \end{aligned}$$

$$\Rightarrow \bar{x} \in S_{(d)}$$

$$\underline{x \in S_{(f)}} \quad \text{iff} \quad \underline{Ax = b}$$

$$\bar{x} \in S_{(f)} \quad \text{iff} \quad A\bar{x} = b$$

does $\alpha x = z$ satisfy $Az = b$?
 $A\alpha x = \alpha Ax = \alpha b$

is $x + \bar{x} \in S_{(f)}$ meaning

does $x + \bar{x} = y$ satisfy $Ay = b$?

$$Ay = A(x + \bar{x}) = Ax + A\bar{x}$$

Step 1:

Take $x \in \text{Span}(S_1 \cup S_2)$, and

show $x \in \text{Span}(S_1) + \text{Span}(S_2)$.

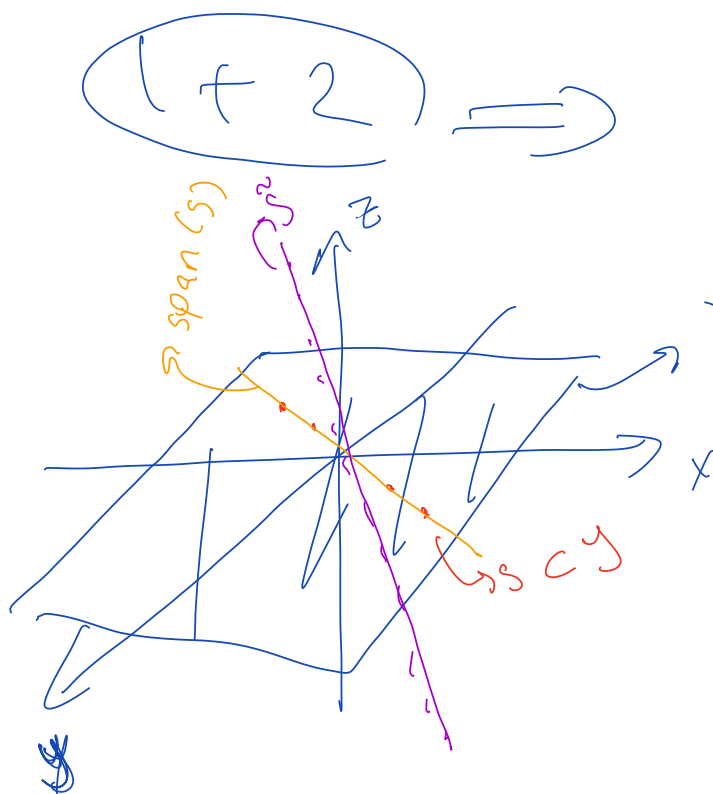
(shows $\text{Span}(S_1 \cup S_2) \subset \text{Span}(S_1) + \text{Span}(S_2)$)

Step 2:

Take $y \in \text{Span}(S_1) + \text{Span}(S_2)$, and

show $y \in \text{Span}(S_1 \cup S_2)$ (shows

$\text{Span}(S_1) + \text{Span}(S_2) \subset \text{Span}(S_1 \cup S_2)$)



$(1 + 2) \Rightarrow$

$$Y = \left\{ \begin{bmatrix} x \\ y \\ z \end{bmatrix} \mid z=0 \right\}$$

want to show
 $\text{span}(S) \subset Y$

step 1 $\rightarrow S \subset Y$
 $\text{span}(S) \subset \text{span}(Y)$

step 2
 Y subspace?
 $(\text{span}(Y) = Y)$