

Ex 1

First consider \mathbb{R} and verify the definition. If $x, y \in \mathbb{R}$, then $x+y \in \mathbb{R}$. Also if $x \in \mathbb{R}$ and $\alpha \in \mathbb{F}$ (which we take to also be \mathbb{R}), then $\alpha x \in \mathbb{R}$.

Next, the $y=0$ axis in \mathbb{R}^2 is all vectors of the form $\begin{bmatrix} x_1 \\ 0 \end{bmatrix}$.

Let $u = \begin{bmatrix} u_1 \\ 0 \end{bmatrix}$, $v = \begin{bmatrix} v_1 \\ 0 \end{bmatrix}$, $\alpha \in \mathbb{R}$. Then

$$u+v = \begin{bmatrix} u_1+v_1 \\ 0 \end{bmatrix}$$

which is on the $y=0$ axis. Further $\alpha u = \begin{bmatrix} \alpha u_1 \\ 0 \end{bmatrix}$ also is.

Ex 2

The "linear combination of columns" view of matrix-vector multiplication is most appropriate. Matrix-matrix multiplication versions 2 and 3 are both appropriate.

Ex 3

$$\text{span}(\{u_1, \dots, u_n\}) = \left\{ y = \sum_{i=1}^n \alpha_i u_i : \alpha_i \in \mathbb{F} \right\}$$

Now take two vectors $x, y \in \text{span}(\{u_1, \dots, u_n\})$, and $\gamma \in \mathbb{F}$.

Then we can write

$$x = \sum_{i=1}^n \alpha_i u_i, \quad y = \sum_{i=1}^n \beta_i u_i$$

for some $\alpha_1, \dots, \alpha_n, \beta_1, \dots, \beta_n$. Therefore

$$\begin{aligned} x+y &= \sum_{i=1}^n \alpha_i u_i + \sum_{i=1}^n \beta_i u_i \\ &= \sum_{i=1}^n (\alpha_i + \beta_i) u_i = \sum_{i=1}^n \delta_i u_i \\ &\quad \underbrace{\hspace{1.5cm}}_{\text{call this } \delta_i} \end{aligned}$$

So $x+y \in \text{span}(\{u_1, \dots, u_n\})$. Further,

$$\begin{aligned} \gamma x &= \gamma \sum_{i=1}^n \alpha_i u_i \\ &= \sum_{i=1}^n \gamma \alpha_i u_i \\ &= \sum_{i=1}^n (\underbrace{\gamma \alpha_i}_{\xi_i}) u_i = \sum_{i=1}^n \xi_i u_i \end{aligned}$$

So $\gamma x \in \text{span}(\{u_1, \dots, u_n\})$, and it is therefore a subspace.

Ex 4

Let $u \in \mathbb{R}^3$ be a vector on the (x, z) plane. Any such vector has the form $u = \begin{bmatrix} u_1 \\ 0 \\ u_3 \end{bmatrix}$. We can write any such vector as

$$u = \frac{1}{2}(u_1 + u_3) \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + \frac{1}{2}(u_1 - u_3) \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

Since u was arbitrary, the vectors span the (x, z) plane.

Ex 5

a) Yes. If not, we could write some c_j as

$$c_j = \begin{bmatrix} a_j \\ b_j \end{bmatrix} = \sum_{i=1}^k \alpha_i \begin{bmatrix} a_i \\ b_i \end{bmatrix}$$

but this would require $a_j = \sum_{i=1}^k \alpha_i a_i$, contradicting the linear independence of the a_i 's.

b) Not necessarily. If b_1, \dots, b_k are linearly independent, then so are c_1, \dots, c_k . For example, let $a_1, \dots, a_k = 1$ ($a_i \in \mathbb{R}$). Then

$$c_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \text{ and } c_2 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \text{ are linearly independent.}$$

Ex 6

Let $x, y \in \mathbb{R}^n$ be orthogonal, and suppose they are linearly dependent.

Then we can write $x = \alpha y$ for some $\alpha \in \mathbb{R}$, and so

$$x^T y = (\alpha y)^T y = \alpha y^T y = \alpha \|y\|^2.$$

Assuming $y \neq 0$, $\|y\|^2 > 0$, contradicting the orthogonality of x and y .

Ex 7

Suppose they are linearly dependent so that $\exists \beta_1, \beta_2, \beta_3 \in \mathbb{R}$

such that $\beta_1 a_1 + \beta_2 a_2 + \beta_3 a_3 = 0$. This would mean

$$\beta_1 - \beta_3 = 0 \quad -\beta_2 + \beta_3 = 0 \quad \beta_2 + \beta_3 = 0$$

The last two imply $2\beta_3 = -0$, so $\beta_3 = 0$, which implies

$\beta_1 = 0$ and $\beta_2 = 0$. Therefore, the only way to have

$\beta_1 a_1 + \beta_2 a_2 + \beta_3 a_3 = 0$ is if $\beta_1 = \beta_2 = \beta_3 = 0$, from which

we conclude that a_1, a_2, a_3 are linearly independent.

Ex 8

$$\begin{aligned}\|x\|^2 &= x^T x = \left(\sum_{i=1}^k \beta_i a_i \right)^T \left(\sum_{j=1}^k \beta_j a_j \right) \\ &= \sum_{i=1}^k \beta_i^2 \underbrace{a_i^T a_i}_1 + \sum_{\substack{i=1 \\ j \neq i}}^k \beta_i \beta_j \underbrace{a_i^T a_j}_0 \\ &= \sum_{i=1}^k \beta_i^2 = \|\beta\|^2\end{aligned}$$

Therefore $\|x\| = \|\beta\|$.

Ex 9

- First, it is obvious that the vectors are linearly independent. Now to show they span \mathbb{R}^n , note that for any $x \in \mathbb{R}^n$,

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = x_1 e_1 + x_2 e_2 + \dots + x_n e_n.$$

- The vectors are already shown to be linearly independent. We could simply note that $\dim(\mathbb{R}^3) = 3$, so the vectors must span \mathbb{R}^3 . More directly, for $x \in \mathbb{R}^3$,

$$\begin{aligned}
 x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} &= x_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + x_2 \left(- \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} + \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) \\
 &+ x_3 \left(\frac{1}{2} \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} \right)
 \end{aligned}$$

$$= a_1 \left(x_1 + x_2 + \frac{1}{2} x_3 \right) + a_2 \left(-x_2 + \frac{1}{2} x_3 \right) + a_3 \left(x_2 + \frac{1}{2} x_3 \right)$$

Ex 10

Among u_1, \dots, u_N , the maximum number of linearly independent vectors is N , so the maximum number of basis vectors for $\text{span}(\{u_1, \dots, u_N\})$ is N .

Ex 11

- $\dim(S) = 1$

- $S + \tau = \left\{ s + t : s \in \text{span} \left(\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right), t \in \text{span} \left(\begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \right) \right\}$

These are all vectors of the form $\begin{bmatrix} a \\ 0 \\ b \end{bmatrix}$, so $S + \tau$ is the (x, z) plane, and $\dim(S + \tau) = 2$.

$$\bullet S \cap \tau = \{ v \in \mathbb{R}^3 : v \in \text{span} \left(\left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right\} \right), v \in \text{span} \left(\left\{ \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \right\} \right) \}$$

Note that these vectors are orthogonal, so $S \cap \tau = \{0\}$.

This has infinite dimension.

Ex 12

Since all vectors are in \mathbb{R}^3 , the maximum dimension is 3, in which case the vectors in $S + \tau$ would span \mathbb{R}^3 . We need to check if any three are linearly independent. We know that τ spans the (x, z) plane, so we need to see if either vector in S is linearly independent of these two. We need to see if

$$\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \alpha_1 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + \alpha_2 \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

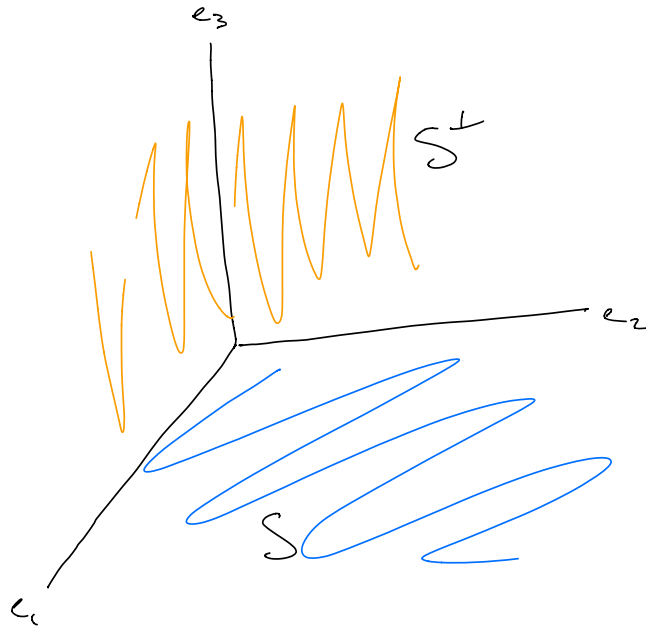
but this is clearly impossible, so $\dim(S + \tau) = 3$ and $S + \tau$ spans \mathbb{R}^3 .

Next note that

$$\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \in \tau$$

So $\begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \in S \cap \tau \Rightarrow S + \tau$ is NOT a direct sum.

Ex 13



Ex 14

Note that $\mathcal{R}(A) = \{y = Ax : x \in \mathbb{R}^n\}$, so $\mathcal{R}(A)$ is a subspace of \mathbb{R}^m . By definition, S^\perp is the subspace of vectors orthogonal to $\mathcal{R}(A)$. Since we have to be able to compute $\langle v, y \rangle$ for $y \in \mathcal{R}(A)$, we must have $v \in \mathbb{R}^m$ as well.

Therefore, vectors in S^\perp have dimension m .

Ex 15

Since $\mathcal{R}(A) \subset \mathbb{R}^m$, its maximum dimension is m . Recall that

$$\mathcal{R}(A) = \text{span}(\text{columns of } A)$$

and A has n columns. Therefore, to have $\dim(\mathcal{R}(A)) = m$, A must have at least m linearly independent columns.