

Subspaces Is $S = \left\{ \begin{bmatrix} x \\ y \\ z \end{bmatrix} \mid x^2 + y^2 = z^2 \right\}$ a subspace?

① Is it closed under scalar multiplication?

If $\begin{bmatrix} x \\ y \\ z \end{bmatrix}$ is in S ($x^2 + y^2 = z^2$)

then for any c in \mathbb{R} $\begin{bmatrix} cx \\ cy \\ cz \end{bmatrix}$ satisfies ~~($x^2 + y^2 = z^2$)~~

$$\begin{aligned} (cx)^2 + (cy)^2 &= c^2x^2 + c^2y^2 \\ &= c^2(x^2 + y^2) \\ &= c^2z^2 \\ &= (cz)^2 \end{aligned}$$

so $(cx)^2 + (cy)^2 = (cz)^2$ so $\begin{bmatrix} cx \\ cy \\ cz \end{bmatrix}$ is in S

Yes closed under scalar mult.

② Closed under addition?

No: For example $\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$ are in S

$$1^2 + 0^2 = 1^2 \quad \checkmark \quad 0^2 + 1^2 = 1^2 \quad \checkmark$$

but $\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$, is not in S because

$$1^2 + 1^2 \neq 2^2$$

Not closed under addition

\Rightarrow Not a subspace

$\&$ $\text{Span}(v_1, v_2, \dots, v_k) =$ all vectors which are linear combinations of v_1, v_2, \dots, v_k
 \uparrow
 IS a subspace
 Contains infinitely many vectors
 dimension is less than or equal to k

Linear independence:

$\left(\begin{bmatrix} 2 \\ -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 4 \\ 0 \\ 1 \end{bmatrix} \right)$ are linearly independent

Proof: If

$$c_1 \begin{bmatrix} 2 \\ -1 \\ 1 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} -3 \\ 4 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Then

$$\begin{cases} 2c_1 - 3c_2 = 0 \\ -c_1 + 4c_2 = 0 \\ c_1 = 0 \\ c_2 = 0 \end{cases} \Rightarrow c_1 = c_2 = 0$$

So the only linear combination giving $\vec{0}$ is when $c_1 = c_2 = 0$
 Thus linearly independent.

on the other hand

$\left[\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right]$ linearly dependent because

If

$$c_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} 2 \\ 3 \end{bmatrix} + c_3 \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{cases} c_1 + 2c_2 = 0 \\ 3c_2 + c_3 = 0 \end{cases} \Rightarrow \begin{cases} c_1 = -2c_2 \\ c_3 = -3c_2 \end{cases} \quad c_2 \text{ free}$$

Say $c_2 = 1$, then $c_1 = -2$, $c_3 = -3$
 combination with some non-zero coefficients \uparrow
 $(-2) \begin{bmatrix} 1 \\ 0 \end{bmatrix} + (1) \begin{bmatrix} 2 \\ 3 \end{bmatrix} + (-3) \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

~~Remember~~
~~CS~~
~~determinants~~



Basis \Rightarrow spans + is linearly independent. $\dim = \#$ vectors in basis

If (v_1, \dots, v_k) span U

can ~~also~~ remove 0 or more vectors to get a basis
 $\Rightarrow \dim U \leq k$

If (v_1, \dots, v_l) are linearly independent vectors in V

can add 0 or more vectors to get a basis.
 $\Rightarrow \dim V \geq l$

Linear independent list	Basis	Spanning list
Could be too small	Just right	Could be too big
	# vectors is dimension	

Any subspace U of V is a subspace of V
 $\dim(U) \leq \dim(V)$

If let $\dim \mathbb{R}^n = n$

If have more than n vectors in \mathbb{R}^n can't be linearly independent (too long list)

$\left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right)$ linearly ind
do not span \mathbb{R}^3

$\left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right)$ span \mathbb{R}^3
not linearly independent

$\left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right)$ do not span \mathbb{R}^3
not linearly independent

$\left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 0 \\ 0 \end{bmatrix} \right)$ do not span \mathbb{R}^3
not linearly independent

span + linearly indep \Rightarrow basis
 $\left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right)$

$\left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right)$ span + lin. ind \Rightarrow basis

1
2
↑
c
k

Matrix A : Find $N(A)$, $C(A)$ basis for each

Reduced row echelon form

$$\begin{bmatrix} 1 & 1 & -1 & -1 \\ 2 & 3 & -1 & -6 \end{bmatrix} \rightarrow \begin{bmatrix} \textcircled{1} & 0 & -2 & 3 \\ 0 & \textcircled{1} & 1 & -4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$x_1 - 2x_3 + 3x_4 = 0$
 $x_2 + x_3 - 4x_4 = 0$

$\uparrow \uparrow$ linearly independent columns for basis of $C(A)$
 $\uparrow \uparrow$ pivots
 $\uparrow \uparrow$ free variables for $N(A)$

$C(A)$ basis = $\left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 3 \end{bmatrix} \right\}$ rank = 2

$N(A)$ basis $\left\{ \begin{bmatrix} 2 \\ -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 4 \\ 0 \\ 1 \end{bmatrix} \right\}$ nullity = 2

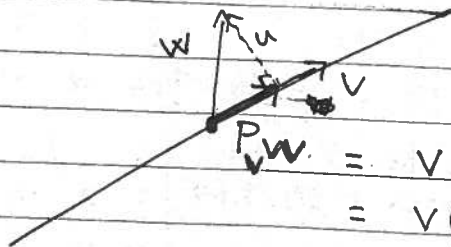
$$N(A) = \begin{bmatrix} 2x_3 - 3x_4 \\ -x_3 + 4x_4 \\ x_3 \\ x_4 \end{bmatrix} = -x_3 \begin{bmatrix} 2 \\ -1 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -3 \\ 4 \\ 0 \\ 1 \end{bmatrix}$$

Orthogonality

If v, w are column vectors $v \cdot w = v^T w$

NOT $v w^T$

Projection



$$\begin{aligned} P_v w &= v(v^T v)^{-1} v^T w \\ &= v(v \cdot v)^{-1} (v \cdot w) \\ &= v \left(\frac{v \cdot w}{v \cdot v} \right) \end{aligned}$$

(If $\|v\|=1$, $v \cdot v = 1^2 = 1$ so then $P_v w = v(v \cdot w)$)

For an orthogonal vector to v : $u = w - P_v w$
(goes into Gram Schmidt procedure)

To project to $\text{Span}(v_1, v_2, \dots, v_k)$

$$A = \begin{bmatrix} | & | & \dots & | \\ v_1 & v_2 & \dots & v_k \\ | & | & \dots & | \end{bmatrix}$$

$$P = A(A^T A)^{-1} A^T$$

To project w to $\text{Span}(v_1, v_2, \dots, v_k)$

$$Pw = A(A^T A)^{-1} A^T w$$

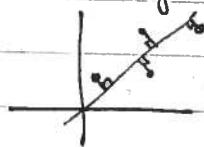
When $Ax = b$ has no solution, the best approximate solution is obtained from solving

$$A^T A x = A^T b$$

(because then $Ax = A(A^T A)^{-1} A^T b = Pb$)

Application: least squares line of best fit

$$y = C + Dx$$



Determinants: Use row operations to make calculations easier

$$\det \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 10 & 11 & 12 \\ 13 & 14 & 15 & 16 \end{bmatrix} = \det \begin{bmatrix} 1 & 2 & 3 & 4 \\ 4 & 4 & 4 & 4 \\ 8 & 8 & 8 & 8 \\ 12 & 12 & 12 & 12 \end{bmatrix}$$

Columns are linearly dependent $\Rightarrow \det = 0$

Subtract R_1 from each of R_2, R_3, R_4

$$= \det \begin{bmatrix} 1 & 2 & 3 & 4 \\ 4 & 4 & 4 & 4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Cofactor expansion using row 3 or 4 = 0