

Recall:

• A subspace V of $\mathbb{R}^n$ is a collection of vectors such that

(1) V contains $\vec{0}$

(2) if V contains $\vec{v}_1$ and $\vec{v}_2$, it also contains $\vec{v}_1 + \vec{v}_2$.

(3) if V contains $\vec{v}$, it contains $k\vec{v}$ for any scalar k .

• Let $\vec{v}_1, \dots, \vec{v}_m$ be vectors.

- $\vec{v}_i$ is redundant if

$\vec{v}_i = c_1 \vec{v}_1 + \dots + c_{i-1} \vec{v}_{i-1}$ for some scalars $c_1, \dots, c_{i-1}$.

- $\vec{v}_1, \dots, \vec{v}_m$ are linearly independent if there is no redundant vector

- they are linearly dependent if there is at least one redundant

Vector:

- $\vec{v}_1, \dots, \vec{v}_m$ is a basis for V if they span V & are linearly independent.
-

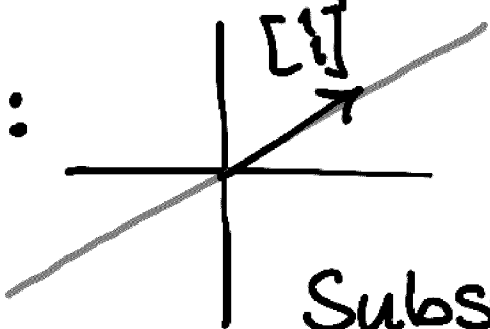
§3.3 Dimension

If V is a subspace of $\mathbb{R}^n$, and $\vec{v}_1, \dots, \vec{v}_m$ is a basis, then any vector $\vec{v}$ in V can be written as a linear combination of the vectors $\vec{v}_1, \dots, \vec{v}_m$ in a unique way. This means

$$\vec{v} = c_1 \vec{v}_1 + \dots + c_m \vec{v}_m$$

has a unique solution $\begin{bmatrix} c_1 \\ \vdots \\ c_m \end{bmatrix}$.

Ex:



$y=x$

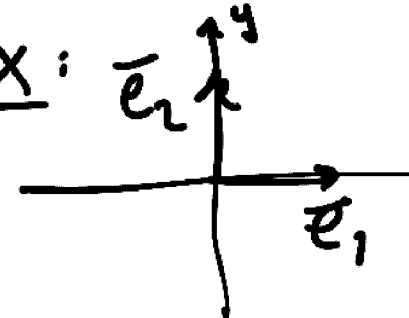
The line is a subspace, and we can

express the line as $\text{span}(\begin{bmatrix} 1 \\ 1 \end{bmatrix})$

$\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ is a basis.

Any vector in the line is of the form $\begin{bmatrix} k \\ k \end{bmatrix} = k \begin{bmatrix} 1 \\ 1 \end{bmatrix}$,

intuitively this subspace is 1-dimensional.

Ex:  $\vec{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$, $\vec{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

in $\mathbb{R}^2$. They span $\mathbb{R}^2$ because

$\mathbb{R}^2 = \text{span}(\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix})$. Any vector

$\begin{bmatrix} x \\ y \end{bmatrix}$ can be written as

$$x \begin{bmatrix} 1 \\ 0 \end{bmatrix} + y \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$

The two vectors $\vec{e}_1$ and $\vec{e}_2$ are also linearly independent because they aren't parallel.

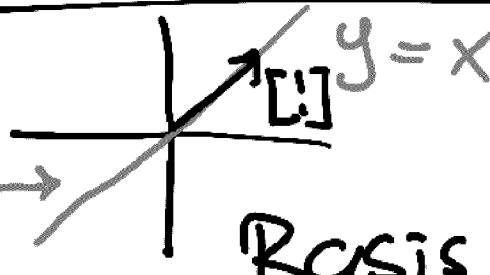
So $\vec{e}_1, \vec{e}_2$ is a basis for $\mathbb{R}^2$.

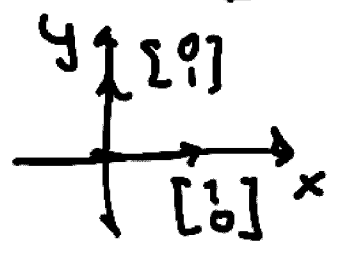
Theorem: If V is a subspace of $\mathbb{R}^n$, then V has a basis.

Moreover, any basis has the same number of vectors.

Definition: If V is a subspace of $\mathbb{R}^n$, then the dimension of V is defined as

$\dim V \stackrel{\text{def}}{=} \# \text{basis vectors in any basis for } V.$

Ex:  $\dim L = 1$
Basis is $[1]$

Ex:  $\vec{e}_1, \vec{e}_2$ is a basis for $\mathbb{R}^2$

$\dim \mathbb{R}^2 = 2.$

Ex: Similarly, in $\mathbb{R}^n$,

$$\vec{e}_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \vec{e}_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \dots, \vec{e}_n = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}$$

forms a basis for $\mathbb{R}^n$, so

$$\dim \mathbb{R}^n = n.$$

Theorem: Let V be a subspace of $\mathbb{R}^n$ with $\dim V = m$.

- (1) We can find at most m linearly independent vectors in V .
 - (2) We need at least m vectors to span V .
 - (3) If m vectors in V are linearly independent, they form a basis of V .
 - (4) If m vectors in V span V , they form a basis of V .
-

Ex: Consider $A = \begin{bmatrix} 1 & 2 & 2 & -5 & 6 \\ -1 & -2 & -1 & 1 & -1 \\ 4 & 8 & 5 & -8 & 9 \\ 3 & 6 & 1 & 5 & -7 \end{bmatrix}$

Let's find both $\ker A$ and $\text{im } A$.

We first compute $\text{ref}(A)$:

$$\begin{bmatrix} 1 & 2 & 2 & -5 & 6 \\ -1 & -2 & -1 & 1 & -1 \\ 4 & 8 & 5 & -8 & 9 \\ 3 & 6 & 1 & 5 & -7 \end{bmatrix} \begin{array}{l} \leftarrow \begin{array}{l} +1 \\ -4 \end{array} \\ \leftarrow -3 \end{array}$$

$$\downarrow$$

$$\begin{bmatrix} 1 & 2 & 2 & -5 & 6 \\ 0 & 0 & 1 & -4 & 5 \\ 0 & 0 & -3 & 12 & -15 \\ 0 & 0 & -5 & 20 & 25 \end{bmatrix} \begin{array}{l} \leftarrow \begin{array}{l} +3 \\ +5 \end{array} \\ \leftarrow \end{array}$$

$$\downarrow$$

$$\begin{bmatrix} 1 & 2 & 2 & -5 & 6 \\ 0 & 0 & 1 & -4 & 5 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{array}{l} \leftarrow -2 \end{array}$$

leading

$$\begin{bmatrix} \textcircled{1} & 2 & 0 & 3 & -4 \\ 0 & 0 & \textcircled{1} & -4 & 5 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} = \text{ref}(A).$$

$x_1 \quad x_2 \quad x_3 \quad x_4 \quad x_5$

Now to find $\ker A$ we need to solve $A\vec{x} = \vec{0}$.

The columns corresponding to non-leading elements need to be assigned parameters.

Set $\begin{cases} x_5 = t \\ x_4 = s \\ x_2 = r \end{cases}$ t, s, r parameters.

$$x_3 = 4x_4 - 5x_5 = 4s - 5t$$

$$\begin{aligned} x_1 &= -2x_2 - 3x_4 + 4x_5 \\ &= -2r - 3s + 4t \end{aligned}$$

$$\text{So } \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} -2r-3s+4t \\ r \\ 4s-5t \\ s \\ t \end{bmatrix}$$

$$= r \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + s \begin{bmatrix} -3 \\ 0 \\ 4 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} 4 \\ 0 \\ -5 \\ 0 \\ 1 \end{bmatrix}.$$

Therefore

$$\ker A = \text{Span} \left(\begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 4 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 0 \\ -5 \\ 0 \\ 1 \end{bmatrix} \right)$$

This will in fact automatically give that $\begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 4 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 0 \\ -5 \\ 0 \\ 1 \end{bmatrix}$ forms a basis for $\ker A$. So

$$\dim(\ker A) = 3.$$

Now for the image,

$\text{im } A = \text{Span of columns in } \text{rref}(A) \text{ containing the leading elements}$
 $= \text{Span}\left(\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}\right)$

& $\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}$ forms a basis, so
 $\dim(\text{im } A) = 2.$

We can always find a basis for $\text{im } A$ by picking out the columns of $\text{rref}(A)$ containing the leading elements. So in fact

$$\dim(\text{im}(A)) = \text{rank } A$$

Rank-nullity theorem

For any $(n \times m)$ -matrix A , we have

$$\dim(\ker A) + \dim(\text{im } A) = m$$

Sometimes $\dim \ker A$ is called the "nullity of A ", so then the equality is

$$(\text{nullity of } A) + (\text{rank of } A) = n$$
