Math 221: LINEAR ALGEBRA

Chapter 5. Vector Space \mathbb{R}^n §5-1. Subspaces and Spanning

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Subspaces of \mathbb{R}^{n}

The null space and the image space

The Eigenspace

Linear Combinations and Spanning Sets

Spanning sets of null(A) and im(A)

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Definitions

- 1. \mathbb{R} denotes the set of real numbers, and is an example of a set of scalars.
- 2. \mathbb{R}^{n} is the set of all n-tuples of real numbers, i.e.,

$$\mathbb{R}^n = \left\{ (x_1, x_2, \ldots, x_n) \mid x_i \in \mathbb{R}, 1 \leq i \leq n \right\}.$$

3. The vector space \mathbb{R}^n consists of the set \mathbb{R}^n written as column matrices, along with the (matrix) operations of addition and scalar multiplication. Unless stated otherwise, \mathbb{R}^n means the vector space \mathbb{R}^n .

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Remark

 \mathbb{R}^n is a concrete example of the abstract vector space will be studied in the next chapter.

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 is a vector in \mathbb{R}^5 , written $\vec{\mathbf{u}} \in \mathbb{R}^5$.

To save space on the page, the same vector \vec{u} may be written instead as a row matrix by taking the transpose of the column:

$$\vec{\mathbf{u}} = \begin{bmatrix} -2, & 3, & 0.7, & 5, & \pi \end{bmatrix}$$

We are interested in nice subsets of \mathbb{R}^n , defined as follows.

Definition (Subspaces)

A subset U of \mathbb{R}^n is a subspace of \mathbb{R}^n if

- S1. The zero vector of \mathbb{R}^n , $\vec{0}_n$, is in U;
- S2. U is closed under addition, i.e., for all $\vec{u}, \vec{w} \in U, \vec{u} + \vec{w} \in U$;
- S3. U is closed under scalar multiplication, i.e., for all $\vec{u} \in U$ and $k \in \mathbb{R},$ $k\vec{u} \in U.$

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Both subset $U = \{\vec{0}_n\}$ and R^n itself are subspaces of \mathbb{R}^n . Any other subspace of \mathbb{R}^n is called a proper subspace of \mathbb{R}^n .

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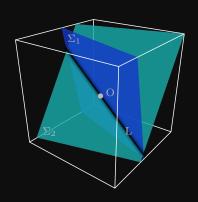
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Notation

If U is a subset of $\mathbb{R}^n,$ we write $U\subseteq\mathbb{R}^n.$



Example

In \mathbb{R}^3 , the line L through the origin that is parallel to the vector

$$\vec{d} = \begin{bmatrix} -5 \\ 1 \\ -4 \end{bmatrix} \text{ has (vector) equation } \begin{bmatrix} x \\ y \\ z \end{bmatrix} = t \begin{bmatrix} -5 \\ 1 \\ -4 \end{bmatrix}, t \in \mathbb{R}, \text{ so}$$

Claim. L is a subspace of
$$\mathbb{R}^3$$
.

- First: $\vec{0}_3 \in L \text{ since } 0\vec{d} = \vec{0}_3$.
- ▶ Suppose $\vec{u}, \vec{v} \in L$. Then by definition, $\vec{u} = s\vec{d}$ and $\vec{v} = t\vec{d}$, for some $s, t \in \mathbb{R}$. Thus

 $L = \left\{ t\vec{d} \mid t \in \mathbb{R} \right\}.$

$$\vec{u} + \vec{v} = s\vec{d} + t\vec{d} = (s+t)\vec{d}.$$

Since $s + t \in \mathbb{R}$, $\vec{u} + \vec{v} \in L$; i.e., L is closed under addition.

Example (continued)

▶ Suppose $\vec{u} \in L$ and $k \in \mathbb{R}$ (k is a scalar). Then $\vec{u} = t\vec{d}$, for some $t \in \mathbb{R}$, so

$$k\vec{u} = k(t\vec{d}) = (kt)\vec{d}.$$

Since kt $\in \mathbb{R}$, k $\vec{u} \in L$; i.e., L is closed under scalar multiplication.

▶ Therefore, L is a subspace of \mathbb{R}^3 .

Example (continued)

▶ Suppose $\vec{u} \in L$ and $k \in \mathbb{R}$ (k is a scalar). Then $\vec{u} = t\vec{d}$, for some $t \in \mathbb{R}$, so

$$k\vec{u} = k(t\vec{d}) = (kt)\vec{d}.$$

Since $kt \in \mathbb{R}$, $k\vec{u} \in L$; i.e., L is closed under scalar multiplication.

▶ Therefore, L is a subspace of \mathbb{R}^3 .

Remark

Note that there is nothing special about the vector \vec{d} used in this example; the same proof works for any nonzero vector $\vec{d} \in \mathbb{R}^3$, so any line through the origin is a subspace of \mathbb{R}^3 .

Example

In \mathbb{R}^3 , let M denote the plane through the origin having equation

$$3x - 2y + z = 0$$
; then M has normal vector $\vec{n} = \begin{bmatrix} 3 \\ -2 \\ 1 \end{bmatrix}$. If $\vec{u} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$, then

$$M = \left\{ \vec{u} \in \mathbb{R}^3 \mid \vec{n} \cdot \vec{u} = 0 \right\},\,$$

where $\vec{n} \cdot \vec{u}$ is the dot product of vectors \vec{n} and \vec{u} .

Claim. M is a subspace of \mathbb{R}^3 .

- First: $\vec{0}_3 \in M$ since $\vec{n} \cdot \vec{0}_3 = 0$.
- ▶ Suppose $\vec{u}, \vec{v} \in M$. Then by definition, $\vec{n} \cdot \vec{u} = 0$ and $\vec{n} \cdot \vec{v} = 0$, so

$$\vec{n} \cdot (\vec{u} + \vec{v}) = \vec{n} \cdot \vec{u} + \vec{n} \cdot \vec{v} = 0 + 0 = 0,$$

and thus $(\vec{u} + \vec{v}) \in M$; i.e., M is closed under addition.

Example (continued)

▶ Suppose $\vec{u} \in M$ and $k \in \mathbb{R}$. Then $\vec{n} \cdot \vec{u} = 0$, so

$$\vec{n} \cdot (k\vec{u}) = k(\vec{n} \cdot \vec{u}) = k(0) = 0,$$

and thus $k\vec{u}\in M;$ i.e., M is closed under scalar multiplication.

▶ Therefore, M is a subspace of \mathbb{R}^3 .

Example (continued)

▶ Suppose $\vec{u} \in M$ and $k \in \mathbb{R}$. Then $\vec{n} \cdot \vec{u} = 0$, so

$$\vec{n}\cdot(k\vec{u})=k(\vec{n}\cdot\vec{u})=k(0)=0,$$

and thus $k\vec{u} \in M$; i.e., M is closed under scalar multiplication.

ightharpoonup Therefore, M is a subspace of \mathbb{R}^3 .

Remark

As in the previous example, there is nothing special about the plane chosen for this example; any plane through the origin is a subspace of \mathbb{R}^3 .

$$\text{Is } U = \left\{ \left[\begin{array}{c} a \\ b \\ c \\ d \end{array} \right] \; \middle| \; a,b,c,d \in \mathbb{R} \quad \text{and} \quad 2a-b=c+2d \right\} \text{ a subspace of } \mathbb{R}^4?$$

Justify your answer.

$$\text{Is } U = \left\{ \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \middle| \begin{array}{l} a,b,c,d \in \mathbb{R} \quad \text{and} \quad 2a-b=c+2d \\ \end{array} \right\} \text{ a subspace of } \mathbb{R}^4?$$

Justify your answer.

Solution

The zero vector of \mathbb{R}^4 is the vector $\begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix}$ with a = b = c = d = 0.

In this case, 2a - b = 2(0) + 0 = 0 and $\vec{c} + 2d = 0 + 2(0) = 0$, so 2a - b = c + 2d. Therefore, $\vec{0}_4 \in U$.

Solution (continued)

Suppose

$$\vec{\mathbf{v}}_1 = \left[egin{array}{c} \mathbf{a}_1 \\ \mathbf{b}_1 \\ \mathbf{c}_1 \\ \mathbf{d}_1 \end{array}
ight] \quad ext{and} \quad \vec{\mathbf{v}}_2 = \left[egin{array}{c} \mathbf{a}_2 \\ \mathbf{b}_2 \\ \mathbf{c}_2 \\ \mathbf{d}_2 \end{array}
ight] \text{ are in U.}$$

Then $2a_1 - b_1 = c_1 + 2d_1$ and $2a_2 - b_2 = c_2 + 2d_2$. Now

$$\vec{v}_1 + \vec{v}_2 = \left[\begin{array}{c} a_1 \\ b_1 \\ c_1 \\ d_1 \end{array} \right] + \left[\begin{array}{c} a_2 \\ b_2 \\ c_2 \\ d_2 \end{array} \right] = \left[\begin{array}{c} a_1 + a_2 \\ b_1 + b_2 \\ c_1 + c_2 \\ d_1 + d_2 \end{array} \right],$$

and

$$\begin{array}{rcl} 2(a_1+a_2)-(b_1+b_2) & = & (2a_1-b_1)+(2a_2-b_2) \\ & = & (c_1+2d_1)+(c_2+2d_2) \\ & = & (c_1+c_2)+2(d_1+d_2). \end{array}$$

Therefore, $\vec{v}_1 + \vec{v}_2 \in U$.

Solution (continued)

Finally, suppose

$$\vec{\mathrm{v}} = \left| egin{array}{c} \mathrm{a} \\ \mathrm{b} \\ \mathrm{c} \\ \mathrm{d} \end{array} \right| \in \mathrm{U} \quad \mathrm{and} \quad \mathrm{k} \in \mathbb{R}$$

Then 2a - b = c + 2d. Now

$$\mathbf{k}\vec{\mathbf{v}} = \mathbf{k} \begin{bmatrix} \mathbf{a} \\ \mathbf{b} \\ \mathbf{c} \\ \mathbf{d} \end{bmatrix} = \begin{bmatrix} \mathbf{k}\mathbf{a} \\ \mathbf{k}\mathbf{b} \\ \mathbf{k}\mathbf{c} \\ \mathbf{k}\mathbf{d} \end{bmatrix},$$

and

$$2ka - kb = k(2a - b) = k(c + 2d) = kc + 2kd.$$

Therefore, $k\vec{v} \in U$.

It follows from the Subspace Test that U is a subspace of \mathbb{R}^4 .

Is
$$U = \left\{ \begin{bmatrix} 1 \\ s \\ t \end{bmatrix} \middle| s, t \in \mathbb{R} \right\}$$
 a subspace of \mathbb{R}^3 ? Justify your answer.

Is $U = \left\{ \begin{bmatrix} 1 \\ s \\ t \end{bmatrix} \middle| s, t \in \mathbb{R} \right\}$ a subspace of \mathbb{R}^3 ? Justify your answer.

Solution

Note that $\vec{0}_3 \notin U$, and thus U is not a subspace of \mathbb{R}^3 .

(You could also show that U is not closed under addition, or not closed under scalar multiplication.)

Is
$$U = \left\{ \begin{bmatrix} r \\ 0 \\ s \end{bmatrix} \middle| r, s \in \mathbb{R} \text{ and } r^2 + s^2 = 0 \right\}$$
 a subspace of \mathbb{R}^3 ?

Justify your answer.

Is
$$U = \left\{ \begin{bmatrix} r \\ 0 \\ s \end{bmatrix} \middle| r, s \in \mathbb{R} \text{ and } r^2 + s^2 = 0 \right\}$$
 a subspace of \mathbb{R}^3 ?

Justify your answer.

Solution

Since $r\in\mathbb{R},\ r^2\geq 0$ with equality if and only if r=0. Similarly, $s\in\mathbb{R}$ implies $s^2\geq 0,$ and $s^2=0$ if and only if s=0. This means $r^2+s^2=0$ if and only if $r^2=s^2=0;$ thus $r^2+s^2=0$ if and only if r=s=0. Therefore U contains only $\vec{0}_3,$ the zero vector, i.e., $U=\{\vec{0}_3\}.$ As we already observed, $\{\vec{0}_n\}$ is a subspace of $\mathbb{R}^n,$ and therefore U is a subspace of $\mathbb{R}^3.$

Subspaces of \mathbb{R}^{r}

The null space and the image space

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The null space and the image space

Definitions (Null Space and Image Space)

Let A be an $m \times n$ matrix. The null space of A is defined as

$$\mathrm{null}(A) = \{ \vec{x} \in \mathbb{R}^n \mid A\vec{x} = \vec{0}_m \},$$

and the image space of A is defined as

$$\operatorname{im}(A) = \{A\vec{x} \mid \vec{x} \in \mathbb{R}^n\}.$$

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and the image space of A is defined as

$$im(A) = \{A\vec{x} \mid \vec{x} \in \mathbb{R}^n\}.$$

Remark

- 1. Since A is $m \times n$ and $\vec{x} \in \mathbb{R}^n$, $A\vec{x} \in \mathbb{R}^m$, so $im(A) \subseteq \mathbb{R}^m$ while $null(A) \subseteq \mathbb{R}^n$.
- 2. Image space is also called column space of A, denoted as col(A):

$$\operatorname{col}(A) = \operatorname{span}\left(\vec{a}_1, \cdots, \vec{a}_n\right) = \operatorname{im}(A).$$

Prove that if A is an $m \times n$ matrix, then null(A) is a subspace of \mathbb{R}^n .

Prove that if A is an $m \times n$ matrix, then null(A) is a subspace of \mathbb{R}^n .

Proof.

S1. Since $A\vec{0}_n = \vec{0}_m, \, \vec{0}_n \in null(A)$.

Prove that if A is an $m \times n$ matrix, then null(A) is a subspace of \mathbb{R}^n .

Proof.

S1. Since $A\vec{0}_n = \overline{\vec{0}_m}, \vec{0}_n \in null(A)$.

S2. Let $\vec{x}, \vec{y} \in null(A)$. Then $A\vec{x} = \vec{0}_m$ and $A\vec{y} = \vec{0}_m$, so

$$A(\vec{x} + \vec{y}) = A\vec{x} + A\vec{y} = \vec{0}_m + \vec{0}_m = \vec{0}_m,$$

and thus $\vec{x} + \vec{y} \in \text{null}(A)$.

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$$A(\vec{x}+\vec{y}) = A\vec{x} + A\vec{y} = \vec{0}_m + \vec{0}_m = \vec{0}_m, \label{eq:alpha}$$

and thus $\vec{x} + \vec{y} \in \text{null}(A)$.

S3. Let $\vec{x} \in \text{null}(A)$ and $k \in \mathbb{R}$. Then $A\vec{x} = \vec{0}_m$, so

$$A(k\vec{x}) = k(A\vec{x}) = k\vec{0}_m = \vec{0}_m,$$

and thus $k\vec{x} \in \text{null}(A)$.

Therefore, null(A) is a subspace of \mathbb{R}^n .

Prove that if A is an $m \times n$ matrix, then im(A) is a subspace of \mathbb{R}^m .

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Proof.

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Prove that if A is an $m \times n$ matrix, then im(A) is a subspace of \mathbb{R}^m .

Proof.

- S1. Since $\vec{0}_n \in \mathbb{R}^n$ and $A\vec{0}_n = \vec{0}_m$, $\vec{0}_m \in im(A)$.
- S2. Let $\vec{x}, \vec{y} \in \text{im}(A)$. Then $\vec{x} = A\vec{u}$ and $\vec{y} = A\vec{v}$ for some $\vec{u}, \vec{v} \in \mathbb{R}^n$, so

$$\vec{x} + \vec{y} = A\vec{u} + A\vec{v} = A(\vec{u} + \vec{v}).$$

Since $\vec{u} + \vec{v} \in \mathbb{R}^n$, it follows that $\vec{x} + \vec{y} \in \text{im}(A)$.

Prove that if A is an $m \times n$ matrix, then im(A) is a subspace of \mathbb{R}^m .

Proof.

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$$\vec{x} + \vec{y} = A\vec{u} + A\vec{v} = A(\vec{u} + \vec{v}).$$

Since $\vec{u} + \vec{v} \in \mathbb{R}^n$, it follows that $\vec{x} + \vec{y} \in \text{im}(A)$.

S3. Let $\vec{x} \in \text{im}(A)$ and $k \in \mathbb{R}$. Then $\vec{x} = A\vec{u}$ for some $\vec{u} \in \mathbb{R}^n$, and thus

$$k\vec{x} = k(A\vec{u}) = A(k\vec{u}).$$

Since $k\vec{u} \in \mathbb{R}^n$, it follows that $k\vec{x} \in im(A)$.

Therefore, im(A) is a subspace of \mathbb{R}^m .

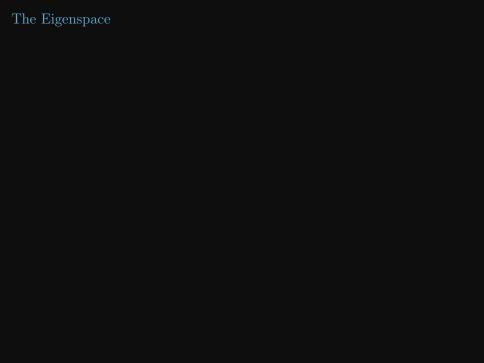
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The Eigenspace

Definition (Eigenspace)

Let A be an n × n matrix and $\lambda \in \mathbb{R}$. The eigenspace of A corresponding to λ is the set

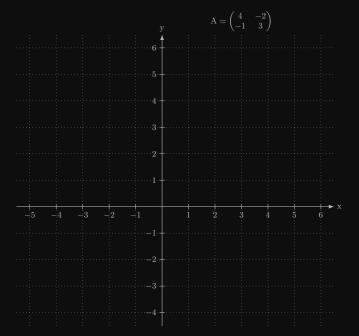
$$\mathrm{E}_{\lambda}(\mathrm{A}) = \{ \vec{\mathrm{x}} \in \mathbb{R}^{\mathrm{n}} \mid \mathrm{A}\vec{\mathrm{x}} = \lambda \vec{\mathrm{x}} \}$$

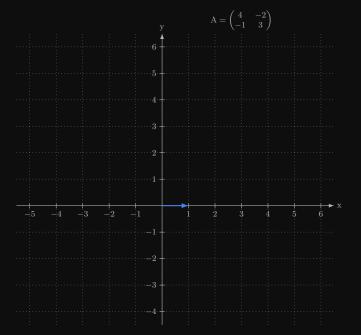
$$A = \begin{pmatrix} 4 & -2 \\ -1 & 3 \end{pmatrix}$$
 has two eigenvalues: $\lambda_1 = 2$ and $\lambda_2 = 5$ with corresponding eigenvectors

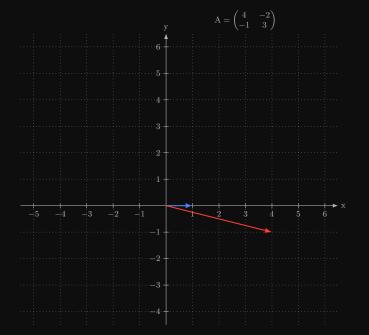
$$\vec{\mathrm{v}}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad \mathrm{and} \quad \vec{\mathrm{v}}_2 = \begin{pmatrix} -1 \\ 1/2 \end{pmatrix}$$

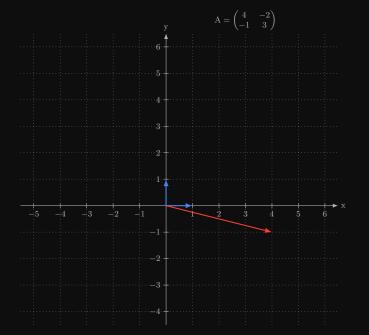
Hence,

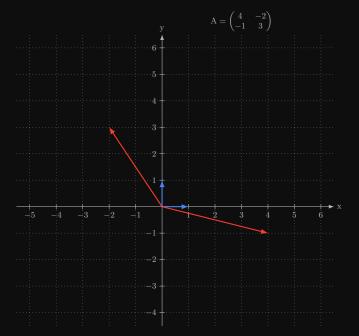
$$\begin{split} E_{\lambda_1}(A) &= E_2(A) = \{t\vec{v}_1|t \in \mathbb{R}\} \\ E_{\lambda_2}(A) &= E_5(A) = \{t\vec{v}_2|t \in \mathbb{R}\} \end{split}$$

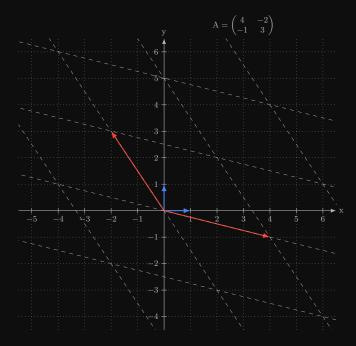


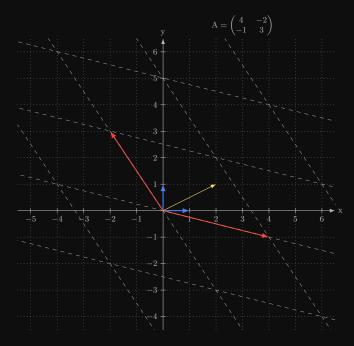


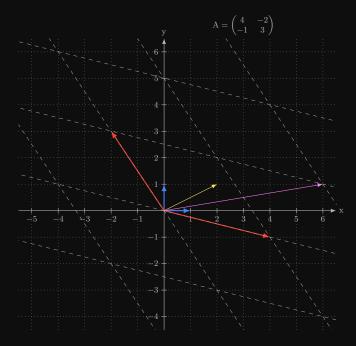


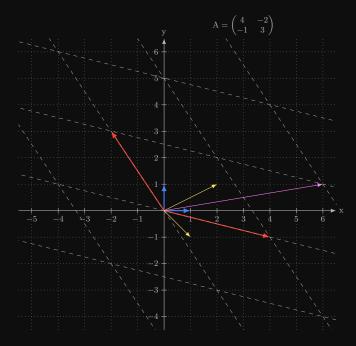


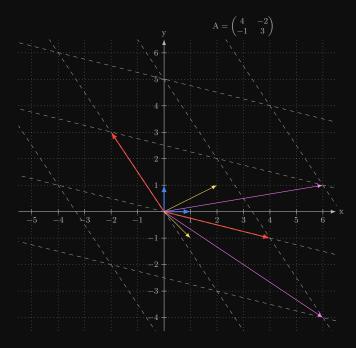


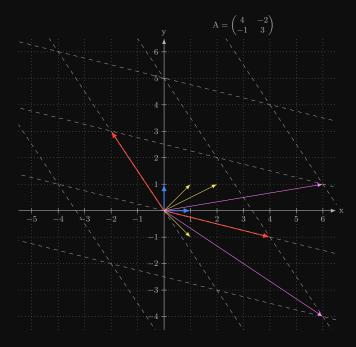


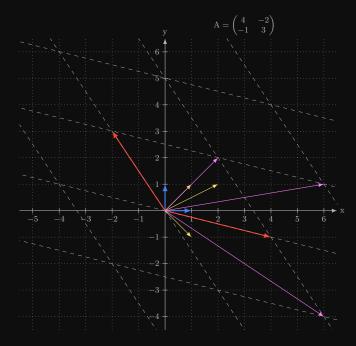


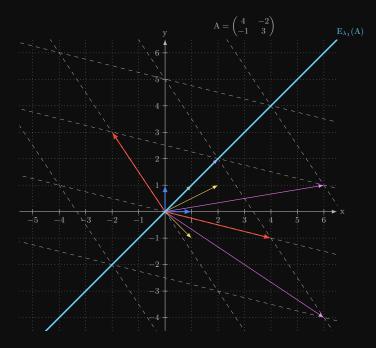


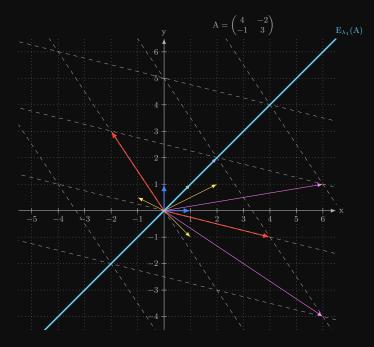


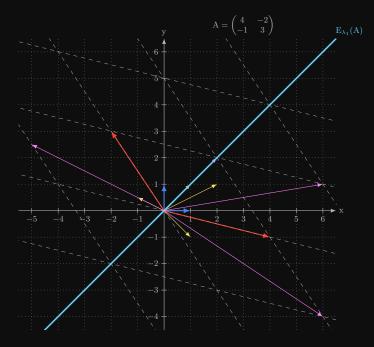


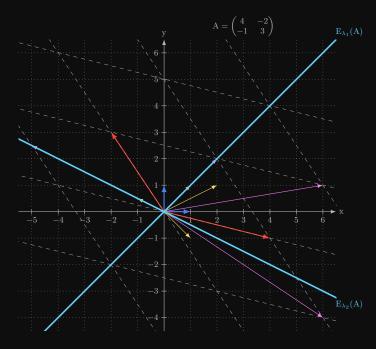


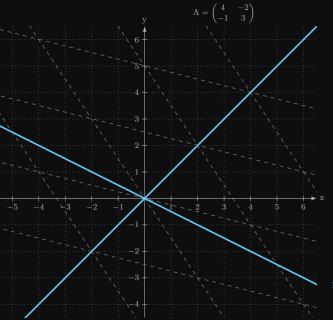












 $E_2(A) = \left\{ t \begin{pmatrix} 1 \\ 1 \end{pmatrix} \middle| t \in \mathbb{R} \right\}$

$$E_5(A) = \left\{ t \begin{pmatrix} -1 \\ 1/2 \end{pmatrix} \middle| t \in \mathbb{R} \right\}$$

Note that

$$\mathrm{E}_{\lambda}(\mathrm{A}) \quad = \quad \{ \vec{\mathrm{x}} \in \mathbb{R}^{\mathrm{n}} \mid \mathrm{A}\vec{\mathrm{x}} = \lambda \vec{\mathrm{x}} \}$$

$$= \left\{ \vec{x} \in \mathbb{R}^n \mid \lambda \vec{x} - A \vec{x} = \vec{0}_n \right\}$$

$$= \quad \left\{ \vec{x} \in \mathbb{R}^n \mid (\lambda I - A) \vec{x} = \vec{0}_n \right\}$$
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$$E_{\lambda}(A) = \text{null}(\lambda I - A).$$

Note that

$$\begin{split} E_{\lambda}(A) &= & \left\{ \vec{x} \in \mathbb{R}^n \mid A\vec{x} = \lambda \vec{x} \right\} \\ &= & \left\{ \vec{x} \in \mathbb{R}^n \mid \lambda \vec{x} - A\vec{x} = \vec{0}_n \right\} \\ &= & \left\{ \vec{x} \in \mathbb{R}^n \mid (\lambda I - A)\vec{x} = \vec{0}_n \right\} \end{split}$$

showing that

$$E_{\lambda}(A) = \text{null}(\lambda I - A).$$

It follows that

- ightharpoonup if λ is **not** an eigenvalue of A, then $E_{\lambda}(A) = \{\vec{0}_n\};$
- ▶ the nonzero vectors of $E_{\lambda}(A)$ are the eigenvectors of A corresponding to λ ;
- ▶ the eigenspace of A corresponding to λ is a subspace of \mathbb{R}^n .

Subspaces of \mathbb{R}^r

The null space and the image space

The Eigenspace

Linear Combinations and Spanning Sets

Spanning sets of null(A) and im(A



Linear Combinations and Spanning Sets

Definition (Linear Combinations and Spanning)

Let $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k \in \mathbb{R}^n$ and $t_1, t_2, \dots, t_k \in \mathbb{R}$. Then the vector

$$\vec{x} = t_1 \vec{x}_1 + t_2 \vec{x}_2 + \dots + t_k \vec{x}_k$$

is called a linear combination of the vectors $\vec{x}_1, \vec{x}_2, \ldots, \vec{x}_k$; the (scalars) $t_1, t_2, \ldots, t_k \in \mathbb{R}$ are the coefficients. The set of all linear combinations of $\vec{x}_1, \vec{x}_2, \ldots, \vec{x}_k$ is called the span of $\vec{x}_1, \vec{x}_2, \ldots, \vec{x}_k$, and is written

$$\mathrm{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\} = \{t_1\vec{x}_1 + t_2\vec{x}_2 + \dots + t_k\vec{x}_k \mid t_1, t_2, \dots, t_k \in \mathbb{R}\}\,.$$

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Additional Terminology. If $U = \text{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$, then

- ightharpoonup U is spanned by the vectors $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k$.
- ▶ the vectors $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k$ span U.
- \blacktriangleright the set of vectors $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$ is a spanning set for U.

Let $\vec{x} \in \mathbb{R}^3$ be a nonzero vector. Then span $\{\vec{x}\} = \{k\vec{x} \mid k \in \mathbb{R}\}$ is a line through the origin having direction vector \vec{x} .

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Example

Let $\vec{x}, \vec{y} \in \mathbb{R}^3$ be nonzero vectors that are not parallel. Then

$$\mathrm{span}\{\vec{x},\vec{y}\} = \{k\vec{x} + t\vec{y} \mid k,t \in \mathbb{R}\}$$

is a plane through the origin containing \vec{x} and $\vec{y}.$

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is a plane through the origin containing \vec{x} and \vec{y} .

How would you find the equation of this plane?

Let
$$\vec{\mathbf{x}} = \begin{bmatrix} 8 \\ 3 \\ -13 \\ 20 \end{bmatrix}$$
, $\vec{\mathbf{y}} = \begin{bmatrix} 2 \\ 1 \\ -3 \\ 5 \end{bmatrix}$ and $\vec{\mathbf{z}} = \begin{bmatrix} -1 \\ 0 \\ 2 \\ -3 \end{bmatrix}$. Is $\vec{\mathbf{x}} \in \operatorname{span}\{\vec{\mathbf{y}}, \vec{\mathbf{z}}\}$?

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Solution

An equivalent question is: can \vec{x} be expressed as a linear combination of \vec{y} and \vec{z} ?

Suppose there exist $a, b \in \mathbb{R}$ so that $\vec{x} = a\vec{y} + b\vec{z}$. Then

$$\begin{bmatrix} 8 \\ 3 \\ -13 \\ 20 \end{bmatrix} = a \begin{bmatrix} 2 \\ 1 \\ -3 \\ 5 \end{bmatrix} + b \begin{bmatrix} -1 \\ 0 \\ 2 \\ -3 \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ 1 & 0 \\ -3 & 2 \\ 5 & -3 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix}.$$

Solve this system of four linear equations in the two variables a and b.

Solution (continued)

$$\begin{bmatrix} 2 & -1 & 8 \\ 1 & 0 & 3 \\ -3 & 2 & -13 \\ 5 & -3 & 20 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 3 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$

Since the system has no solutions, $\vec{x} \notin \text{span}\{\vec{y}, \vec{z}\}.$

Let
$$\vec{\mathbf{w}} = \begin{bmatrix} 8 \\ 3 \\ -13 \\ 21 \end{bmatrix}$$
, $\vec{\mathbf{y}} = \begin{bmatrix} 2 \\ 1 \\ -3 \\ 5 \end{bmatrix}$ and $\vec{\mathbf{z}} = \begin{bmatrix} -1 \\ 0 \\ 2 \\ -3 \end{bmatrix}$. Is $\vec{\mathbf{w}} \in \text{span}\{\vec{\mathbf{y}}, \vec{\mathbf{z}}\}$?

This is almost identical to a previous problem, except that \vec{w} (above) has one entry that is different from the vector \vec{x} of that problem.

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This is almost identical to a previous problem, except that \vec{w} (above) has one entry that is different from the vector \vec{x} of that problem.

Solution

In this case, the system of linear equations is consistent, and gives us $\vec{w} = 3\vec{y} - 2\vec{z}$, so $\vec{w} \in \operatorname{span}\{\vec{y}, \vec{z}\}$.

Theorem

Let $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k \in \mathbb{R}^n$ and let $U = \text{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$. Then

- 1. U is a subspace of \mathbb{R}^n containing each \vec{x}_i , $1 \leq i \leq k$;
- 2. if W is a subspace of \mathbb{R}^n and $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k \in W$, then $U \subseteq W$.

Theorem

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- 1. U is a subspace of \mathbb{R}^n containing each \vec{x}_i , $1 \leq i \leq k$;
- 2. if W is a subspace of \mathbb{R}^n and $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k \in W$, then $U \subseteq W$.

Remark

Property 2 is saying that U is the "smallest" subspace of \mathbb{R}^n that contains $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k$.

Proof. (Part 1 of Theorem)

Since $U = \operatorname{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$ and $0\vec{x}_1 + 0\vec{x}_2 + \dots + 0\vec{x}_k = \vec{0}_n, \vec{0}_n \in U.$

Proof. (Part 1 of Theorem)

Since
$$U = \operatorname{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$$
 and $0\vec{x}_1 + 0\vec{x}_2 + \dots + 0\vec{x}_k = \vec{0}_n, \vec{0}_n \in U$.

Suppose $\vec{x}, \vec{y} \in U$. Then for some $s_i, t_i \in \mathbb{R}, 1 \leq i \leq k$,

$$\vec{x} = s_1 \vec{x}_1 + s_2 \vec{x}_2 + \dots + s_k \vec{x}_k$$

 $\vec{v} = t_1 \vec{x}_1 + t_2 \vec{x}_2 + \dots + t_k \vec{x}_k$

Thus

$$\begin{split} \vec{x} + \vec{y} &= (s_1 \vec{x}_1 + s_2 \vec{x}_2 + \dots + s_k \vec{x}_k) + (t_1 \vec{x}_1 + t_2 \vec{x}_2 + \dots + t_k \vec{x}_k) \\ &= (s_1 + t_1) \vec{x}_1 + (s_2 + t_2) \vec{x}_2 + \dots + (s_k + t_k) \vec{x}_k. \end{split}$$

Since $s_i+t_i\in\mathbb{R}$ for all $1\leq i\leq k,\, \vec{x}+\vec{y}\in U,\, i.e.,\, U$ is closed under addition.

Proof. (Part 1 of Theorem – continued)

Suppose $\vec{x} \in U$ and $a \in \mathbb{R}$. Then for some $s_i \in \mathbb{R}$, 1 < i < k,

$$\vec{x} = s_1 \vec{x}_1 + s_2 \vec{x}_2 + \dots + s_k \vec{x}_k$$

Thus

$$\begin{array}{rcl} a\vec{x} & = & a(s_1\vec{x}_1 + s_2\vec{x}_2 + \dots + s_k\vec{x}_k) \\ & = & (as_1)\vec{x}_1 + (as_2)\vec{x}_2 + \dots + (as_k)\vec{x}_k. \end{array}$$

Since $as_i \in \mathbb{R}$ for all $1 \leq i \leq k,$ $a\vec{x} \in U$. Hence, U is closed under scalar multiplication.

Proof. (Part 1 of Theorem – continued)

Suppose $\vec{x} \in U$ and $a \in \mathbb{R}$. Then for some $s_i \in \mathbb{R}$, $1 \le i \le k$,

$$\vec{x} = s_1\vec{x}_1 + s_2\vec{x}_2 + \cdots + s_k\vec{x}_k$$

Thus

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Since $as_i\in\mathbb{R}$ for all $1\leq i\leq k,$ $a\vec{x}\in U.$ Hence, U is closed under scalar multiplication.

Therefore, U is a subspace of \mathbb{R}^n . Furthermore, since

$$\vec{x}_i = \sum_{i=1}^{i-1} 0 \vec{x}_j + 1 \vec{x}_i + \sum_{i=i+1}^{k} 0 \vec{x}_j,$$

it follows that $\vec{x}_i \in U$ for all i, $1 \le i \le k$.

Proof. (Part 1 of Theorem – continued)

Suppose $\vec{x} \in U$ and $a \in \mathbb{R}$. Then for some $s_i \in \mathbb{R}$, $1 \le i \le k$,

$$\vec{x} = s_1\vec{x}_1 + s_2\vec{x}_2 + \cdots + s_k\vec{x}_k$$

Thus

$$\begin{array}{rcl} a\vec{x} & = & a(s_1\vec{x}_1 + s_2\vec{x}_2 + \dots + s_k\vec{x}_k) \\ & = & (as_1)\vec{x}_1 + (as_2)\vec{x}_2 + \dots + (as_k)\vec{x}_k. \end{array}$$

Since $as_i\in\mathbb{R}$ for all $1\leq i\leq k,$ $a\vec{x}\in U.$ Hence, U is closed under scalar multiplication.

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Let $W\subset \mathbb{R}^n$ be a subspace that contains $\vec{x}_1,\cdots,\vec{x}_n.$ We need to prove that $U\subseteq W.$

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Let $W \subset \mathbb{R}^n$ be a subspace that contains $\vec{x}_1, \cdots, \vec{x}_n$. We need to prove that $U \subseteq W$.

Suppose $\vec{x} \in U$. Then $\vec{x} = s_1\vec{x}_1 + s_2\vec{x}_2 + \dots + s_k\vec{x}_k$ for some $s_i \in \mathbb{R}$, $1 \le i \le k$. Since W contain each \vec{x}_i and W is closed under scalar multiplication, it follows that $s_i\vec{x}_i \in W$ for each $i, 1 \le i \le k$. Furthermore, since W is closed under addition, $\vec{x} = s_1\vec{x}_1 + s_2\vec{x}_2 + \dots + s_k\vec{x}_k \in W$. Therefore, $U \subseteq W$.

Problem (revisited)

Justify your answer.

Is $U = \left\{ \begin{array}{c|c} a \\ b \\ c \\ d \end{array} \right| a, b, c, d \in \mathbb{R} \text{ and } 2a - b = c + 2d \right\}$ a subspace of \mathbb{R}^4 ?

Problem (revisited)

$$\text{Is } U = \left\{ \left[\begin{array}{c} a \\ b \\ c \\ d \end{array} \right] \middle| \begin{array}{c} a,b,c,d \in \mathbb{R} \quad \text{and} \quad 2a-b=c+2d \\ \end{array} \right\} \text{ a subspace of } \mathbb{R}^4?$$

Justify your answer.

Solution (Another)

Let
$$\vec{v} = \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \in U$$
. Since $2a - b = c + 2d$, $c = 2a - b - 2d$, and thus

$$U = \left\{ \left[\begin{array}{c} a \\ b \\ 2a - b - 2d \end{array} \right] \; \middle| \; a, b, d \in \mathbb{R} \right\} = \operatorname{span} \left\{ \left[\begin{array}{c} 1 \\ 0 \\ 2 \\ 0 \end{array} \right], \left[\begin{array}{c} 0 \\ 1 \\ -1 \\ 0 \end{array} \right], \left[\begin{array}{c} 0 \\ 0 \\ -2 \\ 1 \end{array} \right] \right\}.$$

By a previous Theorem, U is a subspace of \mathbb{R}^4 .

Let $\vec{x}, \vec{y} \in \mathbb{R}^n$, $U_1 = \operatorname{span}\{\vec{x}, \vec{y}\}$, and $U_2 = \operatorname{span}\{2\vec{x} - \vec{y}, 2\vec{y} + \vec{x}\}$. Prove that

Let $\vec{x}, \vec{y} \in \mathbb{R}^n$, $U_1 = \operatorname{span}\{\vec{x}, \vec{y}\}$, and $U_2 = \operatorname{span}\{2\vec{x} - \vec{y}, 2\vec{y} + \vec{x}\}$. Prove that $U_1 = U_2$.

Solution

To show that $U_1 = U_2$, prove that $U_1 \subseteq U_2$, and $U_2 \subseteq U_1$. We begin by noting that, by the first part of the previous Theorem, U_1 and U_2 are subspaces of \mathbb{R}^n .

Let $\vec{x}, \vec{y} \in \mathbb{R}^n$, $U_1 = \operatorname{span}\{\vec{x}, \vec{y}\}$, and $U_2 = \operatorname{span}\{2\vec{x} - \vec{y}, 2\vec{y} + \vec{x}\}$. Prove that $U_1 = U_2$.

Solution

To show that $U_1 = U_2$, prove that $U_1 \subseteq U_2$, and $U_2 \subseteq U_1$. We begin by noting that, by the first part of the previous Theorem, U_1 and U_2 are subspaces of \mathbb{R}^n .

Since $2\vec{x} - \vec{y}, 2\vec{y} + \vec{x} \in U_1$, it follows from the second part of the previous Theorem that span $\{2\vec{x} - \vec{y}, 2\vec{y} + \vec{x}\} \subseteq U_1$, i.e., $U_2 \subseteq U_1$.

Also, since

$$\vec{x} = \frac{2}{5} (2\vec{x} - \vec{y}) + \frac{1}{5} (2\vec{y} + \vec{x}) ,$$

$$\vec{y} = -\frac{1}{5} (2\vec{x} - \vec{y}) + \frac{2}{5} (2\vec{y} + \vec{x}) ,$$

 $\vec{x}, \vec{y} \in U_2$. Therefore, by the second part of the previous Theorem, span $\{\vec{x}, \vec{y}\} \subseteq U_2$, i.e., $U_1 \subseteq U_2$. The result now follows.

Show that $\mathbb{R}^n=\mathrm{span}\{\vec{e}_1,\vec{e}_2,\ldots,\vec{e}_n\},$ where \vec{e}_j denote the j^{th} column of $I_n.$

Show that $\mathbb{R}^n = \operatorname{span}\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n\}$, where \vec{e}_j denote the j^{th} column of I_n .

Solution

Let
$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n$$
. Then $\vec{x} = x_1 \vec{e_1} + x_2 \vec{e_2} + \dots + x_n \vec{e_n}$, where $x_1, x_2, \dots, x_n \in \mathbb{R}$. Therefore, $\vec{x} \in \text{span}\{\vec{e_1}, \vec{e_2}, \dots, \vec{e_n}\}$, and thus

 $\mathbb{R}^n \subseteq \operatorname{span}\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n\}.$

Show that $\mathbb{R}^n = span\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n\}$, where \vec{e}_j denote the j^{th} column of I_n .

Solution

Let
$$\vec{\mathbf{x}} = \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \vdots \\ \mathbf{x}_n \end{bmatrix} \in \mathbb{R}^n$$
. Then $\vec{\mathbf{x}} = \mathbf{x}_1 \vec{\mathbf{e}_1} + \mathbf{x}_2 \vec{\mathbf{e}_2} + \dots + \mathbf{x}_n \vec{\mathbf{e}_n}$, where

 $x_1, x_2, \ldots, x_n \in \mathbb{R}$. Therefore, $\vec{x} \in \operatorname{span}\{\vec{e}_1, \vec{e}_2, \ldots, \vec{e}_n\}$, and thus $\mathbb{R}^n \subset \operatorname{span}\{\vec{e}_1, \vec{e}_2, \ldots, \vec{e}_n\}$.

Conversely, since $\vec{e}_i \in \mathbb{R}^n$ for each $i, 1 \le i \le n$ (and \mathbb{R}^n is a vector space), it follows that $\operatorname{span}\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n\} \subseteq \mathbb{R}^n$. The equality now follows.

$$\text{Let } \vec{\mathbf{x}}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \vec{\mathbf{x}}_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \vec{\mathbf{x}}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \end{bmatrix}, \vec{\mathbf{x}}_4 = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}.$$

Does $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\}$ span \mathbb{R}^4 ? (Equivalently, is span $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\} = \mathbb{R}^4$?)

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Does $\{\vec{\mathbf{x}}_1, \vec{\mathbf{x}}_2, \vec{\mathbf{x}}_3, \vec{\mathbf{x}}_4\}$ span \mathbb{R}^4 ? (Equivalently, is span $\{\vec{\mathbf{x}}_1, \vec{\mathbf{x}}_2, \vec{\mathbf{x}}_3, \vec{\mathbf{x}}_4\} = \mathbb{R}^4$?)

Solution

To prove span $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\} = \mathbb{R}^4$, we need to prove two directions:

$$\operatorname{span}\{\vec{x}_1,\vec{x}_2,\vec{x}_3,\vec{x}_4\}\subseteq \mathbb{R}^4\quad \text{and}\quad \mathbb{R}^4\subseteq \operatorname{span}\{\vec{x}_1,\vec{x}_2,\vec{x}_3,\vec{x}_4\}.$$

$$Let \ \vec{x}_1 = \left[\begin{array}{c} 1 \\ 1 \\ 1 \\ 1 \end{array} \right], \vec{x}_2 = \left[\begin{array}{c} 0 \\ 1 \\ 1 \\ 1 \end{array} \right], \vec{x}_3 = \left[\begin{array}{c} 0 \\ 0 \\ 1 \\ 1 \end{array} \right], \vec{x}_4 = \left[\begin{array}{c} 0 \\ 0 \\ 0 \\ 1 \end{array} \right].$$

Does $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\}$ span \mathbb{R}^4 ? (Equivalently, is span $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\} = \mathbb{R}^4$?)

Solution

To prove span $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\} = \mathbb{R}^4$, we need to prove two directions:

$$\operatorname{span}\{\vec{x}_1,\vec{x}_2,\vec{x}_3,\vec{x}_4\}\subseteq\mathbb{R}^4\quad \operatorname{and}\quad \mathbb{R}^4\subseteq\operatorname{span}\{\vec{x}_1,\vec{x}_2,\vec{x}_3,\vec{x}_4\}.$$

For the first relation, since $\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4 \in \mathbb{R}^4$ (and \mathbb{R}^4 is a vector space), span $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\} \subseteq \mathbb{R}^4$.

Solution (continued)

For the second relation, notice that

$$\begin{array}{rcl} \vec{e}_1 & = & \vec{x}_1 - \vec{x}_2 \\ \vec{e}_2 & = & \vec{x}_2 - \vec{x}_3 \\ \vec{e}_3 & = & \vec{x}_3 - \vec{x}_4 \\ \vec{e}_4 & = & \vec{x}_4, \end{array}$$

showing that $\vec{e}_1, \vec{e}_2, \vec{e}_3, \vec{e}_4 \in \text{span}\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\}$. Therefore, since $\text{span}\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\}$ is a vector space,

$$\mathbb{R}^4=\operatorname{span}\{\vec{e}_1,\vec{e}_2,\vec{e}_3,\vec{e}_4\}\subseteq\operatorname{span}\{\vec{x}_1,\vec{x}_2,\vec{x}_3,\vec{x}_4\},$$

and the equality follows.

Show that span $\{\vec{\mathbf{u}}_1, \vec{\mathbf{u}}_2, \vec{\mathbf{u}}_3, \vec{\mathbf{u}}_4\} \neq \mathbb{R}^4$.

$$\text{Let } \vec{\mathbf{u}}_1 = \begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \end{bmatrix}, \vec{\mathbf{u}}_2 = \begin{bmatrix} -1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \vec{\mathbf{u}}_3 = \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}, \vec{\mathbf{u}}_4 = \begin{bmatrix} 1 \\ -1 \\ 1 \\ 1 \end{bmatrix}$$

Show that span $\{\vec{\mathbf{u}}_1, \vec{\mathbf{u}}_2, \vec{\mathbf{u}}_3, \vec{\mathbf{u}}_4\} \neq \mathbb{R}^4$.

Solution

If you check, you'll find that \vec{e}_2 can not be written as a linear combination of $\vec{u}_1, \vec{u}_2, \vec{u}_3$, and \vec{u}_4 .

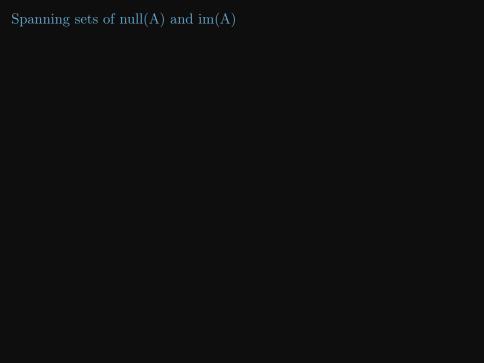
Subspaces of \mathbb{R}^{r}

The null space and the image space

The Eigenspace

Linear Combinations and Spanning Sets

Spanning sets of null(A) and im(A)



Spanning sets of null(A) and im(A)

Lemma

Let A be an $m \times n$ matrix, and let $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$ denote a set of basic solutions to $A\vec{x} = \vec{0}_m$. Then

$$\operatorname{null}(A) = \operatorname{span}\{\vec{x}_1, \cdots, \vec{x}_k\}.$$

Lemma

Let A be an $m\times n$ matrix with columns $\vec{c}_1,\vec{c}_2,\ldots,\vec{c}_n.$ Then

$$\operatorname{im}(A) = \operatorname{span}\{\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n\}.$$

Proof. (of null(A) = span{ $\vec{x}_1, \cdots, \vec{x}_k$ })

"]:" Because $\vec{x}_i \in null(A)$ for each $i,\, 1 \leq i \leq k,$ it follows that

 $\operatorname{span}\{\vec{x}_1,\vec{x}_2,\ldots,\vec{x}_k\}\subseteq\operatorname{null}(A).$

Proof. (of null(A) = span $\{\vec{x}_1, \dots, \vec{x}_k\}$)

"]:" Because $\vec{x}_i \in null(A)$ for each $i,\, 1 \leq i \leq k,$ it follows that

$$\mathrm{span}\{\vec{x}_1,\vec{x}_2,\ldots,\vec{x}_k\}\subseteq \mathrm{null}(A).$$

" \subseteq :" Every solution to $A\vec{x}=\vec{0}_m$ can be expressed as a linear combination of basic solutions, implying that

$$\operatorname{null}(A) \subseteq \operatorname{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}.$$

Therefore, $null(A) = span\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}.$

Proof. (of im(A) = span{ $\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n$ })

" \subseteq :" Suppose $\vec{y} \in \text{im}(A)$. Then (by definition) there is a vector $\vec{x} \in \mathbb{R}^n$ so that $\vec{y} = A\vec{x}$. Write $\vec{x} = \begin{bmatrix} x_1 & x_2 & \dots & x_n \end{bmatrix}^T$. Then

$$ec{y} = Aec{x} = \left[egin{array}{cccc} ec{c}_1 & ec{c}_2 & \dots & ec{c}_n \end{array}
ight] \left[egin{array}{c} x_1 \\ x_2 \\ \vdots \end{array}
ight] = x_1 ec{c}_1 + x_2 ec{c}_2 + \dots + x_n ec{c}_n \end{array}$$

Therefore, $\vec{y} \in \text{span}\{\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n\}$, implying that

$$\operatorname{im}(A) \subseteq \operatorname{span}\{\vec{c}_1,\vec{c}_2,\ldots,\vec{c}_n\}.$$

Proof. (continued)

Notice that for each j, $1 \le j \le n$,

$$\begin{array}{lll} A\vec{e}_{j} & = & \left[\begin{array}{cccc} \vec{c}_{1} & \vec{c}_{2} & \dots & \vec{c}_{n} \end{array} \right] \left[\begin{array}{c} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \\ 0 \end{array} \right] & \leftarrow & jth \; row \\ & \vdots \\ 0 & \end{bmatrix} \\ & = & 0\vec{c}_{1} + 0\vec{c}_{2} + \dots + 0\vec{c}_{j-1} + 1\vec{c}_{j} + 0\vec{c}_{j+1} + \dots + 0\vec{c}_{n} \\ & = & \vec{c}_{j}. \end{array}$$

Thus $\vec{c_j} \in \text{im}(A)$ for each j, $1 \leq j \leq n$. It follows that

$$\operatorname{span}\{\vec{c}_1,\vec{c}_2,\ldots,\vec{c}_n\}\subseteq\operatorname{im}(A),$$

and therefore

$$\operatorname{im}(A) = \operatorname{span}\{\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n\}.$$