

# 25MT103: Linear Algebra

## Unit 4: Real Vector Space

**Dr. D Bhanu Prakash**

Course Page: [dbhanuprakash233.github.io/LA](https://dbhanuprakash233.github.io/LA)

Assistant Professor,  
Department of Mathematics and Statistics.  
Contact: [db\\_maths@vignan.ac.in](mailto:db_maths@vignan.ac.in).  
[dbhanuprakash233.github.io](https://dbhanuprakash233.github.io).



## Real Vector Space - Lecture Slides

# Syllabus

- ☞ Real vector space
- ☞ Subspace
- ☞ Linear dependence and independence
- ☞ Linear span
- ☞ Bases and dimension
- ☞ Row space and column space of a matrix
- ☞ Determining rank of a matrix using row space and column space
- ☞ Row and column spaces of similar matrices

# Outline

- 1 Vectors in a Vector Space
- 2 Linear combination of vectors
- 3 Linear span of Vectors
- 4 Linear dependence and independence
- 5 Real vector spaces
- 6 Subspaces
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# Vectors in a Vector Space

## Definition

Let  $V$  be a real vector space over the field  $\mathbb{R}$ . The elements of  $V$  are called *vectors*. These may be:

- Ordered tuples of real numbers (e.g.  $(x_1, \dots, x_n) \in \mathbb{R}^n$ ),
- Functions  $f : \mathbb{R} \rightarrow \mathbb{R}$  (in function spaces),
- Polynomials  $p(x) = a_0 + a_1x + \dots + a_nx^n$  (in polynomial spaces), or
- Matrices or sequences.

## Notation

Vectors are often denoted by boldface letters ( $\mathbf{v}$ ,  $\mathbf{u}$ ) or arrows ( $\vec{v}$ ), and the zero vector is written  $\mathbf{0}$ .

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# Linear Combination of Vectors

## Definition

A *linear combination* of vectors  $v_1, v_2, \dots, v_k$  in a vector space  $V$  is any vector of the form

$$\alpha_1 v_1 + \alpha_2 v_2 + \cdots + \alpha_k v_k,$$

where  $\alpha_i \in \mathbb{R}$  are scalars.

# Problems: Linear Combination

## Problem 1

Express  $(3, 5)$  as a linear combination of  $(1, 1)$  and  $(1, 2)$ .



## Problems: Linear Combination

### Problem 1

Express  $(3, 5)$  as a linear combination of  $(1, 1)$  and  $(1, 2)$ .

### Solution

We need  $a(1, 1) + b(1, 2) = (3, 5)$ .

$$a + b = 3$$

$$a + 2b = 5.$$

Subtract first from second:  $b = 2$ , hence  $a = 1$ . So  $(3, 5) = 1(1, 1) + 2(1, 2)$ .

# Problems: Linear Combination

## Problem 2

Check if  $(2, 4, 6)$  is a linear combination of  $(1, 0, 1)$ ,  $(0, 1, 1)$ .

# Problems: Linear Combination

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Check if  $(2, 4, 6)$  is a linear combination of  $(1, 0, 1)$ ,  $(0, 1, 1)$ .

## Solution

We need  $a(1, 0, 1) + b(0, 1, 1) = (2, 4, 6)$  giving  $a = 2$ ,  $b = 4$ , and  $a + b = 6$  which holds. Hence yes.

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# Linear Span

## Definition

For a subset  $S \subset V$ , the *span* of  $S$  is

$$\text{span}(S) = \{ \alpha_1 v_1 + \cdots + \alpha_k v_k \mid k \in \mathbb{N}, v_i \in S, \alpha_i \in \mathbb{R} \}.$$

It is the smallest subspace containing  $S$ .

# Problems: Span

## Problem 1

Find  $\text{span}\{(1, 2, 3), (2, 4, 6)\}$  in  $\mathbb{R}^3$ .

# Problems: Span

## Problem 1

Find  $\text{span}\{(1, 2, 3), (2, 4, 6)\}$  in  $\mathbb{R}^3$ .

## Solution

Note  $(2, 4, 6) = 2(1, 2, 3)$  so the span is all multiples of  $(1, 2, 3)$ , i.e. a 1-dimensional subspace  $\{t(1, 2, 3) \mid t \in \mathbb{R}\}$ .

## Problem 2

Does  $\text{span}\{(1, 0, 1), (0, 1, 1)\}$  equal  $\mathbb{R}^3$ ?

# Problems: Span

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## Problem 2

Does  $\text{span}\{(1, 0, 1), (0, 1, 1)\}$  equal  $\mathbb{R}^3$ ?

## Solution

We need to check if an arbitrary  $(x, y, z)$  can be written as  $a(1, 0, 1) + b(0, 1, 1) = (a, b, a + b)$ . This yields  $x = a$ ,  $y = b$ , and  $z = a + b = x + y$ . So  $(x, y, z)$  is in the span iff  $z = x + y$ . Thus the span is the plane  $\{(x, y, z) : z = x + y\}$ , not all  $\mathbb{R}^3$ .



# 3D Visual

Link

<https://trkern.github.io/span3.html>

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# Linear Dependence / Independence

## Definition

A finite set  $\{v_1, \dots, v_k\}$  in  $V$  is *linearly dependent* if there exist scalars  $\alpha_1, \dots, \alpha_k$ , not all zero, such that

$$\alpha_1 v_1 + \dots + \alpha_k v_k = 0.$$

If the only solution is  $\alpha_1 = \dots = \alpha_k = 0$ , the set is *linearly independent*.

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If the only solution is  $\alpha_1 = \dots = \alpha_k = 0$ , the set is *linearly independent*.

## Important properties

- Any set containing the zero vector is linearly dependent.
- Any subset of a linearly independent set is independent.
- If  $k > \dim V$ , any list of  $k$  vectors is dependent.

## Problems: Dependence / Independence

### Problem 1

Determine if the vectors  $(1, 2, 3)$ ,  $(0, 1, 2)$ , and  $(1, 3, 5)$  in  $\mathbb{R}^3$  are linearly independent.

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### Solution

Solve  $a(1, 2, 3) + b(0, 1, 2) + c(1, 3, 5) = (0, 0, 0)$ . This gives the system:

$$a + 0b + c = 0$$

$$2a + b + 3c = 0$$

$$3a + 2b + 5c = 0.$$

From first:  $c = -a$ . Substitute into second:  $2a + b + 3(-a) = 2a + b - 3a = -a + b = 0 \Rightarrow b = a$ . Third:  $3a + 2b + 5c = 3a + 2a + 5(-a) = 0$  — always satisfied. So  $b = a, c = -a$ ; nontrivial solution exists (take  $a = 1$  gives  $(1, 1, -1)$ ). Thus dependent.

## Problems: Dependence / Independence

### Problem 2

Show that the set  $\{(1,0,0), (0,1,0), (0,0,1)\}$  is linearly independent in  $\mathbb{R}^3$ .

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### Problem 2

Show that the set  $\{(1,0,0), (0,1,0), (0,0,1)\}$  is linearly independent in  $\mathbb{R}^3$ .

### Solution

Solve  $a(1,0,0) + b(0,1,0) + c(0,0,1) = 0$ . This yields  $a = b = c = 0$ . Hence independent.



### Question 1

Does any collection of vectors become a vector space?

## Question 2

What is the difference between a set and a space?

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What is the difference between a set and a space?

### Solution

- A **set** is simply a collection of distinct elements with no additional structure.
- A **space** is a set *with structure* — meaning we define extra operations or relations on it.

### Examples:

- Set:  $\{1, 2, 3, 4\}$  (no structure)
- Vector Space:  $\mathbb{R}^2$  under usual addition and scalar multiplication
- Metric Space:  $(\mathbb{R}, d)$  where  $d(x, y) = |x - y|$

**Conclusion:** Every space is a set, but not every set is a space.

## Question 1

Does any collection of vectors become a vector space?

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Does any collection of vectors become a vector space?

### Solution

No. Not every collection of vectors forms a vector space. To be a vector space, a set must satisfy all the **vector space axioms**, including:

- Closure under addition and scalar multiplication
- Existence of a zero vector
- Existence of additive inverses
- Associativity and commutativity of addition
- Distributive and scalar associative laws

If even one property fails, the collection is **not** a vector space.

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# Real Vector Space

## Definition

A *real vector space*  $V$  is a set together with two operations: vector addition  $+: V \times V \rightarrow V$  and scalar multiplication  $\cdot: \mathbb{R} \times V \rightarrow V$ , such that the following axioms hold for all  $u, v, w \in V$  and  $\alpha, \beta \in \mathbb{R}$ :

- ①  $u + v = v + u$  (commutativity),
- ②  $(u + v) + w = u + (v + w)$  (associativity),
- ③ There exists  $0 \in V$  with  $v + 0 = v$  for all  $v$  (Additive identity)
- ④ each  $v$  has an additive inverse  $-v$ ,
- ⑤  $\alpha(v + w) = \alpha v + \alpha w$  (Distributivity)
- ⑥  $(\alpha + \beta)v = \alpha v + \beta v$  (Distributivity)
- ⑦  $(\alpha\beta)v = \alpha(\beta v)$
- ⑧  $1 \cdot v = v$ .

## Important properties

- The zero vector is unique.
- $0 \cdot v = \mathbf{0}$  and  $\alpha \cdot \mathbf{0} = \mathbf{0}$  for all  $\alpha$ .
- $(-1)v = -v$ .



### Example

$\mathbb{R}^n$  with usual addition and scalar multiplication.

## Example

$\mathbb{R}^n$  with usual addition and scalar multiplication.

## Solution

- Vectors in  $\mathbb{R}^n$  can be added:  
$$(x_1, x_2, \dots, x_n) + (y_1, y_2, \dots, y_n) = (x_1 + y_1, x_2 + y_2, \dots, x_n + y_n)$$
- Scalars multiply vectors:  $c(x_1, x_2, \dots, x_n) = (cx_1, cx_2, \dots, cx_n)$
- All vector space axioms hold (closure, zero vector, inverses, etc.)

Hence,  $\mathbb{R}^n$  is the most familiar example of a vector space.

## Counter Example

Give an example of a set that is *not* a vector space.

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### Set

Let  $S = \{(x, y) \in \mathbb{R}^2 : x \geq 0, y \geq 0\}$

## Counter Example

Give an example of a set that is *not* a vector space.

### Set

Let  $S = \{(x, y) \in \mathbb{R}^2 : x \geq 0, y \geq 0\}$

### Solution

- $S$  is closed under addition.
- But not closed under scalar multiplication: for scalar  $-1$ ,  
 $-1 \cdot (1, 1) = (-1, -1) \notin S$ .

**Hence,  $S$  is not a vector space.**

### Example

Can the set of all polynomials with usual addition and scalar multiplication be a vector space?

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### Solution

The set of all real polynomials, denoted  $\mathcal{P} = \{a_0 + a_1x + a_2x^2 + \dots\}$ , is a vector space over  $\mathbb{R}$  because:

- Sum of two polynomials is a polynomial.
- Scalar multiple of a polynomial is a polynomial.
- There exists a zero polynomial (0).

Thus,  $\mathcal{P}$  satisfies all vector space axioms.

### Example

Give a counterexample — a set of polynomials that is *not* a vector space.



### Example

Give a counterexample — a set of polynomials that is *not* a vector space.

### Solution

Let  $S = \{p(x) \in \mathcal{P} : p(0) = 1\}$

- For  $p(x), q(x) \in S$ ,  $(p + q)(0) = p(0) + q(0) = 2 \neq 1$
- So  $S$  is **not closed** under addition

Hence,  $S$  is **not** a vector space.

### Example

Verify that the set  $W = \{f : \mathbb{R} \rightarrow \mathbb{R} \mid f \text{ is continuous}\}$  is a real vector space with pointwise addition and scalar multiplication.

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Verify that the set  $W = \{f : \mathbb{R} \rightarrow \mathbb{R} \mid f \text{ is continuous}\}$  is a real vector space with pointwise addition and scalar multiplication.

### Solution

- 1 Sum of continuous functions is continuous (standard result). Hence closed under addition.
- 2 Scalar multiple of continuous function is continuous. Hence closed under scalar multiplication.
- 3 All axioms (associativity, distributivity, etc.) hold because they hold for real-valued functions pointwise. Therefore  $W$  is a real vector space.

## Examples of Vector Spaces:

- $\mathbb{R}^n, \mathbb{C}^n$
- All polynomials
- All continuous functions  $C[a, b]$

## Counterexamples:

- $\{(x, y) \in \mathbb{R}^2 : x \geq 0, y \geq 0\}$
- $\{p(x) : p(0) = 1\}$

# Summary

How do we check if something is a vector space?

**Answer:** Verify the following 10 axioms (for addition and scalar multiplication):

- 1 Closure under addition
- 2 Commutativity of addition
- 3 Associativity of addition
- 4 Existence of additive identity
- 5 Existence of additive inverse
- 6 Closure under scalar multiplication
- 7 Distributive property (scalar over vectors)
- 8 Distributive property (vector over scalars)
- 9 Compatibility of scalar multiplication
- 10 Existence of multiplicative identity

If all hold — it's a vector space!

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# Subspace

## Definition

A subset  $U$  of a vector space  $V$  is a *subspace* if  $U$  is itself a vector space under the operations inherited from  $V$ . Equivalently,  $U$  is nonempty and closed under addition and scalar multiplication.

## Important properties

- Intersection of subspaces is a subspace.
- Sum  $U + W = \{u + w \mid u \in U, w \in W\}$  is the smallest subspace containing  $U \cup W$ .

# Problems: Subspaces

## Problem 1

Let  $U = \{(x, x, 0) \mid x \in \mathbb{R}\} \subset \mathbb{R}^3$ . Show  $U$  is a subspace.



# Problems: Subspaces

## Problem 1

Let  $U = \{(x, x, 0) \mid x \in \mathbb{R}\} \subset \mathbb{R}^3$ . Show  $U$  is a subspace.

## Solution

- ① Nonempty:  $(0, 0, 0) \in U$ .
- ② Closed under addition:  $(x, x, 0) + (y, y, 0) = (x + y, x + y, 0) \in U$ .
- ③ Closed under scalars:  $\alpha(x, x, 0) = (\alpha x, \alpha x, 0) \in U$ .
- ④ So  $U$  is a subspace.

# Problems: Subspaces

## Problem 2

Is the set  $S = \{(x, y, z) \in \mathbb{R}^3 \mid x^2 + y^2 + z^2 = 1\}$  a subspace of  $\mathbb{R}^3$ ?

## Problems: Subspaces

### Problem 2

Is the set  $S = \{(x, y, z) \in \mathbb{R}^3 \mid x^2 + y^2 + z^2 = 1\}$  a subspace of  $\mathbb{R}^3$ ?

### Solution

This set is the unit sphere. It does not contain the zero vector, and it's not closed under scalar multiplication (e.g.  $2 \cdot (1, 0, 0) = (2, 0, 0)$  not in  $S$ ). Hence not a subspace.

## Problem

Let  $V = \{(x, y, z) \in \mathbb{R}^3 : x + y + z = 0\}$ . Show  $V$  is a real vector space (subspace of  $\mathbb{R}^3$ ).

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Let  $V = \{(x, y, z) \in \mathbb{R}^3 : x + y + z = 0\}$ . Show  $V$  is a real vector space (subspace of  $\mathbb{R}^3$ ).

## Solution

- ① Check closure under addition: take  $u = (x_1, y_1, z_1)$  and  $v = (x_2, y_2, z_2)$  with sums zero. Then

$$(x_1 + x_2) + (y_1 + y_2) + (z_1 + z_2) = (x_1 + y_1 + z_1) + (x_2 + y_2 + z_2) = 0 + 0 = 0,$$

so  $u + v \in V$ .

- ② Check closure under scalar multiplication: for  $\alpha \in \mathbb{R}$ ,

$$\alpha x + \alpha y + \alpha z = \alpha(x + y + z) = \alpha \cdot 0 = 0,$$

so  $\alpha v \in V$ .

- ③ Nonempty:  $(1, -1, 0) \in V$ , so  $V \neq \emptyset$ . Thus  $V$  is a subspace and hence a real vector space.

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# Bases and Dimension

## Definition

A *basis* of a vector space  $V$  is a list of vectors that is linearly independent and spans  $V$ . The *dimension* of  $V$ ,  $\dim V$ , is the number of vectors in any basis (if finite).

# Bases and Dimension

## Definition

A *basis* of a vector space  $V$  is a list of vectors that is linearly independent and spans  $V$ . The *dimension* of  $V$ ,  $\dim V$ , is the number of vectors in any basis (if finite).

## Important properties

- Every finite-dimensional vector space has a basis.
- All bases of a finite-dimensional vector space have the same number of elements.
- If  $\dim V = n$ , any list of  $n$  independent vectors is a basis; any spanning list with  $n$  vectors is a basis.



# Problems: Bases and Dimension

## Problem 1

Find a basis and dimension of  $V = \{(x, y, z) \in \mathbb{R}^3 \mid x + 2y + 3z = 0\}$ .

# Problems: Bases and Dimension

## Problem 1

Find a basis and dimension of  $V = \{(x, y, z) \in \mathbb{R}^3 \mid x + 2y + 3z = 0\}$ .

## Solution

Solve  $x = -2y - 3z$ . Then vectors are  $(x, y, z) = y(-2, 1, 0) + z(-3, 0, 1)$ . So  $V = \text{span}\{(-2, 1, 0), (-3, 0, 1)\}$ . Check independence: only trivial combination gives zero (quick determinant or reasoning), hence they form a basis. Dimension = 2.

# Problems: Bases and Dimension

## Problem 2

Show that any two bases of a finite-dimensional vector space have the same number of elements.

# Problems: Bases and Dimension

## Problem 2

Show that any two bases of a finite-dimensional vector space have the same number of elements.

## Solution

Let  $B$  and  $B'$  be bases. Since  $B'$  spans, each vector of  $B$  is a linear combination of  $B'$ . Using the exchange lemma one shows  $|B| \leq |B'|$ . By symmetry  $|B'| \leq |B|$ . So  $|B| = |B'|$ .

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# Row space and Column space

## Definitions

For a matrix  $A \in \mathbb{R}^{m \times n}$ :

- The *row space*  $\mathcal{R}(A)$  is the subspace of  $\mathbb{R}^n$  spanned by the rows of  $A$  (viewed as row vectors).
- The *column space*  $\mathcal{C}(A)$  is the subspace of  $\mathbb{R}^m$  spanned by the columns of  $A$ .
- The *rank* of  $A$ ,  $\text{rank}(A)$ , is  $\dim \mathcal{R}(A) = \dim \mathcal{C}(A)$ .

# Row space and Column space

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- The *column space*  $\mathcal{C}(A)$  is the subspace of  $\mathbb{R}^m$  spanned by the columns of  $A$ .
- The *rank* of  $A$ ,  $\text{rank}(A)$ , is  $\dim \mathcal{R}(A) = \dim \mathcal{C}(A)$ .

## Important properties

- Row reductions (elementary row operations) do not change the row space (they change which rows generate it) or the row rank; column operations change column space.
- Row rank equals column rank (fundamental theorem of linear algebra).

# Problems: Row/Column Space and Rank

## Problem 1

Find the row space, column space and rank of

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 1 & 1 & 1 \end{pmatrix}.$$

## Solution

Row-reduce  $A$ :

$$\begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 1 & 1 & 1 \end{pmatrix} \xrightarrow{R_2 - 2R_1, R_3 - R_1} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 0 & 0 \\ 0 & -1 & -2 \end{pmatrix}$$



# Problems: Row/Column Space and Rank

## Solution

Swap rows  $R_2 \leftrightarrow R_3$  and scale  $R_2 \rightarrow -R_2$ :

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix}.$$

- Pivots in columns 1 and 2. Rank = 2.
- Row space: span of pivot rows  $\{(1, 2, 3), (0, 1, 2)\}$ .
- Column space: span of  $\{c_1 = (1, 2, 1), c_2 = (2, 4, 1), c_3 = (3, 6, 1)\}$ .

# Problems: Row/Column Space and Rank

## Problem 2

Compute rank of  $B = \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 3 \\ 2 & 3 & 7 \end{pmatrix}$ .

# Problems: Row/Column Space and Rank

## Problem 2

Compute rank of  $B = \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 3 \\ 2 & 3 & 7 \end{pmatrix}$ .

## Solution

Row reduce:

$$\begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 3 \\ 2 & 3 & 7 \end{pmatrix} \xrightarrow{R_3 - 2R_1} \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 3 \\ 0 & 3 & 3 \end{pmatrix} \xrightarrow{R_3 - 3R_2} \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 3 \\ 0 & 0 & -6 \end{pmatrix}.$$

All three pivots; rank = 3.

# Row and Column Spaces of Similar Matrices

## Statement

If  $A$  and  $B$  are similar matrices ( $B = P^{-1}AP$  for invertible  $P$ ), then  $\text{rank}(A) = \text{rank}(B)$  and the row/column spaces are isomorphic via the change of basis given by  $P$ .

## Important properties

- Similarity represents the same linear operator in different bases, so invariant properties like rank, determinant, trace, eigenvalues (multiset) are preserved.

## Problem: Similar Matrices

### Problem 1

Let  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  and  $P = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ . Compute  $B = P^{-1}AP$  and compare ranks and column spaces.

## Problem: Similar Matrices

### Problem 1

Let  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  and  $P = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ . Compute  $B = P^{-1}AP$  and compare ranks and column spaces.

### Solution

Compute  $P^{-1} = \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix}$ . Then

$$B = P^{-1}AP = \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} = A.$$

Here  $B = A$ , ranks and column spaces identical. (This choice of  $P$  centralizes  $A$ .)

## Problem: Similar Matrices

### Problem 2

Explain why rank is invariant under similarity.

## Problem: Similar Matrices

### Problem 2

Explain why rank is invariant under similarity.

### Solution

$B = P^{-1}AP$ . Multiplication by invertible matrices on left/right corresponds to applying invertible linear maps to row/column spaces, which preserves dimensions. Hence rank preserved.



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# Summary and Takeaways

- Vector spaces, subspaces: the foundational language.
- Linear dependence/independence, span, bases, dimension — tightly interconnected.
- Row/column spaces and rank are central in matrix theory; rank preserved under similarity.

# Thank You!

**Dr. D Bhanu Prakash**

[dbhanuprakash233.github.io](https://dbhanuprakash233.github.io)

Mail: [db\\_maths@vignan.ac.in](mailto:db_maths@vignan.ac.in)

I can't change the direction  
of the wind, but I can adjust  
my sails to always reach  
my destination.

(Jimmy Dean)

