



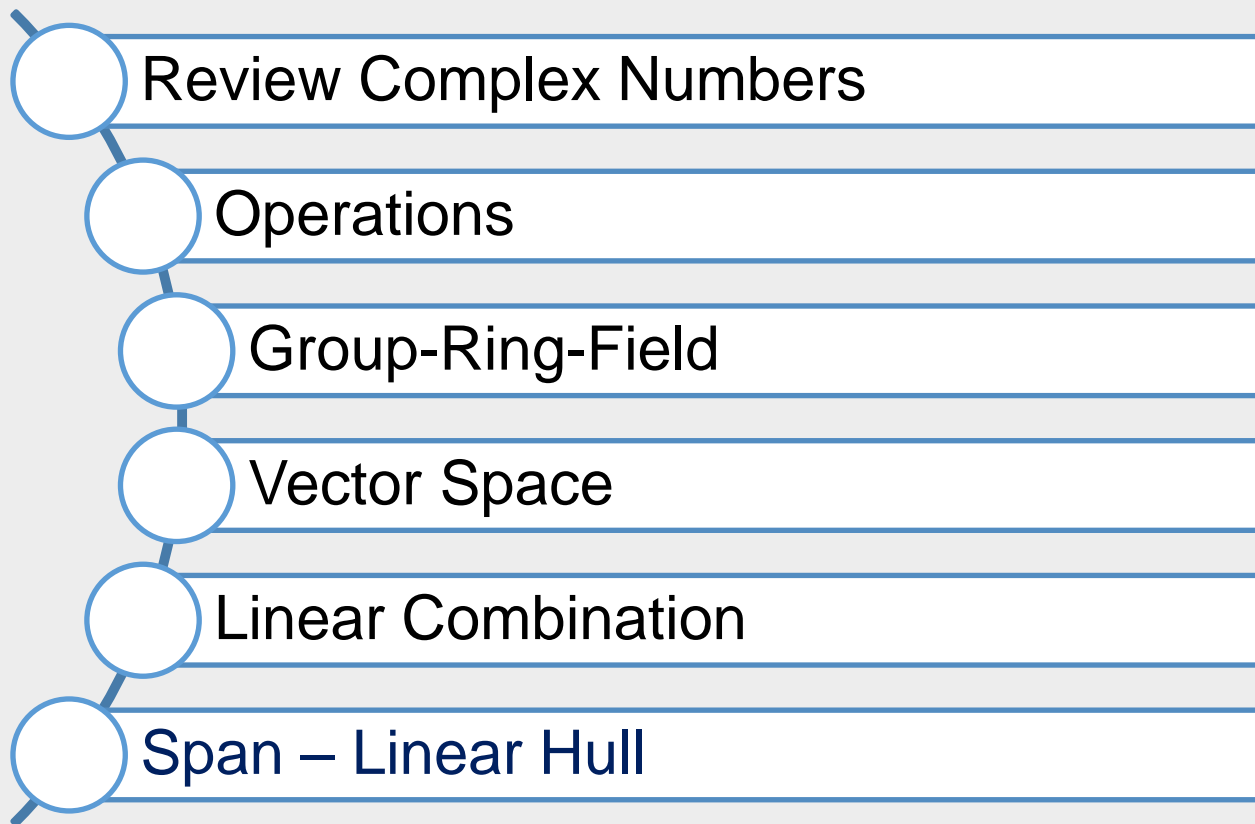
Vector Space

Linear Algebra

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Complex Number Review

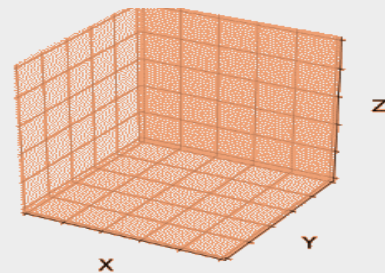
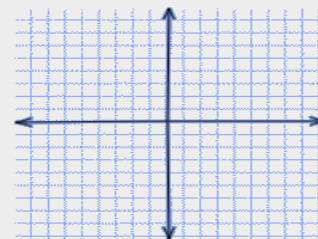
Definition

□ A tuple is an ordered list of numbers.

□ For example: $\begin{bmatrix} 1 \\ 2 \\ 32 \\ 10 \end{bmatrix}$ is a 4-tuple (a tuple with 4 elements).

$$\mathbb{R}^2 = \left\{ \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 0.112 \\ \frac{2}{3} \end{pmatrix}, \begin{pmatrix} \pi \\ e \end{pmatrix}, \dots \right\}$$

$$\mathbb{R}^3 = \left\{ \begin{pmatrix} 17 \\ \pi \\ 2 \end{pmatrix}, \begin{pmatrix} 9 \\ -2 \\ \sqrt{2} \end{pmatrix}, \begin{pmatrix} 1 \\ 22 \\ 2 \end{pmatrix}, \dots \right\}$$





Numbers:

- Real: Nearly any number you can think of is a Real Number!

1	12.38	-0.8625	3/4	$\sqrt{2}$	1998
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- Imaginary: When squared give a negative result.

The “unit” imaginary number (like 1 for Real Numbers) is “ i ”, which is the square root of -1 .

Examples of Imaginary Numbers:

$3i$	$1.04i$	$-2.8i$	$3i/4$	$(\sqrt{2})i$	$1998i$
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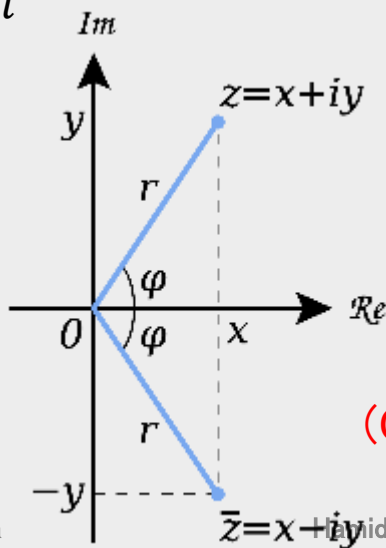
And we keep that little “ i ” there to remind us we need to multiply by $\sqrt{-1}$

Review: Complex Numbers

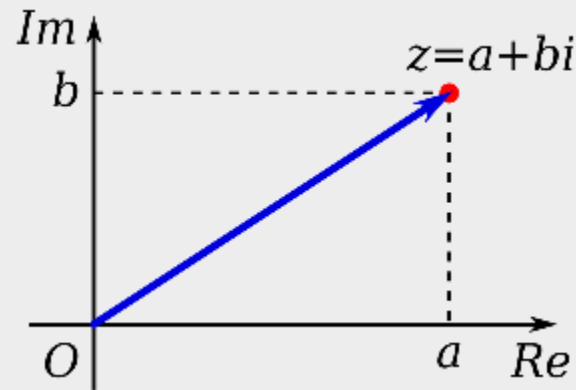


- \mathbb{C} is a plane, where number $(a + bi)$ has coordinates $\begin{bmatrix} a \\ b \end{bmatrix}$
- Imaginary number: bi , $b \in \mathbb{R}$

- Conjugate of $x + yi$ is noted by $\overline{x + yi}$:
 - $x - yi$



(Complex conjugate)





□ Arithmetic with complex numbers $(a + bi)$:

□ $(a + bi) + (c + di)$

□ $(a + bi)(c + di)$

□ $\frac{a+bi}{c+di}$

$$\frac{a + bi}{c + di} = \frac{(a + bi)(c - di)}{(c + di)(c - di)} = \frac{ac + bd}{c^2 + d^2} + \left(\frac{bc - ad}{c^2 + d^2} \right) i$$

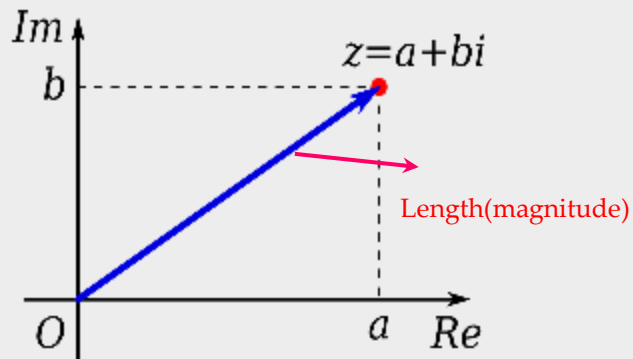


□ Length (magnitude): $||a + bi||^2 = \overline{(a + bi)}(a + bi) = a^2 + b^2$

□ Inner Product:

□ Real: $\langle x, y \rangle = x_1y_1 + x_2y_2 + \dots + x_ny_n$

□ Complex: $\langle x, y \rangle = \overline{x_1}y_1 + \overline{x_2}y_2 + \dots + \overline{x_n}y_n$



Extra resource:

If you want to learn more about complex numbers, [this](#) video is recommended!

Vector Operation



- ❑ Vector–Vector Addition
- ❑ Vector–Vector Subtraction
- ❑ Scalar–Vector Product
- ❑ Vector–Vector Products:
 - $x \cdot y$ is called the **inner product** or **dot product** or **scalar product** of the vectors: $x^T y$ ($y^T x$)
 - $\langle a, b \rangle$ $\langle a|b \rangle$ (a, b) $a \cdot b$
 - $$x^T y \in \mathbb{R} = \begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = \sum_{i=1}^n x_i y_i.$$
 - Transpose of dot product:
 - $(a \cdot b)^T = (a^T b)^T = (b^T a) = (b \cdot a) = b^T a$
 - Length of vector



❑ Commutativity

- The order of the two vector arguments in the inner product does not matter.

$$a^T b = b^T a$$

❑ Distributivity with vector addition

- The inner product can be distributed across vector addition.

$$\begin{aligned}(a + b)^T c &= a^T c + b^T c \\ a^T (b + c) &= a^T b + a^T c\end{aligned}$$



- Bilinear (linear in both a and b)

$$a^T(\lambda b + \beta c) = \lambda a^T b + \beta a^T c$$

- Positive Definite:

$$(a, a) = a^T a \geq 0$$

- 0 only if a itself is a zero vector $a = \mathbf{0}$



□ Associative

- Note: the associative law is that parentheses can be moved around, e.g., $(x+y)+z = x+(y+z)$ and $x(yz) = (xy)z$

1) Associative property of the vector dot product with a scalar (scalar–vector multiplication embedded inside the dot product)

$$\begin{aligned} \text{scalar} \rightarrow \gamma(\mathbf{u}^T \mathbf{v}) &= (\gamma \mathbf{u}^T) \mathbf{v} = \mathbf{u}^T (\gamma \mathbf{v}) = (\mathbf{u}^T \mathbf{v}) \gamma \\ &= (\gamma \mathbf{u})^T \mathbf{v} = \gamma \mathbf{u}^T \mathbf{v} \end{aligned}$$



□ Associative

2) Does vector dot product obey the associative property?

$$\underbrace{\mathbf{u}^T (\mathbf{v}^T \mathbf{w})}_{\substack{\text{vector-scalar product} \\ \text{row vector}}} = \underbrace{(\mathbf{u}^T \mathbf{v})^T \mathbf{w}}_{\substack{\text{scalar-vector product} \\ \text{column vector}}}$$

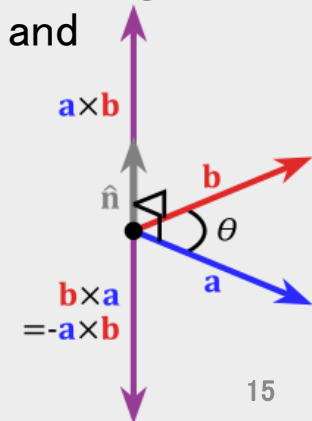
- The cross product is defined only for two 3-element vectors, and the result is another 3-element vector. It is commonly indicated using a multiplication symbol (\times).

$$\|\mathbf{a} \times \mathbf{b}\| = \|\mathbf{a}\| \|\mathbf{b}\| \sin(\theta_{ab})$$

$$\mathbf{a} \times \mathbf{b} = \begin{bmatrix} a_2 b_3 - a_3 b_2 \\ a_3 b_1 - a_1 b_3 \\ a_1 b_2 - a_2 b_1 \end{bmatrix}$$

- It is used often in geometry, for example to create a vector \mathbf{c} that is orthogonal to the plane spanned by vectors \mathbf{a} and \mathbf{b} . It is also used in vector and multivariate calculus to compute surface integrals.

u_1	v_1	
u_2	v_2	
u_3	v_3	$u_2 v_3 - u_3 v_2$
u_1	v_1	$u_3 v_1 - u_1 v_3$
u_2	v_2	$u_1 v_2 - u_2 v_1$





□ Vector-Vector Products:

- Given two vectors $x \in \mathbb{R}^m, y \in \mathbb{R}^n$:

- $x \otimes y = xy^T \in \mathbb{R}^{m \times n}$ is called the outer product of the vectors: $(xy^T)_{ij}$
 $= x_i y_j$
 $xy^T \in \mathbb{R}^{m \times n} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix} \begin{bmatrix} y_1 & y_2 & \cdots & y_n \end{bmatrix} = \begin{bmatrix} x_1 y_1 & x_1 y_2 & \cdots & x_1 y_n \\ x_2 y_1 & x_2 y_2 & \cdots & x_2 y_n \\ \vdots & \vdots & \ddots & \vdots \\ x_m y_1 & x_m y_2 & \cdots & x_m y_n \end{bmatrix}$

Example

- Represent $A \in \mathbb{R}^{m \times n}$ with outer product of two vectors:

$$A = \begin{bmatrix} | & | & \cdots & | \\ x & x & \cdots & x \\ | & | & \cdots & | \end{bmatrix} = \begin{bmatrix} x_1 & x_1 & \cdots & x_1 \\ x_2 & x_2 & \cdots & x_2 \\ \vdots & \vdots & \ddots & \vdots \\ x_m & x_m & \cdots & x_m \end{bmatrix}$$



□ Properties:

- $(u \otimes v)^T = (v \otimes u)$
- $(v + w) \otimes u = v \otimes u + w \otimes u$
- $u \otimes (v + w) = u \otimes v + u \otimes w$
- $c(v \otimes u) = (cv) \otimes u = v \otimes (cu)$
- $(u \cdot v) = \text{trace}(u \otimes v) \quad (u, v \in R^n)$
- $(u \otimes v)w = (v \cdot w)u$



- ❑ Vector–Vector Products:
 - Hadamard
 - Element–wise product

$$c = a \odot b = \begin{bmatrix} a_1 b_1 \\ a_2 b_2 \\ \vdots \\ a_n b_n \end{bmatrix}$$

- ❑ Hadamard product is used in image compression techniques such as JPEG. It is also known as Schur product
- ❑ Hadamard Product is used in LSTM (Long Short–Term Memory) cells of Recurrent Neural Networks (RNNs).



□ Properties:

- $a \odot b = b \odot a$
- $a \odot (b \odot c) = (a \odot b) \odot c$
- $a \odot (b + c) = a \odot b + a \odot c$
- $(\theta a) \odot b = a \odot (\theta b) = \theta(a \odot b)$
- $a \odot \mathbf{0} = \mathbf{0} \odot a = \mathbf{0}$

Binary Operation



Definition

□ Any function from $A \times A \rightarrow A$ is a binary operation.

□ Closure Law:

□ A set is said to be closure under an operation (like addition, subtraction, multiplication, etc.) if that operation is performed on elements of that set and result also lies in set.

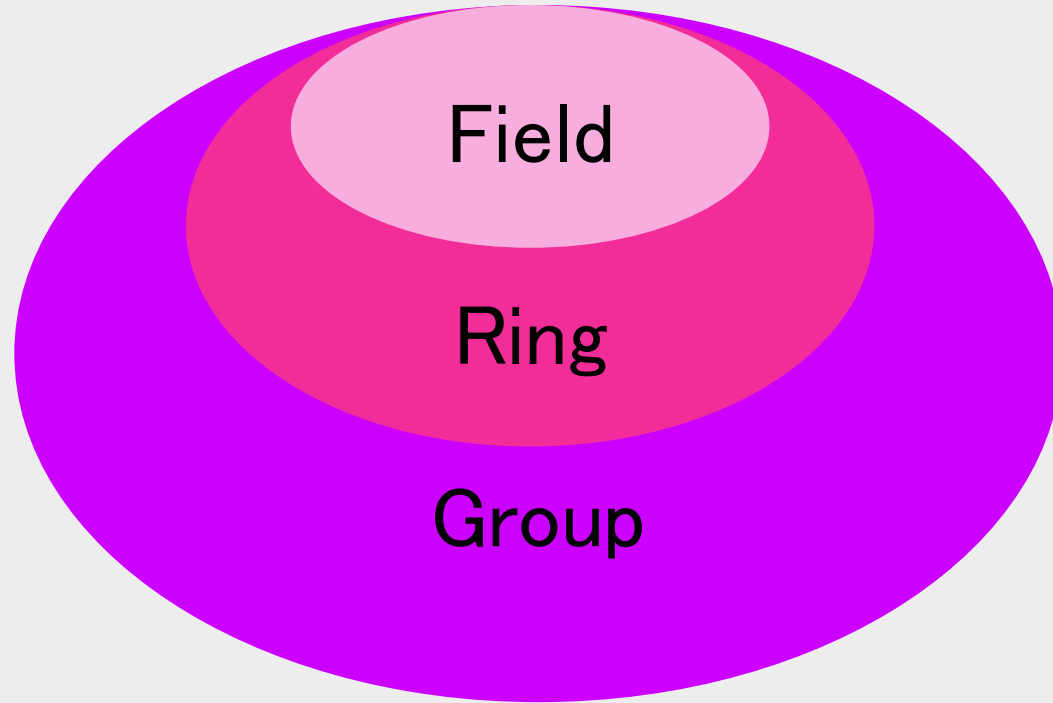
$$\text{if } a \in A, b \in A \rightarrow a * b \in A$$



Example

- ☐ Is “+” a binary operator on natural numbers?
- ☐ Is “ \times ” a binary operator on natural numbers?
- ☐ Is “−” a binary operator on natural numbers?
- ☐ Is “/” a binary operator on natural numbers?

Group–Ring–Field





Definition

- A group G is a pair (S, \circ) , where S is a set and \circ is a binary operation on S such that:
- \circ is **associative**
- **(Identity)** There exists an element $e \in S$ such that:

$$e \circ a = a \circ e = a \quad \forall a \in S$$

- **(Inverses)** For every $a \in S$ there is $b \in S$ such that:

$$a \circ b = b \circ a = e$$

If \circ is commutative, then G is called a **commutative group**!



Definition

□ A **ring** R is a set together with two binary operations $+$ and $*$, satisfying the following properties:

1. $(R, +)$ is a commutative group
 2. $*$ is associative
 3. The **distributive laws** hold in R : (Multiplication is distributive over addition)
- Associative
 - Identity
 - Inverses
 - Commutative


$$(a + b) * c = (a * c) + (b * c)$$

$$a * (b + c) = (a * b) + (a * c)$$



Definition

- A **field** F is a set together with two binary operations $+$ and $*$, satisfying the following properties:

1. $(F, +)$ is a commutative group 
 - Associative
 - Identity
 - Inverses
 - Commutative
2. $(F - \{0\}, *)$ is a commutative group
3. The distributive law holds in F :

$$(a + b) * c = (a * c) + (b * c)$$

$$a * (b + c) = (a * b) + (a * c)$$



- ❑ A field in mathematics is a set of things of elements (not necessarily numbers) for which the basic arithmetic operations (addition, subtraction, multiplication, division) are defined: $(F, +, \cdot)$

Example

$(\mathbb{R}; +, \cdot)$ and $(\mathbb{Q}; +, \cdot)$ serve as examples of fields.
 $(\mathbb{Z}; +, \cdot)$ is an example of a ring which is not a field!

- ❑ Field is a set (F) with two binary operations $(+ , \cdot)$ satisfying following properties:



Properties	Binary Operations	
	Addition (+)	Multiplication (.)
Closure (بسته بودن)	$\exists a + b \in F$	$\exists a.b \in F$
Associative (شرکت پذیری)	$a + (b + c) = (a + b) + c$	$a.(b.c) = (a.b).c$
Commutative (جابه جایی پذیری)	$a + b = b + a$	$a.b = b.a$
Existence of identity $e \in F$	$a + e = a = e + a$	$a.e = a = e.a$
Existence of inverse: For each a in F there <u>must exist</u> b_1 in F	$a + b = e = b + a$	$a.b = e = b.a$ <u>For any nonzero a</u>
Multiplication is distributive over addition $a.(b + c) = a.b + a.c$ $(a + b).c = a.c + b.c$		



Example

Set $B = \{0,1\}$ under following operations is a field?

+	0	1
0	0	1
1	1	0

.	0	1
0	0	0
1	0	1

Example

Which are fields? (two binary operations $+$, $*$)

\mathbb{R}

\mathbb{C}

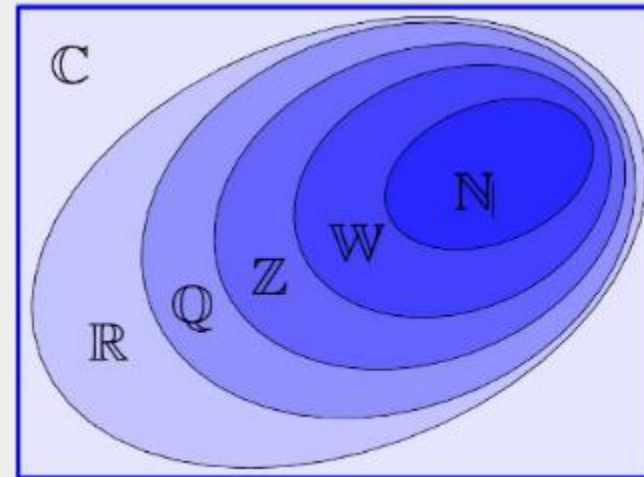
\mathbb{Q}

\mathbb{Z}

\mathbb{W}

\mathbb{N}

$\mathbb{R}^{2 \times 2}$



\mathbb{C} : Complex

\mathbb{R} : Real

\mathbb{Q} : Rational

\mathbb{Z} : Integer

\mathbb{W} : Whole

\mathbb{N} : Natural

Vector Space



- ❑ Building blocks of linear algebra.
- ❑ A **non-empty set V** with **field F** (most of time \mathbb{R} or \mathbb{C}) forms a vector space with two operations:
 1. $+$: Binary operation on V which is $V \times V \rightarrow V$
 2. \cdot : $F \times V \rightarrow V$

Note

In our course, by **default**, field is \mathbb{R} (real numbers).



Definition

A vector space over a field F is the set V equipped with two operations: $(V, F, +, \cdot)$

- i. **Vector addition:** denoted by “+” adds two elements $x, y \in V$ to produce another element $x + y \in V$
- ii. **Scalar multiplication:** denoted by “.” multiplies a vector $x \in V$ with a scalar $\alpha \in F$ to produce another vector $\alpha \cdot x \in V$. We usually omit the “.” and simply write this vector as αx



□ Addition of vector space ($x + y$)

□ **Commutative** $x + y = y + x \quad \forall x, y \in V$

□ **Associative** $(x + y) + z = x + (y + z) \quad \forall x, y, z \in V$

□ **Additive identity** $\exists \mathbf{0} \in V$ such that $x + \mathbf{0} = x, \forall x \in V$

□ **Additive inverse** $\exists (-x) \in V$ such that $x + (-x) = 0, \forall x \in V$



□ Action of the scalars field on the vector space (αx)

□ **Associative** $\alpha(\beta x) = (\alpha\beta)x$ $\forall \alpha, \beta \in F; \forall x \in V$

□ **Distributive over**

scalar addition: $(\alpha + \beta)x = \alpha x + \beta x$ $\forall \alpha, \beta \in F; \forall x \in V$

vector addition: $\alpha(x + y) = \alpha x + \alpha y$ $\forall \alpha \in F; \forall x, y \in V$

□ **Scalar identity** $1x = x$ $\forall x \in V$



Example

Let V be the set of all real numbers with the operations $u \oplus v = u - v$, (\oplus is an ordinary subtraction) and $c \odot u = cu$ (\odot is an ordinary multiplication). Is V a vector space? If it's not, which properties fail to hold?



Example: Fields are \mathbb{R} in this example:

- The n -tuple space,
- The space of $m \times n$ matrices
- The space of functions:

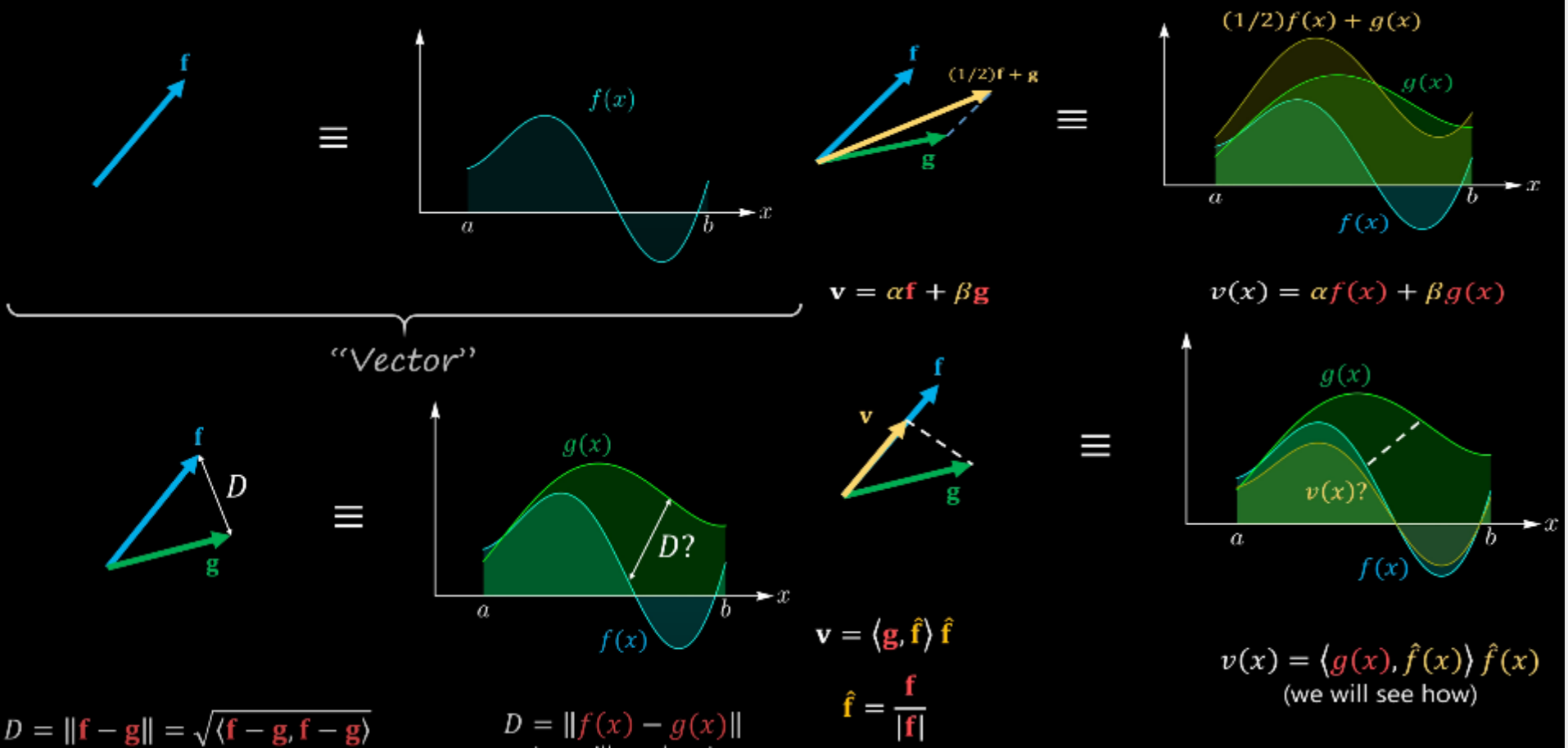
$$(f + g)(x) = f(x) + g(x) \quad \text{and} \quad (cf)(x) = cf(x)$$

$$f(t) = 1 + \sin(2t) \quad \text{and} \quad g(t) = 2 + 0.5t$$

- The space of polynomial functions over a field $f(x)$:

$$p_n(t) = a_0 + a_1t + a_2t^2 + \dots + a_nt^n$$

Vector Space





- Function addition and scalar multiplication

$$(f + g)(x) = f(x) + g(x) \quad \text{and} \quad (af)(x) = af(x)$$

Non-empty set X and any field F \longrightarrow $F^X = \{f: X \rightarrow F\}$

Example

- Set of all polynomials with real coefficients
- Set of all real-valued continuous function on $[0,1]$
- Set of all real-valued function that are differentiable on $[0,1]$



$P_n(\mathbb{R})$: Polynomials with max degree (n)

- ❑ Vector addition
- ❑ Scalar multiplication
- ❑ And other 8 properties!



Example

Which are vector spaces?

- ☐ Set \mathbb{R}^n over \mathbb{R}
- ☐ Set \mathbb{C} over \mathbb{R}
- ☐ Set \mathbb{R} over \mathbb{C}
- ☐ Set \mathbb{Z} over \mathbb{R}
- ☐ Set of all polynomials with coefficient from \mathbb{R} over \mathbb{R}
- ☐ Set of all polynomials of degree at most n with coefficient from \mathbb{R} over \mathbb{R}
- ☐ Matrix: $M_{m,n}(\mathbb{R})$ over \mathbb{R}
- ☐ Function: $f(x): x \rightarrow \mathbb{R}$ over \mathbb{R}



The operations on field F are:

- $+: F \times F \rightarrow F$
- $\times: F \times F \rightarrow F$

The operations on a vector space V over a field F are:

- $+: V \times V \rightarrow V$
- $\cdot: F \times V \rightarrow V$

Linear Combination



- The **linear combinations** of m vectors a_1, \dots, a_m , each with size n is:

$$\beta_1 a_1 + \dots + \beta_m a_m$$

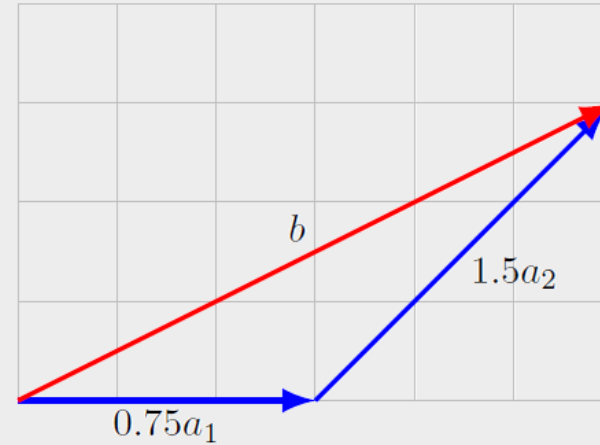
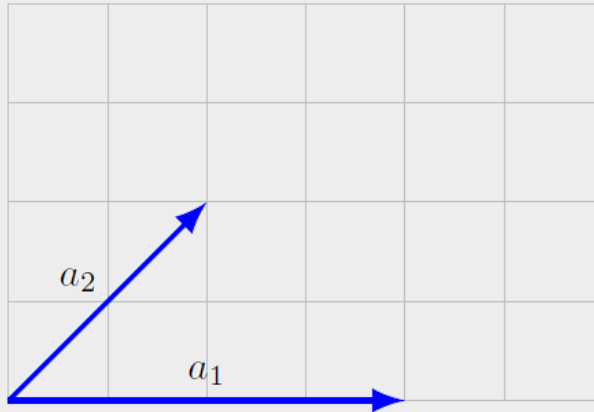
where β_1, \dots, β_m are scalars and called the **coefficients of the linear combination**

- **Coordinates**: We can write any n -vector b as a **linear combination of the standard unit vectors**, as:

$$b = b_1 e_1 + \dots + b_n e_n$$

- Example: What are the coefficients and combination for this vector?

$$\begin{bmatrix} -1 \\ 3 \\ 5 \end{bmatrix}$$



Left. Two 2-vectors a_1 and a_2 . *Right.* The linear combination $b = 0.75a_1 + 1.5a_2$

Special Linear Combinations

- ❑ Sum of vectors
- ❑ Average of vectors

Span – Linear Hull



Definition

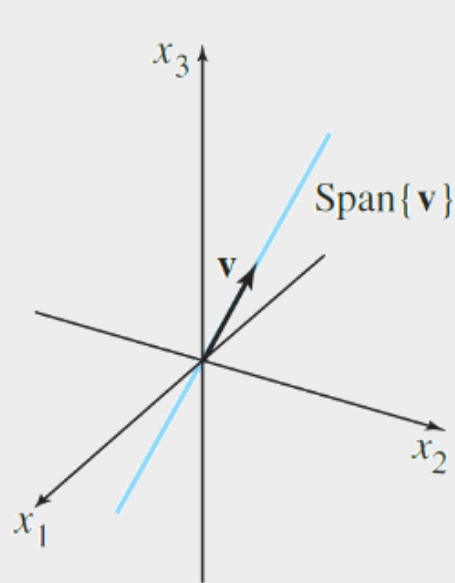
If $v_1, v_2, v_3, \dots, v_p$ are in \mathbb{R}^n , then the set of all linear combinations of v_1, v_2, \dots, v_p is denoted by $\text{Span}\{v_1, v_2, \dots, v_p\}$ and is called the **subset of \mathbb{R}^n spanned (or generated) by v_1, v_2, \dots, v_p** .

That is, $\text{Span}\{v_1, v_2, \dots, v_p\}$ is the collection of all vectors that can be written in the form:

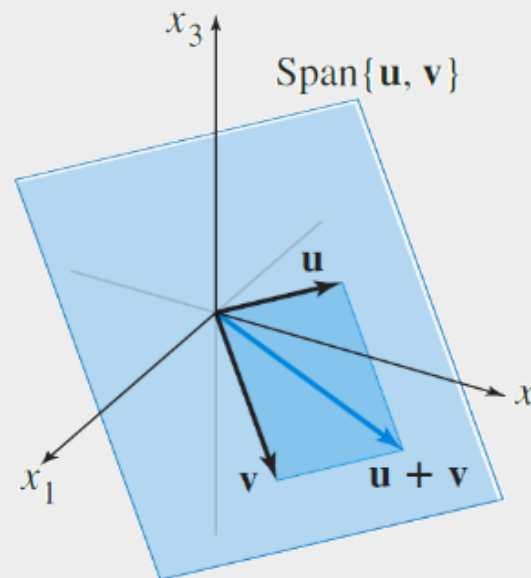
$$c_1 v_1 + c_2 v_2 + \dots + c_p v_p$$

with c_1, c_2, \dots, c_p being scalars.

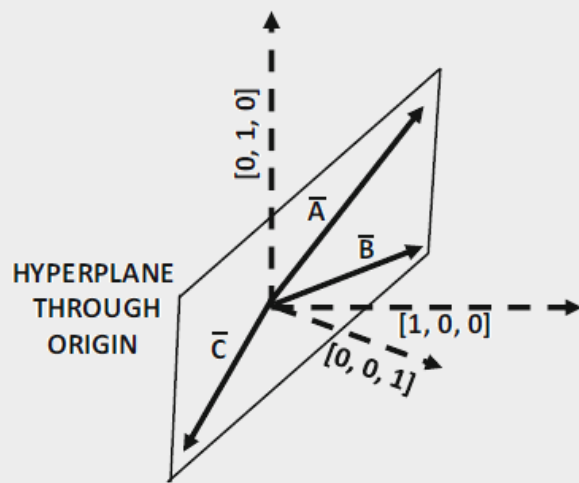
\mathbf{v} and \mathbf{u} are non-zero vectors in \mathbb{R}^3 where \mathbf{v} is not a multiple of \mathbf{u}



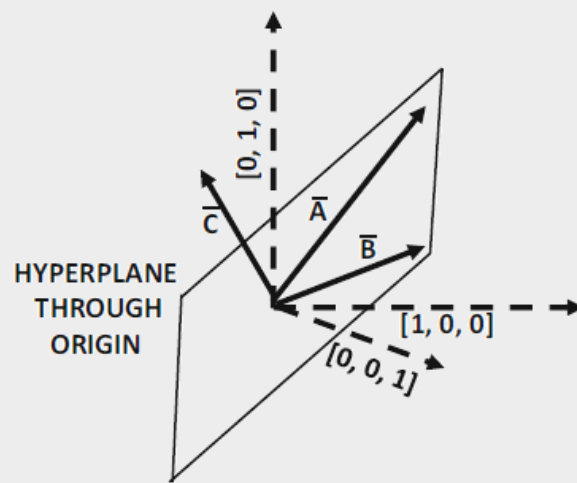
$\text{Span}\{\mathbf{v}\}$ as a line through the origin.



$\text{Span}\{\mathbf{u}, \mathbf{v}\}$ as a plane through the origin.

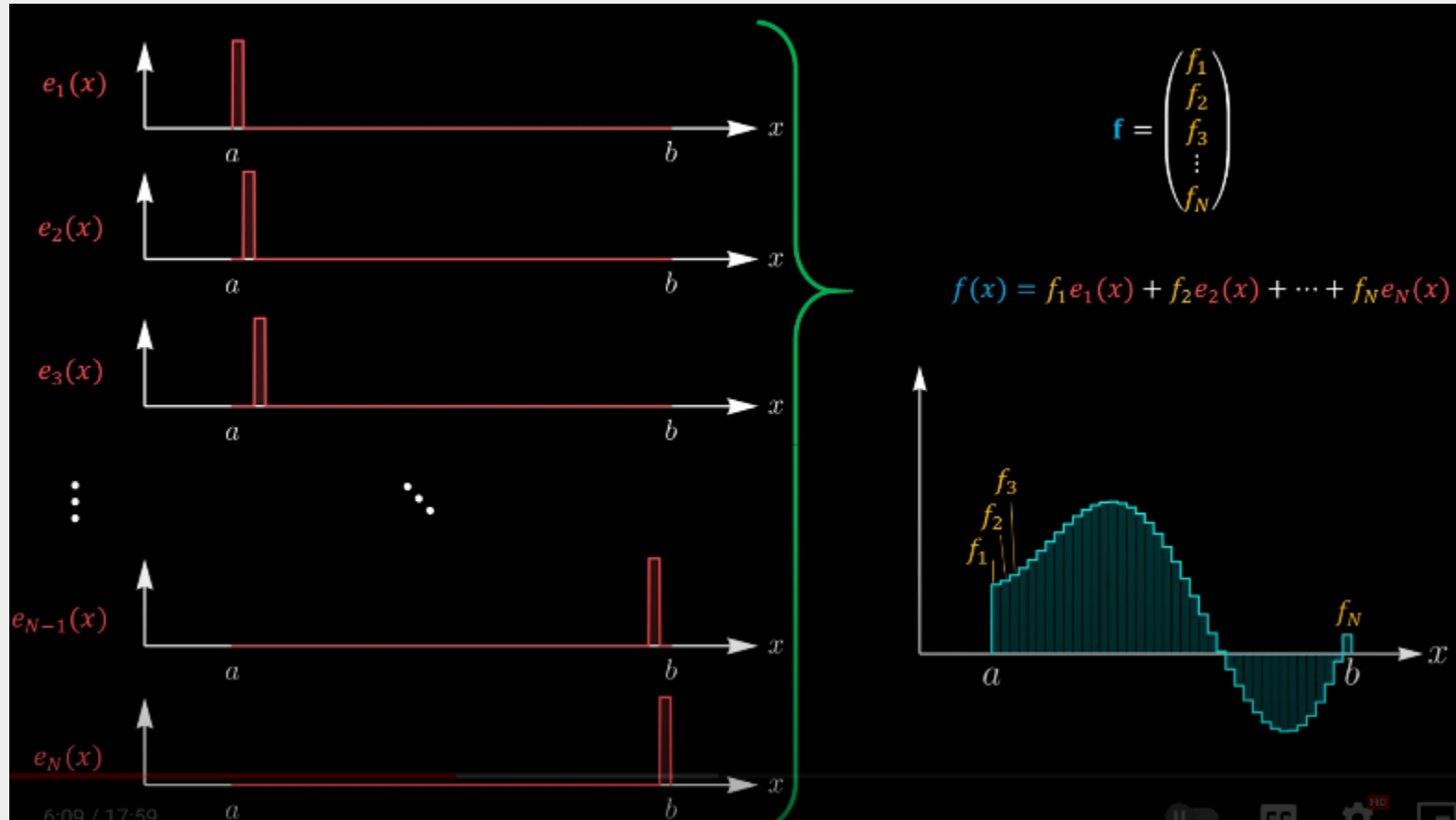


(a) $\text{Span}(\{\vec{A}, \vec{B}\}) = \text{Span}(\{\vec{A}, \vec{B}, \vec{C}\})$
 $\text{Span}(\{\vec{A}, \vec{B}, \vec{C}\}) = \text{All vectors on hyperplane}$



(b) $\text{Span}(\{\vec{A}, \vec{B}\}) \neq \text{Span}(\{\vec{A}, \vec{B}, \vec{C}\})$
 $\text{Span}(\{\vec{A}, \vec{B}, \vec{C}\}) = \text{All vectors in } \mathbb{R}^3$

Figure 2.6: The span of a set of linearly dependent vectors has lower dimension than the number of vectors in the set





Example

- ❑ Is vector b in $\text{Span} \{v_1, v_2, \dots, v_p\}$
- ❑ Is vector v_3 in $\text{Span} \{v_1, v_2, \dots, v_p\}$
- ❑ Is vector 0 in $\text{Span} \{v_1, v_2, \dots, v_p\}$
- ❑ Span of polynomials: $\{(1+x), (1-x), x^2\}$?
- ❑ Is b in $\text{Span} \{a_1, a_2\}$?

$$a_1 = \begin{bmatrix} 1 \\ -2 \\ 3 \end{bmatrix}, a_2 = \begin{bmatrix} 5 \\ -13 \\ -3 \end{bmatrix}, b = \begin{bmatrix} -3 \\ 8 \\ 1 \end{bmatrix}$$



- ❑ Vector–Vector Operations
- ❑ Binary operations
- ❑ Field
- ❑ Vector space
- ❑ Linear combination and introduction to affine combination
- ❑ Span of vectors (linear hull)



- ❑ LINEAR ALGEBRA: Theory, Intuition, Code
- ❑ LINEAR ALGEBRA, KENNETH HOFFMAN.
- ❑ LINEAR ALGEBRA, Jim Hefferon
- ❑ David C. Lay, Linear Algebra and Its Applications
- ❑ Online Courses!
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