

# Linear Algebra Basics

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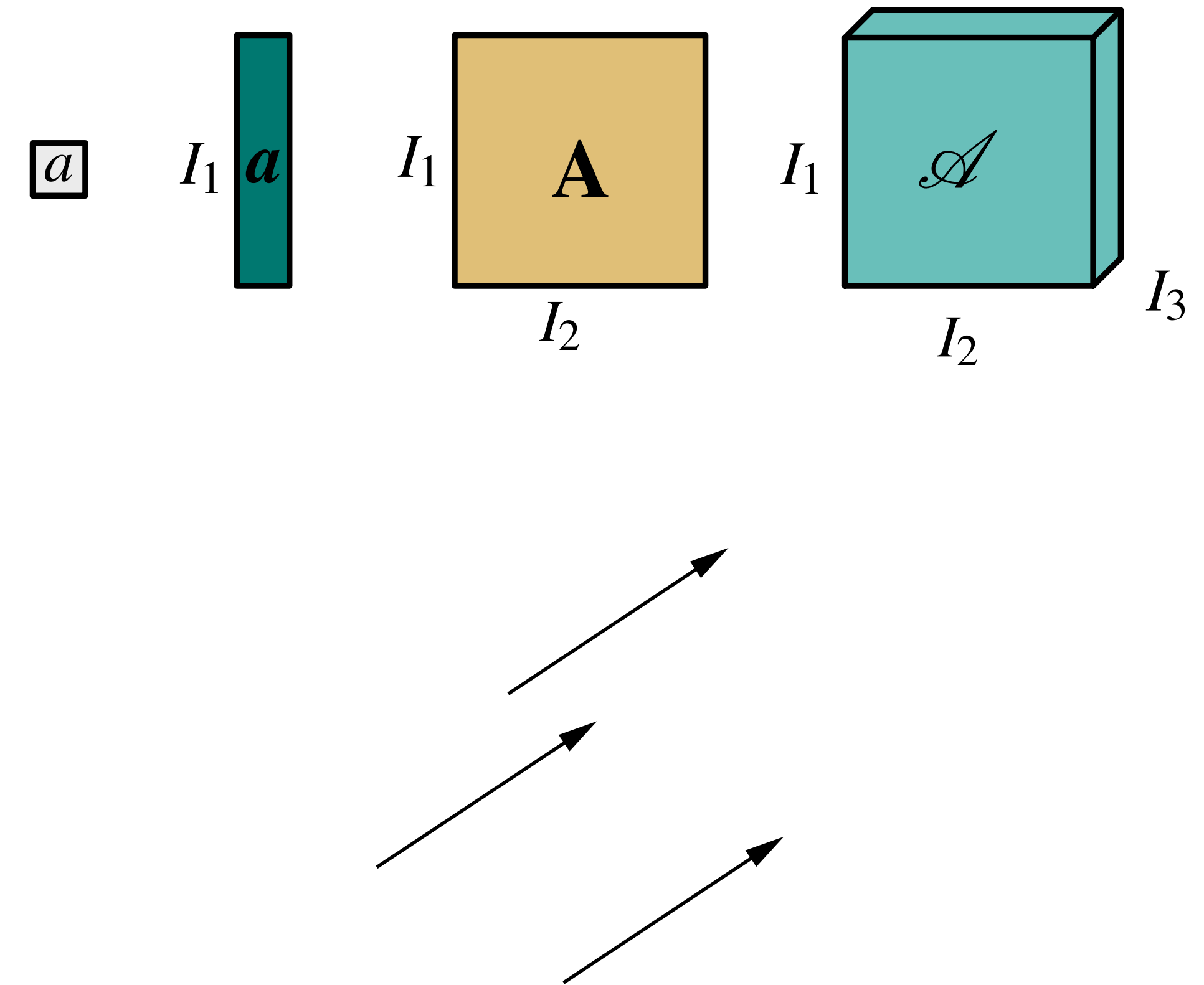
# Overview

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1. Scalar, points and vectors
2. Vector addition and subtraction
3. Vector space
4. Linear independence
5. Affine space
6. Dot and cross products
7. Matrices

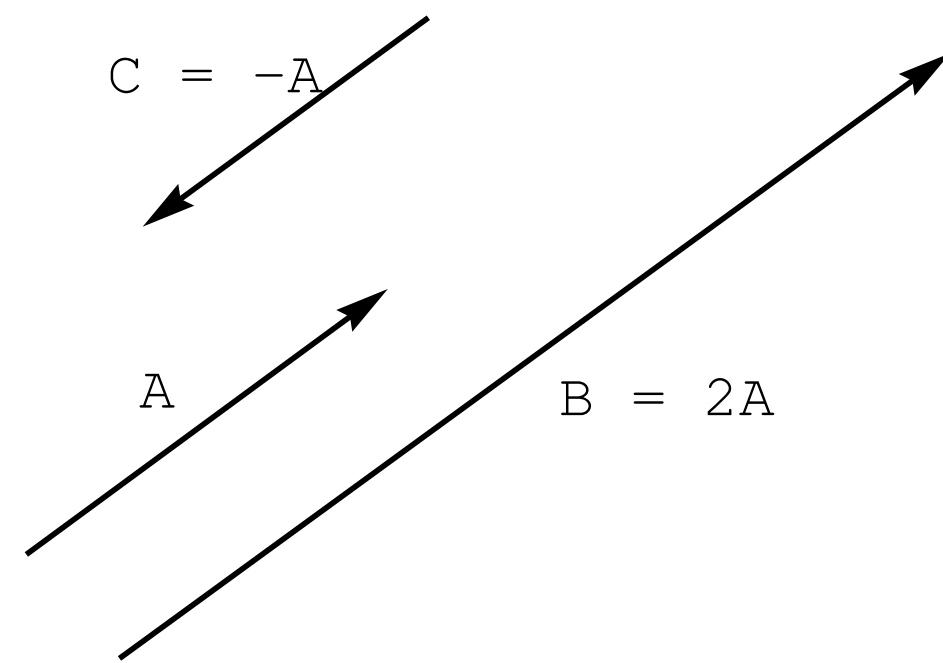
# Scalars, Points and Vectors

- Scalars, vectors, matrices and tensors
  - ▶ Individual numerical measures or numeric data elements can be combined into data records such as vectors and matrices
- Basic types denoting points and directions in space in  $N$ -dimensions
  - ▶  $N$ -tuple of values, common column vector math notation
    - vector  $\mathbf{a} = \begin{pmatrix} a_1 \\ \vdots \\ a_N \end{pmatrix}$  commonly stored as 1D array  $\mathbf{a} = [a_1, a_2, \dots, a_N]$
- A vector is a quantity with orientation and magnitude
  - ▶ High-dimensional data vector (e.g. velocity or force in physics)
- Multiway data arrays for 2D matrices or general  $d$   $ND$  data tensors

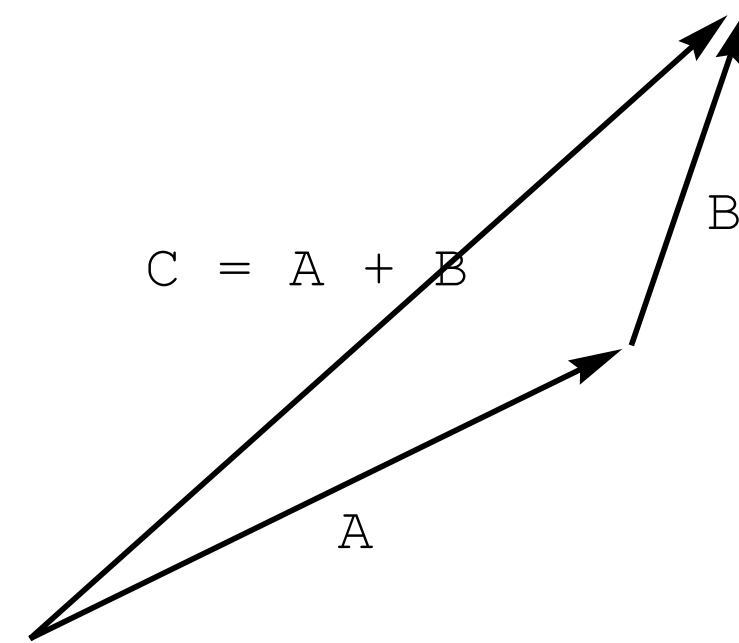


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# Vector Addition and Subtraction

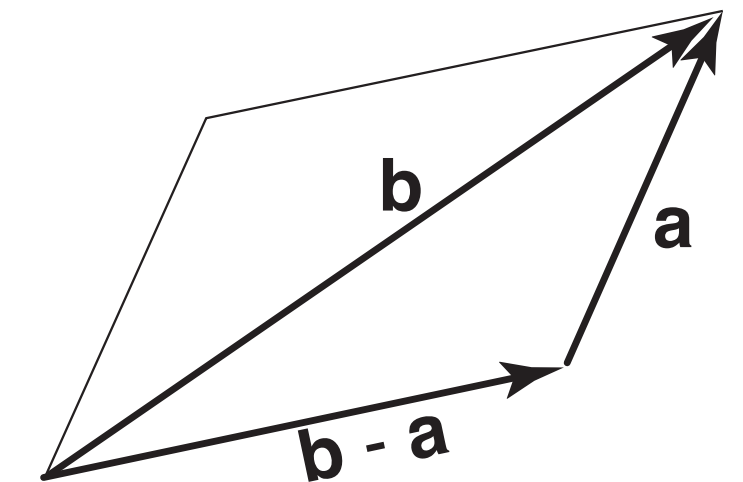
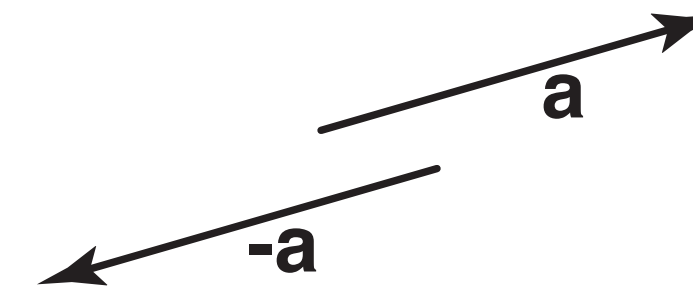
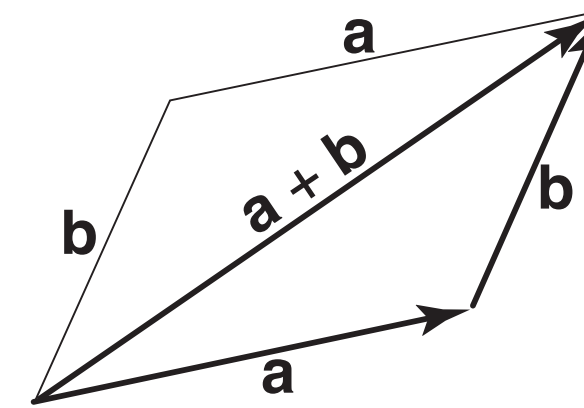


(a)



(b)

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- Vector addition is diagonal of parallelogram
  - ▶ *Head-to-tail* rule of placing vectors
- Scalar multiplication and component-wise vector addition
  - ▶ Scalar multiplication  $b = s \cdot a$  is equivalent to adding  $a$  together  $s$  times

$$a + b = \begin{pmatrix} a_1 + b_1 \\ \vdots \\ a_N + b_N \end{pmatrix}$$

$$b = s \cdot a = \begin{pmatrix} s \cdot a_1 \\ \vdots \\ s \cdot a_N \end{pmatrix}$$

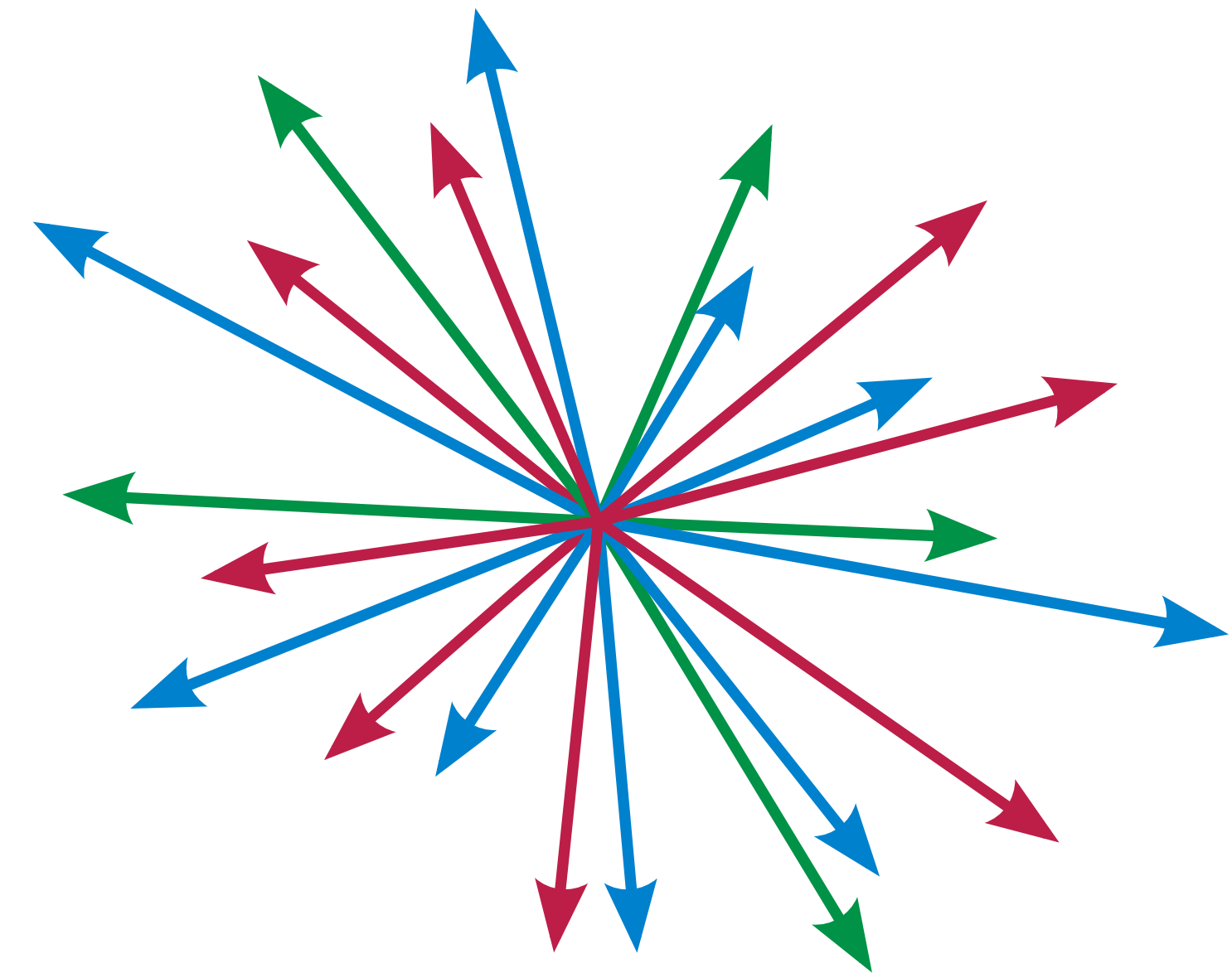
- Subtraction is equivalent to adding a negative vector
  - ▶ Vector negation is inversion of direction
- Vectors are equal if of same length and direction
  - ▶ Subtract to a zero-vector

$$a - b = \begin{pmatrix} a_1 - b_1 \\ \vdots \\ a_N - b_N \end{pmatrix} = a + (-1 \cdot b)$$

# Vector Space

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- In a vector space  $\mathbf{V}$ , addition and multiplication satisfy certain mathematical rules, such as being:
  - ▶ Closed:  $u+v \in \mathbf{V}, \forall u,v \in \mathbf{V}$
  - ▶ Commutative:  $u + v = v + u$
  - ▶ Associative:  $u + (v + w) = (u + v) + w$
  - ▶ Distributive:  $\alpha(u + v) = \alpha u + \alpha v$ ,  $(\alpha + \beta)u = \alpha u + \beta u$  for scalars  $\alpha, \beta$
  - ▶ Zero vector:  $u + 0 = u$
  - ▶ Additive inverse:  $u + (-u) = 0$



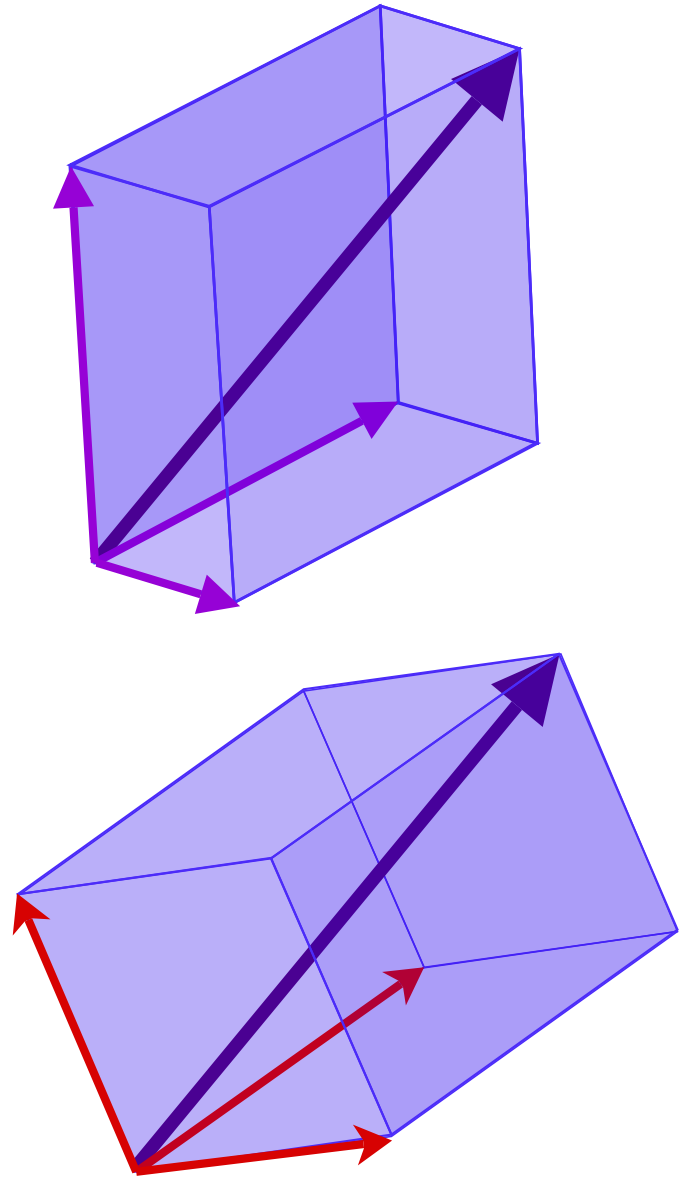
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- A vector space only contains vectors as elements!

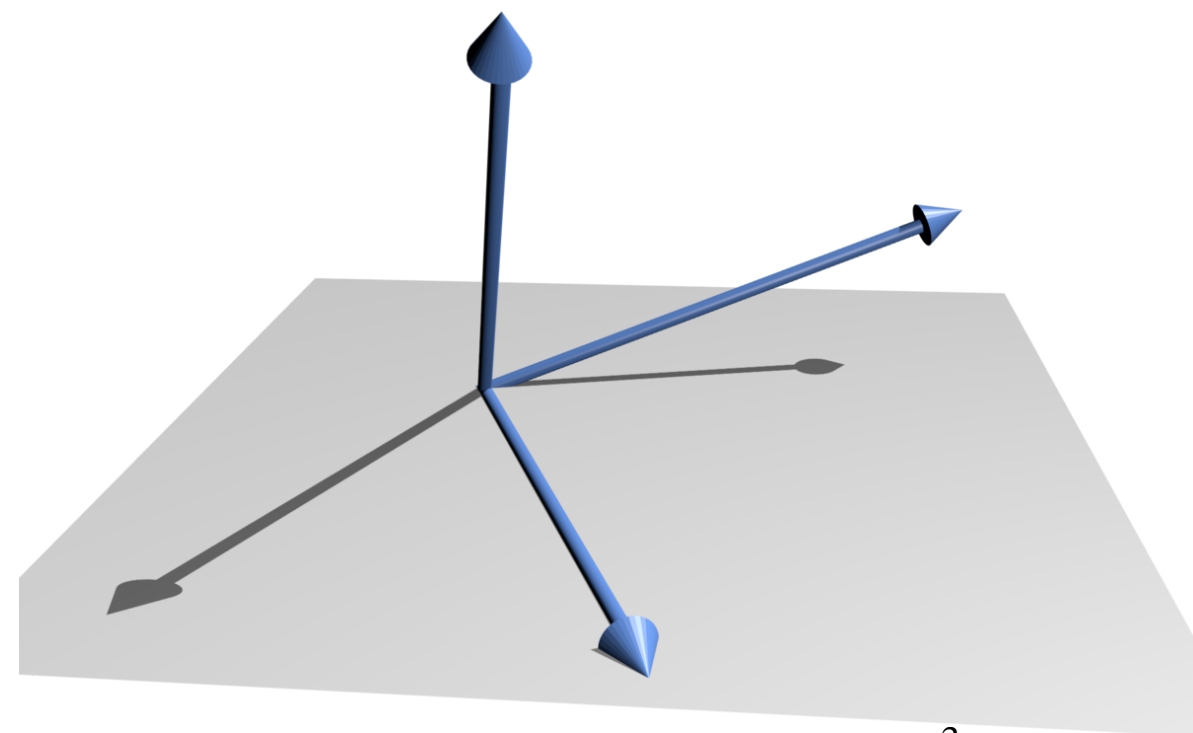


# Linear Independence

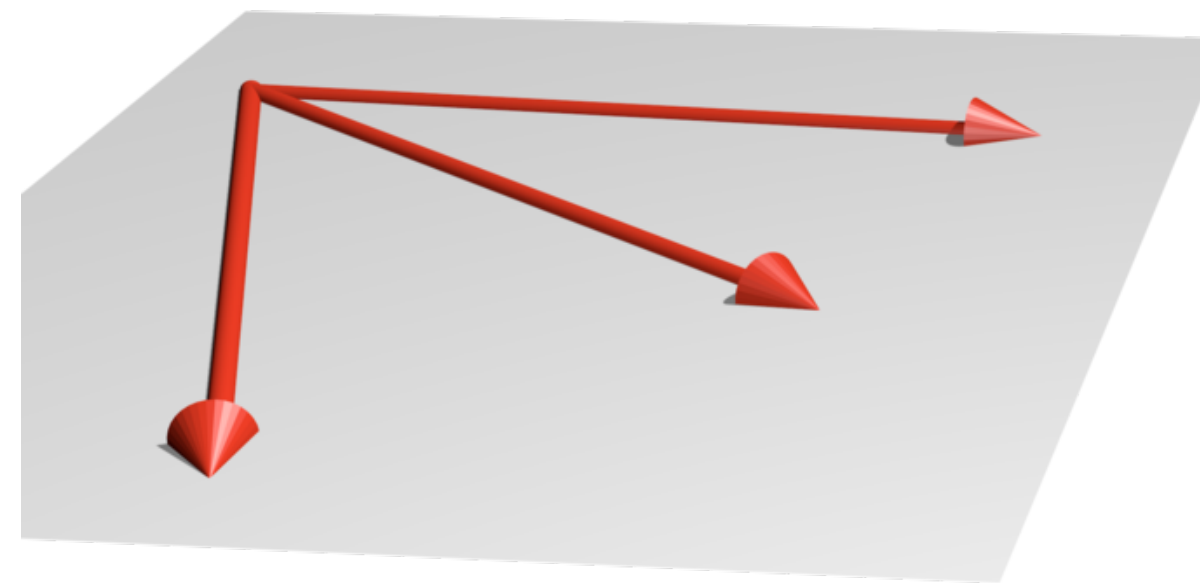
- A **linear combination** of  $N$  vectors  $\mathbf{v}_1 \dots \mathbf{v}_N$  is defined as  $\mathbf{u} = \alpha_1 \cdot \mathbf{v}_1 + \dots + \alpha_N \cdot \mathbf{v}_N$ 
  - ▶ Linearly independent **if**  $\alpha_1 \cdot \mathbf{v}_1 + \alpha_2 \cdot \mathbf{v}_2 \dots + \alpha_N \cdot \mathbf{v}_N = \mathbf{0}$  **only if all**  $\alpha_i = 0$
  - ▶ The dimension is defined by the largest number of linearly independent vectors
- $N$  linearly independent vectors  $\mathbf{v}_1 \dots \mathbf{v}_N$  form a **basis** of an  $N$ -dimensional vector space  $\mathbf{V}$ 
  - ▶  $\forall \mathbf{u} \in \mathbf{V}, \exists \beta_1 \dots \beta_N \in \mathbb{R}$  such that  $\mathbf{u} = \beta_1 \cdot \mathbf{v}_1 + \dots + \beta_N \cdot \mathbf{v}_N$



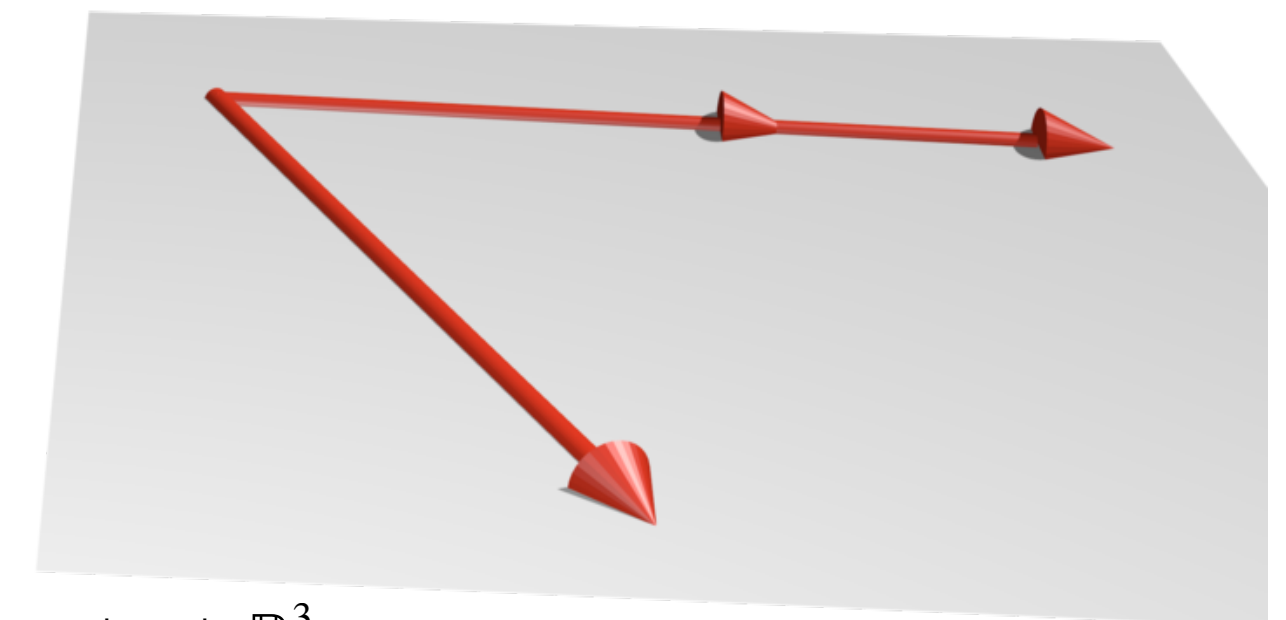
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linearly independent vectors in  $\mathbb{R}^3$

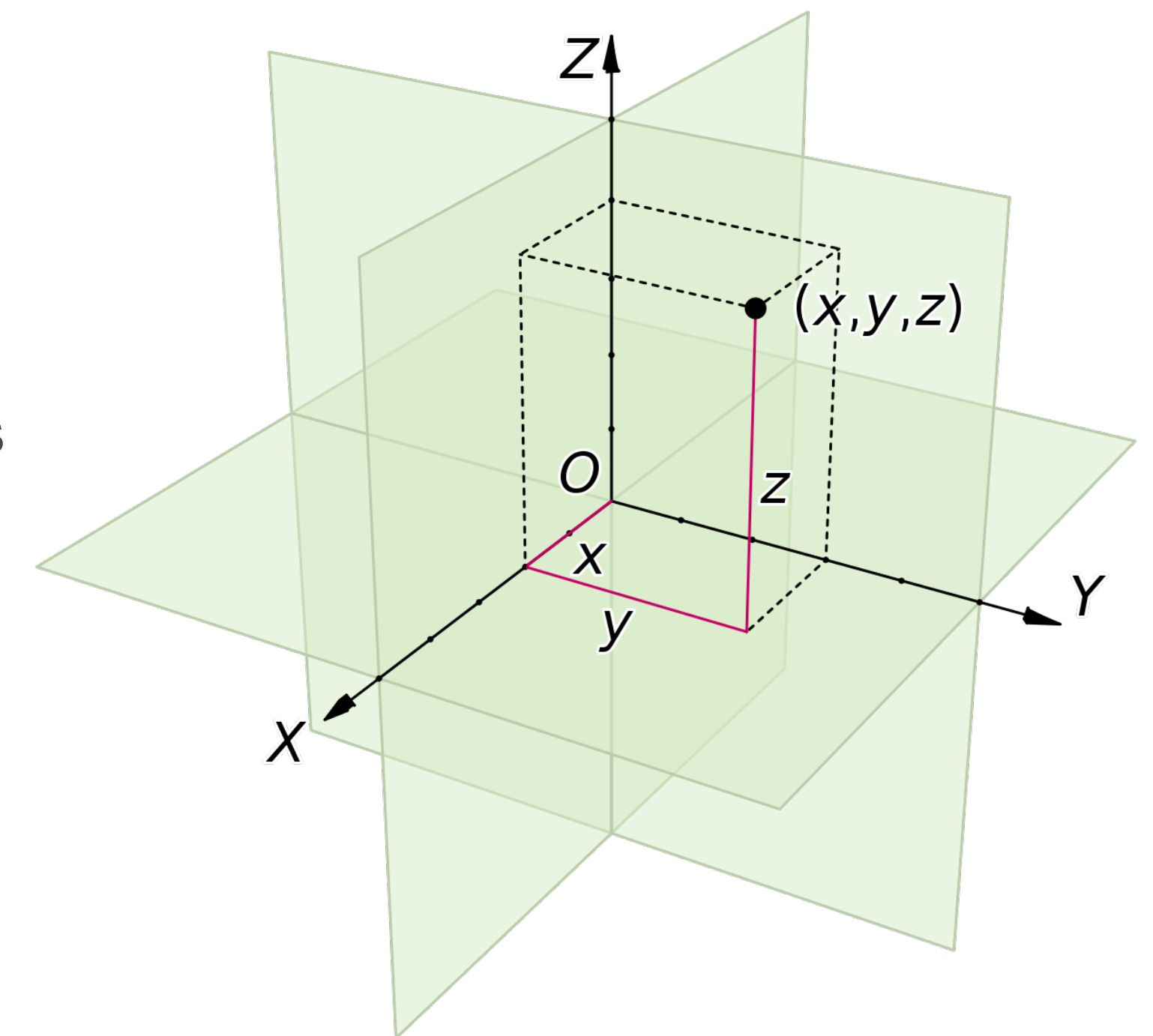


linearly dependent vectors in a plane in  $\mathbb{R}^3$



# Cartesian Coordinate System

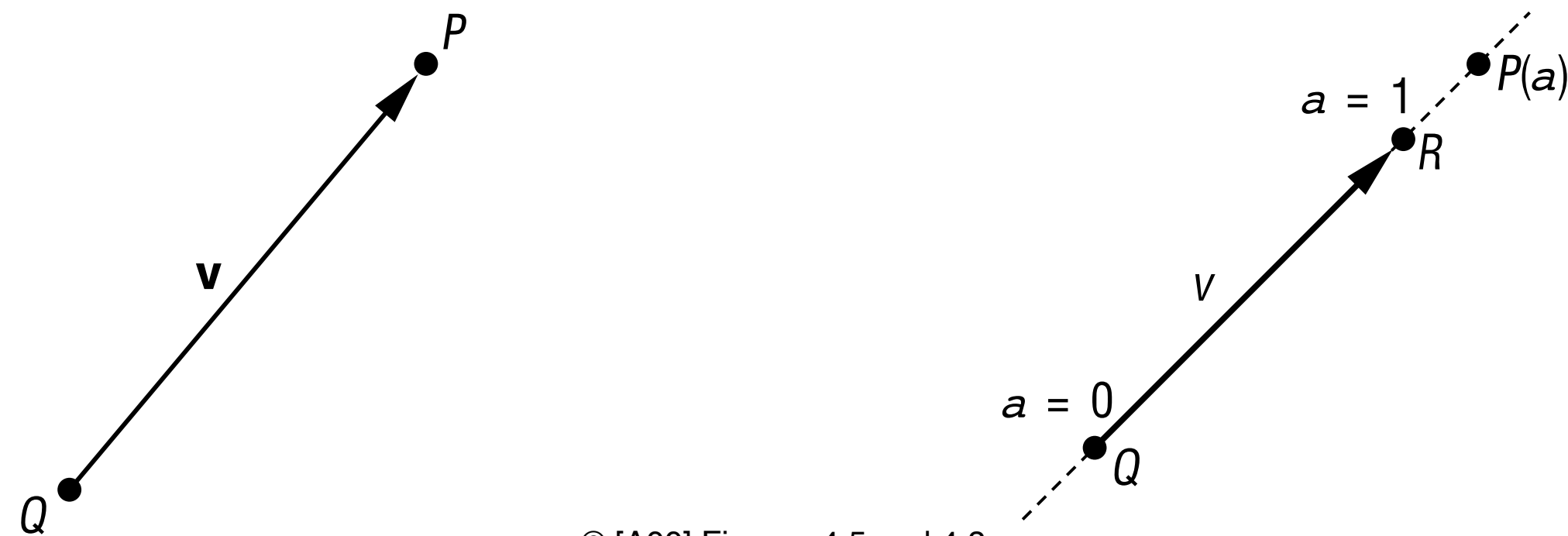
- A Cartesian coordinate system is defined by a set of orthonormal basis vectors
  - ▶ Orthonormal basis iff each vector is of unit length and is orthogonal to all others
- Given by the standard Cartesian  $N$ -dimensional basis vectors  $\mathbf{v}_{i=1,\dots,N} = (\delta_{i1}, \dots, \delta_{ij}, \dots, \delta_{iN})^T$  with components  $\delta_{ij}$ 
  - ▶ Kronecker delta  $\delta_{ij}$  is 0 for  $i \neq j$  and 1 for  $i = j$
  - ▶ For 3D we get the expected  $\mathbf{v}_1 = (1,0,0)$ ,  $\mathbf{v}_2 = (0,1,0)$  and  $\mathbf{v}_3 = (0,0,1)$



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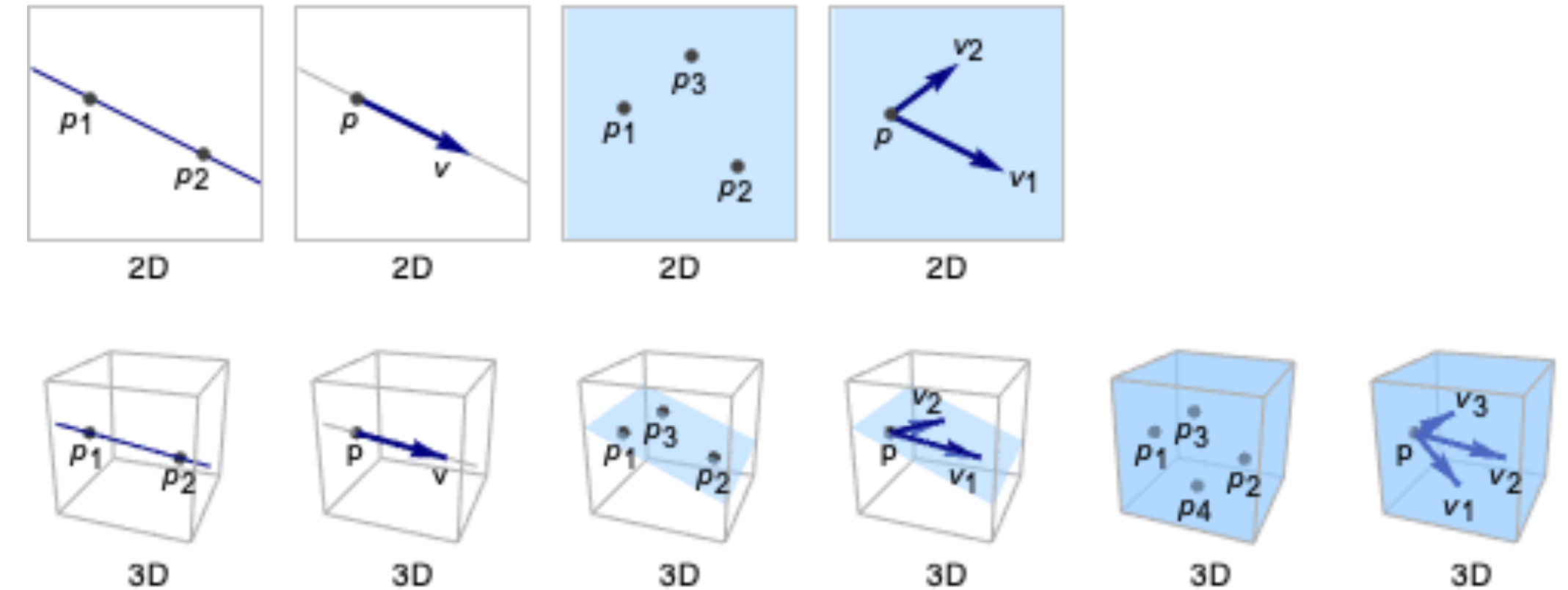


# Affine Space



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- An affine space contains points in addition to scalars and vectors
  - ▶ Points are locations and different from vectors
  - ▶ Operations: point-vector addition and point-point subtraction
    - add a vector to a point to get a new point
    - subtract two points to get the vector in between them
  - ▶ No addition of points or multiplication of points
    - cannot scale a point by scalar multiplication



Source: <https://reference.wolfram.com/language/ref/AffineSpace.html>

- The line  $L(t): P = Q + t \cdot (R - Q)$  is an affine space
  - ▶ Linear interpolation  $(1-t) \cdot Q + t \cdot R$  (line segment  $t \in [0,1]$ )
- A line, a plane or a volume in 2D or 3D space represents an affine (sub-) space

# The Dot-Product

- Given two  $N$ -D vectors  $\mathbf{u}$  and  $\mathbf{v}$  the inner- or dot-product ' $\circ$ ' is defined as  $\mathbf{u} \circ \mathbf{v} = u_1 v_1 + \dots + u_N v_N$

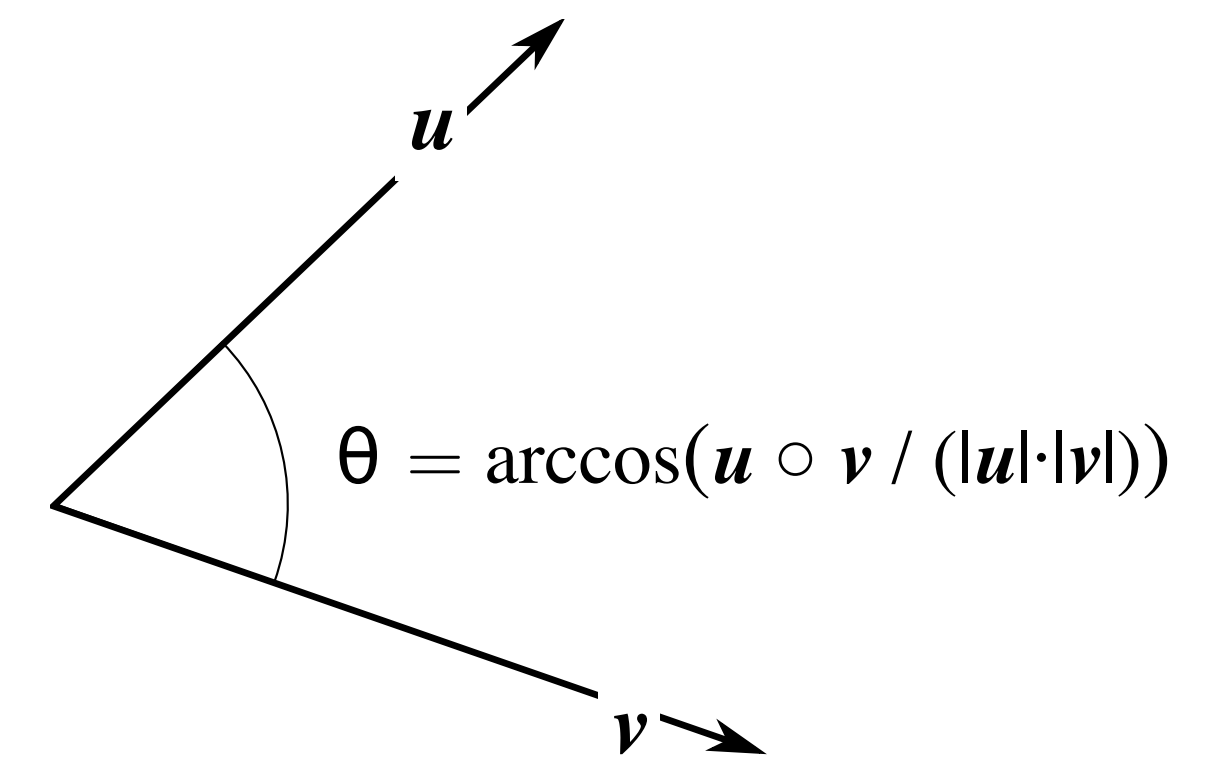
- ▶ Operation on two vectors with scalar result

- The dot-product is

- ▶ *Symmetric*:  $\mathbf{u} \circ \mathbf{v} = \mathbf{v} \circ \mathbf{u}$

- ▶ *Non-degenerate*:  $\mathbf{v} \circ \mathbf{v} = 0 \Leftrightarrow \mathbf{v} = \mathbf{0}$

- ▶ *Bilinear*:  $\mathbf{v} \circ (\mathbf{u} + \alpha \mathbf{w}) = \mathbf{v} \circ \mathbf{u} + \alpha (\mathbf{v} \circ \mathbf{w})$



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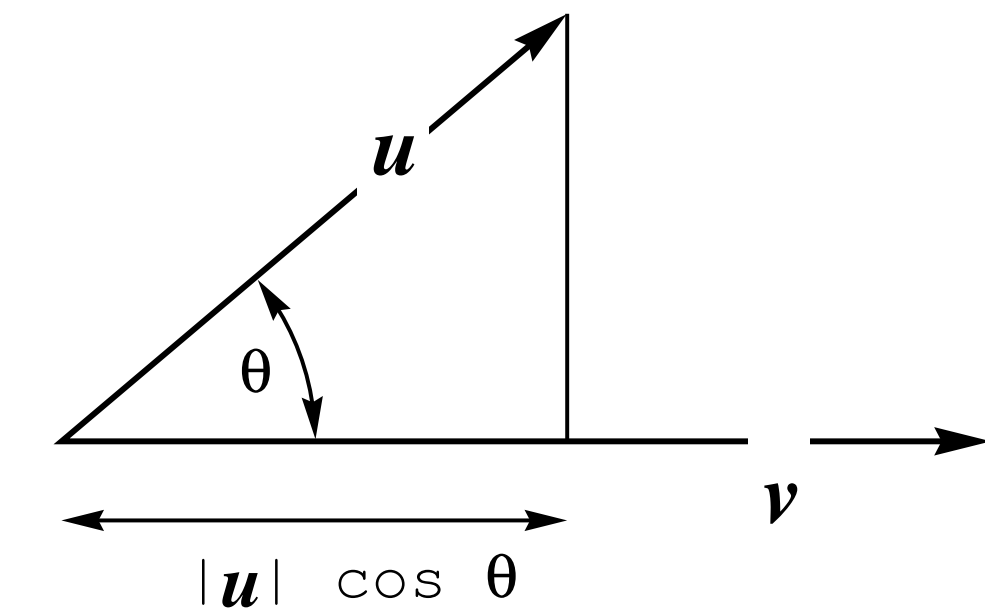
- Relation to angle between vectors:  $\mathbf{u} \circ \mathbf{v} = \cos \theta \cdot |\mathbf{u}| \cdot |\mathbf{v}|$

- ▶  $\mathbf{u}, \mathbf{v}$  point to opposite half-spaces if negative

- ▶ Vectors  $\mathbf{u}$  and  $\mathbf{v}$  are orthogonal if  $\mathbf{u} \circ \mathbf{v} = 0$

- Orthogonal projection of  $\mathbf{u}$  onto  $\mathbf{v}$  given by  $|\mathbf{u}| \cdot \cos \theta = \mathbf{u} \circ \mathbf{v} / |\mathbf{v}|$

- Squared length of vector:  $|\mathbf{u}|^2 = \mathbf{u} \circ \mathbf{u}$



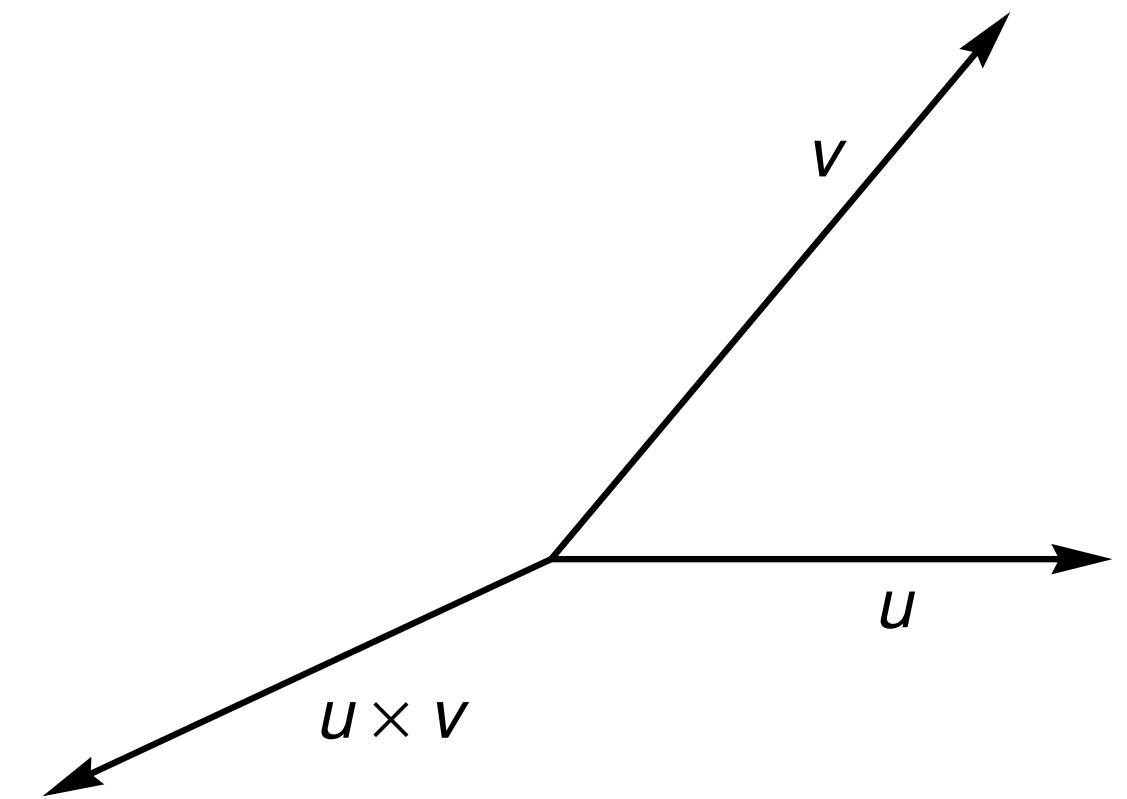
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# The Cross-Product

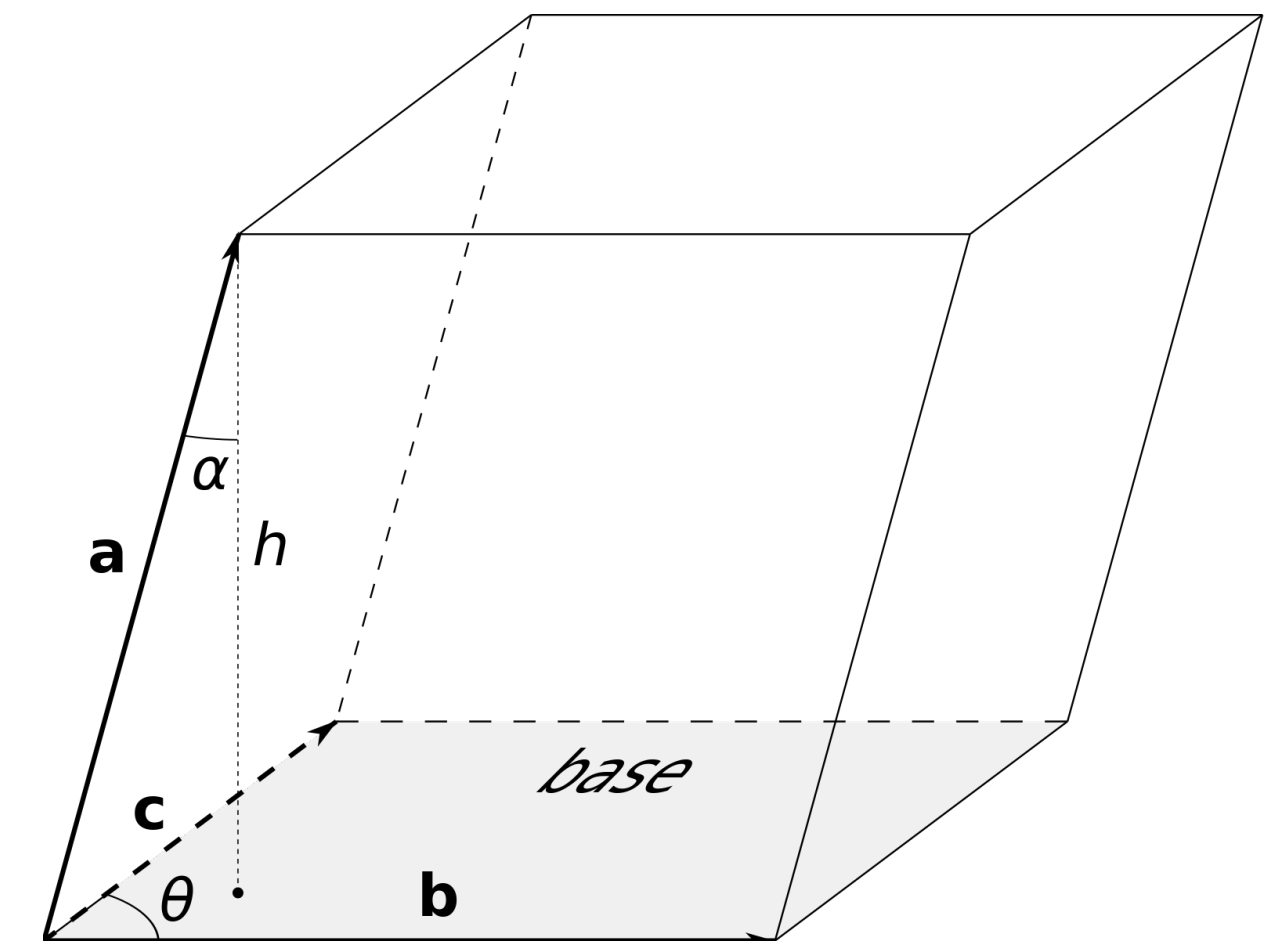
- Given two  $n$ -D vectors  $\mathbf{u}$  and  $\mathbf{v}$  the cross-product  $\mathbf{w} = \mathbf{u} \times \mathbf{v}$  is defined as

$$\begin{pmatrix} w_1 \\ w_2 \\ w_3 \end{pmatrix} = \begin{pmatrix} u_2 v_3 - u_3 v_2 \\ u_3 v_1 - u_1 v_3 \\ u_1 v_2 - u_2 v_1 \end{pmatrix}$$

- The cross-product  $\mathbf{w} = \mathbf{u} \times \mathbf{v}$  is perpendicular to  $\mathbf{u}$  and  $\mathbf{v}$ 
  - Defines right-handed vector system
- Relation to angle between vectors:  $|\mathbf{u} \times \mathbf{v}| = |\sin \theta| \cdot |\mathbf{u}| \cdot |\mathbf{v}|$
- Vectors  $\mathbf{u}$  and  $\mathbf{v}$  are collinear if  $\mathbf{u} \times \mathbf{v} = \mathbf{0}$
- Length  $|\mathbf{u} \times \mathbf{v}|$  is area of parallelogram spanned by  $\mathbf{u}$  and  $\mathbf{v}$



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# Matrices

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- A matrix is a two-dimensional array of numbers
  - ▶  $M \times N$  grid of values organized in  $M$  rows and  $N$  columns

- matrix  $\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & \dots & \dots & a_{2N} \\ & & \vdots & \\ a_{M1} & a_{M2} & \dots & a_{MN} \end{pmatrix}$  commonly stored as 2D array  $A = \begin{bmatrix} [a_{11}, a_{12}, \dots, a_{1N}], \dots, [a_{M1}, a_{M2}, \dots, a_{MN}] \end{bmatrix}$

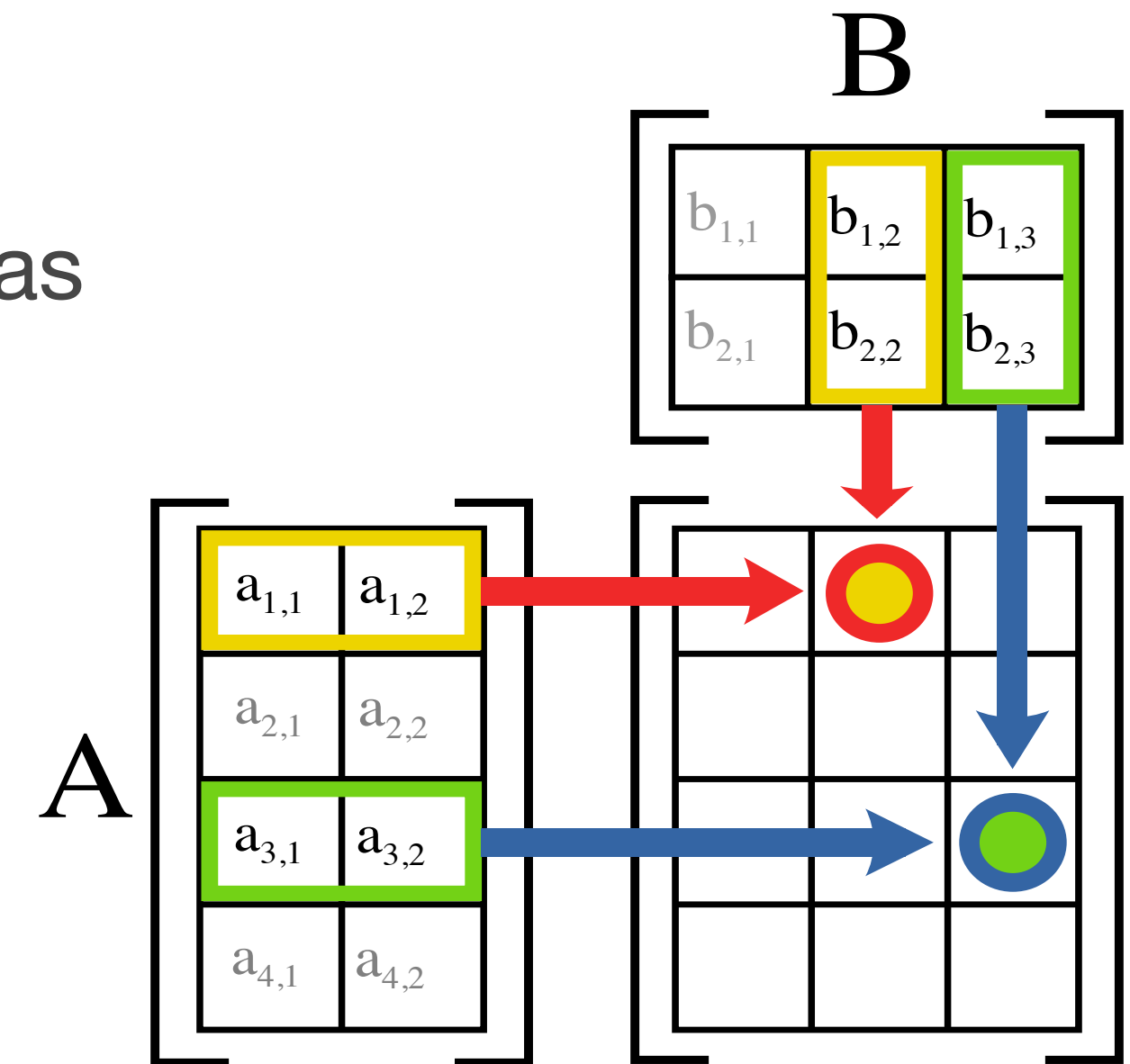
- Special *square*; *diagonal*; *identity*; and *zero* or *null* matrices
  - ▶ Having  $M=N$ ; all  $a_{ij} = 0$  for  $i \neq j$ ; diagonal and all  $a_{ii} = 1$ ; all entries 0
- Two matrices  $\mathbf{A}$  and  $\mathbf{B}$  are equal iff  $a_{ij} = b_{ij}$  for all  $i, j$
- The sum of two matrices  $\mathbf{A}$  and  $\mathbf{B}$  is  $c_{ij} = a_{ij} + b_{ij}$ 
  - ▶ Analogous for the subtraction/difference
- The scalar multiple  $\mathbf{C} = s \cdot \mathbf{A}$  is  $c_{ij} = s \cdot a_{ij}$

# Matrix Multiplication and Inverse

- Is defined component-wise on its elements
- For an  $M \times N$  matrix  $\mathbf{A}$  and  $N \times P$  matrix  $\mathbf{B}$  we get an  $M \times P$  matrix  $\mathbf{C} = \mathbf{AB}$  as

$$c_{ij} = \sum_{s=1}^M a_{is} \cdot b_{sj}$$

- ▶ Element  $ij$  of  $\mathbf{C}$  is dot-product of  $j$ -th column vector from  $\mathbf{B}$  with the  $i$ -th row vector from  $\mathbf{A}$
- ▶ Matrix multiplication is not commutative  $\mathbf{AB} \neq \mathbf{BA}$
- ▶ Matrix multiplication is distributive  $\mathbf{A}(\mathbf{B}+\mathbf{C}) = \mathbf{AB} + \mathbf{AC}$
- The inverse  $\mathbf{A}^{-1}$  has the property that  $\mathbf{A}^{-1} \mathbf{A} = \mathbf{A} \mathbf{A}^{-1} = \mathbf{I}$ 
  - ▶  $\mathbf{I}$  being the identity matrix
  - ▶ Can be found by Gaussian elimination in general
  - ▶ Inverse only exists for square matrices with non-zero determinant



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# Transpose

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- Transposition  $\mathbf{B} = \mathbf{A}^T$  is achieved by mirroring entries about the upper-left to lower-right diagonal

- ▶ Row/column indices and dimensions are swapped  $b_{ij} = a_{ji}$

- Matrix  $\mathbf{A}$  is said to be symmetric if  $\mathbf{A} = \mathbf{A}^T$  is

- An  $N$ -dimensional vector  $\mathbf{a}$  can be considered an  $N \times 1$  matrix

- ▶ Called a column vector  $\mathbf{a} = \begin{pmatrix} a_1 \\ \vdots \\ a_N \end{pmatrix}$

- ▶ Its transpose is then a row vector  $\mathbf{a}^T = (a_1 \ a_2 \ \dots \ a_N)$

- Matrix notation of dot-product between vectors  $\mathbf{a}$  and  $\mathbf{b}$

$$\mathbf{a} \circ \mathbf{b} = \mathbf{a}^T \cdot \mathbf{b} = (a_1 \ \dots \ a_N) \cdot \begin{pmatrix} b_1 \\ \vdots \\ b_N \end{pmatrix}$$

**A**

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix}$$



# Matrix-Vector Multiplication and Outer Vector-Product

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- Matrix-vector multiplication  $\mathbf{A}\mathbf{v}$  corresponds to matrix product rule

$$\mathbf{u} = \mathbf{A} \cdot \mathbf{v} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ & & \ddots & \\ a_{M1} & a_{M2} & \dots & a_{MN} \end{pmatrix} \cdot \begin{pmatrix} v_1 \\ \vdots \\ v_N \end{pmatrix}$$

- ▶ Column vector  $\mathbf{v}$  is multiplied (as dot-product) with each row of  $\mathbf{A}$ 
  - number of columns of  $\mathbf{A}$  must match rows of  $\mathbf{v}$
- ▶ Result vector  $\mathbf{u}$  has  $M$  entries (not  $N$  !)
  - same as number of rows of  $\mathbf{A}$
- Recall dot-product being:  $\mathbf{u}^T \cdot \mathbf{v}$
- Outer product  $\mathbf{u} \cdot \mathbf{v}^T$  of two vectors forms a new matrix

$$\mathbf{u} \cdot \mathbf{v}^T = \begin{pmatrix} u_1 \\ \vdots \\ u_N \end{pmatrix} \cdot (v_1 \ \dots \ v_N) = \begin{pmatrix} u_1 v_1 & u_1 v_2 & \dots & u_1 v_N \\ & \vdots & & \\ u_M v_1 & u_M v_2 & \dots & u_M v_N \end{pmatrix}$$

# Recap

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- **Scalar points and vectors:** operations on vectors, vector and affine spaces
- **Linear combination:** linear independence, coordinate systems
- **Dot/Cross products:** relation to angles between vectors and areas spanned by vectors
- **Matrices:** properties and operations, inverse and transpose, matrix-vector products, dot-product as matrix product, outer vector product
- Required textbook Chapter(s): 3 & 4