

- Exam – when?
- January 7?

### Academic year 2015/2016

**Autumn term: August 31, 2015 – January 18, 2016**

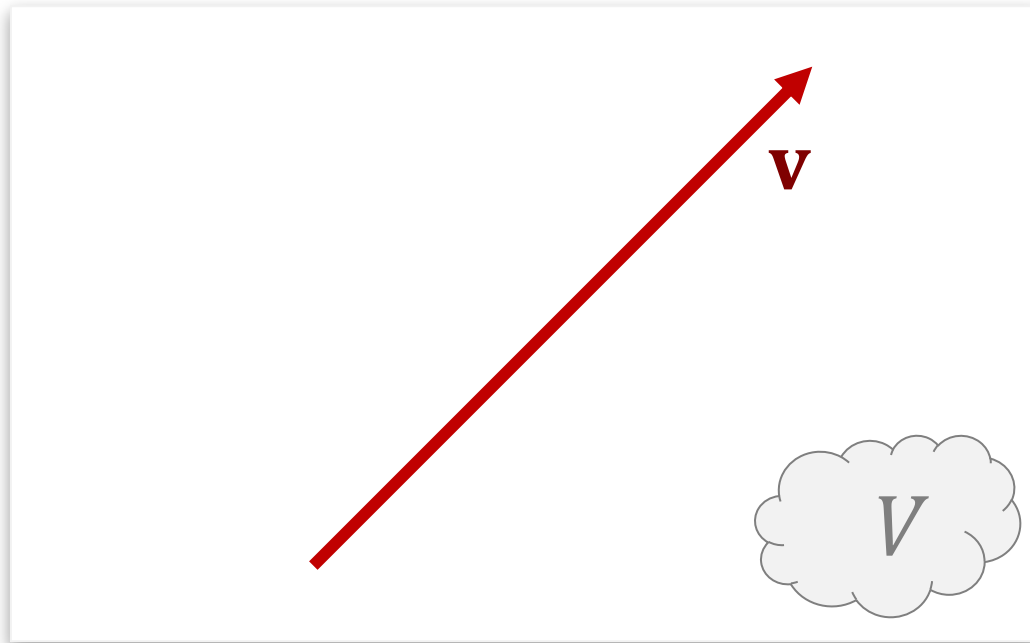
- Study period 1: 15-08-31 – 15-10-16
- Own work: 15-10-19 – 15-10-22
- Exam period 1: 15-10-23 – 15-10-30
- Period 2: 15-11-02 – 15-12-18
- Own work: 15-12-21 – 16-01-05
- Own work/Re-exam 1: 16-01-07 – 16-01-09
- Exam period 2: 16-01-11 – 16-01-18



*Introduction to Visualization and Computer Graphics*  
*DH2320, Fall 2015*  
*Prof. Dr. Tino Weinkauff*

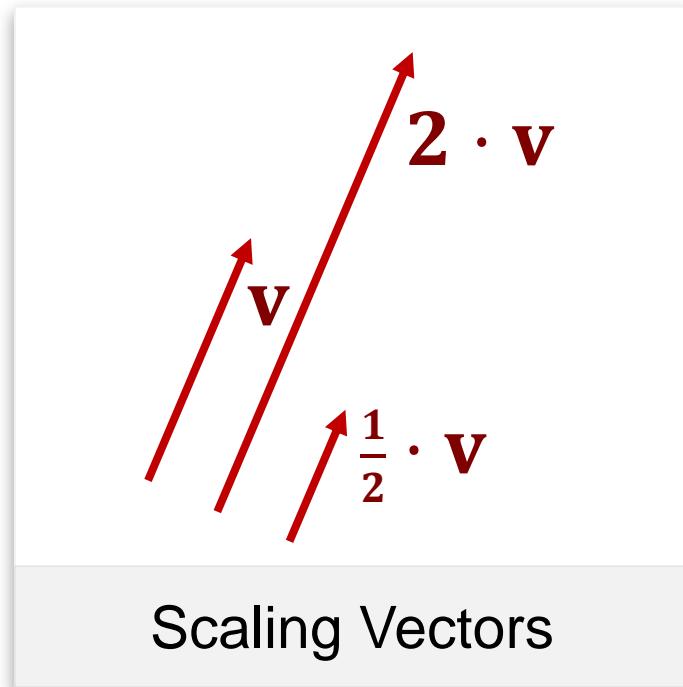
## ***Mathematics***

Vectors and Points

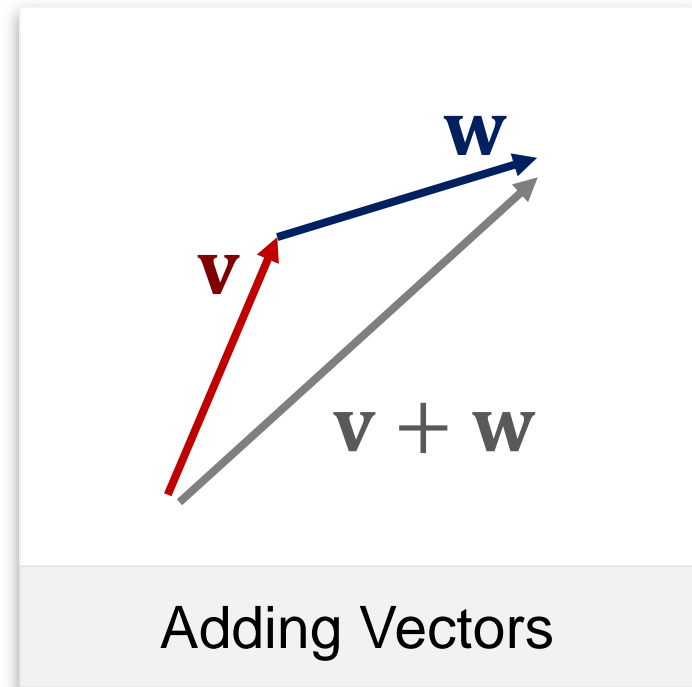


- Vectors
  - A vector is an arrow in space
  - We use bold letters for vectors: **u, v, w, x, y, z** ...
  - Vector space  $V$ : set of possible vectors

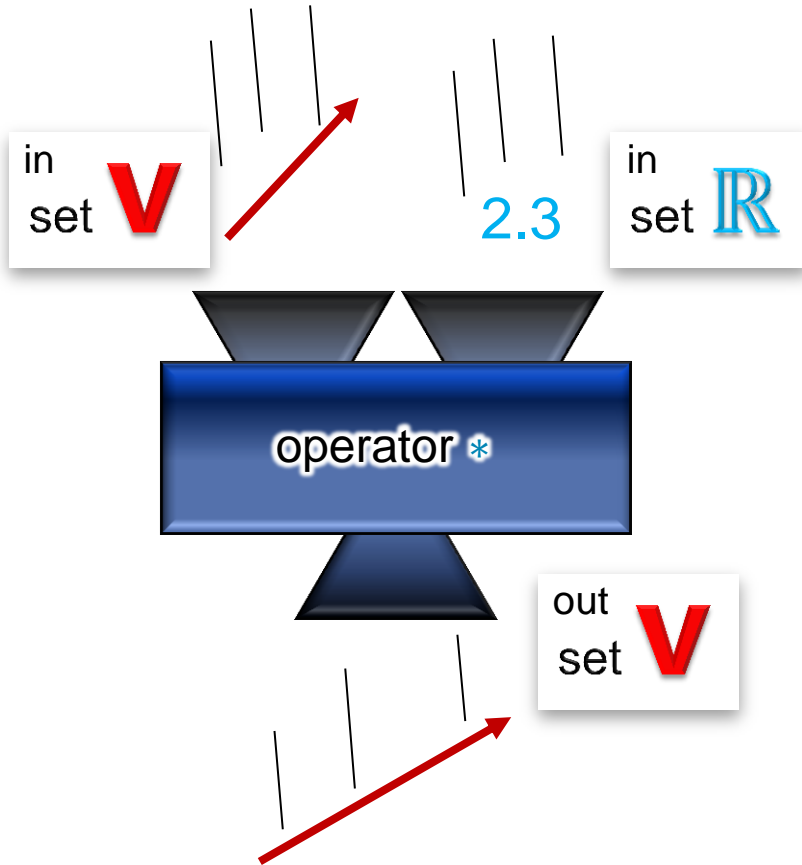
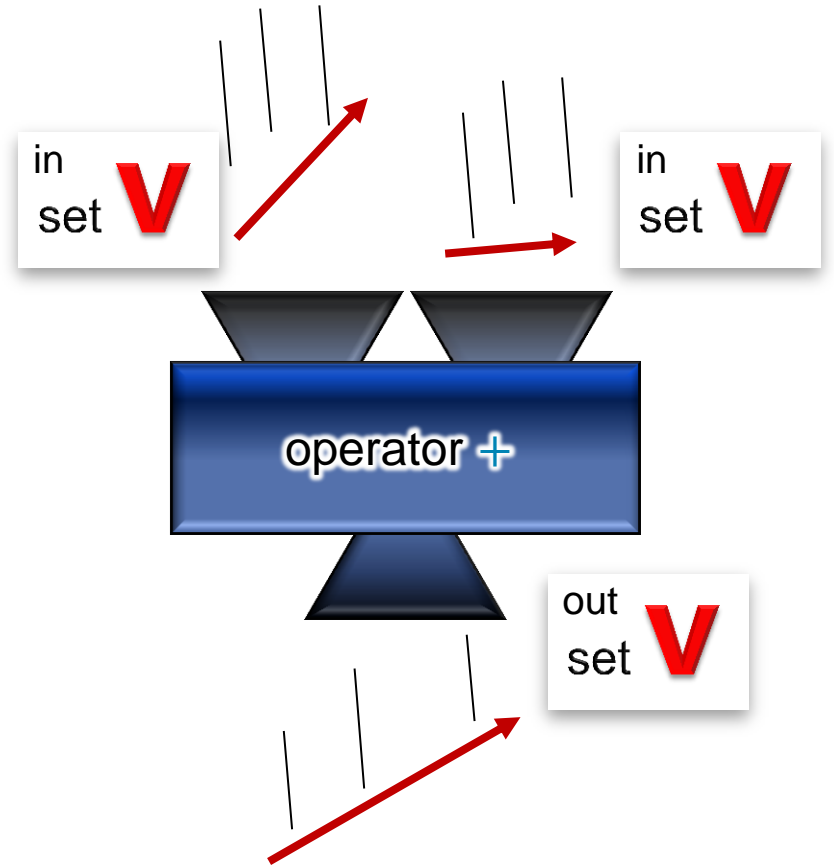
# ***Vector Operations***

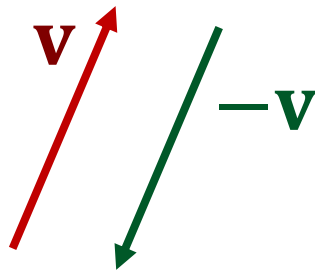


vector-scalar product  
 $\lambda \cdot \mathbf{v}$  ( $\lambda \in \mathbb{R}$ ,  $\mathbf{v} \in V$ )



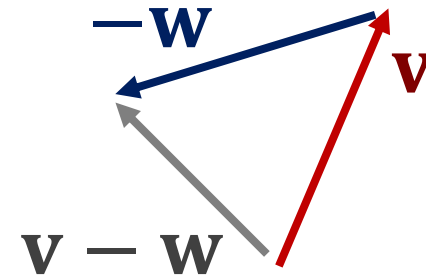
vector-addition  
 $\mathbf{v} + \mathbf{w}$  ( $\mathbf{v}, \mathbf{w} \in V$ )

**Vector-Scalar Multiplication****Vector Addition**



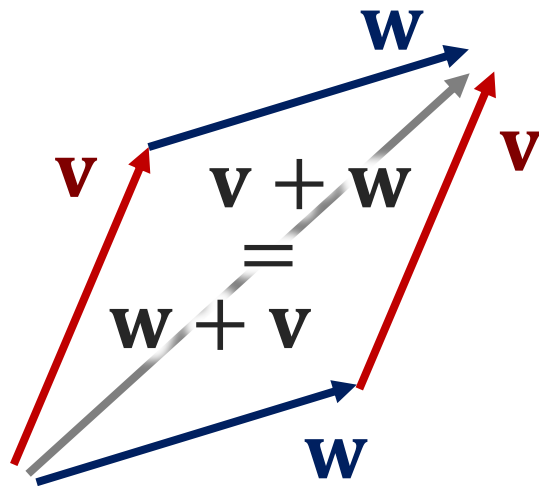
Reversing Vectors

\*) special case of scaling



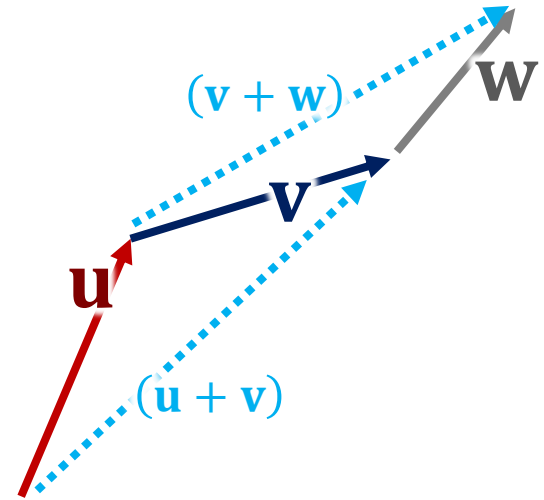
Subtracting Vectors

\*) special case of addition



Commutative

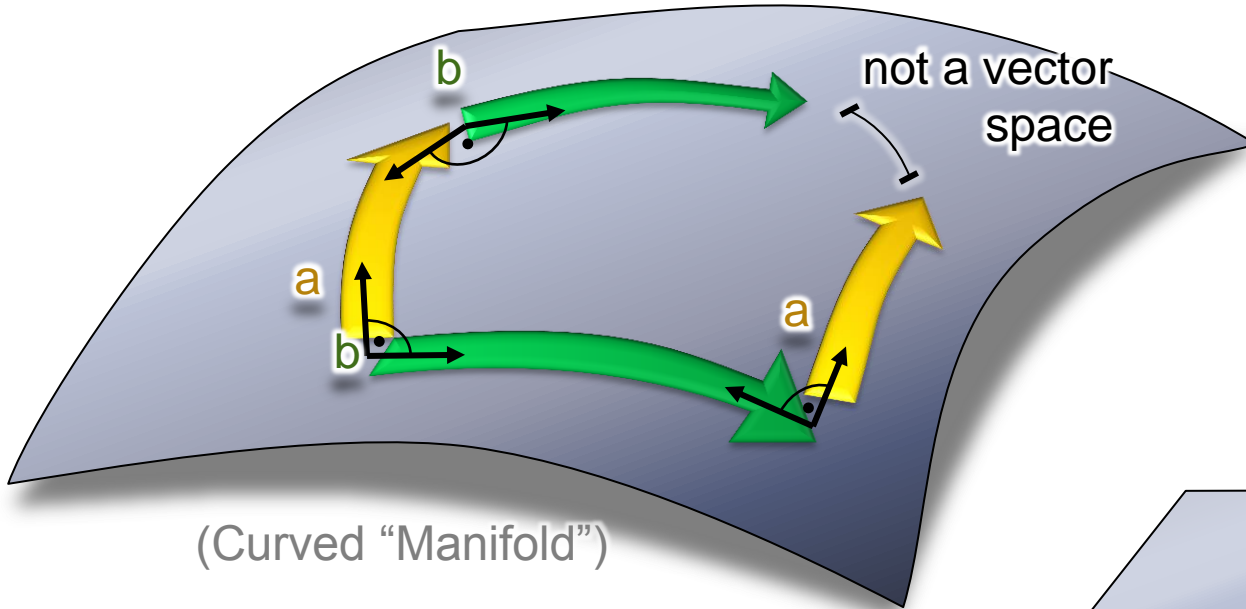
$$\mathbf{v} + \mathbf{w} = \mathbf{w} + \mathbf{v}$$



Associative

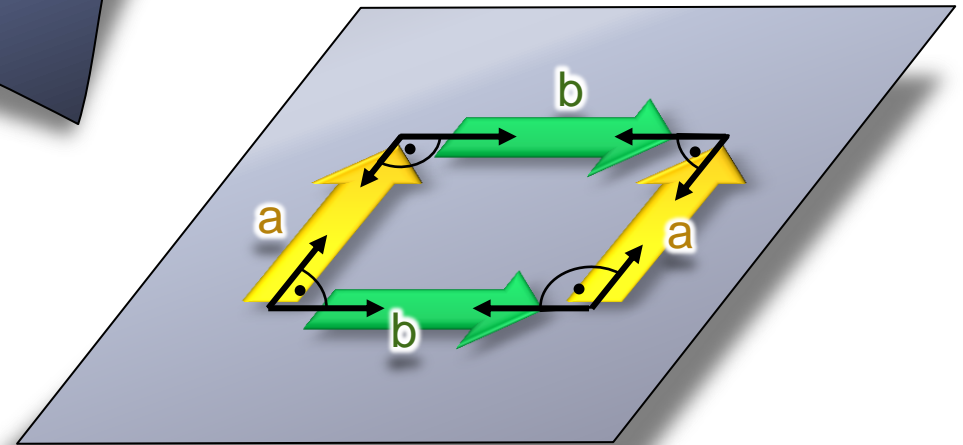
$$\mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}$$



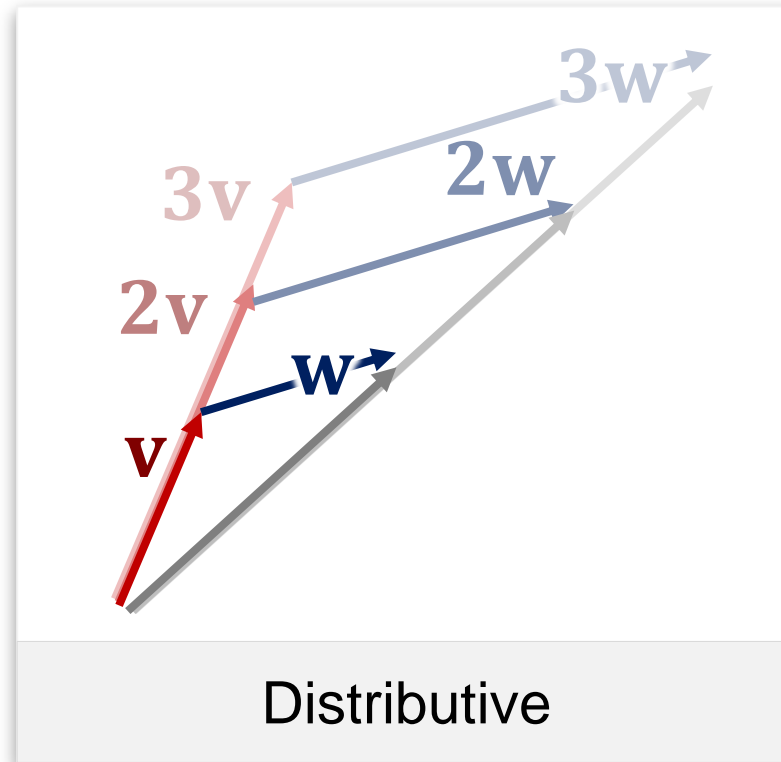


(Curved "Manifold")

\*) associativity still holds

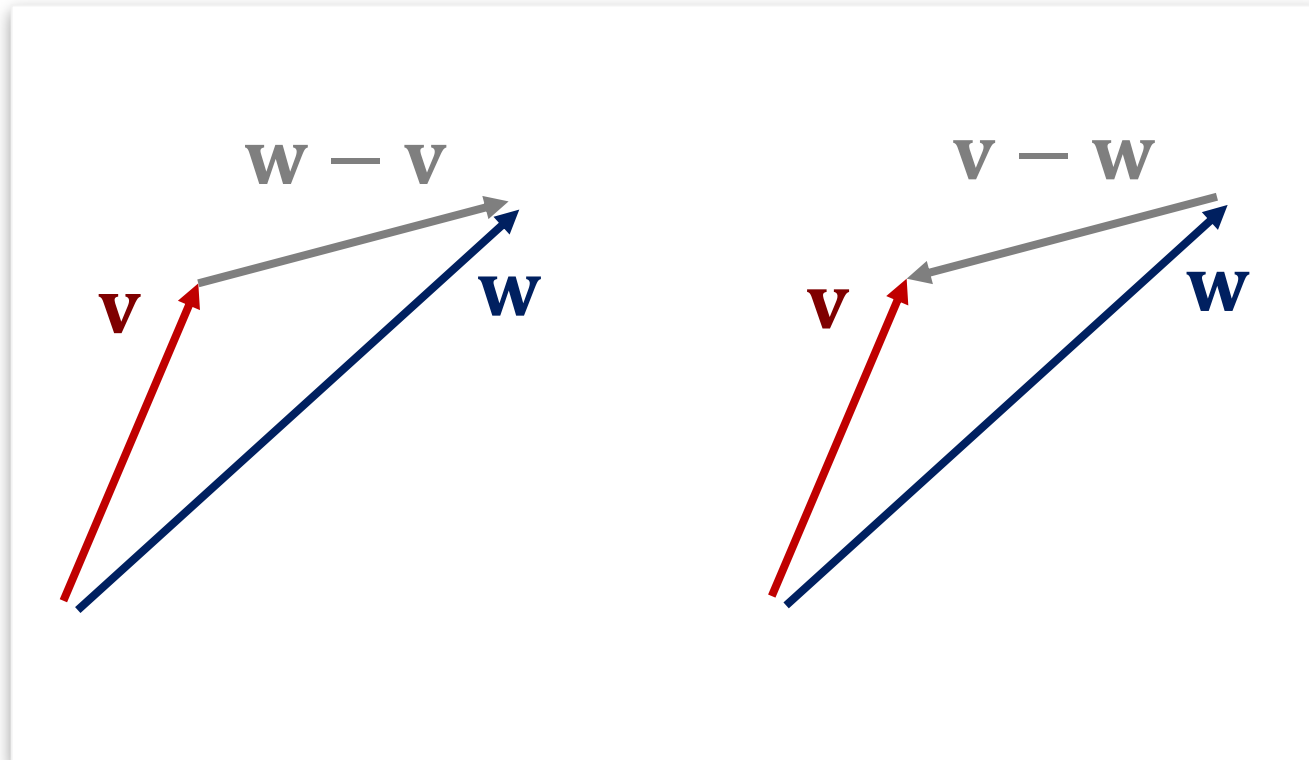


vector space  
(Euclidean geometry)

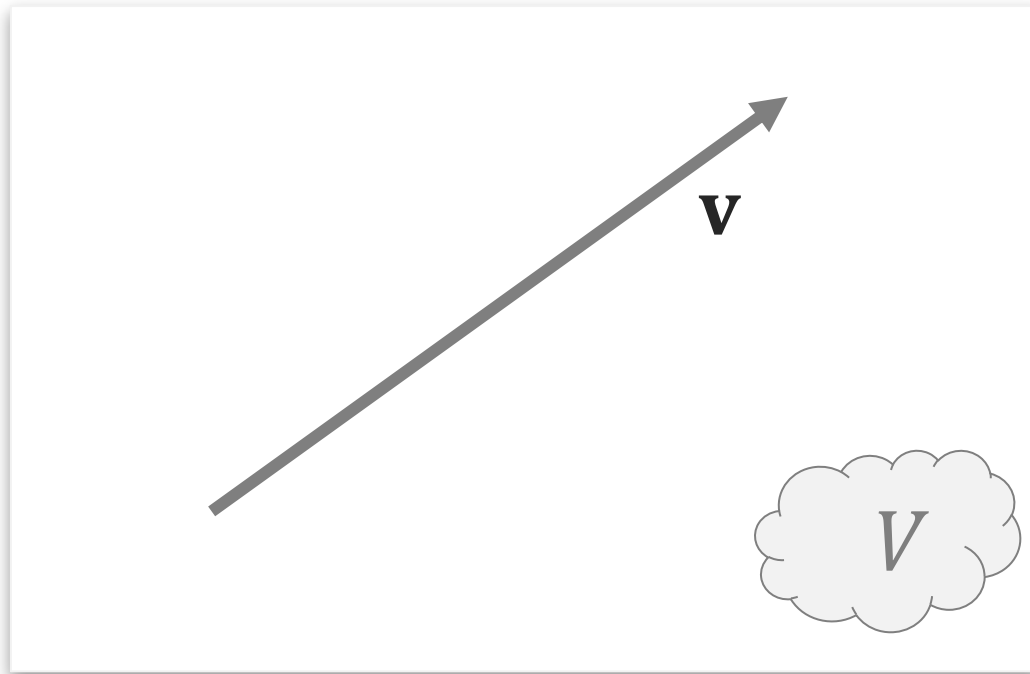


$$\lambda(\mathbf{v} + \mathbf{w}) = \lambda\mathbf{v} + \lambda\mathbf{w}$$

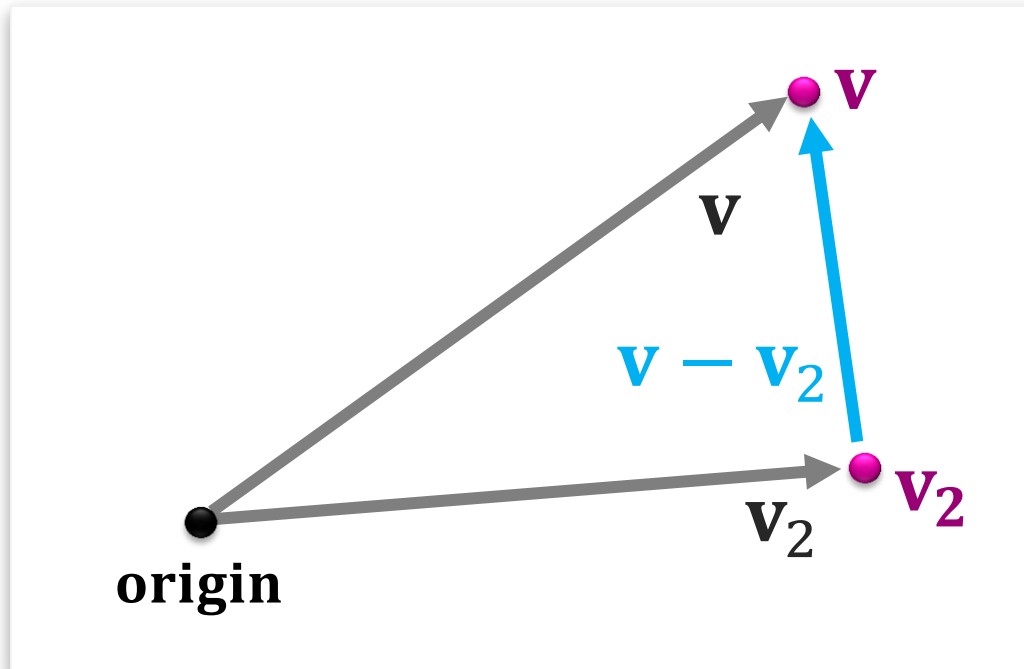
$$\lambda(\mu\mathbf{v}) = \lambda\mu(\mathbf{v})$$



$$\mathbf{v} - \mathbf{w} = -(\mathbf{w} - \mathbf{v})$$

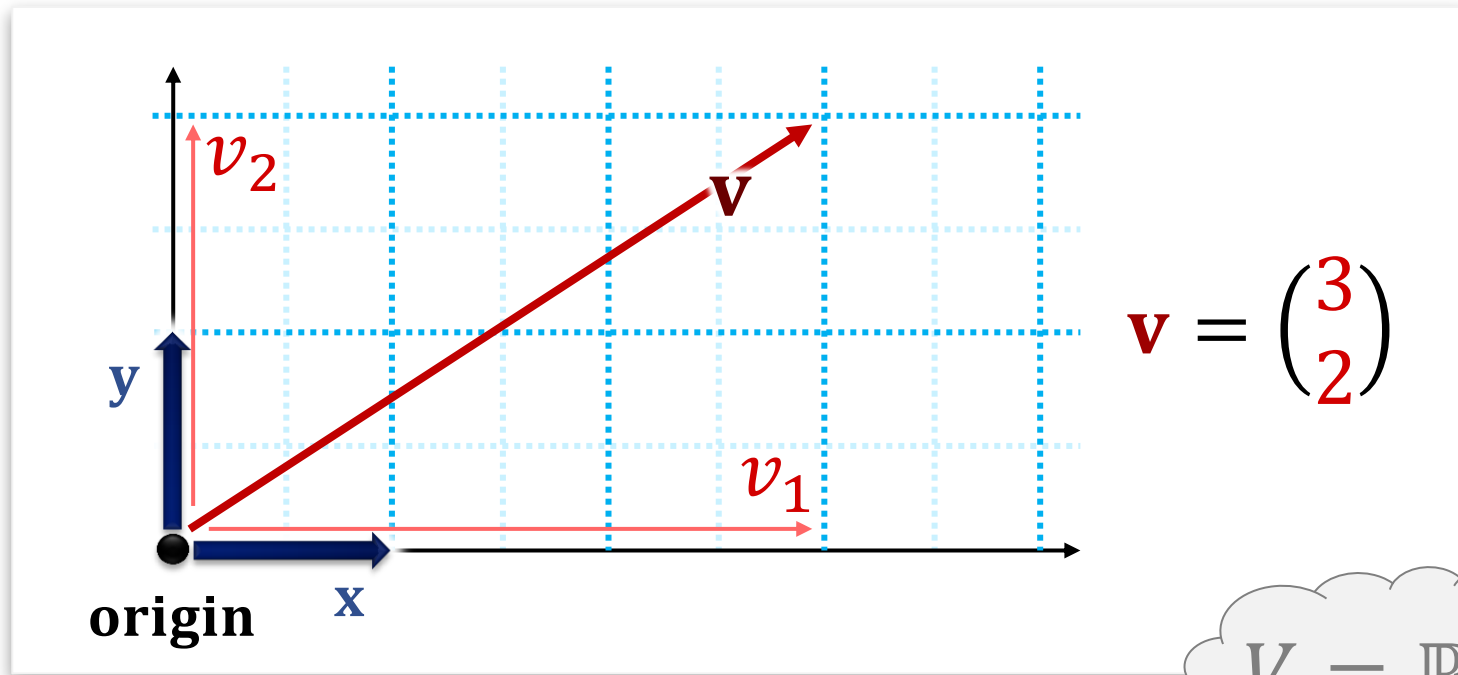


- Vectors
  - A vector is an arrow in space



- Points
  - Fix an origin
  - Store vector from origin to point
  - „Vectors are differences of points“

# ***Algebraic Representation*** ***(Implementation)***

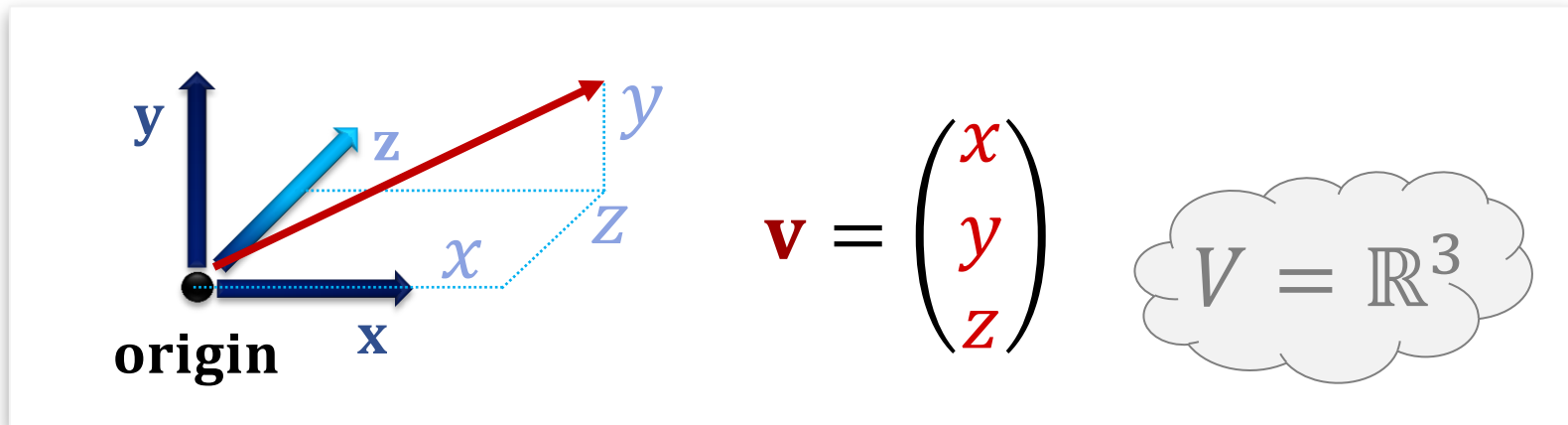


Coordinates!

$$\mathbf{v} = \begin{pmatrix} x\text{-coord.} \\ y\text{-coord.} \end{pmatrix} = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$$

Project on coordinate vectors

- We can add more entries:

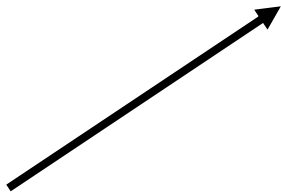


- Or even more entries:

$d = \text{“dimension”}$

$$\mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_d \end{pmatrix} \quad V = \mathbb{R}^d$$

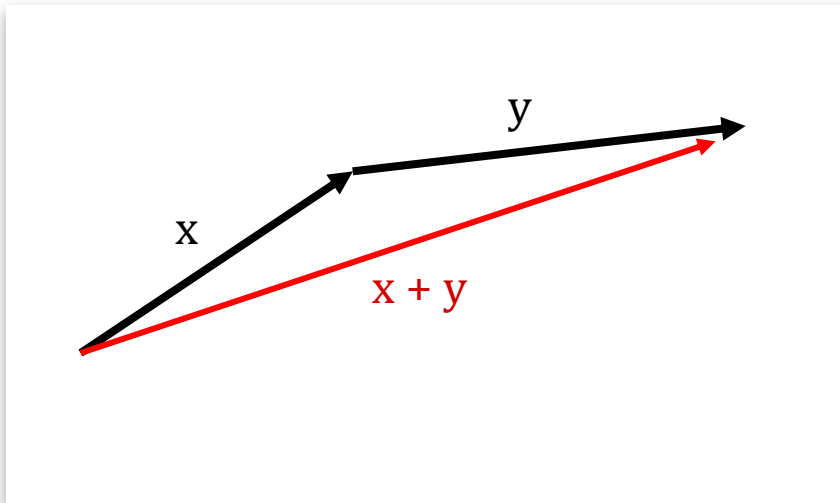




Geometry:  
vectors are arrows in space

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

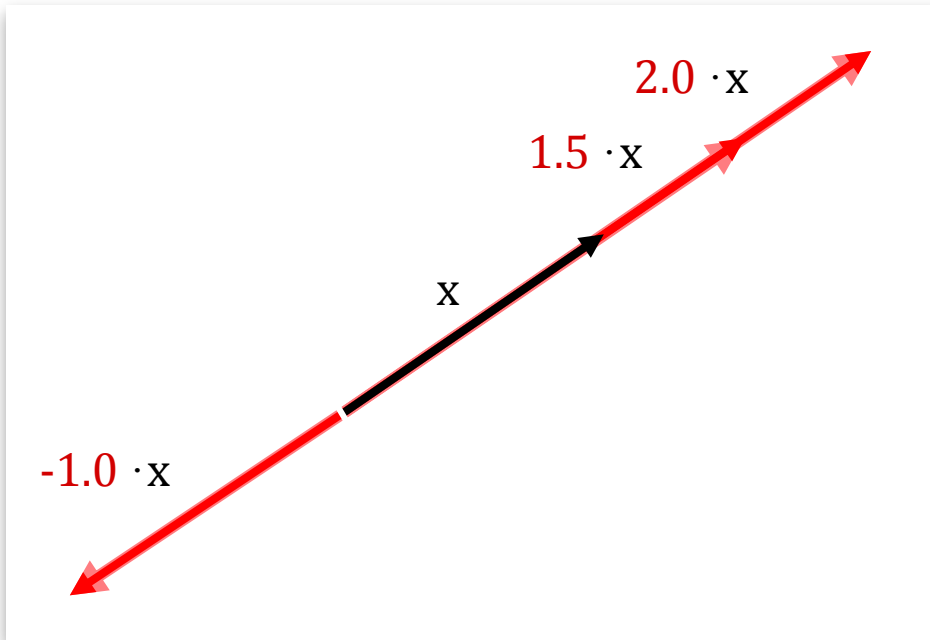
Algebra:  
arrays of numbers



Adding Vectors:  
Concatenation

$$\mathbf{x} + \mathbf{y} = \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \end{pmatrix}$$

Algebra:  
adding numbers



Scalar-Vector Multiplication:  
Scaling (incl. mirroring)

$$\lambda \mathbf{x} = \begin{pmatrix} \lambda x_1 \\ \lambda x_2 \end{pmatrix}$$

Algebra:  
multiplying with real number

- Null vector

$$\mathbf{0} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

- Does not change other vectors in addition
- $\mathbf{v} + \mathbf{0} = \mathbf{0} + \mathbf{v} = \mathbf{v}$  for all vectors  $\mathbf{v}$

- Definition: A *real vector space* of dimension  $d$ 
  - The set of all  $d$ -tuples:

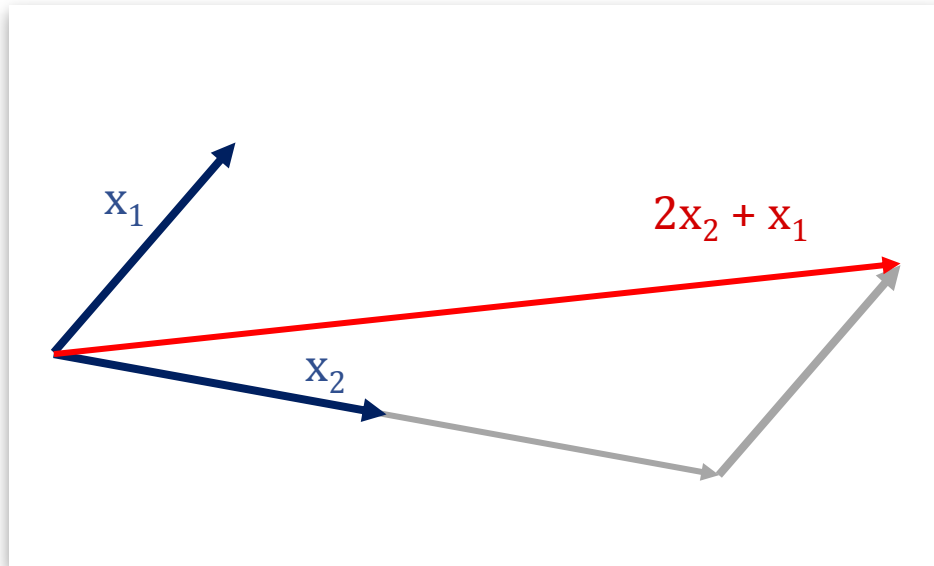
$$\underbrace{\mathbb{R} \times \cdots \times \mathbb{R}}_{d\text{-times}} = \mathbb{R}^d \quad \mathbf{x} \in \mathbb{R}^d : \mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_d \end{pmatrix}$$

- With two operations

$$\mathbf{x} + \mathbf{y} := \begin{pmatrix} x_1 + y_1 \\ \vdots \\ x_d + y_d \end{pmatrix}$$

$$\lambda \cdot \mathbf{x} = \lambda \mathbf{x} := \begin{pmatrix} \lambda x_1 \\ \vdots \\ \lambda x_d \end{pmatrix}$$

$$\mathbf{x}, \mathbf{y} \in \mathbb{R}^d, \quad \lambda \in \mathbb{R}$$



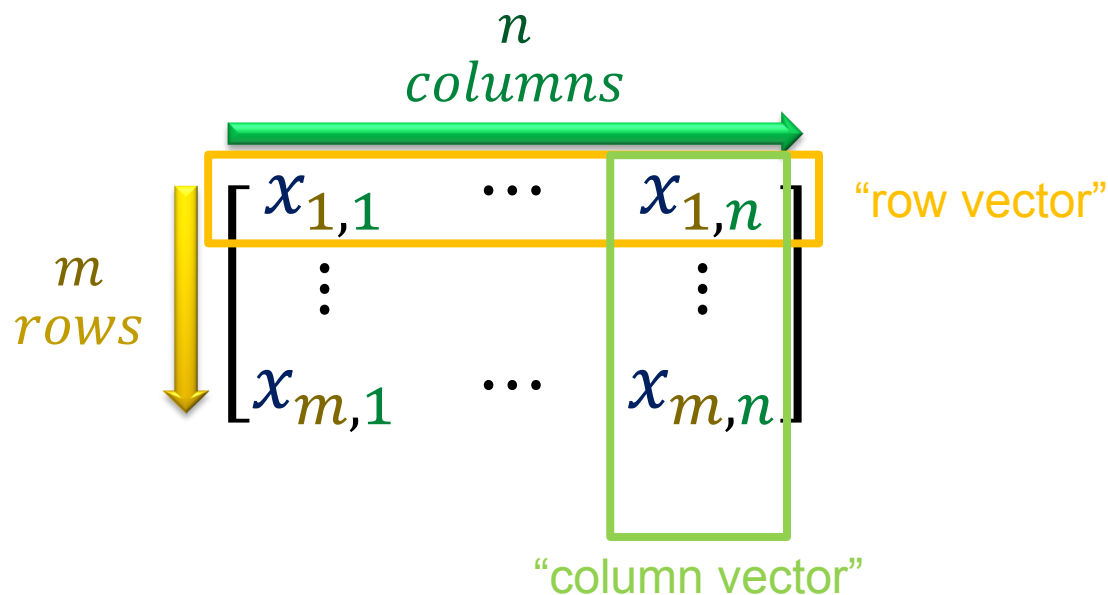
Geometrically

$$\mathbf{p} = \sum_{i=1}^n \lambda_i \mathbf{x}_i$$

Algebraically

The concept of **linear combinations** is the corner stone of graphics and visualization.

# ***Linear Combinations & Matrices***



- Matrix elements

$$x_{\text{row},\text{column}}$$

- Row first, then column

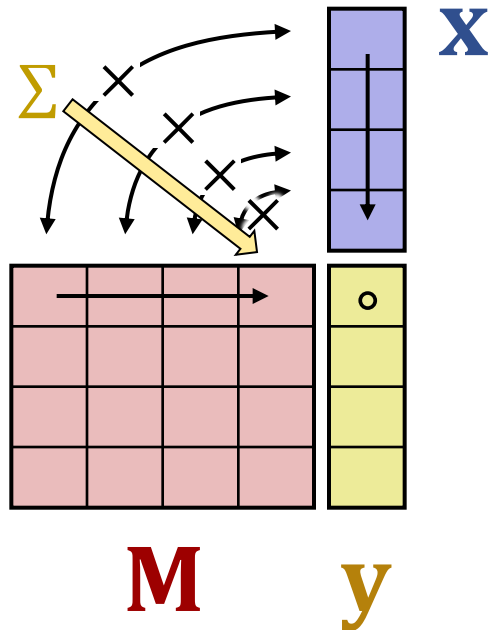
- "y"-coordinate of the array first  
(common convention)



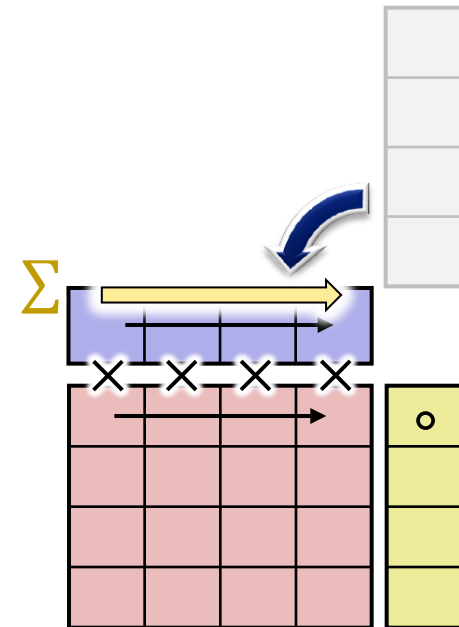
- Algebraic rule:
  - Vector-matrix product:

The diagram illustrates the vector-matrix product  $y = M \cdot x$ . On the left, a yellow vertical vector  $y$  (4x1) is shown. This is followed by an equals sign. To the right of the equals sign is a red 4x4 matrix  $M$ , represented as a grid of 16 cells. This is followed by a dot operator  $\cdot$  and a blue vertical vector  $x$  (4x1). To the right of this visual representation is the algebraic equation  $y = M \cdot x$ , where  $y$  is yellow,  $M$  is red, and  $x$  is blue.

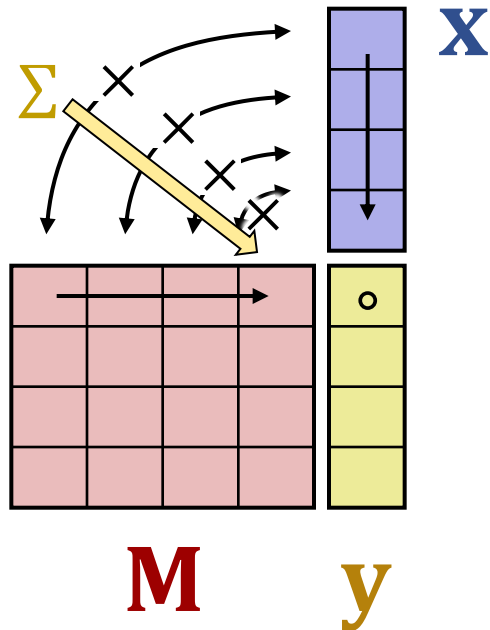
- Algebraic rule:
  - Vector-matrix product:



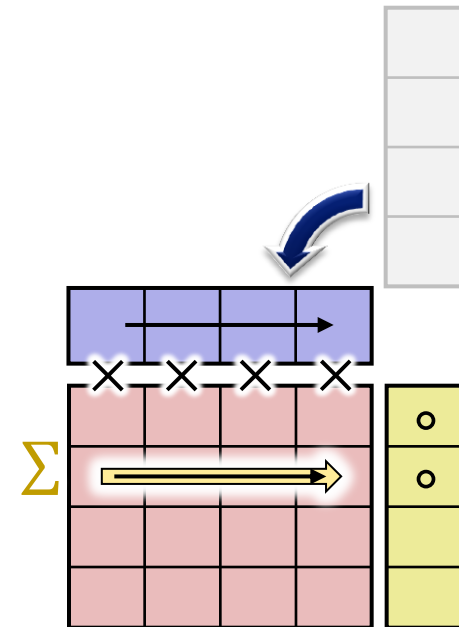
$$y = M \cdot x$$



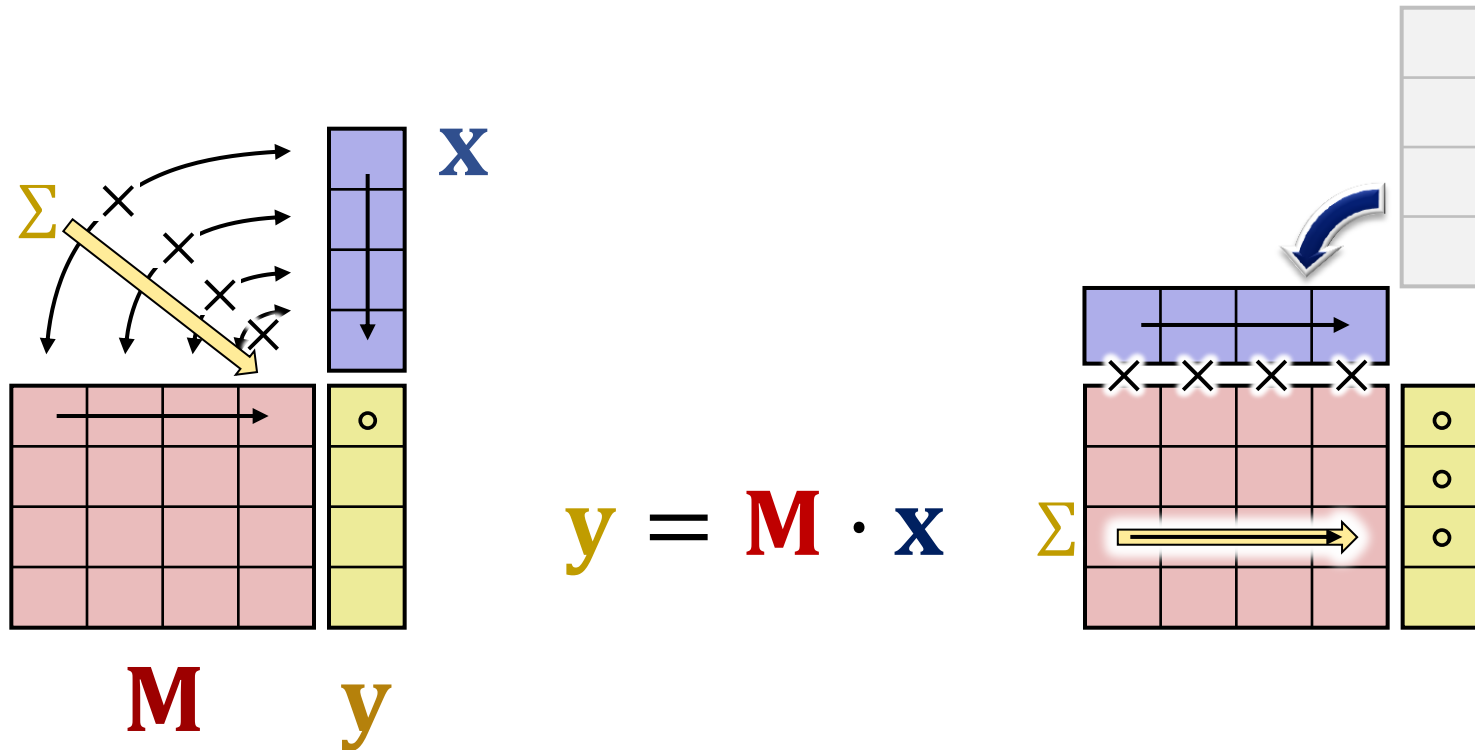
- Algebraic rule:
  - Vector-matrix product:



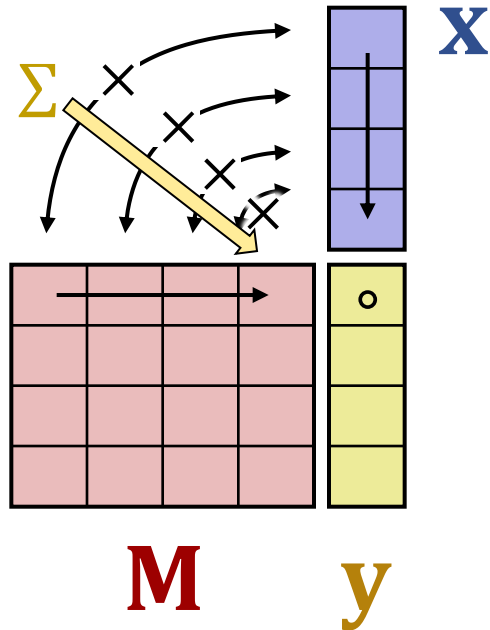
$$y = M \cdot x$$



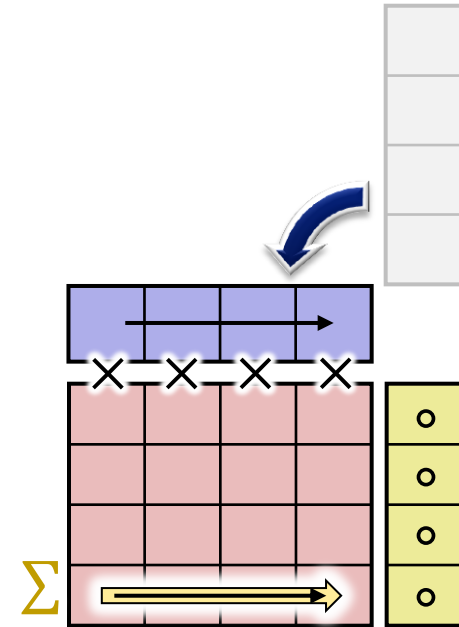
- Algebraic rule:
  - Vector-matrix product:



- Algebraic rule:
  - Vector-matrix product:



$$y = M \cdot x$$



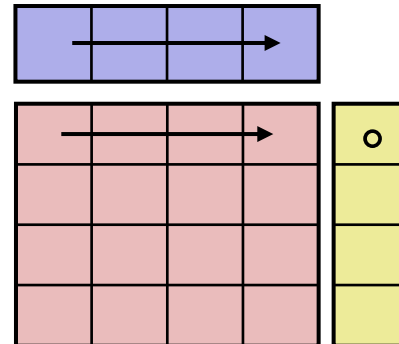
## • Matrix-Vector Multiplication

$$\begin{bmatrix} x_{1,1} & \cdots & x_{1,n} \\ \vdots & & \vdots \\ x_{m,1} & \cdots & x_{m,n} \end{bmatrix} \cdot \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_n \end{bmatrix} \stackrel{:=}{=} \sum_{i=1}^n \lambda_i \begin{bmatrix} x_{1,i} \\ \vdots \\ x_{m,i} \end{bmatrix} = \sum_{i=1}^n \lambda_i \mathbf{x}_i$$

column vectors  
of the matrix

Linear Combination

$$= \begin{bmatrix} \lambda_1 \cdot x_{1,1} + \cdots + \lambda_n \cdot x_{1,n} \\ \vdots \\ \lambda_1 \cdot x_{m,1} + \cdots + \lambda_n \cdot x_{m,n} \end{bmatrix}$$



# ***Standard Transformations***

- Translate a point  $\mathbf{p}$  along a vector  $\mathbf{t}$
- General case:

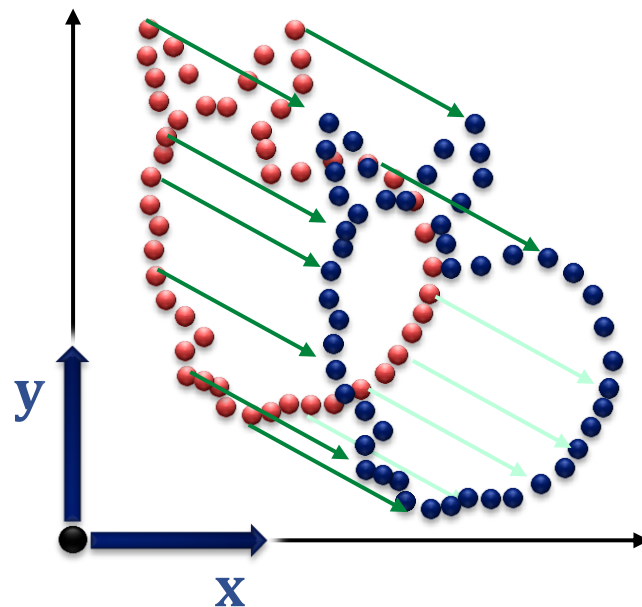
$$\mathbf{p}' = \mathbf{p} + \mathbf{t}$$

- 2D:

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} t_x \\ t_y \end{bmatrix} = \begin{bmatrix} x + t_x \\ y + t_y \end{bmatrix}$$

- 3D:

$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} x \\ y \\ z \end{bmatrix} + \begin{bmatrix} t_x \\ t_y \\ t_z \end{bmatrix} = \begin{bmatrix} x + t_x \\ y + t_y \\ z + t_z \end{bmatrix}$$





- Scale a point  $\mathbf{p}$  in each dimension by the factors  $s_x, s_y, s_z$

- General case:

$$\mathbf{p}' = \mathbf{S} \cdot \mathbf{p} \quad \mathbf{S}: \mathbb{R}^n \rightarrow \mathbb{R}^n,$$

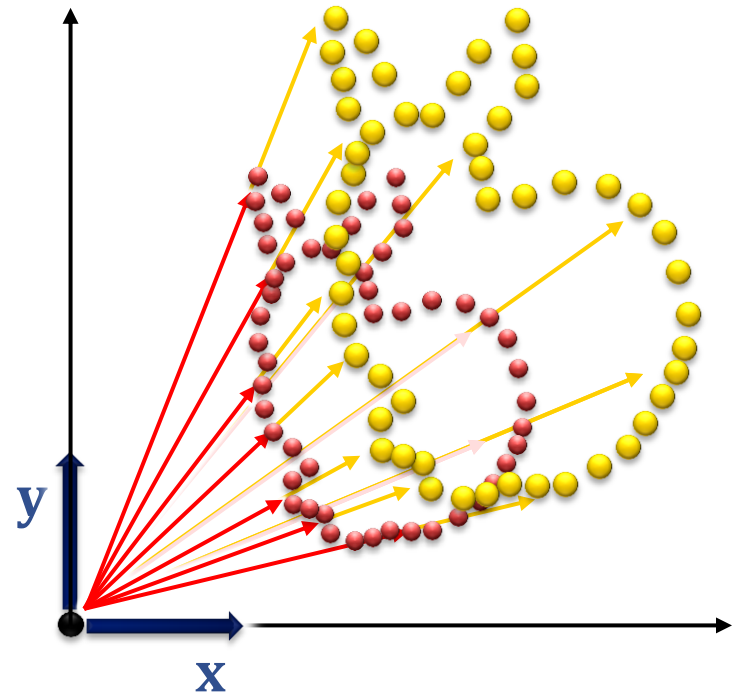
$$\mathbf{S} = \begin{bmatrix} s_1 & 0 & \cdots & 0 \\ 0 & s_2 & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & \cdots & s_n \end{bmatrix}$$

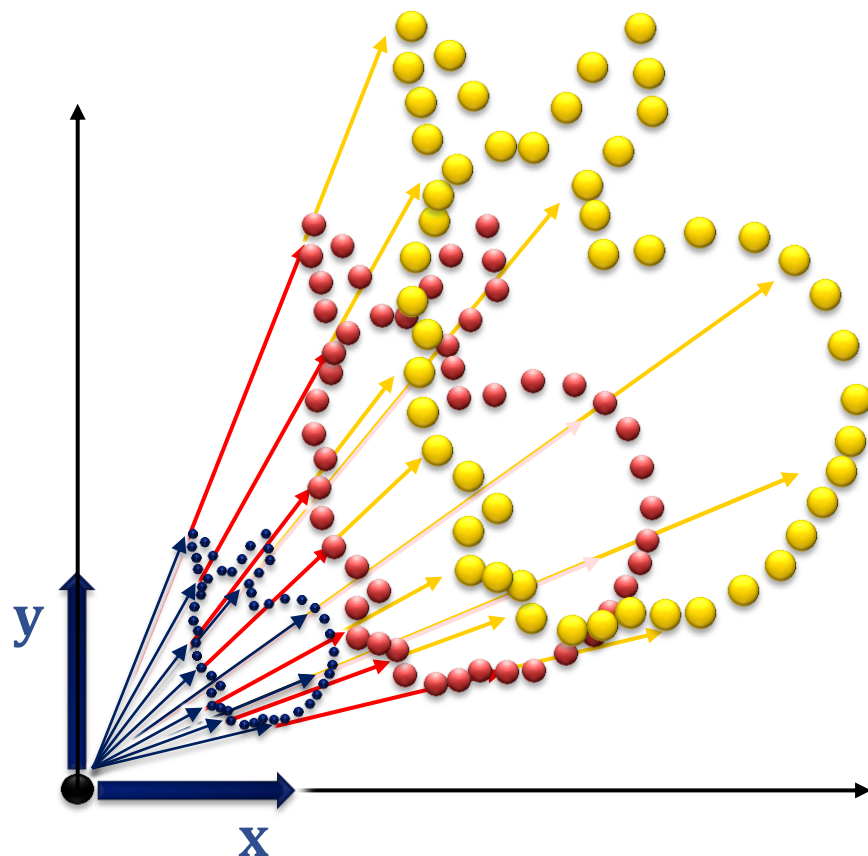
- 2D:

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

- 3D:

$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & s_z \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$





Making something uniformly smaller:

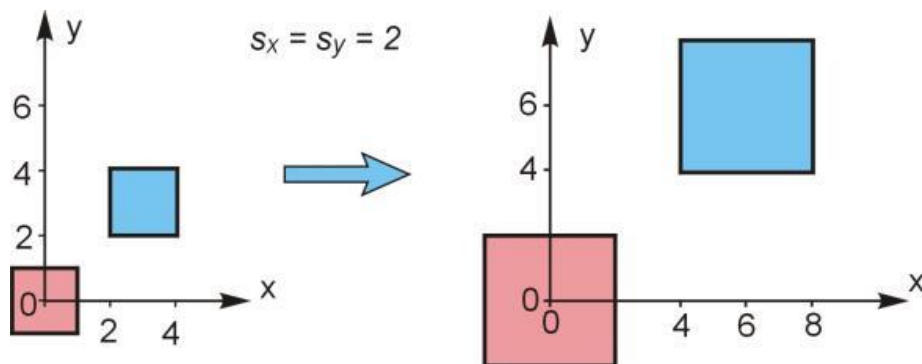
$$s_x = s_y = s_z < 1$$

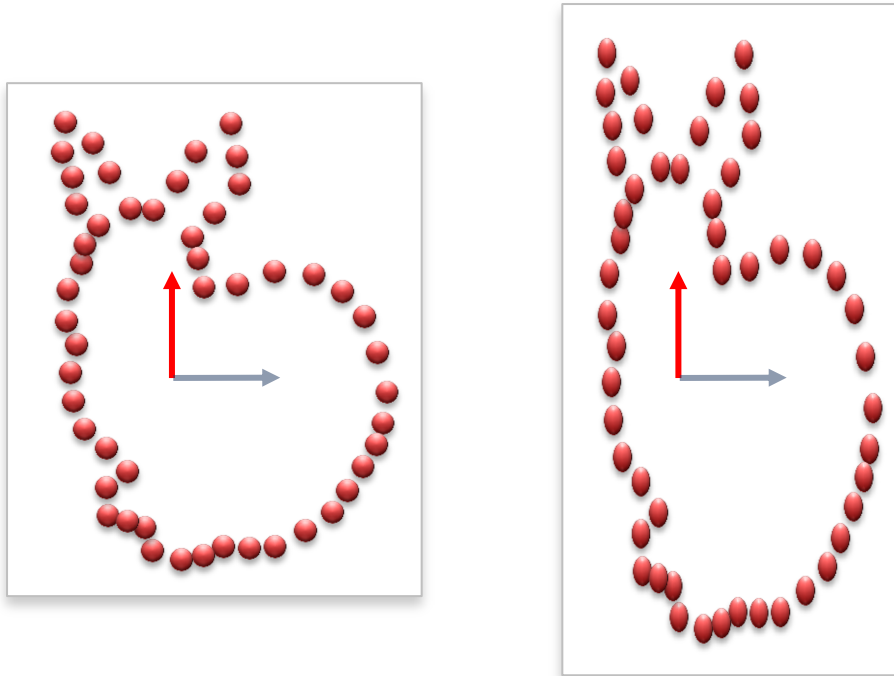
Making something uniformly bigger:

$$s_x = s_y = s_z > 1$$

Note:

Center is at the origin

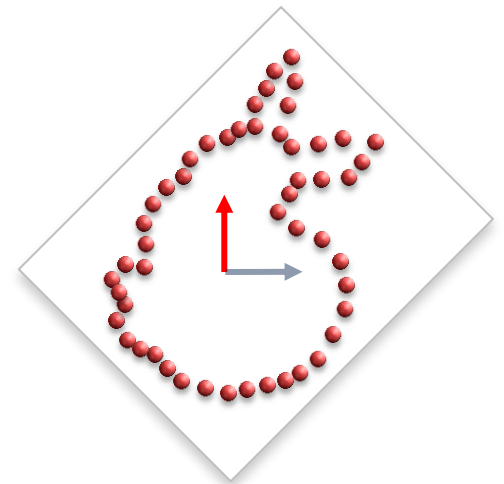
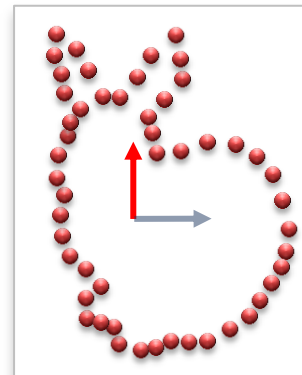




$$S_x \neq S_y \neq S_z$$

- Rotate a point  $\mathbf{p}$  around the origin with an angle  $\alpha$  in counter-clockwise direction
- 2D:

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$



$$x' = r * \cos(\alpha + \phi)$$

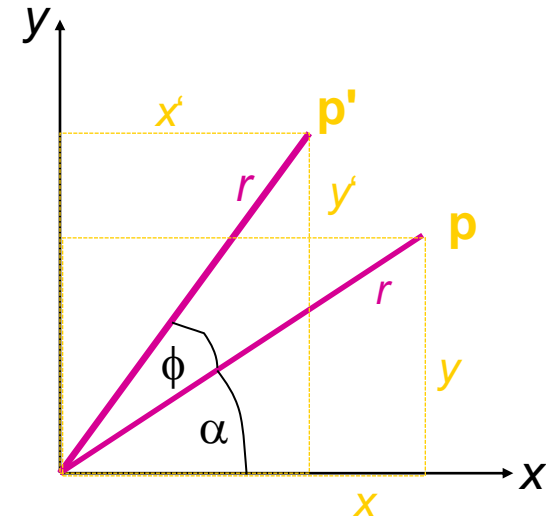
$$x' = \underline{r * \cos \alpha} * \cos \phi - \underline{r * \sin \alpha} * \sin \phi$$

$$x' = x * \cos \phi - y * \sin \phi$$

$$y' = r * \sin(\alpha + \phi)$$

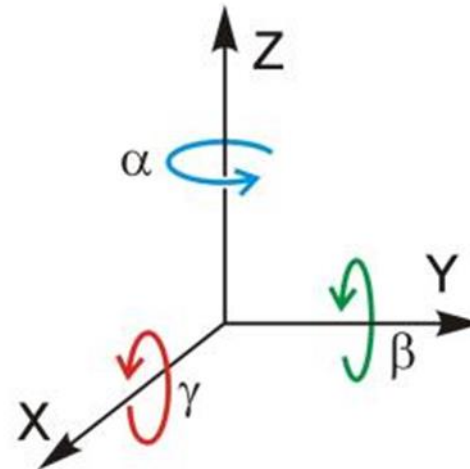
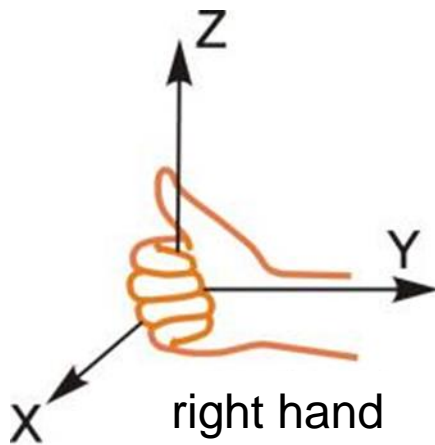
$$y' = \underline{r * \cos \alpha} * \sin \phi + \underline{r * \sin \alpha} * \cos \phi$$

$$y' = x * \sin \phi + y * \cos \phi$$



Remark: The  $\alpha$  from this slide is not the  $\alpha$  from the previous slide!

- Rotate a point  $\mathbf{p}$  around a rotation axis with an angle  $\alpha$  in counter-clockwise direction



- Rotation matrices for the rotation around the coordinate axes:

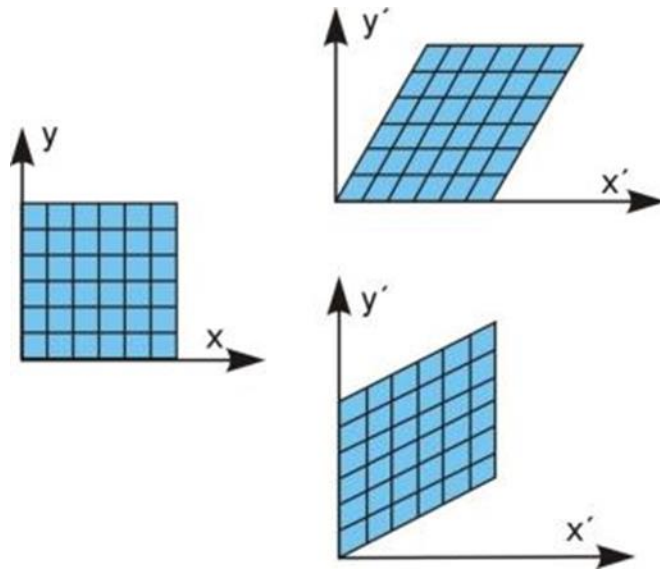
$$\mathbf{R}_x = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & -\sin \alpha \\ 0 & \sin \alpha & \cos \alpha \end{pmatrix}$$

$$\mathbf{R}_y = \begin{pmatrix} \cos \alpha & 0 & \sin \alpha \\ 0 & 1 & 0 \\ -\sin \alpha & 0 & \cos \alpha \end{pmatrix}$$

$$\mathbf{R}_z = \begin{pmatrix} \cos \alpha & -\sin \alpha & 0 \\ \sin \alpha & \cos \alpha & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

- A shear is given as
- 2D:

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 1 & s_y \\ s_x & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$



$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 1 & s_y \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

shear in  $x$ -direction

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ s_x & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

shear in  $y$ -direction

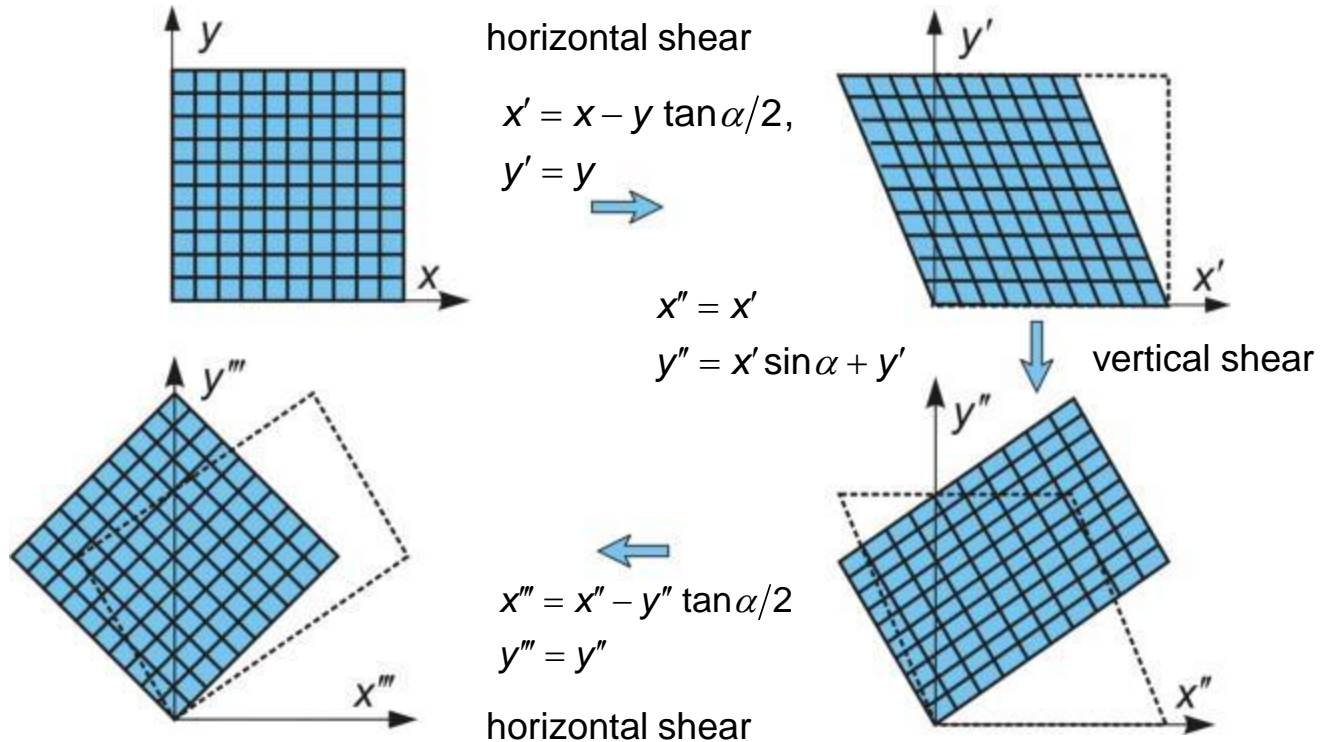


- A shear is given as
- 3D:

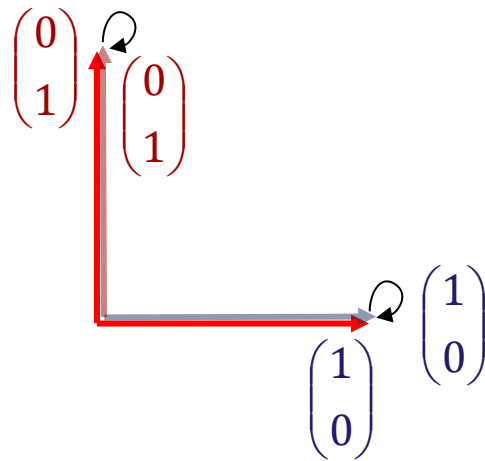
$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} 1 & s_{yx} & s_{zx} \\ s_{xy} & 1 & s_{zy} \\ s_{xz} & s_{yz} & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

- Shears can be used to describe rotations
- Example: Rotation of 2D objects using three subsequent shear transformations

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} 1 & -\tan\alpha/2 \\ 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ \sin\alpha & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & -\tan\alpha/2 \\ 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} x \\ y \end{pmatrix}$$



- The **Identity matrix** keeps points in their original location.



$$\mathbf{M}_{identity} = \mathbf{I} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

## General case

$$\mathbf{I}: \mathbb{R}^n \rightarrow \mathbb{R}^n, \quad \mathbf{I} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$$

# ***Homogeneous Coordinates*** ***(short version)***

- Translations are not linear
  - $\mathbf{x} \rightarrow \mathbf{M}\mathbf{x}$  cannot encode translations
  - **Proof:** Origin cannot be moved:

$$\mathbf{M} \cdot \mathbf{0} = \begin{pmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

- Solution: Just add a constant one
  - Increase dimension  $\mathbb{R}^d \rightarrow \mathbb{R}^{d+1}$
  - Last entry = 1 in vectors
    - “Cheap Trick”, “Evil Hack”

$$\begin{aligned}
 \mathbf{M}' \cdot \mathbf{x} &= \begin{pmatrix} m_{11} & m_{12} & m_{13} & t_1 \\ m_{21} & m_{22} & m_{23} & t_2 \\ m_{31} & m_{32} & m_{33} & t_3 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \\ 1 \end{pmatrix} \\
 &= \begin{pmatrix} \ddots & & \ddots & | \\ & \mathbf{M} & & \mathbf{t} \\ \ddots & & \ddots & | \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} | \\ \mathbf{x} \\ | \\ 1 \end{pmatrix} = \begin{pmatrix} | \\ \mathbf{M}\mathbf{x} + \mathbf{t} \\ | \\ 1 \end{pmatrix}
 \end{aligned}$$

- General case

$$\mathbf{M} \cdot \mathbf{x} = \begin{pmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \\ m_{41} & m_{42} & m_{43} & m_{44} \end{pmatrix} \begin{pmatrix} x \\ y \\ z \\ 1 \end{pmatrix} = \begin{pmatrix} x' \\ y' \\ z' \\ w' \end{pmatrix}$$

- $w'$  might be different from 1
- Convention: Divide by  $w$ -coord. before using

$$\text{Result: } \begin{pmatrix} x'/w' \\ y'/w' \\ z'/w' \\ 1 \end{pmatrix}$$

- General case

$$\mathbf{M} \cdot \mathbf{x} = \begin{pmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \\ m_{41} & m_{42} & m_{43} & m_{44} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ 1 \end{pmatrix} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix} \equiv \begin{pmatrix} y_1/y_4 \\ y_2/y_4 \\ y_3/y_4 \\ 1 \end{pmatrix}$$

- Rules:

- Before using as 3D point, divide by last (4th) entry
- No normalization required during subsequent transformations (matrix-multiplications, see later)



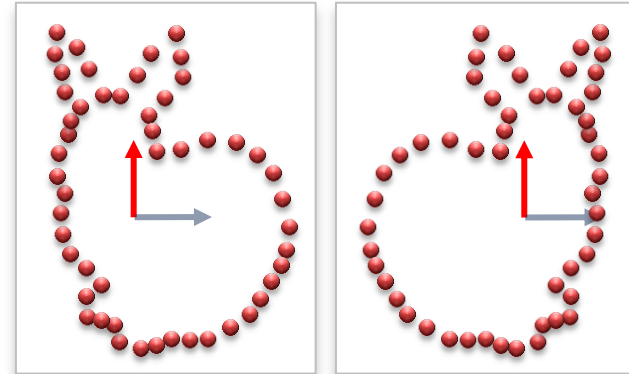
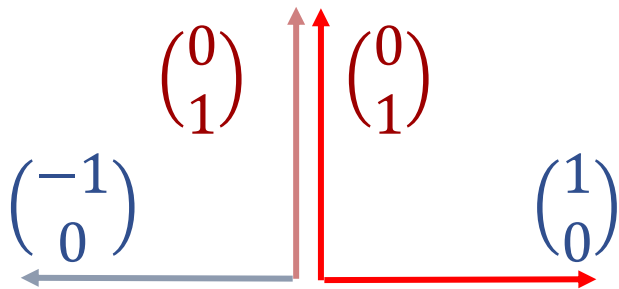
- Projective Geometry
  - Not just an evil hack
  - Deep & interesting theoretical background
  - More on this later
- For simplicity
  - We'll treat it as a computational trick for now
    - Focus on the graphics application
  - Remember for now:
    - We can build “4D Translation matrices” for 3D+1 points
    - We can “divide” by a common linear factor

# ***Overview Standard Transformations with Homogeneous Coordinates***

	Translation	Scaling	Shearing
2D	$\mathbf{T}(t_x, t_y) = \begin{pmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{pmatrix}$	$\mathbf{S}(s_x, s_y) = \begin{pmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & 1 \end{pmatrix}$	$\mathbf{H}_x = \begin{pmatrix} 1 & h_y & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
3D	$\mathbf{T}(t_x, t_y, t_z) = \begin{pmatrix} 1 & 0 & 0 & t_x \\ 0 & 1 & 0 & t_y \\ 0 & 0 & 1 & t_z \\ 0 & 0 & 0 & 1 \end{pmatrix}$	$\mathbf{S}(s_x, s_y, s_z) = \begin{pmatrix} s_x & 0 & 0 & 0 \\ 0 & s_y & 0 & 0 \\ 0 & 0 & s_z & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$	$\mathbf{H} = \begin{pmatrix} 1 & s_1 & s_2 & 0 \\ s_3 & 1 & s_4 & 0 \\ s_5 & s_6 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

2D-Rotation	3D-Rotation
$\mathbf{R}(\phi) = \begin{pmatrix} \cos \phi & -\sin \phi & 0 \\ \sin \phi & \cos \phi & 0 \\ 0 & 0 & 1 \end{pmatrix}$	Rotation around $x$ -axis $\mathbf{R}_x(\phi) = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos \phi & -\sin \phi & 0 \\ 0 & \sin \phi & \cos \phi & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$
	Rotation around $y$ -axis $\mathbf{R}_y(\phi) = \begin{pmatrix} \cos \phi & 0 & \sin \phi & 0 \\ 0 & 1 & 0 & 0 \\ -\sin \phi & 0 & \cos \phi & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$
	Rotation around $z$ -axis $\mathbf{R}_z(\phi) = \begin{pmatrix} \cos \phi & -\sin \phi & 0 & 0 \\ \sin \phi & \cos \phi & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

# ***Further Transformations***



$$\mathbf{M}_{refl} = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$$

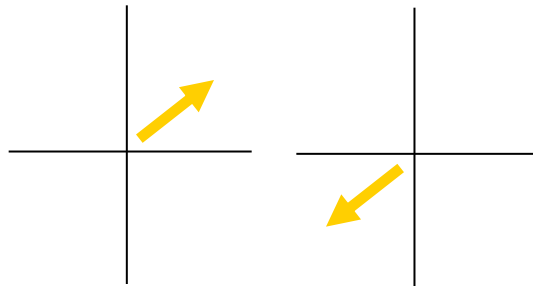
## General case

$$\mathbf{S}_\lambda: \mathbb{R}^n \rightarrow \mathbb{R}^n,$$

$$\mathbf{S}_\lambda = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & -1 & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

Reflection Axis

$$\begin{pmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

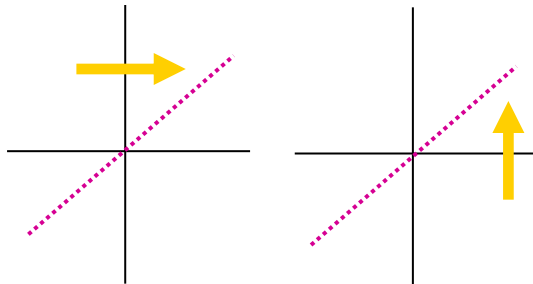


*before*

*after*

reflection over the origin

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$



*before*

*after*

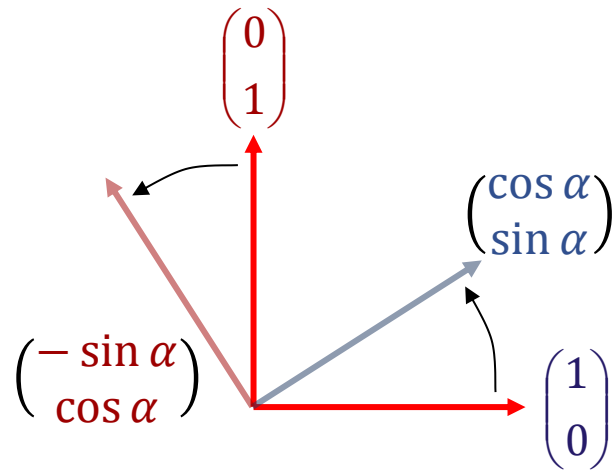
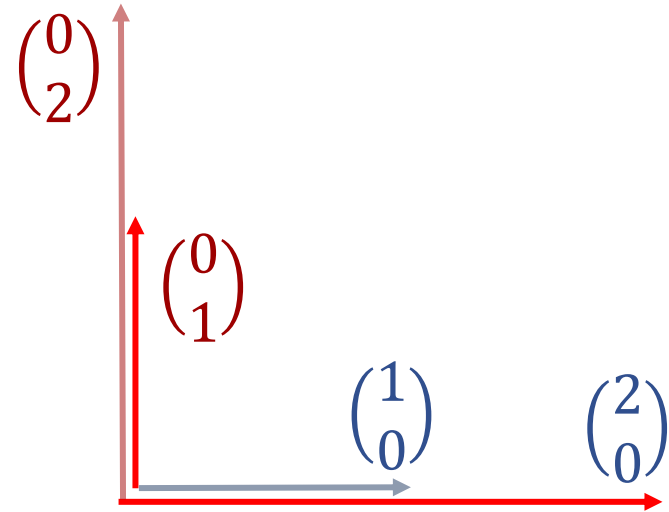
reflection at the line  $y=x$

# ***Combining Transformations***

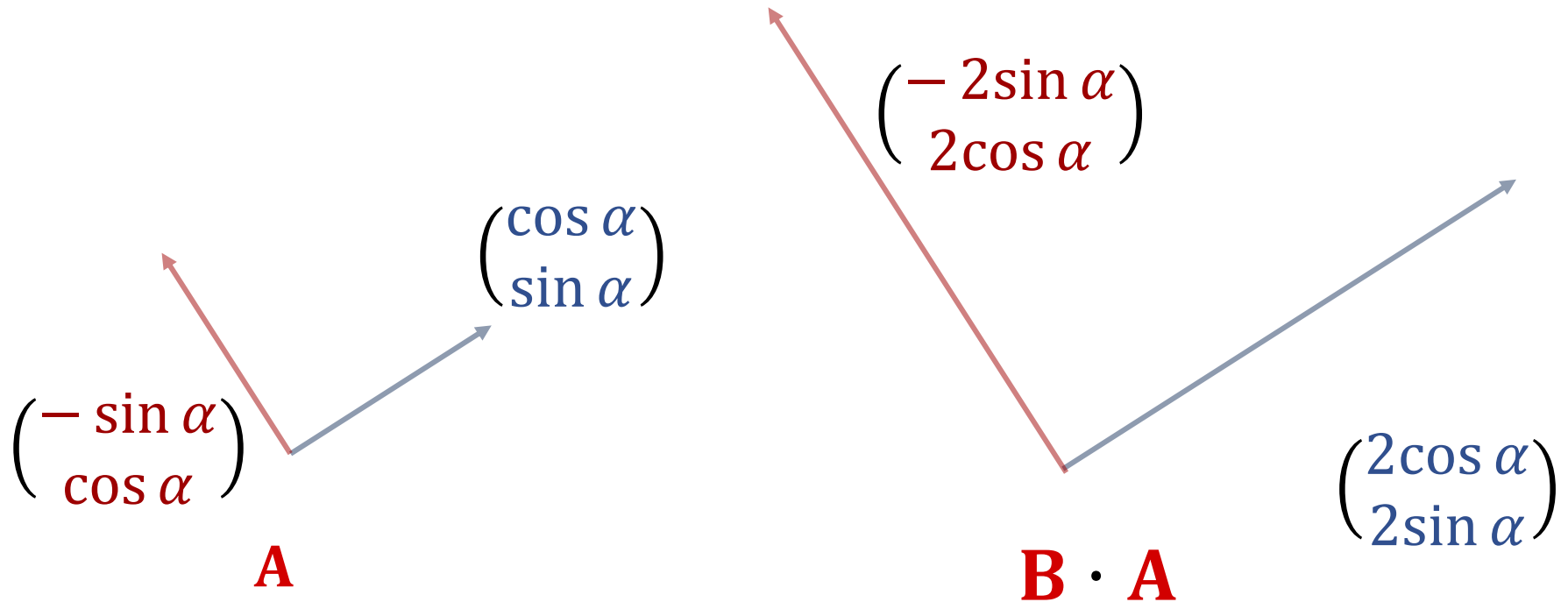


- You can combine all of these
- Example: General axis of rotation
  - First rotate rotation axis to x-axis
  - Rotate around x
  - Rotate back
- Question
  - How to combine multiple transformation matrices?

- Execute multiple transformations, one after another
  - Written as product: matrix multiplication
  - $(\mathbf{B} \cdot \mathbf{A}) \cdot \mathbf{x}$ :
    - Apply  $\mathbf{A}$  to  $\mathbf{x}$  first
    - Then  $\mathbf{B}$
    - $(\mathbf{B} \cdot \mathbf{A})$  is again a matrix

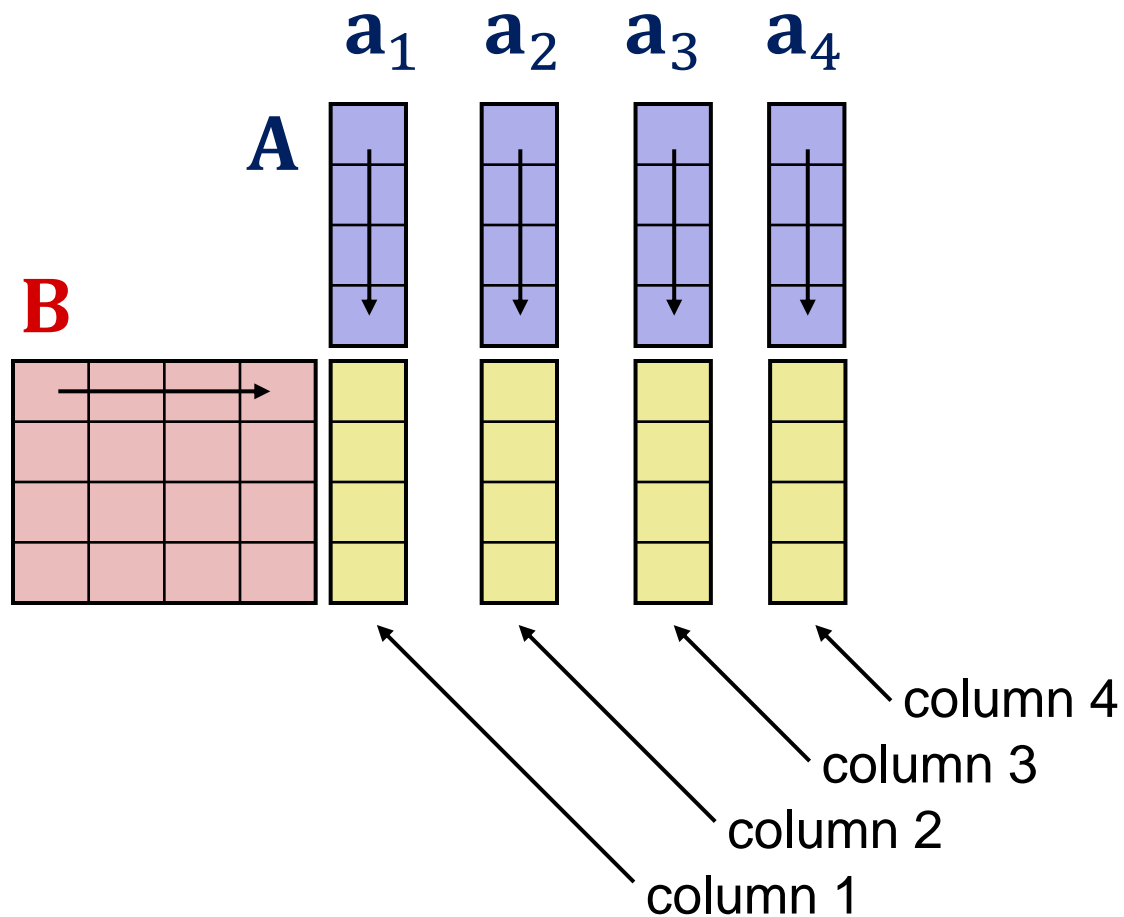
**A****B**

- Consider  $(\mathbf{B} \cdot \mathbf{A})$ :
  - Rotate first (**A**)
  - Then scale (**B**)

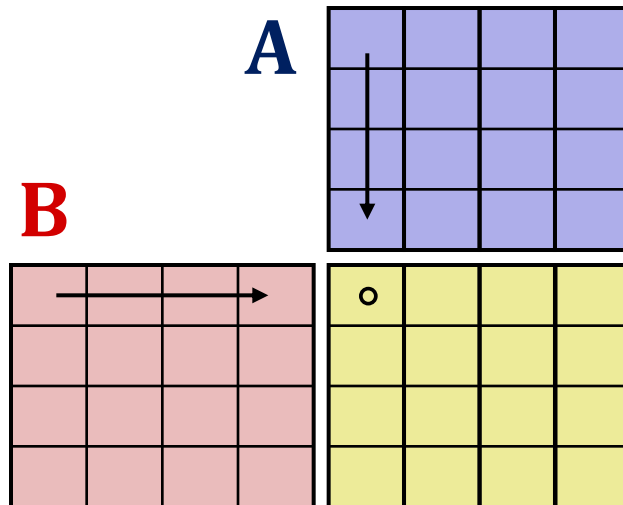


- How to compute  $(\mathbf{B} \cdot \mathbf{A})$ ?
  - Transform basis vectors
  - Transform again

- Matrix product:

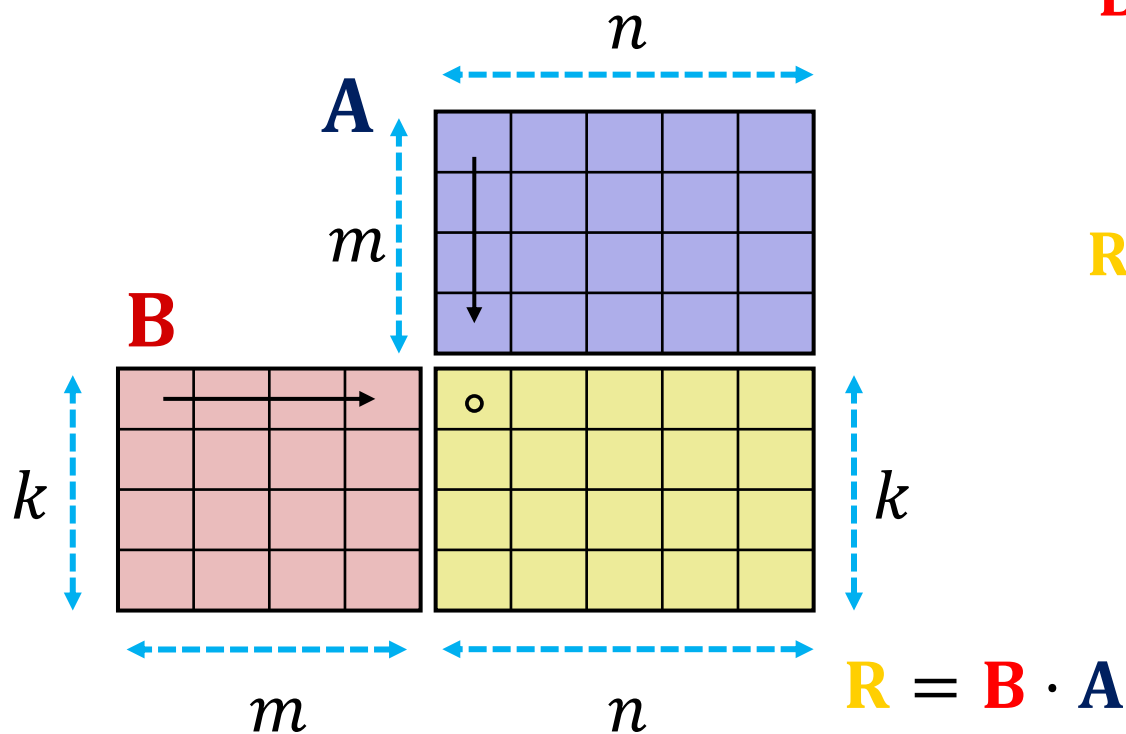


- Matrix product:



- General matrix products:

- $\mathbf{B} \cdot \mathbf{A}$ : possible if  
 $\#Row(\mathbf{A}) = \#Columns(\mathbf{B})$



$$\mathbf{A} = \begin{bmatrix} a_{1,1} & \cdots & a_{1,n} \\ \vdots & & \vdots \\ a_{m,1} & \cdots & a_{m,n} \end{bmatrix}$$

$$\mathbf{B} = \begin{bmatrix} b_{1,1} & \cdots & b_{1,m} \\ \vdots & & \vdots \\ b_{k,1} & \cdots & b_{k,m} \end{bmatrix}$$

$$\mathbf{R} = \begin{bmatrix} r_{1,1} & \cdots & r_{1,n} \\ \vdots & & \vdots \\ r_{k,1} & \cdots & r_{k,n} \end{bmatrix}$$

$$r_{i,j} = \sum_{q=1}^m a_{q,j} \cdot b_{i,q}$$

- Matrix-Multiplication

- Associative

$$(\mathbf{A} \cdot \mathbf{B}) \cdot \mathbf{C} = \mathbf{A} \cdot (\mathbf{B} \cdot \mathbf{C})$$

- Includes vector-multiplication

$$(\mathbf{A} \cdot \mathbf{B}) \cdot \mathbf{v} = \mathbf{A} \cdot (\mathbf{B} \cdot \mathbf{v})$$

- In general, not commutative:

It might be that  $\mathbf{A} \cdot \mathbf{B} \neq \mathbf{B} \cdot \mathbf{A}$

- Linear

$$\mathbf{A} \cdot (\mathbf{v} + \mathbf{w}) = \mathbf{A} \cdot \mathbf{v} + \mathbf{A} \cdot \mathbf{w}$$

$$\mathbf{A} \cdot (\lambda \cdot \mathbf{v}) = \lambda \cdot (\mathbf{A} \cdot \mathbf{v})$$

## Settings

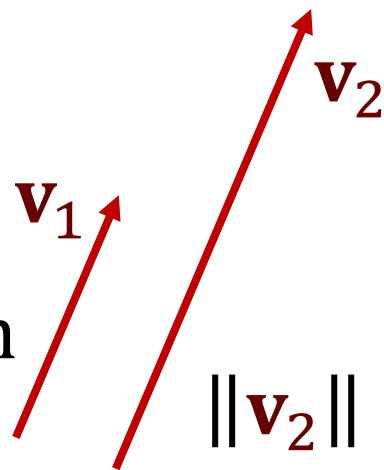
$$\lambda \in \mathbb{R}$$

$\mathbf{A}, \mathbf{B}, \mathbf{C}$  - matrices

$\mathbf{v}, \mathbf{w}$  - vectors



# ***More Vector Operations: Scalar Products***

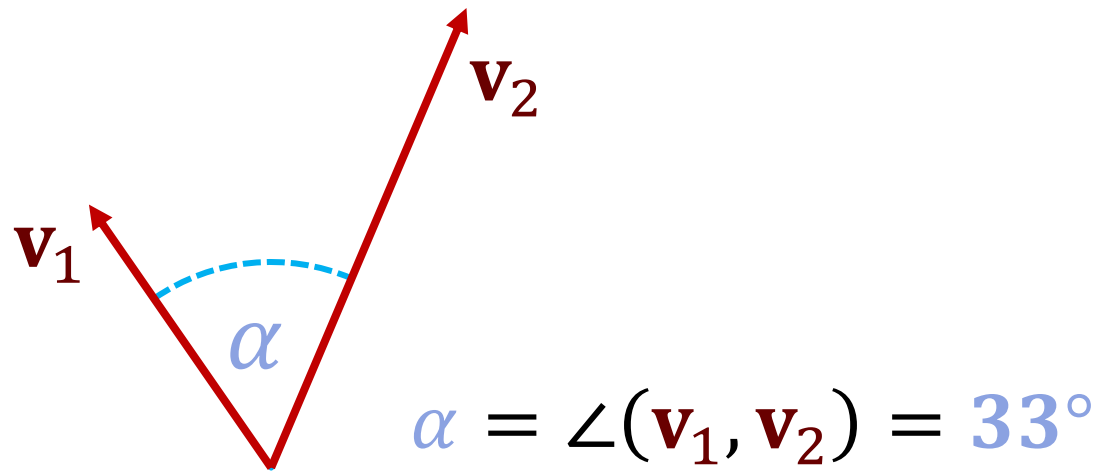


$\|\mathbf{v}_1\| = 2.3\text{cm}$        $\|\mathbf{v}_2\| = 4.2\text{cm}$

Length of Vectors

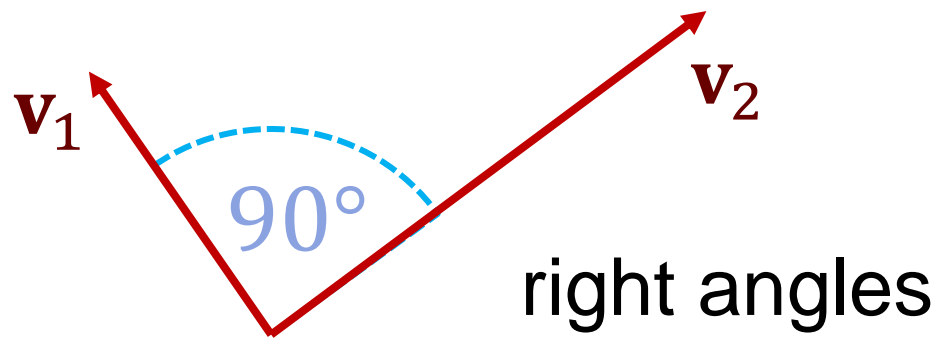
“length” or “norm”

$\|\mathbf{v}\|$  yields real number  $\geq 0$

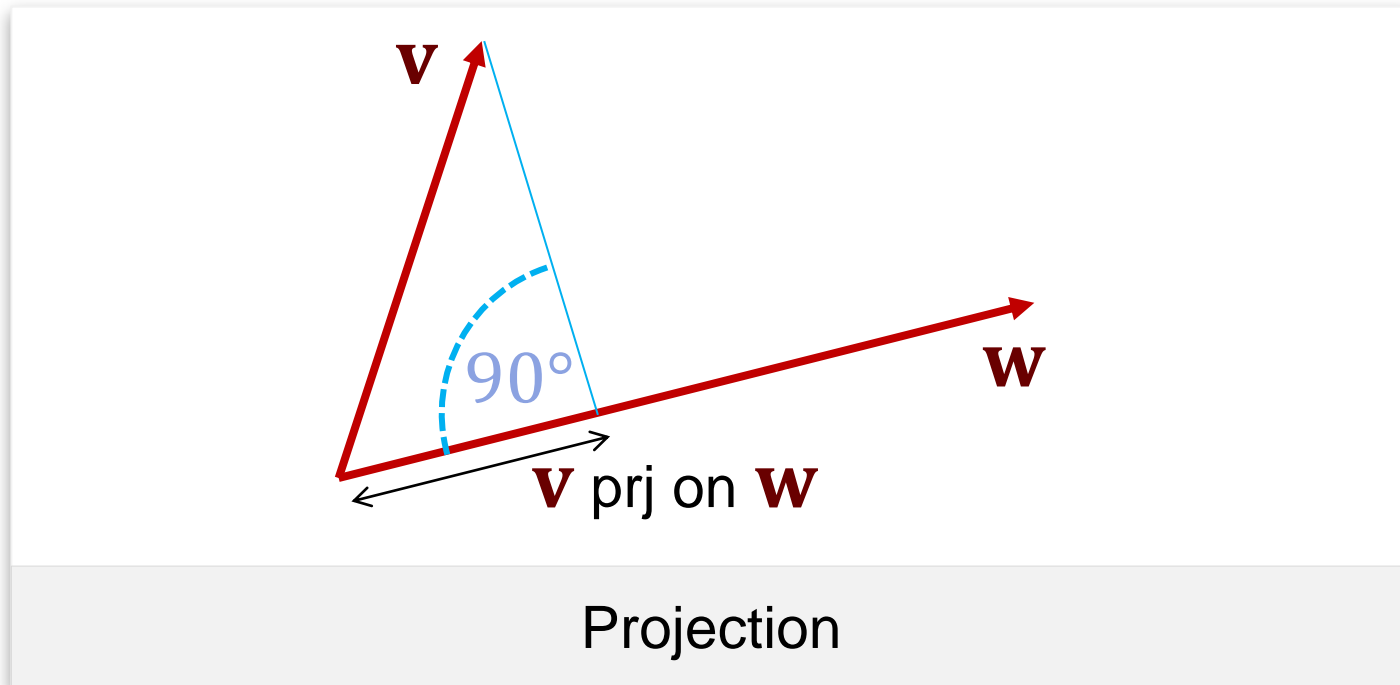


Angle between Vectors

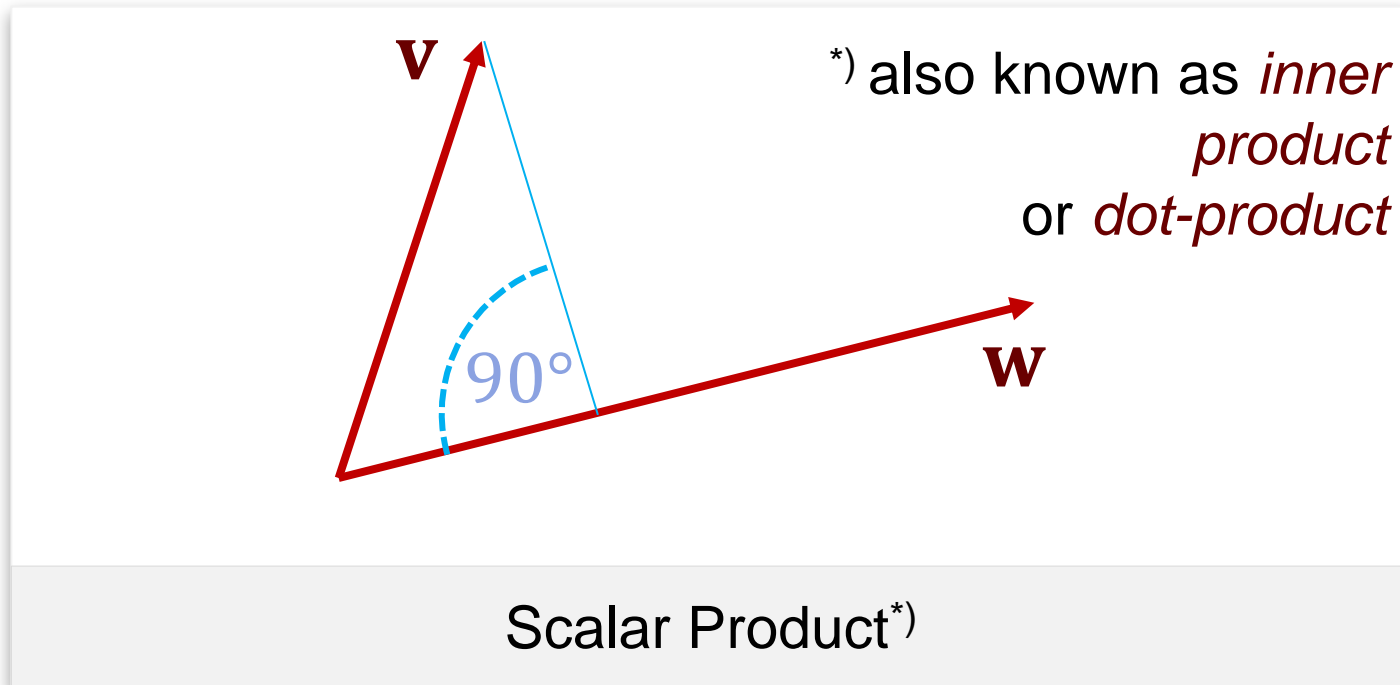
angle  $\angle(\mathbf{v}_1, \mathbf{v}_2)$   
yields real number  
 $[0, \dots, 2\pi) = [0, \dots, 360^\circ)$



Angle between Vectors

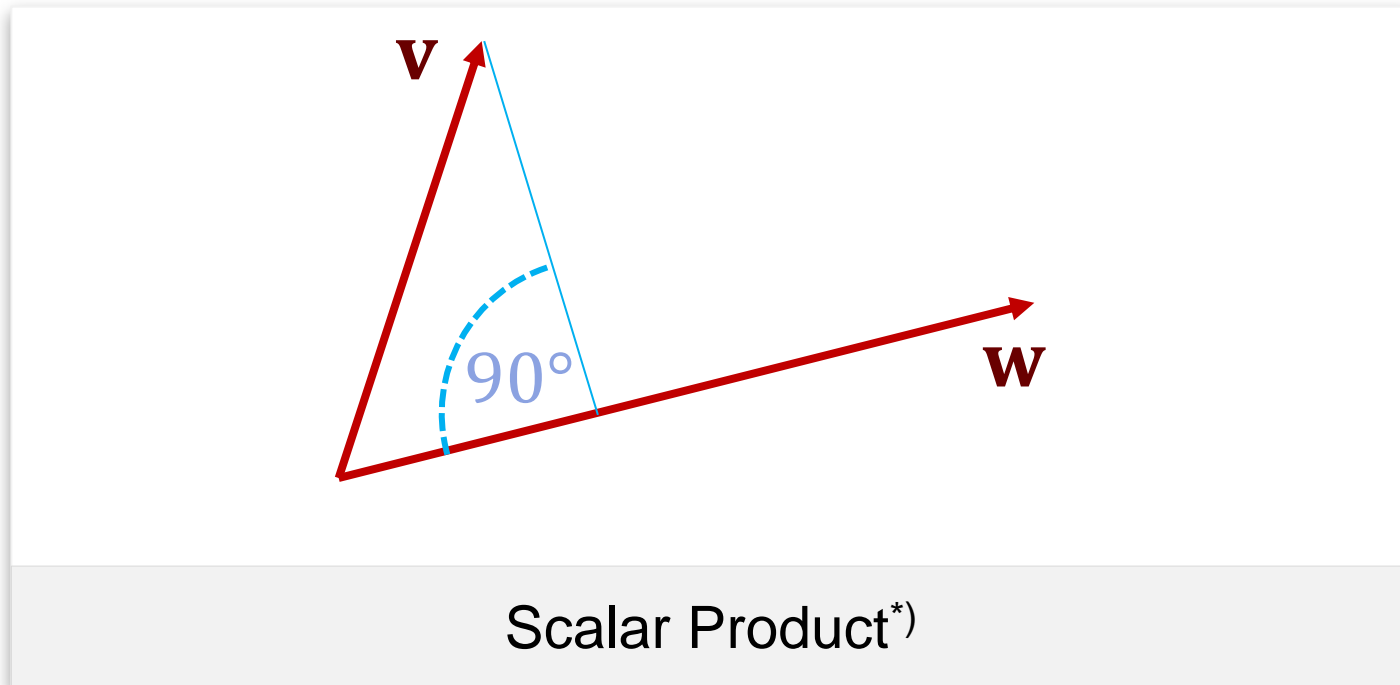


Projection: determine  
length of  $\mathbf{v}$  along direction of  $\mathbf{w}$



$$\mathbf{v} \cdot \mathbf{w} = \|\mathbf{v}\| \cdot \|\mathbf{w}\| \cdot \cos \angle(\mathbf{v}, \mathbf{w})$$

$$\mathbf{v} \cdot \mathbf{w} = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} \cdot \begin{pmatrix} w_1 \\ w_2 \end{pmatrix} := v_1 \cdot w_1 + v_2 \cdot w_2$$



$$\mathbf{v} \cdot \mathbf{w} = \|\mathbf{v}\| \cdot \|\mathbf{w}\| \cdot \cos \angle(\mathbf{v}, \mathbf{w})$$

also:  $\langle \mathbf{v}, \mathbf{w} \rangle$

$$\text{Length: } \|\mathbf{v}\| = \sqrt{\mathbf{v} \cdot \mathbf{v}}$$

$$\text{Angle: } \angle(\mathbf{v}, \mathbf{w}) = \arccos(\mathbf{v} \cdot \mathbf{w})$$

$$\text{Projection: } \text{„}\mathbf{v} \text{ prj on } \mathbf{w}\text{”} = \frac{\mathbf{v} \cdot \mathbf{w}}{\|\mathbf{w}\|}$$

$$\mathbf{v} \cdot \mathbf{w} = \|\mathbf{v}\| \cdot \|\mathbf{w}\| \cdot \cos \angle(\mathbf{v}, \mathbf{w})$$

Comprises: length, projection, angles



- Properties

- Symmetry (commutativity)

$$\langle \mathbf{u}, \mathbf{v} \rangle = \langle \mathbf{v}, \mathbf{u} \rangle$$

- Bilinearity

$$\langle \lambda \mathbf{v}, \mathbf{w} \rangle = \lambda \langle \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{v}, \lambda \mathbf{w} \rangle$$

$$\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle$$

*(symmetry: same for second argument)*

- Positive definite

$$\langle \mathbf{u}, \mathbf{u} \rangle \geq 0, \quad [\langle \mathbf{u}, \mathbf{u} \rangle = \mathbf{0}] \Rightarrow [\mathbf{u} = \mathbf{0}]$$

### Settings

$$\lambda \in \mathbb{R}$$

$$\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathbb{R}^d$$

- Do not mix
  - Scalar-vector product
  - Inner (scalar) product

- In general

$$\langle \mathbf{x}, \mathbf{y} \rangle \cdot \mathbf{z} \neq \mathbf{x} \cdot \langle \mathbf{y}, \mathbf{z} \rangle$$

- Beware of notation:

- $(\mathbf{x} \cdot \mathbf{y}) \cdot \mathbf{z} \neq \mathbf{x} \cdot (\mathbf{y} \cdot \mathbf{z})$

- (no violation of associativity: different operations)

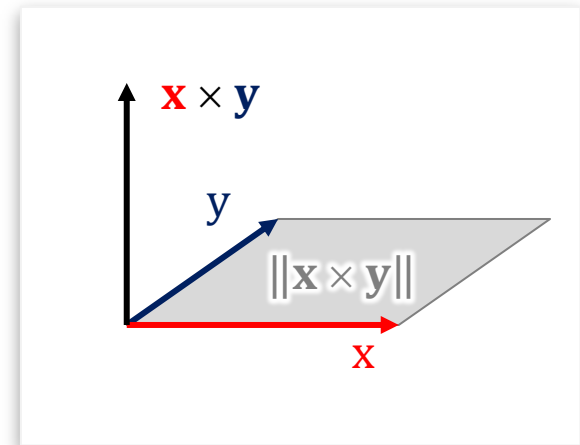
- Cross-Product: Exists Only For 3D Vectors!

- $\mathbf{x}, \mathbf{y} \in \mathbb{R}^3$

- $\mathbf{x} \times \mathbf{y} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \times \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} := \begin{pmatrix} x_2 y_3 - x_3 y_2 \\ x_3 y_1 - x_1 y_3 \\ x_1 y_2 - x_2 y_1 \end{pmatrix}$

- Geometrically: Theorem

- $\mathbf{x} \times \mathbf{y}$  orthogonal to  $\mathbf{x}, \mathbf{y}$
  - Right-handed system  $(\mathbf{x}, \mathbf{y}, \mathbf{x} \times \mathbf{y})$
  - $\|\mathbf{x} \times \mathbf{y}\| = \|\mathbf{x}\| \cdot \|\mathbf{y}\| \cdot \sin \angle(\mathbf{x}, \mathbf{y})$



- Bilinearity

- Distributive:

$$\mathbf{u} \times (\mathbf{v} + \mathbf{w}) = \mathbf{u} \times \mathbf{v} + \mathbf{u} \times \mathbf{w}$$

- Scalar-Mult.:

$$(\lambda \mathbf{u}) \times \mathbf{v} = \mathbf{u} \times (\lambda \mathbf{v}) = \lambda(\mathbf{u} \times \mathbf{v})$$

- **But beware of**

- **Anti-Commutative:**

$$\mathbf{u} \times \mathbf{v} = -\mathbf{v} \times \mathbf{u}$$

- **Not associative;**  
we can have

$$(\mathbf{u} \times \mathbf{v}) \times \mathbf{w} \neq \mathbf{u} \times (\mathbf{v} \times \mathbf{w})$$