

# Chapter 1. Vectors

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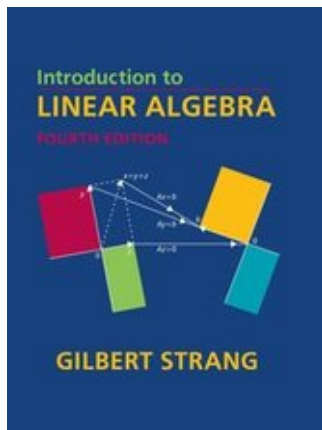


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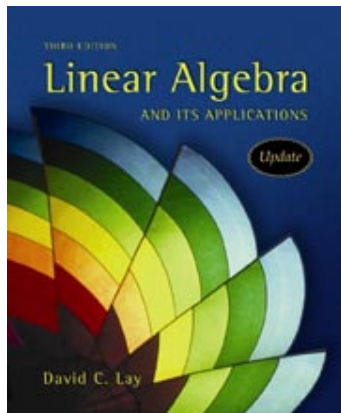
## 1 Vectors

- Vectors and basic operations (a)
- Linear combination (a)
- Inner product or dot product (b)
- Norm, vector length and unit vectors (b)
- Distances and angles (b)
- Multiplication by matrices (b)

# References



G. Strang. Introduction to linear algebra (4th ed). Wellesley Cambridge Press (2009). Chapter 1.



D. Lay. Linear algebra and its applications (3rd ed). Pearson (2006). Chapter 1.

## A little bit of history

Vectors were developed during the XIX<sup>th</sup> century by mathematicians and physicists like [Carl Friedrich Gauss](#) (1799), [William Rowan Hamilton](#) (1837), and [James Clerk Maxwell](#) (1873), mostly as a tool to represent complex numbers, and later as a tool to perform geometrical reasoning. Their modern algebra was formalized by [Josiah Willard Gibbs](#) (1901), a university professor at Yale.



To know more about the history of vectors visit

- <http://www.math.mcgill.ca/labute/courses/133f03/VectorHistory.html>
- [https://www.math.ucdavis.edu/~temple/MAT21D/SUPPLEMENTARY-ARTICLES/Crowe\\_History-of-Vectors.pdf](https://www.math.ucdavis.edu/~temple/MAT21D/SUPPLEMENTARY-ARTICLES/Crowe_History-of-Vectors.pdf)

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# What is a vector?

## Definition 1.1

*Informally, a **vector** is a collection of  $n$  numbers of the same type. We say it has  $n$  components  $(1, 2, \dots, n)$*

We'll see that this definition is terribly simplistic since many other things (like functions, infinite sequences, etc.) can be vectors. But, for the time being, let's stick to this simple definition.

## Example

$\begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \in \mathbb{Z}^3$  is a collection of 3 integer numbers

$\begin{pmatrix} -1.1 \\ 1.1 \end{pmatrix} \in \mathbb{Q}^2$  is a collection of 2 rational numbers

$\begin{pmatrix} -1.1 \\ \sqrt{2} \end{pmatrix} \in \mathbb{R}^2$  is a collection of 2 real numbers

Matlab:

```
[-1.1; sqrt(2)]
```

# Transpose

We distinguish between column vectors (for instance  $\mathbf{v}$  below) and row vectors ( $\mathbf{w}$ ). In the first case, we say  $\mathbf{v}$  is a  $n \times 1$  vector, while in the second, we say  $\mathbf{w}$  is a  $1 \times n$  vector.

$$\mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \\ \dots \\ v_n \end{pmatrix} \text{ and } \mathbf{w} = (w_1 w_2 \dots w_n).$$

## Definition 1.2

The **transpose** is the operation that transforms a column vector into a row vector and viceversa.

## Example

$$(-1 \ 1)^T = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

Matlab:

```
[-1 1]'
```



# Addition of vectors

## Definition 1.3

Given two vectors  $\mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \\ \dots \\ v_n \end{pmatrix}$  and  $\mathbf{w} = \begin{pmatrix} w_1 \\ w_2 \\ \dots \\ w_n \end{pmatrix}$  the **sum** of these two vectors

is another vector defined as  $\mathbf{v} + \mathbf{w} = \begin{pmatrix} v_1 + w_1 \\ v_2 + w_2 \\ \dots \\ v_n + w_n \end{pmatrix}$ . Note that you can only add

two column vectors or two row vectors, but not a column and a row vector.

## Example

$$\begin{pmatrix} -1.1 \\ 1.1 \end{pmatrix} + \begin{pmatrix} -1.1 \\ \sqrt{2} \end{pmatrix} = \begin{pmatrix} -2.2 \\ 1.1 + \sqrt{2} \end{pmatrix}$$

Matlab:

```
[-1.1; 1.1]+[-1.1; sqrt(2)]
```

## Properties 1.1

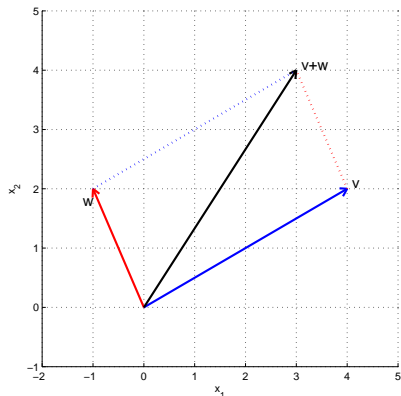
*Commutativity:*

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

# Addition of vectors

## Example

$$\begin{pmatrix} 4 \\ 2 \end{pmatrix} + \begin{pmatrix} -1 \\ 2 \end{pmatrix} = \begin{pmatrix} 3 \\ 4 \end{pmatrix}$$



# Product by scalar

## Definition 1.4

Given a vector  $\mathbf{v}$  and a scalar  $c$ , the **multiplication** of  $c$  and  $\mathbf{v}$  is defined as

$$c\mathbf{v} = \begin{pmatrix} c v_1 \\ c v_2 \\ \dots \\ c v_n \end{pmatrix}$$

## Example

$$2 \begin{pmatrix} -1.1 \\ 1.1 \end{pmatrix} = \begin{pmatrix} -2.2 \\ 2.2 \end{pmatrix}$$

$$- \begin{pmatrix} -1.1 \\ 1.1 \end{pmatrix} = \begin{pmatrix} 1.1 \\ -1.1 \end{pmatrix}$$

Matlab:

```
2*[-1.1; 1.1] -[1.1; 1.1]
```

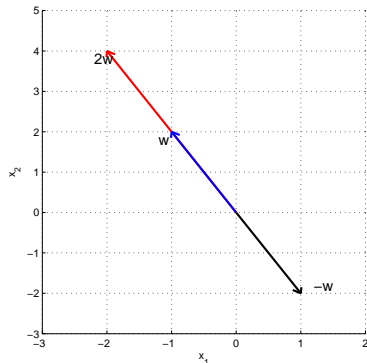
# Product by scalar

## Example

$$\mathbf{w} = \begin{pmatrix} -1 \\ 2 \end{pmatrix}$$

What is the shape of all scaled vectors of the form  $c\mathbf{w}$ ?

If  $\mathbf{w} = \mathbf{0}$ , then it is a single point ( $\mathbf{0}$ ). If  $\mathbf{w} \neq \mathbf{0}$ , then it is the straight line that passes through  $\mathbf{0}$  and  $\mathbf{w}$ .



# Properties

For simplification we will present them as properties for  $\mathbb{R}^n$ , but they apply to all vector spaces. Given any three vectors  $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathbb{R}^n$  and any two scalars  $c, d \in \mathbb{R}$ , we have

## Vector operation properties

Regarding the sum of vectors:

- 1  $\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$  Commutativity
- 2  $(\mathbf{u} + \mathbf{v}) + \mathbf{w} = \mathbf{u} + (\mathbf{v} + \mathbf{w})$  Associativity
- 3  $\mathbf{u} + \mathbf{0} = \mathbf{0} + \mathbf{u} = \mathbf{u}$  Existence of neutral element
- 4  $\mathbf{u} + -\mathbf{u} = -\mathbf{u} + \mathbf{u} = \mathbf{0}$  Existence of symmetric element

Regarding the sum of vectors and scalar product:

- 5  $c(\mathbf{u} + \mathbf{v}) = c\mathbf{v} + c\mathbf{u}$  Distributivity with respect to the sum of vectors
- 6  $(c+d)\mathbf{u} = c\mathbf{u} + d\mathbf{u}$  Distributivity with respect to the sum of scalars

Regarding the scalar product:

- 7  $c(d\mathbf{u}) = (cd)\mathbf{u}$  Associativity
- 8  $1\mathbf{u} = \mathbf{u}$  Existence of neutral element

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# Linear combination

## Definition 2.1

Given a collection of  $p$  scalars ( $x_i$ ,  $i = 1, 2, \dots, p$ ) and  $p$  vectors ( $\mathbf{v}_i$ ), the **linear combination** of the  $p$  vectors using the **weights** given by the  $p$  scalars is defined as

$$\sum_{i=1}^p x_i \mathbf{v}_i = x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \dots + x_p \mathbf{v}_p$$

## Example

$$\frac{1}{2} \begin{pmatrix} -1 \\ 1 \end{pmatrix} - \frac{2}{3} \begin{pmatrix} 2 \\ 2 \end{pmatrix} = \begin{pmatrix} -\frac{5}{6} \\ -\frac{11}{6} \end{pmatrix}$$

Matlab:

```
format rational  
-1/2*[-1; 1]-2/3*[2; 2]
```

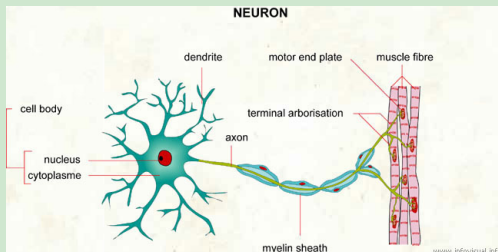
# Linear combination

## Example

A very basic model of the activity of neurons is

$$output = f\left(\sum_i weight_i input_i\right)$$

where  $f(x)$  is a non-linear function. In fact, this is the model used in artificial neuron networks.



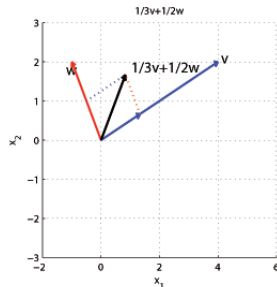
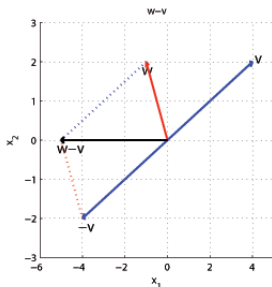
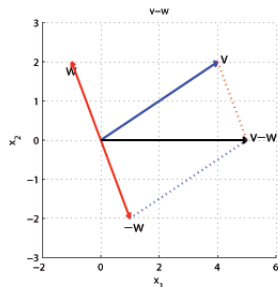
The human brain has in the order of  $10^{11}$  neurons and about  $10^{18}$  connections. See <https://www.youtube.com/watch?v=zLp-edwiGUU>.



# Linear combination

## Example

$$\mathbf{v} = \begin{pmatrix} 4 \\ 2 \end{pmatrix} \quad \mathbf{w} = \begin{pmatrix} -1 \\ 2 \end{pmatrix}$$



We may think of the weight coefficients as the “travelling” instructions. For instance, for the figure in the right, the instructions say: “Travel  $\frac{1}{3}$  of  $\mathbf{v}$  along  $\mathbf{v}$ , then travel  $\frac{1}{2}$  of  $\mathbf{w}$  along  $\mathbf{w}$ ”.

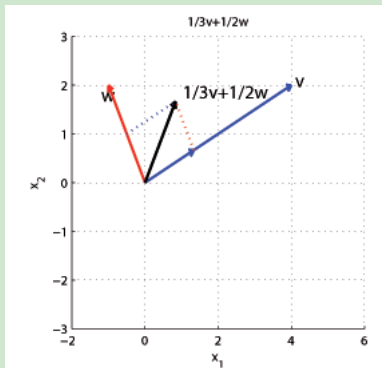
# Linear combination

## What is the shape of all linear combinations of the form $c\mathbf{v} + d\mathbf{w}$

If the two vectors are not collinear (i.e.,  $\mathbf{w} \neq k\mathbf{v}$ ), then it is the whole plane passing by  $\mathbf{0}$ ,  $\mathbf{v}$  and  $\mathbf{w}$ . We can think of it as the sum of all vectors belonging to the line  $\overline{\mathbf{0}\mathbf{v}}$  and  $\overline{\mathbf{0}\mathbf{w}}$ .

The plane generated by  $\mathbf{v}$  and  $\mathbf{w}$  is the set of all vectors that can be generated as a linear combination of both vectors.

$$\Pi = \{\mathbf{r} \mid \mathbf{r} = c\mathbf{v} + d\mathbf{w} \forall c, d \in \mathbf{R}\}$$



# Linear combination

The previous example prompts the following definition:

## Definition 2.2 (Spanned subspace)

*The subspace spanned by the vectors  $\mathbf{v}_i$ ,  $i = 1, 2, \dots, p$ , is the set of all vectors that can be expressed as the linear combination of them. Formally,*

$$\langle \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p \rangle = \text{Span} \{ \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p \} \triangleq \{ \mathbf{v} \in \mathbb{R}^n \mid \mathbf{v} = x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \dots + x_p \mathbf{v}_p \}$$

### Example

Assuming all vectors below are linearly independent:

$\text{Span} \{ \mathbf{v}_1 \}$  is a straight line.

$\text{Span} \{ \mathbf{v}_1, \mathbf{v}_2 \}$  is a plane.

$\text{Span} \{ \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1} \}$  is a hyperplane.

### Properties

$$\mathbf{0} \in \text{Span} \{ \cdot \}$$

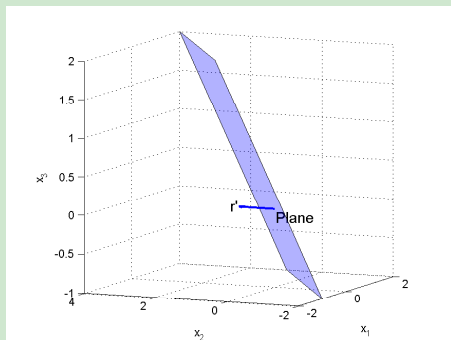
# Linear combination

## Outside the plane

Let  $\mathbf{v} = (1, 1, 0)$  and  $\mathbf{w} = (0, 1, 1)$ . The linear combinations of  $\mathbf{v}$  and  $\mathbf{w}$  fill a plane in 3D. All points belonging to this plane are of the form

$$\Pi = \{\mathbf{r} \mid \mathbf{r} = c(1, 1, 0) + d(0, 1, 1) \forall c, d \in \mathbf{R}\} = \{\mathbf{r} = (c, c + d, d) \forall c, d \in \mathbf{R}\}$$

It is clear that the vector  $\mathbf{r}' = (0, 1, 0) \notin \Pi$ , therefore, it is outside the plane.

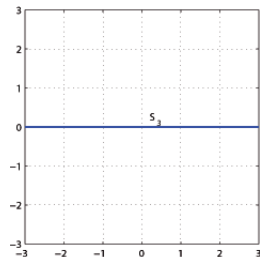
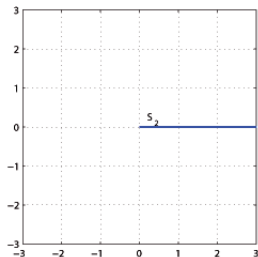
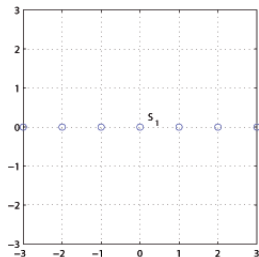


# Linear combination

## Sets of points

Let  $\mathbf{v} = (1, 0)$ .

- 1  $S_1 = \{\mathbf{r} = c\mathbf{v} \forall c \in \mathbb{Z}\}$  is a set of points
- 2  $S_2 = \{\mathbf{r} = c\mathbf{v} \forall c \in \mathbb{R}^+\}$  is a semiline
- 3  $S_3 = \{\mathbf{r} = c\mathbf{v} \forall c \in \mathbb{R}\}$  is a line

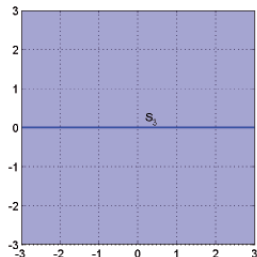
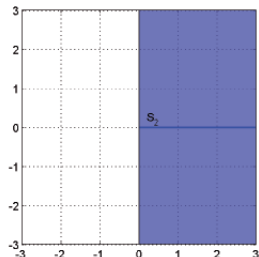
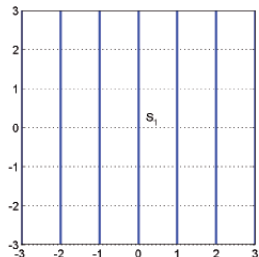


# Linear combination

## Sets of points

Let  $\mathbf{v} = (1, 0)$  and  $\mathbf{w} = (0, 1)$ .

- 1  $S_1 = \{\mathbf{r} = c\mathbf{v} + d\mathbf{w} \mid c \in \mathbb{Z}, d \in \mathbb{R}\}$  is a set of lines
- 2  $S_2 = \{\mathbf{r} = c\mathbf{v} + d\mathbf{w} \mid c \in \mathbb{R}^+, d \in \mathbb{R}\}$  is a semiplane
- 3  $S_3 = \{\mathbf{r} = c\mathbf{v} + d\mathbf{w} \mid c, d \in \mathbb{R}\}$  is a plane



# Linear combination

## Combination coefficients

Let  $\mathbf{v} = (2, -1)$ ,  $\mathbf{w} = (-1, 2)$  and  $\mathbf{b} = (1, 0)$ . Find  $c$  and  $d$  such that  $\mathbf{b} = c\mathbf{v} + d\mathbf{w}$ .

### Solution

We need to find  $c$  and  $d$  such that

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} = c \begin{pmatrix} 2 \\ -1 \end{pmatrix} + d \begin{pmatrix} -1 \\ 2 \end{pmatrix} = \begin{pmatrix} 2c - d \\ 2d - c \end{pmatrix}$$

This gives a simple equation system

$$2c - d = 1$$

$$2d - c = 0$$

whose solution is  $c = \frac{2}{3}$  and  $d = \frac{1}{3}$ . We can easily check it with Matlab:

$$2/3*[2 \ -1]' + 1/3*[-1 \ 2]'$$

## Exercises

From Lay (4th ed.), Chapter 1, Section 3:

- 1.3.1
- 1.3.3
- 1.3.6
- 1.3.7
- 1.3.25
- 1.3.27
- 1.3.29
- 1.3.31



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# Inner product

## Definition 3.1

Given two vectors  $\mathbf{v}$  and  $\mathbf{w}$  the **inner or dot product** between  $\mathbf{v}$  and  $\mathbf{w}$  is defined as

$$\langle \mathbf{v}, \mathbf{w} \rangle = \mathbf{v} \cdot \mathbf{w} \triangleq \mathbf{v}^T \mathbf{w} = \sum_{i=1}^n v_i w_i = v_1 w_1 + v_2 w_2 + \dots + v_n w_n$$

Mathematically, the concept of inner product is much more general, and this operational definition is just a particularization for vectors in  $\mathbb{R}^n$ . Although, the introduced inner product is the most common, it is not the only one that can be defined in  $\mathbb{R}^n$ . But, let's leave these generalization for the moment.

## Example

$$\begin{pmatrix} 4 \\ 2 \end{pmatrix} \cdot \begin{pmatrix} -1 \\ 2 \end{pmatrix} = 4 \cdot (-1) + 2 \cdot 2 = 0$$

Matlab:

```
dot([4; 2], [-1; 2])
```

## Properties 3.1

*Commutativity:*

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

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# Vector norm and vector length

## Definition 4.1

Given a vector  $\mathbf{v}$ , its **length** or **norm** is defined as

$$\|\mathbf{v}\| \triangleq \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle}$$

In the particular case of working with the previously introduced inner product, this definition boils down to

$$\|\mathbf{v}\| \triangleq \sqrt{\mathbf{v}^T \mathbf{v}} = \sqrt{\sum_{i=1}^n v_i^2}$$

that is known as the **Euclidean norm** of vector  $\mathbf{v}$ .

## Properties 4.1

$$\begin{aligned}\|-\mathbf{v}\| &= \|\mathbf{v}\| \\ \|c\mathbf{v}\| &= |c| \|\mathbf{v}\|\end{aligned}$$

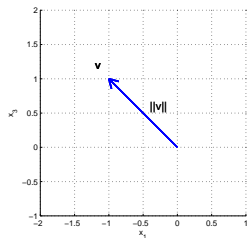
# Vector norm and vector length

## Example

$$\|(-1, 0, 1)\| = \sqrt{(-1)^2 + 0^2 + 1^2} = \sqrt{2}$$

Matlab:

```
norm([-1;0;1])
```



# Unit vectors

## Definition 4.2

$\mathbf{v}$  is **unitary** iff  $\|\mathbf{v}\| = 1$ .

## Example

$$\mathbf{e}_1 = (1, 0)$$

$$\mathbf{e}_2 = (0, 1)$$

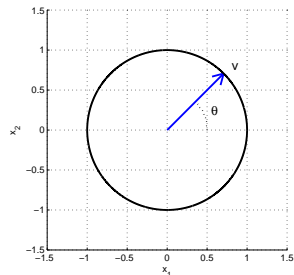
$$\mathbf{e}_\theta = (\cos(\theta), \sin(\theta))$$

Matlab:

```
theta=pi/4;
```

```
e_theta=[cos(theta);sin(theta)];
```

```
norm(e_theta)
```



# Unit vectors

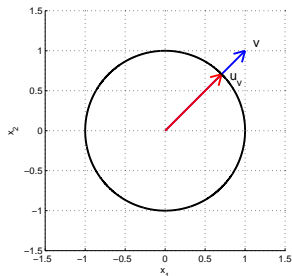
## Definition 4.3 (Construction of a unit vector)

Given any vector  $\mathbf{v}$  (whose norm is not null), we can always construct a unitary vector with the same direction of  $\mathbf{v}$  as  $\mathbf{u}_{\mathbf{v}} = \frac{\mathbf{v}}{\|\mathbf{v}\|}$ .

## Example

$$\mathbf{v} = (1, 1)$$

$$\mathbf{u}_{\mathbf{v}} = \frac{\mathbf{v}}{\|\mathbf{v}\|} = \frac{(1,1)}{\sqrt{2}} = \left( \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$$



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# Distance and angle between two vectors

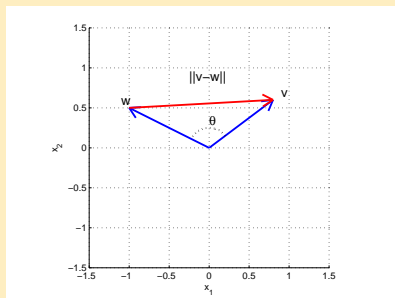
## Definition 5.1

Given two vectors  $\mathbf{v}$  and  $\mathbf{w}$ , the **distance** between both is defined as

$$d(\mathbf{v}, \mathbf{w}) \triangleq \|\mathbf{v} - \mathbf{w}\|$$

and their **angle** is

$$\angle(\mathbf{v}, \mathbf{w}) \triangleq \arccos \frac{\mathbf{v} \cdot \mathbf{w}}{\|\mathbf{v}\| \|\mathbf{w}\|} = \theta$$



## Definition 5.2

Two vectors are **orthogonal** (perpendicular) iff their inner product is 0. We then write  $\mathbf{v} \perp \mathbf{w}$ . In this case,  $\angle(\mathbf{v}, \mathbf{w}) = \frac{\pi}{2}$ .

# Distance and angle between two vectors

## Example

Let  $\mathbf{v} = (-\frac{2}{5}, \frac{2}{3})$  and  $\mathbf{w} = (1, \frac{2}{3})$ . The angle between these two vectors can be calculated as

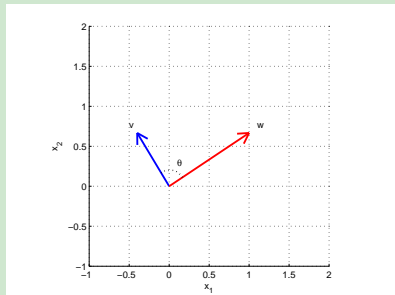
$$\mathbf{v} \cdot \mathbf{w} = (-\frac{2}{5})1 + \frac{2}{3}\frac{2}{3} = \frac{2}{45}$$

$$\|\mathbf{v}\| = \sqrt{(-\frac{2}{5})^2 + (\frac{2}{3})^2} = \frac{\sqrt{136}}{15}$$

$$\|\mathbf{w}\| = \sqrt{(1)^2 + (\frac{2}{3})^2} = \frac{\sqrt{13}}{3}$$

$$\angle(\mathbf{v}, \mathbf{w}) = \arccos \frac{\frac{2}{45}}{\frac{\sqrt{136}}{15} \frac{\sqrt{13}}{3}} = 87.27^\circ$$

$\mathbf{v}$  and  $\mathbf{w}$  are almost orthogonal.



## Example

Let  $\mathbf{v} = (1, 0, 0, 1, 0, 0, 1, 0, 0, 1)$  and  $\mathbf{w} = (0, 1, 1, 0, 1, 1, 0, 1, 1, 0)$ . These two vectors in a 10-dimensional space are orthogonal because

$$\mathbf{v} \cdot \mathbf{w} = 1 \cdot 0 + 0 \cdot 1 + 0 \cdot 1 + 1 \cdot 0 + 0 \cdot 1 + 0 \cdot 1 + 1 \cdot 0 + 0 \cdot 1 + 0 \cdot 1 + 1 \cdot 0 = 0$$

# Distance and angle between two vectors

## Example

Search for a vector that is orthogonal to  $\mathbf{v} = (-\frac{2}{5}, \frac{2}{3})$

### Solution

Let the vector  $\mathbf{w} = (w_1, w_2)$  be such a vector. Since it is orthogonal to  $\mathbf{v}$  it must meet

$$\langle \mathbf{v}, \mathbf{w} \rangle = 0 = (-\frac{2}{5})w_1 + \frac{2}{3}w_2 \Rightarrow w_2 = \frac{3}{5}w_1$$

That is, any vector of the form  $\mathbf{w} = (w_1, \frac{3}{5}w_1) = w_1(1, \frac{3}{5})$  is perpendicular to  $\mathbf{v}$ . This is the line passing by the origin and with direction  $(1, \frac{3}{5})$ . In particular, for  $w_1 = \frac{2}{3}$  we have that  $\mathbf{w} = (\frac{2}{3}, \frac{2}{5})$  and for  $w_1 = -\frac{2}{3}$  we have  $\mathbf{w} = (-\frac{2}{3}, -\frac{2}{5})$ .

This is a general rule in 2D. Given a vector  $\mathbf{v} = (a, b)$ , the vectors  $\mathbf{w} = (b, -a)$  and  $\mathbf{w} = (-b, a)$  are orthogonal to  $\mathbf{v}$ .

$$(a, b) \perp (b, -a) \text{ and } (a, b) \perp (-b, a)$$

# Distance and angle between two vectors

## Theorem 5.1 (Pythagorean theorem)

If  $\mathbf{v} \perp \mathbf{w}$ , then  $\|\mathbf{v} - \mathbf{w}\|^2 = \|\mathbf{v}\|^2 + \|\mathbf{w}\|^2$ .

*Proof*

$$\|\mathbf{v} - \mathbf{w}\|^2 = (\mathbf{v} - \mathbf{w})^T (\mathbf{v} - \mathbf{w}) = \mathbf{v}^T \mathbf{v} - \mathbf{v}^T \mathbf{w} - \mathbf{w}^T \mathbf{v} + \mathbf{w}^T \mathbf{w} = \|\mathbf{v}\|^2 + \|\mathbf{w}\|^2 - 2 \langle \mathbf{v}, \mathbf{w} \rangle$$

But, because  $\mathbf{v} \perp \mathbf{w}$ , we have  $\langle \mathbf{v}, \mathbf{w} \rangle = 0$ , and consequently

$$\|\mathbf{v} - \mathbf{w}\|^2 = \|\mathbf{v}\|^2 + \|\mathbf{w}\|^2 \text{ (q.e.d.)}$$

## Corollary 5.1

- If  $\langle \mathbf{v}, \mathbf{w} \rangle < 0$ , then  $\frac{\pi}{2} < \theta \leq \pi$ .
- If  $\langle \mathbf{v}, \mathbf{w} \rangle > 0$ , then  $0 \leq \theta < \frac{\pi}{2}$ .
- For two unit vectors,  $\mathbf{u}_1$  and  $\mathbf{u}_2$ , we have  $\cos \theta = \langle \mathbf{u}_1, \mathbf{u}_2 \rangle$ , and as a consequence  $-1 \leq \langle \mathbf{u}_1, \mathbf{u}_2 \rangle \leq 1$ .

# Distance and angle between two vectors

## Theorem 5.2 (Cosine formula)

For any two vectors,  $\mathbf{v}$  and  $\mathbf{w}$ , such that  $\|\mathbf{v}\| \neq 0$  and  $\|\mathbf{w}\| \neq 0$ , we have

$$\langle \mathbf{v}, \mathbf{w} \rangle = \|\mathbf{v}\| \|\mathbf{w}\| \cos \theta$$

### Proof

By use of Definition 4.3, we can construct the unit vectors associated to  $\mathbf{v}$  and  $\mathbf{w}$ , that is  $\mathbf{u}_\mathbf{v}$  and  $\mathbf{u}_\mathbf{w}$ . Then by Corollary 5.1 we know that

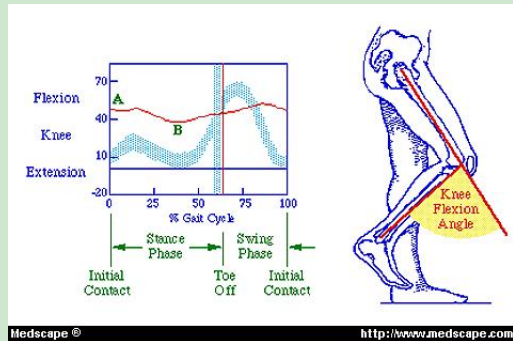
$$\cos \theta = \langle \mathbf{u}_\mathbf{v}, \mathbf{u}_\mathbf{w} \rangle = \left( \frac{\mathbf{v}}{\|\mathbf{v}\|} \right)^T \left( \frac{\mathbf{w}}{\|\mathbf{w}\|} \right) = \frac{1}{\|\mathbf{v}\| \|\mathbf{w}\|} \mathbf{v}^T \mathbf{w} = \frac{\langle \mathbf{v}, \mathbf{w} \rangle}{\|\mathbf{v}\| \|\mathbf{w}\|}$$

From this point it is trivial to deduce that  $\langle \mathbf{v}, \mathbf{w} \rangle = \|\mathbf{v}\| \|\mathbf{w}\| \cos \theta$  (q.e.d.)

# Distance and angle between two vectors

## Example

To compute the knee flexion angle, we need to calculate the dot product between the vectors aligned with the leg before and after the knee.



# Distance and angle between two vectors

## Theorem 5.3 (Cauchy-Schwarz inequality)

For any two vectors,  $\mathbf{v}$  and  $\mathbf{w}$ , it is verified that

$$|\langle \mathbf{v}, \mathbf{w} \rangle| < \|\mathbf{v}\| \|\mathbf{w}\|$$

Proof

From the cosine formula (Theorem 5.2), we know that

$$\begin{aligned} \langle \mathbf{v}, \mathbf{w} \rangle &= \|\mathbf{v}\| \|\mathbf{w}\| \cos \theta \Rightarrow \\ |\langle \mathbf{v}, \mathbf{w} \rangle| &= \|\mathbf{v}\| \|\mathbf{w}\| |\cos \theta| = \|\mathbf{v}\| \|\mathbf{w}\| |\cos \theta| \leq \|\mathbf{v}\| \|\mathbf{w}\| \end{aligned}$$

## Example

Let  $\mathbf{v} = (-\frac{2}{5}, \frac{2}{3})$  and  $\mathbf{w} = (1, \frac{2}{3})$ . We already know that  $\mathbf{v} \cdot \mathbf{w} = \frac{2}{45}$ ,  $\|\mathbf{v}\| = \frac{\sqrt{136}}{15}$ , and  $\|\mathbf{w}\| = \frac{\sqrt{13}}{3}$ . Let us check Cauchy-Schwarz inequality

$$|\frac{2}{45}| < \frac{\sqrt{136}}{15} \frac{\sqrt{13}}{3} \Leftrightarrow 0.0444 < 0.9344$$

# Distance and angle between two vectors

## Example

Show that for any two positive numbers,  $x$  and  $y$ , the geometric mean ( $\sqrt{xy}$ ) is always smaller or equal than the arithmetic mean ( $\frac{x+y}{2}$ ). For instance, the statement is verified for  $x = 2$  and  $y = 3$ :  $\sqrt{6} \leq \frac{5}{2} \Leftrightarrow 2.4495 \leq 2.5$ .

### Proof

Let there be vectors  $\mathbf{v} = (a, b)$  and  $\mathbf{w} = (b, a)$ . Then, by Cauchy-Schwarz inequality we know that

$$|\langle \mathbf{v}, \mathbf{w} \rangle| < \|\mathbf{v}\| \|\mathbf{w}\| \Rightarrow |2ab| \leq a^2 + b^2$$

Since  $x$  and  $y$  are positive numbers, we may consider them to be  $x = a^2$  and  $y = b^2$ . Consequently, we can rewrite the previous expression as

$$2\sqrt{x}\sqrt{y} \leq x + y \Rightarrow \sqrt{xy} \leq \frac{x+y}{2} \text{ (q.e.d.)}$$

In fact, the geometric mean is nothing more than the arithmetic mean in logarithmic units

$$\log(\sqrt{xy}) = \log(xy)^{\frac{1}{2}} = \frac{1}{2}(\log x + \log y) = \frac{\log x + \log y}{2}$$



# Distance and angle between two vectors

## Theorem 5.4 (Triangular inequality)

For any two vectors,  $\mathbf{v}$  and  $\mathbf{w}$ , it is verified that

$$\|\mathbf{v} + \mathbf{w}\| \leq \|\mathbf{v}\| + \|\mathbf{w}\|$$

### Proof

By definition we know that

$$\|\mathbf{v} + \mathbf{w}\|^2 = (\mathbf{v} + \mathbf{w})^T (\mathbf{v} + \mathbf{w}) = \|\mathbf{v}\|^2 + \|\mathbf{w}\|^2 + 2 \langle \mathbf{v}, \mathbf{w} \rangle$$

Applying the Cauchy-Schwarz inequality (Theorem 5.3), we have

$$\|\mathbf{v} + \mathbf{w}\|^2 \leq \|\mathbf{v}\|^2 + \|\mathbf{w}\|^2 + 2\|\mathbf{v}\|\|\mathbf{w}\| = (\|\mathbf{v}\| + \|\mathbf{w}\|)^2$$

Taking the square root we have

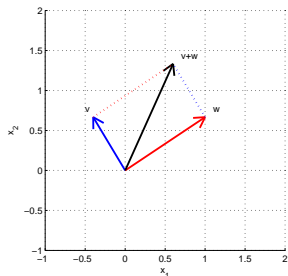
$$\|\mathbf{v} + \mathbf{w}\| \leq \|\mathbf{v}\| + \|\mathbf{w}\|$$

# Distance and angle between two vectors

## Example

Let  $\mathbf{v} = \left(-\frac{2}{5}, \frac{2}{3}\right)$  and  $\mathbf{w} = \left(1, \frac{2}{3}\right)$ . We already know that  $\|\mathbf{v}\| = \frac{\sqrt{136}}{15}$  and  $\|\mathbf{w}\| = \frac{\sqrt{13}}{3}$ . Let us check the triangular inequality

$$\begin{aligned}\mathbf{v} + \mathbf{w} &= \left(\frac{3}{5}, \frac{4}{3}\right) \Rightarrow \|\mathbf{v} + \mathbf{w}\| = \frac{\sqrt{481}}{15} \\ \frac{\sqrt{481}}{15} &\leq \frac{\sqrt{136}}{15} + \frac{\sqrt{13}}{3} \Leftrightarrow 1.4621 \leq 1.9793\end{aligned}$$



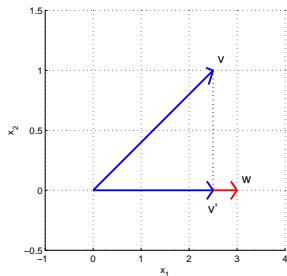
# Distance and angle between two vectors

## Orthogonal projections

Let us consider the orthogonal projection of  $\mathbf{v}$  onto  $\mathbf{w}$ .

$$\mathbf{v}' = \langle \mathbf{v}, \mathbf{w} \rangle \frac{\mathbf{w}}{\|\mathbf{w}\|^2} = \frac{\langle \mathbf{v}, \mathbf{w} \rangle}{\|\mathbf{w}\|} \frac{\mathbf{w}}{\|\mathbf{w}\|}$$

The length of this vector is  $\frac{\langle \mathbf{v}, \mathbf{w} \rangle}{\|\mathbf{w}\|}$



## Example

Let  $\mathbf{v} = (\frac{5}{2}, 1)$  and  $\mathbf{w} = (3, 0)$ . Then,  $\mathbf{v}' = \frac{\frac{5}{2} \cdot 3 + 1 \cdot 0}{3} (1, 0) = (\frac{5}{2}, 0)$ . See the figure above.

## 1 Vectors

- Vectors and basic operations (a)
- Linear combination (a)
- Inner product or dot product (b)
- Norm, vector length and unit vectors (b)
- Distances and angles (b)
- Multiplication by matrices (b)

# Multiplication by matrices

## Example

Let's consider three vectors  $\mathbf{v}_1 = \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}$ ,  $\mathbf{v}_2 = \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}$  and  $\mathbf{v}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ . Let's consider the linear combination

$$\mathbf{y} = x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + x_3\mathbf{v}_3 = x_1 \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix} + x_3 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 - x_1 \\ x_3 - x_2 \end{pmatrix}$$

I can obtain the same result by constructing a matrix

$$A = (\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3) = \begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix}.$$

And making the multiplication

$$\mathbf{y} = A \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = (\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3) \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 - x_1 \\ x_3 - x_2 \end{pmatrix}$$

# Multiplication by matrices

## Example

We can also achieve the same result by calculating  $\mathbf{y}$  as the inner product of the rows of the matrix  $A$  and the weight vector.

$$\mathbf{y} = \begin{pmatrix} \langle (1, 0, 0), (x_1, x_2, x_3) \rangle \\ \langle (-1, 1, 0), (x_1, x_2, x_3) \rangle \\ \langle (0, -1, 1), (x_1, x_2, x_3) \rangle \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 - x_1 \\ x_3 - x_2 \end{pmatrix}$$

Matlab:

```
syms x1 x2 x3
x=[x1; x2; x3]
A=[1 0 0; -1 1 0; 0 -1 1];
y=A*x
```

# Multiplication by matrices

## Matrix multiplication as a linear combination

This is a general rule: a matrix multiplication can be seen as the linear combination of the columns of the matrix.

$$A = (\mathbf{c}_1 \ \mathbf{c}_2 \ \dots \ \mathbf{c}_p) \Rightarrow \mathbf{y} = A\mathbf{x} = \sum_{i=1}^p x_i \mathbf{c}_i$$

## Matrix multiplication as inner products

Also, a matrix multiplication can be seen as the dot product of the weight vector with the rows of the matrix.

$$A = \begin{pmatrix} \mathbf{r}_1^T \\ \mathbf{r}_2^T \\ \dots \\ \mathbf{r}_n^T \end{pmatrix} \Rightarrow \mathbf{y} = A\mathbf{x} = \begin{pmatrix} \langle \mathbf{r}_1, \mathbf{x} \rangle \\ \langle \mathbf{r}_2, \mathbf{x} \rangle \\ \dots \\ \langle \mathbf{r}_n, \mathbf{x} \rangle \end{pmatrix}$$

# Multiplication by matrices

## Properties of multiplication by matrices

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

$$A(c\mathbf{u}) = c(A\mathbf{u})$$



## 1 Vectors

- Vectors and basic operations (a)
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