Linear Algebra

CS425: Computer Graphics I

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Overview

- Basic linear algebra concepts
- Coordinate systems
- Coordinate frame

Euclidian space

- A *n*-dimensional real Euclidian space is denoted \mathbb{R}^n .
- A vector v in this space is an n-tuple (an ordered list of real numbers).

$$\mathbf{v} \in \mathbb{R}^n \Longleftrightarrow \mathbf{v} = \begin{pmatrix} v_0 \\ v_1 \\ \vdots \\ v_{n-1} \end{pmatrix}$$
, with $v_i \in \mathbb{R}, i = 0, \dots, n-1$

• v_0, \dots, v_{n-1} : elements, coefficients, or components of vector **v**.

Operations

$$\mathbf{u} + \mathbf{v} = \begin{pmatrix} u_0 \\ u_1 \\ \vdots \\ u_{n-1} \end{pmatrix} + \begin{pmatrix} v_0 \\ v_1 \\ \vdots \\ v_{n-1} \end{pmatrix} = \begin{pmatrix} u_0 + v_0 \\ u_1 + v_1 \\ \vdots \\ u_{n-1} + v_{n-1} \end{pmatrix} \in \mathbb{R}$$
 Addition

$$\alpha \mathbf{u} = \begin{pmatrix} \alpha u_0 \\ \alpha u_1 \\ \vdots \\ \alpha u_{n-1} \end{pmatrix} \in \mathbb{R}$$
 Multiplication by scalar

Properties

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

ii.
$$\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$$
 (commutativity)

iii.
$$\mathbf{0} + \mathbf{v} = \mathbf{v}$$
 (zero identity)

iv.
$$\mathbf{v} + (-\mathbf{v}) = \mathbf{0}$$
 (additive inverse)

$$\mathsf{v.} \quad (ab)\mathbf{u} = a(b\mathbf{u})$$

vi.
$$(a + b)\mathbf{u} = a\mathbf{u} + b\mathbf{u}$$
 (distributive law)

vii.
$$a(\mathbf{u} + \mathbf{v}) = a\mathbf{u} + a\mathbf{v}$$
 (distributive law)

viii.
$$1\mathbf{u} = \mathbf{u}$$

Dot product

$$\mathbf{u} \cdot \mathbf{v} = \sum_{i=0}^{n-1} u_i \, v_i$$
 (dot product)

Properties:

i.
$$\mathbf{u} \cdot \mathbf{u} \ge 0$$
, with $\mathbf{u} \cdot \mathbf{u} = 0$ if and only if $\mathbf{u} = (0,0,...,0) = \mathbf{0}$

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

iii.
$$(\alpha \mathbf{u}) \cdot \mathbf{v} = \alpha (\mathbf{u} \cdot \mathbf{v})$$

iv.
$$\mathbf{u} \cdot \mathbf{v} = \mathbf{v} \cdot \mathbf{u}$$
 (commutative)

$$\mathbf{v} \cdot \mathbf{u} \cdot \mathbf{v} = 0 \Leftrightarrow \mathbf{u} \perp \mathbf{v}$$

Norm of a vector

$$\|\mathbf{u}\| = \sqrt{\mathbf{u} \cdot \mathbf{u}} = \sqrt{\sum_{i=0}^{n-1} u_i^2}$$
 (norm)

Properties:

i.
$$\|\mathbf{u}\| = 0 \Leftrightarrow \mathbf{u} = (0,0,...,0) = \mathbf{0}$$

ii.
$$\|\alpha \mathbf{u}\| = |\alpha| \|\mathbf{u}\|$$

iii.
$$||\mathbf{u} + \mathbf{v}|| \le ||\mathbf{u}|| + ||\mathbf{v}||$$
 (triangle inequality)

iv.
$$\|\mathbf{u} \cdot \mathbf{v}\| \le \|\mathbf{u}\| \|\mathbf{v}\|$$
 (Cauchy-Schawrtz inequality)

- A set of vectors $\mathbf{v_0}$, $\mathbf{v_1}$,..., $\mathbf{v_{n-1}}$ is *linearly independent* if: $\alpha_0 \mathbf{v_0} + \alpha_1 \mathbf{v_1} + \cdots + \alpha_{n-1} \mathbf{v_{n-1}} = 0$ if and only if $\alpha_0 = \alpha_1 = \cdots = \alpha_{n-1} = 0$
- If a set of vectors is linearly independent, we cannot represent one in terms of the other.
- If a set of vectors is linearly dependent, at least one can be written in terms of the other.

- Example: $\mathbf{v_0} = (4,3)$ and $\mathbf{v_1} = (8,6)$ are not linearly independent, since $\alpha_0 = 2$ and $\alpha_1 = -1$ satisfy.
- Example: $\mathbf{v_0} = (4,3)$ and $\mathbf{v_1} = (2,6)$ are linearly independent, since the only scalars to satisfy previous equation are $\alpha_0 = 0$ and $\alpha_1 = 0$.

If a set of vectors $\mathbf{v_0}, \mathbf{v_1}, ..., \mathbf{v_{n-1}} \in \mathbb{R}^n$ is linearly independent and any vector $\mathbf{u} \in \mathbb{R}^n$ can be written as a combination of:

$$\mathbf{u} = \sum_{i=0}^{n-1} u_i \mathbf{v}_i$$

Then we can say that the vectors \mathbf{v}_0 , ..., \mathbf{v}_{n-1} span the Euclidian space \mathbb{R}^n .

- In addition, if $u_0, ..., u_{n-1}$ are uniquely determined by \mathbf{u} for all $\mathbf{u} \in \mathbb{R}^n$, then $\mathbf{v}_0, ..., \mathbf{v}_{n-1}$ is called a basis of \mathbb{R}^n .
- What this means: every vector can be described **uniquely** by n scalars $(u_0, ..., u_{n-1})$, and the basis vectors $\mathbf{v}_0, ..., \mathbf{v}_{n-1}$.
- (1,0), (0,1)
 - Linearly independent? Yes.
 - Spans \mathbb{R}^2 ? Yes.
 - Basis for \mathbb{R}^2 ? Yes.
- (1,0), (1,1), (0,2)
 - · Linearly independent? No.
 - Spans \mathbb{R}^2 ? Yes.
 - Basis for \mathbb{R}^2 ? No.

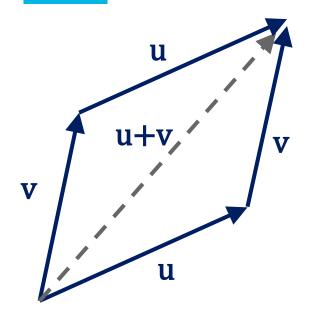
- (1,1,0), (0,1,1)
 - Linearly independent? Yes.
 - Spans \mathbb{R}^3 ? No.
 - Basis for \mathbb{R}^3 ? No.
- (1,1),(1,-1)
 - Linearly independent? Yes.
 - Spans \mathbb{R}^2 ? Yes.
 - Basis for \mathbb{R}^2 ? Yes.

Coordinate systems

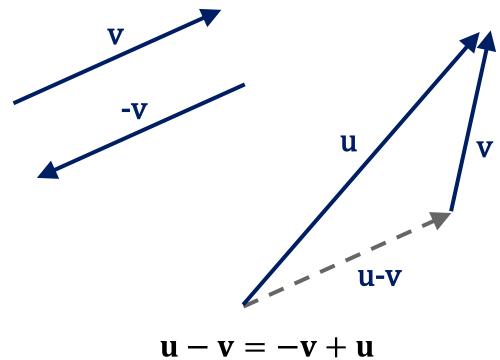
- Consider a basis $\mathbf{v}_0, \dots, \mathbf{v}_{n-1}$
- A vector can be written as $\mathbf{v} = \alpha_0 \mathbf{v}_0 + \alpha_1 \mathbf{v}_1 + \cdots + \alpha_{n-1} \mathbf{v}_{n-1}$
- List of scalars is the representation of v with respect to given basis.
- A vector describes a direction and a length (norm).
- We can write this as:

$$\mathbf{v} = (\alpha_0 \quad \alpha_1 \quad \dots \quad \alpha_{n-1})^T = \begin{pmatrix} \alpha_0 \\ \alpha_1 \\ \dots \\ \alpha_{n-1} \end{pmatrix}$$

Operations

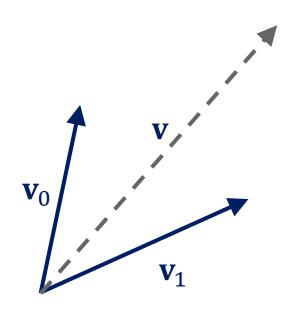


$$\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$$
 (commutativity)

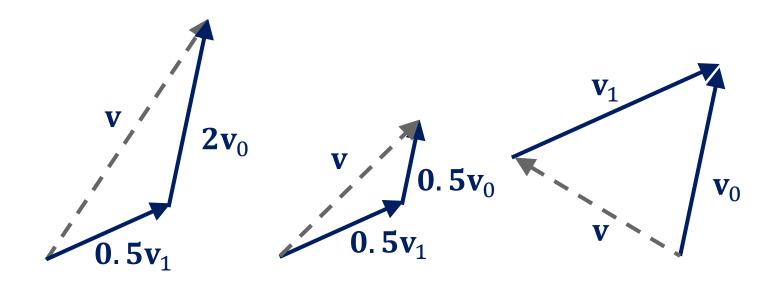


$$\mathbf{u} - \mathbf{v} = -\mathbf{v} + \mathbf{u}$$

Coordinates



 $\mathbf{v} = \alpha_0 \mathbf{v}_0 + \alpha_1 \mathbf{v}_1$ \mathbf{v}_0 and \mathbf{v}_1 form a 2D basis



$$v = 2v_0 + 0.5v_1$$
 $v = 0.5v_0 + 0.5v_1$ $v = v_0 - v_1$

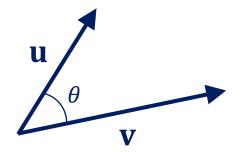
Length

- Length of a vector is denoted as ||v||
- A vector can be normalized, to change its length to 1, without affecting its direction: $\mathbf{u} = \frac{\mathbf{v}}{\|\mathbf{v}\|}$

Dot product

 The dot product is related to the length of two vectors and the angle between them.

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta$$

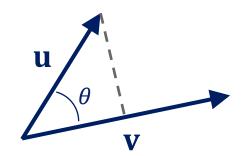


 If both are normalized, it is directly the cosine of the angle between them.

Projection

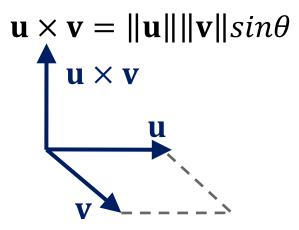
 The length of the projection of u onto v can be computed using the dot product

$$\mathbf{u} \to \mathbf{v} = \|\mathbf{u}\| cos\theta = \frac{\mathbf{u} \cdot \mathbf{v}}{\|\mathbf{v}\|}$$



Cross product

- Defined only for 3D vectors.
- The resulting vector is perpendicular to both **u** and **v**, the direction depends on the right hand rule.
- The magnitude is equal to the area of the parallelogram formed by u and v.

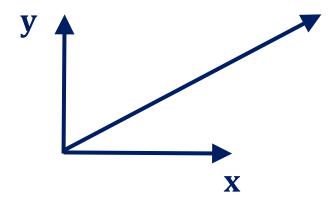


Properties

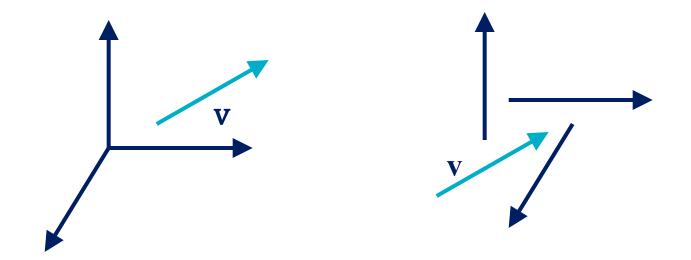
- i. $\mathbf{u} \times \mathbf{v} = -\mathbf{v} \times \mathbf{u}$ (anti-commutativity)
- ii. $(a\mathbf{u} + b\mathbf{v}) \times \mathbf{w} = a(\mathbf{u} \times \mathbf{w}) + b(\mathbf{v} \times \mathbf{w})$ (linearity)
- iii. $\mathbf{0} + \mathbf{v} = \mathbf{v}$ (zero identity)
- iv. $\mathbf{u} \times (\mathbf{v} \times \mathbf{w}) = (\mathbf{u} \cdot \mathbf{w})\mathbf{v} (\mathbf{u} \cdot \mathbf{v})\mathbf{w}$ (vector triple product)

Cartesian coordinates

• $\mathbf{x} = (1,0)$ and $\mathbf{y} = (0,1)$ form a canonical, Cartesian basis.



Coordinate systems



Which one is correct? Both, vectors don't have fixed location.

Coordinate systems

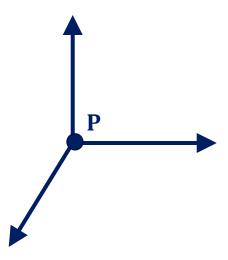
 We will always use orthonormal bases, which are formed by pairwise orthogonal unit vectors:

$$\mathbf{u} \cdot \mathbf{v} = \begin{cases} 0, i \neq j \\ 1, i = j \end{cases}$$

 This means that every basis vector must have a length of one, and also that each pair of basis vectors must be orthogonal.

Coordinate frames

- Note that a coordinate system is insufficient to represent points.
- We can add an origin to the basis vectors to form a frame.



Coordinate frames

- Frame determined by $(\mathbf{P}_0, \mathbf{v}_0, \mathbf{v}_1, \mathbf{v}_2)$.
- Within this frame, every vector can be written as:

$$\mathbf{v} = \alpha_0 \mathbf{v}_0 + \alpha_1 \mathbf{v}_1 + \dots + \alpha_{n-1} \mathbf{v}_{n-1}$$

• Within this frame, every point can be written as:

$$\mathbf{P} = \mathbf{P}_0 + \beta_0 \mathbf{v}_0 + \beta_1 \mathbf{v}_1 + \dots + \beta_{n-1} \mathbf{v}_{n-1}$$

Points and vectors

Consider the point and the vector:

$$\mathbf{v} = \alpha_0 \mathbf{v}_0 + \alpha_1 \mathbf{v}_1 + \dots + \alpha_{n-1} \mathbf{v}_{n-1}$$

$$\mathbf{P} = \mathbf{P}_0 + \beta_0 \mathbf{v}_0 + \beta_1 \mathbf{v}_1 + \dots + \beta_{n-1} \mathbf{v}_{n-1}$$

Similar representations:

$$\mathbf{v} = \begin{pmatrix} \alpha_0 \\ \alpha_1 \\ \vdots \\ \alpha_{n-1} \end{pmatrix} \quad \mathbf{P} = \begin{pmatrix} \beta_0 \\ \beta_1 \\ \vdots \\ \beta_{n-1} \end{pmatrix}$$

But: a vector has no position.

References

- Real-time Rendering, 3rd Ed. by Tomas Akenine-Möller, Eric Haines, and Naty Hoffman (Appendix A)
- Interactive Computer Graphics 7th Ed. by Ed Angel and Dave Shreiner (Chapter 3)