

Topics in Computer Graphics  
Chap 2: Introductory Material  
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University of Seoul  
School of Computer Science  
Minho Kim

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# Affine Space

- ▶ *Coordinate-free or coordinate-independent methods*  
→ “affine geometry”
- ▶ Distinction between points and vectors
  - ▶ *Points* are elements of 3D Euclidean (or point) space  $\mathbb{E}^3$ .  
→ A.k.a. “affine space”
  - ▶ *Vectors* are elements of 3D linear (or vector) space  $\mathbb{R}^3$ .
- ▶ Operations
  - ▶ Vector + vector  $\in \mathbb{R}^3$
  - ▶ Point + vector  $\in \mathbb{E}^3$
  - ▶ Point + point not allowed

# Barycentric Combinations

- ▶ A.k.a. *affine combinations*
- ▶ In general, a linear combination of points

$$\sum_{j=0}^n \alpha_j \mathbf{b}_j, \quad \mathbf{b}_j \in \mathbb{E}^3$$

is not allowed. (Why?)

- ▶ But allowed/defined when  $\sum_{j=0}^n \alpha_j = 1$ .

$$\sum_{j=0}^n \alpha_j \mathbf{b}_j, \quad \mathbf{b}_j \in \mathbb{E}^3, \quad \sum_{j=0}^n \alpha_j = 1.$$

(Why?)

$$\sum_{j=0}^n \alpha_j \mathbf{b}_j = \mathbf{b}_0 + \sum_{j=1}^n \alpha_j (\mathbf{b}_j - \mathbf{b}_0)$$

- ▶  $\mathbf{b}_0 \in \mathbb{E}^3$  and  $\mathbf{b}_j - \mathbf{b}_0 \in \mathbb{R}^3$
- ▶ Examples: centroid of a triangle, midpoint of a line, etc.

# Convex Combinations

$$\sum_{j=0}^n \alpha_j \mathbf{b}_j, \quad \mathbf{b}_j \in \mathbb{E}^3, \quad \sum_{j=0}^n \alpha_j = 1, \quad \alpha_j \geq 0 \quad \forall j.$$

- ▶ A convex combination of points is always inside of the *convex hull* of those points.
- ▶ For any two points in the set, the straight line connecting them is also contained in the set.
- ▶ Affine maps preserve convexity.

## Other Combinations

- ▶ What if the sum of coefficients is 0?  
For  $\mathbf{p}_j \in \mathbb{E}^3$ ,

$$\sum_{j=0}^n \sigma_j \mathbf{p}_j \in \mathbb{R}^3.$$

- ▶ For any form  $\mathbf{a} = \sum \beta_j \mathbf{b}_j$ , if  $\mathbf{a}$  is supposed to be a point, we must be able to split the sum into three groups:

$$\mathbf{a} = \sum_{\sum \beta_j = 1} \beta_j \mathbf{b}_j + \sum_{\sum \beta_j = 0} \beta_j \mathbf{b}_j + \sum_{\text{remaining } \beta_s} \beta_j \mathbf{b}_j$$

- ▶  $\mathbf{b}_j$ s in  $\sum_{\sum \beta_j = 1} \beta_j \mathbf{b}_j$  are points (mandatory)
- ▶  $\mathbf{b}_j$ s in  $\sum_{\sum \beta_j = 0} \beta_j \mathbf{b}_j$  are either points or vectors (optional)
- ▶  $\mathbf{b}_j$ s in  $\sum_{\text{remaining } \beta_s} \beta_j \mathbf{b}_j$  are vectors (optional)

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# Affine Maps

## Definition

A map  $\Phi$  that maps  $\mathbb{E}^3$  into itself is called an affine map if it leaves barycentric combinations invariant.

- ▶ A.k.a. affine transformation
- ▶ If

$$\mathbf{x} = \sum \alpha_j \mathbf{a}_j, \quad \sum \alpha_j = 1, \mathbf{x}, \mathbf{a}_j \in \mathbb{E}^3,$$

and  $\Phi$  is an affine map, then also

$$\Phi \mathbf{x} = \Phi \left( \sum \alpha_j \mathbf{a}_j \right) = \sum \alpha_j \Phi \mathbf{a}_j, \quad \Phi \mathbf{x}, \Phi \mathbf{a}_j \in \mathbb{E}^3.$$

- ▶ Example: The midpoint of two points will be mapped to the midpoint of the affine image of the points.

## Affine Maps (cont'd)

Any affine map is of the form

$$\Phi \mathbf{x} = A\mathbf{x} + \mathbf{v}, \quad A \in \mathbb{R}^{3 \times 3}, \mathbf{v} \in \mathbb{R}^3.$$

- ▶ Proof: Show that the form preserves a barycentric combination.
- ▶ The inverse is true as well: Every map of the form above represents an affine map.

## Affine Maps (cont'd)

- ▶ Examples: The identity, translation, scaling, rotation, shear, parallel projection
- ▶ What is the different from the linear transformations?  
→ “translation” added
- ▶ Euclidean maps (a.k.a. rigid body motions)
  - ▶ Characterized by orthonormal matrices  $A$  ( $A^T A = I$ )
  - ▶ Leaves lengths and angles unchanged
  - ▶ Rotations or translations.
- ▶ Affine maps can be composed.
- ▶ Every affine map can be composed of translations, rotations, shears, and scalings.
- ▶ *Rank* of  $A$ : dimension of the image
- ▶ An affine map from  $\mathbb{E}^2$  ( $\mathbb{E}^3$ ) to  $\mathbb{E}^2$  ( $\mathbb{E}^3$ ) is uniquely determined by a nondegenerate triangle (tetrahedron) and its image.
- ▶ Affine maps of vectors → Same as the linear map  $A$ :

$$\Phi(\mathbf{w}) = A\mathbf{w}, \quad \mathbf{w} \in \mathbb{R}^3.$$

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# Norm Ellipse

1. An ellipse with center at the origin is given by a quadratic form  $\mathbf{x}^T A \mathbf{x} = 1$ . where  $A$  is a symmetric matrix with two nonnegative eigenvalues. (Why?)
2. We're given a 2D point set  $\mathbf{p}_1, \dots, \mathbf{p}_L$  whose centroid is located at the origin.:  $\sum_{j=1}^L \mathbf{p}_j = \mathbf{0}$ .
3. If a point  $\mathbf{p}_i$  were on the ellipse defined by  $A$ , then all points would satisfy  $\mathbf{p}_i^T A \mathbf{p}_i = 1, \quad i = 1, \dots, L$ .
4. Define  $\mathbf{P} := [\mathbf{p}_1 \quad \dots \quad \mathbf{p}_L] \in \mathbb{R}^{2 \times L}$ .
5. Then  $\mathbf{P}^T A \mathbf{P} = I \in \mathbb{R}^{L \times L}$
6.  $\mathbf{P} \mathbf{P}^T A \mathbf{P} \mathbf{P}^T = \mathbf{P} \mathbf{P}^T$
7. Defining  $B := \mathbf{P} \mathbf{P}^T \in \mathbb{R}^{2 \times 2}$  and assuming it is invertible,  $A = B^{-1}$ .

## Norm Ellipse (cont'd)

8. An ellipse is uniquely defined by the points in an affinely invariant way. → “norm ellipse”
  - The axes of the ellipse defined by  $A$  represent the distribution of the points.
  - The axes are given by the eigenvectors of  $A$ .
  - The lengths of the axes are determined by the corresponding eigenvalues.
  - Application: image registration

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# Function Spaces

- ▶ Example #1:  $C[a, b]$ : the set of all real-valued continuous functions defined over the interval  $[a, b]$  of the real axis

- ▶ By defining

$$(\alpha f + \beta g)(t) = \alpha f(t) + \beta g(t),$$

$C[a, b]$  forms a *linear space* over the reals.

- ▶  $f_1, \dots, f_n \in C[a, b]$  are *linearly independent* if  $\sum c_i f_i = 0$  for all  $t \in [a, b]$  implies  $c_1 = \dots = c_n = 0$ .
- ▶ Example #2:  $C^k[a, b]$ : the set of all real-valued functions defined over  $[a, b]$  that are  $k$ -times continuously differentiable.



# Function Spaces (cont'd)

- ▶ Example #3:  $\mathcal{P}^n$ : the set of all polynomials of degree  $n$ .
  - ▶ The dimension of  $\mathcal{P}^n$  is  $n + 1$ . (Why?)
  - ▶ A basis of  $\mathcal{P}^n$  is the *monomials*  $\{1, t, t^2, \dots, t^n\}$ . (Why?)
- ▶ Example #4: Piecewise linear functions
  - ▶ Forms a linear function space.
  - ▶ Basis: *hat functions*  $H_i(t)$ 
    - any piecewise linear function  $f$  with  $f(t_j) = f_j$  can always be written as

$$f(t) = \sum_{j=0}^n f_j H_j(t).$$

- ▶ *Linear operators*
  - ▶ Assigns a function  $\mathcal{A}f$  to a given function  $f$   
 $\mathcal{A} : C[a, b] \rightarrow C[a, b]$
  - ▶  $\mathcal{A}(\alpha f + \beta g) = \alpha \mathcal{A}f + \beta \mathcal{A}g, \quad \alpha, \beta \in \mathbb{R}.$
  - ▶ Example: derivative operator