## Lecture 4-5 Transformations, Projections, and Viewing



# **1.1 Points and Vectors**

Points

A solid object with infinitely small size

- $\rightarrow$  a mathematical abstraction
- Use points to define locations, to describe trajectories of objects, and model geometric shapes...
- Vectors
  - Length (magnitude) + direction : e.g. velocity
  - Add two vectors → parallelogram rule of analytical geometry
  - Multiply a vector by a scalar  $\rightarrow$  change magnitude not direction
  - Vectors and their operations (addition and scalar multiplication) → a vector space

# **1.1 Points and Vectors**

Points and Vectors

Pick an origin o, each point p corresponds to a vector x

(since each p and o defines a length and a direction)

 $\rightarrow$  produces vectors by point difference

The correspondence depends on the origin

 $\Box y = x + \delta, \delta = q - o$ 



# 1.1 Point Space→Vector Space

- inner product (dot product): xy
- norm (length) of a vector:  $|x| = \sqrt{x \cdot x}$
- a unit vector  $\leftarrow$  length equals unity
- orthogonal vectors  $\leftarrow$  dot product is zero
- orthonormal bases (largest sets of unit vectors, pairwise orthogonal)

In such a basis  $\{e_1, e_2, \dots, e_n\}$ :

- $x = x_1 e_1 + x_2 e_2 + \dots + x_n e_n$
- the inner product:

$$x.y = X^{t}Y = x_{1}y_{1} + x_{2}y_{2} + \dots + x_{n}y_{n}$$

and the length becomes:

$$x = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

# 1.2 frame

- Points correspond to Vectors, given a fixed origin
- Vectors correspond to column matrices, given a fixed basis
- → Represent points using column matrices
- A pair (origin, basis) is called a <u>frame</u>, or <u>coordinate</u> <u>system</u>
  - For a fixed frame, points → column matrices
     (elements of the matrix are called the coordinates of
     the point in that frame)

# 1.3 cross product

 Useful especially in 3D, defined in a right-handed, orthonormal, 3-D basis:

 $\boldsymbol{x} \times \boldsymbol{y} = (x_2 y_3 - x_3 y_2) \boldsymbol{e}_1 + (x_3 y_1 - x_1 y_3) \boldsymbol{e}_2 + (x_1 y_2 - x_2 y_1) \boldsymbol{e}_3$ 



> Cross product of two parallel vectors is zero

- Otherwise, its magnitude = the area of the parallelogram, its direction is perpendicular to both vectors
- For completing a 3D orthonormal basis when two of its vectors are known

# 2. Transformations

- Moving, scaling, and deforming objects are fundamental operations in geometric modeling.
- If objects are considered as sets of points, what we need are transformations that map points onto other points

...Start with some basic transformations...

Computing Transformation is important because we want:

- To compute/represent transformations when necessary
  - □ For rendering, interactive visualization
  - Other visual computing applications
- To analyze shapes under transformation
  - Find invariant shape properties under various transformations (applications: shape descriptors for shape retrieval, object recognition, object tracking/localization...)

geometry is the study of invariants under transformations

## 2.1 Linear Transformations

- A transformation is linear if it distributes over linear combinations, i.e. T(ax+by) = aT(x)+bT(y)
- Computing Linear Transformation → changing between two bases
- For a fixed basis E, each transformation to a new basis F corresponds to a square matrix
- Solving a square matrix that maps a basis to another basis is not difficult

## 2.1 Linear Transformations

- A transformation is linear if it distributes over linear combinations, i.e. T(ax + by) = aT(x) + bT(y)
- Suppose we have two bases

$$E = \begin{bmatrix} \boldsymbol{e}_1 & \cdots & \boldsymbol{e}_n \end{bmatrix}$$
$$F = \begin{bmatrix} \boldsymbol{f}_1 & \cdots & \boldsymbol{f}_n \end{bmatrix}$$

• And we want to find the linear transformation:



- $T_{ef}(e_i) = f_i, \quad i = 1, \dots, n.$
- What is the effect of such a transformation on an arbitrary vector?

$$y = T_{ef}(x)$$
  $x = EX^{e}$ 

• Following the linearity definition:

$$\mathbf{y} = \mathbf{T}_{ef}(\mathbf{x}) = \begin{bmatrix} \mathbf{T}_{ef}(\mathbf{e}_1) & \cdots & \mathbf{T}_{ef}(\mathbf{e}_n) \end{bmatrix} X^e = \begin{bmatrix} \mathbf{f}_1 & \cdots & \mathbf{f}_n \end{bmatrix} X^e$$

# 2.1 Linear Transformations (cont.)

 $\mathbf{y} = \mathbf{T}_{ef}(\mathbf{x}) = \begin{bmatrix} \mathbf{T}_{ef}(\mathbf{e}_1) & \cdots & \mathbf{T}_{ef}(\mathbf{e}_n) \end{bmatrix} X^e = \begin{bmatrix} f_1 & \cdots & f_n \end{bmatrix} X^e$ 

 $y = \begin{bmatrix} EF_1^e & \cdots & EF_n^e \end{bmatrix} X^e = E \begin{bmatrix} F_1^e & \cdots & F_n^e \end{bmatrix} X^e$ 

• Meaning the components of y in basis E are  $M^e = \begin{bmatrix} F_1^e & \cdots & F_n^e \end{bmatrix}$  $Y^e = M^e X^e$ 

- These two equations show:
  - For a fixed basis E, <u>each transformation corresponds to</u> <u>a square matrix</u> (correspondence between linear transformation and square matrices)
  - They give us computational tools for evaluating transformation effect on a vector (simply multiply the corresponding matrix)
  - The way to construct this matrix mapping a basis to another basis is convenient

# 2.1 Linear Transformations (Example)

A Rotation Example:

- Coordinate frames are usually attached to objects, suppose we want to orient left rectangle such that it aligns with the right rectangle
- E.g. find transformation T such that
  - T(e1)=f1, T(e2)=f2





 Solve a linear system, based on elementary trigonometry

$$F_1^e = \begin{bmatrix} \cos\theta \\ \sin\theta \end{bmatrix}, \quad F_2^e = \begin{bmatrix} -\sin\theta \\ \cos\theta \end{bmatrix}$$
$$M^e = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$$

## 2.1 Specific Linear Transformations

- Several most common linear 2D transformations
  - Scaling
  - Rotation
  - Shear
  - Reflection -- scaling with negative factors
  - Orthographic projection





• Rotation in 2D:





• Shear: let one of the off-diagonal elements of the scaling transformation matrix be non-zero



# 2.1.3 Reflection and Orthographic Projection

• Reflection = scaling with negative factors



- Does not map a basis onto another basis
- Singular transformation (can't be inverted, lost all depth information along a direction)

# 2.2 Translation

- simply
  - $\begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} \delta_1 \\ \delta_2 \end{bmatrix} = \begin{bmatrix} x + \delta_1 \\ y + \delta_2 \end{bmatrix}$



 Not a linear transformation (can't be computed by 2\*2 matrix multiplication)

#### Affine Transformations

- = translations + non-degenerate linear transformations
- Rigid Transformations = translations + rotations
  - □ If a transformation preserves distance → called isometries

# 2.3 Rotation/Scaling Center

- They are about the origin
  - Rotation: the whole space rotates
  - Scaling: the object will be farther or closer to the origin depending on the scales
- To transform around the shape center, or an any given point p: (3-step Composition)
  - 1) <u>Translate p to the origin</u>
  - 2) <u>Conduct the transformation</u>
  - 3) Translate the object back





# 2.4 Homogeneous Coordinates

- Translations and linear transformations can be treated more uniformly if we introduce a different system of coordinates: homogeneous coordinates.
- Use the 2D example in the following, but can be generalized to n-D straightforwardly.

1. introduce an additional component and associate with the vector x the column matrix:

$$X^* = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} X \\ 1 \end{bmatrix}$$

 $X^* \rightarrow$  homogeneous coordinates

# 2.4 Homogeneous Coordinates (cont.)

- 2. Also add a third row and column to the linear transformation matrix  $M^* = \begin{bmatrix} a & b & 0 \\ c & d & 0 \\ 0 & 0 & 1 \end{bmatrix}$ i.e.  $M^* = \begin{bmatrix} M & 0 \\ 0 & 1 \end{bmatrix}$
- 3. Multiply this augmented matrices, and get:

$$M^*X^* = \begin{bmatrix} M & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ 1 \end{bmatrix} = \begin{bmatrix} MX \\ 1 \end{bmatrix} = \begin{bmatrix} Y \\ 1 \end{bmatrix} = Y$$

Nothing changed so far.

• When elements of the third column become non-zero:

e.g.  $M^{*} = \begin{bmatrix} 1 & 0 & a \\ 0 & 1 & b \\ 0 & 0 & 1 \end{bmatrix}$ • We get:  $Y^{*} = M^{*}X^{*} = \begin{bmatrix} 1 & 0 & a \\ 0 & 1 & b \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} x+a \\ y+b \\ 1 \end{bmatrix}$ 

# 2.4 Homogeneous Coordinates (cont.)

- Now we have uniform treatment of translation and rotation
  - only one procedure needed to process both
  - one matrix-multiplication hardware works for both
  - Also deal with projections (for displaying 3D objects, later)

## 2.4 Homogeneous Coordinates (cont.) -- Geometric Interpretation

(1) generalize the coordinates of an Euclidean point p:

Increase the dimension of the original space by 1
 The original standard Euclidean plane is at w=1



 $P^* = y$ 

# 2.4 Homogeneous Coordinates (cont.) -- Geometric Interpretation

(2) Connect p with the origin (get line L)

- Each p corresponds to one line L, any point on L differ with p by a scaling
- We can always normalize the coordinate

to 
$$\begin{bmatrix} x/w \\ y/w \\ 1 \end{bmatrix}$$



# 2.4 Homogeneous Coordinates (cont.)

-- Geometric Interpretation

- The set of all lines through the origin of our auxiliary 3D space is called the projective plane.
- The elements of the projective plane are called <u>projective points</u>. (note they are actually lines)
- Each Euclidean point p has a corresponding line L and projective point p\*
- Therefore, we can manipulate Euclidean points through operations on their projective counterparts.

(see the diagram on the right:

an affine transform  ${\sf T}$ 

= Imbed + projective transform + normalize)



## **Composition of Transformations**

With homogeneous coordinates, affine transformations and their composition can be represented by matrices and their product

#### Examples:

$$\begin{bmatrix} 1 & d_{x2} \\ 1 & d_{y2} \\ 1 \end{bmatrix} \bullet \begin{bmatrix} 1 & d_{x1} \\ 1 & d_{y1} \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & d_{x1} + d_{x2} \\ 1 & d_{y1} + d_{y2} \\ 1 \end{bmatrix} = \begin{bmatrix} s_{x2} & s_{y2} \\ s_{y2} & s_{y2} \\ 1 \end{bmatrix} \bullet \begin{bmatrix} s_{x1} & s_{x2} \\ s_{y2} & s_{y2} \\ 1 \end{bmatrix} = \begin{bmatrix} s_{x1} \cdot s_{x2} & s_{y2} \\ s_{y1} \cdot s_{y2} \\ 1 \end{bmatrix}$$

- Rigid Transformation
  - $\Box$   $\leftarrow$  product of an arbitrary sequence of rotation and translation matrices
  - preserving length, angles
- Affine transformation
  - $\square \leftarrow$  product of an arbitrary sequence of rotation, translation, scale, and shear matrices
  - preserving parallelism

## Composition of Transformations (cont.)

- □ Combine fundamental R, S, and T matrices → produce desired general affine transformation
- Gain efficiency by applying a single composed transformation, rather than a series of transformations

Examples: Rotating a house about a point  $P_1$  by an angle



#### 2D Window-To-Viewport Transformation

- We get objects (2D) represented in world-coordinate system
- We need to map them onto screen coordinates
- □ A common question:
  - Given a rectangular region in world coordinates (world-coordinate window)
  - A corresponding rectangular region in screen coordinates (viewport)
  - To find the transformation



Sometimes, we want shape-preserving: make  $s_u = s_v$