


1

CS-184: Computer Graphics

Lecture #4: 2D Transformations

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University of California, Berkeley



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Today

- 2D Transformations
 - "Primitive" Operations
 - Scale, Rotate, Shear, Flip, Translate
 - Homogenous Coordinates
 - SVD
 - Start thinking about rotations...

2

Introduction

3

- Transformation:
An operation that changes one configuration into another
- For images, shapes, *etc.*
A geometric transformation maps positions that define the object to other positions
Linear transformation means the transformation is defined by a linear function... which is what matrices are good for.

3

Some Examples

4

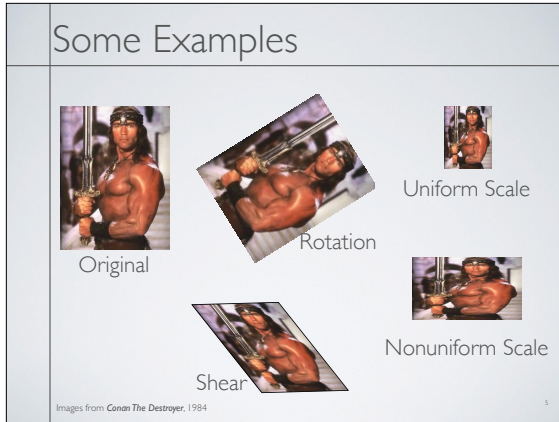


Original

4

Some Examples

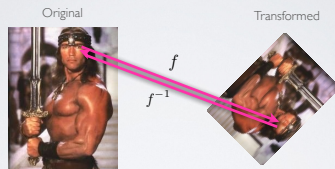
5



Mapping Function

6

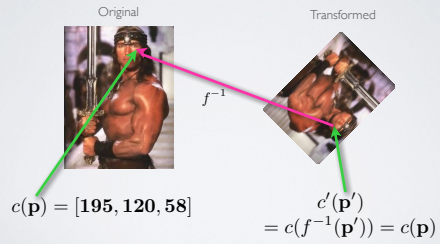
$f(\mathbf{p}) = \mathbf{p}'$ Maps points in original image $\mathbf{p} = (x, y)$ to point in transformed image $\mathbf{p}' = (x', y')$



Mapping Function

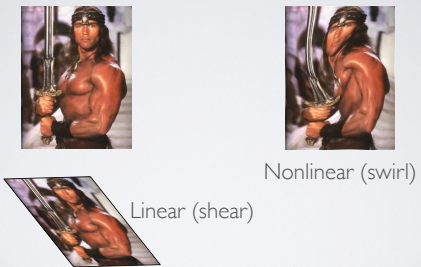
7

$f(\mathbf{p}) = \mathbf{p}'$ Maps points in original image $\mathbf{p} = (x, y)$ to point in transformed image $\mathbf{p}' = (x', y')$



Linear -vs- Nonlinear

8



Geometric -vs- Color Space

9



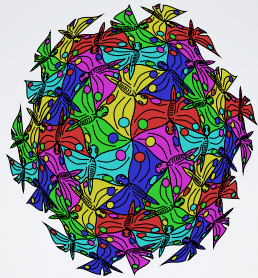
Color Space Transform
(edge finding)



Linear Geometric
(flip)

Instancing

10



M.C. Escher from Ghostscript 8.0 Distribution

Instancing

11



Carlo Sequin

11

Instancing

12

- Reuse geometric descriptions
- Saves memory

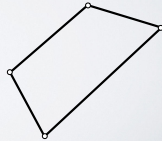


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Linear is Linear

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- Polygons defined by points
- Edges defined by interpolation between two points
- Interior defined by interpolation between all points
- *Linear interpolation*

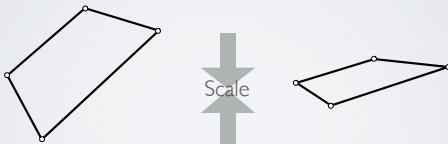


13

Linear is Linear

14

- Composing two linear function is still linear
- Transform polygon by transforming vertices



14

Linear is Linear

15

- Composing two linear function is still linear
- Transform polygon by transforming vertices

$$f(x) = a + bx \quad g(f) = c + df$$

$$g(x) = c + df(x) = c + ad + bdx$$

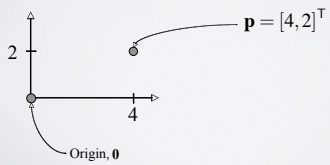
$$g(x) = a' + b'x$$

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Points in Space

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- Represent point in space by vector in \mathbf{R}^n
 - Relative to some origin!
 - Relative to some coordinate axes!
 - The choice of coordinate system is arbitrary and should be *convenient*.
- Later we'll add something extra...

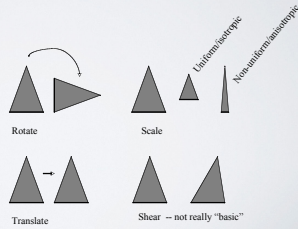


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Basic Transformations

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- Basic transforms are: rotate, scale, and translate
- Shear is a composite transformation!



Linear Functions in 2D

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$$x' = f(x, y) = c_1 + c_2x + c_3y$$

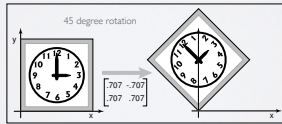
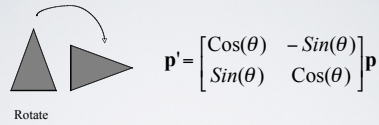
$$y' = f(x, y) = d_1 + d_2x + d_3y$$

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} t_x \\ t_y \end{bmatrix} + \begin{bmatrix} M_{xx} & M_{xy} \\ M_{yx} & M_{yy} \end{bmatrix} \cdot \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\mathbf{x}' = \mathbf{t} + \mathbf{M} \cdot \mathbf{x}$$

Rotations

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Rotations

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- Rotations are positive counter-clockwise
- Consistent w/ right-hand rule
- Don't be different...
- Note:
 - rotate by zero degrees give identity
 - rotations are modulo 360 (or 2π)

Rotations

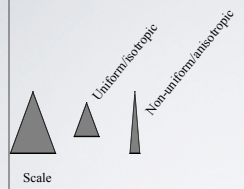
21

- Preserve lengths and distance to origin
- Rotation matrices are orthonormal
- $\text{Det}(\mathbf{R}) = 1 \neq -1$
- In 2D rotations commute...
 - But in 3D they won't!

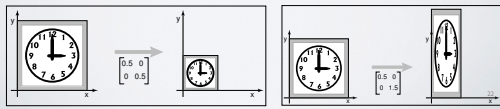
21

Scales

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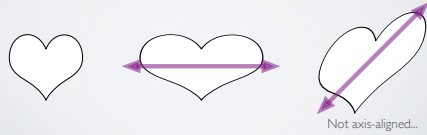
$$\mathbf{p}' = \begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \mathbf{p}$$



Scales

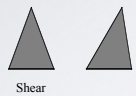
23

- Diagonal matrices
 - Diagonal parts are scale in X and scale in Y directions
 - Negative values flip
 - Two negatives make a positive (180 deg. rotation)
 - Really, axis-aligned scales

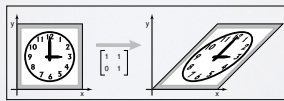


Shears

24



$$\mathbf{P}' = \begin{bmatrix} 1 & H_{yx} \\ H_{xy} & 1 \end{bmatrix} \mathbf{P}$$



Shears

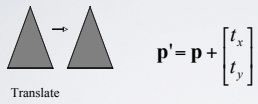
25

- Shears are not really primitive transforms
- Related to non-axis-aligned scales
- More shortly....

Translation

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- This is the not-so-useful way:



Note that its not like the others.

Arbitrary Matrices

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- For everything but translations we have:

$$\mathbf{x}' = \mathbf{A} \cdot \mathbf{x}$$

- Soon, translations will be assimilated as well
- What does an arbitrary matrix mean?

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Singular Value Decomposition

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- For any matrix, \mathbf{A} , we can write SVD:

$$\mathbf{A} = \mathbf{Q}\mathbf{S}\mathbf{R}^T$$

where \mathbf{Q} and \mathbf{R} are orthonormal and \mathbf{S} is diagonal

- Can also write Polar Decomposition

$$\mathbf{A} = \mathbf{P}\mathbf{R}\mathbf{S}\mathbf{R}^T$$

where \mathbf{P} is also orthonormal $\mathbf{P} = \mathbf{Q}\mathbf{R}^T$

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Decomposing Matrices

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- We can force **P** and **R** to have **Det=1** so they are rotations
- Any matrix is now:
 - Rotation:Rotation:Scale:Rotation
 - See, shear is just a mix of rotations and scales

Composition

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- Matrix multiplication composites matrices

$$\mathbf{p}' = \mathbf{B}\mathbf{A}\mathbf{p}$$

"Apply **A** to **p** and then apply **B** to that result."

$$\mathbf{p}' = \mathbf{B}(\mathbf{A}\mathbf{p}) = (\mathbf{B}\mathbf{A})\mathbf{p} = \mathbf{C}\mathbf{p}$$

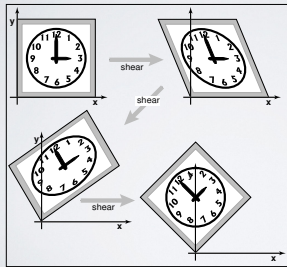
- Several translations composted to one
- Translations still left out...

$$\mathbf{p}' = \mathbf{B}(\mathbf{A}\mathbf{p} + \mathbf{t}) = \cancel{\mathbf{A}}\mathbf{p} + \mathbf{B}\mathbf{t} = \mathbf{C}\mathbf{p} + \mathbf{u}$$

Composition

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Transformations built up from others



SVD builds from scale and rotations

Also build other ways

i.e. 45 deg rotation built from shears

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Homogeneous Coordinates

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- Move to one higher dimensional space

- Append a 1 at the end of the vectors

$$\mathbf{p} = \begin{bmatrix} p_x \\ p_y \end{bmatrix} \quad \tilde{\mathbf{p}} = \begin{bmatrix} p_x \\ p_y \\ 1 \end{bmatrix}$$

- For *directions* the extra coordinate is a zero

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Homogeneous Translation

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$$\tilde{\mathbf{p}}' = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} p_x \\ p_y \\ 1 \end{bmatrix}$$

$$\tilde{\mathbf{p}}' = \tilde{\mathbf{A}}\tilde{\mathbf{p}}$$

The tildes are for clarity to distinguish homogenized from non-homogenized vectors.

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Homogeneous Others

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$$\tilde{\mathbf{A}} = \begin{bmatrix} \mathbf{A} & \mathbf{0} \\ \mathbf{0} & 1 \end{bmatrix}$$

Now everything looks the same...
Hence the term "homogenized!"

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Compositing Matrices

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- Rotations and scales always about the origin
- How to rotate/scale about another point?

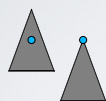


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Rotate About Arb. Point

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- Step 1: Translate point to origin



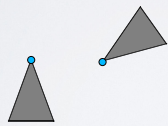
Translate (-C)

36

Rotate About Arb. Point

37

- Step 1: Translate point to origin
- Step 2: Rotate as desired



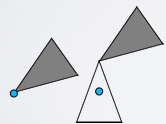
Translate (-C)
Rotate (θ)

37

Rotate About Arb. Point

38

- Step 1: Translate point to origin
- Step 2: Rotate as desired
- Step 3: Put back where it was



Translate (-C)
Rotate (θ)
Translate (C)

$$\tilde{\mathbf{p}}' = \mathbf{-T} \mathbf{R} \mathbf{T} \tilde{\mathbf{p}} = \mathbf{A} \tilde{\mathbf{p}}$$

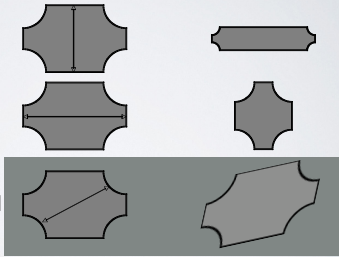
Don't negate the 1...

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Scale About Arb. Axis

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- Diagonal matrices scale about coordinate axes only;

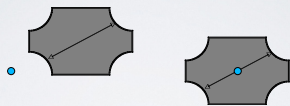


Not axis-aligned

Scale About Arb. Axis

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- Step 1: Translate axis to origin



Scale About Arb. Axis

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- Step 1: Translate axis to origin
- Step 2: Rotate axis to align with one of the coordinate axes



Scale About Arb. Axis

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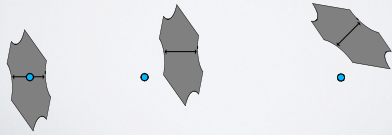
- Step 1: Translate axis to origin
- Step 2: Rotate axis to align with one of the coordinate axes
- Step 3: Scale as desired



Scale About Arb. Axis

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- Step 1: Translate axis to origin
- Step 2: Rotate axis to align with one of the coordinate axes
- Step 3: Scale as desired
- Steps 4&5: Undo 2 and 1 (reverse order)



Order Matters!

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- The order that matrices appear in matters

$$\mathbf{A} \cdot \mathbf{B} \neq \mathbf{B} \mathbf{A}$$

- Some special cases work, but they are special
- But matrices are associative

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

- Think about efficiency when you have many points to transform...

Matrix Inverses

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- In general: \mathbf{A}^{-1} undoes effect of \mathbf{A}
- Special cases:
 - Translation: negate t_x and t_y
 - Rotation: transpose
 - Scale: invert diagonal (axis-aligned scales)
- Others:
 - Invert matrix
 - Invert SVD matrices

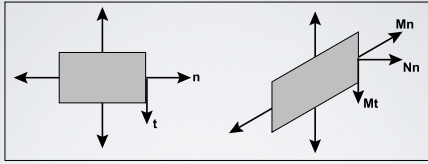
Point Vectors / Direction Vectors

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- Points in space have a 1 for the "w" coordinate
- What should we have for $\mathbf{a} - \mathbf{b}$?
 - $\mathbf{w} = 0$
 - Directions not the same as positions
 - Difference of positions is a direction
 - Position + direction is a position
 - Direction + direction is a direction
 - Position + position is nonsense

Some Things Require Care

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For example normals do not transform normally

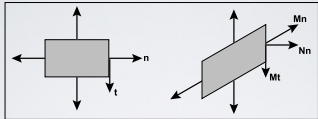
$$\mathbf{M}(\mathbf{a} \times \mathbf{b}) \neq (\mathbf{M}\mathbf{a}) \times (\mathbf{M}\mathbf{b})$$

47

Some Things Require Care

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For example normals transform abnormally



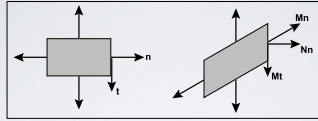
$$\mathbf{n}^T \mathbf{t} = 0 \quad \mathbf{t}_M = \mathbf{M}\mathbf{t} \quad \text{find } \mathbf{N} \text{ such that } \mathbf{n}_N^T \mathbf{t}_M = 0$$

48

Some Things Require Care

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For example normals transform abnormally



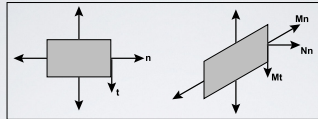
$$\mathbf{n}^T \mathbf{t} = 0 \quad \mathbf{t}_M = \mathbf{M} \mathbf{t} \quad \text{find } \mathbf{N} \text{ such that } \mathbf{n}_N^T \mathbf{t}_M = 0$$

$$\mathbf{n}^T \mathbf{t} = \mathbf{n}^T \mathbf{I} \mathbf{t} = \mathbf{n}^T \mathbf{M}^{-1} \mathbf{M} \mathbf{t} = 0$$

Some Things Require Care

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For example normals transform abnormally



$$\mathbf{n}^T \mathbf{t} = 0 \quad \mathbf{t}_M = \mathbf{M} \mathbf{t} \quad \text{find } \mathbf{N} \text{ such that } \mathbf{n}_N^T \mathbf{t}_M = 0$$

$$\mathbf{n}^T \mathbf{t} = \mathbf{n}^T \mathbf{I} \mathbf{t} = \mathbf{n}^T \mathbf{M}^{-1} \mathbf{M} \mathbf{t} = 0$$

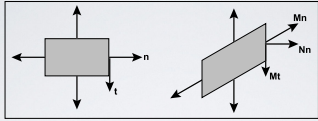
$$(\mathbf{n}^T \mathbf{M}^{-1}) \mathbf{t}_M = 0$$

$$\mathbf{n}_N^T = \mathbf{n}^T \mathbf{M}^{-1}$$

Some Things Require Care

51

For example normals transform abnormally



$$\mathbf{n}^T \mathbf{t} = 0 \quad \mathbf{t}_M = \mathbf{M} \mathbf{t} \quad \text{find } \mathbf{N} \text{ such that } \mathbf{n}_N^T \mathbf{t}_M = 0$$

$$\mathbf{n}^T \mathbf{t} = \mathbf{n}^T \mathbf{I} \mathbf{t} = \mathbf{n}^T \mathbf{M}^{-1} \mathbf{M} \mathbf{t} = 0$$

$$(\mathbf{n}^T \mathbf{M}^{-1}) \mathbf{t}_M = 0$$

$$\mathbf{n}_N^T = \mathbf{n}^T \mathbf{M}^{-1}$$

$$\mathbf{n}_N = (\mathbf{n}^T \mathbf{M}^{-1})^T$$

$$\mathbf{N} = (\mathbf{M}^{-1})^T \quad \text{See book for details}$$