## CS-184: Computer Graphics

Lecture \#4: 2D Transformations

Prof. James O'Brien University of California, Berkeley
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

## Today

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


## Introduction

## - 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.
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$


Linear -vs- Nonlinear


Nonlinear (swirl)
Linear (shear)

## Geometric -vs- Color Space



Linear Geometric
(flip)


Color Space Transform (edge finding)
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

## Instancing




## Linear is Linear

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



## Linear is Linear

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

$$
\begin{aligned}
& f(x)=a+b x \quad g(f)=c+d f \\
& g(x)=c+d f(x)=c+a d+b d x
\end{aligned}
$$

$$
g(x)=a^{\prime}+b^{\prime} x
$$

$\qquad$
$\qquad$
$\square$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\longrightarrow$
$\qquad$
$工$ $\longrightarrow$
$\qquad$
$\qquad$
$\longrightarrow$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

## Points in Space

- Represent point in space by vector in $R^{n}$
- Relative to some origin!
- Relative to some coordinate axes!
- Later we'll add something extra...



## Basic Transformations

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



## Linear Functions in 2D

$$
\begin{gathered}
x^{\prime}=f(x, y)=c_{1}+c_{2} x+c_{3} y \\
y^{\prime}=f(x, y)=d_{1}+d_{2} x+d_{3} y \\
{\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{l}
t_{x} \\
t_{y}
\end{array}\right]+\left[\begin{array}{ll}
M_{x x} & M_{x y} \\
M_{y x} & M_{y y}
\end{array}\right] \cdot\left[\begin{array}{l}
x \\
y
\end{array}\right]} \\
\mathbf{x}^{\prime}=\mathbf{t}+\mathbf{M} \cdot \mathbf{x}
\end{gathered}
$$

$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\square$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

## Rotations

- 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 $2 \pi$ )


## Rotations

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


19

## Scales

## - Diagonal matrices

- Diagonal parts are scale in X and scale in Y directions
- Negative values flip
- Two negatives make a positive (I80 deg. rotation)
- Really, axis-aligned scales
.

$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$


## Shears

$\bigwedge_{\text {Shear }} \quad \mathbf{p}^{\prime}=\left[\begin{array}{cc}1 & H_{y x} \\ H_{x y} & 1\end{array}\right] \mathbf{p}$


## Shears

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

21
$\qquad$
$\qquad$
$\longrightarrow$
$\qquad$
$\qquad$
$\qquad$
$\longrightarrow$
$\qquad$
$\qquad$
$\qquad$
$\longrightarrow$

## Translation

- This is the not-so-useful way:


Translate

Note that its not like the others.

## Arbitrary Matrices

- For everything but translations we have:

$$
\mathbf{x}^{\prime}=\mathbf{A} \cdot \mathbf{x}
$$

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


## Singular Value Decomposition

- For any matrix, A, we can write SVD:

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

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

- Can also write Polar Decomposition
$\mathbf{A}=\mathbf{Q R S R}^{\boldsymbol{\top}}$
where $\mathbf{Q}$ is still orthonormal


## Decomposing Matrices

- We can force $\mathbf{Q}$ and $\mathbf{R}$ 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
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$工$
$\qquad$
$\qquad$
$\qquad$
$\longrightarrow$
$\qquad$
$\qquad$
$\qquad$
$\qquad$


## Composition

- Matrix multiplication composites matrices

$$
\mathbf{p}^{\prime}=\mathbf{B A} \mathbf{p}
$$

"Apply A to $\mathbf{p}$ and then apply B to that result."

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

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

$$
\mathbf{p}^{\prime}=\mathbf{B}(\mathbf{A p}+\mathbf{t})=\mathbf{B A p}+\mathbf{B} \mathbf{t}=\mathbf{C} \mathbf{p}+\mathbf{u}
$$

## Composition

- Matrix multiplication composites matrices

$$
\mathbf{p}^{\prime}=\mathbf{B A p}
$$

"Apply A to $\mathbf{p}$ and then apply $\mathbf{B}$ to that result."

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

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

$$
\mathbf{p}^{\prime}=\mathbf{B}(\mathbf{A p}+\mathbf{t})=\mathbf{w}+\mathbf{B t}=\mathbf{C p}+\mathbf{u}
$$

$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
工.
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\longrightarrow$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

## Composition



Transformations built up from others

SVD builds from scale and rotations

Also build other ways
i.e. 45 deg rotation built from shears

## Homogeneous Coordiantes

- Move to one higher dimensional space
- Append a 1 at the end of the vectors

$$
\mathbf{p}=\left[\begin{array}{l}
p_{x} \\
p_{y}
\end{array}\right] \quad \widetilde{\mathbf{p}}=\left[\begin{array}{c}
p_{x} \\
p_{y} \\
1
\end{array}\right]
$$

$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\square$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

## Homogeneous Translation

$$
\begin{gathered}
\widetilde{\mathbf{p}}^{\prime}=\left[\begin{array}{lll}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{c}
p_{x} \\
p_{y} \\
1
\end{array}\right] \\
\widetilde{\mathbf{p}}^{\prime}=\widetilde{\mathbf{A}} \widetilde{\mathbf{p}}
\end{gathered}
$$

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

## Homogeneous Others

$$
\widetilde{\mathbf{A}}=\left[\right]
$$

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

## Compositing Matrices

Rotations and scales always about the origin

- How to rotate/scale about another point?
-vs-


Rotate About Arb. Point

- Step I:Translate point to origin



## Rotate About Arb. Point

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

Translate (-C)
Rotate $(\theta)$

## Rotate About Arb. Point

- Step I:Translate point to origin
- Step 2: Rotate as desired
- Step 3: Put back where it was
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$


$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\longrightarrow$
$\qquad$
$\qquad$
$\qquad$
$\longrightarrow$


## Rotate About Arb. Point

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

Translate (-C)
Rotate ( $\theta$ )


Translate (C)
$\widetilde{\mathbf{p}}^{\prime}=(-\mathbf{T}) \mathbf{R T} \widetilde{\mathbf{p}}=\mathbf{A} \widetilde{\mathbf{p}}$

## Rotate About Arb. Point

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


Scale About Arb.Axis

- Diagonal matrices scale about coordinate axes only:


36

## Scale About Arb.Axis

- Step I:Translate axis to origin

$\qquad$
$\square$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$


## Scale About Arb.Axis

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



## Scale About Arb.Axis

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

$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$


## Scale About Arb.Axis

- Step I: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 I (reverse order)

$\square$ 40


## Order Matters!

- The order that matrices appear in matters

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

- 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...
$(\mathbf{A} \cdot \mathbf{B}) \cdot \mathbf{C}=\mathbf{A} \cdot(\mathbf{B} \cdot \mathbf{C})$
$\square$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$


## Matrix Inverses

- 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

- Points in space have a 1 for the " $w$ " coordinate
- What should we have for $\mathbf{a}-\mathbf{b}$ ?
- $w=0$
- 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

Somethings Require Care


For example normals do not transform normally

$$
\begin{gathered}
\mathbf{M}(\mathbf{a} \times \mathbf{b}) \neq(\mathbf{M a}) \times(\mathbf{M b}) \\
\mathbf{M}(\mathbf{R e}) \neq \mathbf{R}(\mathbf{M e})
\end{gathered}
$$

$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

