## CS4670: Computer Vision

Noah Snavely

## Lecture 8: Geometric transformations



## Reading

- Szeliski: Chapter 3.6


## Announcements

- Project 2 out today, due Oct. 4
- (demo at end of class today)


## Image alignment



Why don't these image line up exactly?

## What is the geometric relationship between these two images?



Answer: Similarity transformation (translation, rotation, uniform scale)

What is the geometric relationship between these two images?


## What is the geometric relationship between these two images?



Very important for creating mosaics!

## Image Warping

- image filtering: change range of image

$$
\text { - } g(x)=h(f(x))
$$





- image warping: change domain of image
- $g(x)=f(h(x))$




## Image Warping

- image filtering: change range of image

$$
\cdot g(x)=h(f(x))
$$



- image warping: change domain of image



## Parametric (global) warping

- Examples of parametric warps:

aspect


## Parametric (global) warping



$$
p=(x, y)
$$

$$
\mathbf{p}^{\prime}=\left(x^{\prime}, y^{\prime}\right)
$$

- Transformation $T$ is a coordinate-changing machine:

$$
\mathrm{p}^{\prime}=T(\mathrm{p})
$$

- What does it mean that $T$ is global?
- Is the same for any point $p$
- can be described by just a few numbers (parameters)
- Let's consider linear xforms (can be represented by a 2D matrix):

$$
\mathbf{p}^{\prime}=\mathbf{T} \mathbf{p} \quad\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\mathbf{T}\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

## Common linear transformations

- Uniform scaling by $s$ :


$$
\mathbf{S}=\left[\begin{array}{ll}
s & 0 \\
0 & s
\end{array}\right]
$$

What is the inverse?

## Common linear transformations

- Rotation by angle $\theta$ (about the origin)


$$
\mathbf{R}=\left[\begin{array}{cc}
\cos \theta & -\sin \theta \\
\sin \theta & \cos \theta
\end{array}\right]
$$

What is the inverse?
For rotations:

$$
\mathbf{R}^{-1}=\mathbf{R}^{T}
$$

## 2x2 Matrices

- What types of transformations can be represented with a $2 \times 2$ matrix?
2D mirror about Y axis?

$$
\begin{aligned}
& x^{\prime}=-x \\
& y^{\prime}=y
\end{aligned} \quad \mathbf{T}=\left[\begin{array}{cc}
-1 & 0 \\
0 & 1
\end{array}\right]
$$

2D mirror across line $y=x$ ?

$$
\begin{aligned}
x^{\prime} & =y \\
y^{\prime} & =x
\end{aligned} \quad \mathbf{T}=\left[\begin{array}{ll}
0 & 1 \\
1 & 0
\end{array}\right]
$$

## 2x2 Matrices

- What types of transformations can be represented with a $2 \times 2$ matrix?

2D Translation?

$$
\begin{aligned}
x^{\prime} & =x+t_{x} \\
y^{\prime} & =y+t_{y}
\end{aligned}
$$

NO!

## All 2D Linear Transformations

- Linear transformations are combinations of ...
- Scale,
- Rotation,
- Shear, and

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{ll}
a & b \\
c & d
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

- Mirror
- Properties of linear transformations:
- Origin maps to origin
- Lines map to lines
- Parallel lines remain parallel
- Ratios are preserved
- Closed under composition

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{ll}
a & b \\
c & d
\end{array}\right]\left[\begin{array}{ll}
e & f \\
g & h
\end{array}\right]\left[\begin{array}{ll}
i & j \\
k & l
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

## Homogeneous coordinates

Trick: add one more coordinate:

$$
(x, y) \Rightarrow\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

homogeneous image coordinates


Converting from homogeneous coordinates

$$
\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right] \Rightarrow(x / w, y / w)
$$

## Translation

- Solution: homogeneous coordinates to the rescue

$$
\begin{aligned}
& \mathbf{T}=\left[\begin{array}{lll}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right] \\
& {\left[\begin{array}{ccc}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]=\left[\begin{array}{c}
x+t_{x} \\
y+t_{y} \\
1
\end{array}\right]}
\end{aligned}
$$

## Affine transformations

$$
\mathbf{T}=\left[\begin{array}{ccc}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]
$$

any transformation with
last row [ 0001 ] we call an
affine transformation

$$
\left[\begin{array}{lll}
a & b & c \\
d & e & f \\
0 & 0 & 1
\end{array}\right]
$$

## Basic affine transformations

$$
\begin{gathered}
{\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{lll}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]} \\
\text { Translate }
\end{gathered}
$$

$$
\begin{gathered}
{\left[\begin{array}{c}
\boldsymbol{x}^{\prime} \\
\boldsymbol{y}^{\prime} \\
1
\end{array}\right]=} \\
\text { Scale }
\end{gathered}
$$

$$
\begin{aligned}
{\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right] } & =\left[\begin{array}{ccc}
\cos \theta & -\sin \theta & 0 \\
\sin \theta & \cos \theta & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right] \\
& \text { 2D in-plane rotation }
\end{aligned}
$$

$$
\begin{gathered}
{\left[\begin{array}{l}
\boldsymbol{x}^{\prime} \\
\boldsymbol{y}^{\prime} \\
1
\end{array}\right]=} \\
\text { Shear }
\end{gathered} \frac{\left[\begin{array}{ccc}
1 & \boldsymbol{s} \boldsymbol{h}_{\boldsymbol{x}} & 0 \\
\boldsymbol{s} \boldsymbol{h}_{\boldsymbol{y}} & 1 & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
\boldsymbol{x} \\
\boldsymbol{y} \\
1
\end{array}\right]}{}
$$

## Affine Transformations

- Affine transformations are combinations of ...
- Linear transformations, and
- Translations

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
w
\end{array}\right]=\left[\begin{array}{lll}
a & b & c \\
d & e & f \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right]
$$

- Properties of affine transformations:
- Origin does not necessarily map to origin
- Lines map to lines
- Parallel lines remain parallel
- Ratios are preserved
- Closed under composition


## Is this an affine transformation?



## Where do we go from here?

$\underset{\text { affine transformation }}{\left[\begin{array}{lll}a & b & c \\ d & e & f \\ 0 & 0 & 1\end{array}\right]} \underset{\substack{\text { what happens when we } \\ \text { mess with this row? }}}{\left[\begin{array}{l}\text { when } \\ \hline\end{array}\right]}$

## Projective Transformations aka

## Homographies aka Planar Perspective Maps

$$
\mathbf{H}=\left[\begin{array}{lll}
a & b & c \\
d & e & f \\
g & h & 1
\end{array}\right]
$$

Called a homography (or planar perspective map)


## Homographies

- Example on board


## Image warping with homographies



## Homographies



## Projective Transformations

- Projective transformations ...
- Affine transformations, and
- Projective warps

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right]=\left[\begin{array}{lll}
a & b & c \\
d & e & f \\
g & h & i
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right]
$$

- Properties of projective transformations:
- Origin does not necessarily map to origin
- Lines map to lines
- Parallel lines do not necessarily remain parallel
- Ratios are not preserved
- Closed under composition


## 2D image transformations



| Name | Matrix | \# D.O.F. | Preserves: | Icon |
| :--- | :---: | :---: | :--- | :---: |
| translation | $[\boldsymbol{I} \mid \boldsymbol{t}]_{2 \times 3}$ | 2 | orientation $+\cdots$ | $\square$ |
| rigid (Euclidean) | $[\boldsymbol{R} \mid \boldsymbol{t}]_{2 \times 3}$ | 3 | lengths $+\cdots$ | $\square$ |
| similarity | $[s \boldsymbol{R} \mid \boldsymbol{t}]_{2 \times 3}$ | 4 | angles $+\cdots$ | $\square$ |
| affine | $[\boldsymbol{A}]_{2 \times 3}$ | 6 | parallelism $+\cdots$ | $\square$ |
| projective | $[\tilde{\boldsymbol{H}}]_{3 \times 3}$ | 8 | straight lines | $\square$ |

These transformations are a nested set of groups

- Closed under composition and inverse is a member


## Image Warping

- Given a coordinate xform $\left(x^{\prime}, y^{\prime}\right)=\boldsymbol{T}(x, y)$ and a source image $f(x, y)$, how do we compute an xformed image $g\left(x^{\prime}, y^{\prime}\right)=f(T(x, y))$ ?



## Forward Warping

- Send each pixel $f(x)$ to its corresponding location $\left(x^{\prime}, y^{\prime}\right)=\boldsymbol{T}(x, y)$ in $g\left(x^{\prime}, y^{\prime}\right)$
- What if pixel lands "between" two pixels?



## Forward Warping

- Send each pixel $f(x, y)$ to its corresponding location $\boldsymbol{x}^{\prime}=\boldsymbol{h}(\boldsymbol{x}, \boldsymbol{y})$ in $\boldsymbol{g}\left(\boldsymbol{x}^{\prime}, \boldsymbol{y}^{\prime}\right)$
- What if pixel lands "between" two pixels?
- Answer: add "contribution" to several pixels, normalize later (splatting)
- Can still result in holes



## Inverse Warping

- Get each pixel $g\left(x^{\prime}, y^{\prime}\right)$ from its corresponding location $(x, y)=T^{-1}(x, y)$ in $f(x, y)$
- Requires taking the inverse of the transform
- What if pixel comes from "between" two pixels?



## Inverse Warping

- Get each pixel $\boldsymbol{g}\left(\boldsymbol{x}^{\prime}\right)$ from its corresponding location $\boldsymbol{x}^{\prime}=\boldsymbol{h}(\boldsymbol{x})$ in $\boldsymbol{f}(\boldsymbol{x})$
- What if pixel comes from "between" two pixels?
- Answer: resample color value from interpolated (prefiltered) source image



## Interpolation

- Possible interpolation filters:
- nearest neighbor
- bilinear
- bicubic (interpolating)
- sinc
- Needed to prevent "jaggies" and "texture crawl"
(with prefiltering)

