

16-385 Computer Vision
http://www.cs.cmu.edu/~16385/ Spring 2019, Lecture 7

## Course announcements

- Homework 2 is posted on the course website.
- It is due on February $27^{\text {th }}$ at 23:59 pm.
- Start early because it is much larger and more difficult than homework 1.
- Note the updated deadline - homeworks are now back in sync with lectures, but homework 4 will be turned into a single-week homework.


## Overview of today's lecture

- Reminder: image transformations.
- 2D transformations.
- Projective geometry 101.
- Transformations in projective geometry.
- Classification of 2D transformations.
- Determining unknown 2D transformations.
- Determining unknown image warps.


## Slide credits

Most of these slides were adapted from:

- Kris Kitani (16-385, Spring 2017).

Reminder: image transformations

## What is an image?

$$
f(\boldsymbol{x})
$$

grayscale image

What is the range of the image function $f$ ?
 the image function?


$$
\text { domain } \boldsymbol{x}=\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

A (grayscale) image is a 2D function.

## What types of image transformations can we do?



## What types of image transformations can we do?



## Warping example: feature matching



## Warping example: feature matching



## Warping example: feature matching



- object recognition
- 3D reconstruction
- augmented reality
- image stitching

How do you compute the transformation?

## Warping example: feature matching

Given a set of matched feature points:

and a transformation:

find the best estimate of the parameters

## 2D transformations

## 2D transformations


translation

affine

rotation

perspective

aspect

cylindrical

## 2D planar transformations

$y$


## 2D planar transformations



How would you implement scaling?

- Each component multiplied by a scalar
- Uniform scaling - same scalar for each component


## 2D planar transformations

$$
\begin{gathered}
x^{\prime}=a x \\
y^{\prime}=b y
\end{gathered}
$$

Scale
What's the effect of using different scale factors?

- Each component multiplied by a scalar
- Uniform scaling - same scalar for each component


## 2D planar transformations

$$
\begin{aligned}
x^{\prime} & =a x \\
y^{\prime} & =b y
\end{aligned}
$$

matrix representation of scaling:
Scale

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\underbrace{\left[\begin{array}{ll}
a & 0 \\
0 & b
\end{array}\right]}_{\text {scaling matrix S }}\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

- Each component multiplied by a scalar
- Uniform scaling - same scalar for each component


## 2D planar transformations

$y$


How would you implement shearing?

## 2D planar transformations

$$
\begin{aligned}
x^{\prime} & =x+a \cdot y \\
y^{\prime} & =b \cdot x+y
\end{aligned}
$$

Shear
or in matrix form:

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{ll}
1 & a \\
b & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

## 2D planar transformations

$y$


How would you implement rotation?
rotation around the origin

$$
\boldsymbol{x}=\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

## 2D planar transformations

$y$


## 2D planar transformations

## Polar coordinates...



$$
\begin{aligned}
& x=r \cos (\phi) \\
& y=r \sin (\phi) \\
& x^{\prime}=r \cos (\phi+\theta) \\
& y^{\prime}=r \sin (\phi+\theta)
\end{aligned}
$$

Trigonometric Identity...
$x^{\prime}=r \cos (\phi) \cos (\theta)-r \sin (\phi) \sin (\theta)$
$y^{\prime}=r \sin (\phi) \cos (\theta)+r \cos (\phi) \sin (\theta)$

Substitute...
$x^{\prime}=x \cos (\theta)-y \sin (\theta)$
$y^{\prime}=x \sin (\theta)+y \cos (\theta)$

## 2D planar transformations

$y$

$$
\begin{gathered}
x^{\prime}=\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right] \quad \begin{array}{l}
x^{\prime}=x \cos \theta-y \sin \theta \\
y^{\prime}=x \sin \theta+y \cos \theta \\
\text { or in matrix form: }
\end{array} \\
{\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{cc}
\cos \theta & -\sin \theta \\
\sin \theta & \cos \theta
\end{array}\right]\left[\begin{array}{l}
x \\
\text { rotation around } \\
\text { the origin }
\end{array}\right]}
\end{gathered}
$$

## 2D planar and linear transformations

$$
\begin{aligned}
& x^{\prime}=f(x ; p) \\
& \downarrow \\
& {\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=M\left[\begin{array}{l}
x \\
y
\end{array}\right]} \\
& \text { parameters } p
\end{aligned}
$$

## 2D planar and linear transformations

Scale
$\mathbf{M}=\left[\begin{array}{cc}s_{x} & 0 \\ 0 & s_{y}\end{array}\right]$
Rotate
$\mathbf{M}=\left[\begin{array}{cc}\cos \theta & -\sin \theta \\ \sin \theta & \cos \theta\end{array}\right]$

Shear

$$
\mathbf{M}=\left[\begin{array}{cc}
1 & s_{x} \\
s_{y} & 1
\end{array}\right]
$$

## 2D translation

$y$ $\xrightarrow{\text { How would you implement translater }}$

## 2D translation

$y$

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

What about matrix representation?

$$
\mathbf{M}=\left[\begin{array}{ll}
? & ? \\
? & ?
\end{array}\right]
$$

## 2D translation

$y$

Not possible.

## Projective geometry 101

## Homogeneous coordinates

heterogeneous homogeneous coordinates coordinates



- Represent 2D point with a 3D vector


## Homogeneous coordinates

heterogeneous homogeneous
coordinates coordinates

## $\left[\begin{array}{l}x \\ y\end{array}\right] \Rightarrow\left[\begin{array}{l}x \\ y \\ 1\end{array}\right] \stackrel{\text { def }}{=}\left[\begin{array}{c}a x \\ a y \\ a\end{array}\right]$

- Represent 2D point with a 3D vector
- 3D vectors are only defined up to scale


## 2D translation

$y$


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

What about matrix representation using homogeneous coordinates?

## 2D translation

$y$


## 2D translation using homogeneous coordinates

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{c}
x \\
y \\
1
\end{array}\right]=\left[\begin{array}{c}
x+t_{x} \\
y+t_{y} \\
1
\end{array}\right]
$$



## Homogeneous coordinates

Conversion:

- heterogeneous $\rightarrow$ homogeneous

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

- homogeneous $\rightarrow$ heterogeneous

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

- point at infinity


## Special points:

$$
\left[\begin{array}{lll}
x & y & 0
\end{array}\right]
$$

- undefined

$$
\left[\begin{array}{lll}
0 & 0 & 0
\end{array}\right]
$$

- scale invariance

$$
\left[\begin{array}{lll}
x & y & w
\end{array}\right]^{\top}=\lambda\left[\begin{array}{lll}
x & y & w
\end{array}\right]^{\top}
$$

## Projective geometry



What does scaling X correspond to?

Transformations in projective geometry

## 2D transformations in heterogeneous coordinates

Re-write these transformations as $3 \times 3$ matrices:

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\underbrace{\left[\begin{array}{l}
?
\end{array}\right]}_{\text {scaling }}\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{l}
?
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

rotation
shearing

$$
\begin{aligned}
& {\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=[ } \\
& \text { lid }
\end{aligned}
$$

## 2D transformations in heterogeneous coordinates

Re-write these transformations as $3 \times 3$ matrices:

$$
\left[\begin{array}{c}
{\left[\begin{array}{c}
\boldsymbol{x}^{\prime} \\
\boldsymbol{y}^{\prime} \\
1
\end{array}\right]} \\
\qquad \begin{array}{ccc}
{\left[\begin{array}{ccc}
\boldsymbol{s}_{\boldsymbol{x}} & 0 & 0 \\
0 & \boldsymbol{s}_{\boldsymbol{y}} & 0 \\
0 & 0 & 1
\end{array}\right]}
\end{array}\left[\begin{array}{l}
\boldsymbol{x} \\
\boldsymbol{y} \\
1
\end{array}\right] \\
\text { scaling }
\end{array}\right.
$$

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

rotation
I

shearing

## 2D transformations in heterogeneous coordinates

Re-write these transformations as $3 \times 3$ matrices:

$$
\left[\begin{array}{c}
{\left[\begin{array}{c}
\boldsymbol{x}^{\prime} \\
\boldsymbol{y}^{\prime} \\
1
\end{array}\right]} \\
\qquad \begin{array}{ccc}
{\left[\begin{array}{ccc}
\boldsymbol{s}_{\boldsymbol{x}} & 0 & 0 \\
0 & \boldsymbol{s}_{\boldsymbol{y}} & 0 \\
0 & 0 & 1
\end{array}\right]}
\end{array}\left[\begin{array}{l}
\boldsymbol{x} \\
\boldsymbol{y} \\
1
\end{array}\right] \\
\text { scaling }
\end{array}\right.
$$

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

$$
]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

rotation

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
1 & \beta_{x} & 0 \\
\beta_{y} & 1 & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

shearing

## 2D transformations in heterogeneous coordinates

Re-write these transformations as $3 \times 3$ matrices:

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

$$
\left[\begin{array}{c}
{\left[\begin{array}{c}
\boldsymbol{x}^{\prime} \\
\boldsymbol{y}^{\prime} \\
1
\end{array}\right]}
\end{array}=\underset{\text { scaling }}{\left[\begin{array}{ccc}
\boldsymbol{s}_{\boldsymbol{x}} & 0 & 0 \\
0 & \boldsymbol{s}_{\boldsymbol{y}} & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
\boldsymbol{x} \\
\boldsymbol{y} \\
1
\end{array}\right]}\right.
$$

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\underbrace{\left[\begin{array}{ccc}
\cos \Theta & -\sin \Theta & 0 \\
\sin \Theta & \cos \Theta & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]}_{\text {rotation }}
$$

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\underset{\text { shearing }}{\left[\begin{array}{ccc}
1 & \beta_{x} & 0 \\
\beta_{y} & 1 & 0 \\
0 & 0 & 1
\end{array}\right]}\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

## Matrix composition

## Transformations can be combined by matrix multiplication:

$$
\begin{aligned}
{\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right] } & =\left(\left[\begin{array}{lll}
1 & 0 & t x \\
0 & 1 & t y \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{ccc}
\cos \Theta & -\sin \Theta & 0 \\
\sin \Theta & \cos \Theta & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{ccc}
s x & 0 & 0 \\
0 & s y & 0 \\
0 & 0 & 1
\end{array}\right]\right) \\
\mathrm{p}^{\prime} & =?
\end{aligned}\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right]
$$

## Matrix composition

Transformations can be combined by matrix multiplication:

$$
\begin{aligned}
{\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right] } & =\left(\left[\begin{array}{lll}
1 & 0 & t x \\
0 & 1 & t y \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{ccc}
\cos \Theta & -\sin \Theta & 0 \\
\sin \Theta & \cos \Theta & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{ccc}
s x & 0 & 0 \\
0 & s y & 0 \\
0 & 0 & 1
\end{array}\right]\right)\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right] \\
\mathbf{p}^{\prime} & =\operatorname{translation}\left(\mathrm{t}_{x}, \mathrm{t}_{y}\right) \quad \operatorname{rotation}(\theta)
\end{aligned}
$$

## Classification of 2D transformations

## Classification of 2D transformations



## Classification of 2D transformations

| Name | Matrix | \# D.O.F. |
| :--- | :---: | :---: |
| translation | $[\boldsymbol{I} \mid \boldsymbol{t}]$ | $?$ |
| rigid (Euclidean) | $[\boldsymbol{R} \mid \boldsymbol{t}]$ | $?$ |
| similarity | $[s \boldsymbol{R} \mid t]$ | $?$ |
| affine | $[\boldsymbol{A}]$ | $?$ |
| projective | $[\tilde{\boldsymbol{H}}]$ | $?$ |

# Classification of 2D transformations 

Translation: $\left[\begin{array}{ccc}1 & 0 & t_{1} \\ 0 & 1 & t_{2} \\ 0 & 0 & 1\end{array}\right]$

How many degrees of freedom?


## Classification of 2D transformations



Are there any values that are related?


## Classification of 2D transformations



How many degrees of freedom?


## Classification of 2D transformations



## Classification of 2D transformations



## Classification of 2D transformations



## Classification of 2D transformations



Are there any values that are related?


## Classification of 2D transformations

multiply these four by scale s

Similarity: uniform scaling + rotation

+ translation


How many degrees of freedom?


## Classification of 2D transformations



## Classification of 2D transformations



Are there any values that are related?


## Classification of 2D transformations



Are there any values that are related?


## Classification of 2D transformations



How many degrees of freedom?


## Affine transformations

Affine transformations are combinations of

- arbitrary (4-DOF) linear transformations; and
- translations

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

Properties of affine transformations:

- origin does not necessarily map to origin
- lines map to lines
- parallel lines map to parallel lines

- ratios are preserved
- compositions of affine transforms are also affine transforms


## Affine transformations

Affine transformations are combinations of

- arbitrary (4-DOF) linear transformations; and
- translations

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

Properties of affine transformations:

- origin does not necessarily map to origin
- lines map to lines
- parallel lines map to parallel lines

- ratios are preserved
- compositions of affine transforms are also affine transforms


## How to interpret affine transformations here?



## Projective transformations

Projective transformations are combinations of

- affine transformations; and
- projective wraps

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

How many degrees of freedom?
Properties of projective transformations:

- origin does not necessarily map to origin
- lines map to lines
- parallel lines do not necessarily map to parallel lines
- ratios are not necessarily preserved
- compositions of projective transforms are also projective transforms


## Projective transformations

Projective transformations are combinations of

- affine transformations; and
- projective wraps

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

8 DOF: vectors (and therefore
Properties of projective transformations: matrices) are defined up to scale)

- origin does not necessarily map to origin
- lines map to lines
- parallel lines do not necessarily map to parallel lines
- ratios are not necessarily preserved
- compositions of projective transforms are also projective transforms


## How to interpret projective transformations here?

mage plane


## Determining unknown 2D transformations

## Determining unknown transformations

Suppose we have two triangles: ABC and DEF.


## Determining unknown transformations

Suppose we have two triangles: ABC and DEF.

- What type of transformation will map A to D, B to E, and C to F?



## Determining unknown transformations

Suppose we have two triangles: ABC and DEF.

- What type of transformation will map $A$ to $D, B$ to $E$, and $C$ to $F$ ?
- How do we determine the unknown parameters?

Affine transform:

| uniform scaling + shearing |
| ---: |
| + rotation + translation | \(\quad\left[\begin{array}{ccc}a_{1} \& a_{2} \& a_{3} <br>

a_{4} \& a_{5} \& a_{6} <br>

0 \& 0 \& 1\end{array}\right] \quad\)| How many degrees of |
| :--- |
| freedom do we have? |

## Determining unknown transformations

Suppose we have two triangles: ABC and DEF.

- What type of transformation will map $A$ to $D, B$ to $E$, and $C$ to $F$ ?
- How do we determine the unknown parameters?

unknowns
$\boldsymbol{x}^{\prime}=\mathbf{M} \boldsymbol{x}$

point correspondences
- One point correspondence gives how many equations?
- How many point correspondences do we need?


## Determining unknown transformations

Suppose we have two triangles: ABC and DEF .

- What type of transformation will map $A$ to $D, B$ to $E$, and $C$ to $F$ ?
- How do we determine the unknown parameters?

$\boldsymbol{x}^{\prime}=\mathbf{M} \boldsymbol{x}$

point correspondences



## Least Squares Error

$$
E_{\mathrm{LS}}=\sum_{i}\left\|f\left(\boldsymbol{x}_{i} ; \boldsymbol{p}\right)-\boldsymbol{x}_{\boldsymbol{x}}^{\prime}\right\|^{2}
$$



Least Squares Error
What is this?

$$
\boldsymbol{H}_{\mathrm{LS}}=\sum_{i} \mid \overbrace{\substack{\text { What is } \\ \text { this? }}} \overbrace{\substack{\text { What is }}}^{\substack{\boldsymbol{x} \\ \text { this? }}}
$$


Least Squares Error


Least Squares Error

$$
E_{\mathrm{LS}}=\sum_{i} \frac{\left\|f\left(\boldsymbol{x}_{i} ; \boldsymbol{p}\right)-\boldsymbol{x}_{i}^{\prime}\right\|^{2}}{\vdots}
$$



## Least Squares Error

$$
\begin{aligned}
E_{\mathrm{LS}}= & \sum_{i}\left\|\boldsymbol{f}\left(\boldsymbol{x}_{i} ; \boldsymbol{p}\right)-\boldsymbol{x}_{i}^{\prime}\right\|^{2} \\
& \text { What is the free variable? } \\
& \text { What do we want to optimize? }
\end{aligned}
$$



Find parameters that minimize squared error

$$
\hat{\boldsymbol{p}}=\underset{\boldsymbol{p}}{\arg \min } \sum_{i}\left\|\boldsymbol{f}\left(\boldsymbol{x}_{i} ; \boldsymbol{p}\right)-\boldsymbol{x}_{i}^{\prime}\right\|^{2}
$$

General form of linear least squares
(Warning: change of notation. $x$ is a vector of parameters!)

$$
\begin{aligned}
E_{\mathrm{LLS}} & =\sum_{i}\left|\boldsymbol{a}_{i} \boldsymbol{x}-\boldsymbol{b}_{i}\right|^{2} \\
& =\|\mathbf{A} \boldsymbol{x}-\boldsymbol{b}\|^{2} \quad \text { (matrix form) }
\end{aligned}
$$

## Determining unknown transformations

Affine transformation:

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{lll}
p_{1} & p_{2} & p_{3} \\
p_{4} & p_{5} & p_{6}
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

Why can we drop the last line?

Vectorize transformation parameters:

Stack equations from point correspondences:


General form of linear least squares
(Warning: change of notation. $x$ is a vector of parameters!)

$$
\begin{aligned}
E_{\mathrm{LLS}} & =\sum_{i}\left|\boldsymbol{a}_{i} \boldsymbol{x}-\boldsymbol{b}_{i}\right|^{2} \\
& =\|\mathbf{A} \boldsymbol{x}-\boldsymbol{b}\|^{2}
\end{aligned}
$$

This function is quadratic.
How do you find the root of a quadratic?

## Solving the linear system

Convert the system to a linear least-squares problem:

$$
E_{\mathrm{LLS}}=\|\mathbf{A} \boldsymbol{x}-\boldsymbol{b}\|^{2}
$$

Expand the error:

$$
E_{\mathrm{LLS}}=\boldsymbol{x}^{\top}\left(\mathbf{A}^{\top} \mathbf{A}\right) \boldsymbol{x}-2 \boldsymbol{x}^{\top}\left(\mathbf{A}^{\top} \boldsymbol{b}\right)+\|\boldsymbol{b}\|^{2}
$$

Minimize the error:

$$
\text { Set derivative to } 0\left(\mathbf{A}^{\top} \mathbf{A}\right) \boldsymbol{x}=\mathbf{A}^{\top} \boldsymbol{b}
$$

$$
\text { Solve for } x \quad \boldsymbol{x}=\left(\mathbf{A}^{\top} \mathbf{A}\right)^{-1} \mathbf{A}^{\top} \boldsymbol{b} \longleftarrow \quad \begin{gathered}
\text { Note: You almost never want to } \\
\text { compute the inverse of a matrix. }
\end{gathered}
$$

Linear least squares estimation only works when the transform function is ?

Linear least squares estimation only works when the transform function is linear! (duh)

Also doesn't deal well with outliers

## Determining unknown image warps

## Determining unknown image warps

Suppose we have two images.

- How do we compute the transform that takes one to the other?



## Forward warping

Suppose we have two images.

- How do we compute the transform that takes one to the other?


1. Form enough pixel-to-pixel correspondences between two images
$\longleftarrow$ later lecture
2. Solve for linear transform parameters as before
3. Send intensities $f(x, y)$ in first image to their corresponding location in the second image

## Forward warping

Suppose we have two images.

- How do we compute the transform that takes one to the other?

what is the problem with this?

1. Form enough pixel-to-pixel correspondences between two images
2. Solve for linear transform parameters as before
3. Send intensities $f(x, y)$ in first image to their corresponding location in the second image

## Forward warping

Pixels may end up between two points

- How do we determine the intensity of each point?



## Forward warping

Pixels may end up between two points

- How do we determine the intensity of each point?
$\checkmark$ We distribute color among neighboring pixels $\left(x^{\prime}, y^{\prime}\right)$ ("splatting")

- What if a pixel $\left(x^{\prime}, y^{\prime}\right)$ receives intensity from more than one pixels $(x, y)$ ?


## Forward warping

Pixels may end up between two points

- How do we determine the intensity of each point?
$\checkmark$ We distribute color among neighboring pixels $\left(x^{\prime}, y^{\prime}\right)$ ("splatting")

- What if a pixel $\left(x^{\prime}, y^{\prime}\right)$ receives intensity from more than one pixels $(x, y)$ ?
$\checkmark$ We average their intensity contributions.


## Inverse warping

Suppose we have two images.

- How do we compute the transform that takes one to the other?

$$
f(x, y)
$$


what is the problem with this?

1. Form enough pixel-to-pixel correspondences between two images $\longleftarrow$
2. Solve for linear transform parameters as before, then compute its inverse
3. Get intensities $g\left(x^{\prime}, y^{\prime}\right)$ in in the second image from point $(x, y)=T^{-1}\left(x^{\prime}, y^{\prime}\right)$ in first image

## Inverse warping

Pixel may come from between two points

- How do we determine its intensity?



## Inverse warping

Pixel may come from between two points

- How do we determine its intensity?
$\checkmark$ Use interpolation



## Bilinear interpolation

Grayscale example


In matrix form (with adjusted coordinates)

$$
f(x, y) \approx\left[\begin{array}{ll}
1-x & x
\end{array}\right]\left[\begin{array}{ll}
f(0,0) & f(0,1) \\
f(1,0) & f(1,1)
\end{array}\right]\left[\begin{array}{c}
1-y \\
y
\end{array}\right] .
$$

In Matlab:
call interp2

## Forward vs inverse warping

Suppose we have two images.

- How do we compute the transform that takes one to the other?


Pros and cons of each?

## Forward vs inverse warping

Suppose we have two images.

- How do we compute the transform that takes one to the other?

- Inverse warping eliminates holes in target image
- Forward warping does not require existence of inverse transform


## References

Basic reading:

- Szeliski textbook, Section 3.6.

Additional reading:

- Hartley and Zisserman, "Multiple View Geometry in Computer Vision," Cambridge University Press 2004. a comprehensive treatment of all aspects of projective geometry relating to computer vision, and also a very useful reference for the second part of the class.
- Richter-Gebert, "Perspectives on projective geometry," Springer 2011.
a beautiful, thorough, and very accessible mathematics textbook on projective geometry (available online for free from CMU's library).

