## Image Warping



Computational Photography
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## Reminder: Proj 2 due monday

- Much more difficult than project 1 - get started asap if not already
- Must compute SSD cost for every pixel (slow but not horribly slow using filtering method; see tips at end of project page)
- Learn how to debug visual algorithms: imshow, plot, dbstop if error, keyboard and break points are your friends
- Suggestion: For "quilt_simple", first set upper-left patch to be upper-left patch in source and iteratively find minimum cost patch and overlay --- should reproduce original source image, at least for part of the output


## Review from last class: Gradient Domain

## Editing

General concept: Solve for pixels of new image that satisfy constraints on the gradient and the intensity

- Constraints can be from one image (for filtering) or more (for blending)


## Project 3: Reconstruction from Gradients

1. Preserve $x-y$ gradients
2. Preserve intensity of one pixel

Source pixels: s
Variable pixels: v

1. minimize $\left(v(x+1, y)-v(x, y)-(s(x+1, y)-s(x, y))^{\wedge} 2\right.$
2. minimize $\left(v(x, y+1)-v(x, y)-(s(x, y+1)-s(x, y))^{\wedge} 2\right.$
3. minimize $(v(1,1)-s(1,1))^{\wedge} 2$

## Project 3 (extra): NPR

- Preserve gradients on edges
- e.g., get canny edges with edge(im, 'canny')
- Reduce gradients not on edges
- Preserve original intensity



## Colorization using optimization

- Solve for uv channels (in Luv space) such that similar intensities have similar colors
- Minimize squared color difference, weighted by intensity similarity

$$
J(U)=\sum_{\mathbf{r}}\left(U(\mathbf{r})-\sum_{\mathbf{s} \in N(\mathbf{r})} w_{\mathrm{r} s} U(\mathbf{s})\right)^{2}
$$

- Solve with sparse linear system of equations



## Gradient-domain editing

Many image processing applications can be thought of as trying to manipulate gradients or intensities:

- Contrast enhancement
- Denoising
- Poisson blending
- HDR to RGB
- Color to Gray
- Recoloring
- Texture transfer

See Perez et al. 2003 and GradientShop for many examples

## Gradient-domain processing



Saliency-based Sharpening

## Gradient-domain processing



Non-photorealistic rendering

## Project 3: Gradient Domain Editing

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## Project 3: Reconstruction from Gradients

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## Project 3 (extra): NPR

- Preserve gradients on edges
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## Take-home questions

1) I am trying to blend this bear into this pool. What problems will I have if I use:
a) Alpha compositing with feathering
b) Laplacian pyramid blending
c) Poisson editing?


## Take-home questions

2) How would you make a sharpening filter using gradient domain processing? What are the constraints on the gradients and the intensities?

# Next two classes: warping and morphing 

- Today
- Global coordinate transformations
- Image alignment
- Tuesday
- Interpolation and texture mapping
- Meshes and triangulation
- Shape morphing


## Photo stitching: projective alignment

Find corresponding points in two images


Solve for transformation that aligns the images

## Capturing light fields

Estimate light via projection from spherical surface onto image


## Morphing

Blend from one object to other with a series of local transformations


## Image Transformations

image filtering: change range of image

$$
g(x)=T(f(x))
$$




image warping: change domain of image

$$
g(x)=f(T(x))
$$




## Image Transformations

image filtering: change range of image

$$
g(x)=T(f(x))
$$


image warping: change domain of image

$$
g(x)=f(T(x))
$$



## Parametric (global) warping


$\mathbf{p}=(x, y)$


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

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)

For linear transformations, we can represent $T$ as a matrix

$$
\begin{array}{r}
\mathrm{p}^{\prime}=\mathbf{M p} \\
{\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\mathbf{M}\left[\begin{array}{l}
x \\
y
\end{array}\right]}
\end{array}
$$

## Parametric (global) warping

Examples of parametric warps:


## Scaling

- Scaling a coordinate means multiplying each of its components by a scalar
- Uniform scaling means this scalar is the same for all components:



## Scaling

- Non-uniform scaling: different scalars per component:




## Scaling

- Scaling operation: $x^{\prime}=a x$

$$
y^{\prime}=b y
$$

- Or, in matrix form:

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

What is the transformation from $\left(x^{\prime}, y^{\prime}\right)$ to $(x, y)$ ?

## 2-D Rotation



## 2-D Rotation



```
Polar coordinates...
\(x=r \cos (\phi)\)
\(y=r \sin (\phi)\)
\(x^{\prime}=r \cos (\phi+\theta)\)
\(y^{\prime}=r \sin (\phi+\theta)\)
Trig 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...
\(x^{\prime}=x \boldsymbol{\operatorname { c o s }}(\theta)-y \boldsymbol{\operatorname { s i n }}(\theta)\)
\(y^{\prime}=x \boldsymbol{\operatorname { s i n }}(\theta)+y \cos (\theta)\)
```


## 2-D Rotation

This is easy to capture in matrix form:

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\underbrace{\left[\begin{array}{cc}
\cos (\theta) & -\sin (\theta) \\
\sin (\theta) & \cos (\theta)
\end{array}\right]}_{\mathbf{R}}\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

Even though $\sin (\theta)$ and $\cos (\theta)$ are nonlinear functions of $\theta$,
$-x^{\prime}$ is a linear combination of $x$ and $y$
$-y^{\prime}$ is a linear combination of $x$ and $y$

What is the inverse transformation?

- Rotation by - $\theta$
- For rotation matrices $\quad \mathbf{R}^{-1}=\mathbf{R}^{T}$


## 2x2 Matrices

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

2D Identity?

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

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

2D Scale around ( 0,0 )?

$$
\begin{aligned}
& x^{\prime}=s_{x} * x \\
& y^{\prime}=s_{y} * y
\end{aligned} \quad\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{cc}
s_{x} & 0 \\
0 & s_{y}
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

## 2x2 Matrices

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

2D Rotate around ( 0,0 )?

$$
\begin{aligned}
& x^{\prime}=\cos \Theta^{*} x-\sin \Theta^{*} y \\
& y^{\prime}=\sin \Theta^{*} x+\cos \Theta^{*} y
\end{aligned} \quad\left[\begin{array}{l}
x^{\prime} \\
\boldsymbol{y}^{\prime}
\end{array}\right]=\left[\begin{array}{cc}
\cos \Theta & -\sin \Theta \\
\sin \Theta & \cos \Theta
\end{array}\right]\left[\begin{array}{l}
\boldsymbol{x} \\
\boldsymbol{y}
\end{array}\right]
$$

2D Shear?

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

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{cc}
1 & k_{x} \\
k_{y} & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

## $2 \times 2$ Matrices

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

2D Mirror about Y axis?

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

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

2D Mirror over (0,0)?

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

## $2 \times 2$ 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

Q: How can we represent translation in matrix form?

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

## Homogeneous Coordinates

Homogeneous coordinates

- represent coordinates in 2 dimensions with a 3 -vector



## Homogeneous Coordinates

2D Points $\rightarrow$ Homogeneous Coordinates

- Append 1 to every 2D point: (xy) $\rightarrow$ ( x y 1 ) Homogeneous coordinates $\rightarrow$ 2D Points
- Divide by third coordinate (x y w) $\rightarrow$ ( $\mathrm{x} / \mathrm{w} \mathrm{y} / \mathrm{w}$ ) Special properties
- Scale invariant: (x y w) = k * (x y w)
- $(x, y, 0)$ represents a point at infinity
- $(0,0,0)$ is not allowed



## Homogeneous Coordinates

Q: How can we represent translation in matrix form?

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

A: Using the rightmost column:

$$
\text { Translation }=\left[\begin{array}{ccc}
1 & 0 & \boldsymbol{t}_{\boldsymbol{x}} \\
0 & 1 & \boldsymbol{t}_{\boldsymbol{y}} \\
0 & 0 & 1
\end{array}\right]
$$

## Translation Example

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




## Basic 2D transformations as $3 \times 3$ matrices

$$
\begin{array}{cc}
{\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]} & {\left[\begin{array}{c}
\boldsymbol{x}^{\prime} \\
\boldsymbol{y}^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
\boldsymbol{s}_{\boldsymbol{x}} & 0 & 0 \\
0 & s_{y} & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
\boldsymbol{x} \\
\boldsymbol{y} \\
1
\end{array}\right]} \\
\text { Translate } & \text { Scale } \\
{\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\underset{\text { Rotate }}{\left[\begin{array}{ccc}
\cos \Theta & -\sin \Theta & 0 \\
\sin \Theta & \cos \Theta & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{c}
x \\
y \\
1
\end{array}\right]}} & {\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
1 & \beta_{x} & 0 \\
\beta_{y} & 1 & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{c}
x \\
y \\
1
\end{array}\right]} \\
\text { Shear }
\end{array}
$$

## Matrix Composition

Transformations can be combined by matrix multiplication

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

Does the order of multiplication matter?

## Affine Transformations

Affine transformations are combinations of

- Linear transformations, and
- Translations

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

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


## Projective Transformations

Projective transformations are combos of

- Affine transformations, and
- Projective warps

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

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
- Models change of basis
- Projective matrix is defined up to a scale (8 DOF)


## 2D image transformations



| Name | Matrix | \# D.O.F. | Preserves: | Icon |
| :--- | :---: | :---: | :--- | :---: |
| translation | $[\boldsymbol{I} \mid \boldsymbol{t}]_{2 \times 3}$ |  | $\square$ |  |
| $\operatorname{rigid}$ (Euclidean) | $[\boldsymbol{R} \mid \boldsymbol{t}]_{2 \times 3}$ |  | $\square$ |  |
| similarity | $[s \boldsymbol{R} \mid \boldsymbol{t}]_{2 \times 3}$ |  | $\square$ |  |
| affine | $[\boldsymbol{A}]_{2 \times 3}$ |  | $\square$ |  |
| projective | $[\tilde{\boldsymbol{H}}]_{3 \times 3}$ |  | $\square$ |  |

These transformations are a nested set of groups

- Closed under composition and inverse is a member


## Recovering Transformations



- What if we know $f$ and $g$ and want to recover the transform T?
- willing to let user provide correspondences
- How many do we need?


## Translation: \# correspondences?



- How many Degrees of Freedom?
- How many correspondences needed for translation?
- What is the transformation matrix?

$$
\mathbf{M}=\left[\begin{array}{ccc}
1 & 0 & p_{x}^{\prime}-p_{x} \\
0 & 1 & p_{y}^{\prime}-p_{y} \\
0 & 0 & 1
\end{array}\right]
$$

## Euclidian: \# correspondences?



- How many DOF?
- How many correspondences needed for translation+rotation?


## Affine: \# correspondences?



- How many DOF?
- How many correspondences needed for affine?

Affine transformation estimation

- Math
- Matlab demo


## Projective: \# correspondences?



- How many DOF?
- How many correspondences needed for projective?


## Take-home Question

1) Suppose we have two triangles: $A B C$ and $A^{\prime} B^{\prime} C^{\prime}$. What transformation will map $A$ to $A^{\prime}$, $B$ to $B^{\prime}$, and $C$ to $C^{\prime}$ ? How can we get the parameters?


## Take-home Question

2) Show that distance ratios along a line are preserved under 2d linear transformations.


Hint: Write down $x 2$ in terms of $x 1$ and $x 3$, given that the three points are co-linear

Next class: texture mapping and morphing

