# Projective Geometry and Camera Models 

Computer Vision<br>CS 543 / ECE 549<br>University of Illinois

Derek Hoiem

## HWs

- HW 1 back today
- Top segmentation scores:
- (0.618, 0.659) - Zigang Xiao: multiscale filters
- (0.623, 0.653) - Fu Ouyang: oriented filters, + morphological operators to close gaps
- Common mistakes
- Forgetting to convert to double when computing Laplacian pyramids
- Mistakes in creating oriented filters
- Incorrectly applied non-max suppression
- Solutions are posted
- HW 2 due next Tues
- HW 3 should be out Thursday


## Think about your final projects

- Strongly encouraged to work in groups of 2-4 (but if you have a good reason to work by self, could be ok)
- Projects don't need to be of publishable originality but should evince independent effort to learn about a new topic, try something new, or apply to an application of interest
- Proposals will be due after Spring Break
- Project ideas from Cinda Hereen

1. Classroom attendance: I teach in Siebel 1404, and I'd like to be able to track attendance by taking photos of the room (students who opt out can just put their heads down). Extra facts in the problem: a) I have photo rosters; b) i'm willing to use a tagging system wherein students can verify our assertions; c) the results don't have to be exact.
2. Medication tracking: I would like for a person to be able to chart his/her medicine consumption by taking a photo of the med bottles. Extra facts: a. it has to work on a mobile device (limited computation), b. it could be a matching problem--could photograph known meds and use the photos as the labels.
3. Thermometer reading: In a related problem, I'd like to be able to take a picture of a thermometer (digital or analogue) and record it's reading.

## One last note on registration

- Thin-plate splines: combines global affine warp with smooth local deformation
$E_{T P S}(f)=\sum_{a=1}^{K}\left\|y_{a}-f\left(v_{a}\right)\right\|^{2}+\lambda \iint\left[\left(\frac{\partial^{2} f}{\partial x^{2}}\right)^{2}+2\left(\frac{\partial^{2} f}{\partial x \partial y_{\lambda}}\right)^{2}+\left(\frac{\partial^{2} f}{\partial y^{2}}\right)^{2}\right] d x d y$
Diff of predicted vs. actual position
Smoothness cost for local warps

$$
\begin{array}{ll}
f\left(v_{a}, d, w\right)=v_{a} \cdot d+\phi\left(v_{a}\right) \cdot w \\
\text { Affine warp } & \begin{array}{l}
\text { Local deformation } \\
\text { according to distance from } \\
\text { control points }
\end{array}
\end{array}
$$

- Robust non-rigid point matching: http://noodle.med.yale.edu/~chui/tps-rpm.html (includes code, demo, paper)
- Thin-plate spline registration with robustness to outliers


Fig. 12. Large Deformation-Caterpillar Example. From left to right, matching frame 1 to frame 5, 7, 11 and 12. Top: Original location. Middle: matched result. Bottom: deformation found.

## Next two classes: Single-view Geometry



## Today's class

Mapping between image and world coordinates

- Pinhole camera model
- Projective geometry
- Vanishing points and lines
- Projection matrix


## Image formation



Let's design a camera

- Idea 1: put a piece of film in front of an object
- Do we get a reasonable image?


## Pinhole camera



Idea 2: add a barrier to block off most of the rays

- This reduces blurring
- The opening known as the aperture


## Pinhole camera



## Camera obscura: the pre-camera

- First idea: Mo-Ti, China (470BC to 390BC)
- First built: Alhacen, Iraq/Egypt (965 to 1039AD)


Illustration of Camera Obscura


Freestanding camera obscura at UNC Chapel Hill

## Camera Obscura used for Tracing



Lens Based Camera Obscura, 1568

## First Photograph

Oldest surviving photograph

- Took 8 hours on pewter plate


Joseph Niepce, 1826

Photograph of the first photograph


Stored at UT Austin

Niepce later teamed up with Daguerre, who eventually created Daguerrotypes

## Dimensionality Reduction Machine (3D to 2D)

## 3D world

2D image


Point of observation

## Projection can be tricky...



## Projection can be tricky...



Making of 3D sidewalk art: http://www.youtube.com/watch?v=3SNYtdOAyt0

## Projective Geometry

## What is lost?

- Length



## Length is not preserved



Figure by David Forsyth

## Projective Geometry

## What is lost?

- Length
- Angles



## Projective Geometry

## What is preserved?

- Straight lines are still straight



## Vanishing points and lines

Parallel lines in the world intersect in the image at a "vanishing point"


## Vanishing points and lines



- The projections of parallel 3D lines intersect at a vanishing point
- The projection of parallel 3D planes intersect at a vanishing line
- If a set of parallel 3D lines are also parallel to a particular plane, their vanishing point will lie on the vanishing line of the plane
- Not all lines that intersect are parallel
- Vanishing point <-> 3D direction of a line
- Vanishing line <-> 3D orientation of a surface


## Vanishing points and lines



## Vanishing points and lines



## Note on estimating vanishing points



Use multiple lines for better accuracy
... but lines will not intersect at exactly the same point in practice
One solution: take mean of intersecting pairs
... bad idea!
Instead, minimize angular differences

## Vanishing objects



## Projection: world coordinates $\rightarrow$ image coordinates



## Homogeneous coordinates

## Conversion

Converting to homogeneous coordinates

$$
\begin{array}{cc}
(x, y) \Rightarrow\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right] & (x, y, z) \Rightarrow\left[\begin{array}{l}
x \\
y \\
z \\
1
\end{array}\right] \\
\text { homogeneous image } & \text { homogeneous scene } \\
\text { coordinates } & \text { coordinates }
\end{array}
$$

Converting from homogeneous coordinates

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

## Homogeneous coordinates

Invariant to scaling

$$
\begin{aligned}
& k\left[\begin{array}{c}
x \\
y \\
k
\end{array}\right]=\left[\begin{array}{c}
k x \\
w y \\
k y
\end{array}\right]
\end{aligned} \underset{\text { Homogeneous }}{\text { Coordinates }} \underset{\text { Cortesian }}{\left[\begin{array}{c}
\frac{k x}{k w} \\
\frac{k y}{k w}
\end{array}\right]=\left[\begin{array}{c}
\frac{x}{w} \\
\frac{y}{w}
\end{array}\right]} \underset{\text { Coordinates }}{\text { Con }}
$$

Point in Cartesian is ray in Homogeneous

Basic geometry in homogeneous coordinates

- Line equation: $\mathrm{ax}+\mathrm{by}+\mathrm{c}=0$

$$
\text { line }_{i}=\left[\begin{array}{l}
a_{i} \\
b_{i} \\
c_{i}
\end{array}\right]
$$

- Append 1 to pixel coordinate to get homogeneous coordinate

$$
p_{i}=\left[\begin{array}{l}
u_{i} \\
v_{i} \\
1
\end{array}\right]
$$

- Line given by cross product of two points

$$
\text { line }_{i j}=p_{i} \times p_{j}
$$

- Intersection of two lines given by cross product of the lines

$$
q_{i j}=\text { line }_{i} \times \text { line }_{j}
$$

Another problem solved by homogeneous coordinates

Intersection of parallel lines


## Projection matrix



$$
\mathbf{x}=\mathbf{K}\left[\begin{array}{ll}
\mathbf{R} & \mathbf{t}
\end{array}\right] \mathbf{X}
$$

x: Image Coordinates: $(u, v, 1)$
K: Intrinsic Matrix (3x3)
R: Rotation (3x3)
t: Translation (3x1)
X: World Coordinates: (X,Y,Z, 1)

Interlude: when have I used this stuff?

## When have I used this stuff?

Object Recognition (CVPR 2006)


## When have I used this stuff?

## Single-view reconstruction (SIGGRAPH 2005)



## When have I used this stuff?

Getting spatial layout in indoor scenes (ICCV 2009)


## When have I used this stuff?

Inserting synthetic objects into images: http://vimeo.com/28962540



## Projection matrix



Intrinsic Assumptions Extrinsic Assumptions

- Unit aspect ratio
- Optical center at $(0,0)$
- No skew
$\mathbf{x}=\mathbf{K}\left[\begin{array}{ll}\mathbf{I} & \mathbf{0}\end{array}\right] \mathbf{X} \Rightarrow w\left[\begin{array}{l}u \\ v \\ 1\end{array}\right]=\left[\begin{array}{llll:l}f f & 0 & 0 & 0 \\ 0 & - & 0 \\ 0 & 0 & 1 & 0\end{array}\right]\left[\begin{array}{l}x \\ y \\ z \\ 1\end{array}\right]$


# Remove assumption: known optical center 

Intrinsic Assumptions Extrinsic Assumptions

- Unit aspect ratio
- No skew

\author{

- No rotation
}
- Camera at (0,0,0)

$$
\mathbf{x}=\mathbf{K}\left[\begin{array}{ll}
\mathbf{I} & \mathbf{0}
\end{array}\right] \mathbf{X} \Rightarrow w\left[\begin{array}{c}
u \\
v \\
1
\end{array}\right]=\left[\begin{array}{ccc:c}
f & 0 & u_{0} & 0 \\
0 & f & v_{0} & 0 \\
0 & 0 & 0 & y \\
y & 0 & 1 & 0 \\
z \\
1
\end{array}\right]
$$

## Remove assumption: square pixels

$$
\begin{aligned}
& \text { Intrinsic Assumptions Extrinsic Assumptions } \\
& \text { - No skew } \\
& \text { - No rotation } \\
& \text { - Camera at ( } 0,0,0 \text { ) } \\
& \mathbf{X}=\mathbf{K}\left[\begin{array}{ll}
\mathbf{I} & \mathbf{0}
\end{array}\right] \mathbf{X} \Rightarrow w\left[\begin{array}{l}
u \\
v \\
1
\end{array}\right]=\left[\begin{array}{ccc:c}
\alpha \alpha & 0 & u_{0} & 0 \\
0 & \beta & v_{0} & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 \\
z \\
z
\end{array}\right]
\end{aligned}
$$

## Remove assumption: non-skewed pixels

## Intrinsic Assumptions Extrinsic Assumptions <br> - No rotation <br> - Camera at ( $0,0,0$ ) <br> $$
\mathbf{x}=\mathbf{K}\left[\begin{array}{ll} \mathbf{I} & \mathbf{0} \end{array}\right] \mathbf{X} \Rightarrow w\left[\begin{array}{c} u \\ v \\ 1 \end{array}\right]=\left[\begin{array}{ccc:c} \alpha & s & u_{0} & 0 \\ 0 & \beta & v_{0} & 0 \\ 0 & 0 & 0 & 1 \\ y & y \\ z \\ z \end{array}\right]
$$

Note: different books use different notation for parameters

## Oriented and Translated Camera



## Allow camera translation

## Intrinsic Assumptions Extrinsic Assumptions

- No rotation
$\mathbf{x}=\mathbf{K}\left[\begin{array}{ll}\mathbf{I} & \mathbf{t}\end{array}\right] \mathbf{X} \Rightarrow w\left[\begin{array}{l}u \\ v \\ 1\end{array}\right]=\left[\begin{array}{lll}\alpha & 0 & u_{0} \\ 0 & \beta & v_{0} \\ 0 & 0 & 1\end{array}\right]\left[\begin{array}{llll}1 & 0 & 0 & t_{x} \\ 0 & 1 & 0 & t_{y} \\ 0 & 0 & 1 & t_{z}\end{array}\right]\left[\begin{array}{c}x \\ y \\ z \\ 1\end{array}\right]$


## 3D Rotation of Points

Rotation around the coordinate axes, counter-clockwise:


## Allow camera rotation

$$
\begin{aligned}
& \mathbf{x}=\mathbf{K}\left[\begin{array}{ll}
\mathbf{R} & \mathbf{t}
\end{array}\right] \mathbf{X} \\
& w\left[\begin{array}{c}
u \\
v \\
1
\end{array}\right]=\left[\begin{array}{lll}
\alpha & s & u_{0} \\
0 & \beta & v_{0} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{llll}
r_{11} & r_{12} & r_{13} & t_{x} \\
r_{21} & r_{22} & r_{23} & t_{y} \\
r_{31} & r_{32} & r_{33} & t_{z}
\end{array}\right]\left[\begin{array}{c}
x \\
y \\
z \\
1
\end{array}\right]
\end{aligned}
$$

## Degrees of freedom

## $\mathbf{x}=\mathbf{K}\left[\begin{array}{ll}\mathbf{R} & \mathbf{t}\end{array}\right] \mathbf{X}$

$$
w\left[\begin{array}{c}
u \\
v \\
1
\end{array}\right]=\left[\begin{array}{lll}
\alpha & s & u_{0} \\
0 & \beta & v_{0} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{llll}
r_{11} & r_{12} & r_{13} & t_{x} \\
r_{21} & r_{22} & r_{23} & t_{y} \\
r_{31} & r_{32} & r_{33} & t_{z}
\end{array}\right]\left[\begin{array}{c}
x \\
y \\
z \\
1
\end{array}\right]
$$

## Vanishing Point $=$ Projection from Infinity

$$
\mathbf{p}=\mathbf{K}\left[\begin{array}{ll}
\mathbf{R} & \mathbf{t}
\end{array}\right]\left[\begin{array}{c}
x \\
y \\
z \\
0
\end{array}\right] \Rightarrow \mathbf{p}=\mathbf{K} \mathbf{R}\left[\begin{array}{l}
x \\
y \\
z
\end{array}\right] \Rightarrow \mathbf{p}=\mathbf{K}\left[\begin{array}{c}
x_{R} \\
y_{R} \\
z_{R}
\end{array}\right]
$$

$$
w\left[\begin{array}{l}
u \\
v \\
1
\end{array}\right]=\left[\begin{array}{ccc}
f & 0 & u_{0} \\
0 & f & v_{0} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x_{R} \\
y_{R} \\
z_{R}
\end{array}\right] \Rightarrow \begin{gathered}
u=\frac{f x_{R}}{z_{R}}+u_{0} \\
v=\frac{f y_{R}}{z_{R}}+v_{0}
\end{gathered}
$$

## Orthographic Projection

- Special case of perspective projection
- Distance from the COP to the image plane is infinite

- Also called "parallel projection"
- What's the projection matrix?

$$
w\left[\begin{array}{l}
u \\
v \\
1
\end{array}\right]=\left[\begin{array}{llll}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
z \\
1
\end{array}\right]
$$

## Scaled Orthographic Projection

- Special case of perspective projection
- Object dimensions are small compared to distance to camera

- Also called "weak perspective" $\quad w\left[\begin{array}{l}u \\ v \\ 1\end{array}\right]=\left[\begin{array}{cccc}f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 0 & s\end{array}\right]\left[\begin{array}{l}x \\ y \\ z \\ 1\end{array}\right]$


## Example

Far field: object appearance doesn't change as objects translate


Near field: object appearance changes as objects translate

## Beyond Pinholes: Radial Distortion

- Common in wide-angle lenses or for special applications (e.g., security)
- Creates non-linear terms in projection
- Usually handled by through solving for non-linear terms and then correcting image


No Distortion


Barrel Distortion


Pincushion Distortion


Corrected Barrel Distortion

## Things to remember

- Vanishing points and vanishing lines

- Pinhole camera model and camera projection matrix


$$
\mathbf{x}=\mathbf{K}\left[\begin{array}{ll}
\mathbf{R} & \mathbf{t}
\end{array}\right] \mathbf{X}
$$

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

## Next class

- Applications of camera model and projective geometry
- Recovering the camera intrinsic and extrinsic parameters from an image
- Recovering size in the world
- Projecting from one plane to another

