# Projective Geometry and Camera Models 

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

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## HWs

- HW 1 back today
- Solutions are posted
- Frequent mistake on question about shadow on specular surface

- HW 2 due next Tues

- HW 3 should be out by end of week


## Top edge methods

- Huy Le (0.673):
- Find edge magnitudes using sobel
- Suppress using canny on R, G, B (OR edge maps) max pooling
- Get max filter response of applying RFS filters (from Oxford) to boundary map



## Top edge methods

- Austin Walters (0.673):
- convert to LAB
- compute gradient magnitudes
- take max over channels

- Yang Xu (0.646)
- oriented filter within RGB+HSV
- group edges using connected components
- threshold based on edge length/intensity



## 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


## Last notes on registration

- Thin-plate splines: combines global affine warp with smooth local deformation

$$
\begin{gathered}
E_{T P S}(f)=\sum_{a=1}^{K}\left\|y_{a}-f\left(v_{a}\right)\right\|^{2}+\lambda \iint\left[\left(\frac{\partial^{2} f}{\partial x^{2}}\right)^{2}+2\left(\frac{\partial^{2} f}{\partial x \partial y}\right)^{2}+\left(\frac{\partial^{2} f}{\partial y^{2}}\right)^{2}\right] d x d y \\
\text { Diff of predicted vs. actual position } \quad \text { Smoothness cost for local warps }
\end{gathered}
$$

There is a closed form solution for parameter estimation and warping

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

- 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: Alhazen, 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

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

Point in Cartesian is ray in Homogeneous

Basic geometry in homogeneous coordinates

- Line equation: $a x+b y+c=0$

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

- Append 1 to pixel coordinate to get homogeneous coordinate

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

- Line given by cross product of two points

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

- 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



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

$\mathbf{x : ~ I m a g e ~ C o o r d i n a t e s : ~ ( ~} u, v, 1$ )
K: Intrinsic Matrix (3x3)
R: Rotation ( $3 \times 3$ )
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


## When have I used this stuff?

Creating detailed and complete 3D scene models from a single view (ongoing)

## Projection matrix



Intrinsic Assumptions Extrinsic Assumptions

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


## Remove assumption: known optical center

Intrinsic Assumptions Extrinsic Assumptions

- Unit aspect ratio
- No skew
- No rotation
- Camera at ( $0,0,0$ )

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

## Remove assumption: square pixels

$$
\begin{array}{ll}
\quad \begin{array}{l}
\text { Intrinsic Assumptions } \\
\bullet \text { No skew }
\end{array} & \begin{array}{l}
\text { Extrinsic Assumptions } \\
\\
\\
\\
\text { • No rotation }
\end{array} \\
\mathbf{~} \text { Camera at }(0,0,0)
\end{array}
$$

## 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[$$
\begin{array}{l} u \\ v \\ 1 \end{array}
$$\right]=\left[$$
\begin{array}{ccc:c} {\left[\begin{array}{ccc} \alpha & - & u_{0} \end{array}
$$\right.

 \& 0 <br>10 \& \beta \& v_{0} \& 0 <br>
0 \& 0 \& 1 \& 0 <br>
\hdashline y <br>
y <br>
z <br>
1
\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

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

## Degrees of freedom

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

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

## Vanishing Point $=$ Projection from Infinity

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

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

## Scaled Orthographic Projection

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

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


## 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

