

Projective Geometry and Camera Models

Computer Vision
CS 543 / ECE 549
University of Illinois

Derek Hoiem

HWs

- HW 1 back yesterday
 - Solutions are posted
 - Shadow on specular surface



- HW 2 due next Mon night



- HW 3 should be out by next Tuesday

Top-scoring edge methods

- Lengyue Chen (Overall: 0.681):
 - RGB to LAB
 - Smooth each channel (L by $\sigma=2.8$, others by 3)
 - Average gradient magnitudes
 - No non-max suppression
- Anusri Papari (0.669):
 - HSV + RGB gradient mags, $\sigma=2.5$
- Max Feinberg (0.662)
 - HSV + smoothing + gradient



L. Chen Result



Think about your final projects

- Strongly encouraged to work in groups of 3-4 (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

$$E_{TPS}(f) = \sum_{a=1}^K \|y_a - f(v_a)\|^2 + \lambda \int \int \left[\left(\frac{\partial^2 f}{\partial x^2}\right)^2 + 2\left(\frac{\partial^2 f}{\partial x \partial y}\right)^2 + \left(\frac{\partial^2 f}{\partial y^2}\right)^2 \right] dx dy$$

Diff of predicted vs. actual position

Smoothness cost for local warps

There is a closed form solution for parameter estimation and warping

$$f(v_a, d, w) = v_a \cdot d + \phi(v_a) \cdot w$$

Affine warp

Local deformation
according to distance from
control points

- 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

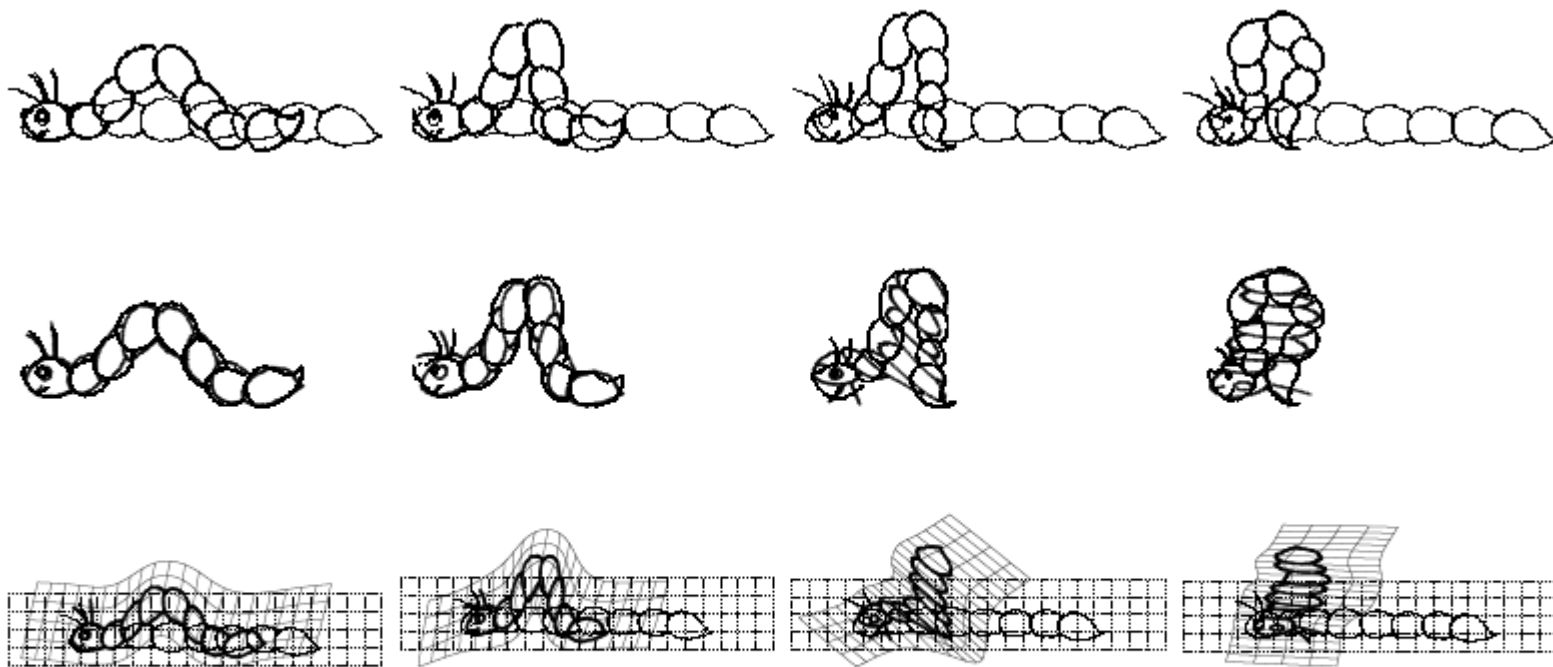
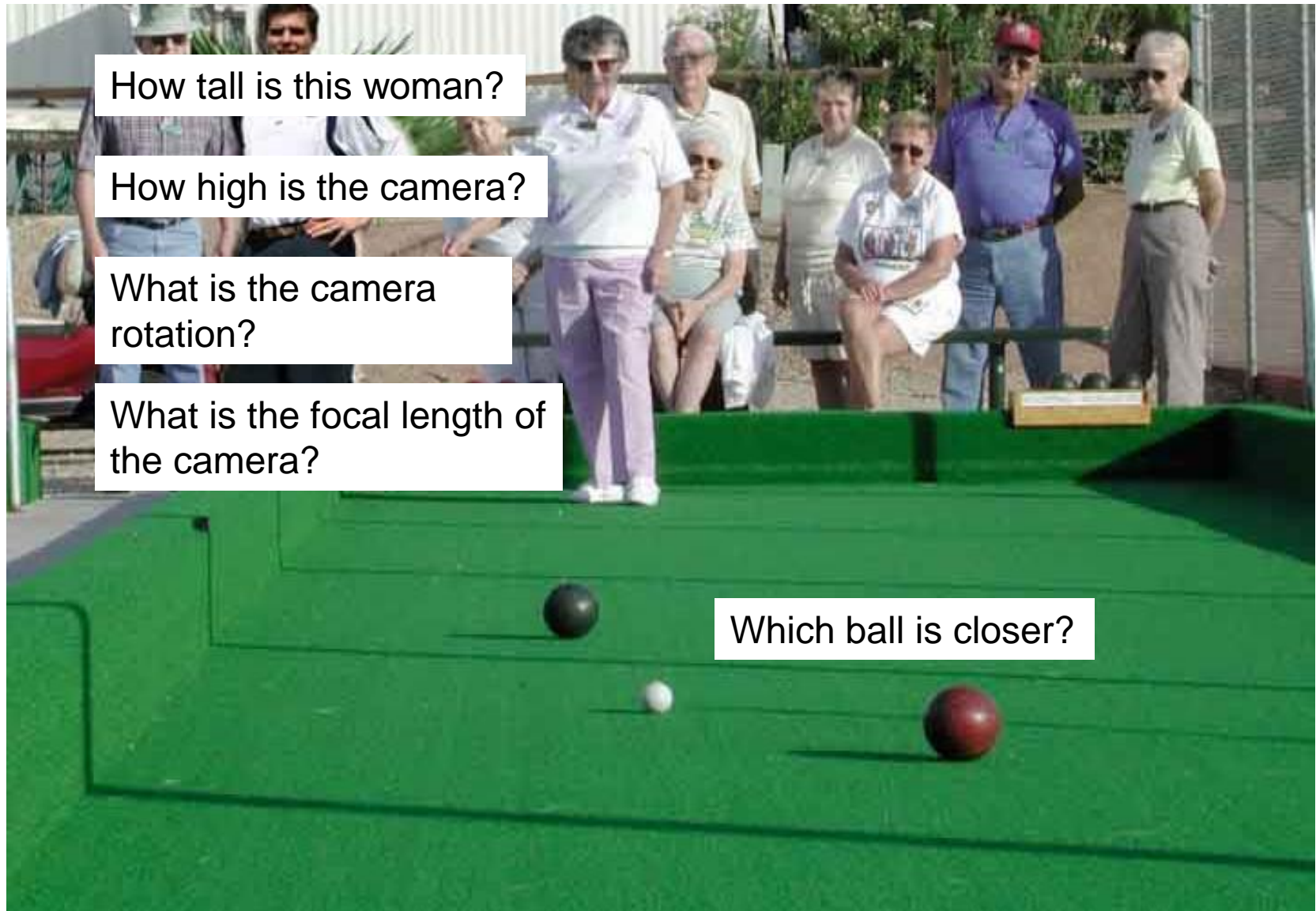


Fig. 12. Large Deformation—Caterpillar Example. From left to right, matching frame 1 to frame 5, 7, 11 and 12. Top: Original location. Middle: matched result. Bottom: deformation found.

Next two classes: Single-view Geometry



How tall is this woman?

How high is the camera?

What is the camera rotation?

What is the focal length of the camera?

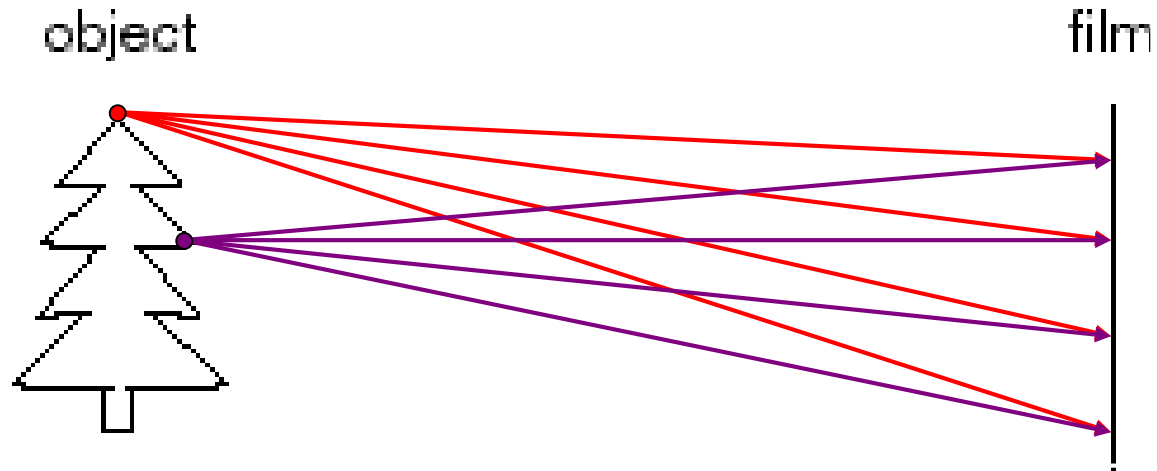
Which ball is closer?

Today's class

Mapping between image and world coordinates

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

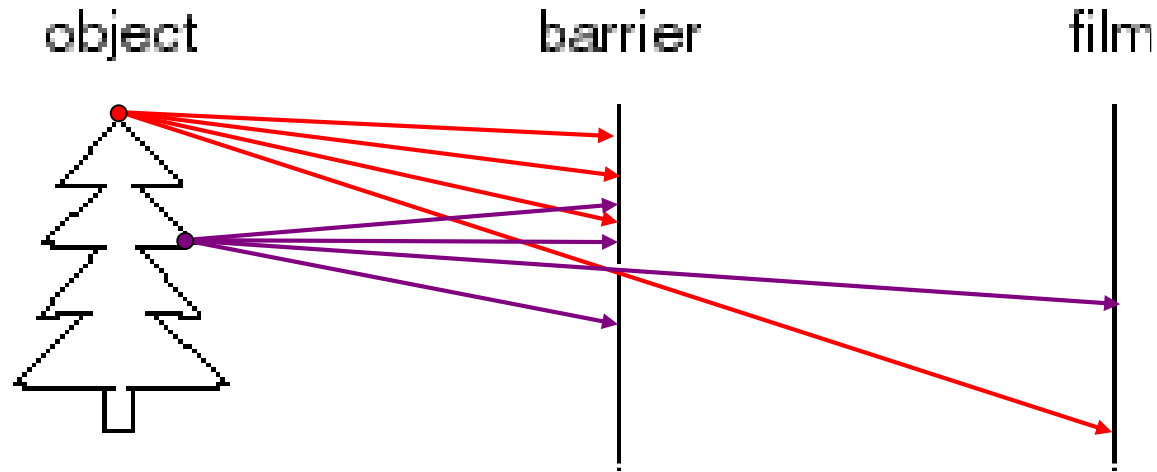
Image formation



Let's design a camera

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

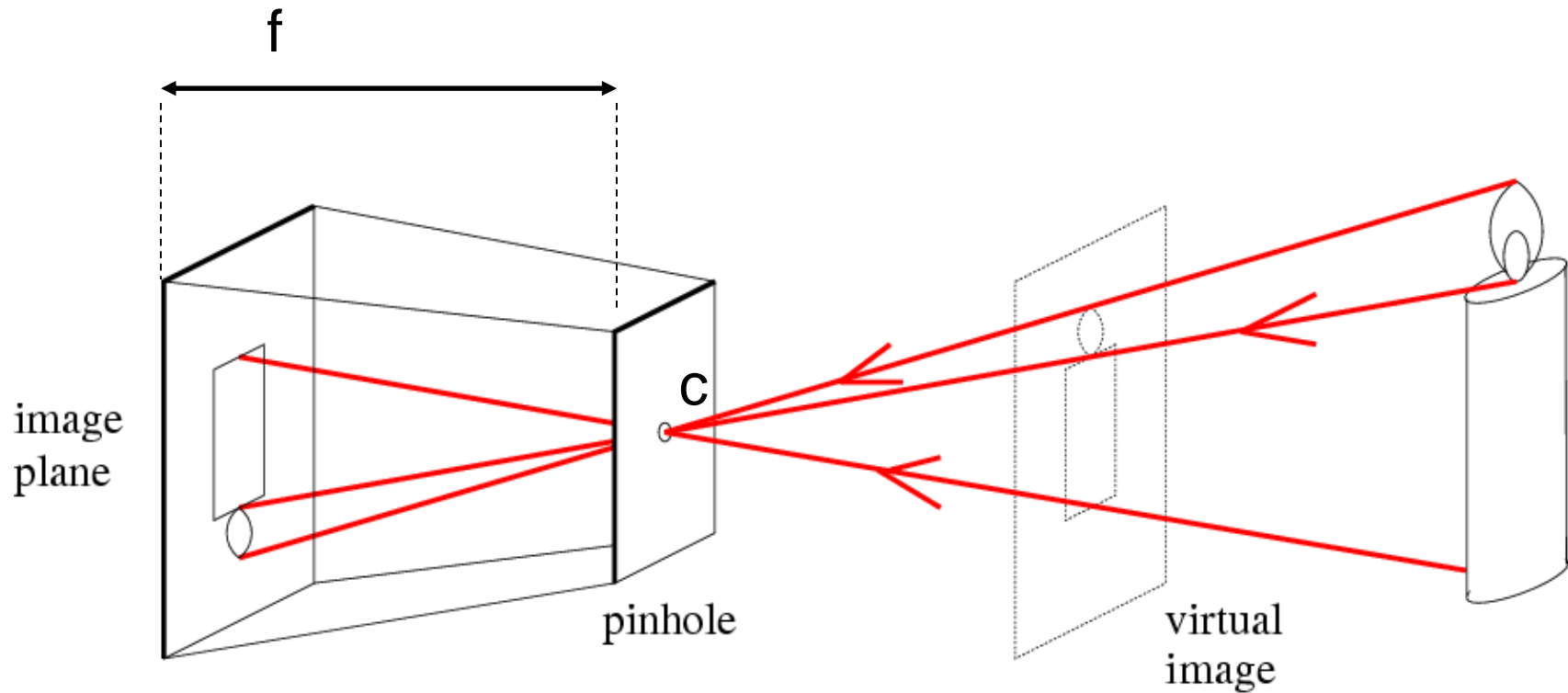
Pinhole camera



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

- This reduces blurring
- The opening known as the **aperture**

Pinhole camera



f = focal length

c = center of the camera

Camera obscura: the pre-camera

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

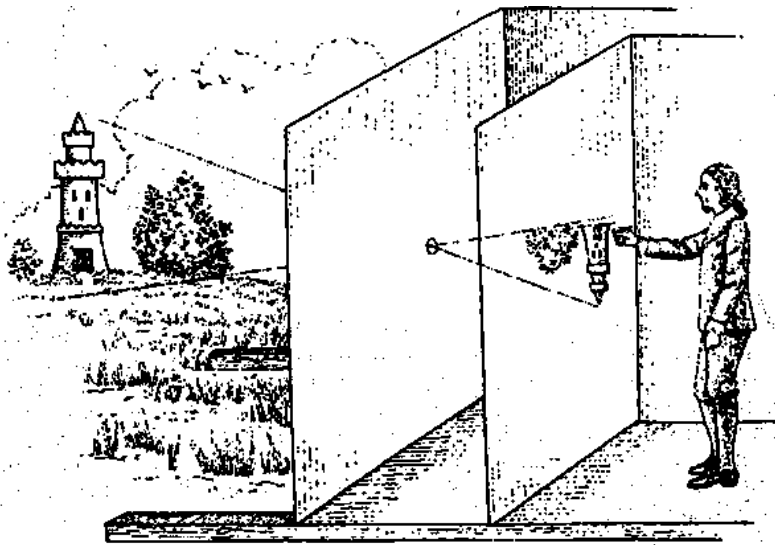


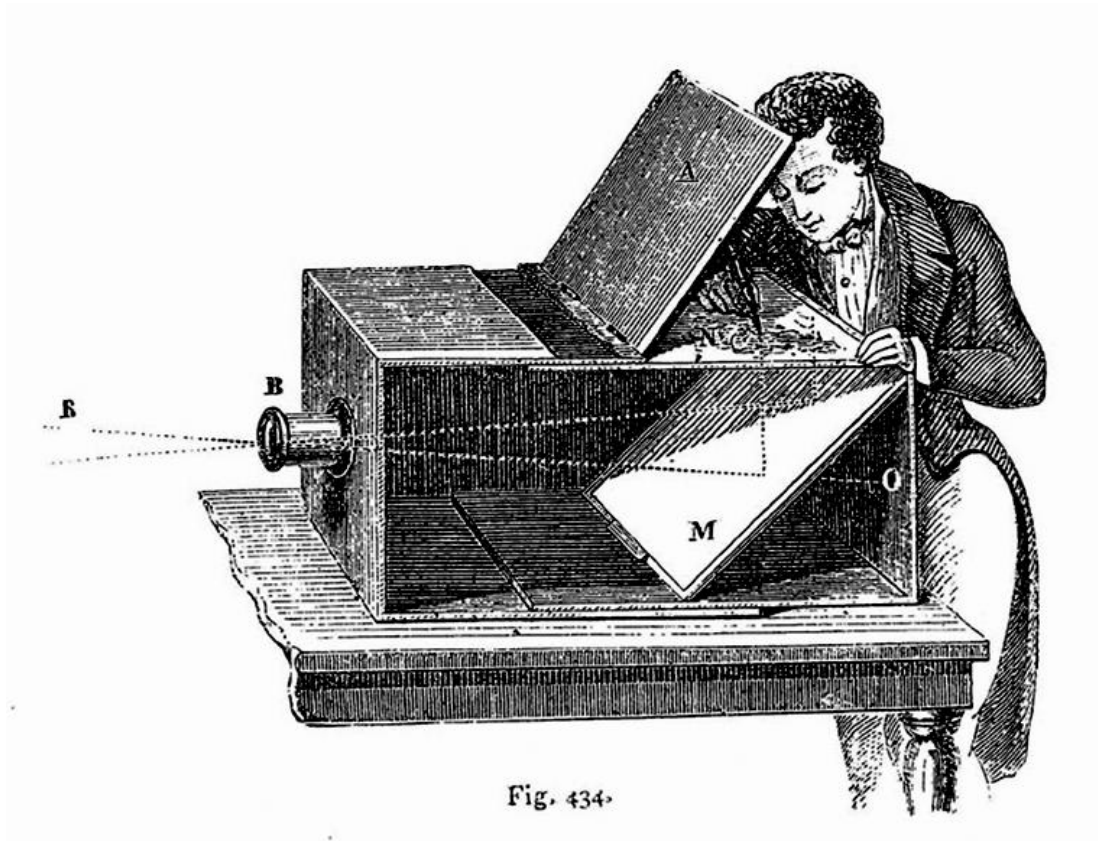
Illustration of Camera Obscura



Freestanding camera obscura at UNC Chapel Hill

Photo by Seth Ilys

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

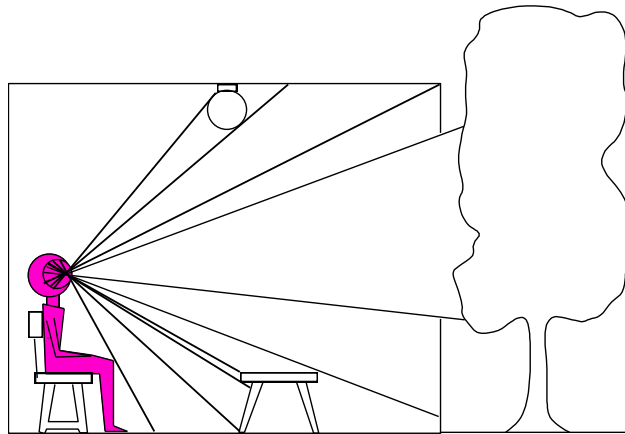


Stored at UT Austin

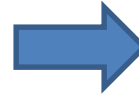
Niepce later teamed up with Daguerre, who eventually created Daguerrotypes

Dimensionality Reduction Machine (3D to 2D)

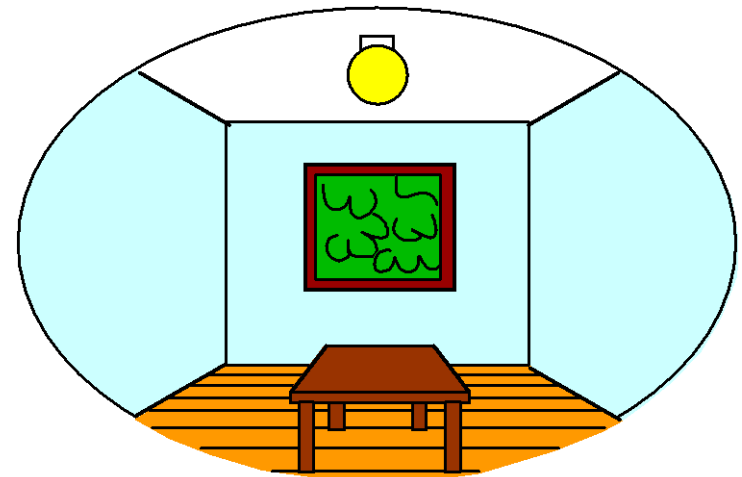
3D world



Point of observation



2D image



Projection can be tricky...



Projection can be tricky...

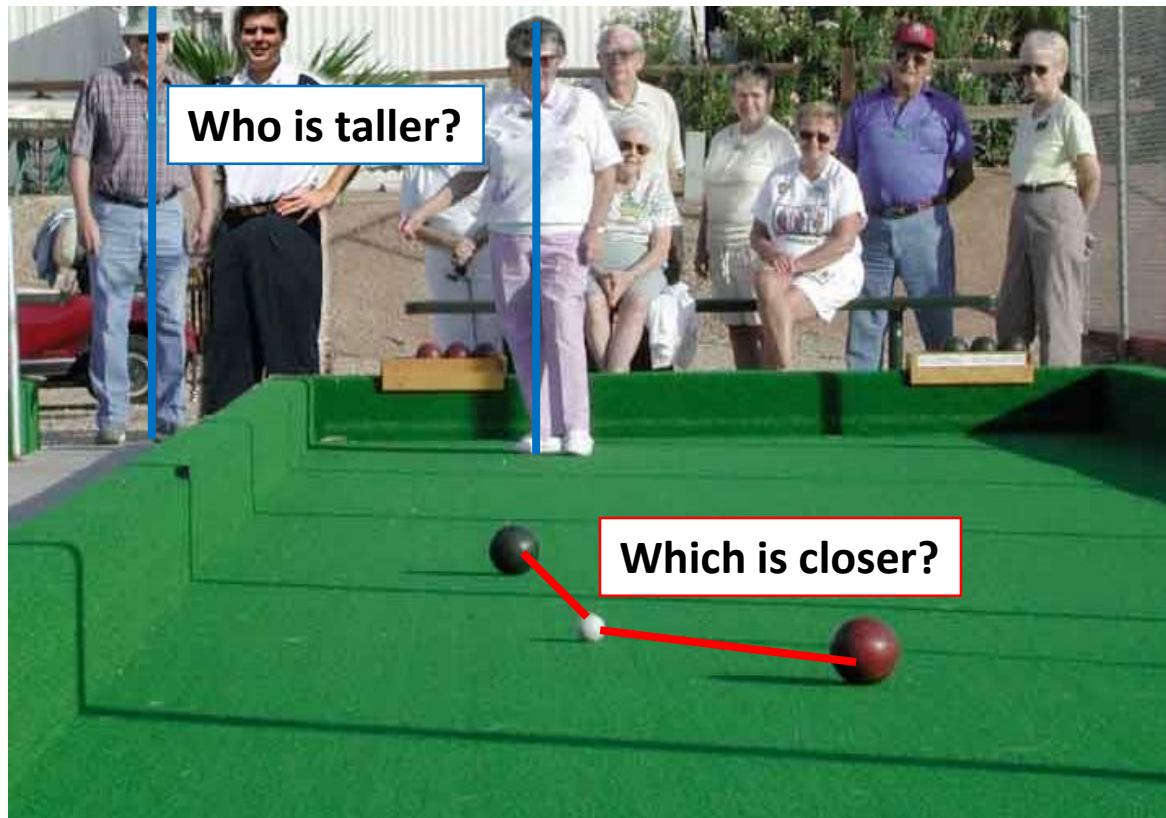


Making of 3D sidewalk art: <http://www.youtube.com/watch?v=3SNYtd0Ayt0>

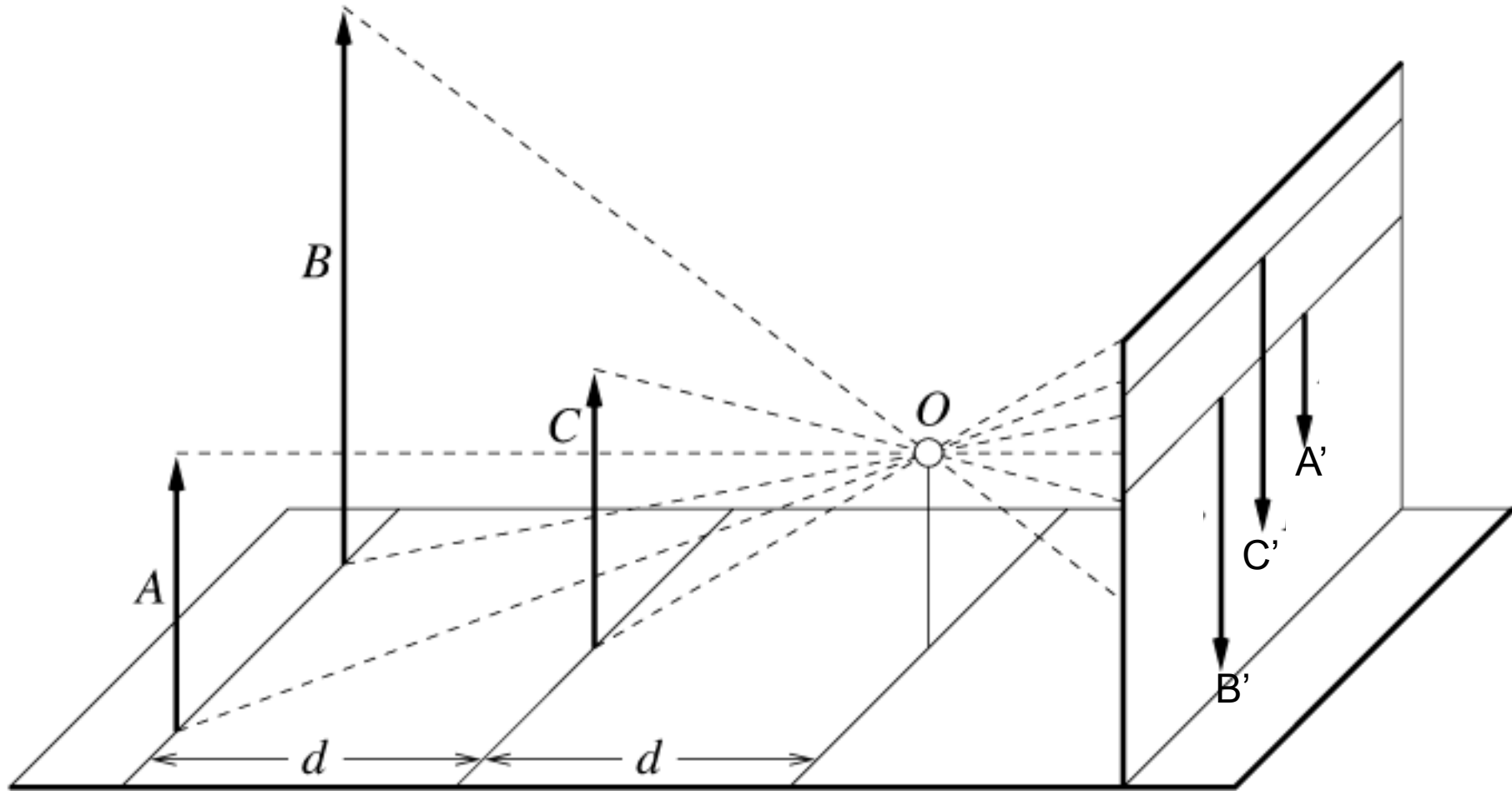
Projective Geometry

What is lost?

- Length



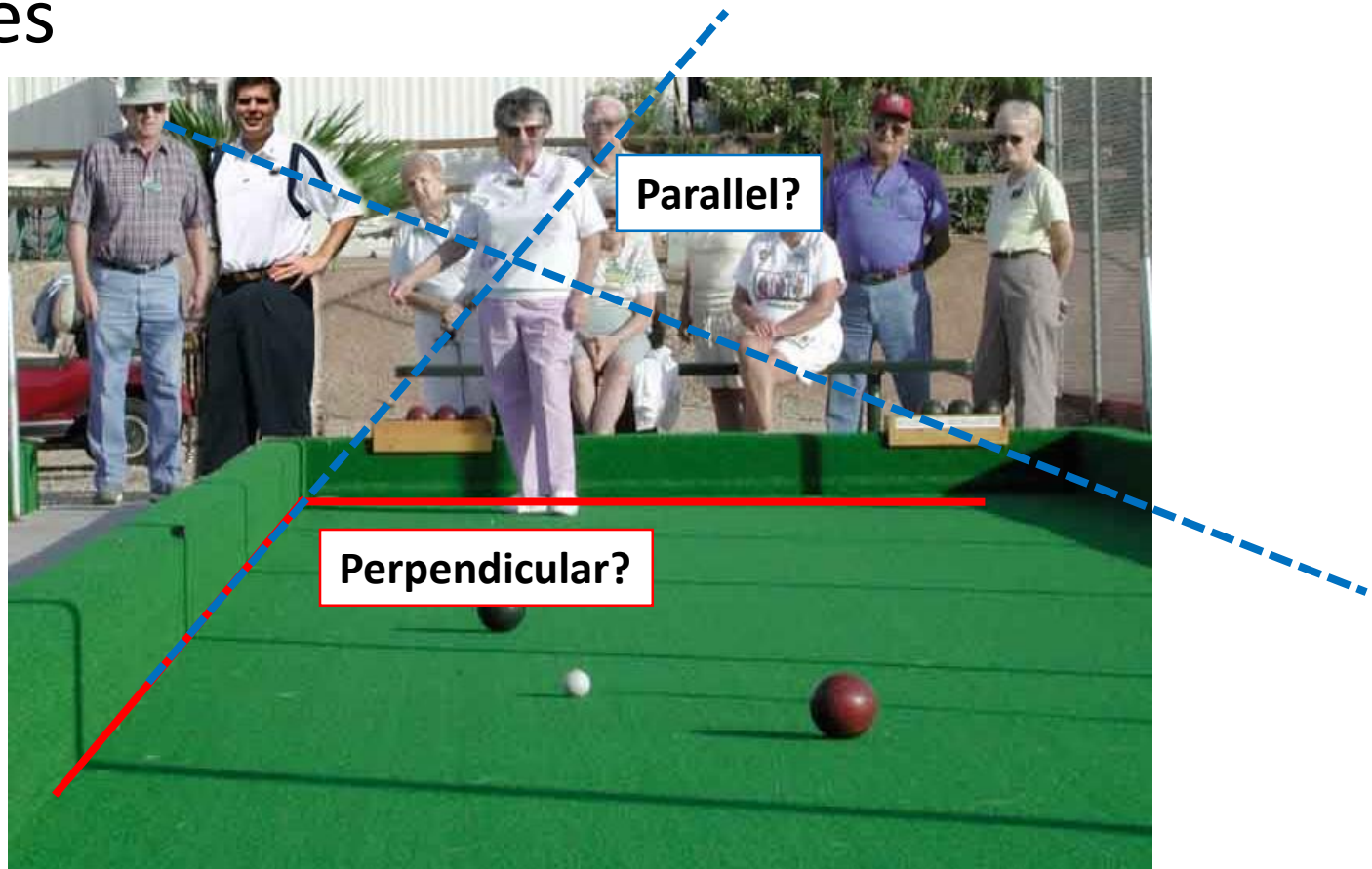
Length is not preserved



Projective Geometry

What is lost?

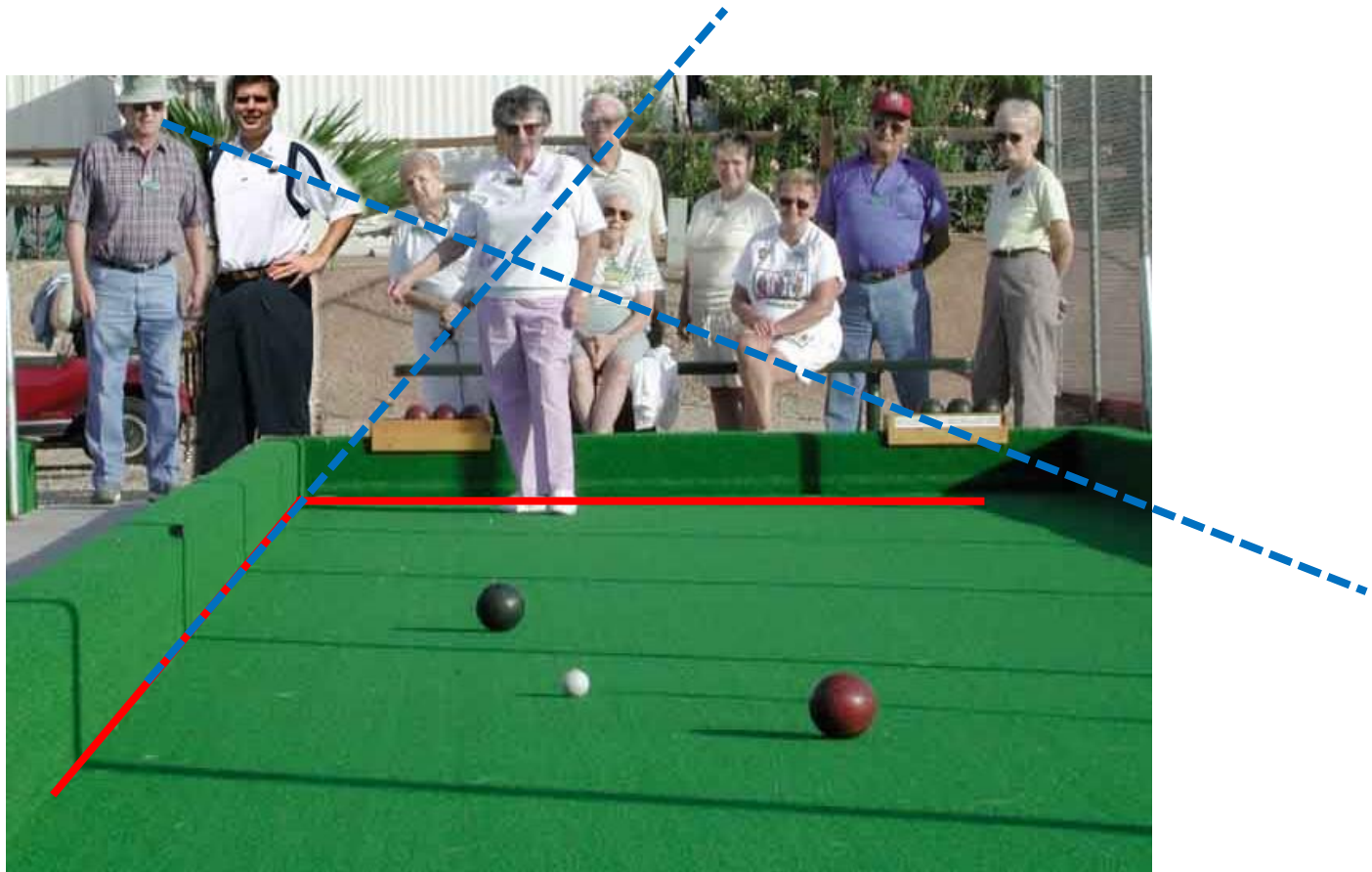
- Length
- Angles



Projective Geometry

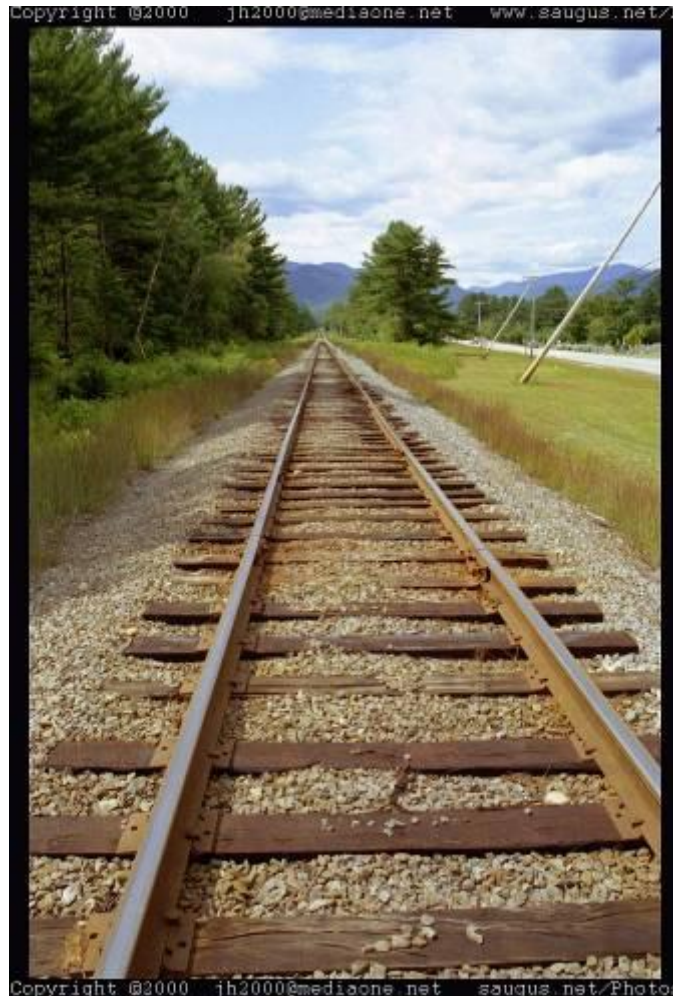
What is preserved?

- Straight lines are still straight

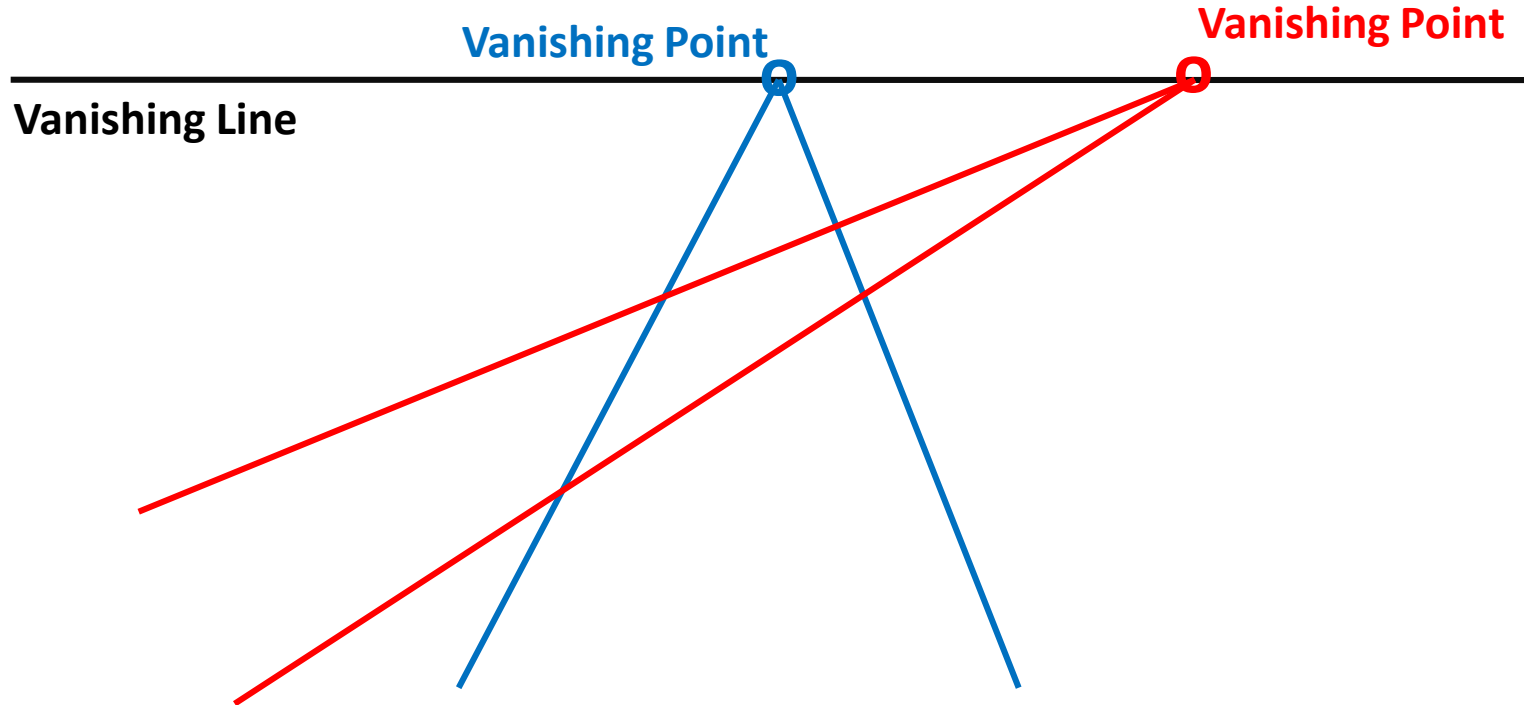


Vanishing points and lines

Parallel lines in the world intersect in the image at a “vanishing point”

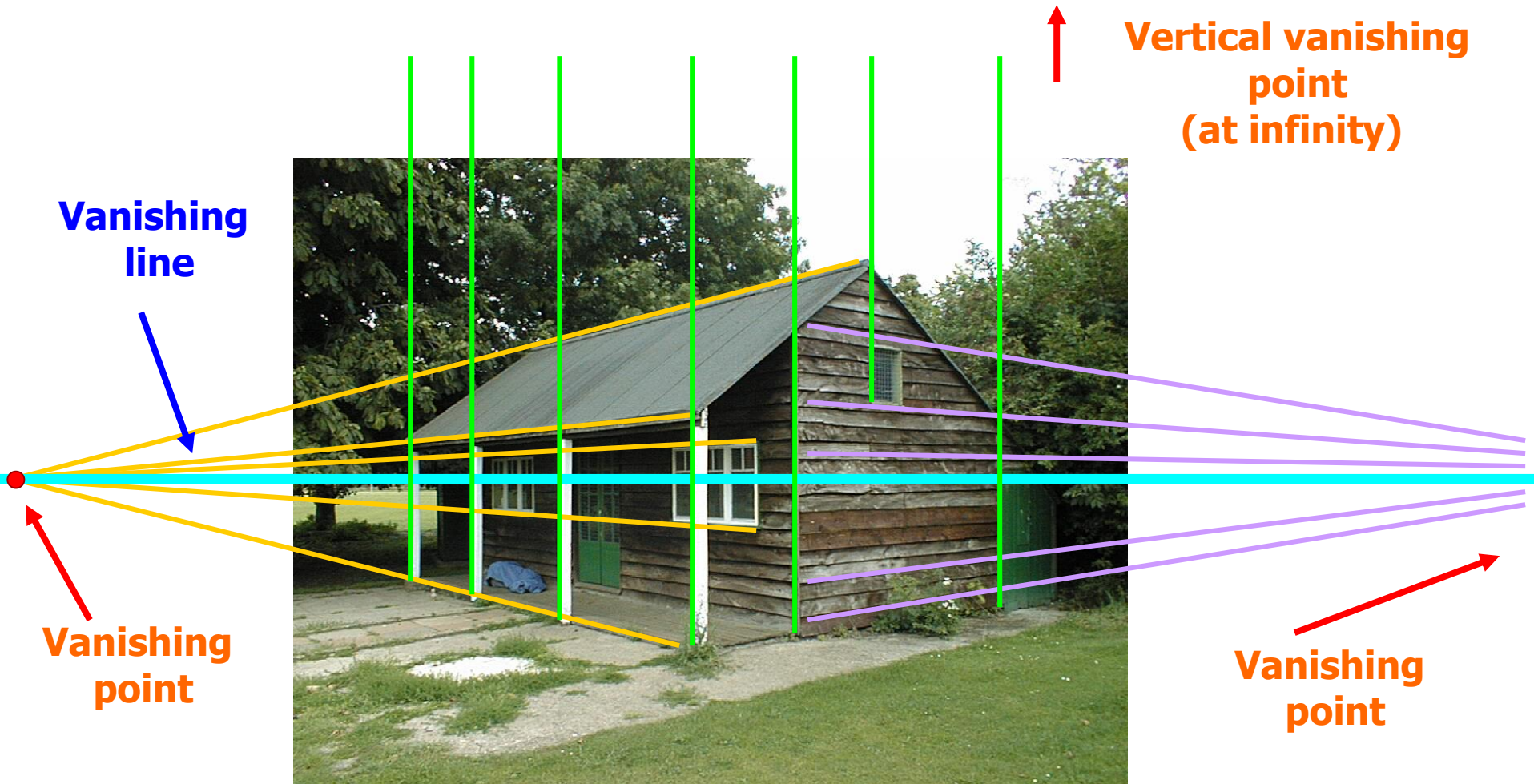


Vanishing points and lines



- 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 \leftrightarrow 3D direction of a line
- Vanishing line \leftrightarrow 3D orientation of a surface

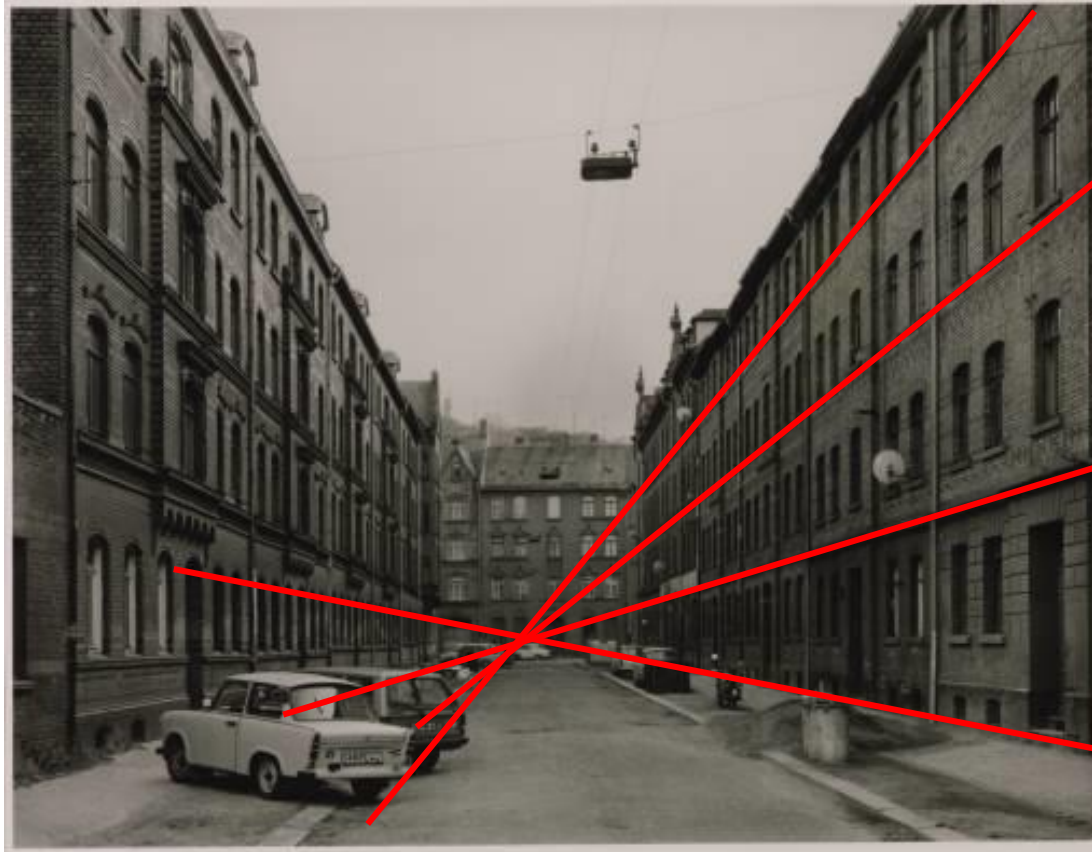
Vanishing points and lines



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

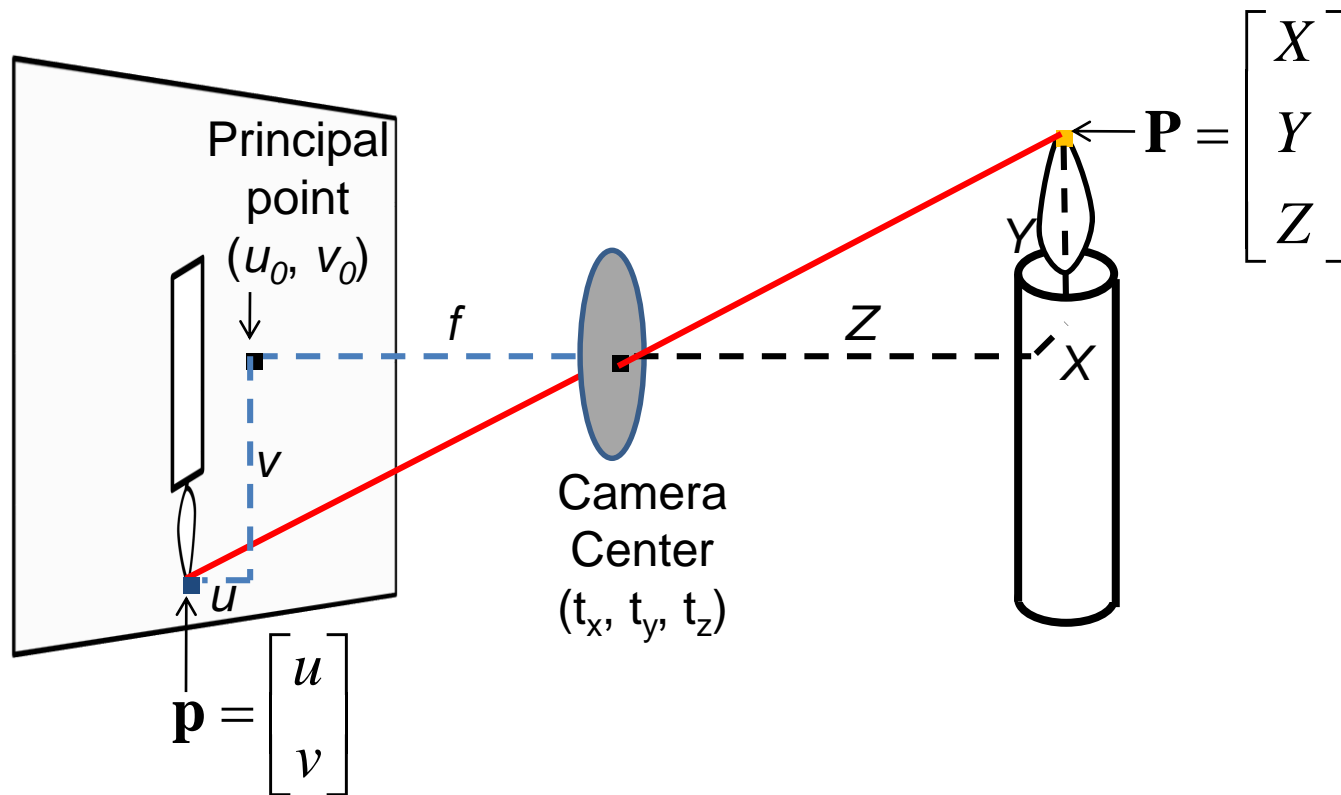
... bad idea!

Instead, minimize angular differences

Vanishing objects



Projection: world coordinates \rightarrow image coordinates



Homogeneous coordinates

Conversion

Converting to *homogeneous* coordinates

$$(x, y) \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

homogeneous image
coordinates

$$(x, y, z) \Rightarrow \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

homogeneous scene
coordinates

Converting *from* homogeneous coordinates

$$\begin{bmatrix} x \\ y \\ w \end{bmatrix} \Rightarrow (x/w, y/w)$$

$$\begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} \Rightarrow (x/w, y/w, z/w)$$

Homogeneous coordinates

Invariant to scaling

$$k \begin{bmatrix} x \\ y \\ w \end{bmatrix} = \begin{bmatrix} kx \\ ky \\ kw \end{bmatrix} \Rightarrow \begin{bmatrix} \frac{kx}{kw} \\ \frac{ky}{kw} \\ \frac{kw}{kw} \end{bmatrix} = \begin{bmatrix} \frac{x}{w} \\ \frac{y}{w} \\ 1 \end{bmatrix}$$

Homogeneous
Coordinates

Cartesian
Coordinates

Point in Cartesian is ray in Homogeneous

Basic geometry in homogeneous coordinates

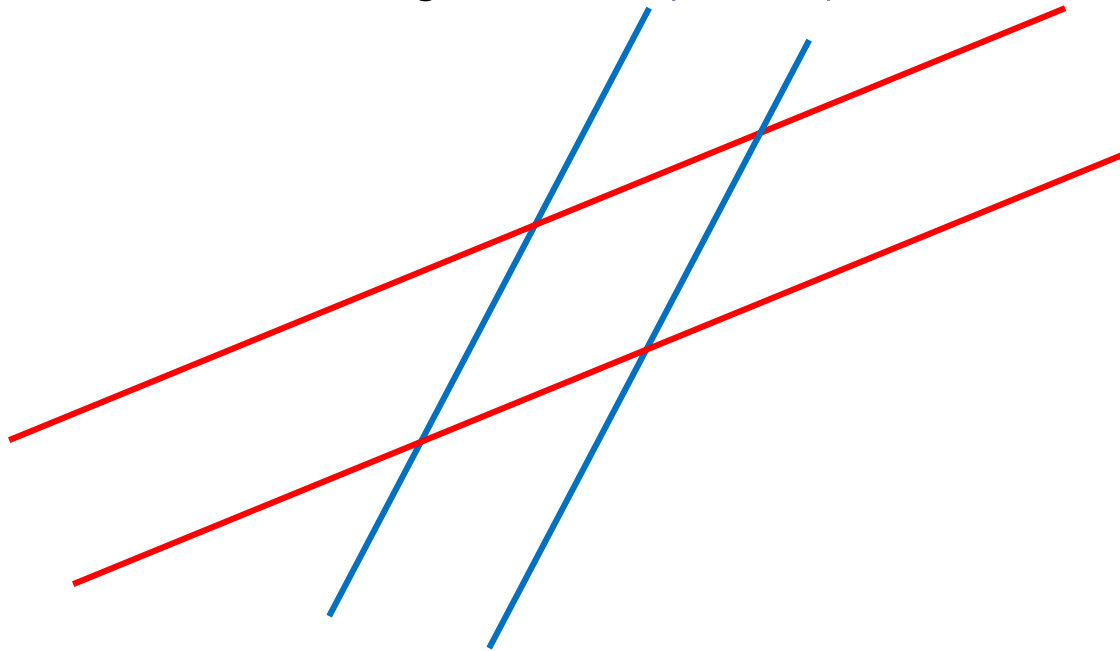
- Line equation: $ax + by + c = 0$ $line_i = \begin{bmatrix} a_i \\ b_i \\ c_i \end{bmatrix}$
- Append 1 to pixel coordinate to get homogeneous coordinate $p_i = \begin{bmatrix} u_i \\ v_i \\ 1 \end{bmatrix}$
- Line given by cross product of two points $line_{ij} = p_i \times p_j$
- Intersection of two lines given by cross product of the lines $q_{ij} = line_i \times line_j$

Another problem solved by homogeneous coordinates

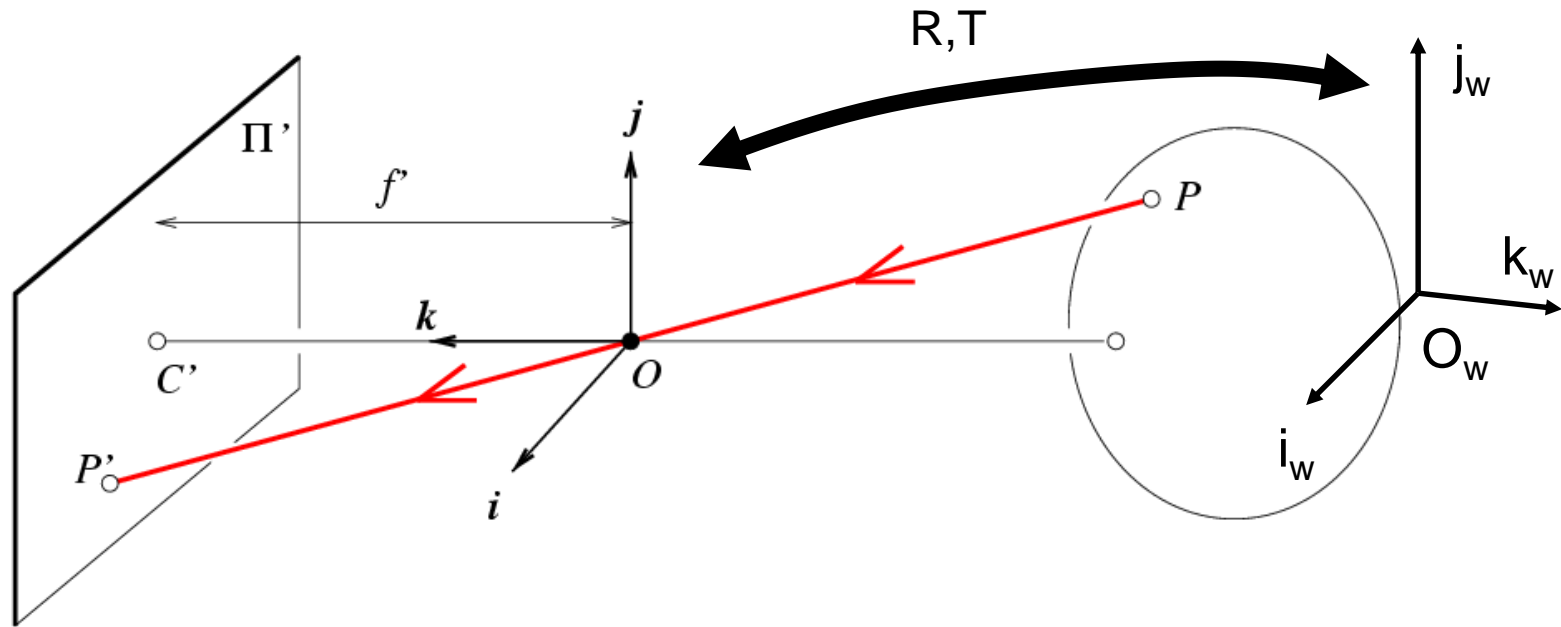
Intersection of parallel lines

Cartesian: (Inf, Inf)
Homogeneous: $(1, 1, 0)$

Cartesian: (Inf, Inf)
Homogeneous: $(1, 2, 0)$



Projection matrix



$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \mathbf{X}$$

\mathbf{x} : Image Coordinates: $(u, v, 1)$

\mathbf{K} : Intrinsic Matrix (3×3)

\mathbf{R} : Rotation (3×3)

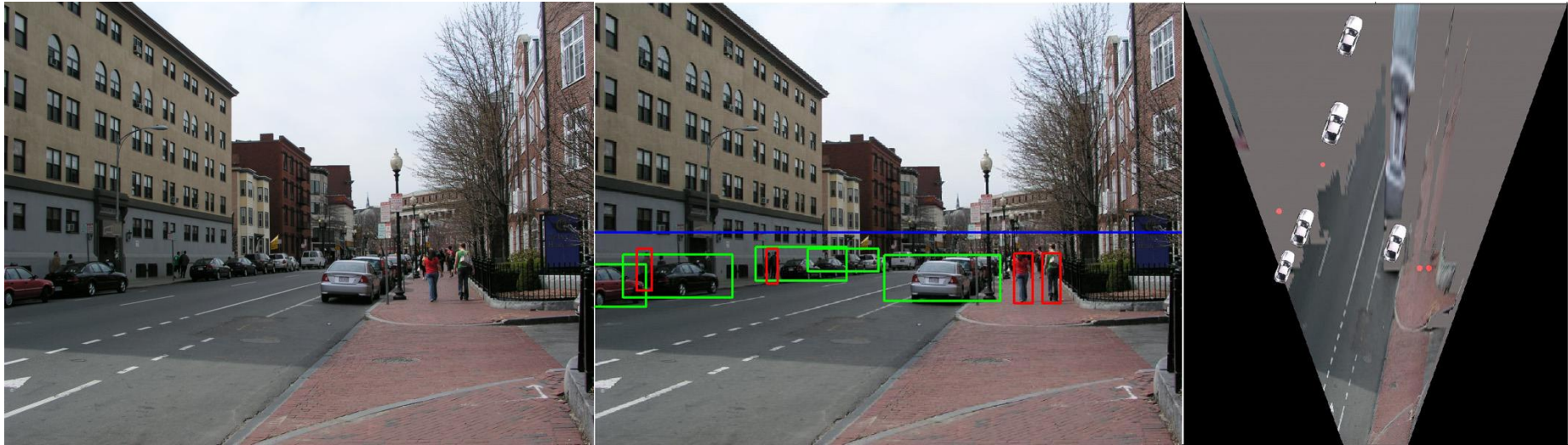
\mathbf{t} : Translation (3×1)

\mathbf{X} : World Coordinates: $(X, Y, Z, 1)$

Interlude: when have I used this stuff?

When have I used this stuff?

Object Recognition (CVPR 2006)



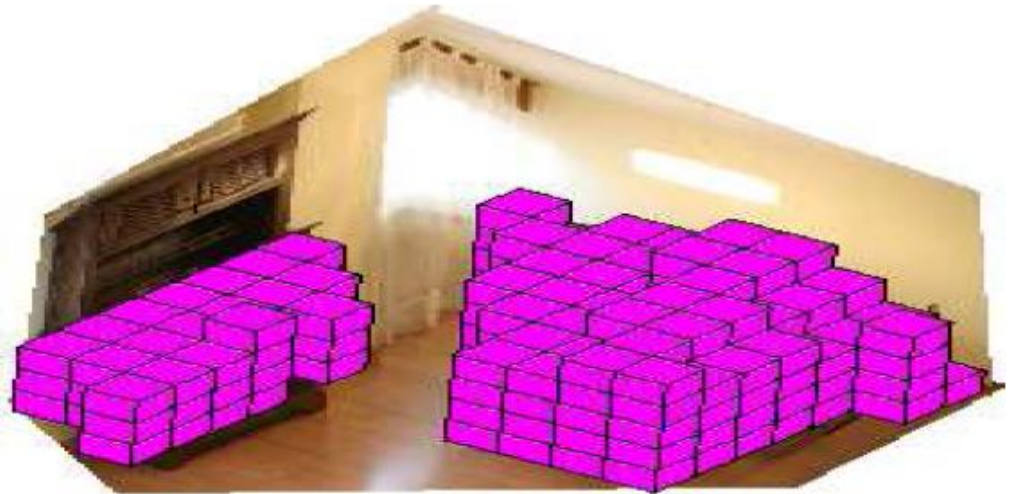
When have I used this stuff?

Single-view reconstruction (SIGGRAPH 2005)



When have I used this stuff?

Getting spatial layout in indoor scenes (ICCV 2009)



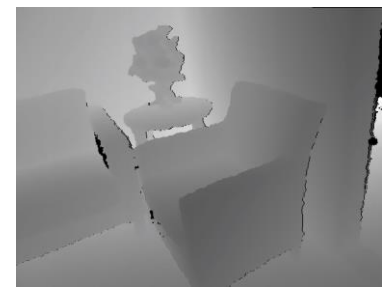
When have I used this stuff?

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

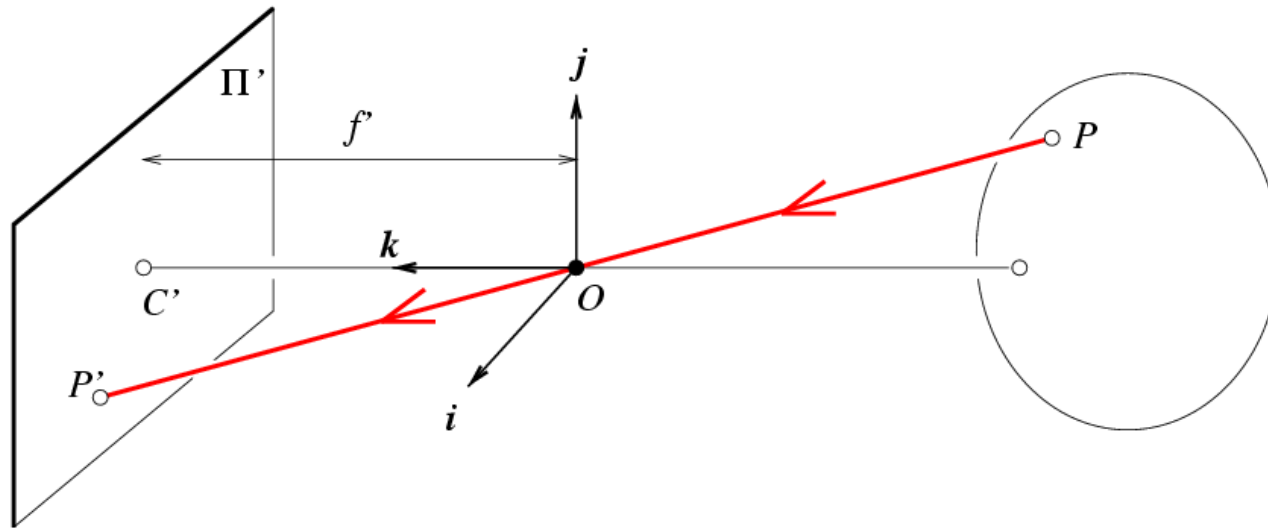


When have I used this stuff?

Multiview 3D reconstruction at Reconstruct



Projection matrix



Intrinsic Assumptions

- Unit aspect ratio
- Principal point at $(0,0)$
- No skew

Extrinsic Assumptions

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

$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{I} & \mathbf{0} \end{bmatrix} \mathbf{X} \Rightarrow w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

K

Remove assumption: known principal point

Intrinsic Assumptions

- Unit aspect ratio
- No skew

Extrinsic Assumptions

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

$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{I} & \mathbf{0} \end{bmatrix} \mathbf{X} \quad \Rightarrow \quad w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & u_0 \\ 0 & f & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

Remove assumption: square pixels

Intrinsic Assumptions

- No skew

Extrinsic Assumptions

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

$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{I} & \mathbf{0} \end{bmatrix} \mathbf{X} \Rightarrow w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \alpha & 0 & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

Remove assumption: non-skewed pixels

Intrinsic Assumptions

- Pinhole camera model

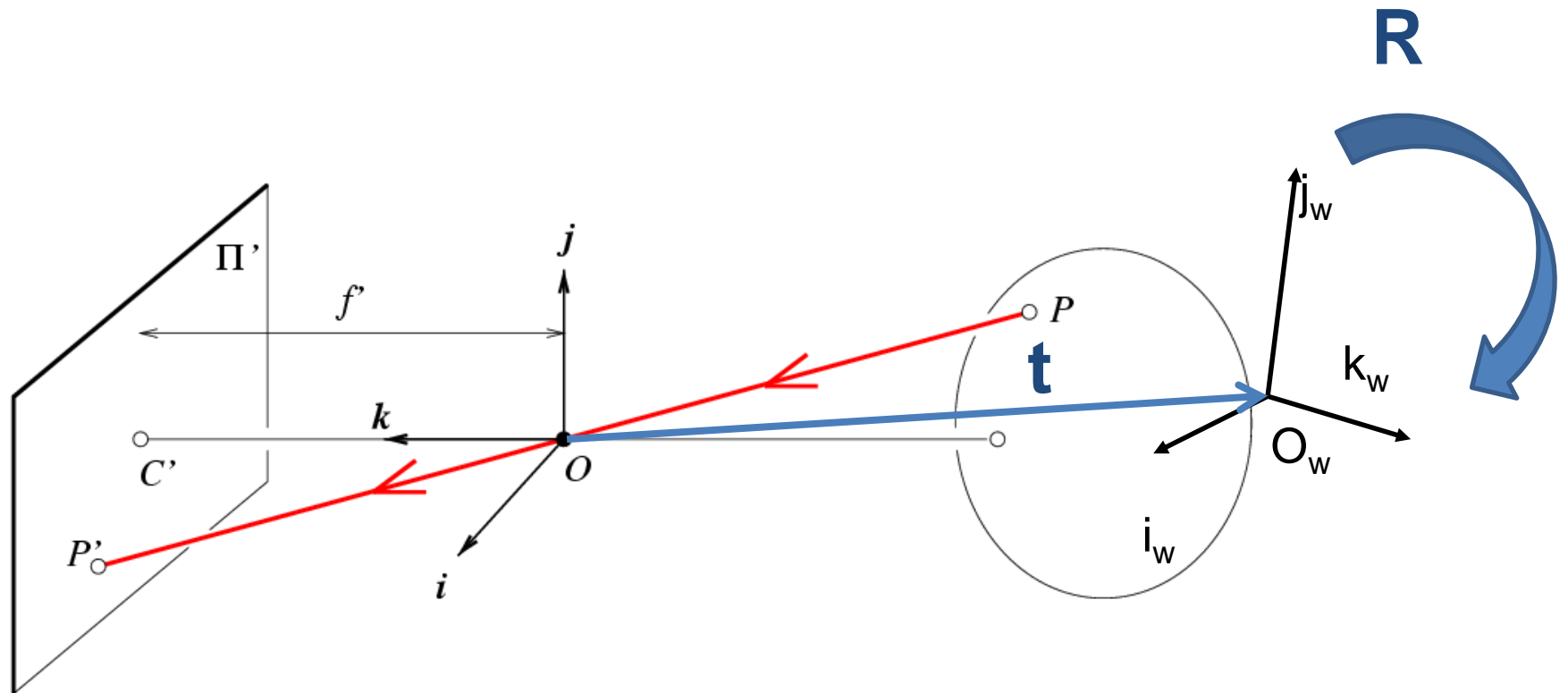
Extrinsic Assumptions

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

$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{I} & \mathbf{0} \end{bmatrix} \mathbf{X} \Rightarrow w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \alpha & s & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

Note: different books use different notation for parameters

Oriented and Translated Camera



Allow camera translation

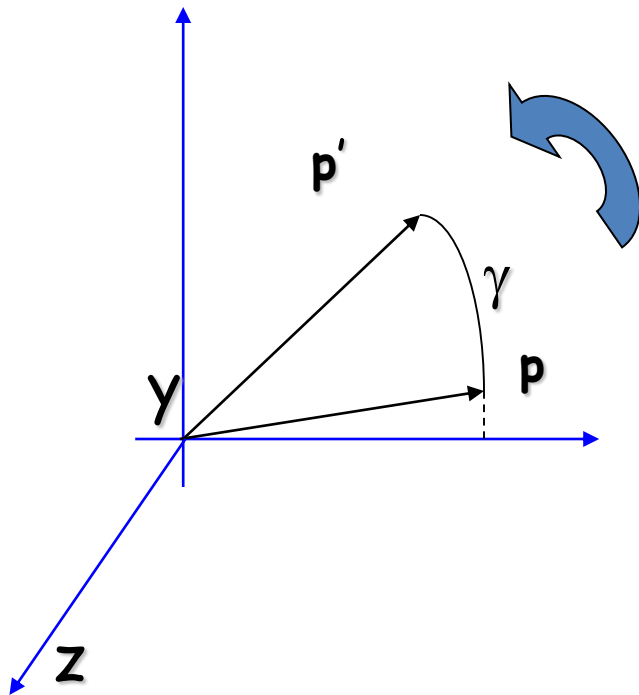
Intrinsic Assumptions Extrinsic Assumptions

- Pinhole camera model
- No rotation

$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{I} & \mathbf{t} \end{bmatrix} \mathbf{X} \quad \Rightarrow \quad w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \alpha & 0 & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & t_x \\ 0 & 1 & 0 & t_y \\ 0 & 0 & 1 & t_z \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

3D Rotation of Points

Rotation around the coordinate axes, **counter-clockwise**:



$$R_x(\alpha) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & -\sin \alpha \\ 0 & \sin \alpha & \cos \alpha \end{bmatrix}$$

$$R_y(\beta) = \begin{bmatrix} \cos \beta & 0 & \sin \beta \\ 0 & 1 & 0 \\ -\sin \beta & 0 & \cos \beta \end{bmatrix}$$

$$R_z(\gamma) = \begin{bmatrix} \cos \gamma & -\sin \gamma & 0 \\ \sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Allow camera rotation

$$\mathbf{x} = \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}$$



$$w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \alpha & s & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} t_x \\ t_y \\ t_z \\ 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

Degrees of freedom

$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \mathbf{X}$$



$$w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{matrix} 5 \\ \begin{bmatrix} \alpha & s & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{bmatrix} \end{matrix} \begin{matrix} 6 \\ \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \end{matrix} \begin{bmatrix} t_x \\ t_y \\ t_z \\ \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \end{bmatrix}$$

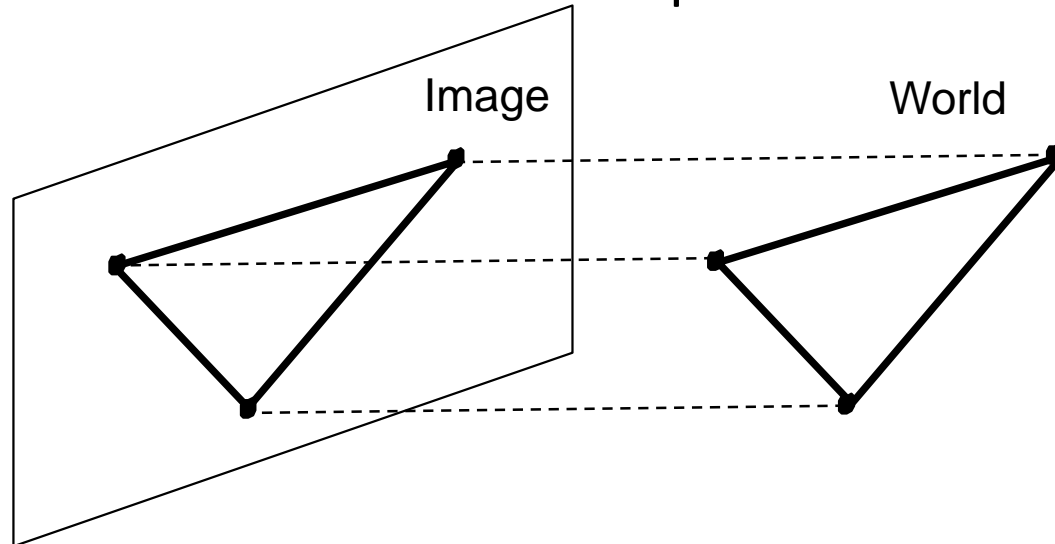
Vanishing Point = Projection from Infinity

$$\mathbf{p} = \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \begin{bmatrix} x \\ y \\ z \\ 0 \end{bmatrix} \Rightarrow \mathbf{p} = \mathbf{KR} \begin{bmatrix} x \\ y \\ z \end{bmatrix} \Rightarrow \mathbf{p} = \mathbf{K} \begin{bmatrix} x_R \\ y_R \\ z_R \end{bmatrix}$$

$$w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & u_0 \\ 0 & f & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_R \\ y_R \\ z_R \end{bmatrix} \Rightarrow \begin{aligned} u &= \frac{fx_R}{z_R} + u_0 \\ v &= \frac{fy_R}{z_R} + v_0 \end{aligned}$$

Scaled Orthographic Projection

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

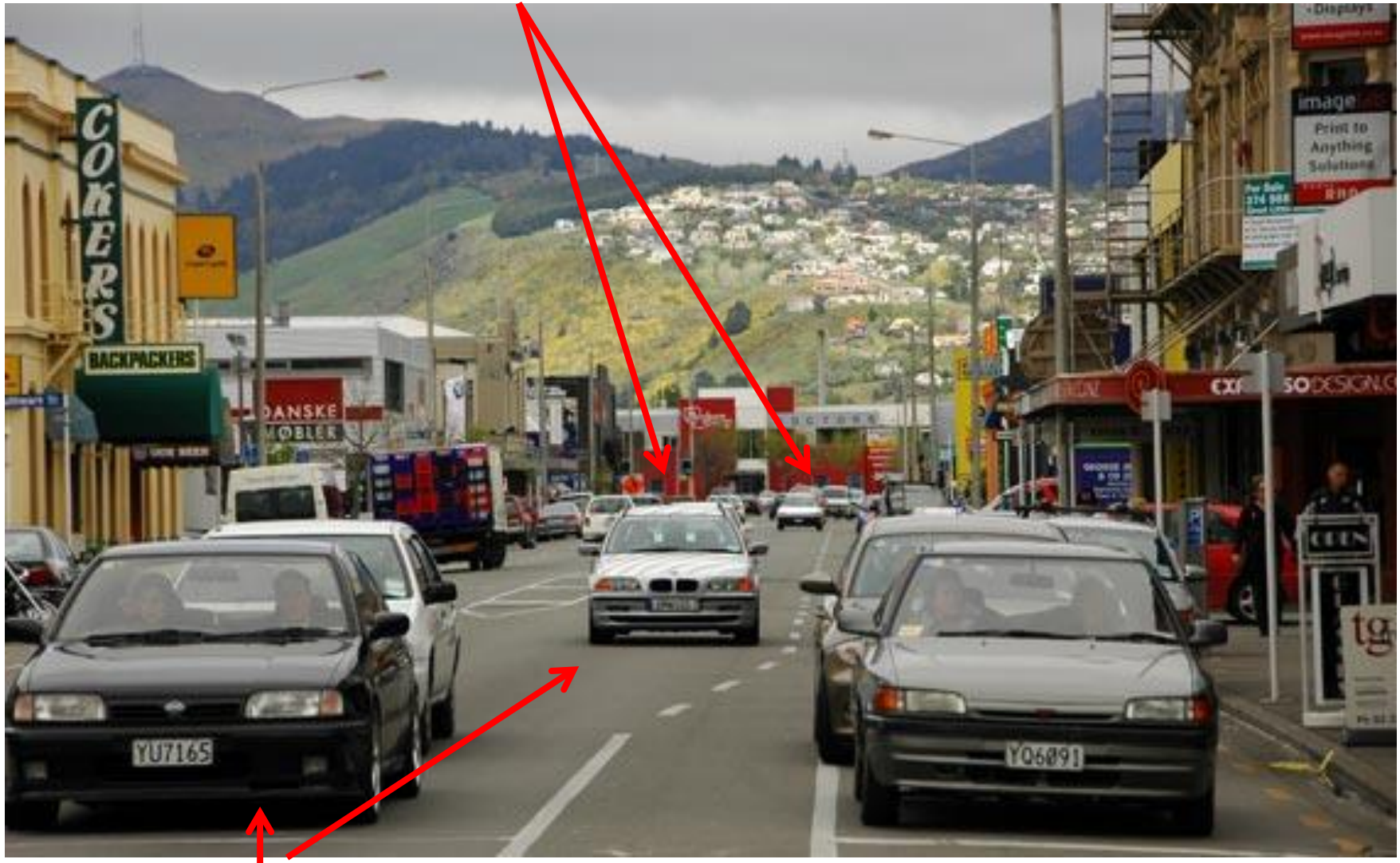


- Also called “weak perspective”

$$w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 0 & s \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

Example

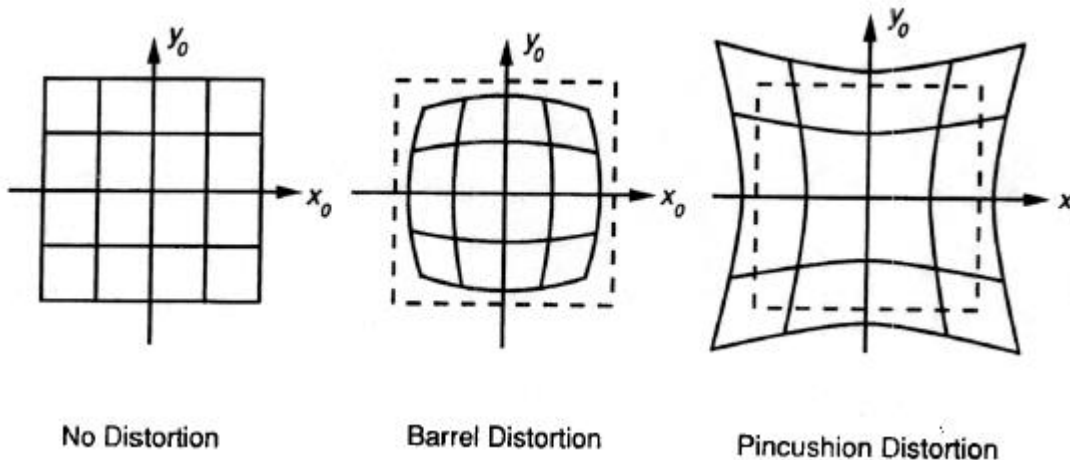
Far field: object appearance doesn't change as objects translate



Near field: object appearance changes as objects translate

Beyond Pinholes: Radial Distortion

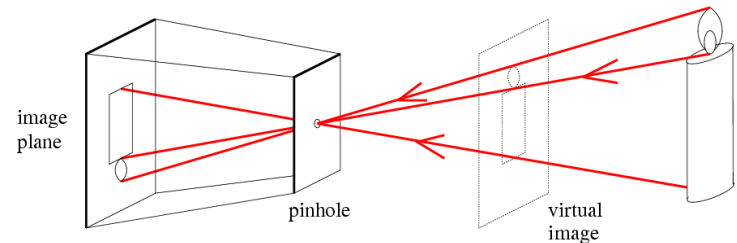
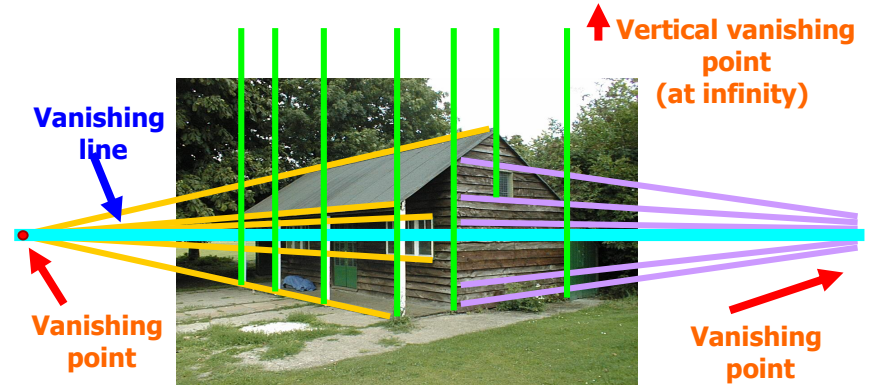
- Common in wide-angle lenses or for special applications (e.g., security)
- Creates non-linear terms in projection
- Usually handled by through solving for non-linear terms and then correcting image



Corrected Barrel Distortion

Things to remember

- Vanishing points and vanishing lines
- Pinhole camera model and camera projection matrix
- Homogeneous coordinates



$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \mathbf{X}$$

$$(x, y) \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

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