

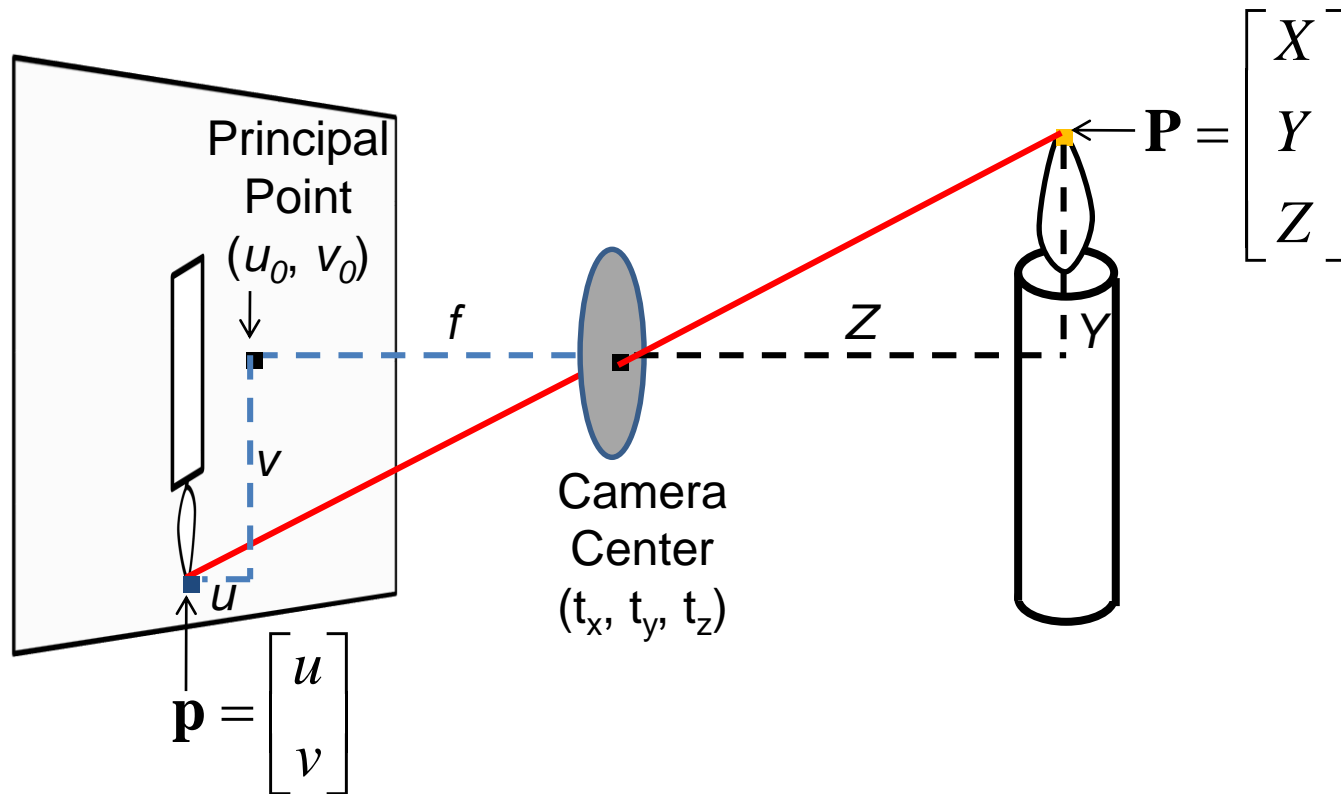
Single-view Metrology and Camera Calibration



Computer Vision

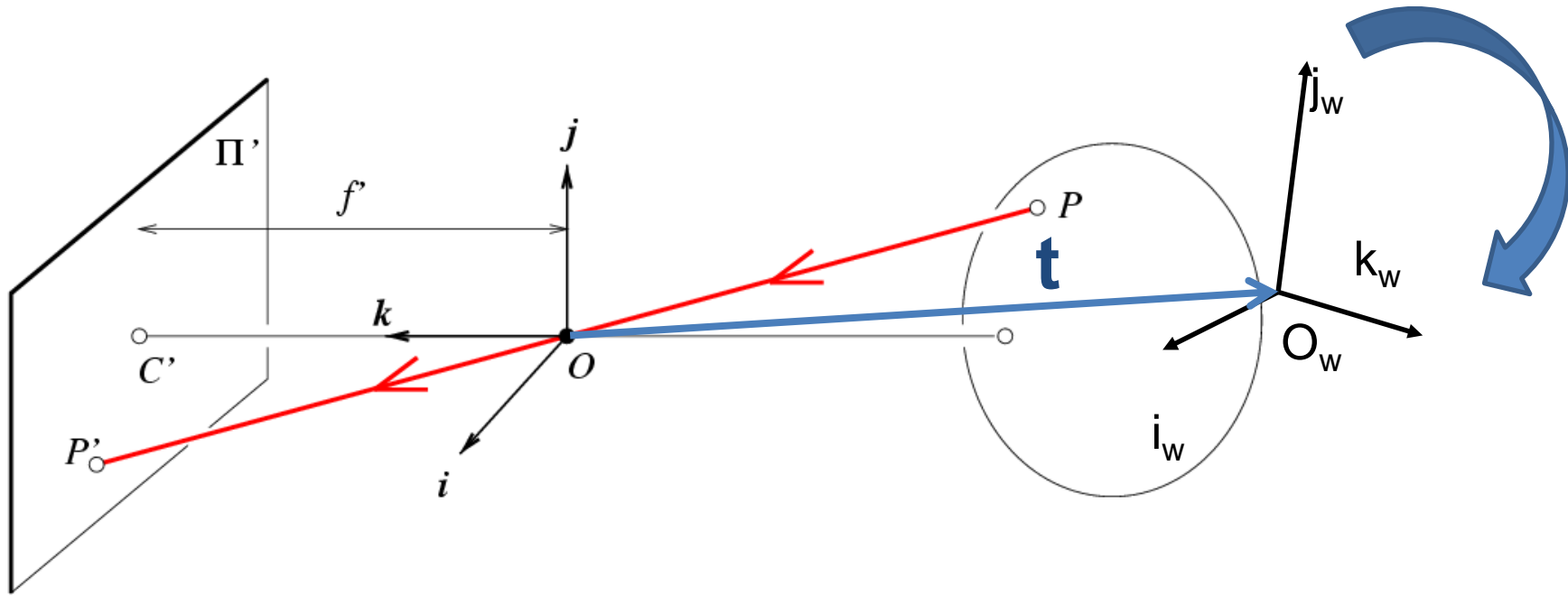
Derek Hoiem, University of Illinois

Last Class: Pinhole Camera



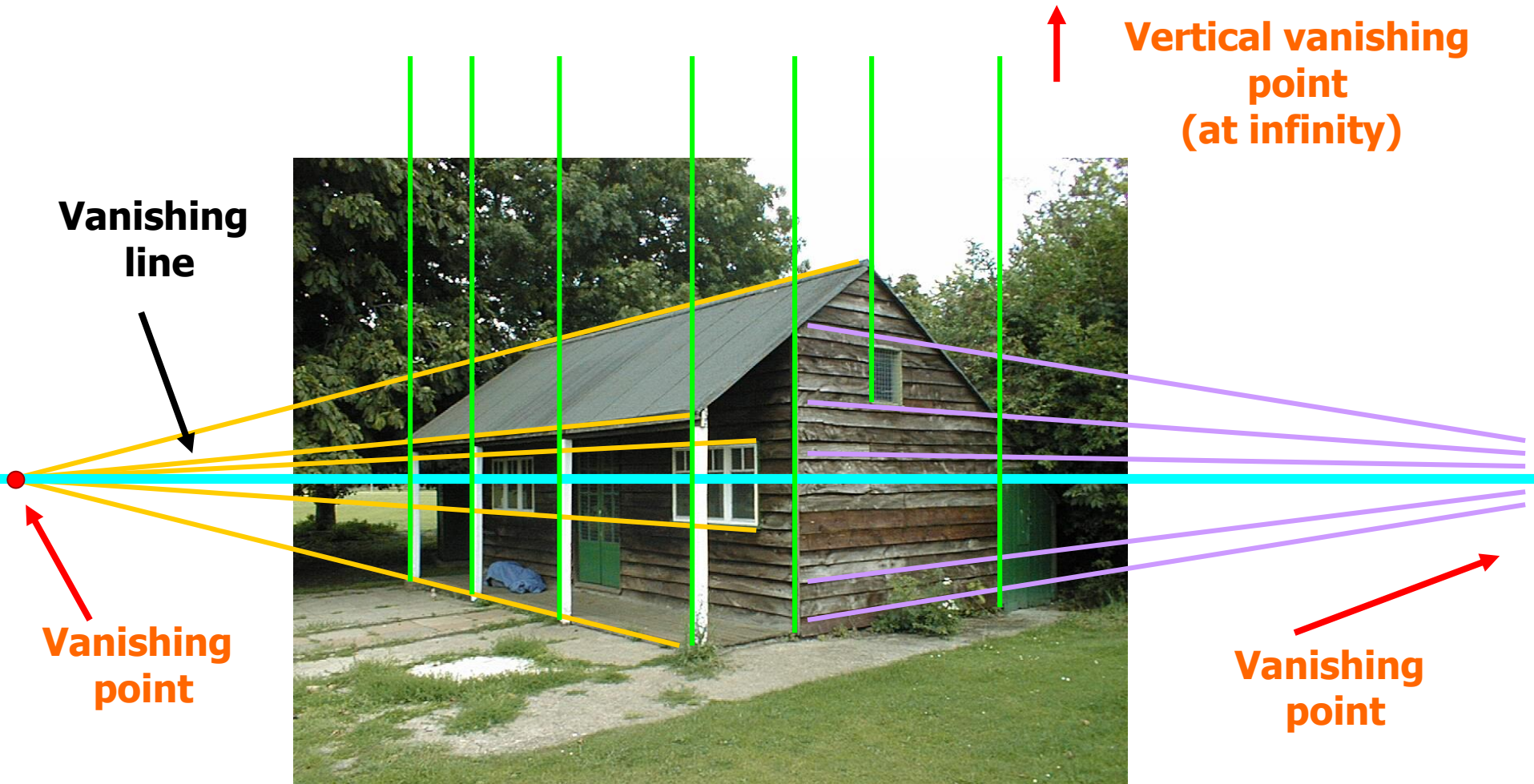
Last Class: Projection Matrix

R



$$\mathbf{x} = \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X} \rightarrow_w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & s & u_0 \\ 0 & cf & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Last class: Vanishing Points



This class

- How can we calibrate the camera?
- How can we measure the size of objects in the world from an image?
- What about other camera properties: focal length, field of view, depth of field, aperture, f-number?

How to calibrate the camera?

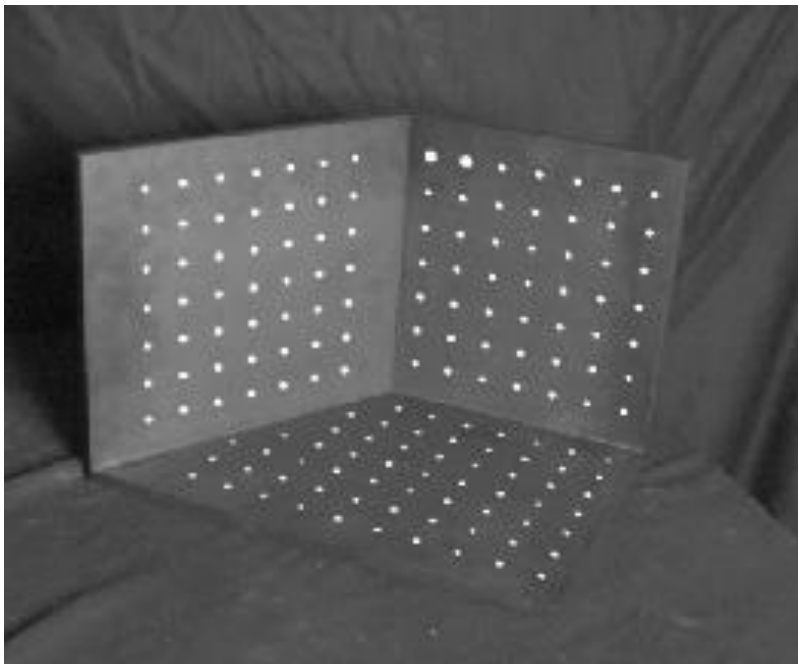
$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \mathbf{X}$$

$$\begin{bmatrix} wu \\ wv \\ w \end{bmatrix} = \begin{bmatrix} * & * & * & * \\ * & * & * & * \\ * & * & * & * \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Calibrating the Camera

Method 1: Use an object (calibration grid) with known geometry

- Correspond image points to 3d points
- Get least squares solution (or non-linear solution)



$$\begin{bmatrix} wu \\ wv \\ w \end{bmatrix} = \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Linear method

- Solve using linear least squares

$$\begin{bmatrix} wu \\ wv \\ w \end{bmatrix} = \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} X_1 & Y_1 & Z_1 & 1 & 0 & 0 & 0 & 0 & -u_1 X_1 & -u_1 Y_1 & -u_1 Z_1 & -u_1 \\ 0 & 0 & 0 & 0 & X_1 & Y_1 & Z_1 & 1 & -v_1 X_1 & -v_1 Y_1 & -v_1 Z_1 & -v_1 \\ & & & & & & \vdots & & & & & \\ X_n & Y_n & Z_n & 1 & 0 & 0 & 0 & 0 & -u_n X_n & -u_n Y_n & -u_n Z_n & -u_n \\ 0 & 0 & 0 & 0 & X_n & Y_n & Z_n & 1 & -v_n X_n & -v_n Y_n & -v_n Z_n & -v_n \end{bmatrix} \begin{bmatrix} m_{11} \\ m_{12} \\ m_{13} \\ m_{14} \\ m_{21} \\ m_{22} \\ m_{23} \\ m_{24} \\ m_{31} \\ m_{32} \\ m_{33} \\ m_{34} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{bmatrix} \quad \mathbf{Ax=0} \text{ form}$$

Calibration with linear method

- Advantages
 - Easy to formulate and solve
 - Provides initialization for non-linear methods
- Disadvantages
 - Doesn't directly give you camera parameters
 - Doesn't model radial distortion
 - Can't impose constraints, such as known focal length
 - Doesn't minimize projection error
- Non-linear methods are preferred
 - Define error as difference between projected points and measured points
 - Minimize error using Newton's method or other non-linear optimization

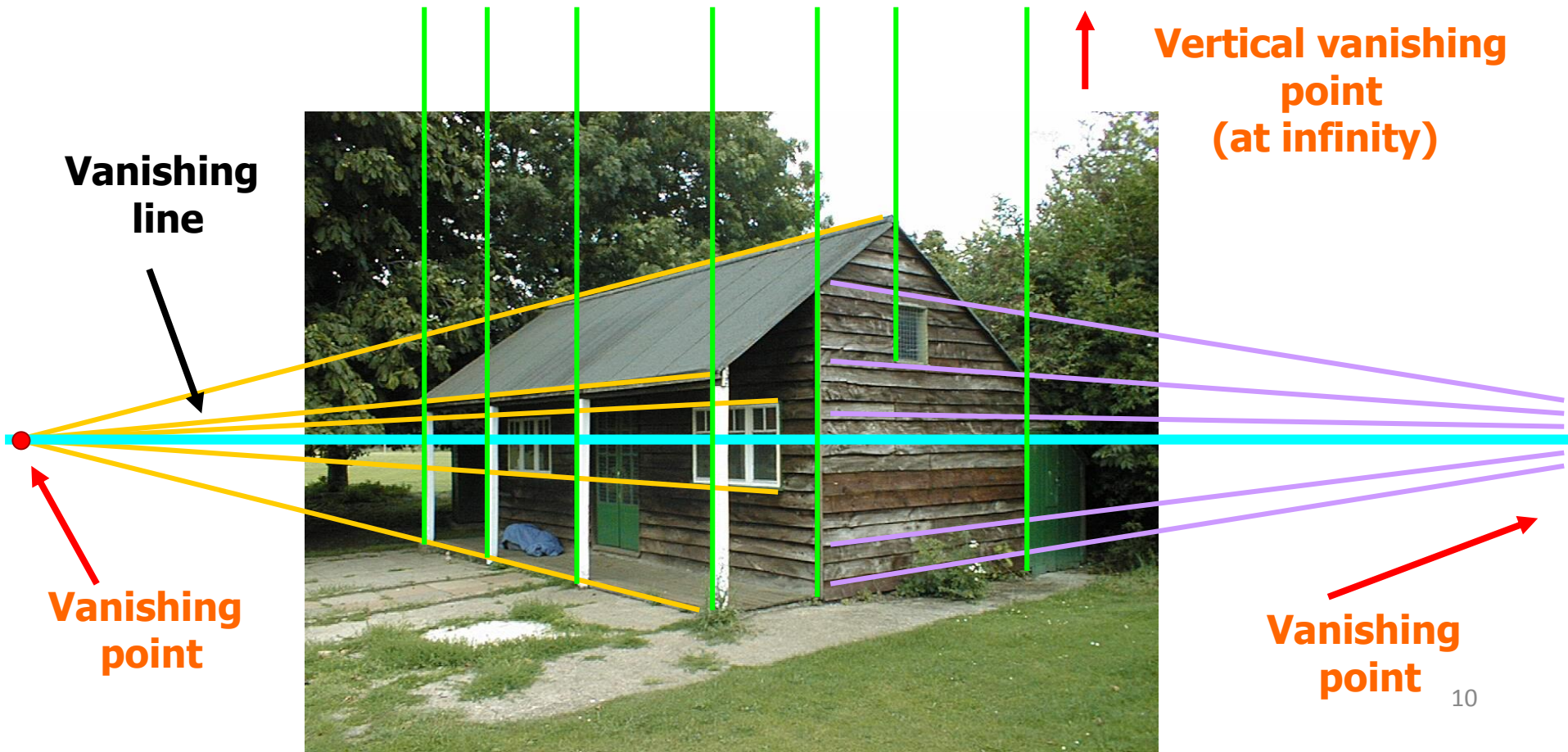
Can solve for explicit camera parameters:

<http://ksimek.github.io/2012/08/14/decompose/>

Calibrating the Camera

Method 2: Use vanishing points

- Find vanishing points corresponding to orthogonal directions



Calibration by orthogonal vanishing points

- Intrinsic camera matrix
 - Use orthogonality as a constraint
 - Model K with only f, u_0, v_0

$$\mathbf{p}_i = \mathbf{KRX}_i$$

For vanishing points

$$\mathbf{X}_i^T \mathbf{X}_j = 0$$

- What if you don't have three finite vanishing points?
 - Two finite VP: solve f , get valid u_0, v_0 closest to image center
 - One finite VP: u_0, v_0 is at vanishing point; can't solve for f

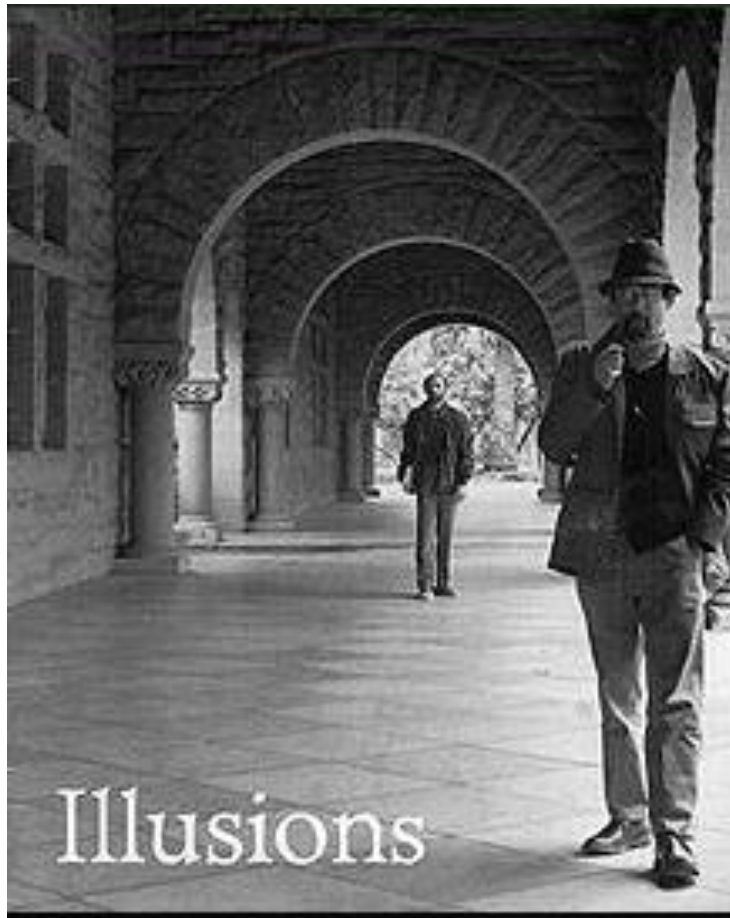
Calibration by vanishing points

- Intrinsic camera matrix

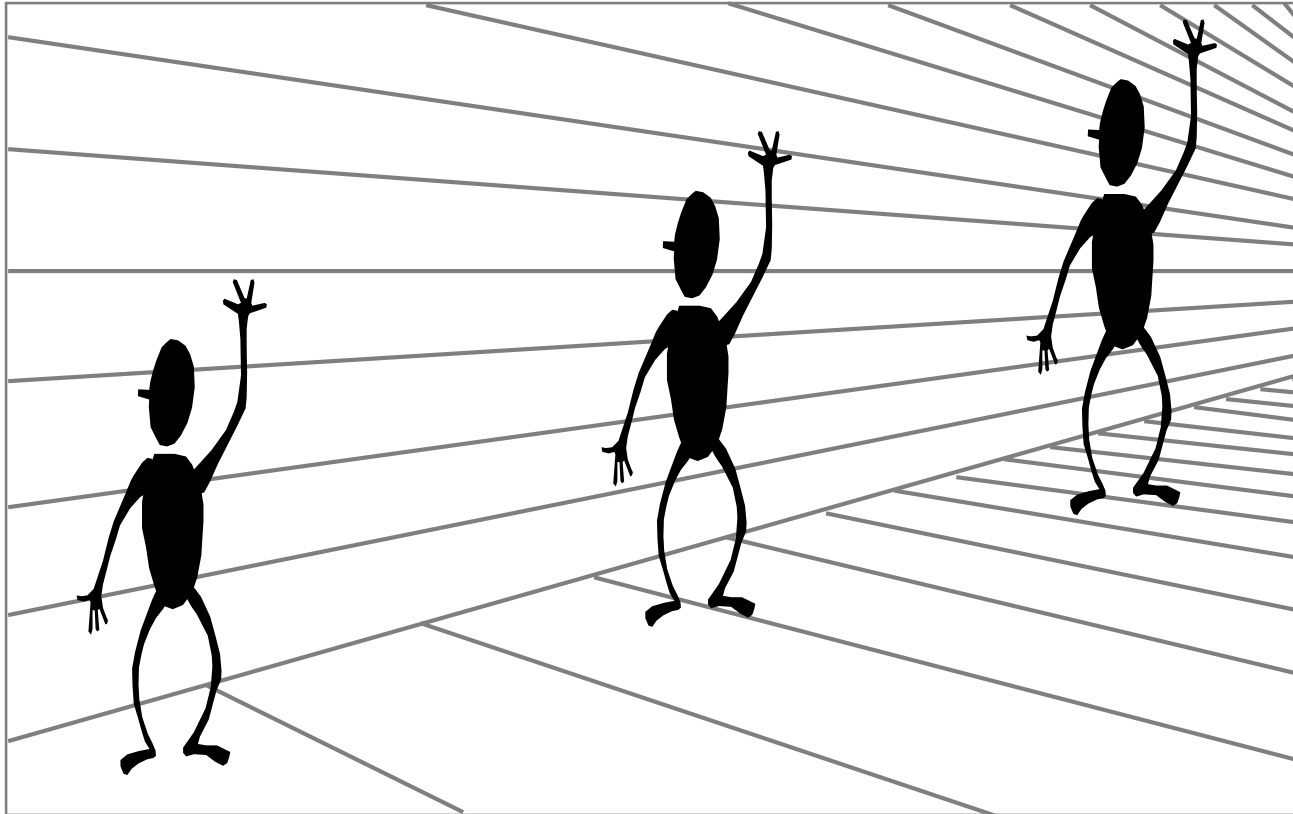
$$\mathbf{p}_i = \mathbf{KRX}_i$$

- Rotation matrix
 - Set directions of vanishing points
 - e.g., $\mathbf{X}_1 = [1, 0, 0]$
 - Each VP provides one column of \mathbf{R}
 - Special properties of \mathbf{R}
 - $\text{inv}(\mathbf{R}) = \mathbf{R}^T$
 - Each row and column of \mathbf{R} has unit length

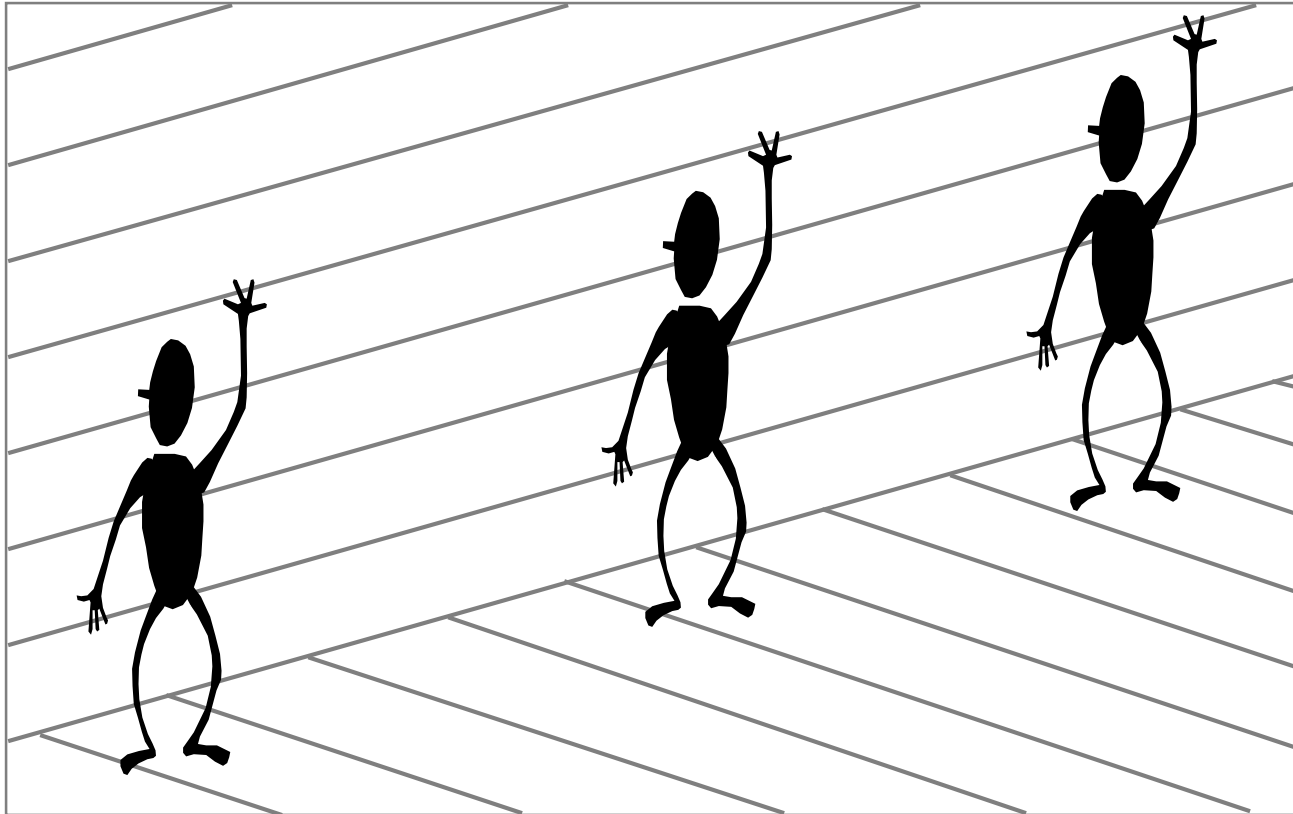
How can we measure the size of 3D objects from an image?



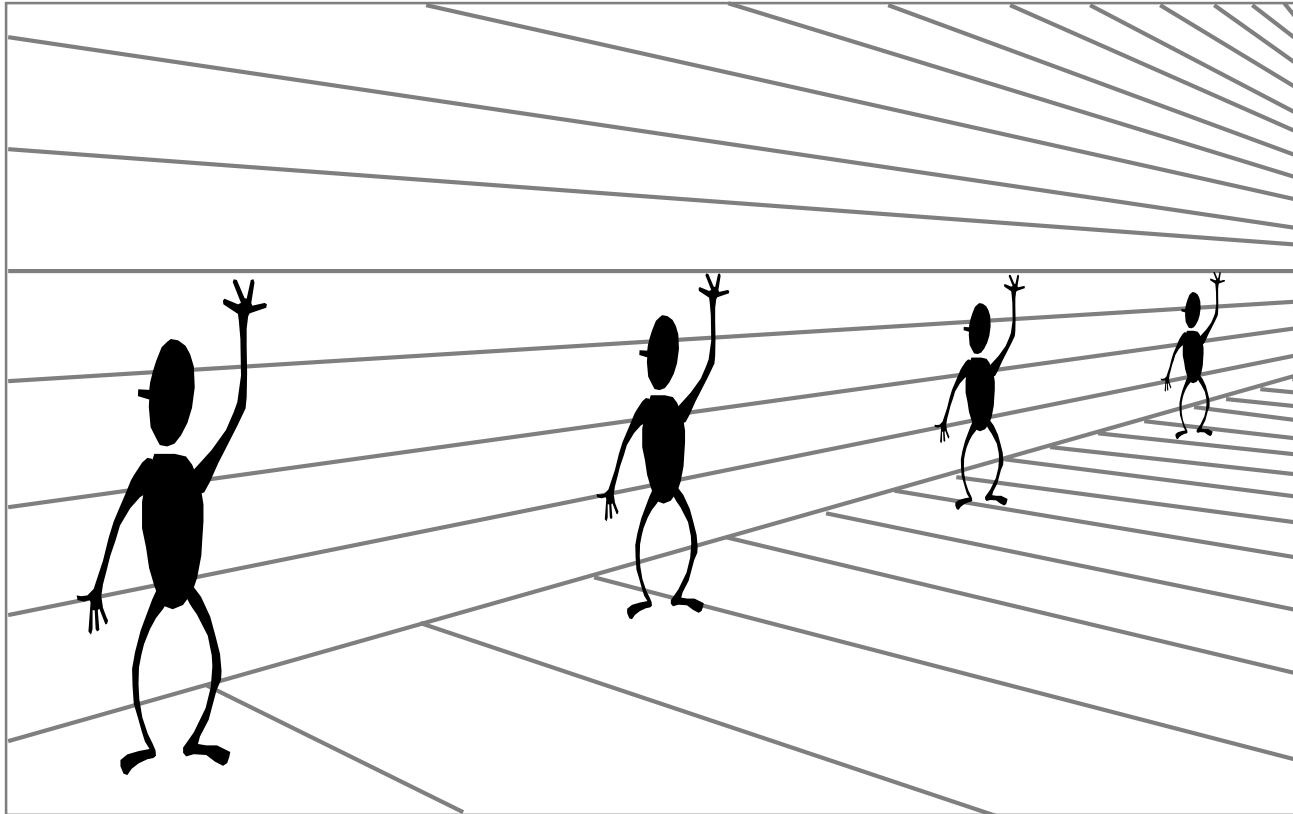
Perspective cues



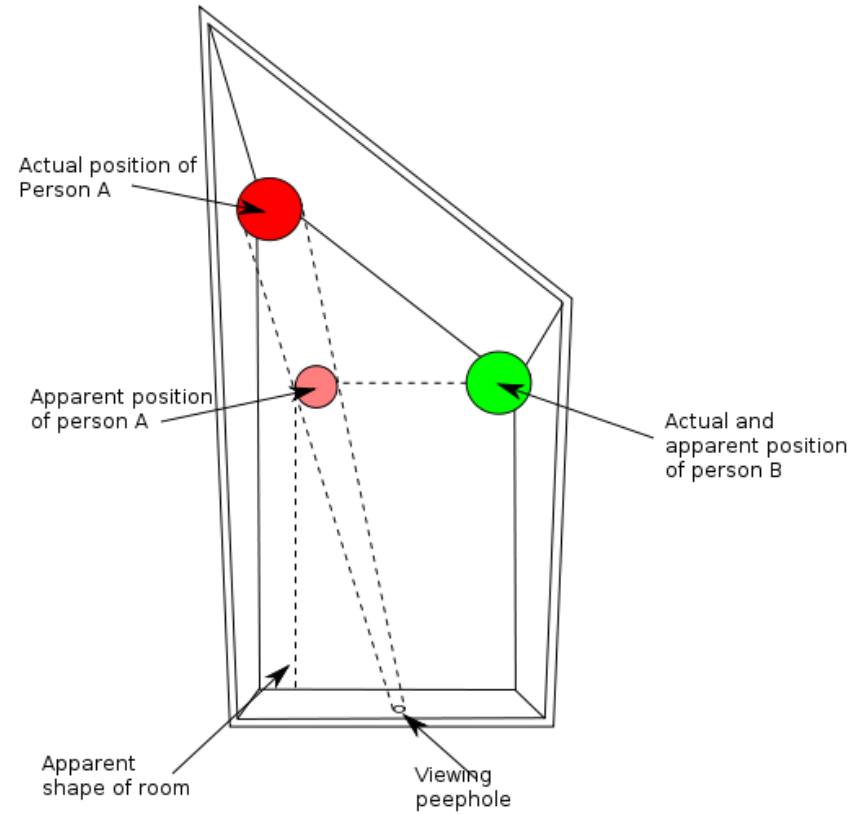
Perspective cues



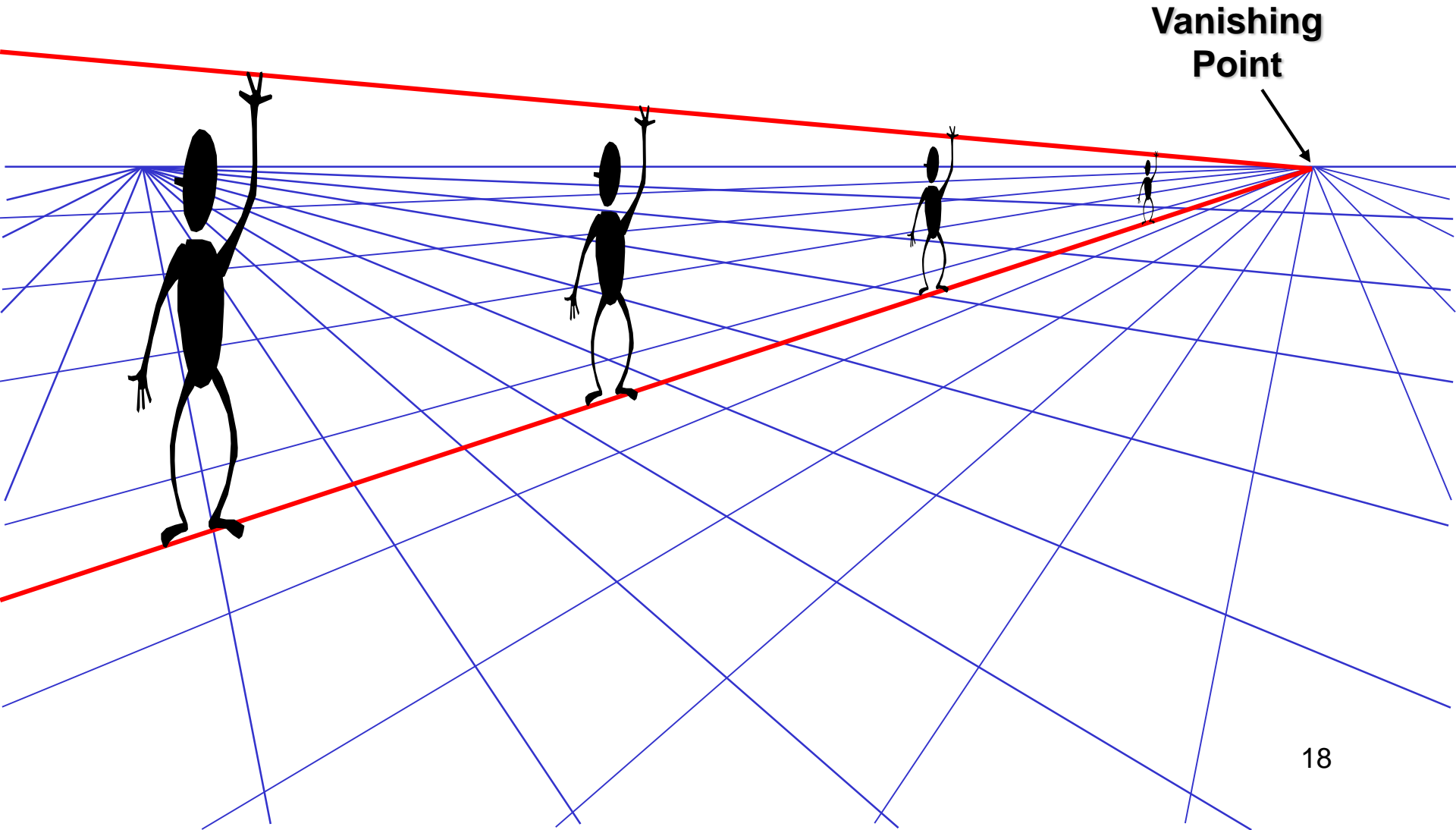
Perspective cues



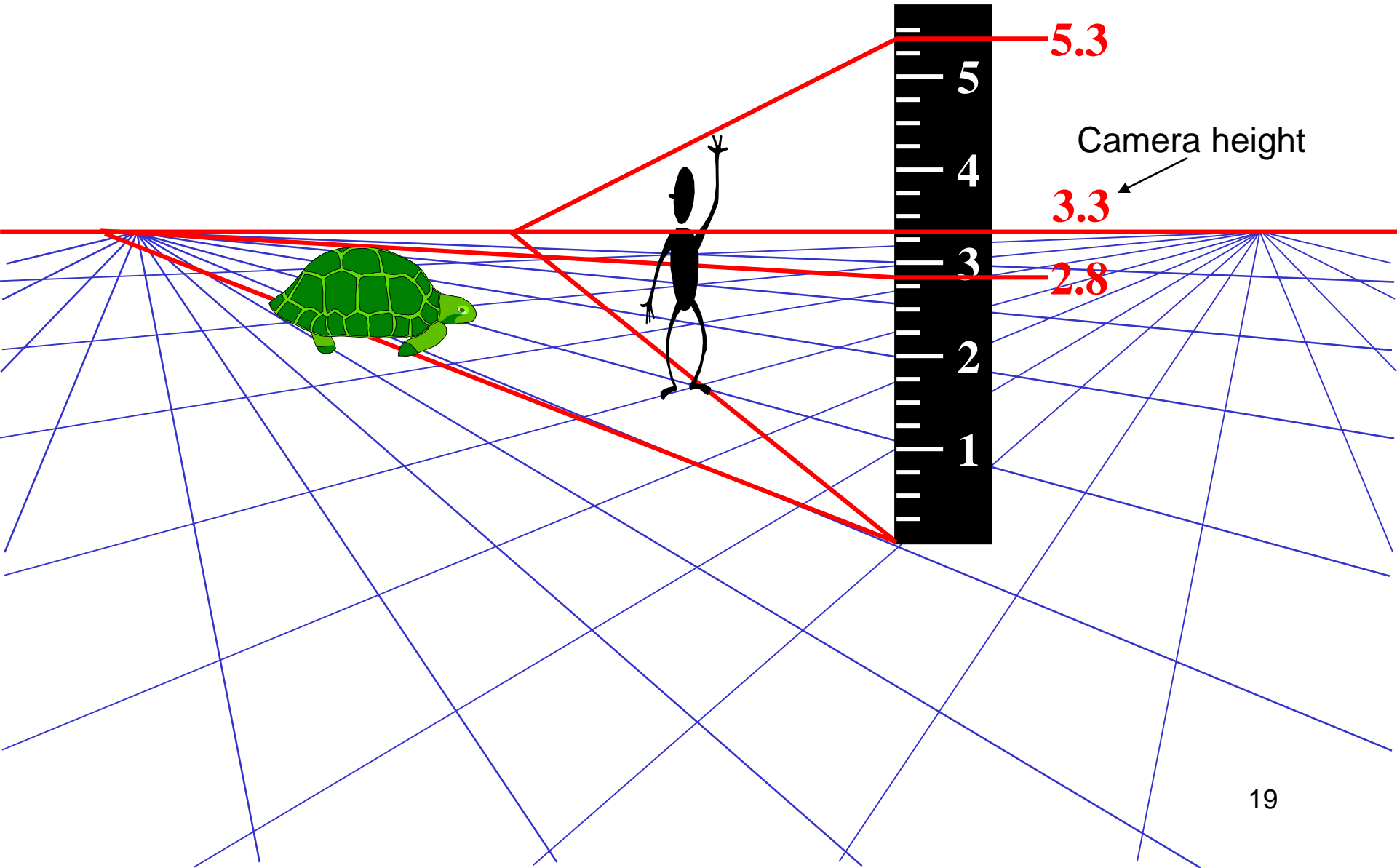
Ames Room



Comparing heights



Measuring height



Which is higher – the camera or the man in the parachute?

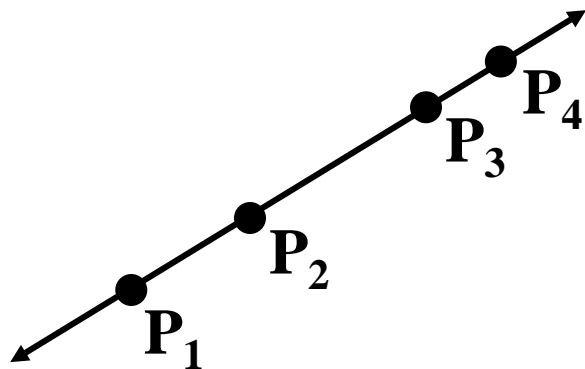


The cross ratio

A Projective Invariant

- Something that does not change under projective transformations (including perspective projection)

The cross-ratio of 4 collinear points



$$\frac{\| \mathbf{P}_3 - \mathbf{P}_1 \| \| \mathbf{P}_4 - \mathbf{P}_2 \|}{\| \mathbf{P}_3 - \mathbf{P}_2 \| \| \mathbf{P}_4 - \mathbf{P}_1 \|}$$

$$\mathbf{P}_i = \begin{bmatrix} X_i \\ Y_i \\ Z_i \\ 1 \end{bmatrix}$$

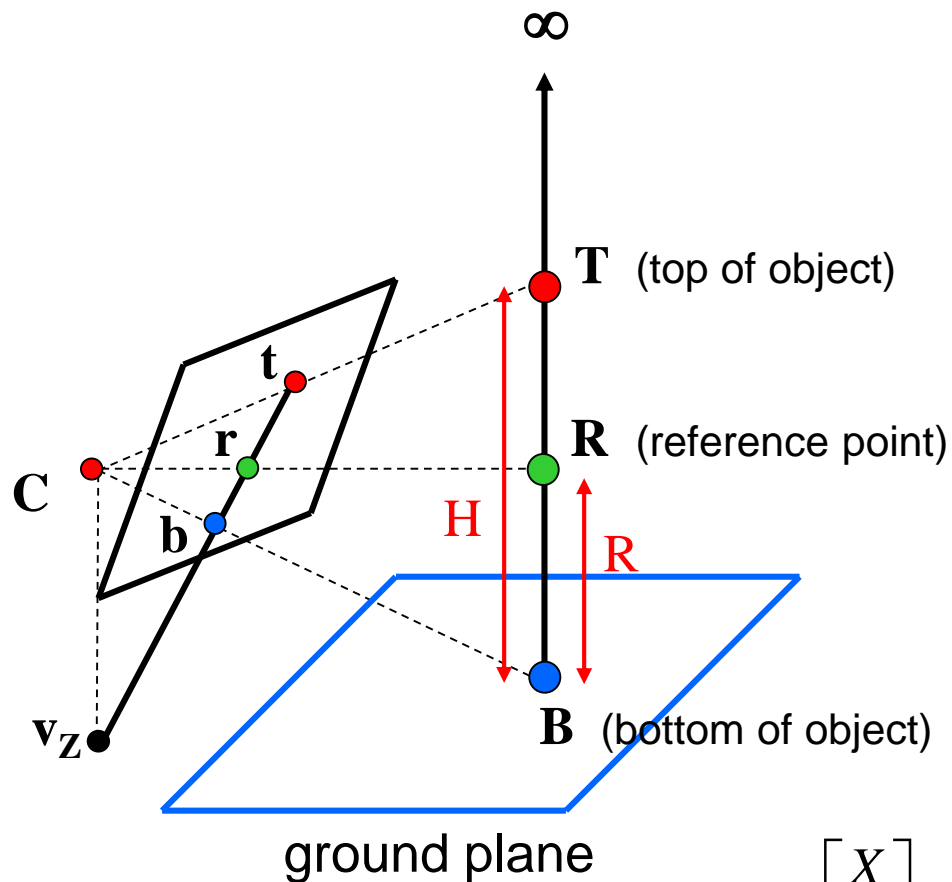
Can permute the point ordering

$$\frac{\| \mathbf{P}_1 - \mathbf{P}_3 \| \| \mathbf{P}_4 - \mathbf{P}_2 \|}{\| \mathbf{P}_1 - \mathbf{P}_2 \| \| \mathbf{P}_4 - \mathbf{P}_3 \|}$$

- $4! = 24$ different orders (but only 6 distinct values)

This is the fundamental invariant of projective geometry

Measuring height



scene points represented as $\mathbf{P} = \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$

image points as $\mathbf{p} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$

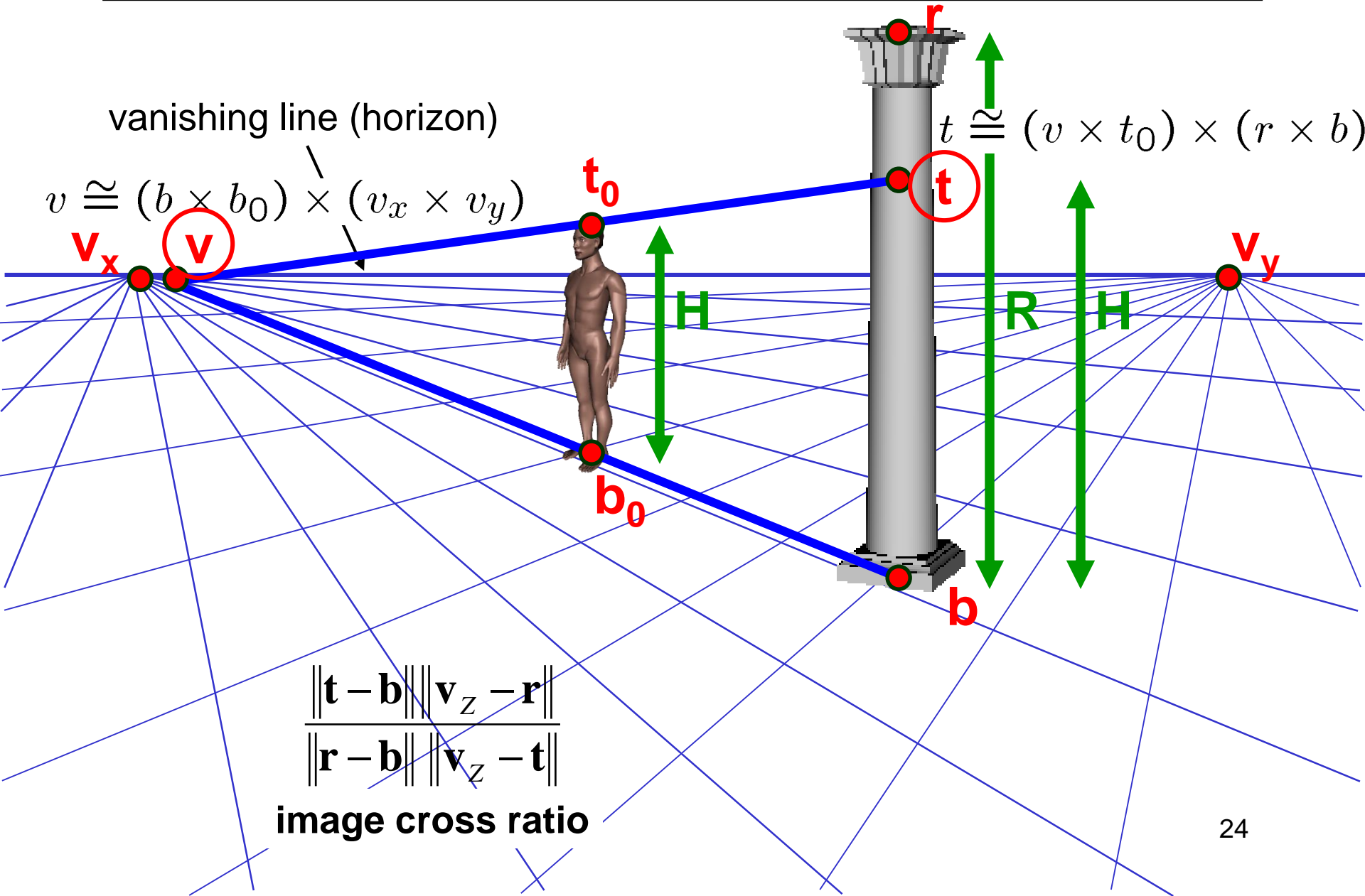
$$\frac{\|\mathbf{B} - \mathbf{T}\| \|\infty - \mathbf{R}\|}{\|\mathbf{B} - \mathbf{R}\| \|\infty - \mathbf{T}\|} = \frac{H}{R}$$

scene cross ratio

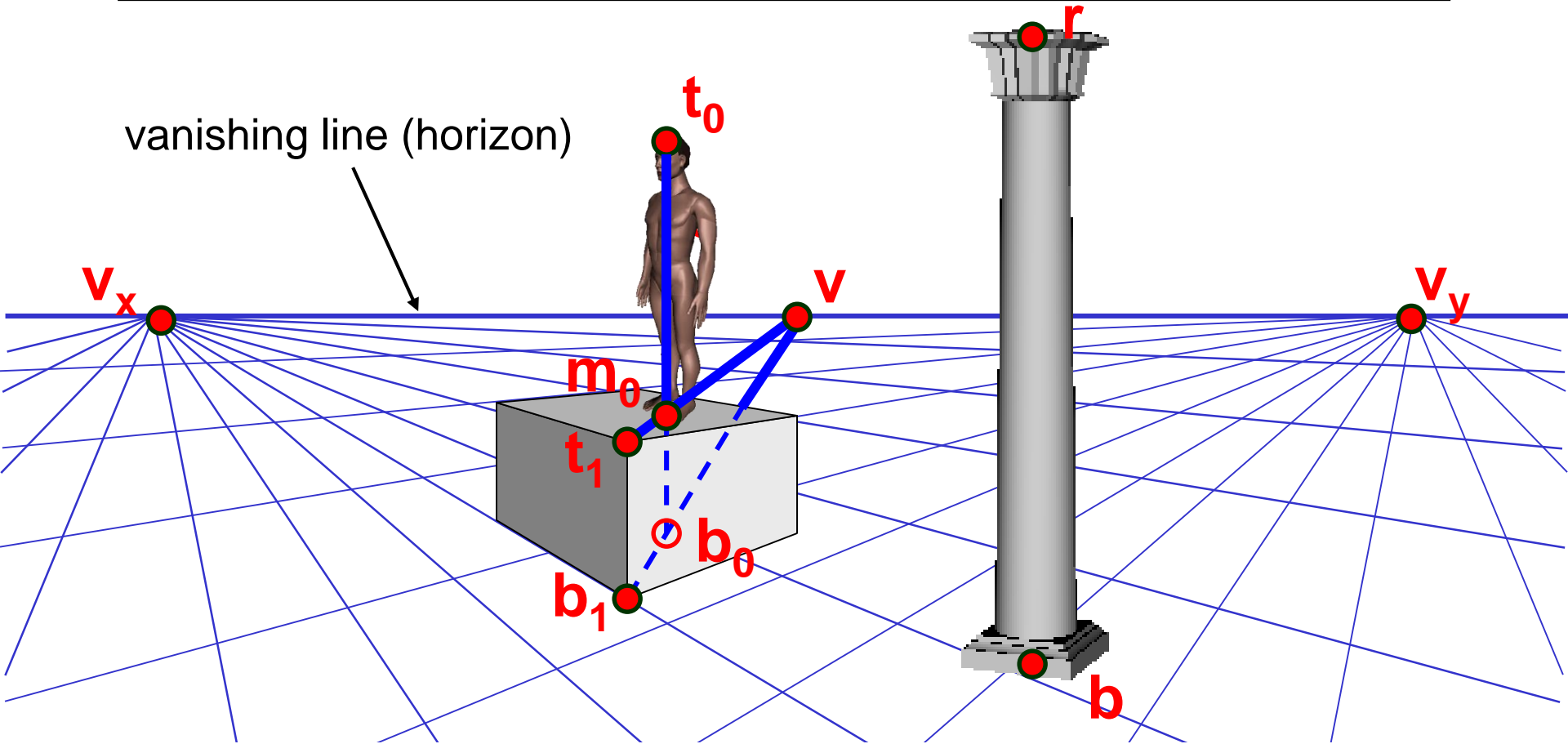
$$\frac{\|\mathbf{b} - \mathbf{t}\| \|\mathbf{v}_Z - \mathbf{r}\|}{\|\mathbf{b} - \mathbf{r}\| \|\mathbf{v}_Z - \mathbf{t}\|} = \frac{H}{R}$$

image cross ratio

Measuring height



Measuring height

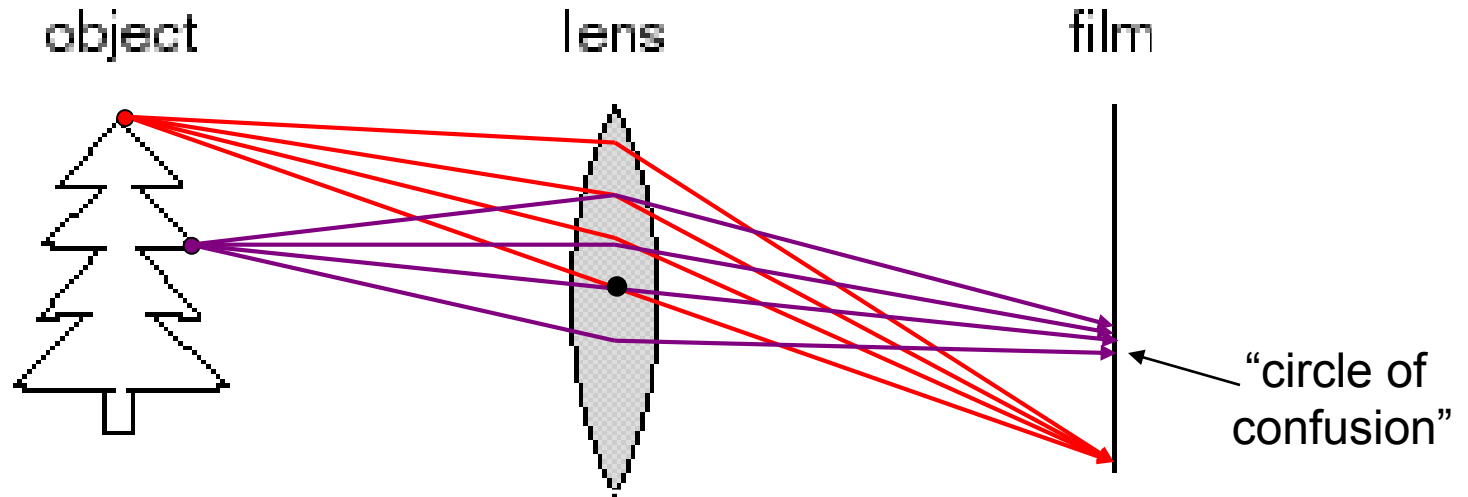


What if the point on the ground plane \mathbf{b}_0 is not known?

- Here the guy is standing on the box, height of box is known
- Use one side of the box to help find \mathbf{b}_0 as shown above

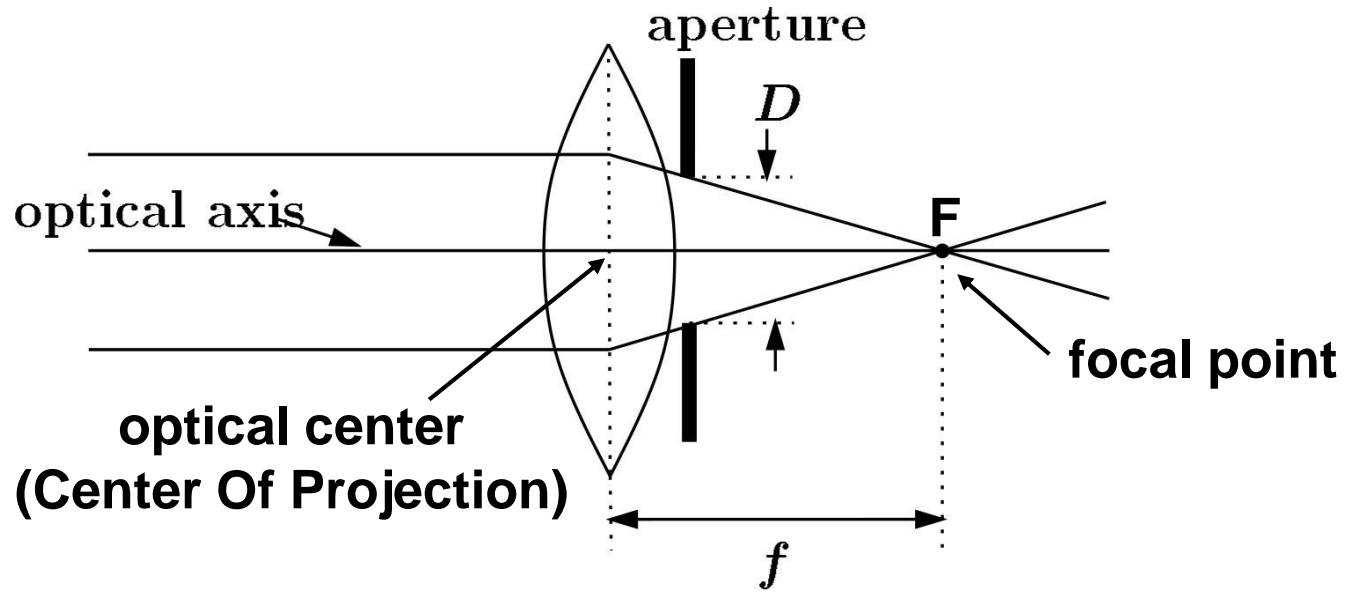
What about focus, aperture, DOF, FOV, etc?

Adding a lens



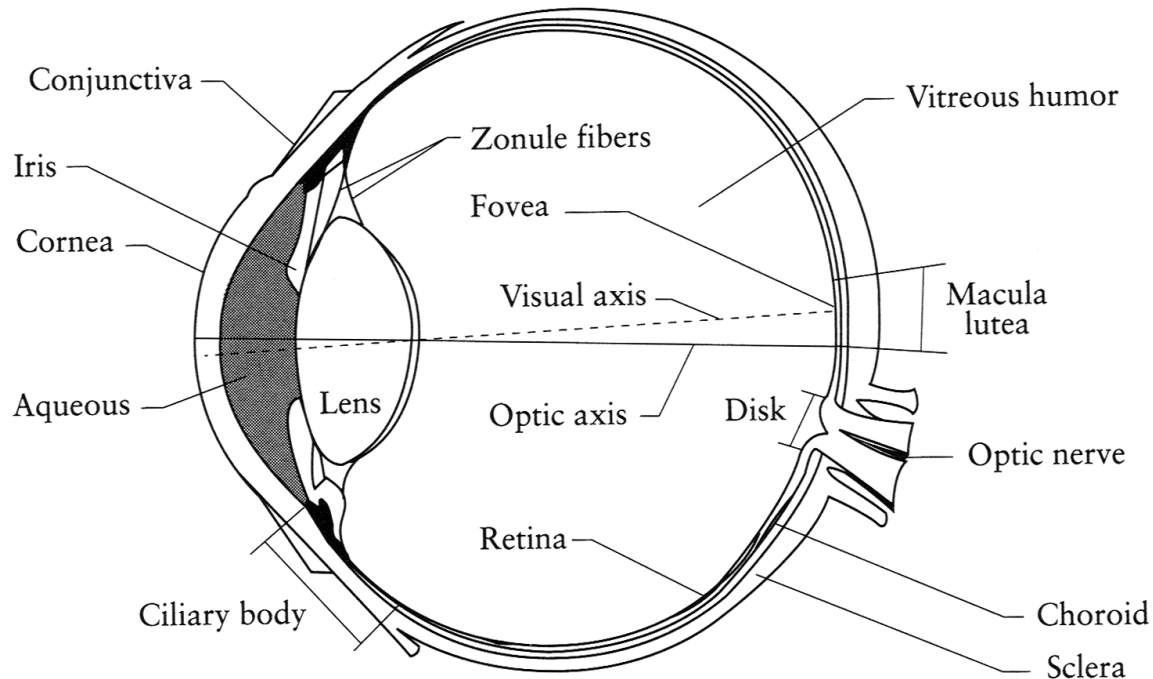
- A lens focuses light onto the film
 - There is a specific distance at which objects are “in focus”
 - other points project to a “circle of confusion” in the image
 - Changing the shape of the lens changes this distance

Focal length, aperture, depth of field



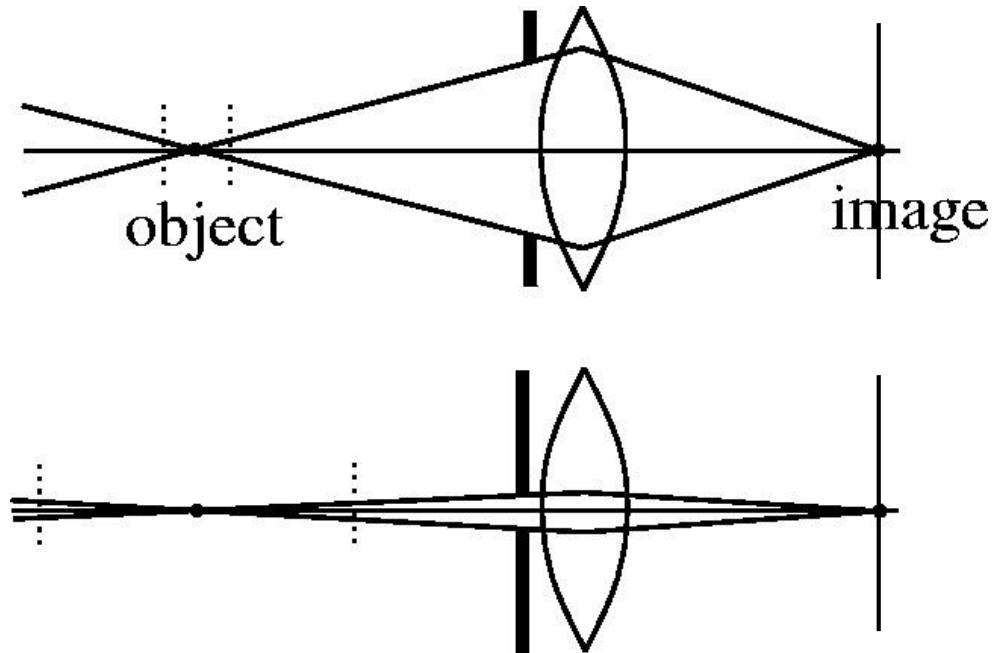
- A lens focuses parallel rays onto a single focal point
- focal point at a distance f beyond the plane of the lens
 - Aperture of diameter D restricts the range of rays

The eye



- The human eye is a camera
 - **Iris** - colored annulus with radial muscles
 - **Pupil** (aperture) - the hole whose size is controlled by the iris
 - **Retina** (film): photoreceptor cells (rods and cones)

Depth of field



$f/5.6$



$f/32$

Changing the aperture size or focal length affects depth of field

Varying the aperture

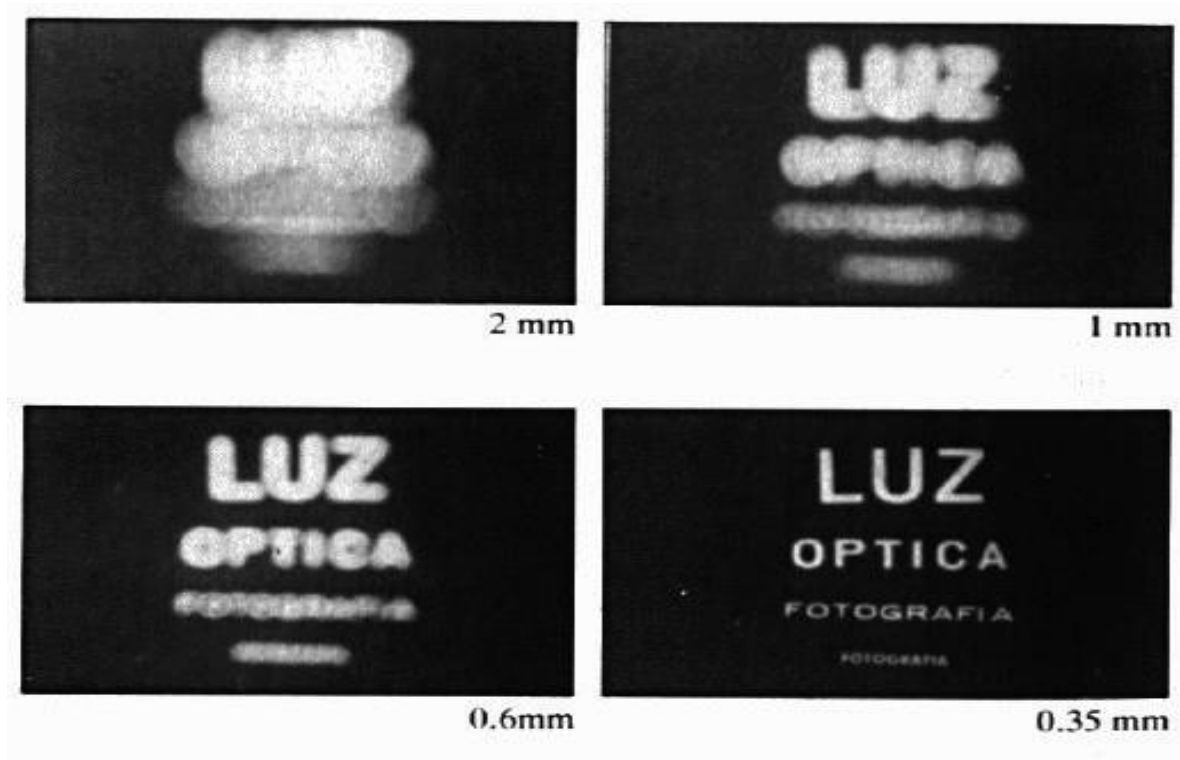


Large aperture = small DOF



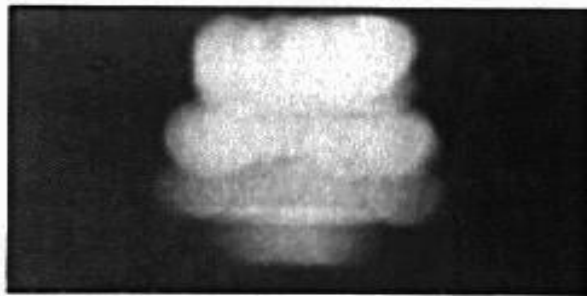
Small aperture = large DOF

Shrinking the aperture

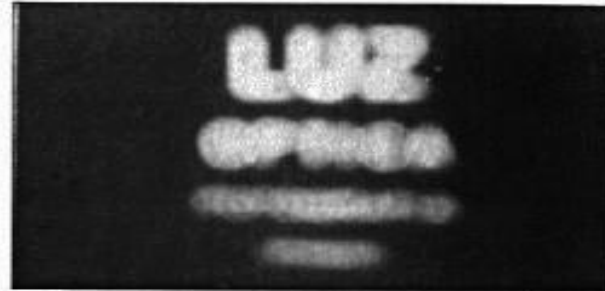


- Why not make the aperture as small as possible?
 - Less light gets through
 - Diffraction effects

Shrinking the aperture



2 mm



1 mm



0.6mm



0.35 mm



0.15 mm



0.07 mm

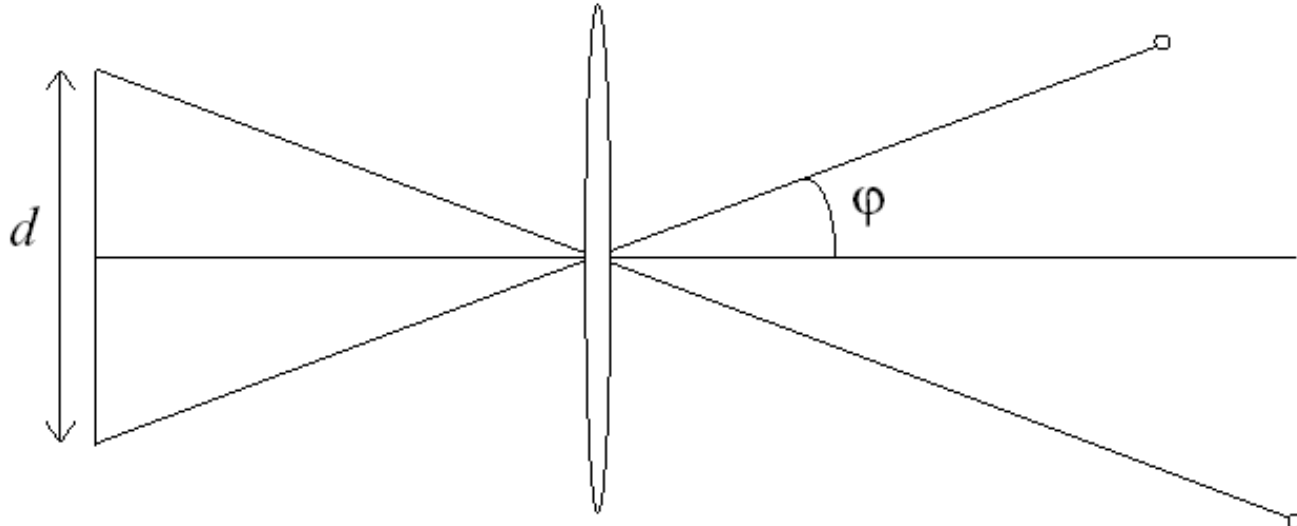
Relation between field of view and focal length

Field of view (angle width)

Film/Sensor Width

$$fov = \tan^{-1} \frac{d}{2f}$$

Focal length



Dolly Zoom or “Vertigo Effect”

<http://www.youtube.com/watch?v=NB4bikrNzMk>

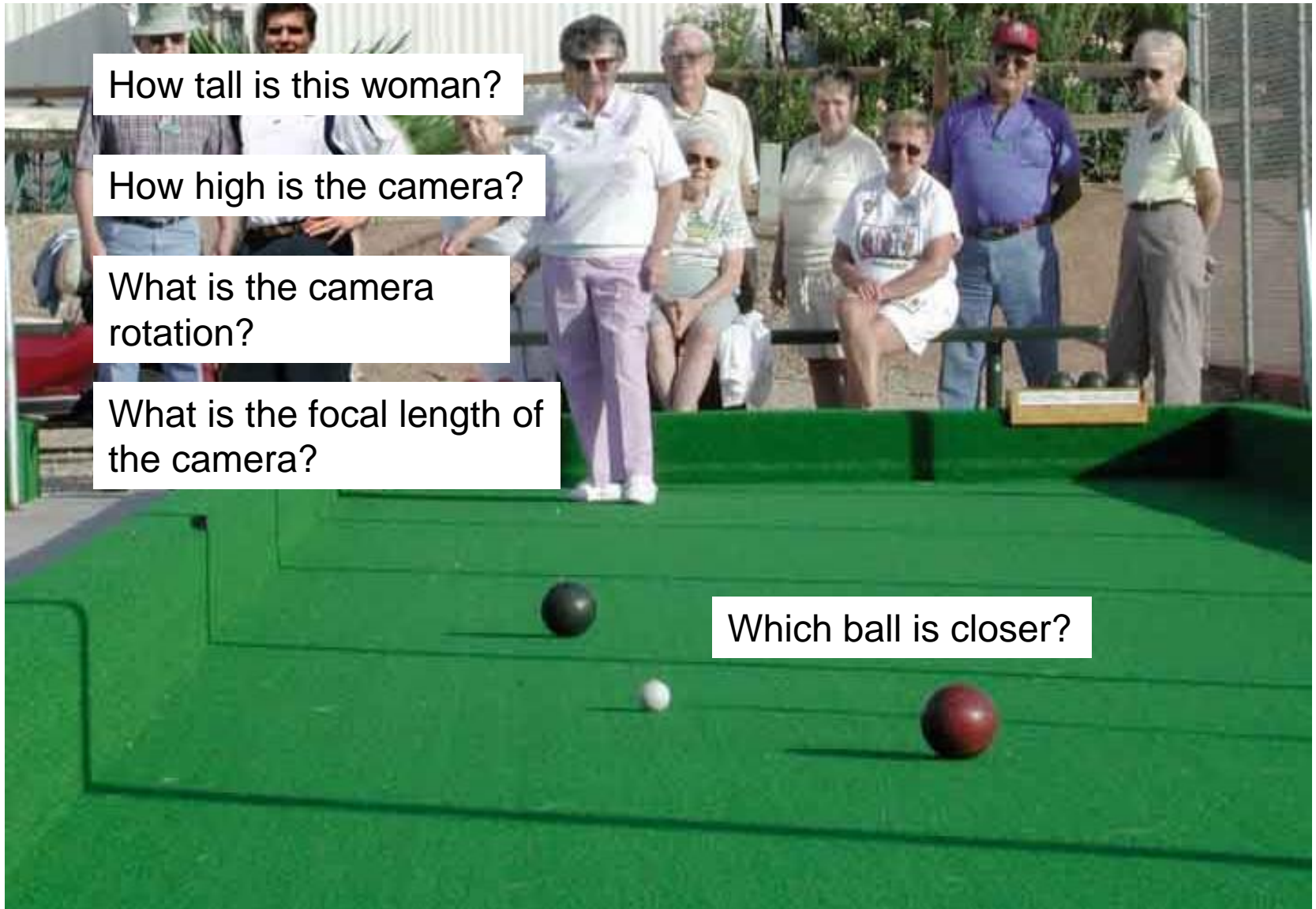


How is this done?

Zoom in while
moving away

http://en.wikipedia.org/wiki/Focal_length

Review



Next class

- Image stitching

