

Single-View Geometry

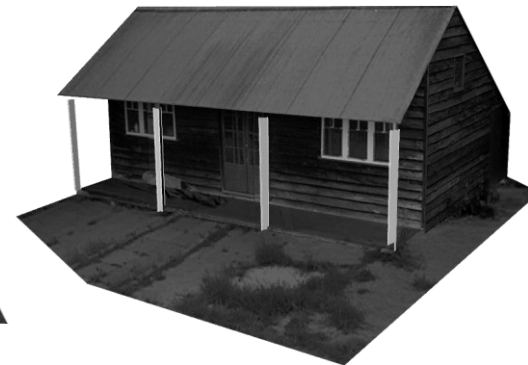
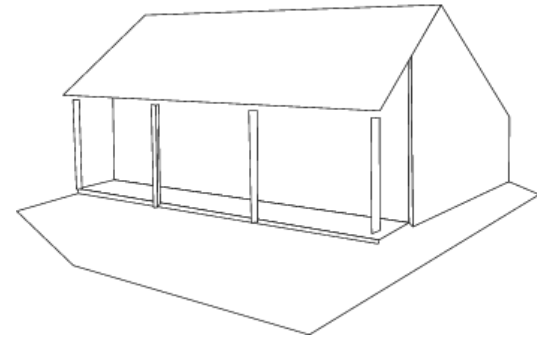
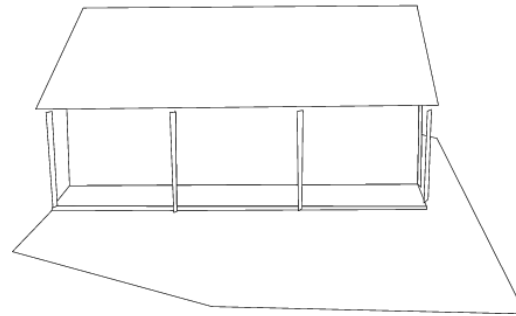
EECS 442 – David Fouhey and Justin Johnson
Winter 2021, University of Michigan

<https://web.eecs.umich.edu/~justincj/teaching/eecs442/WI2021/>

Updates

- HW6 is Optional. We'll still help you through it and think it's valuable but if you don't have time, no need to do it.
- Many project proposals turned in yesterday. We'll try to respond quickly

Application: Single-view modeling



A. Criminisi, I. Reid, and A. Zisserman,
[Single View Metrology](#), IJCV 2000

Application: Measuring Height



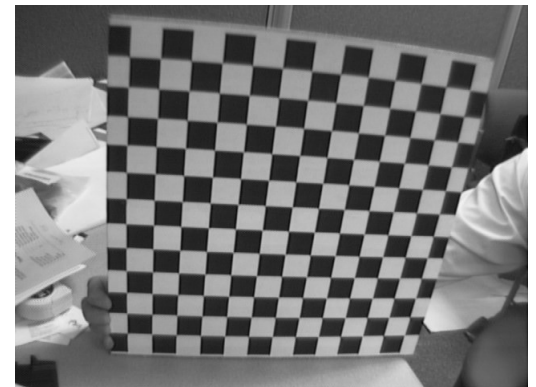
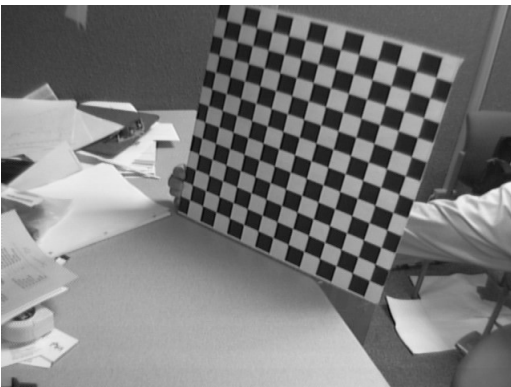
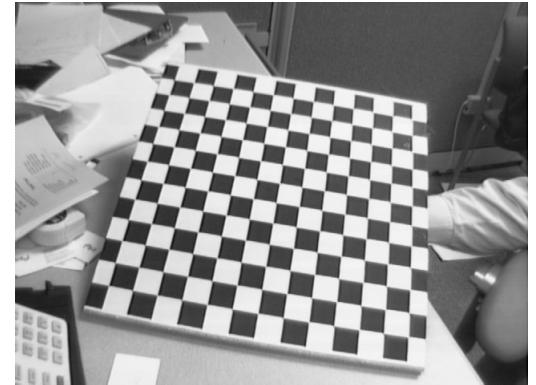
Application: Measuring Height



- CSI before CSI
- Covered criminal cases talking to random scientists (e.g., footwear experts)
- How do you tell how tall someone is if they're not kind enough to stand next to a ruler?

Application: Camera Calibration

Calibration a HUGE pain



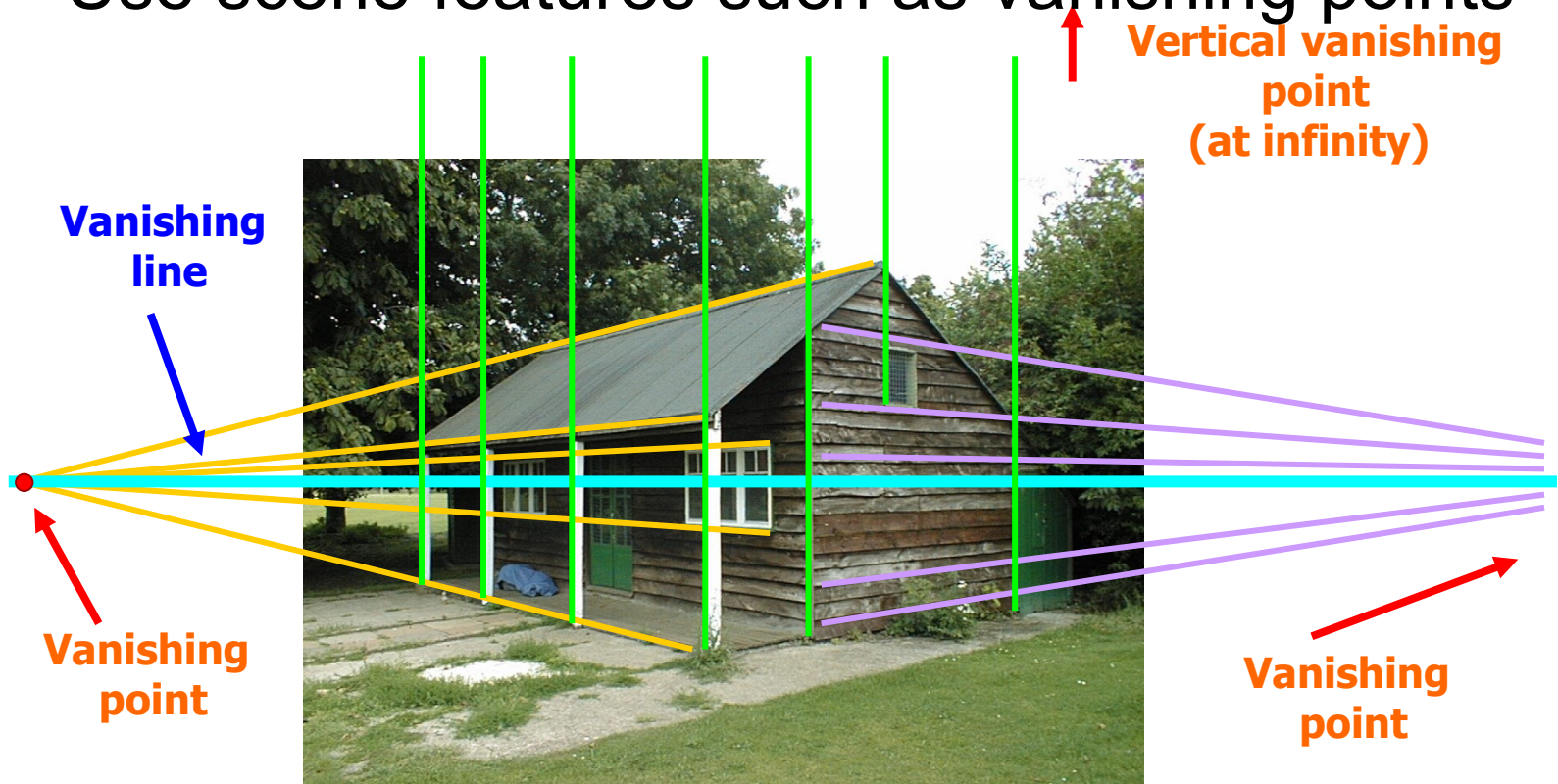
Application: Camera Calibration

- What if 3D coordinates are unknown?
- Use scene features such as vanishing points

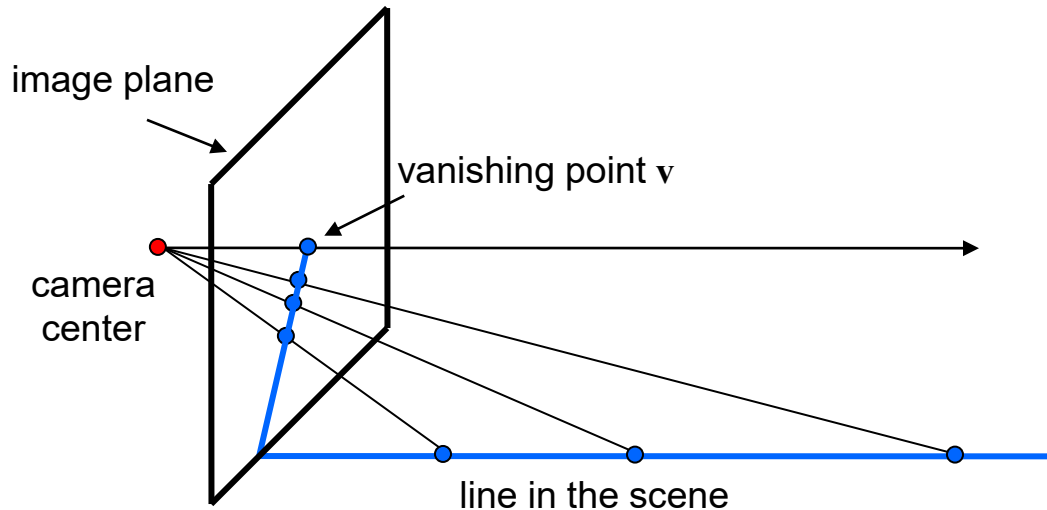


Camera calibration revisited

- What if 3D coordinates are unknown?
- Use scene features such as vanishing points



Recall: Vanishing points

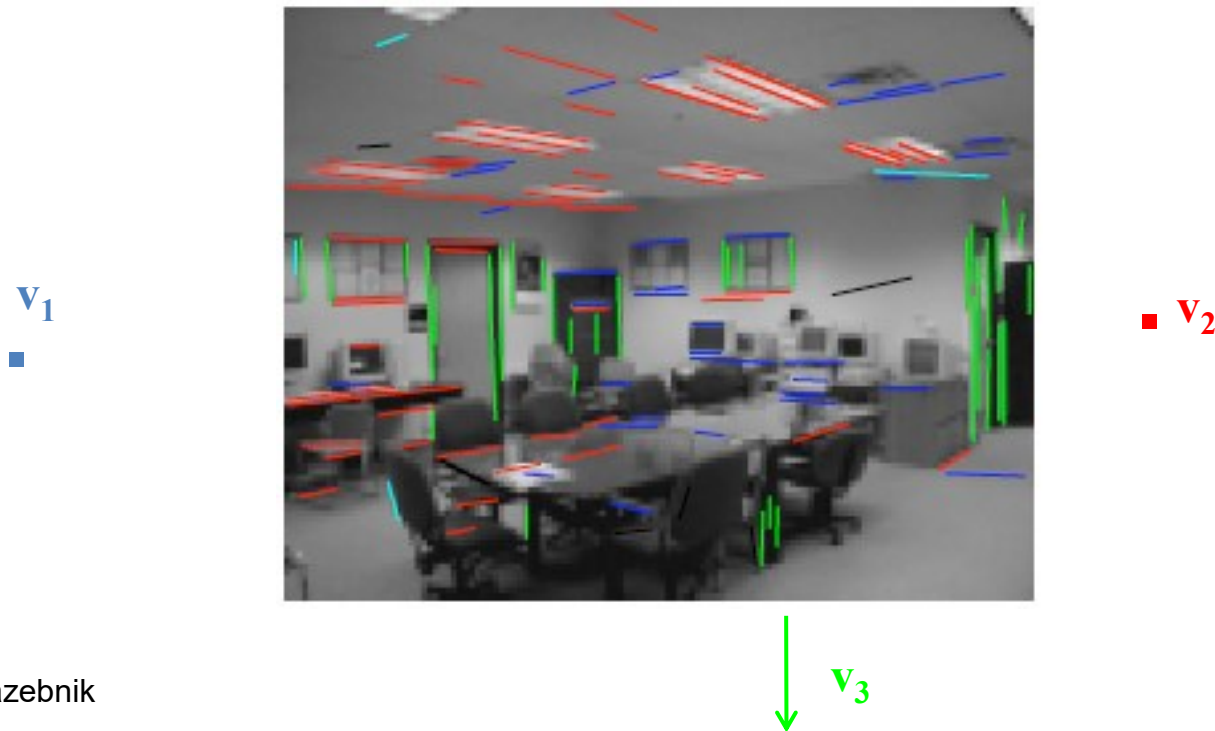


All lines having the same *direction* share the same vanishing point

Calibration from vanishing points

Consider a scene with 3 orthogonal directions
 \mathbf{v}_1 , \mathbf{v}_2 are *finite* vps, \mathbf{v}_3 *infinite* vp

Want to align world coordinates with directions



Calibration from vanishing points

$$\mathbf{P}_{3 \times 4} \equiv [\mathbf{p}_1 \quad \mathbf{p}_2 \quad \mathbf{p}_3 \quad \mathbf{p}_4]$$

It turns out that

$$\mathbf{p}_1 \equiv \mathbf{P} [1,0,0,0]^T \quad \text{VP in X direction}$$

$$\mathbf{p}_2 \equiv \mathbf{P} [0,1,0,0]^T \quad \text{VP in Y direction}$$

$$\mathbf{p}_3 \equiv \mathbf{P} [0,0,1,0]^T \quad \text{VP in Z direction}$$

$$\mathbf{p}_4 \equiv \mathbf{P} [0,0,0,1]^T \quad \text{Projection of origin}$$

Note the usual \equiv (i.e., all of this is up to scale) as well as where the 0 is

Calibration from vanishing points

Let's align the world coordinate system with the three orthogonal vanishing directions:

$$\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \quad \mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \quad \mathbf{e}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\lambda \mathbf{v}_i = \mathbf{K}[\mathbf{R}, \mathbf{t}] \begin{bmatrix} \mathbf{e}_i \\ 0 \end{bmatrix}$$

$$\lambda \mathbf{v}_i = \mathbf{K} \mathbf{R} \mathbf{e}_i$$

Drop the \mathbf{t}

$$\mathbf{R}^{-1} \mathbf{K}^{-1} \lambda \mathbf{v}_i = \mathbf{e}_i$$

Inverses

Calibration from vanishing points

So $e_i = R^{-1}K^{-1}\lambda v_i$, but who cares?

What are some properties of axes?

Know $e_i^T e_j = 0$ for $i \neq j$, so K, R have to satisfy

$$(R^{-1}K^{-1}\lambda_j v_j)^T (R^{-1}K^{-1}\lambda_i v_i) = 0$$

$$(R^T K^{-1}\lambda_j v_j)^T (R^T K^{-1}\lambda_i v_i) = 0 \quad R^{-1} = R^T$$

$$\lambda_i \lambda_j (R^T K^{-1} v_j)^T (R^T K^{-1} v_i) = 0 \quad \text{Move scalars}$$

$$v_j K^{-T} R R^T K^{-1} v_i = 0 \quad \text{Clean up}$$

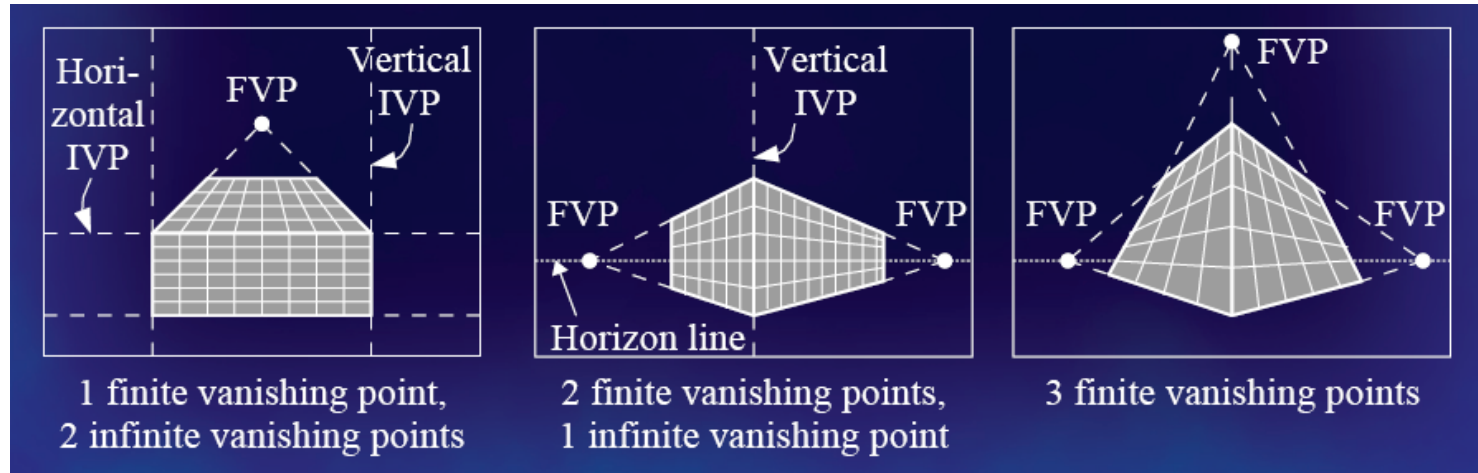
$$v_j K^{-T} K^{-1} v_i = 0 \quad R R^T = I$$

Calibration from vanishing points

- Intrinsic (focal length f , principal point u_0, v_0) have to ensure that the rays corresponding to vanishing points for 3 mutually orthogonal directions are orthogonal

$$\mathbf{v}_j \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{v}_i = \mathbf{0}$$

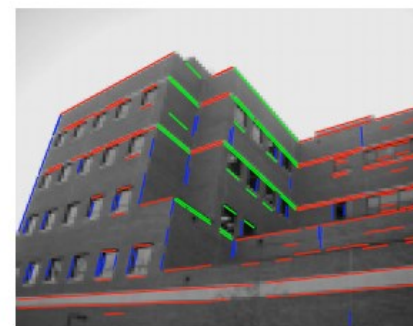
Calibration from vanishing points



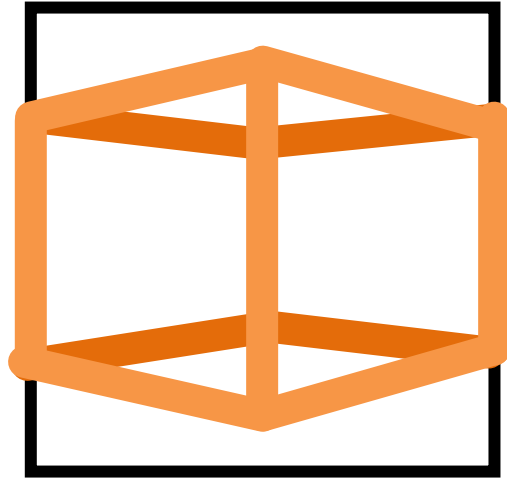
Cannot recover focal length, principal point is the third vanishing point



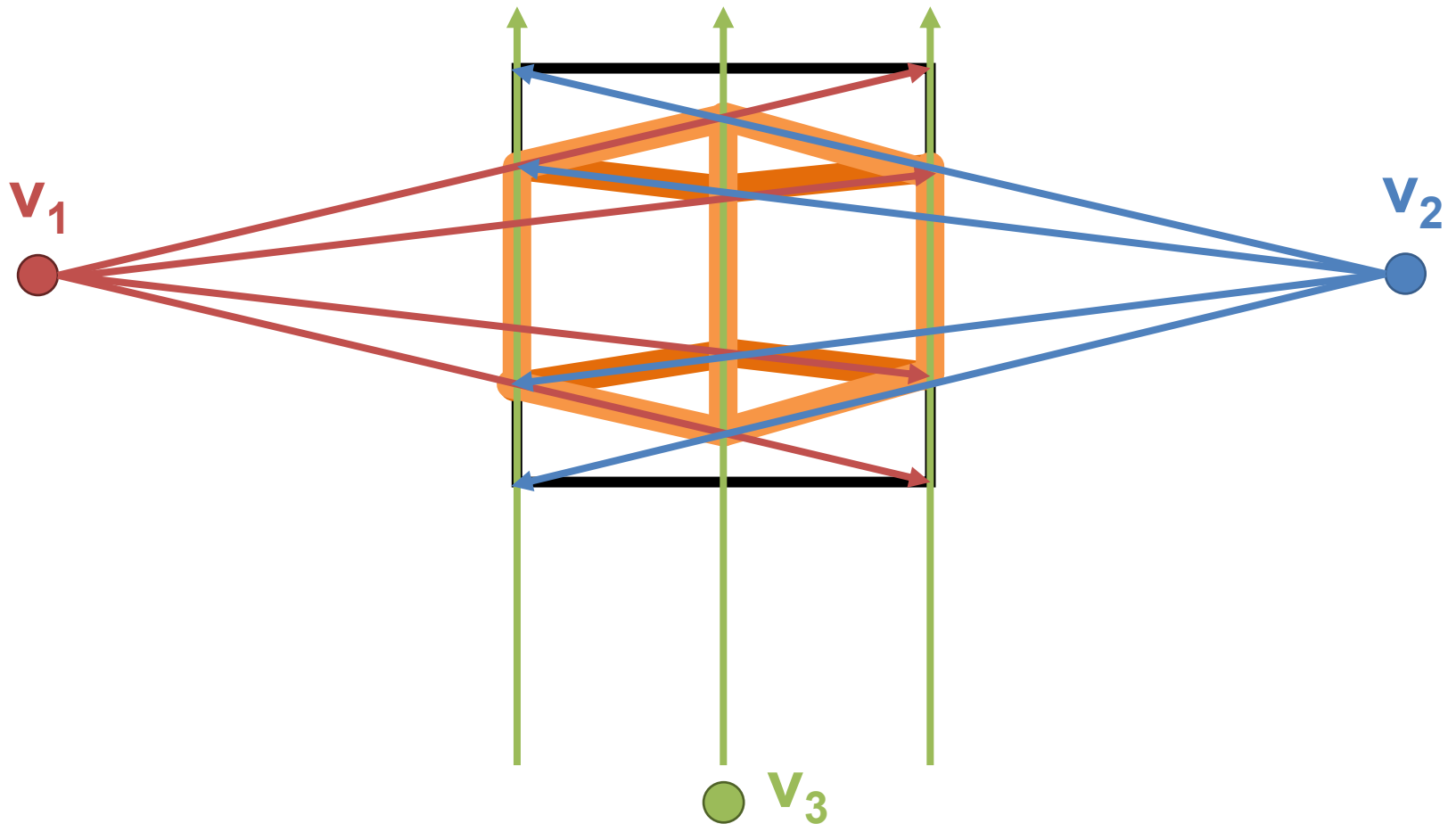
Can solve for focal length, principal point



Directions and vanishing points

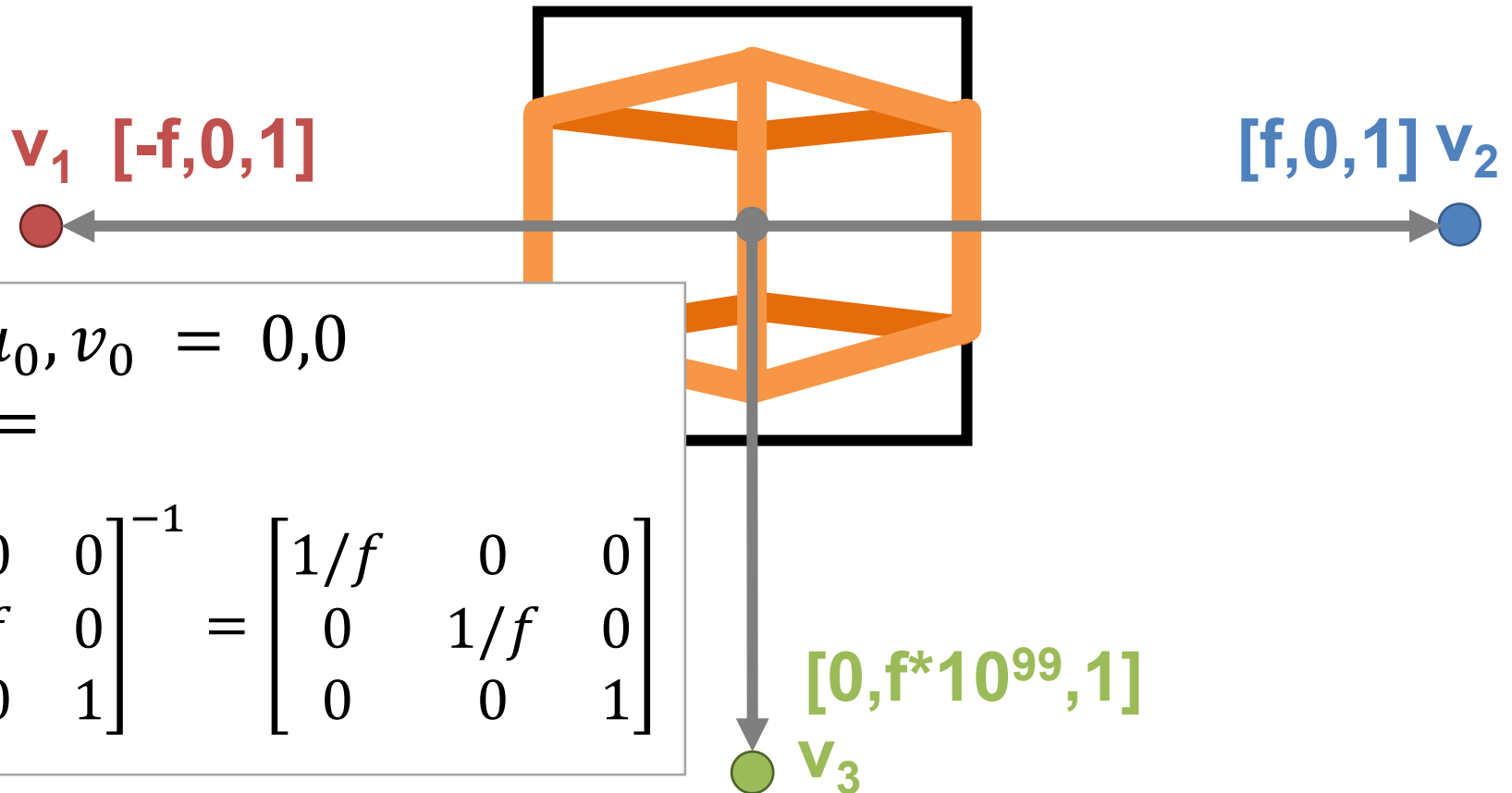


Directions and vanishing points



Directions and vanishing points

If \mathbf{v} vanishing point, and \mathbf{K} the camera intrinsics, $\mathbf{K}^{-1}\mathbf{v}$ is the corresponding direction.



Directions and vanishing points

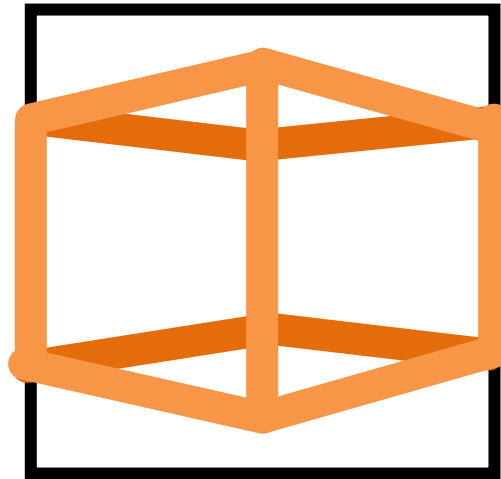
If I normalize each $K^{-1}\mathbf{v}_i$, I get:

$$\left[-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}\right], \left[\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}\right], [0, 1, 0]$$

$$\mathbf{v}_1 \quad [-f, 0, 1]$$



$$K^{-1}\mathbf{v}_1 = [-1, 0, 1]$$



$$[f, 0, 1] \mathbf{v}_2$$



$$K^{-1}\mathbf{v}_2 = [1, 0, 1]$$

$$K^{-1}\mathbf{v}_3 = [0, 10^{99}, 1]$$

$$[0, f \cdot 10^{99}, 1]$$



\mathbf{v}_3

$$K^{-1} = \begin{bmatrix} 1/f & 0 & 0 \\ 0 & 1/f & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Rotation from vanishing points

Know that $\lambda_i \mathbf{v}_i = \mathbf{K} \mathbf{R} \mathbf{e}_i$ and have \mathbf{K} , but want \mathbf{R}

$$\text{So: } \lambda \mathbf{K}^{-1} \mathbf{v}_i = \mathbf{R} \mathbf{e}_i$$

What does $\mathbf{R} \mathbf{e}_i$ look like?

$$\mathbf{R} \mathbf{e}_1 = [\mathbf{r}_1 \quad \mathbf{r}_2 \quad \mathbf{r}_3] \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \mathbf{r}_1$$

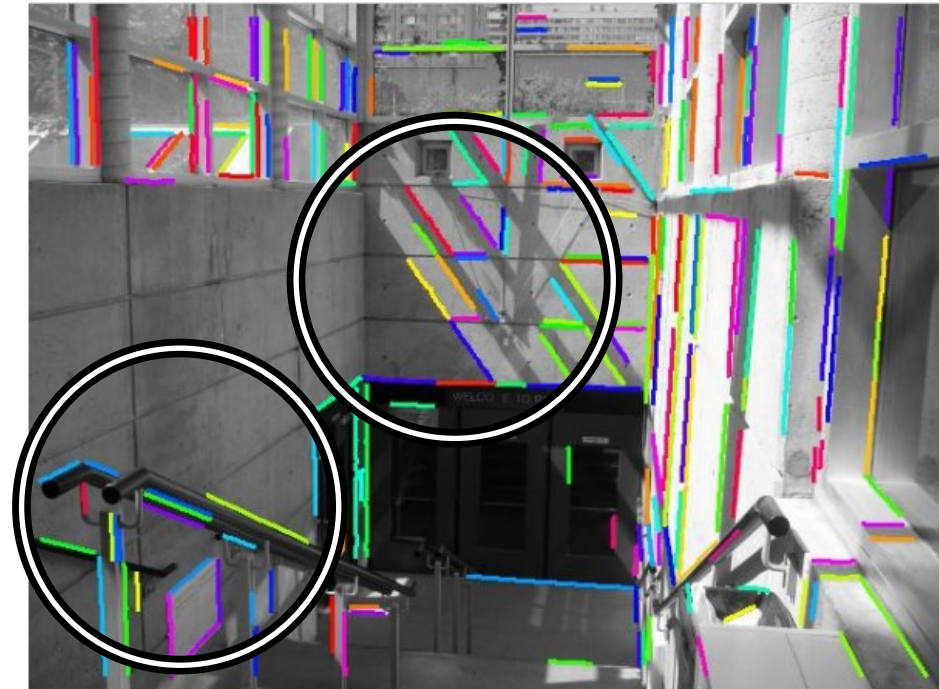
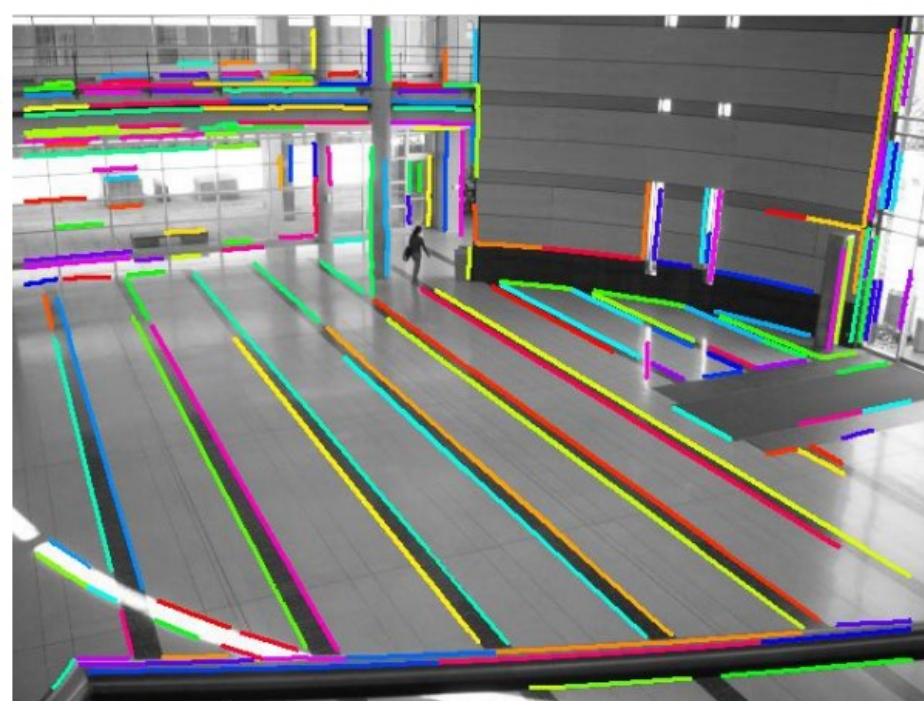
The i th column of \mathbf{R} is a scaled version of

$$\mathbf{r}_i = \lambda \mathbf{K}^{-1} \mathbf{v}_i$$

Calibration from vanishing points

- Solve for K (focal length, principal point) using 3 orthogonal vanishing points
- Get rotation directly from vanishing points once calibration matrix known
- Pros:
 - Could be totally automatic!
- Cons:
 - Need 3 vanishing points, estimated accurately, AND orthogonal with at least two finite!

Finding Vanishing Points



What might go wrong with the circled points?

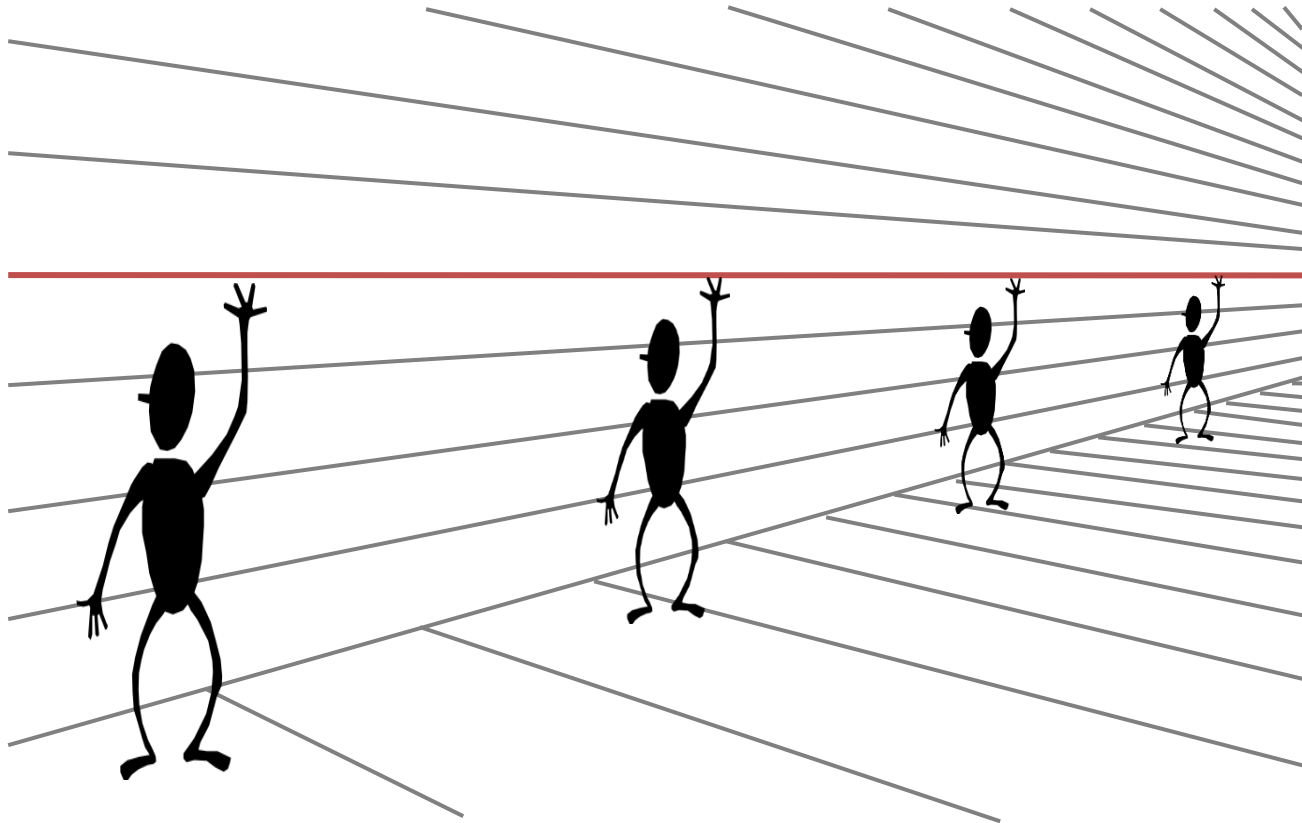
Finding Vanishing Points

- Find long edges $E = \{e_1, \dots, e_n\}$
- All $\binom{n}{2}$ intersections of edges $v_{ij} = e_i \times e_j$ are potential vanishing points
- Try all triplets of popular vanishing points, check if the camera's focal length, principal point "make sense"
- **What are some options for this?**

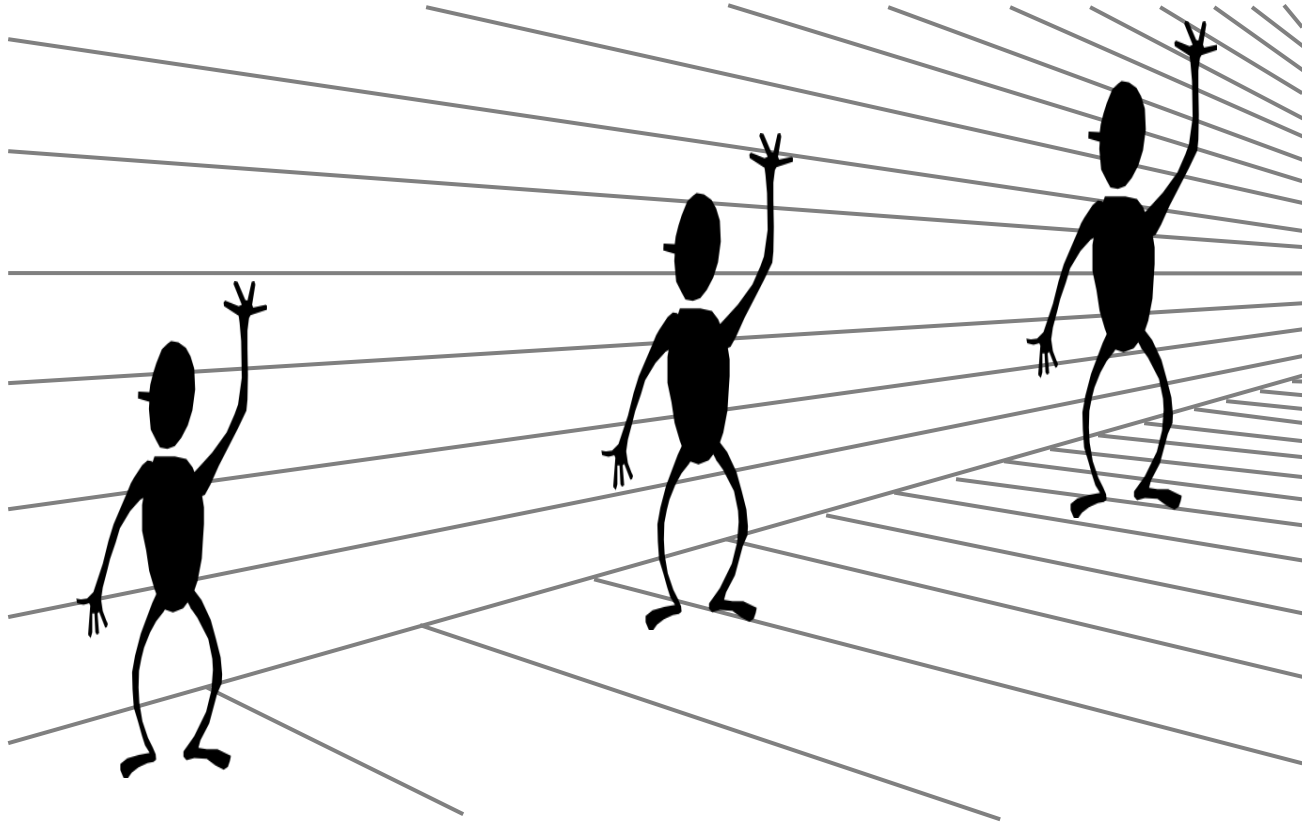
Finding Vanishing Points



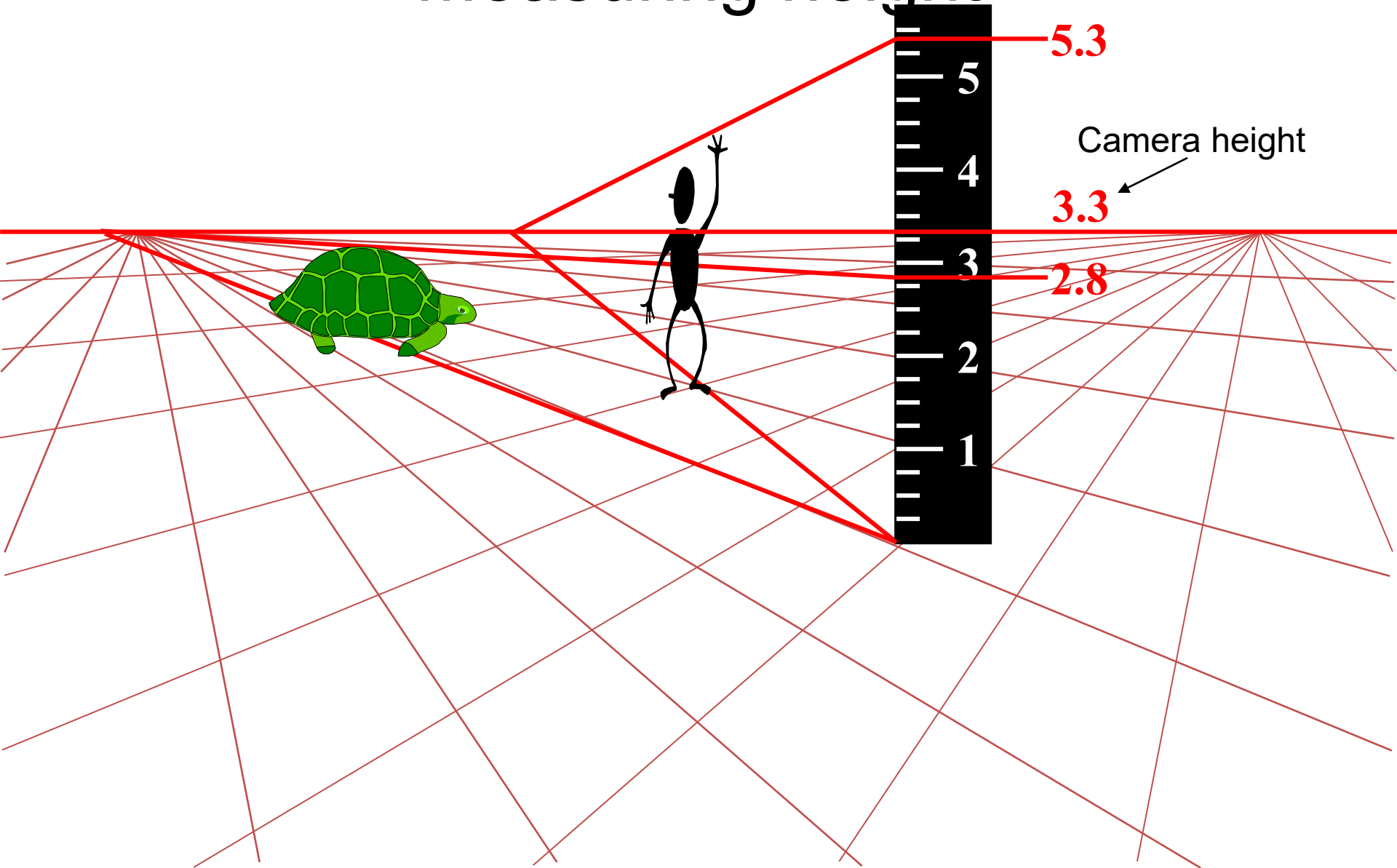
Measuring height



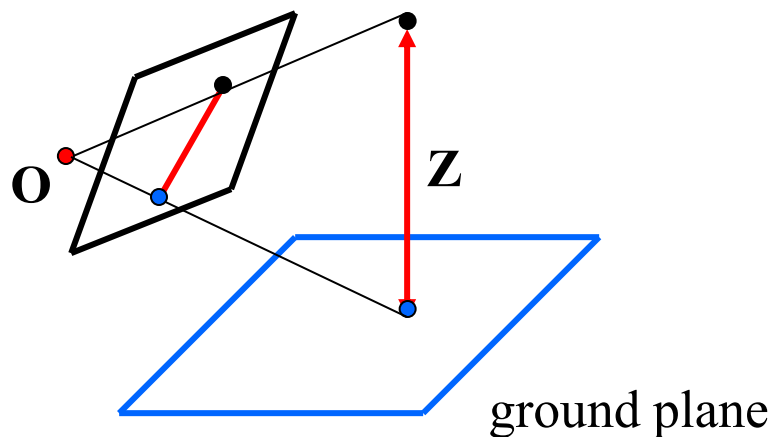
Measuring height



Measuring height



Measuring height without a ruler



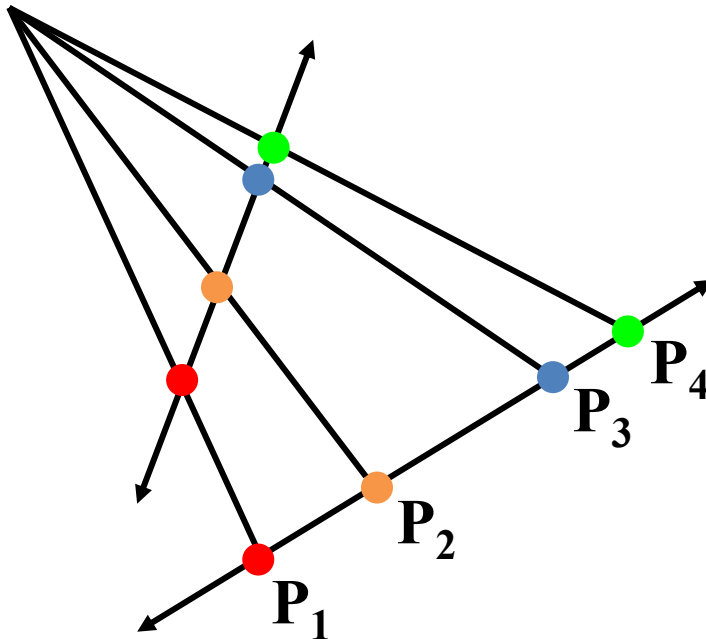
Compute Z from image measurements: We'll need more than vanishing points to do this

Projective invariant

- We need to use a *projective invariant*: a quantity that does not change under projective transformations (including perspective projection)

Projective invariant

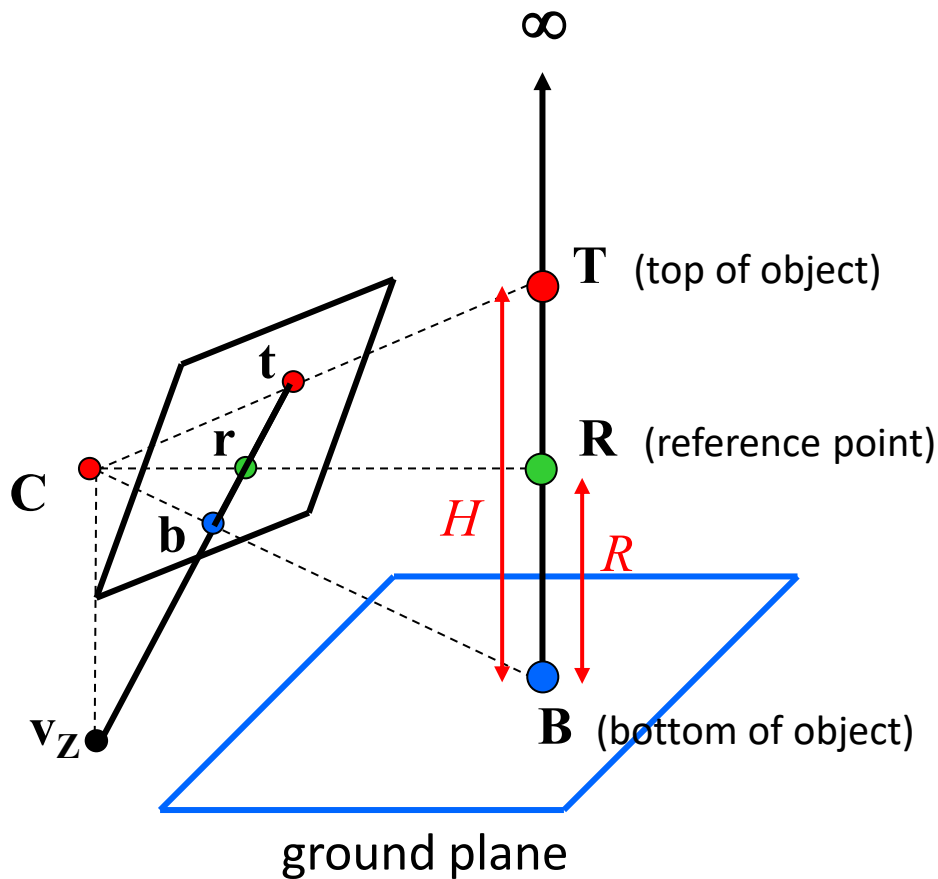
- We need to use a *projective invariant*: a quantity that does not change under projective transformations (including perspective projection)
- The cross-ratio of four points:



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

This is one of the cross-ratios (can reorder arbitrarily)

Measuring height



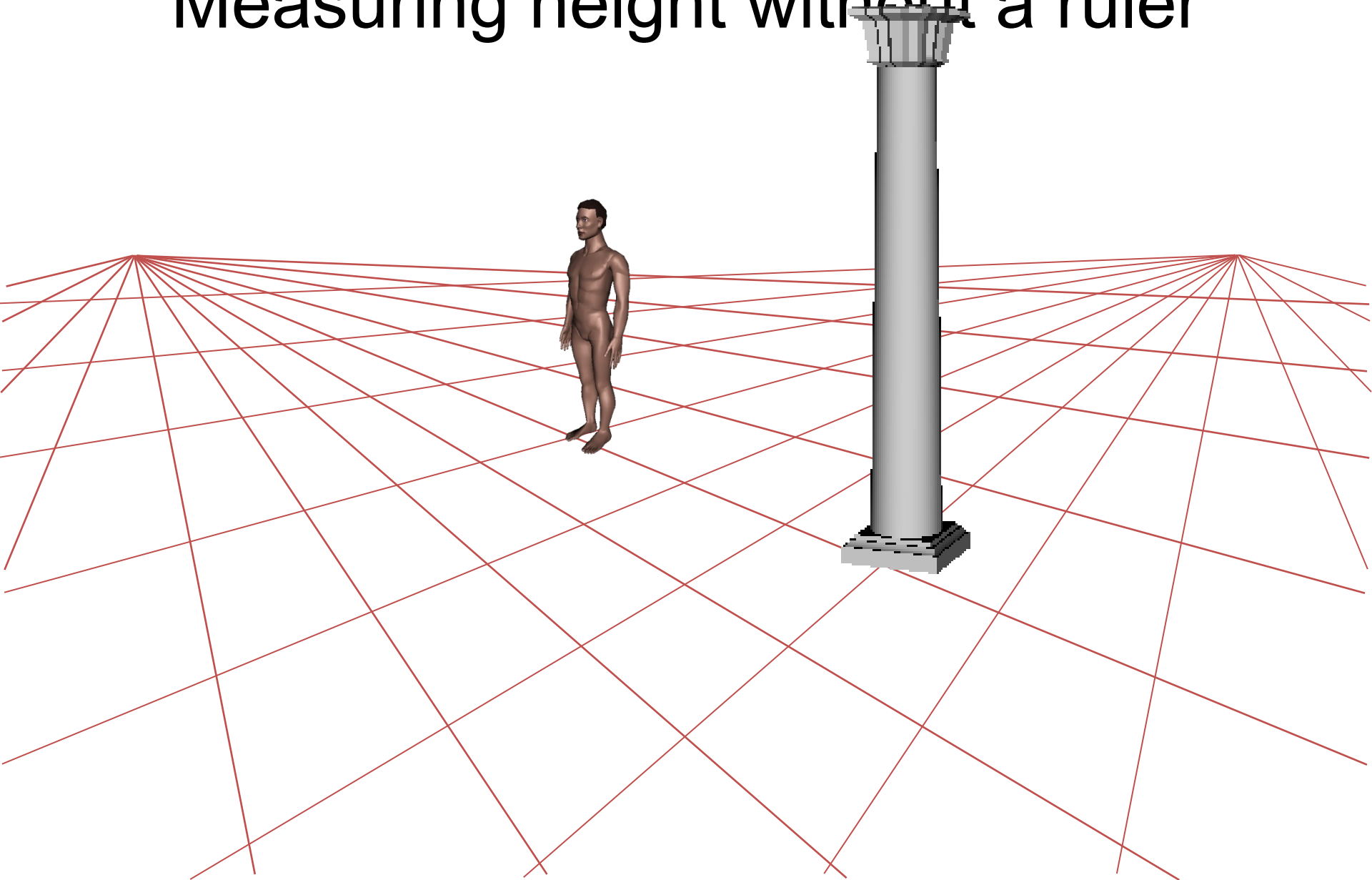
$$\frac{\|T - B\| \|\infty - R\|}{\|R - B\| \|\infty - T\|} = \frac{H}{R}$$

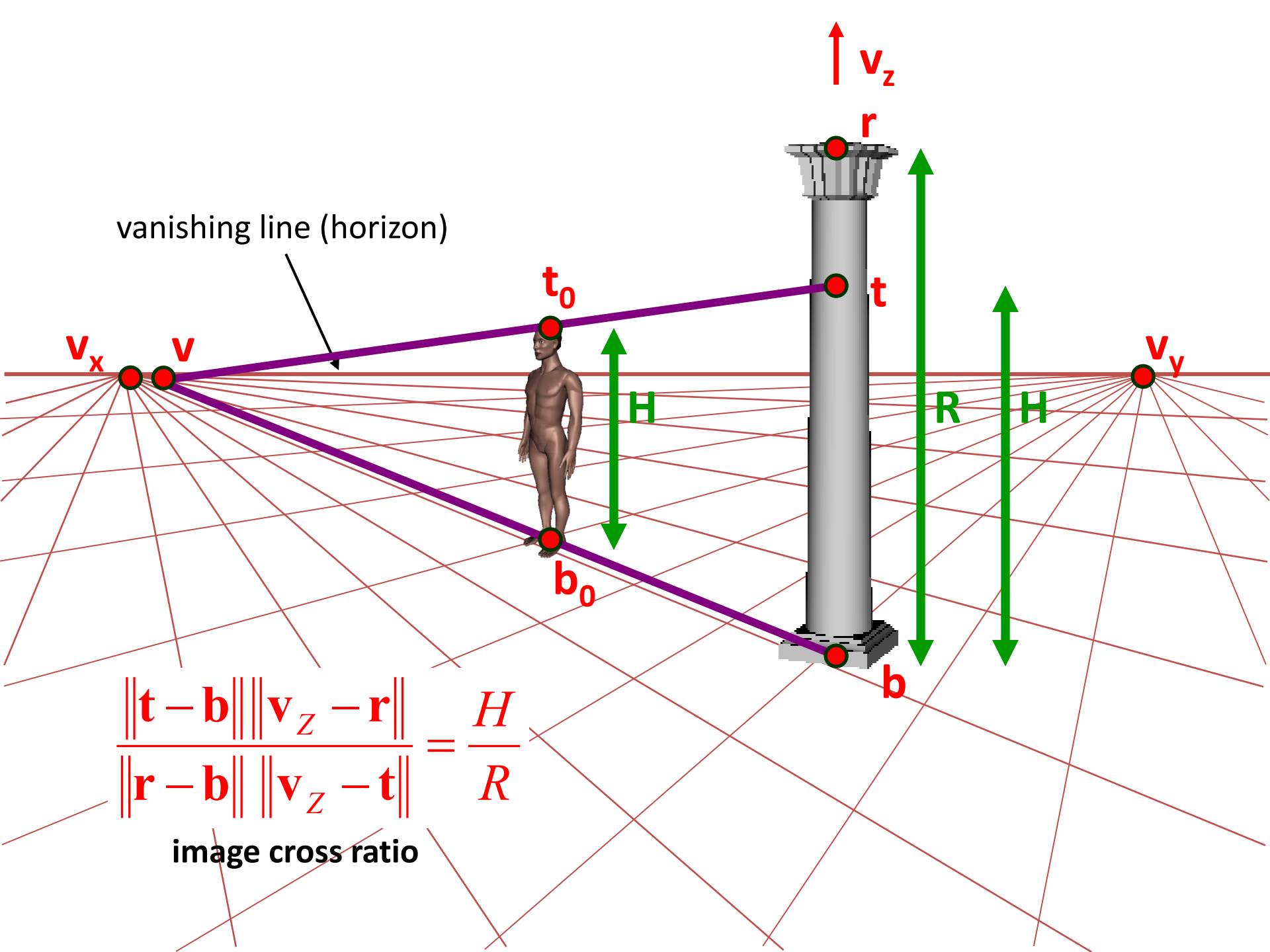
scene cross ratio

$$\frac{\|t - b\| \|v_Z - r\|}{\|r - b\| \|v_Z - t\|} = \frac{H}{R}$$

image cross ratio

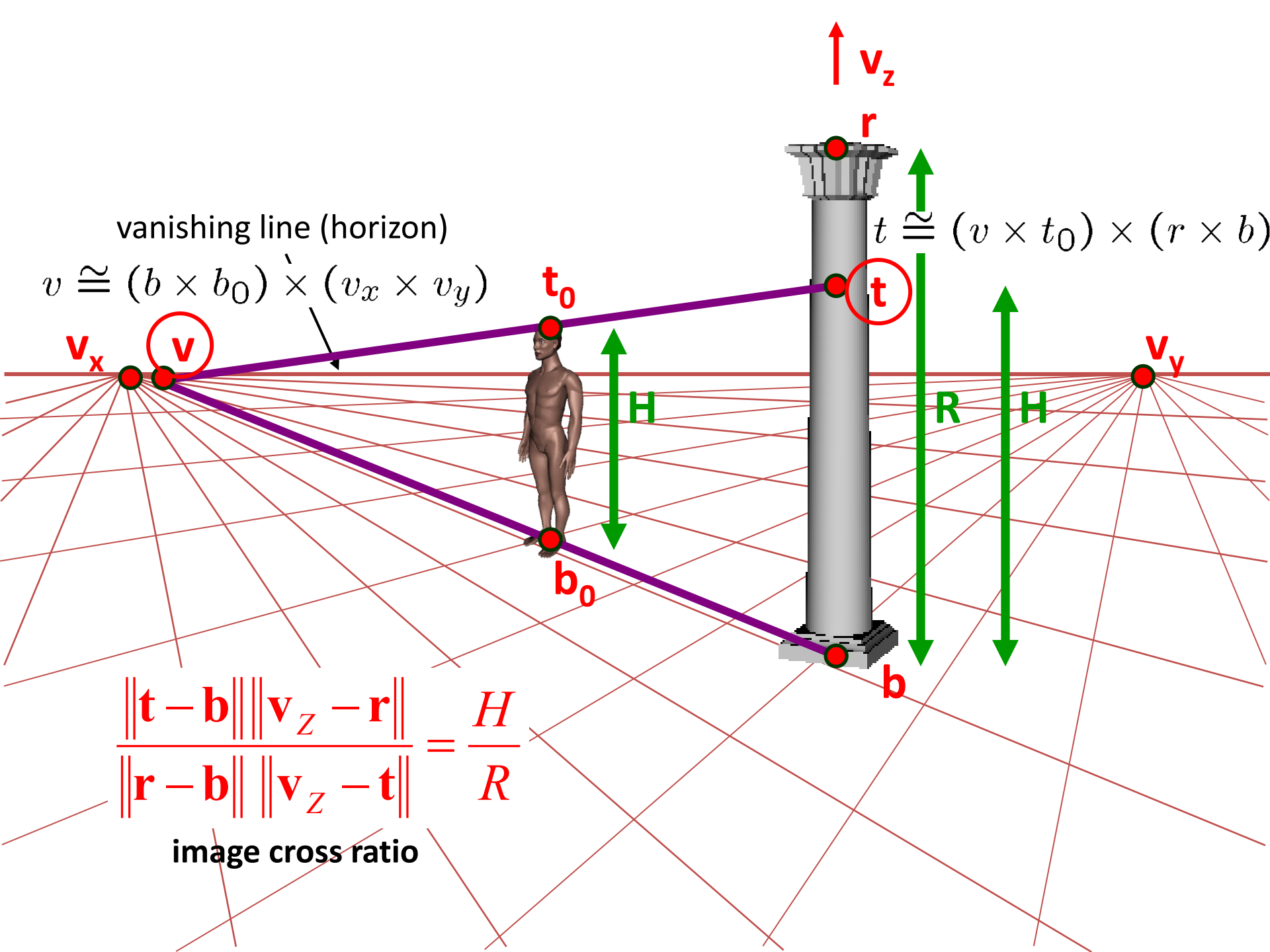
Measuring height without a ruler





Remember This?

- Line equation: $ax + by + c = 0$
- Vector form: $\mathbf{l}^T \mathbf{p} = 0$, $\mathbf{l} = [a, b, c]$, $\mathbf{p} = [x, y, 1]$
- Line through two points?
 - $\mathbf{l} = \mathbf{p}_1 \times \mathbf{p}_2$
- Intersection of two lines?
 - $\mathbf{p} = \mathbf{l}_1 \times \mathbf{l}_2$
- Intersection of two parallel lines is at infinity



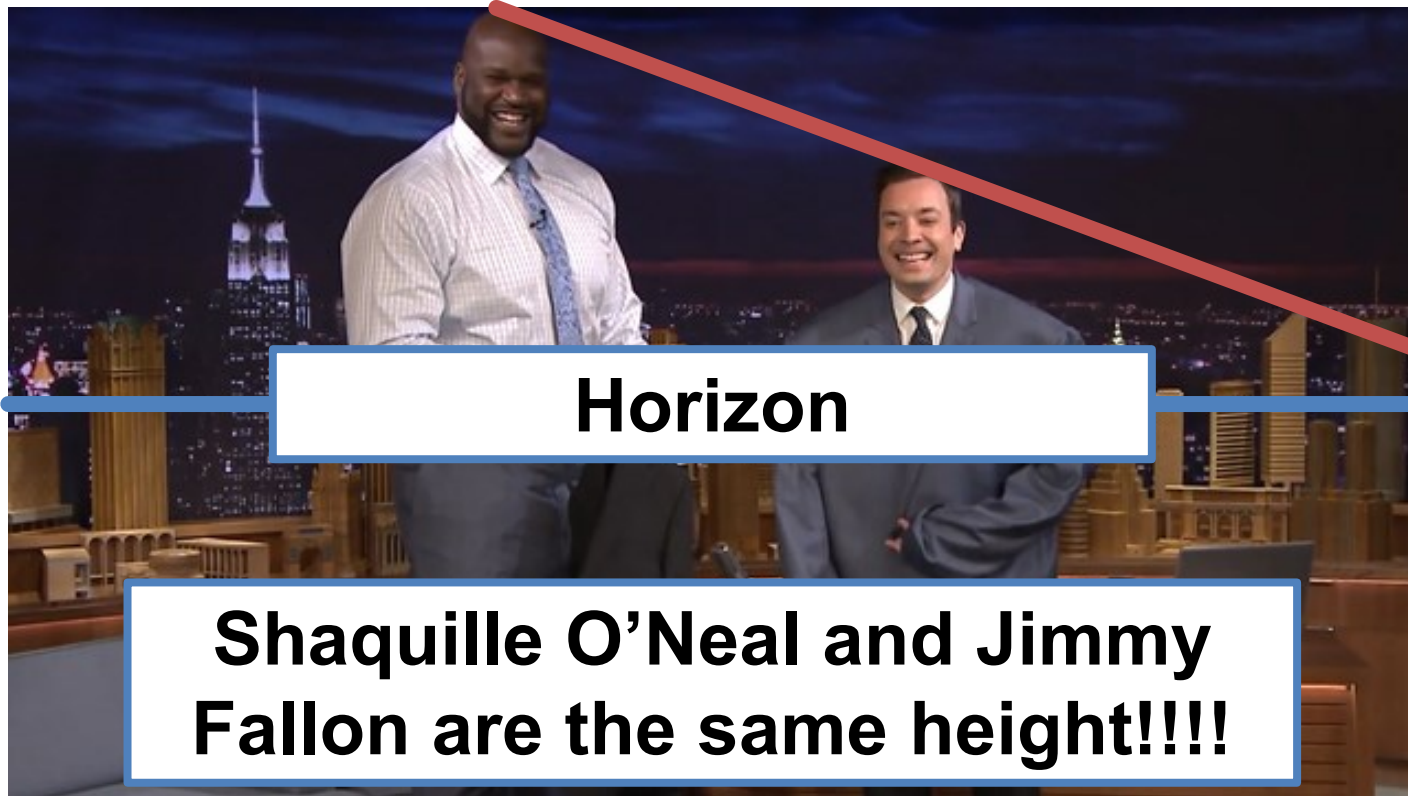
Example Gone Wrong



Know length of red → can figure out height of blue because they intersect at vanishing point v

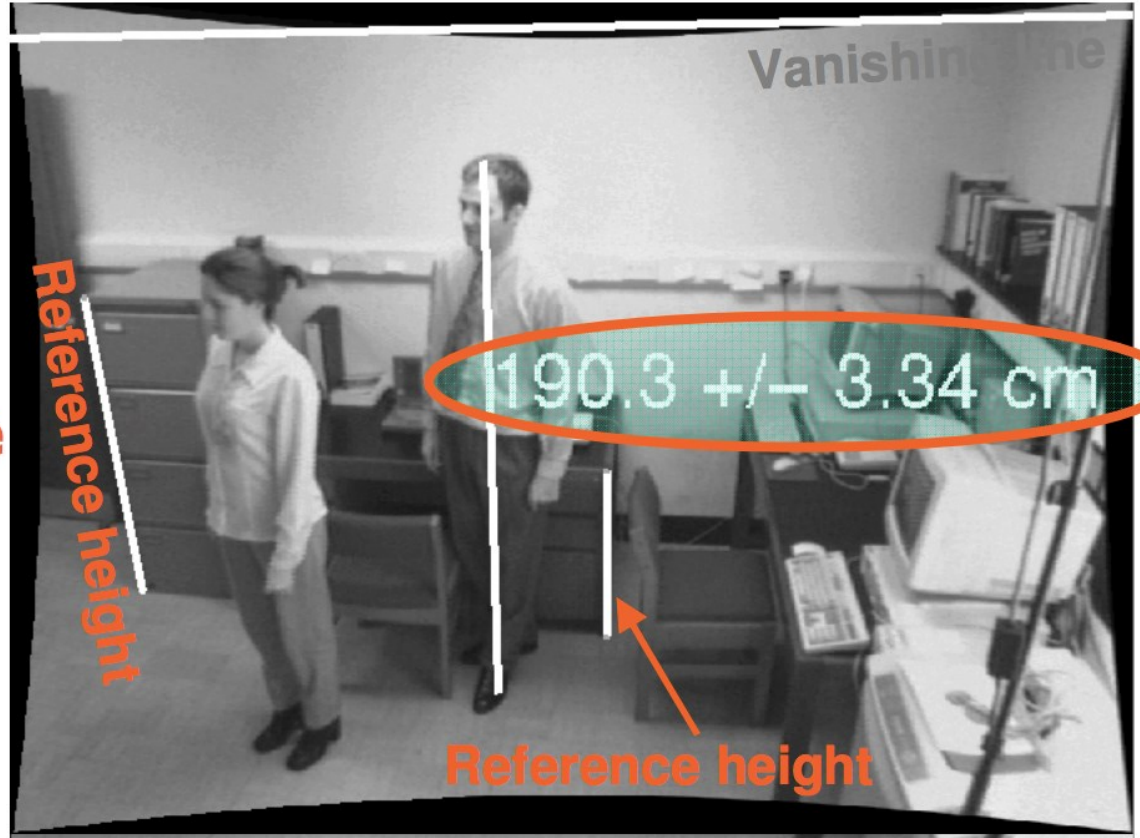
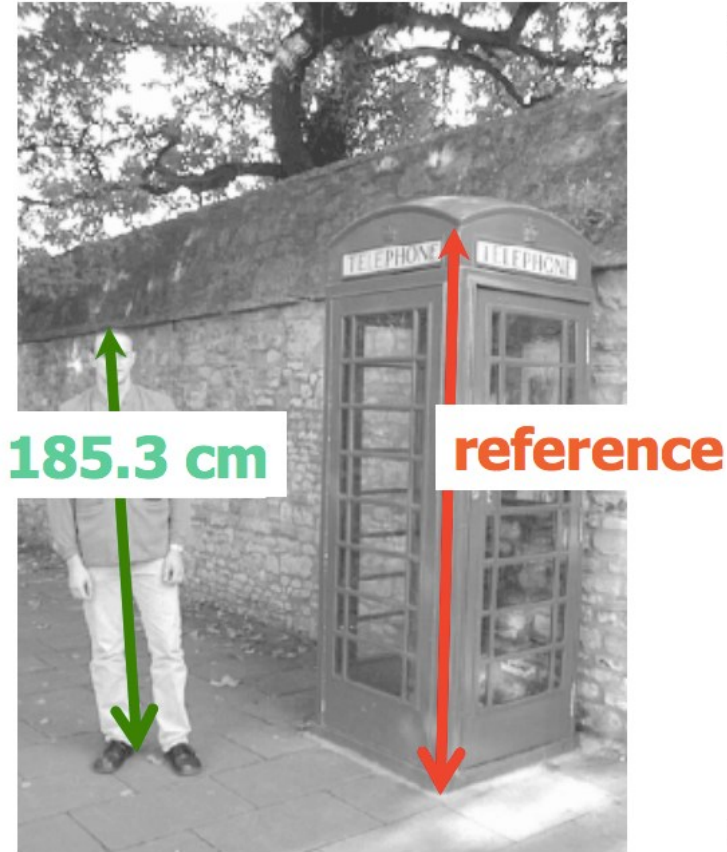
Wrong! Any two lines always intersect!
Need to point to same 3D direction / VP.

Example Gone Wrong



Wrong! Need to connect feet to the horizon (at infinity – thank homogenous coordinates), and then to Jimmy's head.

Examples



A. Criminisi, I. Reid, and A. Zisserman, [Single View Metrology](#), IJCV 2000

Slide credit: S. Lazebnik

Figure from [UPenn CIS580 slides](#)

Another example

- Are the heights of the two groups of people consistent with one another?



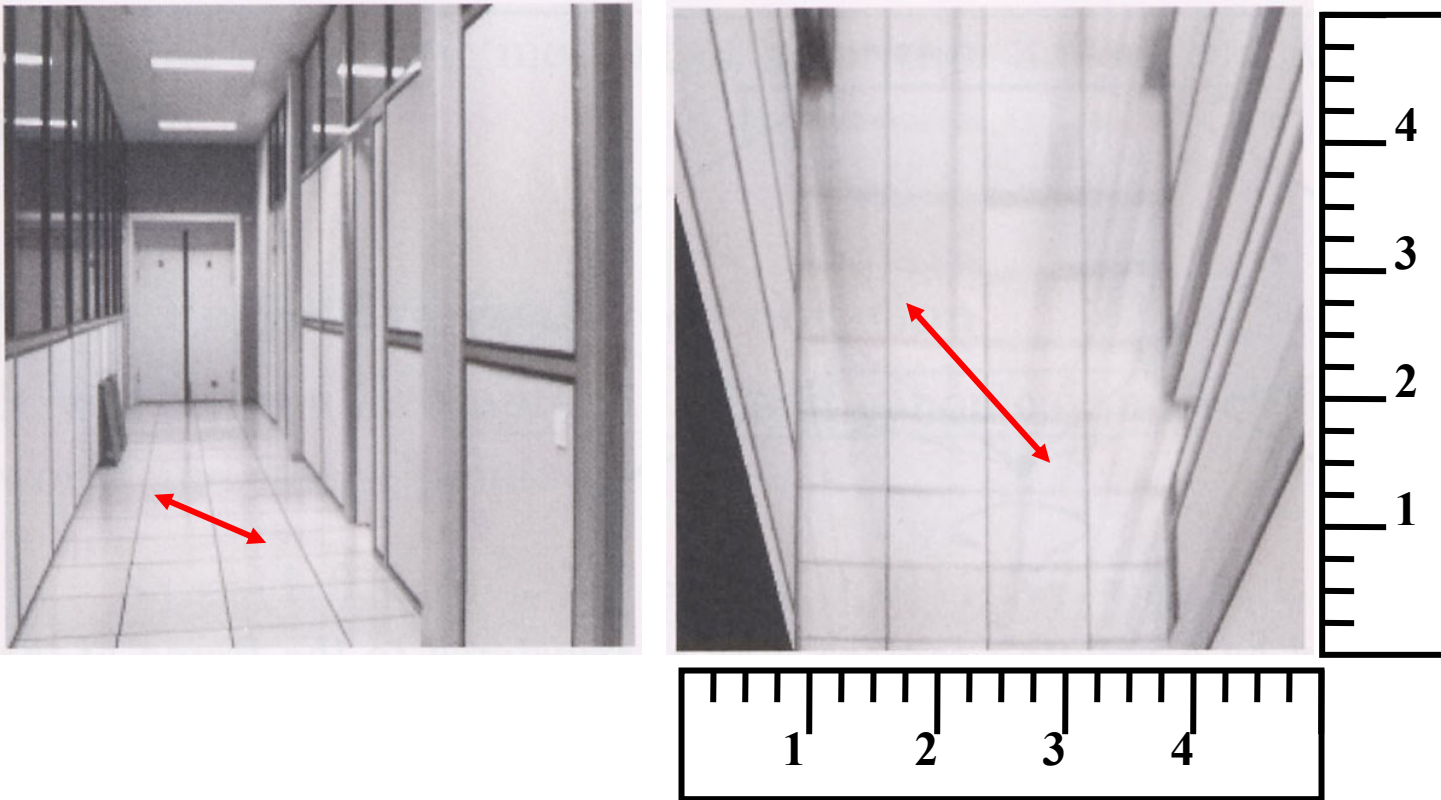
Piero della Francesca, *Flagellation*, ca. 1455

A. Criminisi, M. Kemp, and A. Zisserman, [Bringing Pictorial Space to Life: computer techniques for the analysis of paintings](#),

Slide credit: S. Lazebnik

Proc. Computers and the History of Art, 2002

Measurements on planes



Measurements on planes

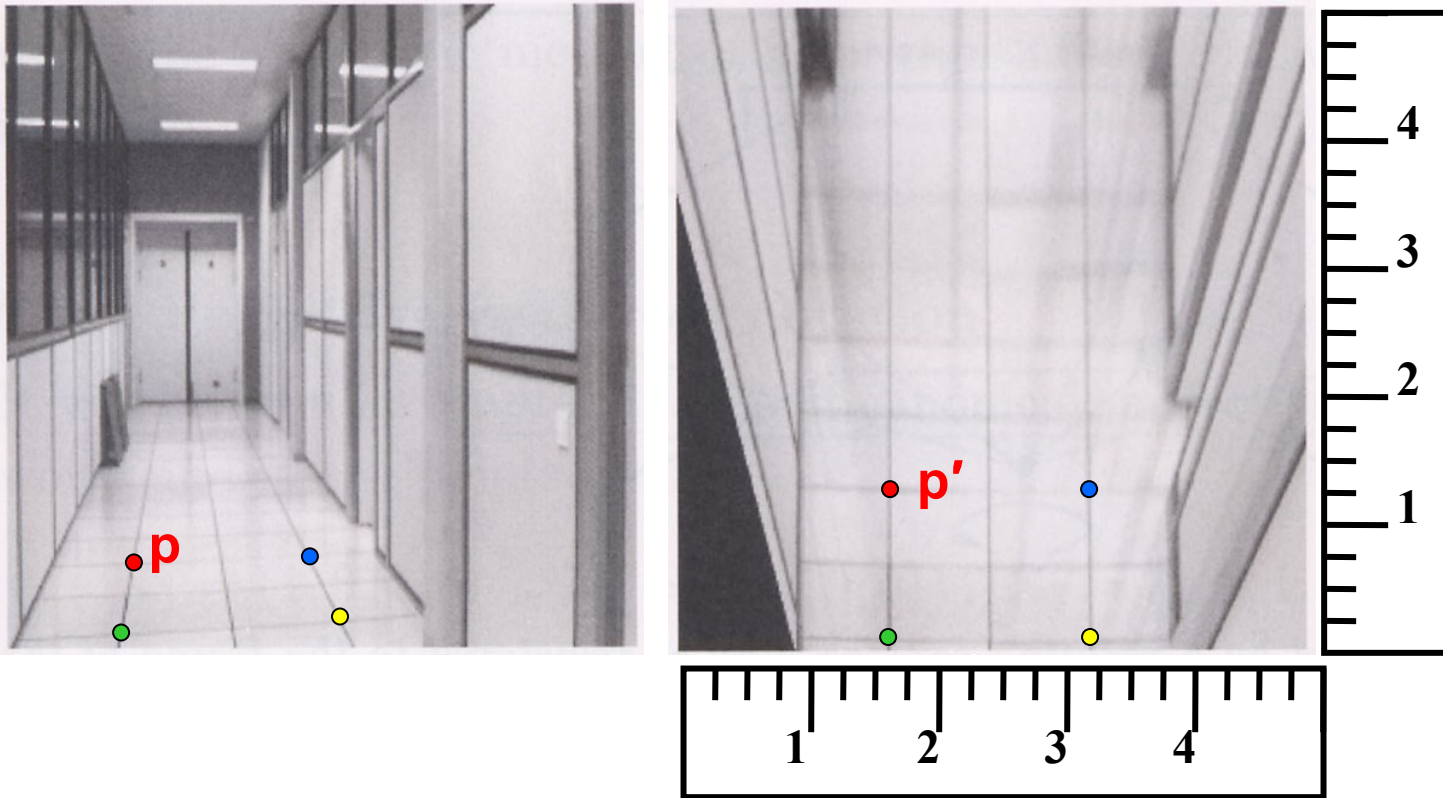
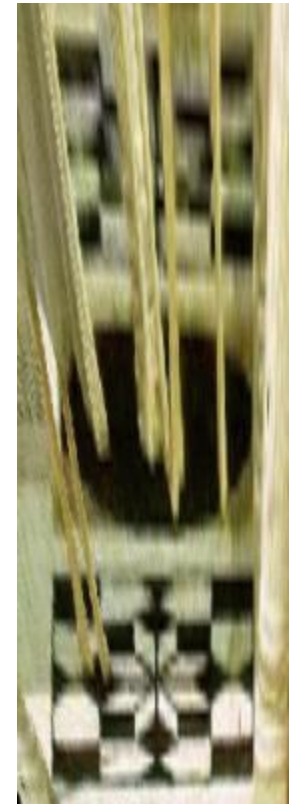
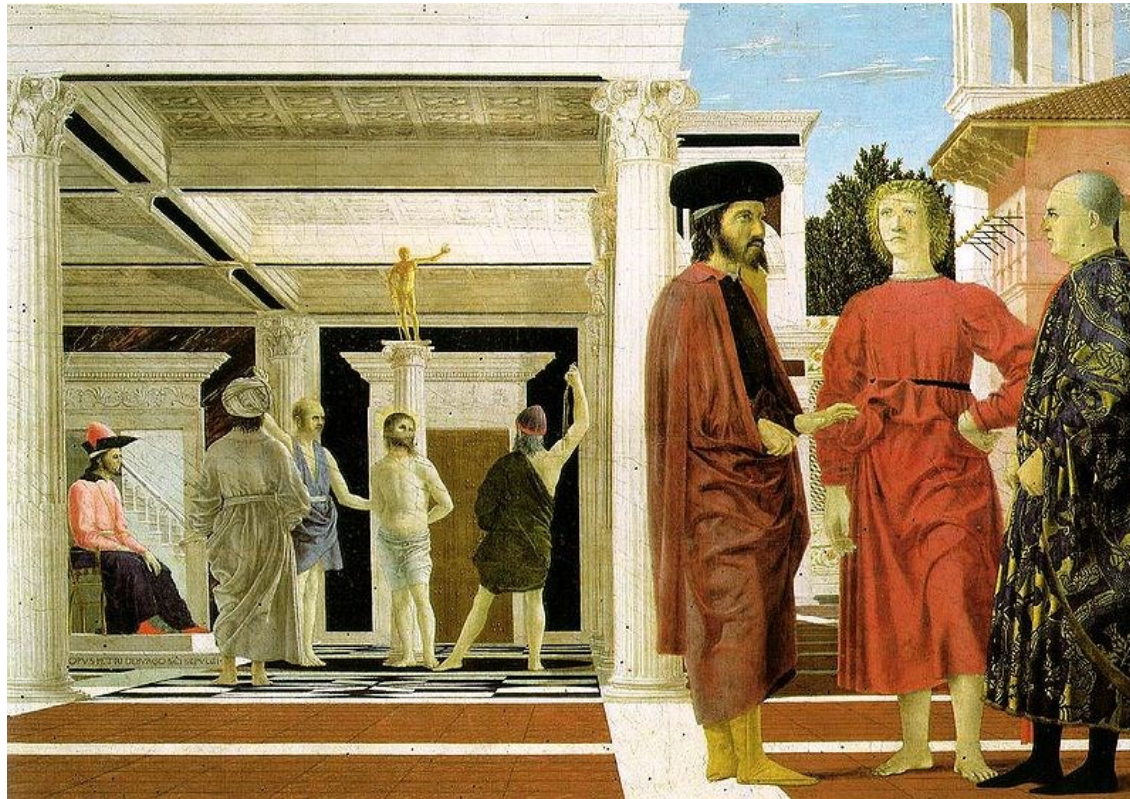
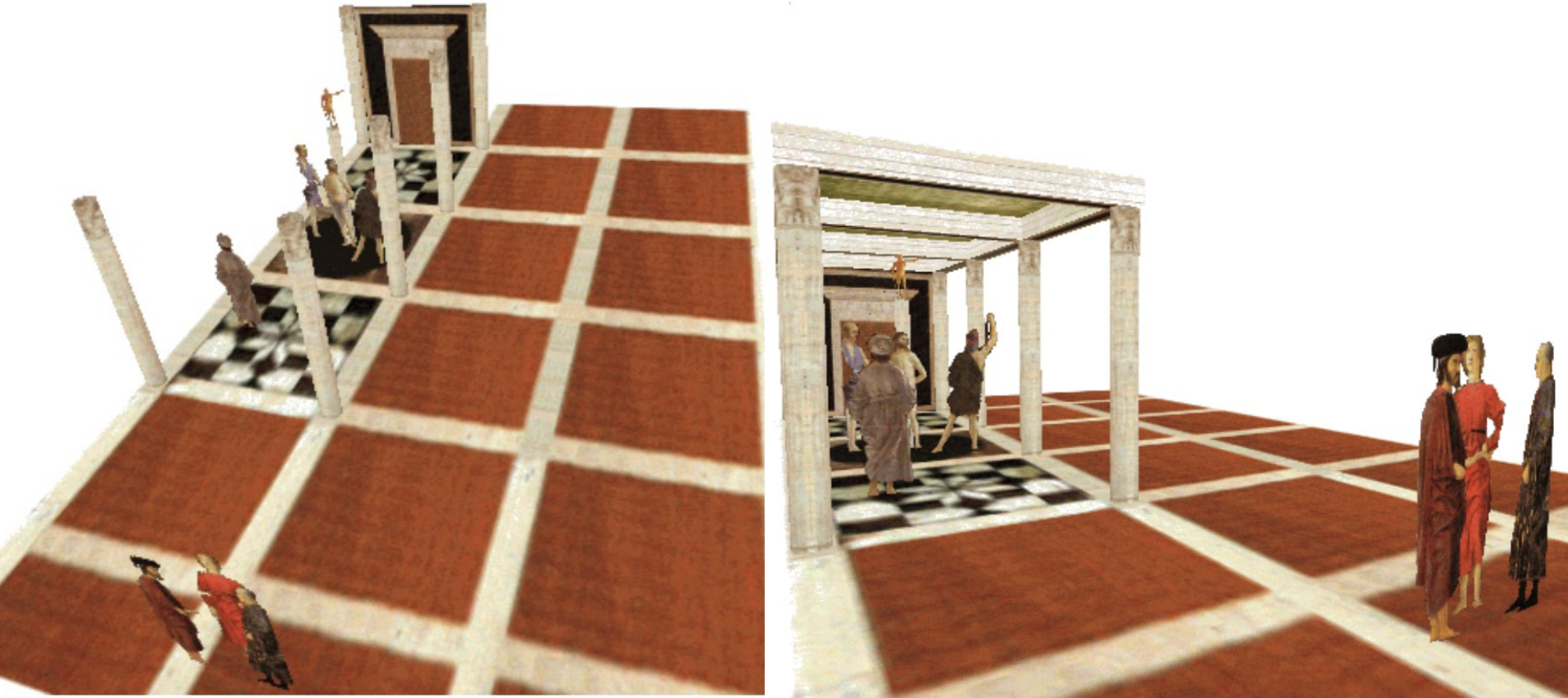


Image rectification: example



Piero della Francesca, *Flagellation*, ca. 1455

Application: 3D modeling from a single image



A. Criminisi, M. Kemp, and A. Zisserman, [Bringing Pictorial Space to Life: computer techniques for the analysis of paintings](#),

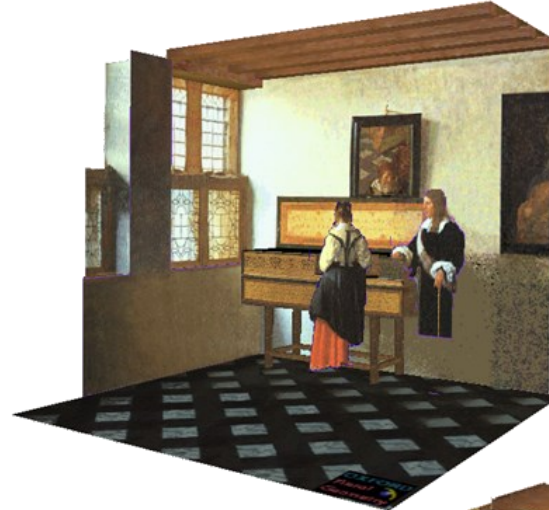
Slide credit: S. Lazebnik

Proc. Computers and the History of Art, 2002

Application: 3D modeling from a single image



J. Vermeer, *Music Lesson*, 1662

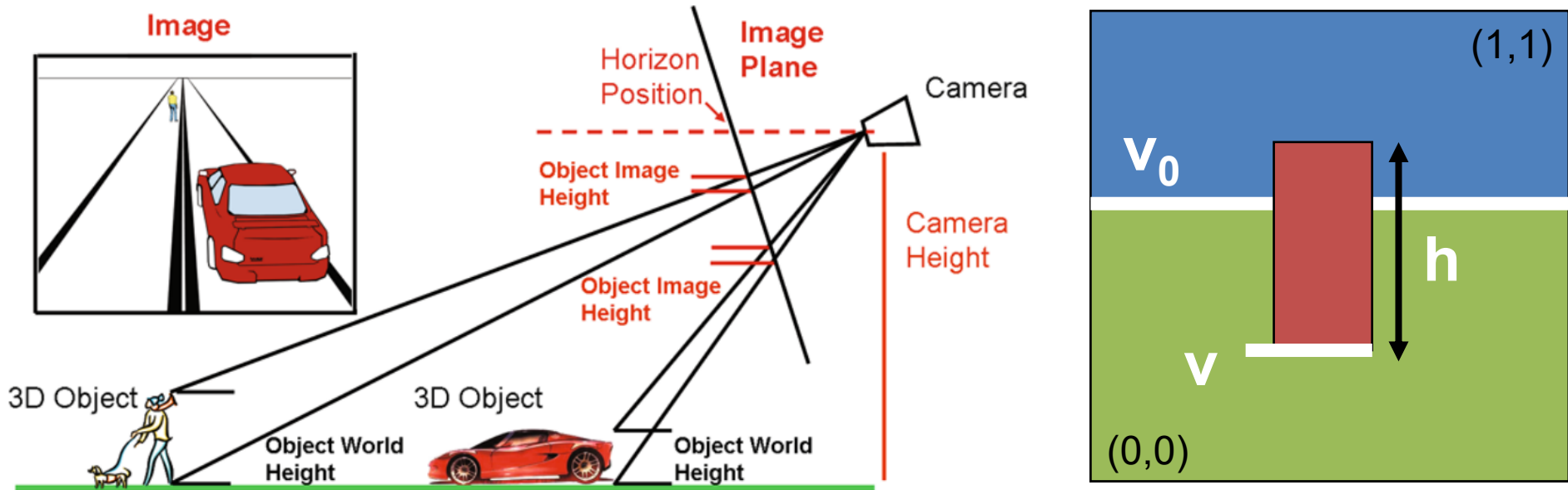


A. Criminisi, M. Kemp, and A. Zisserman, [Bringing Pictorial Space to Life: computer techniques for the analysis of paintings](#),

Slide credit: S. Lazebnik

Proc. Computers and the History of Art, 2002

Application: Object Detection



“Reasonable” approximation:

$$y_{object} \approx \frac{h y_{camera}}{v_0 - v}$$

Application: Object detection

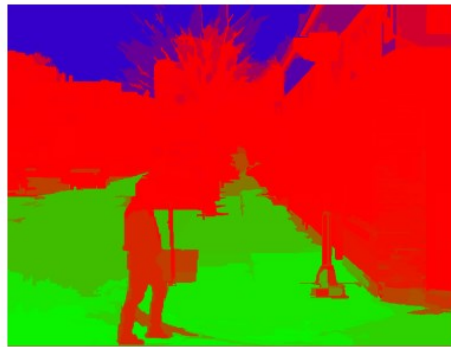


(a) input image

Application: Object detection



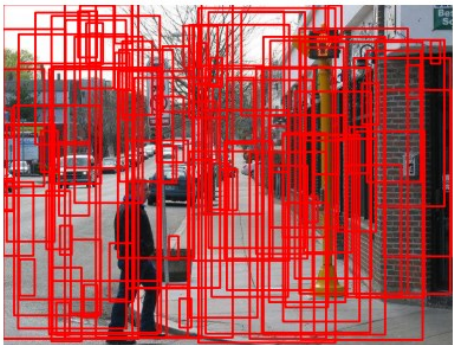
(a) input image



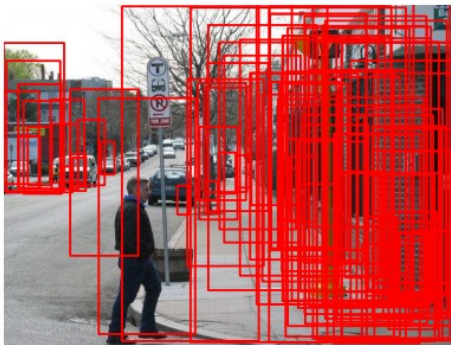
(c) surface orientation estimate



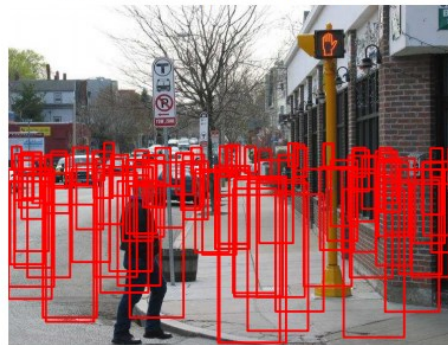
(e) $P(\text{viewpoint} \mid \text{objects})$



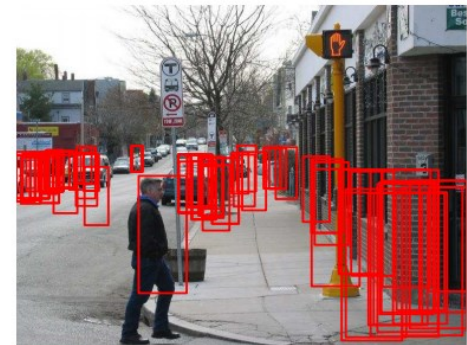
(b) $P(\text{person}) = \text{uniform}$



(d) $P(\text{person} \mid \text{geometry})$

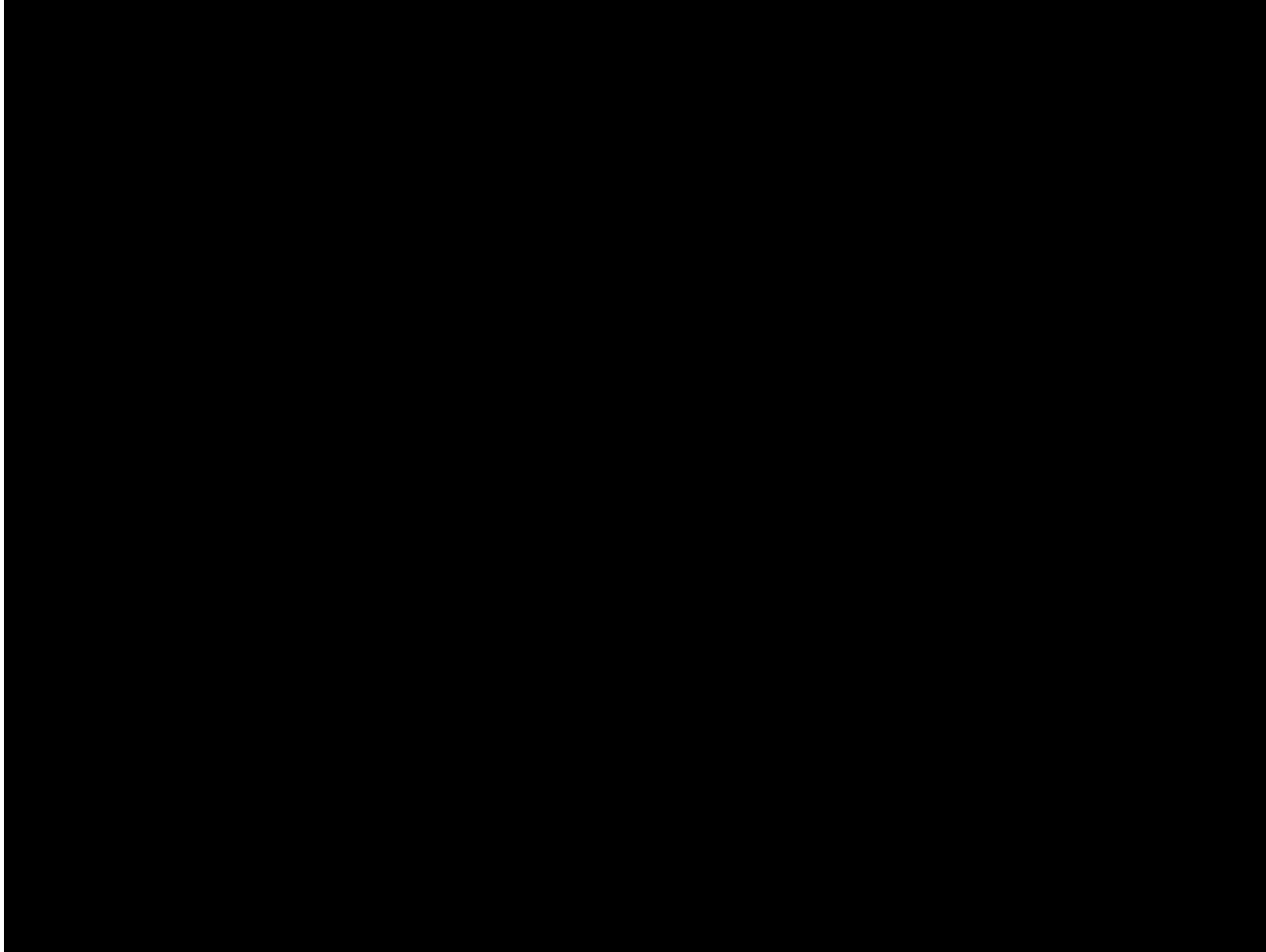


(f) $P(\text{person} \mid \text{viewpoint})$



(g) $P(\text{person} \mid \text{viewpoint, geometry})$

Application: Image Editing



K. Karsch and V. Hedau and D. Forsyth and D. Hoiem, [Rendering Synthetic Objects into Legacy Photographs](#), *SIGGRAPH Asia* 2011

Application: Estimating Layout



V. Hedau, D. Hoiem, D. Forsyth
Recovering the spatial layout of cluttered rooms ICCV 2009