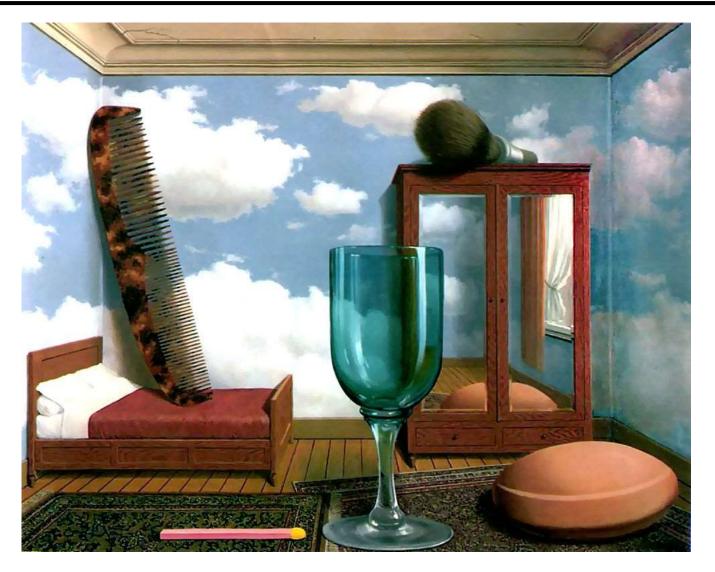
Single-view metrology

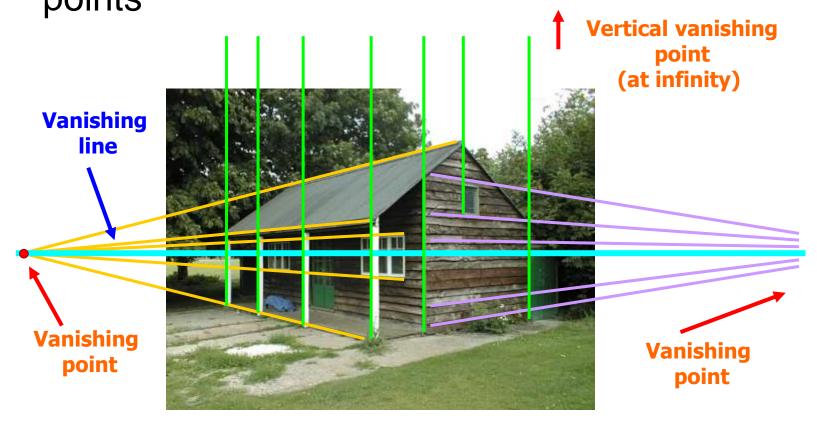


Magritte, Personal Values, 1952

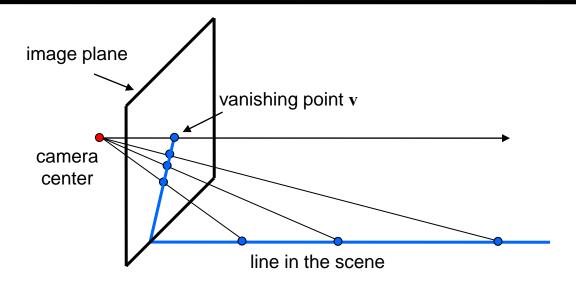
Camera calibration revisited

What if world coordinates of reference 3D points are not known?

We can use scene features such as vanishing points

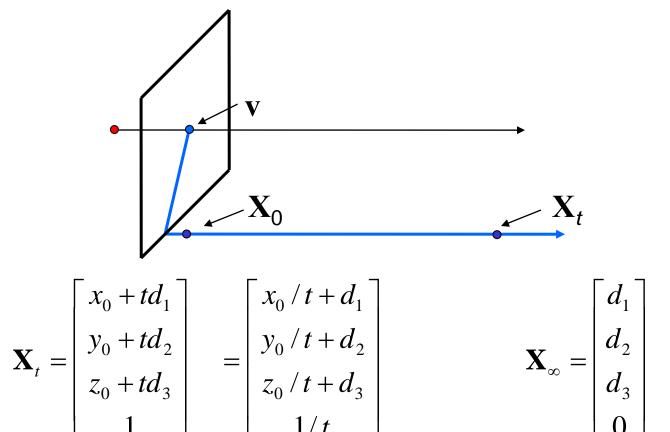


Recall: Vanishing points



 All lines having the same direction share the same vanishing point

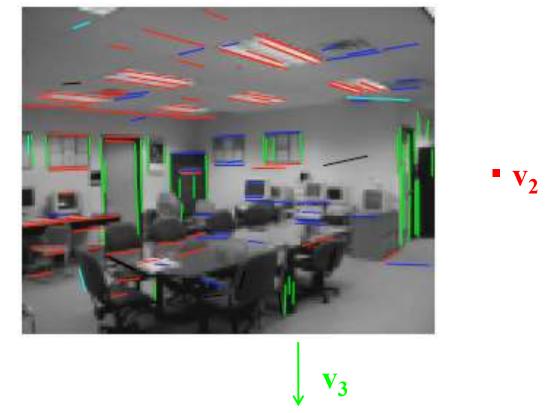
Computing vanishing points



- \mathbf{X}_{∞} is a *point at infinity,* \mathbf{v} is its projection: $\mathbf{v} = \mathbf{P}\mathbf{X}_{\infty}$
- The vanishing point depends only on line direction
- All lines having direction ${f d}$ intersect at ${f X}_{\infty}$

 \mathbf{v}_1

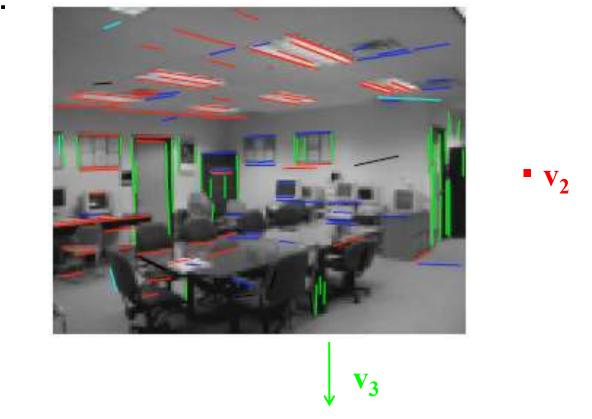
Consider a scene with three orthogonal vanishing directions:



Note: v₁, v₂ are finite vanishing points and v₃ is an infinite vanishing point

 \mathbf{v}_1

Consider a scene with three orthogonal vanishing directions:



 We can align the world coordinate system with these directions

- $\mathbf{p_1} = \mathbf{P}(1,0,0,0)^{\mathrm{T}}$ the vanishing point in the x direction
- Similarly, p₂ and p₃ are the vanishing points in the y and z directions
- $\mathbf{p_4} = \mathbf{P}(0,0,0,1)^T$ projection of the origin of the world coordinate system
- Problem: we can only know the four columns up to independent scale factors, additional constraints needed to solve for them

 Let us align the world coordinate system with three orthogonal vanishing directions in the scene:

$$\mathbf{e_1} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad \mathbf{e_2} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad \mathbf{e_3} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \qquad \lambda_i \mathbf{v}_i = \mathbf{K} \begin{bmatrix} \mathbf{R} \mid \mathbf{t} \end{bmatrix} \begin{bmatrix} \mathbf{e}_i \\ 0 \end{bmatrix} = \mathbf{K} \mathbf{R} \mathbf{e}_i$$

$$\mathbf{e}_{i} = \lambda_{i} \mathbf{R}^{T} \mathbf{K}^{-1} \mathbf{v}_{i}, \quad \mathbf{e}_{i}^{T} \mathbf{e}_{j} = 0$$

$$\mathbf{v}_{i}^{T} \mathbf{K}^{-T} \mathbf{R} \mathbf{R}^{T} \mathbf{K}^{-1} \mathbf{v}_{j} = \mathbf{v}_{i}^{T} \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{v}_{j} = 0$$

 Each pair of vanishing points gives us a constraint on the focal length and principal point zero skew, unit aspect ratio

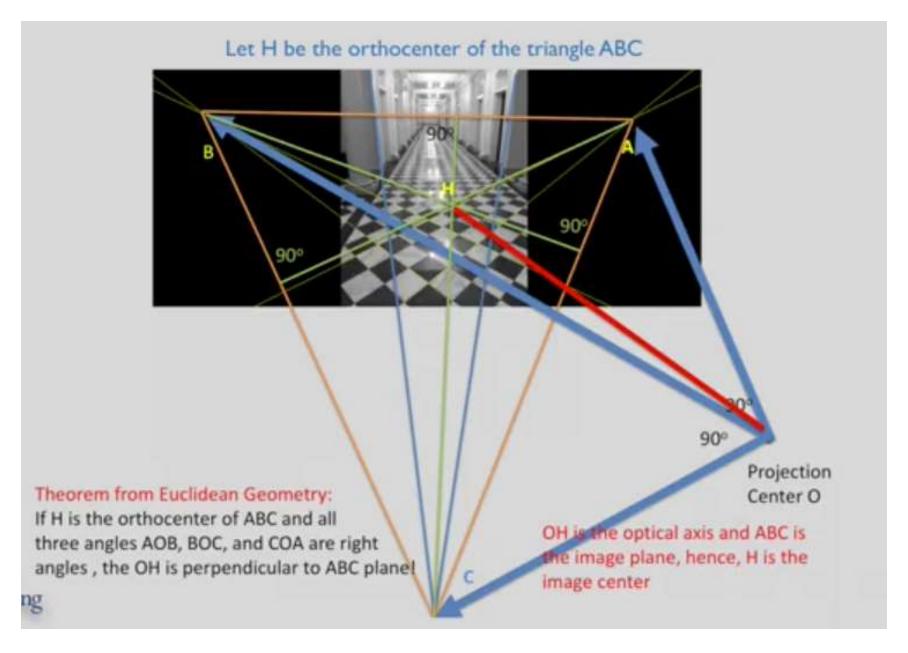
$$K = \begin{bmatrix} f & 0 & u_0 \\ 0 & f & v_0 \\ 0 & 0 & 1 \end{bmatrix} K^{-1} = \begin{bmatrix} 1/f & 0 & -u_0/f \\ 0 & 1/f & -v_0/f \\ 0 & 0 & 1 \end{bmatrix}$$

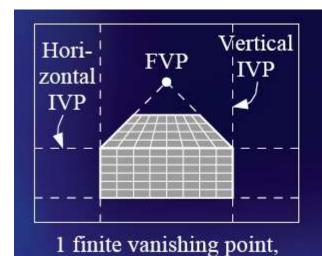
$$v_{i}^{T}K^{-T}K^{-1}v_{j} = 0$$

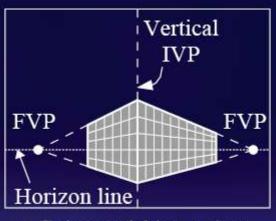
$$v_{j}^{T}K^{-T}K^{-1}v_{k} = 0$$

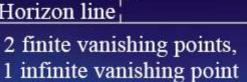
$$v_{i}^{T}K^{-T}K^{-1}v_{k} = 0$$

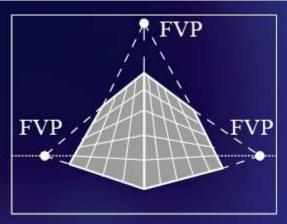
- 3 finite vanishing points: get f, u0, v0
- 2 finite and one infinite: u0,v0 as point on vf1 vf2 closest to image center, get f
- 2 infinite vanishing points: f cant be recovered u0, v0 is at the third vanishing point







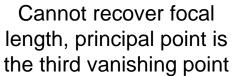




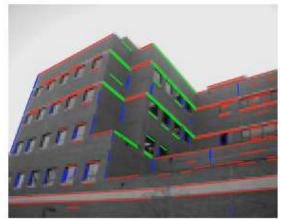
3 finite vanishing points



2 infinite vanishing points







Can solve for focal length, principal point

Rotation from vanishing points

$$\lambda_{i} \mathbf{v}_{i} = \mathbf{K} \begin{bmatrix} \mathbf{R} \mid \mathbf{t} \end{bmatrix} \begin{bmatrix} \mathbf{e}_{i} \\ 0 \end{bmatrix} = \mathbf{K} \mathbf{R} \mathbf{e}_{i}$$

$$/_{1} \mathbf{K}^{-1} \mathbf{v}_{1} = \mathbf{R} \mathbf{e}_{1} = \begin{bmatrix} \mathbf{r}_{1} & \mathbf{r}_{2} & \mathbf{r}_{3} \end{bmatrix} \begin{array}{c} \stackrel{c}{\theta} & 1 & \stackrel{\dot{\mathsf{U}}}{\mathsf{U}} \\ \stackrel{c}{\theta} & 0 & \stackrel{\dot{\mathsf{U}}}{\mathsf{U}} = \mathbf{r}_{1} \\ \stackrel{c}{\theta} & 0 & \stackrel{\dot{\mathsf{U}}}{\mathsf{U}} = \mathbf{r}_{1} \end{array}$$

Thus, $\lambda_i \mathbf{K}^{-1} \mathbf{v}_i = \mathbf{r}_i$.

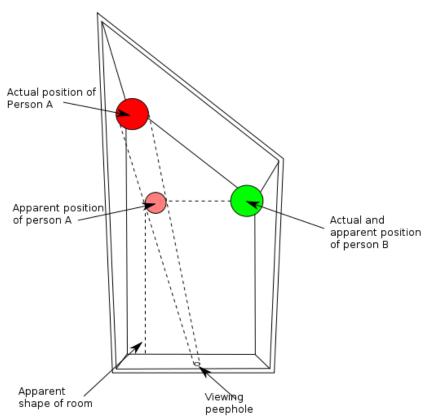
Get λ_i by using the constraint $||\mathbf{r}_i||^2=1$.

Calibration from vanishing points: Summary

- Solve for K (focal length, principal point) using three orthogonal vanishing points
- Get rotation directly from vanishing points once calibration matrix is known
- Advantages
 - No need for calibration chart, 2D-3D correspondences
 - Could be completely automatic
- Disadvantages
 - Only applies to certain kinds of scenes
 - Inaccuracies in computation of vanishing points
 - Problems due to infinite vanishing points

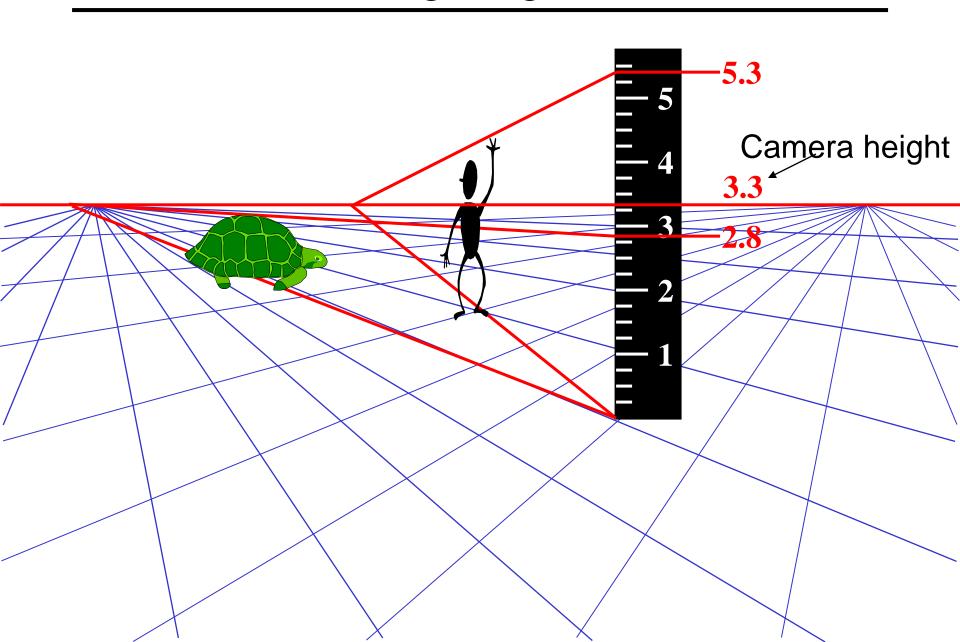
Making measurements from a single image



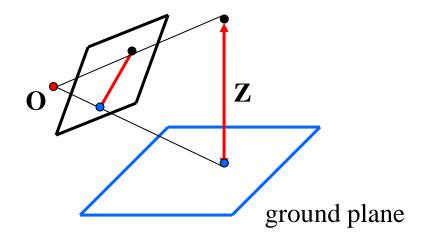


http://en.wikipedia.org/wiki/Ames_room

Recall: Measuring height



Measuring height without a ruler



Compute Z from image measurements

Need more than vanishing points to do this

Projective invariant

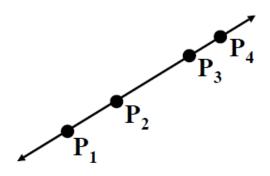
- We need to use a projective invariant: a quantity that does not change under projective transformations (including perspective projection)
 - What are some invariants for similarity, affine transformations?

The cross ratio

A Projective Invariant

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

The cross-ratio of 4 collinear points



$$\frac{\|\mathbf{P}_3 - \mathbf{P}_1\| \|\mathbf{P}_4 - \mathbf{P}_2\|}{\|\mathbf{P}_3 - \mathbf{P}_2\| \|\mathbf{P}_4 - \mathbf{P}_1\|} \qquad \mathbf{P}_i = \begin{vmatrix} X_i \\ Y_i \\ Z_i \end{vmatrix}$$

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

Can permute the point ordering

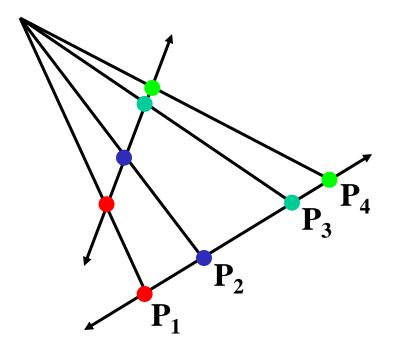
$$\frac{\|\mathbf{P}_{1} - \mathbf{P}_{3}\| \|\mathbf{P}_{4} - \mathbf{P}_{2}\|}{\|\mathbf{P}_{1} - \mathbf{P}_{2}\| \|\mathbf{P}_{4} - \mathbf{P}_{3}\|}$$

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

This is the fundamental invariant of projective geometry

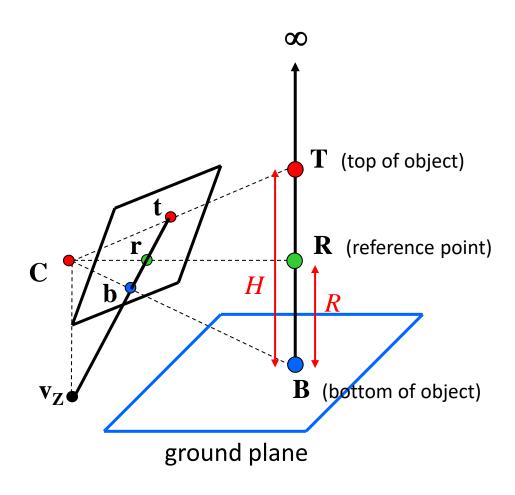
Projective invariant

- 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:



$$\frac{\|\mathbf{P}_{3} - \mathbf{P}_{1}\| \|\mathbf{P}_{4} - \mathbf{P}_{2}\|}{\|\mathbf{P}_{3} - \mathbf{P}_{2}\| \|\mathbf{P}_{4} - \mathbf{P}_{1}\|}$$

Measuring height



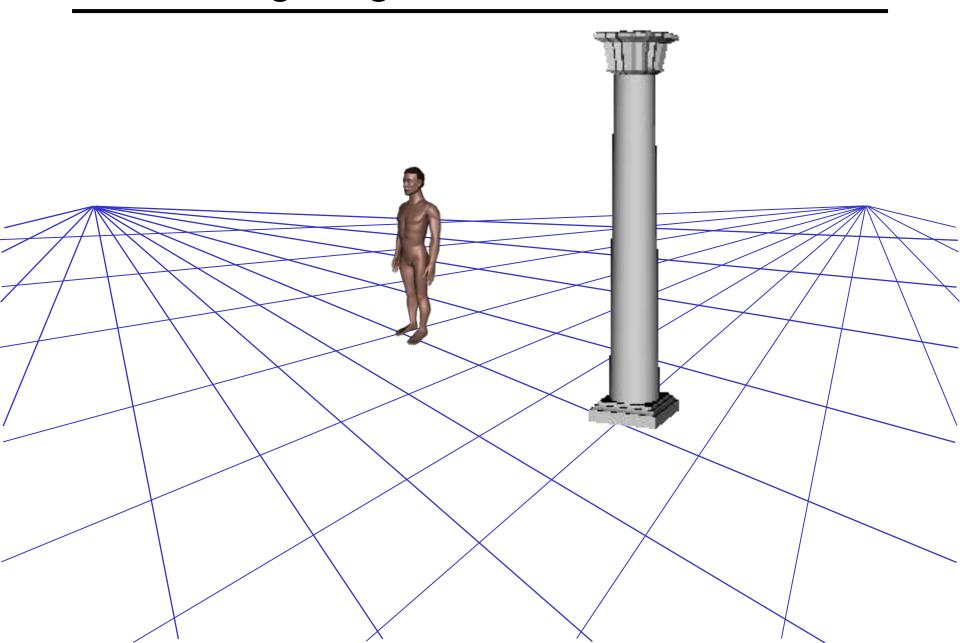
$$\frac{\|\mathbf{T} - \mathbf{B}\| \|\infty - \mathbf{R}\|}{\|\mathbf{R} - \mathbf{B}\| \|\infty - \mathbf{T}\|} = \frac{H}{R}$$

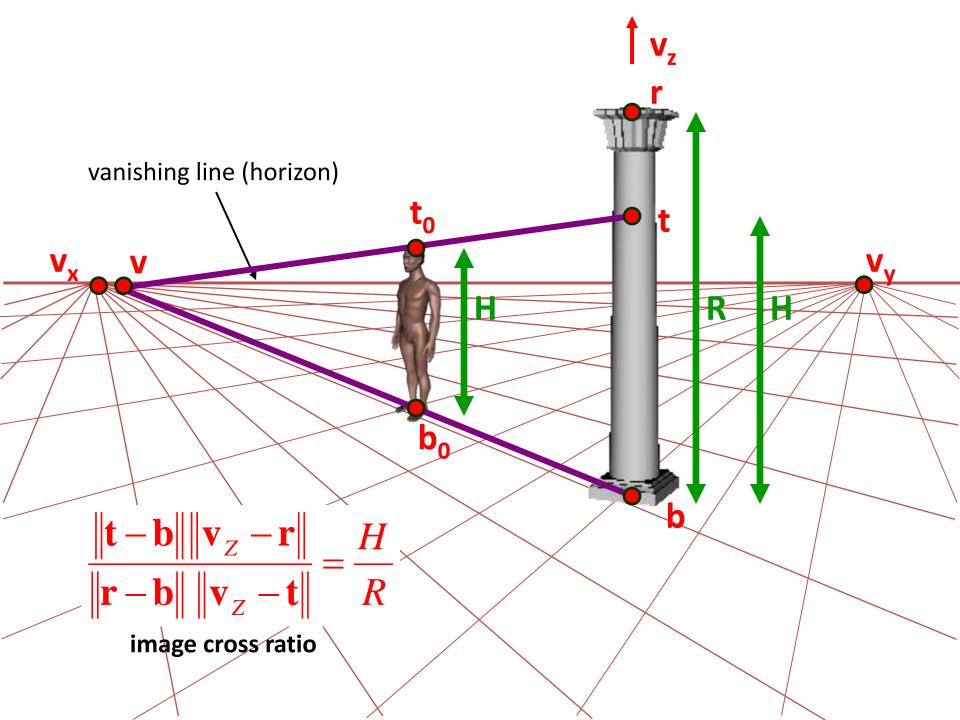
scene cross ratio

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

image cross ratio

Measuring height without a ruler



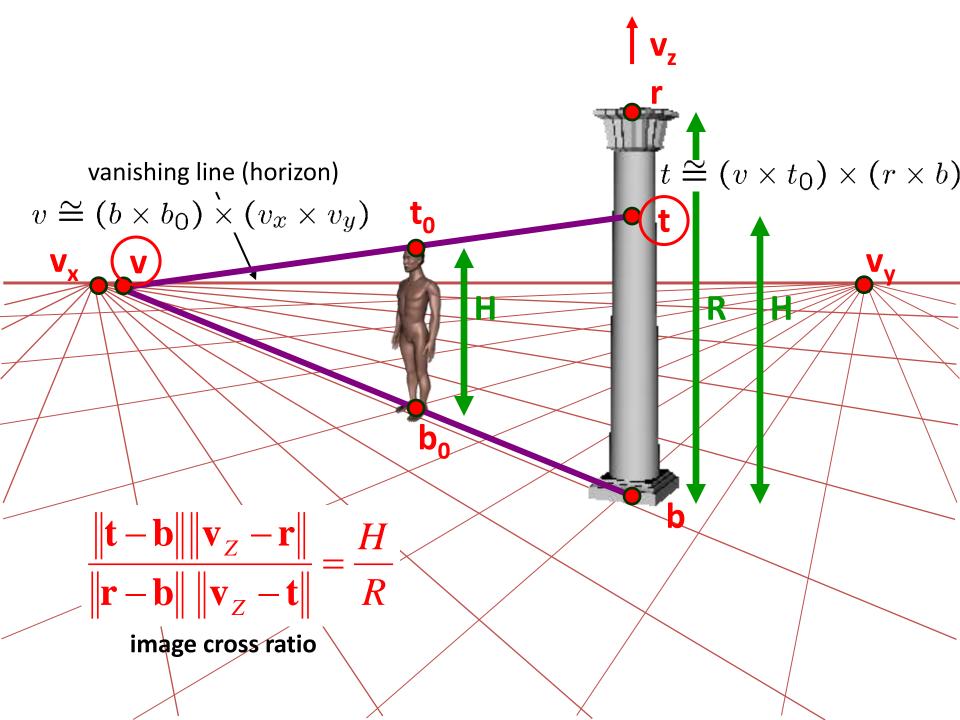


2D lines in homogeneous coordinates

• Line equation: ax + by + c = 0

$$\mathbf{l}^T \mathbf{x} = 0$$
 where $\mathbf{l} = \begin{vmatrix} a \\ b \\ c \end{vmatrix}$, $\mathbf{x} = \begin{vmatrix} x \\ y \\ 1 \end{vmatrix}$

- Line passing through two points: $\mathbf{l} = \mathbf{x}_1 \times \mathbf{x}_2$
- Intersection of two lines: $\mathbf{x} = \mathbf{l}_1 \times \mathbf{l}_2$



Measurements on planes

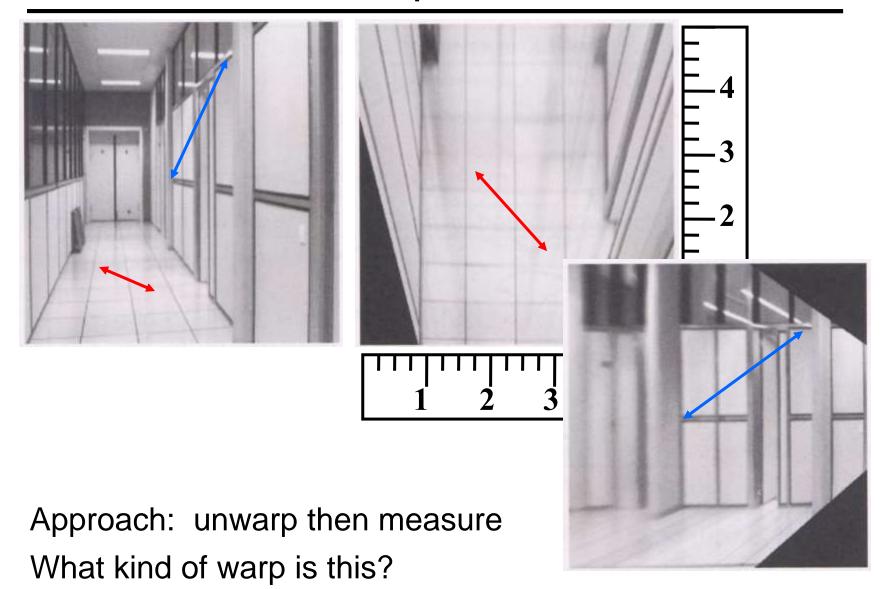
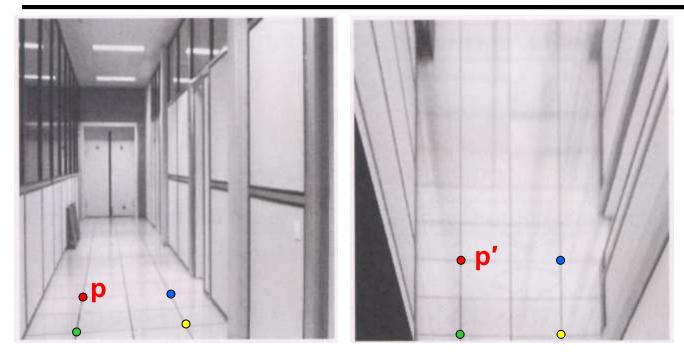


Image rectification



To unwarp (rectify) an image

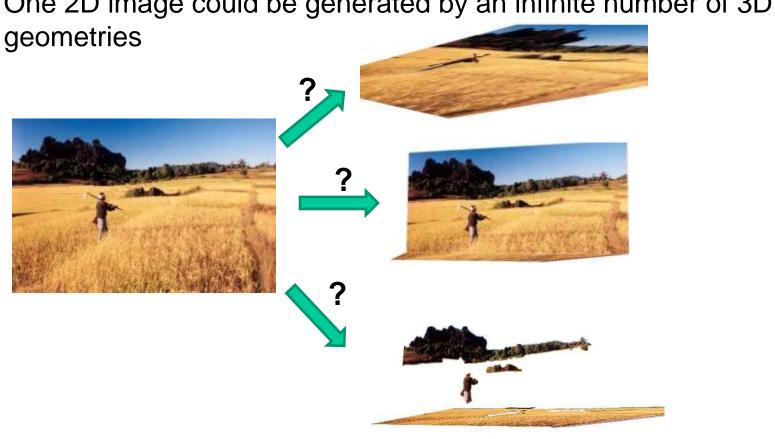
- solve for homography H given p and p'
- how many points are necessary to solve for H?

Today's class: 3D Reconstruction



The challenge

One 2D image could be generated by an infinite number of 3D



The solution

Make simplifying assumptions about 3D geometry







Unlikely

Likely

Today's class: Two Models

Box + frontal billboards

Ground plane + non-frontal billboards



"Tour into the Picture" (Horry et al. SIGGRAPH '97)

Create a 3D "theatre stage" of five planes



Specify foreground objects through bounding polygons



Use camera transformations to navigate through the scene



The idea

Many scenes can be represented as an axis-aligned box volume (i.e. a stage)

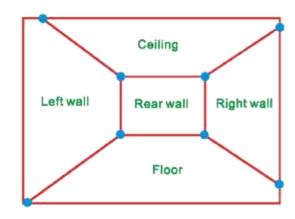
Key assumptions

All walls are orthogonal Camera view plane is parallel to back of volume

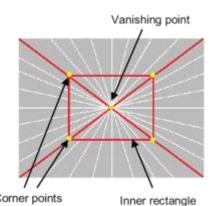
How many vanishing points does the box have?

Three, but two at infinity Single-point perspective

Can use the vanishing point to fit the box to the particular scene



Step 1: specify scene geometry



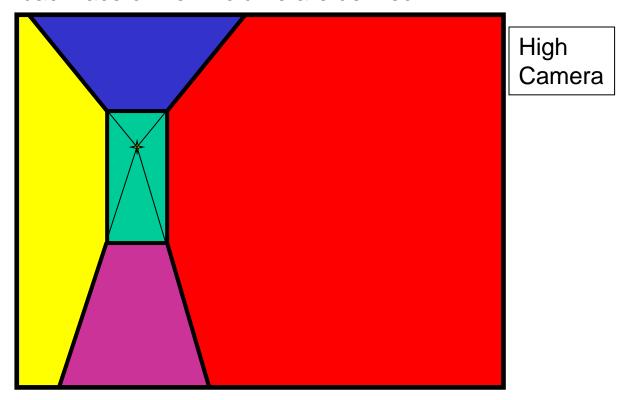


User controls the inner box and the vanishing point placement (# of DOF?)

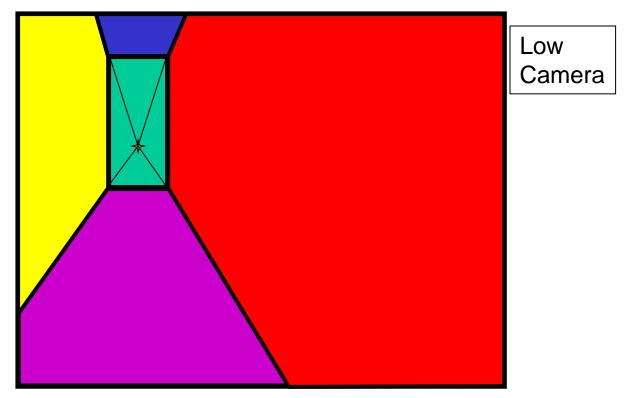
Q: If we assume camera is looking straight at back wall, what camera parameter(s) does the vanishing point position provide?

A: Vanishing point direction is perpendicular to image plane, so the vp is the principal point

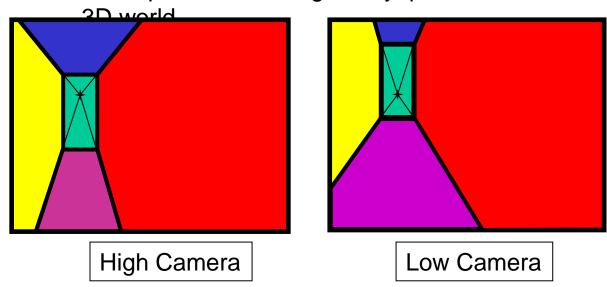
Example of user input: vanishing point and back face of view volume are defined



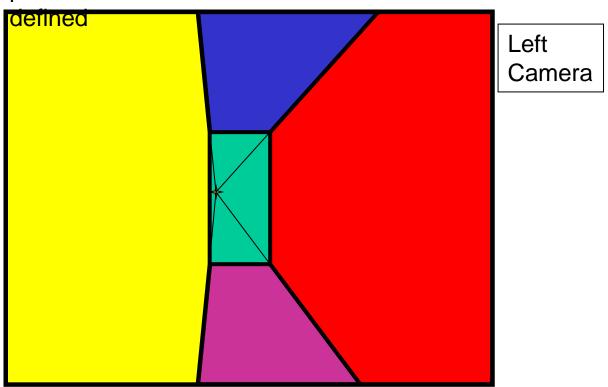
Example of user input: vanishing point and back face of view volume are defined



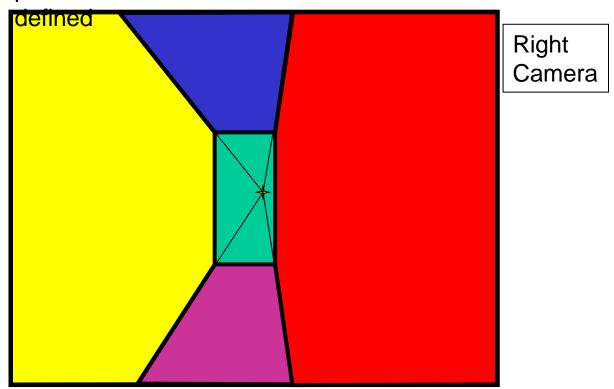
Comparison of how image is subdivided based on two different camera positions. You should see how moving the box corresponds to moving the eyepoint in the



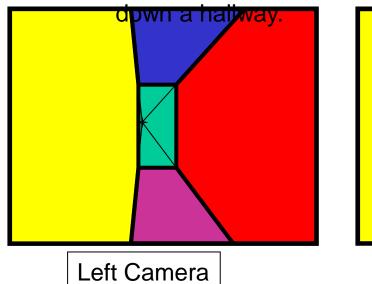
Another example of user input: vanishing point and back face of view volume are

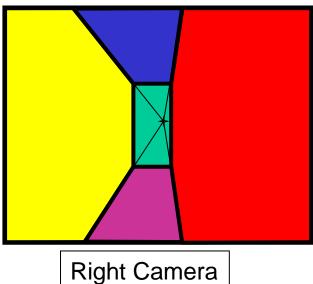


Another example of user input: vanishing point and back face of view volume are



Comparison of two camera placements – left and right. Corresponding subdivisions match view you would see if you looked

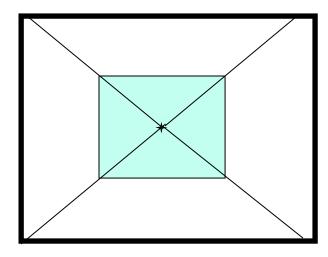




Question

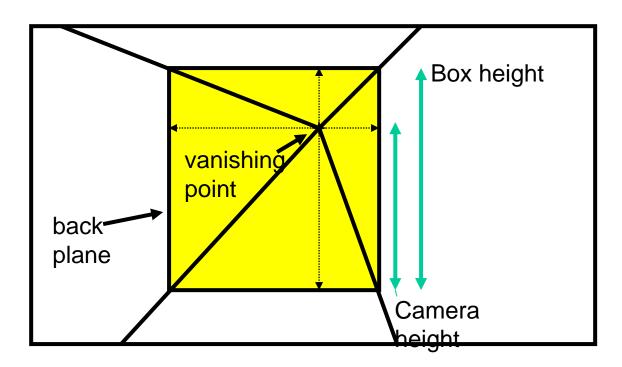
Think about the camera center and image plane...

- What happens when we move the box?
- What happens when we move the vanishing point?



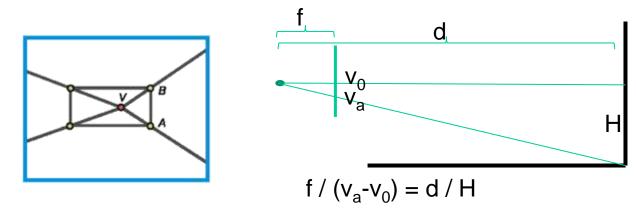
2D to 3D conversion

Use ratios



Box width / height in 3D is proportional to width over height in the image because back plane is parallel to image plane

Get depth using similar triangles

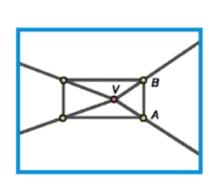


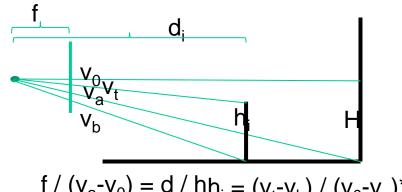
Can compute by similar triangles (CVA vs. CV'A') Need to know focal length f (or FoV)

Note: can compute position of any object on the ground

- Simple unprojection
- What about things off the ground?

Get depth using similar triangles





$$f / (v_a-v_0) = d / hh_i = (v_t-v_b) / (v_0-v_a)*H$$

 $f / (v_b-v_t) = d_i / h_i$

Can compute by similar triangles (CVA vs. CV'A') Need to know focal length f (or FoV)

Can compute 3D position of any object on the ground w/ unprojection

What about things off the ground?

Step 2: map image textures into frontal view

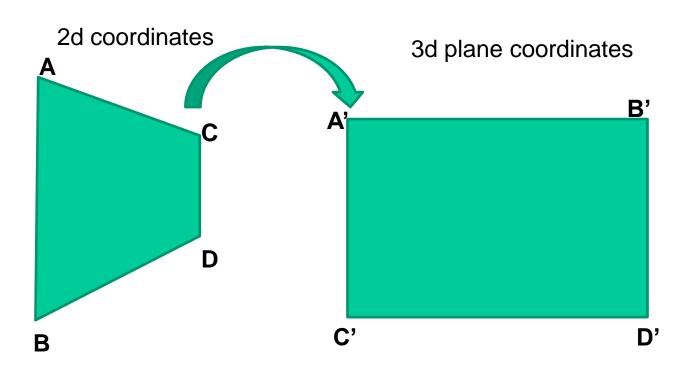
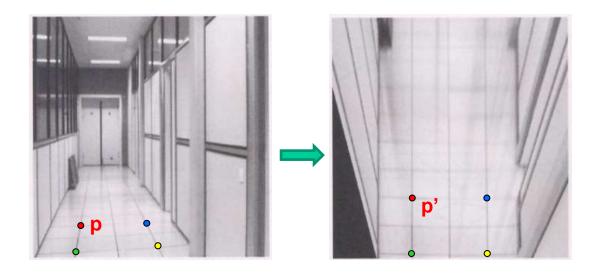


Image rectification by homography



To unwarp (rectify) an image solve for homography **H** given **p** and **p'**: w**p'=Hp**

Computing homography

Assume we have four matched points: How do we compute homography **H**?

Direct Linear Transformation (DLT)

$$\mathbf{p'} = \mathbf{Hp} \qquad \mathbf{p'} = \begin{bmatrix} w'u' \\ w'v' \\ w' \end{bmatrix} \qquad \mathbf{p} = \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} \qquad \mathbf{H} = \begin{bmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & h_9 \end{bmatrix}$$

$$\begin{bmatrix} -u & -v & -1 & 0 & 0 & 0 & uu' & vu' & u' \\ 0 & 0 & 0 & -u & -v & -1 & uv' & vv' & v' \end{bmatrix} \mathbf{h} = \mathbf{0}$$

$$\mathbf{h} = \begin{bmatrix} h_2 \\ h_3 \\ h_4 \\ h_5 \\ h_6 \\ h_7 \\ h_8 \\ h_9 \end{bmatrix}$$

Computing homography

Direct Linear Transform

$$\begin{bmatrix} -u_1 & -v_1 & -1 & 0 & 0 & 0 & u_1u_1' & v_1u_1' & u_1' \\ 0 & 0 & 0 & -u_1 & -v_1 & -1 & u_1v_1' & v_1v_1' & v_1' \\ & & \vdots & & & & \\ 0 & 0 & 0 & -u_n & -v_n & -1 & u_nv_n' & v_nv_n' & v_n' \end{bmatrix} \mathbf{h} = \mathbf{0} \Rightarrow \mathbf{A}\mathbf{h} = \mathbf{0}$$

Apply SVD: $USV^T = A$

$$\mathbf{h} = \mathbf{V}_{\text{smallest}} \text{ (column of } \mathbf{V}^{T} \text{ corresponds to smallest singular value)}$$

$$\mathbf{h} = \begin{bmatrix} h_{1} & h_{2} & h_{3} \\ h_{2} & \vdots \\ h_{9} \end{bmatrix} \text{ } \mathbf{H} = \begin{bmatrix} h_{1} & h_{2} & h_{3} \\ h_{4} & h_{5} & h_{6} \\ h_{7} & h_{8} & h_{9} \end{bmatrix} \text{ } \mathbf{V}^{T} \text{ corresponds to smallest singular value)}$$

$$\mathbf{U}, S, Vt = \text{scipy.linalg.svd} \text{ (A)}$$

$$\text{# last column corr. to smallest singular value}$$

$$\mathbf{h} = Vt[:,-1];$$

Explanation of SVD, solving systems of linear equations, derivation of solution here

Solving for homography (another formulation)

$$\begin{bmatrix} -u_1 & -v_1 & -1 & 0 & 0 & 0 & u_1u_1' & v_1u_1' & u_1' \\ 0 & 0 & 0 & -u_1 & -v_1 & -1 & u_1v_1' & v_1v_1' & v_1' \\ & & & \vdots & & & & \\ 0 & 0 & 0 & -u_n & -v_n & -1 & u_nv_n' & v_nv_n' & v_n' \end{bmatrix} \mathbf{h} = \mathbf{0} \Rightarrow \mathbf{A}\mathbf{h} = \mathbf{0}$$

$$\mathbf{A}$$

$$\mathbf{a}$$

$$\mathbf{a}$$

$$\mathbf{a}$$

$$\mathbf{a}$$

$$\mathbf{a}$$

$$\mathbf{a}$$

$$\mathbf{b}$$

$$\mathbf{a}$$

$$\mathbf{b}$$

$$\mathbf{a}$$

$$\mathbf{c}$$

$$\mathbf{n}$$

Defines a least squares problem: $\min \|Ah - 0\|^2$

- Since h is only defined up to scale, solve for unit vector h
- Solution: $\hat{\mathbf{h}}$ = eigenvector of $\mathbf{A}^T\mathbf{A}$ with smallest eigenvalue
 - Can derive using Lagrange multipliers method
- · Works with 4 or more points

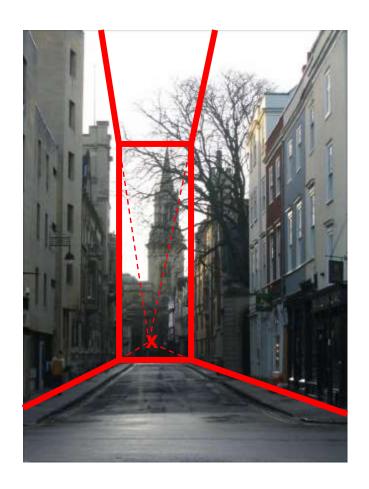
Tour into the picture algorithm

1. Set the box corners



Tour into the picture algorithm

- 1. Set the box corners
- 2. Set the VP
- 3. Get 3D coordinates
 - Compute height, width, and depth of box
- 4. Get texture maps
 - homographies for each face
- 5. Create file to store plane coordinates and texture maps



Result

Render from new views







http://www.cs.cmu.edu/afs/cs.cmu.edu/academic/class/15463-f08/www/proj5/www/dmillett/

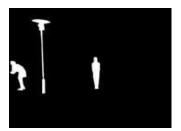
Foreground Objects

Use separate billboard for each



For this to work, three separate images used:

- Original image.
- Mask to isolate desired foreground images.
- Background with



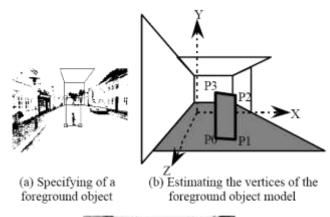


Foreground Objects

Add vertical rectangles for each foreground object

Can compute 3D coordinates P0, P1 since they are on known plane.

P2, P3 can be computed as before (similar triangles)

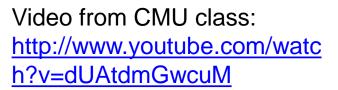




(c) Three foreground object models

Foreground Result

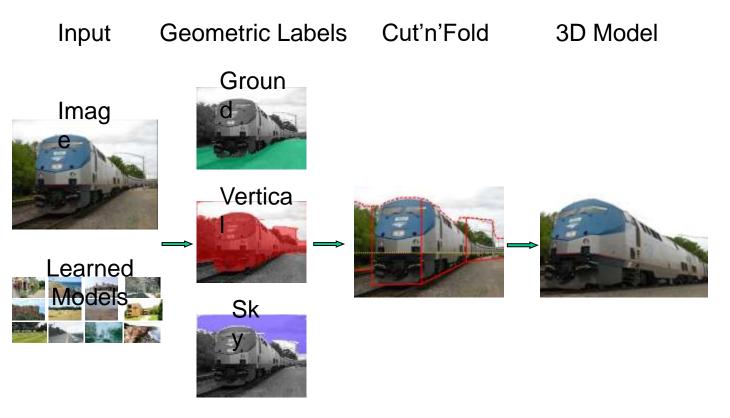




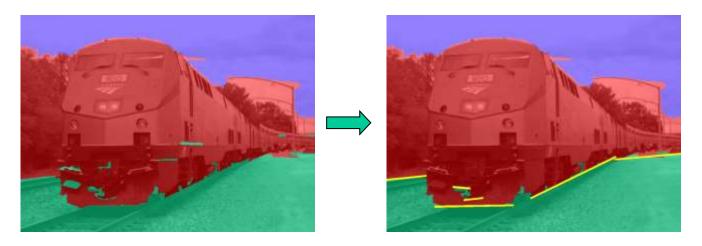




Automatic Photo Pop-up

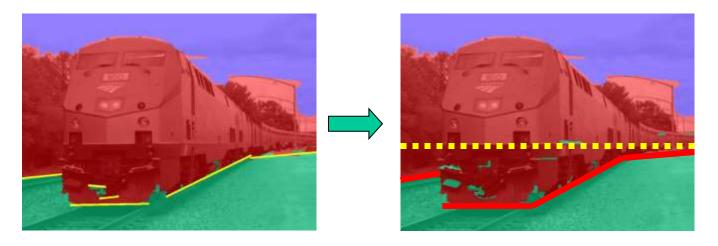


Hoiem et al. 2005



Fit ground-vertical boundary

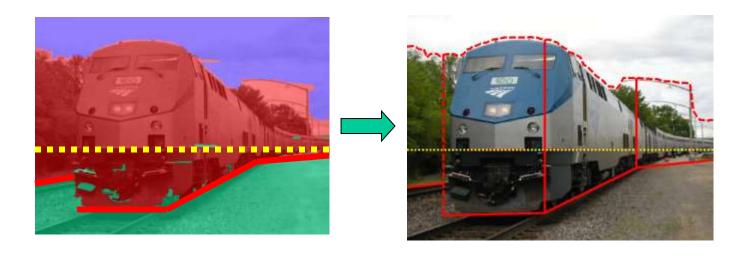
Iterative Hough transform



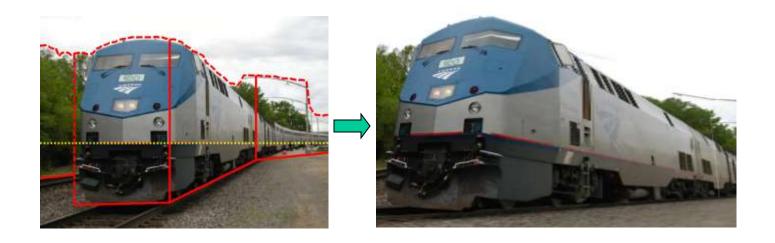
Form polylines from boundary segments

- Join segments that intersect at slight angles
- Remove small overlapping polylines

Estimate horizon position from perspective cues



- "Fold" along polylines and at corners
- "Cut" at ends of polylines and along vertical-sky boundary



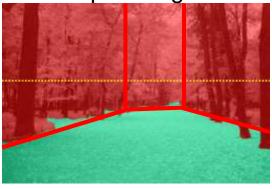
Construct 3D model Texture map

Results

http://www.cs.illinois.edu/homes/dhoiem/projects/popup/



Input Image



Cut and Fold





Automatic Photo Popup

Results





Input Image





Automatic Photo Popup

Comparison with Manual Method



Input Image



[Liebowitz et al. 1999]



Automatic Photo Pop-up (15 sec)!

Failures



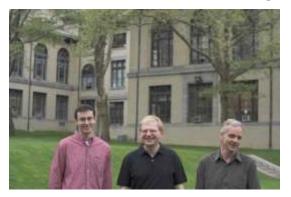


Labeling Errors



Failures

Foreground Objects

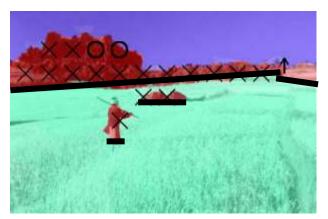




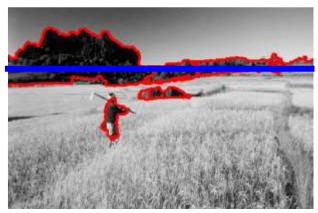




Adding Foreground Labels



Recovered Surface Labels + Ground-Vertical Boundary Fit



Object Boundaries + Horizon



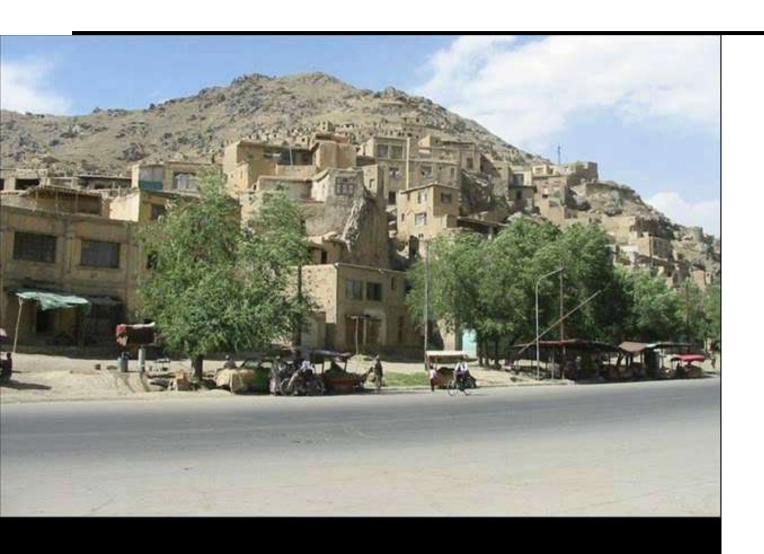
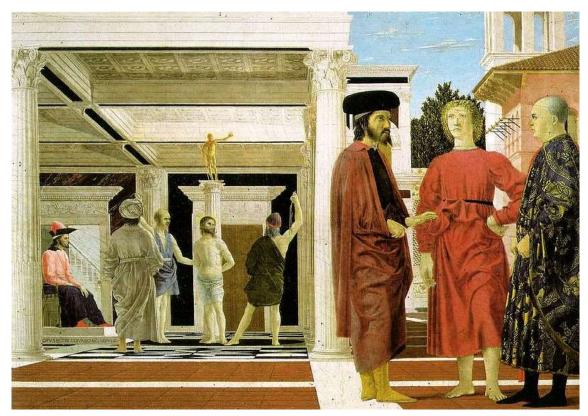
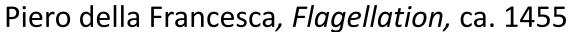


Image rectification: example







Application: 3D modeling from a single image



J. Vermeer, Music Lesson, 1662

A. Criminisi, M. Kemp, and A. Zisserman, <u>Bringing Pictorial</u>
<u>Space to Life: computer techniques for the analysis of paintings</u>, *Proc. Computers and the History of Art*, 2002

http://research.microsoft.com/en-us/um/people/antcrim/ACriminisi_3D_Museum.wmv

Application: 3D modeling from a single image



D. Hoiem, A.A. Efros, and M. Hebert, "Automatic Photo Pop-up", SIGGRAPH 2005.

http://dhoiem.cs.illinois.edu/projects/popup/popup_movie_450_250.mp4

Application: Image editing

Inserting synthetic objects into images:

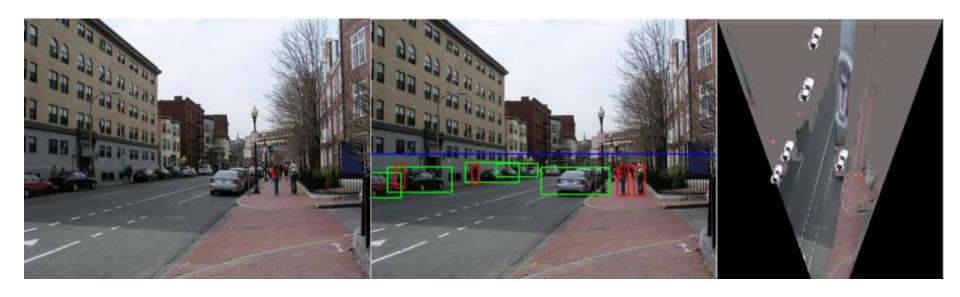
http://vimeo.com/28962540





K. Karsch and V. Hedau and D. Forsyth and D. Hoiem, "Rendering Synthetic Objects into Legacy Photographs," *SIGGRAPH Asia* 2011

Application: Object recognition



D. Hoiem, A.A. Efros, and M. Hebert, "Putting Objects in Perspective", CVPR 2006