# **CS5670: Computer Vision**

#### Single-View Modeling



# **Single-View Modeling**



Ames Room

- Readings
  - Mundy and Zisserman. Geometric Invariance in Computer Vision, Appendix: Projective Geometry for Machine Vision, MIT Press, 1992, (read 23.1-23.5, 23.10)
    - available online: <u>http://www.cs.cmu.edu/~ph/869/papers/zisser-mundy.pdf</u>

#### **Ames Room**





#### Announcements

- Project 3: Autostitch (Panorama Stitching)
  - Due on Friday, March 17, by 8pm
  - To be done in groups of 2
  - If you need help finding a team member, let us know

#### **Forced perspective in film**



How Lord of the Rings used forced perspective shots with a moving camera <u>https://www.youtube.com/watch?v=QWMFpxkGO\_s</u>

# Projective geometry—what's it good for?

- Uses of projective geometry
  - Drawing
  - Measurements
  - Mathematics for projection
  - Undistorting images
  - Camera pose estimation
  - Object recognition



#### Paolo Uccello

### **Applications of projective geometry**



Vermeer's Music Lesson





Reconstructions by Criminisi et al.

# Making measurements in images

WARBY PARKER

#### Measure your pupillary distance (PD)

Your PD is the distance between your pupils. To measure it, follow the instructions below — once you submit your photo, our team of experts will determine your PD and email you once we've applied it to your order.





Wearing glasses? Take 'em off before you get started.



Hold up any card with a magnetic strip (we use this for scale).



Look straight ahead and snap a photo.

#### **Measurements on planes**



# **Point and line duality**

- A line I is a homogeneous 3-vector
- It is  $\perp$  to every point (ray) **p** on the line: **I**·**p**=0



What is the line **I** spanned by points **p**<sub>1</sub> and **p**<sub>2</sub>?

- I is  $\perp$  to  $\mathbf{p_1}$  and  $\mathbf{p_2} \implies \mathbf{I} = \mathbf{p_1} \times \mathbf{p_2}$
- I can be interpreted as a *plane normal*



What is the intersection of two lines I<sub>1</sub> and I<sub>2</sub>?

• **p** is  $\perp$  to  $\mathbf{I_1}$  and  $\mathbf{I_2} \implies \mathbf{p} = \mathbf{I_1} \times \mathbf{I_2}$ 

Points and lines are *dual* in projective space

### Example



What is the line passing through points **p** and **q**?

 $\mathbf{p}\times\mathbf{q}$ 

#### Example



#### Example



What is the line passing through points **p** and **r**?

$$\mathbf{p} \times \mathbf{r} = \begin{bmatrix} 100\\200\\1 \end{bmatrix} \times \begin{bmatrix} 150\\150\\1 \end{bmatrix} = \begin{bmatrix} 200 \cdot 1 - 150 \cdot 1\\150 \cdot 1 - 100 \cdot 1\\100 \cdot 150 - 150 \cdot 200 \end{bmatrix} = \begin{bmatrix} 50\\50\\-15000 \end{bmatrix} \sim \begin{bmatrix} 1\\1\\-300 \end{bmatrix}$$
i.e., all points (x, y) such that  $x + y = 300$ 

### **Question time**



Consider the above image, with four points **p**, **q**, **r**, **s**, labeled (assume these are 2D homogeneous points).

What is a simple expression for the point of intersection between the line through **p** and **r** and the line through **q** and **s**?

### **Question time**



Consider the above image, with four points **p**, **q**, **r**, **s**, labeled (assume these are 2D homogeneous points).

What is a simple expression for the point of intersection between the line through **p** and **r** and the line through **q** and **s**?

#### Answer: $(\mathbf{p} \times \mathbf{r}) \times (\mathbf{q} \times \mathbf{s})$

# **Ideal points and lines**





- Ideal point ("point at infinity")
  - $-p \cong (x, y, 0)$  parallel to image plane
  - It has infinite image coordinates
- Ideal line
  - $I \cong (a, b, 0)$  parallel to image plane
  - Corresponds to a line in the image (finite coordinates)
    goes through image origin (*principal point*)

# **3D projective geometry**

- These concepts generalize naturally to 3D
  - Homogeneous coordinates
    - Projective 3D points have four coords:  $\mathbf{P} = (X, Y, Z, W)$
  - Duality
    - A plane **N** is also represented by a 4-vector
    - Points and planes are dual in 3D: N P=0
    - Three points define a plane, three planes define a point

#### **3D to 2D: perspective projection**

Projection:



#### Figure 23.4

A perspective view of a set of parallel lines in the plane. All of the lines converge to a single vanishing point.

# Vanishing points (1D)



- Vanishing point
  - projection of a point at infinity
  - can often (but not always) project to a finite point in the image
    camera center

image plane —

ground plane

# Vanishing points (2D)





- Properties
  - Any two parallel lines (in 3D) have the same vanishing point  ${\bf v}$
  - The ray from  ${\bf C}$  through  ${\bf v}$  is parallel to the lines
  - An image may have more than one vanishing point
    - in fact, every image point is a potential vanishing point



- Properties  $\mathbf{v} = \mathbf{\Pi} \mathbf{P}_{\infty}$ 
  - $\mathbf{P}_{\infty}$  is a point at *infinity*, **v** is its projection
  - Depends only on line *direction*
  - Parallel lines  $\mathbf{P}_0$  + t $\mathbf{D}$ ,  $\mathbf{P}_1$  + t $\mathbf{D}$  intersect at  $\mathbf{P}_{\infty}$

#### **One-point perspective**



#### **Two-point perspective**



# **Three-point perspective**



#### **Questions?**

# **Vanishing lines**



- Multiple Vanishing Points
  - Any set of parallel lines on the plane define a vanishing point
  - The union of all of these vanishing points is the horizon line
    - also called *vanishing line*
  - Note that different planes (can) define different vanishing

# **Vanishing lines**



- Multiple Vanishing Points
  - Any set of parallel lines on the plane define a vanishing point
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    - also called *vanishing line*
  - Note that different planes (can) define different vanishing

# **Computing vanishing lines**





- Properties
  - I is intersection of horizontal plane through **C** with image plane
  - Compute I from two sets of parallel lines on ground plane
  - All points at same height as C project to I
    - points higher than C project above I
  - Provides way of comparing height of objects in the scene



#### Lots of fun with vanishing points



### **Perspective cues**



#### **Perspective cues**



#### **Perspective cues**



# **Comparing heights**





# **Computing vanishing points (from lines)**



Intersect  $p_1q_1$  with  $p_2q_2$ 

$$v = (p_1 \times q_1) \times (p_2 \times q_2)$$

Least squares version

- Better to use more than two lines and compute the "closest" point of intersection
- See notes by <u>Bob Collins</u> for one good way of doing this:
  - <u>http://www-2.cs.cmu.edu/~ph/869/www/notes/vanishing.txt</u>



# Measuring height without a ruler



Compute Z from image measurements

• Need more than vanishing points to do this

# The cross ratio

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

The cross-ratio of 4 collinear points





 $\left| \begin{array}{c} Y_i \\ Z_i \end{array} \right|$ 

Can permute the point ordering

• 4! = 24 different orders (but only 6 distinct values) This is the fundamental invariant of projective geometry  $\frac{\left\|\mathbf{P}_{1}-\mathbf{P}_{3}\right\|\left\|\mathbf{P}_{4}-\mathbf{P}_{2}\right\|}{\left\|\mathbf{P}_{1}-\mathbf{P}_{2}\right\|\left\|\mathbf{P}_{4}-\mathbf{P}_{3}\right\|}$ 

# **Measuring height**



# Finding the vertical (z) vanishing point







- Here the person 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



St. Jerome in his Study, H. Steenwick

Bringing Pictorial Space to Life: Computer Techniques for the Analysis of Paintings. Antonio Criminisi, Martin Kemp, Andrew Zisserman. 2002.





Flagellation, Piero della Francesca



video by Antonio Criminisi





Flagellation. Piero della Francesca. c1453.

# **Related problem: camera calibration**

- Goal: estimate the camera parameters
  - Version 1: solve for 3x4 projection matrix

- Version 2: solve for camera parameters separately
  - intrinsics (focal length, principal point, pixel size)
  - extrinsics (rotation angles, translation)
  - radial distortion

# Vanishing points and projection matrix

• 
$$\boldsymbol{\pi}_1 = \boldsymbol{\Pi} \begin{bmatrix} 1 & 0 & 0 \end{bmatrix}^T = \boldsymbol{v}_x$$
 (X vanishing point)

• similarly, 
$$\boldsymbol{\pi}_2 = \boldsymbol{v}_Y, \ \boldsymbol{\pi}_3 = \boldsymbol{v}_Z$$

• 
$$\boldsymbol{\pi}_4 = \boldsymbol{\Pi} \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}^T$$
 = projection of world origin

$$\mathbf{\Pi} = \begin{bmatrix} \mathbf{v}_X & \mathbf{v}_Y & \mathbf{v}_Z & \mathbf{0} \end{bmatrix}$$

Not So Fast! We only know v's up to a scale factor

$$\mathbf{\Pi} = \begin{bmatrix} a \mathbf{v}_X & b \mathbf{v}_Y & c \mathbf{v}_Z & \mathbf{0} \end{bmatrix}$$

• Can fully specify by providing 3 reference points with known coordinates

# **Calibration using a reference object**

- Place a known object in the scene
  - identify correspondence between image and scene
  - compute mapping from scene to image

lssues

- must know geometry very accurately
- must know 3D -> 2D correspondence



#### **AR codes**



# **Estimating the projection matrix**

- Place a known object in the scene
  - identify correspondence between image and scene
  - compute mapping from scene to image

$$\begin{bmatrix} u_i \\ v_i \\ 1 \end{bmatrix} \cong \begin{bmatrix} m_{00} & m_{01} & m_{02} & m_{03} \\ m_{10} & m_{11} & m_{12} & m_{13} \\ m_{20} & m_{21} & m_{22} & m_{23} \end{bmatrix} \begin{bmatrix} X_i \\ Y_i \\ Z_i \\ 1 \end{bmatrix}$$



# **Alternative: multi-plane calibration**



Images courtesy Jean-Yves Bouguet

#### Advantage

- Only requires a plane
- Don't have to know positions/orientations
- Good code available online! (including in OpenCV)
  - Matlab version by Jean-Yves Bouget: <u>http://www.vision.caltech.edu/bouguetj/calib\_doc/index.html</u>
  - Amy Tabb's camera calibration software: <u>https://github.com/amy-tabb/basic-camera-calibration</u>

# Single-image depth prediction using deep learning



Image



Li and Snavely. Megadepth: Learning single-view depth prediction from internet photos. CVPR 2018.

# **MiDaS depth prediction**

Ranftl et al. Towards Robust Monocular Depth Estimation: Mixing Datasets for Zero-shot Cross-dataset Transfer.



https://gradio.app/g/AK391/MiDaS

https://github.com/intel-isl/MiDaS

#### Single-image depth prediction





Miangoleh<sup>\*</sup>, Dille<sup>\*</sup>, Mai, Paris, and Aksoy. Boosting Monocular Depth Estimation Models to High-Resolution via Content-Adaptive Multi-Resolution Merging.

# **Deep geometry prediction**

• More on this topic later!

#### **Questions?**