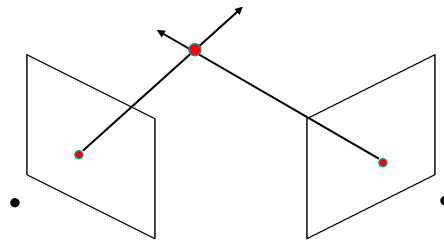


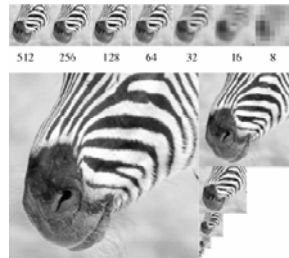
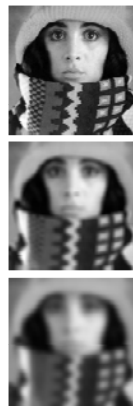
Epipolar geometry & stereo vision

Tuesday, Oct 20

Kristen Grauman
UT-Austin



Recap: Features and filters



Transforming and
describing images;
textures, colors, edges

Recap: Grouping & fitting



Parallelism



Symmetry



Continuity

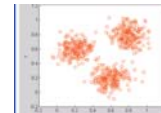
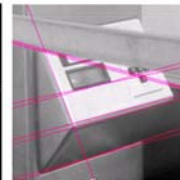
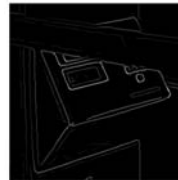


Closure

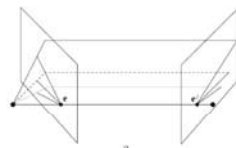
Clustering,
segmentation,
fitting; what parts
belong together?



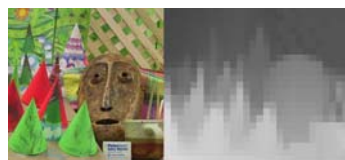
[fig from Shi et al]



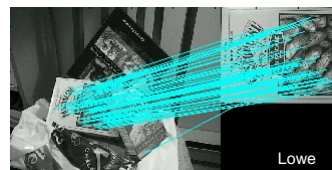
Multiple views



Hartley and Zisserman



Multi-view geometry,
matching, invariant
features, stereo vision



Lowe



Fei-Fei Li

Why multiple views?

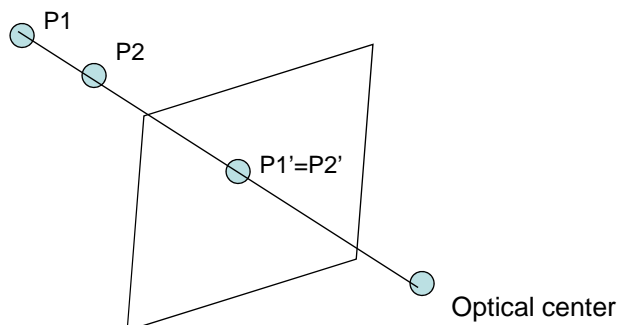
- Structure and depth are inherently ambiguous from single views.



Images from Lana Lazebnik

Why multiple views?

- Structure and depth are inherently ambiguous from single views.



- What cues help us to perceive 3d shape and depth?

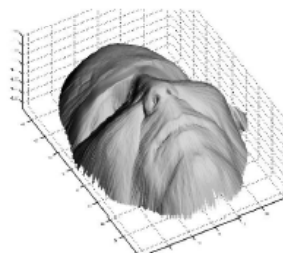
Shading



a)



b)



c)

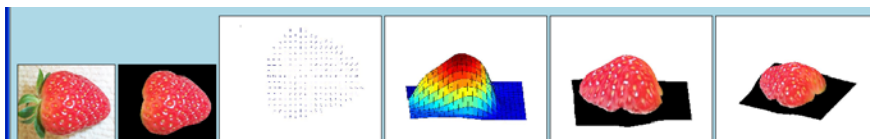
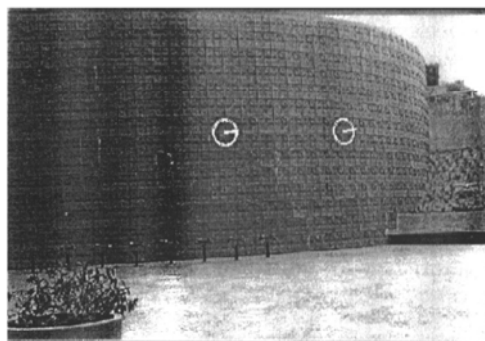
[Figure from Prados & Faugeras 2006]

Focus/Defocus



[Figure from H. Jin and P. Favaro, 2002]

Texture



[From [A.M. Loh, The recovery of 3-D structure using visual texture patterns](#), PhD thesis]

Perspective effects



Image credit: S. Seitz

Motion



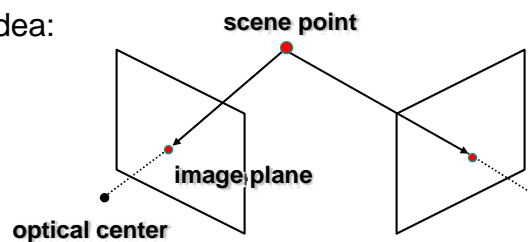
Figures from L. Zhang

<http://www.brainconnection.com/teasers/?main=illusion/motion-shape>

Estimating scene shape

- “Shape from X”: Shading, Texture, Focus, Motion...
- **Stereo:**
 - shape from “motion” between two views
 - infer 3d shape of scene from two (multiple) images from different viewpoints

Main idea:

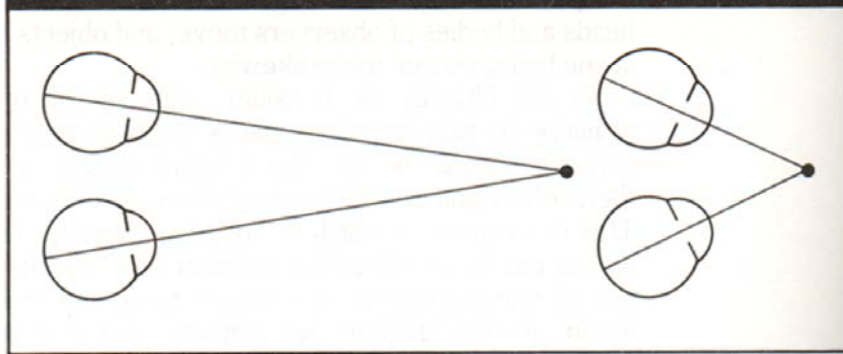


Outline

- Human stereopsis
- Stereograms
- Epipolar geometry and the epipolar constraint
 - Case example with parallel optical axes
 - General case with calibrated cameras

Human stereopsis: disparity

FIGURE 7.1

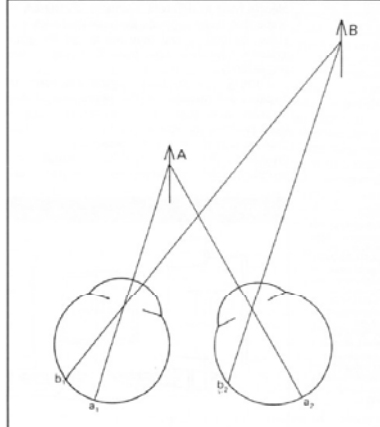


From Bruce and Green, Visual Perception,
Physiology, Psychology and Ecology

Human eyes **fixate** on point in space – rotate so that corresponding images form in centers of fovea.

Human stereopsis: disparity

FIGURE 7.3

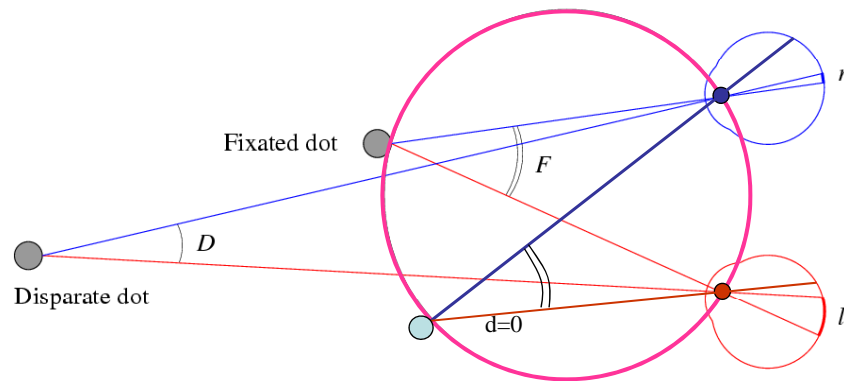


From Bruce and Green, Visual Perception,
Physiology, Psychology and Ecology

Adapted from David Forsyth, UC Berkeley

Disparity occurs when eyes fixate on one object; others appear at different visual angles

Human stereopsis: disparity



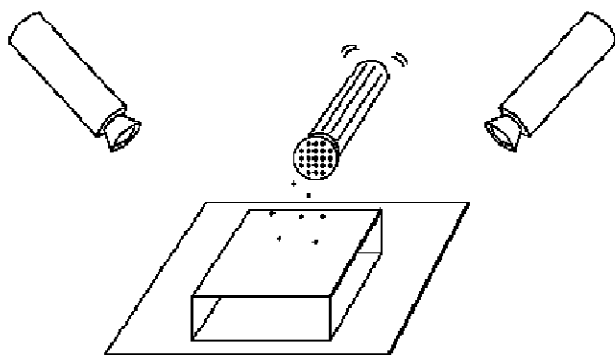
Disparity: $d = r - l = D - F$.

Forsyth & Ponce

Random dot stereograms

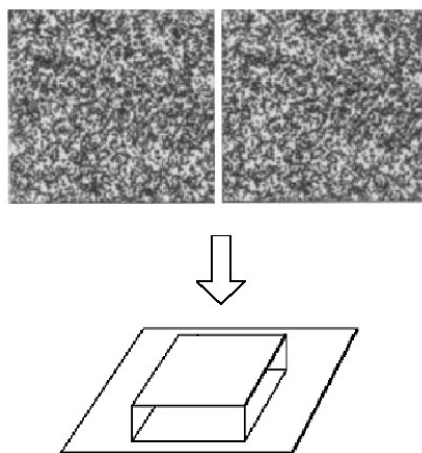
- Julesz 1960: Do we identify local brightness patterns before fusion (monocular process) or after (binocular)?
- To test: pair of synthetic images obtained by randomly spraying black dots on white objects

Random dot stereograms



Forsyth & Ponce

Random dot stereograms



Random dot stereograms

- When viewed monocularly, they appear random; when viewed stereoscopically, see 3d structure.
- Conclusion: human binocular fusion not directly associated with the physical retinas; must involve the central nervous system
- Imaginary “*cyclopean retina*” that combines the left and right image stimuli as a single unit

Stereo photography and stereo viewers

Take two pictures of the same subject from two slightly different viewpoints and display so that each eye sees only one of the images.



Invented by Sir Charles Wheatstone, 1838



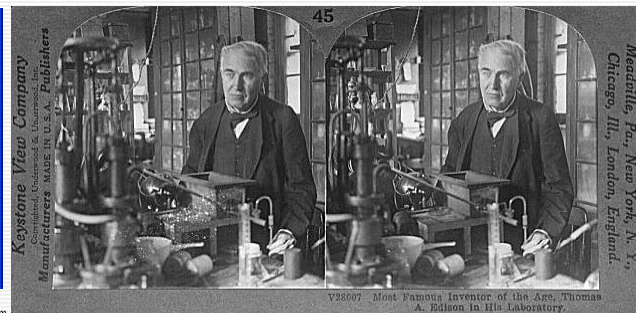
Image courtesy of fisher-price.com



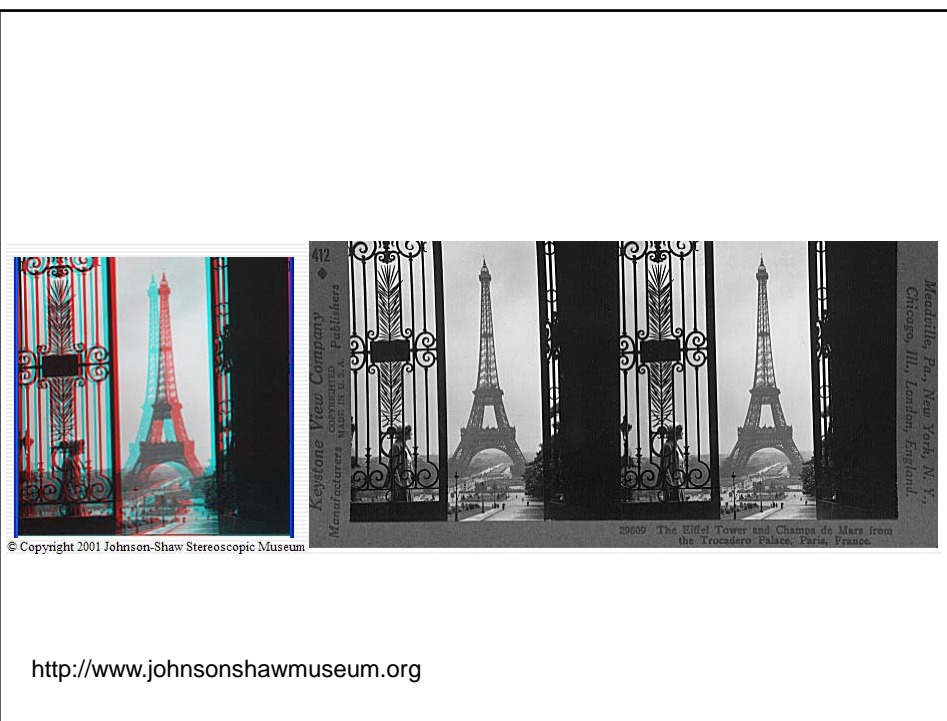
Public Library, Stereoscopic Looking Room, Chicago, by Phillips, 1923



© Copyright 2001 Johnson-Shaw Stereoscopic Museum



<http://www.johnsonshawmuseum.org>



Autostereograms



Exploit disparity as depth cue using single image.

(Single image random dot stereogram, Single image stereogram)

Images from magiceye.com

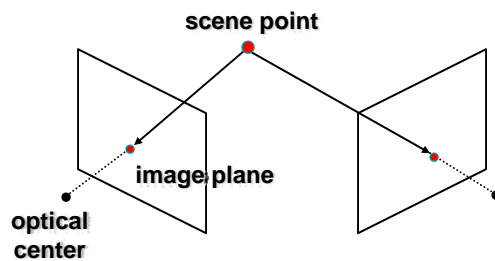
Autostereograms



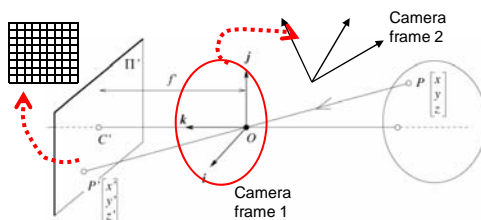
Images from magiceye.com

Estimating depth with stereo

- **Stereo**: shape from “motion” between two views
- We’ll need to consider:
 - Info on camera pose (“calibration”)
 - Image point correspondences



Camera parameters



Extrinsic parameters:
Camera frame 1 \leftrightarrow Camera frame 2

Intrinsic parameters:
Image coordinates relative to camera \leftrightarrow Pixel coordinates

- *Extrinsic* params: rotation matrix and translation vector
- *Intrinsic* params: focal length, pixel sizes (mm), image center point, radial distortion parameters

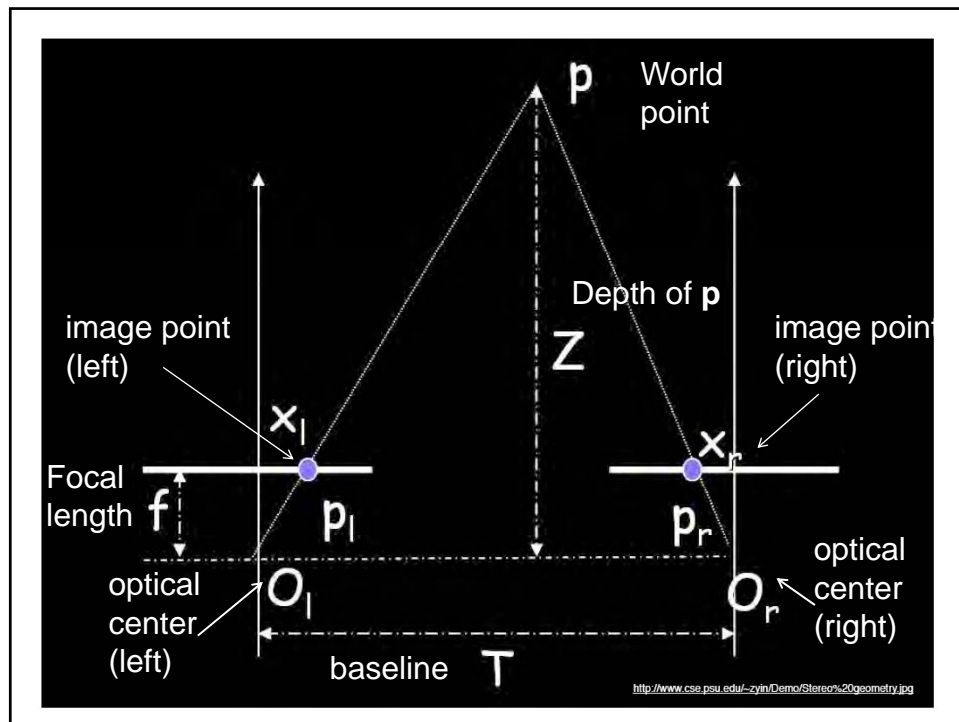
We'll assume for now that these parameters are given and fixed.

Outline

- Human stereopsis
- Stereograms
- Epipolar geometry and the epipolar constraint
 - Case example with parallel optical axes
 - General case with calibrated cameras

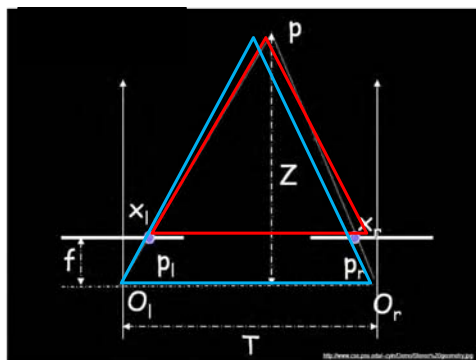
Geometry for a simple stereo system

- First, assuming parallel optical axes, known camera parameters (i.e., calibrated cameras):



Geometry for a simple stereo system

- Assume parallel optical axes, known camera parameters (i.e., calibrated cameras). We can triangulate via:



Similar triangles (p_l, P, p_r) and (O_l, P, O_r) :

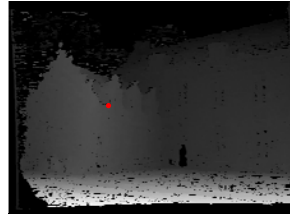
$$\frac{T + x_l - x_r}{Z - f} = \frac{T}{Z}$$

$$Z = f \frac{T}{x_r - x_l}$$

disparity \rightarrow

$$x_r - x_l$$

Depth from disparity

image $I(x,y)$ Disparity map $D(x,y)$ image $I'(x',y')$ 

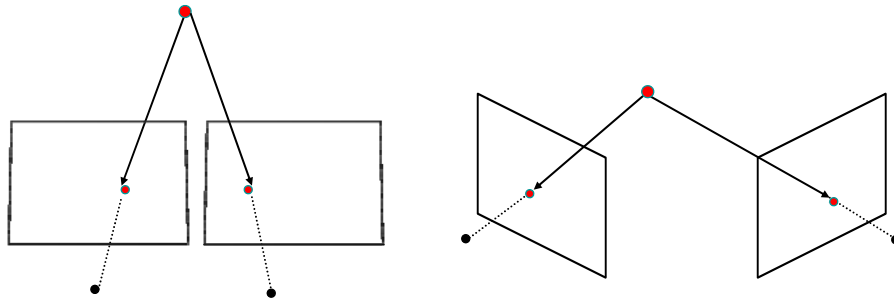
$$(x',y')=(x+D(x,y), y)$$

Outline

- Human stereopsis
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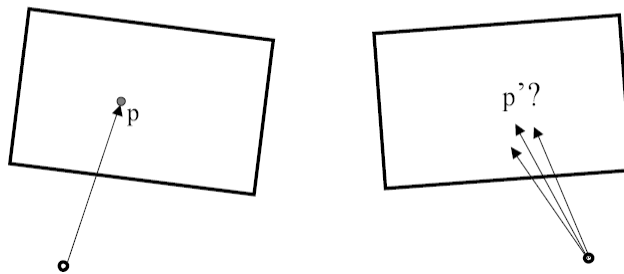
General case, with calibrated cameras

- The two cameras need not have parallel optical axes.



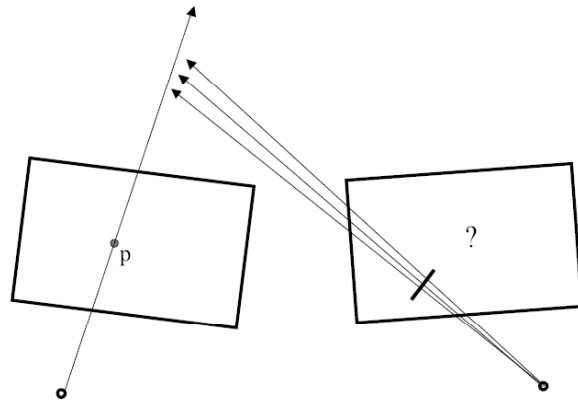
Vs.

Stereo correspondence constraints

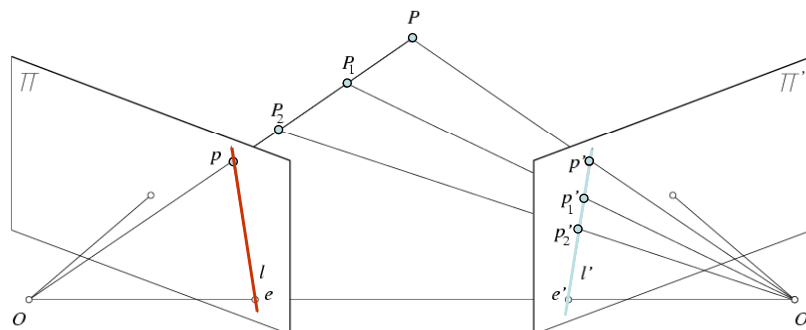


- Given p in left image, where can corresponding point p' be?

Stereo correspondence constraints



Epipolar constraint



Geometry of two views constrains where the corresponding pixel for some image point in the first view must occur in the second view:

- It must be on the line carved out by a plane connecting the world point and optical centers.

Why is this useful?

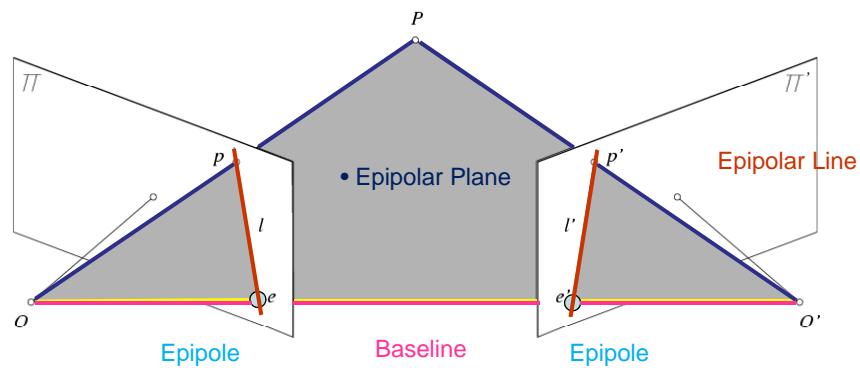
Epipolar constraint



This is useful because it reduces the correspondence problem to a 1D search along an epipolar line.

Image from Andrew Zisserman

Epipolar geometry

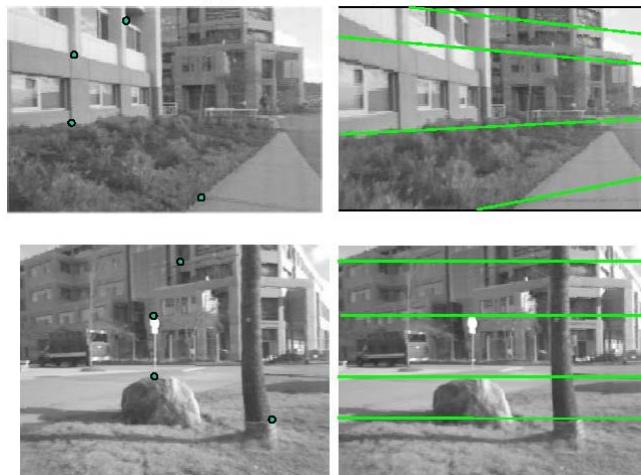


<http://www.ai.sri.com/~luong/research/Meta3DViewer/EpipolarGeo.html>

Epipolar geometry: terms

- **Baseline:** line joining the camera centers
 - **Epipole:** point of intersection of baseline with the image plane
 - **Epipolar plane:** plane containing baseline and world point
 - **Epipolar line:** intersection of epipolar plane with the image plane
-
- All epipolar lines intersect at the epipole
 - An epipolar plane intersects the left and right image planes in epipolar lines

Example



Example: converging cameras

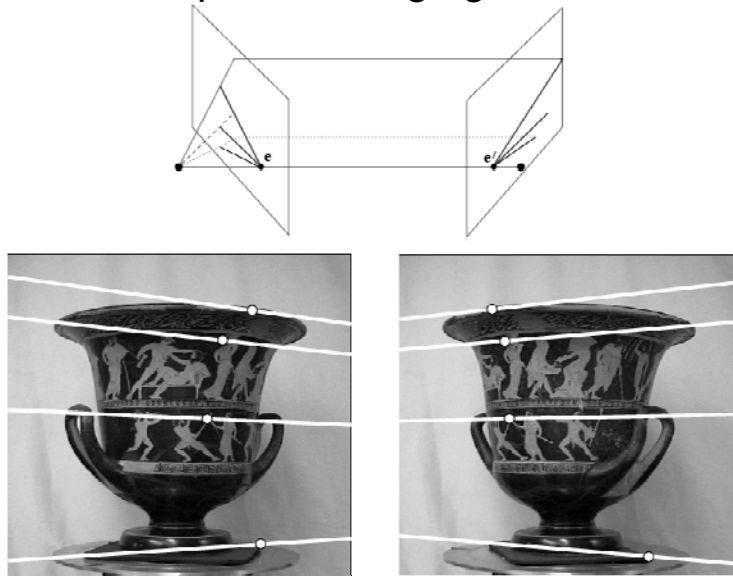


Figure from Hartley & Zisserman

Example: parallel cameras

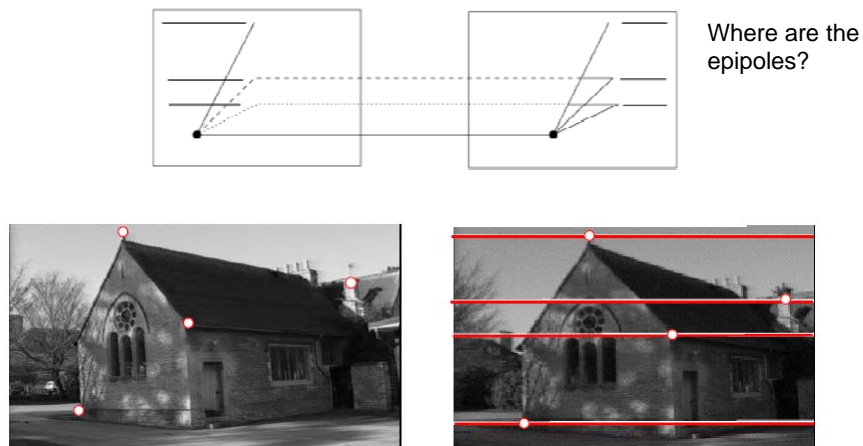
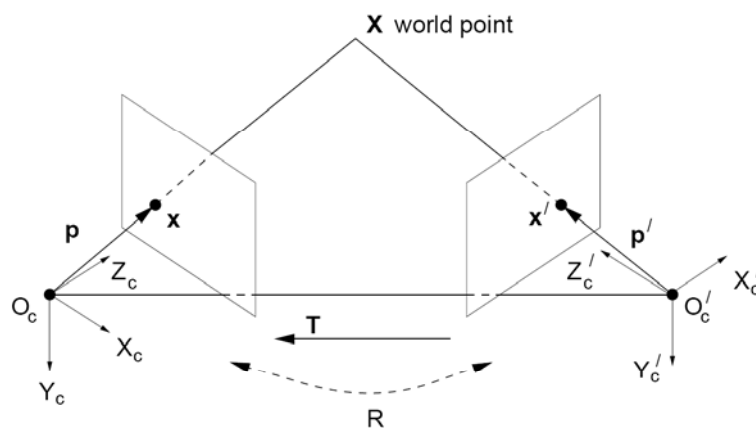


Figure from Hartley & Zisserman

- So far, we have the explanation in terms of geometry.
- Now, how to express the epipolar constraints algebraically?

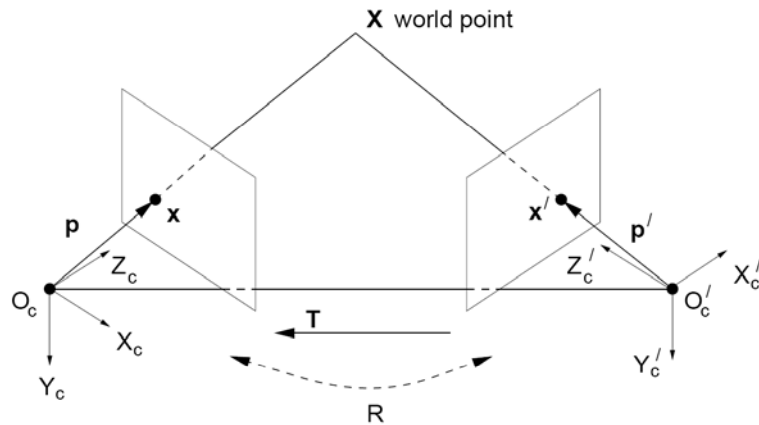
Stereo geometry, with calibrated cameras



If the stereo rig is calibrated, we know :
 how to **rotate** and **translate** camera reference frame 1 to
 get to camera reference frame 2.

Rotation: 3 x 3 matrix R ; translation: 3 vector T .

Stereo geometry, with calibrated cameras



If the stereo rig is calibrated, we know :
 how to **rotate** and **translate** camera reference frame 1 to
 get to camera reference frame 2. $\mathbf{X}'_c = \mathbf{R}\mathbf{X}_c + \mathbf{T}$

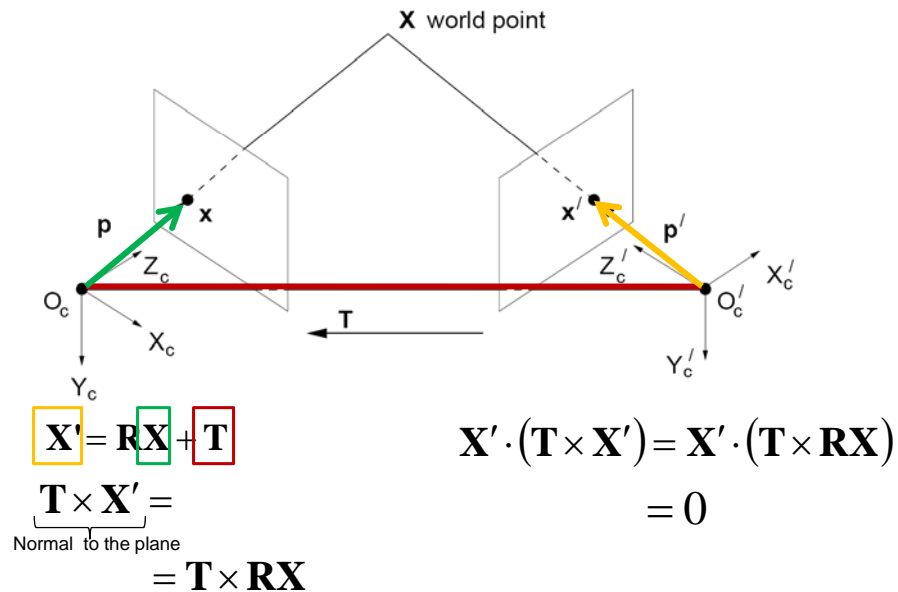
An aside: cross product

$$\vec{a} \times \vec{b} = \vec{c} \qquad \begin{aligned} \vec{a} \cdot \vec{c} &= 0 \\ \vec{b} \cdot \vec{c} &= 0 \end{aligned}$$

Vector cross product takes two vectors and returns a third vector that's perpendicular to both inputs.

So here, c is perpendicular to both a and b, which means the dot product = 0.

From geometry to algebra



Another aside: Matrix form of cross product

$$\vec{a} \times \vec{b} = \begin{bmatrix} 0 & -a_3 & a_2 \\ a_3 & 0 & -a_1 \\ -a_2 & a_1 & 0 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \vec{c}$$

$$\vec{a} \cdot \vec{c} = 0$$

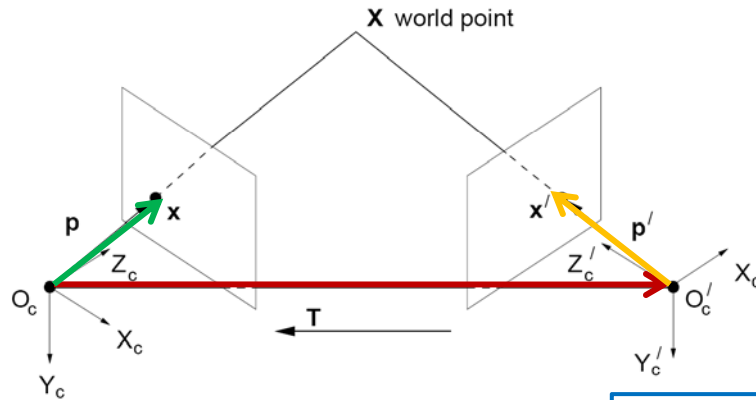
$$\vec{b} \cdot \vec{c} = 0$$

Can be expressed as a matrix multiplication.

$$[a_x] = \begin{bmatrix} 0 & -a_3 & a_2 \\ a_3 & 0 & -a_1 \\ -a_2 & a_1 & 0 \end{bmatrix}$$

$$\vec{a} \times \vec{b} = [a_x] \vec{b}$$

From geometry to algebra



$$\mathbf{X}' = \mathbf{R}\mathbf{X} + \mathbf{T}$$

$$\mathbf{X}' \cdot (\mathbf{T} \times \mathbf{X}') = \mathbf{X}' \cdot (\mathbf{T} \times \mathbf{R}\mathbf{X}) = 0$$

$$\underbrace{\mathbf{T} \times \mathbf{X}'}_{\text{Normal to the plane}} = \mathbf{T} \times \mathbf{R}\mathbf{X} + \mathbf{T} \times \mathbf{T} = \mathbf{T} \times \mathbf{R}\mathbf{X}$$

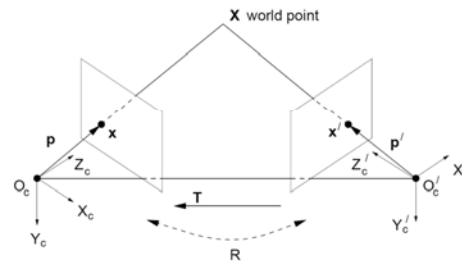
Essential matrix

$$\mathbf{X}' \cdot (\mathbf{T} \times \mathbf{R}\mathbf{X}) = 0$$

$$\mathbf{X}' \cdot (\mathbf{T}_x \mathbf{R}\mathbf{X}) = 0$$

Let $\mathbf{E} = \mathbf{T}_x \mathbf{R}$

$$\mathbf{X}'^T \mathbf{E} \mathbf{X} = 0$$

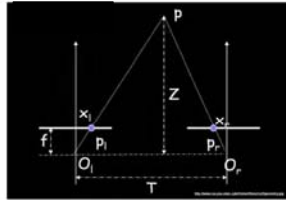


\mathbf{E} is called the **essential matrix**, and it relates corresponding image points between both cameras, given the rotation and translation.

If we observe a point in one image, its position in other image is constrained to lie on line defined by above.

Note: these points are in camera coordinate systems.

Essential matrix example: parallel cameras



$$\mathbf{R} = \mathbf{I}$$

$$\mathbf{T} = [-d, 0, 0]^T$$

$$\mathbf{E} = [\mathbf{T}_x] \mathbf{R} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & d \\ 0 & -d & 0 \end{pmatrix}$$

$$\mathbf{p} = [x, y, f]$$

$$\mathbf{p}' = [x', y', f']$$

$$\mathbf{p}'^T \mathbf{E} \mathbf{p} = 0$$

$$\begin{bmatrix} x' & y' & f' \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & d \\ 0 & -d & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ f \end{bmatrix} = 0$$

For the parallel cameras, image of any point must lie on same horizontal line in each image plane.

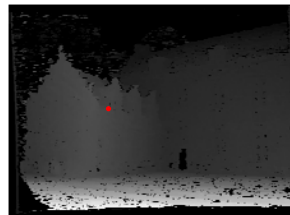
$$\Leftrightarrow \begin{bmatrix} x' & y' & f' \end{bmatrix} \begin{bmatrix} 0 \\ df \\ -dy \end{bmatrix} = 0$$

$$\Leftrightarrow y = y'$$

image $I(x, y)$

Disparity map $D(x, y)$

image $I'(x', y')$

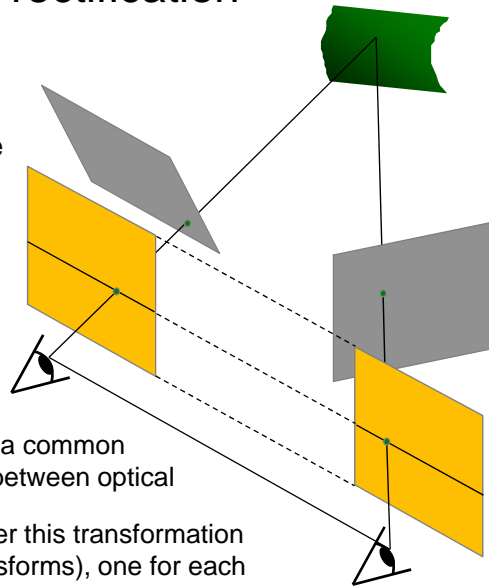


$$(x', y') = (x + D(x, y), y)$$

What about when cameras' optical axes are not parallel?

Stereo image rectification

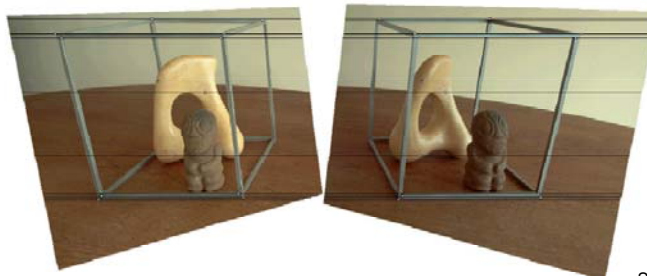
In practice, it is convenient if image scanlines (rows) are the epipolar lines.



reproject image planes onto a common plane parallel to the line between optical centers
 pixel motion is horizontal after this transformation
 two homographies (3x3 transforms), one for each input image reprojection

Adapted from Li Zhang

Stereo image rectification: example



Source: Alyosha Efros

Summary

- Depth from stereo: main idea is to triangulate from corresponding image points.
- Epipolar geometry defined by two cameras
 - We've assumed known extrinsic parameters relating their poses
- Epipolar constraint limits where points from one view will be imaged in the other
 - Makes search for correspondences quicker
- Terms: epipole, epipolar plane / lines, disparity, rectification, intrinsic/extrinsic parameters, essential matrix, baseline

Coming up

- Thursday:
 - Computing correspondences
 - Non-geometric stereo constraints
 - Weak calibration