

# Homographies and Panoramas

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*with a lot of slides stolen from  
Steve Seitz and Rick Szeliski*

© Andrew Campbell

CS180: Intro to Computer Vision and Comp. Photo  
Angjoo Kanazawa and Alexei Efros, UC Berkeley, Fall 2023

# Logistics

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Project 3 due today!

Project 4 released this Wednesday

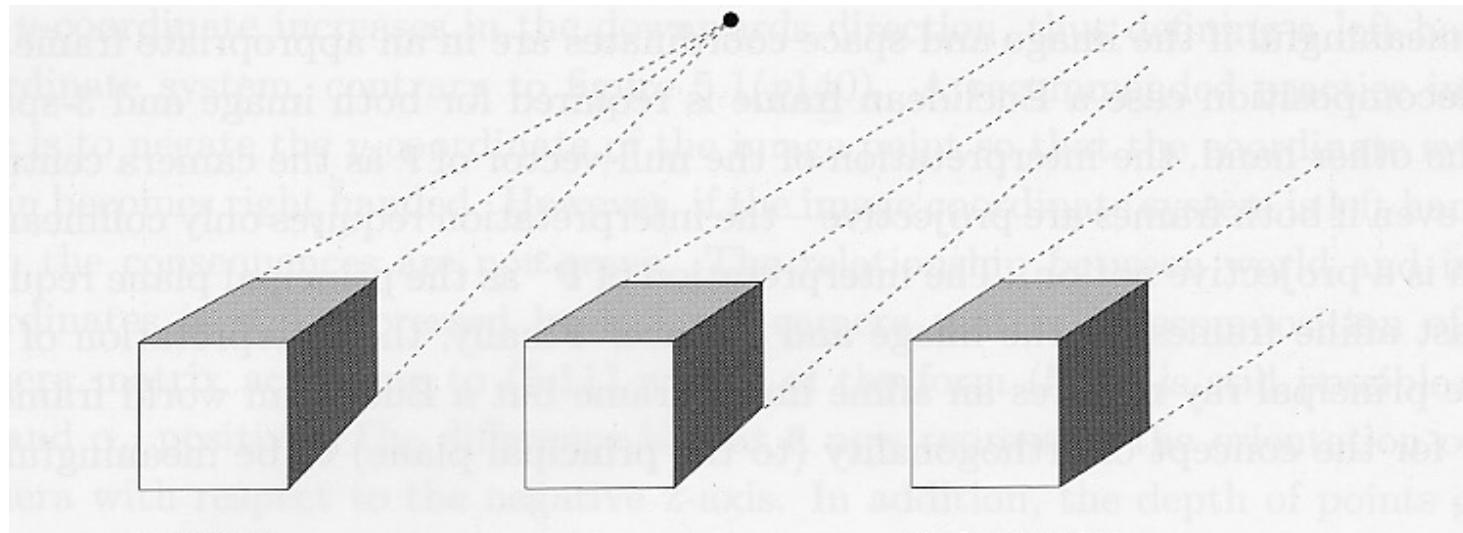
All is due 10/25 (with some midpoint checks)!! It is quite challenging!  
Make sure to start early!

Try to preserve slip days for emergencies.

Project 2 voting is out on Ed!!

# Recap

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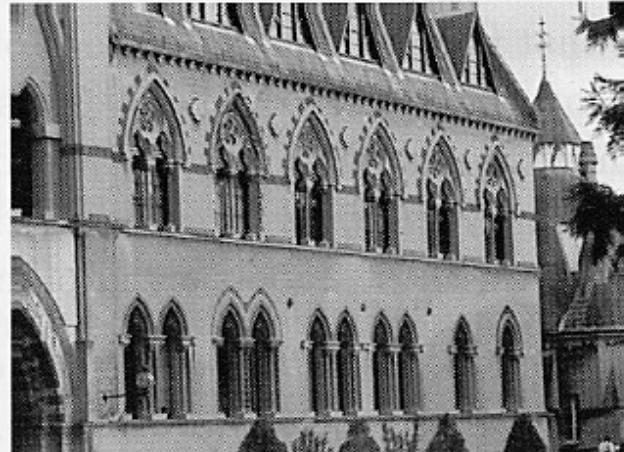
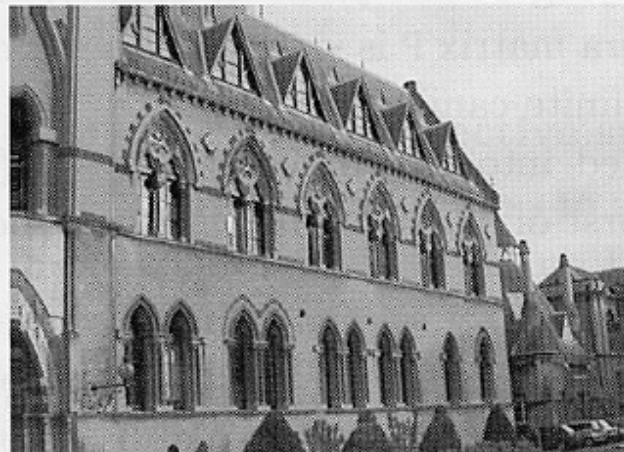


**perspective**

**weak perspective**

increasing focal length →

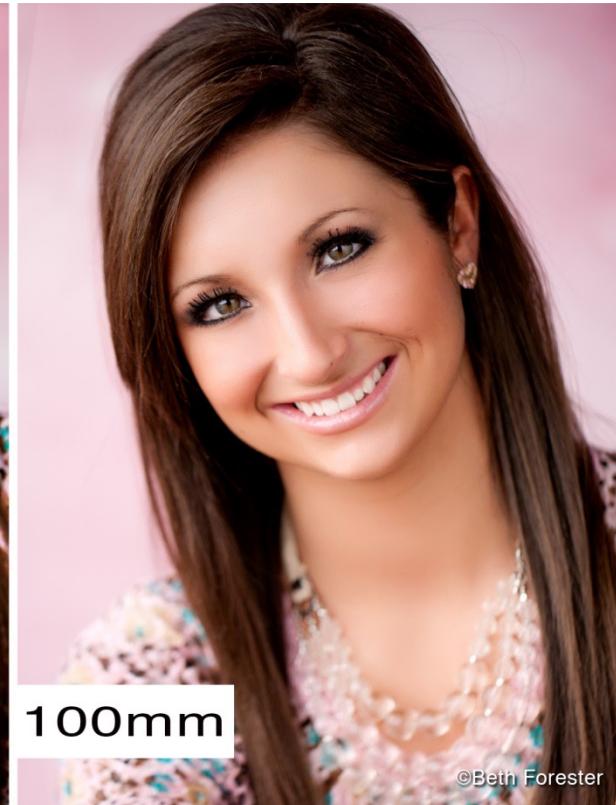
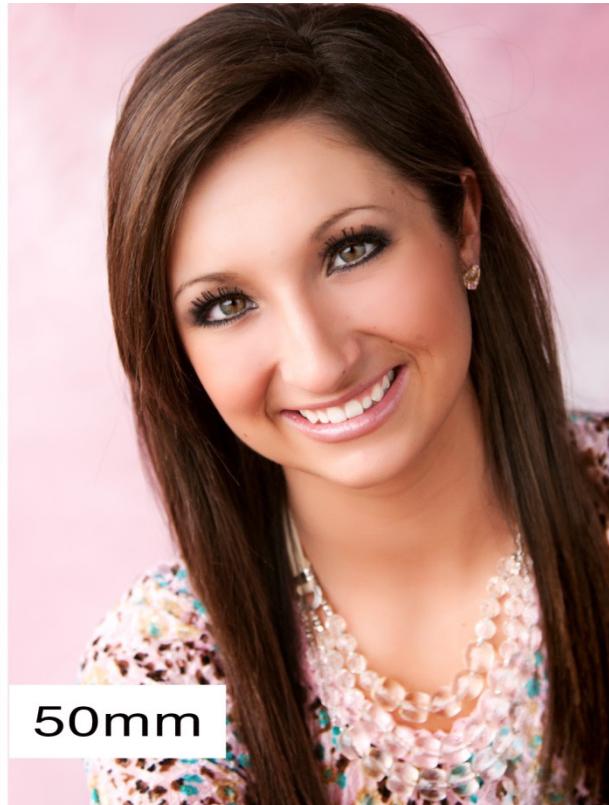
increasing distance from camera →



From Zisserman & Hartley

# Focal length / distance in portraiture

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©Beth Forester

# Perspective Distortion

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Not due to lens flaws

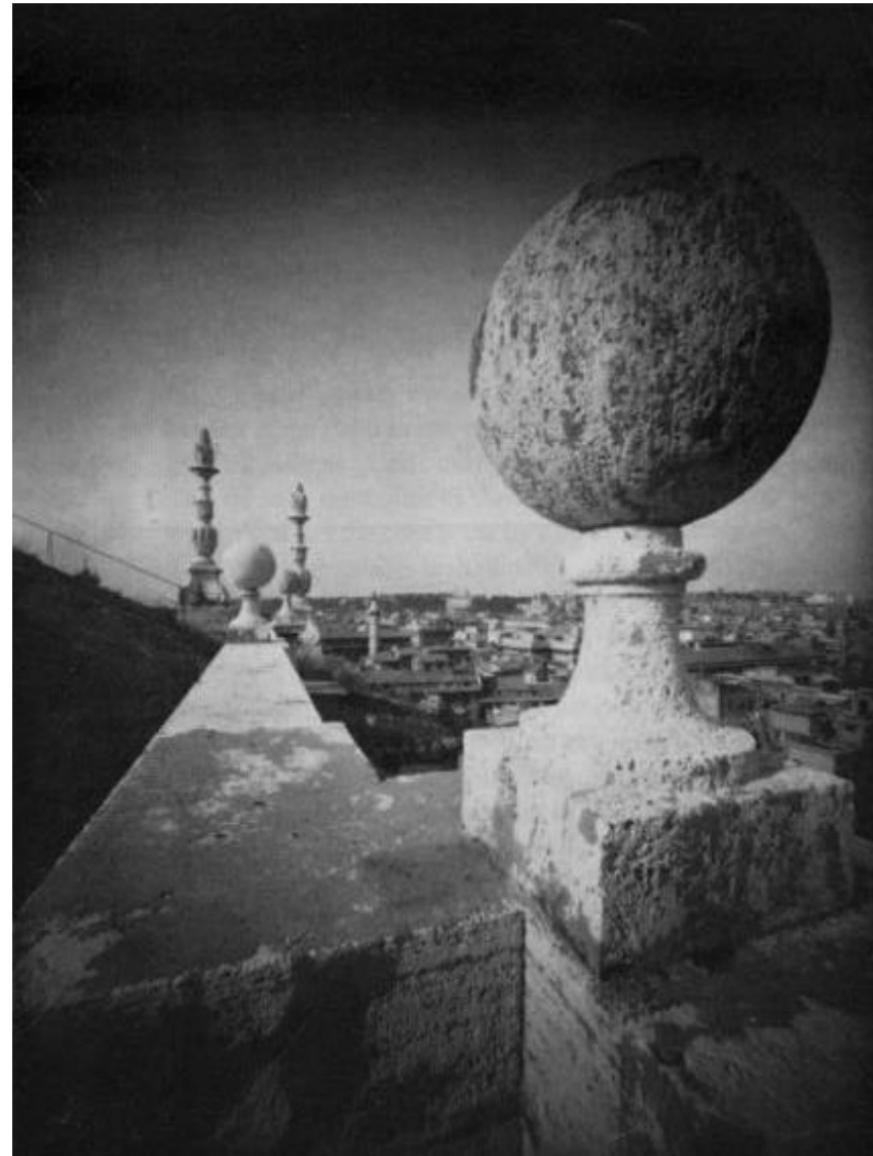
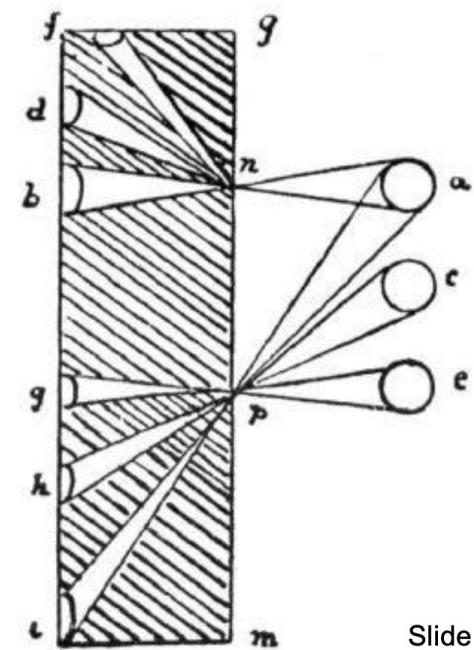
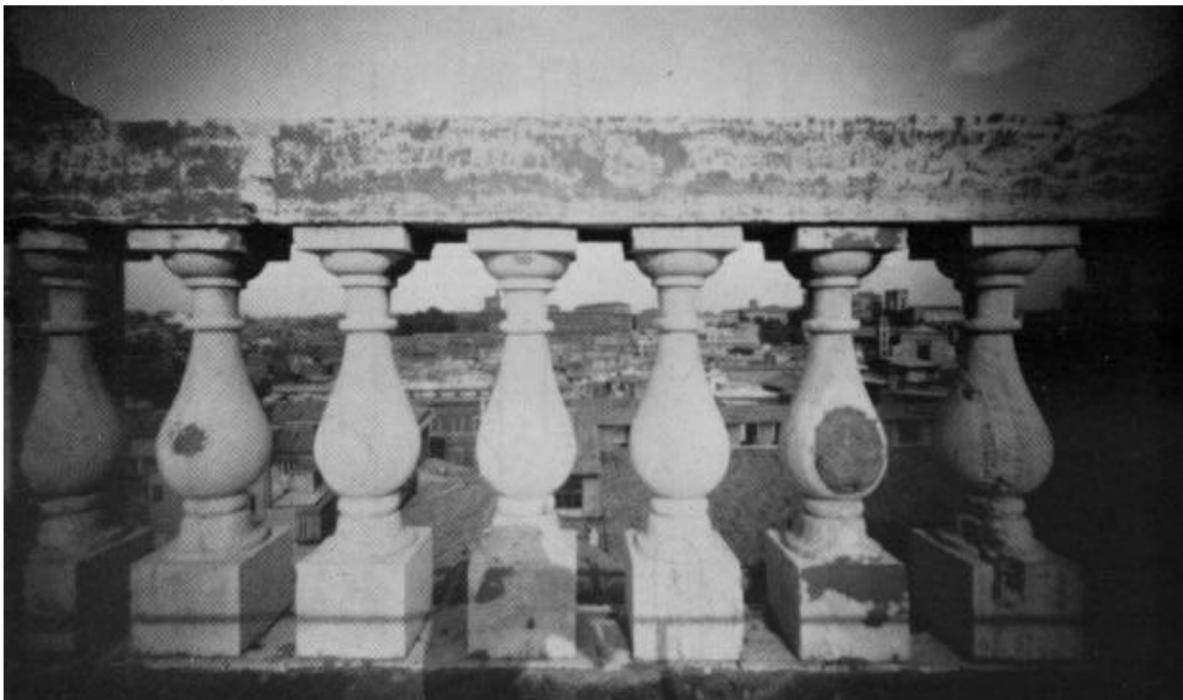


Image source: F. Durand

# Problem pointed out by Da Vinci

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The exterior columns appear bigger



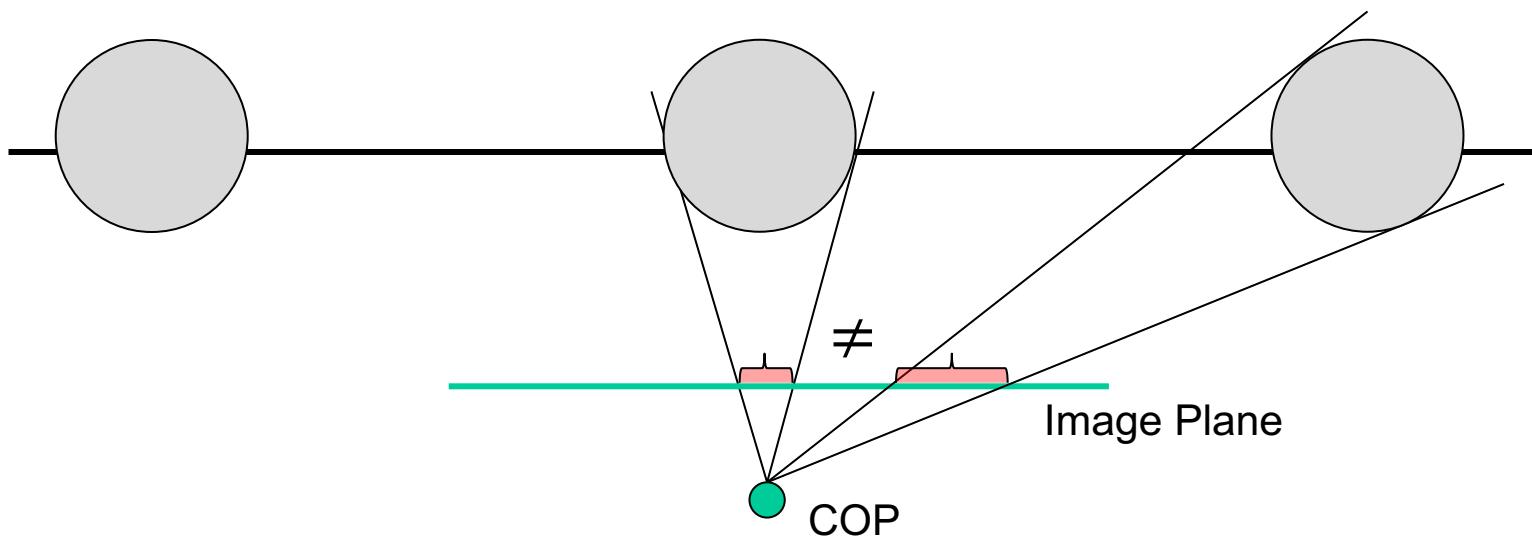
Slide by F. Durand

Image Plane

# Perspective Distortion

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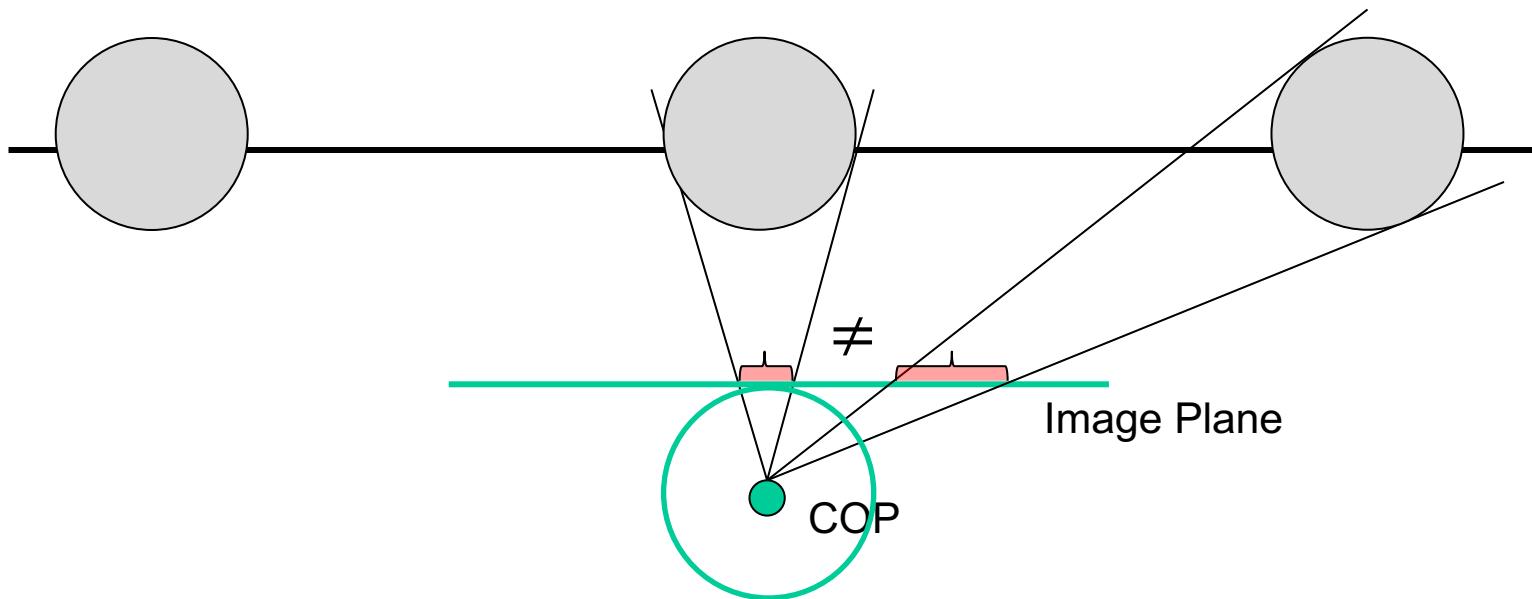
Recall Perspective Projection:  $x' = f \frac{x}{z}$      $y' = f \frac{y}{z}$



# Perspective Distortion

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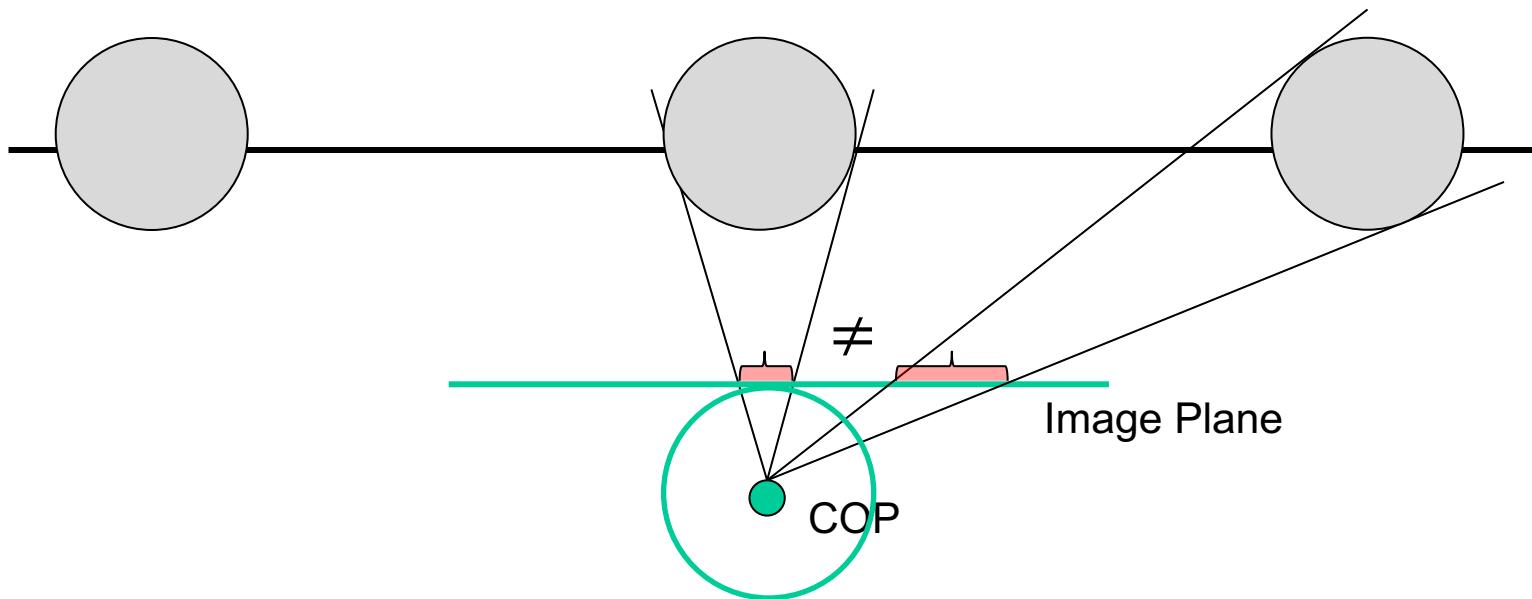
With a spherical projection plane



# Perspective Distortion

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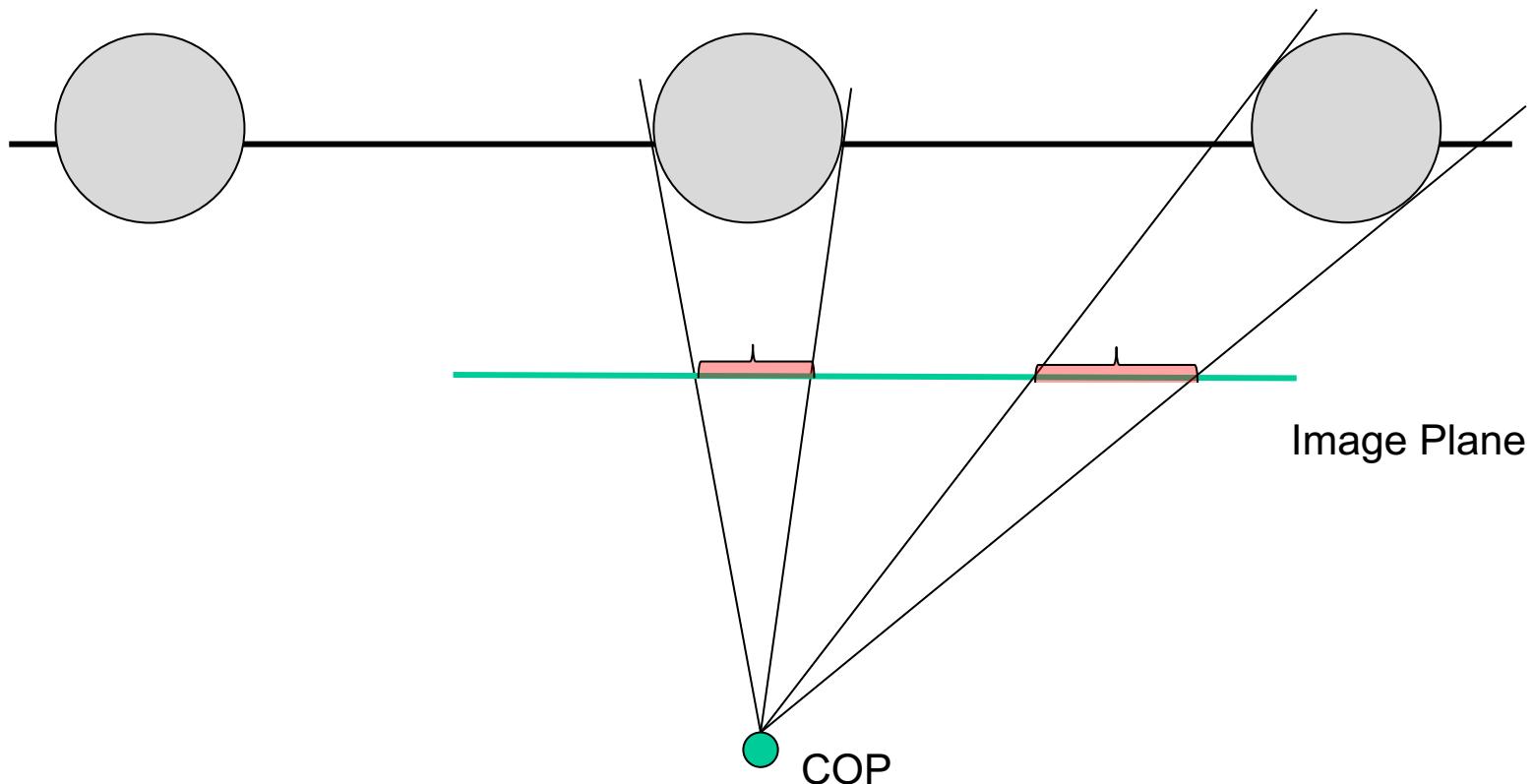
With a spherical projection plane



# Perspective Distortion

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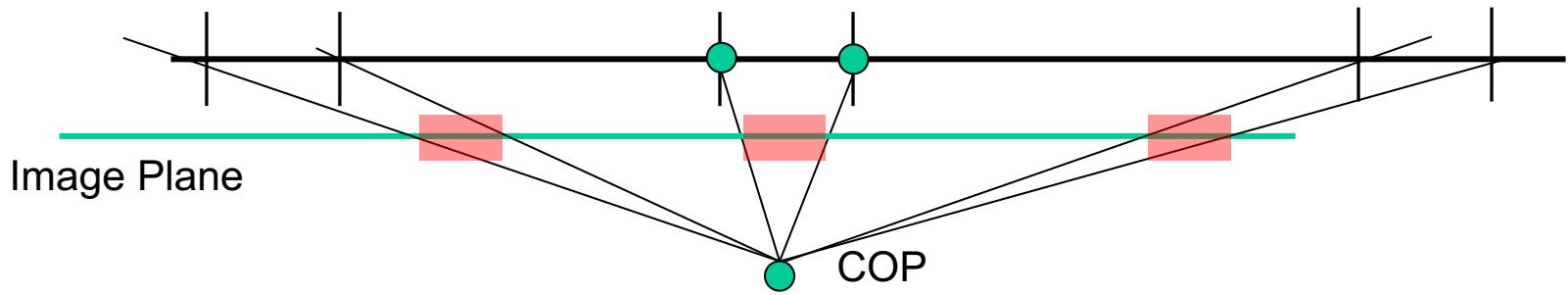
Less noticeable with long focal length (i.e. you see distortion more with wide-angle camera)



# Perspective Distortion

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It's about the change in depths

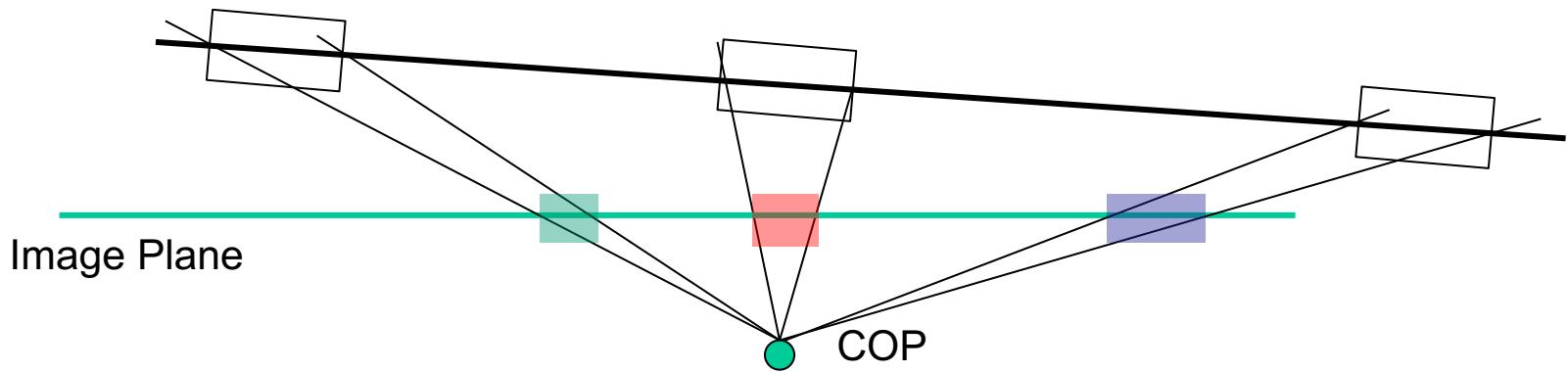


But this is a very special case..

# Perspective Distortion

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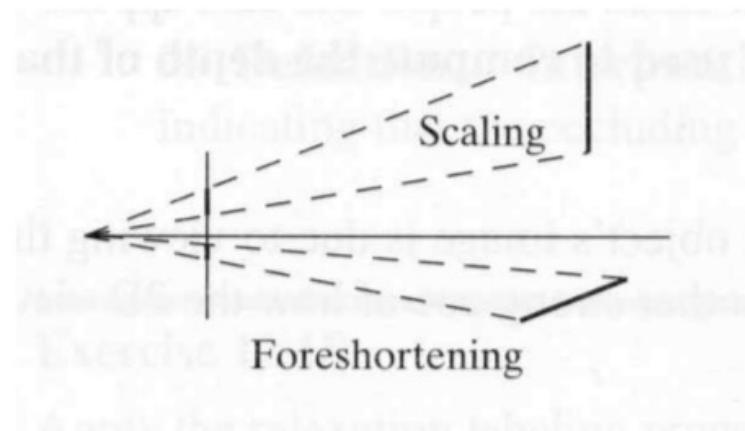
More likely..



# Foreshortening

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- When a line (or surface) is parallel to the image plane, the effect of perspective projection is *scaling*.
- When a line (or surface) is not parallel to the image plane, we use the term *foreshortening* to describe the projective distortion (i.e., the dimension parallel to the optical axis is compressed relative to the frontal dimension).



# Fixing Perspective Distortion

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(a) A wide-angle photo with distortions on subjects' faces.

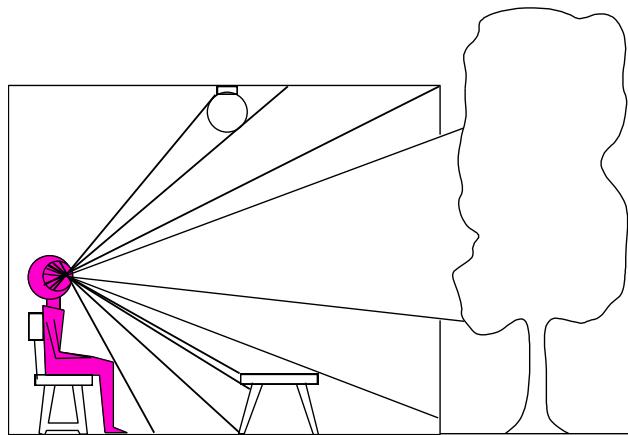


(b) Distortion-free photo by our method.

# What do we see?

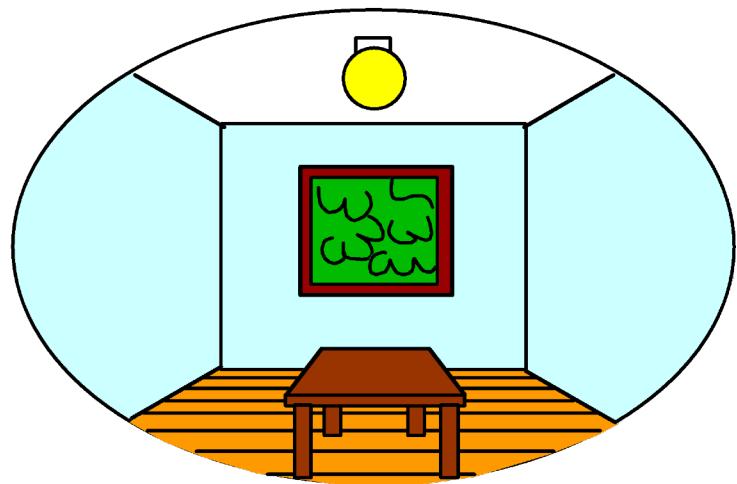
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*3D world*



Point of observation

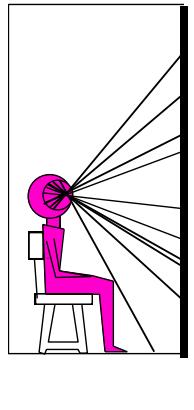
*2D image*



# What do we see?

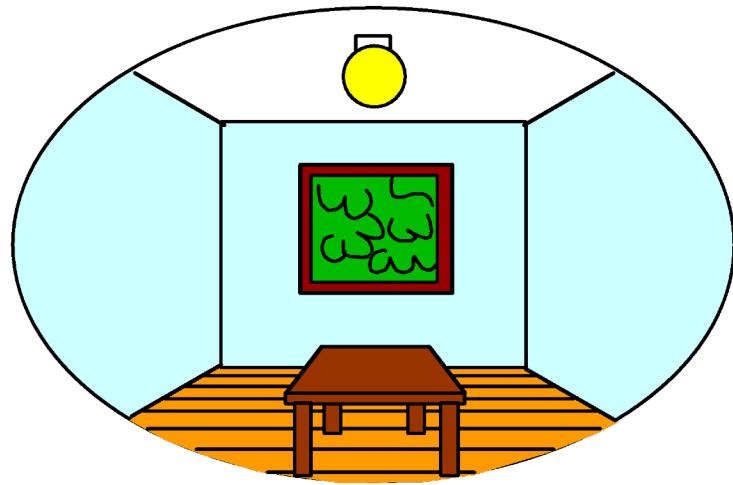
---

*3D world*



Painted  
backdrop

*2D image*



# On Simulating the Visual Experience

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Just feed the eyes the right data

- No one will know the difference!

Philosophy:

- Ancient question: “Does the world really exist?”

Science fiction:

- Many, many, many books on the subject, e.g. *Slowglass* from “[Light of Other Days](#)”
- “Latest” take: *The Matrix*

Physics:

- *Slowglass* might be possible?

Computer Science:

- Virtual Reality

To simulate we need to know:

What does a person see?

# The Plenoptic Function

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Figure by Leonard McMillan

Q: What is the set of all things that we can ever see?

A: The Plenoptic Function (Adelson & Bergen)

Let's start with a stationary person and try to parameterize everything that he can see...

# Grayscale snapshot

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$$P(\theta, \phi)$$

is intensity of light

- Seen from a single view point
- At a single time
- Averaged over the wavelengths of the visible spectrum

(can also do  $P(x,y)$ , but spherical coordinate are nicer)

# Color snapshot

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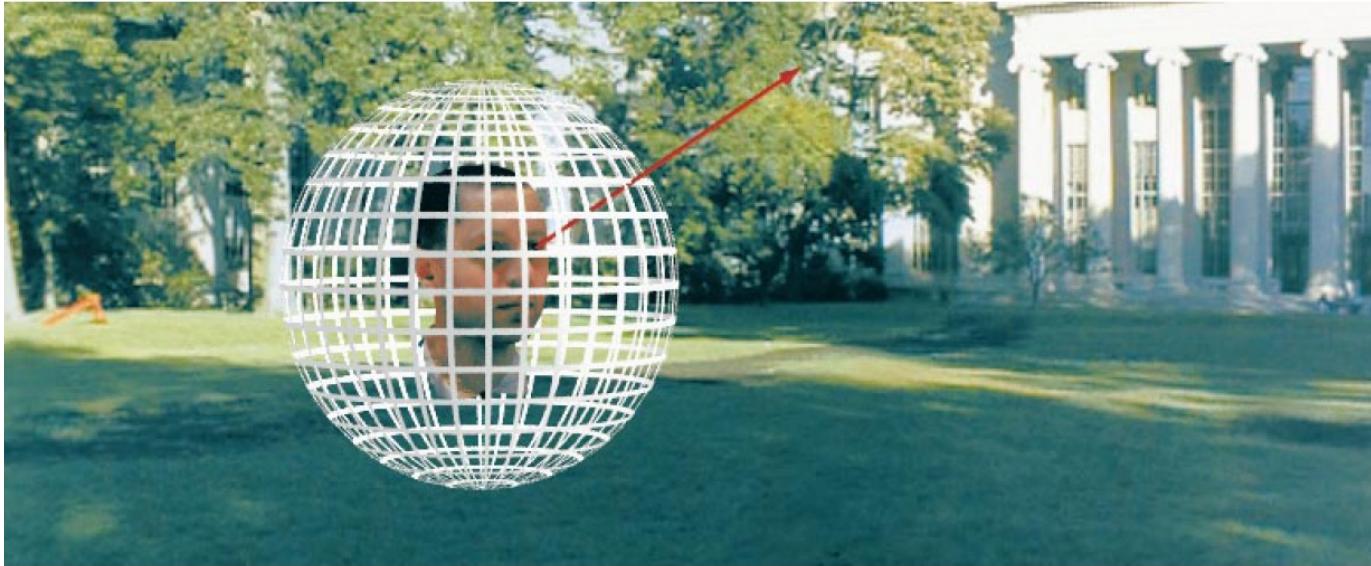
$$P(\theta, \phi, \lambda)$$

is intensity of light

- Seen from a single view point
- At a single time
- As a function of wavelength

# A movie

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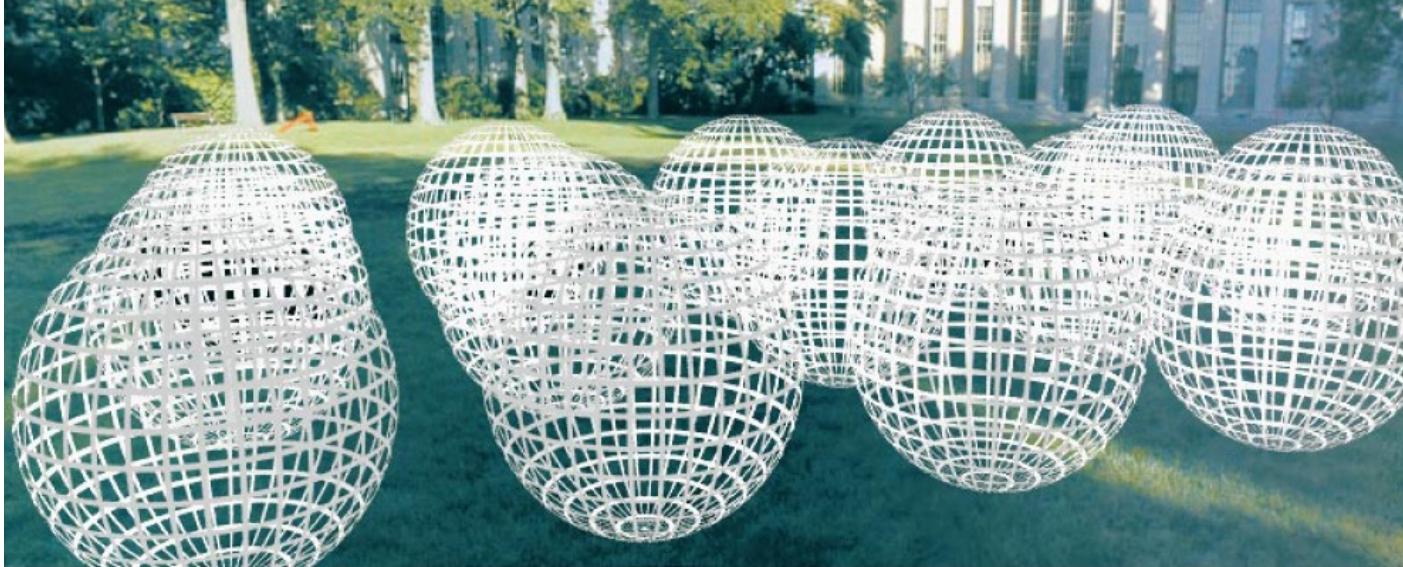
$$P(\theta, \phi, \lambda, t)$$

is intensity of light

- Seen from a single view point
- Over time
- As a function of wavelength

# Holographic movie

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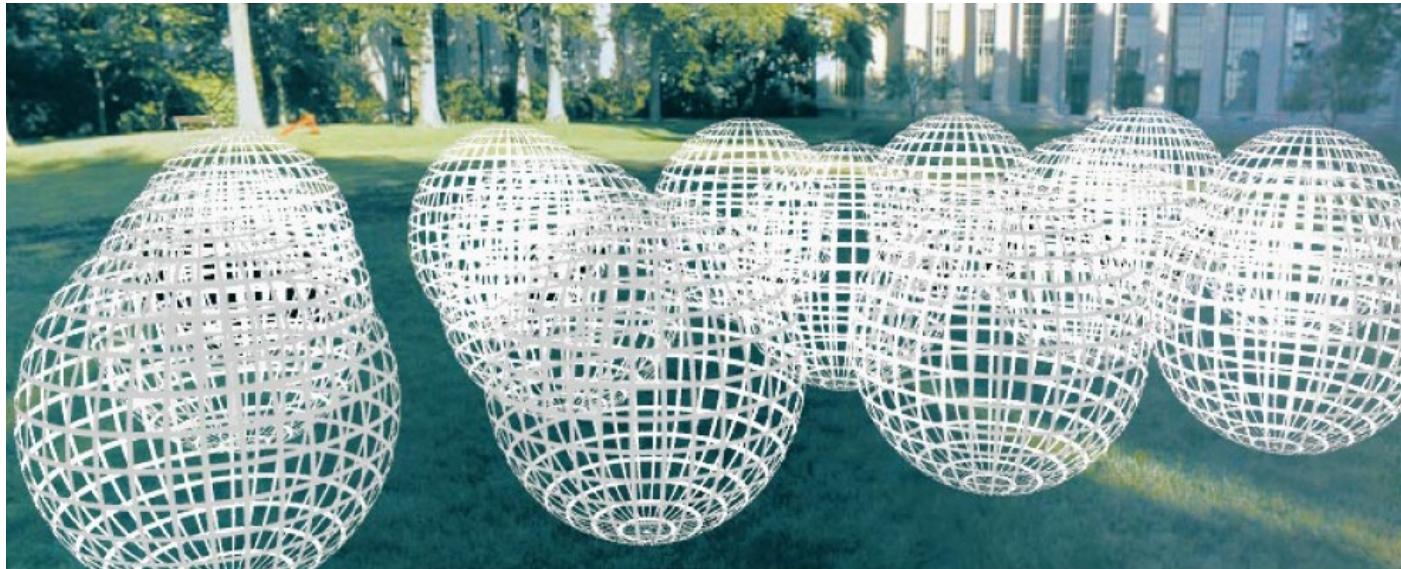
$$P(\theta, \phi, \lambda, t, V_x, V_y, V_z)$$

is intensity of light

- Seen from ANY viewpoint
- Over time
- As a function of wavelength

# The Plenoptic Function

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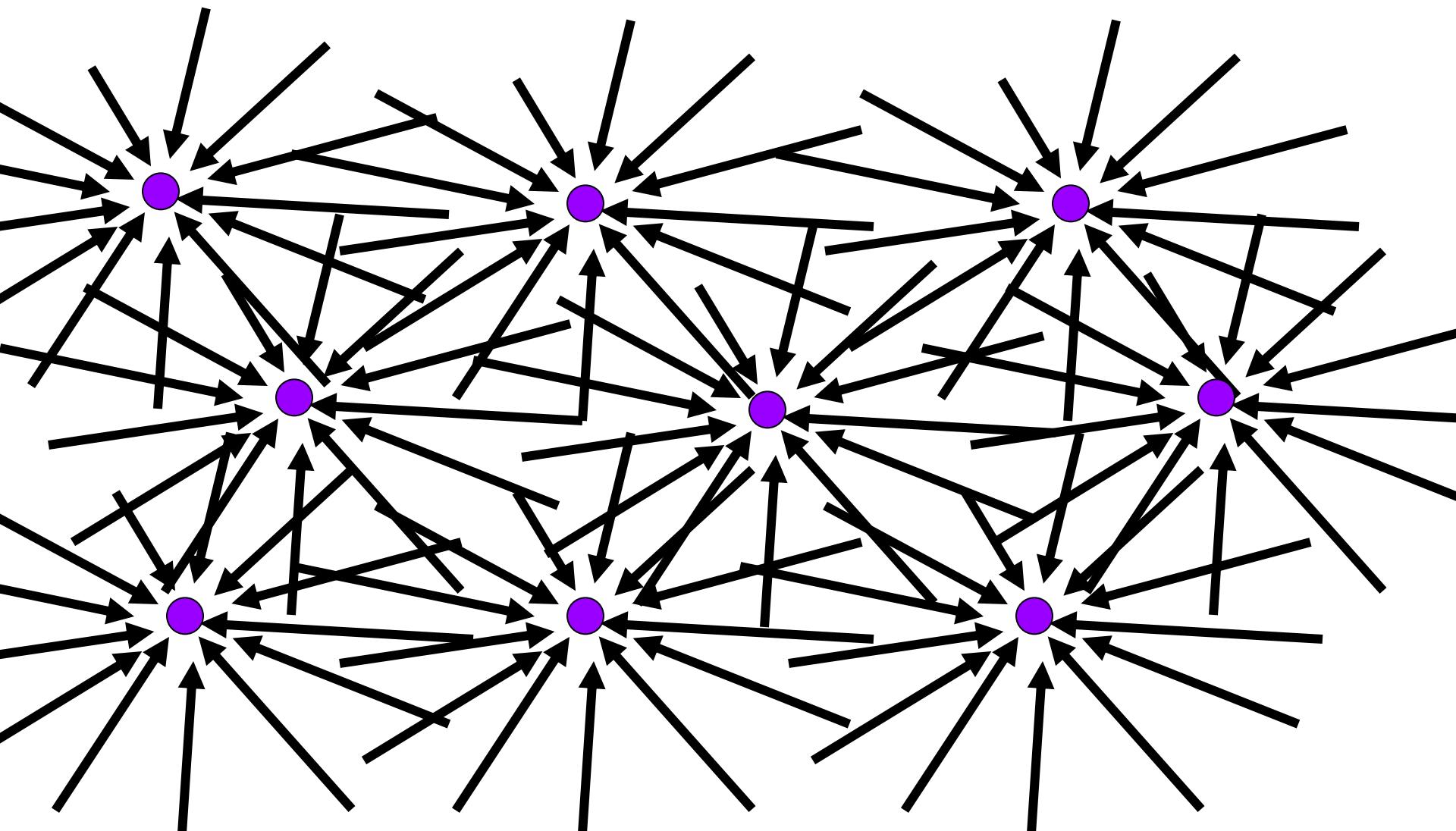


$$P(\theta, \phi, \lambda, t, V_x, V_y, V_z)$$

- Can reconstruct every possible view, at every moment, from every position, at every wavelength
- Contains every photograph, every movie, everything that anyone has ever seen! it completely captures our visual reality! Not bad for a function...

# Sampling Plenoptic Function (top view)

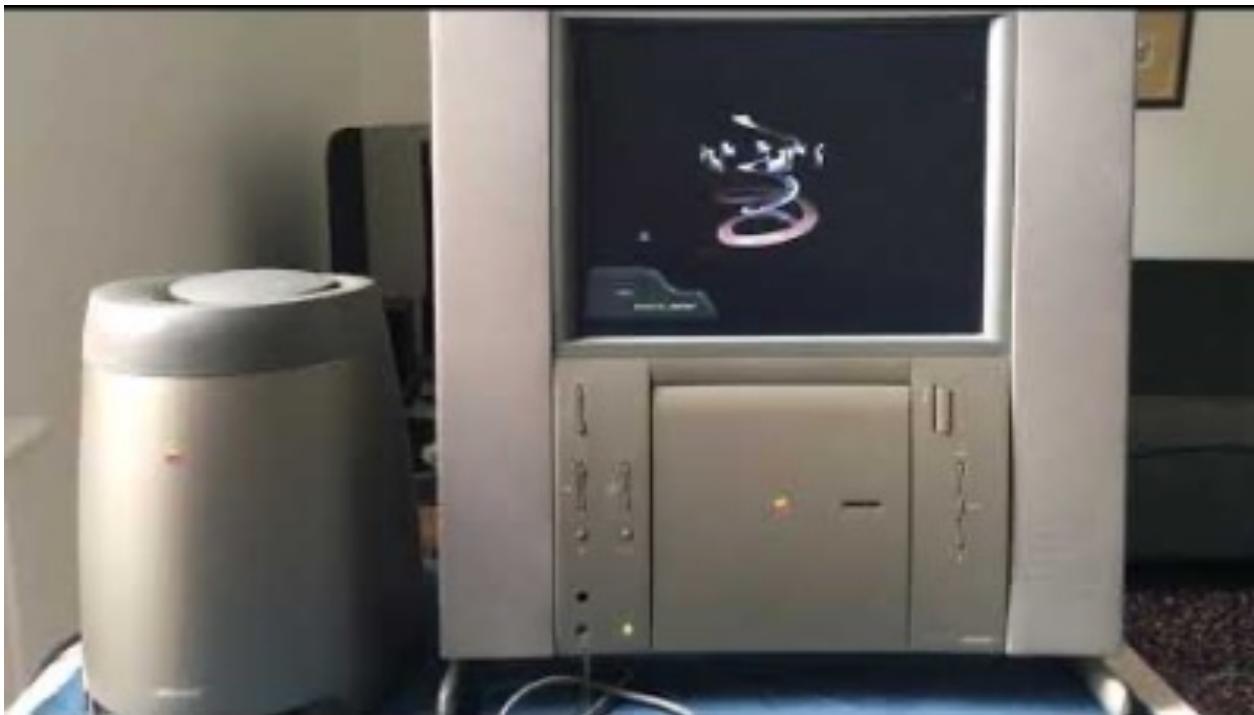
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Just lookup -- Quicktime VR

# QuickTime VR 1995

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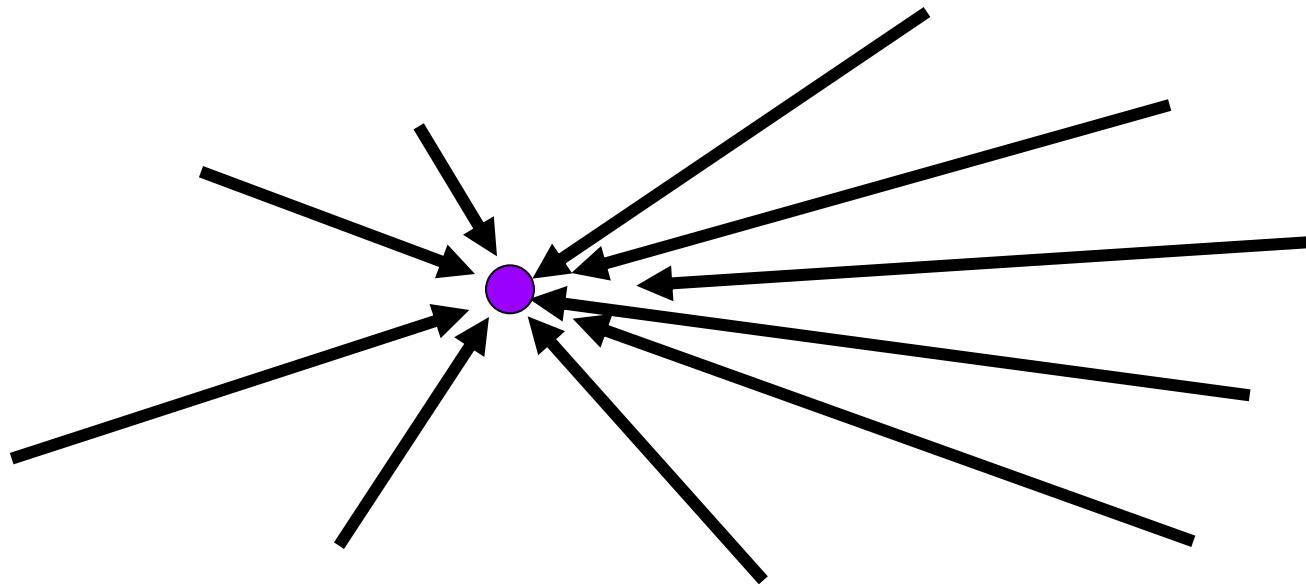


# Apple Quicktime VR

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# What is an image?

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# Spherical Panorama

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See also: 2003 New Years Eve  
<http://www.panoramas.dk/New-Year/times-square.html>

All light rays through a point form a ponorama

Totally captured in a 2D array --  $P(\theta, \phi)$

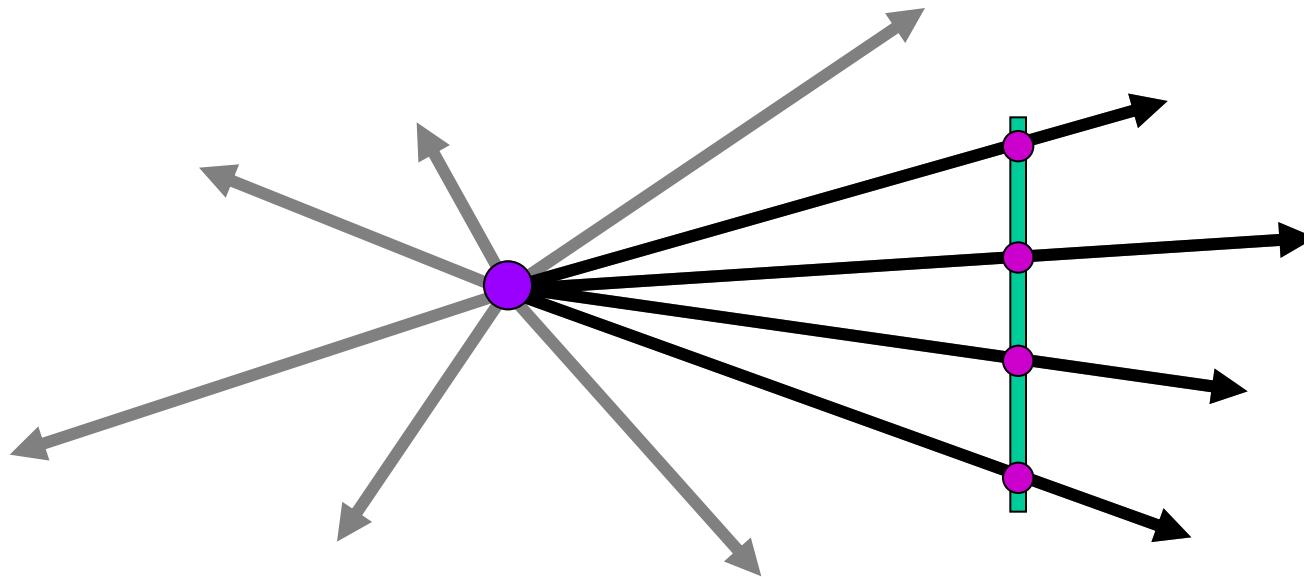
Where is the geometry???

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[https://www.360cities.net/curated\\_sets/90-new-year's-eve-celebrations](https://www.360cities.net/curated_sets/90-new-year's-eve-celebrations)

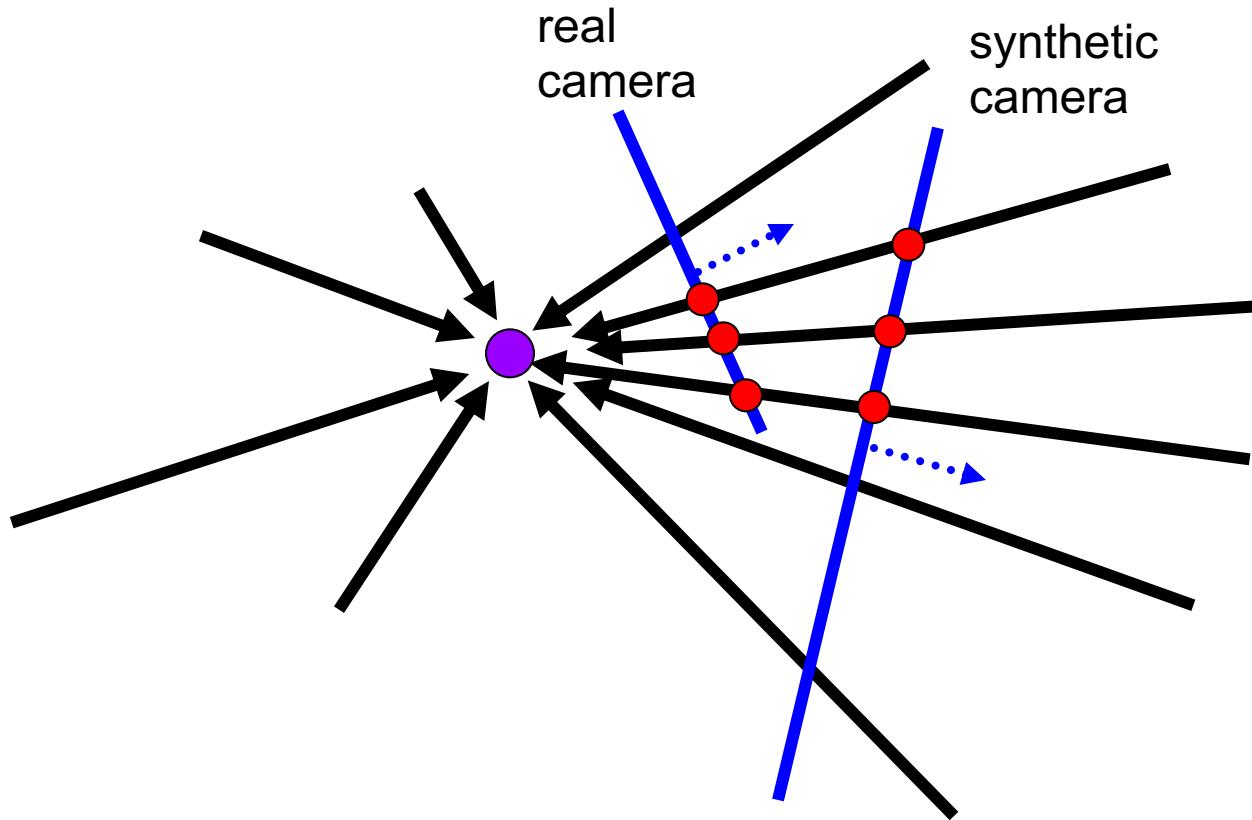
# What is an Image?

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# A pencil of rays contains all views

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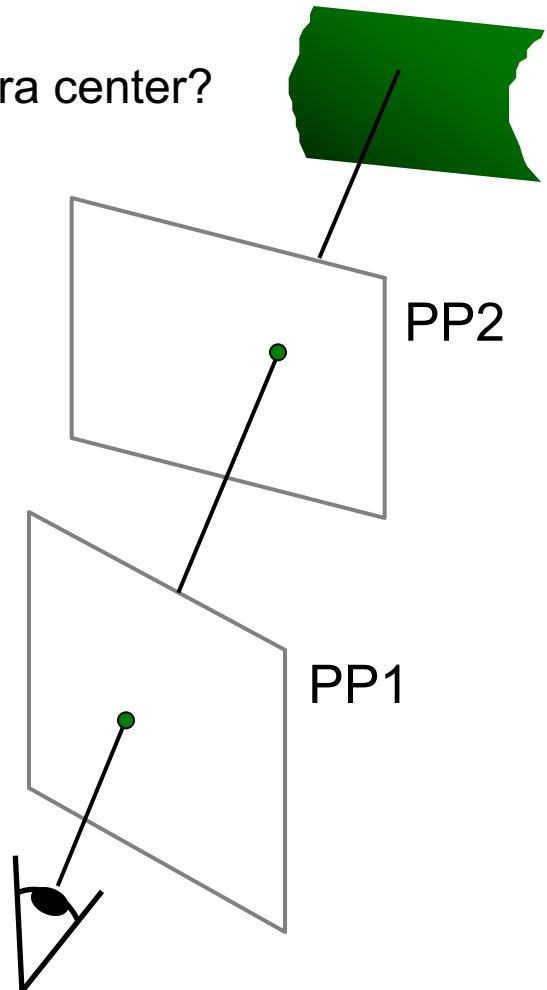
Can generate any synthetic camera view  
as long as it has **the same center of projection!**

# Image reprojection

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## Basic question

- How to relate two images from the same camera center?
  - how to map a pixel from PP1 to PP2



## Answer

- Cast a ray through each pixel in PP1
- Draw the pixel where that ray intersects PP2

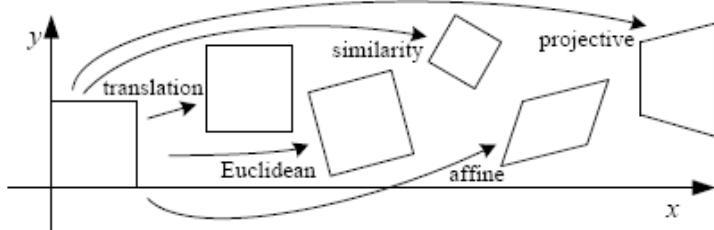
But don't we need to know the geometry of the two planes in respect to the eye?

Observation:

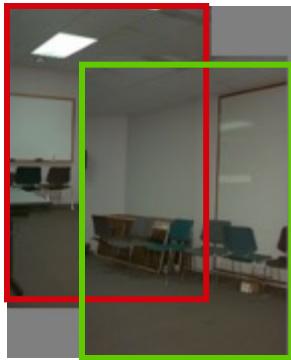
Rather than thinking of this as a 3D reprojection,  
think of it as a 2D **image warp** from one image to another

# Back to Image Warping

Which t-form is the right one for warping PP1 into PP2?  
e.g. translation, Euclidean, affine, projective

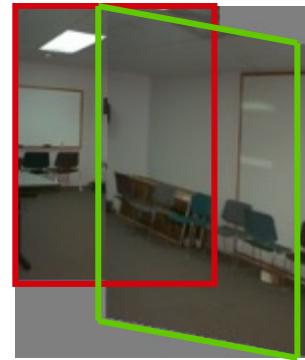


Translation



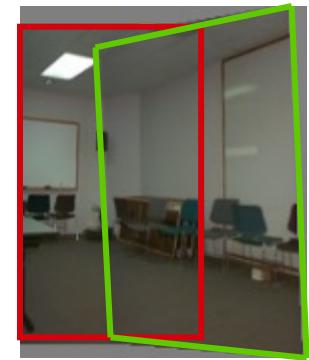
2 unknowns

Affine



6 unknowns

Perspective



8 unknowns

# Homography

A: Projective – mapping between any two PPs with the same center of projection

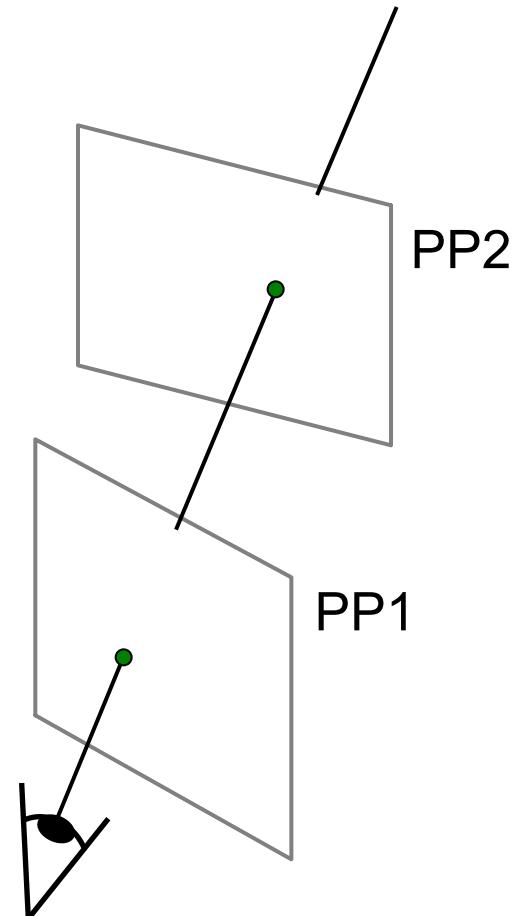
- rectangle should map to arbitrary quadrilateral
- parallel lines aren't
- but must preserve straight lines
- same as: unproject, rotate, reproject

called Homography

$$\begin{bmatrix} wx' \\ wy' \\ w \end{bmatrix} = \begin{bmatrix} * & * & * \\ * & * & * \\ * & * & * \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$
$$\mathbf{p}' \quad \mathbf{H} \quad \mathbf{p}$$

To apply a homography  $\mathbf{H}$

- Compute  $\mathbf{p}' = \mathbf{H}\mathbf{p}$  (regular matrix multiply)
- Convert  $\mathbf{p}'$  from homogeneous to image coordinates



# Image warping with homographies

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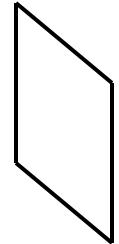
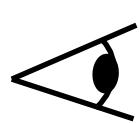
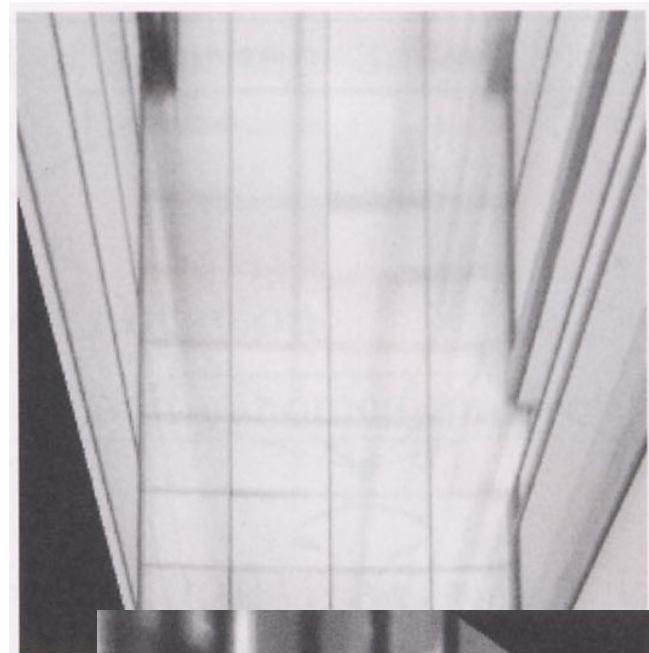
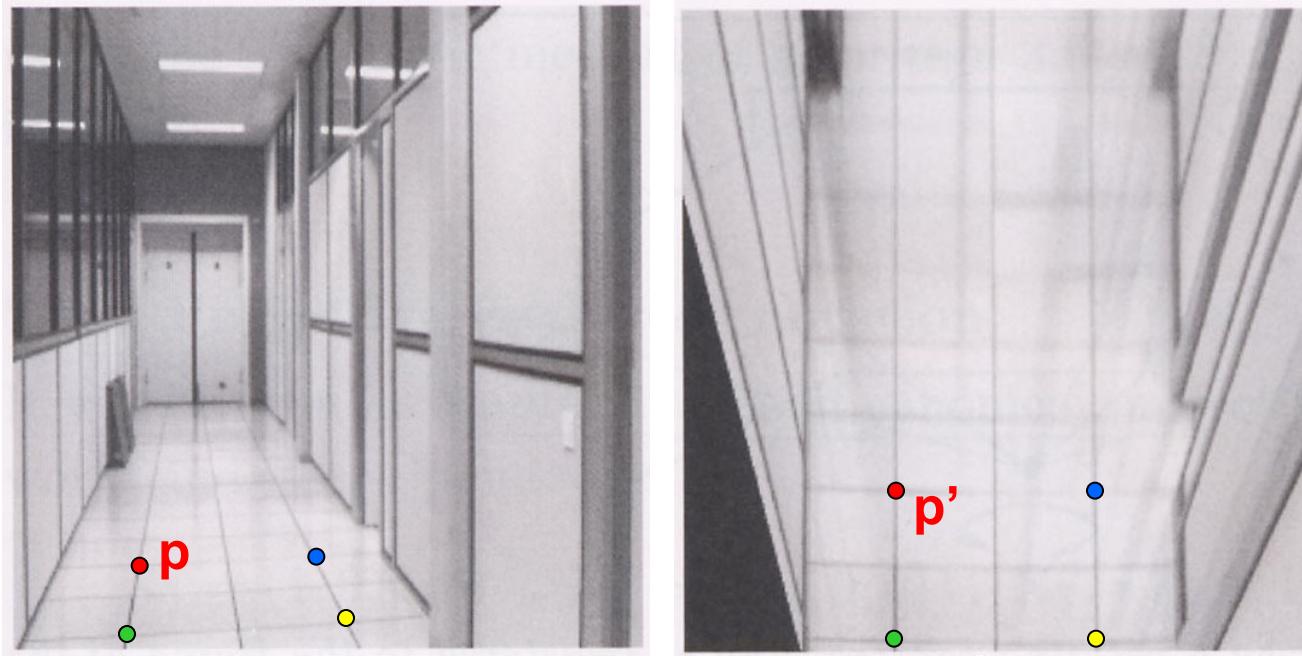


image plane in front

black area  
where no pixel  
maps to

# Image rectification

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To un warp (rectify) an image

- Find the homography  $\mathbf{H}$  given a set of  $\mathbf{p}$  and  $\mathbf{p}'$  pairs
- How many correspondences are needed?
- Tricky to write  $\mathbf{H}$  analytically, but we can solve for it!
  - Find such  $\mathbf{H}$  that “best” transforms points  $\mathbf{p}$  into  $\mathbf{p}'$
  - Use least-squares!

# Least Squares Example

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Say we have a set of data points  $(p_1, p_1')$ ,  $(p_2, p_2')$ ,  
 $(p_3, p_3')$ , etc. (e.g. person's height vs. weight)

We want a nice compact formula (a line) to predict  $p'$   
from  $p$ :  $px_1 + x_2 = p'$

We want to find  $x_1$  and  $x_2$

How many  $(p, p')$  pairs do we need?

$$p_1x_1 + x_2 = p_1'$$

$$p_2x_1 + x_2 = p_2'$$

$$\begin{bmatrix} p_1 & 1 \\ p_2 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} p_1' \\ p_2' \end{bmatrix}$$

$$Ax = b$$

# Least Squares Example

---

Say we have a set of data points  $(p_1, p_1')$ ,  $(p_2, p_2')$ ,  
 $(p_3, p_3')$ , etc. (e.g. person's height vs. weight)

We want a nice compact formula (a line) to predict  $p'$   
from  $p$ :  $px_1 + x_2 = p'$

We want to find  $x_1$  and  $x_2$

How many  $(p, p')$  pairs do we need?

$$p_1x_1 + x_2 = p_1'$$

$$p_2x_1 + x_2 = p_2'$$

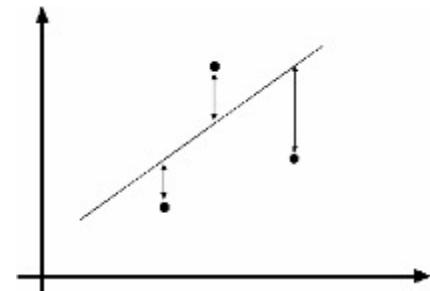
$$\begin{bmatrix} p_1 & 1 \\ p_2 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} p_1' \\ p_2' \end{bmatrix}$$

$$Ax = b$$

What if the data is noisy?

$$\begin{bmatrix} p_1 & 1 \\ p_2 & 1 \\ p_3 & 1 \\ \dots & \dots \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} p_1' \\ p_2' \\ p_3' \\ \dots \end{bmatrix}$$

$$\min \|Ax - b\|^2$$



overconstrained

# Least-Squares

---

- Solve:

$$A \ x = b$$

$$(N,d)(d,1) = (N,1)$$

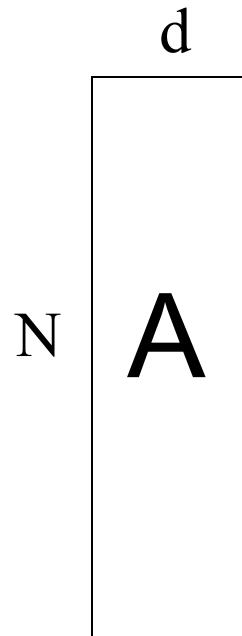
- Normal equations

$$A^T A \ x = A^T b$$

$$(d,N)(N,d)(d,1) = (d,N)(N,1)$$

- Solution:

$$x = (A^T A)^{-1} A^T b$$



$\text{rank}(A) \leq \min(d, N)$   
assume  $\text{rank}(A) = d$   
implies  $\text{rank}(A^T A) = d$   
 $A^T A$  is invertible

# Solving for homographies

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$$\mathbf{p}' = \mathbf{H}\mathbf{p}$$

$$\begin{bmatrix} wx' \\ wy' \\ w \end{bmatrix} = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Can set scale factor  $i=1$ . So, there are 8 unknowns.

Set up a system of linear equations:

$$\mathbf{Ah} = \mathbf{b}$$

where vector of unknowns  $\mathbf{h} = [a,b,c,d,e,f,g,h]^T$

Need at least 8 eqs, but the more the better...

Solve for  $\mathbf{h}$ . If overconstrained, solve using least-squares:

$$\min \|A\mathbf{h} - \mathbf{b}\|^2$$

Can be done in Matlab using “\” command

- see “help lmdivide”

# Fun with homographies

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Original image



St.Petersburg  
photo by A. Tikhonov

Virtual camera rotations



# Analysing patterns and shapes

What is the shape of the b/w floor pattern?



Homography



The floor (enlarged)

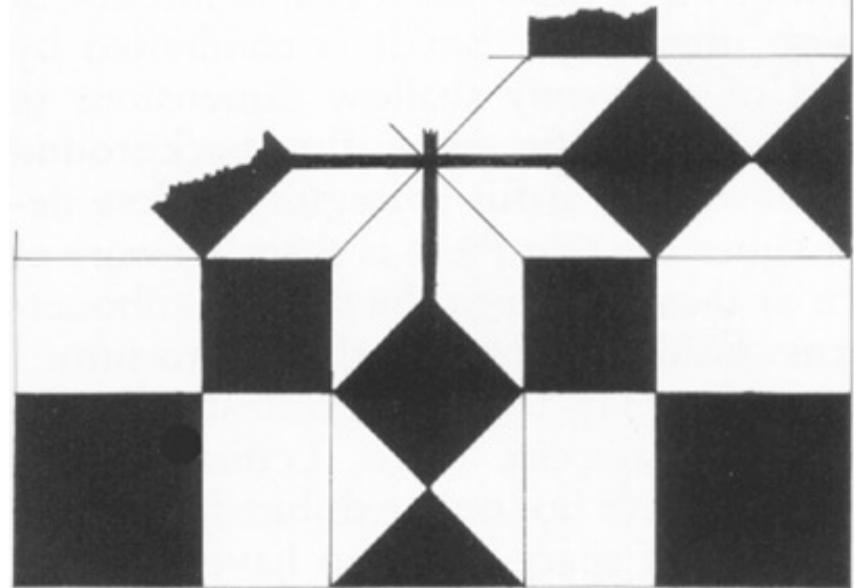


Automatically  
rectified floor

# Analysing patterns and shapes

---

Automatic rectification



From Martin Kemp *The Science of Art*  
*(manual reconstruction)*

**2 patterns have been discovered !**

# Analysing patterns and shapes

---



What is the (complicated) shape of the floor pattern?



Automatically rectified floor

***St. Lucy Altarpiece, D. Veneziano***

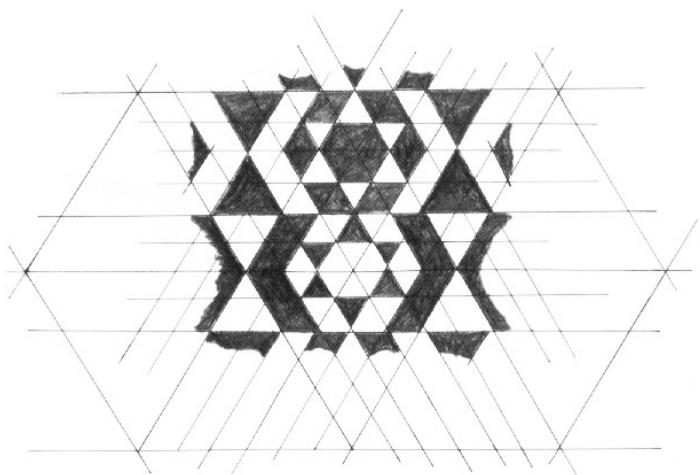
Slide from Criminisi

# Analysing patterns and shapes

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Automatic  
rectification



From Martin Kemp, *The Science of Art*  
*(manual reconstruction)*

# Mosaics: stitching images together

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virtual wide-angle camera

# Why Mosaic?

---

Are you getting the whole picture?

- Compact Camera FOV =  $50 \times 35^\circ$

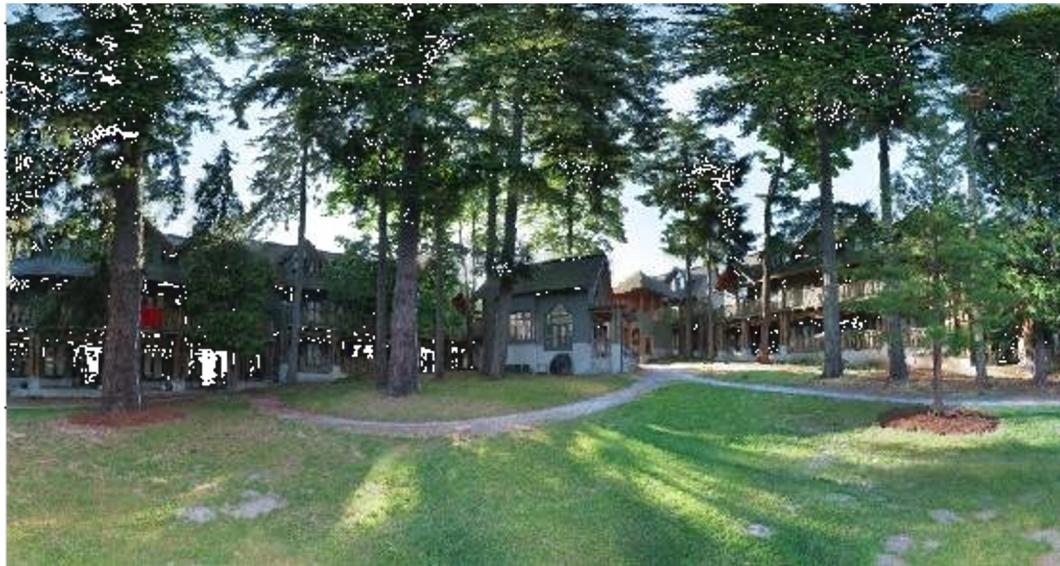


# Why Mosaic?

---

Are you getting the whole picture?

- Compact Camera FOV =  $50 \times 35^\circ$
- Human FOV =  $200 \times 135^\circ$

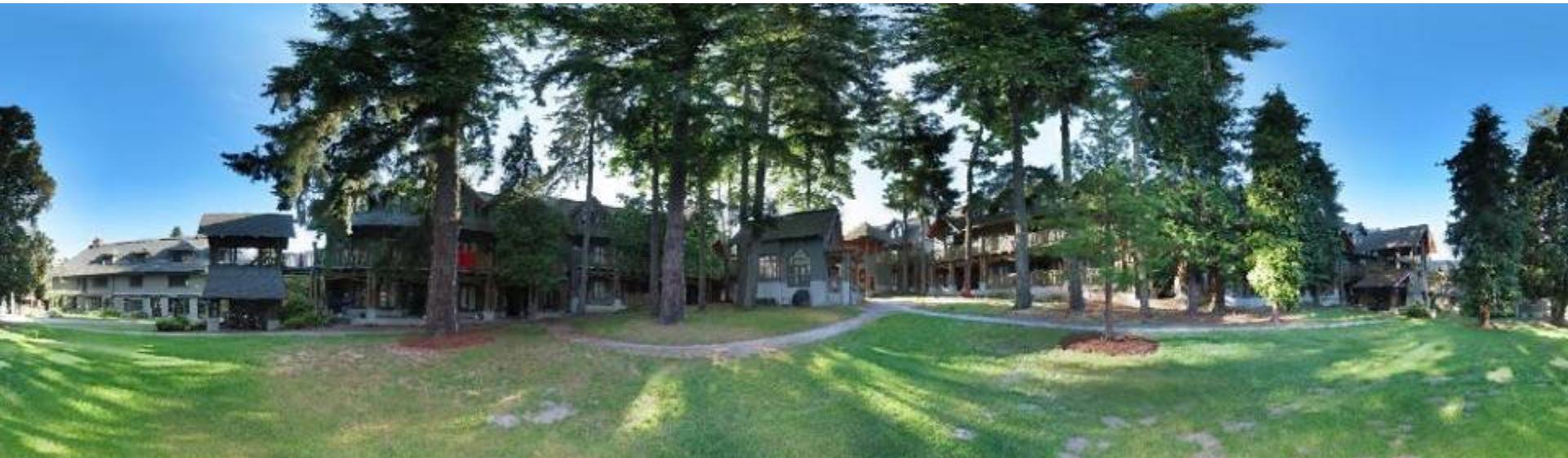


# Why Mosaic?

---

Are you getting the whole picture?

- Compact Camera FOV =  $50 \times 35^\circ$
- Human FOV =  $200 \times 135^\circ$
- Panoramic Mosaic =  $360 \times 180^\circ$



# Naïve Stitching

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left on top



right on top

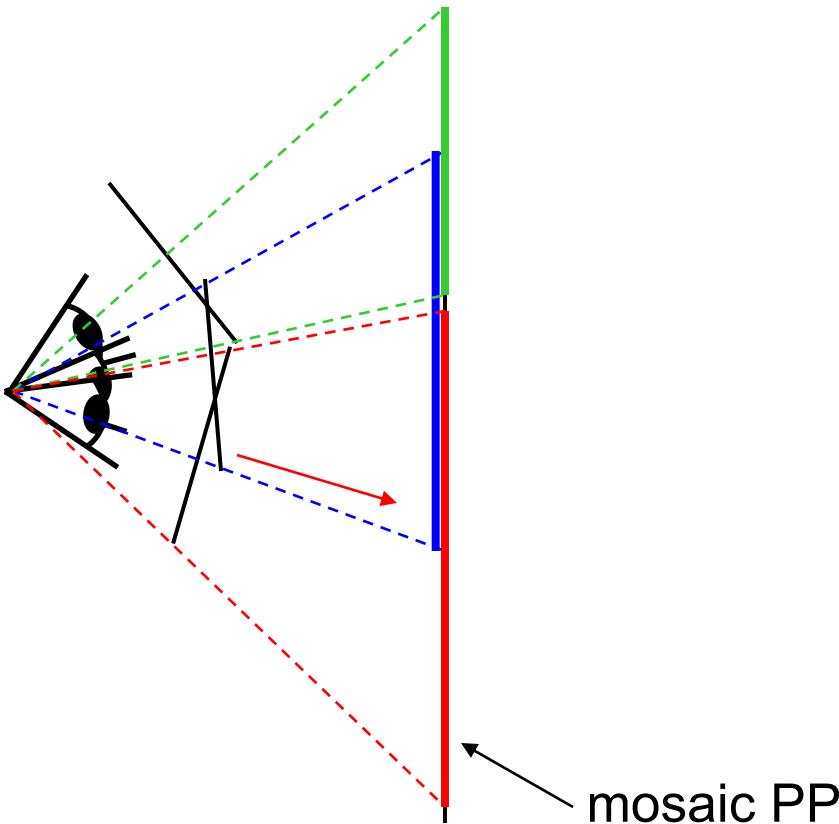


Translations are not enough to align the images



# Image reprojection

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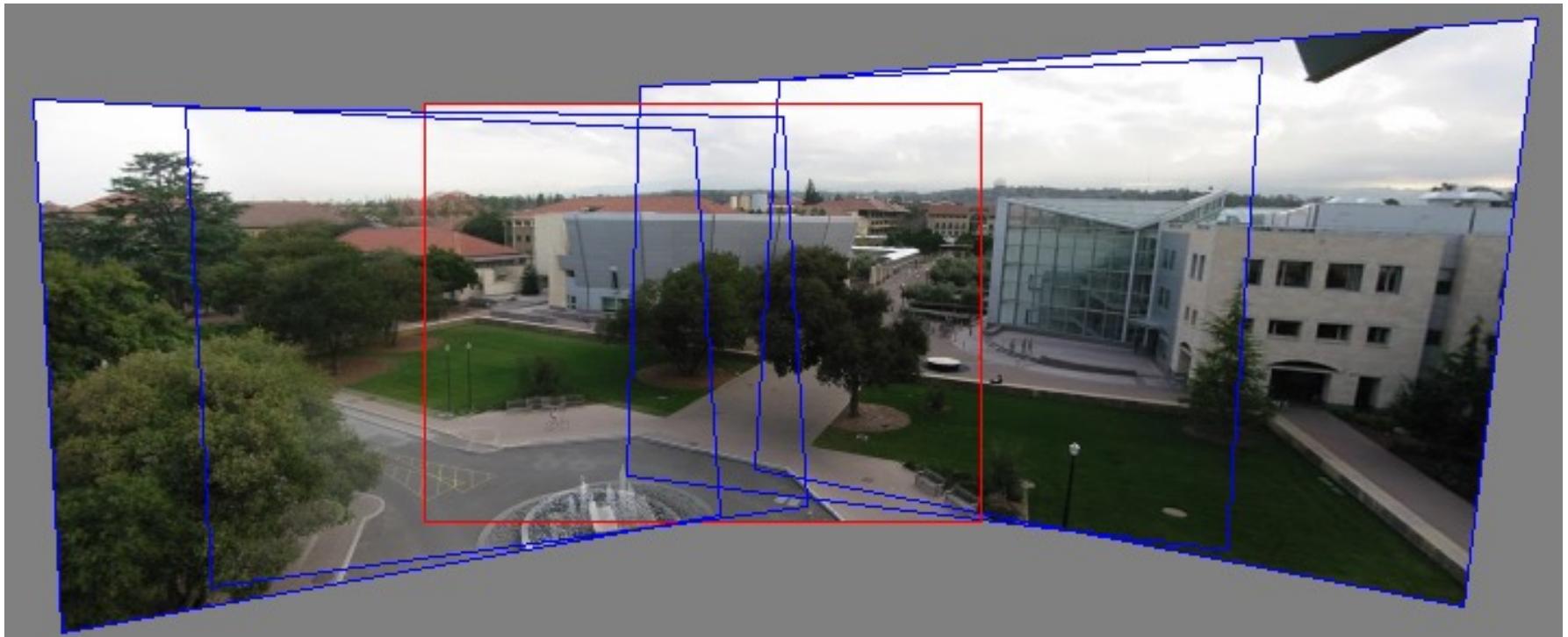


The mosaic has a natural interpretation in 3D

- The images are reprojected onto a common plane
- The mosaic is formed on this plane
- Mosaic is a *synthetic wide-angle camera*

# Panoramas

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1. Pick one image (red)
2. Warp the other images towards it (usually, one by one)
3. blend

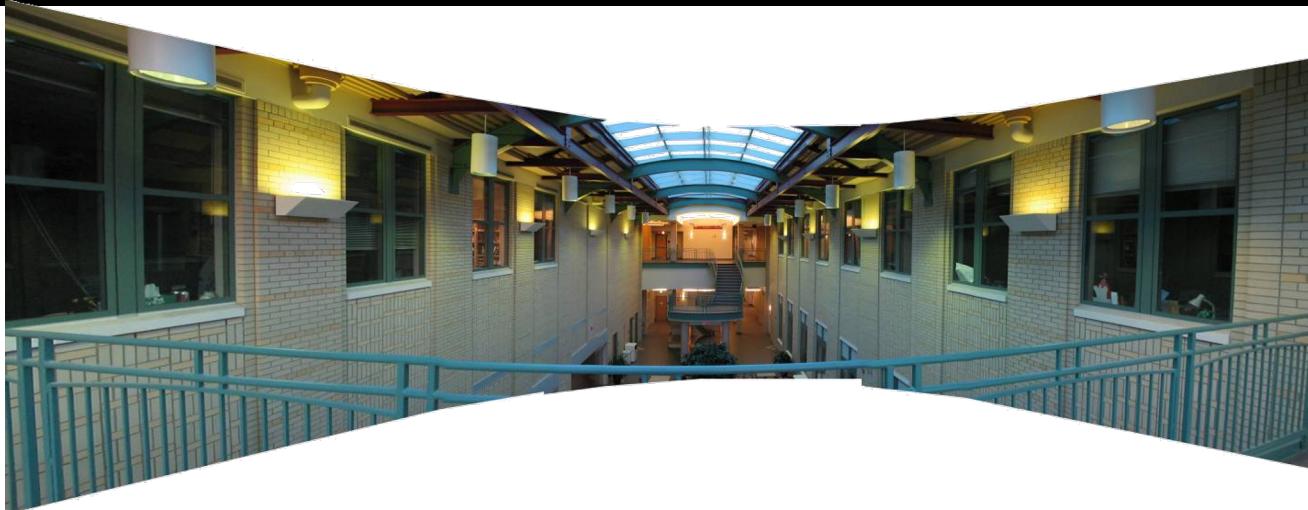
# Holbein, *The Ambassadors*

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# Programming Project #4 (part 1)

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## Homographies and Panoramic Mosaics

- Capture photographs (and possibly video)
  - Might want to use tripod
- Compute homographies (define correspondences)
  - will need to figure out how to setup system of eqs.
- (un)warp an image (undo perspective distortion)
- Produce panoramic mosaics (with blending)
- Do some of the Bells and Whistles

# Example homography final project

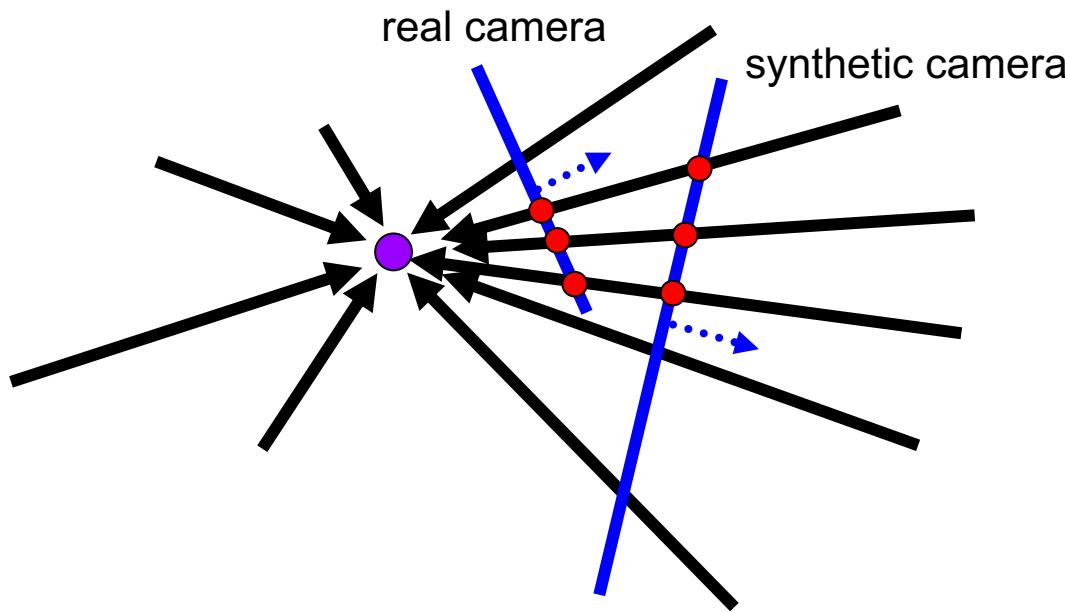
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# Think about this: When is this not true?

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We can generate any synthetic camera view as long as it has **the same center of projection!**



What happens if there are two center of projection?  
(you move your head)