



# CSC496: Deep learning in computer vision

## Prof. Bei Xiao

### Lecture 3: Image formation, Linear filtering

Slides mostly from  
MIT 6.819 Lecture 2

# Today's class

- Virtual box installation
- Image Formation with pin-hole camera; Lens
- Python/Numpy primer (finish the exercises in the tutorial)
- Homework 1: A simple vision problem: Due Wed September 18<sup>th</sup>.
- Homework 2: Pin-hole camera (out next week). Might need a real camera (however crappy). Library might be able to lend you one. Cell phone camera also works!
- <https://www.instructables.com/id/Pointless-Object-Cell-Phone-Camera-Obscura/>

# Learning outcomes

- Be familiar using common computing tools to process, manipulating, extract information from images. Being able to describe basic image formation, measurements, and analysis.
- Understand the geometric relationship between 2d image and the 3D world.
- Be familiar with basic machine learning methods in context of computer vision.
- Gained experiences with one or two common deep learning frameworks (e.g. Tensor flow, Keras, PyTorch) and GPU computing (CUDA).
- Grasp the principles of state-of-the-art deep neural networks; and
- Developed practical skills to build computer vision applications

# When do we get to deep learning?

- In about two weeks, September 26. Please be patient! And yes, you will learn A LOT about deep learning. But some foundation is very important!!
- Before that we need to cover some basics in vision:
  - 1. Image formation
  - 2. Linear filter (signal processing)
  - 3. Multiscale Pyramids
  - 4. Statistical Modeling of Images
  - 5. Get hands wet with Python/numpy/tensorflow

# Today's class

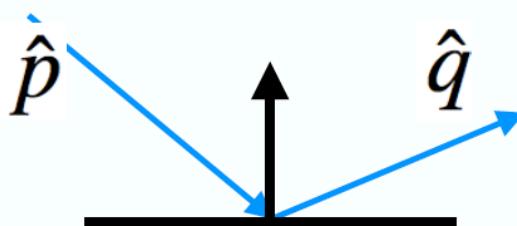
- Image Formation with pin-hole camera
- Linear filters
- Virtual box installation
- Python/Numpy primer (finish the exercises in the tutorial)
- Homework 1: A simple vision problem: Due Wed September 18<sup>th</sup>.

# Imaging

- How do we form an image with pin hole?
- How do we form an image on lenses?

# Light striking a surface

- Bidirectional reflectance distribution function

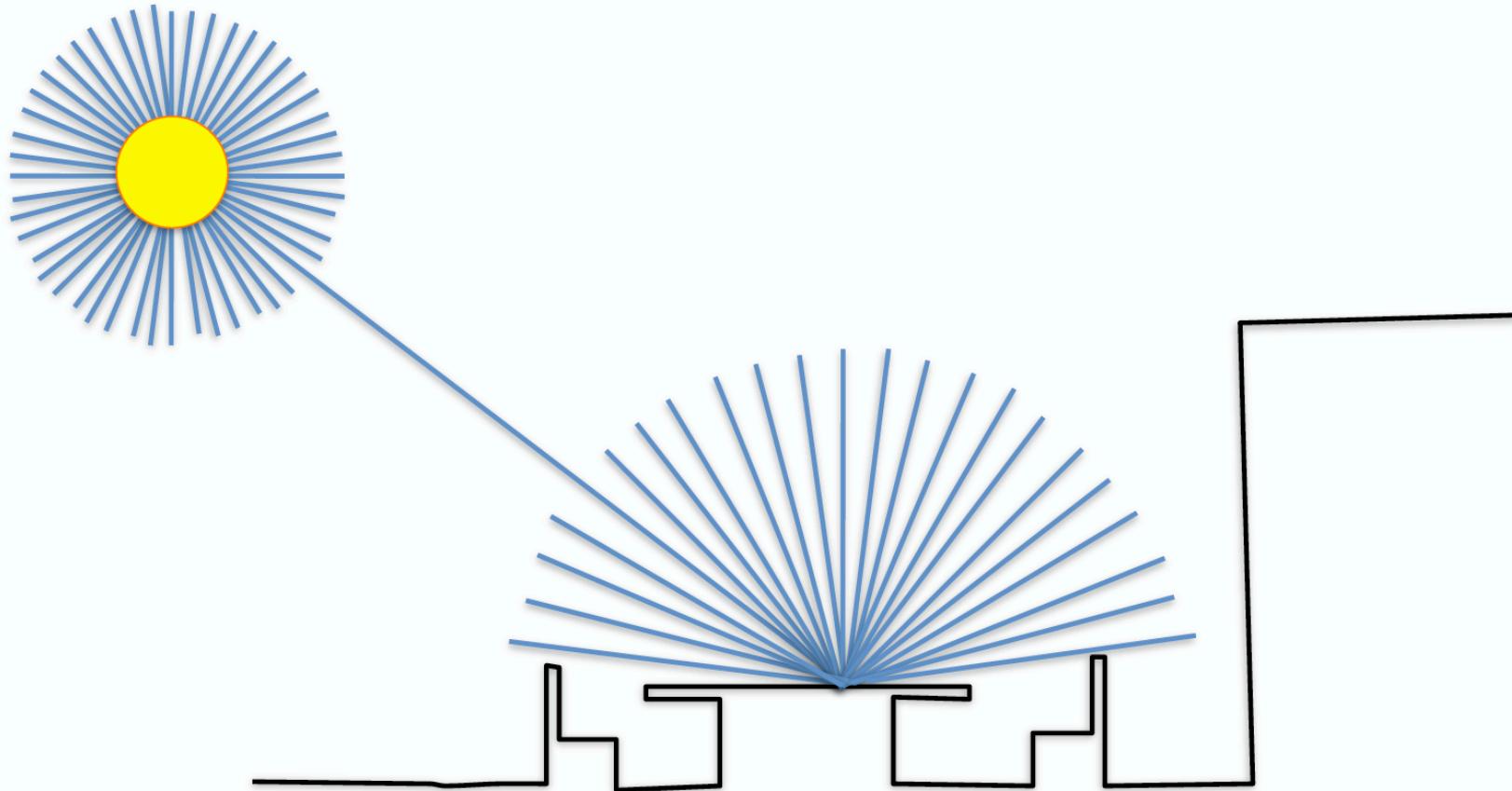


$$I_{\text{out}} = F(I_{\text{in}}, \hat{n}, \lambda, \hat{p}, \hat{q})$$

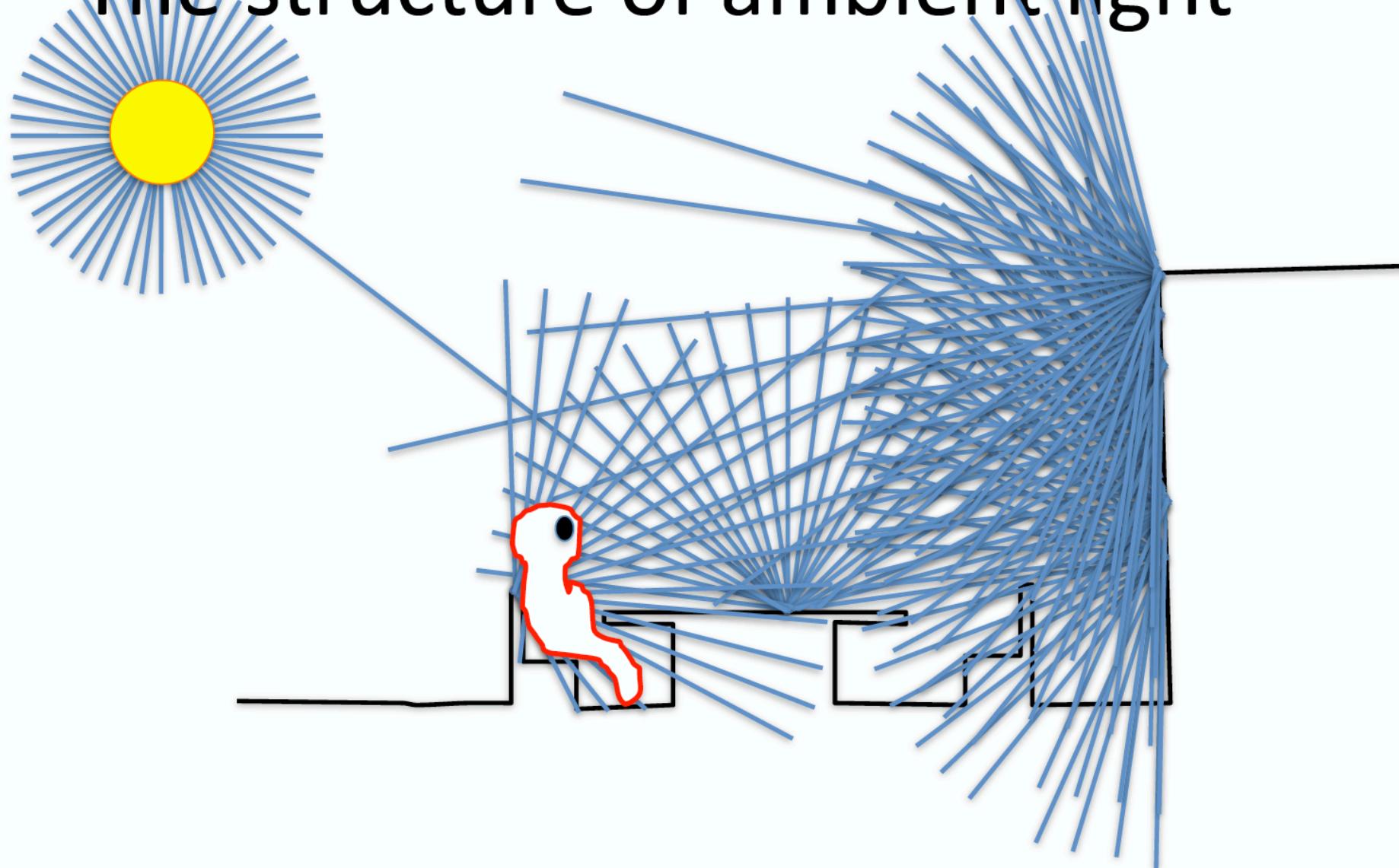
- For a Lambertian surface:

$$I_{\text{out}} = F_L(I_{\text{in}}, \hat{n}, \hat{p}) = AI_{\text{in}}(\lambda) \cos(\hat{n} \cdot \hat{p})$$

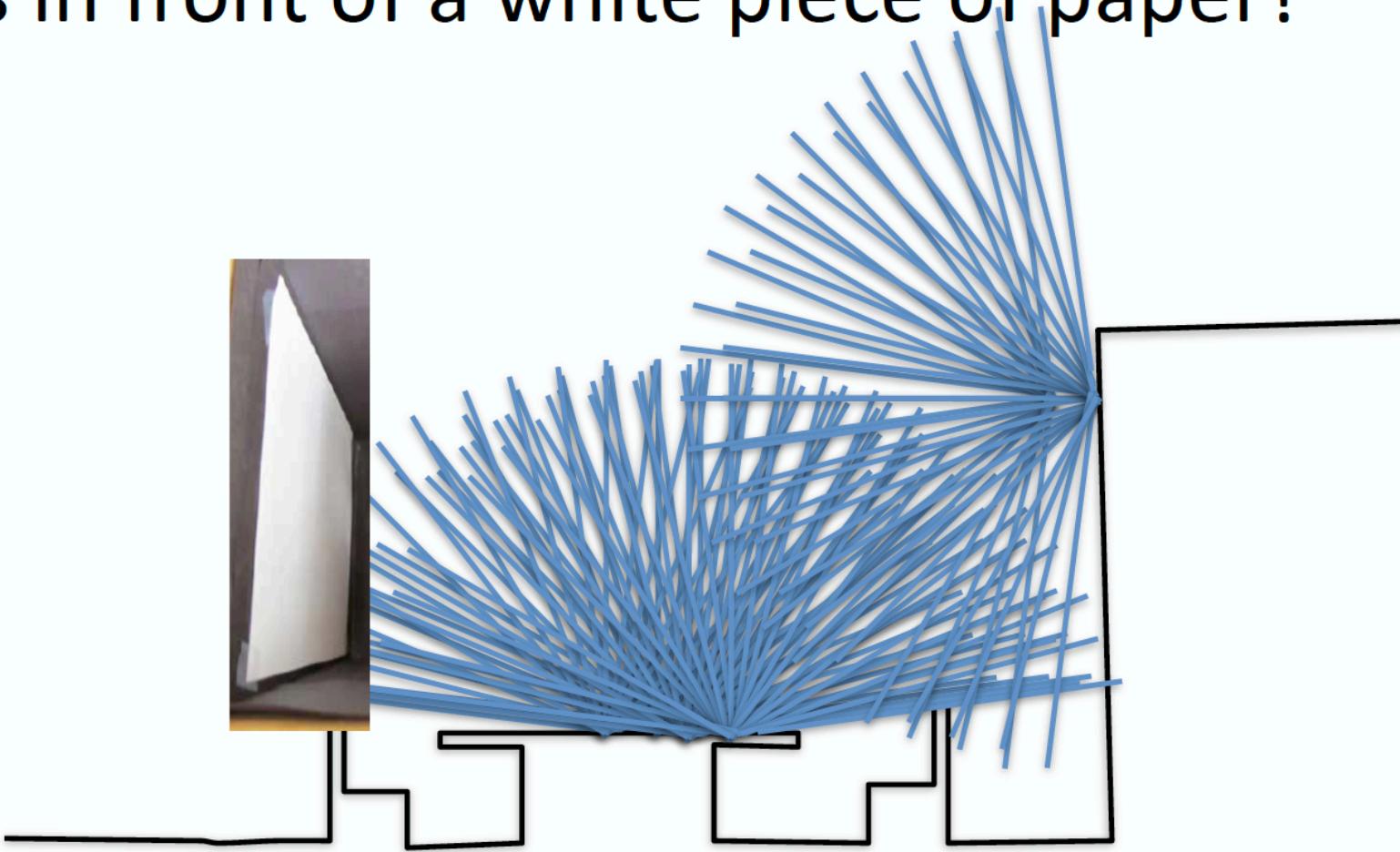
# The structure of ambient light



# The structure of ambient light

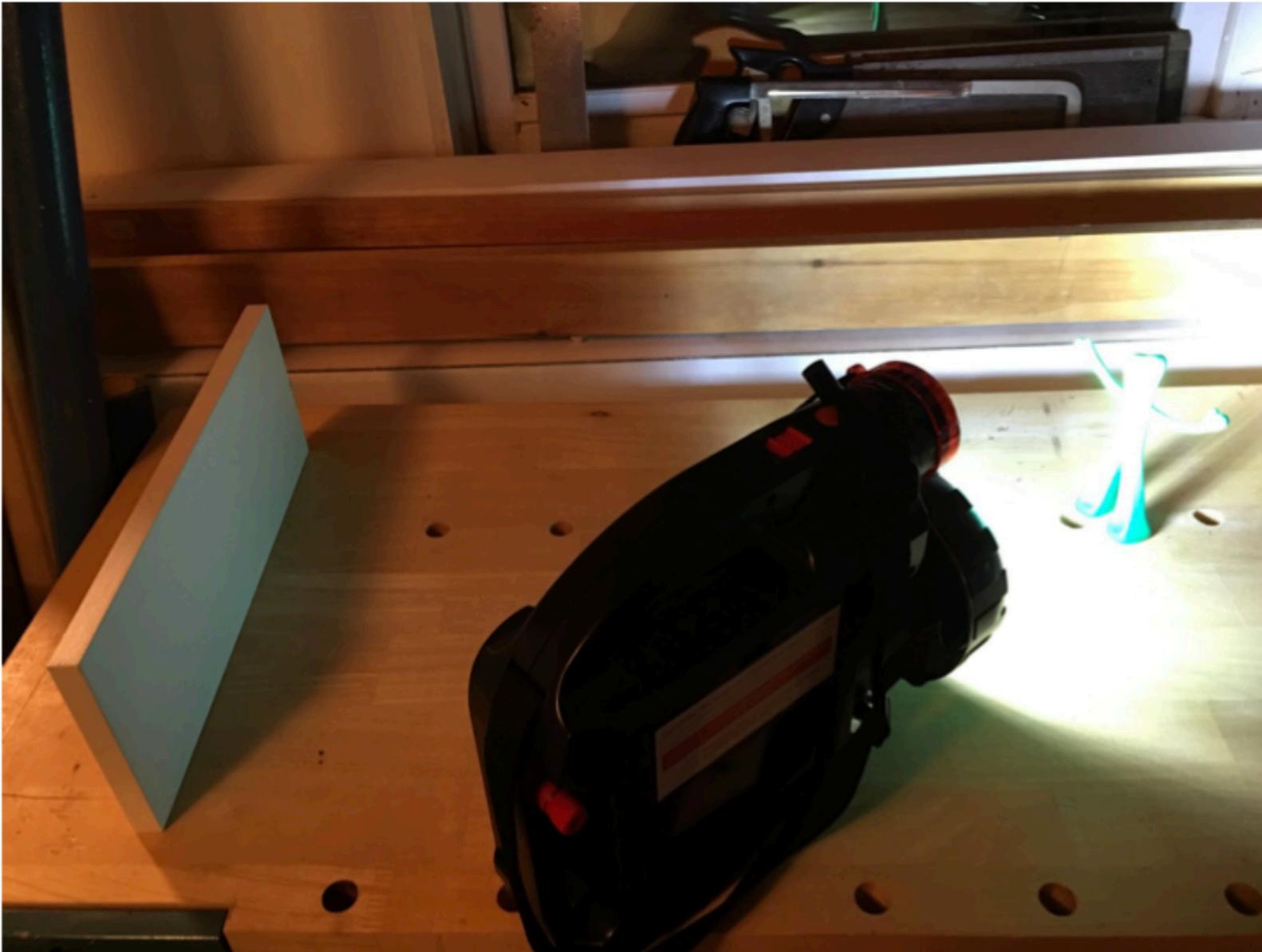


Why don't we generate an image when an object is in front of a white piece of paper?

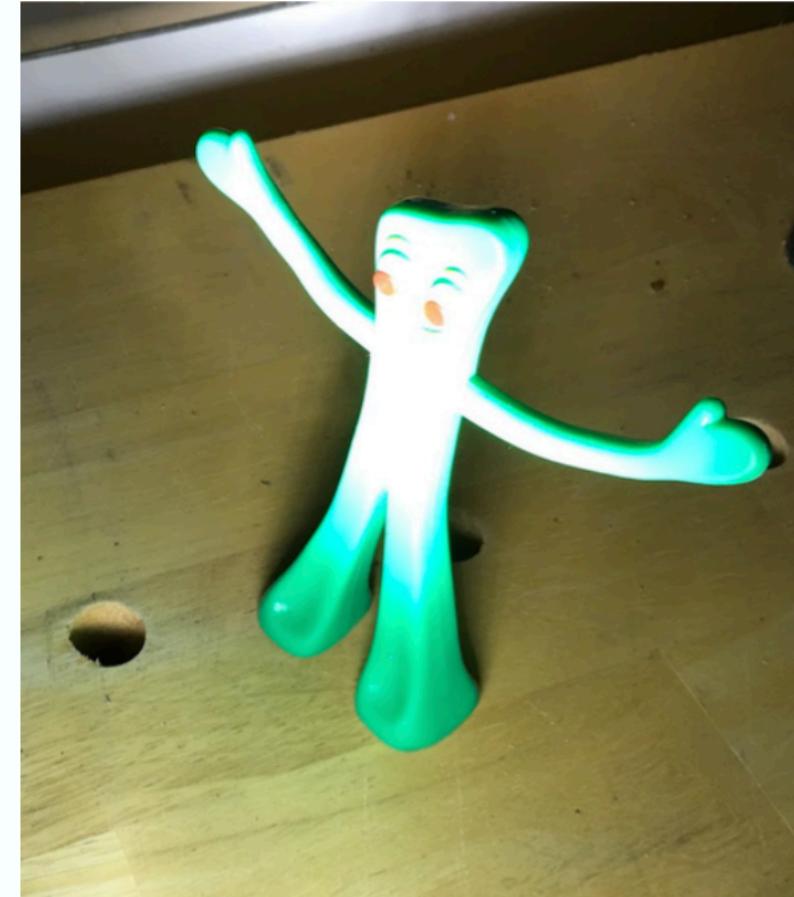


Why is there no picture appearing on the paper?

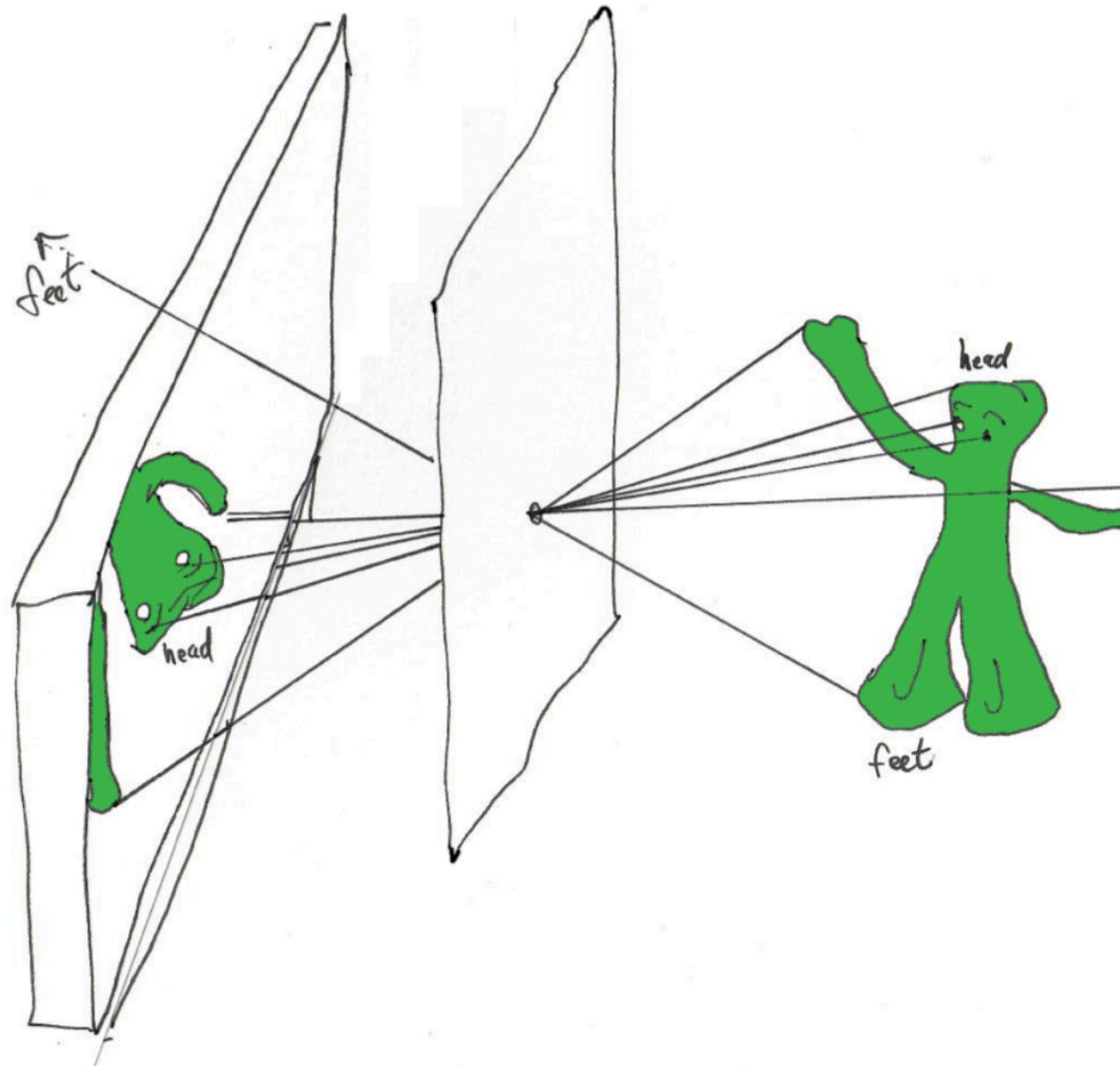
# Let's check, do we get an image?



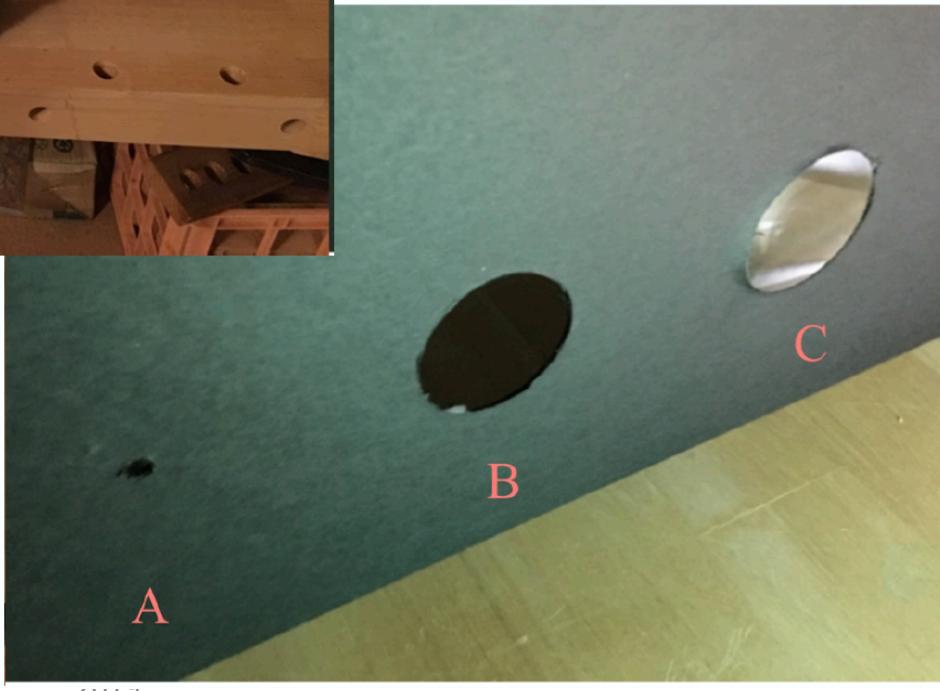
Let's check, do we get an image? No



# image is inverted



Let's try putting different occluders in between



# light on wall past pinhole



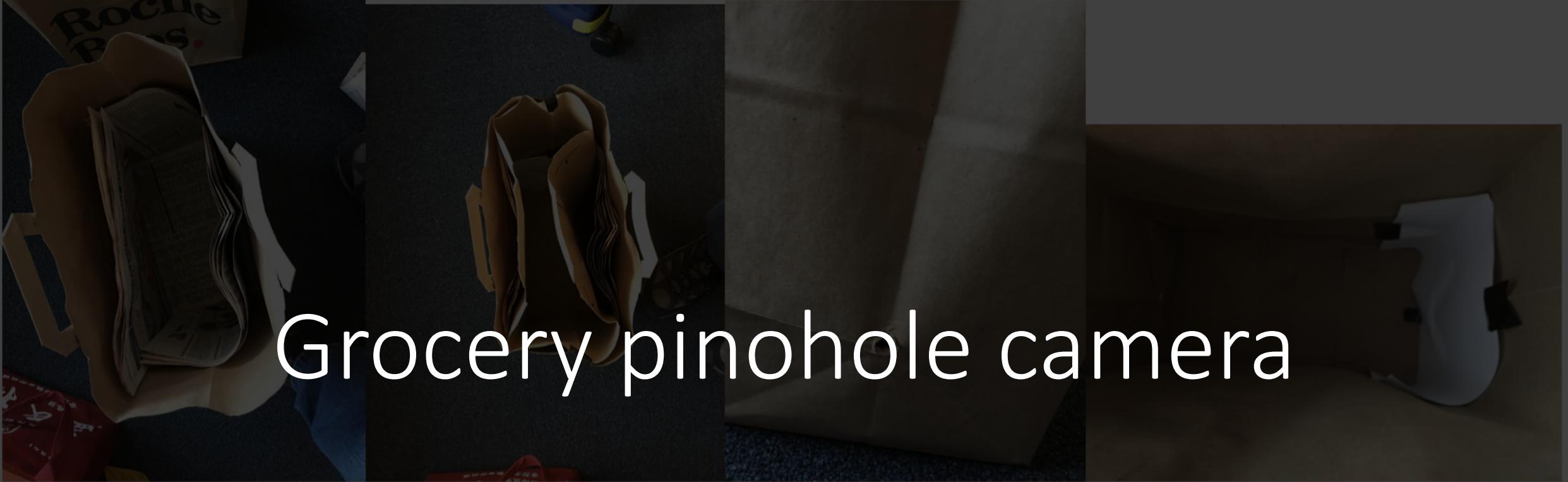
Light through hole A

# Ways to make a pin-hole camera

- Method 1:
  - Pick a box and poke a hole and remove a side for viewing
- Method 2:
  - The second is a paper bag, padded to be opaque, with a hole in it, inside of which one sticks his head.

Needed elements:

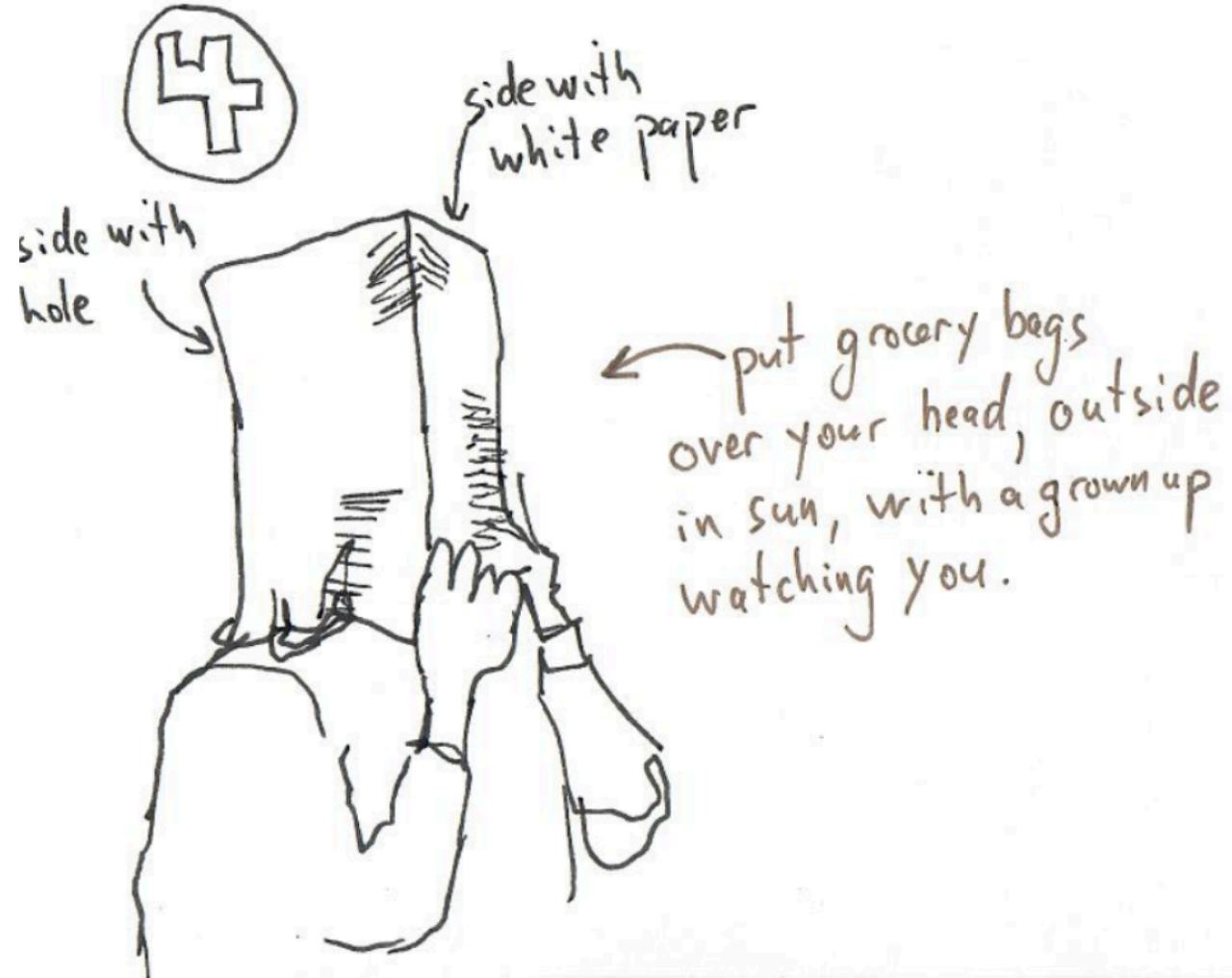
- An aperture to let light through, mechanisms to block stray light, projection screen, and some method to view or record the image on the projection screen (a camera)



# Grocery pinhole camera



# grocery bag pinhole camera



# Grocery bag pinhole camera

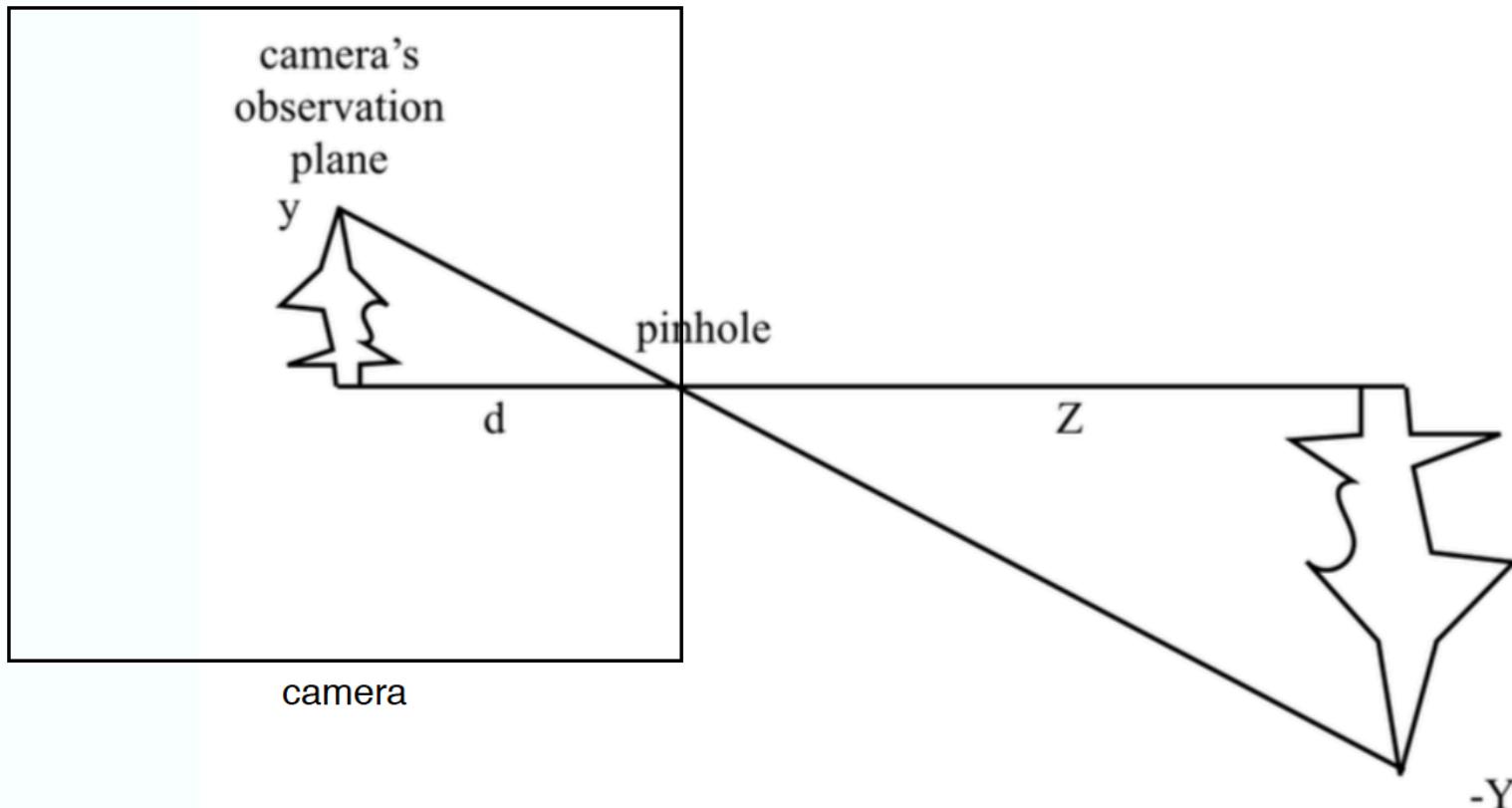
View outside the bag:

<https://www.youtube.com/watch?v=FZyCFxsyx8o>

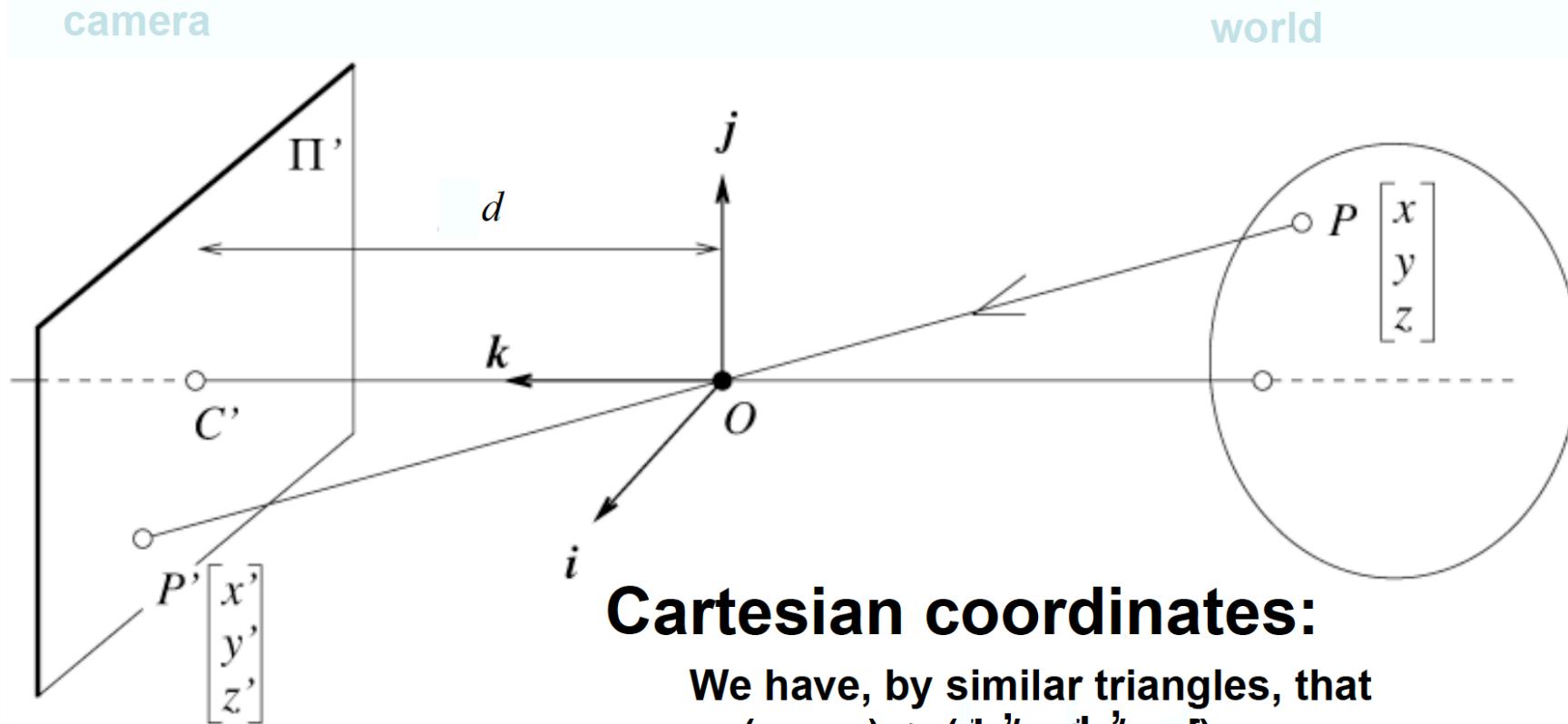
View inside the bag:

<https://www.youtube.com/watch?v=-rhZaAM3F44&feature=youtu.be>

# Perspective projection



# Perspective projection



## Cartesian coordinates:

We have, by similar triangles, that

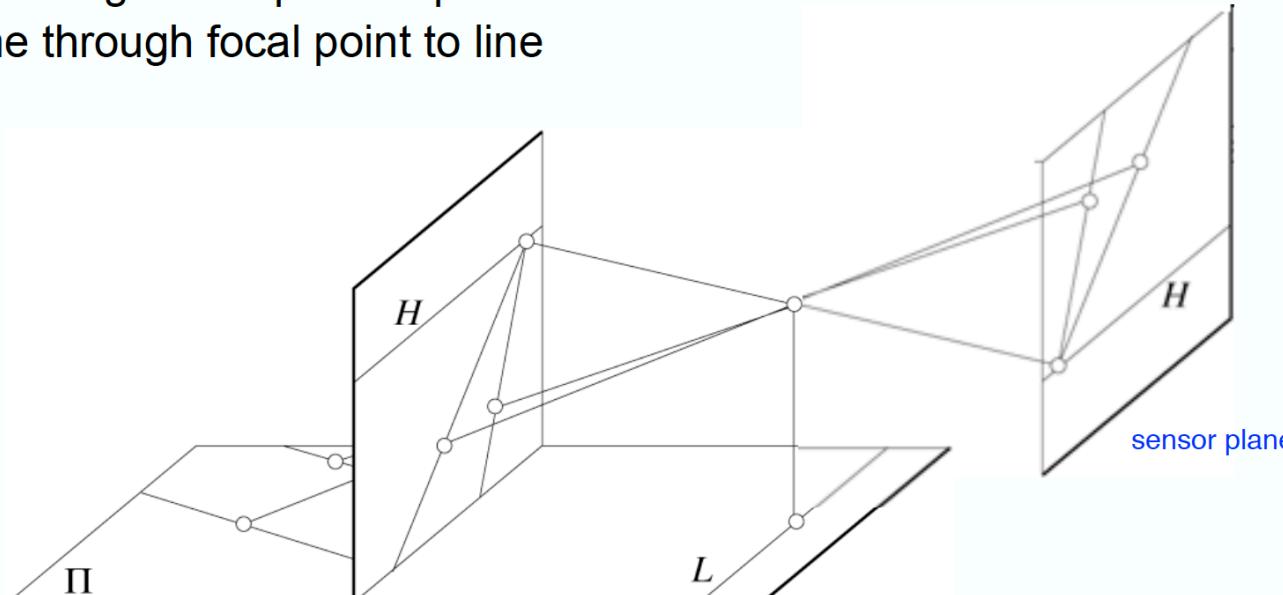
$$(x, y, z) \rightarrow (dx/z, dy/z, d)$$

Ignore the third coordinate, and get

$$(x, y, z) \rightarrow (d \frac{x'}{z}, d \frac{y'}{z})$$

# Geometric properties of projection

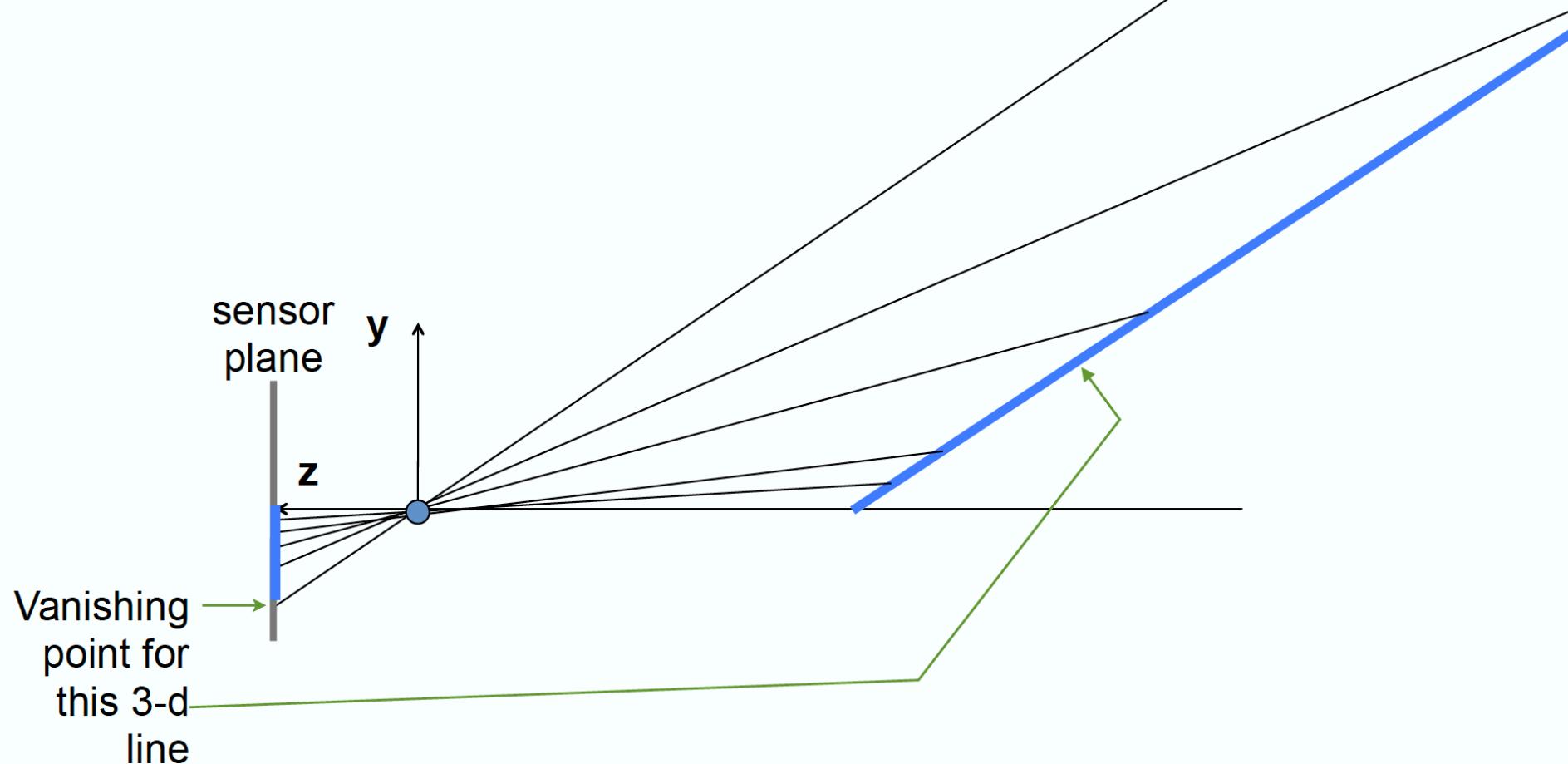
- Points go to points
- Lines go to lines
- Planes go to whole image or half-planes.
- Polygons go to polygons
- Degenerate cases
  - line through focal point to point
  - plane through focal point to line



you can also draw the sensor plane here, for simpler visualization

Figure from: Forsyth and Ponce.

# Vanishing point



## Line in 3-space

$$x(t) = x_0 + at$$

$$y(t) = y_0 + bt$$

$$z(t) = z_0 + ct$$

## Perspective projection of that line

$$x'(t) = \frac{fx}{z} = \frac{f(x_0 + at)}{z_0 + ct}$$

$$y'(t) = \frac{fy}{z} = \frac{f(y_0 + bt)}{z_0 + ct}$$

In the limit as  $t \rightarrow \pm\infty$   
we have (for  $c \neq 0$ ):



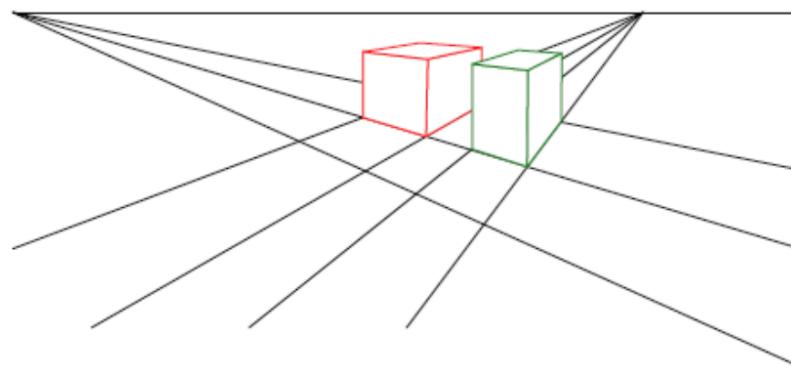
This tells us that any set of parallel lines (same  $a, b, c$  parameters) project to the same point (called the vanishing point).

$$x'(t) \longrightarrow \frac{fa}{c}$$

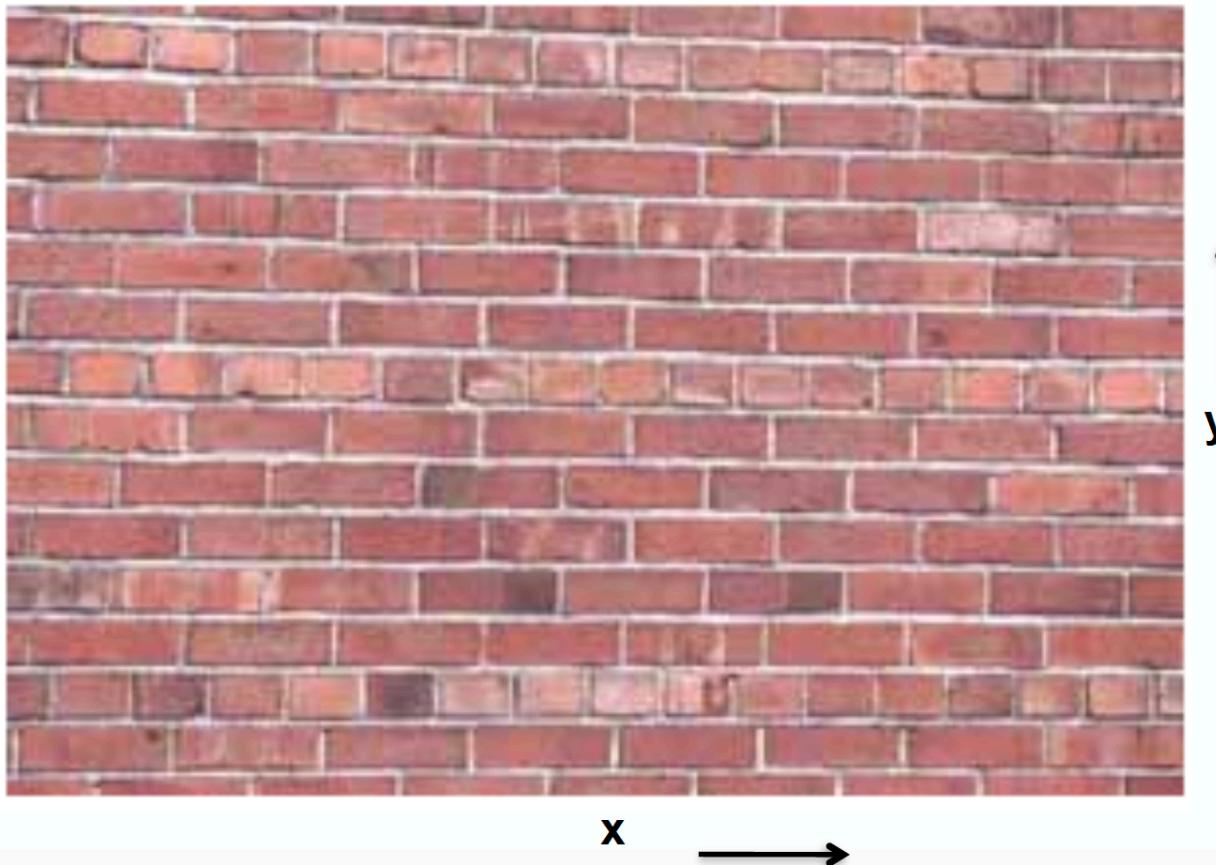
$$y'(t) \longrightarrow \frac{fb}{c}$$

# Vanishing points

- Each set of parallel lines (=direction) meets at a different point
  - The *vanishing point* for this direction
- Sets of parallel lines on the same plane lead to *collinear* vanishing points.
  - The line is called the *horizon* for that plane



# What if you photograph a brick wall head-on?



**Brick wall line in 3-space**

$$x(t) = x_0 + at$$

$$y(t) = y_0$$

$$z(t) = z_0$$

**Perspective projection of that line**

$$x'(t) = \frac{f \cdot (x_0 + at)}{z_0}$$

$$y'(t) = \frac{f \cdot y_0}{z_0}$$

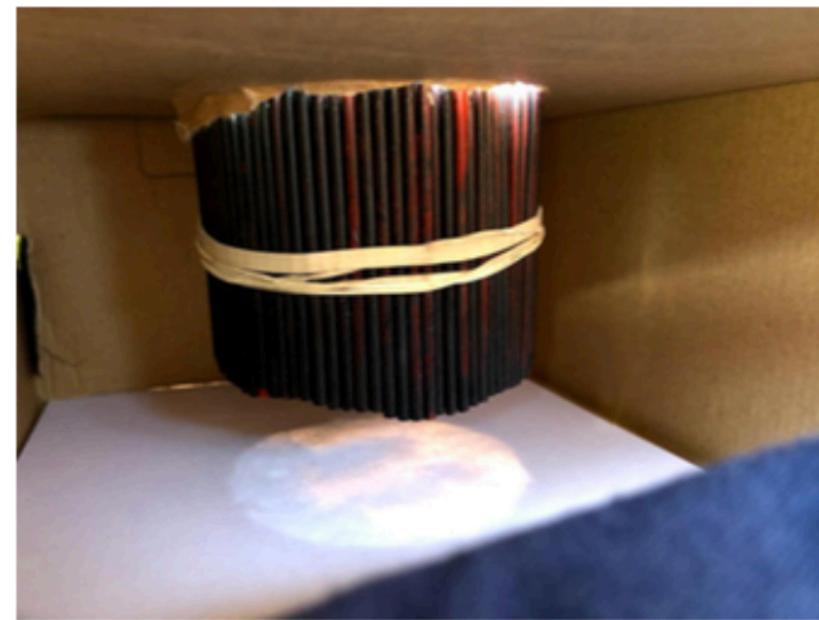
**All bricks have same  $z_0$ . Those in same row have same  $y_0$**

**Thus, a brick wall, photographed head-on, gets rendered as set of parallel lines in the image plane.**

# Straw camera



(a)

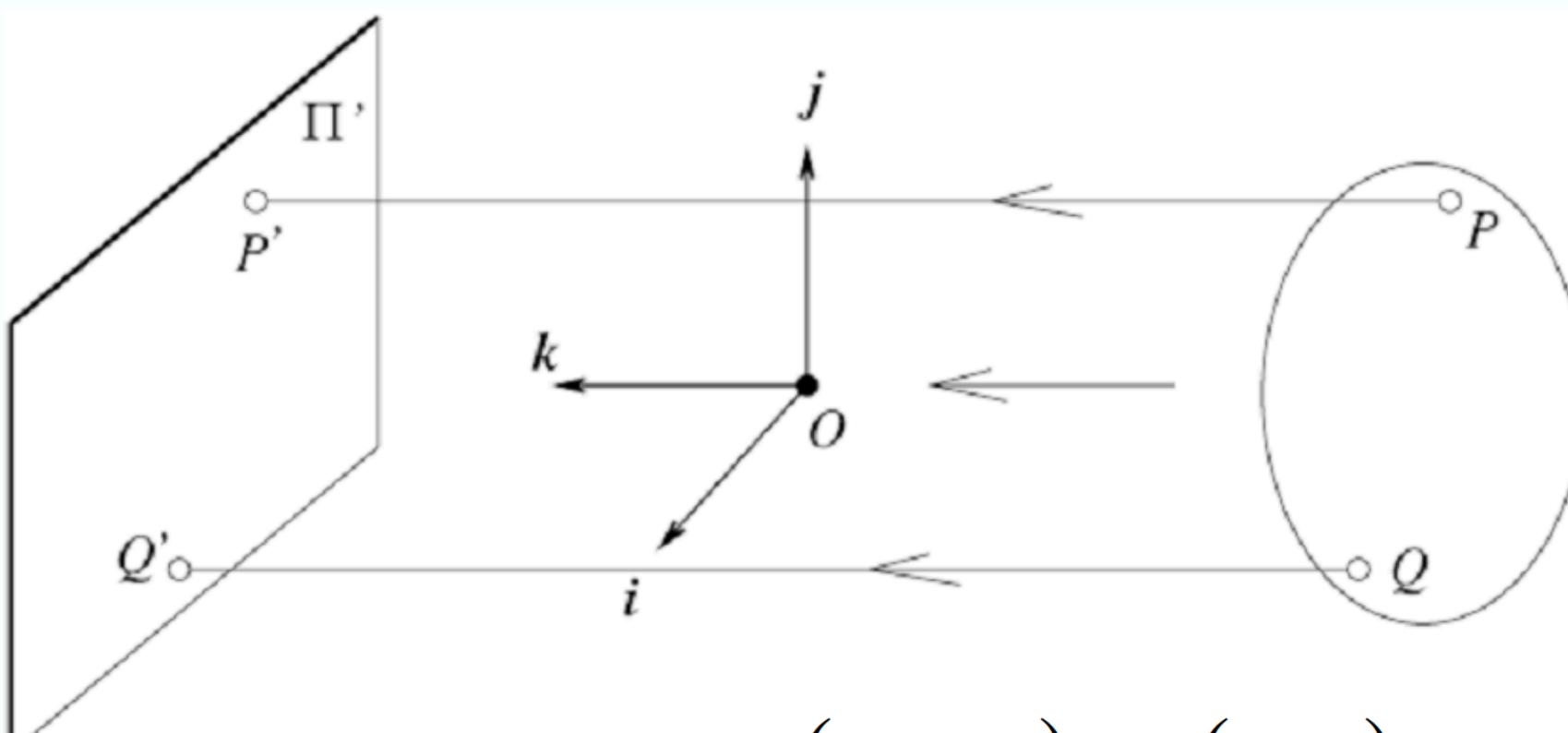


(b)

# Straw camera



# Other projection models: Orthographic projection

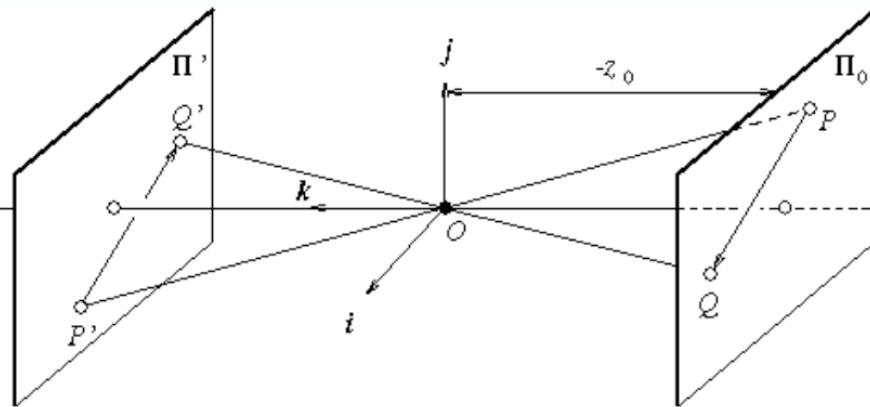


$$(x, y, z) \rightarrow (x, y)$$

# Other projection models: Weak perspective

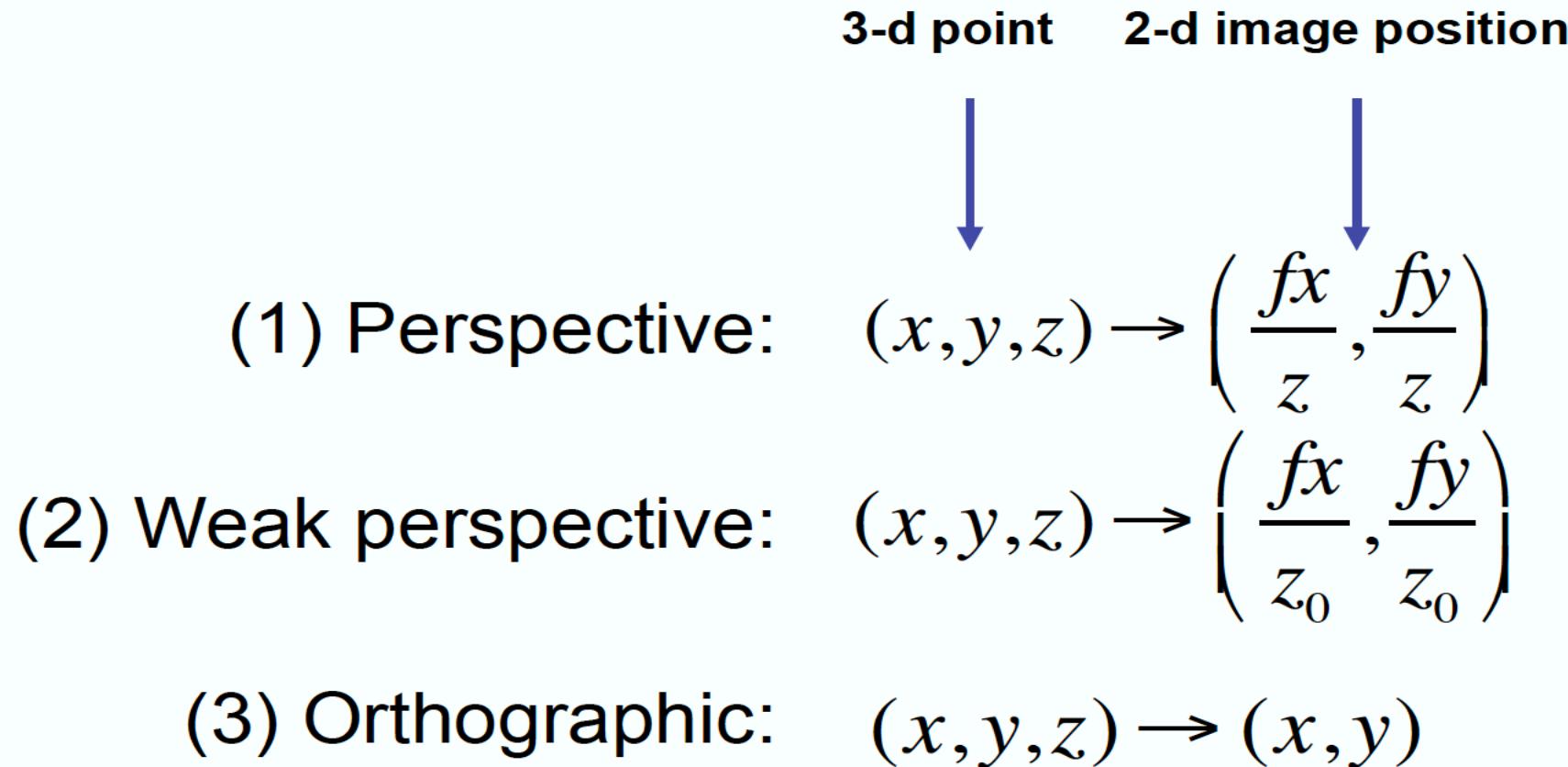
- Issue

- perspective effects, but not over the scale of individual objects
- collect points into a group at about the same depth, then divide each point by the depth of its group
- Adv: easy
- Disadv: only approximate



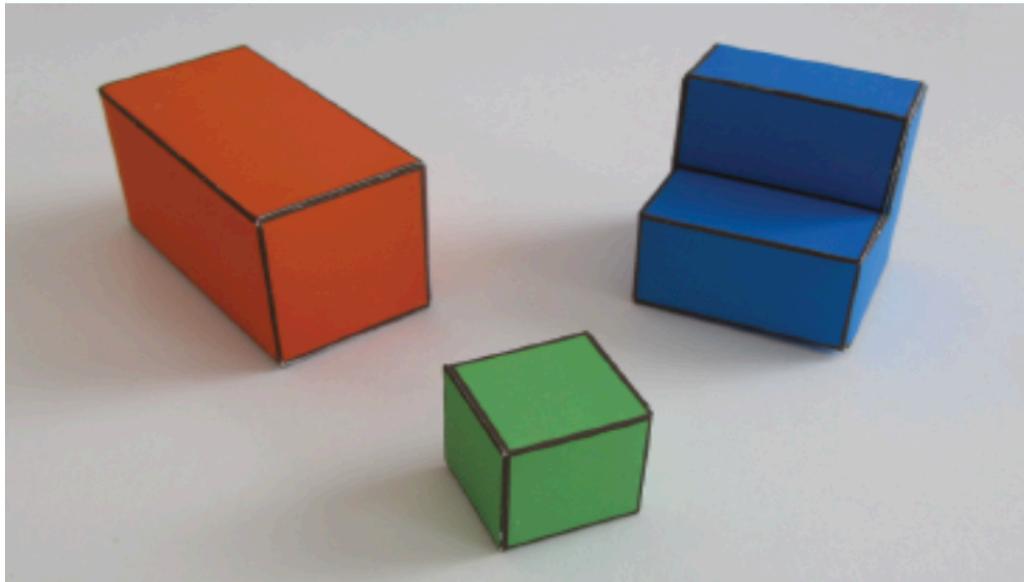
$$(x, y, z) \rightarrow \left( \frac{fx}{z_0}, \frac{fy}{z_0} \right)$$

# Three camera projections

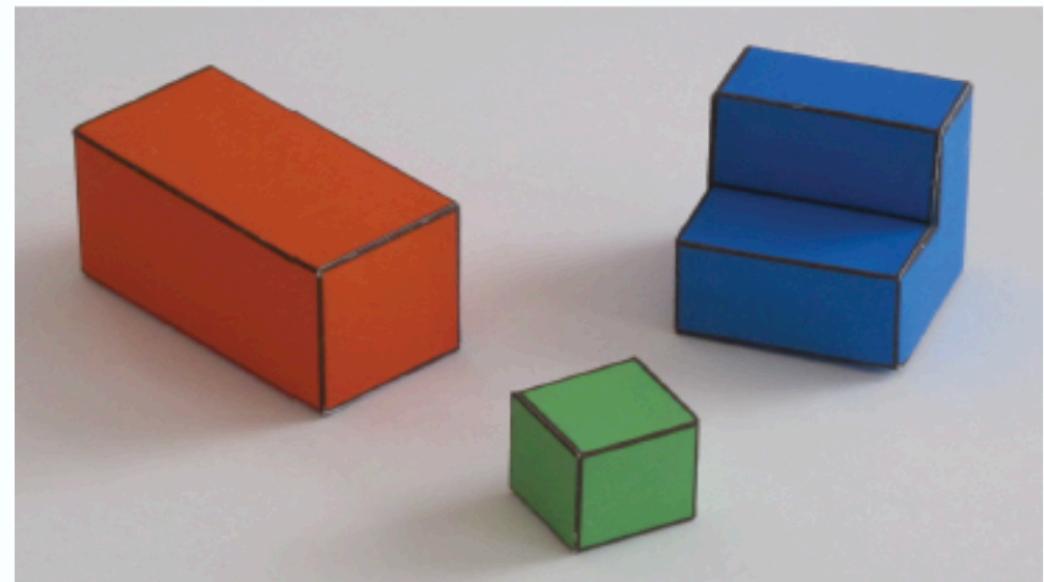


# which is perspective, which orthographic?

Perspective projection



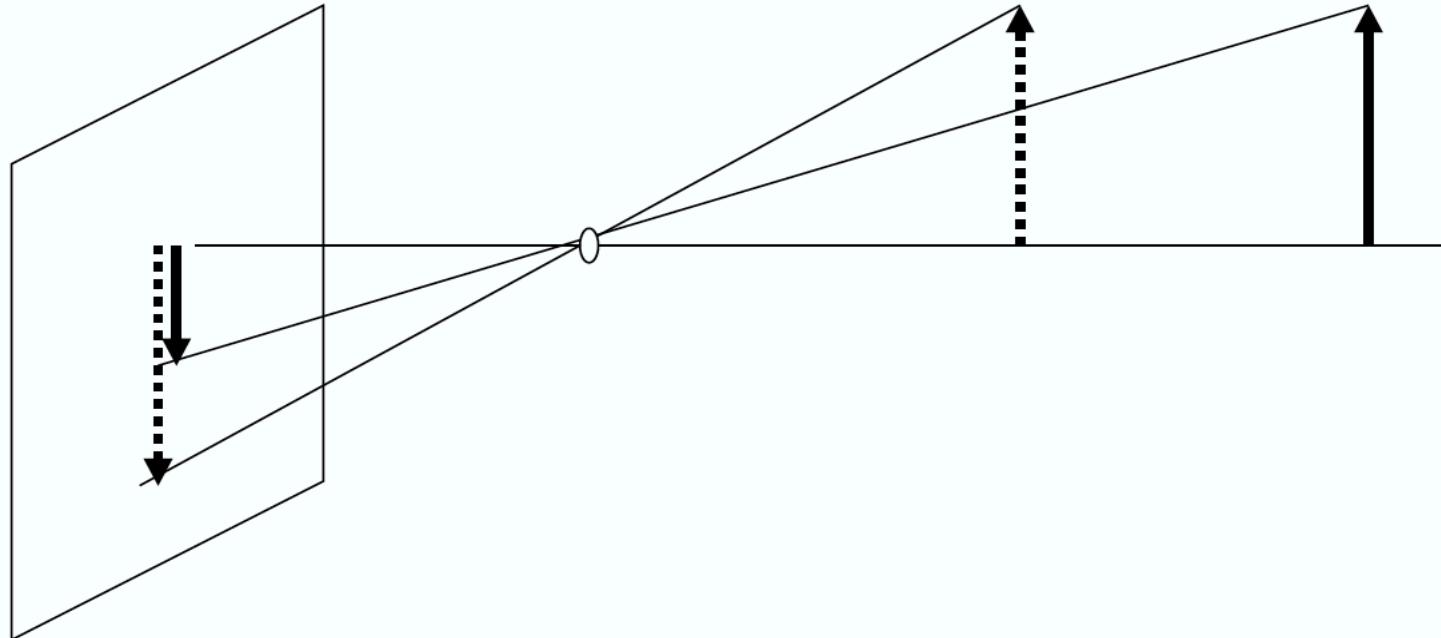
Parallel (orthographic) projection



# Example images from pinhole camera



# Measuring distance

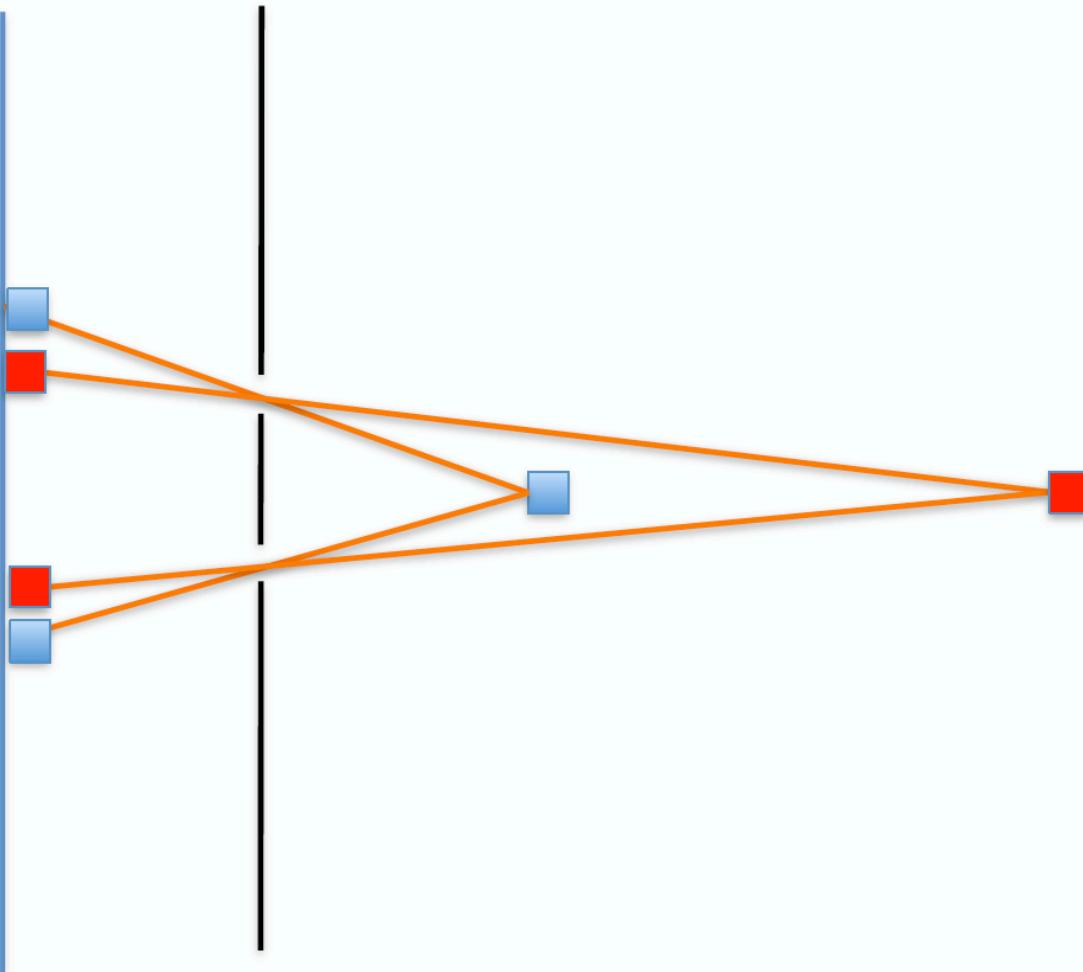


- Object size decreases with distance to the pinhole
- There, given a single projection, if we know the size of the object we can know how far it is.
- But for objects of unknown size, the 3D information seems to be lost.

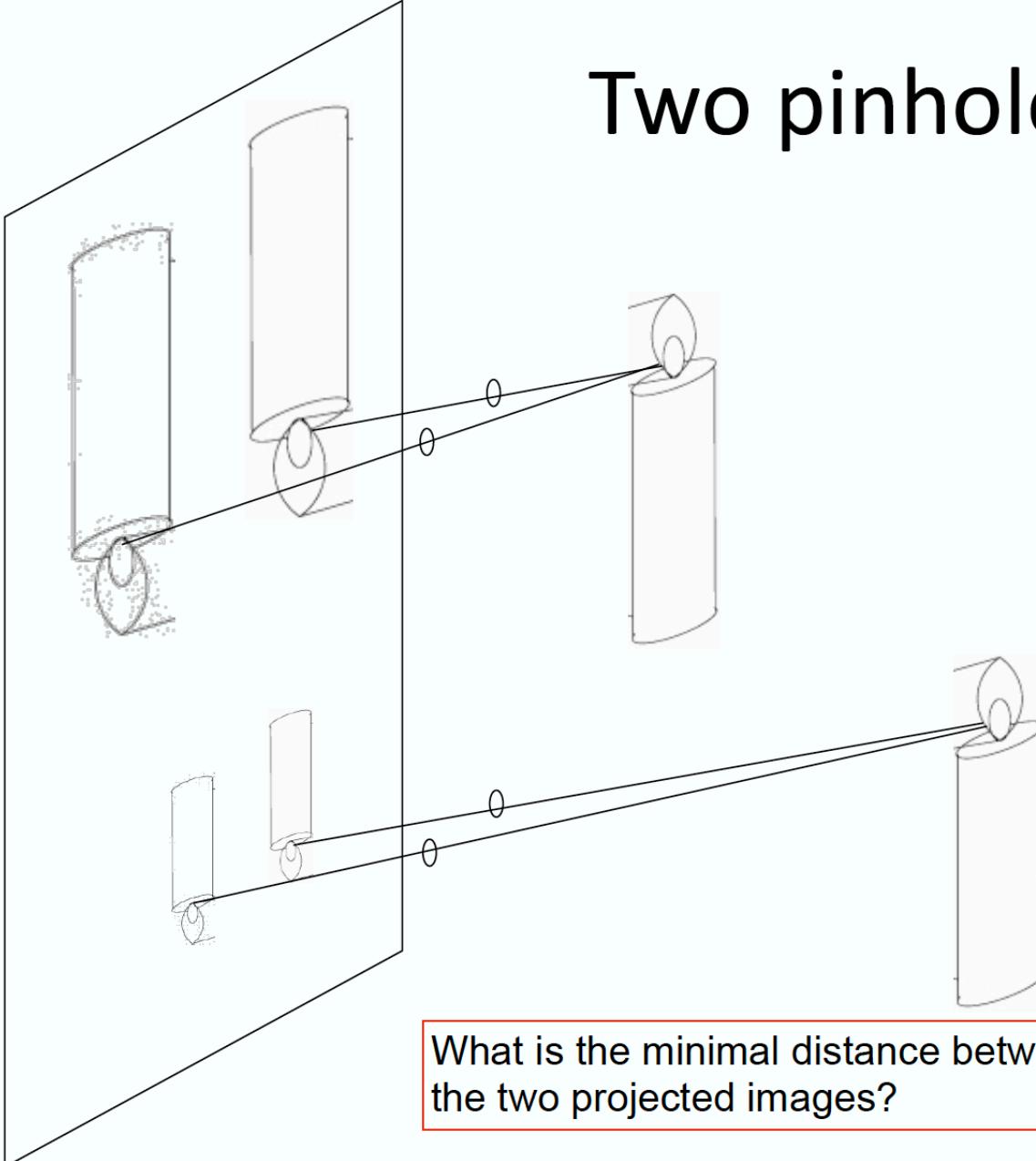
# Playing with pinholes



# Two pinholes



# Two pinholes

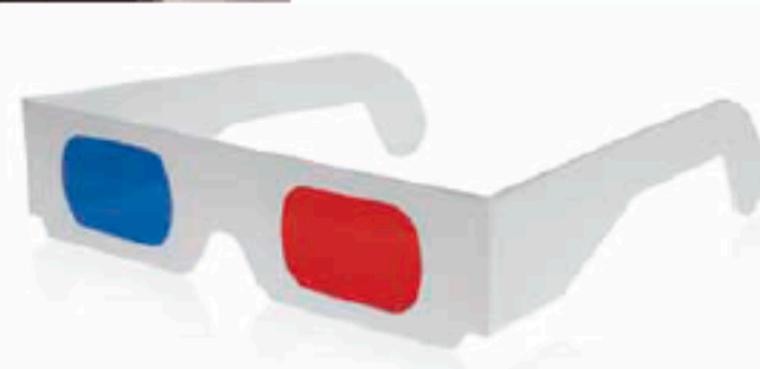
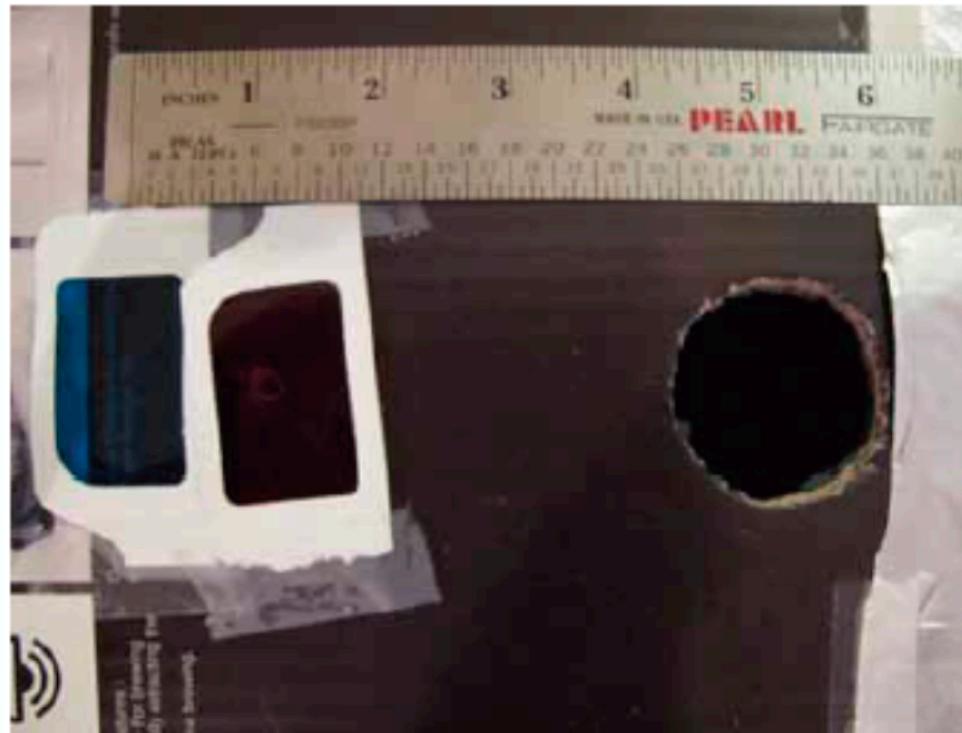


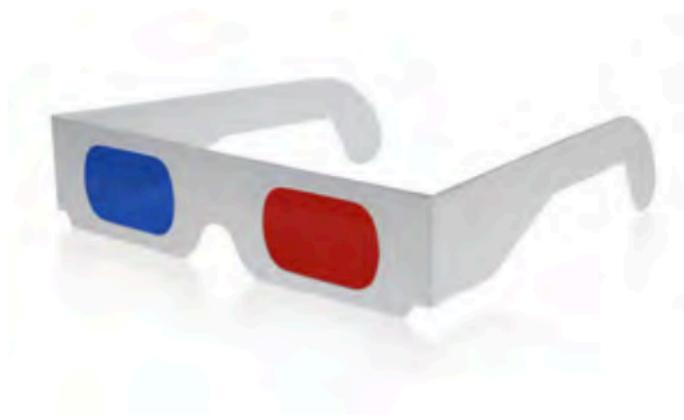
What is the minimal distance between  
the two projected images?

# Anaglyph image

Anaglyph images are a particular method to produce 3D images. Anaglyph images are formed by superimposing a pair of stereo images into two different color channels. They provide a 3D effect when viewed with color glasses.

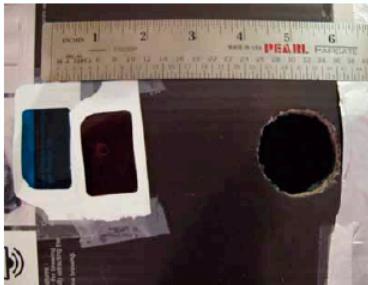
# Anaglyph pinhole camera





We used one of the glasses to get the filters and placed them in front of the two holes. Figure 3 shows two pictures of the two holes (top left) and the filters positioned in front of each hole (top right). On the bottom of that figure is a resulting 3D image, with the two views superimposed. That image should give you a 3D percept when viewed with anaglyph glasses.

# Anaglyph pinhole camera



front of camera



image of a point of light

# Anaglyph pinhole camera



# Problem set 2: Make your own pin-hole camera

- Build the camera: e.g. shoe box, make sure no light is leaking through.
- Decide where to put the hole: Make sure that the distance between the pinhole face and the screen face -- the focal length of your pinhole camera -- is not less than the minimum focus distance of your camera.
- Make pinkhole(s): I recommend cutting a larger hole in the cardboard box and attaching a thinner sheet of card stock, and having holes in the card stock.
- Cover the inside wall facing the pinhole with white copy paper.
- Create a hole for the digital camera

# Problem set 2: Make your own pin-hole camera



# Take-home reading and next class

- Lecture notes: cameras.pdf (uploaded on blackboards)
- Next class: linear filter and convolution