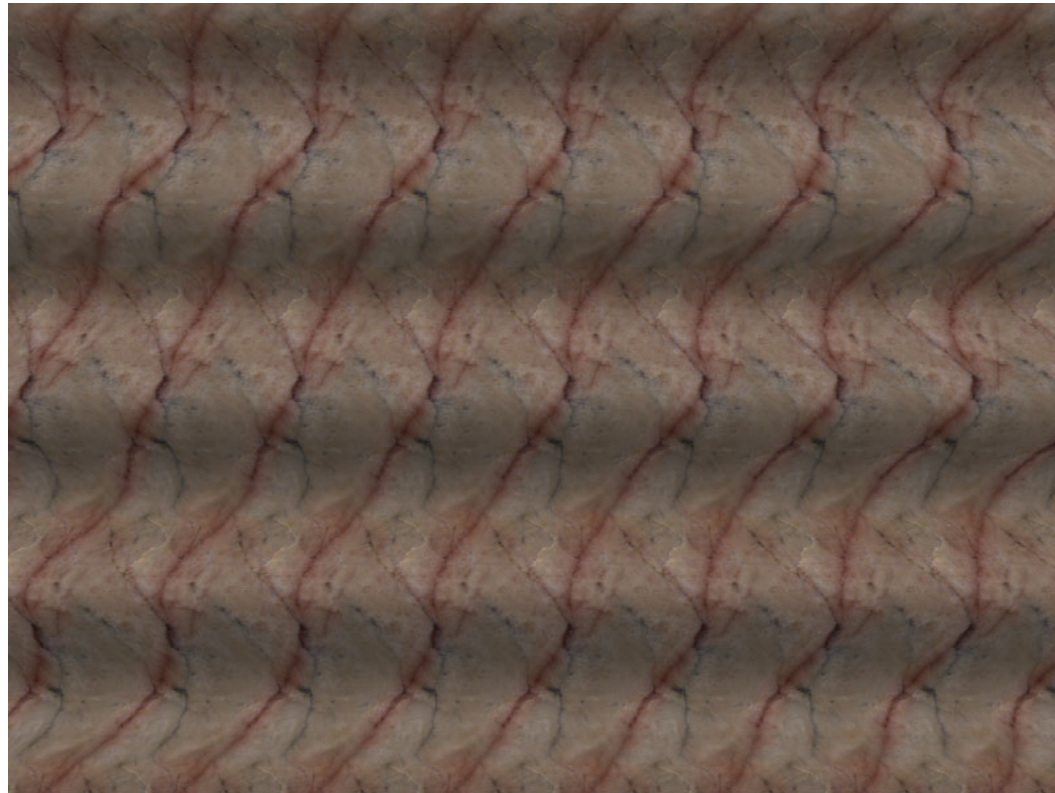


# CS4670/5670: Computer Vision

Kavita Bala

## Lec 25: Stereo

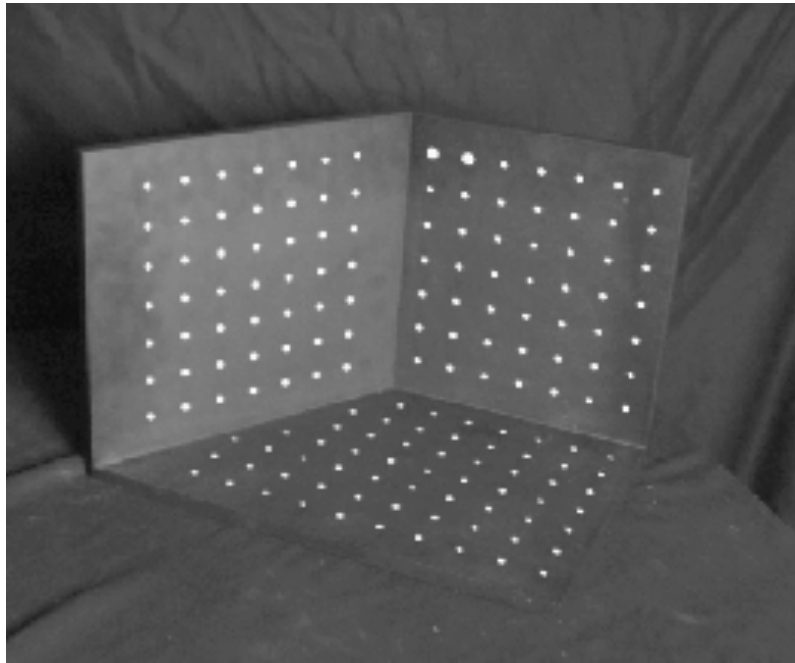


# Announcements

- Prelim grades out
  - Solutions on CMS
  - Regrades on Problem 4 and 6
    - Please drop off prelim in hand back room or in office hours with TAs

# Calibration using a reference object

- Place a known object in the scene
  - identify correspondence between image 2D and scene 3D
  - compute mapping from scene to image



## Issues

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

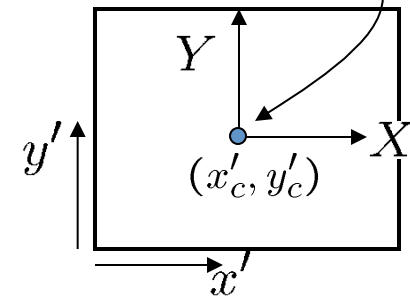
# Camera parameters

A camera is described by several parameters

- Translation  $\mathbf{T}$  of the optical center from the origin of world coords
- Rotation  $\mathbf{R}$  of the image plane
- focal length  $f$ , principal point  $(x'_c, y'_c)$ , pixel size  $(s_x, s_y)$
- blue parameters are called “extrinsics,” red are “intrinsic”

Projection equation

$$\mathbf{X} = \begin{bmatrix} sx \\ sy \\ s \end{bmatrix} = \begin{bmatrix} * & * & * & * \\ * & * & * & * \\ * & * & * & * \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} = \mathbf{\Pi} \mathbf{X}$$



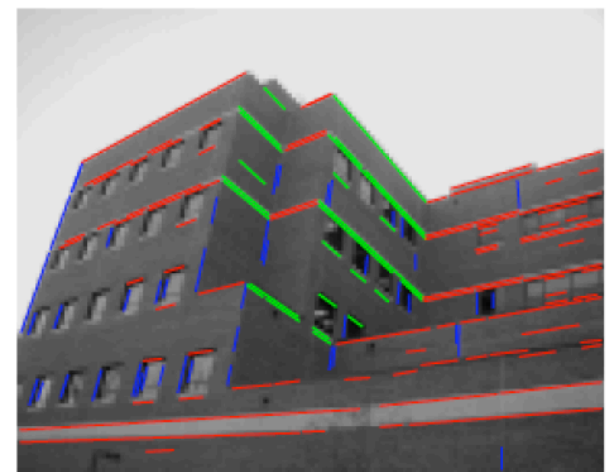
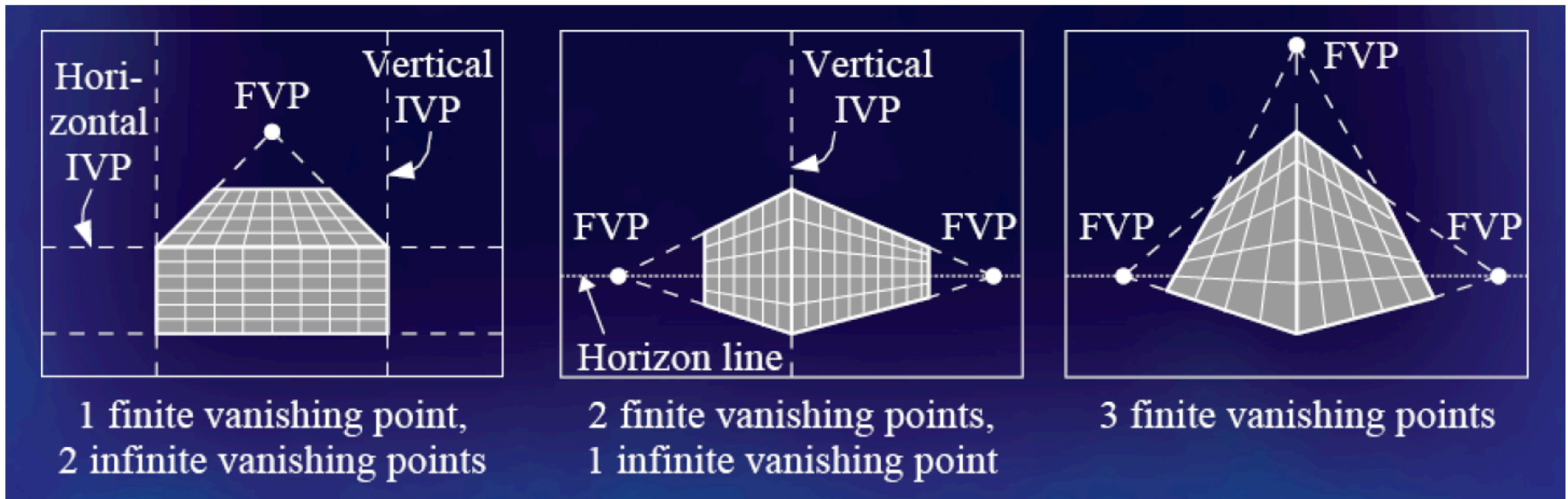
- The projection matrix models the cumulative effect of all parameters
- Useful to decompose into a series of operations

$$\mathbf{\Pi} = \begin{bmatrix} -fs_x & 0 & x'_c \\ 0 & -fs_y & y'_c \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{R}_{3 \times 3} & \mathbf{0}_{3 \times 1} \\ \mathbf{0}_{1 \times 3} & 1 \end{bmatrix} \begin{bmatrix} \mathbf{I}_{3 \times 3} & \mathbf{T}_{3 \times 1} \\ \mathbf{0}_{1 \times 3} & 1 \end{bmatrix}$$

intrinsic                  projection                  rotation                  translation                  identity matrix

- The definitions of these parameters are **not** completely standardized
  - especially intrinsic—varies from one book to another

# Calibration from vanishing points



# Calibration from 3 vanishing points

3 orthogonal world space axes

$$e_i = [1, 0, 0, 0]^T e_j = [0, 1, 0, 0]^T e_k = [0, 0, 1, 0]^T$$

Vanishing points and relation to axes

$$v_i = K R e_i, v_j = K R e_j, v_k = K R e_k$$

$$e_i^T e_j = 0$$

$$v_i^T K^{-T} R R^T K^{-1} v_j = v_i^T K^{-T} K^{-1} v_j = 0$$

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

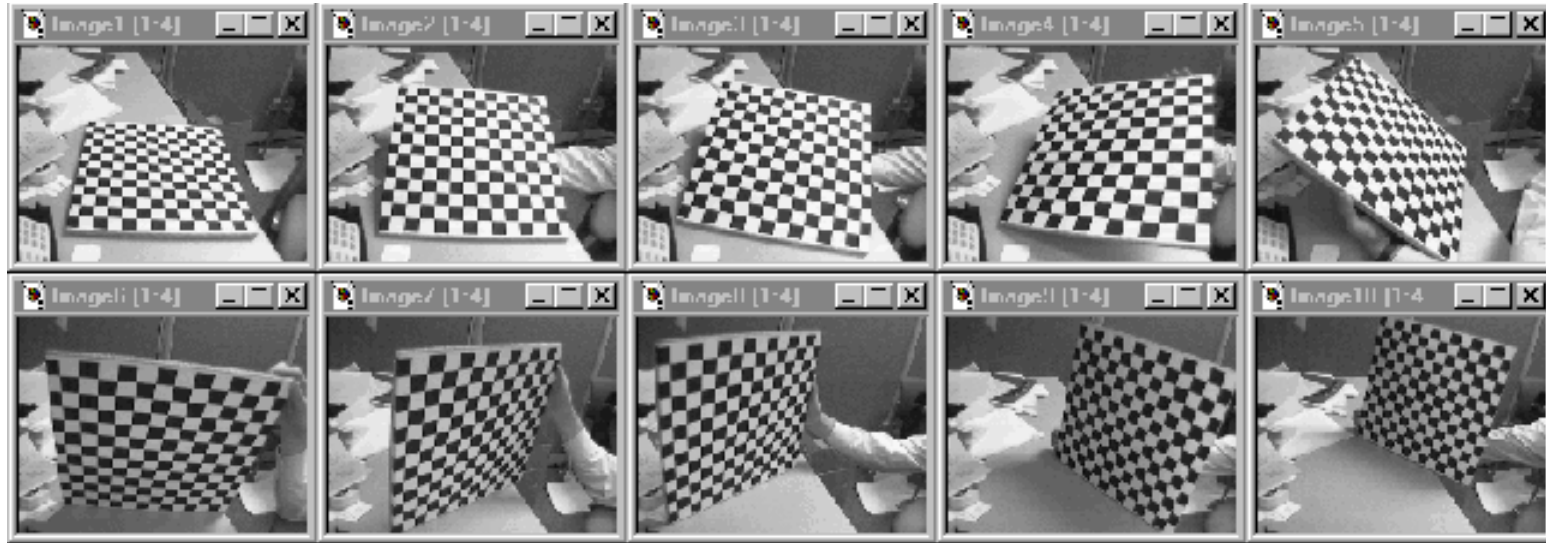
$$v_i^T K^{-T} K^{-1} v_j = 0$$

$$v_j^T K^{-T} K^{-1} v_k = 0$$

$$v_i^T K^{-T} K^{-1} v_k = 0$$

- 3 finite vanishing points: get  $f$ ,  $u_0$ ,  $v_0$

# Alternative: multi-plane calibration



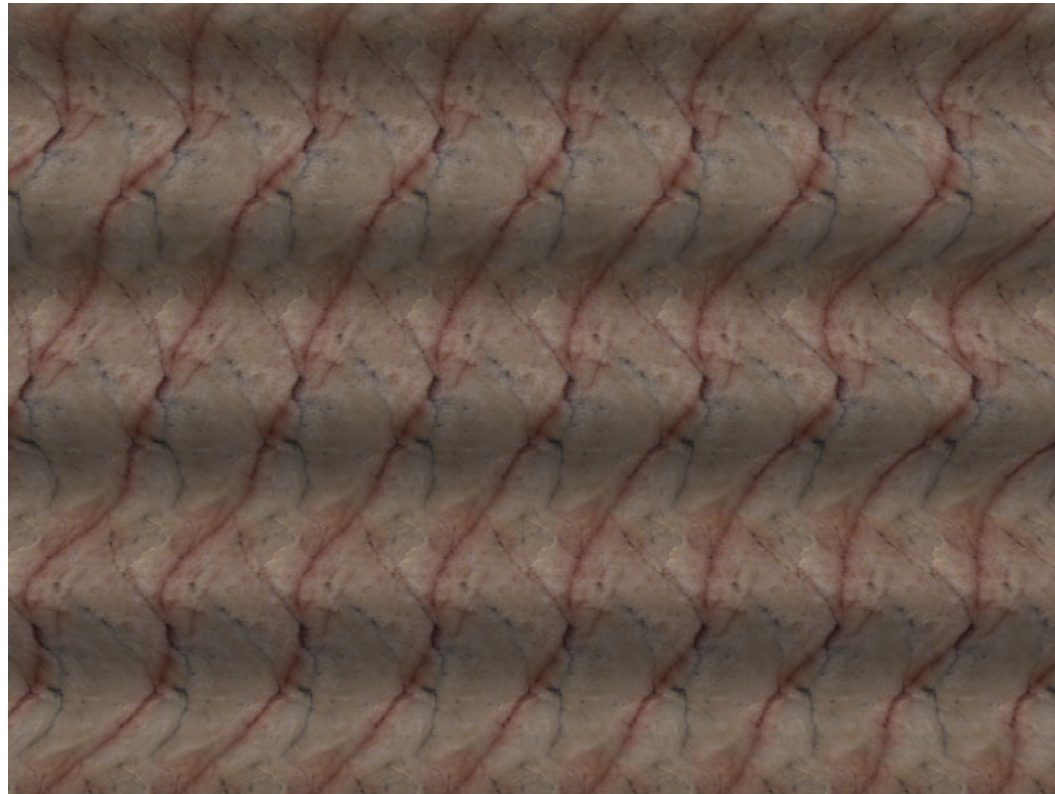
Images courtesy Jean-Yves Bouguet, Intel Corp.

## Advantage

- Only requires a plane
- Don't have to know positions/orientations
- Good code available online! (including in OpenCV)
  - Matlab version by Jean-Yves Bouguet:  
[http://www.vision.caltech.edu/bouguetj/calib\\_doc/index.html](http://www.vision.caltech.edu/bouguetj/calib_doc/index.html)
  - Zhengyou Zhang's web site: <http://research.microsoft.com/~zhang/Calib/>



# Stereo



Single image stereogram, by [Niklas Een](#)

# Announcements

- Readings
  - Szeliski, Chapter 7.1 and 7.2
  - [Fundamental matrix song](#)

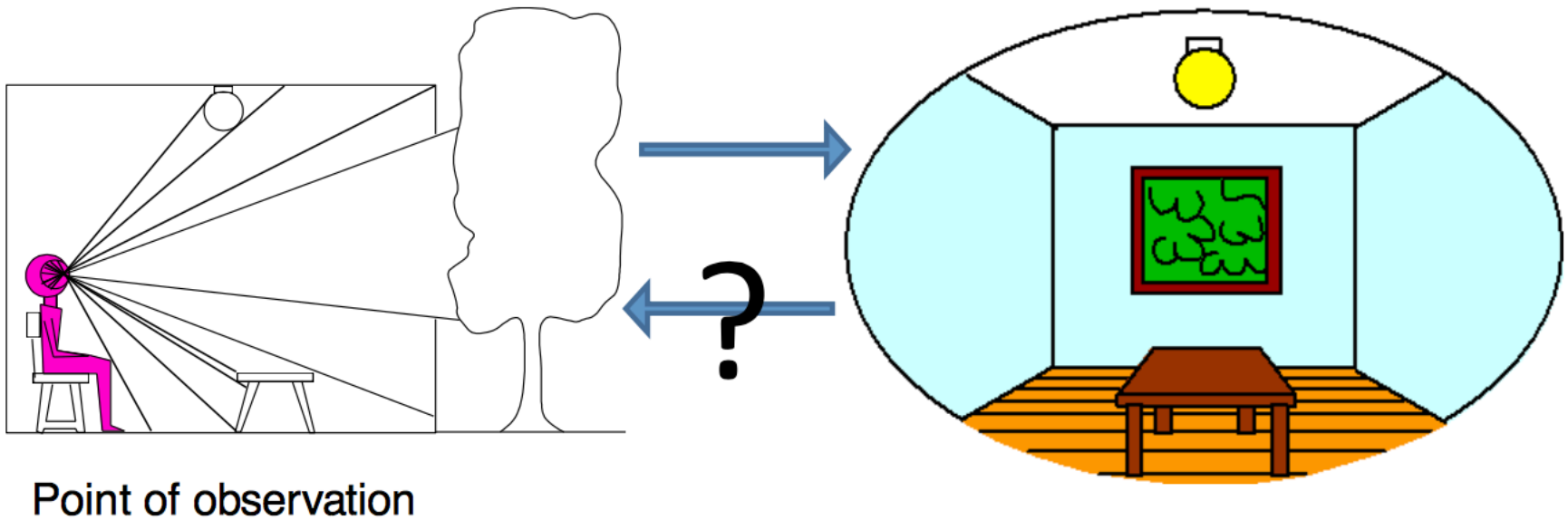
# Next two lectures

- Intro to stereo vision
- Epipolar geometry
- Fundamental matrix

# Recovering 3D from images

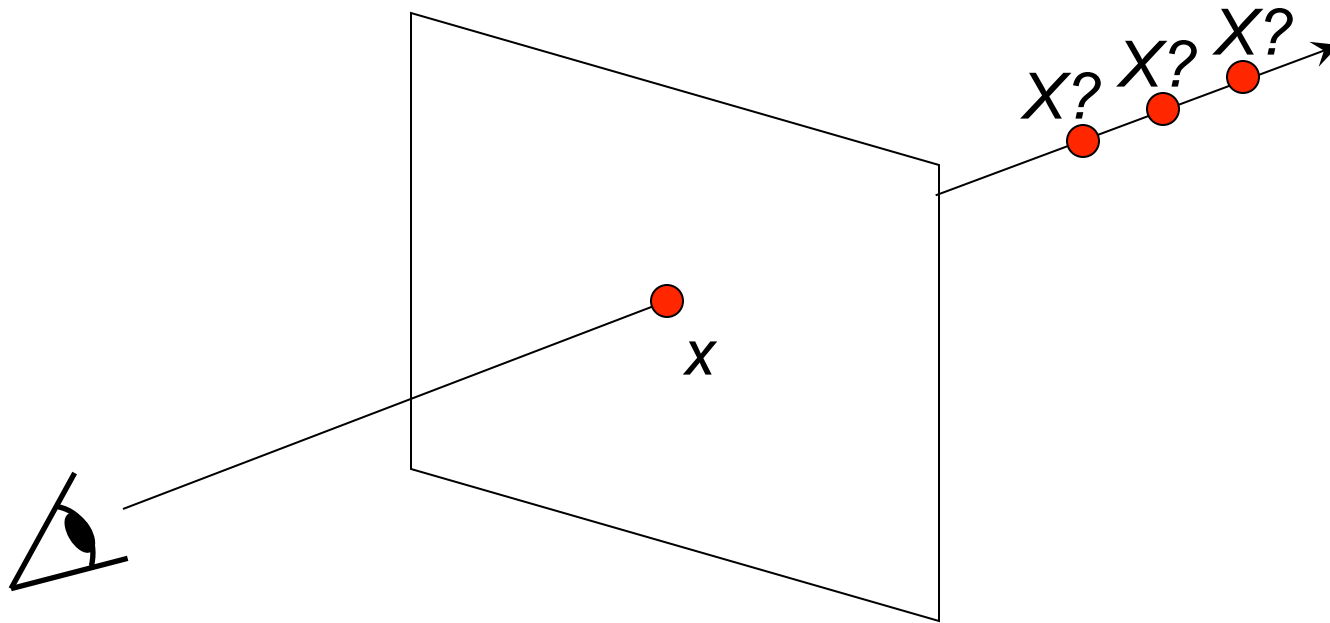
*Real 3D world*

*2D image*



# Projective Geometry

- Recovery of structure from one image is inherently ambiguous

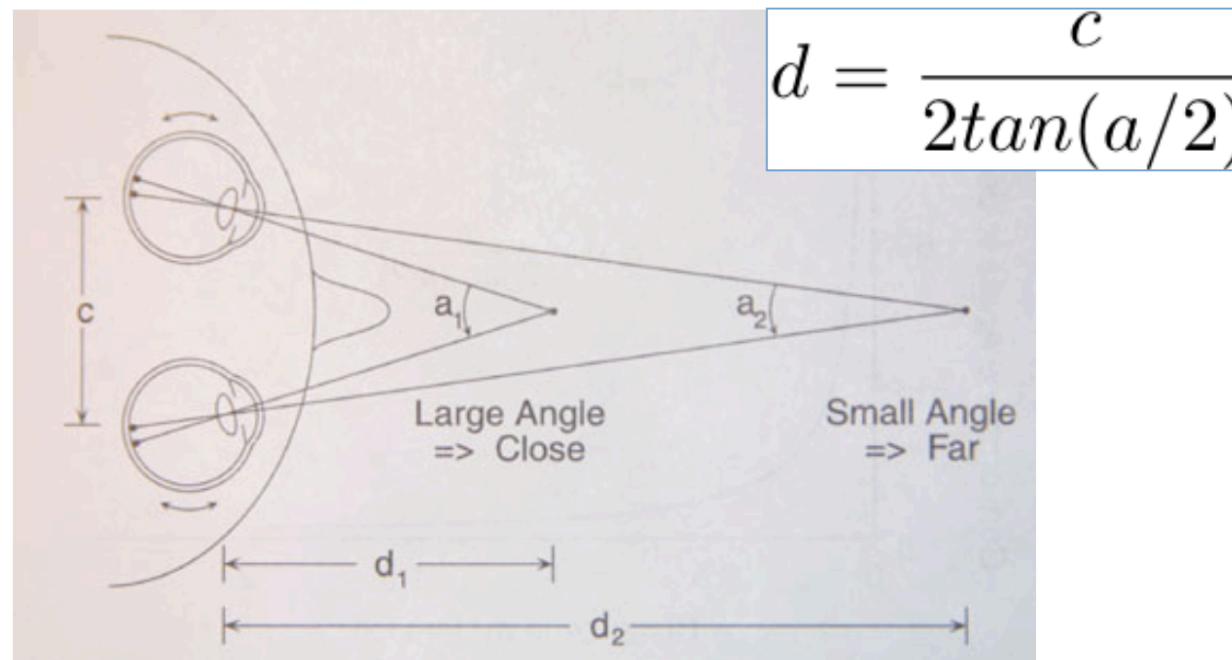


# Visual Cues for 3D

- Shading
  - Texture
  - Shadows
  - Focus
  - Motion
- 
- Shape from X: X is shading, texture, motion, ....

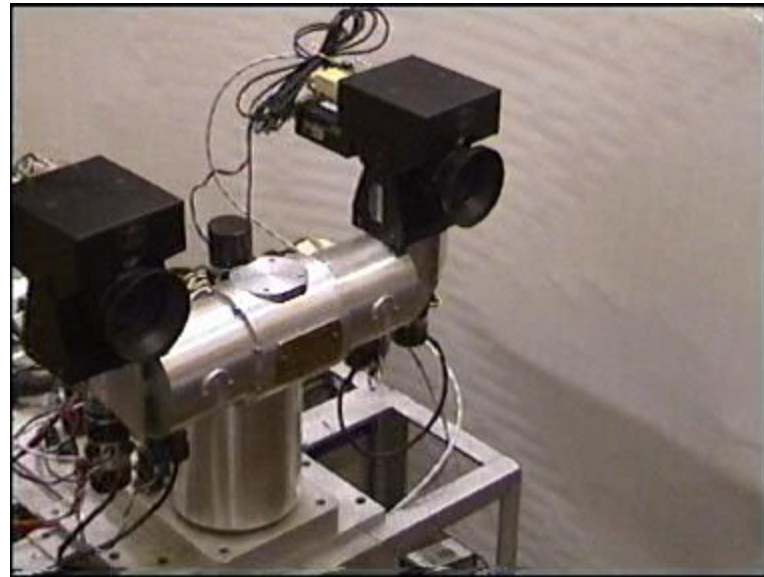
# Stereo Reconstruction

- Shape from 2 (or more) images
- Motivated by human visual system

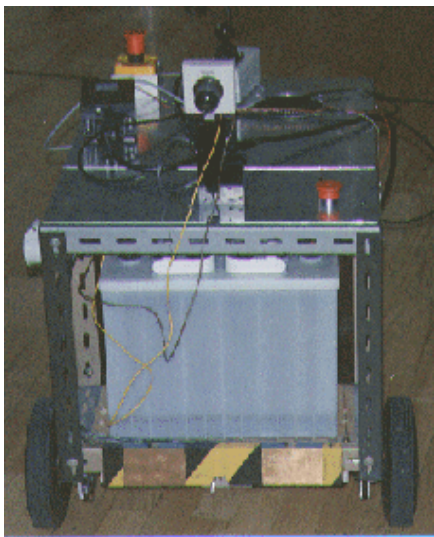


*Human performance: up to 6-8 feet*

Stereo head



Camera on a mobile vehicle

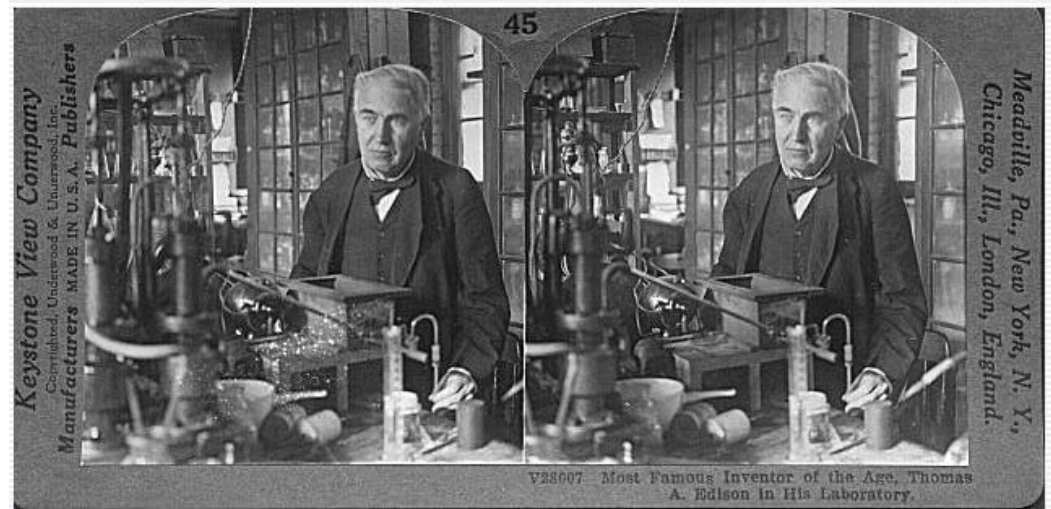


(COURTESY SONY)



# Stereograms

- Invented by Sir Charles Wheatstone, 1838

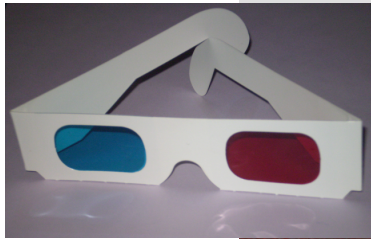


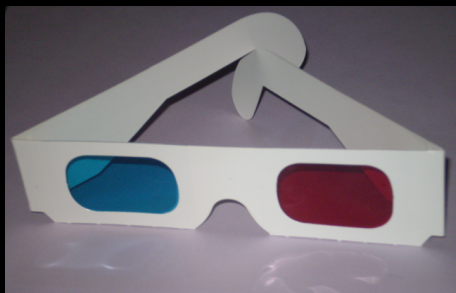


*Underwood & Underwood, Publishers.  
New York, London, Toronto-Canada, Ottawa-Kansas.*

*Works and Littleton, Wash. D.C.  
H. H. Washington, D.C.*

The Stereograph as an Educator—Underwood Patent Extension Cabinet in a home Library. (2)  
Copyright 1901 by Underwood & Underwood.

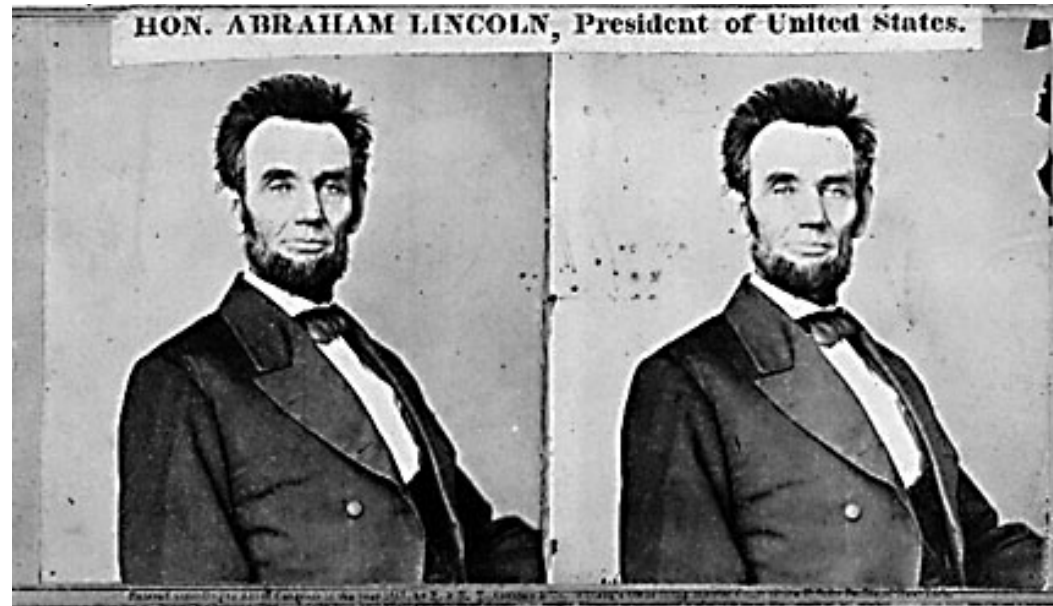




Mark Twain at Pool Table", no date, UCR Museum of Photography



# Stereo

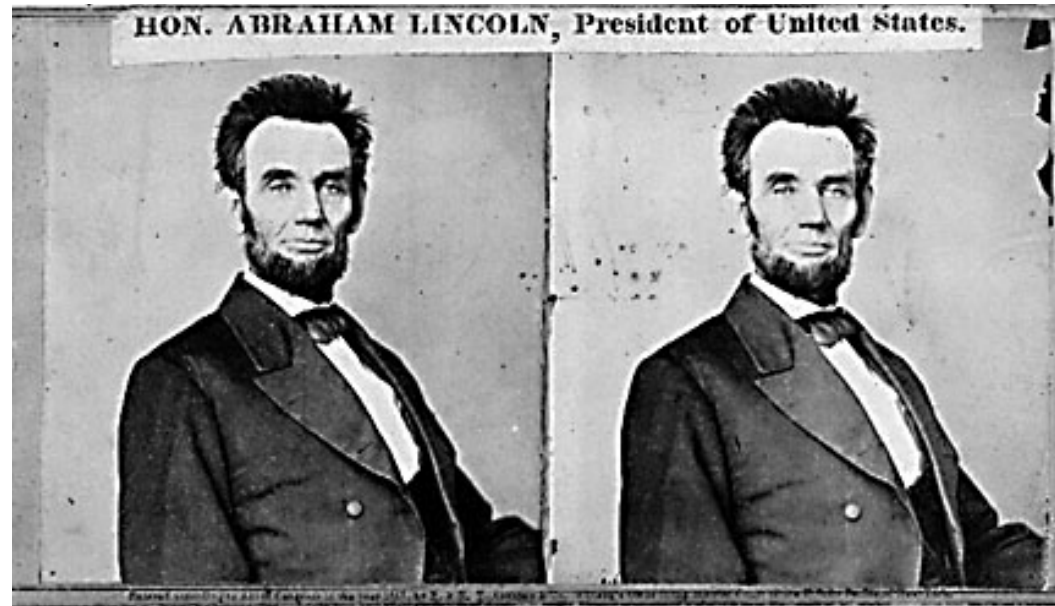


- Given two images from different viewpoints
  - How can we compute the depth of each point in the image?
  - Based on *how much each pixel moves* between the two images



- <https://github.com/404>

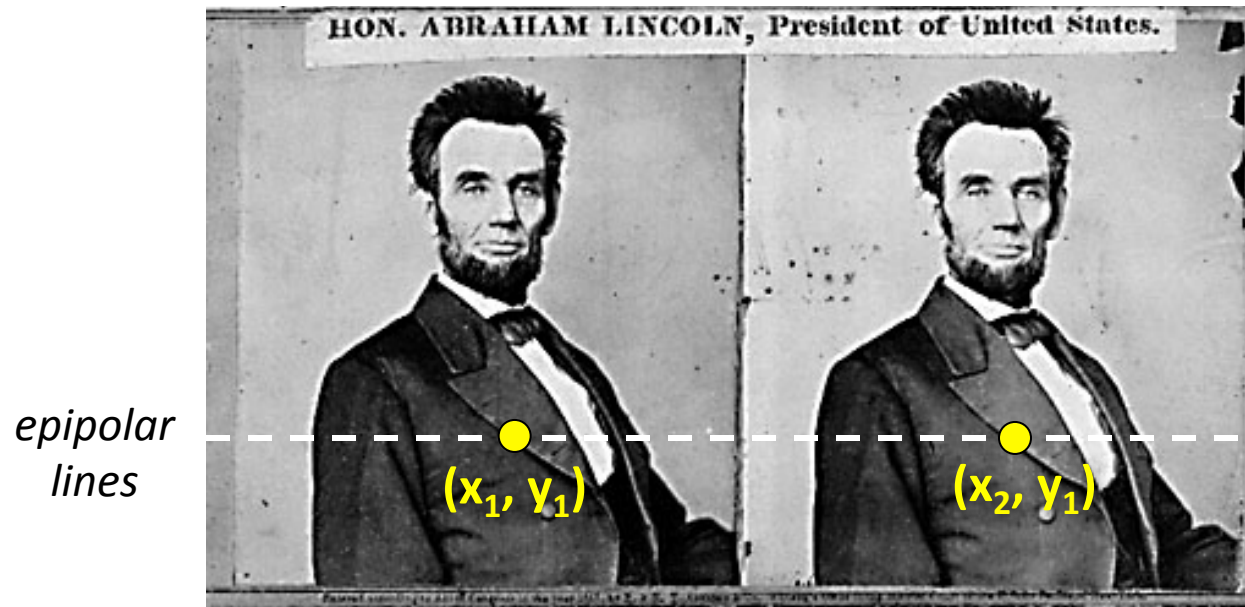
# Stereo



Two images captured by a purely horizontal translating camera  
(*rectified* stereo pair)



# Epipolar geometry

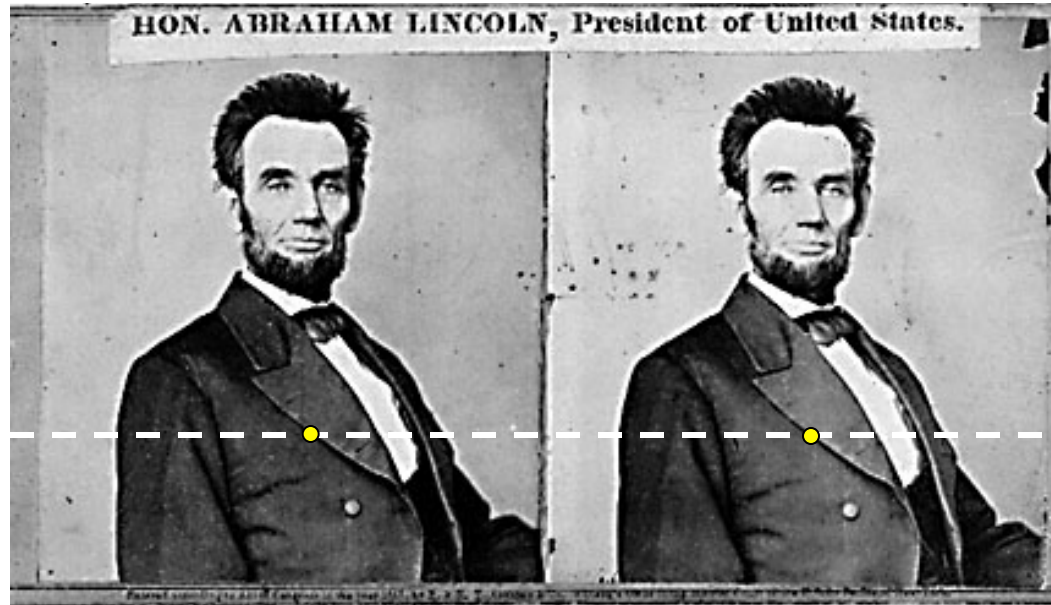


Two images captured by a purely horizontal translating camera  
(*rectified* stereo pair)

$x_2 - x_1 =$  the *disparity* of pixel  $(x_1, y_1)$   
Disparity is inversely proportional to depth

# Disparity and Depth

# Your basic stereo algorithm



For each epipolar line

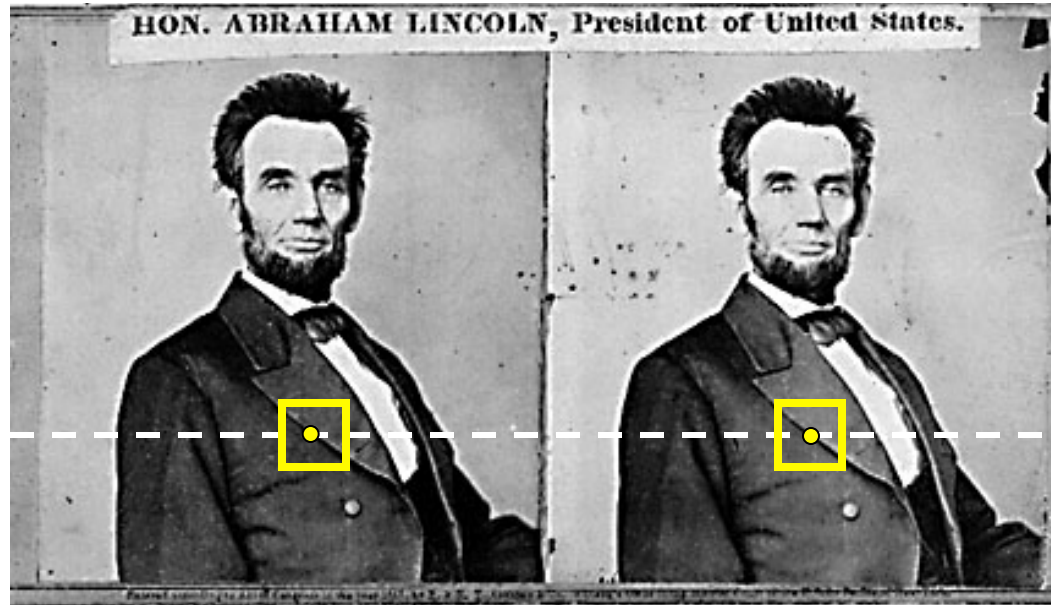
For each pixel in the left image

- compare with every pixel on same epipolar line in right image
- pick pixel with minimum match cost
- Compute disparity

# Stereo matching algorithms

- Match Pixels in conjugate epipolar Lines
  - Assume brightness constancy
  - This is a tough problem
  - Numerous approaches
    - A good survey and evaluation: <http://www.middlebury.edu/stereo/>

# Your basic stereo algorithm



For each epipolar line

For each pixel in the left image

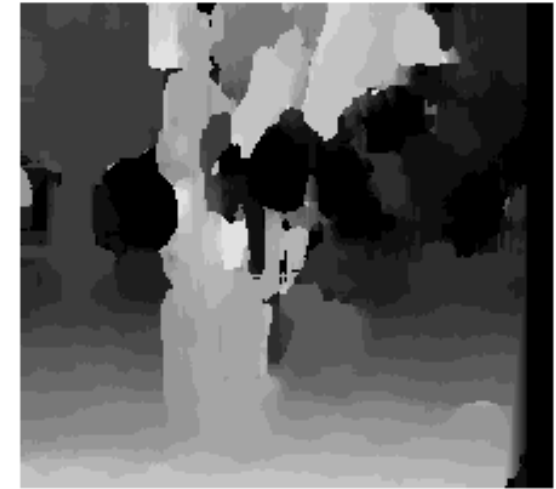
- compare with every pixel on same epipolar line in right image
- pick pixel with minimum match cost

Improvement: match *windows*

# Window size



$W = 3$



$W = 20$

## Effect of window size

- Smaller window
  - + more detail
  - more noise
- Larger window
  - + less noise
  - less detail

## Better results with *adaptive window*

- T. Kanade and M. Okutomi, [A Stereo Matching Algorithm with an Adaptive Window: Theory and Experiment](#), Proc. International Conference on Robotics and Automation, 1991.
- D. Scharstein and R. Szeliski. [Stereo matching with nonlinear diffusion](#). International Journal of Computer Vision, 28(2):155-174, July 1998

# Stereo results

- Data from University of Tsukuba
- Similar results on other images without ground truth

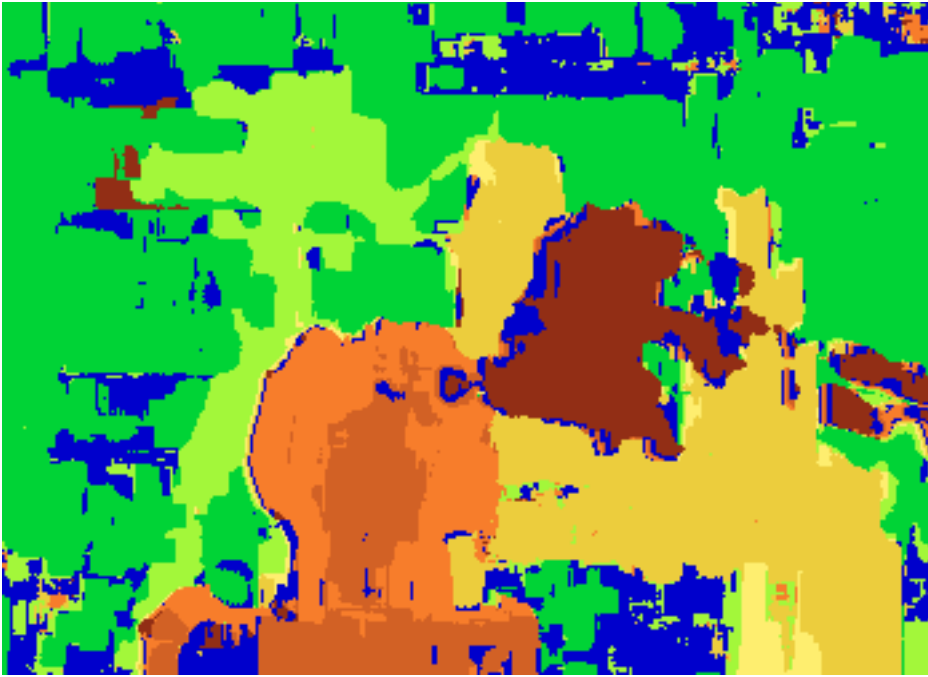


Scene



Ground truth

# Results with window search



Window-based matching  
(best window size)



Ground truth



# Better methods exist...



State of the art method

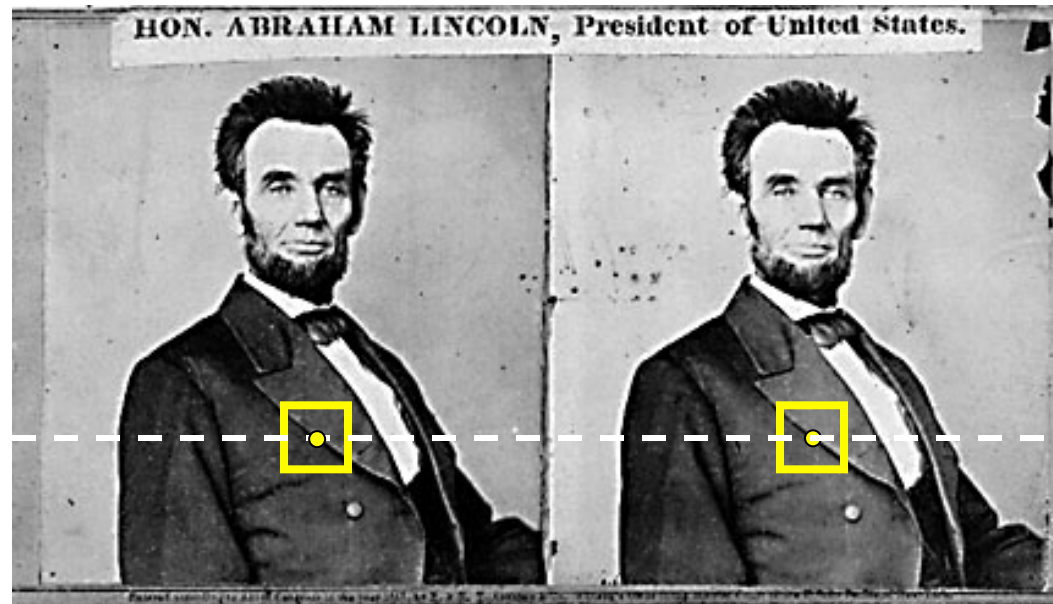
Boykov et al., [Fast Approximate Energy Minimization via Graph Cuts](#),  
International Conference on Computer Vision, September 1999.



Ground truth

For the latest and greatest: <http://www.middlebury.edu/stereo/>

# Back to stereo



- Where do epipolar lines come from?

- Have a good break!