

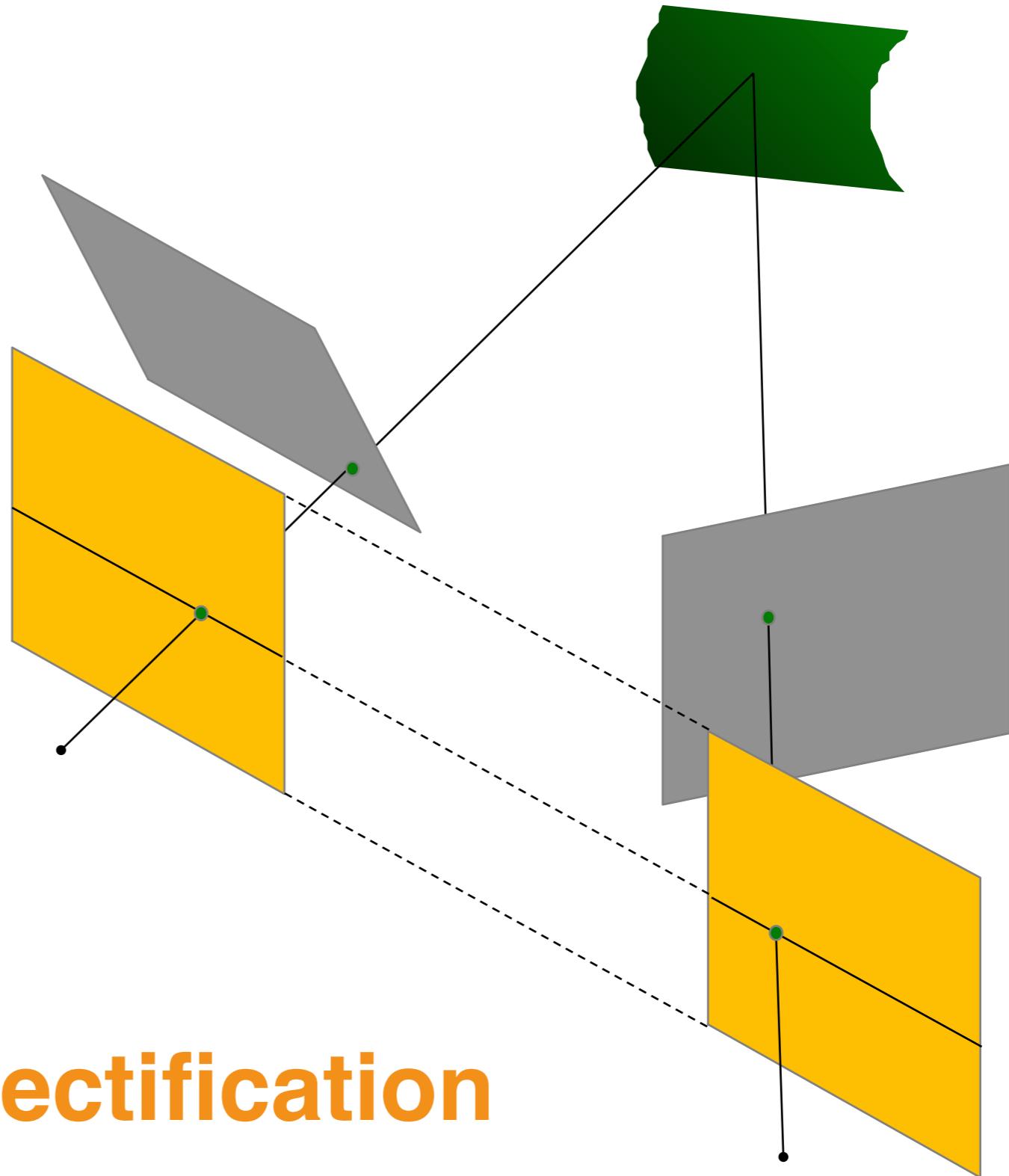


# Stereo Matching

16-385 Computer Vision (Kris Kitani)  
**Carnegie Mellon University**

## *What is stereo rectification?*

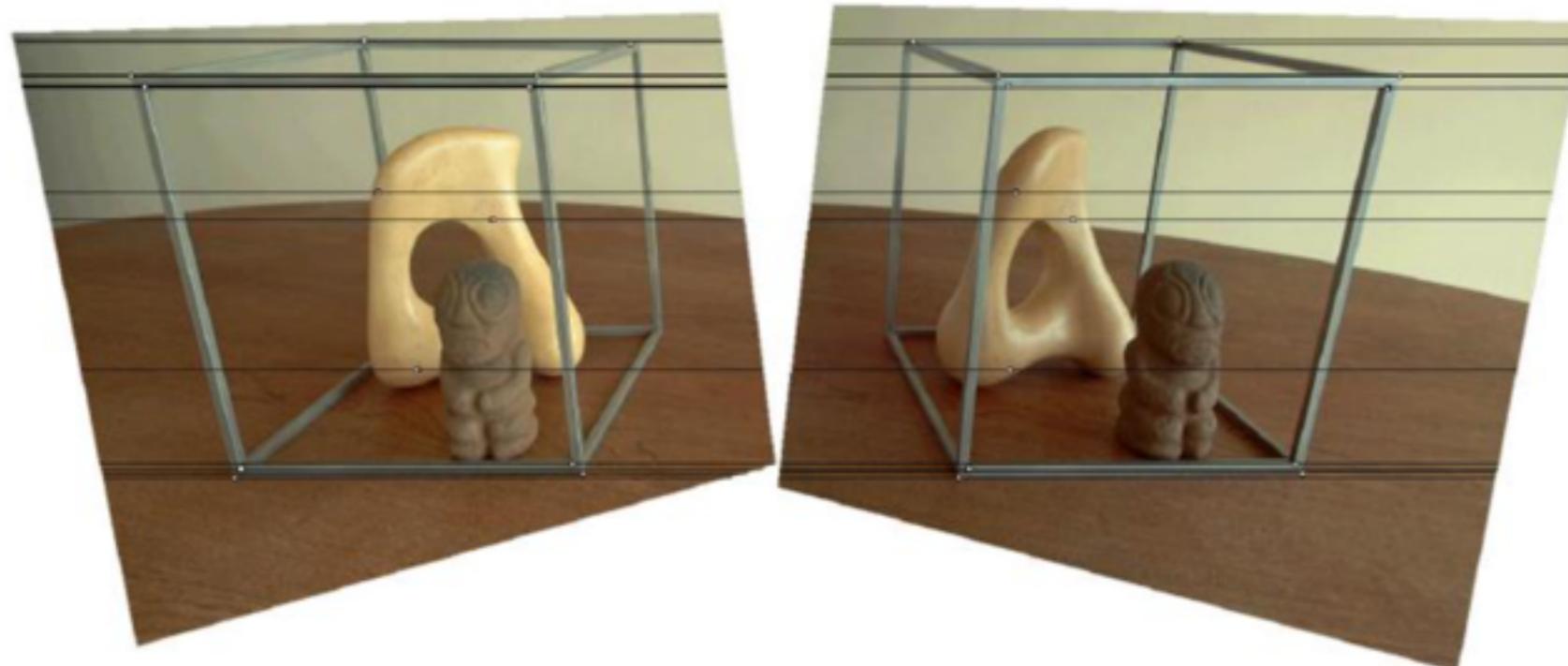
Reproject image planes onto a common plane parallel to the line between camera centers



**Recall: Stereo Rectification**



What can we do after  
rectification?

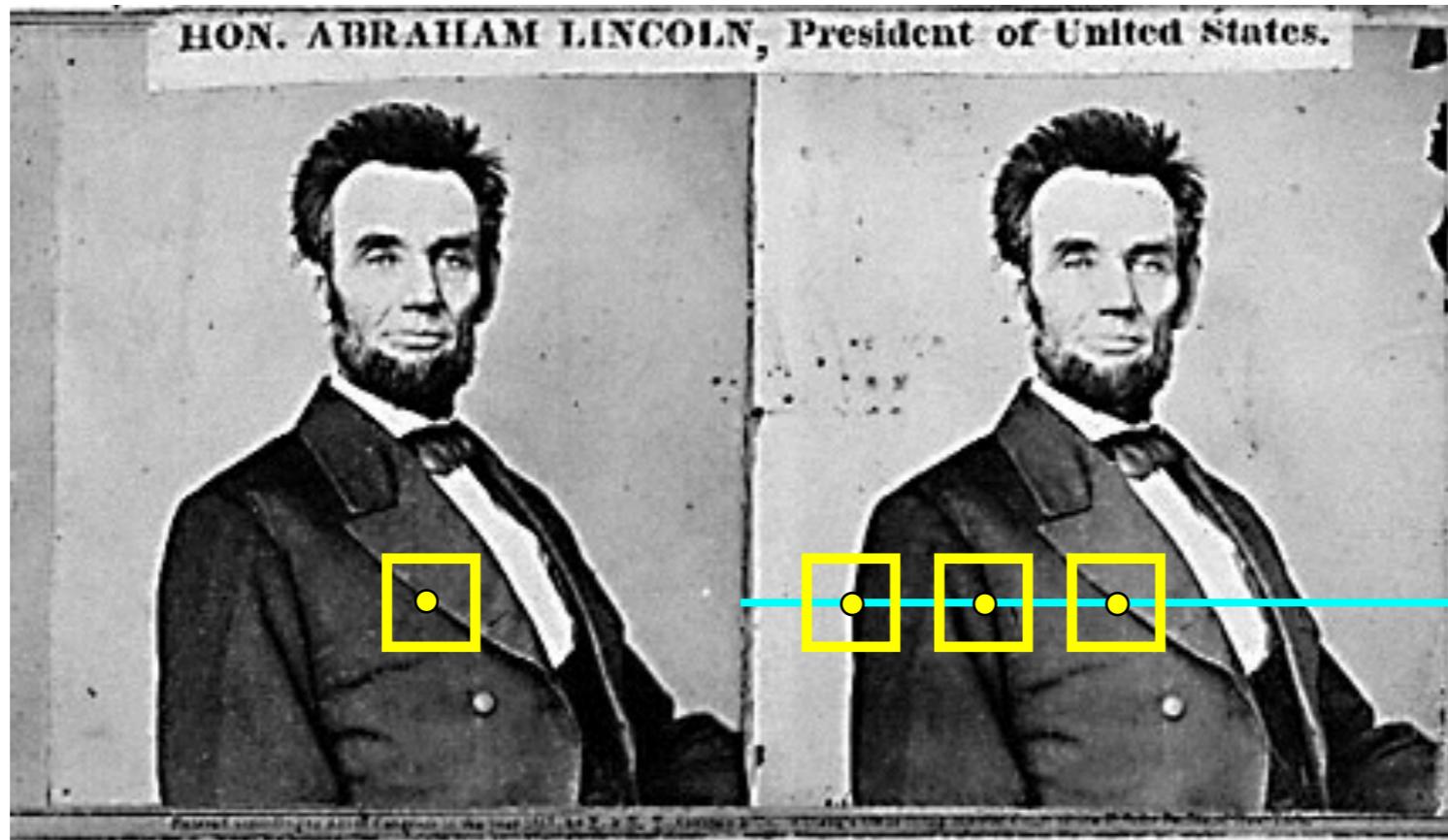




Depth Estimation via Stereo Matching



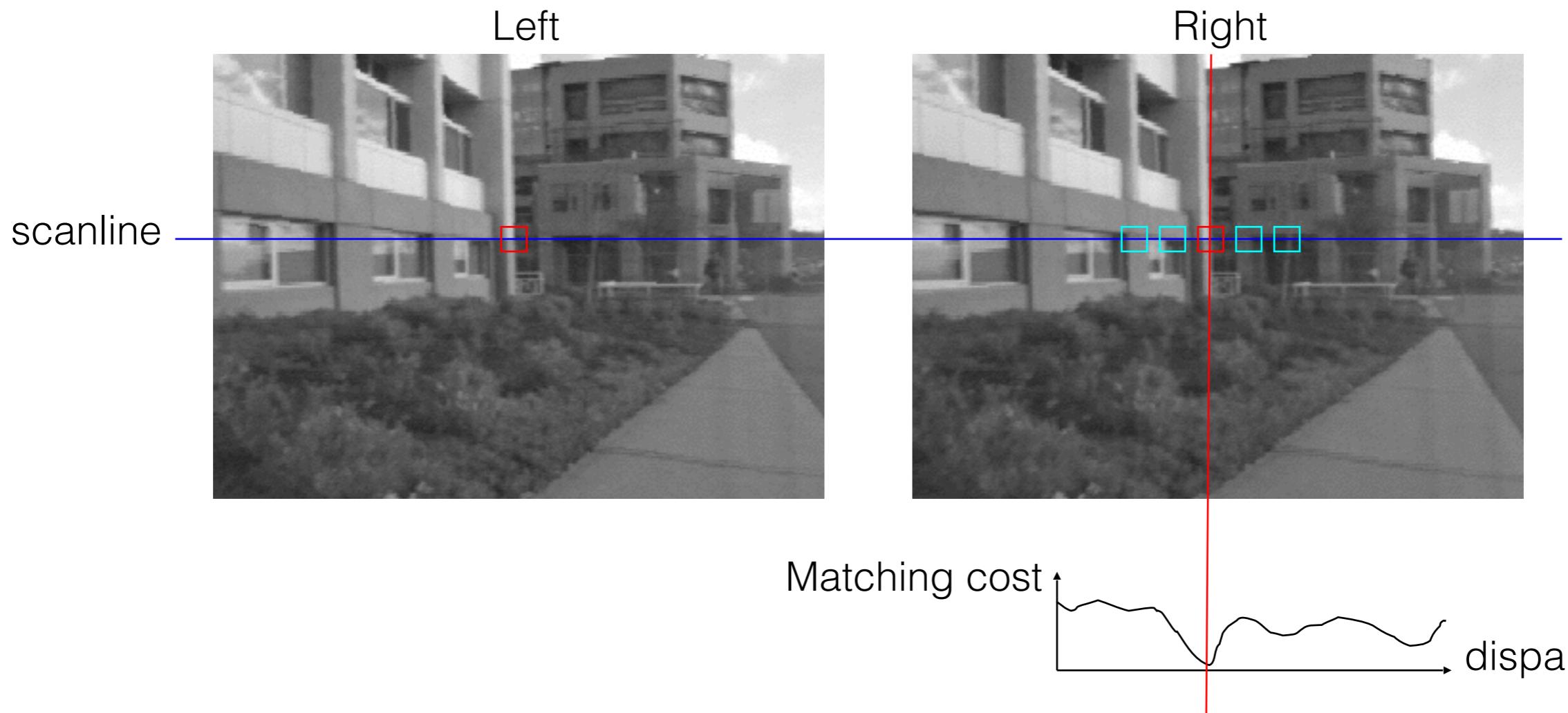




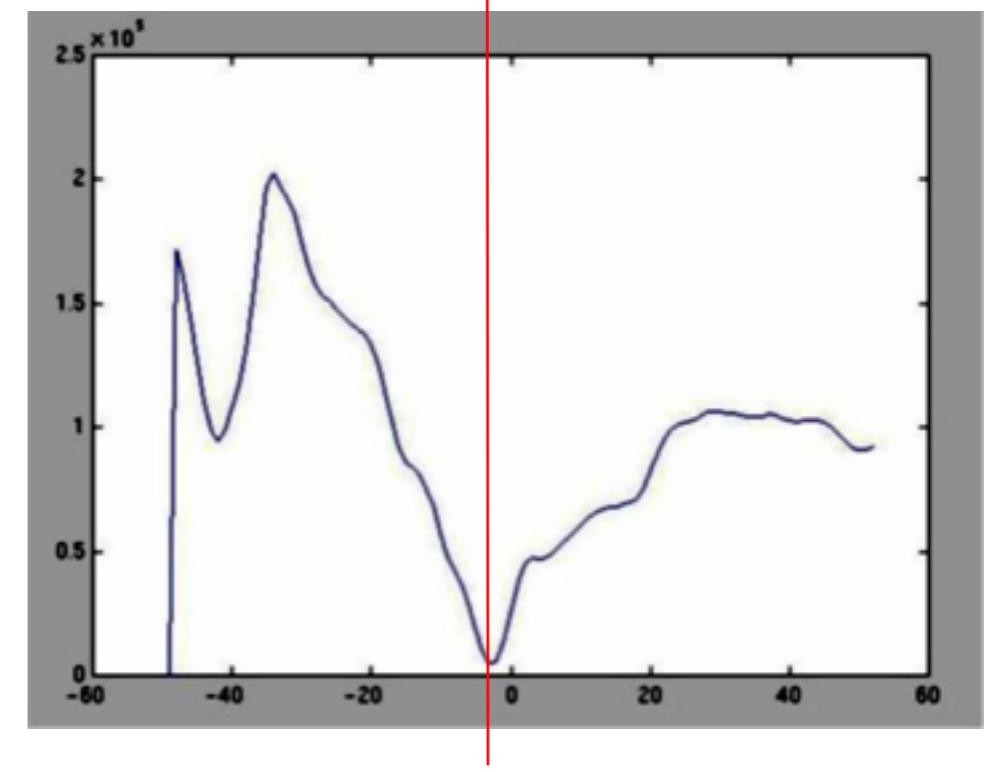
1. Rectify images  
(make epipolar lines horizontal)
2. For each pixel
  - a. Find epipolar line
  - b. Scan line for best match
  - c. Compute depth from disparity

$$Z = \frac{bf}{d}$$

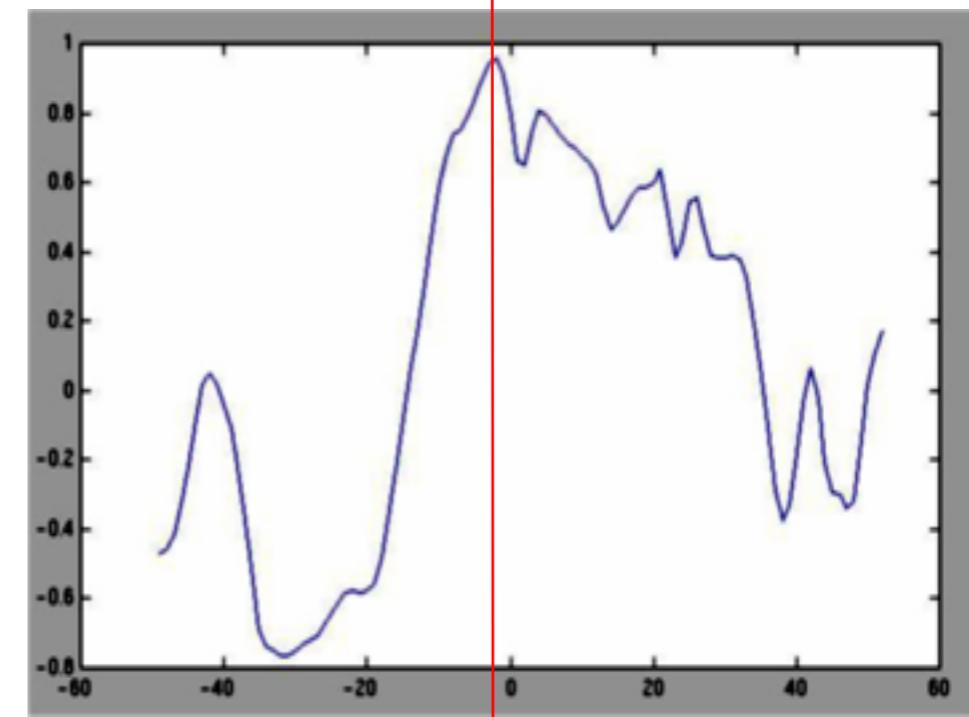
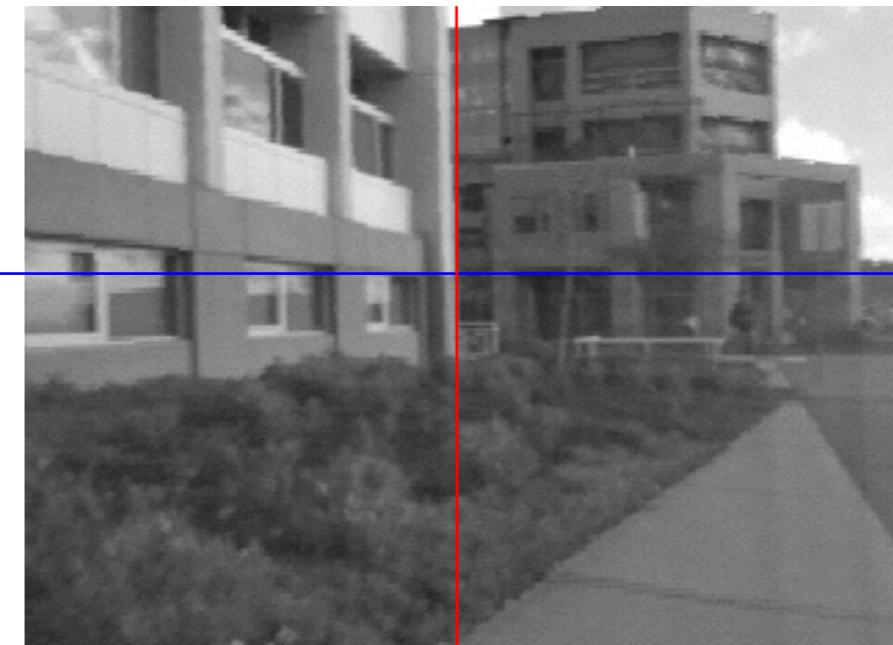
# Stereo Block Matching



- Slide a window along the epipolar line and compare contents of that window with the reference window in the left image
- Matching cost: SSD or normalized correlation



SSD



Normalized cross-correlation

# Similarity Measure

# Formula

Sum of Absolute Differences (SAD)

$$\sum_{(i,j) \in W} |I_1(i,j) - I_2(x+i, y+j)|$$

Sum of Squared Differences (SSD)

$$\sum_{(i,j) \in W} (I_1(i,j) - I_2(x+i, y+j))^2$$

Zero-mean SAD

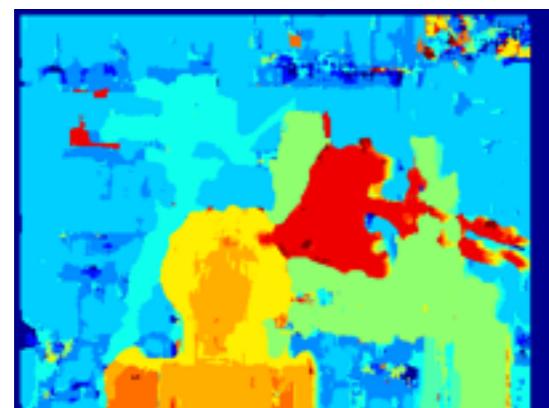
$$\sum_{(i,j) \in W} |I_1(i,j) - \bar{I}_1(i,j) - I_2(x+i, y+j) + \bar{I}_2(x+i, y+j)|$$

Locally scaled SAD

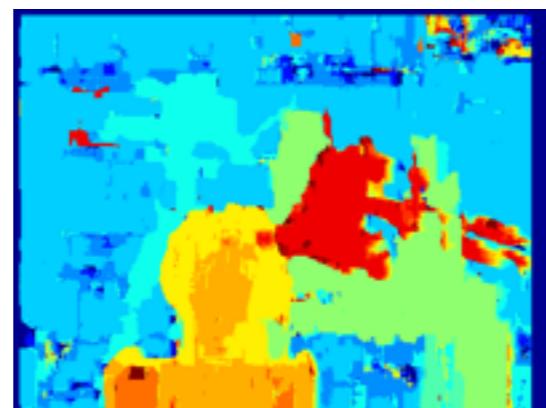
$$\sum_{(i,j) \in W} |I_1(i,j) - \frac{\bar{I}_1(i,j)}{\bar{I}_2(x+i, y+j)} I_2(x+i, y+j)|$$

Normalized Cross Correlation (NCC)

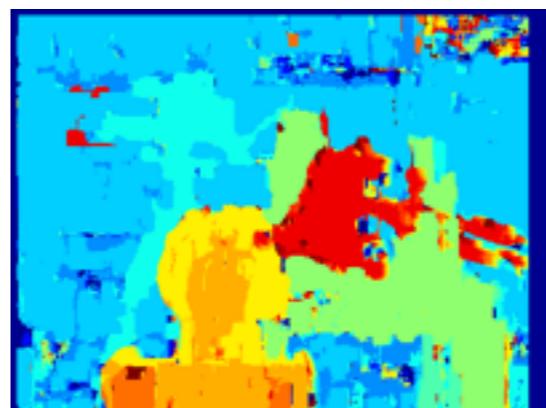
$$\frac{\sum_{(i,j) \in W} I_1(i,j) \cdot I_2(x+i, y+j)}{\sqrt{2} \sqrt{\sum_{(i,j) \in W} I_1^2(i,j) \cdot \sum_{(i,j) \in W} I_2^2(x+i, y+j)}}$$



SAD



SSD



NCC

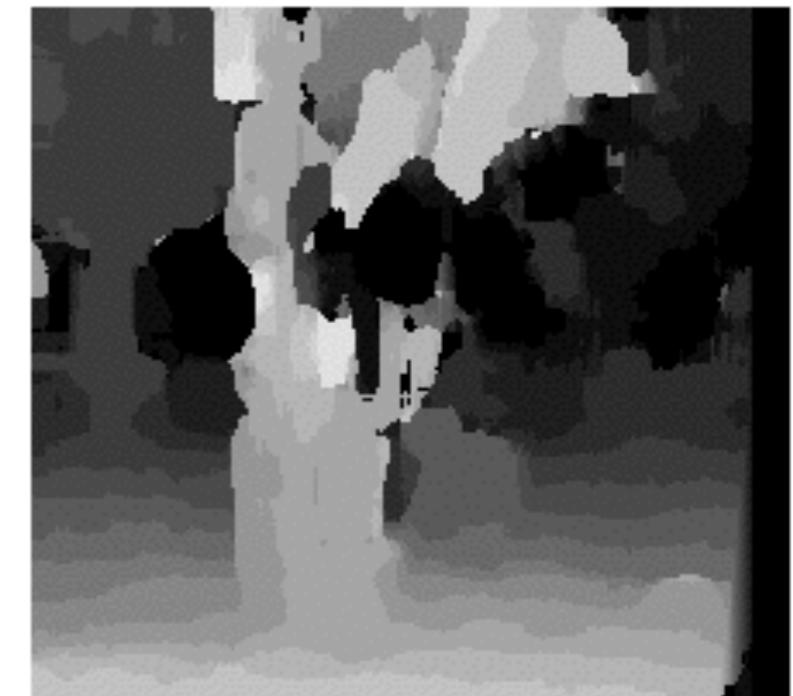


Ground truth

# Effect of window size



$W = 3$

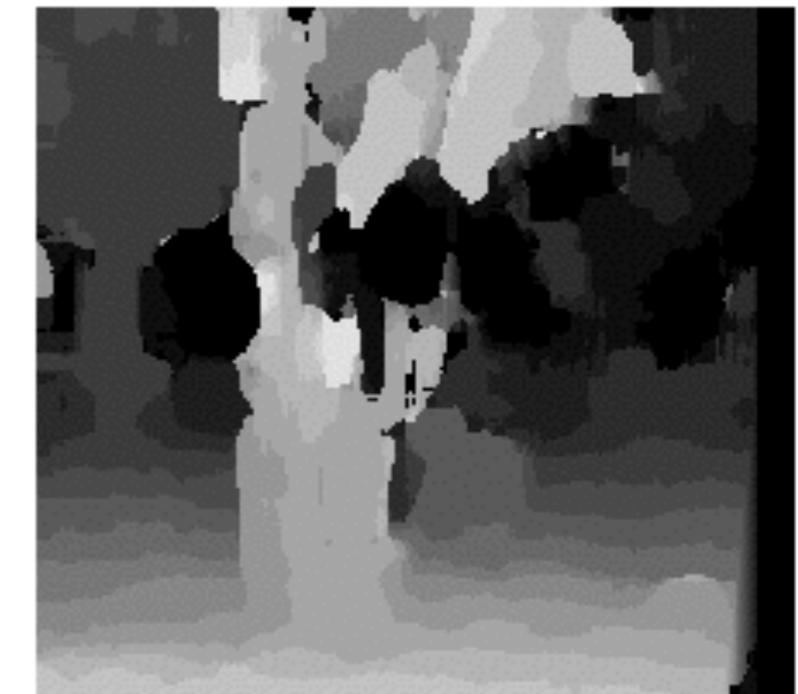


$W = 20$

# Effect of window size



$W = 3$



$W = 20$

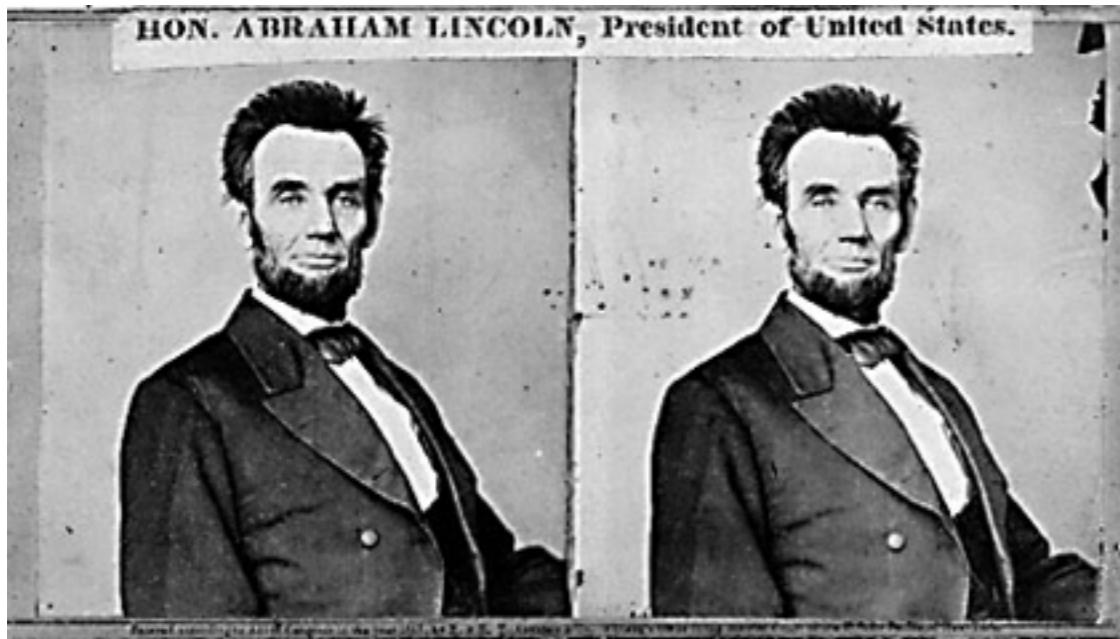
## Smaller window

- + More detail
- More noise

## Larger window

- + Smoother disparity maps
- Less detail
- Fails near boundaries

## When will stereo block matching fail?



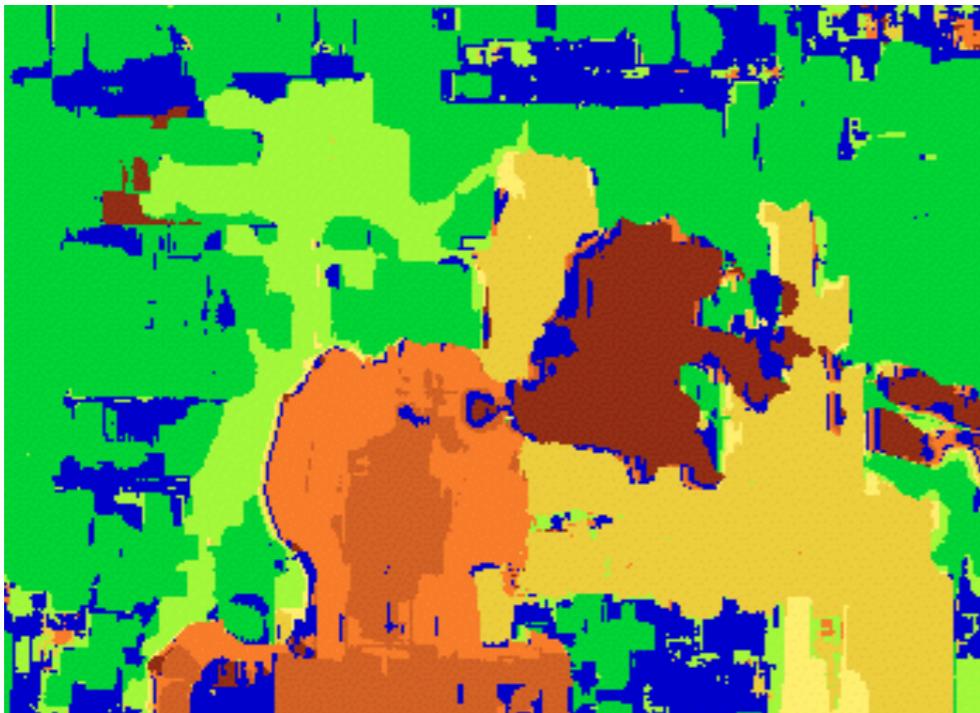
## *When will stereo block matching fail?*



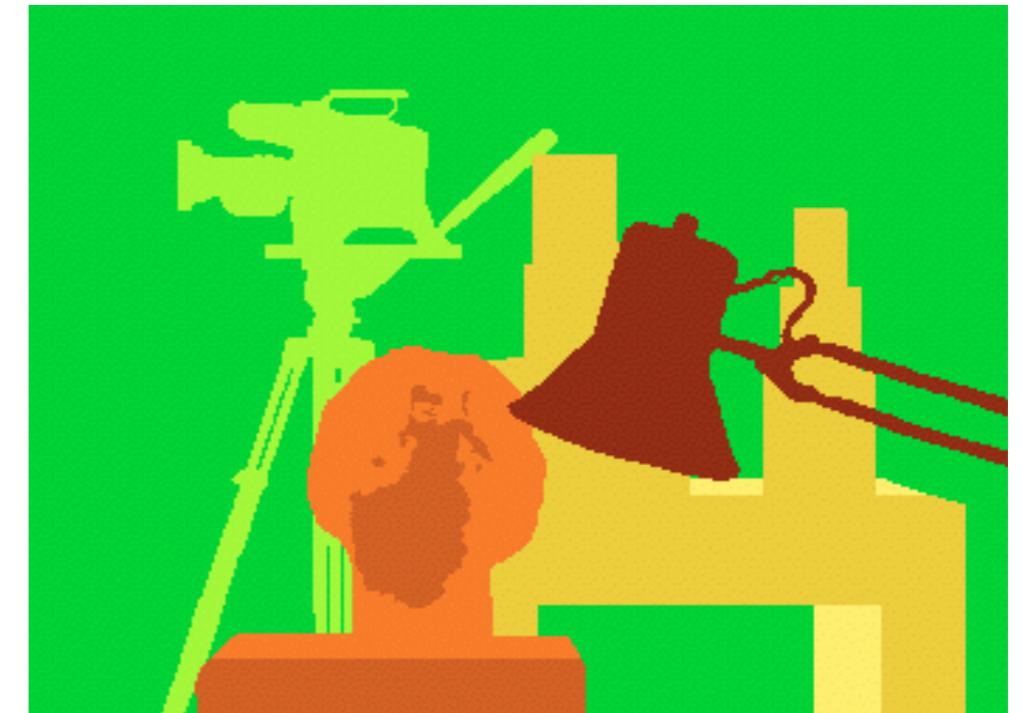
# Improving Stereo Block Matching



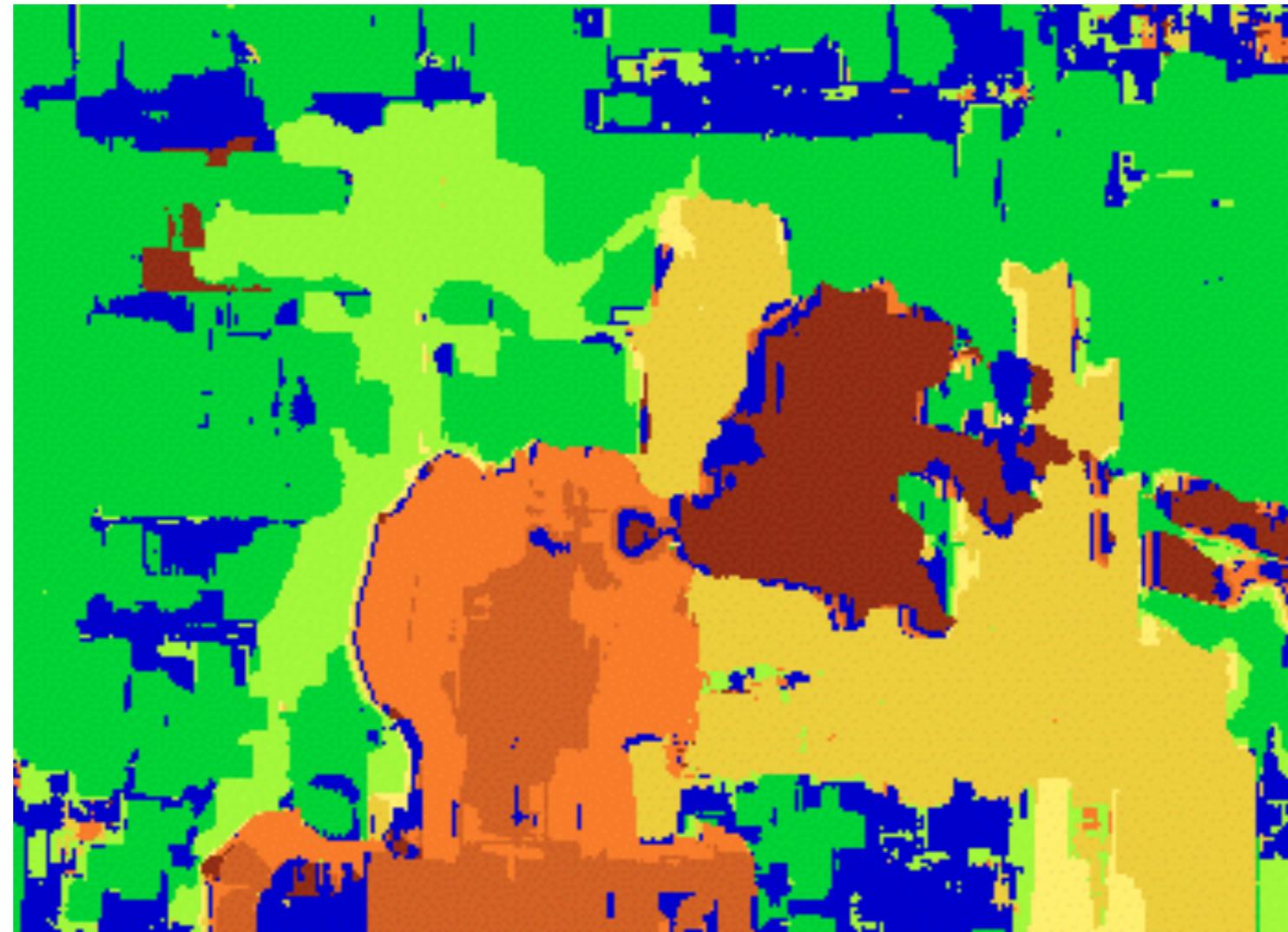
Block matching



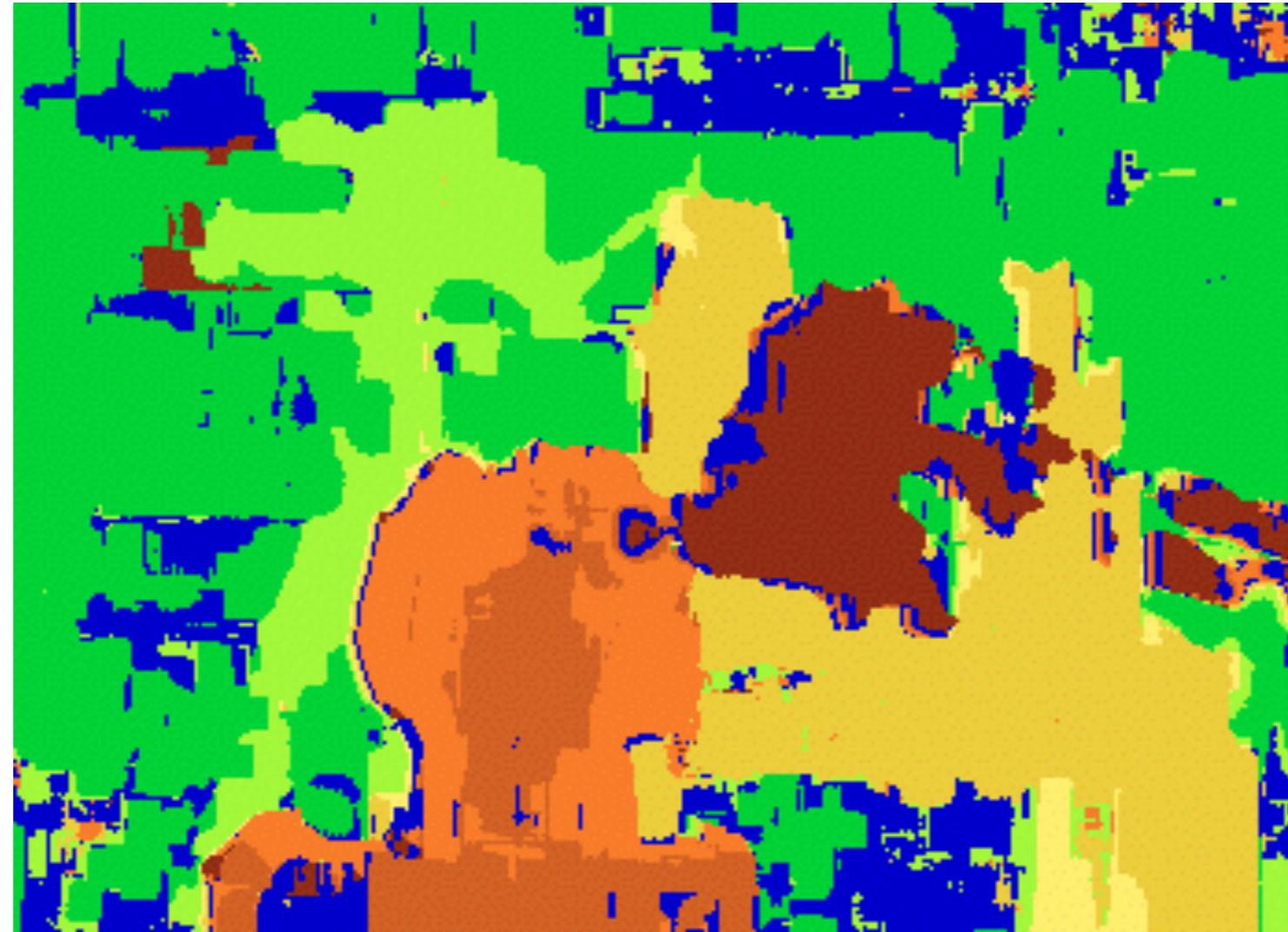
Ground truth



*What are some problems with the result?*



*How can we improve depth estimation?*



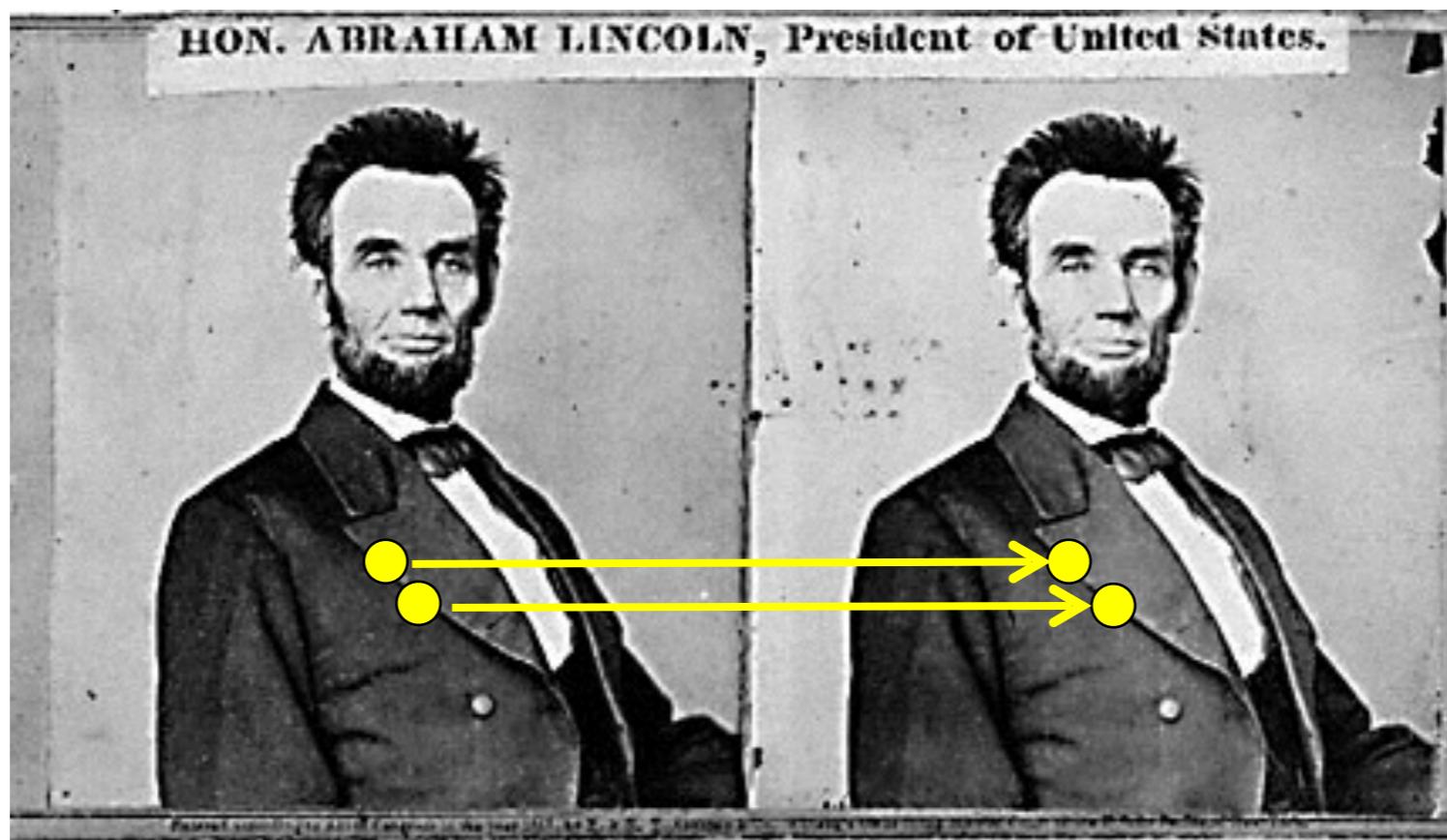
*How can we improve depth estimation?*

Too many discontinuities.  
We expect disparity values to change slowly.

Let's make an assumption:  
**depth should change smoothly**

Stereo matching as ...

# Energy Minimization



What defines a good stereo correspondence?

1. **Match quality**
  - Want each pixel to find a good match in the other image
2. **Smoothness**
  - If two pixels are adjacent, they should (usually) move about the same amount

energy function  
(for one pixel)

$$E(d) = \underbrace{E_d(d)}_{\text{data term}} + \lambda \underbrace{E_s(d)}_{\text{smoothness term}}$$

energy function  
(for one pixel)

$$E(d) = \underbrace{E_d(d)}_{\text{data term}} + \lambda \underbrace{E_s(d)}_{\text{smoothness term}}$$

Want each pixel to find a good  
match in the other image  
(block matching result)

Adjacent pixels should (usually)  
move about the same amount  
(smoothness function)

$$E(d) = E_d(d) + \lambda E_s(d)$$

$$E_d(d) = \sum_{(x,y) \in I} C(x, y, d(x, y))$$

data term

SSD distance between windows  
centered at  $I(x, y)$  and  $J(x + d(x, y), y)$

$$E(d) = E_d(d) + \lambda E_s(d)$$

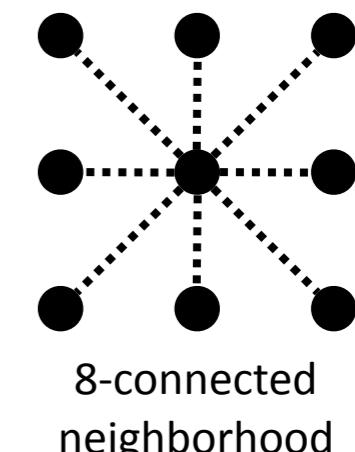
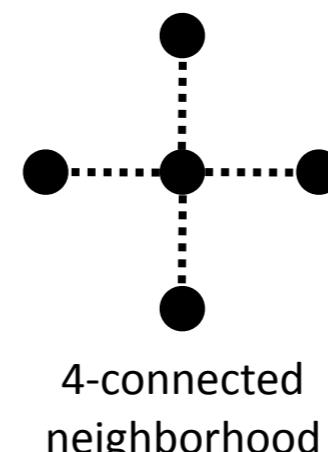
$$E_d(d) = \sum_{(x,y) \in I} C(x, y, d(x, y))$$

SSD distance between windows  
centered at  $I(x, y)$  and  $J(x + d(x, y), y)$

$$E_s(d) = \sum_{(p,q) \in \mathcal{E}} V(d_p, d_q)$$

smoothness term

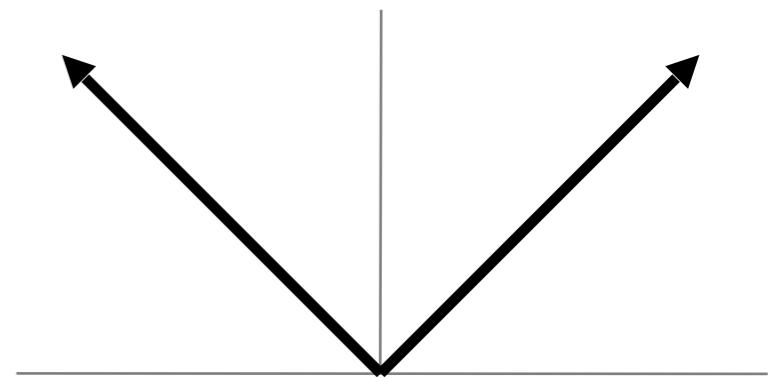
$\mathcal{E}$  : set of neighboring pixels



$$E_s(d) = \sum_{\substack{\text{smoothness term} \\ (p,q) \in \mathcal{E}}} V(d_p, d_q)$$

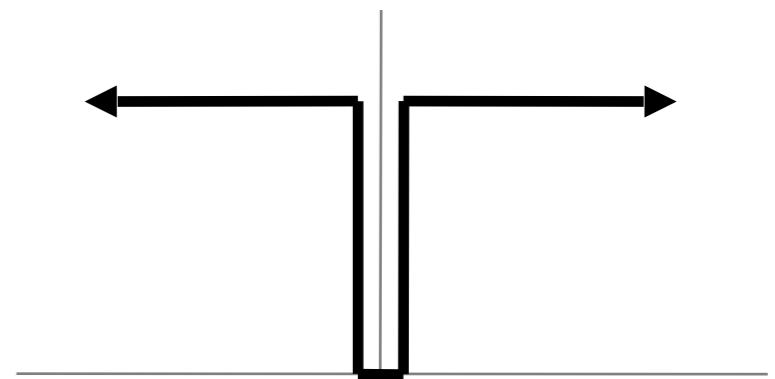
$$V(d_p, d_q) = |d_p - d_q|$$

$L_1$  distance



$$V(d_p, d_q) = \begin{cases} 0 & \text{if } d_p = d_q \\ 1 & \text{if } d_p \neq d_q \end{cases}$$

“Potts model”



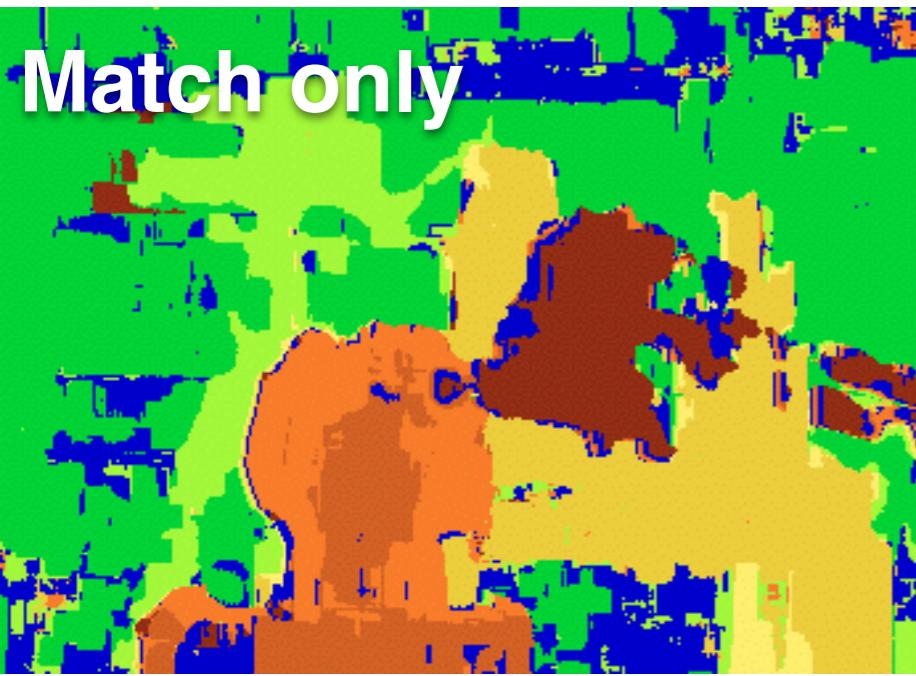
# Dynamic Programming

$$E(d) = E_d(d) + \lambda E_s(d)$$

Can minimize this independently per scanline  
using dynamic programming (DP) 

$D(x, y, d)$  : minimum cost of solution such that  $d(x, y) = d$

$$D(x, y, d) = C(x, y, d) + \min_{d'} \{D(x - 1, y, d') + \lambda |d - d'|\}$$



Y. Boykov, O. Veksler, and R. Zabih, Fast Approximate Energy Minimization via Graph Cuts, PAMI 2001