7. Stereo Reconstruction

Reconstruction from 2 images

Stereo Reonstruction

Fuse a calibrated binocular stereo pair to produce a depth image image 1

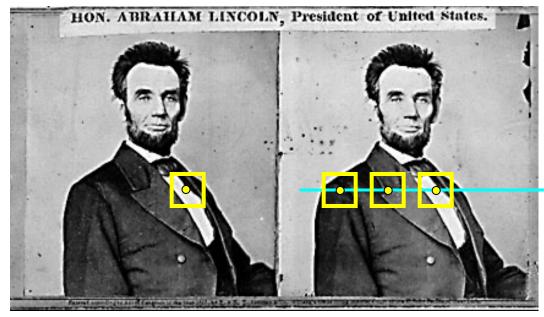




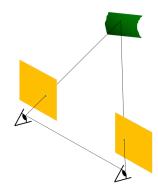
Dense depth map



Basic stereo matching algorithm



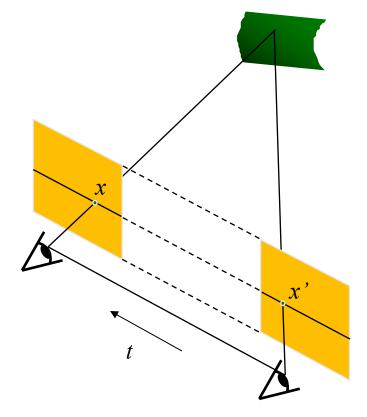
- For each pixel in the first image
 - Find corresponding epipolar line in the right image
 - Search along epipolar line and pick the best match
 - Triangulate the matches to get depth information



Simplest Case: Parallel images

- Image planes of cameras are parallel to each other and to the baseline
- Camera centers are at same height
- Focal lengths are the same
- Then, epipolar lines fall along the horizontal scan lines of the images

Simplest Case: Parallel images



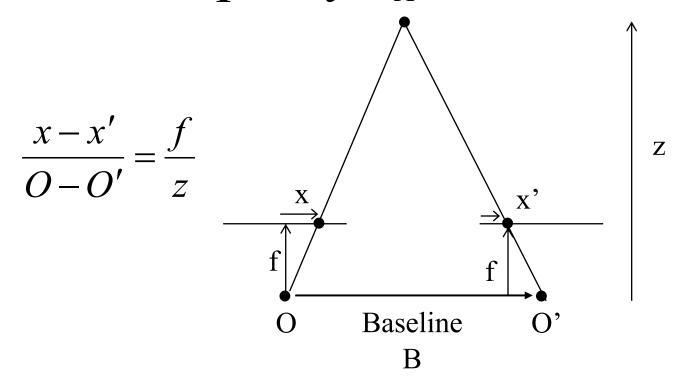
Epipolar constraint:

$$x^T E x' = 0, \quad E = t \times R$$

$$R = I \qquad t = (T, 0, 0)$$

$$E = t \times R = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -T \\ 0 & T & 0 \end{bmatrix}$$

Depth from disparity



$$disparity = x - x' = \frac{B \cdot f}{z}$$

Disparity is inversely proportional to depth.

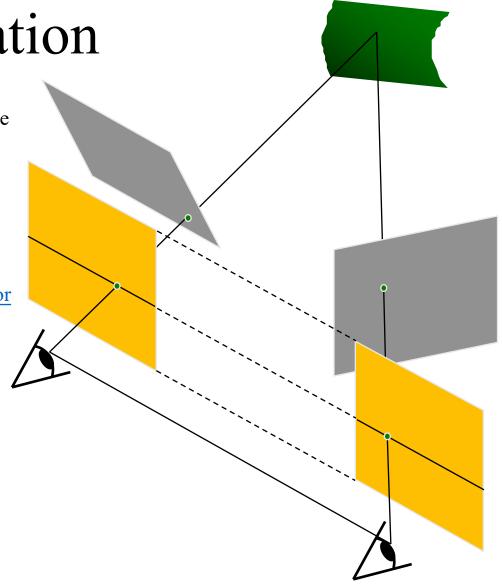
Stereo image rectification

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

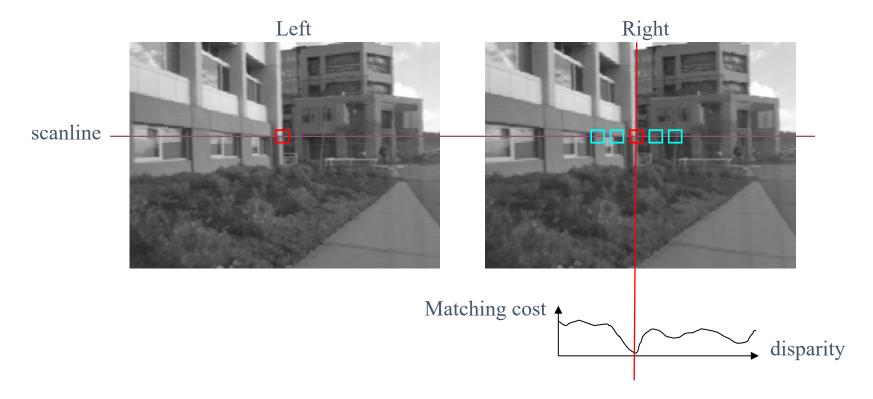
• Pixel motion is horizontal after this transformation

• Two homographies (3x3 transform), one for each input image reprojection

C. Loop and Z. Zhang. Computing Rectifying Homographies for Stereo Vision. IEEE Conf. Computer Vision and Pattern Recognition, 1999.

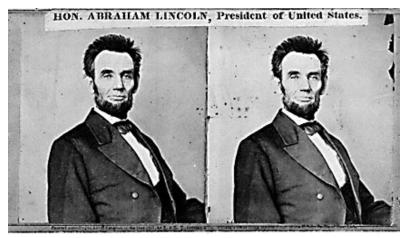


Correspondence search



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

Failures of correspondence search



Textureless surfaces



Occlusions, repetition







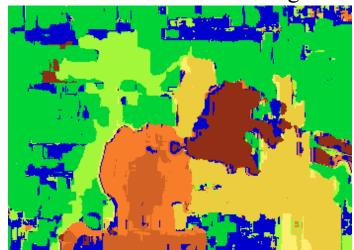
Non-Lambertian surfaces, specularities

Results with window search



Data

Window-based matching



Ground truth



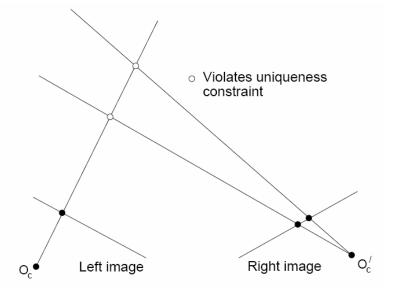
Improve window-based matching: Graph-Cut

So far, matches are independent for each point. Add constraints or priors

Uniqueness

• For any point in one image, there should be at most one matching point in the

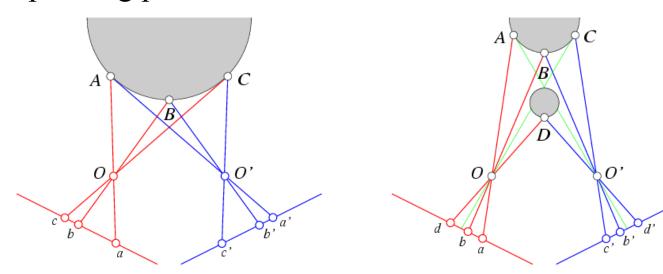
other image



Improve window-based matching

So far, matches are independent for each point. Add constraints or priors

- Ordering
 - Corresponding points should be in the same order in both views



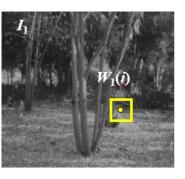
Ordering constraint doesn't hold

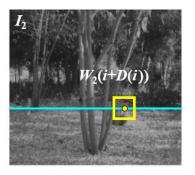
Improve window-based matching

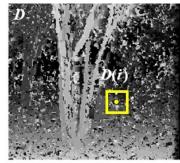
So far, matches are independent for each point. Add constraints or priors

• Smoothness: We expect disparity values to change slowly (for the

most part)





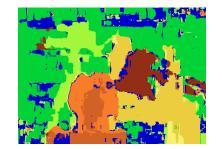


$$E = E_{\text{data}}(D; I_1, I_2) + \beta E_{\text{smooth}}(D)$$

$$E_{\text{data}} = \sum_{i} \left(W_1(i) - W_2(i + D(i)) \right)^2 \qquad E_{\text{smooth}} = \sum_{\text{neighbors } i,j} \left\| D(i) - D(j) \right\|^2$$

• Energy functions of this form can be minimized using *graph cuts*

Improve window-based matching



- Before:
- Unique, Ordering and Smoothness (Added using Graph-cut)

