## **Binocular stereo**

 Given a calibrated binocular stereo pair, produce a depth image

image 1

image 2





#### Dense depth map





## 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



## 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



## Essential matrix for parallel images



Epipolar constraint:  $x^{T} E x' = 0, \quad E = [t_{\times}]R$ 

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

$$E = [t_{\times}]R = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -T \\ 0 & T & 0 \end{bmatrix}$$



## Essential matrix for parallel images



The y-coordinates of corresponding points are the same!

# Depth from disparity



Disparity is inversely proportional to depth!

## **Basic stereo matching algorithm**



- If necessary, rectify the two stereo images to transform epipolar lines into scanlines
- For each pixel x in the first image
  - Find corresponding epipolar scanline in the right image
  - Examine all pixels on the scanline and pick the bes match x'
  - Compute disparity x-x' and set depth(x) = 1/(x-x')

## **Correspondence** problem



 Multiple matching hypotheses satisfy the epipolar constraint, but which one is correct?

# **Correspondence** problem

- Let's make some assumptions to simplify the matching problem
  - The baseline is relatively small (compared to the depth of scene points)
  - Then most scene points are visible in both views
  - Also, matching regions are similar in appearance





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# Correspondence search with similarity constraint



- 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

# Correspondence search with similarity constraint



# Correspondence search with similarity constraint



## Effect of window size



W = 3

W = 20

- Smaller window
  - + More detail
  - More noise
- Larger window
  - + Smoother disparity maps
  - Less detail



# The similarity constraint



- Corresponding regions in two images should be similar in appearance
- ...and non-corresponding regions should be different
- When will the similarity constraint fail?



## Limitations of similarity constraint



### **Textureless surfaces**



#### Occlusions, repetition



### Non-Lambertian surfaces, specularities

## Results with window search Data



## Window-based matching

### Ground truth





## Non-local constraints

## Uniqueness

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







