

Stereo (Part 2)

Introduction to Computer Vision
CSE 152
Lecture 14

Announcements

- Midterm exam grading will be completed this week
- Homework 3 will be assigned today
- Reading:
 - Section 7.1 Triangulation
 - Section 7.2 Two-frame structure from motion
 - Section 11.1 Epipolar geometry

Two-View Geometry

Essential Matrix E

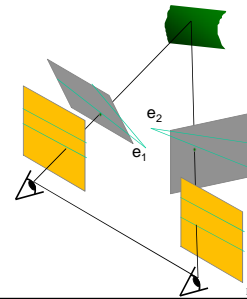
- Rank 2
- Calibrated
- Normalized coordinates
- 5 degrees of freedom
 - Camera rotation
 - Direction of camera translation
- Similarity reconstruction

Fundamental Matrix F

- Rank 2
- Uncalibrated
- Image coordinates
- 7 degrees of freedom
 - Homogeneous matrix to scale
 - $\det F = 0$
- Projective reconstruction

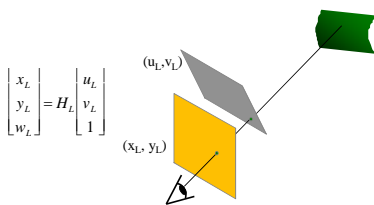
Rectification

Given a pair of images, transform both images so that epipolar lines are scan lines.



Rectification

Under perspective projection, the mapping from a plane to a plane is given by a projective transformation (aka homography).



Rectification

Under perspective projection, the mapping from a plane to a plane is given by a projective transformation (aka homography).

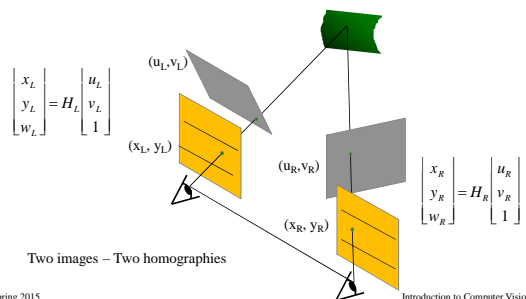


Image pair rectification

Simplify stereo matching by warping the images

Apply projective transformation so that epipolar lines correspond to horizontal scanlines

H should map epipole e to $(1,0,0)$, a point at infinity

H should minimize image distortion

$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = He$$

Note that rectified images usually not rectangular
See Text for complete method

CSE 152, Spring 2015 Introduction to Computer Vision

Rectification

Given a pair of images, transform both images so that epipolar lines are scan lines.

Input Images

CSE 152, Spring 2015 Introduction to Computer Vision

Rectification

Given a pair of images, transform both images so that epipolar lines are scan lines.

Rectified Images

See Section 7.3.7 for specific method

CSE 152, Spring 2015 Introduction to Computer Vision

Rectification

CSE 152, Spring 2015 Introduction to Computer Vision

Features on same epipolar line

CSE 152, Spring 2015 Introduction to Computer Vision

Using epipolar & constant Brightness constraints for stereo matching

For each epipolar line

For each pixel in the left image

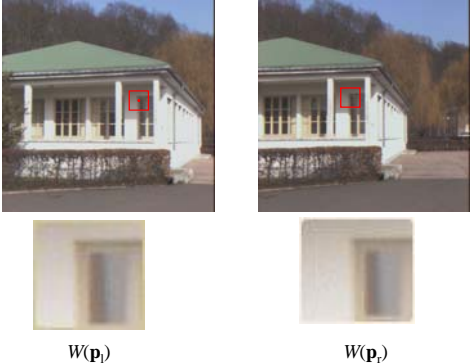
- compare with every pixel on same epipolar line in right image
 - pick pixel with minimum match cost
 - This will never work, so:

match **windows**

(Seitz)

CSE 152, Spring 2015 Introduction to Computer Vision

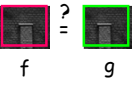
Finding Correspondences



$W(p_1)$ $W(p_2)$

CSE 152, Spring 2015 Introduction to Computer Vision

Comparing Windows:



$$SSD = \sum_{[i,j] \in R} (f(i,j) - g(i,j))^2$$

$$C_{fg} = \sum_{[i,j] \in R} f(i,j)g(i,j)$$

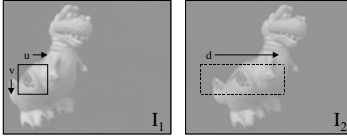
Most popular

For each window, match to closest window on epipolar line in other image.

(Camps)

CSE 152, Spring 2015 Introduction to Computer Vision

Correspondence Search Algorithm



```

For i = 1:nrows
  for j = 1:ncols
    best(i,j) = -1
    for k = mindisparity:maxdisparity
      c = Match_Metric(I1(i,j), I2(i,j+k), winsize)
      if (c > best(i,j))
        best(i,j) = c
        disparities(i,j) = k
      end
    end
  end
end
  
```

Complexity: $O(\text{nrows} * \text{ncols} * \text{disparities} * \text{winx} * \text{winy})$

CSE 152, Spring 2015 Introduction to Computer Vision

Match Metric Summary

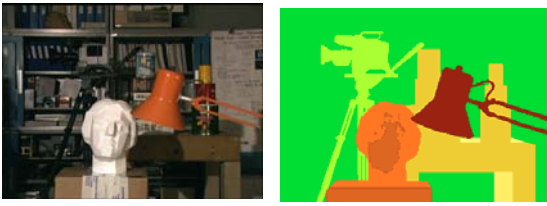
MATCH METRIC	DEFINITION
Normalized Cross-Correlation (NCC)	$\frac{\sum_{u,v} (I_1(u,v) - \bar{I}_1)(I_2(u+d,v) - \bar{I}_2)}{\sqrt{\sum_{u,v} (I_1(u,v) - \bar{I}_1)^2} \sqrt{\sum_{u,v} (I_2(u+d,v) - \bar{I}_2)^2}}$
Sum of Squared Differences (SSD)	$\sum_{u,v} (I_1(u,v) - I_2(u+d,v))^2$
Normalized SSD	$\frac{\sum_{u,v} (I_1(u,v) - \bar{I}_1)(I_2(u+d,v) - \bar{I}_2)}{\sqrt{\sum_{u,v} (I_1(u,v) - \bar{I}_1)^2} \sqrt{\sum_{u,v} (I_2(u+d,v) - \bar{I}_2)^2}}$
Sum of Absolute Differences (SAD)	$\sum_{u,v} I_1(u,v) - I_2(u+d,v) $
Zero Mean SAD	$\sum_{u,v} I_1(u,v) - \bar{I}_1 - (I_2(u+d,v) - \bar{I}_2) $
Rank	$I_1(u,v) = \sum_{m,n} I_2(m,n) < I_1(u,v)$
Census	$I_1(u,v) = \text{BITSTRING}_{m,n} (I_2(m,n) < I_1(u,v))$ $\sum_{m,n} \text{HAMMING}(I_1(u,v), I_2(u+d,v))$

These two are actually the same

CSE 152, Spring 2015 Introduction to Computer Vision

Stereo results

– Data from University of Tsukuba



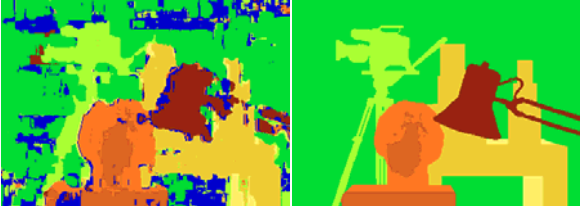
Scene Ground truth

Color coded. For each pixel, green is further and red is closer

(Seitz)

CSE 152, Spring 2015 Introduction to Computer Vision

Results with window correlation



Window-based matching (best window size) Ground truth

(Seitz)

CSE 152, Spring 2015 Introduction to Computer Vision

Results with better method



Using global optimization

Boykov et al., [Fast Semantics-Based Image Segmentation via Graph Cuts](#),
International Conference on Computer Vision, September 1999.

(Seitz)

Ground truth

CSE 152, Spring 2015

Introduction to Computer Vision

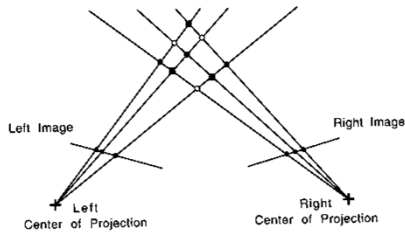
Some Issues

- Epipolar ordering
- Ambiguity
- Window size
- Window shape
- Lighting
- Half occluded regions

CSE 152, Spring 2015

Introduction to Computer Vision

A challenge: Multiple Interpretations

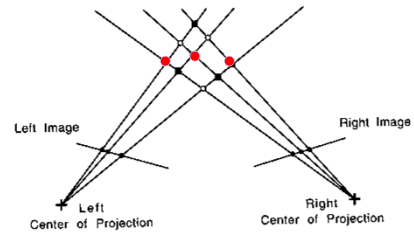


Each feature on left epipolar line match one and only one feature on right epipolar line.

CSE 152, Spring 2015

Introduction to Computer Vision

Multiple Interpretations

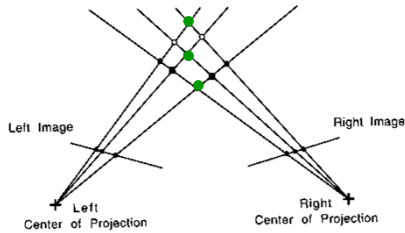


Each feature on left epipolar line match one and only one feature on right epipolar line.

CSE 152, Spring 2015

Introduction to Computer Vision

Multiple Interpretations

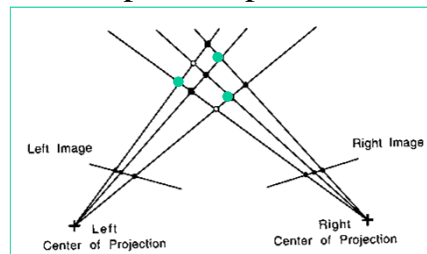


Each feature on left epipolar line match one and only one feature on right epipolar line.

CSE 152, Spring 2015

Introduction to Computer Vision

Multiple Interpretations



Each feature on left epipolar line match one and only one feature on right epipolar line.

CSE 152, Spring 2015

Introduction to Computer Vision

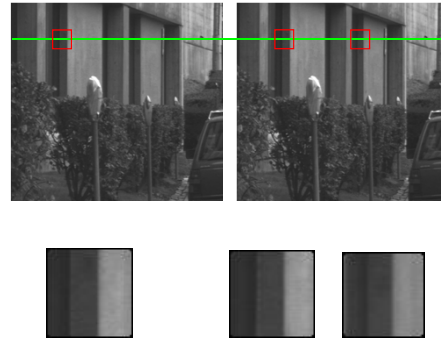
Some Issues

- Epipolar ordering
- **Ambiguity**
- Window size
- Window shape
- Lighting
- Half occluded regions

CSE 152, Spring 2015

Introduction to Computer Vision

Ambiguity



CSE 152, Spring 2015

Introduction to Computer Vision

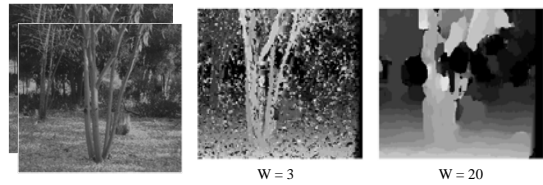
Some Issues

- Epipolar ordering
- Ambiguity
- **Window size**
- Window shape
- Lighting
- Half occluded regions

CSE 152, Spring 2015

Introduction to Computer Vision

Window size



- Effect of window size

Better results with *adaptive window*

- T. Kanade and M. Okutomi, [A Stereo Matching Algorithm with an Adaptive Window](#), *Theory and Experiments*, Proc. International Conference on Robotics and Automation, 1991.
- D. Scharstein and R. Szeliski, [Stereoscopic Image Matching with a Dynamic Programming Approach](#), *International Journal of Computer Vision*, 28(2):155-174, July 1998

(Seitz)

CSE 152, Spring 2015

Introduction to Computer Vision

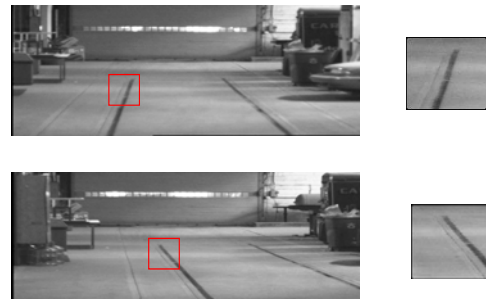
Some Issues

- Epipolar ordering
- Ambiguity
- Window size
- **Window shape**
- Lighting
- Half occluded regions

CSE 152, Spring 2015

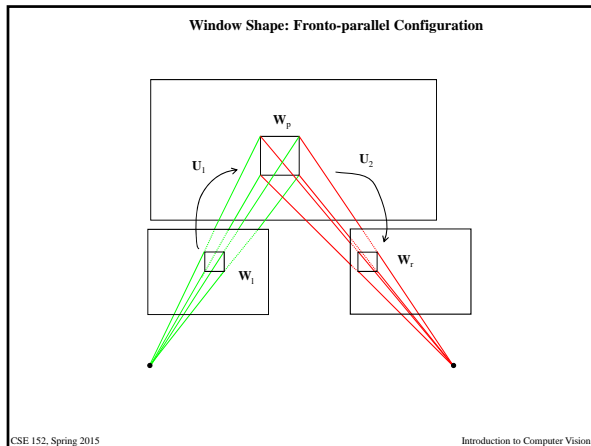
Introduction to Computer Vision

Window Shape and Forshortening

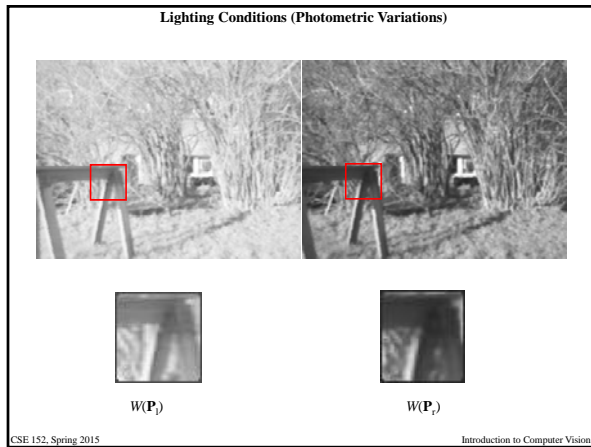


CSE 152, Spring 2015

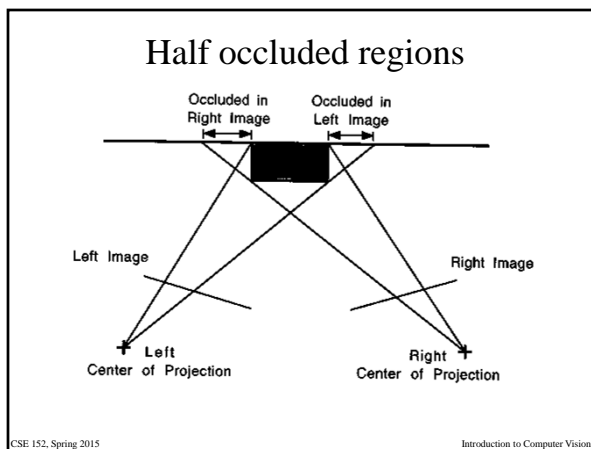
Introduction to Computer Vision



- ## Some Issues
- Epipolar ordering
 - Window size
 - Ambiguity
 - Window shape
 - **Lighting**
 - Half occluded regions
- CSE 152, Spring 2015 Introduction to Computer Vision



- ## Some Issues
- Epipolar ordering
 - Ambiguity
 - Window size
 - Window shape
 - Lighting
 - **Half occluded regions**
- CSE 152, Spring 2015 Introduction to Computer Vision

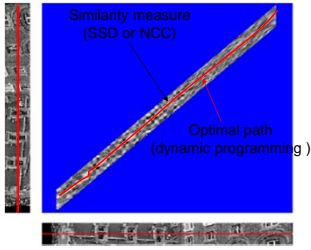


Summary of Stereo Constraints

CONSTRAINT	BRIEF DESCRIPTION
1-D Epipolar Search	Arbitrary images of the same scene may be rectified based on epipolar geometry such that stereo matches lie along one-dimensional scanlines. This reduces the computational complexity and also reduces the likelihood of false matches.
Monotonic Ordering	Points along an epipolar scanline appear in the same order in both stereo images, assuming that all objects in the scene are approximately the same distance from the cameras.
Image Brightness Constancy	Assuming Lambertian surfaces, the brightness of corresponding points in stereo images are the same.
Match Uniqueness	For every point in one stereo image, there is at most one corresponding point in the other image.
Disparity Continuity	Disparities vary smoothly (i.e. disparity gradient is small) over most of the image. This assumption is violated at object boundaries.
Disparity Limit	The search space may be reduced significantly by limiting the disparity range, reducing both computational complexity and the likelihood of false matches.
Fronto-Parallel Surfaces	The implicit assumption made by area-based matching is that objects have fronto-parallel surfaces (i.e. depth is constant within the region of local support). This assumption is violated by sloping and creased surfaces.
Feature Similarity	Corresponding features must be similar (e.g. edges must have roughly the same length and orientation).
Structural Grouping	Corresponding feature groupings and their connectivity must be consistent.

CSE 152, Spring 2015 (From G. Hager) Introduction to Computer Vision

Stereo matching



- Constraints
- epipolar
 - ordering
 - uniqueness
 - disparity limit
 - disparity gradient limit
- Trade-off
- Matching cost (data)
 - Discontinuities (prior)

(From Pollefeys)

(Cox et al. CVGIP'96; Koch'96; Falkenhagen '97;
Van Meerbergen, Vergauwen, Pollefeys, VanGool IJCV'02)

CSE 152, Spring 2015

Introduction to Computer Vision