

Lecture 10: CS677

Sept 21, 2017

1

Admin and Review

- HW2 due Sept 26; 9AM
- Exam 1, Oct 10, class period
 - Details to follow next week
- Previous class
 - Hough Transform
 - Harris corner detector
 - SIFT features
- Today's objective
 - Homography
 - Stereo geometry
 - Stereo matching

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2

Other Feature Detectors (in OpenCV)

- SURF: similar to SIFT but faster
- BRIEF: shorter descriptors
- FAST: faster corner detector
- ORB (FAST + BRIEF): claims to be like SIFT/SURF but faster, smaller and without patent issues
- More in:
 - [https://en.wikipedia.org/wiki/Feature_detection_\(computer_vision\)](https://en.wikipedia.org/wiki/Feature_detection_(computer_vision))
- For our course, knowledge of SIFT is sufficient

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3

Homography

- Consider images of a points, P , in a given plane Π .
- In general, $p = 1/z MP$, where M is a 3×4 matrix
- For points on a plane, we can choose world coordinate system so Z axis is normal to plane, $z=0$ for all points
- There is still some matrix M , which projects P to p
 - M includes the rotations and translations to transform world coordinates to camera coordinates
 - In computing MP , note that third coordinate of P is 0 so we can drop the third column of M , giving a 3×3 matrix, say H , that maps $(X,Y,1)$ to (wx', wy', w)
- Note that H is invertible
 - Geometrically: given image point, construct a ray and intersect with known plane.

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4

Homography Mapping

- Now consider images of the point in two different cameras
- Each is related to P by a homography matrix, hence both are also related by another homography matrix (3 x 3).
 - Alternately, we can invert one image to give P, then project again
- $\mathbf{X}_2 = \mathbf{H} \mathbf{X}_1$, where $\mathbf{X}_1, \mathbf{X}_2$ are image points in homogeneous coordinates, \mathbf{H} is a 3 x 3 matrix.

$$\begin{bmatrix} x_2' \\ y_2' \\ w_2' \end{bmatrix} = \begin{bmatrix} H_{11} & H_{12} & H_{13} \\ H_{21} & H_{22} & H_{23} \\ H_{31} & H_{32} & H_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ y_1 \\ 1 \end{bmatrix} \text{ and } \begin{bmatrix} x_2 \\ y_2 \end{bmatrix} = \begin{bmatrix} x_2' \\ y_2' \\ w_2' \end{bmatrix}$$

- Note: homography also relates two views of a *non-planar* scene if only the camera orientation changes (no translation)
 - Second camera essentially maps image points of first camera which are, of course, on a plane (draw figure)

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— In spherical coordinates, just a translation (draw figure) 5

Creating a Mosaic

- Two (or more) images, obtained by camera rotations only are related by a homography
- If we can estimate parameters, we can transform to a common coordinate system and create a panorama (mosaic)
- We can compute homography parameters if we can match sufficient number of points between the two images
- H has nine parameters but scaling is arbitrary, fix $H_{33} = 1$; we are left with 8 unknown parameters
- For each point match between two images, we get two linear equations (from the matrix equation on the previous slide)
- 4 point matches should suffice to compute all parameters.
- How to find four correct matches?

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6

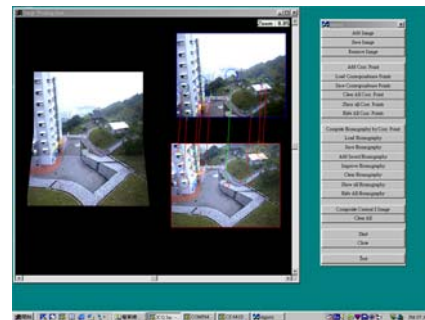
Matching using SIFT Features

- How to match SIFT keypoints?
 - Similarity of description vectors (128-D)
 - Use Euclidean distance or angle between vectors
- Global match
 - Local matches may be ambiguous (a point may match well with several other points in another image)
 - Global matching may help reduce the ambiguities
- Global constraints
 - Assuming points lie in a plane: homography
 - Useful for image mosaicing (assume background is planar)
 - For examples see, <http://www.cse.ust.hk/~cstws/research/641D/mosaic/>
 - General stereo pairs: epipolar geometry

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7

Example from UST.HK



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8

Global Matching: RANSAC Approach

- Random Sample and Consensus (RANSAC)
- Select set of matches (4 for homography), at random
- Compute the homography matrix
 - A linear solution is possible, in principle, subject to some numerical issues
- Find other matches consistent with this transformation
 - If enough points are found, it is a good estimate, refit the transformation to the expanded set, otherwise discard
- Repeat the above process and choose one that is the “best”
 - Best may be defined by fit error or number of point matches
- RANSAC is a general process, can be used for fitting any model
 - One alternative to Generalized Hough Transform

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9

RANSAC

- How many trials are needed before we find the desired model?
- Depends on the number of “outliers”
 - Suppose that w is probability that a match is good
 - If choose n matches randomly ($n=4$ for homography), probability that all samples are good is $(1-w^n)$
 - If we repeat k times, probability that all trials result in bad samples is $(1-w^n)^k$
 - Say $w=.5$ (half the matches are outliers), $n=4$
 - $w^4=.0625$, $1-w^4=.9375$
 - $k=1$, $z=.9375$; $k=10$, $z=.52$; $k=100$, $z=.0015$
 - Say $w=.1$ (90% of the matches are outliers), $n=4$
 - $w^4=.0001$, $1-w^4=.9999$
 - $k=1$, $z=.9999$; $k=10,000$ $z=.36$; $k=100,000$, $z=.00004$

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10

Localize Objects, an Example (from Lowe IJCV04)

- Locate Objects on the left, in the right image



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11

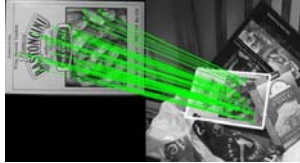
A Recognition Example (from Lowe IJCV04)



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12

Example from OpenCV



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13

Inference of 3-D

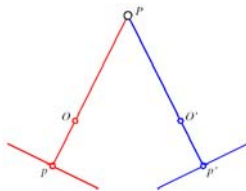
- From a single image
 - Intensity (shape from shading)
 - Texture (requires regular texture, not covered in class/book)
 - 2-D shape (details not covered in class/book)
- From multiple images, static scene
 - From multiple fixed cameras or from camera motion
 - Stereo: Two fixed cameras
 - *Structure from motion*: Estimate camera motion and 3-D scene structure
- From object motion, stationary camera
 - Must infer object motion in addition to 3-D surface
- Most difficult case is when both the sensor and some of the objects in the scene move

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14

Steps of Stereo Analysis

- Point correspondence between views
 - Difficulties/details covered later
- Triangulation
 - Assuming calibrated cameras, each view provides a ray, say R and R' , on which the 3-D point must lie: compute intersection

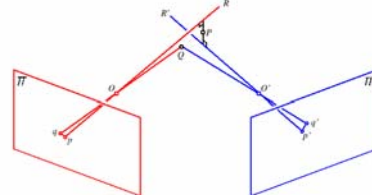


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15

Triangulation: with measurement errors

- Solve for P as mid point of line minimizing *skew* distance between R and R'
- An algebraic method to solve $Zp = MP$ and $Z'p' = M'P$
 - 3 unknowns, 4 linear equations; See FP, p. 201 for details
- Find Q that minimizes $d^2(p,q) + d^2(p',q')$
 - Requires non-linear optimization



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16

Some Examples

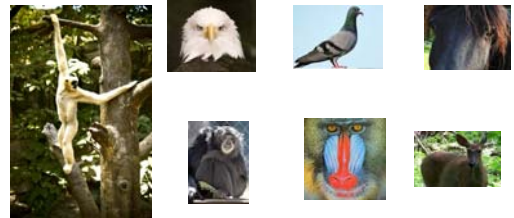


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17

Stereo in Animals

- For humans, effective mostly for a distance of few meters
- Few animals possess it: birds of prey, *brachiating* mammals...



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18

Next Class

- FP: Chapter 7, sections 7.1, 7.2, 7.4, 7.5

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19